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**STUDY AND ANALYSIS OF KNOWLEDGBASE OF MOLECULAR
SYSTEMS AND TO DEVELOP MODEL FOR PREDICTION OF
MOLECULAR STRUCTURE**

A Thesis Submitted to

SAURASHTRA UNIVERSITY

For the award of the degree of

**DOCTOR OF PHILOSOPHY
IN
COMPUTER SCIENCE**

IN THE FACULTY OF SCIENCE

Submitted by

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AUGUST – 2009

Guide's Certificate

*I hereby certify that Mr. Binod Kumar has completed his thesis for the doctorate degree entitled “**Study and Analysis of Knowledgebase of Molecular Systems and to Develop Model for Prediction of Molecular Structure**”. I further certify that the research work done by him is of his own and original and carried out under my guidance and supervision. For the thesis that he is submitting, he has not been conferred any degree, diploma or distinction by either Saurashtra University or any other University according to best of my knowledge.*

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Researcher Certificate

I certify that the development model for Molecular Structure Prediction and strategies derived by analysis and described in the thesis has been based on the literature survey , bibliographical references, and research paper from International journals and National journals and various conferences proceedings and through study literature available on various websites in respect of related areas.

Apart from these, all analysis, hypothesis, inferences and interpretations of data and strategy have been my own and original creation. The model has been developed is my own and original creation. Moreover, I declare that for the work done in the thesis, either the Saurashtra University or any other University has not conferred any degree, diploma or distinction on me before.

Place:

Date:

Binod Kumar

Acknowledgement

Research is to see what everybody else has seen, and to think what nobody else has thought. This work is also no exception. It is my pleasure to convey my gratitude to all those who have directly or indirectly contributed to make this work successful.

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Chapter -1
INTRODUCTION

Chapter 1

INTRODUCTION

1.1 Introduction

This research work aims to analyze experimental data about biochemical properties and their corresponding kinetics. In this research the attempt has been made to analyze protein and DNA structure using tools such as DAMBE and Jemboss. Some Molecular Visualization or Analysis tools are already developed that reads, analyses, and cross-correlates experimental information which is useful for chemist, Organist Chemist, Biochemist and Druggist.

Under this research the analysis of different chemical and biochemical substances including drugs using tools like ACD/ChemSketch and NMR Prediction have been performed. The information obtained by the way of analysis that facilitates for in depth understanding of structures and that makes possible for a quantification of new chemical structure.

In this research using ACD/ChemSketch compounds are stored in databases and SMILE code (Simplified Molecular Input Line Specification) is generated. A SMILE defines the molecules in the form of alphanumeric chains. In this research work chemical shift of every carbon atom of the molecule have been displayed by using NMR Prediction.

Under this research CML codes of molecules have been developed and that codes have been used for molecular information like symmetry, and atom and bond attributes. Here multiple observations of the same molecule like conformational analysis and NMR prediction have been performed.

Using Pubchem/NCBI additional miscellaneous information such as bioactivity analysis by structure & activity similarity and revised compound selection after addition of similar compounds have been analyzed.

Under the research work geometric optimization of molecules, chemical structure visualization and calculation of electronic absorption spectra of chemical structure have been performed using ArgusLab tool. In this research Single Entry Point Calculation, Molecular Orbital calculation on grids for plotting HOMO and LUMO and ESP Mapped Density calculations have been also performed.

Under the research work of different types of analysis like prediction of protein secondary structure, isoelectric point calculation etc. have been performed on nucleotide and protein sequence using DAMBE and Jemboss tools.

The aim of this research work is to develop a model for the prediction of molecular structure. In research work bioinformatics and cheminformatics approaches on molecule has been covered. In this research an integrated bioinformatics and cheminformatics approach has been discussed that enables retrieval and visualization of biological relationships across heterogeneous data sources. So, now it is getting importance to integrate biological information on large molecules and their interaction networks with programs chemical information on small drug molecules.

Bioinformaticists and Chemoinformaticists have working independently in their respective fields. But now development of small molecule drugs and small drug molecules with known properties has been utilized to study the functions of large networks of biological molecules in the fields of chemical biology.

The objective of this research work is to assist the organic and biochemist in each step of the synthesis planning process for prediction of molecular structure. This research work provides a series of methods and tools for chemical or biochemical applications. Built-in catalogs of fine chemicals or biochemical provide suitable starting materials for a

synthesis or molecular structure prediction target. Using *similarity searches* or *substructure* searches the connection between the target compound and available starting materials has been achieved.

This research work aims to search strategic bonds in target molecule for synthesis procedures. Structural criteria of each bond within the query molecule are also taken into account. In this research data mining tools has been used to predict physical properties of structures. In research work analysis on knowledgebase molecular system has been performed and a model has been developed that uses information to make decisions and suggest new strategies for chemistry and biochemistry problems.

The knowledgebase molecular system has three components:

A. Knowledgebase as Chemical Memory: An attempt has been made to concentrate on knowledge based data with an increasing number of chemical systems. Taking advantage of data sharing, each calculation increases the level of ‘experience’ of expert system extending the knowledge base upon which new hypothesis and chemical concept has been derived.

B. Data Mining: A component for increasing the chemical knowledge is extracting chemically meaningful data out of large scale chemical simulations with minimum human effort. The challenges lies in distinguishing data that is irrelevant for specific question under specific investigation for those that are important. To carry out this task an attempt has been made to concentrate on a knowledgebase system that process the molecular orbital and trace changes and similarities between molecules. Under this research visualization techniques have been used to enlarge scope of analysis.

C. Towards Artificial Chemical Intelligence: The final part of this research to formulate hypothesis based on data provided by molecules .Under this research work an attempt has been made for prediction of molecular structure. Finally, a research work

result has been collected and then analyzed using analysis tools and then evaluated the result.

1.2 The Research Area, Problem Domain and Literature Survey

Bioinformatics and management of scientific data are critical to support life science discovery. As computational models of proteins, cells and organisms become increasingly realistic much biology research has migrated from the wetlab to the computer. Successfully accomplishing the translation of biology *in silico*, however, requires access to a huge amount of information from across the research community. Much of information is currently available from publicly accessible data sources and more is being added daily. Unfortunately, scientists are not currently able to identify easily and exploit this information because of the variety of semantics, interfaces and data formats used by the underlying data sources. Providing biochemist, medical researcher and computer scientist with integrated access to all information they need a consistent format requires overcoming a large number of technical, social and political challenges.

In the last decade, biologist have experienced a fundamental revolution from traditional research and development (R&D) consisting in discovering and understanding genes , metabolic pathways and cellular mechanisms to large scale computer-based R&D that simulates the disease , the physiology , the molecular mechanism and pharmacology. This represents a shift away from life science's empirical roots in which it was an interactive process. Today it is systematic thematic and predictive with genomics, informatics and automation all playing a role. This fusion of biology and information science is expected to continue and expand for predictable futures. The first consequence of this revolution is the explosion of available data that biomolecular researchers have to exploit. For example, an average pharmaceutical company currently uses information from at least 40 databases [1] , each containing large amounts of data (e.g. as of June 2002, GenBank [2,3] provides access to 20,649,000,000 bases in 17,471,000 sequences) that can be analyzed using a variety of complex tools such as FASTA, BLAST etc.

Over past several years, bioinformatics has become both an all encompassing term for every thing relating to computer science and biology and an every trendy one. There are variety of reasons for this including : (1) As computational biology evolves and expands , the need for solutions to the data integration problems it faces increases; (2) the media are beginning to understand the implications of genomics revolution that has been going on the last 15 or more years ; (3) the recent headlines and debates surrounding the cloning of animals and humans ; and (4) to appear cutting edge , many companies have relabeled the work the work as they are doing as bioinformatics instead of geneticists , biologists or computer science.

The analysis of data sets is one of the most important tasks in investigation of properties of chemical or biochemical compounds. Especially in Drug Design, methods are used to characterize complete sets of chemical or biochemical compounds instead of describing individual molecule. Data Mining, i.e. the exploration of large amounts of data in search for consistent patterns, correlation or other systematic relationships, can be helpful tool to evaluate “hidden” information in a set of molecules. Finding the adequate information for representation of new chemical structures is one of the most important problems in chemical data mining.

With the progressive specialization in services and extensive use of computational methods the steady increase of data is barely manageable even by a team of scientist. Thereby the interest in specific information is pushed into backward while global information of complete sets of data is becoming more and more important. Thus, the recognition of superior information for complete data sets becomes one of the most important tasks for information management in science.

In Chemistry or Biochemistry the investigation of molecular structures and of their properties is one of the most important areas. In chemistry an own language and namespace for molecular exists, that is still in development stage. With increase of computational information processing several conventions and formats for chemical information have been developed.

But, in one of the most important communication media of modern times, the internet, the chemical language has been used only in a few applications. While a couple of databases were accessible via WWW, no service exists, that allows a data mining of chemical datasets by the use of this specific language.

The task of Data Mining in chemical or biochemical context is to evaluate “hidden” information in set of chemical data. One of the differences of Data Mining compared to conventional database queries is the production of new information that is used to characterize chemical data in a more general way. Generally, it is not possible to hold all of the potentially required information in a data set of chemical structures. Thus, the extraction of relevant information and production of reliable secondary information are important.

The similarity of two compounds concerning their biological activity is one of central tasks in the development of pharmaceutical products. A typical application is retrieval of structures with defined biological activity from a database. Biological activity is of special interest the development of drugs. The diversity of structures in a data set of drugs has been the interest for the synthesis of new compounds. With increasing variety of data set, the chance to find a new way of synthesis for a compound with similar biological property is increasing.

Therefore, finding the adequate information for representation of chemical structures is one of the basic problems in chemical data mining. Several methods have been developed in the last decades for the description of molecules including their chemical or biochemical properties.

1.3 Relevance of the Research

Data Mining Service Chemistry (DMSC) [4] is a project for the development and exploration of chemical data sets. With this service it is possible to analyze chemical or

biochemical data sets for molecular patterns and systematic relationships using the methods like Statistical analyses and neural networks of individual molecules.

System for Drug Discovery (QIS D2) [5] is a unique adaptive learning system designed to predict potential large-scale drug characteristics such as toxicity and efficacy. BioSpice is a set of software tools designed to represent and simulate cellular processes.

A new computer program is developed that describes, GRINSP (geometrically restrained inorganic structure prediction) [6], which allows the exploration of the possibilities of occurrence of 3-, 4-, 5- and 6-connected three-dimensional networks.

A global optimization method [7] is presented for predicting the minimum energy structure of small protein-like molecules. This method begins by collecting a large number of molecular conformations, each obtained by finding a local minimum of a potential energy function from a random starting point. The information from these conformers is then used to form a convex quadratic global underestimating function for the potential energy of all known conformers.

GenomeThreader [8] implements several data types in a reusable manner. Compared to its predecessor GeneSeqer, it is considerably faster, easier to maintain, and extensible. It is widely used for gene structure prediction.

The general approach [9] for the prediction of possible crystal structures consists of the global exploration of the energy landscape of the chemical system, with typical methods being simulated annealing or genetic algorithms. In the case of simulated annealing, combinations of model potentials and Ab initio calculations for the energy evaluation are state of the art.

The characteristics [10] of a free web-based spectral database for the chemical research community, containing ^{13}C NMR spectra data from more than 4000 natural compounds, and with a continuous increasing. This database allows flexible searching via chemical

structure, substructure, name, and family of compounds, as well as spectral features as chemical shift, allowing the structural elucidation of known and unknown compounds by comparison of ^{13}C NMR data.

In this research work planning has been made to provide a centralized access to a wide variety of data mining methods, like statistical processing and prediction of molecular structure. With this service it is possible to submit data sets or to compile a data set by extracting structures from chemical databases via Internet. For submitting or compiled data sets descriptors have been calculated with an extensive set of options. On the basis of these descriptors, several methods of data analysis have been performed on the data set.

1.4 Details of Remaining Chapters

This thesis is meant to be a major step in my personal interest in prediction of molecular structure.

Second chapter of this thesis provides an overview of tools like ACD/ChemSketch, NMR Prediction, Argus Lab, DAMBE and Jemboss. **ACD/Labs** is used for developed molecular structures, reactions, and schematic diagrams and calculated chemical properties of different substances (chemical and biochemical). **NMR Prediction** tool is used to perform estimation of $^1\text{H-NMR}$ and $^{13}\text{C-NMR}$ of different substances. **ArgusLab** tool is used to build chemical structure and optimized its geometry. **DAMBE** tool is used to manipulate and analyze molecular sequence data. **Jemboss** can perform activities on sequences like predicting protein secondary structure etc. **CML** is designed to represent molecular information. **SMILES** (Simplified Molecular Input Line Entry System) is a line notation for entering and representing molecules.

Third chapter of this thesis provides an overview of pair wise sequence alignment and multiple sequence alignment. In this chapter alignment score and gap penalty between sequences has been calculated. Multiple sequence alignment is useful in finding patterns

in nucleotide sequences and for identifying structural and functional domains in protein families. The method of converting MSA to a phylogenetic tree has been used to reduce the problem of a multiple alignment to an iterative process of pair-wise alignments.

Forth chapter of this thesis provides an overview of sequence alignment tools like BLAST and FASTA. Here their working methods and the syntax used by these tools has been discussed. FASTA uses algorithm to search for similarities between one sequence and any group of sequences of same type (nucleic acid or protein) as the query sequence. BLAST uses a heuristic algorithm that seeks local as opposed to global alignments and is therefore able to detect relationships among sequences that share only isolated regions of similarity.

Fifth chapter of this thesis provides an overview of protein structure and Cheminformatics. The subunits of a protein are amino acids. The primary structure is the sequence of residues in the polypeptide chain. Secondary structure is a local regularly occurring structure in proteins and is mainly formed through hydrogen bonds between backbone atoms. Tertiary structure describes the packing of alpha-helices, beta-sheets and random coils with respect to each other on the level of one whole polypeptide chain. Ab Initio method and Heuristic methods have been used for protein structure prediction.

Sixth chapter of this thesis shows the strong interaction between representation and the methods used for data analysis: molecular representation need to capture relevant information and be compatible with the statistical methods used to analyze the data. The chapters review molecular representations and put focus on model validation using statistics, visualization methods, and standardization approaches.

1.5 References

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Chapter -2

Computational techniques, Tools and Technologies To support Bioinformatics

Chapter 2

COMPUTATIONAL TECHNIQUES, TOOLS AND TECHNOLOGIES TO SUPPORT BIOINFORMATICS

2.1 Introduction

Under this research work tools like ACD/ChemSketch, NMR Prediction, Argus Lab, DAMBE and Jemboss have been discussed.

ACD/Labs [1] has been used for developed molecular structures, reactions, and schematic diagrams and calculated chemical properties of different substances (chemical and biochemical). **NMR Prediction** [2] tool has been used to perform estimation of **¹H-NMR** and **¹³C-NMR** of different substances. The proton shift estimation program has been invoked by this tool and the result has been displayed written to the drawing. Sometimes the drawing is changed to allow the display of certain shifts. **ArgusLab** [3] tool has been used to build chemical structure and its geometry has been optimized. It is being used for visualization of frontier p molecular orbitals of chemical structure.

DAMBE [4] tool has been used to manipulate and analyze molecular sequence data. DAMBE is used for calculation of genetic distances or phylogenetic reconstruction. **Jemboss** [5] has been used for interactively editing sequence alignment. Different activities on sequences have been performed by this tool like Editing Functions, Locking Sequences, Trim Sequences, Colour Schemes, Scoring Matrix, Consensus Sequence, Identity Table and Consensus Plot etc.

2.2 ACD/ChemSketch

2.2.1 Introduction

ACD/ChemSketch is the powerful all-purpose chemical drawing and graphics package from ACD/Labs developed to help chemists quickly and easily draw molecular

structures, reactions, and schematic diagrams, calculate chemical properties, and design professional reports and presentations. ACD/Labs has been fully dedicated to building integrated solutions that enable data transfer and connection with in chemical organizations.

ChemBasic is a simple, convenient, and functionally rich *programming language* for presentation and manipulation of molecular structure related objects and all the contents of ACD /Labs current and future programs. ChemBasic is founded on, and fully integrated with, ACD/Labs existing functionality. At the same time, ChemBasic has all of the things a programming language should have: numeric and string variables, arrays, flow control and conditional operators, input output procedures, etc.

ChemBasic inherits from *generic BASIC* and some of its extensions. Most evident is a product of Microsoft's Visual Basic for Applications (VBA). ChemBasic is designed as object oriented language. This means that all the chemistry related things are described as objects—that is, specific data structures which correspond to molecules, conformations, etc. I can design multi item input forms using ChemBasic programs using ACD/Forms Manager.

2.2.2 ACD/ChemSketch includes

- Structure mode for drawing chemical structures and calculating their properties.
- Draw mode or text and graphics processing.
- Additional modules that extend the ChemSketch possibilities (most of them should be purchased separately).

Structure mode. General information

In the Structure mode, following actions can be performed:

- Chemical structures can be drawn using the buttons located on the Structure toolbar, Atoms toolbar and References toolbar.

- For the selected structure the molar refractivity, molar volume, parachor, index of refraction, surface tension, density, and some other physicochemical properties can be calculated.
- Chemical structures can be finding according to their systematic or non-systematic names, therapeutic category or inhibited enzyme by using the integrated ACD/Dictionary .
- Most favorable tautomeric forms of the drawn structure can be checked and can be automatically corrected the structure by using the integrated Tautomeric Forms function on the Structure toolbar.
- An optimized 3D model of a 2D structure can be get.

In **Draw mode**, the following actions can be performed:

- Graphical objects such as lines, arrows, rectangles, ellipses, arcs, polylines, and polygons can be drawn by using the Drawing toolbar buttons.
- Objects can be manipulated.
- Location of objects on the page with a ruler and gridlines can be controlled.

2.2.3 Structure Representation

Antialiasing has been supported by ACD/ChemSketch that displays chemical structures drawn with smooth lines. Antialiasing is a computer rendering technique that blurs the hard edges and adds shaded pixels to create the appearance of smoothness. This addresses the common issue with printers and computer monitors, when, due to the relatively low resolution, the tilted lines appear “stairlike” instead of smooth straight lines or curves. For example, compare the two pictures below:

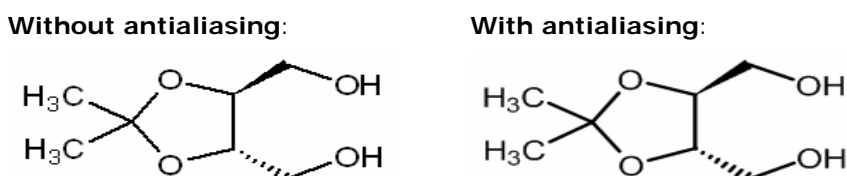


Figure 2.1: Stairlike curves

Treat some bonds to metal atoms as coordination bonds

ACD/Labs support the usage of a special *coordination bond* to represent a specific bonding between a ligand and a metal center in coordination structures. Such a bond indicates a connection but does not affect the valence of the corresponding atoms. However, often use of the regular *single bond* to represent a coordination that leads to formal violation of valence rules. Such a violation is marked in ACD/ChemSketch by “crossed atoms”.

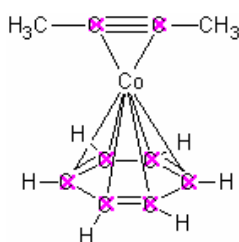
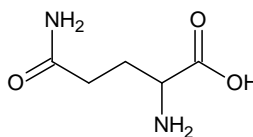


Figure 2.2 : Coordination Bond Representation

2.2.4 IUPAC International Chemical Identifier (InChI)

The IUPAC International Chemical Identifier (InChI™) is a non-proprietary identifier enabling unambiguous identification of chemical substances for electronic handling of chemical structural information. InChI codes significantly expand the use of InChI encoding for structure specification and searching over the Internet. For example:



InChI=1/C5H10N2O3/c6-3(5(9)10)1-2-4(7)8/h3H,1-2,6H2,(H2,7,8)(H,9,10)

Figure 2.3:2,5-diamino-5-oxopentanoic acid

InChI generation options include an option for InChIKey generation:

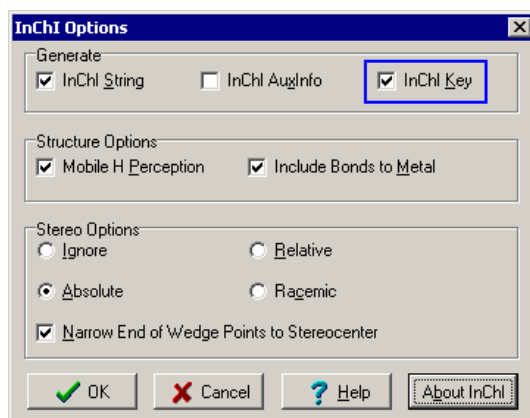


Figure 2.4: InChIKey Option

For quick access of InChI generation, a special button "Generate InChI" has been added to the top toolbar:



Figure 2.5: Generate InChIKey Button

2.3 NMR Prediction

2.3.1 Introduction

This software performs different estimation of a structure. It estimates $^1\text{H-NMR}$. It invokes the proton shift estimation program and displays the results written to the drawing. Sometimes the drawing is changed to allow the display of certain shifts. It estimates $^{13}\text{C-NMR}$. It invokes the carbon-13 shift estimation program and displays the results written to the drawing. **Show Protocol** command displays detailed information about the most recently invoked shift estimation. **Calculate 3D Coordinates** command displays the currently drawn structure as a 3D display in its own window. The molecule can be rotated by moving the mouse.

2.3.2 Taking example of Glutamyl

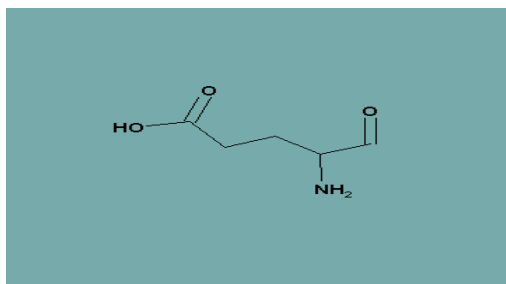


Figure 2.6: Example of Glutamyl

$^1\text{H-NMR}$ spectra of Glutamyl

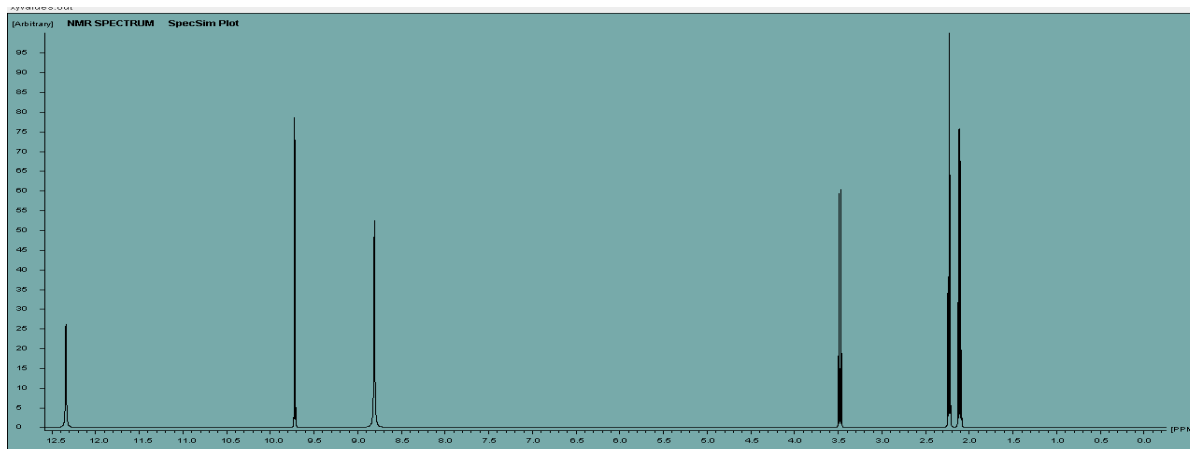


Figure 2.7: $^1\text{H-NMR}$ spectra of Glutamyl

Table 2.1: Shift Prediction Protocol

Node	Shift	Base + Inc.	Comment (ppm rel. to TMS)
NH2	8.81	2.00	amine
		6.81	general corrections
CH	3.48	1.50	methine
		1.13	1 alpha -N
		0.86	1 alpha -C=O
		-0.01	1 beta -C
CH	9.72	9.60	CHO
		0.12	1 -C
CH2	2.11	1.37	methylene
		0.22	1 beta -N
		0.29	1 beta -C=O
		0.23	1 beta -C(=O)O
CH2	2.23	1.37	methylene
		0.90	1 alpha -C(=O)O
		-0.04	1 beta -C
OH	12.34	11.00	carboxylic acid

2.4 ArgusLab

2.4.1 Introduction

Argus Lab performing following capabilities:

- Build chemical structure and optimize its geometry.
- Visualize frontier p molecular orbital's of chemical structure.
- Calculate the electronic absorption spectra of chemical structure.
- Use a surface to visualize the spin-density in a molecule with unpaired spins.
- Make a surface that maps the electrostatic potential to the electron density.
- Using surfaces to see what happens to the electron density when a molecule absorbs light.

2.4.2 Building of Benzene

Benzene structure can be built from scratch and its geometry can be optimized. After addition of atoms from editor Benzene molecule can be generated and bonds can be made automatically. Following structure can be shown.

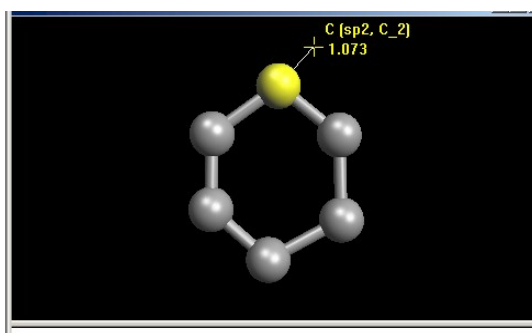


Figure 2.8: Building of Benzene

All bonds can be shown in that Benzene structure.

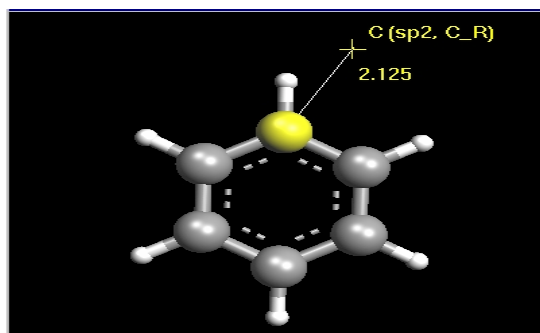


Figure 2.9: Bonds in Benzene Structure

Visualize the Building of Benzene

ArgusLab with generated MO grid files.

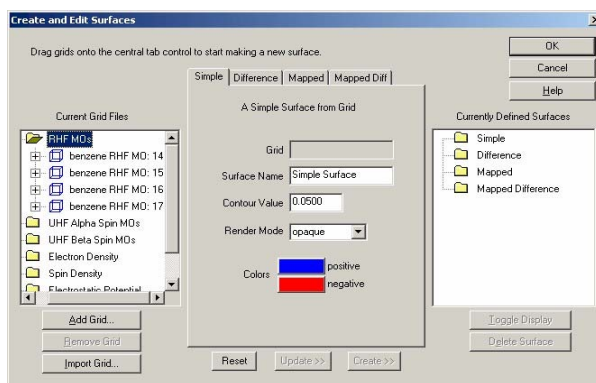


Figure 2.10: Building of Benzene Visualization

Visualization of MOs of Benzene

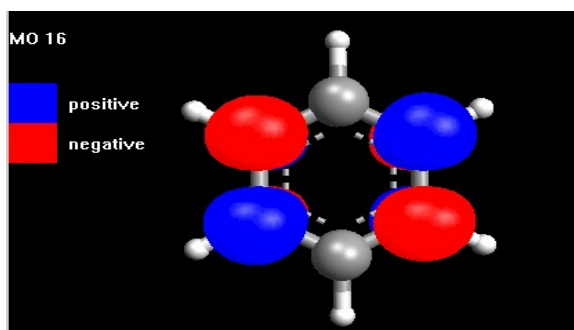


Figure 2.11: Visualization of MOs of Benzene

Calculating the electronic UV/Visible absorption spectrum of Benzene

The electronic excited states of benzene can be calculated using the semi-empirical ZINDO method which is parameterized for low-energy excited states of organic and organo-metallic molecules.

Calculating the ZINDO Electronic Spectra of a Molecule

The calculation consists of a ground-state closed shell SCF calculation followed by a configuration interaction calculation, using single-excited configurations, to solve for the excited states. Currently, only singlet excited states can be calculated.

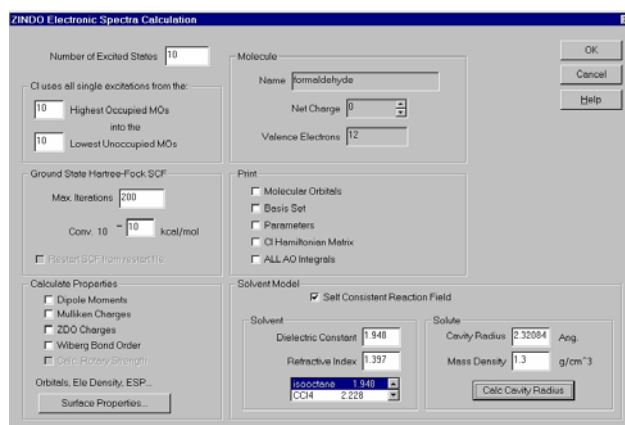


Figure 2.12: Calculating the ZINDO Electronic Spectra of a Molecule

Making an electrostatic potential-mapped electron density surface

ArgusLab can generate Mapped surfaces. These are surfaces where one property is mapped onto a surface created by another property. The most popular example of this is to map the electrostatic potential (ESP) onto a surface of the electron density. In an ESP-mapped density surface, the electron density surface gives the shape of the surface while the value of the ESP on that surface gives the colors.

The electrostatic potential is the potential energy felt by a positive "test" charge at a particular point in space. If the ESP is negative, this is a region of stability for the positive test charge. Conversely, if the ESP is positive, this is a region of relative instability for the positive test charge. Thus, an ESP-mapped density surface can be used to show regions of a molecule that might be more favorable to nucleophilic or electrophilic attack, making these types of surfaces useful for qualitative interpretations of chemical reactivity.

Steps for calculating the following surface:

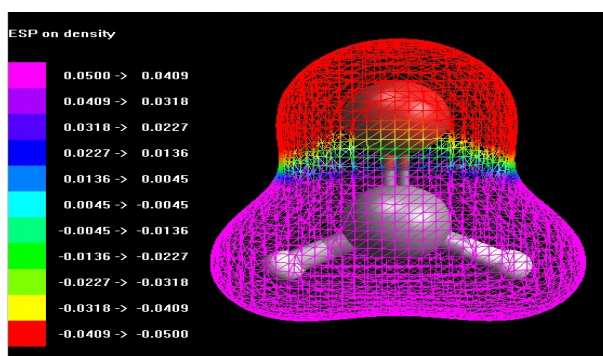


Figure 2.13: The surface in a mesh rendering to make it easier to see the underlying molecular structure

This is an ESP-mapped density surface of formaldehyde. The colors are the value of the ESP at the points on the electron density surface. The color map is given on the left. The large red region around the oxygen-end of the molecule. There is enhanced electron density here. The red color indicates the most negative regions of the electrostatic potential where a positive test charge would have favorable interaction energy. The hydrogen-end of the molecule, with the magenta color, shows regions of relatively unfavorable energy for the ESP.

Making the Surface: Generate the grid data

All surfaces are constructed from grid data that is generated from a calculation. To generate the grid data, a single-point energy calculation of formaldehyde can be run.

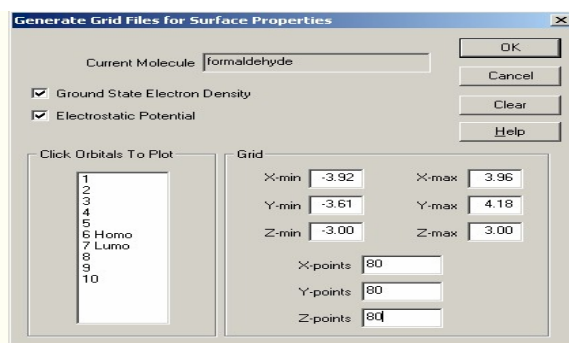


Figure 2.14: Generate the Grid Data

Seeing the lone pairs on the oxygen

Some of the surface's settings can be altered to visualize the lone pair electron density on the oxygen.

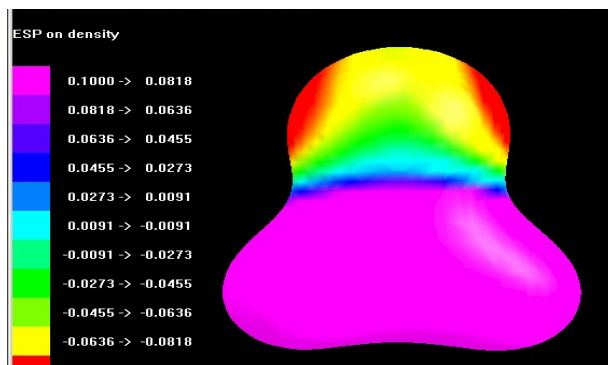


Figure 2.15: The lone pairs on the oxygen

Using surfaces to see change in the electron density when a molecule absorbs light

Here the first excited state of simple molecule formaldehyde (CH_2O) has been examined. The highest occupied molecular orbital (HOMO) of formaldehyde is a non-bonding type MO that is in the plane of the molecule. The lowest unoccupied molecular orbital (LUMO) is a p MO perpendicular to the plane of the molecule. The first excited state of formaldehyde is an $n \rightarrow p^*$ transition that is composed almost exclusively of the HOMO \rightarrow LUMO transition.

Calculate the electronic absorption spectra of formaldehyde

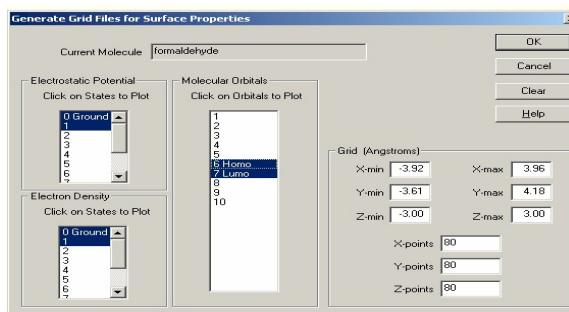


Figure 2.16 : Visualizing the frontier MOs

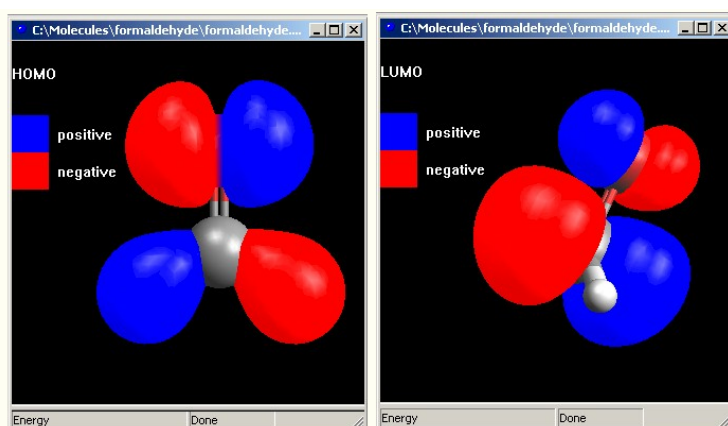


Figure 2.17 : Visualizing the frontier MOs(Diagram)

Electron Density Difference

Different surface can be made to show the difference of the excited state minus the ground state electron density.

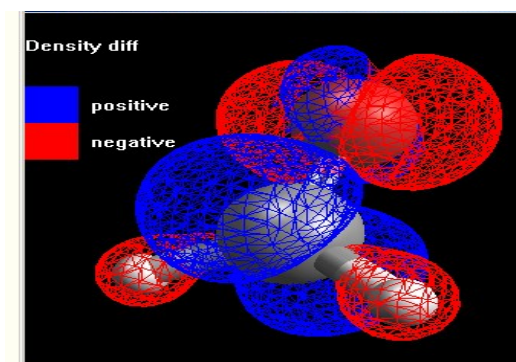


Figure 2.18: Electron Density Difference of benzene

Mapping the ESP difference onto the electron density

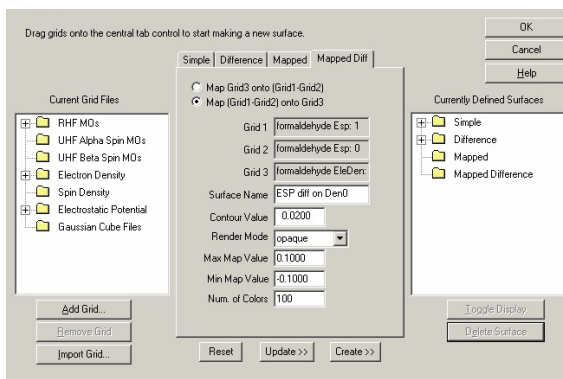


Figure 2.19: Mapping the ESP difference onto the electron density

2.5 DAMBE

DAMBE stands for Data Analysis in Molecular Biology and Evolution. It is an integrated software package for retrieving, organizing, manipulating aligning and analyzing molecular sequence data. Allele frequency data can also be used by DAMBE for calculating genetic distances or phylogenetic reconstruction.

2.5.1 Main Feature

DAMBE's main features can be classified into the following five categories:

- 1 Database and network functions:
 - (a) Molecular sequences can be directly read from GenBank or other networked computers;
 - (b) Specific sequences from GenBank sequences can be extracted by using information contained in the FEATURES table of GenBank sequence files;
- 2 Sequence conversion and manipulation utilities:
 - (a) It can be automatically detected and can be converted to 18 most commonly used molecular data formats;
 - (b) Complementary sequences can be getting.
 - (c) Protein-coding nucleotide sequences can be translated into amino acid sequences, with 12 different genetic codes implemented;
 - (d) Sequence can be aligned,
 - (e) Site-wise unresolved nucleotide, amino acid or codon sites can be deleted.
 - (f) Particular sites can be extracted, e.g., first, second or third codon positions, for particular analyses;
- 3 Sequence analysis can be focused on, factors affecting the frequency parameters in substitution models:
 - (a) Nucleotide and Dinucleotide frequencies

- (b) Codon frequencies
 - (c) Amino acid frequencies
 - (d) Amino acid properties can be plotted along the sequence; with the following properties implemented:
 - Polarity
 - Polar requirement
 - Chemical composition of the side chain
 - Volume
 - Hydrophathy
 - Isoelectric point
 - Aromaticity
- 4 Basic comparative sequence analysis can be performed that focus on factors affecting the rate ratio parameters in substitution models:
- (a) Nucleotide substitution pattern
 - (b) Codon substitution pattern
 - (c) Amino acid substitution pattern
 - (d) Substitution saturation
- 5 Advanced comparative sequence analysis can be performed
- (a) Phylogenetic reconstruction based on the distance, maximum parsimony and maximum likelihood methods
 - (b) Reconstruction of ancestral sequences
 - (c) Testing the molecular clock hypothesis
 - (d) Evaluating relative statistical support of alternative phylogenetic hypotheses (e.g., alternative phylogenetic trees)
 - (e) Fitting probability distributions to substitution data over sites.

2.5.2 Sequence Analysis

This command computes the nucleotide and dinucleotide frequencies.

A part of a sample output (for one sequence) is shown below:

```

=====
                A      C      G      U      Other      Sum (ACGU)
=====
FLAHAOHF
Freq           339    211    209    210         0         969
Prop.          .22     35     .22     .22     .22
=====
                A      C      G      U      Sum
=====

Obs. A         117     75     72     75         339
Exp.           119     74     73     74

Obs. C          88     52     26     44         210
Exp.           74     46     45     46

Obs. G          79     32     53     45         209
Exp.           73     46     45     45

Obs. U          55     52     57     46         210
Exp.           74     46     45     46

Subtotal       339     211     208     210         968

```

The output is of two parts for each sequence, the first part lists the nucleotide frequencies, with "Other" stands for all characters that are not "acgtu", e.g., "-?". The second part lists the di-nucleotide frequencies and the expected frequencies when there is no association or repulsion between nucleotides (i.e., the probability of two nucleotides sitting next to each other depends entirely on their frequencies). The di-nucleotides are counted from the beginning to the end of the sequence, with the nucleotides on the left column being the first, and those on the top row being the second, of the dinucleotide. From the first part of the output, it has been interpreted that A is being used more frequently than other nucleotides.

2.5.3 Codon Frequency

This opens a dialog box for computing codon frequencies and codon usage bias. A part of a default sample output, based on a segment of the Influenza A viruses, is shown below:

Output from sequences in file C:\MS\virus\virus.rst on

Sequence length = 969 (After excluding '?', '-' and 'n'.)

Number of codons = 323

From pooled sequences

AA	Codon	Mean Number (Sum)	RSCU
A	GCA	8.0 (32)	1.87
	GCC	3.3 (13)	0.75
	GCG	2.3 (9)	0.56
	GCU	3.5 (14)	0.82
V	GUA	6.5 (26)	1.29
	GUC	4.0 (16)	0.80
	GUG	0.3 (21)	1.04
	GUU	4.5 (18)	0.88

The codon usage table is based on the following sequences:

- 1 FLAHAOHF
- 2 FLAHA1N
- 3 IAU11858
- 4 IVHATG391

	CodSite	A	C	G	U	Sum
1	Freq.	433	221	357	281	1292
	Prop.	.34	.17	.28	.22	1
2	Freq.	428	318	240	306	1292
	Prop.	.33	.25	.19	.24	1
3	Freq.	461	319	228	284	1292
	Prop.	.36	.25	.18	.22	1

The output is in two parts. The first is a table of codon frequencies categorized into codon families, and the second lists nucleotide frequencies separately for each of the three codon positions designated as CodSite in the output.

2.5.4 Nonsynonymous codon substitution:

The sequence pairs available for selection on the left list depends on what input file format that is being used. If input format is NOT the RST format, then the number of possible sequence pairs is simply $N*(N-1)/2$. A partial sample output for a set of elongation factor 1-sequences (for only one pair-wise comparison between two chelicerate species) is shown below:

N	Cod1	Cod2	AA1	AA2	DG	DM
=====						
node#8 vs. node#9						
037	AGG	AGU	R	S	109.00	2.73
106	ACC	GCC	T	A	58.00	0.90
150	UCA	CCA	S	P	73.00	0.55
241	AAG	GAG	K	E	53.00	1.05
244	GUU	CUU	V	L	32.00	0.91
357	GAA	GAC	E	D	61.00	1.46

Mean					64.33	1.27
Num NS: 6						
node#11 vs. Bra90058						
26	UAC	UUC	Y	F	22.00	0.48
106	GCC	AAC	A	N	110.00	1.77
117	ACU	UCU	T	S	58.00	0.89
147	GCC	ACC	A	T	58.00	0.90
148	AAG	AAC	K	N	94.00	1.83
149	AUG	UUG	M	L	14.00	0.41
156	AAC	GCC	N	A	110.00	1.77
171	GAA	GAC	E	D	61.00	1.46
213	AUG	AUC	M	I	10.00	0.29
272	AAC	AGC	N	S	46.00	1.31
282	UCU	UAC	S	Y	143.00	3.32

Mean					66.00	1.31
Num NS: 11						

Pair-wise comparisons along the tree are either between internal nodes, or between an internal node and a terminal node. This information is shown at the beginning of each pair-wise comparison. The first column shows the sequential numbering of codons along the DNA sequences (after deleting unresolved codons). The second and third columns show which codon pairs are involved in the substitution, and the fourth and fifth columns show the corresponding amino acids.

2.6 Jemboss

2.6.1 Introduction

Jemboss is developed by the EMBOSS team and is a graphical interface to the European Molecular Biology Open Software Suite, EMBOSS. Jemboss incorporates the 200+ applications of both the EMBOSS and EMBASSY packages.

The job manager is used to monitor the status of batch processes. These are those EMBOSS applications that are computationally intensive. Instead of waiting for the results these processes are submitted as batch, which frees the interface for other analyses to be carried out. This product includes code licensed from RSA Data Security.

2.6.2 Local and Remote File Manager

The users local and the remote file systems can be displayed. The local files are those stored on the computer that Jemboss is being run on. The remote files are the users files located on the server machine that runs the EMBOSS applications.

The activities performed by file manager are:

- Drag and Drop Files
- Transferring Files
- Refresh' File Manager
- Multiple File Selection

2.6.3 Jemboss Results Manager

Applications in Jemboss can be run '**interactively**' or in '**batch**' mode. Interactive applications wait for the process to finish and the results pop up on the screen. Batch process run in the background so that other tasks can be performed in Jemboss while the application is running. In both cases the results are stored on the server machine and can be retrieved at any time.

2.6.4 Sequence List

This window allows us to store their commonly used sequences.

An EMBOSS list file contains "references" to sequences, for example the file has been looked like: opsd_abyko.fasta, sw: opsd_xenla, sw: opsd_c* and @another_list etc.

The sequence length has been calculated by 'Calculate sequence attributes' under the 'Tools' menu. The sequence start and end positions has been displayed.

2.6.5 Jemboss Alignment Editor

The Jemboss Alignment Editor has been used interactively to edit a sequence alignment (read in fasta or MSF format). It can also be used from the command line to produce image files of the alignment (e.g. within a script).

Following activities has been performed by alignment editor:

- Loading Sequences
- Editing Functions
- Locking Sequences
- Trim Sequences
- Colour Schemes
- Scoring Matrix
- Consensus Sequence

2.7 Chemical Markup Language (CML)

2.7.1 Introduction

This is a variety of XML [6] designed to represent molecular information. It has been used to store chemical formulas and to display the molecules in graphical formats.

CML [7] has been developed to carry molecules, crystallographic data and reactions using an XML language. A universal, platform and application independent format for storing and exchanging chemical information has been offered by CML. CML outlines a variety of general purpose 'data-holder' elements and a smaller number of more specifically chemical elements (e.g. <molecule>, <reaction>, <crystal>) used to indicate chemical 'objects'. For example, a <molecule> will contain a <list> of <atom>s, which in turn have three <float>s specifying Cartesian coordinates for each atom.

CML provides no default conventions for labeling data elements and puts few restrictions on element ordering. The design of CML and contains minimal preconceptions as to the type of chemical information that has been stored using it.

2.7.2 Reading XML Documents [8]

Here is an example from the CML Schema:

```
<cml>
  <molecule id="m1">
    <atomArray>
      <atom elementType="N"/>
      <atom elementType="O"/>
    </atomArray>
  </molecule>
</cml>
```

The first tag is <cml>. This is the top level tag. The next tag is <molecule id="m1">. The CML Schema reference says that the <molecule> tag is "a container for atoms, bonds and submolecules." The 'id' attribute is used as a unique identifier so that the

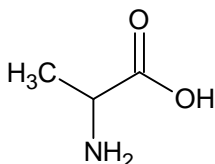
molecule can be referred to from elsewhere. Similarly, the <atomArray> is "a container for a list of atoms." The tag <atom elementType="N"/> specifies a nitrogen atom and the tag <atom elementType="O"/> an oxygen atom.

The tag </atomArray> closes the <atomArray> element. Tags must always be closed with a </...> pattern in XML to create well formed documents. Also, tags must be fully enclosed within other tags and cannot overlap. For example, <a> is well formed XML but <a> is not. If a tag does not have anything inside it then the shorthand <.../> can be used to indicate both opening and closing an empty tag.

2.7.3 Examples of the molecules with CML

In this research substance like Alanine, Amino butyric Acid, Asparagine and Glutamine have been studied.

(A) Alanine



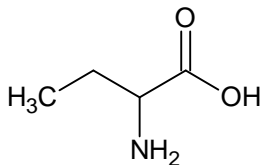
```
<list xmlns:cml="http://www.xml-cml.org/schema/cml2/core" xmlns:stm="http://www.xml-cml.org/schema/stmml" xmlns:ichi="http://www.iupac.org/foo/ichi" xmlns="http://www.xml-cml.org/schema/cml2/core" title="/var/wwwtmp/mn_convert29053.xml">
  <molecule convention="CACTVS">
    <metadataList>
      <metadata name="dc:title">chemical structure data</metadata>
      <metadata name="dc:creator">wwwrun</metadata>
      <metadata name="dc:date">2009-06-03</metadata>
    </metadataList>
    <atomArray>
      <atom id="1">
        <string builtin="elementType">C</string>
        <float builtin="x2">16.9045</float>
        <float builtin="y2">-7.5353</float>
        <float builtin="x3">0</float>
        <float builtin="y3">0</float>
        <float builtin="z3">0</float>
      </atom>
      <atom id="2">
        <string builtin="elementType">O</string>
        <float builtin="x2">16.9045</float>
        <float builtin="y2">-6.2053</float>
        <float builtin="x3">0</float>
        <float builtin="y3">0</float>
        <float builtin="z3">0</float>
      </atom>
      <atom id="3">
        <string builtin="elementType">C</string>
        <float builtin="x2">15.7527</float>
        <float builtin="y2">-8.2003</float>
        <float builtin="x3">0</float>
        <float builtin="y3">0</float>
        <float builtin="z3">0</float>
      </atom>
    </atomArray>
  </molecule>
</list>
```

```

</atom>
<atom id="4">
  <string builtin="elementType">N</string>
  <float builtin="x2">15.7527</float>
  <float builtin="y2">-9.5303</float>
  <float builtin="x3">0</float>
  <float builtin="y3">0</float>
  <float builtin="z3">0</float>
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<atom id="5">
  <string builtin="elementType">C</string>
  <float builtin="x2">14.6008</float>
  <float builtin="y2">-7.5353</float>
  <float builtin="x3">0</float>
  <float builtin="y3">0</float>
  <float builtin="z3">0</float>
</atom>
<atom id="6">
  <string builtin="elementType">O</string>
  <float builtin="x2">18.0563</float>
  <float builtin="y2">-8.2003</float>
  <float builtin="x3">0</float>
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  <float builtin="z3">0</float>
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</atomArray>
<bondArray>
<bond id="1">
  <string builtin="atomRef">1</string>
  <string builtin="atomRef">2</string>
  <string builtin="order">2</string>
</bond>
<bond id="2">
  <string builtin="atomRef">1</string>
  <string builtin="atomRef">3</string>
  <string builtin="order">1</string>
</bond>
<bond id="3">
  <string builtin="atomRef">3</string>
  <string builtin="atomRef">4</string>
  <string builtin="order">1</string>
</bond>
<bond id="4">
  <string builtin="atomRef">3</string>
  <string builtin="atomRef">5</string>
  <string builtin="order">1</string>
</bond>
<bond id="5">
  <string builtin="atomRef">6</string>
  <string builtin="atomRef">1</string>
  <string builtin="order">1</string>
</bond>
</bondArray>
</molecule>
</list>

```

(B) Amino Butyric Acid



```

<list xmlns:cml="http://www.xml-cml.org/schema/cml2/core" xmlns:stm="http://www.xml-
cml.org/schema/stmml" xmlns:ichi="http://www.iupac.org/foo/ichi" xmlns="http://www.xml-
cml.org/schema/cml2/core" title="/var/wwwtmp/mn_convert28995.xml">
  <molecule conversion="CACTVS">
    <metadataList>
      <metadata name="dc:title">chemical structure data</metadata>
      <metadata name="dc:creator">wwwrun</metadata>
      <metadata name="dc:date">2009-06-03</metadata>
    </metadataList>

```

```

<atomArray>
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  <float builtin="y2">-9.5383</float>
  <float builtin="x3">0</float>
  <float builtin="y3">0</float>
  <float builtin="z3">0</float>
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<atom id="2">
  <string builtin="elementType">O</string>
  <float builtin="x2">16.7272</float>
  <float builtin="y2">-8.2083</float>
  <float builtin="x3">0</float>
  <float builtin="y3">0</float>
  <float builtin="z3">0</float>
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  <float builtin="x2">15.5755</float>
  <float builtin="y2">-10.2033</float>
  <float builtin="x3">0</float>
  <float builtin="y3">0</float>
  <float builtin="z3">0</float>
</atom>
<atom id="4">
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  <float builtin="y2">-11.5333</float>
  <float builtin="x3">0</float>
  <float builtin="y3">0</float>
  <float builtin="z3">0</float>
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  <string builtin="elementType">C</string>
  <float builtin="x2">14.4236</float>
  <float builtin="y2">-9.5383</float>
  <float builtin="x3">0</float>
  <float builtin="y3">0</float>
  <float builtin="z3">0</float>
</atom>
<atom id="6">
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  <float builtin="x2">17.8791</float>
  <float builtin="y2">-10.2033</float>
  <float builtin="x3">0</float>
  <float builtin="y3">0</float>
  <float builtin="z3">0</float>
</atom>
<atom id="7">
  <string builtin="elementType">C</string>
  <float builtin="x2">13.2718</float>
  <float builtin="y2">-10.2034</float>
  <float builtin="x3">0</float>
  <float builtin="y3">0</float>
  <float builtin="z3">0</float>
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<bond id="1">
  <string builtin="atomRef">1</string>
  <string builtin="atomRef">2</string>
  <string builtin="order">2</string>
</bond>
<bond id="2">
  <string builtin="atomRef">1</string>
  <string builtin="atomRef">3</string>
  <string builtin="order">1</string>
</bond>
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  <string builtin="atomRef">4</string>
  <string builtin="order">1</string>
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```

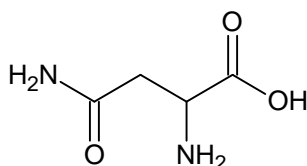


```

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</bond>
</bondArray>
</molecule>
</list>

```

(C)Asparagine

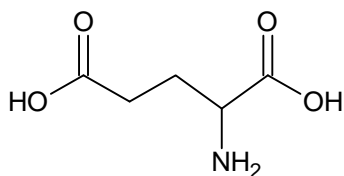


```

<list xmlns:cml="http://www.xml-cml.org/schema/cml2/core" xmlns:stm="http://www.xml-
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cml.org/schema/cml2/core" title="/var/wwwtmp/mm_convert29126.xml">
<molecule convention="CACTVS">
<metadataList>
<metadata name="dc:title">chemical structure data</metadata>
<metadata name="dc:creator">wwwrun</metadata>
<metadata name="dc:date">2009-06-03</metadata>
</metadataList>
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```

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</molecule>
</list>
```

(D) Glutamine

```

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cml.org/schema/cml2/core" title="/var/wwwtmp/mn_convert29157.xml">
<molecule convention="CACTVS">
<metadataList>
<metadata name="dc:title">chemical structure data</metadata>
<metadata name="dc:creator">wwwrun</metadata>
<metadata name="dc:date">2009-06-03</metadata>
</metadataList>
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</atom>
<atom id="8">

```

```

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<float builtin="y3">0</float>
<float builtin="z3">0</float>
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</bond>
<bond id="6">
<string builtin="atomRef">4</string>
<string builtin="atomRef">6</string>
<string builtin="order">1</string>
</bond>
<bond id="7">
<string builtin="atomRef">6</string>
<string builtin="atomRef">7</string>
<string builtin="order">1</string>
</bond>
<bond id="8">
<string builtin="atomRef">9</string>
<string builtin="atomRef">1</string>
<string builtin="order">1</string>
</bond>
<bond id="9">
<string builtin="atomRef">10</string>
<string builtin="atomRef">7</string>
<string builtin="order">1</string>
</bond>
</bondArray>
</molecule>
</list>

```

2.8 SMILES - A Simplified Chemical Language

2.8.1 Introduction

SMILES [9] (Simplified Molecular Input Line Entry System) is a line notation for entering and representing molecules and reactions. Some examples are:

Table 2.2: SMILE examples

SMILES	Name	SMILES	Name
CC	ethane	[OH3+]	hydronium ion
O=C=O	carbon dioxide	[2H]O[2H]	deuterium oxide
C#N	hydrogen cyanide	[235U]	uranium-235
CCN(CC)CC	triethylamine	F/C=C/F	E-difluoroethene
CC(=O)O	acetic acid	F/C=C\F	Z-difluoroethene
C1CCCCC1	cyclohexane	N[C@@H](C)C(=O)O	L-alanine
c1ccccc1	benzene	N[C@H](C)C(=O)O	D-alanine

Reaction SMILES	Name
[I-].[Na+].C=CCBr>>[Na+].[Br-].C=CCI	displacement reaction
(C(=O)O).(OCC)>>(C(=O)OCC).(O)	intermolecular esterification

SMILES are a true language, although with a simple vocabulary (atom and bond symbols) and only a few grammar rules. SMILES representations of structure can in turn be used as "words" in the vocabulary of other languages designed for storage of chemical information (information about chemicals) and chemical intelligence (information about chemistry).

2.8.2 Canonicalization

A SMILE denotes a molecular structure as a graph with optional chiral indications. This is essentially the two-dimensional picture chemists draw to describe a molecule. SMILES describing only the labeled molecular graph (i.e. atoms and bonds, but no chiral or isotopic information) are known as generic SMILES.

It can be shown in the following examples.

Table 2.3: Canonicalization of SMILE

Input SMILES	Unique SMILES
OCC	CCO
[CH3][CH2][OH]	CCO
C-C-O	CCO
C(O)C	CCO
OC(=O)C(Br)(Cl)N	NC(Cl)(Br)C(=O)O
ClC(Br)(N)C(=O)O	NC(Cl)(Br)C(=O)O
O=C(O)C(N)(Br)Cl	NC(Cl)(Br)C(=O)O

2.8.3 SMILES Specification Rules

SMILES notation consists of a series of characters containing no spaces. Hydrogen atoms may be omitted (hydrogen-suppressed graphs) or included (hydrogen-complete graphs). Aromatic structures may be specified directly.

There are five generic SMILES encoding rules, corresponding to specification of atoms, bonds, branches, ring closures, and disconnections.

2.8.3.1 Atoms

Atoms are represented by their atomic symbols: this is the only required use of letters in SMILES. Each non-hydrogen atom is specified independently by its atomic symbol enclosed in square brackets, []. The second letter of two-character symbols must be entered in lower case. Atoms in aromatic rings are specified by lower case letters, e.g., aliphatic carbon is represented by the capital letter C, aromatic carbon by lower case c.

Table 2.4: SMILE atoms

C	methane	(CH4)
P	phosphine	(PH3)
N	ammonia	(NH3)
S	hydrogen sulfide	(H2S)
O	water	(H2O)
Cl	hydrochloric acid	(HCl)

2.8.3.2 Bonds

Single, double, triple, and aromatic bonds are represented by the symbols -, =, #, and :, respectively. Adjacent atoms are assumed to be connected to each other by a single or aromatic bond (single and aromatic bonds may always be omitted). Examples are:

Table 2.5: SMILE bonds

CC	ethane	(CH ₃ CH ₃)
C=O	formaldehyde	(CH ₂ O)
C=C	ethene	(CH ₂ =CH ₂)
O=C=O	carbon dioxide	(CO ₂)
COC	dimethyl ether	(CH ₃ OCH ₃)
C#N	hydrogen cyanide	(HCN)
CCO	ethanol	(CH ₃ CH ₂ OH)
[H][H]	molecular hydrogen	(H ₂)

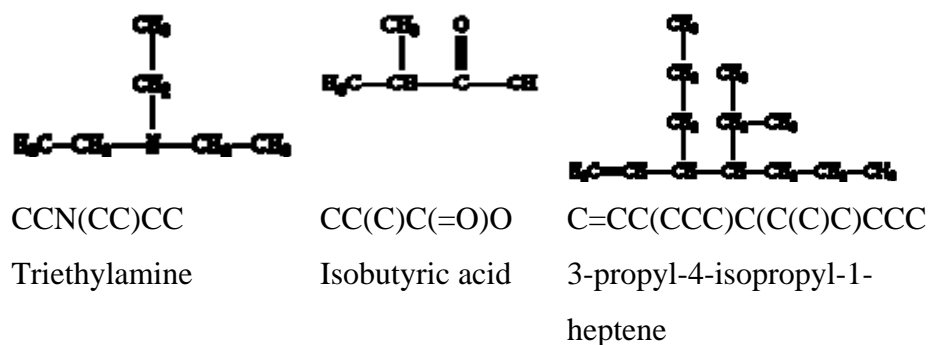
For linear structures, SMILES notation corresponds to conventional diagrammatic notation except that hydrogens and single bonds are generally omitted. For example, 6-hydroxy-1,4-hexadiene can be represented by many equally valid SMILES, including the following three:

Structure	Valid SMILES
	C=CCC=CCO
CH ₂ =CH-CH ₂ -CH=CH-CH ₂ -OH	C=C-C-C=C-C-O
	OCC=CCC=C

2.8.3.3 Branches

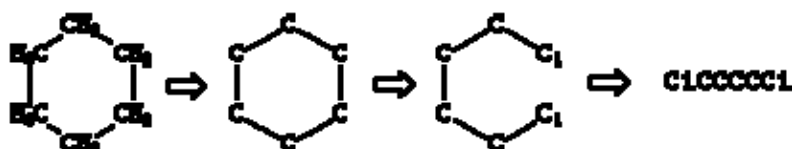
Branches are specified by enclosing them in parentheses, and can be nested or stacked. In all cases, the implicit connection to a parenthesized expression (a "branch") is to the left. Examples are:

Table 2.6: SMILE branches

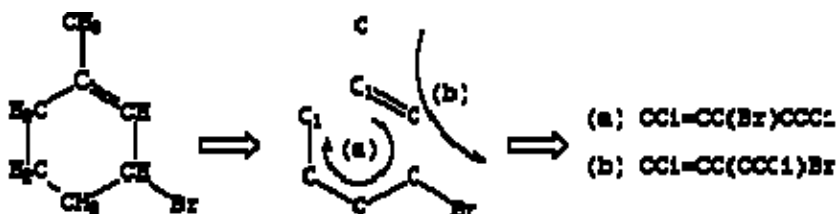


2.8.3.4 Cyclic Structures

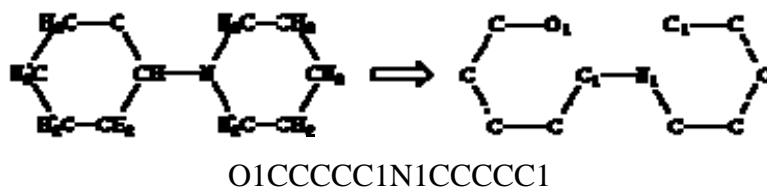
Cyclic structures are represented by breaking one bond in each ring. The bonds are numbered in any order, designating ring opening (or ring closure) bonds by a digit immediately following the atomic symbol at each ring closure. This leaves a connected non-cyclic graph which is written as a non-cyclic structure using the three rules described above. Cyclohexane is a typical example:



There are usually many different, but equally valid descriptions of the same structure, e.g., the following SMILES notations for 1-methyl-3-bromo-cyclohexene-1:



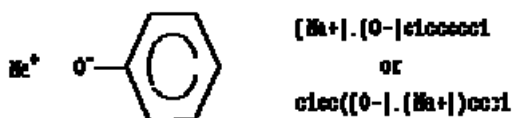
Digits denoting ring closures has been reused. As an example, the digit 1 used twice in the specification:



The ability to re-use ring closure digits makes it possible to specify structures with 10 or more rings. Structures that require more than 10 ring closures to be open at once are exceedingly rare.

2.8.3.5 Disconnected Structures

Disconnected compounds are written as individual structures separated by a "." (period). If desired, the SMILES of one ion may be imbedded within another as shown in the example of sodium phenoxide.



Matching pairs of digits following atom specifications imply that the atoms are bonded to each other. The bond may be explicit (bond symbol and/or direction preceding the ring closure digit) or implicit (a nondirectional single or aromatic bond). This is true whether or not the bond ends up as part of a ring.

Adjacent atoms separated by dot (.) imply that the atoms are not bonded to each other. This is true whether or not the atoms are in the same connected component.

For example, C1.C1 specifies the same molecule as CC(ethane) .

2.10 References

[1] www.acdlabs.com

[2] www.upstream.ch

[3] www.arguslab.com

[4] <http://dambe.bio/uottawa.ca>

[5] <http://mEMBOSS/jemboss/jar/resources/readme.html>

[6] <http://www.xml-cml.org/>

[7] <http://zyon.org/xxl/CML1.0/Output/index.html>

[8] <http://cml.sourceforge.net/>

[9] <http://www.daylight.com>

Chapter -3

Alignment of PAirs and
Mul tipl e SequenceS and
Phyl ogenetic AnAI ysis

Chapter 3

ALIGNMENT OF PAIRS AND MULTIPLE SEQUENCES AND PHYLOGENETIC ANALYSIS

3.1 Introduction

In this chapter Pair wise Sequence Alignment and Multiple Sequence Alignment has been discussed. In this chapter alignment score and Gap Penalty between sequences has been calculated. The gap penalty formula can be extended to include a penalty for alignments for the gaps at the end of a sequence of equal length. Multiple sequence alignment is useful in finding patterns in nucleotide sequences and for identifying structural and functional domains in protein families.

Multiple sequence alignment (MSA) is an extension of the similarity concepts to determine levels of homology (relatedness) between members of a series of globally related sequences are aligned together in column.

3.2 Sequence Description

Patterns provide appropriate representations of conserved regions in biosequences. In most cases one is given a set of sequences (DNA or proteins) and is looking for patterns that appear in some minimum number (or percentage) of these sequences. The exact definition of a pattern varies from algorithm to algorithm. In general, a pattern is a member of a well defined subset C of all the possible regular expressions over \sum^1 (the set C is called a pattern language). Being a regular expression, every pattern P defines a language $L(P)$ in the natural way: a string belongs to $L(P)$ if it is recognized by the automaton of P . A sequence $s \in \sum^*$ is said to "match" a given pattern P if s contains some substring that belongs to $L(P)$.

For an illustration, consider the following set of strings over the English alphabet:

$$S = \{\text{LARGE, FINGER, AGE}\}$$

In this case the pattern "L..GE" has support 2 since it is matched by the first two strings of S ('.' is called the don't-care character and is used to indicate position that can be occupied by an arbitrary alphabet character). The term support denotes the number of input strings matching a given pattern. As another example, the pattern "A*GE" has also support 2 (it is matched by the first and the last strings). Here, the character '*' is used to match substrings of arbitrary length.

3.3 Pair wise Sequence Alignment

Pair wise sequence alignment [1] involves the matching of two sequences, one pair of elements at a time. The challenge in pair wise sequence alignment is to find the optimum alignment of two sequences with some degree of similarity. This optimum condition is based on a score that reflects the number of paired characters in two sequences and number and length of gaps required to adjust the sequences so the maximum number of characters are in alignment. For example, consider the ideal case of identical nucleotide sequences, (A) and (B)

A) **ATTCGGCATTTCAGTGCTAGA**

B) **ATTCGGCATTTCAGTGCTAGA**

Assuming that the alignment scoring algorithm counts one point per pair of aligned characters, then the score for each of the 20 pairs, or 20 points. Now, consider the case when several of character pairs aren't aligned:

C) **ATTCGGCATTTCAGTGCTAGA**

D) **ATTCGGCATTGCTAGA**

In this case, the score is 11, because only 11 pairs of characters in sequences (C) and (D) are aligned. By moving last six characters ahead in sequence (D) by adding four gaps, the sequences become:

E) ATTCGGCATT CAGTGCTAGA

F) ATTCGGCATT - - - -GCTAGA

Now, the score, based on the original algorithm of character pairing, is 16. However, because the score would have been 11 without the inserted gaps, a penalty should be extracted for each gap inserted into the sequence to favor alignments that can be made with as few gaps as possible. Assuming a gap penalty of -0.5 per gap, the alignment score becomes $10 + 6 + (4 \times -0.5)$ or 14.

Another scenario is that in which the areas of similarity and difference are not obvious. Consider the sequences (G) and (H):

G) ATTCGGCATT CAGAGCGAGA

H) ATTCGACATT GCTAGTGGTA

Unlike the previous cases, there are no relatively long runs of character pairings, and the matching pairs are separated by unaligned characters. The alignment score is 1 point per aligned pair or 13. One attempt at visual alignment by adding four gaps into sequence (H) results in:

G) ATTCGGCATT CAGAGCTAGA

I) ATTCGACATT - - - -GCTAGTGGTA

This alignment results in a score of 12, or 14 alignments minus 2 points for the 4 gaps introduced into sequence (H), transforming it to sequence (I). In addition, a penalty of -0.5 per character pair is scored for an inexact match. In case of sequences (G) and (I), there are 6 inexact matches. In case of sequences (G) and (I), there are 6 inexact matches,

for a penalty of $(6 \times -0.5 = -3)$. Using this new alignment scoring algorithm, and ignoring the length difference between the two sequences, the alignment score for the (G)-(I) alignment becomes:

$$\begin{aligned}\text{Alignment Score} &= 14 \text{ alignments} + 4 \text{ gaps} + 6 \text{ inexact matches} \\ &= 14 + (4 \times -0.5) + (6 \times -0.5) \\ &= 14 - 2 - 3 \\ &= 9\end{aligned}$$

In this example, adding gaps results in a lower alignment score, illustrating how the relative worth of exact matches, inexact matches and gaps determines the eventual alignment of two sequences.

Although a simple gap penalty of - 0.5 point per gap has been used to illustrate the role of alignment scores on sequence alignment, gap penalty is typically calculated as:

$$\text{Penalty}_{\text{gap}} = \text{Cost}_{\text{opening}} + \text{Cost}_{\text{extension}} \times \text{Length}_{\text{gap}}$$

In this formula, $\text{Penalty}_{\text{gap}}$ is the total gap penalty, $\text{Cost}_{\text{opening}}$ is the cost of opening is the cost of opening a gap in a sequence, $\text{Cost}_{\text{extension}}$ is the cost of extending an existing gap by one character, and $\text{Length}_{\text{gap}}$ is the length of the gap in characters. The minimum value of $\text{Length}_{\text{gap}}$ is one. Returning to sequence pair (E)-(F), assuming that $\text{Cost}_{\text{opening}}$ is (- 0.5) and $\text{Cost}_{\text{extension}}$ is (- 0.5), the gap penalty becomes:

$$\begin{aligned}\text{Penalty}_{\text{gap}} &= \text{Cost}_{\text{opening}} + \text{Cost}_{\text{extension}} \times \text{Length}_{\text{gap}} \\ &= -0.5 + (-0.5 \times 4) \\ &= -2.5\end{aligned}$$

With the expanded method of computing gap penalty, the score becomes $10 + 6 - 2.5 = 13.5$ points. The gap penalty formula can be extended to include a penalty for alignments for the gaps at the end of a sequence of equal length.

3.3.1 Local versus Global Alignment

Sequence pair (E) – (F) is an example of global alignment- that is , an attempt to line up the two sequences matching as many characters as possible, for the entire length of each segment. Global alignment considers all characters in a sequence, and bases alignment on the total score, even at the expense of stretches in the sequence that share similarity as shown in figure. Global alignment is used to help determine whether two protein sequences are the same family.

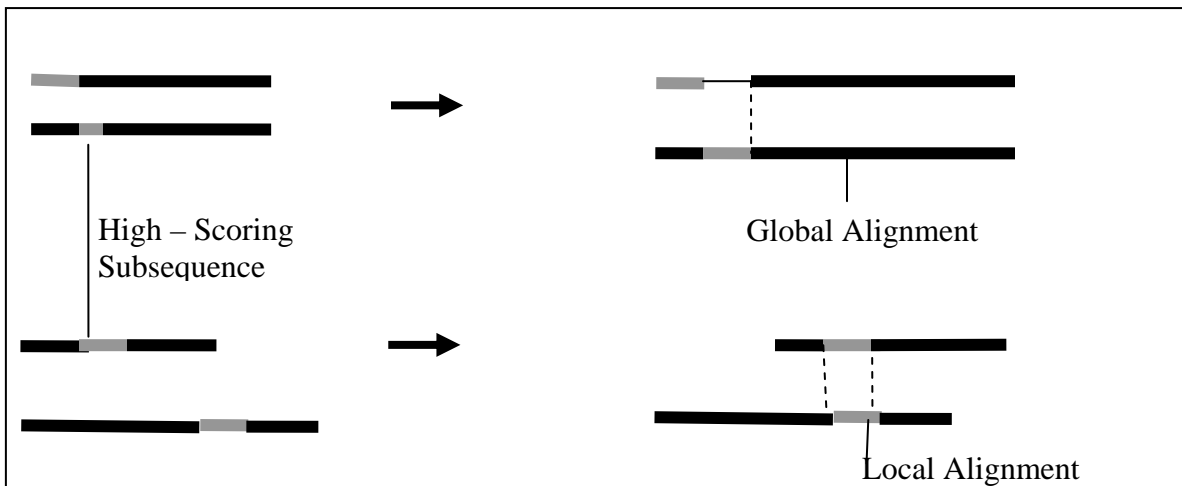


Figure 3.1: Local (top) versus Global (bottom) Alignment. In local alignment, the alignment of local, high – scoring sequences takes precedence over the overall alignment. In global alignment, the best overall alignment is sought, regardless of whether local, high-scoring subsequences are in alignment or not.

3.3.2 Methods of Sequence Alignment

There are several approaches for conducting sequence alignments. Many of these methods are heuristics methods. One of them is Dynamic Programming (DP) method for sequence alignment.

Dynamic Programming

In genetics, sequence alignment is an important application where dynamic programming is essential. Typically, the problem consists of transforming one sequence into another using edit operation that replace, insert, or remove an element. Each operation has an associated cost, and the goal is to find the sequence of edits with the lowest total cost.

The problem can be stated naturally as a recursion, a sequence A is optimally edited into a sequence B by either:

1. inserting the first character of B, and performing an optimal alignment of A and the tail of B
2. deleting the first character of A, and performing the optimal alignment of the tail of A and B
3. replacing the first character of A with the first character of B, and performing optimal alignments of the tails of A and B.

The partial alignments can be tabulated in a matrix, where cell (i, j) contains the cost of the optimal alignment of A [1...i] to B [1...j]. The cost in cell (i, j) can be calculated by adding the cost of the relevant operations to the cost of its neighboring cells, and selecting the optimum.

Dynamic programming is a form of recursion in which intermediate results are saved in a matrix. The comparison can solve complex mathematical equations, with the results of one equation feeding the input of another. With dynamic programming, the intermediate results can be recorded and next equation can be solved with regard to following equations.

The function has been used to illustrate the value of dynamic programming in sequence alignment:

$$MaxValue = f(A_i, B_j)$$

In this equation, *MaxValue* is function of variables A_i and B_j , where i and j are indices to the variables defined in tree structure illustrated in **Figure 3.3**. That is, the possible values of A_i are represented by A_1 through A_5 , and possible values of B_j are represented by B_1 through B_{11} . The best solution to *MaxValue* depends on the equation that defines *MaxValue*. Following possible value of *MaxValue* have been considered as example:

$$MaxValue = (A_i \times B_j)$$

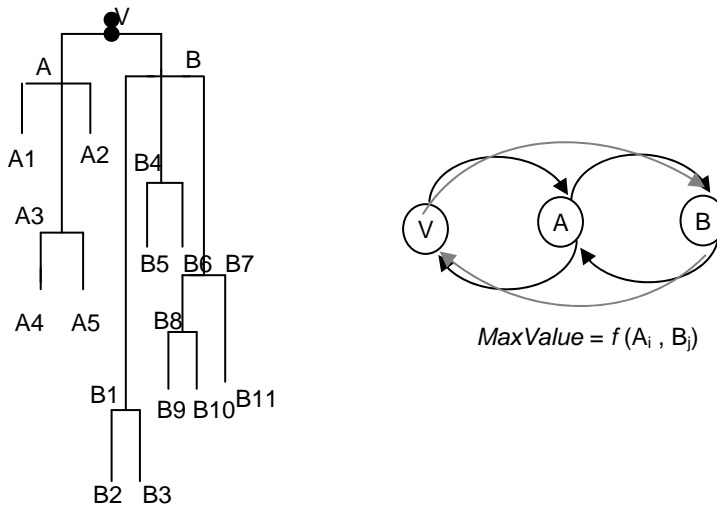


Figure 3.2: Dynamic Programming Problems. Values for A and B are defined in the tree structure. Maximizing *MaxValue* requires evaluating the equation for every combination of i and j .

In this example, the solution is simply the largest value of A and the largest values of B. However, following definition of *MaxValue* has been considered:

$$MaxValue = 3 \sqrt{\frac{14 \times A^2}{\log(A^2 + B^2)}}$$

In Brute- Force Methods of solving for *MaxValue* every combinations of A and B has been required and each value has been found out recursively and defined in the tree structure.

For evaluating every value of B in the *MaxValue* equation needs evaluating every value of A that has been illustrated in **Figure 3.3**. For example, assume that the values of A_i and B_j are defined as:

$$A = \begin{bmatrix} 2 \\ 3 \\ 8 \\ 4 \\ 1 \end{bmatrix} \qquad B = \begin{bmatrix} 11 \\ 1 \\ 0 \\ 3 \\ 8 \\ 1 \\ 7 \\ 5 \\ 3 \\ 2 \end{bmatrix}$$

Solving for the first value of A_i ($A_1 = 2$) and ignoring the specific equation for *MaxValue* for Clarity:

MaxValue(1,1) =	$f(A_1, B_1) =$	$f(2,9) =$	5
MaxValue(1,2) =	$f(A_1, B_2) =$	$f(2,11) =$	3
MaxValue(1,3) =	$f(A_1, B_3) =$	$f(2,1) =$	0
MaxValue(1,4) =	$f(A_1, B_4) =$	$f(2,0) =$	2
MaxValue(1,5) =	$f(A_1, B_5) =$	$f(2,3) =$	8
MaxValue(1,6) =	$f(A_1, B_6) =$	$f(2,8) =$	0
MaxValue(1,7) =	$f(A_1, B_7) =$	$f(2,1) =$	-2
MaxValue(1,8) =	$f(A_1, B_8) =$	$f(2,7) =$	1
MaxValue(1,9) =	$f(A_1, B_9) =$	$f(2,5) =$	2
MaxValue(1,10) =	$f(A_1, B_{10}) =$	$f(2,3) =$	8
MaxValue(1,11) =	$f(A_1, B_{11}) =$	$f(2,2) =$	4

If the branches of A and B have hundreds of sub-branches, representing hundreds of values, then problem is likely computationally infeasible.

Dynamic programming can address this computational and time dilemma by creating a matrix to store the values for A_i , B_j and $MaxValue$ for each combination of i and j . For example, consider the solution matrix for $MaxValue$ in **Figure 3.3**. The solution set to $MaxValue$ computed earlier for A_1 appears in first row of the matrix. Examining only this first row, it can be seen that there are two solutions to $MaxValue$, B_5 and B_{10} , each of which results in a value of 8.

		B_j										
		1	2	3	4	5	6	7	8	9	10	11
A_i	1	5	3	0	2	8	0	-2	1	2	8	4
	2	0	3	0	6	11	0	-6	5	7	4	0
	3	9	0	12	2	0	0	0	5	0	0	0
	4	1	7	5	5	11	0	-1	1	7	5	4
	5	9	0	0	2	5	0	5	5	1	4	1

Figure 3.3: Solution Matrix for $MaxValue$ for A_i and B_j . The solution to $MaxValue$ is A_3 and B_3 with $MaxValue = 12$

$MaxValue$ has been considered as aligned score for pair wise alignment of two sequences. $MaxValue$ takes into account gap penalties, correct alignments and imperfect alignments. After the matrix is filled in using the alignment score to determine $MaxValue$, the highest scoring path is followed back to the beginning of the alignment to define the best alignment of elements in sequence, including gaps.

Example of Global Sequence Alignment using Dynamic Programming

The following is an example of global sequence alignment using Needleman/Wunsch techniques. For this example, the two sequences to be globally aligned are

G A A T T C A G T T A (sequence #1)

G G A T C G A (sequence #2)

So $M = 11$ and $N = 7$ (the length of sequence #1 and sequence #2, respectively)

A simple scoring scheme is assumed where

- $S_{i,j} = 1$ if the residue at position i of sequence #1 is the same as the residue at position j of sequence #2 (match score); otherwise
- $S_{i,j} = 0$ (mismatch score)
- $w = 0$ (gap penalty)

Three steps in dynamic programming

1. Initialization
2. Matrix fill (scoring)
3. Traceback (alignment)

Initialization Step

The first step in the global alignment dynamic programming approach is to create a matrix with $M + 1$ columns and $N + 1$ rows where M and N correspond to the size of the sequences to be aligned.

Since this example assumes there is no gap opening or gap extension penalty, the first row and first column of the matrix can be initially filled with 0.

		G	A	A	T	T	C	A	G	T	T	A
	0	0	0	0	0	0	0	0	0	0	0	0
G	0											
G	0											
A	0											
T	0											
C	0											
G	0											
A	0											

Matrix Fill Step

One possible (inefficient) solution of the matrix fill step finds the maximum global alignment score by starting in the upper left hand corner in the matrix and finding the maximal score $M_{i,j}$ for each position in the matrix. In order to find $M_{i,j}$ for any i,j it is minimal to know the score for the matrix positions to the left, above and diagonal to i, j . In terms of matrix positions, it is necessary to know $M_{i-1,j}$, $M_{i,j-1}$ and $M_{i-1,j-1}$.

For each position, $M_{i,j}$ is defined to be the maximum score at position i,j ; i.e.

$$M_{i,j} = \text{MAXIMUM} [$$

- $M_{i-1,j-1} + S_{i,j}$ (match/mismatch in the diagonal),
- $M_{i,j-1} + w$ (gap in sequence #1),
- $M_{i-1,j} + w$ (gap in sequence #2)]

In the example, $M_{i-1,j-1}$, $M_{i,j-1}$ and $M_{i-1,j}$ has been considered red, green and blue respectively.

Using this information, the score at position 1, 1 in the matrix can be calculated. Since the first residue in both sequences is a G, $S_{1,1} = 1$, and by the assumptions stated at the beginning, $w = 0$. Thus, $M_{1,1} = \text{MAX} [M_{0,0} + 1, M_{1,0} + 0, M_{0,1} + 0] = \text{MAX} [1, 0, 0] = 1$.

A value of 1 is then placed in position 1, 1 of the scoring matrix.

	G	A	A	T	T	T	C	A	G	T	T	A
0	0	0	0	0	0	0	0	0	0	0	0	0
G	0	1										
G	0											
A	0											
T	0											
C	0											
G	0											
A	0											

Since the gap penalty (w) is 0, the rest of row 1 and column 1 can be filled in with the value 1. Take the example of row 1. At column 2, the value is the max of 0 (for a mismatch), 0 (for a vertical gap) or 1 (horizontal gap). The rest of row 1 can be filled out similarly until we get to column 8. At this point, there is a G in both sequences (light blue). Thus, the value for the cell at row 1 column 8 is the maximum of 1 (for a match), 0 (for a vertical gap) or 1 (horizontal gap). The value will again be 1. The rest of row 1 and column 1 can be filled with 1 using the above reasoning.

	G	A	A	T	T	T	C	G	T	T	A
0	0	0	0	0	0	0	0	0	0	0	0
G	0	1	1	1	1	1	1	1	1	1	1
G	0	1									
A	0	1									
T	0	1									
C	0	1									
G	0	1									
A	0	1									

Then after column 2 will be considered. The location at row 2 will be assigned the value of the maximum of 1(mismatch), 1(horizontal gap) or 1 (vertical gap). So its value is 1.

At the position column 2 row 3, there is an A in both sequences. Thus, its value will be the maximum of 2(match), 1 (horizontal gap), 1 (vertical gap) so its value is 2.

Moving along to position column 2 row 4, its value will be the maximum of 1 (mismatch), 1 (horizontal gap), 2 (vertical gap) so its value is 2. For all of the remaining

positions except the last one in column 2, the choices for the value will be the exact same as in row 4 since there are no matches. The final row will contain the value 2 since it is the maximum of 2 (match), 1 (horizontal gap) and 2(vertical gap).

	G	A	A	T	T	C	A	G	T	T	A
	0	0	0	0	0	0	0	0	0	0	0
G	0	1	1	1	1	1	1	1	1	1	1
G	0	1	1								
A	0	1	2								
T	0	1	2								
C	0	1	2								
G	0	1	2								
A	0	1	2								

Using the same techniques as described for column 2, column 3 has to be filled.

	G	A	A	T	T	C	A	G	T	T	A
	0	0	0	0	0	0	0	0	0	0	0
G	0	1	1	1	1	1	1	1	1	1	1
G	0	1	1	1							
A	0	1	2	2							
T	0	1	2	2							
C	0	1	2	2							
G	0	1	2	2							
A	0	1	2	3							

After filling in all of the values the score matrix is as follows:

	G	A	A	T	T	C	A	G	T	T	A
	0	0	0	0	0	0	0	0	0	0	0
G	0	1	1	1	1	1	1	1	1	1	1
G	0	1	1	1	1	1	1	2	2	2	2
A	0	1	2	2	2	2	2	2	2	2	3
T	0	1	2	2	3	3	3	3	3	3	3
C	0	1	2	2	3	3	3	4	4	4	4
G	0	1	2	2	3	3	3	4	4	5	5
A	0	1	2	3	3	3	3	4	5	5	6

Trace back Step

After the matrix fill step, the maximum alignment score for the two test sequences is 6. The traceback step determines the actual alignment(s) that result in the maximum score. With a simple scoring algorithm such as one that is used here, there are likely to be multiple maximal alignments.

The traceback step begins in the M, J position in the matrix, i.e. the position that leads to the maximal score. In this case, there is a 6 in that location.

Traceback takes the current cell and looks to the neighbor cells that could be direct predecessors. This means it looks to the neighbor to the left (gap in sequence #2), the diagonal neighbor (match/mismatch), and the neighbor above it (gap in sequence #1). The algorithm for traceback chooses as the next cell in the sequence one of the possible predecessors. In this case, the neighbors are marked in red. They are all also equal to 5.

	G	A	A	T	T	C	A	G	T	T	A
0	0	0	0	0	0	0	0	0	0	0	0
G	0	1	1	1	1	1	1	1	1	1	1
G	0	1	1	1	1	1	1	2	2	2	2
A	0	1	1	2	2	2	2	2	2	2	3
T	0	1	2	2	3	3	3	3	3	3	3
C	0	1	2	2	3	3	4	4	4	4	4
G	0	1	2	2	3	3	4	4	5	5	5
A	0	1	2	3	3	3	4	5	5	5	6

Since the current cell has a value of 6 and the scores are 1 for a match and 0 for anything else, the only possible predecessor is the diagonal match/mismatch neighbor. If more than one possible predecessor exists, any can be chosen. This gives us a current alignment of

```

(Seq #1)      A
              |
(Seq #2)      A
    
```

So now we look at the current cell and determine which cell is its direct predecessor. In this case, it is the cell with the red 5.

	G	A	A	T	T	C	A	G	T	T	A	
0	0	0	0	0	0	0	0	0	0	0	0	
G	0	1	1	1	1	1	1	1	1	1	1	
G	0	1	1	1	1	1	1	2	2	2	2	
A	0	1	2	2	2	2	2	2	2	2	2	
T	0	1	2	2	3	3	3	3	3	3	3	
C	0	1	2	2	3	3	4	4	4	4	4	
G	0	1	2	2	3	3	4	4	5	5	5	
A												6

The alignment as described in the above step adds a gap to sequence #2, so the current alignment is

```

(Seq #1)   T A
           |
(Seq #2)   _ A
    
```

Once again, the direct predecessor produces a gap in sequence #2.

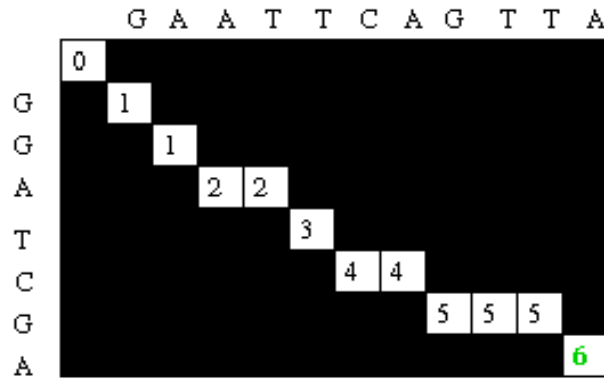
	G	A	A	T	T	C	A	G	T	T	A	
0	0	0	0	0	0	0	0	0	0	0	0	
G	0	1	1	1	1	1	1	1	1	1	1	
G	0	1	1	1	1	1	1	2	2	2	2	
A	0	1	2	2	2	2	2	2	2	2	2	
T	0	1	2	2	3	3	3	3	3	3	3	
C	0	1	2	2	3	3	4	4	4	4	4	
G	0	1	2	2	3	3	4	4	5	5	5	
A												6

After this step, the current alignment is

```

(Seq #1)   T T A
           |
           |
(Seq #2)   _ _ A
    
```

Continuing on with the traceback step, a position in column 0 row 0 has been found which indicates completion of traceback. One possible maximum alignment is:

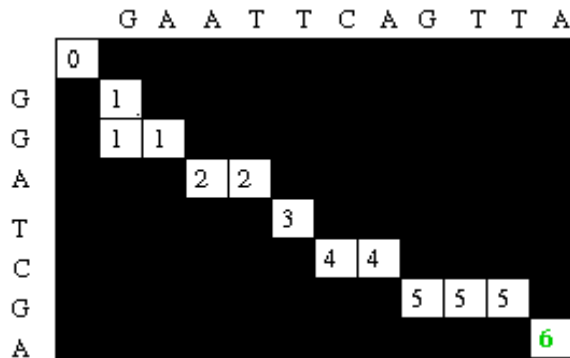


Giving an alignment of:

```

G A A T T C A G T T A
| | | | |
G G A _ T C _ G _ _ A
    
```

An alternate solution is:



Giving an alignment of :

```

G _ A A T T C A G T T A
|   | | | |
G G _ A _ T C _ G _ _ A
    
```

There are more alternative solutions each resulting in a maximal global alignment score of 6. Since this is an exponential problem, only a single solution has been printed by most of dynamic programming algorithms.

3.4 Multiple Sequence Alignment

Multiple sequence alignment [2], in which three or more sequences must be aligned, is useful in finding patterns in nucleotide sequences and for identifying structural and functional domains in protein families.

Multiple sequence alignment is as an extension of the pair-wise alignment. The first step in multiple sequence alignment is pair-wise alignment of all the sequences. For example four sequences- S_1 , S_2 , S_3 and S_4 have been considered. The alignment of these sequences involves 5 pair-wise comparisons (S_1 and S_2 , S_1 and S_3 , S_1 and S_4 , S_2 and S_3 , S_2 and S_4 and S_3 and S_4). This is shown in **Figure 3.4**. The result of this alignment has been represented as a tree or a dendrogram.



Figure 3.4: Pair-wise alignment of multiple sequences (6 pairwise comparisons then cluster analysis)

Then process of aligning the multiple sequences step-wise has been followed. The first step has been used to align the pair of most similar sequences, followed by less similar and so on. The gaps in the alignments have been used to optimize the alignment. This has been shown in **Figure 3.5**.

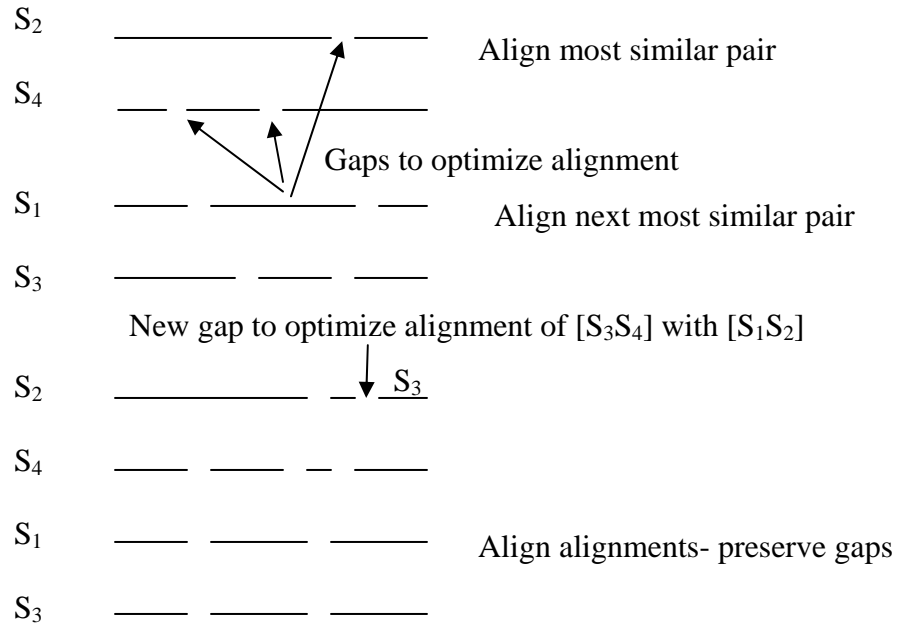


Figure 3.5: Step-wise alignment of sequences

3.4.1 Methods of Multiple Sequence Alignment

There are several approaches for conducting multiple sequence alignment. The most common approach to approach to multiple sequence alignment is Progressive Alignment.

Progressive Alignment Method

The approach of progressive alignment is to begin with an alignment of most alike sequences and then builds upon the alignment using other sequences. Progressive alignments work by first aligning the most alike sequences using dynamic programming and then progressively adding less related sequences to the initial alignment.

Cluster pair-wise alignment is a simple score between extensions of sequence alignment. In a pair-wise alignment, the comparison score between any two positions in those clusters is the arithmetic average of scores for all possible symbol comparisons at those positions. When gaps have inserted into a cluster to produce an alignment, they have been

inserted at the same position in all of the sequences of the cluster. The full multiple alignments are obtained once all the sequences have been clustered into one. This hierarchical clustering can be plotted as a dendrogram.

Since the alignment is calculated on a progressive basis, the order of the initial sequences can affect the final alignment. Different comparison matrices or gap weights affects the multiple alignments.

Consider the 5 sequences given in **Table 3.1** that need to be aligned.

Table 3.1: Set of 5 sequences

S ₁	T	C	Y	G	I	F	V	L		
S ₂	T	C	G	I	F	V	L			
S ₃	S	C	Y	G	I	F	V	L	S	G
S ₄	T	C	F	G	I	F	V	L		
S ₅	A	C	G	I	F	V	L	S	G	

All pair-wise comparisons are performed, resulting in a matrix of scores. The scores have been represented in **Table 3.2**.

Table 3.2: Scores for the Pair-wise Comparisons

	S ₁	S ₂	S ₃	S ₄	S ₅
S ₁		26	38	38	26
S ₂	26		26	26	32
S ₃	38	26		36	36
S ₄	38	26	36		26
S ₅	26	32	36	26	

The most closely-related pair of sequences is aligned first. In this example S_1 and S_3 and S_1 and S_4 have the same score, so they can be used as the first pair. The interactions are shown in **Table 3.3**.

Table 3.3: Steps in Aligning 5 Sequences Given Above

There are four steps.

Step 1 – start with S_1 and S_3

S_1	T	C	Y	G	I	F	V	L	-	-
S_3	S	C	Y	G	I	F	V	L	S	G

Step 2 – add S_4

S_1	T	C	Y	G	I	F	V	L	-	-
S_3	S	C	Y	G	I	F	V	L	S	G
S_4	T	C	F	G	I	F	V	L	-	-

Step 3 – add S_2

S_1	T	C	Y	G	I	F	V	L	-	-
S_3	S	C	Y	G	I	F	V	L	S	G
S_4	T	C	F	G	I	F	V	L	-	-
S_2	T	C	-	G	I	F	V	L	-	-

Step 4 – add S_5

S_1	T	C	Y	G	I	F	V	L	-	-
S_3	S	C	Y	G	I	F	V	L	S	G
S_4	T	C	F	G	I	F	V	L	-	-
S_2	T	C	-	G	I	F	V	L	-	-
S_5	A	C	G	I	F	V	L	S	G	-

The alignment can be thought of as occurring in a “star” configuration, where the sequence with the greatest similarity to the others is at the centre and the rays of the star represent the pair-wise distances to the remaining sequences. Each time a sequence is added, gaps are inserted either in newly added sequence or in the entire alignment to optimize alignment.

The first problem with progressive methods is that they depend upon the initial pair-wise sequence alignments. If the sequences are closely related then the likelihood is good that the initial alignment contains relatively few errors. However, if initial sequences are distantly related, then there will be more errors in the alignment, which will propagate through the rest of the alignments. The second problem is that suitable scoring matrices and gap penalties must be chosen to apply to the sequences as a set.

3.4.2 Application of Multiple Alignments

The basic information from a multiple alignment of protein sequences is the position and nature of the conserved regions in each member of the group. Conserved sequence regions correspond to functionally and structurally important parts of the protein. Hypotheses about functional importance or specific roles can then be directly tested by mutagenesis and truncation experiments.

Multiple sequence alignments can be used to find regions of similar sequence in all of the sequences that defines a conserved consensus pattern. If the alignment is strong, MSA can also be used to derive the possible evolutionary relationships among the sequences.

Multiple alignments are powerful tools for identifying new members of the aligned group. It is possible to query databases of multiple alignments with single sequences and to query sequence databases with multiple alignments. It has been observed that such searches are more sensitive and selective than sequence-to-sequence searches.

Multiple alignments of many sequences and those with different sequence weights are difficult to visualize. Sequence logos are a graphical way for presenting multiple alignments. A different graphical view of multiply aligned sequences is by a tree relating their sequence similarity. This is very useful when the aligned sequences are of several functional subtypes and we wish to know to which one our sequence/s belongs. A way to estimate the significance of a tree is by bootstrap values. These values have been used to find number of times branching point have been observed with different models of the

input data. The higher the fraction of the bootstrap value (number of observations/number of trials) indicates that the sequences emerging from that branch point cluster together.

Multiple alignments are powerful tools for identifying new members of the aligned group. It is possible to query databases of multiple alignments with single sequences and to query sequence databases with multiple alignments. It has been seen that such searches are more sensitive and selective than sequence-to-sequence searches.

The **Blimps** program has been used to query both protein and nucleotide sequence databases with protein blocks and vice versa. The queries are single sequences or blocks. The program is available on the WWW and by e-mail server for searching multiple alignment databases with single sequences.

The **MAST** program [3] has been used to query sequence databases with blocks. Protein or nucleotide databases are queried with protein blocks. The query can be a single block or all the blocks of a protein family.

The **LAMA** program has been used to search blocks databases with block queries. Queries can be obtained from the Blocks database, BlockMaker program or by reformatting multiple alignments.

3.5 Phylogenetic Analysis

The sequencing of DNA and proteins has become easy and fast with the use of automated tools. Similarity searches and multiple alignments of sequences have been used to find the relationship between the sequences.

Based on given a set of sequences, the evolutionary relationship among genes has been reconstructed. To reconstruct the evolutionary relationship, a branched structure termed a phylogeny or tree has been created. A phylogeny illustrates the relationship between the sequences. Analysis of phylogeny of a family has been done to know about molecular

evolution. Molecular evolution is based on mutations imparting the DNA to derive during evolution. The number of types of changes in residues of a MSA can be used to start a phylogenetic analysis. Each column in a MSA denotes mutations that occur at one site during evolution of sequence family.

Phylogenetic Trees

The method [4] of converting MSA to a phylogenetic tree is used to reduce the problem of a multiple alignment to an iterative process of pair-wise alignments. The purpose works as follows:

- Compute all pair-wise distance between given sequence
- Compute a tree by single linkage clustering by using methods like UPGMA or Nearest Neighbor.
- Align the sequences in an orderly fashion.

An evolutionary relationship has been represented using phylogenetic trees. A tree is 2D graph showing evolutionary relationships among organisms. This separate source of sequences has been referred as taxa, defined as phylogenetically distinct units on the tree. The tree is composed of nodes representing the taxa and branches representing the relationships among the taxa. An example of a rooted tree of 4 taxa is shown in **Figure 3.4**.

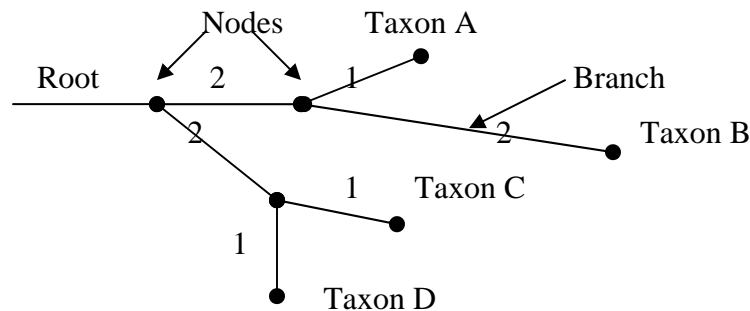


Figure 3.6: Example of a rooted tree of 4 taxa showing branch lengths proportional to the number of changes in branch

The most used terms in phylogenetic analysis are given in **Table 3.4**.

Table 3.4: Terms Used in Phylogenetic Analysis

<i>Node</i>	<i>A Node Represents a Taxonomic Units it can be a Taxon</i>
Branch	Defines the relationship between the taxa in terms of descent and ancestry.
Topology	Is the branching pattern
Branch length	Often represents the number of changes that have occurred in that branch.
Root	Is the common ancestor of all taxa.
Distance scale	Scale which represents the number of differences between sequences

3.5.1 Methods of Phylogenetic Analysis

There are two approaches to deriving phylogenetic trees. One approach makes no reference to any historical model of relationships. Proceed by measuring a set of distances between species and generate the tree by a hierarchical *Clustering procedure*. This is called the **phenetic** approach. The alternative, the *cladistic approach*, is to consider possible pathways of evolution, infer the features of ancestor at each node and choose an optimal tree according to some model of evolutionary change. Phenetics is based on similarity; cladistics is based on genealogy.

Clustering Methods

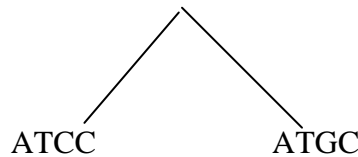
Clustering approaches to determination of phylogenetic relationships are explicitly non-historical. A simple clustering procedure works as follows: Given a set of species, determine for all pairs a measure of similarity or difference between them. To create a tree from the set of dissimilarities, first choose the two most closely related species and insert a node to represent their common ancestor. Then replace the two selected species by a set containing both and replace the distances from the pair to the others by the average of distances of the two selected species to the others. Now we have a set of pairwise dissimilarities not between individual species but between sets of species. (Regard each remaining individual species as a set containing only one element.) Then repeat the process, as the following example.

Example: Consider four species characterized by homologous sequences ATCC, ATGC, TTCG and TCGG. Taking the number of differences as the measure of dissimilarity between each pair of species, use a simple clustering procedure to derive a phylogenetic tree.

The distance matrix is:

	ATCC	ATGC	TTCG	TCGG
ATCC	0	1	2	4
ATGC		0	3	3
TTCG			0	2
TCGG				0

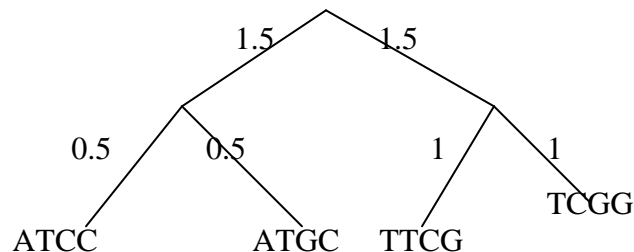
Because the matrix is symmetric, we need fill in only the upper half. The smallest nonzero distance is 1, between ATCC and ATGC. Therefore our first cluster is {ATCC, ATGC}. The tree will contain the fragment:



The reduced distance matrix is:

	(ATCC, ATGC)	TTCG	TCGG
(ATCC, ATGC)	0	$\frac{1}{2}(2+3)=2.5$	$\frac{1}{2}(4+3)=3.5$
TTCG		0	2
TCGG			0

The next cluster is {TTCG, TCGG}, distance 2. Finally, linking the clusters {ATCC, ATGC} and {TTCG, TCGG} gives the tree:



Branch lengths have been assigned according to the rule:

Branch length of edge between nodes X and Y = $\frac{1}{2}$ distance between X and Y.

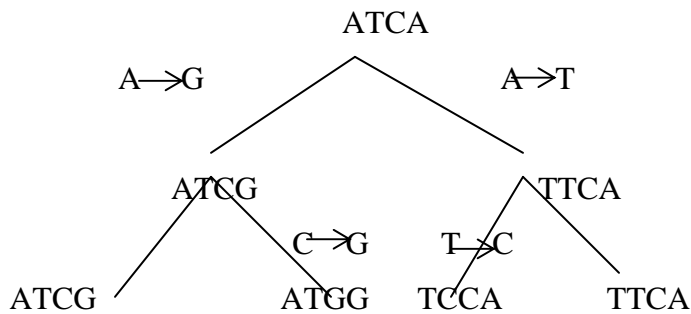
Whether the branch lengths are truly proportional to divergence times of the taxa represented by the nodes must be determined from external evidence.

This process of tree building is called the UPGMA method (Unweighted Pair Group Method with Arithmetic mean).

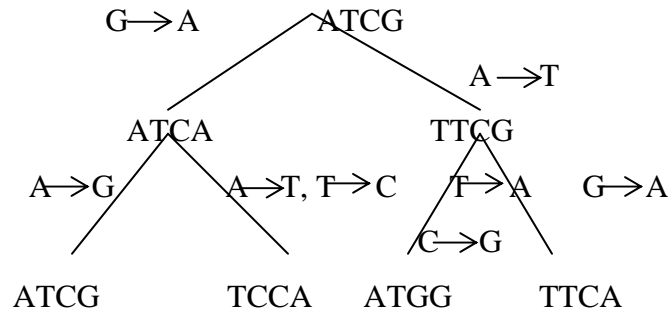
Cladistic Methods

Cladistic methods [5] deal explicitly with the patterns of ancestry implied by the possible trees relating a set of taxa. Their aim is to select the correct tree by utilizing an explicit model of the evolutionary process. The most popular cladistic methods in molecular phylogeny are the *maximum parsimony* and *maximum likelihood* approaches. They are specialized to sequence to sequence data, starting from a multiple sequence alignment.

The *maximum parsimony* [6] method defines an optimal tree as the one that postulates the fewest mutations. For instance, given species characterized by homologous sequences ATGC, ATGG, TCCA and TTCA, the tree postulates four mutations:



An alternative tree postulates seven mutations.



Note that the second tree implies that the $G \rightarrow A$ mutation in the fourth position occurred twice independently. The former tree is optimal according to the maximum parsimony method, because no other tree involves fewer mutations.

The *maximum likelihood* method assigns quantitative probabilities to mutational events, rather than merely counting them. Like maximum parsimony, maximum likelihood reconstructs ancestors at all nodes of each tree considered; but it also assigns branch lengths based on probabilities of mutational events. For each possible tree topology, the assumed substitution rates are varied to find the parameters that give the highest likelihood of producing the observed sequences. The optimal tree is the one with the highest likelihood of generating the observed data.

3.5.2 Computational Considerations

Cladistic methods- maximum parsimony and maximum likelihood requires large amounts of computer time. By considering Cladistic methods the total number of possible trees, increases very rapidly with the number of species.

Calculated phylogenies are often approximations, methods for testing them are:

1. Comparison of phylogenies obtained from different characters describing the same set of taxa. If trees produced from different characters share a subtree,

perhaps that portion of the phylogeny has been determined reliably and other portions have not.

2. Analysis of subsets of taxa should give the same answer – respect to the subset- as appears within the full tree.
3. Formal statistical test, involving returning the calculation on subsets of the original data, are known as **jackknifing** and **bootstrapping**:
 - **Jackknifing** is calculation with data sets samples randomly from the original data. For phylogeny calculations from multiple sequence alignments. Select different subsets of the positions in the alignment and return the calculation. Finding that each subset gives the same phylogenetic tree lends it credibility. If such subset gives a different tree, none of them is trustworthy.
 - **Bootstrapping** is similar to Jackknifing except that the positions chosen at random may include multiple copies of the same position to form data sets of the same size as the original to preserve statistical properties of the sampling.
4. If there are very long edges, then this has been considered seriously because there is the possibility of unequal variation in evolutionary rate that may have disturbed the calculation. So outgroup taxa has been introduced to check this.

3.5 References

- [1] Bryan Bergeron .“ Bioinformatics Computing” Eastern Economy Edition, 306-310,2003.
- [2] S.C. Rastogi, N.Mendiratta, P. Rastogi, “Bioinformatics Methods and Applications”, PHI, 93-100, 2006.
- [3] http://bioinformatics.weizmann.ac.il/blocks/process_blocks.html
- [4] Arthur M. Lesk. “Introduction to Bioinformatics” Oxford University Press, 203-209, 2005.
- [5] <http://evolution.genetics.washington.edu/phylip/software.html>
- [6] Whelan , S. Lio , P. & Goldman, N. , Molecular Phylogenetics : State –of-the-art methods for looking into past, Trends in Genetics,17 , 262 -272 , 2001

Chapter -4

Similarities Search and
Sequence Alignment

Chapter 4

SIMILARITIES SEARCH AND SEQUENCE ALIGNMENT

4.1 FASTA Algorithm

The FASTA algorithm [1] is a heuristic method for string comparison. FASTA compares a query string against a single text string. When searching the whole database for matches to a given query, the query using the FASTA algorithm to every string in the database has been compared.

When looking for an alignment, a few segments have been found in which there has absolute identity between the two compared strings. The algorithm is using this property and focuses on these identical regions.

The stages in the FASTA algorithm are as follows:

1. An integer parameter called *ktup* (short for *k respective tuples*), has been specified and *ktup*-length matching substrings of the two strings has been looked. The standard recommended *ktup* values are six for DNA sequence matching and two for protein sequence matching. The matching *ktup*-length substrings are referred to as *hot spots*. Consecutive hot spots are located along the dynamic programming matrix diagonals. This stage can be done efficiently by using a lookup table or a hash to store all the *ktup*-length substrings from one string, and then search the table with the *ktup*-length substrings from the other string.
2. In this stage 10 best *diagonal runs* of hot spots in the matrix has been found out. A diagonal run is a sequence of nearby hot spots on the same diagonal.

In order to evaluate the diagonal runs, FASTA gives each hot spot a positive score, and the space between consecutive hot spots in a run is given a negative score that decreases with the increasing distance. The score of the diagonal run is the sum of the hot spots scores and the interspot scores. FASTA finds the 10 highest scoring diagonal runs under this evaluating scheme.

3. A diagonal run specifies a pair of aligned substrings. The alignment is composed of matches (the hot spots) and mismatches (from the interspot regions), but it does not contain any indels because it is derived from a single diagonal. The runs have been using an amino acid (or nucleotide) substitution matrix, and pick the best scoring run. The single best subalignment found in this stage is called *init*₁. Apart from computing *init*₁, a filtration has been performed and the diagonal runs achieving relatively low scores have been discarded.
4. Until now any indels in the subalignments have not been allowed. Now it has been tried to combine “good” diagonal runs from close diagonals, thus achieving a subalignment with indels allowed. Then after “good” subalignments have been taken from the previous stage (subalignments whose score is above some specified cutoff) and attempt to combine them into a single larger high-scoring alignment that allows some spaces.
5. In this step FASTA computes an alternative local alignment score, in addition to *init*_n. A diagonal segment has been defined by *init*₁ in the dynamic programming matrix. A narrow diagonal band in the matrix has been considered, centered along this segment. The optimal local alignment in this band has been computed, using the ordinary dynamic programming algorithm. Assuming that the best local alignment is indeed within the defined band, the local alignment algorithm essentially merges diagonal runs found in the previous stages to achieve a local alignment which may contain indels. The band width is dependent on the *ktup* choice. The best local alignment computed in this stage is called *opt*.

- In the last stage, the database sequences are ranked according to $init_n$ scores or opt scores, and the full dynamic programming algorithm is used to align the query sequence against each of the highest ranking result sequences.

Although FASTA is a heuristic, and as such it is possible to show instances in which the alignments found by the algorithm are not optimal, it is claimed that the resulting alignment scores well compare to the optimal alignment, while the FASTA algorithm is much faster than the ordinary dynamic programming alignment algorithm.

4.1.2 FASTA Implementation

FASTA at the EBI is one of the most popular FASTA implementation. The FASTA input form is given in **Figure 4.1**.

Address: <http://www.ebi.ac.uk/Tools/fasta33/index.html>

Help Index

- General Help
- Formats
- Gaps
- Matrix
- References
- FASTA Help
- MView Help
- VisualFASTA Help

View all FASTA's at EBI

FASTA Programmatic Access

Database Information

- UniProt
- UniParc

Similar Applications

- FASTA
- BLAST
- MPsrch
- ScanPS
- SSEARCH

FASTA Related Literature

Search for FASTA related literature in Medline... [more](#)

FASTA and SSEARCH - Protein Similarity Search

Provides sequence similarity searching against protein databases using the FASTA and SSEARCH programs. **SSEARCH** does a rigorous Smith-Waterman search for similarity between a query sequence and a database. **FASTA** can be very specific when identifying long regions of low similarity especially for highly diverged sequences. You can also conduct sequence similarity searching against [nucleotide databases](#) or complete [proteome/genome](#) databases using the [FASTA programs](#).

[Download Software](#)

PROGRAM	DATABASES	RESULTS	SEARCH TITLE	YOUR EMAIL
FASTA	Protein	interactive	Sequence	binod.istar.1970

MATRIX	GAP OPEN	GAP EXTEND	KTUP	EXPECTATION UPPER VALUE	EXPECTATION LOWER VALUE
BLOSUM62	-10	-2	2	10.0	default

DNA STRAND:
 HISTOGRAM:
 MOLECULE TYPE:

SCORES	ALIGNMENTS	SEQUENCE RANGE	DATABASE RANGE	FILTER	STATISTICAL ESTIMATES
50	50	START-END	START-END	none	Regress

Enter or Paste a Sequence in any format:

```

MFQAFPGDYDSGSRCS SSPSAESQYLSSVDSFGSPPTAAASQECAGLGEMPGSFVPTVTAITTSQDLQ
WLVQPTLIS SMAQSQGQPLASQPPAVDPYDMPGTSYSTPGLSAYSTGGASGSGGPSTSTTTSGPVSAR
PARARPRRPREETLTP EEEKRRVRRRNKLA AAKCMNRRRELTDRLQ AETDQLEEEKAELESEIAEL
QKEKERLEFVLVAHKPGCKIPYEEGPGGPLAEVRDLPGSTSAKEDFGWLLPPPPPPPLPFQSSRDA
PPNLTASLFT HSEVQVLGDFP VVSPSYTSSFVLTCPVSAFAGA QRTSGSEQPSDPLNSP SLLAL
    
```

Figure 4.1: FASTA page at EBI

The Histogram

The histogram [2] compares the predicted extreme value distribution of local similarity scores, represented by asterisks with the actual number obtained, represented by equal signs. Each bar represents the number of local alignment having the z-opt score indicated in the first column. The second column is the actual number of sequences with that z-opt scores. The third column is the predicted number of alignments having z-scores in that interval.

For histogram for this particular sequence is given in **Figure 4.2**.

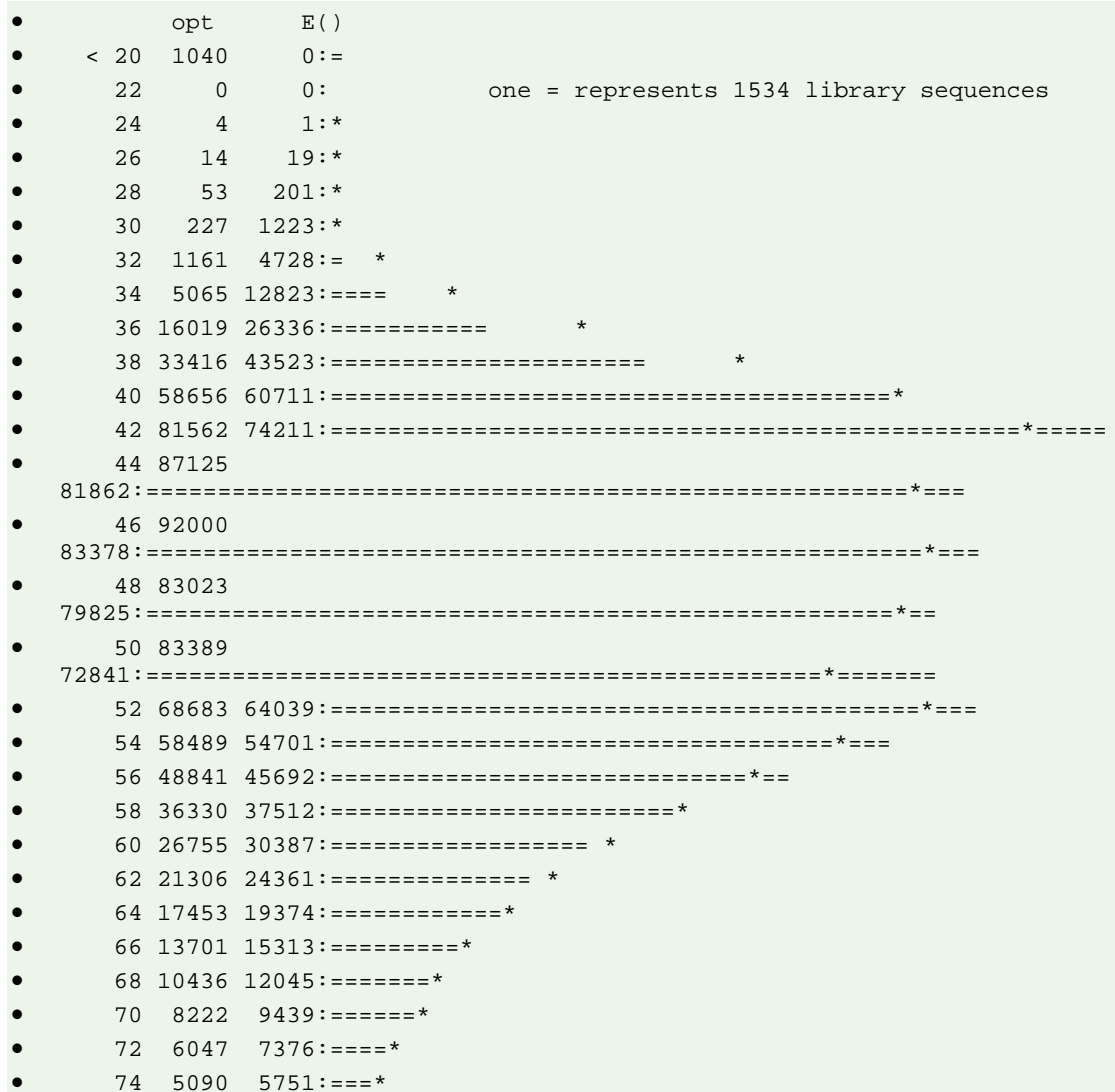


Figure 4.2: Histogram from FASTA output

The Sequence Listing

The next part of the output is a listing of the best scoring sequences in the database. The best alignments are reported first and the worst hits last.

The first column identifies the database sequence reported by database, database accession number and database identifier. The next column reports the total length of database sequence. The final scores are reported in the opt column and the E () value for that particular database sequence is reported in the last column.

FASTA Results

SUBMISSION PARAMETERS			
Title	Sequence	Database	uniprot
Sequence length	338	Sequence type	p
Program	fasta	Version	35.04 Mar. 8, 2009
Expectation upper value	10.0	Matrix	BL50
Sequence range	1-	Number of scores	50
Number of alignments	50	Word size	2
Open gap penalty	-10	Gap extension penalty	-2
Histogram	false		

Show Annotation
FASTA Result
MView
VisualFasta
XML
SUBMIT ANOTHER JOB

Clear all
Check all
Invert selection
Reset
Show Alignments
Download
fasta

Alignment	DB:ID	Source	Length	Identity%	Similar%	Overlap	E()
1 <input type="checkbox"/>	UNIPROT:A2RSD4_MOUSE	FBJ osteosarcoma oncogene	338	100.0	100.0	338	6.8e-104
2 <input type="checkbox"/>	UNIPROT:FOSB_MOUSE	Protein fosB OS=Mus musculus	338	100.0	100.0	338	6.8e-104
3 <input type="checkbox"/>	UNIPROT:FOSB_HUMAN	Protein fosB OS=Homo sapiens	338	95.9	98.8	338	2.7e-100
4 <input type="checkbox"/>	UNIPROT:A8K9K5_HUMAN	cDNA FLJ76347, highly simi	338	95.9	98.8	338	2.7e-100
5 <input type="checkbox"/>	UNIPROT:Q6FGK6_HUMAN	FOSB protein (Fragment) OS	338	95.5	98.5	337	1.2e-99
6 <input type="checkbox"/>	UNIPROT:A6QLQ3_BOVIN	FOSB protein OS=Bos taurus	341	95.0	98.2	341	1.4e-99
7 <input type="checkbox"/>	UNIPROT:FOSB_CANFA	Protein fosB OS=Canis famili	338	94.4	97.9	338	4.1e-98

Figure 4.3: Best Scoring Sequences

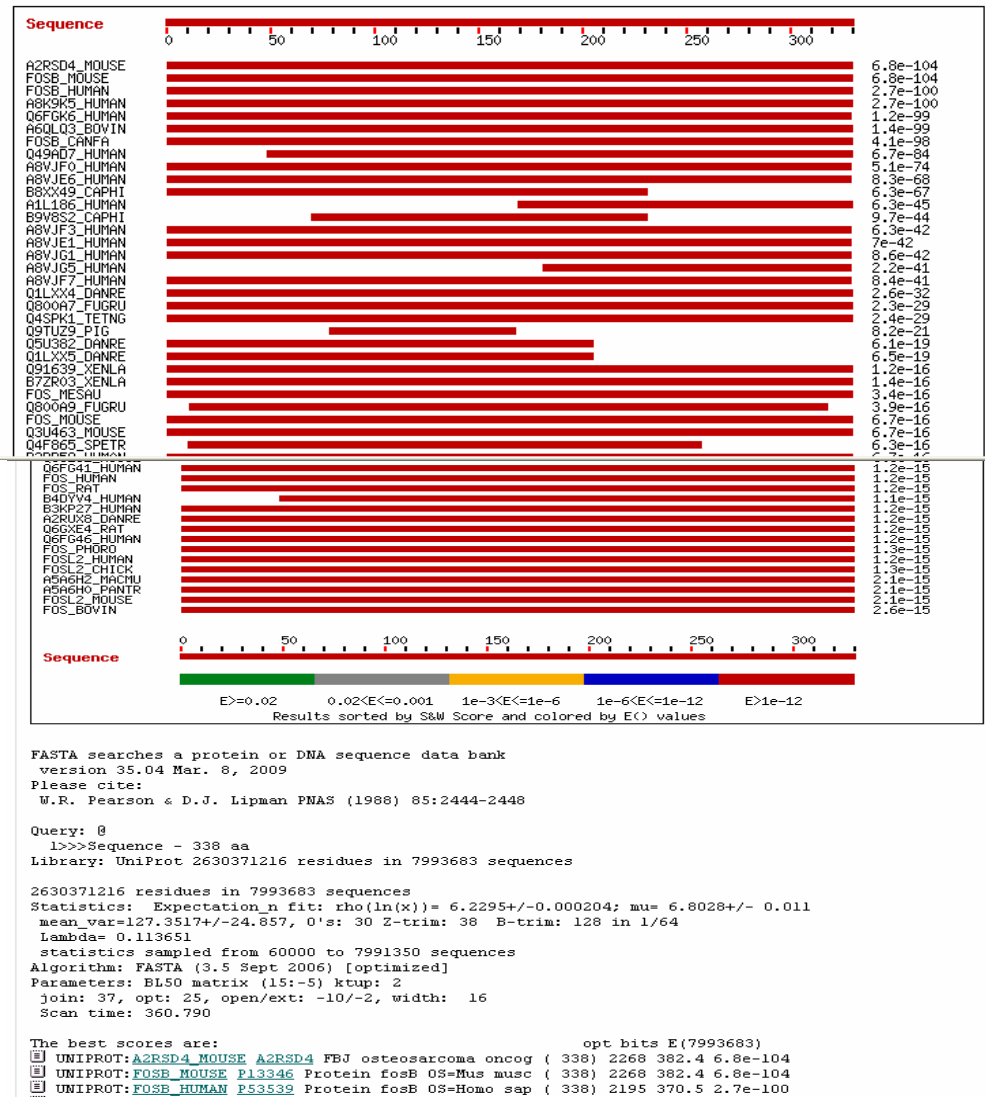


Figure 4.4: Visual FASTA Result

The Local Alignments

After the list of hits, the actual local alignments identified are displayed. The `initn` and `initl` columns report the local similarity scores calculated at different stages of the FASTA procedure. For each alignment, the various scores and $E()$ values are reported again along with some new information. The Smith-Waterman score reported is the same as the `opt` score. The percent identity and length of the alignments are displayed.

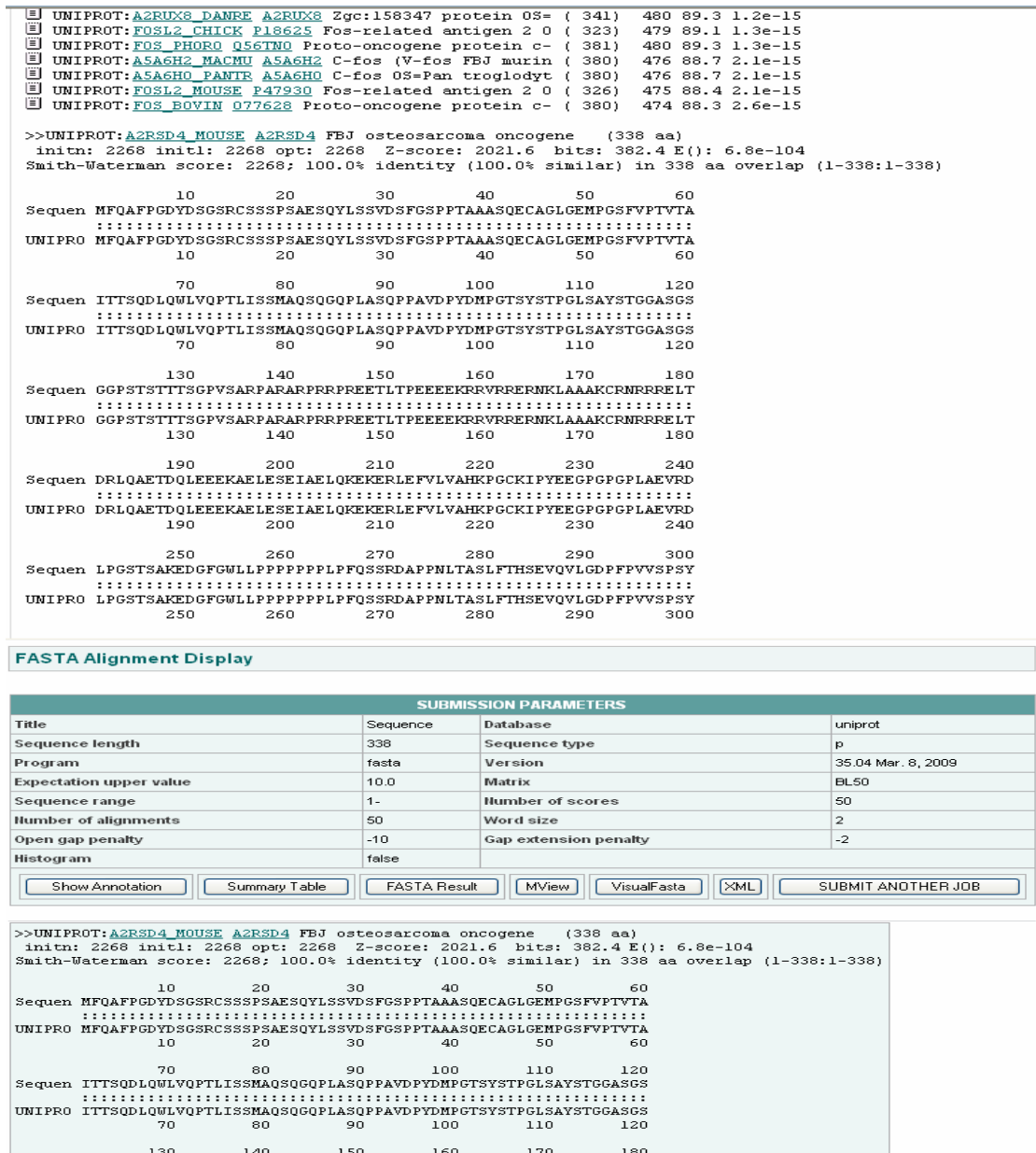


Figure 4.5: Local Alignment Score

Significance of the E-values

FASTA calculates an E-value (expectation of significance). E () values represent the number of sequences having a given alignment occurred by chance. In the above example of FASTA search, the best hits mostly have E () values of zero (0). This can be interpreted as: “we can expect zero sequences to have the score this alignment has, strictly by chance”.

E () values are calculated from the probability derived from the extreme value distribution for each z-opt score interval, and the number of sequences in the database. Therefore, as more sequences are added to a database, the E () values can change and sequences identified in one search may not be found in a later one.

4.2. BLAST Algorithm

BLAST is one of the most widely used bioinformatics programs, because it addresses a fundamental problem and the algorithm emphasizes speed over sensitivity. This emphasis on speed is vital to making the algorithm practical on the huge genome databases currently available, although subsequent algorithms can be even faster.

BLAST is about 50 times faster than dynamic programming; however, it cannot guarantee the optimal alignments of the query and database sequences as in the dynamic programming, but just works to find the related sequences in a database search. BLAST is more time efficient than FASTA by searching only for the more significant patterns in the sequences, but with comparative sensitivity.

Algorithm

To run, BLAST requires a query sequence to search for, and a sequence to search against (also called the target sequence) or a sequence database containing multiple such sequences. BLAST finds subsequences in the database which are similar to subsequences in the query. In typical usage, the query sequence is much smaller than the database, e.g., the query may be one thousand nucleotides while the database is several billion nucleotides.

The main idea of BLAST is that there are often high-scoring segment pairs (HSP) contained in a statistically significant alignment. BLAST searches for high scoring sequence alignments between the query sequence and sequences in the database using a heuristic approach that approximates the Smith-Waterman algorithm. The exhaustive Smith-Waterman approach is too slow for searching large genomic databases such as

GenBank. Therefore, the BLAST algorithm uses a heuristic approach that is less accurate than the Smith-Waterman but over 50 times faster.

An overview of the BLASTP algorithm (a protein to protein search) is as follows:

1. **Remove low-complexity region or sequence repeats in the query sequence.**

Low-complexity region means a region of a sequence is composed of few kinds of elements. These regions might give high scores that confuse the program to find the actual significant sequences in the database, so they should be filtered out. The regions is to be marked with an X (protein sequences) or N (nucleic acid sequences) and then be ignored by the BLAST program. To filter out the low-complexity regions, the SEG program is used for protein sequences and the program DUST is used for DNA sequences.

2. **Make a k-letter word list of the query sequence.**

The words is listed of length 3 in the query protein sequence by taking $k=3$ (k is usually 11 for a DNA sequence) “sequentially”, until the last letter of the query sequence is included. The method can be illustrated in **Figure 4.6**.

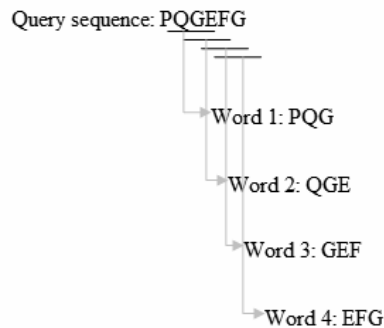


Figure 4.6: The method to establish the k-letter query word list

3. **List the possible matching words.**

This step is one of the main differences between BLAST and FASTA. FASTA cares about all of the common words in the database and query sequences that are listed in step 2; however, BLAST cares about only the high-scoring words. The

scores are created by comparing the word in the list in step 2 with all the 3-letter words. By using the scoring matrix (substitution matrix) to score the comparison of each residue pair, there are 20^3 possible match scores for a 3-letter word.

4. **Organize the remaining high-scoring words into an efficient search tree.**
This is for the purpose that the program can rapidly compare the high-scoring words to the database sequences.
5. **Repeat step 1 to 4 for each k-letter word in the query sequence.**
6. **Scan the database sequences for exact match with the remaining high-scoring words.**

The BLAST program scans the database sequences for the remaining high-scoring word, such as PEG, of each position. If an exact match is found, this match is used to seed a possible ungapped alignment between the query and database sequences.

7. **Extend the exact matches to high-scoring segment pair (HSP).**
 - The original version of BLAST stretches a longer alignment between the query and the database sequence in left and right direction, from the position where exact match is scanned. The extension does not stop until the accumulated total score of the HSP begins to decrease. A simplified example is presented in **Figure 4.7**.

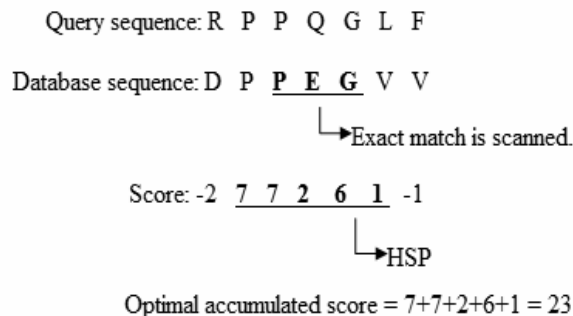


Figure 4.7: The process to extension the exact match.

- BLAST2 adopts a lower neighborhood word score threshold to maintain the same level of sensitivity for detecting sequence similarity. Therefore, the possible matching words list in step 3 becomes longer. Next, the exact matched region, within distance A from each other on the same diagonal in **Figure 4.8**, has been joined as a longer new region. Finally, the new regions are then extended as the same method in the original version of BLAST, and the HSPs' (High-scoring segment pair) scores of the extended regions are then created by using a substitution matrix as before.

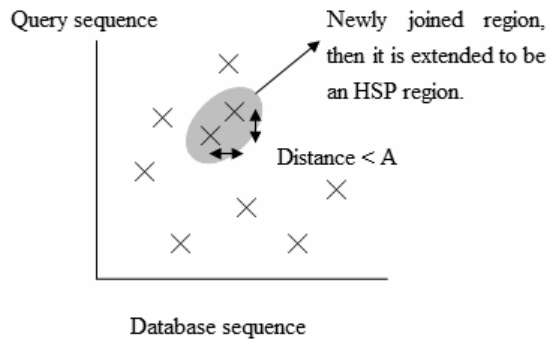


Figure 4.8: The positions of the exact matches.

8. **List all of the HSPs in the database whose score is high enough to be considered.**

The HSPs have been listed whose scores are greater than the empirically determined cutoff score S . By examining the distribution of the alignment scores modeled by comparing random sequences, a cutoff score S can be determined such that its value is large enough to guarantee the significance of the remained HSPs.

9. **Evaluate the significance of the HSP score.**

BLAST next assesses the statistical significance of each HSP score by exploiting the Gumbel extreme value distribution (EVD). In accordance with the Gumbel EVD, the probability p of observing a score S equal to or greater than x is given by the equation

$$p(S \geq x) = 1 - \exp(-e^{-\lambda(x-\mu)})$$

$$\text{Where } \mu = \frac{[\log Km'n']}{\lambda}$$

The statistical parameters λ and K are estimated by fitting the distribution of the ungapped local alignment scores, of the query sequence and a lot of shuffled versions (Global or local shuffling) of a database sequence, to the Gumbel extreme value distribution. λ and K depend upon the substitution matrix, gap penalties, and sequence composition (the letter frequencies). The m' and n' is the effective length of the query and database sequence, respectively. The original sequence length is shortened to the effective length to compensate for the edge effect. They can be calculated as:

$$m' \approx m - \frac{(\ln Km'n')}{H}$$

$$n' \approx n - \frac{(\ln Km'n')}{H}$$

Where H is the average expected score per aligned pair of residues in an alignment of two random sequences. Altschul and Gish gave the typical values, $\lambda = 0.318$, $K = 0.13$, and $H = 0.40$, for ungapped local alignment. The expect score E of a database match is the number of times that an unrelated database sequence would obtain a score S higher than x by chance. The expectation E obtained in a search for a database of D sequences is given by

$$E \approx 1 - e^{-p(x>s)D}$$

Furthermore, when $p < 0.1$, E could be approximated by the Poisson distribution as: $E \approx pD$

10. Make two or more HSP regions into a longer alignment.

Sometimes, it has been found that two or more HSP regions in one database sequence that can be made into a longer alignment. This provides additional evidence of the relation between the query and database sequence. There are two methods, the Poisson method and the sum-of scores method, to compare the significance of the newly combined HSP regions. Suppose that there are two

combined HSP regions with the sets of score (65, 40) and (52, 45), respectively. The Poisson method gives more significance to the set with the lower score of each set is higher ($45 > 40$). However, the sum-of-scores method prefers the first set, because $65+40$ (105) is greater than $52+45$ (97). The original BLAST uses the Poisson method; gapped BLAST and the WU-BLAST use the sum-of scores method.

11 Show the gapped Smith-Waterman local alignments of the query and each of the matched database sequences.

- The original BLAST only generates ungapped alignments including the initially found HSPs individually, even when there is more than one HSP found in one database sequence.
- BLAST2 versions produce a single alignment with gaps that can include all of the initially found HSP regions. Note that the computation of the score and its corresponding E score is involved with the adequate gap penalties.

12 Report matches whose expect score is lower than a threshold parameter E.

4.2.1 BLAST: Output

The protein sequence has been used for BLAST output [4]. First there is a short description of the program options chosen. Then there is a list of all of database sequences that match our query sequence. Several numbers are assigned to each of these sequences that represent the quality of the match.

```

#####
! Program: garnier
! Rundate: Sun 9 Apr 2009 10:53:46
! Commandline: garnier
! -sequence 'C:\Documents and Settings\Binodkumar\Local Settings\Temp\garnie
! -idc 0
! -rformat tagseq
! -auto
! Report_format: tagseq
! Report_file: ovax_chick.garnier
-----
# HitCount: 52
#
# DCH = 0, DCS = 0
#
# Please cite:
# Garnier, Osguthorpe and Robson (1978) J. Mol. Biol. 120:97-120
#
#
#-----
      . 10 . 20 . 30 . 40 . 50
      QIKDLLVSSSTDLDTLLVLVNAIYFKGMWKTAFNAEDTREMPPFHVTKQES
helix HH          HHHHH H HHHHHHHHHHHHHHHHH
sheet EEEEE EEEEEEE
turns          T          TT          T
coil      CCCC C          C CC          C
      . 60 . 70 . 80 . 90 . 100
      KPVQMMCMNNSFNVALPAEKMKILELPFASGDL SMLVLLPDEVSDLERI
helix HHH          HHHHHHHHHHHHHHHHHHHHH HHHHHHHHH
sheet EEE          EEEEE
turns          TT T
coil C          CC C
      . 110 . 120 . 130 . 140 . 150
      EKTINFEKLTWNTNPTMEKRRVKVYLPQMKIEEKYNTSVMALGMTDL
helix HHHHHHH          HHHHHHH HHHHHHHHHHHHHHH HHHHH
sheet HHHHHHH          EEE          E EE
turns          TT
coil      CCC CCCCC
      . 160 . 170 . 180 . 190 . 200
      FIPSANLTGISSAESLKISQAVHGAFMELSEDGIEMAGSTGVIEDIKHSP
helix          HHHHHHHHHHH HHHHHHHHHHH HH
sheet EE          EE
turns          T
coil CCCCCCCCC          C          CCCC CCCC
      . 210 . 220 . 230
      ESEQFRADHPFLFLIKHNPTNTIVYFGRYWSP
helix HHHHHHH HHHH
sheet          EE EEEEE
turns T          T TT TTTT
coil      C          CCC          CC
#-----
#
# Residue totals: H:131 E: 38 T: 18 C: 45
# percent: H: 60.6 E: 17.6 T: 8.3 C: 20.8
#
#-----
# Total_sequences: 1
# Total_hitcount: 52
#-----

```

Figure 4.9: Output from BLAST Query

BLAST Basic Local Alignment Search Tool

Home Recent Results Saved Strategies Help

NCBI/BLAST/blastn suite

blastn blastp blastx tblastn tblastx

Enter Query Sequence BLASTII programs search nucleotide databases using a nucleotide query. [more...](#)

Enter accession number, gi, or FASTA sequence Clear Query subrange From To

>EST_Clone_DW1
 ATAGTACAACTTAGGGCTTTTATTTCAGGCAGTAAAGTAAAGGAACAGCAAAGTGGGAGGGCTACACCAT
 CACCATGGCAACAGAAAGCCTCAAAAACATAAAGTCCCTCGACTTATGTCGGGTAGACTCTTCTAGCTC
 AGGAGAAACACATTTTAACTGGCTGAGGACAAGGCCAGGCAGCCTGGCCACTGCGGAAAGGGCAGNTGG
 ACGCGGGCTCTGGTCAGTCTTGAAGTCTTGGTGA666CTTCAGCAGTCTCTGCTTCTCAGACCA

Or, upload file Browse...

Job Title
 Enter a descriptive title for your BLAST search

Align two or more sequences

Choose Search Set

Database Human genomic + transcript Mouse genomic + transcript Others (nr etc.):
 Nucleotide collection (nr/nt)

Organism Optional
 Enter organism name or id--completions will be suggested

Enter organism common name, binomial, or tax id. Only 20 top taxa will be shown.

Entrez Query Optional
 Enter an Entrez query to limit search

Figure 4.10: Query sequence section of Nucleotide Blast Form.

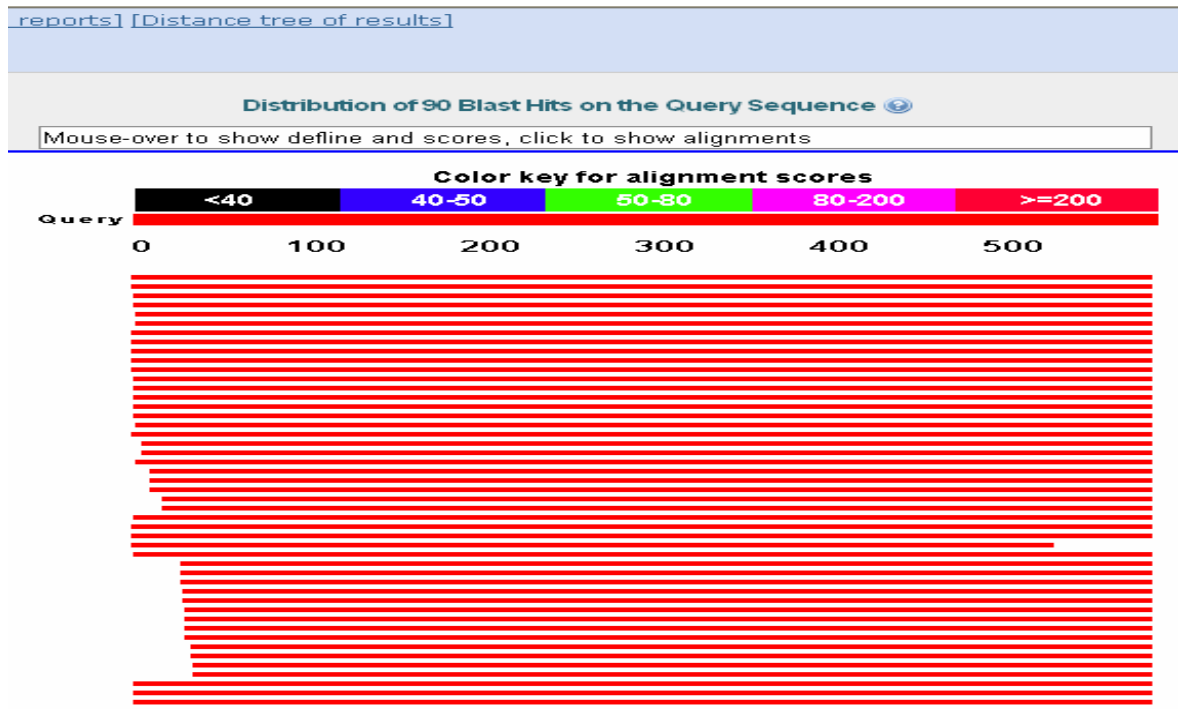


Figure 4.11: Graphic Summary of Nucleotide Blast Form

4.2.2 BLAST Services

The BLAST web server, hosted by the NCBI, allows with a web browser to perform similarity searches against constantly updated databases of proteins and DNA that include most of the newly sequenced organisms.

BLAST is actually a family of programs (all included in the blastall executable). These include:

1. Nucleotide-nucleotide BLAST (blastn)

This program, given a DNA query, returns the most similar DNA sequences from the DNA database that the user specifies.

2. Protein-protein BLAST (blastp)

This program, given a protein query, returns the most similar protein sequences from the protein database that the user specifies.

3. Position-Specific Iterative BLAST (PSI-BLAST)

This program is used to find distant relatives of a protein. First, a list of all closely related proteins is created. These proteins are combined into a general "profile" sequence, which summarizes significant features present in these sequences. A query against the protein database is then run using this profile, and a larger group of proteins is found. This larger group is used to construct another profile, and the process is repeated. By including related proteins in the search, PSI-BLAST is much more sensitive in picking up distant evolutionary relationships than a standard protein-protein BLAST.

4. Nucleotide 6-frame translation-protein (blastx)

This program compares the six-frame conceptual translation products of a nucleotide query sequence (both strands) against a protein sequence database.

5. Nucleotide 6-frame translation-nucleotide 6-frame translation (tblastx)

This program is the slowest of the BLAST family. It translates the query nucleotide sequence in all six possible frames and compares it against the six-frame translations of a nucleotide sequence database. The purpose of tblastx is to find very distant relationships between nucleotide sequences.

6. Protein-nucleotide 6-frame translation (tblastn)

This program compares a protein query against the all six reading frames of a nucleotide sequence database.

7. Large numbers of query sequences (megablast)

When comparing large numbers of input sequences via the command-line BLAST, "megablast" is much faster than running BLAST multiple times. It concatenates many input sequences together to form a large sequence before searching the BLAST database, and then post-analyze the search results to glean individual alignments and statistical values.

Table 4.1: BLAST Program Options

Program	Query Sequence	Database	Alignment Type
Blastp	Protein	Protein	Gapped
Blastn	Nucleic acid	Nucleic acid	Gapped
Blastx	Translated nucleic Acid	Protein	Each frame gapped
Tblastn	Protein	Translated nucleic Acid	Each frame gapped
Tblastx	Translated nucleic Acid2	Translated nucleic Acid	Ungapped

4.2.3 FILTERING and GAPPED BLAST

Filtering is the process of removing the undesired sequences from the query sequence prior to the search. BLAST filters regions of low-complexity. If my sequence contains large regions of “low complexity” it may not significant hits to the database.

GAPPED- BLAST

Gapped –BLAST is BLAST 2.0[6]. It represents BLAST plus a new heuristic for gapped alignments. It allows the introduction of gaps (deletions and insertions) into alignments. With a gapped alignment tool, homologous domains do not have to be broken into several segments. The programs, blastn and blastp offer fully gapped alignments. blastx

and tblastn have ‘in-frame’ gapped alignments and use sum statistics to link alignments from different frames.

Output of BLAST 2.0 is as follows:

- Information on Query sequence and Databases used
- Histogram, like the FASTA histogram
- Scores in bits-E value: number of hits expected to be reported by chance
- Alignments found(default =50)
- Parameters used in BLAST search.

4.2.4 FASTA and BLAST Algorithms Comparison

Table 4.2: Comparison of BLAST and FASTA

<p>1.FASTA offers many of the same functionalities as BLAST. Although BLAST tools are faster, FASTA provides more accurate sequence alignments. BLAST uses a different algorithm, but its results are similar to those found by FASTA. BLAST uses a general set of rules to compare specific regions of similarity for a given search string. BLAST is more than a precision matching tool. It provides a method for comparing the structures and functions of samples to genetic sequences and proteins</p>
<p>2.FASTA is superior to BLAST for translated DNA–protein comparison and DNA database searches because it calculates a single alignment that allows frame shifts. In contrast, BLAST performs forward-frame searches separately. By treating forward-reading frames as a single sequence, FASTA makes it much easier to produce high-quality alignments that extend the length of the protein sequence, resulting in improved sensitivity.</p>
<p>3.FASTA is a little more flexible than BLAST for DNA sequence searches. It provides small word sizes to accommodate polymerase chain reaction primers having short sequences. And FASTA uses several different scoring matrixes to help identify sequences of varying lengths.</p>

4.3 References

- [1] <http://www.ebi.ac.uk/fasta33/genomics.html>
- [2] <http://www.ebi.ac.uk/snpfasta3/index.html>
- [3] <http://www.ncbi.nlm.nih.gov/BLAST>
- [4] www.ch.embnet.org/software/bBLAST.html
- [5] <http://www.ncbi.nlm.nih.gov/BLAST/>
- [6] S.C. Rastogi, N.Mendiratta, P. Rastogi, “Bioinformatics Methods and Applications”, PHI, 144-145, 2006.

Chapter -5

Protein Structure and
Cheminformatics

bond usually is rigid (**Figure 2**). The existence of an amino group (**N-Terminal**) at one end of the chain and a carboxy group (**C-Terminal**) at the other end designs a direction to the chain. Conventionally the beginning of a polypeptide is its N-Terminal.

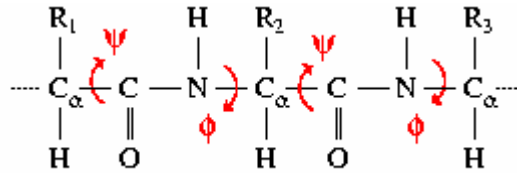


Figure 5.2: Torsion (or dihedral) angles of the backbone

5.2 Different Levels of Protein Structure

The wide variety of 3-dimensional protein structures corresponds to the diversity of functions proteins fulfill.

Proteins fold in three dimensions. Protein structure is organized hierarchically from so-called *primary structure* to *quaternary structure*. Higher-level structures are *motifs* and *domains*.

Above all the wide variety of conformations is due to the huge amount of different sequences of amino acid residues. The **primary structure** is the sequence of residues in the polypeptide chain.

Secondary structure is a local regularly occurring structure in proteins and is mainly formed through hydrogen bonds between backbone atoms. So-called random coils, loops or turns don't have a stable secondary structure. There are two types of stable secondary structures: **Alpha helices and beta-sheets** (**Figure 3 and Figure 4**). Alpha-helices and beta-sheets are preferably located at the core of the protein, where as loops prefer to reside in outer regions.

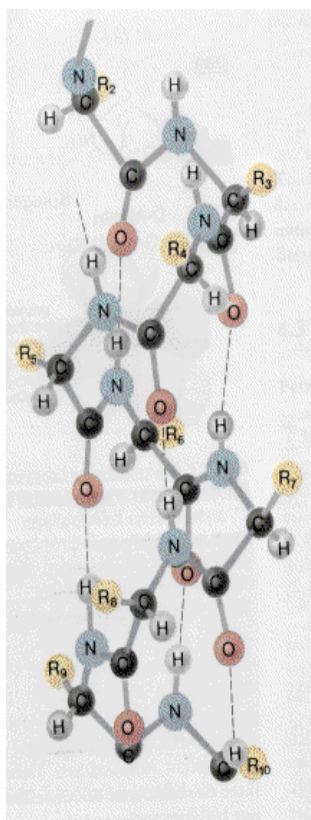


Figure 5.3: An alpha helix: The backbone is formed as a helix. An ideal alpha helix consists of 3.6 residues per complete turn. There are hydrogen bonds between the carboxy group of amino acid n and the amino group of another amino acid $n+4$ [1][2]. The mean phi angle is -62 degrees and the mean psi angle is -41 degrees [3].

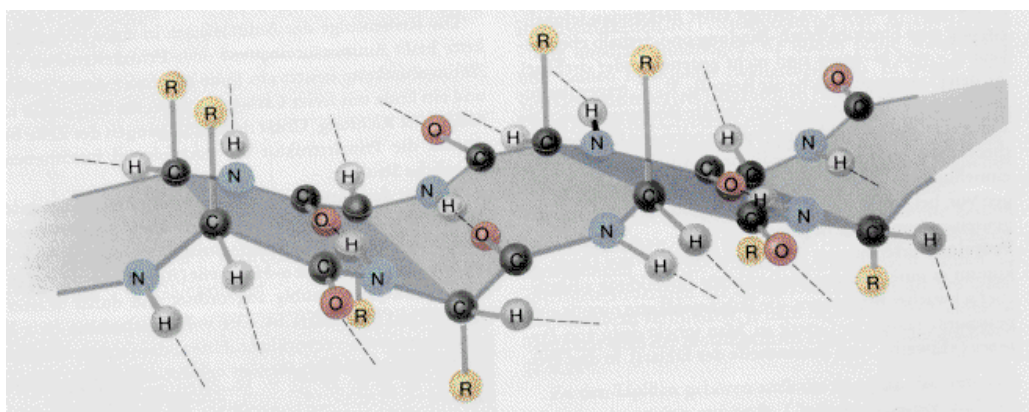


Figure 5.4: An antiparallel beta sheet. Beta sheets are created, when atoms of beta strands are hydrogen bond. Beta sheets may consist of parallel strands, antiparallel strands or out of a mixture of parallel and antiparallel strands [4].

Tertiary structure describes the packing of alpha-helices, beta-sheets and random coils with respect to each other on the level of one whole polypeptide chain. **Figure 5** shows the tertiary structure of Chain B of Protein Kinase C Interacting Protein.



Figure 5.5: Secondary structure of Protein. Helices are visualized as ribbons and extended strands of betasheets by broad arrows

Quaternary structure only exists, if there is more than one polypeptide chain present in a complex protein. Then quaternary structure describes the spatial organization of the chains. **Figure 5.6** shows both, Chain A and Chain B of Protein Kinase C Interacting Protein forming the quaternary structure.

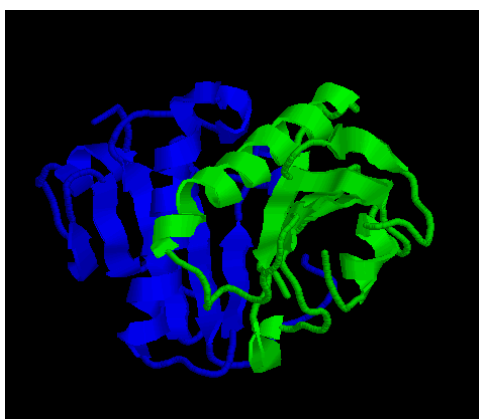


Figure 5.6: Quaternary structure of Protein.

5.3 Prediction Methods

Protein's 3D structure helps us to understand its functionality and provides means for planning experiments and drug design. Experimental methods given by X-ray crystallography and NMR spectroscopy to determine protein structure. The Brookhaven Protein Data Bank (PDB) is the repository for those structures. Files including atom coordinates which are suited for visualization by graphical molecule viewers like rasmol can be obtained at this site. PDB is also searchable with a sequence as a query, e.g. with the BLAST service located at NCBI with a polypeptide as a query.

The various prediction methods are based on the assumption, that the three-dimensional protein structure is determined by its primary structure.

Structure prediction methods are divided two categories:

1. Ab Initio Methods
2. Heuristic Methods

1. Ab Initio Methods

Ab Initio methods of determining protein structure are based on sequence data and molecular dynamics. One assumption is that a protein's secondary structure can be completely defined as a function of bond lengths, bond angles and torsion angles. The overall process of predicting tertiary protein structure from known sequence is illustrated in **Figure 5.7**. Given a sequence of amino acids, the first step is to generate a secondary structure by using bond lengths, angles and torsion angles. The next phase of process, generating the tertiary structure, involves methods such as molecular dynamics to create a library of tertiary protein structure.

Molecular dynamics calculates the force on each atom and move that atom a distance in small unit of time. The process is repeated until a pre-determined time limit is reached.

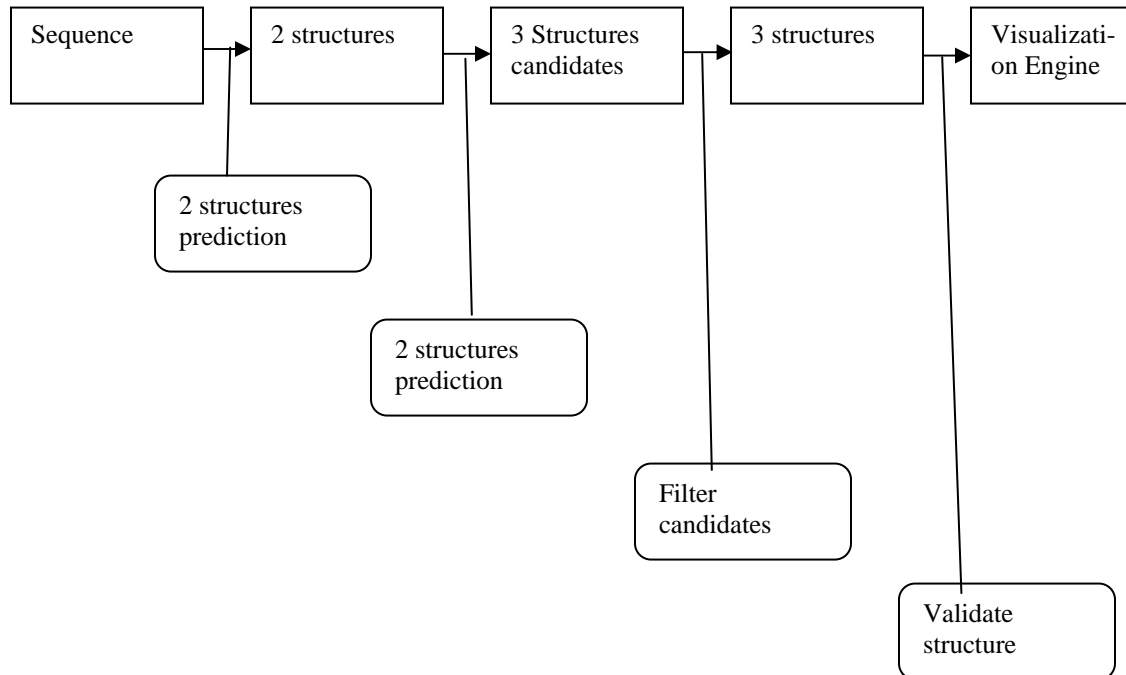


Figure 5.7: General Ab Initio Protein Structure Process

From a library of 3D structure candidates, the most promising structures are filtered from less capable structures. A common method of filtering to identify the most stable molecular conformations is based on the assumption that the native conformation of a protein is the conformation with the lowest energy. Once the top protein structure candidate is identified, it is validated and visualized. In validation process there is comparison of the predicted protein structure with a structure derived from NMR experiments. Validation process involves assigning a figure of merit to the predicted structure, based on comparison to the gold standard. The most often-used figure of merit in protein structure comparison is the root mean squared deviation (RMSD). The calculation for RMSD, expressed in Angstroms (**Figure 5.8**).

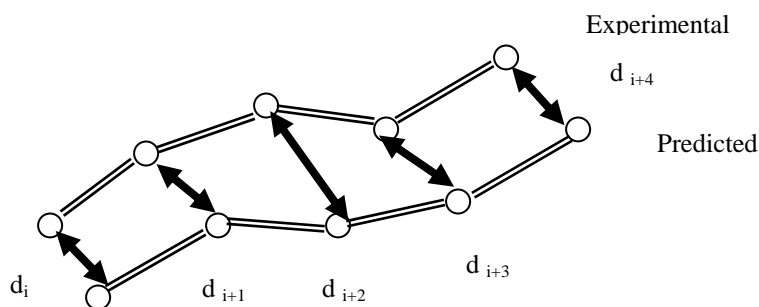


Figure 5.8: RMSD Calculation.

Where,

$$RMSD = \sqrt{\frac{\sum d_i^2}{N}}$$

N = Number of atoms

d = the distance in Angstroms between corresponding atoms in experimental and predicted protein structures.

Identical structures with perfect match would have an RMSD of 0; matching short to moderate –length protein structures have RMSD in the 1-3 Angstrom range. An RMSD of 5 or 6 Angstroms may be intolerable in a molecule with only 50 residues, but perfectly acceptable in large protein molecules for applications such as searching structure databases for known protein structures. However, even a relative measure, RMSD is valuable when working within a single family of proteins because the size of structures will be about same.

Visualization of the protein structure is performed through protein engines on the web, such as Rasmol or SWISS-PDBViewer.

2. Heuristic Methods

Heuristic methods use a database of protein structures to make prediction about the structure of newly sequenced proteins. In heuristic methods most newly sequenced proteins share structural similarities with proteins whose structures and sequences are known, and that these structures can serve as templates for new sequences. It is also assumed that relatively substantial changes in amino acid sequence may not alter the protein structure.

The main heuristic method of predicting protein structure from amino acid sequence data is comparative modeling i.e. to find similarities in amino acid sequence. Comparative modeling assumes that protein with similar amino acid sequences share the same basic 3D structure.

The basis for comparative modeling is typically the PDB, which contains description of 3D structures of proteins and other molecules as determined by NMR and X-ray crystallography experiments. Protein structures defined within Protein Database Modeling (PDM) and virtually every other protein structure database are based on assumptions that may not be completely valid. For example, the common assumption that amino acid sequences result in similar protein structures is known to have exceptions.

Comparative modeling is an iterative, multi-phase process. In **Figure 5.9**, given protein sequence data, the main phases of process are template selection, alignment, model building and evaluation. 3D visualization is often performed as part of the evaluation phase. The key activities in each phase of the comparative modeling process are outlined here.

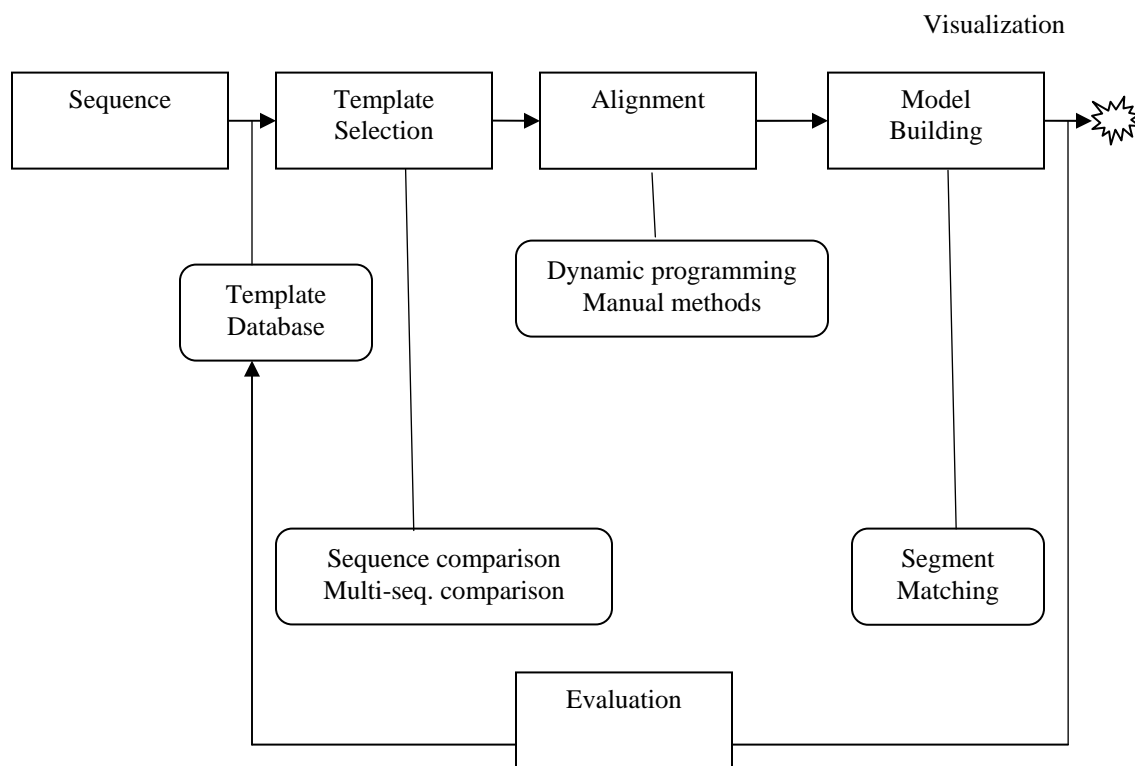


Figure 5.9: Comparative Modeling Process

Template Selection

Template selection involves searching a template database for the closest match or matches to the new (target) molecules, based on the target's amino acid sequence. The goal of template selection is to discover a link between the target protein and a known protein structure. For this usually PDB are used. Selecting an appropriate group of database entries from the database to serve as structure templates is based on sequences comparisons or threading.

Pairwise sequence comparison involves searching selections of the template candidate for amino acid sequences that are similar to sequences in the target protein. Multiple sequence comparison relies on an iterative algorithm that expands the template search to include all candidate templates from the template database.

Threading involves aligning the sequence of the target protein with the 3D structure of a template to determine whether the amino acid sequences is spatially and chemically similar to the template.

Alignment

The main aim of the alignment phase of comparative modeling is to align the sequence of polypeptides in the target sequence with that of the template structure in order to position the target and template in the same 3D orientation. Many of the alignment procedures are based on dynamic programming techniques.

Model Building

Actual model building begins only after identification of the libraries of templates that match the target protein. The structure of one of the template exactly fits the target protein, signifying that the structure of that target is identified to that of the template.

Rigid body assembly approach uses large segments of the template that are dissected at natural folds and reassembled over the superimposed structure of the target molecule. The accuracy of model building through rigid assembly is increased because there is an increased chance of availability of sub-assembly of molecule that matches the sequence in a corresponding area in the target protein as shown in **Figure 5.10**.

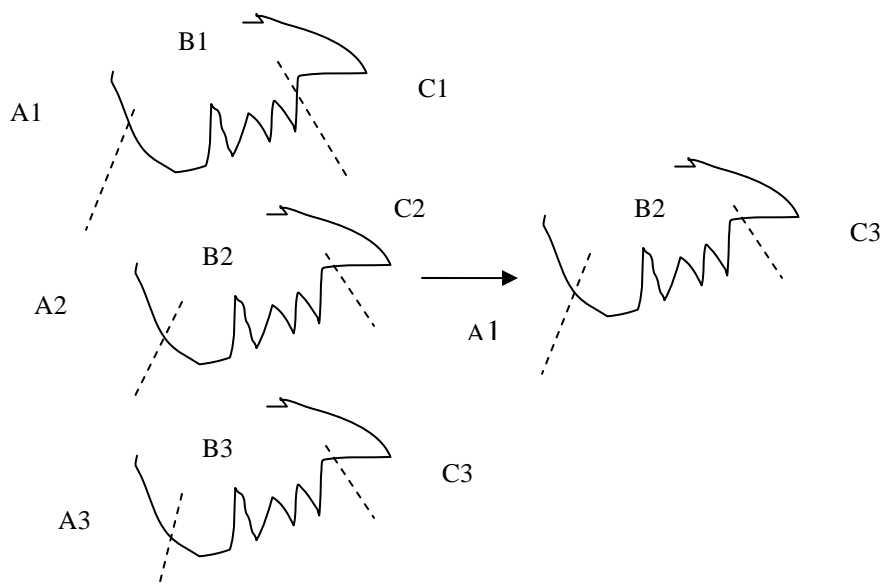


Figure 5.10: Rigid Body Assembly of Protein Structure

The aim of segment matching is to identify areas on structure templates that match areas in the target protein with similar sequences. These short matching segments in template are used as guiding positions in the target molecule, as shown in **Figure 5.11**.

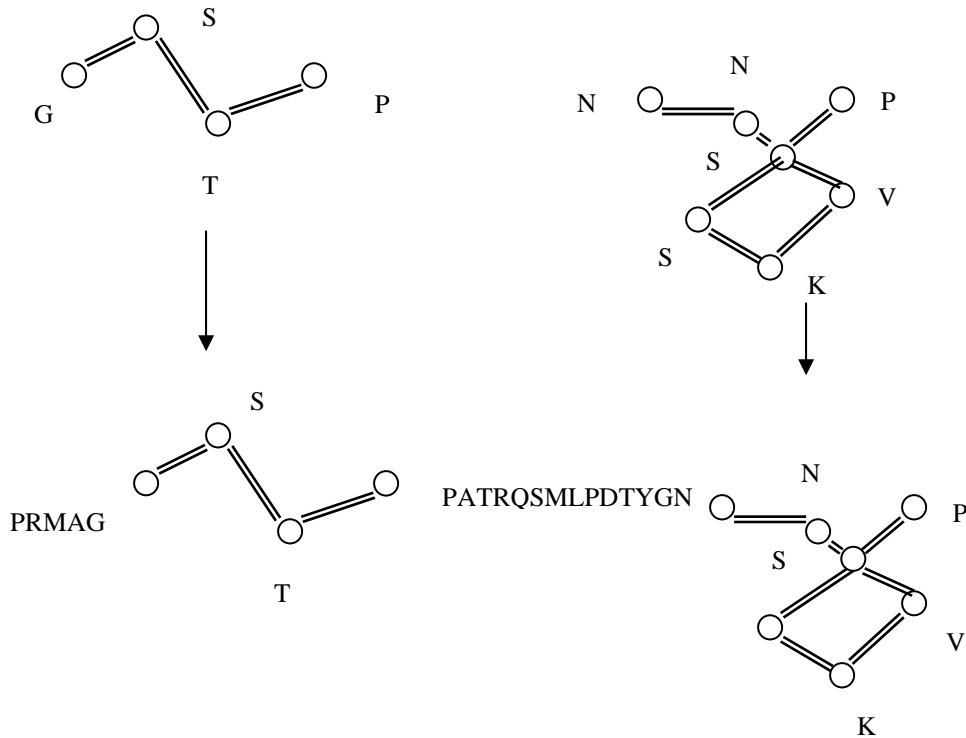


Figure 5.11: Short Segment Assembly of Protein Structure

Evaluation

In evaluating comparative modeling, even best methods like ab initio methods rarely achieve accuracies approaches 70 percent. The modeling process is repeated dozen of times before a reasonable target structured is constructed. In this evaluation process visualization tool is used to validate gross measures.

For quantitative evaluation, a measure of target-template similarity can be used. The greater the similarity of the model with the closest template, as measured by RMSD, the more likely the model is an accurate prediction of the actual structure.

5.4 Secondary Structure Prediction

Linus Pauling [5] suggested that amino acid chains could assume regular local structures, namely alpha helices and beta strands. In between these secondary structure elements there are turns or loops. There is a long tradition of attempts to predict local secondary structure based on sequence. State-of-the-art secondary structure prediction generally observes the frequencies of occurrences of k-tuples in particular secondary structures. Based on this statistic prediction can be made for a new sequence.

Chou and Fasman [6] apply a basic log-odds approach for the occurrences of single amino acid residues in the sequence, while the GOR method [7] which is based on information theory uses all possible pair frequencies within a sliding window.

As long as one restricts to the problem to the prediction for a single sequence there seems to be an inherent limit in prediction accuracy of around 65%. Multiply aligned sequences offer a means to surpass this limit. The PHD-method [8] uses evolutionary information from multiple sequence alignments in a multi level system of neural networks. So, the average accuracy of PHD-method is greater than 72%.

The GOR method is used to estimate for the prediction of secondary structure of protein.

5.5 The Protein Folding Problem

It has long been known that the structure of a protein is determined purely by the amino acid sequence [9], and the structure of the protein determines the function. The function of a protein depends entirely on the ability of the protein to fold rapidly and reliably to its native structure.

This folding process satisfy two conditions - one thermodynamic, and one kinetic. The thermodynamic consideration is that the protein adopts a single, stable, folded conformation. The kinetic requirement is that the protein must fold to the native state on an appropriate timescale. It has been suggested that for a protein of 100 amino acids, a purely random conformational search would require around 10³⁶ s or around 10²⁹

years,[10] and yet proteins are able to fold on a timescale of milliseconds to seconds. This suggests that only a small amount of conformational space is sampled during the folding process and this in turn implies the existence of kinetic folding pathways, [11]. This paradox of how proteins fold rapidly and reliably to their native conformation is known as the protein folding problem.

The computational difficulty of protein folding is classified as an NP-complete problem. If a problem is NP-complete, it means that a particular solution can be checked in a polynomial time but to solve the whole problem requires an exponential time algorithm. A problem is in NP if it has a nondeterministic polynomial time solution. This means that the solution can be checked within polynomial time. As the exponential function in an NP-complete problem increases as much more rapid rate than a polynomial, these problems are untraceable.

5.6 Cheminformatics

5.6.1 Introduction

Cheminformatics (also known as **chemoinformatics** and **chemical informatics**) is the use of computer and informational techniques, applied to a range of problems in the field of chemistry. These in silico techniques are used in pharmaceutical companies in the process of drug discovery. These methods can also be used in chemical and allied industries in various other forms.

Chemoinformatics [12] is the mixing of those information resources to transform data into information and information into knowledge for the intended purpose of making better decisions faster in the area of drug lead identification and optimization. Cheminformatics combines the scientific working fields of chemistry and computer science for example in the area of chemical graph theory and mining the chemical space. It is to be expected that the chemical space contains at least 10^{60} molecules. Cheminformatics can also be applied to data analysis for various industries like paper and

pulp, dyes and such allied industries. Enzymes are a subset of receptor-like proteins that are directly responsible for catalyzing the biochemical reactions. DNA polymerase and related enzymes are crucial for cell division and replication. Enzymes are genetically programmed to be absolutely specific for their appropriate molecular targets.

The most important concept in drug design is to understand the methods by which the active site of a receptor selectively restricts the binding of inappropriate structures. Any potential molecule that can bind to a receptor is called a ligand. In order for a ligand to bind, it must contain a specific combination of atoms that presents the correct size, shape, and charge composition in order to bind and interact with the receptor **Figure 5.12** schematically shows a typical ligand-receptor binding interaction.

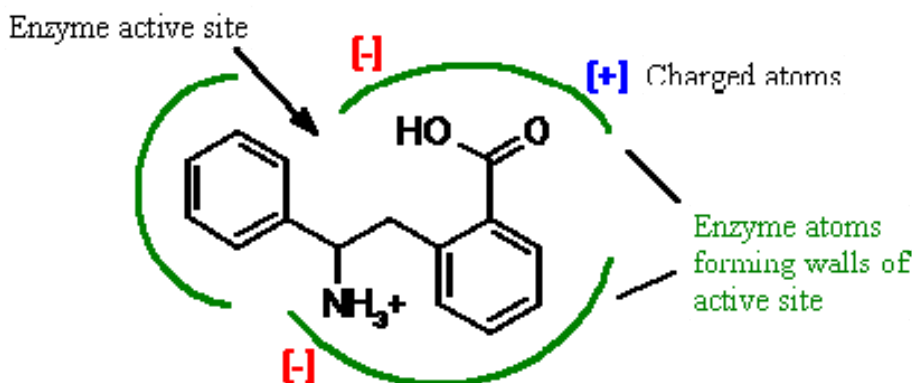


Figure 5.12: Enzyme substrate complementary interactions.

Ligand-receptor interaction must have complementary size and shape. This is termed steric complementarity. As is the case with an actual key, if a different molecule varies by even a single atom in the wrong place, it may not fit properly, and will most likely not interact with the receptor. However, the more closely the fit between the ligand and receptor, the more tightly the interaction becomes.

The main driving force for ligand and receptor binding is hydrophobic interaction. In order for ligand and receptor to interact, there must be a driving force that compels the ligand to leave the water and bind to the receptor. The hydrophobicity of a ligand is what causes this. Hydrophobicity stands for 'water fearing' and is a measure of how 'greasy' a

compound is. It can be roughly approximated by the percentage of hydrogen and carbon in the molecule. As shown in **Figure 5.13**, the active site may contain a mixture of hydrophobic pockets and regions that are more polar.

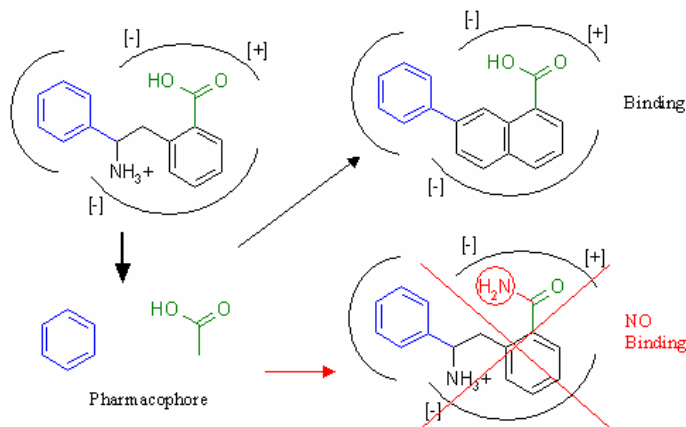


Figure 5.13: Pharmacophore and Receptor Binding

There are numerous potential interactions between ligand and receptor. Depending upon the size of the active site, there may be a numerous steric, electrostatic, and hydrophobic contact. The specific interactions that are crucial for ligand recognition and binding by the receptor are termed the pharmacophore. Usually, these are the interactions that directly factor into the structural integrity of a receptor or are involved in the mechanism of its action.

This is shown in **Figure 5.13** above. In the upper left frame of this figure, there is native ligand bound within the active site. Through biochemical investigation, phenyl ring (blue) and the carboxylic acid group (green) are vital to receptor interaction has been determined. Thus, it has been interpreted that these two groups must be the pharmacophore that a ligand must present to the receptor for binding.

5.6.2 The Challenge of Drug Design

There are difficulties in designing drugs towards specific target receptors.

Table 5.1. Major Tasks and concerns in Drug development [13].

1. Characterize medical condition and determine receptor targets.
2. Achieve active site complementarity: steric, electrostatic, and hydrophobic.
3. Consider biochemical mechanism of receptor.
4. Adhere to laws of chemistry.
5. Synthetic feasibility.
6. Biological considerations.
7. Patent considerations.

When a medical condition exists where a drug could be beneficial, extensive scientific study must first be done in order to determine the biological and biochemical problems that underlie the disease process.

Once a receptor target has been established and well characterized, the process of ligand design begins. The first consideration is that the designed ligand must complement the active site of the receptor target. Steric, electrostatic, and hydrophobic complementarity must be established. The pharmacophore must be presented to the receptor in order for recognition and binding to occur. Otherwise, the designed ligands have no chance of interacting with the receptor.

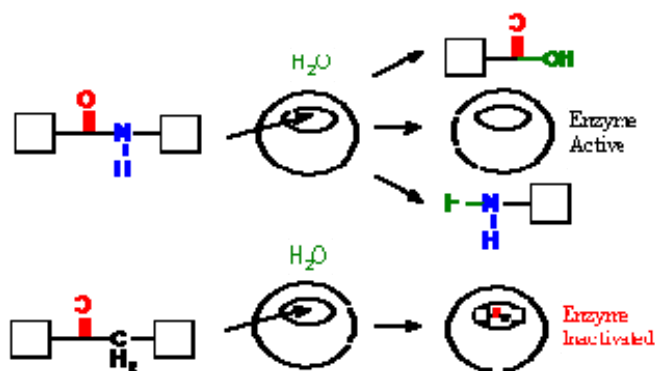


Figure 5.14: Designing ligands to offset enzyme mechanism.

In addition to adequately binding the receptor, the biochemical mechanism of the receptor target must be taken into consideration. This is shown in **Figure 5.14**. In this the biochemical mechanism of a protease has been schematically represented. A protease is an enzyme that cleaves proteins and peptides. In the top part of the figure, a specific group of atoms colored in red and blue, called a peptide bond has been recognized by protease. If the peptide bond is present at a specific position in the active site when the ligand binds, it is cleaved by the protease with the addition of water (H_2O) to form two separate fragments.

Having characterized the active site region and the mechanism of action of the target receptor, the challenge then becomes one of designing a suitable ligand. The optimal combination of atoms and functional groups to complement the receptor is often the natural ligand of the receptor.

There are biological considerations to the development of new drugs. The liver is the major organ of detoxification in the human body. Any drug that is taken undergoes a number of chemical reactions in the liver as the body attempts to neutralize foreign substances. Various chemical structures are highly toxic to biological systems, and these have been also well characterized.

5.6.3 The Drug Discovery Pipeline

The development [19] of any potential drug begins with years of scientific study to determine the biochemistry behind a medical problem for which pharmaceutical intervention is possible. The result is the determination of specific receptor targets that must be modulated to alter their activity in some way. Once these targets have been identified, the goal is then to find compounds that have to interact with the receptors.

The modern day drug discovery pipeline is outlined in **Figure 5.14**. The first step is to determine an estimate for the receptor. An estimation is a chemical or biological test that turns positive when a suitable binding agent interacts with the receptor. Usually, this test

is some form of colorimetric estimation, in which an indicator turns a specific color when complementary ligands are present. This estimate is then used in mass screening, which is a technique whereby hundreds of thousands of compounds can be tested in a matter of days to weeks. Entire corporate database of known compounds has been first screened by pharmaceutical company. The reason is that if a successful match is found, the database compound is usually very well characterized. Then after, a synthetic method has been known for this compound. This enables the company to rapidly prototype a candidate ligand.

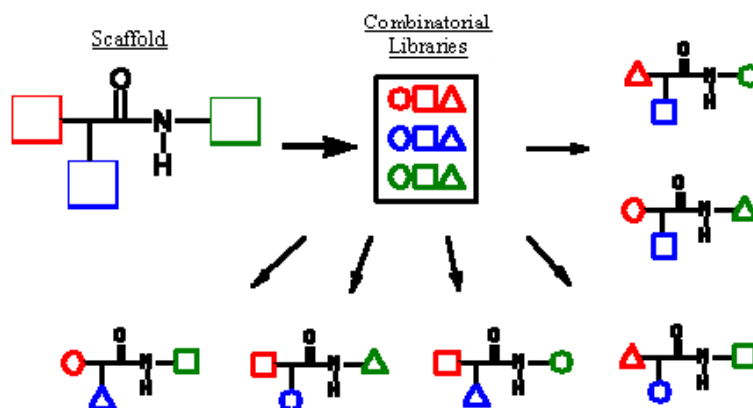


Figure 5.15: Combinatorial chemistry schematic.

Combinatorial chemistry is a very powerful technique that has been applied in the refinement of the lead compound. Combinatorial chemistry is a synthetic tool to rapidly generate thousands of lead compound derivatives for testing. As shown above in **Figure 5.15**, subsite groups (shown in red, green, and blue) are potential sites for derivatization. These subsites are then reacted with combinatorial libraries to generate a multitude of derivative structures, each with different substituent groups. By carefully selecting libraries based upon the study of the active site, the derivatization process towards optimizing ligand receptor interaction has been targeted.

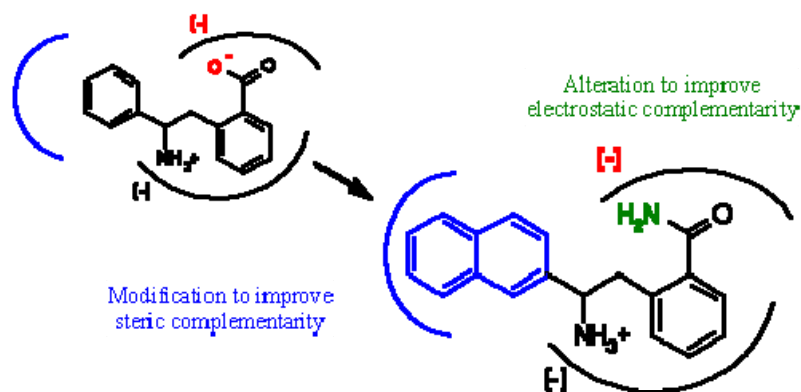


Figure 5.16: Structured Based Drug Design

Structure based design, often called rational drug design, and is much more focused than combinatorial chemistry. As shown above in **Figure 5.16**, it involves using the biochemical laws of ligand-receptor association discussed above to postulate ligand refinements to improve binding. Functional groups on the ligand can have been changed in order to expand electrostatic complementarity with the receptor. However, the danger in altering any portion of the ligand is the effect on the remaining ligand structures. Modifying even a single atom in the middle of the ligand can drastically change the shape of the overall structure. Even though complementarity in one portion of the ligand might be improved by the chemical revision, the overall binding might be severely compromised. This is the difficulty in any ligand refinement procedure.

5.6.4 Computer-Aided Drug Design (CADD)

Computer graphics technology has achieved the ability to generate vector models of chemical structures and manipulate them in real-time. The ability to study computer models of ligand structures and their binding interactions with a receptor has been offered by Computer graphics technology.

The time and effort required for drug synthesis and testing has been avoided by simply generating novel compounds using the computer. Testing has been replaced by calculating the ligand-receptor binding affinity using the physical laws of chemistry. The

concept of generating **virtual lead compounds** entirely through computer simulation has been termed as Denovo Design.

Computer-Aided Drug Design (CADD) [14] is a specialized discipline that uses computational methods to simulate drug-receptor interactions. CADD methods are heavily dependent on bioinformatics tools, applications and databases.

5.6.5 Difficulties Implementing Denovo Design

Although computers have become exponentially faster, the complete number of calculations needed to accurately predict the binding of a denovo generated ligand to its receptor in a useful timeframe still requires significant approximations. In denovo design, a whole ligand from scratch has been generated and it has been docked within the receptor. A ligand is flexible structure, and can guess an excess of different conformations and orientations. The predicted binding structure has to be similar with the calculated one. Failure in this attempt has damaged the utility of denovo structure generating software.

The second most significant problem in computer aided denovo design [15] is the generation of undesired chemical structures. There are a nearly infinite number of potential combinations of atoms. However, the vast majority of these structures are of no use. As discussed above, undesired structures are rejected due to toxicity, chemical instability, or synthetic difficulty. Nearly all denovo design software packages are overwhelmed by this problem, especially with respect to synthetic feasibility.

5.6.7 Benefits of CADD

CADD methods and bioinformatics tools offer significant benefits for drug discovery programs.

- **Cost Savings:** Many biopharmaceutical companies now use computational methods and bioinformatics tools to reduce this cost burden. Virtual screening, lead

optimization and predictions of bioavailability and bioactivity can help guide experimental research.

- **Time-to-Market:** The predictive power of CADD can help drug research programs choose only the most promising drug candidates. By focusing drug research on specific lead candidate's biopharmaceutical companies can get drugs to market more quickly.
- **Insight:** Molecular models of drug compounds can disclose atomic scale binding properties that are difficult to envision in any other way. Researcher's shows new molecular models to find out protein targets for new binding for new improved compound.

5.7 References

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Chapter -6

Conformational Study of
Molecules using Tools

Chapter 6
CONFORMATIONAL STUDY OF MOLECULE
USING TOOLS

6.1 Introduction

Under this research work various chemical and biochemical compounds have been analyzed including drugs using tools like ACD/ChemSketch, NMR Prediction and ArgusLab.

Under the research work of **Activity No.-1** molecules have been analyzed using tool like ACD/ChemSketch and NMR Prediction. In this research using ACD/ChemSketch compounds are stored in databases and SMILE code (Simplified Molecular Input Line Specification) is generated. A SMILE defines the molecules in the form of alphanumeric chains. In this research work chemical shift of every carbon atom of the molecule have been displayed by using NMR Prediction. Using Pubchem/NCBI additional miscellaneous information such as bioactivity analysis by structure & activity similarity and revised compound selection after addition of similar compounds have been found out.

Under the research work of **Activity No.-2** geometry of molecules have been optimized, chemical structure has been visualized and electronic absorption spectra of chemical structure has been calculated by using ArgusLab tool.

Under the research work of **Activity No.-3** different types of analysis like prediction of protein secondary structure, isoelectric point calculation etc. have been performed on nucleotide Sequence and protein sequence using DAMBE and Jemboos tools.

6.2 Experimental Work

6.2.1 Activity No -1

In this research work Alanine (Amino acid) has been used on ACD/ChemSketch editor and its SMILE code has been generated. After that this structured has been transferred to NMRPrediction editor. Similar activity has been performed with Amino butyric acid, Asparagine and Glutamine.

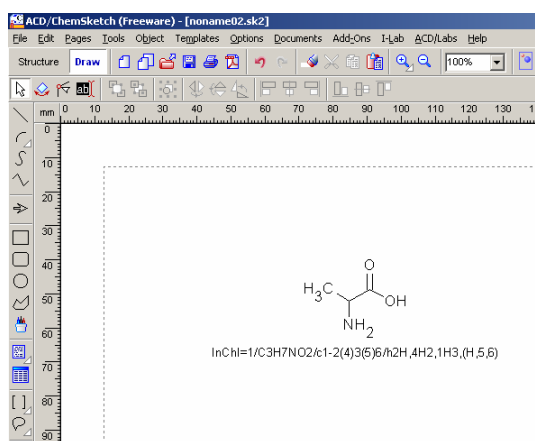


Figure 6.1: Alanine on ACD/ChemSketch editor

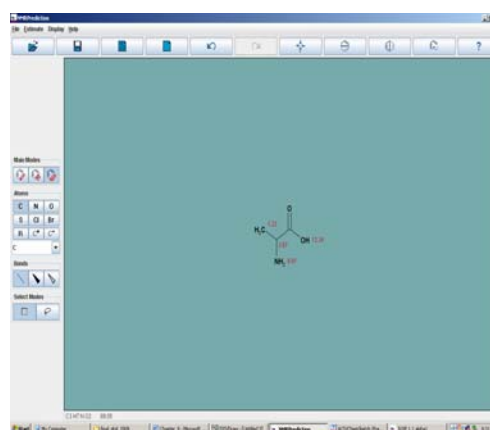


Figure 6.2: ^{13}C NMR of of Alanine

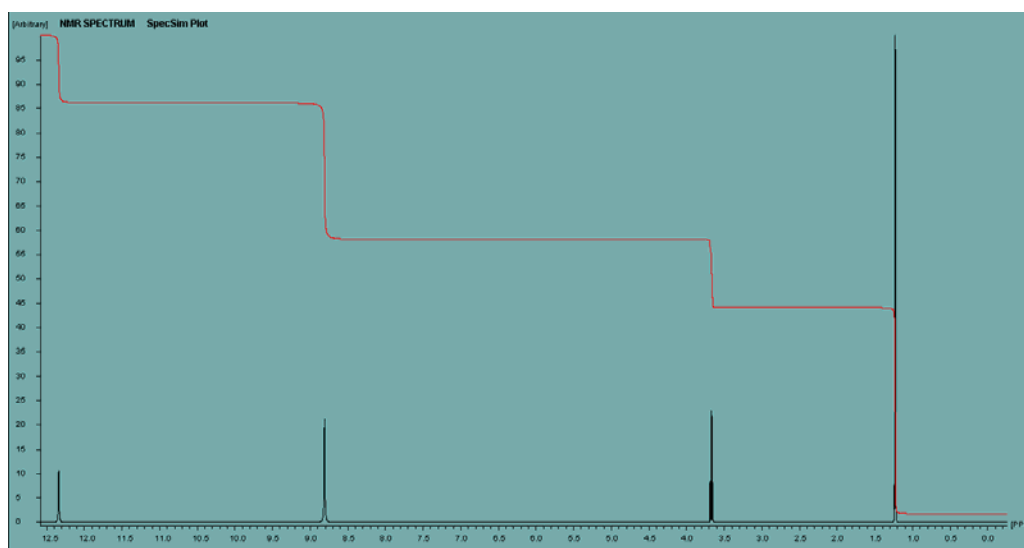


Figure 6.3: Estimation of ^1H NMR of Alanine

Table 6.1: Shift Prediction Protocol of Alanine using NMRPrediction

Node	Shift	Base + Inc.	Comment (ppm rel. to TMS)
C	174.7	166.0	1-carboxyl
		11.0	1 -C-C
		-2.2	general corrections
CH	51.5	-2.3	aliphatic
		21.8	1 alpha -C(=O)-O
		9.1	1 alpha -C
		28.3	1 alpha -N
		-5.4	general corrections
CH3	19.6	-2.3	aliphatic
		9.1	1 alpha -C
		2.0	1 beta -C(=O)-O
		11.3	1 beta -N
		-0.5	general corrections

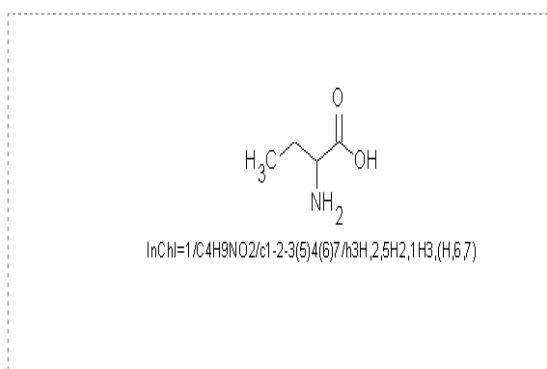


Figure 6.4: 2-aminobutanoic acid on ACDCHEM editor

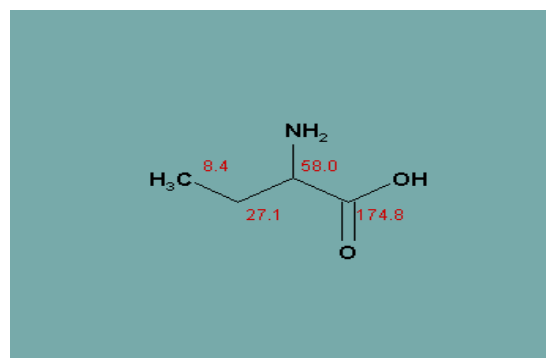


Figure 6.5: ^{13}C NMR of Aminobutanoic acid

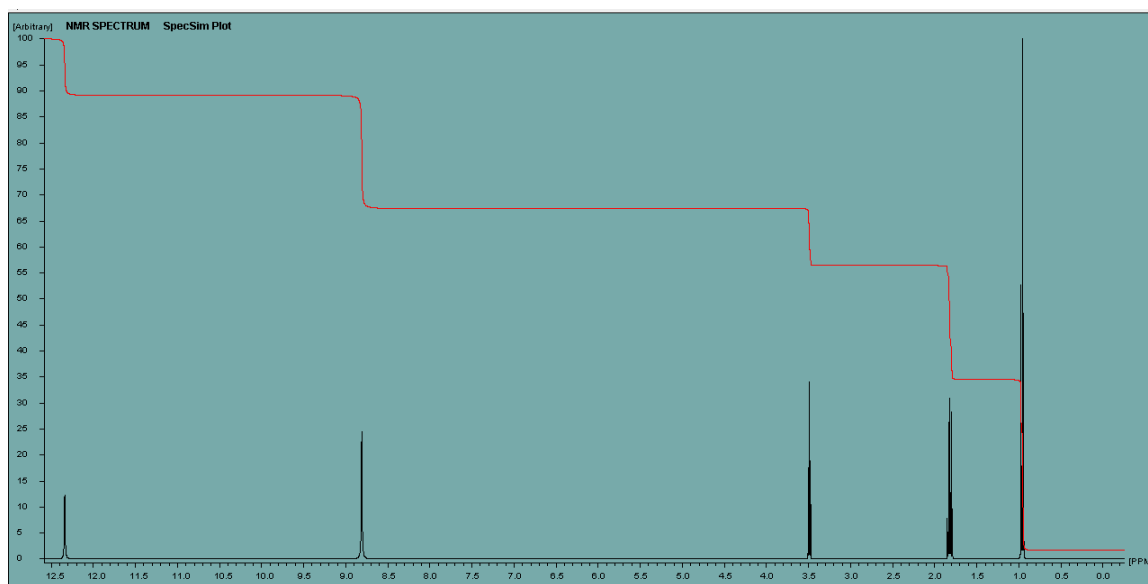


Figure 6.6: Estimation of ^1H NMR of Aminobutanoic acid

Table 6.2: Shift Prediction Protocol of 2-aminobutanoic acid using NMRPrediction

Node	Shift	Base + Inc.	Comment (ppm rel. to TMS)
CH	3.49	1.50	methine
		1.13	1 alpha -N
		0.87	1 alpha -C(=O)O
		-0.01	1 beta -C
NH2	8.81	2.00	amine
		6.81	general corrections
CH2	1.82	1.37	methylene
		0.00	1 alpha -C
		0.22	1 beta -N
		0.23	1 beta -C(=O)O
OH	12.34	11.00	carboxylic acid
		1.34	general corrections
CH3	0.96	0.86	methyl
		0.10	1 beta -C-R

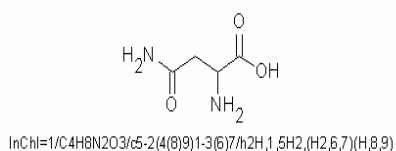


Figure 6.7: Asparagine on ACD/ChemSketch editor

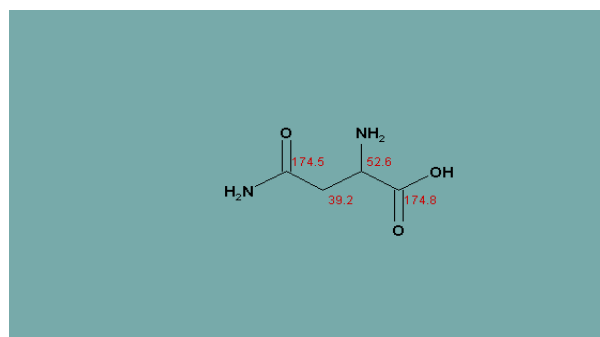


Figure 6.8: ^{13}C NMR of Asparagine

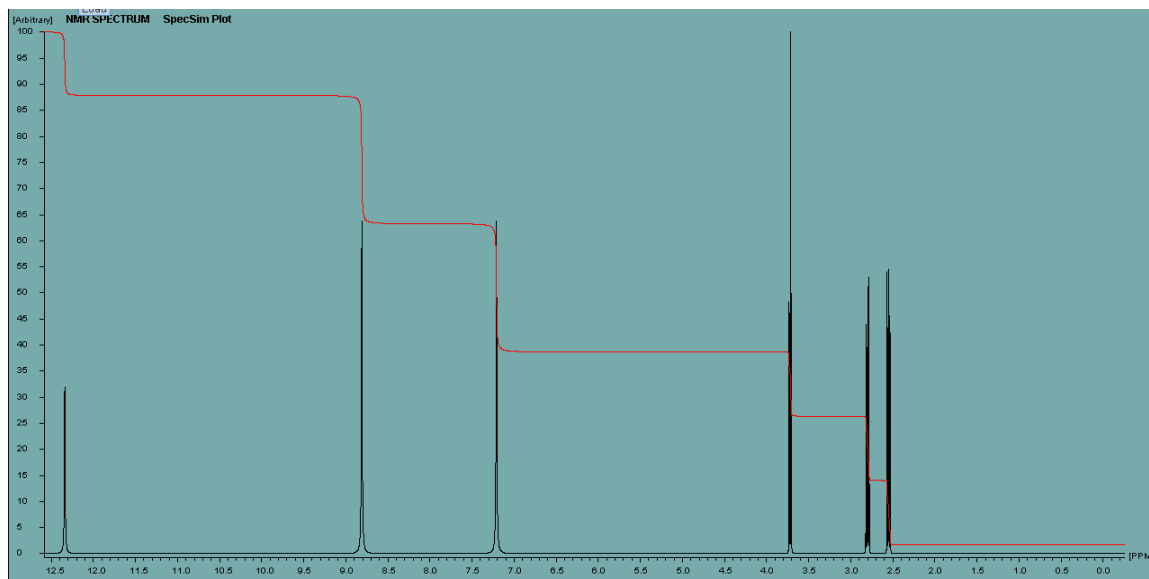


Figure 6.9: Estimation of ^1H NMR of Asparagine

Table 6.3: Shift Prediction Protocol of Asparagine using NMRPrediction

Node	Shift	Base + Inc.	Comment (ppm rel. to TMS)
CH ₂	2.80, 2.55	1.37	methylene
		0.85	1 alpha -C(=O)N
		0.22	1 beta -N
		0.23	1 beta -C(=O)O
		1.50	methine
CH	3.72	1.13	1 alpha -N
		0.87	1 alpha -C(=O)O
		0.22	1 beta -C=O
		6.00	prim. amide
NH ₂	7.21	1.21	general corrections
NH ₂	8.81	2.00	amine
		6.81	general corrections
OH	12.34	11.00	carboxylic acid
		1.34	general corrections

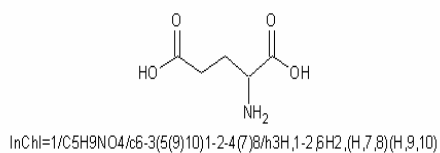


Figure 6.10: Glutamine on ACD/ChemSketch editor

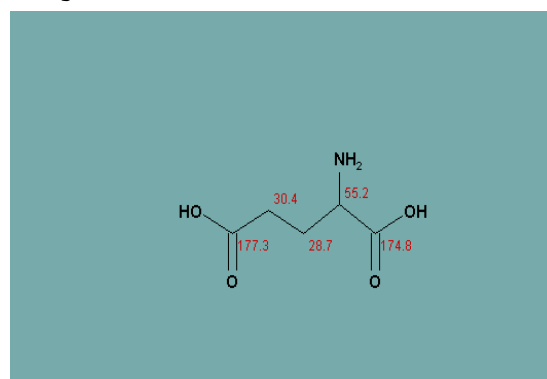


Figure 6.11: ¹³C NMR of Glutamine

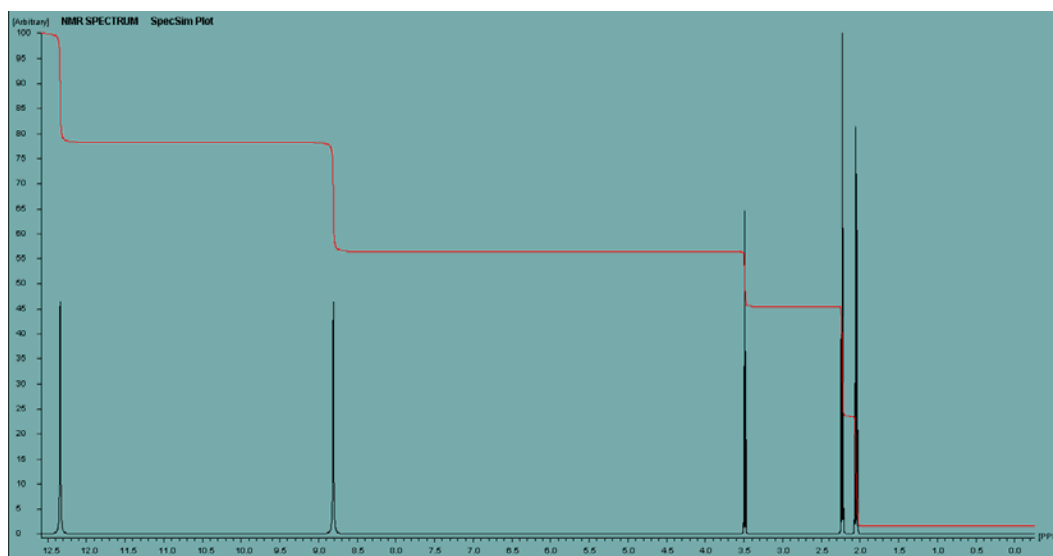
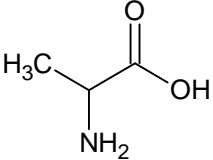
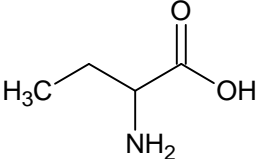
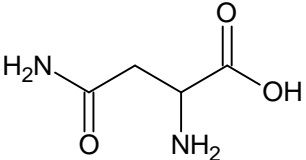
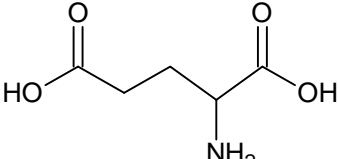


Figure 6.12: Estimation of ¹H NMR of Glutamine

Table 6.4: Shift Prediction Protocol of Glutamine using NMRPrediction

Node	Shift	Base + Inc.	Comment (ppm rel. to TMS)
CH2	2.23	1.37 0.90 -0.04	methylene 1 alpha -C(=O)O 1 beta -C
CH2	2.05	1.37 0.23 0.22	methylene 1 beta -C(=O)O 1 beta -N
CH	3.49	0.23 1.50 1.13 0.87	1 beta -C(=O)O methine 1 alpha -N 1 alpha -C(=O)O
OH	12.34	-0.01 11.00	1 beta -C carboxylic acid
OH	12.34	1.34 11.00	general corrections carboxylic acid
NH2	8.81	1.34 2.00 6.81	general corrections amine general corrections

Table 6.5: SMILE Code of various structures

Structure	SMILE Code
 <p>Alanine</p>	<chem>CC(N)C(O)=O</chem>
 <p>Amino butyric Acid</p>	<chem>CCC(N)C(O)=O</chem>
 <p>Asparagine</p>	<chem>NC(CC(N)=O)C(O)=O</chem>
 <p>Glutamine</p>	<chem>NC(CCC(O)=O)C(O)=O</chem>

After that web based structure search queries have been performed on these compounds using Pubchem/NCBI. Here activities like Bioactivity Analysis by Structure & Activity Similarity of molecule , Bioactivity Analysis by Structure & Activity Similarity of molecule from Normalized score Percentile , Bioactivity Analysis by Activity & protein target Similarity of molecule from Normalized score Percentile , Bioactivity Analysis by concise Data Table of molecule , Bioactivity Analysis by addition of similar compounds of molecule and Revised compound selection after addition of similar compounds of molecule have been performed.

The screenshot displays the PubChem search results for Alanine (CID: 602). The interface includes the NCBI and PubChem logos, a search bar with the text 'PubChem Compound', and navigation options like 'Limits', 'Preview/Index', 'History', 'Clipboard', and 'Details'. The search results are displayed in a 'Summary' view, showing the compound name 'alanine; DL-ALANINE; D-alanine ...', its IUPAC name '2-aminopropanoic acid', molecular weight (MW: 89.093180 g/mol), and molecular formula (MF: C₃H₇NO₂). The bioactivity data indicates 'Tested in BioAssays: All: 0, Active: 0'. A 'Recent Activity' box on the right shows 'Your browsing activity is empty.' The page also features a 'My NCBI' section with 'Sign In' and 'Register' links, and a 'Tools' section with 'Related Structures', 'BioAssays', 'Literature', and 'Other Links'.

Figure 6.13: Search query using Pubchem/NCBI of Alanine

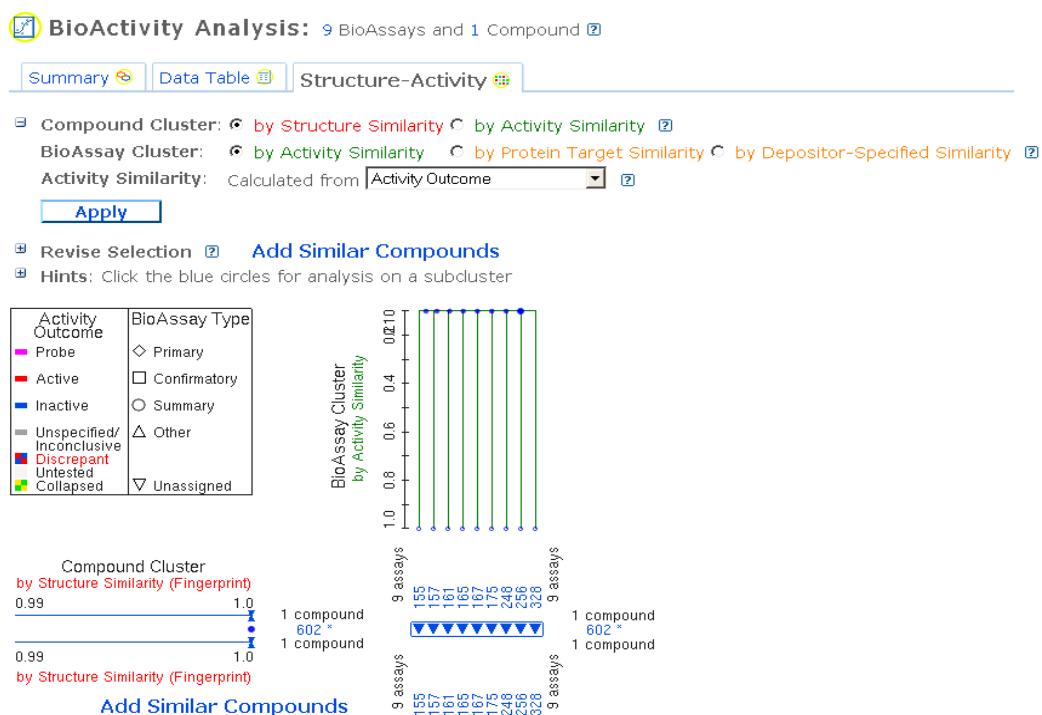


Figure 6.14: Bioactivity Analysis by Structure & Activity Similarity of Aniline

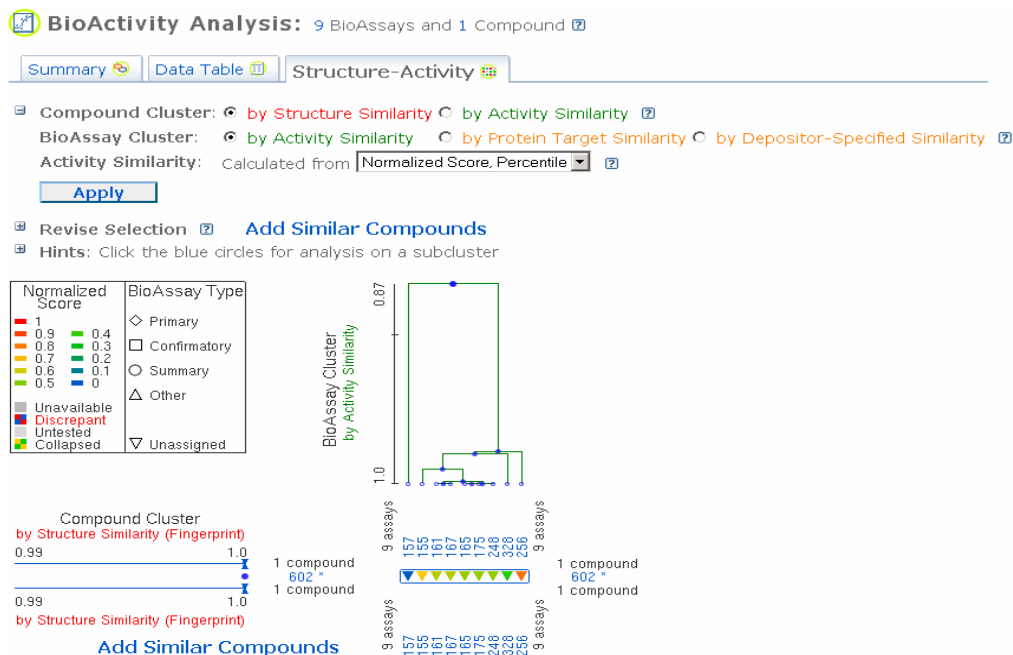


Figure 6.15: Bioactivity Analysis by Structure & Activity Similarity of Aniline from Normalized score Percentile

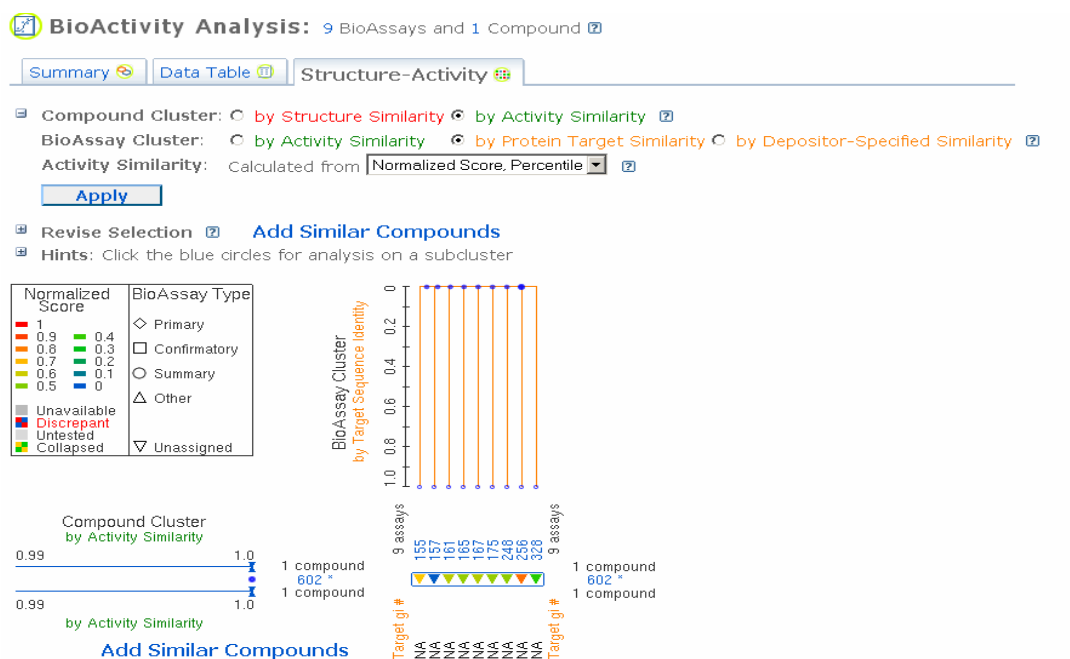


Figure 6.16: Bioactivity Analysis by Activity & protein target Similarity of Aniline from Normalized score Percentile

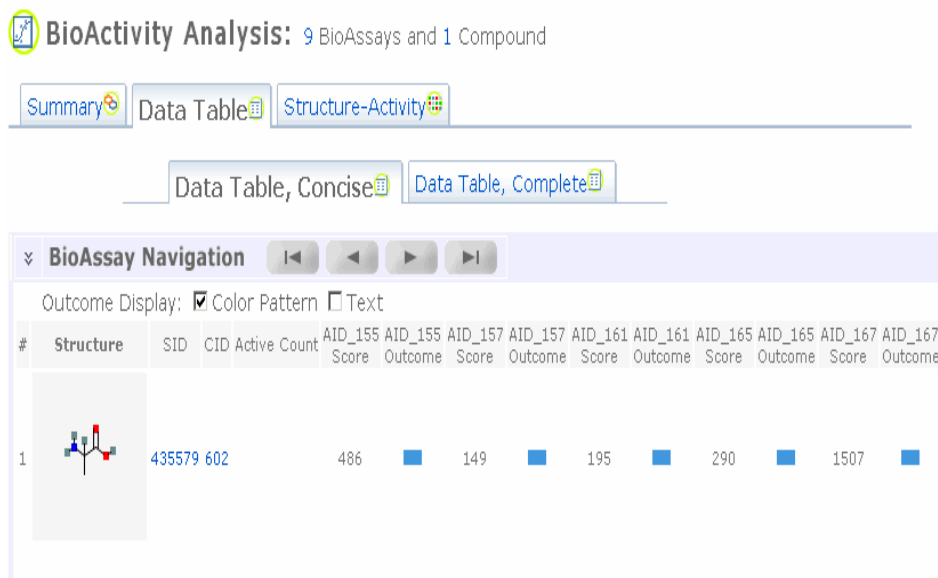


Figure 6.17: Bioactivity Analysis by concise Data Table of Aniline

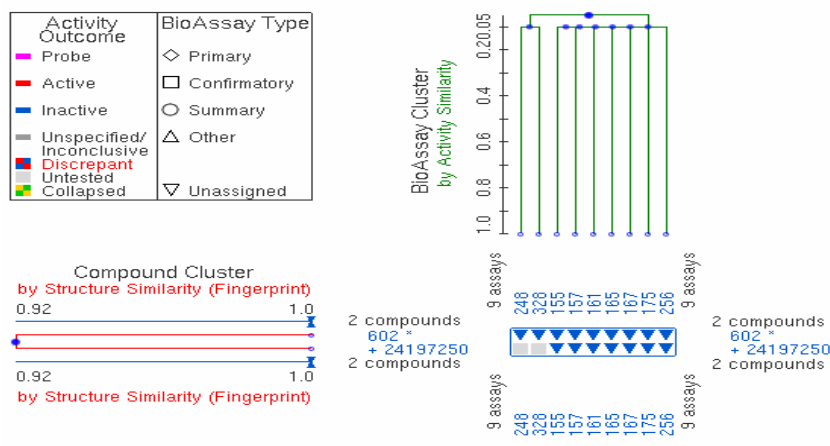


Figure 6.18: Bioactivity Analysis by addition of similar compounds of Aniline

Table 6.6: 9- assays by addition of similar compound of Aniline

CID/ Activity /AID	248	328	155	157	161	165	167	175	256
602	Inact	Inact	Inact	Inact	Inact	Inact	Inact	Inact	Inact
24197250			Inact	Inact	Inact	Inact	Inact	Inact	Inact

Inact = Inactive

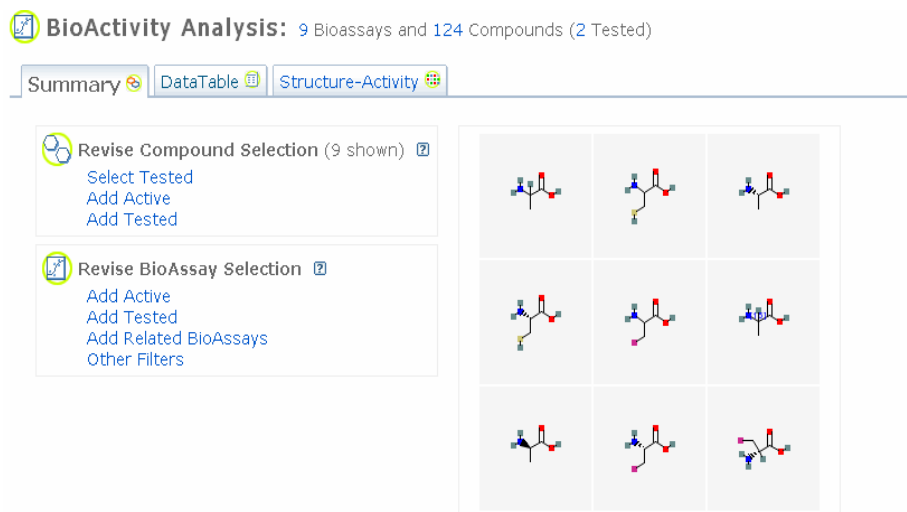


Figure 6.19: Revised compound selection after addition of similar compounds of Aniline

Tools: Links: [Related Structures](#), [BioAssays](#), [Literature](#), [Other Links](#)

All: 1 | BioAssay: 1 | Protein3D: 1 | Rule of 5: 1

1: CID: 6657 [Related Structures, BioAssays, Literature, Other Links](#)

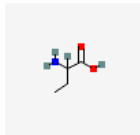
 Butyric; DL-2-Aminobutyric acid; 2-Aminobutyric acid ...
 IUPAC: 2-aminobutanoic acid
 MW: 103.119760 g/mol | MF: C₄H₉NO₂
 Tested in BioAssays: All: 7, Active: 0; [BioActivity Analysis](#)

Figure 6.20: Search query using Pubchem/NCBI of Amino butyric Acid

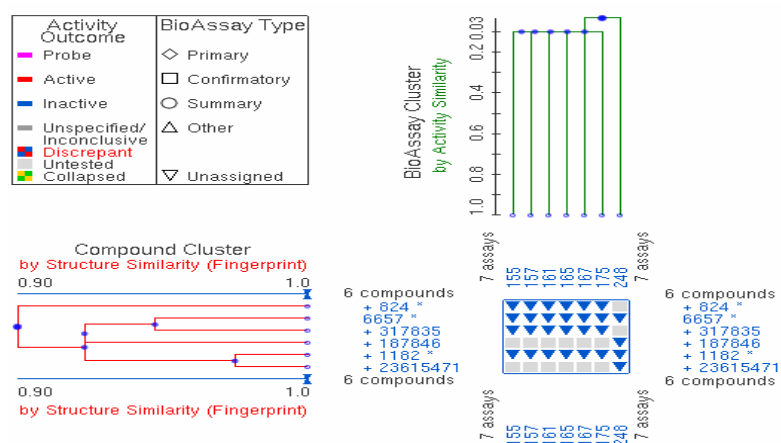


Figure 6.21: Bioactivity Analysis by addition of similar compounds of Amino butyric Acid

Table 6.7: 7- assays by addition of similar compound Amino butyric Acid

CID/Activity/AID	155	157	161	165	167	175	248
824	Inact	Inact	Inact	Inact	Inact	Inact	
6657	Inact	Inact	Inact	Inact	Inact	Inact	Inact
317835	Inact	Inact	Inact	Inact	Inact	Inact	
187846							Inact
1182	Inact	Inact	Inact	Inact	Inact	Inact	Inact
23615471							Inact

Inact = Inactive

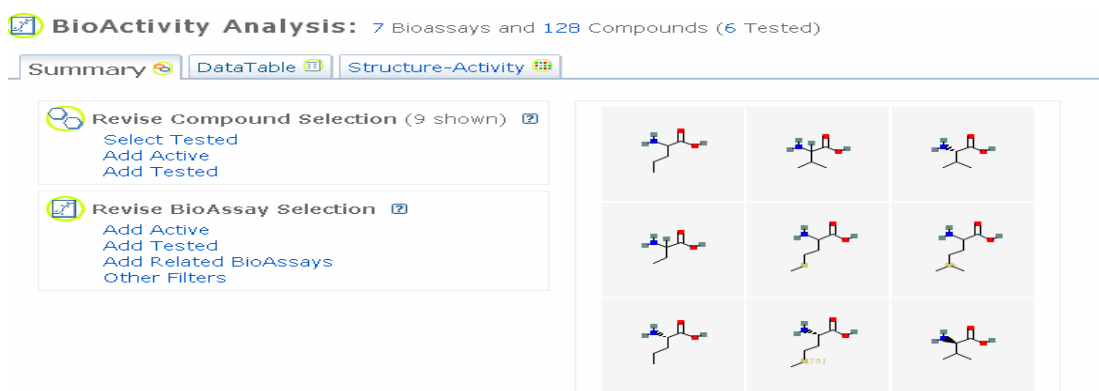


Figure 6.22: Revised compound selection after addition of similar compounds of Amino butyric Acid

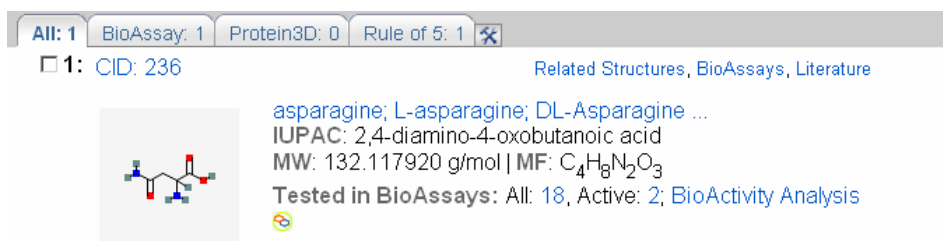


Figure 6.23: Search query using Pubchem/NCBI of Asparagine

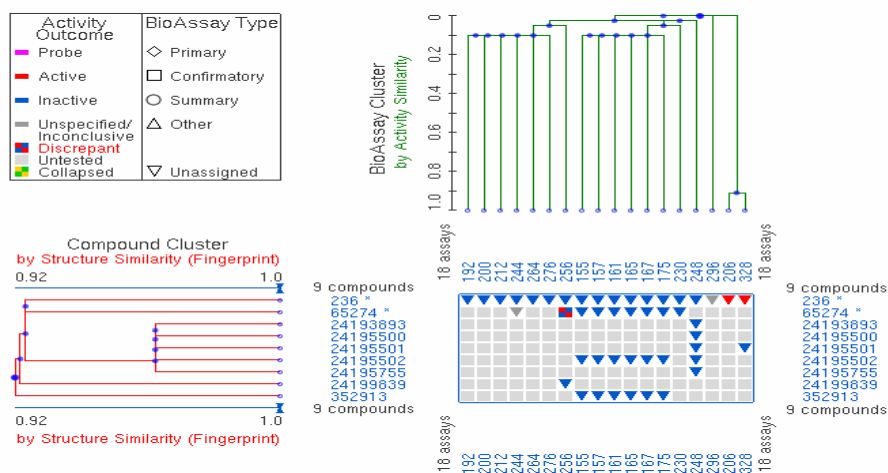


Figure 6.24: Bioactivity Analysis by addition of similar compounds of Asparagine

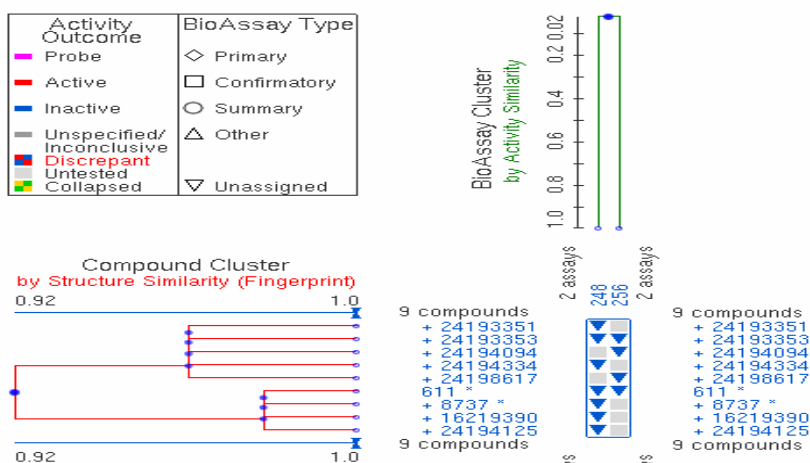


Figure 6.27: Bioactivity Analysis by addition of similar compounds of Glutamine

Table 6.9: 2- assays by addition of similar compound Glutamine

CID/Activity/AID	248	256
24193351	Inactive	
24193353	Inactive	Inactive
24194094		Inactive
24194334	Inactive	
24198617		Inactive
611	Inactive	Inactive
8737	Inactive	
16219390	Inactive	
24194125	Inactive	

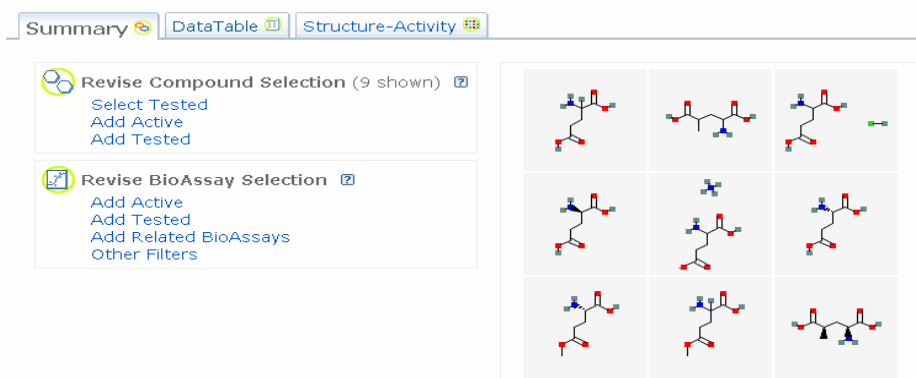


Figure 6.28: Revised compound selection after addition of similar compounds of Glutamine

6.2.2 Activity No -2

In this research work Alanine, Amino butyric acid, Asparagine and Glutamine have been analyzed on ArgusLab tool and calculations like Single Entry Point Calculation, Geometric Optimization and UV Electronic Spectra have been performed.

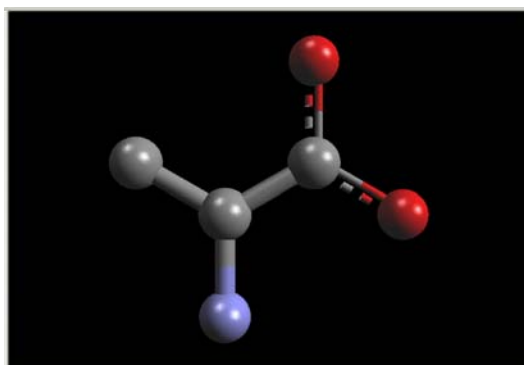


Figure No 6.29 : 3D-Molecular structure of Alanine on ArgusLab

Table No 6.10: Single Entry Point Calculation of Alanine Using ArgusLab

***** Validated Experiment & Chemical System Settings *****

Calculation started: Thu Jun 04 10:37:09 2009

Title:E:\myphdthesis_2009\ACd_results\Ala

Max. SCF cycles 100
SCF convergence 1.5936e-013 au. for energy

Input Atomic Information

1	C	16.904500	-7.535300	0.000000
2	O	16.904500	-6.205300	0.000000
3	C	15.752700	-8.200300	0.000000
4	N	15.752700	-9.530300	0.000000
5	C	14.600800	-7.535300	0.000000
6	O	18.056300	-8.200300	0.000000

Constructing Chemical System(s)

Basis Set

basis functions : 24
shells : 12
primitives : 72

Memory for Main Chemical System
Max. number 2-ele. ints. = 1596

Chapter 6: Conformational Study of Substance through Different Tools

Memory Requirements (bytes)

Core 121248
Scratch 10368
System charge 0.000000

***** SCF *****

Core repulsion 79.3611 au

Calculating one electron matrix
Diagonalizing starting one-ele. matrix
Performing SCF

Cycle	Energy (au)	Difference
1	-36.971063	
2	-32.276391119	4.69467
3	-33.620242051	-1.34385
4	-26.283142152	7.3371
5	-33.505498559	-7.22236
6	-26.230341192	7.27516
7	-33.507013214	-7.27667
8	-26.229486478	7.27753
9	-33.508071400	-7.27858
10	-26.229663085	7.27841
11	-33.508488479	-7.27883
12	-26.229742535	7.27875
13	-33.508637619	-7.2789
14	-26.229770710	7.27887
15	-33.508689823	-7.27892
16	-26.229780460	7.27891
17	-33.508708012	-7.27893
18	-26.229783833	7.27892
19	-33.508714344	-7.27893
20	-26.229785002	7.27893
21	-33.508716547	-7.27893
22	-26.229785408	7.27893
23	-33.508717315	-7.27893
24	-26.229785549	7.27893
25	-33.508717582	-7.27893
26	-26.229785599	7.27893
27	-33.508717675	-7.27893
28	-26.229785616	7.27893
29	-33.508717707	-7.27893
30	-26.229785622	7.27893
31	-33.508717718	-7.27893
32	-26.229785624	7.27893
33	-33.508717722	-7.27893
34	-26.229785624	7.27893
35	-33.508717723	-7.27893
36	-26.229785625	7.27893
37	-33.508717724	-7.27893
38	-26.229785625	7.27893
39	-33.508717724	-7.27893
40	-26.229785625	7.27893
41	-33.508717724	-7.27893
42	-26.229785625	7.27893
43	-33.508717724	-7.27893
44	-26.229785625	7.27893
45	-33.508717724	-7.27893
46	-26.229785625	7.27893
47	-33.508717724	-7.27893
48	-26.229785625	7.27893
49	-33.508717724	-7.27893
50	-26.229785625	7.27893
51	-33.508717724	-7.27893
52	-26.229785625	7.27893
53	-33.508717724	-7.27893
54	-26.229785625	7.27893
55	-33.508717724	-7.27893
56	-26.229785625	7.27893
57	-33.508717724	-7.27893
58	-26.229785625	7.27893
59	-33.508717724	-7.27893
60	-26.229785625	7.27893
61	-33.508717724	-7.27893

62	-26.229785625	7.27893
63	-33.508717724	-7.27893
64	-26.229785625	7.27893
65	-33.508717724	-7.27893
66	-26.229785625	7.27893
67	-33.508717724	-7.27893
68	-26.229785625	7.27893
69	-33.508717724	-7.27893
70	-26.229785625	7.27893
71	-33.508717724	-7.27893
72	-26.229785625	7.27893
73	-33.508717724	-7.27893
74	-26.229785625	7.27893
75	-33.508717724	-7.27893
76	-26.229785625	7.27893
77	-33.508717724	-7.27893
78	-26.229785625	7.27893
79	-33.508717724	-7.27893
80	-26.229785625	7.27893
81	-33.508717724	-7.27893
82	-26.229785625	7.27893
83	-33.508717724	-7.27893
84	-26.229785625	7.27893
85	-33.508717724	-7.27893
86	-26.229785625	7.27893
87	-33.508717724	-7.27893
88	-26.229785625	7.27893
89	-33.508717724	-7.27893
90	-26.229785625	7.27893
91	-33.508717724	-7.27893
92	-26.229785625	7.27893
93	-33.508717724	-7.27893
94	-26.229785625	7.27893
95	-33.508717724	-7.27893
96	-26.229785625	7.27893
97	-33.508717724	-7.27893
98	-26.229785625	7.27893
99	-33.508717724	-7.27893
100	-26.229785625	7.27893

Maximum number of iterations reached: SCF NOT CONVERGED!

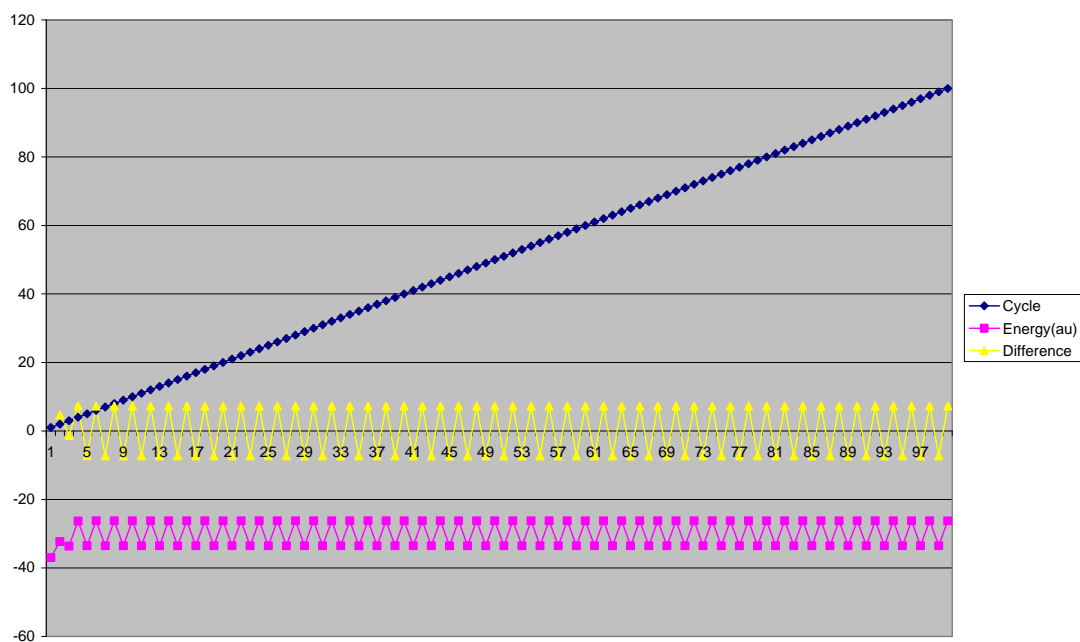


Diagram 6.30: Energy vs. Difference S.E.C of Alanine

Writing final SCF to disk

Final SCF Energy = -26.2297856247 au

Chapter 6: Conformational Study of Substance through Different Tools

```
Final SCF Energy = -16459.4538 kcal/mol

**** Heat of Formation ****
11889.4613 kcal/mol

Wiberg Atom-Atom Bond Orders
*****

      1      2      3      4      5      6
1  0.000000
2  0.331606  0.000000
3  0.241269  0.010621  0.000000
4  0.000639  0.000104  0.000185  0.000000
5  0.003243  0.000301  0.001393  0.000002  0.000000
6  0.327085  0.009206  0.009767  0.000086  0.000241  0.000000

Atomic spin densities
*****
      1  C  0.6170
      2  O  0.0225
      3  C  0.3284
      4  N  0.0005
      5  C  0.0044
      6  O  0.0273
S2 operator
*****
exact          0.750000
calculated     0.750201

Ground State Dipole (debye)
      X      Y      Z      length
106.85663544  55.51461066  -0.00000000  120.41682828

Mulliken Atomic Charges
*****
      1  C  2.6734
      2  O  4.0964
      3  C -3.8542
      4  N -3.0018
      5  C -3.9958
      6  O  4.0820

Properties elapsed time 0 sec.

Total Elapsed Time 0 sec.
```

Table No 6.11: Geometry Optimization of Analine using ArgusLab.

```
***** Validated Experiment & Chemical System Settings *****

Calculation started: Thu Jun 04 11:06:01 2009

Title:E:\myphdthesis_2009\ACd_results\Ala

Max. SCF cycles          100
SCF convergence          1.5936e-013 au. for energy

Max. geom cycles         10
Convergence criteria:
max. grad. component < 0.000084 au.
```

Chapter 6: Conformational Study of Substance through Different Tools

AM1 param file C:\Program Files\ArgusLab\params\am1.prm
SCF saved every 1000 cycles

Input Atomic Information

1	C	16.435771	-7.307230	0.037217
2	O	17.070817	-6.090942	-0.014829
3	C	15.266201	-8.194435	-0.019871
4	N	16.383638	-9.920985	0.002328
5	C	14.143594	-7.191811	-0.001207
6	O	18.671478	-8.501396	-0.003638

Constructing Chemical System(s)

Basis Set

basis functions : 24
shells : 12
primitives : 36

Memory for Main Chemical System
Max. number 2-ele. ints. = 1596

Memory Requirements (bytes)

Core 325080
Scratch 10368
System charge 0.000000

***** SCF *****

Core repulsion 66.1182 au

Calculating one electron matrix
Diagonalizing starting one-ele. matrix
Performing SCF

Cycle	Energy (au)	Difference
1	-36.702730	
2	-31.180576166	5.52215
3	-34.977490714	-3.79691
4	-26.267268379	8.71022
5	-36.048536714	-9.78127
6	-27.952390982	8.09615
7	-36.298981727	-8.34659
8	-27.939103265	8.35988
9	-35.978322254	-8.03922
10	-27.902782233	8.07554
11	-35.690378260	-7.7876
12	-27.894461588	7.79592
13	-35.629035870	-7.73457
14	-27.895280755	7.73376
15	-35.613901033	-7.71862
16	-27.896032136	7.71787
17	-35.609077364	-7.71305
18	-27.896355285	7.71272
19	-35.607344655	-7.71099
20	-27.896481549	7.71086
21	-35.606693537	-7.71021
22	-27.896530159	7.71016
23	-35.606444862	-7.70991
24	-27.896548859	7.7099
25	-35.606349333	-7.7098
26	-27.896556058	7.70979
27	-35.606312557	-7.70976
28	-27.896558832	7.70975
29	-35.606298388	-7.70974
30	-27.896559901	7.70974
31	-35.606292928	-7.70973

Chapter 6: Conformational Study of Substance through Different Tools

32	-27.896560313	7.70973
33	-35.606290823	-7.70973
34	-27.896560472	7.70973
35	-35.606290012	-7.70973
36	-27.896560533	7.70973
37	-35.606289699	-7.70973
38	-27.896560557	7.70973
39	-35.606289579	-7.70973
40	-27.896560566	7.70973
41	-35.606289532	-7.70973
42	-27.896560569	7.70973
43	-35.606289514	-7.70973
44	-27.896560571	7.70973
45	-35.606289508	-7.70973
46	-27.896560571	7.70973
47	-35.606289505	-7.70973
48	-27.896560571	7.70973
49	-35.606289504	-7.70973
50	-27.896560571	7.70973
51	-35.606289503	-7.70973
52	-27.896560572	7.70973
53	-35.606289503	-7.70973
54	-27.896560572	7.70973
55	-35.606289503	-7.70973
56	-27.896560572	7.70973
57	-35.606289503	-7.70973
58	-27.896560572	7.70973
59	-35.606289503	-7.70973
60	-27.896560572	7.70973
61	-35.606289503	-7.70973
62	-27.896560572	7.70973
63	-35.606289503	-7.70973
64	-27.896560572	7.70973
65	-35.606289503	-7.70973
66	-27.896560572	7.70973
67	-35.606289503	-7.70973
68	-27.896560572	7.70973
69	-35.606289503	-7.70973
70	-27.896560572	7.70973
71	-35.606289503	-7.70973
72	-27.896560572	7.70973
73	-35.606289503	-7.70973
74	-27.896560572	7.70973
75	-35.606289503	-7.70973
76	-27.896560572	7.70973
77	-35.606289503	-7.70973
78	-27.896560572	7.70973
79	-35.606289503	-7.70973
80	-27.896560572	7.70973
81	-35.606289503	-7.70973
82	-27.896560572	7.70973
83	-35.606289503	-7.70973
84	-27.896560572	7.70973
85	-35.606289503	-7.70973
86	-27.896560572	7.70973
87	-35.606289503	-7.70973
88	-27.896560572	7.70973
89	-35.606289503	-7.70973
90	-27.896560572	7.70973
91	-35.606289503	-7.70973
92	-27.896560572	7.70973
93	-35.606289503	-7.70973
94	-27.896560572	7.70973
95	-35.606289503	-7.70973
96	-27.896560572	7.70973
97	-35.606289503	-7.70973
98	-27.896560572	7.70973
99	-35.606289503	-7.70973
100	-27.896560572	7.70973

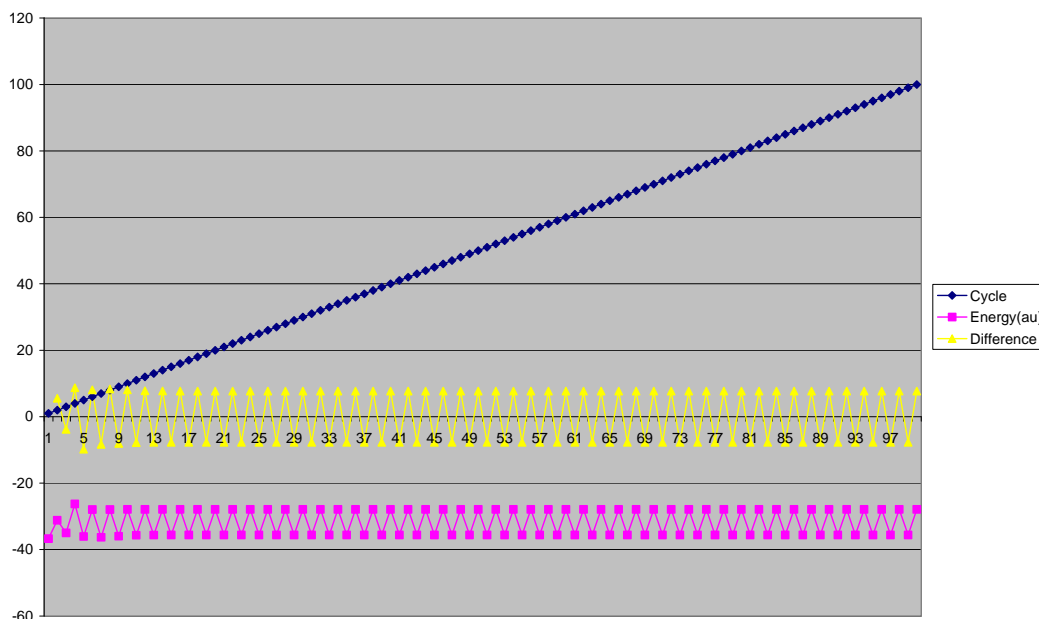


Figure 6.31: Energy vs. Difference of geometric optimization of Aniline

Maximum number of iterations reached: SCF NOT CONVERGED!

Writing final SCF to disk

Final SCF Energy = -27.8965605715 au
 Final SCF Energy = -17505.3718 kcal/mol

Saving the final SCF to the restart file
 E:\myphdthesis_2009\ACd_results\Ala.restartscf

SCF elapsed time 1 sec.

Table 6.12: Geometric for different components of Aniline

***** Geometry Optimization *****

Checkpointing coordinate to E:\myphdthesis_2009\ACd_results\Ala.cor

Geometry Search using BFGS update

Cycle	Energy(au)	delE (au)	Grad Norm	Max Grad(i)	alpha
start	-27.896561	0.0000e+000	2.662552	1.270036	
1	-36.561033	-8.6645e+000	1.698515	1.124240	1.5023e-001
2	-37.477903	-9.1687e-001	1.227344	0.768170	1.1775e-001
3	-29.966887	7.5110e+000	3.034984	1.893725	2.5710e-031
4	-36.328840	-6.3620e+000	1.450727	0.862822	1.3180e-001
5	-29.337689	6.9912e+000	1.962616	1.270998	2.1751e-031
6	-34.996943	-5.6593e+000	1.393716	0.687540	2.0381e-001
7	-35.809607	-8.1266e-001	1.285275	0.586843	1.4350e-001
8	-30.962245	4.8474e+000	2.617189	1.509296	2.4551e-031
9	-37.516635	-6.5544e+000	1.251350	0.814439	1.5284e-001
10	-30.178555	7.3381e+000	1.996299	1.450952	2.5216e-031

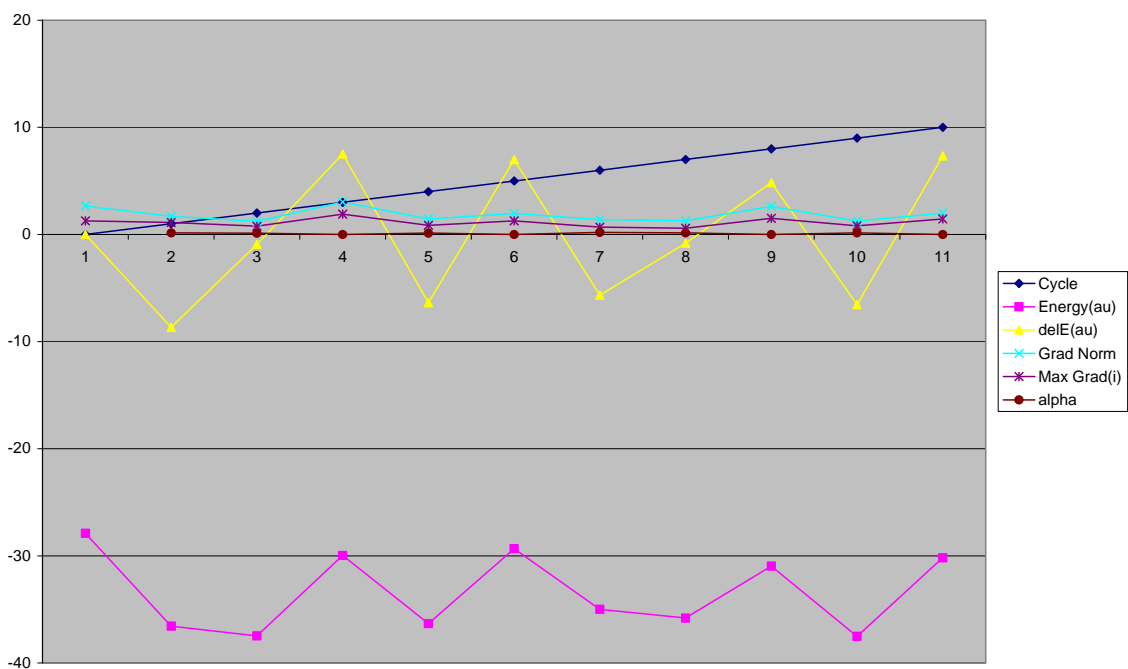


Figure 6.32: Geometric for different components of Aniline

>>>Geometry optimization did not converge<<<

Maximum cycles reached, optimization terminated

***** Final Geometry *****

C	16.58012803	-7.44007254	0.04973112	6
O	16.95387408	-6.07861932	-0.02119125	8
C	15.45848448	-8.29244282	-0.02981828	6
N	16.27002874	-9.85642611	0.00468287	7
C	14.07596372	-7.09355715	0.00047493	6
O	18.63302097	-8.44568207	-0.00387938	8

Final Geom Energy = -30.1785551555 au
 Final Geom Energy = -18937.3464 kcal/mol

Geometry Optimization elapsed time 1 min. 54 sec.

***** Heat of Formation *****
 6661.9791 kcal/mol

Atomic spin densities

1	C	0.5857
2	O	0.0239
3	C	0.3962
4	N	0.0000
5	C	-0.0059
6	O	0.0000

S2 operator

exact	0.750000
calculated	0.751934

Properties elapsed time 0 sec.

Total Elapsed Time 1 min. 55 sec.

Quick Plot HOMO : Aniline

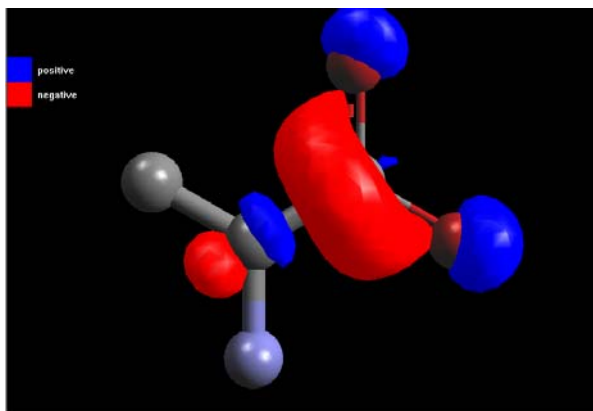


Diagram 6.33: Quick Plot HOMO of Aniline

Table 6.13 : Calculating Molecular Orbitals on grids for plotting HOMO of Aniline

***** Validated Experiment & Chemical System Settings *****

Calculation started: Thu Jun 04 15:34:36 2009

Max. SCF cycles 200
SCF convergence 1.5936e-009 au. for energy

AM1 param file C:\Program Files\ArgusLab\params\am1.prm
SCF saved every 1000 cycles

Two-electron integrals
 buffer size 1000
 storage random list in core

Property integrals one center
Dipole integrals length operator

Input Atomic Information

1	C	16.904500	-7.535300	0.000000
2	O	16.904500	-6.205300	0.000000
3	C	15.752700	-8.200300	0.000000
4	N	15.752700	-9.530300	0.000000
5	C	14.600800	-7.535300	0.000000
6	O	18.056300	-8.200300	0.000000

Plotting the following orbitals to grid files:14

Constructing Chemical System(s)

Basis Set

basis functions : 24
shells : 12
primitives : 72

Memory for Main Chemical System
Max. number 2-ele. ints. = 1596

Memory Requirements (bytes)

Core 973208
Scratch 10368

Chapter 6: Conformational Study of Substance through Different Tools

System charge 0.000000

Total number of 2-ele integrals 870

Integrals elapsed time 0 sec.

***** SCF *****

Core repulsion 79.3611 au

Calculating one electron matrix

Diagonalizing starting one-ele. matrix

Performing SCF

Cycle	Energy (au)	Difference
1	-36.971063	
2	-32.276391119	4.69467
3	-33.620242051	-1.34385
4	-26.283142152	7.3371
5	-33.505498559	-7.22236
6	-26.230341192	7.27516
7	-33.507013214	-7.27667
8	-26.229486478	7.27753
9	-33.508071400	-7.27858
10	-26.229663085	7.27841
11	-33.508488479	-7.27883
12	-26.229742535	7.27875
13	-33.508637619	-7.2789
14	-26.229770710	7.27887
15	-33.508689823	-7.27892
16	-26.229780460	7.27891
17	-33.508708012	-7.27893
18	-26.229783833	7.27892
19	-33.508714344	-7.27893
20	-26.229785002	7.27893
21	-33.508716547	-7.27893
22	-26.229785408	7.27893
23	-33.508717315	-7.27893
24	-26.229785549	7.27893
25	-33.508717582	-7.27893
26	-26.229785599	7.27893
27	-33.508717675	-7.27893
28	-26.229785616	7.27893
29	-33.508717707	-7.27893
30	-26.229785622	7.27893
31	-33.508717718	-7.27893
32	-26.229785624	7.27893
33	-33.508717722	-7.27893
34	-26.229785624	7.27893
35	-33.508717723	-7.27893
36	-26.229785625	7.27893
37	-33.508717724	-7.27893
38	-26.229785625	7.27893
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40	-26.229785625	7.27893
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199	-33.508717724	-7.27893
200	-26.229785625	7.27893

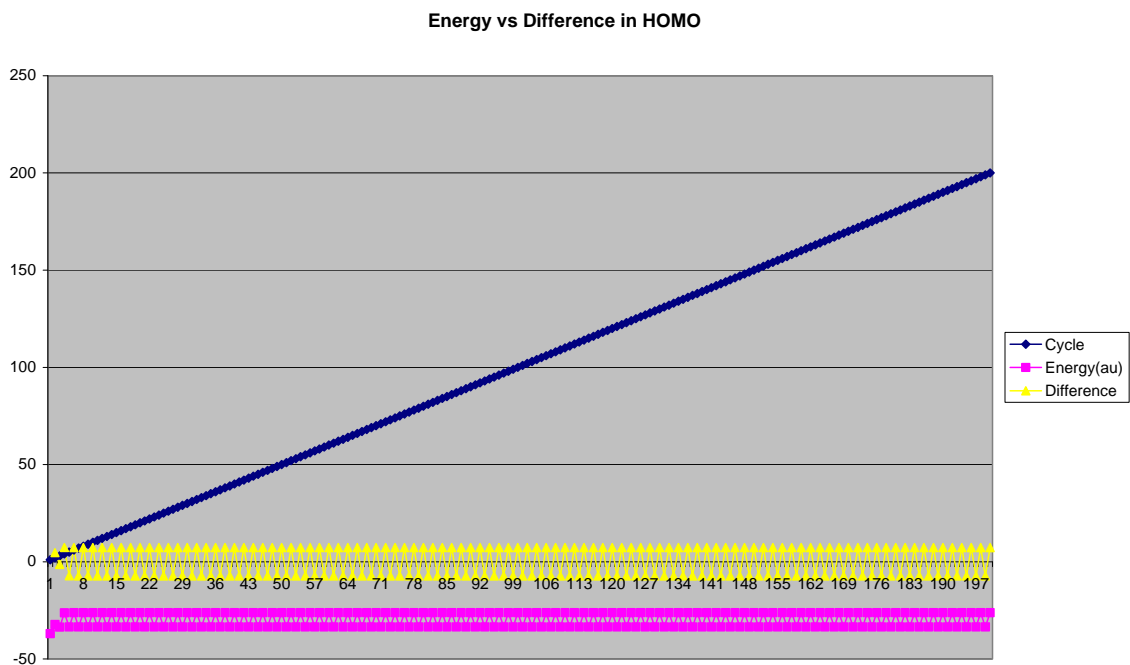


Figure 6.34: Energy vs. Difference in HOMO of Aniline

```

Maximum number of iterations reached: SCF NOT CONVERGED!

Writing final SCF to disk

Final SCF Energy = -26.2297856247 au
Final SCF Energy = -16459.4538 kcal/mol

Saving the final SCF to the restart file
E:\myphdthesis_2009\ACd_results\Ala.restartscf

SCF elapsed time 1 sec.

***** Heat of Formation *****
11889.4613 kcal/mol

Calculating Molecular Orbitals on grids for plotting.

Atomic spin densities
*****
1 C 0.6170
2 O 0.0225
3 C 0.3284
4 N 0.0005
5 C 0.0044
6 O 0.0273

S2 operator
*****
exact 0.750000
calculated 0.750201

Properties elapsed time 1 sec.
Total Elapsed Time 2 sec.
    
```


Quick Plot LUMO: Alanine

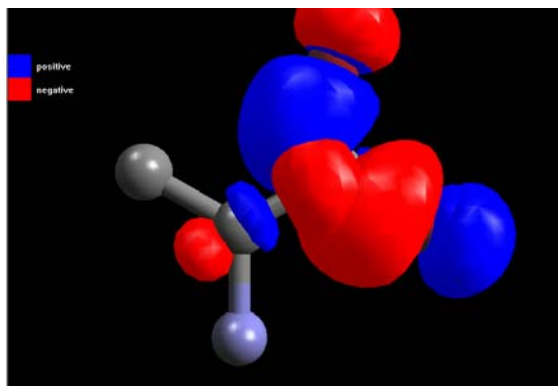


Diagram 6.35: Quick Plot LUMO of Alanine

Table 6.14 : Calculating Molecular Orbitals on grids for plotting LUMO of Alanine

***** Validated Experiment & Chemical System Settings *****

Calculation started: Thu Jun 04 15:53:38 2009

Max. SCF cycles 200
SCF convergence 1.5936e-009 au. for energy

AM1 param file C:\Program Files\ArgusLab\params\am1.prm
SCF saved every 1000 cycles

Input Atomic Information

1	C	16.904500	-7.535300	0.000000
2	O	16.904500	-6.205300	0.000000
3	C	15.752700	-8.200300	0.000000
4	N	15.752700	-9.530300	0.000000
5	C	14.600800	-7.535300	0.000000
6	O	18.056300	-8.200300	0.000000

Plotting the following orbitals to grid files:15

Constructing Chemical System(s)

Basis Set

basis functions : 24
shells : 12
primitives : 72

Memory for Main Chemical System
Max. number 2-ele. ints. = 1596

Memory Requirements (bytes)

Core 973208
Scratch 10368
System charge 0.000000

Chapter 6: Conformational Study of Substance through Different Tools

***** SCF *****

Core repulsion 79.3611 au

Calculating one electron matrix

Diagonalizing starting one-ele. matrix

Performing SCF

Cycle	Energy (au)	Difference
1	-36.971063	
2	-32.276391119	4.69467
3	-33.620242051	-1.34385
4	-26.283142152	7.3371
5	-33.505498559	-7.22236
6	-26.230341192	7.27516
7	-33.507013214	-7.27667
8	-26.229486478	7.27753
9	-33.508071400	-7.27858
10	-26.229663085	7.27841
11	-33.508488479	-7.27883
12	-26.229742535	7.27875
13	-33.508637619	-7.2789
14	-26.229770710	7.27887
15	-33.508689823	-7.27892
16	-26.229780460	7.27891
17	-33.508708012	-7.27893
18	-26.229783833	7.27892
19	-33.508714344	-7.27893
20	-26.229785002	7.27893
21	-33.508716547	-7.27893
22	-26.229785408	7.27893
23	-33.508717315	-7.27893
24	-26.229785549	7.27893
25	-33.508717582	-7.27893
26	-26.229785599	7.27893
27	-33.508717675	-7.27893
28	-26.229785616	7.27893
29	-33.508717707	-7.27893
30	-26.229785622	7.27893
31	-33.508717718	-7.27893
32	-26.229785624	7.27893
33	-33.508717722	-7.27893
34	-26.229785624	7.27893
35	-33.508717723	-7.27893
36	-26.229785625	7.27893
37	-33.508717724	-7.27893
38	-26.229785625	7.27893
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153	-33.508717724	-7.27893
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200	-26.229785625	7.27893

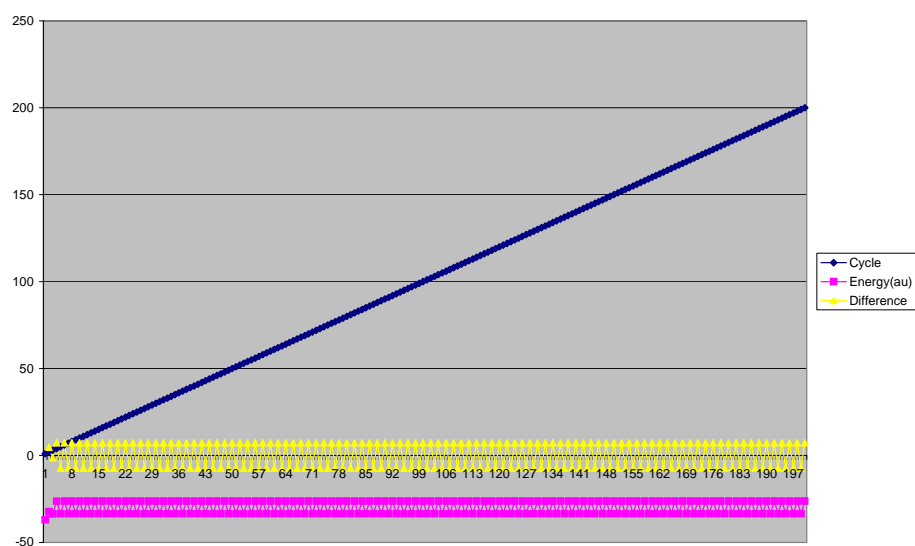


Figure 6.36: Energy vs. Difference by LUMO of Aniline

```
Maximum number of iterations reached: SCF NOT CONVERGED!

Writing final SCF to disk

Final SCF Energy = -26.2297856247 au
Final SCF Energy = -16459.4538 kcal/mol

Saving the final SCF to the restart file
E:\myphdthesis_2009\ACd_results\Ala.restartscf

SCF elapsed time 0 sec.

***** Heat of Formation *****
11889.4613 kcal/mol

Calculating Molecular Orbitals on grids for plotting.

Atomic spin densities
*****
1 C 0.6170
2 O 0.0225
3 C 0.3284
4 N 0.0005
5 C 0.0044
6 O 0.0273

S2 operator
*****

exact 0.750000
calculated 0.750201

Properties elapsed time 1 sec.
Total Elapsed Time 1 sec.
```

Quick Plot ESP Mapped Density : Aniline

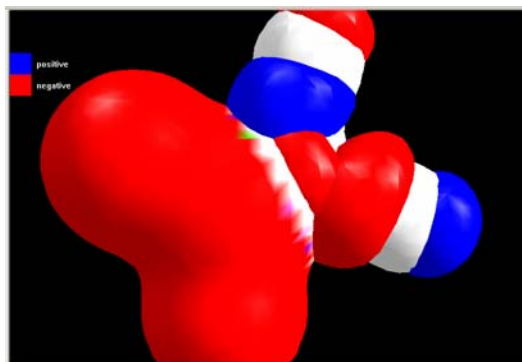


Figure 6.37: Quick Plot ESP Mapped Density of Aniline

Table 6.15 : Calculating Molecular Orbitals on grids for plotting ESP Mapped Density of Aniline

```
***** Validated Experiment & Chemical System Settings *****

Max. SCF cycles 200
SCF convergence 1.5936e-009 au. for energy
```

Chapter 6: Conformational Study of Substance through Different Tools

```
Two-electron integrals
  buffer size          1000
Input Atomic Information
*****
  1  C   16.904500  -7.535300  0.000000
  2  O   16.904500  -6.205300  0.000000
  3  C   15.752700  -8.200300  0.000000
  4  N   15.752700  -9.530300  0.000000
  5  C   14.600800  -7.535300  0.000000
  6  O   18.056300  -8.200300  0.000000
```

```
Plotting electron density for the following states to grid files:0
Plotting electrostatic potential for the following states to grid files:0
```

Constructing Chemical System(s)

Basis Set

```
basis functions : 24
shells          : 12
primitives      : 72
```

Memory for Main Chemical System
Max. number 2-ele. ints. = 1596

Memory Requirements (bytes)

```
Core           973208
Scratch        10368
System charge  0.000000
```

***** SCF *****

Core repulsion 79.3611 au

Calculating one electron matrix
Diagonalizing starting one-ele. matrix
Performing SCF

Cycle	Energy (au)	Difference
1	-36.971063	
2	-32.276391119	4.69467
3	-33.620242051	-1.34385
4	-26.283142152	7.3371
5	-33.505498559	-7.22236
6	-26.230341192	7.27516
7	-33.507013214	-7.27667
8	-26.229486478	7.27753
9	-33.508071400	-7.27858
10	-26.229663085	7.27841
11	-33.508488479	-7.27883
12	-26.229742535	7.27875
13	-33.508637619	-7.2789
14	-26.229770710	7.27887
15	-33.508689823	-7.27892
16	-26.229780460	7.27891
17	-33.508708012	-7.27893
18	-26.229783833	7.27892
19	-33.508714344	-7.27893
20	-26.229785002	7.27893
21	-33.508716547	-7.27893
22	-26.229785408	7.27893
23	-33.508717315	-7.27893
24	-26.229785549	7.27893
25	-33.508717582	-7.27893
26	-26.229785599	7.27893
27	-33.508717675	-7.27893
28	-26.229785616	7.27893
29	-33.508717707	-7.27893
30	-26.229785622	7.27893
31	-33.508717718	-7.27893
32	-26.229785624	7.27893

Chapter 6: Conformational Study of Substance through Different Tools

33	-33.508717722	-7.27893
34	-26.229785624	7.27893
35	-33.508717723	-7.27893
36	-26.229785625	7.27893
37	-33.508717724	-7.27893
38	-26.229785625	7.27893
39	-33.508717724	-7.27893
40	-26.229785625	7.27893
41	-33.508717724	-7.27893
42	-26.229785625	7.27893
43	-33.508717724	-7.27893
44	-26.229785625	7.27893
45	-33.508717724	-7.27893
46	-26.229785625	7.27893
47	-33.508717724	-7.27893
48	-26.229785625	7.27893
49	-33.508717724	-7.27893
50	-26.229785625	7.27893
51	-33.508717724	-7.27893
52	-26.229785625	7.27893
53	-33.508717724	-7.27893
54	-26.229785625	7.27893
55	-33.508717724	-7.27893
56	-26.229785625	7.27893
57	-33.508717724	-7.27893
58	-26.229785625	7.27893
59	-33.508717724	-7.27893
60	-26.229785625	7.27893
61	-33.508717724	-7.27893
62	-26.229785625	7.27893
63	-33.508717724	-7.27893
64	-26.229785625	7.27893
65	-33.508717724	-7.27893
66	-26.229785625	7.27893
67	-33.508717724	-7.27893
68	-26.229785625	7.27893
69	-33.508717724	-7.27893
70	-26.229785625	7.27893
71	-33.508717724	-7.27893
72	-26.229785625	7.27893
73	-33.508717724	-7.27893
74	-26.229785625	7.27893
75	-33.508717724	-7.27893
76	-26.229785625	7.27893
77	-33.508717724	-7.27893
78	-26.229785625	7.27893
79	-33.508717724	-7.27893
80	-26.229785625	7.27893
81	-33.508717724	-7.27893
82	-26.229785625	7.27893
83	-33.508717724	-7.27893
84	-26.229785625	7.27893
85	-33.508717724	-7.27893
86	-26.229785625	7.27893
87	-33.508717724	-7.27893
88	-26.229785625	7.27893
89	-33.508717724	-7.27893
90	-26.229785625	7.27893
91	-33.508717724	-7.27893
92	-26.229785625	7.27893
93	-33.508717724	-7.27893
94	-26.229785625	7.27893
95	-33.508717724	-7.27893
96	-26.229785625	7.27893
97	-33.508717724	-7.27893
98	-26.229785625	7.27893
99	-33.508717724	-7.27893
100	-26.229785625	7.27893
101	-33.508717724	-7.27893
102	-26.229785625	7.27893
103	-33.508717724	-7.27893
104	-26.229785625	7.27893
105	-33.508717724	-7.27893
106	-26.229785625	7.27893
107	-33.508717724	-7.27893
108	-26.229785625	7.27893
109	-33.508717724	-7.27893
110	-26.229785625	7.27893
111	-33.508717724	-7.27893
112	-26.229785625	7.27893
113	-33.508717724	-7.27893

114	-26.229785625	7.27893
115	-33.508717724	-7.27893
116	-26.229785625	7.27893
117	-33.508717724	-7.27893
118	-26.229785625	7.27893
119	-33.508717724	-7.27893
120	-26.229785625	7.27893
121	-33.508717724	-7.27893
122	-26.229785625	7.27893
123	-33.508717724	-7.27893
124	-26.229785625	7.27893
125	-33.508717724	-7.27893
126	-26.229785625	7.27893
127	-33.508717724	-7.27893
128	-26.229785625	7.27893
129	-33.508717724	-7.27893
130	-26.229785625	7.27893
131	-33.508717724	-7.27893
132	-26.229785625	7.27893
133	-33.508717724	-7.27893
134	-26.229785625	7.27893
135	-33.508717724	-7.27893
136	-26.229785625	7.27893
137	-33.508717724	-7.27893
138	-26.229785625	7.27893
139	-33.508717724	-7.27893
140	-26.229785625	7.27893
141	-33.508717724	-7.27893
142	-26.229785625	7.27893
143	-33.508717724	-7.27893
144	-26.229785625	7.27893
145	-33.508717724	-7.27893
146	-26.229785625	7.27893
147	-33.508717724	-7.27893
148	-26.229785625	7.27893
149	-33.508717724	-7.27893
150	-26.229785625	7.27893
151	-33.508717724	-7.27893
152	-26.229785625	7.27893
153	-33.508717724	-7.27893
154	-26.229785625	7.27893
155	-33.508717724	-7.27893
156	-26.229785625	7.27893
157	-33.508717724	-7.27893
158	-26.229785625	7.27893
159	-33.508717724	-7.27893
160	-26.229785625	7.27893
161	-33.508717724	-7.27893
162	-26.229785625	7.27893
163	-33.508717724	-7.27893
164	-26.229785625	7.27893
165	-33.508717724	-7.27893
166	-26.229785625	7.27893
167	-33.508717724	-7.27893
168	-26.229785625	7.27893
169	-33.508717724	-7.27893
170	-26.229785625	7.27893
171	-33.508717724	-7.27893
172	-26.229785625	7.27893
173	-33.508717724	-7.27893
174	-26.229785625	7.27893
175	-33.508717724	-7.27893
176	-26.229785625	7.27893
177	-33.508717724	-7.27893
178	-26.229785625	7.27893
179	-33.508717724	-7.27893
180	-26.229785625	7.27893
181	-33.508717724	-7.27893
182	-26.229785625	7.27893
183	-33.508717724	-7.27893
184	-26.229785625	7.27893
185	-33.508717724	-7.27893
186	-26.229785625	7.27893
187	-33.508717724	-7.27893
188	-26.229785625	7.27893
189	-33.508717724	-7.27893
190	-26.229785625	7.27893
191	-33.508717724	-7.27893
192	-26.229785625	7.27893
193	-33.508717724	-7.27893
194	-26.229785625	7.27893

195	-33.508717724	-7.27893
196	-26.229785625	7.27893
197	-33.508717724	-7.27893
198	-26.229785625	7.27893
199	-33.508717724	-7.27893
200	-26.229785625	7.27893

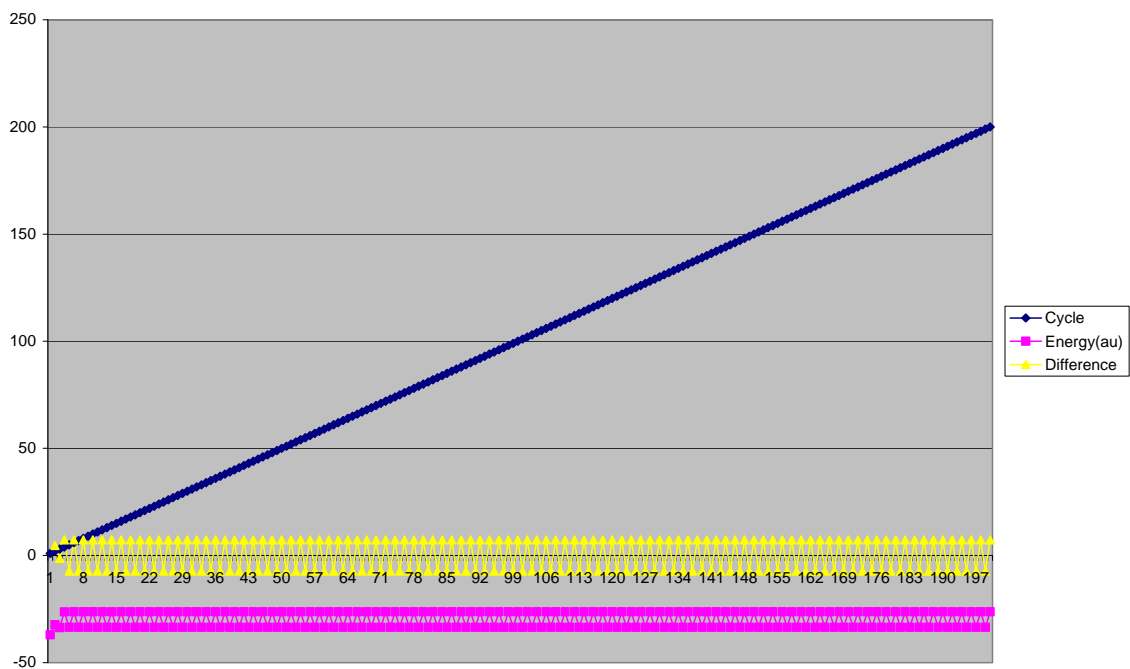


Figure 6.38: Energy vs. Difference /cycle by ESP Mapped Density of Aniline

Final SCF Energy = -26.2297856247 au
 Final SCF Energy = -16459.4538 kcal/mol

SCF elapsed time 1 sec.

***** Heat of Formation *****
 11889.4613 kcal/mol

Calculating ground-state density on grid.

Calculating ground-state electrostatic potential on grid.

Atomic spin densities

```
*****
1   C   0.6170
2   O   0.0225
3   C   0.3284
4   N   0.0005
5   C   0.0044
6   O   0.0273
```

S2 operator

```
exact          0.750000
calculated     0.750201
```

Properties elapsed time 17 sec.
 Total Elapsed Time 18 sec.

6.2.2.1 Data Analysis of Experimental Work

In this research work like Alanine other compounds amino butyric acid, Asparagine and Glutamine have been analyzed on ArgusLab tool and calculations like Single Entry Point Calculation and Geometric Optimization have been performed.

(A) Amino butyric acid

Single Entry Point Calculation: Amino butyric acid

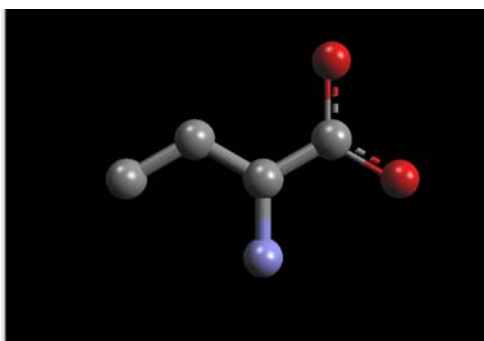


Figure 6.39: Structure of Amino butyric acid

Table 6.16: Single Entry Point Calculation of Amino butyric acid

```
***** Validated Experiment & Chemical System Settings *****

Input Atomic Information
*****
  1  C   16.727200  -9.538300  0.000000
  2  O   16.727200  -8.208300  0.000000
  3  C   15.575500 -10.203300  0.000000
  4  N   15.575500 -11.533300  0.000000
  5  C   14.423600  -9.538300  0.000000
  6  O   17.879100 -10.203300  0.000000
  7  C   13.271800 -10.203400  0.000000

Constructing Chemical System(s)

Basis Set
*****
basis functions : 28
shells          : 14
primitives      : 84

Memory for Main Chemical System
Max. number 2-ele. ints. = 2212

Memory Requirements (bytes)
*****
Core              154592
Scratch           13888
System charge    0.000000
```

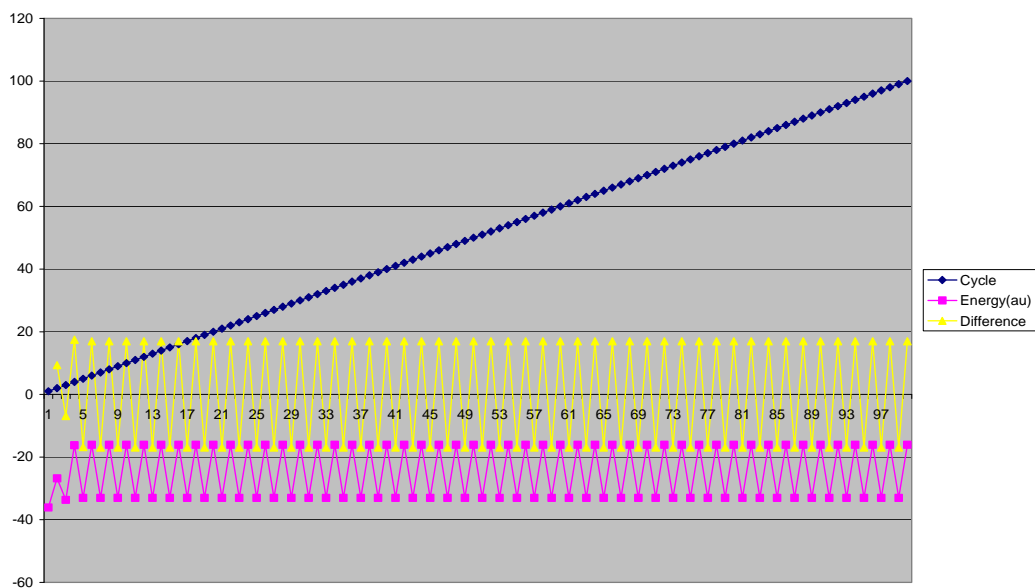


Figure 6.40: Energy vs. Difference /cycle by SEC of Amino butyric acid

```

***** SCF *****

Writing final SCF to disk

Final SCF Energy = -16.0505499405 au
Final SCF Energy = -10071.8812 kcal/mol

SCF elapsed time 1 sec.

***** Heat of Formation *****
21233.9419 kcal/mol

Wiberg Atom-Atom Bond Orders
*****
      1      2      3      4      5      6
1  0.000000
2  0.159277      0.000000
3  0.593365      0.132542      0.000000
4  0.001599      0.002860      0.003560      0.000000
5  0.000804      0.005289      0.002427      0.000046      0.000000
6  0.052133      0.006354      0.140044      0.004958      0.000753      0.000000
7  0.000011      0.000003      0.000014      0.000000      0.000000      0.000006

Atomic spin densities
*****
      1  C  0.0616
      2  O  0.7115
      3  C  0.1654
      4  N  0.0033
      5  C  0.0068
      6  O  0.0514
      7  C  0.0000

S2 operator
*****
exact          0.750000
calculated     0.751744

Ground State Dipole (debye)
      X      Y      Z      length
179.55334737  64.31200335  -0.00000000  190.72346034
    
```

Mulliken Atomic Charges

1	C	3.7788
2	O	4.7938
3	C	-3.4233
4	N	-2.9980
5	C	-4.0014
6	O	5.8506
7	C	-4.0005

Properties elapsed time 0 sec.

Total Elapsed Time 1 sec.

Geometry Optimization: Amino butyric acid

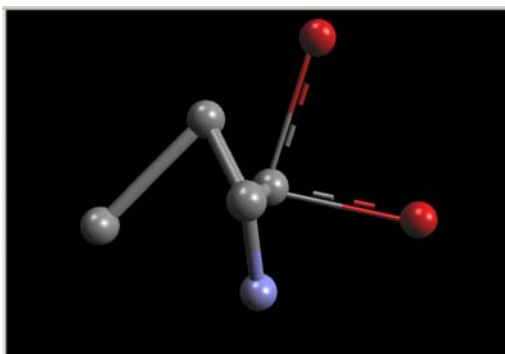


Figure 6.41: Geometry Optimization of Amino butyric acid

Table 6.9 : Geometry Optimization calculation of Amino butyric acid

***** Validated Experiment & Chemical System Settings *****

Calculation started: Thu Jun 04 16:50:46 2009

Title:E:\myphdthesis_2009\ACd_results\Abu
Max. geom cycles 100
Convergence criteria:
max. grad. component < 0.000084 au.

Input Atomic Information

1	C	16.727200	-9.538300	0.000000
2	O	16.727200	-8.208300	0.000000
3	C	15.575500	-10.203300	0.000000
4	N	15.575500	-11.533300	0.000000
5	C	14.423600	-9.538300	0.000000
6	O	17.879100	-10.203300	0.000000
7	C	13.271800	-10.203400	0.000000

Constructing Chemical System(s)

Basis Set

basis functions	: 28
shells	: 14
primitives	: 42

```

Memory for Main Chemical System
Max. number 2-ele. ints. = 2212

Memory Requirements (bytes)
*****
Core          440544
Scratch       13888
System charge 0.000000

Total number of 2-ele integrals 1232

Integrals elapsed time 0 sec.
    
```

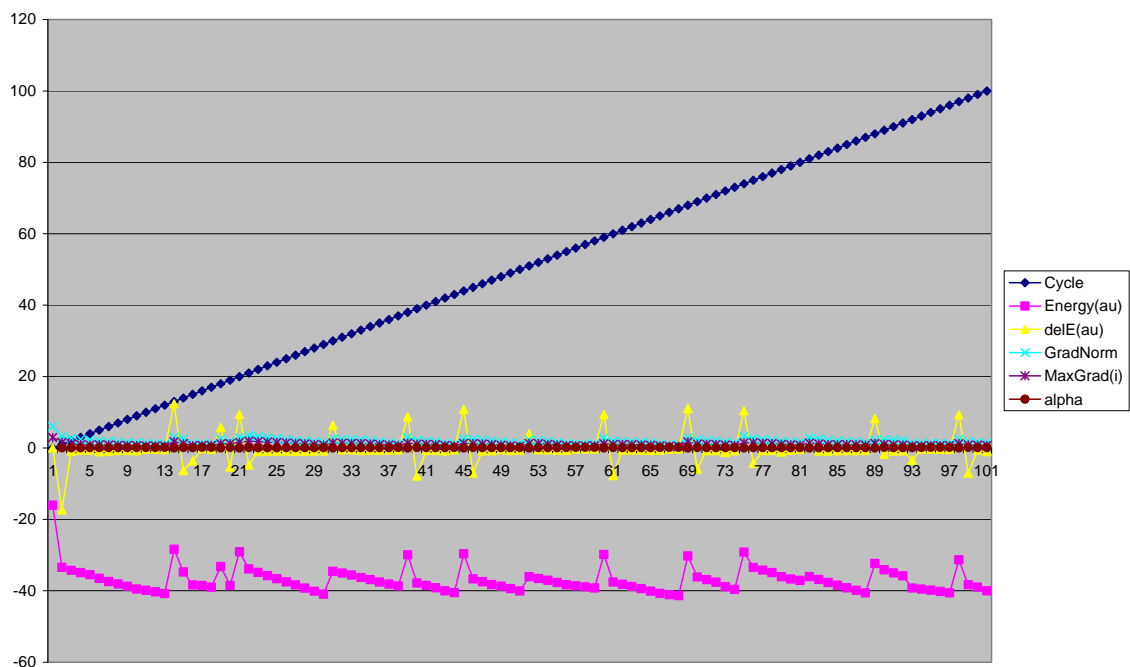


Figure 6.42: Geometry Optimization Energy for various components of Amino butyric acid

```

***** SCF *****

Core repulsion      99.4321 au
Final SCF Energy =  -16.0321304196 au
Final SCF Energy = -10060.3228 kcal/mol

***** Final Geometry *****

C  15.94062971   -9.99263718   -0.48994553    6
O  16.76806159   -7.34951479    0.07850090    8
C  15.52407585  -10.21360278    0.64044515    6
N  15.72470082  -11.86193116   -0.07280711    7
C  14.77876094   -8.77341644   -0.10464275    6
O  18.56438623  -10.55937950    0.03787747    8
C  12.87928488  -10.67771816   -0.08942813    6

Final Geom Energy = -39.9881072545 au
Final Geom Energy = -25092.9388 kcal/mol

Geometry Optimization elapsed time 4 min. 53 sec.

***** Heat of Formation *****

10076.7315 kcal/mol
    
```

```
Atomic spin densities
*****
      1  C   1.0005
      2  O   0.0005
      3  C  -0.0225
      4  N   0.0189
      5  C   0.0019
      6  O   0.0008
      7  C  -0.0000

S2 operator
*****
exact                0.750000
calculated           0.790950

Properties elapsed time 0 sec.
Total Elapsed Time 4 min. 54 sec.
```

Quick Plot HOMO: Amino butyric acid

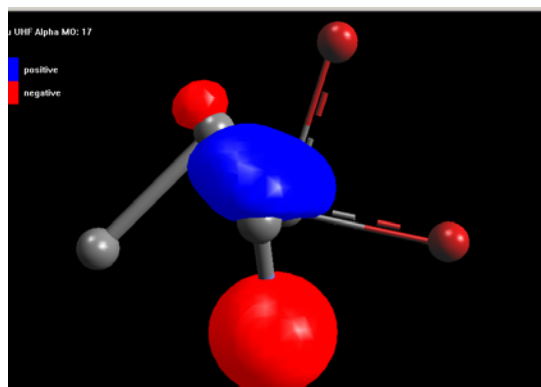


Figure 6.43: Quick Plot HOMO of Amino butyric acid

Table 6.17: Quick Plot HOMO calculation of Amino butyric acid

```
Calculation started: Thu Jun 04 17:29:12 2009

Max. SCF cycles          200
SCF convergence          1.5936e-009 au. for energy

Input Atomic Information
*****
      1  C   15.940630  -9.992637  -0.489946
      2  O   16.768062  -7.349515   0.078501
      3  C   15.524076 -10.213603   0.640445
      4  N   15.724701 -11.861931  -0.072807
      5  C   14.778761  -8.773416  -0.104643
      6  O   18.564386 -10.559379   0.037877
      7  C   12.879285 -10.677718  -0.089428

Plotting the following orbitals to grid files:16
```

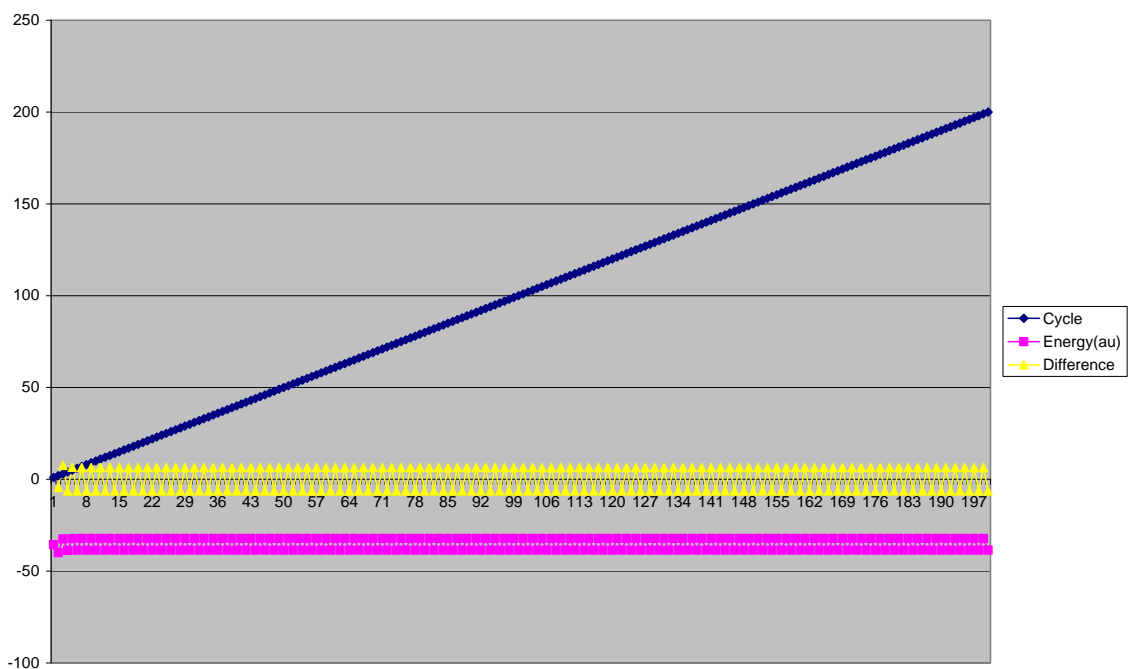


Figure 6.44: Energy vs. Difference /cycle Quick Plot HOMO calculation of Amino butyric acid

```

Constructing Chemical System(s)
  Basis Set
  *****

  basis functions : 28
  shells          : 14
  primitives      : 84

  Memory for Main Chemical System
  Max. number 2-ele. ints. = 2212

  Memory Requirements (bytes)
  *****

  Core          1002304
  Scratch       13888
  System charge 0.000000

**** SCF ****

  Core repulsion 79.7563 au
  Final SCF Energy = -38.5635015264 au
  Final SCF Energy = -24198.9844 kcal/mol

**** Heat of Formation ****
  7106.8388 kcal/mol

  Calculating Molecular Orbitals on grids for plotting.

  Atomic spin densities
  *****

  1  C  -0.0184
  2  O   0.0000
  3  C   0.0197
  4  N   0.0012
  5  C   0.9972
  6  O  -0.0000
  7  C   0.0002
    
```

S2 operator

exact 0.750000
calculated 0.751080

Properties elapsed time 1 sec.

Total Elapsed Time 2 sec.

Quick Plot LUMO : Amino butyric acid

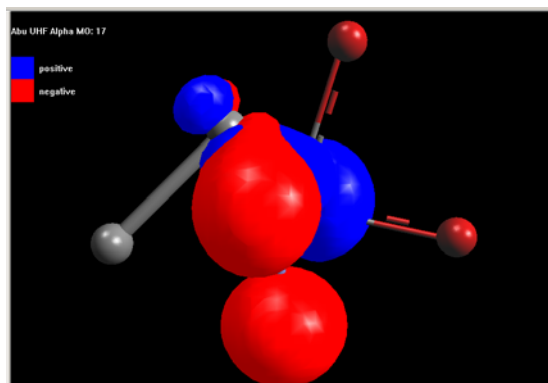


Figure 6.45: Quick Plot LUMO of Amino butyric acid

Table 6.18 : Quick Plot LUMO calculation of Amino butyric acid

***** Validated Experiment & Chemical System Settings *****

Calculation started: Thu Jun 04 17:50:51 2009
Max. SCF cycles 200
SCF convergence 1.5936e-009 au. for energy

Input Atomic Information

1	C	15.940630	-9.992637	-0.489946
2	O	16.768062	-7.349515	0.078501
3	C	15.524076	-10.213603	0.640445
4	N	15.724701	-11.861931	-0.072807
5	C	14.778761	-8.773416	-0.104643
6	O	18.564386	-10.559379	0.037877
7	C	12.879285	-10.677718	-0.089428

Constructing Chemical System(s)

Basis Set

basis functions : 28
shells : 14
primitives : 84

Memory for Main Chemical System
Max. number 2-ele. ints. = 2212

Memory Requirements (bytes)

Core 1002304
Scratch 13888
System charge 0.000000


```

Total number of 2-ele integrals 2212
***** SCF *****

Core repulsion      79.7563 au
Final SCF Energy =  -38.5635015264 au
Final SCF Energy =  -24198.9844 kcal/mol

SCF elapsed time 1 sec.

***** Heat of Formation *****
7106.8388 kcal/mol
    
```

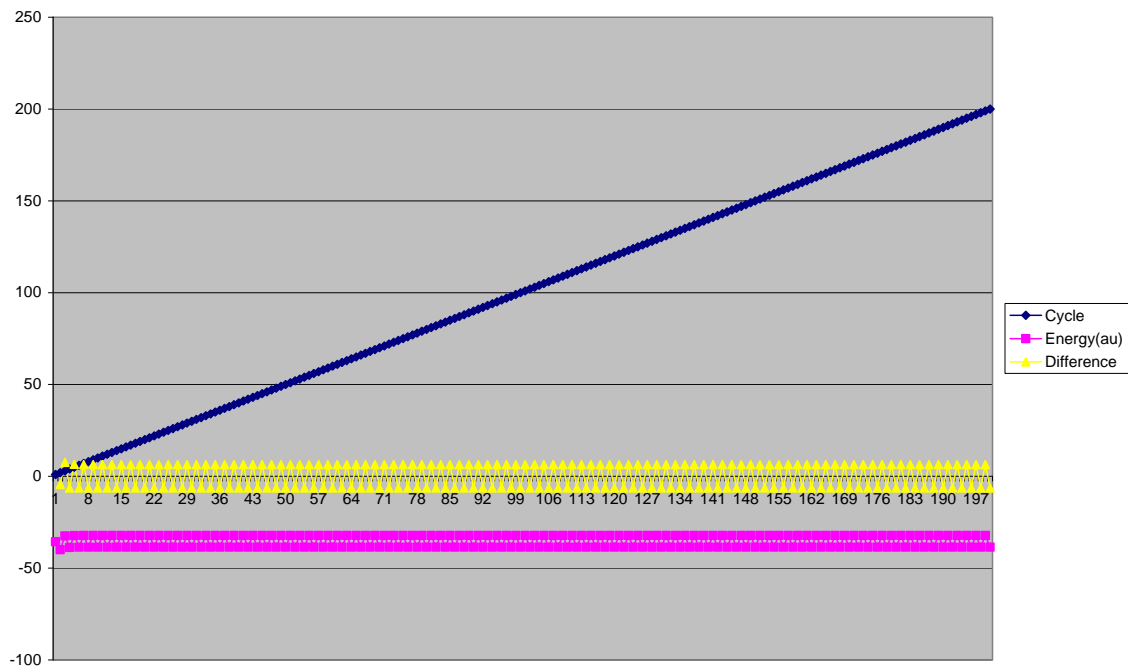


Figure 6.46: Energy vs. Difference/ cycle by Quick Plot LUMO of Amino butyric acid

```

Calculating Molecular Orbitals on grids for plotting.
Atomic spin densities
*****
    1  C  -0.0184
    2  O   0.0000
    3  C   0.0197
    4  N   0.0012
    5  C   0.9972
    6  O  -0.0000
    7  C   0.0002

S2 operator
*****
exact          0.750000
calculated     0.751080

Properties elapsed time 1 sec.
    
```

Quick Plot ESP Mapped Density: Amino butyric acid

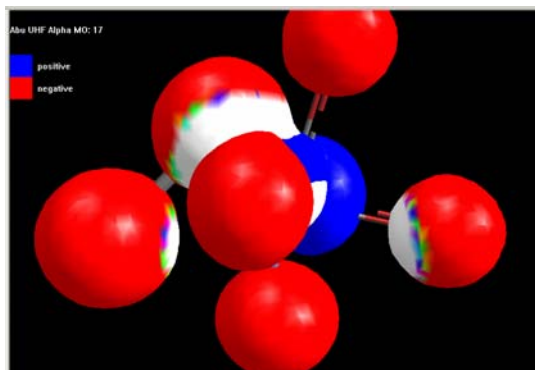


Figure 6.47: Quick Plot ESP Mapped Density of Amino butyric acid

Table 6.19: Quick Plot ESP Mapped Density calculation of Amino butyric acid

***** Validated Experiment & Chemical System Settings *****

Calculation started: Thu Jun 04 18:40:01 2009

Title:
SCF convergence 1.5936e-009 au. for energy

Input Atomic Information

1	C	15.940630	-9.992637	-0.489946
2	O	16.768062	-7.349515	0.078501
3	C	15.524076	-10.213603	0.640445
4	N	15.724701	-11.861931	-0.072807
5	C	14.778761	-8.773416	-0.104643
6	O	18.564386	-10.559379	0.037877
7	C	12.879285	-10.677718	-0.089428

Constructing Chemical System(s)

Basis Set

basis functions : 28
shells : 14
primitives : 84

Memory for Main Chemical System
Max. number 2-ele. ints. = 2212

Memory Requirements (bytes)

Core 1002304
Scratch 13888
System charge 0.000000

Total number of 2-ele integrals 2212

Integrals elapsed time 0 sec.

***** SCF *****

Core repulsion 79.7563 au
 Calculating one electron matrix
 Diagonalizing starting one-ele. matrix
 Performing SCF
 Final SCF Energy = -38.5635015264 au
 Final SCF Energy = -24198.9844 kcal/mol
 SCF elapsed time 1 sec.

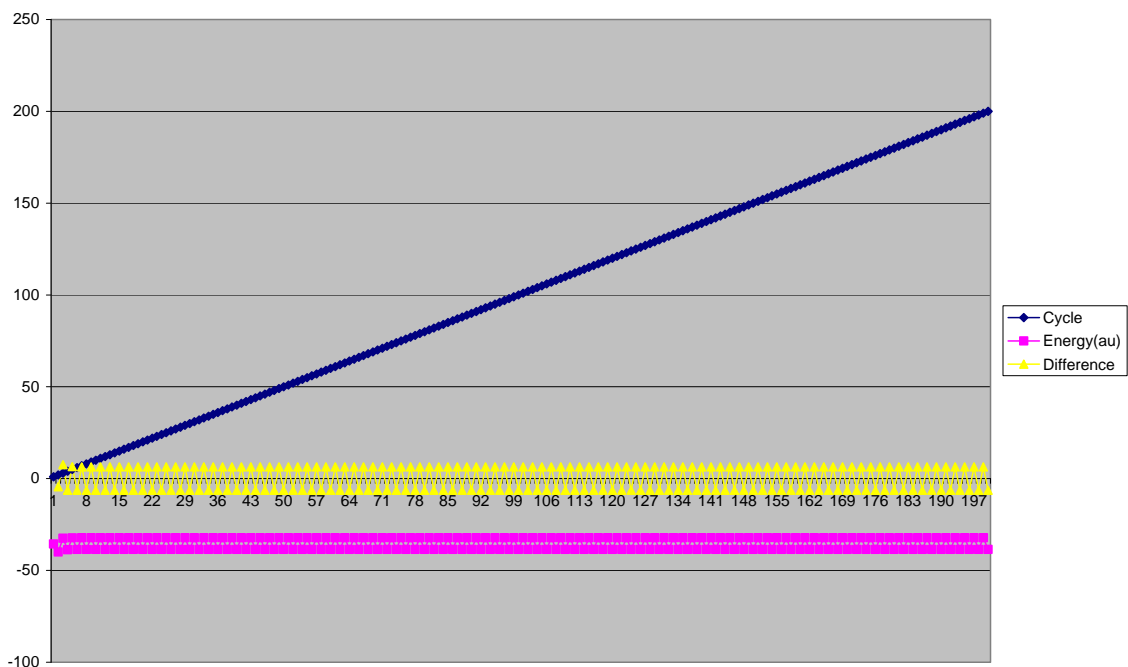


Figure 6.48: Energy vs. Difference / cycle by Quick Plot ESP Mapped Density of Amino butyric acid

***** Heat of Formation *****
 7106.8388 kcal/mol

Atomic spin densities

1	C	-0.0184
2	O	0.0000
3	C	0.0197
4	N	0.0012
5	C	0.9972
6	O	-0.0000
7	C	0.0002

S2 operator

exact	0.750000
calculated	0.751080

Properties elapsed time 19 sec.
 Total Elapsed Time 20 sec.

Asparagine

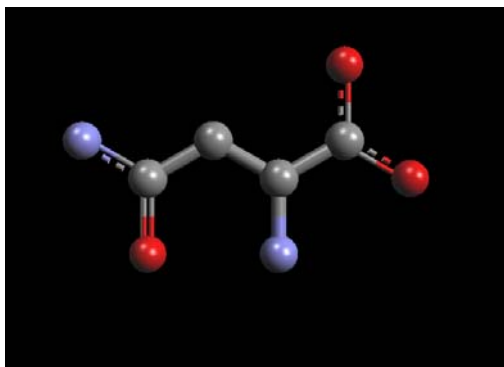


Figure 6.49: Structure of Asparagine

Single Entry Point Calculation: Asparagine

Table 6.20: Single Entry Point Calculation of Asparagine

```
Input Atomic Information
*****
  1  C   21.569300  -8.224400  0.000000
  2  C   23.873000  -8.224400  0.000000
  3  C   20.417600  -8.889400  0.000000
  4  C   22.721200  -8.889400  0.000000
  5  O   23.873000  -6.894400  0.000000
  6  O   20.417600 -10.219400  0.000000
  7  N   19.265800  -8.224400  0.000000
  8  N   22.721200 -10.219400  0.000000
  9  O   25.024800  -8.889400  0.000000

Constructing Chemical System(s)

Basis Set
*****
      basis functions : 36
      shells          : 18
      primitives      : 108

      Memory for Main Chemical System
      Max. number 2-ele. ints. = 3744

***** SCF *****
      Core repulsion    165.108 au
      Final SCF Energy = -24.2129529102 au
      Final SCF Energy = -15193.8710 kcal/mol

***** Heat of Formation *****
      28241.3138 kcal/mol
```

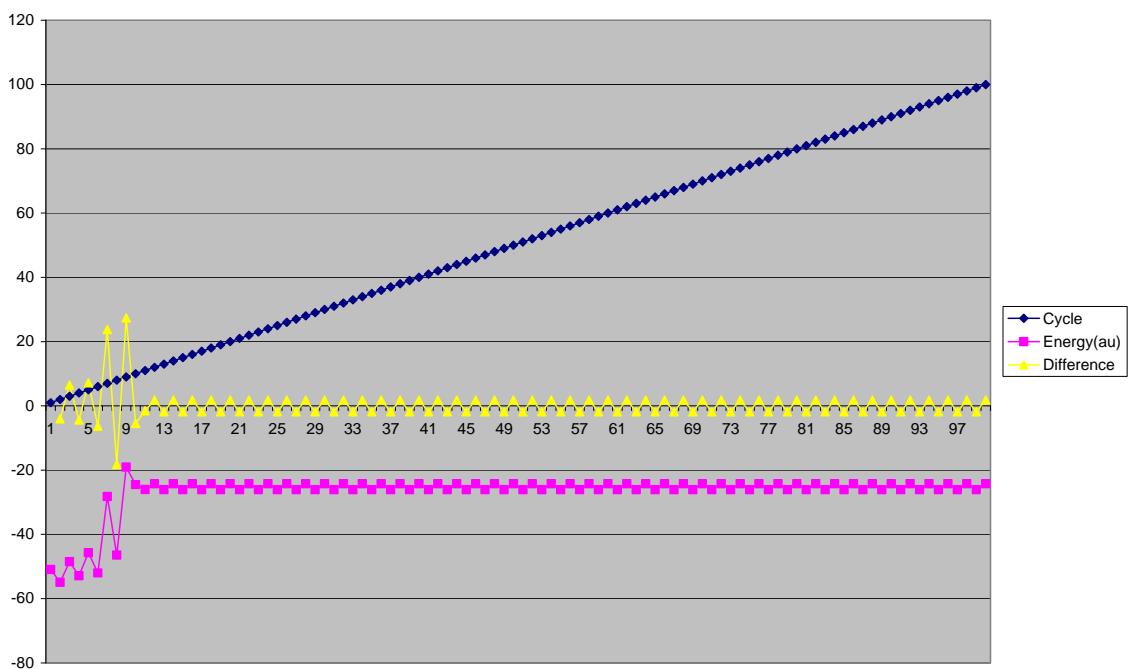


Figure 6.50: Energy vs. Difference of Single Entry Point Calculation of Asparagine

Properties elapsed time 0 sec.

Total Elapsed Time 1 sec.

Geometry Optimization: Asparagine

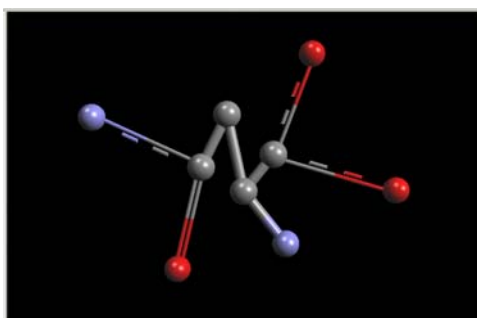


Figure 6.51: Geometry Optimization of Asparagine

Table 6.21 : Geometry Optimization calculation of Asparagine

Max. SCF cycles	50
SCF convergence	1.5936e-013 au. for energy
Max. geom cycles	100
Convergence criteria:	
max. grad. component <	0.000084 au.

```

Input Atomic Information
*****
      1  C   21.569300  -8.224400  0.000000
      2  C   23.873000  -8.224400  0.000000
      3  C   20.417600  -8.889400  0.000000
      4  C   22.721200  -8.889400  0.000000
      5  O   23.873000  -6.894400  0.000000
      6  O   20.417600 -10.219400  0.000000
      7  N   19.265800  -8.224400  0.000000
      8  N   22.721200 -10.219400  0.000000
      9  O   25.024800  -8.889400  0.000000

Constructing Chemical System(s)

Basis Set
*****
      basis functions : 36
      shells          : 18
      primitives      : 54

      Memory for Main Chemical System
      Max. number 2-ele. ints. = 3744

Memory Requirements (bytes)
*****
      Core              727504
      Scratch           22464
**** SCF ****

      Core repulsion   165.108 au
      Final SCF Energy = -24.2078555875 au
      Final SCF Energy = -15190.6724 kcal/mol

      Final Geom Energy = -49.2464605506 au
      Final Geom Energy = -30902.6484 kcal/mol
      Geometry Optimization elapsed time 10 min. 41 sec.

**** Heat of Formation ****
      15290.0969 kcal/mol

Total Elapsed Time 10 min. 42 sec.
    
```

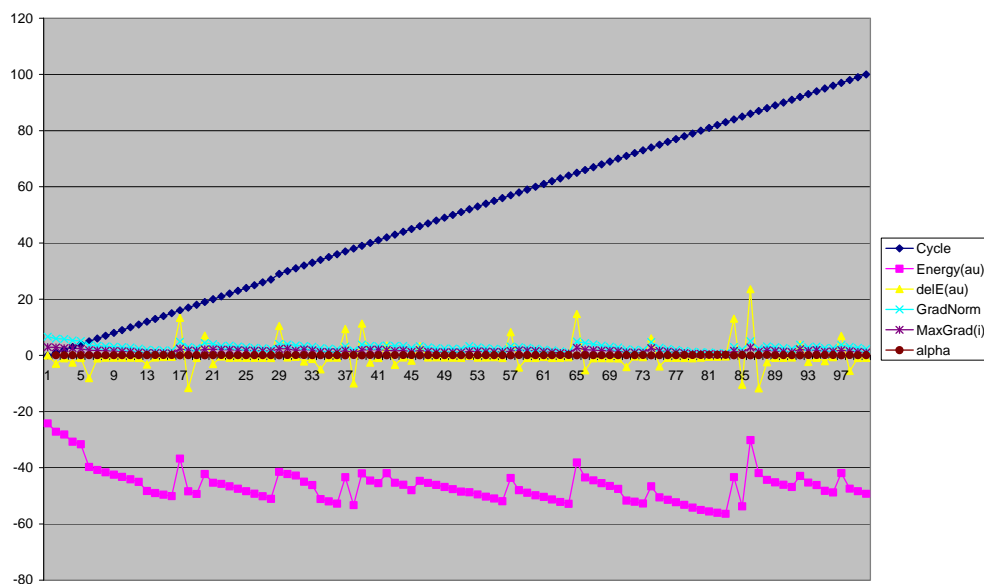


Figure 6.52: Geometry Optimization Energy for various components of Asparagine

QuickPlot HOMO: Asparagine

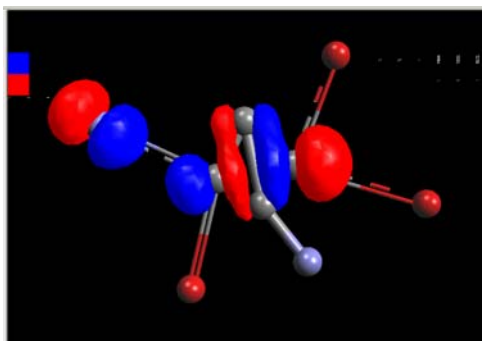


Figure 6.53: QuickPlot HOMO of Asparagine

Table 6.22: QuickPlot HOMO of Asparagine

```
***** Validated Experiment & Chemical System Settings *****

Max. SCF cycles          200
SCF convergence          1.5936e-009 au. for energy

Input Atomic Information
*****
   1  C   21.754545  -7.442867  0.000000
   2  C   22.902106  -8.435586 -0.000002
   3  C   21.117400  -8.750163 -0.000003
   4  C   22.167307  -9.347993  0.000005
   5  O   23.854067  -5.931479  0.000000
   6  O   20.535689 -11.277524  0.000000
   7  N   18.418534  -7.512176  0.000000
   8  N   23.206285 -10.652366 -0.000000
   9  O   25.927568  -9.324447  0.000000

Plotting the following orbitals to grid files:21
Constructing Chemical System(s)

Basis Set
*****
      basis functions : 36
      shells          : 18
      primitives      : 108

Memory for Main Chemical System
Max. number 2-ele. ints. = 3744

Memory Requirements (bytes)
*****
      Core            1051640
      Scratch         22464

**** SCF ****
      Core repulsion  133.616 au
      Final SCF Energy = -67.3585267794 au
      Final SCF Energy = -42268.1518 kcal/mol
**** Heat of Formation ****
      1167.0330 kcal/mol
Total Elapsed Time 2 sec.
```

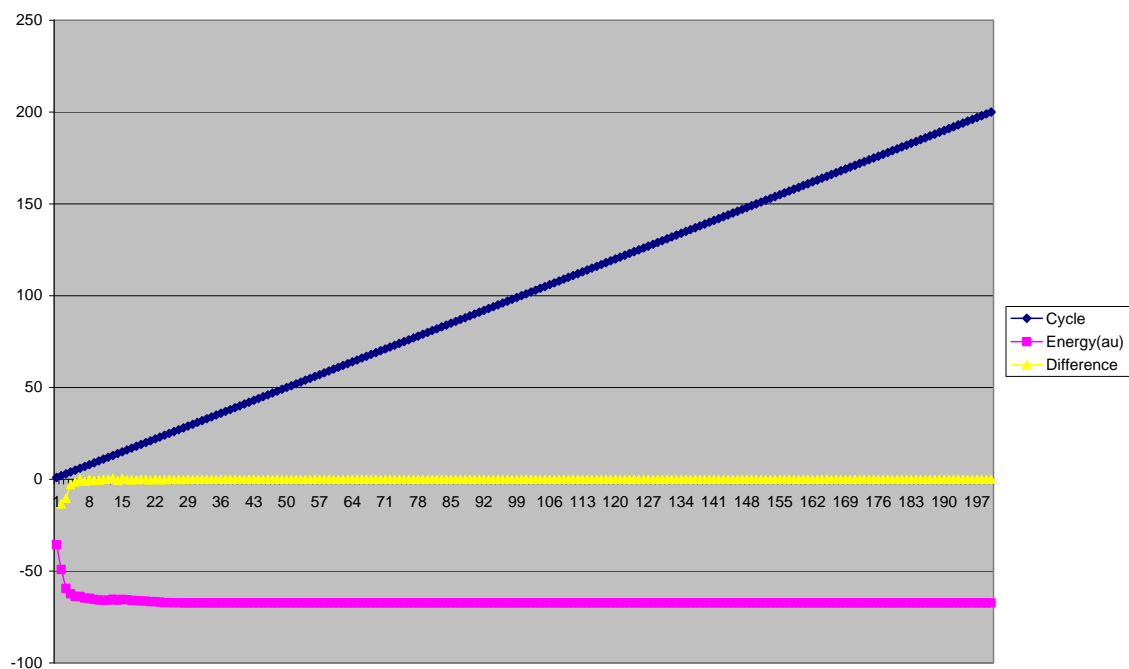


Figure 6.54 : Energy vs. Difference /cycle for QuickPlot HOMO of Asparagine

QuickPlot LUMO: Asparagine

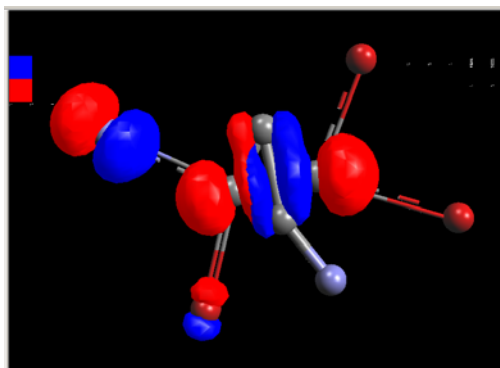


Figure 6.55: QuickPlot LUMO of Asparagine

Table 6.23: QuickPlot LUMO calculation of Asparagine

```

***** Validated Experiment & Chemical System Settings *****
Max. SCF cycles          200
SCF convergence          1.5936e-009 au. for energy

Input Atomic Information
*****
  1  C   21.754545  -7.442867   0.000000
  2  C   22.902106  -8.435586  -0.000002
  3  C   21.117400  -8.750163  -0.000003
  4  C   22.167307  -9.347993   0.000005
  5  O   23.854067  -5.931479   0.000000
  6  O   20.535689 -11.277524   0.000000
    
```



```

7  N  18.418534  -7.512176  0.000000
8  N  23.206285 -10.652366 -0.000000
9  O  25.927568  -9.324447  0.000000

Constructing Chemical System(s)

Basis Set
*****
basis functions : 36
shells         : 18
primitives     : 108

Memory for Main Chemical System
Max. number 2-ele. ints. = 3744

Memory Requirements (bytes)
*****
Core           1051640
Scratch        22464

***** SCF *****

Core repulsion  133.616 au
Final SCF Energy = -67.3585267794 au
Final SCF Energy = -42268.1518 kcal/mol

***** Heat of Formation *****
1167.0330 kcal/mol

Total Elapsed Time 2 sec.

```

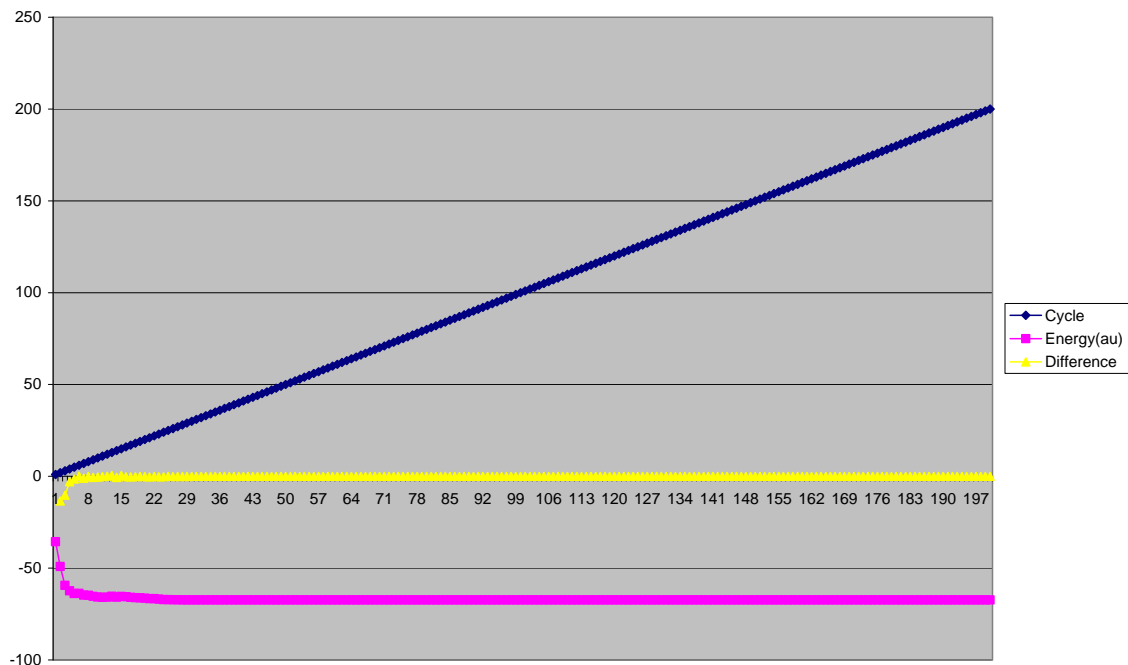


Figure 6.56: Energy vs. Difference /cycle for QuickPlot LUMO of Asparagine

Quick Plot ESP Mapped Density: Asparagine

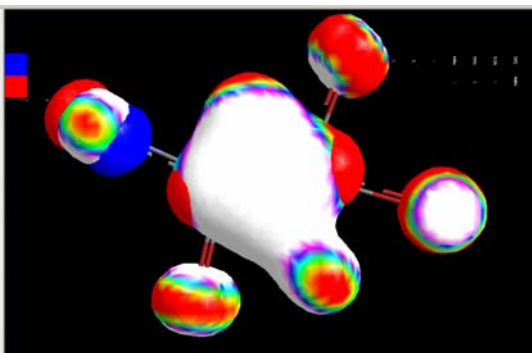


Figure 6.57: Quick Plot ESP Mapped Density of Asparagine

Table 6.24: Quick Plot ESP Mapped Density calculation of Asparagine

***** Validated Experiment & Chemical System Settings *****

Max. SCF cycles 200
 SCF convergence 1.5936e-009 au. for energy

Input Atomic Information

1	C	21.754545	-7.442867	0.000000
2	C	22.902106	-8.435586	-0.000002
3	C	21.117400	-8.750163	-0.000003
4	C	22.167307	-9.347993	0.000005
5	O	23.854067	-5.931479	0.000000
6	O	20.535689	-11.277524	0.000000
7	N	18.418534	-7.512176	0.000000
8	N	23.206285	-10.652366	-0.000000
9	O	25.927568	-9.324447	0.000000

Constructing Chemical System(s)

Basis Set

basis functions : 36
 shells : 18
 primitives : 108

Memory for Main Chemical System
 Max. number 2-ele. ints. = 3744

Memory Requirements (bytes)

Core 1051640
 Scratch 22464

Total number of 2-ele integrals 2935

***** SCF *****

Core repulsion 133.616 au
 Final SCF Energy = -67.3585267794 au
 Final SCF Energy = -42268.1518 kcal/mol

***** Heat of Formation *****
 1167.0330 kcal/mol

Total Elapsed Time 26 sec.

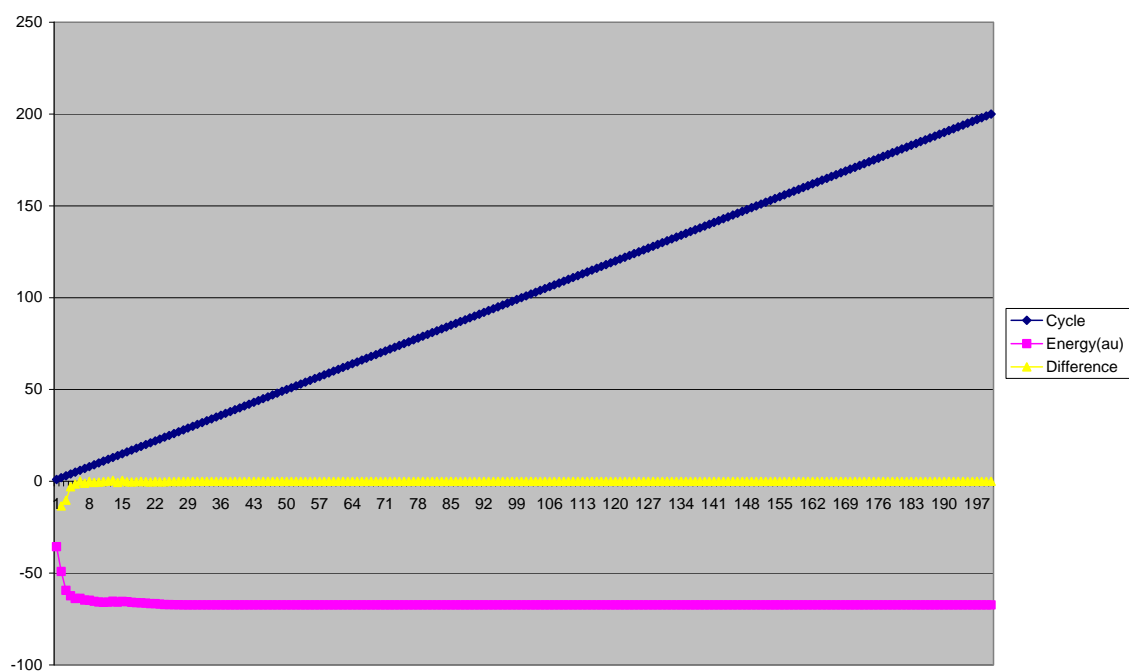


Figure 6.58: Energy vs. Difference for Quick Plot ESP Mapped Density calculation of Asparagine
Glutamine

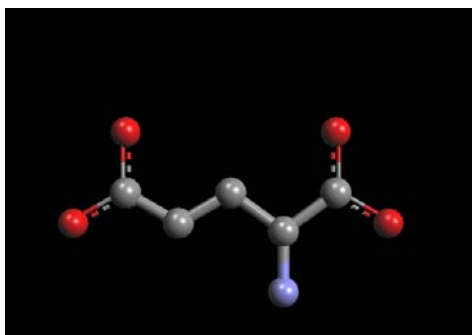


Figure 6.59: Structure of Glutamine

Single Entry Point Calculation: Glutamine

Table 6.25: Single Entry Point Calculation of Glutamine

```

***** Validated Experiment & Chemical System Settings *****
Max. SCF cycles           100
SCF convergence           1.5936e-013 au. for energy

Basis Set
*****
basis functions : 40
shells          : 20
primitives      : 120
    
```

```
Memory for Main Chemical System
Max. number 2-ele. ints. = 4660

Memory Requirements (bytes)
*****

Core          251216
Scratch       27520

Total number of 2-ele integrals 2554

***** SCF *****

Core repulsion    188.429 au
Final SCF Energy = -34.1405108556 au
Final SCF Energy = -21423.5133 kcal/mol

***** Heat of Formation *****
                27536.8764 kcal/mol

Atomic spin densities
*****

   1   C   0.0000
   2   C   0.0006
   3   O   0.0000
   4   C   0.0000
   5   O   0.0006
   6   C   0.0002
   7   C   0.0379
   8   O   0.0001
   9   O   0.0000
  10   N   0.9605

S2 operator
*****
exact          0.750000
calculated     0.750000

Total Elapsed Time 1 sec.
```

Geometry Optimization: Glutamine

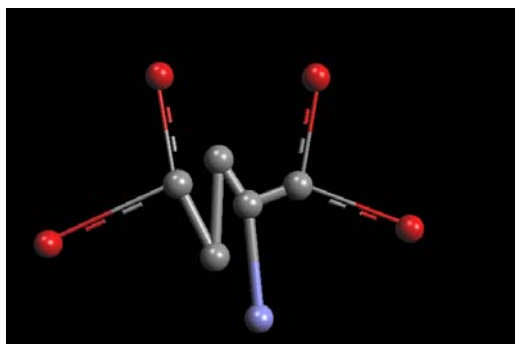


Figure 6.60: Geometry Optimization of Glutamine

Table 6.26: Geometry Optimization calculation of Glutamine

Constructing Chemical System(s)

```
Basis Set
*****

basis functions : 40
shells          : 20
primitives      : 60
```

```

Memory for Main Chemical System
Max. number 2-ele. ints. = 4660

Memory Requirements (bytes)
*****
Core          899000
Scratch       27520

Total number of 2-ele integrals 2770

Final SCF Energy = -17.6840713073 au
Final SCF Energy = -11096.9323 kcal/mol

***** Final Geometry *****

C   15.31615076  -13.95376286   0.00000000   6
C   18.08684663  -13.98614029   0.00000000   6
O   14.89461756  -11.46192091   0.00000000   8
C   16.17329576  -15.62707299   0.00000000   6
O   18.50404616  -11.50082702  -0.00000000   8
C   16.26399478  -13.38874488   0.00000000   6
C   16.93275916  -14.42568040  -0.00000000   6
O   20.66644582  -14.98357984  -0.00000000   8
O   12.32919477  -15.33565085   0.00000000   8
N   17.19304862  -17.05401998   0.00000000   7

Final Geom Energy = -59.9845308085 au
Final Geom Energy = -37640.8953 kcal/mol

***** Heat of Formation *****
11319.6093 kcal/mol

Total Elapsed Time 12 min. 16 sec.

```

QuickPlot HOMO: Glutamine

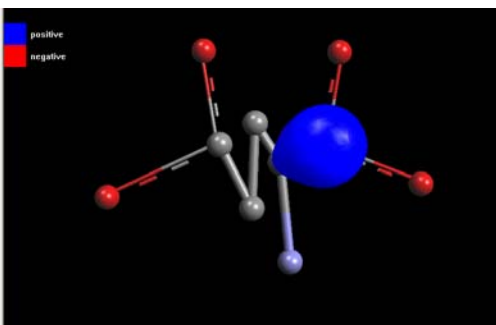


Figure 6.61: QuickPlot HOMO calculation of Glutamine

Table 6.27: QuickPlot HOMO calculation of Glutamine

```

Constructing Chemical System(s)

Basis Set
*****
basis functions : 40
shells          : 20
primitives      : 120

```

```

Memory for Main Chemical System
Max. number 2-ele. ints. = 4660

Memory Requirements (bytes)
*****
Core                1115224
Scratch             27520

***** SCF *****

Core repulsion      159.976 au
Final SCF Energy = -59.1523636153 au
Final SCF Energy = -37118.7021 kcal/mol

S2 operator
*****

exact               0.750000
calculated          0.750001

Total Elapsed Time 4 sec.

```

QuickPlot LUMO: Glutamine

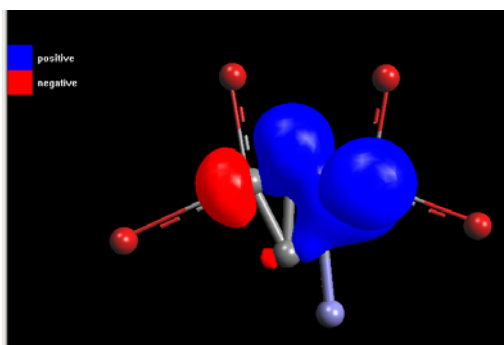


Figure 6.62: QuickPlot LUMO calculation of Glutamine

Table 6.28 QuickPlot LUMO calculation of Glutamine

```

Input Atomic Information
*****

  1  C   15.316151 -13.953763  0.000000
  2  C   18.086847 -13.986140  0.000000
  3  O   14.894618 -11.461921  0.000000
  4  C   16.173296 -15.627073  0.000000
  5  O   18.504046 -11.500827 -0.000000
  6  C   16.263995 -13.388745  0.000000
  7  C   16.932759 -14.425680 -0.000000
  8  O   20.666446 -14.983580 -0.000000
  9  O   12.329195 -15.335651  0.000000
 10  N   17.193049 -17.054020  0.000000

constructing Chemical System(s)

Basis Set
*****

basis functions : 40
shells          : 20
primitives      : 120
Final SCF Energy = -59.1523636153 au
Final SCF Energy = -37118.7021 kcal/mol

```

```
Memory for Main Chemical System
Max. number 2-ele. ints. = 4660

Memory Requirements (bytes)
*****
Core          1115224
Scratch       27520

S2 operator
*****

exact          0.750000
calculated     0.750001

Properties elapsed time 1 sec.
Total Elapsed Time 3 sec.
```

Quick Plot ESP Mapped Density: Glutamine

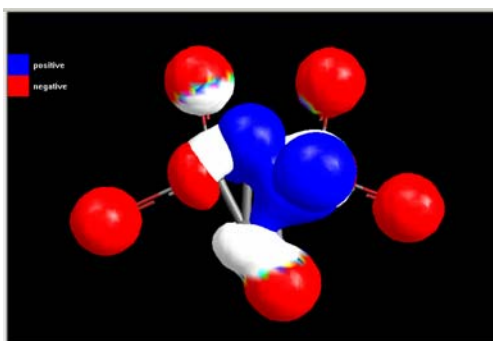


Figure 6.63: Quick Plot ESP Mapped Density of Glutamine

Table 6.29: Quick Plot ESP Mapped Density of Glutamine

```
Constructing Chemical System(s)

Basis Set
*****

basis functions : 40
shells          : 20
primitives      : 120

Memory for Main Chemical System
Max. number 2-ele. ints. = 4660

Memory Requirements (bytes)
*****

Core          1115224
Scratch       27520

**** SCF ****

Core repulsion   159.976 au

Final SCF Energy = -59.1523636153 au
Final SCF Energy = -37118.7021 kcal/mol

SCF elapsed time 2 sec.

**** Heat of Formation ****
11841.6877 kcal/mol
```

6.2.3 Activity No -3

In the second phase of research work experiment has been performed using tools like DAMBE and Jumboss. Under this research work molecular sequence of Nucleotides and Proteins has been analyzed.

6.2.3.1 Sequence Analysis Using Jemboss

Creation of Sequence from Multiple Alignments

In this phase of my research work I have created nucleotide sequence from multiple alignments using *tropomyosin.fasta* file using Jemboss software.

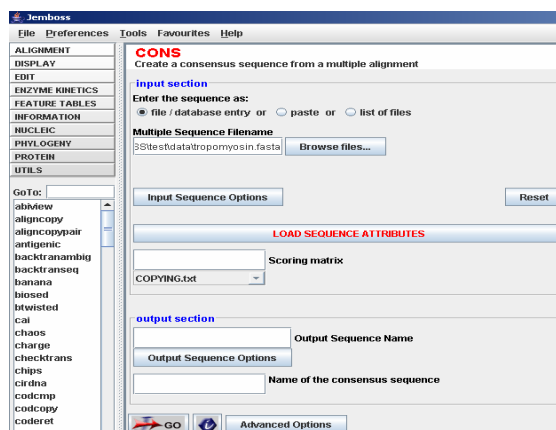


Figure 6.64 : Creation of Nucleotide Sequence from Multiple Alignment using Jemboss

```
>EMBOSS_001
CCGGCCGCCAGCAGCACTAATGTGCTGGAGGCGAACTCACCATATGCTCCGGCACCCC
AAGGTGGGGGGGAGGGGGCGCACAGGAGGCGCAGCGGCTGCAGGAAGAAGAGGGCGAG
AGGGAGGTGATGGAGGGAGGGGCGAGAGCGGGCGGCAAGCAGGAGAGCTAACGGCTGATC
ACGGCGGCGTCAGCGAATGAGAGGAGGGCTGGAACGGCCAGGTGGCGGAGCGAGGACGCG
GAAGCGGAAACTGAGAAGAAGAAGGGGAGAGGGCCGAAGCGTAGTAAGAGAGGCAAAG
AACAAAGAA---GAAG--G--A--GAAG--G--GA-A---A--A-----AA---G--
GAGGA-G-----AG---G--G-A-----C-G-T---C--G-----
-----G--G-A--G-C--G-GG-GG-----AA-G--GA-G-G--A-G--GA-
AA-----A---G--A--A-----T-----G-A-----A-----TG-T
-----G--GA---A---GAA--A-T-----T--A-G---A-----
-----T-CT---G-A---AA---A-----A-A---T---GA--A-----AC-A-
-----A---C--CC-C--GAAC-A--AGA--C---C---C---AG-----G--TCAG
AGCC--AACTAC---ATCAGTAA-CT-C--AA---GACGAG-AT-A-C--CTCC-ACACT
C--GA-CT--GTG-TGGC--CTTCTC-T--ATGAATGA-CT--AGGTGTTGCC-CTG--C
AACG-C-CAA-C---C-G---TGG-----C--T---GAGGTGTATGAAGGCCAG
GTGTCCGGAATGCCCAACGACCAAGCCCCCTGCAAGTGGCTGTGAAGACGCTGCCTGAA
GTGTGC
```

Figure 6.65 : Creation of Nucleotide Sequence from Multiple Alignment

Drawing a Threshold Dot Plot of two Sequences

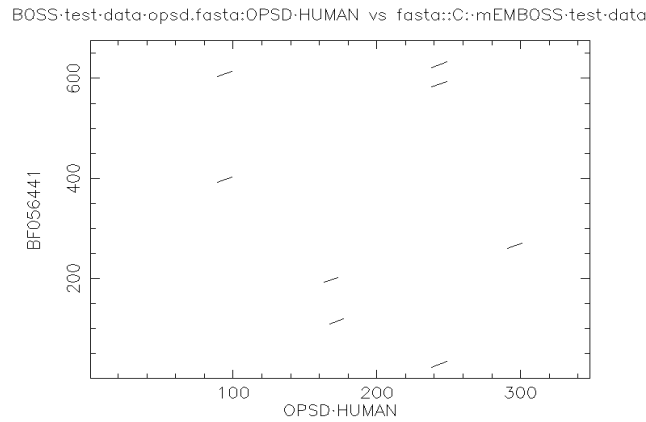


Figure 6.66 : Draw a threshold Dot Plot of two Sequences

Displaying Restriction Enzyme binding site in nucleotide sequence

```

F056441
F056441; 7k05a04.x1 NCI_CGAP_GC6 Homo sapiens cDNA clone
MAGE:3443238 3' similar to SW:TPM4_HUMAN P07226 TROPOMYOSIN,
TBROBLAST NON-MUSCLE TYPE ;, mRNA sequence.

                                     BsrDI
                                     |   CviRI
Tsp4CI                               |   |   TspEI
|   |   CviRI                       |   |   |   MwoI           BssKI
|   |   TfiI                         |   |   |   BstAPI         EcoRII
|   |   HinfI                        |   |   |   |   MaeI       |SecI
\   \   \                             \   \   \   \   \   \   \
acagttgcaagaatcctaaagtgtggattttattccattgcacacaatttgctagtgtatttc
      10          20          30          40          50          60
-----|-----|-----|-----|-----|-----|-----|-----|
tgtcaacgcttcttagatttcacacctaataaaggtaacgtgttaaacgatcacataaag
/   /   /   /   /   /   /   /   /   /   /   /   /   /   /
|   |   CviRI HinfI           BsrDI |   |   |   MaeI
Tsp4CI   TfiI
                                     |   |   TspEI
                                     |   BstAPI
                                     |   MwoI
                                     CviRI

T V A R I * S V D F I P L H N L L V Y F
Q L Q E S K V W I L F H C T I C * C I S
S C K N L K C G F Y S I A Q F A S V F P
-----|-----|-----|-----|-----|-----|-----|
V T A L I * L T S K I G N C L K S T Y K
X L Q L F R F H P N * E M A C N A L T N
C N C S D L T H I K N W Q V I Q * H I E

                                     SmlI
ScrFI                               AflIII           AluBI
BseBI                               |MseI           CviJI
\                                   \ \             \
ctgggtagtggtgctgaataaataggaataaatgctacttaaggaaaaataagagag
      70          80          90          100         110         120
-----|-----|-----|-----|-----|-----|-----|
gaccatcacaccagacttattttatccttatttacgatgaattcctttttattctctc

```

Figure 6.67: Display restriction Enzyme binding site in nucleotide sequence

Calculation of Codon adaptation Index

Sequence: BF056441 CAI: 0.208
Sequence: BE848719 CAI: 0.223
Sequence: BF022813 CAI: 0.139
Sequence: BF452255 CAI: 0.136
Sequence: BG089808 CAI: 0.161
Sequence: BG147728 CAI: 0.128
Sequence: BI817778 CAI: 0.107
Sequence: AF186109 CAI: 0.184
Sequence: AF186110 CAI: 0.171
Sequence: AF310722 CAI: 0.176
Sequence: AF362886 CAI: 0.189
Sequence: AF362887 CAI: 0.191
Sequence: AF087679 CAI: 0.176

Figure 6.68 : Calculation of Codon adaptation Index

Calculation of isochores in DNA Sequence

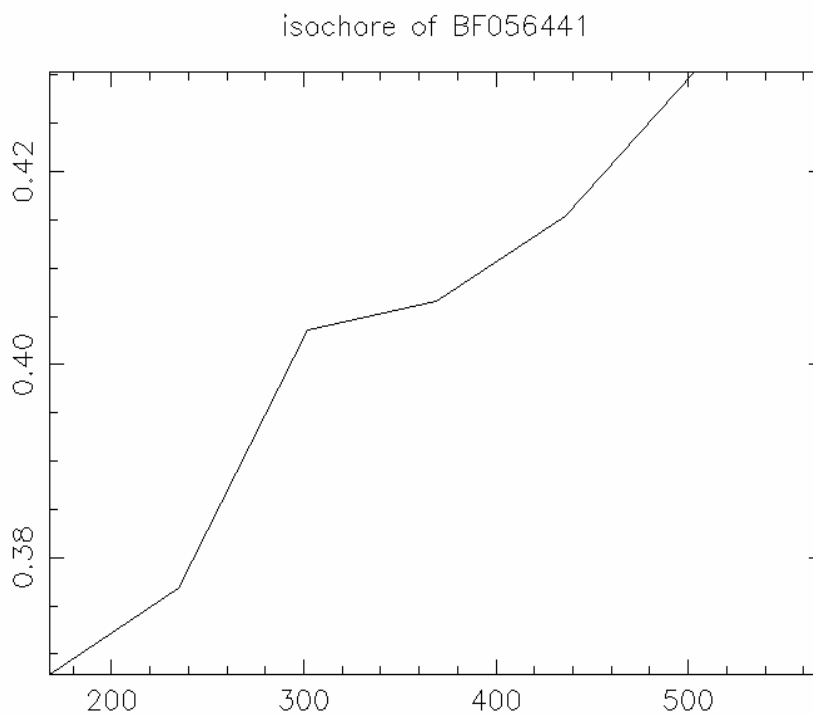


Figure 6.69 : Calculation of isochores in DNA Sequence

Finding of siRNA duplexes in mRNA

Table 6.30 : Finding of siRNA duplexes in mRNA

```

#-----
#
# Sequence: BF056441   from: 1   to: 675
# HitCount: 130
#
# No CDS region was found in the feature table.
# No CDS region was indicated by setting -sbegin.
# There will therefore be no penalty for siRNAs found in the first 100 bases.
#
#=====

```

Start	End	Strand	Score	GC%	Sense_siRNA	Antisense_siRNA
170	192	+	9.000	50.0	CUGGUGUCUCAAAGCUUCUCdTdT	GAGAAGCUUUGAGCACCAGdTdT
277	299	+	9.000	50.0	GAGAAAGAGAAAAGACCCACGdTdT	CGUGGGUCUUCUCUUCUCdTdT
386	408	+	9.000	50.0	GCCCACGUUCUCUUCUUGdTdT	CAAAGAAGAGAACGUGGGdTdT
114	136	+	8.000	45.0	GAGAGCUGAAAAAGCUGGdTdT	ACCAGCUUUUCAGCUCUCdTdT
282	304	+	8.000	55.0	GAGAAAGACCCACGGAGCUCdTdT	AGCUCGUGGGUCUUCUCdTdT
319	341	+	8.000	40.0	CAAGACUCUUCUGUUUGdTdT	GCAAAACAGAGAGUCUUGdTdT
415	437	+	8.000	45.0	GUUUCUCUCCAGGUCaucdTdT	GAUGACCUGGAAGAGAAAcdTdT
457	479	+	8.000	45.0	CCGUUCUCUCUGCAAAUUCdTdT	GAAUUUGCAGAGAGAACGGdTdT
472	494	+	8.000	55.0	AUUCAGCACGGGUCcAGcTdT	GCUGAGACCCGUGCGAAUdTdT
597	619	+	8.000	45.0	UUCUGAGUUCUUCcAGGdTdT	CCUGGAGAAGAACUCAGAAdTdT
118	140	+	7.000	50.0	GCUGAAAAAGCUGGUGCCAdTdT	UGGCACCAGCUUUUCAGCdTdT
124	146	+	7.000	40.0	AAAGCUGGUGCCAuuUGAAdTdT	UUCAAAUGGCACCAGCUUdTdT
125	147	+	7.000	40.0	AAGCUGGUGCCAuuUGAAAdTdT	UUUCAAAUGGCACCAGCUdTdT
126	148	+	7.000	40.0	AGCUGGUGCCAuuUGAAAAdTdT	UUUCAAAUGGCACCAGCUdTdT
127	149	+	7.000	40.0	GCUGGUGCCAuuUGAAAAAdTdT	UUUUUCAAAUGGCACCAGCdTdT
160	182	+	7.000	40.0	UGAGAUUUAAACUGGUGCUCdTdT	GAGCACCAGUUAAAUUCAdTdT
220	242	+	7.000	40.0	UUUGCUUGACAUUUCcAGCdTdT	GCUGGAAAUGUCAAGCAAdTdT
241	263	+	7.000	40.0	AGCGAAGAUGGCCAAUAAcAdTdT	UGUUUUUGCCAUCUUCGCUdTdT
242	264	+	7.000	40.0	GCGAAGAUGGCCAAUAAcAdTdT	UUGUUUUUGCCAUCUUCGCUdTdT
287	309	+	7.000	60.0	AGACCCACGGAGCUCcAGAdTdT	UCUGGAGCUCCGUGGGUCdTdT
352	374	+	7.000	40.0	GUUCGUUUAGUGUCUGAUcTdT	GAUCAGACACUAAACGAACdTdT

Calculation of fractional GC Content of Nucleic Acid sequences

Table 6.31: Calculation of fractional GC Content of Nucleic Acid sequences

#Sequence	GC content
BF056441	0.40
BE848719	0.46
BF022813	0.59
BF452255	0.57
BG089808	0.54
BG147728	0.55
BI817778	0.62
AF186109	0.55
AF186110	0.54
AF310722	0.56
AF362886	0.49
AF362887	0.50
AF087679	0.52

6.2.3.2 Nucleotide Sequence Using DAMBE

Under this research work nucleotide sequence has been created from multiple alignments through *tropomyosin.fasta* file using DAMBE software.

Running FASTA algorithm to align locally two sequences

Table 6.32: Running FASTA algorithm to align locally two sequences

```
Query: ACCGCGATGACGAATA
Target: GAATACGACTGACGATGGA
N-tuple: 1
```

```
Result of the local alignment
Number of matched words of length 1: 7
Query: ACCGCGATGACGAATA
Target: GAATACGACTGACGATGGA
```

Computational details:

Part I. Hash table of query

0	0	6	9	12	13	15
1	1	2	4	10		
2	3	5	8	11		
3	7	14				

Part II. Target seq. table

0	-3	-5	-8	-11		
1	1	-5	-8	-11	-12	-14
2	2	-4	-7	-10	-11	-13
3	-4	-11				
4	4	-2	-5	-8	-9	-11
5	4	3	1	-5		
6	3	1	-2	-5		
7	7	1	-2	-5	-6	-8
8	7	6	4	-2		
9	2	-5				
10	7	5	2	-1		
11	11	5	2	-1	-2	-4
12	11	10	8	2		
13	10	8	5	2		
14	14	8	5	2	1	-1
15	8	1				
16	13	11	8	5		
17	14	12	9	6		
18	18	12	9	6	5	3

Part III. Frequency table

-15	0
-14	1
-13	1
-12	1
-11	5
-10	1
-9	1
-8	4
-7	1
-6	1
-5	7
-4	3
-3	1
-2	5
-1	3
0	0
1	6

2 7
 3 3
 4 3
 5 6
 6 3
 7 3

Nucleotide Frequency Calculation

Table 6.33: Nucleotide Frequency Calculation

```

Output from sequences in file C:\mEMBOSS\test\data\tropomyosin.fasta on Saturday, May 23, 2009

Part Ia. Nucleotide frequencies.
=====
SeqName          A      C      G      T  Sum(ACGT)  X2  ProbX2    PA    PC    PG    PT
=====
emb1:BF056441    187   141   129   218     675  30.274  0.0000  0.2770  0.2089  0.1911  0.3230
=====

Note: The Chisquare tests above are for testing if Pa = Pc = Pg = Pt = 0.25, with 3 degree of freedom.

Part Ib. The test of heterogeneity of nucleotide frequencies among OTUs was cancelled by user request.
    
```

Relative CpG, TpG and CpA abundance and GC%

Table 6.34 :Relative CpG, TpG and CpA abundance and GC%

Relative CpG, TpG and CpA abundance and GC%

```

=====
SeqName          RA (CpG)      RA (TpG+CpA)  GC%
=====
emb1:BF056441    0.4688        1.7385         0.4000
emb1:BE848719    0.4070        1.7275         0.4642
emb1:BF022813    1.7365        3.0724         0.5919
emb1:BF452255    1.2272        2.3809         0.5706
emb1:BG089808    0.9545        1.8657         0.5380
emb1:BG147728    1.0379        2.3310         0.5506
emb1:BI817778    1.5579        2.3169         0.6239
emb1:AF186109    0.9357        1.6668         0.5503
emb1:AF186110    0.5796        1.4454         0.5413
emb1:AF310722    0.6665        1.3112         0.5621
emb1:AF362886    1.4989        4.1446         0.4903
emb1:AF362887    1.0261        2.7307         0.5023
emb1:AF087679    0.7944        1.2812         0.5158
=====
    
```

RA is the odds-ratio measure, i.e., $F(XY)/[F(X)*F(Y)]$ for quantifying the relative abundance of dinucleotides (for nucleotide sequences) or di-aa (for protein sequences).

For nucleotide sequences, F(X) and F(Y) are the frequencies of X and Y, respectively, and F(XY) is the frequency of dinucleotide XY.

For amino acid sequences, they are amino acid and di-aa frequencies.

XY is considered high (or low) when $RA > 1.25$ (or < 0.78). See Hollander, M., and D. A. Wolfe. 1973. Nonparametric statistical methods. Wiley, New York.

For an application of RA, see Karlin, S., W. Doerfler, and L. R. Cardon. 1994. Why is CpG suppressed in the genomes of virtually all small eukaryotic viruses but not in those of large eukaryotic viruses? *J Virol* 68:2889-2897.

Part II: Distribution of CpG in individual sequences:

Summary statistics of the inter-CpG distances

SeqName	N	Mean	STD	CV	Skew	Kurt
embl:BF056441	7	67.4286	54.7809	0.8124	0.8692	0.5025
embl:BE848719	9	60.7778	51.2732	0.8436	0.9363	0.9433
embl:BF022813	14	18.0417	30.2834	1.6785	3.2732	11.6885
embl:BF452255	15	19.4583	30.9234	1.5892	2.9717	9.8669
embl:BG089808	15	17.9630	26.4175	1.4707	2.7305	7.9922
embl:BG147728	12	21.2000	33.9017	1.5991	2.7200	7.8178
embl:BI817778	21	16.5357	15.1645	0.9171	1.8831	4.0354
embl:AF186109	20	21.1212	33.3164	1.5774	3.3791	13.5583
embl:AF186110	25	29.7667	34.7768	1.1683	2.7001	9.3338
embl:AF310722	28	21.7778	30.5977	1.4050	3.1991	13.1335
embl:AF362886	7	43.4286	52.4622	1.2080	1.6513	2.6721
embl:AF362887	10	41.4000	50.7701	1.2263	2.2652	5.7178
embl:AF087679	21	23.4857	26.0151	1.1077	1.6260	2.0147

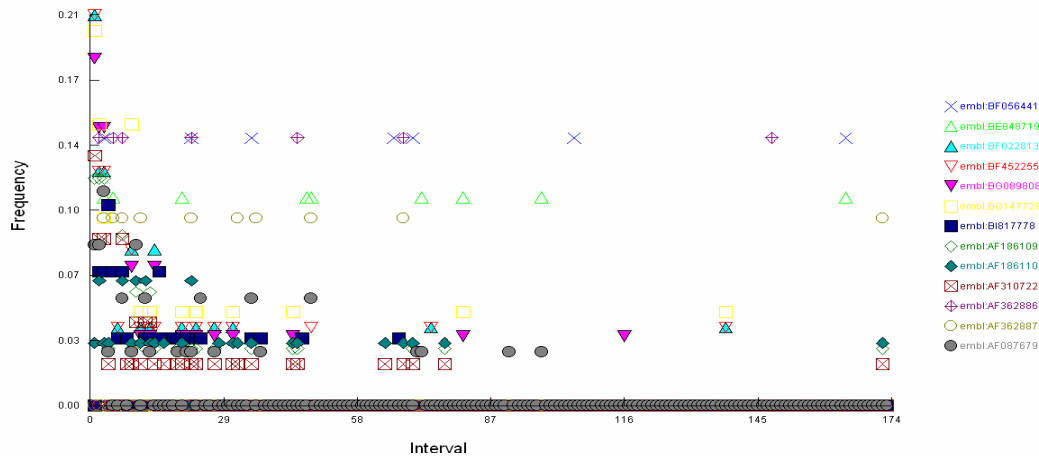


Figure 6.72: Graph of Relative CpG, TpG and CpA abundance and GC%

Di-nucleotide Substitution Pattern

Table 6.35: Di-nucleotide Substitution Pattern

DiNuc	AA	AC	AG	AT	CA	CC	CG	CT	GA	GC	GG	GT	TA	TC	TG	TT
AA	68.000	6.000	35.000	8.000	17.000	8.000	0.000	14.000	30.000	16.000	12.000	2.000	3.000	4.000	6.000	7.000
AC	6.000	34.000	17.000	11.000	5.000	4.000	3.000	0.000	2.000	9.000	1.000	2.000	3.000	5.000	0.000	4.000
AG	35.000	17.000	102.000	12.000	2.000	0.000	7.000	5.000	6.000	6.000	11.000	7.000	11.000	4.000	30.000	10.000
AT	8.000	11.000	12.000	32.000	9.000	2.000	2.000	10.000	4.000	7.000	5.000	11.000	0.000	6.000	2.000	5.000
CA	17.000	5.000	2.000	9.000	82.000	7.000	7.000	10.000	15.000	1.000	5.000	8.000	8.000	6.000	5.000	3.000
CC	8.000	4.000	0.000	2.000	7.000	48.000	4.000	8.000	0.000	9.000	8.000	4.000	4.000	8.000	1.000	1.000
CG	0.000	3.000	7.000	2.000	7.000	4.000	8.000	6.000	4.000	0.000	3.000	4.000	0.000	0.000	11.000	4.000
CT	14.000	0.000	5.000	10.000	10.000	8.000	6.000	110.000	4.000	2.000	2.000	8.000	0.000	2.000	4.000	27.000
GA	30.000	2.000	6.000	4.000	15.000	0.000	4.000	4.000	68.000	14.000	18.000	14.000	10.000	6.000	6.000	12.000
GC	16.000	9.000	6.000	7.000	1.000	9.000	0.000	2.000	14.000	54.000	14.000	13.000	6.000	13.000	5.000	5.000
GG	12.000	1.000	11.000	5.000	5.000	8.000	3.000	2.000	18.000	14.000	32.000	5.000	3.000	3.000	8.000	4.000
GT	2.000	2.000	7.000	11.000	8.000	4.000	4.000	8.000	14.000	13.000	5.000	36.000	2.000	2.000	2.000	3.000
TA	3.000	3.000	11.000	0.000	8.000	4.000	0.000	0.000	10.000	6.000	3.000	2.000	22.000	5.000	7.000	7.000
TC	4.000	5.000	4.000	6.000	6.000	8.000	0.000	2.000	6.000	13.000	3.000	2.000	5.000	104.000	11.000	16.000
TG	6.000	0.000	30.000	2.000	5.000	1.000	11.000	4.000	6.000	5.000	8.000	2.000	7.000	11.000	66.000	17.000
TT	7.000	4.000	10.000	5.000	3.000	1.000	4.000	27.000	12.000	5.000	4.000	3.000	7.000	16.000	17.000	104.000

Creation of Phylogenetics Tree of Nucleotide sequence based on Distance Method

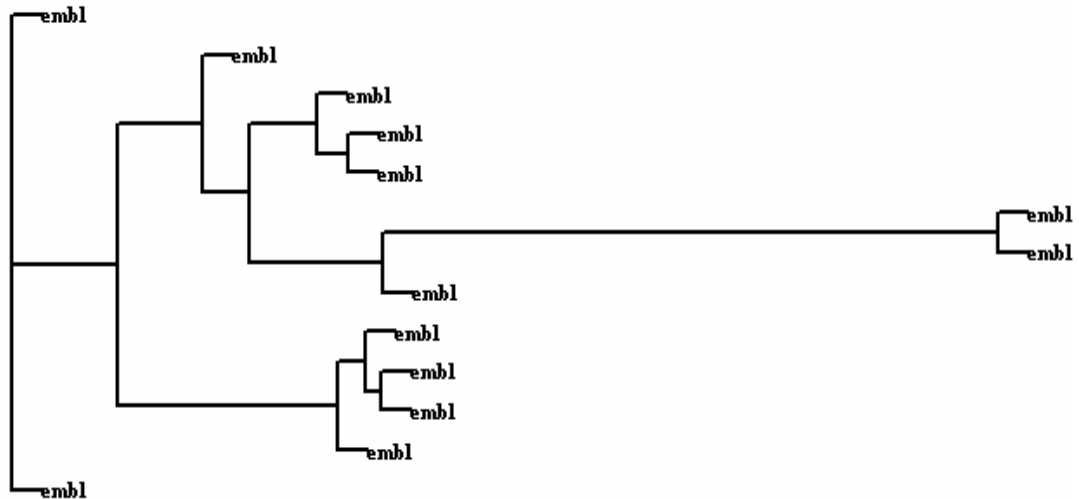


Figure 6.75: Creation of Phylogenetics Tree

Phylogenetics with DAMBE

```

Phylogenetics with DAMBE

I. Distance options:
Genetic distance: MLCompositeTN93
Distance matrix:

13
embl:AF362886
embl:BI817778 0.10926
embl:BF056441 0.16187 0.34219
embl:BE848719 0.15956 0.43393 0.06991
embl:AF362887 0.06521 0.22293 0.15434 0.14749
embl:BF022813 0.03321 0.17257 0.23391 0.23395 0.10590
embl:BG147728 0.10176 0.16636 0.23323 0.20145 0.06607 0.01352
embl:BF452255 0.08620 0.16987 0.22369 0.22928 0.09530 0.00000 0.01066
embl:BG089808 0.13301 0.16920 0.24180 0.25178 0.09866 0.00729 0.01565 0.00592
embl:AF186109 0.05730 0.13858 0.19510 0.17294 0.03262 0.10667 0.09176 0.09759 0.10287
embl:AF087679 0.09643 0.16259 0.20991 0.20248 0.07973 0.08898 0.09421 0.08630 0.09957 0.04254
embl:AF186110 0.05041 0.12009 0.18171 0.23756 0.02972 0.10873 0.08484 0.09738 0.10350 0.00315 0.07633
embl:AF310722 0.04889 0.13765 0.18796 0.23428 0.02942 0.10575 0.09176 0.09693 0.10231 0.00000 0.06836 0.00229

Composite lnL maximized after 26 iterations
k1 = 15.6653
k2 = 27.4965
lnL = -31.78876

II. Tree options
Tree-building method: FastME
Outgroup: embl:AF362886
Branch evaluation: Balanced
Initial tree: GME
Branch swapping: Yes.

Best tree:
(embl:BI817778:0.105152, ((embl:AF087679:0.034556, ((embl:AF186109:-0.005268, (embl:AF186110:0.002106, embl:AF310722:0.000185):0.005699):0.0128:
    
```

Figure 6.76: Phylogenetics with DAMBE

6.2.3.3 Protein Sequence Using Jembooss

In this phase of research work protein structure has been analyzed using *tropomyosin.fasta* file through Jembooss software.

Prediction of Protein Secondary structure using GOR Method

```
#####
# Program: garnier
# Rundate: Sun 23 May 2009 16:53:17
# Commandline: garnier
# -sequence C:\mEMBOSS\test\data\tropomyosin.fasta
# -idc 0
# -rformat tagseq
# -auto
# Report_format: tagseq
# Report_file: bf056441.garnier
#####

#=====
#
# Sequence: BF056441      from: 1   to: 675
# HitCount: 125
#
# DCH = 0, DCS = 0
#
# Please cite:
# Garnier, Osguthorpe and Robson (1978) J. Mol. Biol. 120:97-120
#
#=====

      . 10   . 20   . 30   . 40   . 50
      acagttgcaagaatctaaagtgtggattttattccattgcacaatttgct
helix          HHHHH
sheet E      E      EEEE      EE      EEEE
turns TTTT TT      T      TTTTTTTTTT      TTTT
coil   C      C CCCCCC
      . 60   . 70   . 80   . 90   . 100
      agtgtatttcctgggtagtgtggtgctgaataaataggaataaatgctac
helix          HHHHHHHHHHHHHHHH
sheet          EEEEE      EE      EE E      EEE
turns TTTT      TTT      TTTT      TTT
coil C      CCC      CC C
      . 110  . 120  . 130  . 140  . 150
      ttaagggaaaaataagagagactgaaaaagctggtgccatttgaaaaaaaa
```

Figure 6.77: Prediction of Protein Secondary structure using GOR Method

Generation of residue / base frequency plot

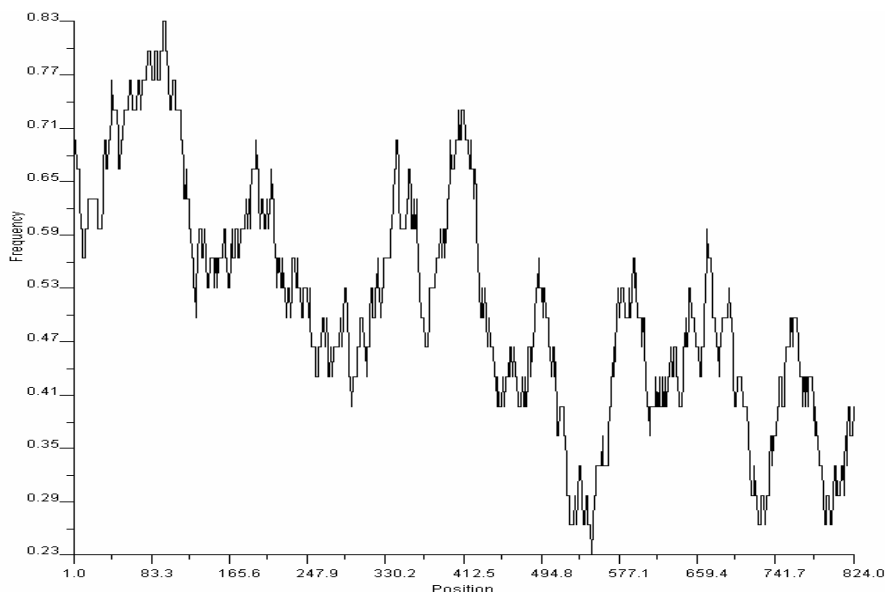


Figure 6.81 : Generation of residue / base frequency plot

Calculation of isoelectric points of protein

Table 6.37: Calculation of isoelectric points of protein

IEP of BF056441 from 1 to 675
 Isoelectric Point = 4.9650

pH	Bound	Charge
1.00	143.00	1.00
1.50	142.99	0.99
2.00	142.98	0.98
2.50	142.93	0.93
3.00	142.80	0.80
3.50	142.56	0.56
4.00	142.28	0.28
4.50	142.10	0.10
5.00	141.99	-0.01
5.50	141.87	-0.13
6.00	141.56	-0.44
6.50	140.60	-1.40
7.00	137.65	-4.35
7.50	129.11	-12.89
8.00	107.92	-34.08
8.50	71.06	-70.94
9.00	34.16	-107.84
9.50	12.93	-129.07
10.00	4.36	-137.64
10.50	1.41	-140.59
11.00	0.45	-141.55
11.50	0.14	-141.86
12.00	0.04	-141.96
12.50	0.01	-141.99
13.00	0.00	-142.00
13.50	0.00	-142.00
14.00	0.00	-142.00

IEP of BE848719 from 1 to 698
 Isoelectric Point = 4.8956

pH	Bound	Charge
1.00	195.00	1.00
1.50	194.99	0.99
2.00	194.98	0.98
2.50	194.93	0.93

Isoelectric Point Plot Charge vs PH

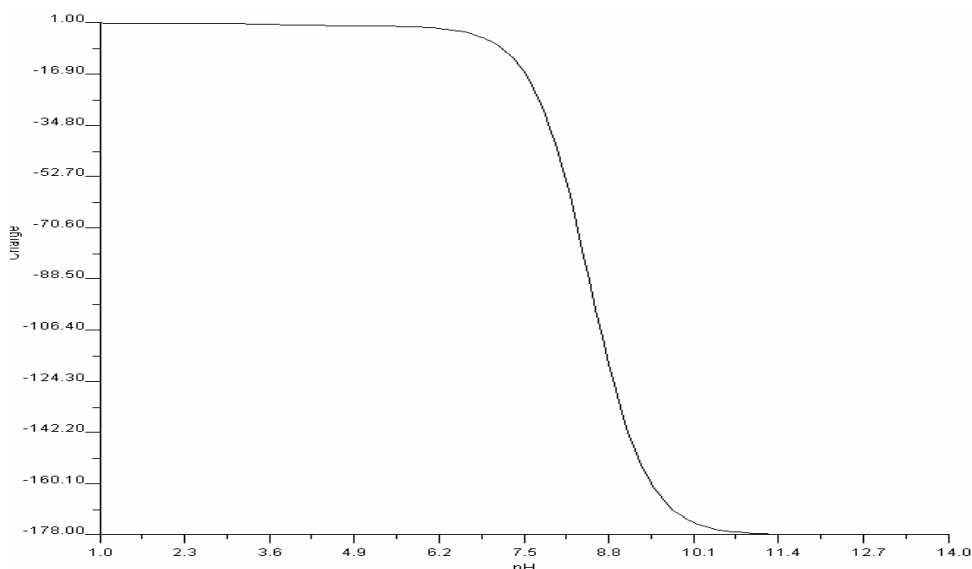


Figure 6.82: Isoelectric Point Plot Charge vs PH

Calculation of Statistics of protein properties

Table 6.38: Calculation of Statistics of protein properties

PEPSTATS of BF056441 from 1 to 675

Molecular weight = 57252.78 Residues = 675
 Average Residue Weight = 84.819 Charge = 0.0
 Isoelectric Point = 4.9650
 A280 Molar Extinction Coefficient = 0
 A280 Extinction Coefficient 1mg/ml = 0.00
 Improbability of expression in inclusion bodies = 0.586

Residue	Number	Mole%	DayhoffStat
A = Ala	187	27.704	3.221
B = Asx	0	0.000	0.000
C = Cys	141	20.889	7.203
D = Asp	0	0.000	0.000
E = Glu	0	0.000	0.000
F = Phe	0	0.000	0.000
G = Gly	129	19.111	2.275
H = His	0	0.000	0.000
I = Ile	0	0.000	0.000
J = ---	0	0.000	0.000
K = Lys	0	0.000	0.000
L = Leu	0	0.000	0.000
M = Met	0	0.000	0.000
N = Asn	0	0.000	0.000
O = ---	0	0.000	0.000
P = Pro	0	0.000	0.000
Q = Gln	0	0.000	0.000
R = Arg	0	0.000	0.000
S = Ser	0	0.000	0.000
T = Thr	218	32.296	5.294
U = ---	0	0.000	0.000
V = Val	0	0.000	0.000

```

W = Trp      0      0.000      0.000
X = Xaa     0      0.000      0.000
Y = Tyr     0      0.000      0.000
Z = Glx     0      0.000      0.000
    
```

Property	Residues	Number	Mole%
Tiny	(A+C+G+S+T)	675	100.000
Small	(A+B+C+D+G+N+P+S+T+V)	675	100.000
Aliphatic	(A+I+L+V)	187	27.704
Aromatic	(F+H+W+Y)	0	0.000
Non-polar	(A+C+F+G+I+L+M+P+V+W+Y)	457	67.704
Polar	(D+E+H+K+N+Q+R+S+T+Z)	218	32.296
Charged	(B+D+E+H+K+R+Z)	0	0.000
Basic	(H+K+R)	0	0.000
Acidic	(B+D+E+Z)	0	0.000

Plot of amino acid properties of proteins in parallel

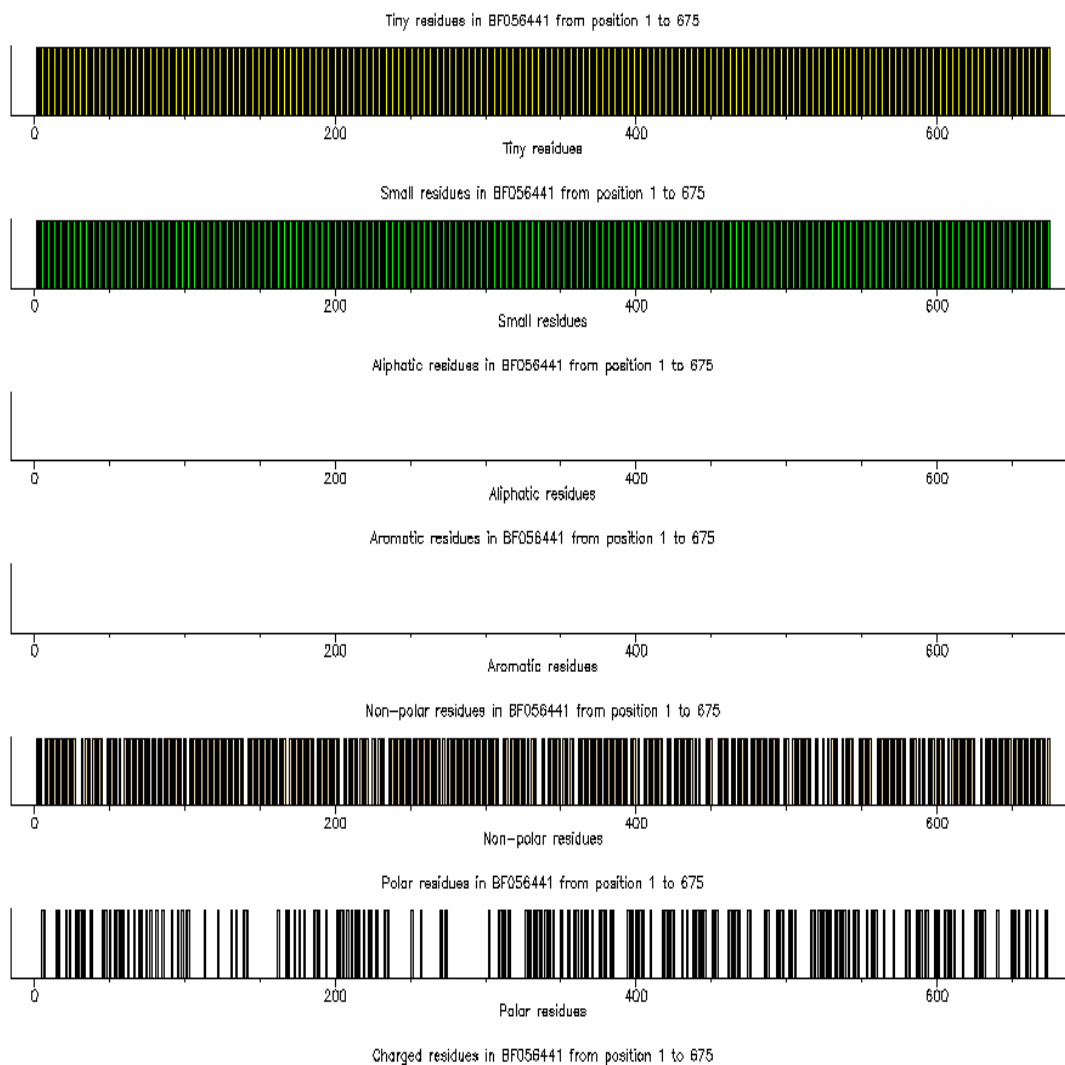


Figure 6.83 : Plot of Histogram of general properties

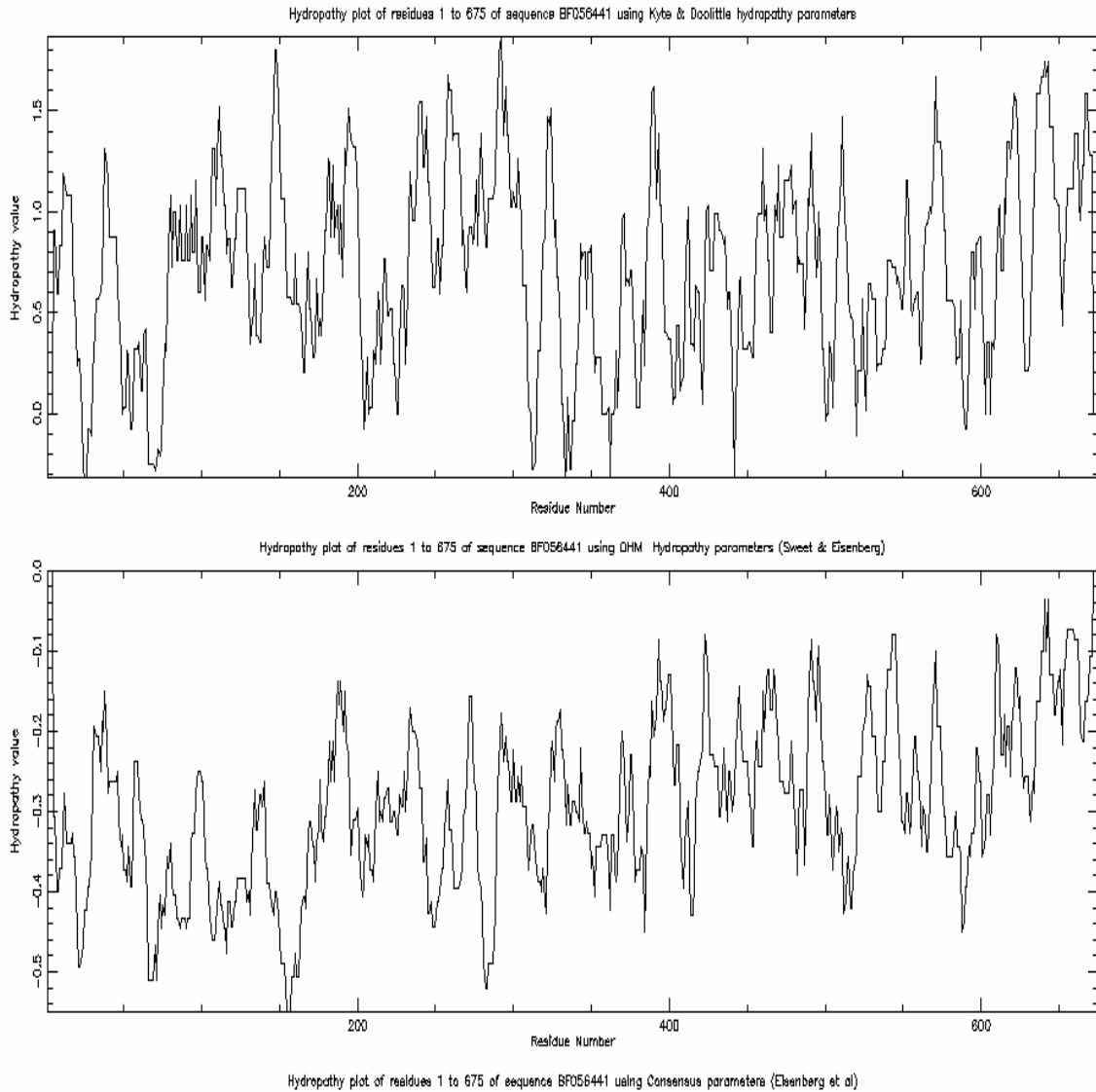


Figure 6.84: Plot graph of hydropathy

WaterMan- Eggert Local Alignment of two Protein Sequences

```
#####
# Program: matcher
# Rundate: Sun 23 May 2009 20:40:49
# Commandline: matcher
#   -asequence C:\MEMBOSS\test\data\tropomyosin.fasta
#   -sprotein1
#   -bsequence C:\MEMBOSS\test\data\opds.fasta
#   -sprotein2
#   -alternatives 1
#   -gapopen 0
#   -gapextend 0
#   -aformat markx0
#   -auto
# Align_format: markx0
# Report_file: bf056441.matcher
#####
```


Chapter 6: Conformational Study of Substance through Different Tools

```
# Aligned_sequences: 2
# 1: BF056441
# 2: OPSD_HUMAN
# Matrix: EBLOSUM62
# Gap_penalty: 0
# Extend_penalty: 0
#
# Length: 624
# Identity:      88/624 (14.1%)
# Similarity:   105/624 (16.8%)
# Gaps:         519/624 (83.2%)
# Score: 487
#
#=====
          10          20
BF0564 GT-TGCAAG-----A-ATCTAAAG---T-----GTGGA-----T
          : :          : :          : :          : :
OPSD_H GTE-----GPNFYVPFSNAT-----GVVRSPFYEPQYYL-----AEPWQF-
          10          20          30

          30          40          50
BF0564 TTTATTCCATTGCA--CAA-----TTTG-----CT---AGT-----
          : :          : :          : :          : :
OPSD_H -----SML-AAYMFLLIIVL---GFPINFL-TLYV--TVQHKK
          40          50          60

          60          70          80
BF0564 --GT-----A-TTCCTGGGTA-----GTG-TGGTGCTGAAT--A
          : :          : :          : :          : :
OPSD_H LR-TPLNYILLNLAV-----ADLFMVLG-GFT--S-----TLY-
          70          80          90

          90          100          110
BF0564 AATA--G---G-AATAAATGCTAC---TTAAG--GAAAAAAT-AAGAG-
          : :          : :          : :          : :
OPSD_H --TSLHGYFVFGP-----TG---CNLE---GFF-----A-TL--G-GE
          100          110          120

          120          130          140          150
BF0564 -A--GCT---GAAA-----AAGCTGGTGC---CATTGAAAAA-----
          : :          : :          : :          : :
OPSD_H IALW--SLVVL---AIERYVVV-----CKPM-----SN
          130          140

          160          170          180
BF0564 ---AAG---GGA--AG-GA-AT---GA-GATTAACTGGTGCTCAA---A
          : :          : :          : :          : :
OPSD_H FRF--GENH--AIM-GV-AF-TWVM-AL-----A-----CAAPPLA
          150          160          170

          190          200          210
BF0564 G-CTTCT-----CCG--ATACAAAATATTGGTCATG---T-----
          : :          : :          : :          : :
OPSD_H GW----SRYIPE--GLQ---C-----SC--GIDYYTLKPEVNNE
          180          190          200

          220          230
BF0564 A-----T-----TC-ATAATTG---CT---TGACATTTCC
          : :          : :          : :          : :
OPSD_H SFVIYMFVVHFTIPMIIFF-CY-----GQLVF-TVKE--A-A-----
          210          220          230

          240          250          260          270
BF0564 A---GCAAAGCGAAGATGGCAAT--AACAAAAGGA---ACTTCT---
          : :          : :          : :          : :
OPSD_H AQQQE--SA-----T-----TQK-----AEKEV-----TRMVI
          240          250

          280          290          300
BF0564 ---TA---C---AAGAGAAGAGAAAGACCCA-CGGA---GCT--CCA
          : :          : :          : :          : :
OPSD_H IMVI-AFLICWVPY-----A---SV---AFYIF--THQ---
          260          270

          310          320
BF0564 GA--G---TTTCTGT--TGGA--A-CAAGA-----C--
          : :          : :          : :          : :
OPSD_H GSNFGPIFM-----TIP---AFFAK-SA-AIYNPVIYIMMNKQFRNCML
          280          290          300          310
```

```

          330          340          350          360
BF0564 TCTTCTGTT-TTGCTTATATACAGTTAAG----TTCG---TTTAGTGTCT
          :  :  :  :  :  :  :  :  :  :  :  :  :  :  :  :  :  :
OPSD_H -----TTI---C-----C-----GKNPL---GDDE---A---S---
          320                                330

          370          380
BF0564 GAT-CCA-GT-GTCT--GA-TGTA
          :  :  :  :  :  :  :  :  :  :  :  :  :  :  :  :  :
OPSD_H -ATV--SK-TE-T-SQV-AP---A
          340
    
```

Figure 6.85: WaterMan- Eggert Local Alignment of two Protein Sequences

Needleman-Wunsch Global Alignment of two sequences

```

#####
# Program: needle
# Rundate: Sun 23 May 2009 20:46:00
# Commandline: needle
#   -asequence C:\MEMBOSS\test\data\tropomyosin.fasta
#   -sprotein1
#   -bsequence C:\MEMBOSS\test\data\opds.fasta
#   -sprotein2
#   -gapopen 0.0
#   -gapextend 0.0
#   -brief
#   -aformat srspair
#   -auto
# Align format: srspair
# Report file: bf056441.needle
#####

#=====
#
# Aligned sequences: 2
# 1: BF056441
# 2: OPSD_HUMAN
# Matrix: EBLOSUM62
# Gap_penalty: 0.0
# Extend_penalty: 0.0
#
# Length: 894
# Identity:      88/894 ( 9.8%)
# Similarity:   105/894 (11.7%)
# Gaps:         765/894 (85.6%)
# Score: 486.0
#
#
#=====

BF056441      1 acagttgcaagaatctaagtgtggatTTTtattccattgcacaatttgcT      50
OPSD_HUMAN    0 -----

BF056441     51 agtgtatttctctgggtagtgtggtgctgaataaataggaataaatgctac    100
OPSD_HUMAN    0 -----

BF056441    101 ttaagaaaaaataagagagctgaaaaagctggtgccatttgaaaaaaaa    150
OPSD_HUMAN    0 -----

BF056441    151 aaggaaggaatgagatttaactggtgctcaaagcttctccgatacaaaa    200
OPSD_HUMAN    0 -----

BF056441    201 tatttggatcatgtattcataatttgcTtgacatttccagcaaagcgaaga    250
OPSD_HUMAN    0 -----

BF056441    251 tggcaataacaaaaggaacttcttacaagagaagagaagaccacgga-    299
OPSD_HUMAN    1 -----MNG-----T-----E      5
    
```

Chapter 6: Conformational Study of Substance through Different Tools

```

BF056441      300 gctcc-----agagtttctgttgg--a-----ac-----a--a      323
      |   |   |   |   |   |   |   |   |   |   |   |   |   |
OPSD_HUMAN    6  G---PNFYVPPFSNA-----T-GVVRSPFEYPOYYLA-EPWQFSMLA      41

BF056441      324 gactcttctgttt-----t-gc--t--t--ata-----tac-----      347
      |   |   |   |   |   |   |   |   |   |   |   |   |   |
OPSD_HUMAN   42  -A-----YMFLLIVLG-FPINFLTLYVTVQHKKLRT--PLNYIL      76

BF056441      348 ---agtta---a-gttcgtt-tagtg--tct--gatcc--a-g-tgtct      380
      |   |   |   |   |   |   |   |   |   |   |   |   |   |
OPSD_HUMAN   77  LNLA--VADLFMVLG--G--FTS-T-LYT-SLHG----YFVFGPTG-C-      110

BF056441      381 gatgtaa---gccc--acgttctcttcttggcctg-ggca--agtttct      422
      |   |   |   |   |   |   |   |   |   |   |   |   |   |
OPSD_HUMAN  111  -----NLEG---FFA-----T-LGG--EIA-----      124

BF056441      423 ct--tcc---aggtc-----atcaa--tt---gtcttttcc---agtt      452
      |   |   |   |   |   |   |   |   |   |   |   |   |   |
OPSD_HUMAN  125  --LWS--LVVLA---IERVVVC--KPMSNFRFG-----ENHA---      153

BF056441      453 tt--gcaaccgttctctc-tgca-a-attc-agcacgggtctcagcctct      496
      |   |   |   |   |   |   |   |   |   |   |   |   |   |
OPSD_HUMAN  154  --IMG-VA-----FT---WVMA--LA-CA-----A-----      169

BF056441      497 ttc---agtt-t-----gt--cagacagaagttaatt---tc-----      525
      |   |   |   |   |   |   |   |   |   |   |   |   |   |
OPSD_HUMAN  170  ---PPLAG--WSRYIPBG-LQC--SC---G-----IDYYT-LKPEVN      199

BF056441      526 t-tc-----t--t-----catattt-gtcctcctt--t-tc---a      550
      |   |   |   |   |   |   |   |   |   |   |   |   |   |
OPSD_HUMAN  200  NES-FVIYMFVVHFTIPMIIIFFC-----YG-----QLVFT-VKEA      233

BF056441      551 gaa---tacttttc--ag--atgc-----agcctc--c---ag      576
      |   |   |   |   |   |   |   |   |   |   |   |   |   |
OPSD_HUMAN  234  -AAQQQESA---TT-QKA-EKEVT--RMVIIMVIA-----FLICWVPYA-      269

BF056441      577 ag-att---tca--gatt-g---t--agtt--ac-aattctga--g--      605
      |   |   |   |   |   |   |   |   |   |   |   |   |   |
OPSD_HUMAN  270  S-VA--FYIFT--HQG-SNFGPIFMTIPA---FFA-KSA-----AIYNPV      304

BF056441      606 ----t---tct--tct-ccaggtcaccacattttattca---gac--      637
      |   |   |   |   |   |   |   |   |   |   |   |   |   |
OPSD_HUMAN  305  IYIMMNKQFRNC-MLT-TICC--G-----KNPLG--DD      331

BF056441      638 -acctccgcacgcttctc-tgcc-tc-tcagc-ta-acccttc      675
      |   |   |   |   |   |   |   |   |   |   |   |   |   |
OPSD_HUMAN  332  EA--S---A-----T-VS---KT-ET-S--QVAPA-----      348

```

```

#####
#
# Aligned_sequences: 2
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# 2: OPSD_XENLA
# Matrix: EBLOSUM62
# Gap_penalty: 0.0
# Extend_penalty: 0.0
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# Length: 897
# Identity:      84/897 ( 9.4%)
# Similarity:   106/897 (11.8%)
# Gaps:         765/897 (85.3%)
# Score: 484.0
#
#
#####

```

```

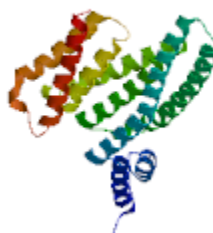
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OPSD_XENLA    0  -----
BF056441     51  agtgtatttctctgggtagtgtggtgctgaataaataggaataaatgctac      100
OPSD_XENLA    0  -----
BF056441    101  ttaaggaaaaaataagagagctgaaaaagctggtgccatttgaaaaaaaa      150
OPSD_XENLA    0  -----
BF056441    151  aaggggaaggaatgagatttaactggtgctcaaagcttctccgatacaaaa      200

```

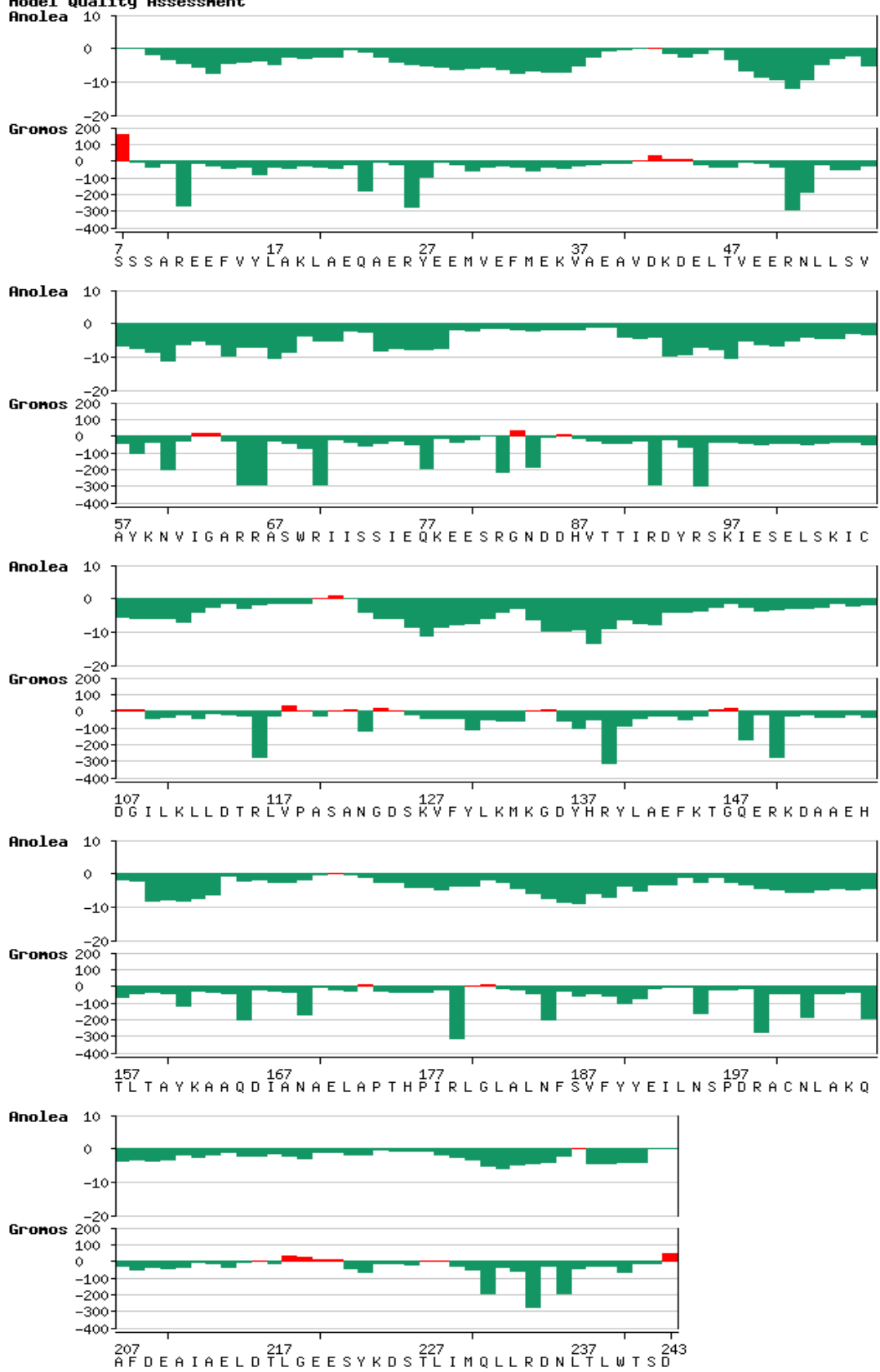
OPSD_XENLA	0	-----	0
BF056441	201	tatttggtcatgtattcataatttgcttgacatttcagcaagcgaaga	250
OPSD_XENLA	0	-----	0
BF056441	251	tggcaataacaaaag-gaac---t--tct-tacaagagaagagaa-ag-	290
OPSD_XENLA	1	MNG---T-----EG---PNFYVPMs-NKT-----GVVRS-P	23
BF056441	291	-----accc-----acggagctcc-aga-----gtttc-----	312
OPSD_XENLA	24	FDYPQYYLA---EPWQYS---A-----LA-AYMFLILLG---LPINFM	57
BF056441	313	tg--t-----tggaaaca-----g---actct--t-ctgttttg	338
OPSD_XENLA	58	T-LFVTIQHKKLR-----PLNYILLNLVFA---NHFMVLC-----G	90
BF056441	339	cttata-tacagttaag--ttc-gttta---g-tgtct--gatcc--	373
OPSD_XENLA	91	-----FT---VT---MYTS-MHG---YFIFGPTG-C-YIEG---FF	116
BF056441	374	agt-gtctg--a--tg-ta-agcc-----c--acgttctcttctttg	406
OPSD_XENLA	117	A-TLG---GEVALWS-LVVLA---VERYIVVCKPMA-----	144
BF056441	407	gcctgg--gca--a-gtttctcttcc-agg-tc--a-tcaattgtc	442
OPSD_XENLA	145	----NFRFG--ENHAIMG-----VA--FT-WIMALSC-A-----	168
BF056441	443	ttttccagtttt---gca-acc-----gttct--ctctgcaaat--tc	477
OPSD_XENLA	169	-----A-----PPLFG--WS--RYIPEG---MQCSC-G---VDYYT-	193
BF056441	478	agc-----acggg-----tc---t--ca-gcctcttt--	498
OPSD_XENLA	194	--LKPEVNNES---FVIYMFIVHFT-IPLIVIFFC-YG-----RLL	227
BF056441	499	cagtttgtcagac---agaag---t-tta--atttc--ttc-----t	529
OPSD_XENLA	228	C-----T-----VKEA-AA-QQESLTT-QKA---EKEVT-RMVVIMV	258
BF056441	530	t---c---a-tatt---t-gtcctccttttca-gaatactt-t--	559
OPSD_XENLA	259	VFFLICWVPYAYVA--FYIFTHQG-----SNFG-----PVFM	288
BF056441	560	tcag--atgc--agcc-tccaga-g-a--t-t---tcagattgtag-	592
OPSD_XENLA	289	T--VPA---FFA---KS--S-AIYNPVIYIVLNKQFRNC-----L	317
BF056441	593	-tta-caattctgagttcttctcca---g---gtcaccacattttattc	633
OPSD_XENLA	318	ITT-LC---C---G-----KNPFGDEDG--S--S-A---A-T-	338
BF056441	634	agacacc-tccgc-acgcttctctgccc-tctcagc-ta-acccttc	675
OPSD_XENLA	339	----S--KT---EA-----S-S---VS-S-S--QVSPA-----	354

Figure 6.86: Needleman-Wunsch Global Alignment of two sequences

SWISS-MODEL Repository Model

		Model 3D Structure				
		Based on template: 2o98 [SMTL] [PDB] [SCOP] [CATH]				
		Sequence identity:		87%		
		Residue range:		7 to 243		
Alignment						
TARGET	7	SSSAREEF	VYLAKLAEQA	ERYEEMVEFM	EKVAEAVDKD	ELTVEERNLL
2o98B	4	aptareen	vymaklaeqa	eryeemvefm	ekvsnslgse	eltveernll
TARGET		hhhhh	hhhhhhhhh	hhhhhhhh	hhhhh	hhhhhhh
2o98B		hhhhh	hhhhhhhhh	hhhhhhhh	hh	hhhhhhh
TARGET	55	SVAYKNVIGA	RRASWRIISS	IEQKEESRGN	DDHVTTIRDY	RSKIESELK
2o98B	52	svayknviga	rraswriiss	ieqkeesrgn	eehvnsirey	rskienelsk
TARGET		hhhhhhhhhh	hhhhhhhhhh	hhhhh	hhhhhhhhhh	hhhhhhhhhh
2o98B		hhhhhhhhhh	hhhhhhhhhh	hhhhh	hhhhhhhhhh	hhhhhhhhhh
TARGET	105	ICDGILKLLD	TRLVPASANG	DSKVFYLMKM	GDYHRYLAEF	KTGQERKDA
2o98B	102	icdgilkllld	aklipsaasg	dskvfylkkm	gdyhrylaef	ktgaerkeaa
TARGET		hhhhhhhhhh	hhhhh	hhhhhhhh	hhhhhh	hhhhhhh
2o98B		hhhhhhhhhh	hhhhh	hhhhhhhh	hhhhhh	hhhhhhh
TARGET	155	EHTLTAYKAA	QDIANAELAP	THPIRLGLAL	NFSVFYYEIL	NSPDRACNLA
2o98B	152	estltaykaa	qdiattelap	thpirglal	nfsvfyyeil	nspdracnla
TARGET		hhhhhhhhhh	hhhhh	hhhhhh	hhhhhhhh	hhhhhhh
2o98B		hhhhhhhhhh	hhhhh	hhhhhh	hhhhhhhh	hhhhhhh
TARGET	205	KQAFDEAIAE	LDTLGEESYK	DSTLIMQLLR	DNLTLWTS	
2o98B	202	kqafdeaiae	ldtlgeesyk	dstlimqllr	dnltlwtsd	
TARGET		hhhhhhhh	hhhhh	hhhhhhhh	hhhhh	
2o98B		hhhhhhhh	hhhhh	hhhhhhhh	hhhhh	

Model Quality Assessment



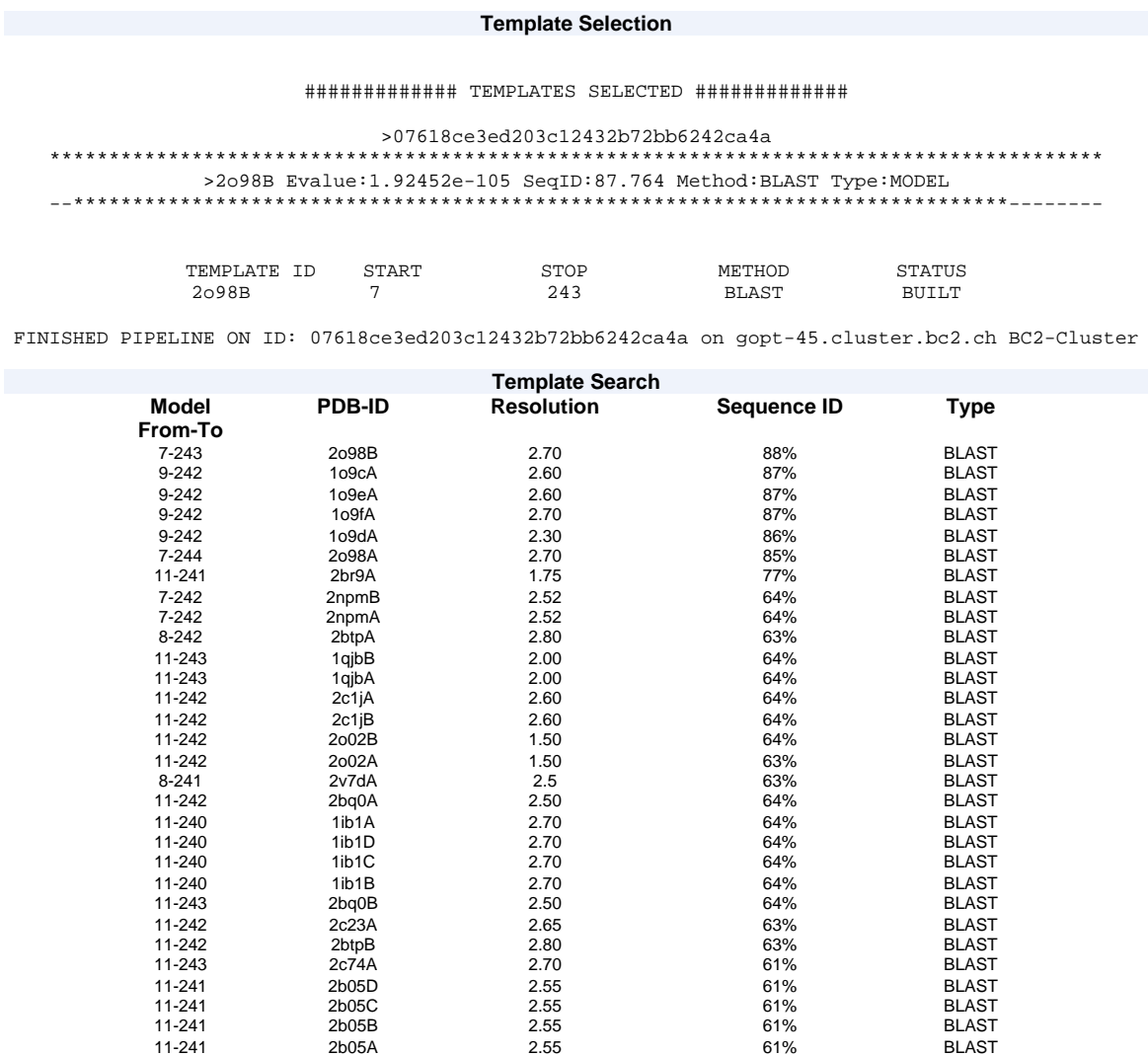


Figure 6. 87: SWISS-MODEL Repository Model

6.3 Data Analysis and Experimental Outcome

In the first phase of experimental work of research (**Activity No-1**) analyses on compounds like Alanine, Amino butyric Acid, Asparagine and Glutamine have been performed using NMRPrediction and ACD/ChemSketch.

This research work describes the characteristics of a free web-based spectral database for chemical research community, containing ^{13}C NMR spectra data from natural compounds. This database allows flexible searching via chemical structure, substructure, name and family of compounds as well as spectral features as chemical shifts, allowing the structural elucidation of known and unknown compounds by comparison of ^{13}C NMR data.

In this experiment script calculates and represents the ^{13}C NMR spectra of the compound. The chemical shifts value obtained in different NMR experiments can be entered with the carbon's hybridization type. This experiment permits to carry out the enquiry with the required number of carbons, from one carbon to the totality of the compound's carbon. It is possible to specify the required deviation (+/-), to the limit in a detailed way. So it limits the search distinctly and therefore a reasonable and manageable compound can be obtained.

If the skeleton of the studied substance is known, and if some distinctive chemical shifts of most important signals are also available, a search by shifts in each particular position of the molecule can be carried out. So it can be obtained the compounds of the family whose shifts, in those position match with those with problem compound.

In this research using ACD/ChemSketch compounds are stored in databases and SMILE codes (Simplified Molecular Input Line Specification) have been generated. A SMILE defines the molecules in the form of alphanumeric chains. This format of structural specification has been used for sharing chemical structure information.

Under this research CML codes of molecules have been developed and that codes have been used for molecular information like symmetry, and atom and bond attributes. Here multiple observations of the same molecule (e.g. conformational analysis and NMR prediction) have been performed.

After that web based structure search queries have been performed on these compounds using web based Pubchem/NCBI. Here activities like bioactivity analysis by structure & activity similarity of molecule , bioactivity analysis by structure & activity similarity of molecule from normalized score percentile , bioactivity analysis by activity & protein target similarity of molecule from normalized score percentile , bioactivity analysis by addition of similar compounds of molecule and revised compound selection after addition of similar compounds of molecule have been performed.

In the second phase of experimental work of research (**Activity No-2**) analyses on same compounds like Alanine, Amino butyric Acid, Asparagine and Glutamine have been performed using ArgusLab tool. Under these experimental work calculations like single entry point calculation, geometry optimization, quick plot HOMO, Quick plot LUMO and quick plot ESP mapped density have been performed.

The outcomes of Activity No-2 have been derived in form of like heat of formation, atom-atom bond orders, atomic spin densities, ground state dipole, SCF plot between energy vs. difference per cycle, final SCF energy, geometric search, comparison between exact and calculated of s2 operator, calculation of ground-state density on grid, calculation of ground-state electrostatic potential on grid and total elapsed time etc. The above all outcomes of compounds like Alanine, Amino butyric Acid, Asparagine and Glutamine have been compared.

It is possible to carry out a combined and simultaneous use of outcomes of Activity No-1 and Activity No-2 like SMILE, chemical shifts, CML, bioactivity analysis of structures, atom-atom bond orders, atomic spin densities, SCF energy and geometric search etc. that

undoubtedly amplifies the search capacity and increases the possibilities of finding compounds and to predict their molecular structure.

PROPOSED MODEL FOR MOLECULAR STRUCTURE PREDICTION

In this research a Model for Molecular Structure Prediction has been developed. This model has been used for prediction of molecular structure. The basic components of this model have been shown in this **Figure 6.88**.

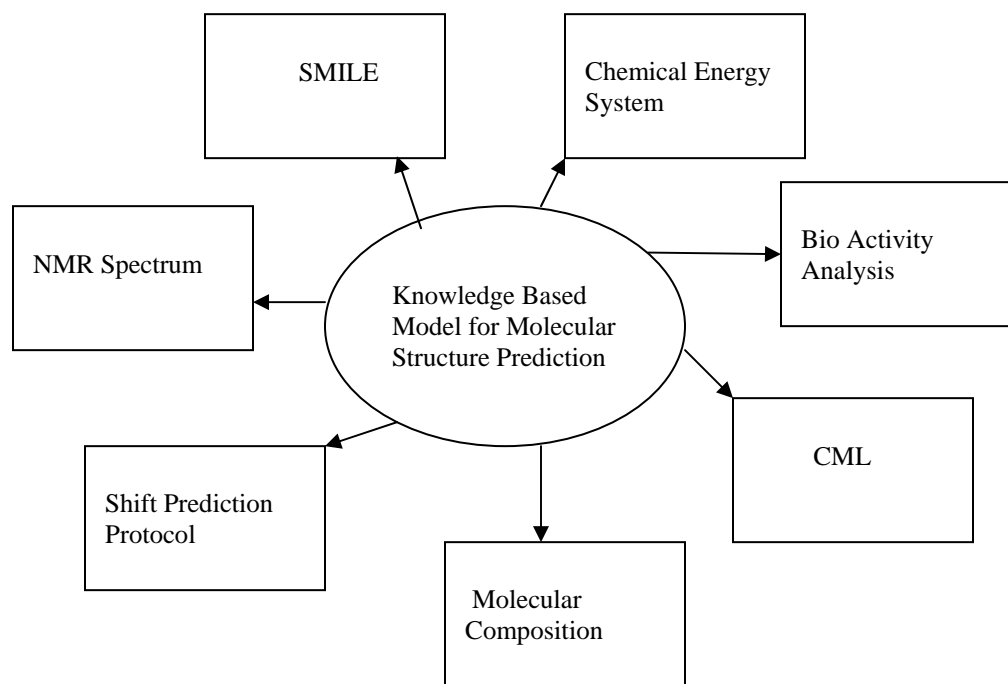


Figure 6.88: Basic Components of Knowledge Based Model for Molecular Structure Prediction

The combinations of all above basic components enhance the capability of this model in predicting molecular structure.

Figure 6.89 is basic view of the model that has been developed. In this model all basic components of compounds like Alanine, Amino butyric Acid, Asparagine and Glutamine have been analyzed and these have been shown in the form of Menu. Here SMILE structure of all above molecules has been generated.

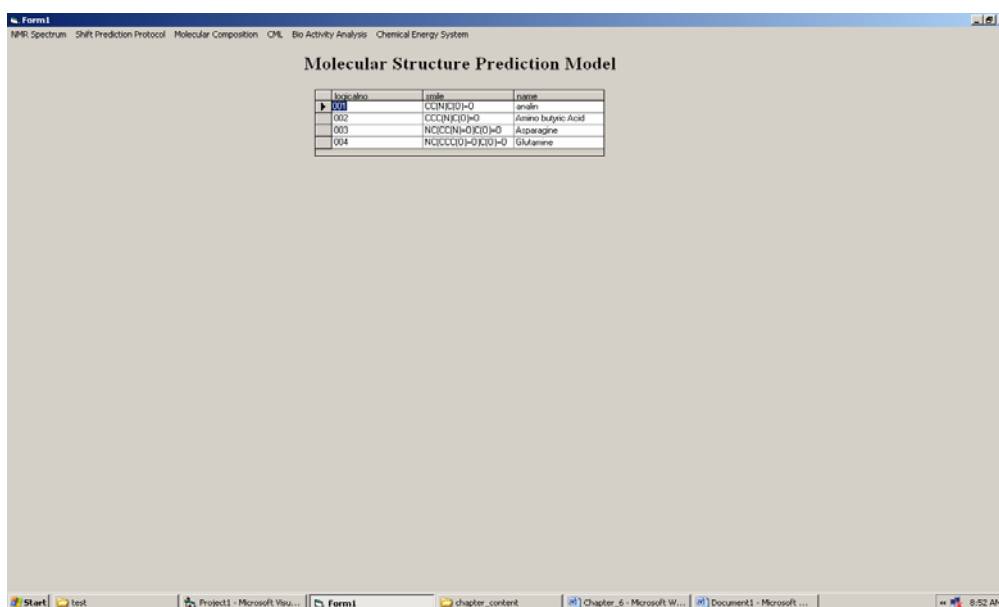


Figure 6.89: Basic view of Knowledge Based Model for Molecular Structure Prediction

In Figure 6.90 different nodes of Aniline like $-C$, CH and $-CH_3$ have been generated in the model in form of 1H -NMR curve.

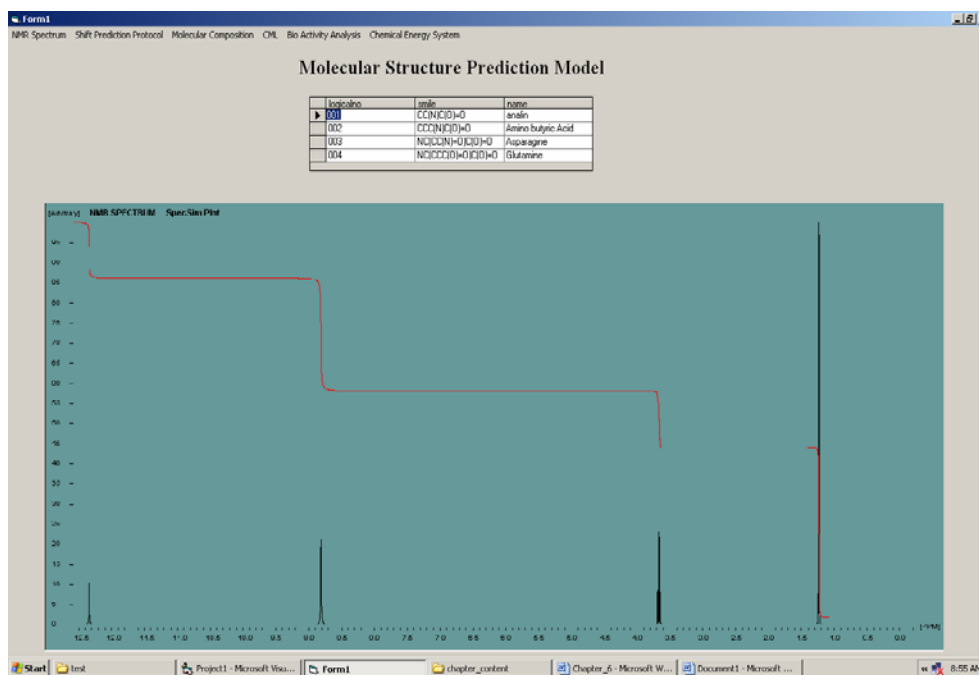


Figure 6.90: Nodes of Aniline in 1H -NMR in model

In Figure 6.91 chemical shift values of different nodes of aniline has been generated in model.

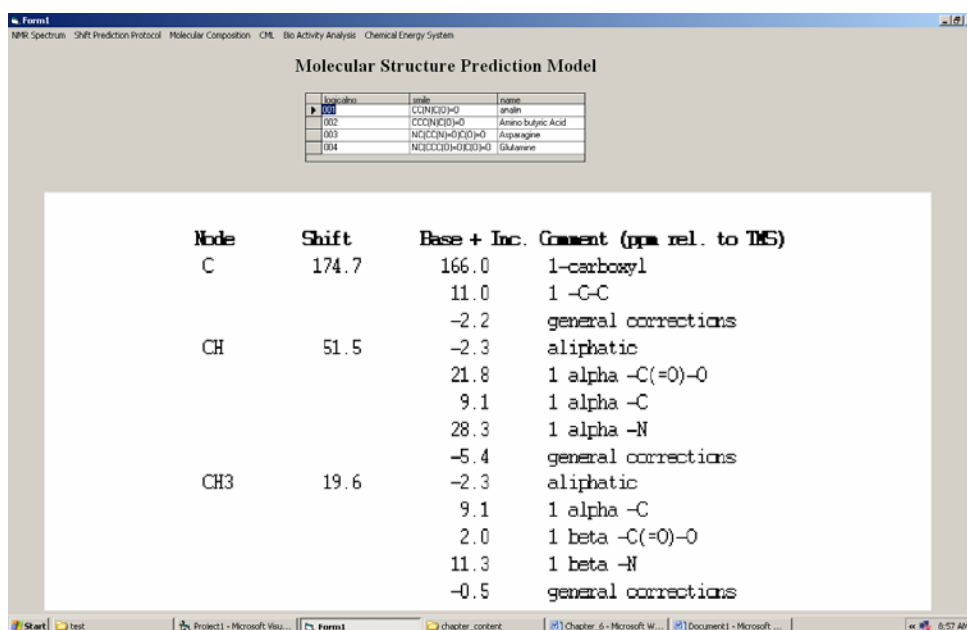


Figure 6.91: Chemical shift values of different nodes of aniline in model

In Figure 6.92 bond orders of the aniline has been generated in model using ^{13}C - NMR.

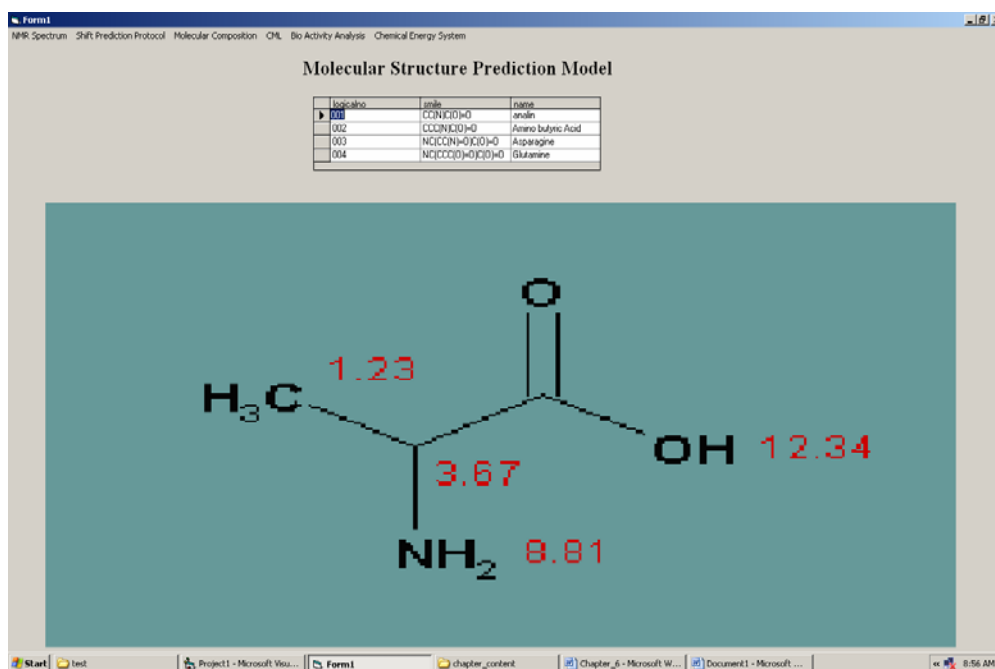


Figure 6.92: Bond order of the aniline in model using ^{13}C - NMR

In **Figure 6.95** different curves like activity outcome, compound cluster have been generated in the model after addition of similar compounds of aniline.

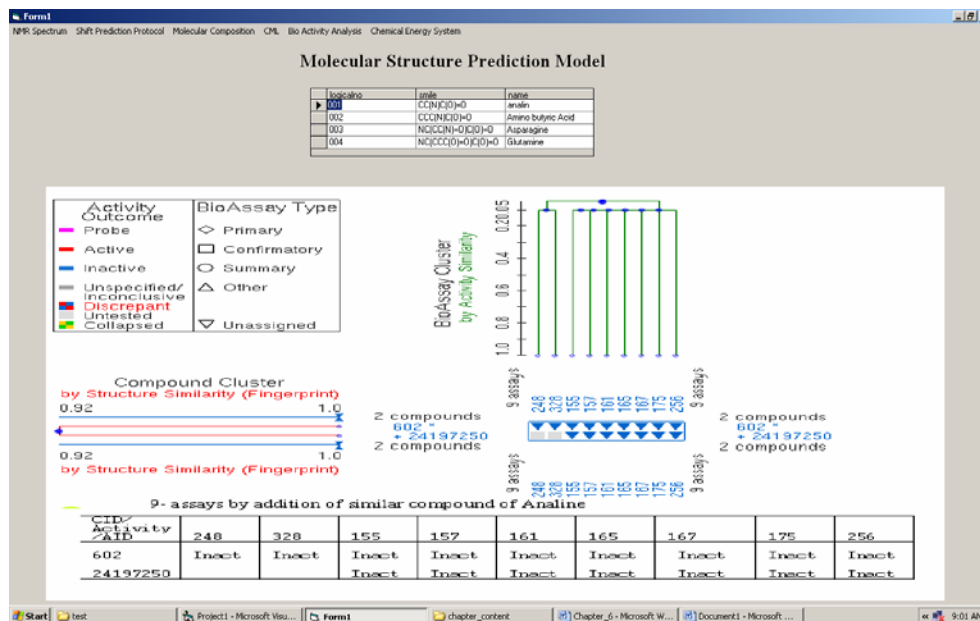


Figure 6.95: Bio activity analysis of aniline after addition of similar compounds in the model

In **Figure 6.96** revised compound selection of aniline has been generated in the model.

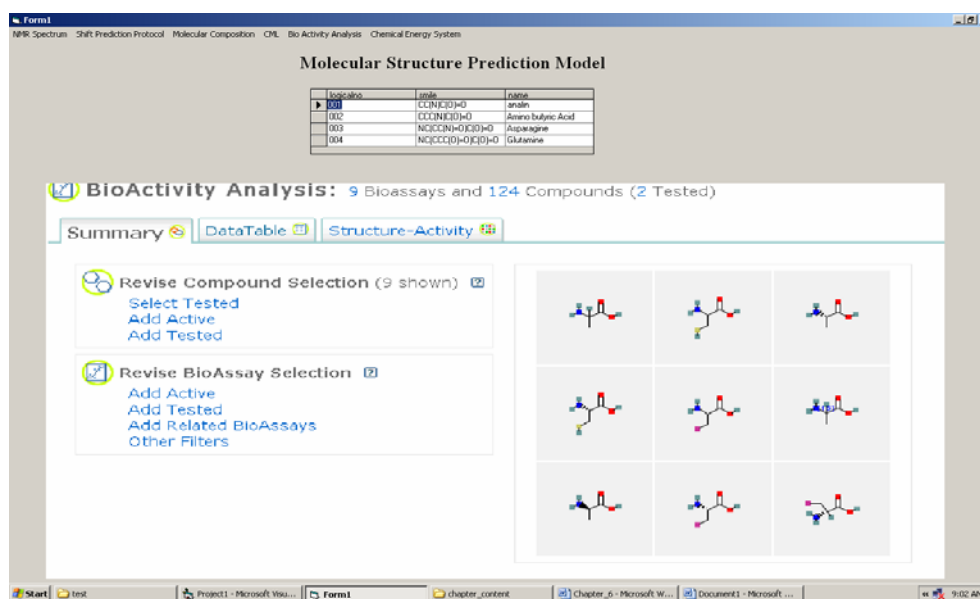


Figure 6.96: Revised compound selection of aniline in the model

In **Figure 6.97** geometric optimization of analine for different components have generated and it ahs been shown in the model.

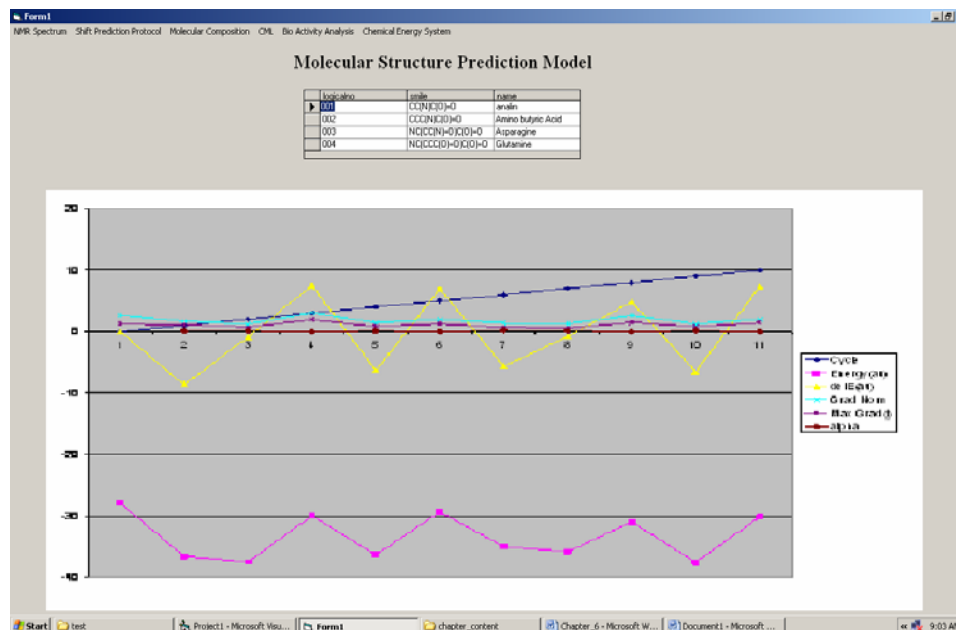


Figure 6.97: Geometric Optimization of analine for different components in the model

In this experimental part of research work different types of energy calculations like Heat of Formation (SEP), Geometric Opt.(Final SEF Energy), Geometric Opt.(Final Gemt. Energy), Geometric Opt.(Heat of Formation), HOMO(Heat of Formation), LUMO(Final SCF Energy), LUMO(Heat of Formation), ESP(Heat of Formation) and ESP(Final SCF) on Alanine, Amino butyric Acid, Asparagine and Glutamine have performed and their result have been presented in form of graph in **Figure 6.98**. All energies values have been shown in **Table 6.39**.

Table 6.39 : Energy Comparative Table

	Aniline	Amino Butyric acid	Asparagine	Glutamine
Heat of Formation(SEP)	11889.4613	-10071.8812	28241.3138	27536.8764
Geometric Opt. (Final SEF Energy)	-17505.372	-10060.3228	-15190.672	-11096.9323
Geometric Opt. (Final Gemt. Energy)	-18937.346	-25092.9388	-30902.648	-37640.8953
Geometric Opt. (Heat of Formation)	6661.9791	10076.7315	15290.0969	11319.6093
HOMO (Heat of Formation)	11889.4613	-24198.9844	1167.033	11841.6877
LUMO (Final SCF Energy)	-16459.454	-24198.9844	-42268.152	-37118.7021
LUMO (Heat of Formation)	11889.4613	7106.8388	1167.033	11841.6877
ESP (Heat of Formation)	11889.4613	7106.8388	1167.033	11841.6877
ESP(Final SCF)	-16459.454	-24198.9844	-42268.152	-37118.7021

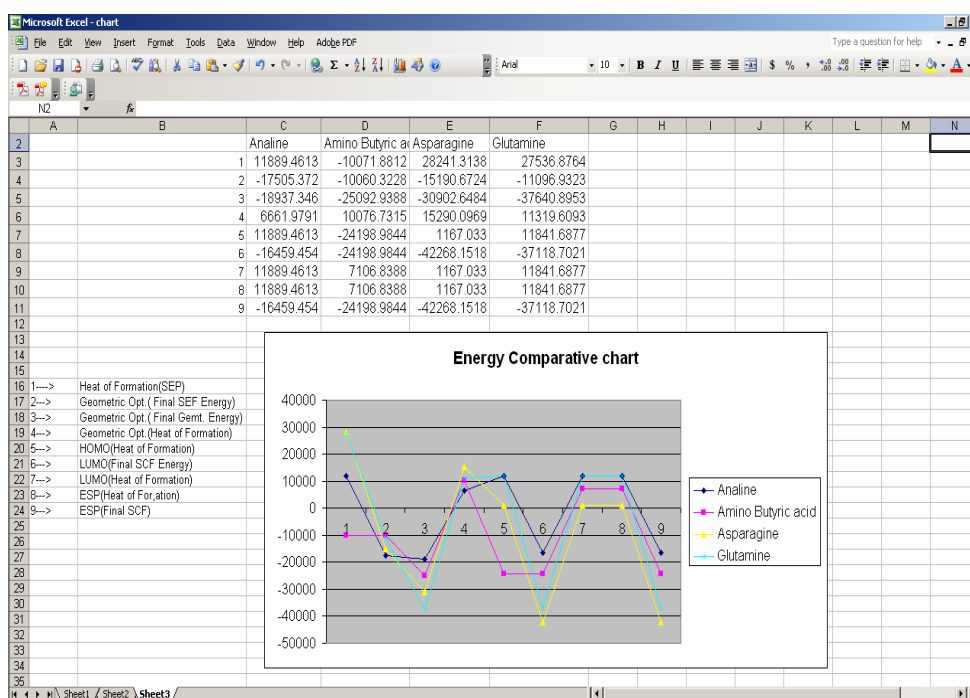


Figure 6.98: Energy Comparative Chart

From this chart it has been interpreted that there has been minor increase in all types of energies due to increase in complexity of structure as complexity in structure is getting increased from aniline to glutamine.

In the last phase of experimental work of research (**Activity No-3**) various analyses on nucleotide and protein sequences have been performed by using DAMBE and Jemboss tools. The outcomes of this experimental part of research have been analyzed and evaluated. Multiple sequence alignment is an extension of pair-wise alignments. Phylogenetic trees are useful representation and method for multiple alignments. The pattern-matching approaches using scores for gaps and inexact matching are statistically valid for assessing the degree of string similarity. It is easy to rationalize the need for gaps because of computational infeasibility of solving long string comparisons without the provision for gaps. However, even a short gap in polypeptide sequence can disturb secondary and tertiary structures of protein and probably alter its function as well. Heuristics approaches, such as match matrices, attempt to add some sense of biological relevance to the mathematical equations that define the relative similarity of nucleotide and polypeptide sequences.

6.4 Conclusions and Future Scope of Research

The outcomes of this work provides an innovative platform to solve complex chemical problems such as structure elucidation, required the joint efforts of computer science and chemistry specialists. The results obtained are a good reason to expect success with novel approaches for current research challenges, as the field of chemoinformatics matures and a closer collaboration with bioinformatics is developed.

By increasing the number of stored compounds, by adding more searches and use information visualization techniques could give more insight in the analysis process. Supervised and unsupervised machine learning methods that could lead to interesting predictions for the different substructures. Use of genetic software techniques could lead to automatically design molecules.

In the extension of this research work, computational neural networks could be used to predict the mapping between protein sequence and secondary structure. By adding neural network units that detect periodicities in the input sequence, secondary structure and tertiary structure prediction accuracy could be increased.