# Underlying Mathematics in Diversification of Human Olfactory Receptors in Different Loci 

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#### Abstract

As per conservative estimate, approximately (51-105) Olfactory Receptors (ORs) loci are present in human genome occurring in clusters. These clusters are apparently unevenly spread as mosaics over 21 pair of human chromosomes. Olfactory Receptor (OR) gene families which are thought to have expanded for the need to provide recognition capability for huge number of pure and complex odorants. ORs form the largest known multigene family in the human genome. Recent studies have shown that 388 full length and 414 OR pseudo-genes are present in these OR genomic clusters. In this paper, the authors report a classification method for all human ORs based on their sequential quantitative information like presence of poly strings of nucleotides bases, long range correlation and so on. An L-System generated sequence has been taken as an input into a star-model of specific subfamily members and resultant sequence has been mapped to a specific OR based on the classification scheme using fractal parameters like Hurst exponent and fractal dimensions.


Categories and Subject Descriptors: Computational Genomics, Olfaction and Molecular Biology
General Terms: Theory and Results
Additional Key Words and Phrases: L-System, Olfactory Receptors (ORs), Star Model, Hurst exponent, Poly-String Mean and Standard Deviation

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## 1. Introduction

Humans can detect a very large number of odor molecules (water or air borne; pure or mixed) and give output as distinct sets. Nasal epitheliums house a large repertoire of seven transmembrane domain G protein coupled olfactory receptor (ORs). These receptors have been described to have diverse protein sequences. Researchers have classified these receptors based on simple computer based alignment tools them into several families and subfamilies. HORDE database used in this study have used a nomenclature which is not sufficient enough to cover all the kinds of receptors that have been found in nature inside human nose. More importantly, SNPs and genomic variant studies have not been performed to assign the classification very accurately. One of many advantages of the OR sequences is that they do not possess introns. Therefore, one can use them as model systems for exons as well. By a conservative estimate, it is known that there are 388 intact and 414 OR pseudogenes (Menashe I et al. [2006] and Yoshihito Niimura et al. [2003]). Researchers have reported that they are unevenly distributed among 51-105
different loci on 21 human chromosomes where each locus contains 1-3 genes (Gustavo Glusman et al [2000]). The present authors do not believe for many reasons that these are uneven distributions. It is a strong conviction of the authors that there is beautiful organization followed by some mathematical governess in diversification of human ORs in different loci. Using a number of mathematical methodologies, in this paper, the authors report that it is possible to find relationship between apparently unrelated ORs located in different chromosome. This hitherto has not been studied and this is the first report to the best of knowledge of authors.

## 2. Methods and Results:

## (2A) Methods and Materials:

Any DNA sequence is clearly consisting of four textures of A, T, C, and G (nucleotide cluster) as shown in the color template (Appendix-I). There are several poly-strings having lengths $1,2,3$, etc in all the textures of any base A, T, C and G . The frequencies of different poly-strings of different lengths for each nucleotide cluster for each OR have been calculated. Poly-string mean and standard deviation for each nucleotide cluster of A, T, C and G have been enumerated. A decreasing ordering of poly-string mean and standard deviation of different nucleotide textures were arranged. Results show that all ORs could be classified into 21 different classes using these two deterministic statistical parameters (Data available as supplementary material-I $(A) \&(B)$ ).

Each of A, T, C and G of all OR sequences have been encoded by the following two bit information.
$A \rightarrow 00, T \rightarrow 11, C \rightarrow 01, G \rightarrow 10$. The reason for such encoding is that A pairs with T and G with C .
As for an example: AGTCG have been encoded into 0010110110.
Hurst exponent for each of the encoded human OR sequences were then calculated (Data available as supplementary material-II).
Without loss of generality, it was assumed that all subfamily members (exons) namely OR1D2, OR1D4, and OR1D5 subfamily members of D family (as HORDE nomenclature). Then the star model of those sequences was been extracted (Sk. S. Hassan et al. [2010]). An L-System generated sequence can be taken as an input into the star model to get a full length sequence (methodology is shown in (Sk. S. Hassan et al. [2010]). The resultant sequence could be classified based on the parameters poly-string mean and standard deviation. Now the Hurst exponent was employed for exact quantification of any one or more ORs of the mapped classes.

## (2B) Results and Analysis:

## (I) Mapping Between Resultant Sequence and a Full Length OR

The star model for the ORs namely OR1D2, OR1D4, OR1D5 have been extracted which is shown below. The process of extracting a star model is shown in (Sk. S. Hassan et al. [2010]). In the following star model there are 108 mismatches which are shown as hyphened in the following star model.

```
ATGGATGGAG--AACCAGAGTGA----TCA-AGTTCCTTCTCCTGGGGAT . . . . }5
-TCAGAGAGTCCTGAGCAGCAGC-GATCCTGTTTTGGATGTTCCTGTCCA . . . }10
TGTACCTGGTCACGGTG-TGGGAAATGTGCTCATCATCCTGGCCATCAGC. . . }15
TCTGATTCCC-CCTGCACACCCCC-TGTACTTCTTCCTGGCCAACCTCTC. . . }20
CTTCACTGACCTCTTCTTTGTCACCAACACAATCCCCAAGATGCTGGTGA . . . 250
AC-TCCAGTCCCA-AACAAAGCCATCTCCTATGCAGGGTGTCTGACACAG . . . }30
CTCTACTTCCTGGTCTCCTTGGTG-CCCTGGACAACCTCATCCTGGC-GT . . . }35
GATGGC-TATGA-CGCTATGTGGCCA-CTGCTGCCCCCTCCACTA---CA . . . }40
CAGCCATGAGCCCT--GCTCTGT-TCTT-CTCCT-TCCTTGTGTTGGG-- . . . 450
CT-TC-GT-CTCTATGGCCTC-T-C-CACC-TCCTC-TGACCAG-GTGAC . . . }50
CTTCTGTGGG-C--GA-A-ATCCACTAC-TCTTCTGTGA-ATGTA--T-- . . . }55
TGCTG-GG-TGGCATGTTCCAACA--CA-AT-A-TCACACAG-G-TGATT. . . }60
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GCCAC-GGCTGCTTCATCTTCCTCA--CCCTT-GG-TTC-TGA-CA--TC. . . }65
CTATGT-CG-ATT-TCAGA-CCATCCT---AAT-CCCTC-G-CTCTAAGA . . . }70
AATACAAA-CCTTCTC-ACCTGTGCCTCCCATTTGGGTG--GTCTCCCTC . . . }75
TT-TATGGGA--CTT--TATGGT-TACCT--AGCCCCTCCATACCTACTC. . . }80
--TGAAGGACTCAGTAGCCACAGTGATGTATGCTGTG-TGACACC-ATGA. . . }85
TGAA-CC-TTCATCTACAG-CTGAGGAACAA-GACATGCATGGGGCTC-G . . . }90
GGAAGA-TCCTA---A-AC-CTTT-AGAGGC--A-A . . . }93
```

Fig.1A: Star model of OR1D subfamily of OR gene sequences.

An L-System (shown below) is used to generate a sequence and that is being inputted into the above star model as proposed in (Sk. S. Hassan et al. [2010]). Consequently we got the following resultant sequence (RS-1) shown in fig 1 B .

## L-System:

Axiom: A
Production Rules: $\mathrm{A} \rightarrow \mathrm{CTG}, \mathrm{C} \rightarrow \mathrm{CCA}, \mathrm{T} \rightarrow \mathrm{TGC}, \mathrm{G} \rightarrow \mathrm{GAC}$.
ATGGATGGAGCCAACCAGAGTGAGTCCTCACAGTTCCTTCTCCTGGGGATGTCAGAGAGTCCTGAGCAGCAGCAGATCCTGTTTTTGGATGTT CCTGTCCATGTACCTGGTTCACGGTGCTGGGAAATGTGCTCATCATCCTGGCCATCAGCTCTGATTCCCCCCTGCACACCCCCGTGTACTTC TTCCTGGCCAACCTCTCCTTCACTGACCTCTTCTTTGTCACCAACACAATCCCCAAGATGCTGGTGAACCTCCAGTCCCAGAACAAAGCCAT CTCCTATGCAGGGTGTCTGACACAGCTCTACTTCCTGGTCTCCTTGGTGACCCTGGACAACCTCATCCTGGCCGTGATGGCCTATGATCGCT ATGTGGCCAGCTGCTGCCCCCTCCACTACGCCACAGCCATGAGCCCTGCGCTCTGTCTCTTCCTCCTGTCCTTGTGTTGGGCGCTGTCAGTC CTCTATGGCCTCCTGCCCACCGTCCTCATGACCAGCGTGACCTTCTGTGGGCCTCGAGACATCCACTACGTCTTCTGTGACATGTACCTGGT GCTGCGGTTGGCATGTTCCAACAGCCACATGAATCACACAGCGCTGATTGCCACGGGCTGCTTCATCTTCCTCACTCCCTTGGGATTCCTGA CCAGGTCCTATGTCCCCATTGTCAGACCCATCCTGGGAATACCCTCCGCCTCTAAGAAATACAAAGCCTTCTCCACCTGTGCCTCCCATTTG GGTGGAGTCTCCCTCTTATATGGGACCCTTCCTATGGTTTACCTGGAGCCCCTCCATACCTACTCCCTGAAGGACTCAGTAGCCACAGTGAT GTATGCTGTGGTGACACCCATGATGAACCCGTTCATCTACAGCCTGAGGAACAAGGACATGCATGGGGCTCAGGGAAGACTCCTACGCAGAC CCTTTGAGAGGCAAACA
[Fig 1B: The resultant sequence (RS-1)]
We have found the following data corresponding to the Resultant Sequence (RS-1):

| Nucleotide Texture (NT) | Poly-String Mean <br> $(\boldsymbol{P S M})$ | Poly String Standard Deviation <br> $(\boldsymbol{P S}$-SD $)$ |
| :---: | :---: | :---: |
| $\mathbf{A}$ | 1.176471 | 0.459008 |
| $\mathbf{T}$ | 1.195000 | 1.195000 |
| $\mathbf{C}$ | 1.594872 | 0.838175 |
| $\mathbf{G}$ | 1.344156 | 0.658411 |

[Table-I: Poly String mean and SD for the resultant sequence]
Therefore the decreasing ordering based on mean and SD is $C>G>A>T$ (CGAT) and $C>G>T>A$ (CGTA) respectively. Also the Hurst exponent of the encoded two bit sequence corresponding to the RS-1 is $\mathbf{0 . 6 1 3 1 9 1}$. Consequently the RS-1 is mapped into the class CGAT and CGTA based on mean and SD respectively (See the supplementary material-I). Now, the sequences OR1D2, OR1D4, OR1D5, and OR1N2 belong to the union of two classes CGAT and CGTA. According to Hurst exponent the resultant sequence is mapped into OR1N2 (See the supplementary material-II).

In the same manner as above some other results are given below:

| Subfamily Members | L-System | Data |  |  | Hurst Exponent of RS | Mapped OR |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| OR1D2, OR1D4, OR1D5, OR1E1 | $\begin{aligned} & \text { Axiom: A } \\ & \text { A->CTGCTG } \\ & \text { C->CCACCA } \\ & \text { T->TGCTGC } \\ & \text { G->GACGAC } \end{aligned}$ | NT | PSM | PS-SD | 0.608632 | OR1R1P |
|  |  | A | 1.182390 | 0.460453 |  |  |
|  |  | T | 1.208556 | 0.531702 |  |  |
|  |  | C | 1.656388 | 1.080867 |  |  |
|  |  | G | 1.229508 | 0.524590 |  |  |
| OR1D2, OR1D4, <br> OR1D5, OR1E1, <br> OR1E2  | $\begin{aligned} & \text { Axiom: A } \\ & \text { A }->\text { ATGC } \\ & \text { C->CGCG } \\ & \text { T->GTCG } \\ & \text { G }->\text { CCCA } \end{aligned}$ | NT | PSM | PS-SD | 0.578986 | OR1ABAP |
|  |  | A | 1.256000 | 0.618437 |  |  |
|  |  | T | 1.200000 | 0.445769 |  |  |
|  |  | C | 1.672646 | 1.094428 |  |  |
|  |  | G | 1.349693 | 0.678628 |  |  |
| OR1D2, OR1D4,OR1D5, OR1E1,OR1E2, OR1E6 | $\begin{aligned} & \text { Axiom: A } \\ & \text { A }->\text { ATGC } \\ & \text { C->CGCG } \\ & \text { T->GTCG } \\ & \text { G->CCCA } \end{aligned}$ | NT | PSM | PS-SD | 0.58051 | OR1J2 |
|  |  | A | 1.221311 | 0.535805 |  |  |
|  |  | T | 1.351485 | 0.605440 |  |  |
|  |  | C | 1.538462 | 0.939753 |  |  |
|  |  | G | 1.356643 | 0.662751 |  |  |
| OR1F1, OR1F2,OR1F12 | $\begin{aligned} & \text { Axiom: A } \\ & \text { A->GCGC } \\ & \text { C->CCTC } \\ & \text { T->AAAG } \\ & \text { G->GAAC } \end{aligned}$ | NT | PSM | PS-SD | 0.546249 | OR1Q1 |
|  |  | A | 1.253247 | 0.553169 |  |  |
|  |  | T | 1.237180 | 0.717118 |  |  |
|  |  | C | 1.663934 | 1.083487 |  |  |
|  |  | G | 1.265306 | 0.563354 |  |  |
| $\begin{aligned} & \text { OR1F1, OR1F2, } \\ & \text { OR1F12, OR1I1 } \end{aligned}$ | $\begin{aligned} & \text { Axiom: A } \\ & \text { A }->\text { GCAC } \\ & \text { C->CTCC } \\ & \text { T->GGGC } \\ & \text { G->CTTC } \end{aligned}$ | NT | PSM | PS-SD | 0.511355 | OR1F2 |
|  |  | A | 1.114286 | 0.397954 |  |  |
|  |  | T | 1.237374 | 0.681070 |  |  |
|  |  | C | 1.807087 | 1.215949 |  |  |
|  |  | G | 1.279221 | 0.540523 |  |  |

[Table-II: Detail results]
In the above table, it is observed that a star model of a set of subfamily members together with the contribution of an L-System would lead to a resultant sequence which is mapped to either a pseudo-gene or a full length gene. It is worth noting that all star models of full-length genes have been considered. Interestingly the contribution of LSystem input into the star model would map to either a full length gene or a pseudo-gene. Also an interesting significant observation is that a full gene could be mapped to either full length gene or pseudo gene and same is true for pseudo gene too.

## 3. Conclusion

In this paper a classification scheme is explored based on fractal and deterministic statistical parameters. Starting from a particular set of sequences which are taken from a subfamily of ORs, a unique resultant sequence corresponding to an L-system could be generated and mapped to another OR which may or may not be in the same loci where from the subfamily members were chosen. In this regard a conclusion could be drawn that this is how diversification of human ORs were done, which is governed by the mathematical principle as described in the paper, of course there might be different principle which really followed by nature to make the diversification of ORs in the
different loci but this is our humble try to understand nature, we believe nature is beyond of all our artificial engineering.

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## Appendix-I

The color template of DNA as shown below: Coding scheme is as follows: A=Red, T=Blue, G=Green and C=Yellow.


## Appendix-II (Supplementary materials)

The classification tables for Poly-String Mean and Standard deviation are given as supplementary material-I. Also list of Hurst exponent for all encoded 2 bit information corresponding to all OR sequences is also attached as supplementary material-II.

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