

# Multidimensional Analysis of Snow Cover Data in South Dakota Diversity of Landscapes

Nature Proceedings: doi:10.1038/npre.2008.24451 : Posted 27 Oct 2008

Photo picture taken:  
Thursday, March 27, 2008,  
7:45 am

Shmagin, B.A. & C.A. Johnston  
Water Research Institute      Department of Biology & Microbiology

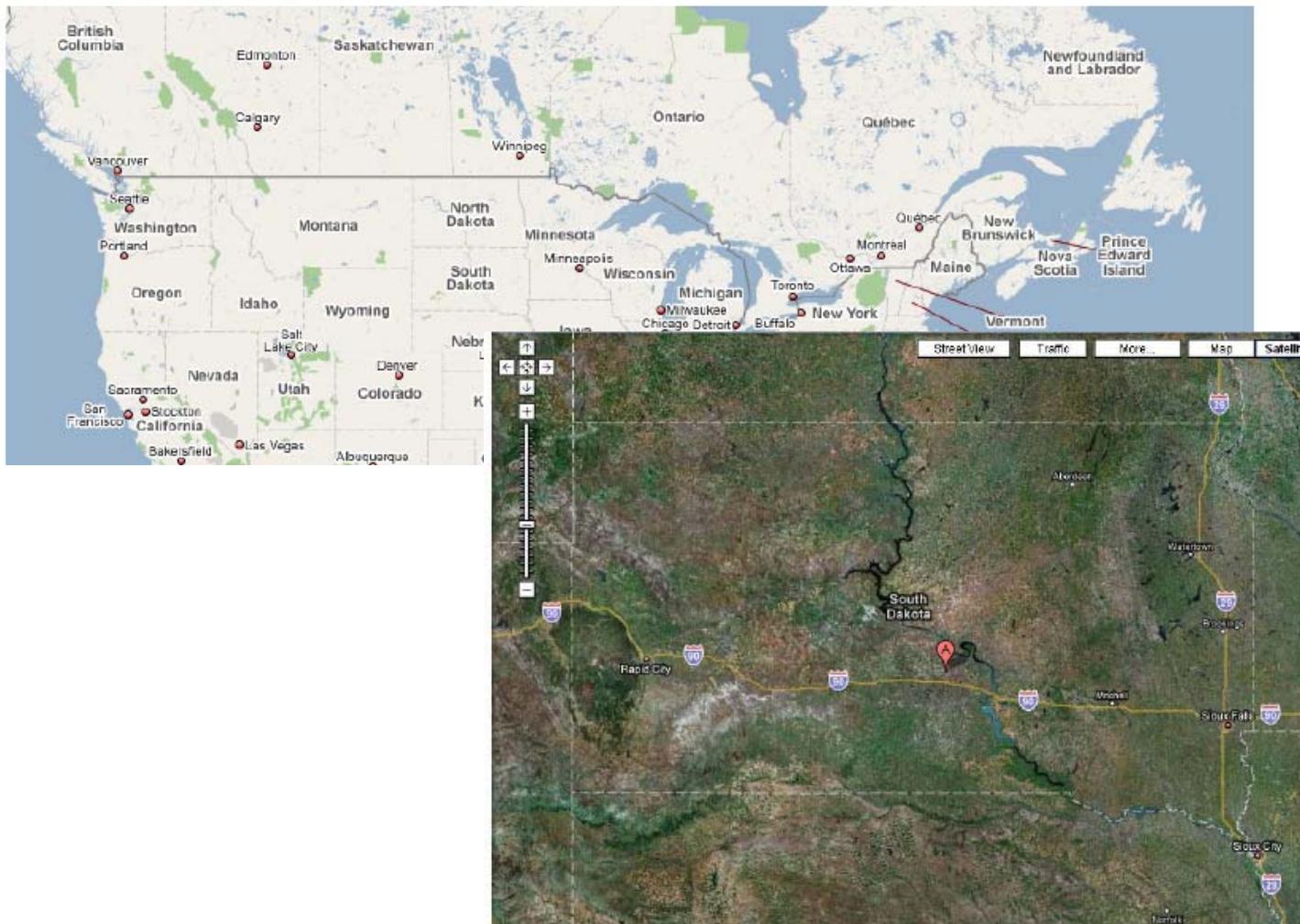
South Dakota State University,  
Brookings, SD



# Topics

- Introduction: maps & snow fall as the most reliable characteristic
- Data & philosophy of analysis
- Analysis of snow fall for period 1955-79
- Regime of snow fall in SD
- Improving the knowledge & creating maps of natural resources of SD
- Questions

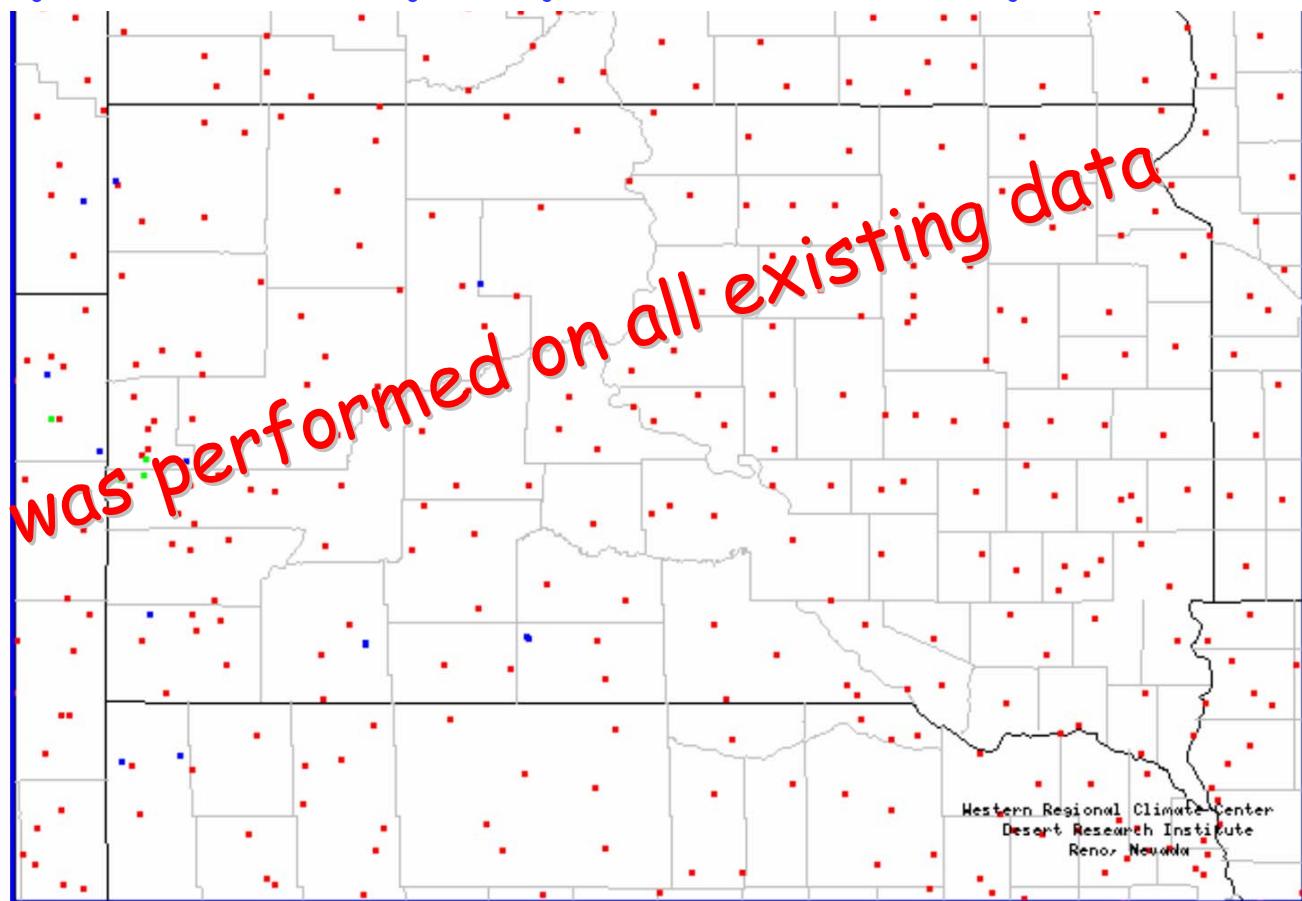
# Maps as a universal tool



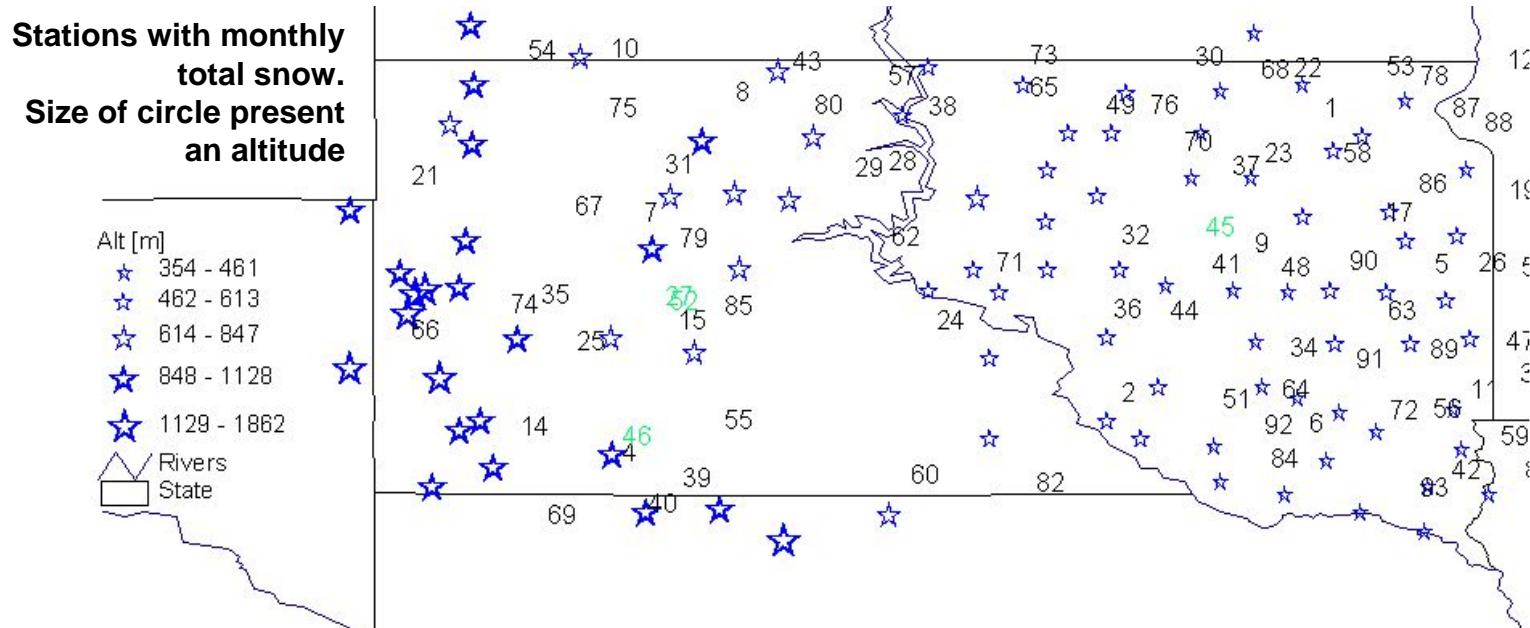
[http://hprcc.unl.edu/cgi-bin/cli\\_perl\\_lib/cliMAIN.pl?sd1076](http://hprcc.unl.edu/cgi-bin/cli_perl_lib/cliMAIN.pl?sd1076)

# Data & philosophy of analysis

The analysis was performed on all existing data



# Data on monthly snow fall in SD

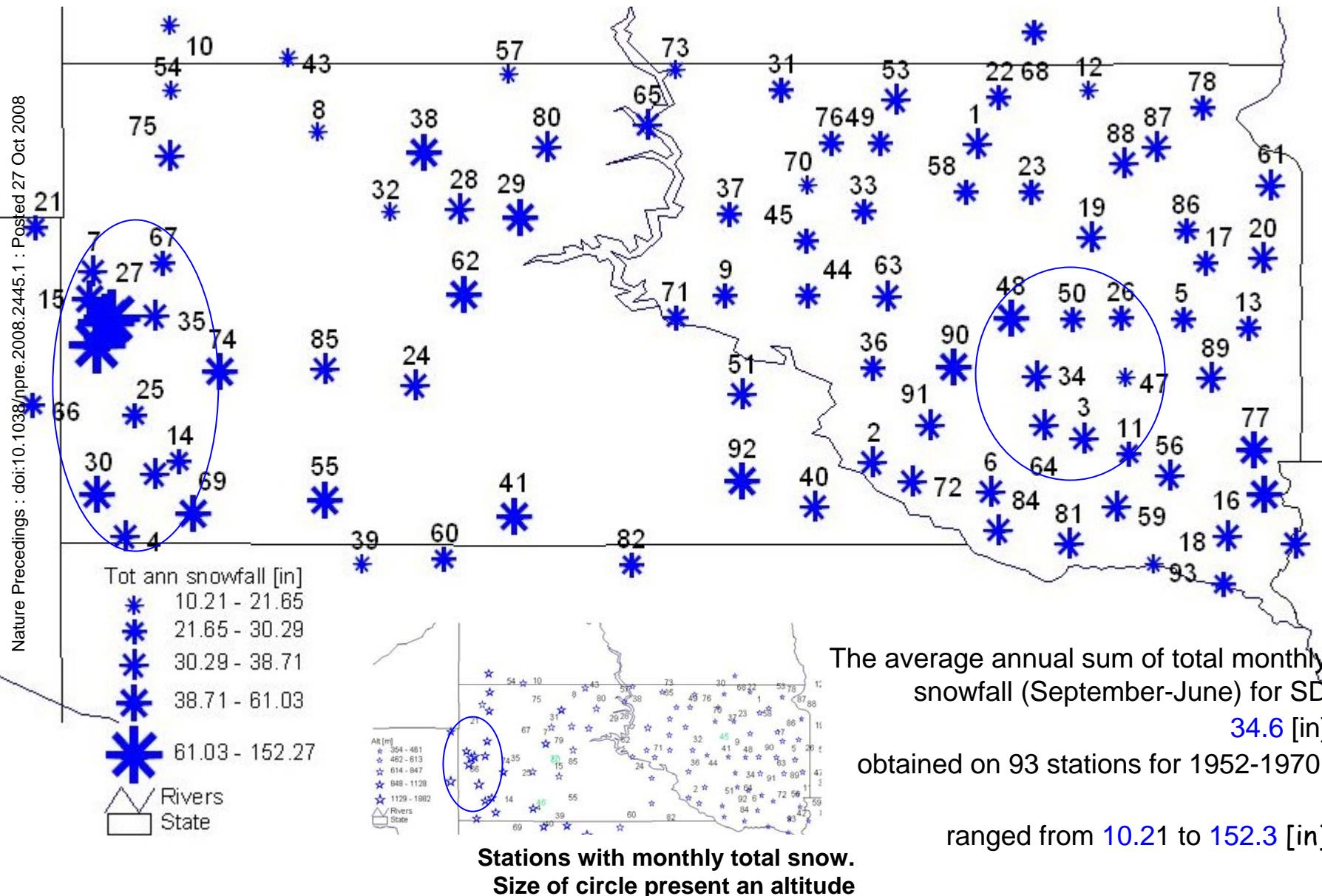


Initial matrix:  $X_{(n \times p)}$  or  $X_{(93 \times 11)}$  there are:

n=93 – number of rows or stations, those stations have mutual time interval of 18 winters of observations  
(1952-53 – 1969-70),

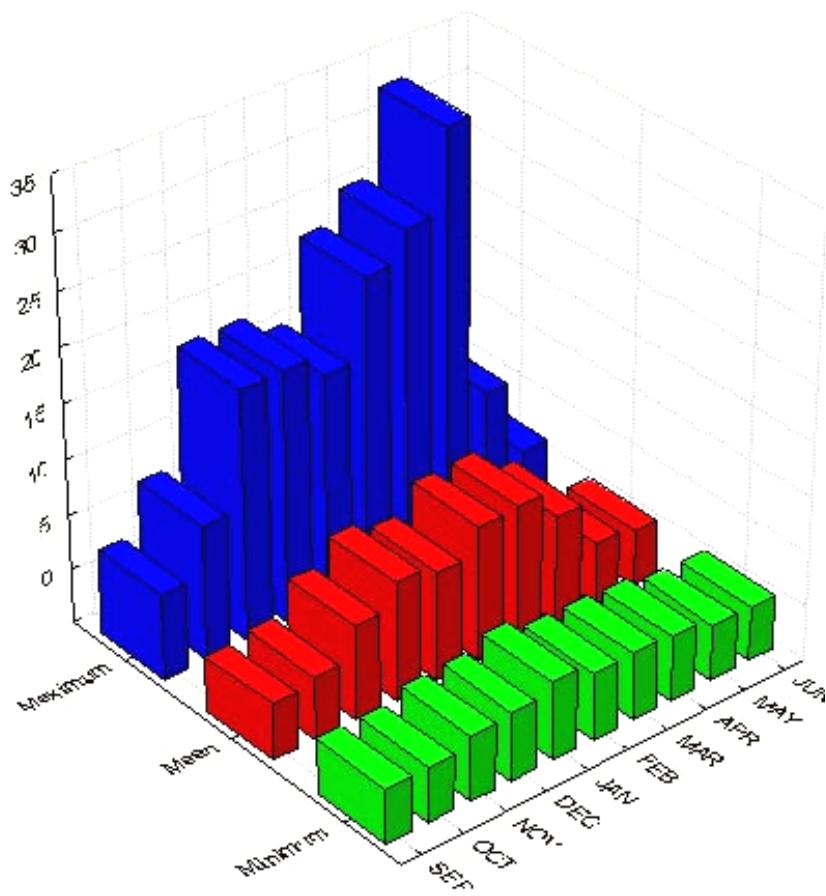
p=11 – number of variables or 10 monthly proportions & annual total snowfall for the winter season

# Average annual snow fall in SD

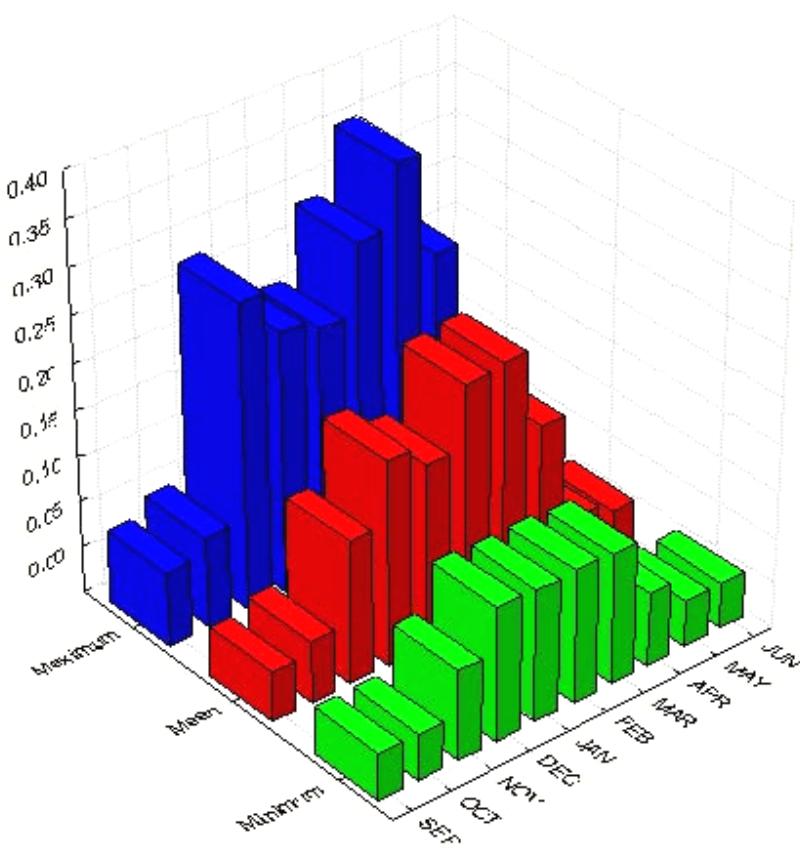


# Average annual snow fall in SD

3D Sequential Graph



3D Sequential Graph



The average total monthly snowfall [in] obtained on 93 stations for 1952-1970 (left)  
& monthly proportion of total annual

# Data & philosophy of analysis

**Multivariate analysis** is the simultaneous statistical consideration of relationships among many measured properties of a given system (Gould 1996, p. 42;).

A **factor** is a portion of a quantity, usually an integer or polynomial that, when multiplied by other factors, gives the entire quantity.

The determination of factors is called factorization (or sometimes "factoring"). It is usually desired to break factors down into the smallest possible pieces so that no factor is itself factorable.

**Factor analysis** allows the determination of common axes influencing sets of independent measured sets.

The main applications of factor analytic techniques are:

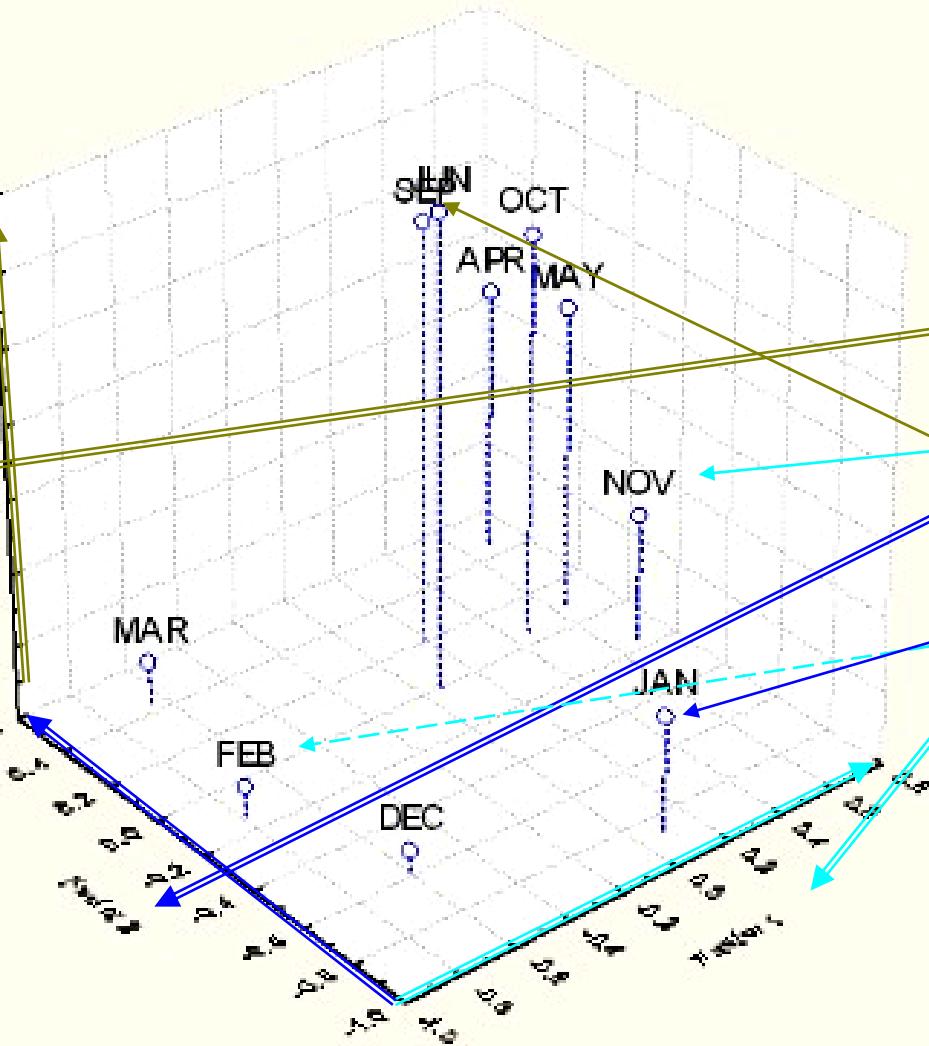
- (1) to reduce the number of variables &
- (2) to *detect structure in the relationships between variables*.

(From: Wolfram *MathWorld*)

# Structure of monthly snow fall

Nature Precedings : doi:10.1038/npre.2008.2445.1 : Posted 27 Oct 2008

3D Scatterplot Factor Loadings (Varimax normalized)



Factor Loading

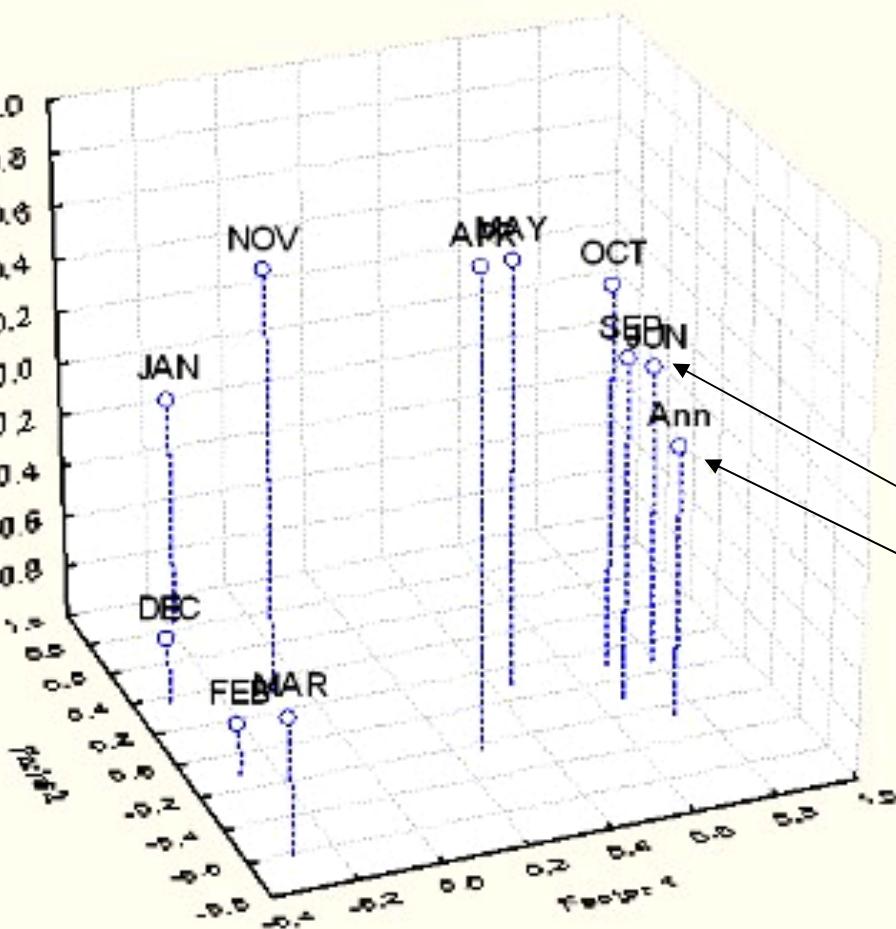
	Factor 1	Factor 2	Factor 3
SEP			0.74
OCT	0.46		0.69
NOV		0.69	
DEC	-0.59		-0.35
JAN		-0.58	-0.86
FEB	-0.79		-0.33
MAR	-0.65	0.43	-0.29
APR	0.69	0.46	0.31
MAY	0.65		0.42
JUN			0.88
Expl.Var	3.03	1.58	2.40
Prp.Totl	0.30	0.16	0.24

# Annual & monthly snow fall in SD

Factor Loadings, Factor 1 vs. Factor 2 vs. Factor 3

Rotation: Varimax normalized

Extraction: Principal components

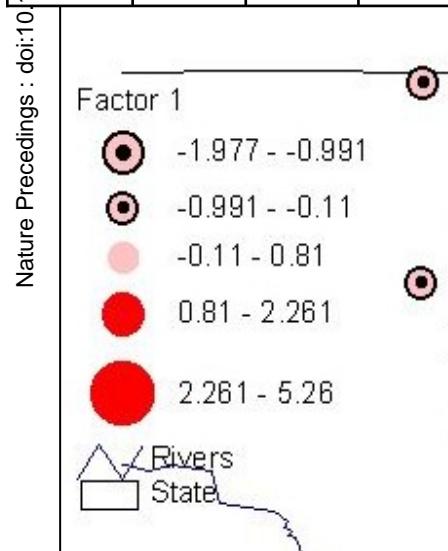


## Factor Loading

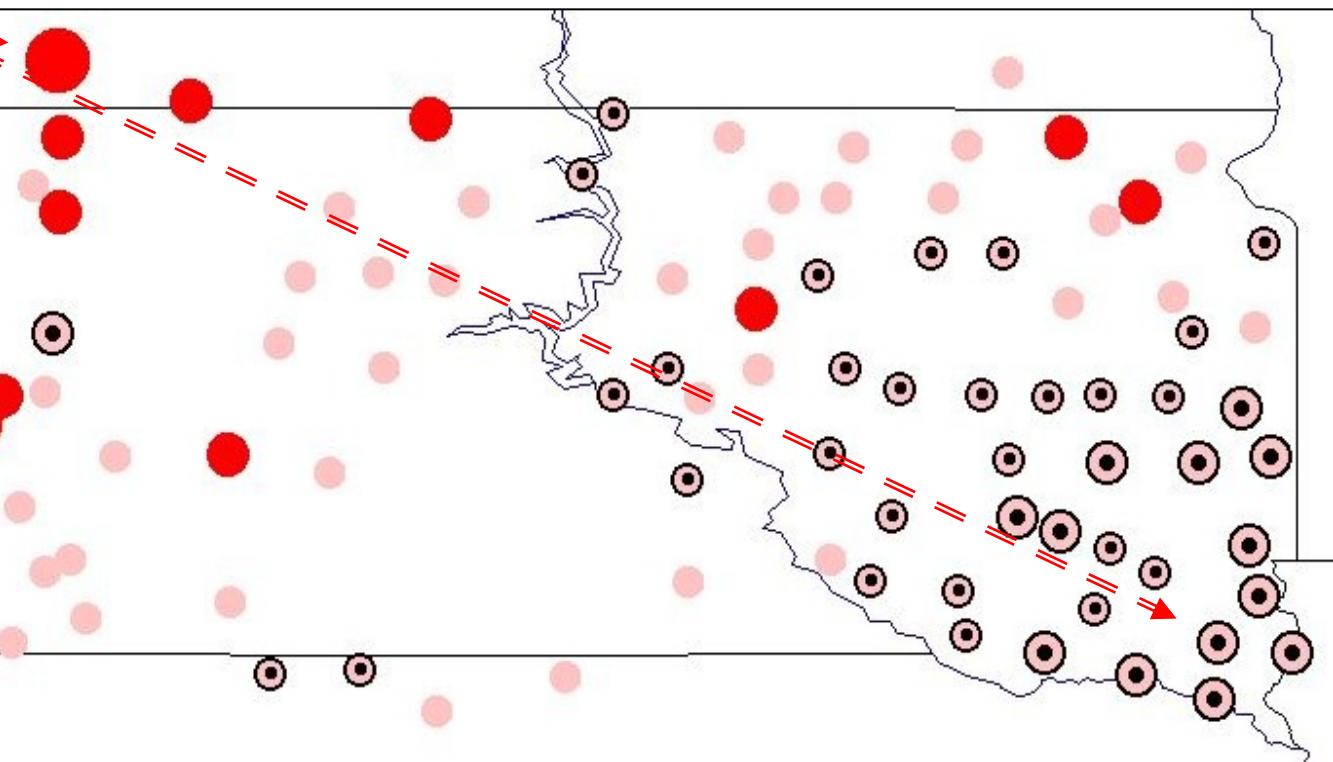
	Factor 1	Factor 2	Factor 3
SEP	<b>0.66</b>		0.31
OCT	<b>0.69</b>		0.47
NOV		0.32	<b>0.62</b>
DEC	-0.31	0.37	<b>-0.76</b>
JAN		<b>0.85</b>	
FEB	-0.28		<b>-0.81</b>
MAR	-0.30	<b>-0.60</b>	-0.48
APR	0.27		<b>0.82</b>
MAY	0.45		<b>0.63</b>
JUN	<b>0.79</b>		
Ann	<b>0.74</b>	-0.25	
Expl.Var	<b>(2.66)</b>	1.49	<b>(3.26)</b>
Prp.Totl	0.24	0.14	0.30

# Distribution of monthly snow fall

	Factor 1	Factor 2	Factor 3
SEP			<b>0.74</b>
OCT	0.46		<b>0.69</b>
NOV	<b>0.69</b>		
DEC	-0.59	<b>-0.58</b>	-0.35
JAN		<b>-0.86</b>	
FEB	-0.79		-0.33
MAR	-0.65	0.43	-0.29
APR	<b>0.69</b>	0.46	0.31
MAY	<b>0.65</b>		0.42
JUN			<b>0.88</b>
Expl.Var	3.03	1.58	2.40
Pr.Ptotl	0.30	0.16	0.24



Factor 1 Scores distribution  
Size of circle present Factor Scores



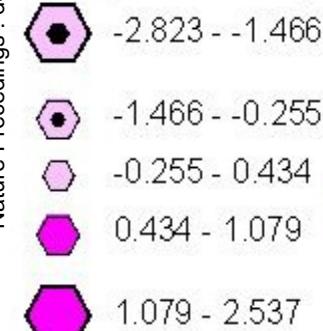
# Distribution of monthly snow fall

## Factor 2 Scores distribution

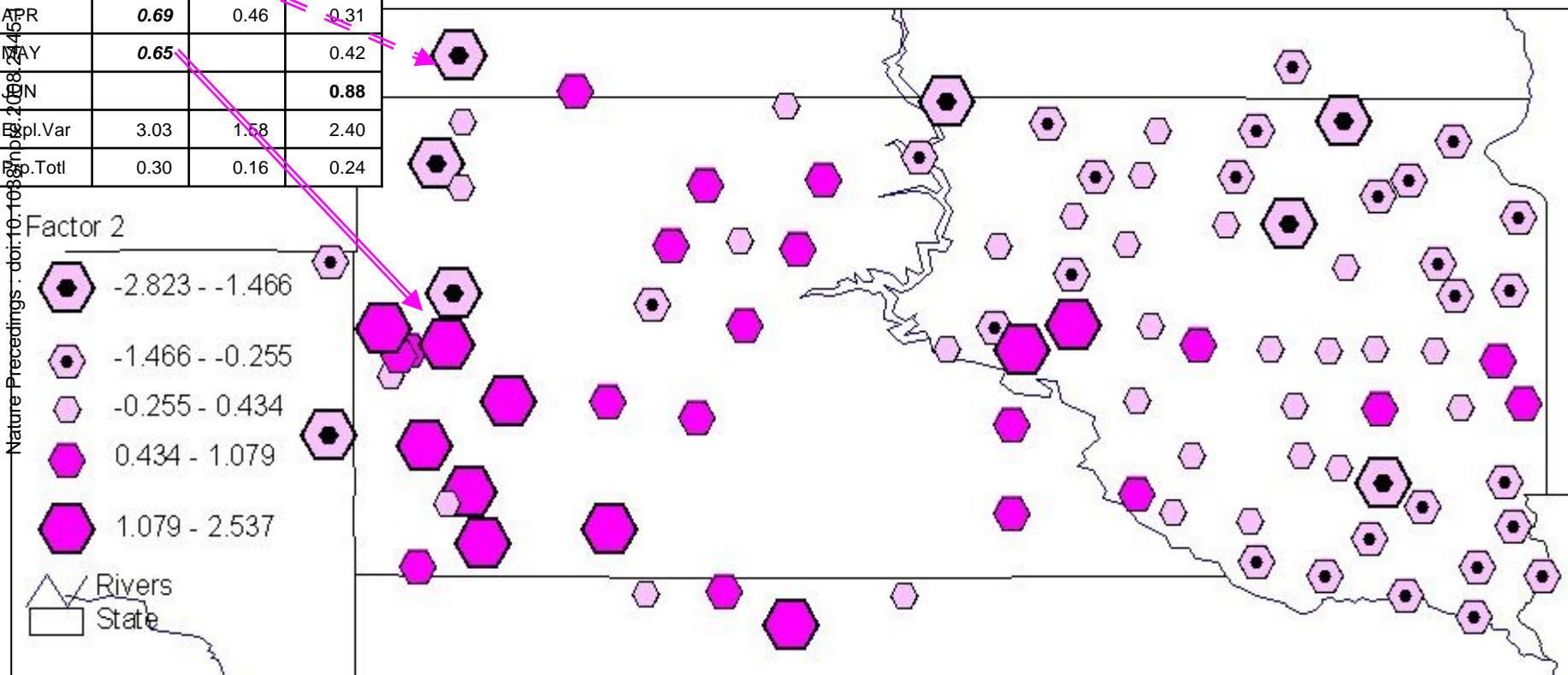
Size of circle present Factor Scores

	Factor 1	Factor 2	Factor 3
SEP			<b>0.74</b>
OCT	0.46		<b>0.69</b>
NOV	<b>0.69</b>		
DEC	-0.59	-0.58	-0.35
JAN		-0.86	
FEB	-0.79		-0.33
MAR	-0.65	0.43	-0.29
APR	<b>0.69</b>	0.46	0.31
MAY	<b>0.65</b>		0.42
JUN			<b>0.88</b>
Expl.Var	3.03	1.58	2.40
Prop.Totl	0.30	0.16	0.24

Factor 2



Rivers  
State



# Distribution of monthly snow fall

## Factor 3 Scores distribution

Size of circle present Factor Scores

	Factor 1	Factor 2	Factor 3
SEP			<b>0.74</b>
OCT	0.46		<b>0.69</b>
NOV	<b>0.69</b>		
DEC	-0.59	<b>-0.58</b>	-0.35
JAN		<b>-0.86</b>	
FEB	-0.79		-0.33
MAR	<b>-0.65</b>	0.43	-0.29
APR	<b>0.69</b>	0.46	0.31
MAY	<b>0.65</b>		0.42
JUN			<b>0.88</b>
Expl.Var	3.03	1.58	2.40
Pop.Totl	0.30	0.16	0.24

Factor 3

doi:10.1088/0964-0796/20/1/008  
Nature Precedings : doi:10.1088/0964-0796/20/1/008

-1.977 - -0.991

-0.991 - -0.11

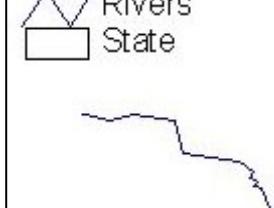
-0.11 - 0.81

0.81 - 2.261

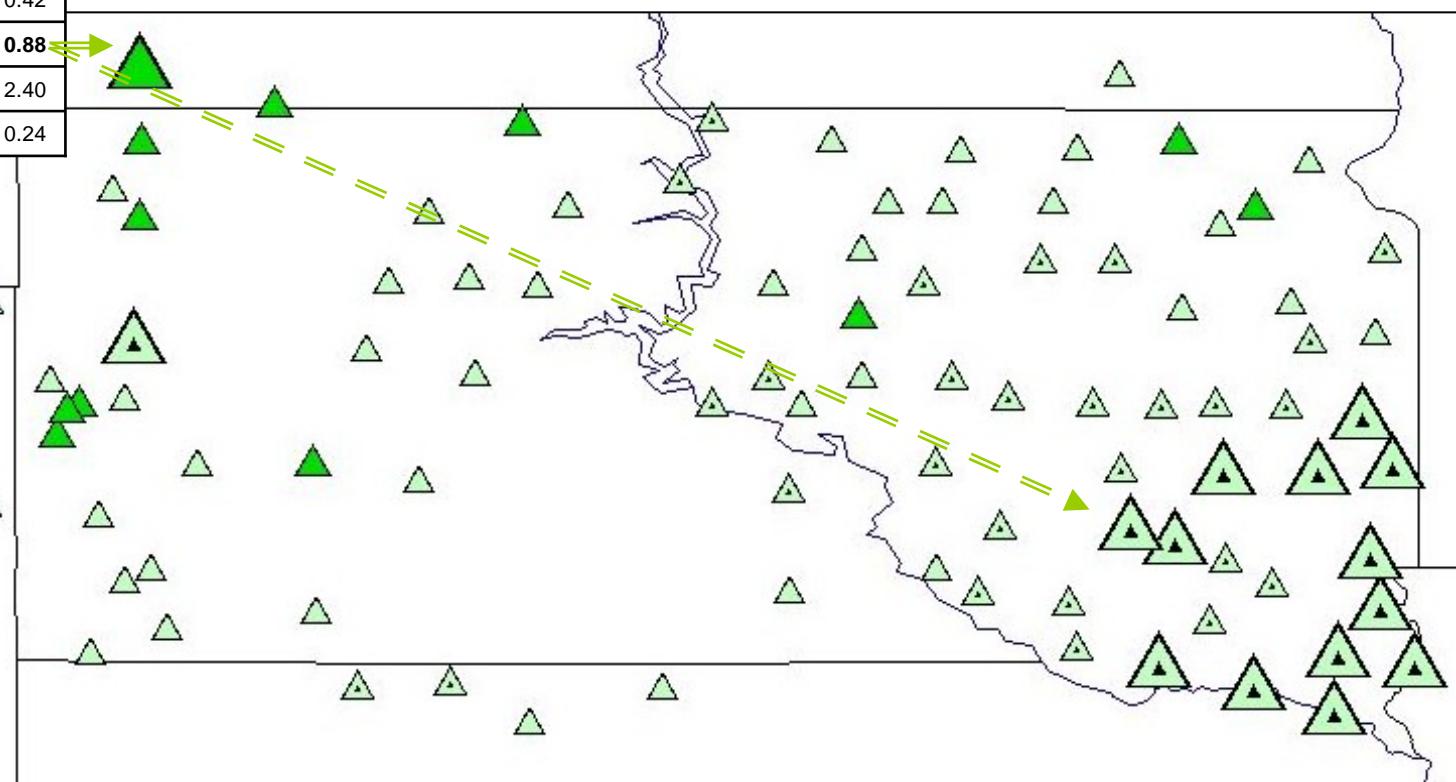
2.261 - 5.26

Rivers

State

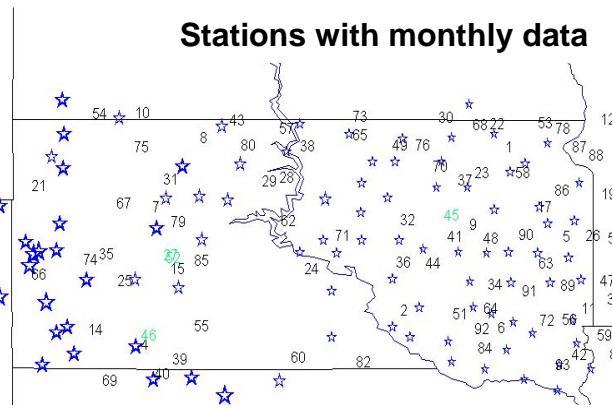


Rivers  
State

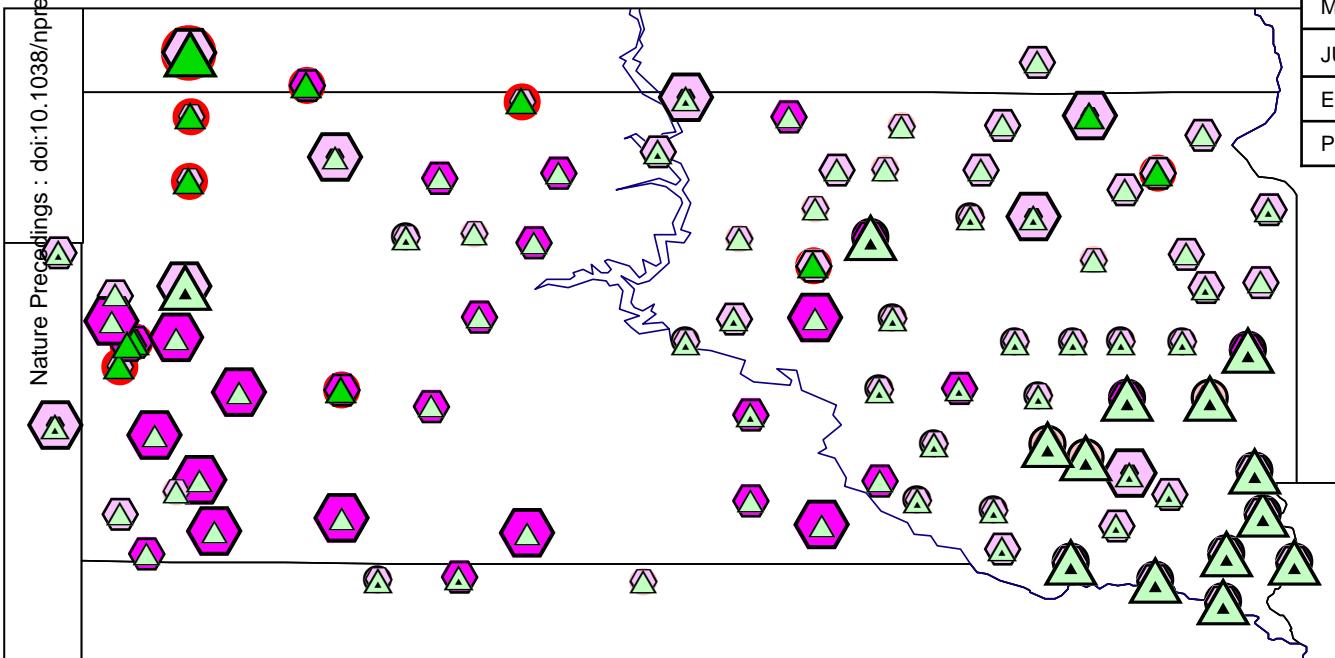


# Structure of monthly snow fall

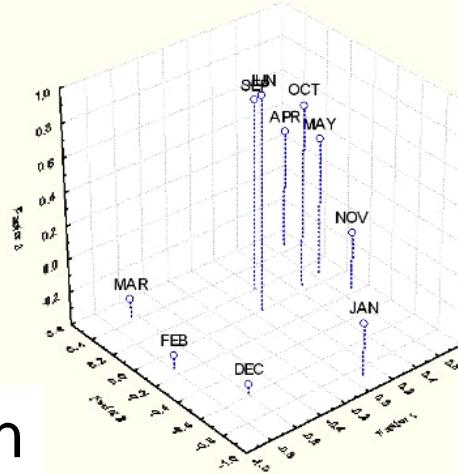
Size of circle present  
an altitude



Factor Scores distribution



3D Scatterplot Factor Loadings (Varimax normalized)



	[in]	Factor 1	Factor 2	Factor 3
SEP	.15			0.74
OCT	.66	0.46		0.69
NOV	3.6	0.69		
DEC	5.7	-0.59	-0.58	-0.35
JAN	4.9		-0.86	
FEB	7.2	-0.79		-0.33
MAR	7.6	-0.65	0.43	-0.29
APR	4.9	0.69	0.46	0.31
MAY	.57	0.65		0.42
JUN	.04			0.88
Expl.Var		3.03	1.58	2.40
Prp.Totl		0.30	0.16	0.24

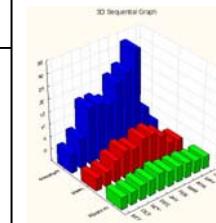
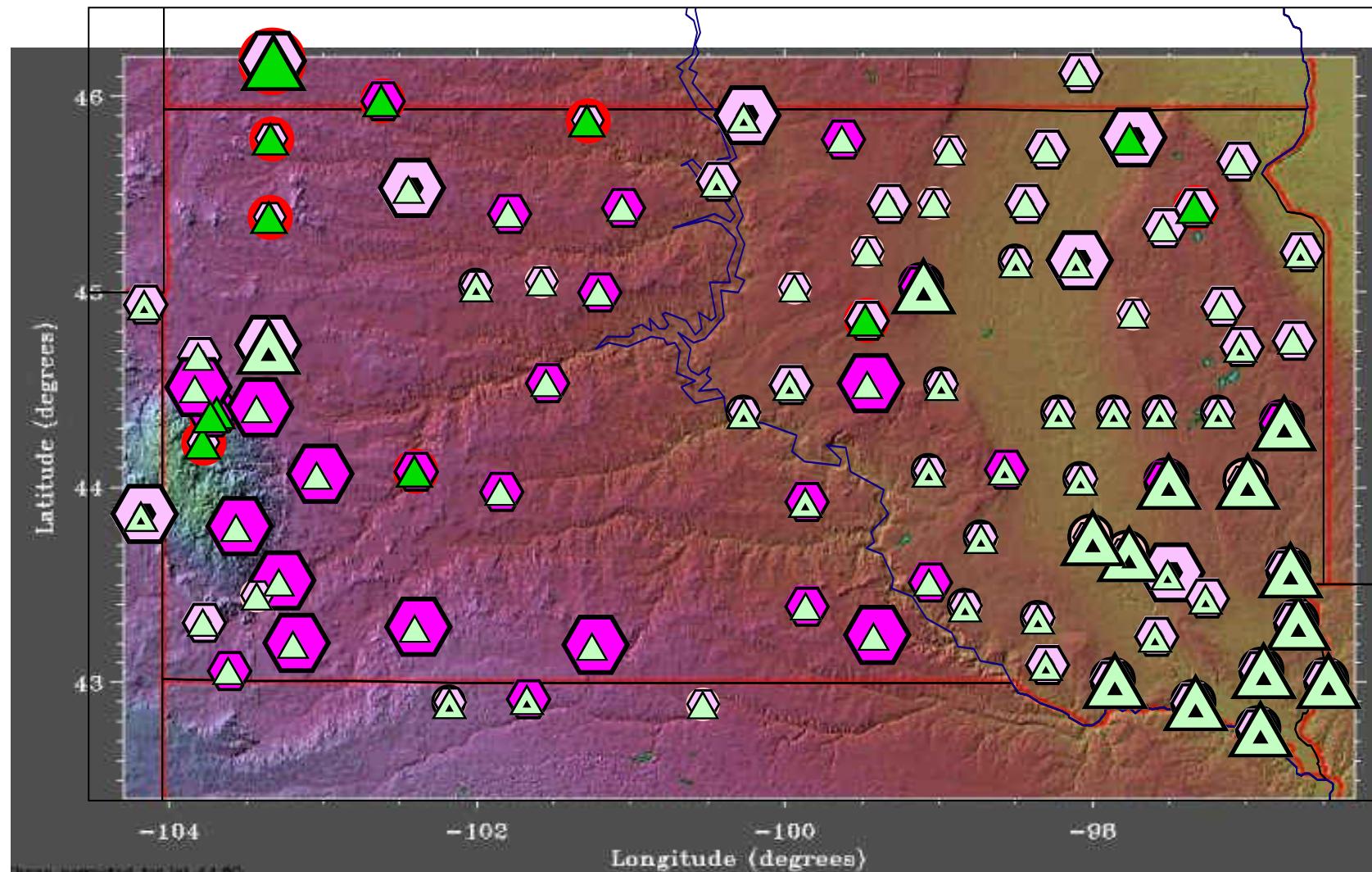
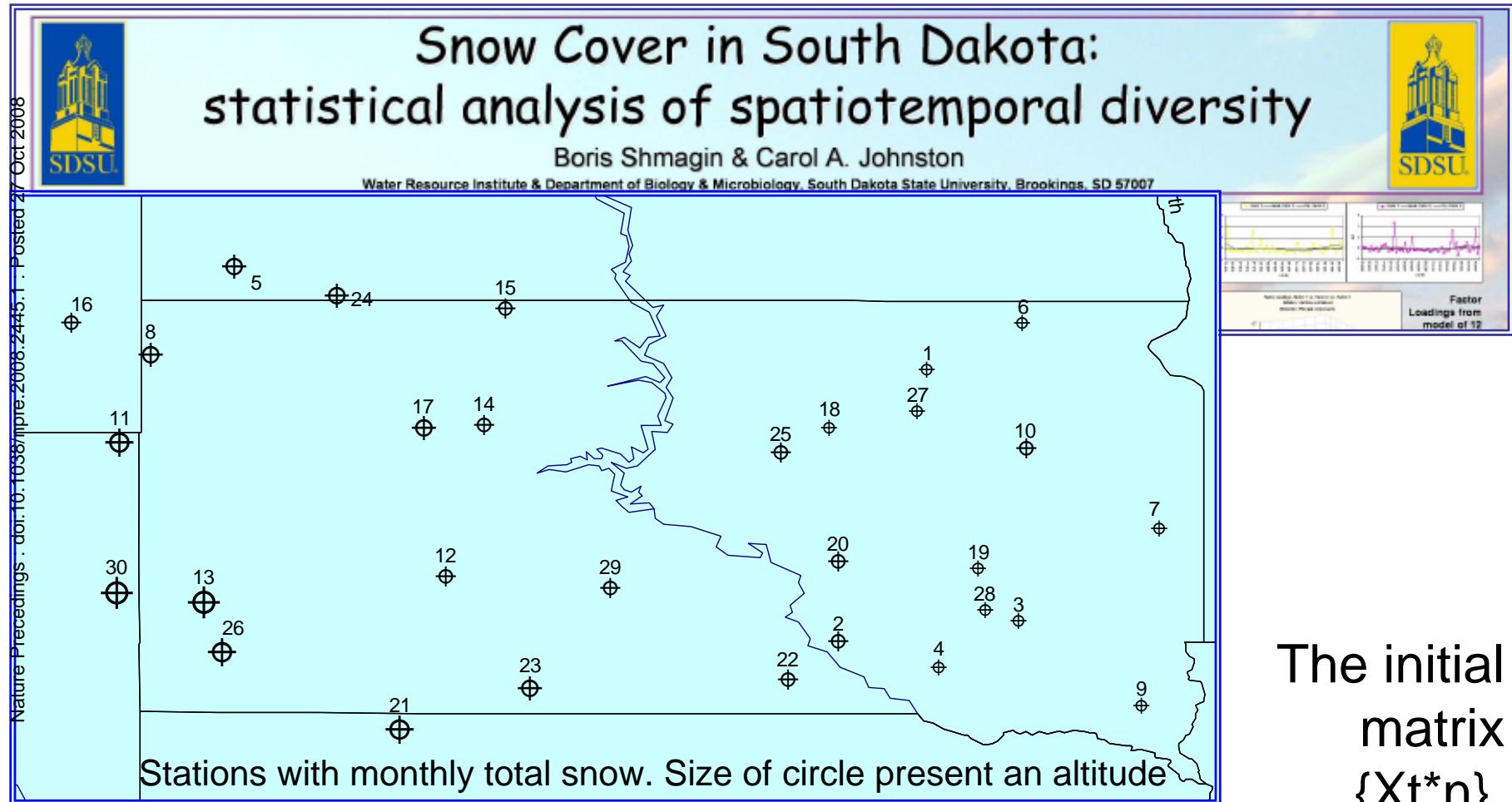


Table of  
average total  
annual snow fall  
& Factor  
Loading

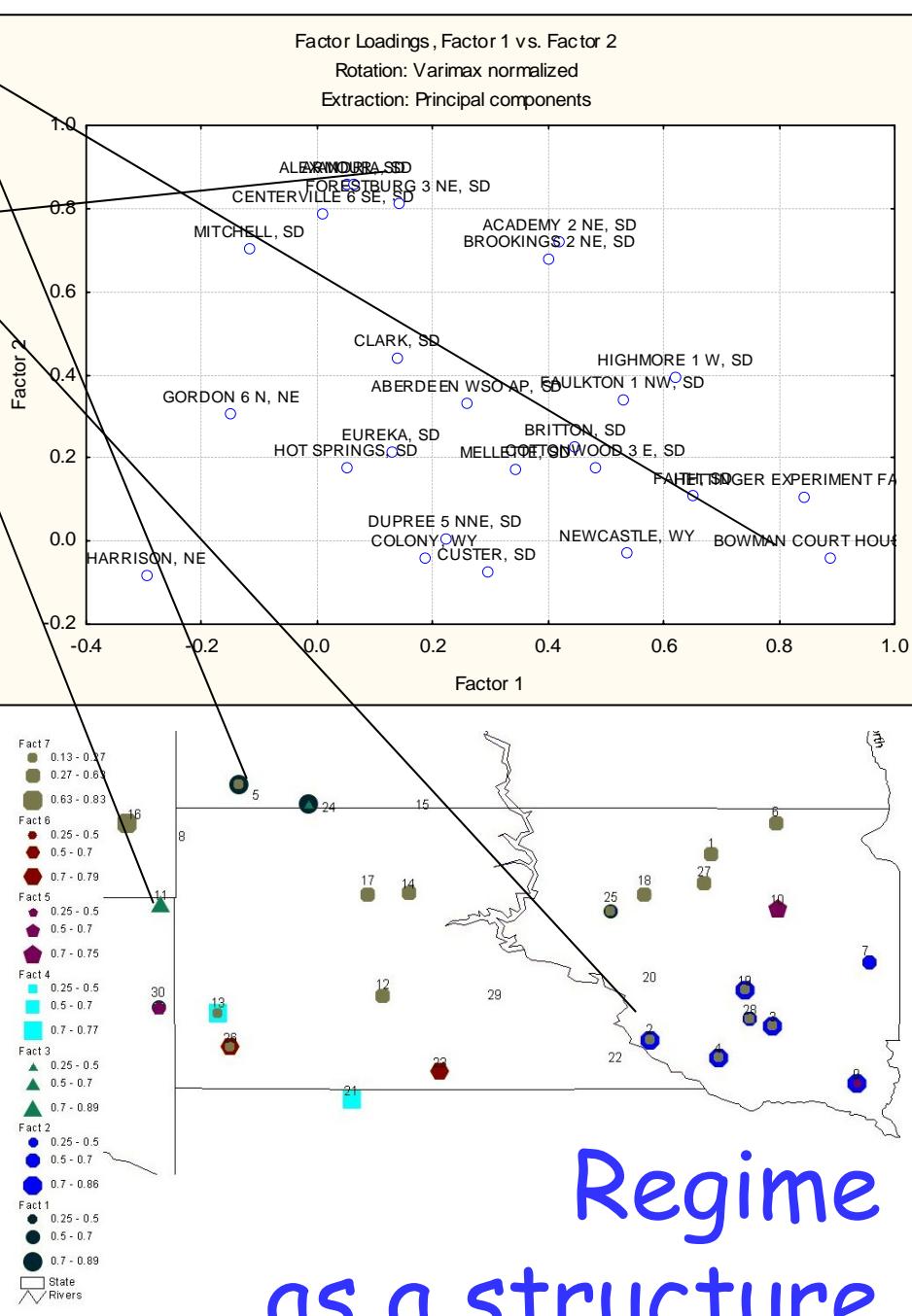
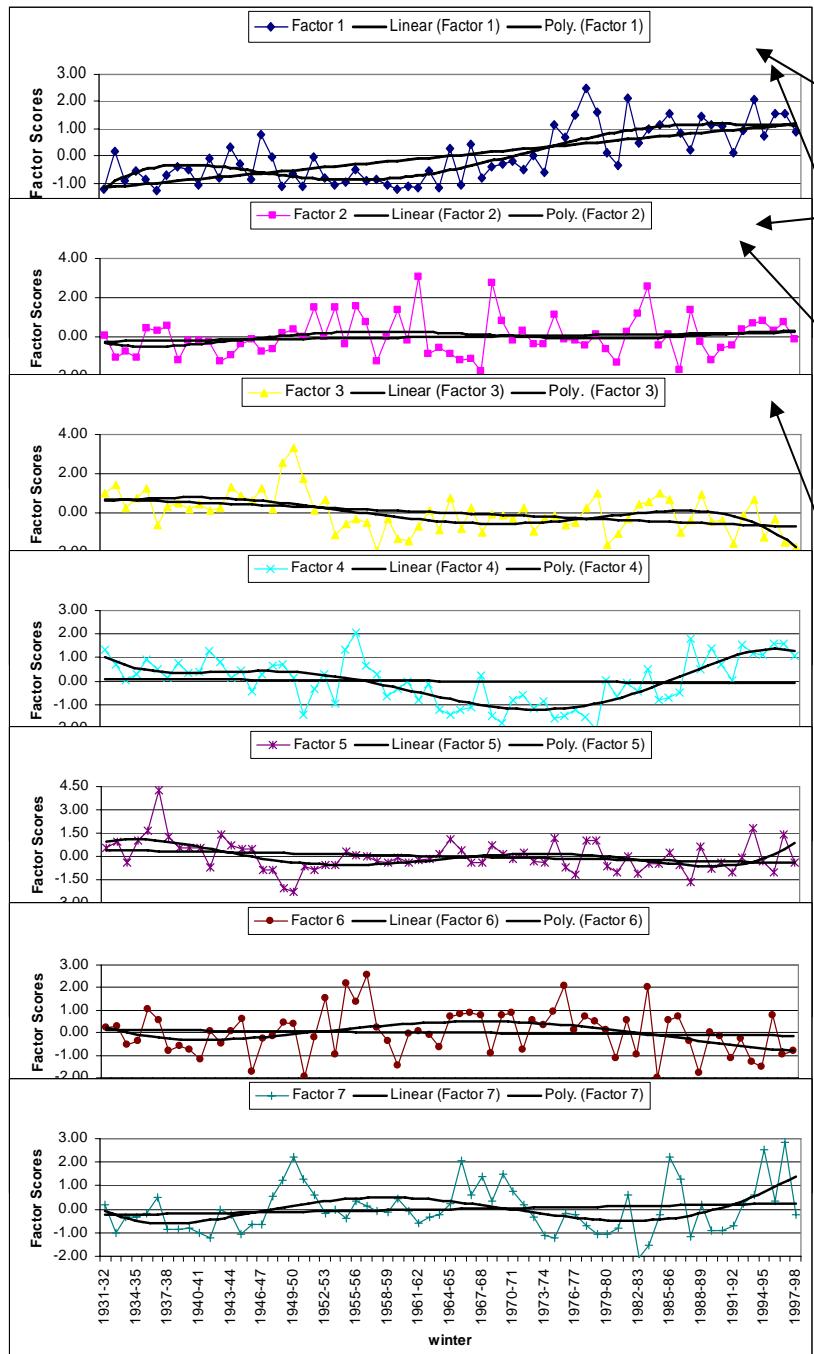
# Snow fall & landscapes of SD



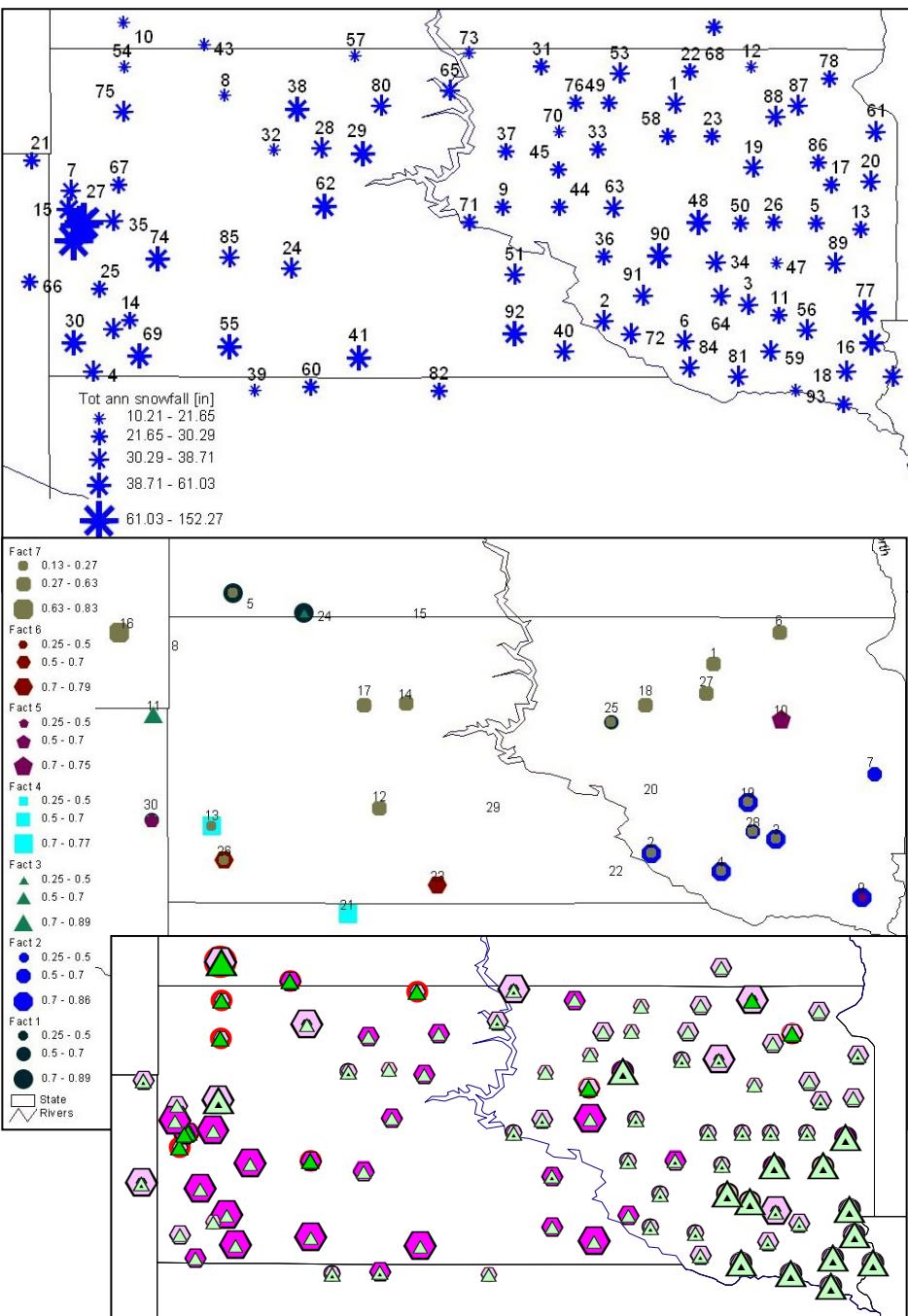
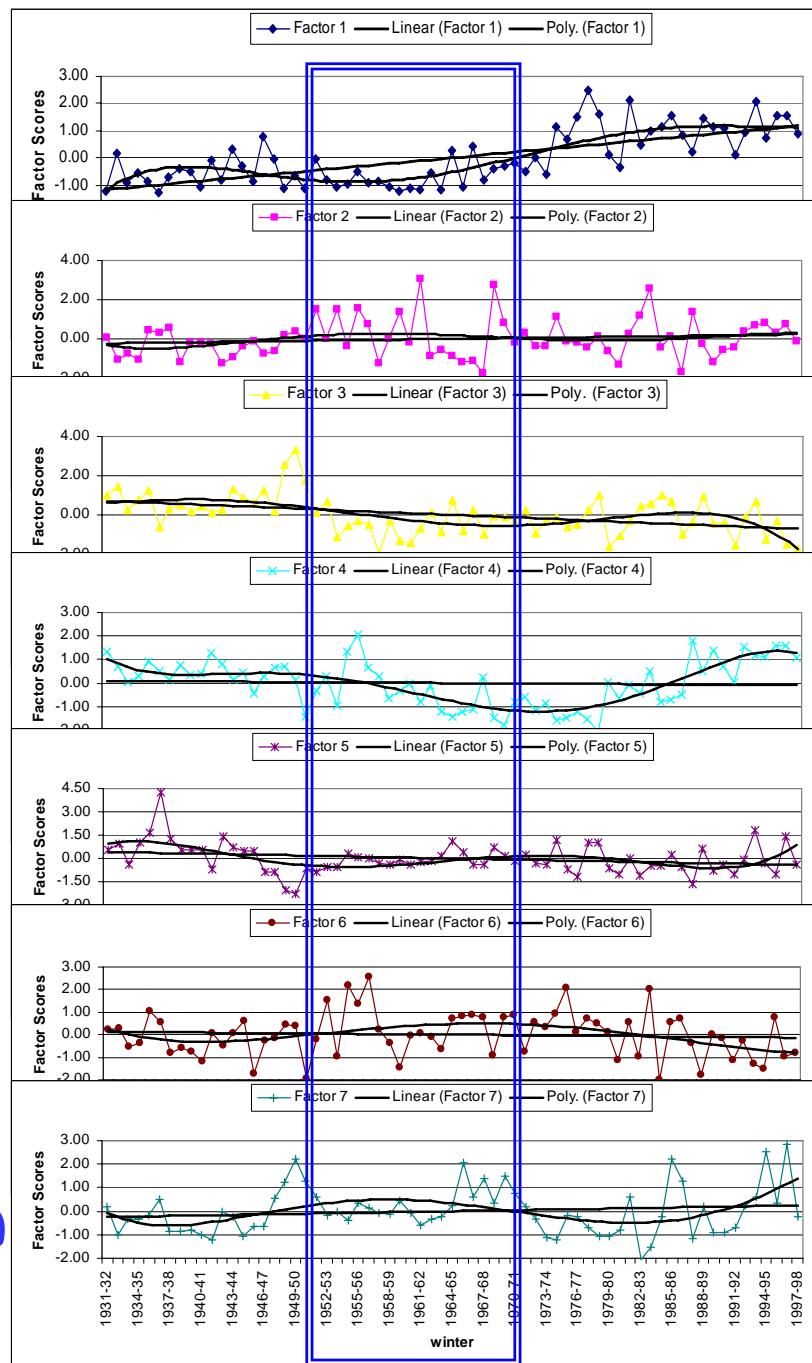
# Regime of snow fall in SD



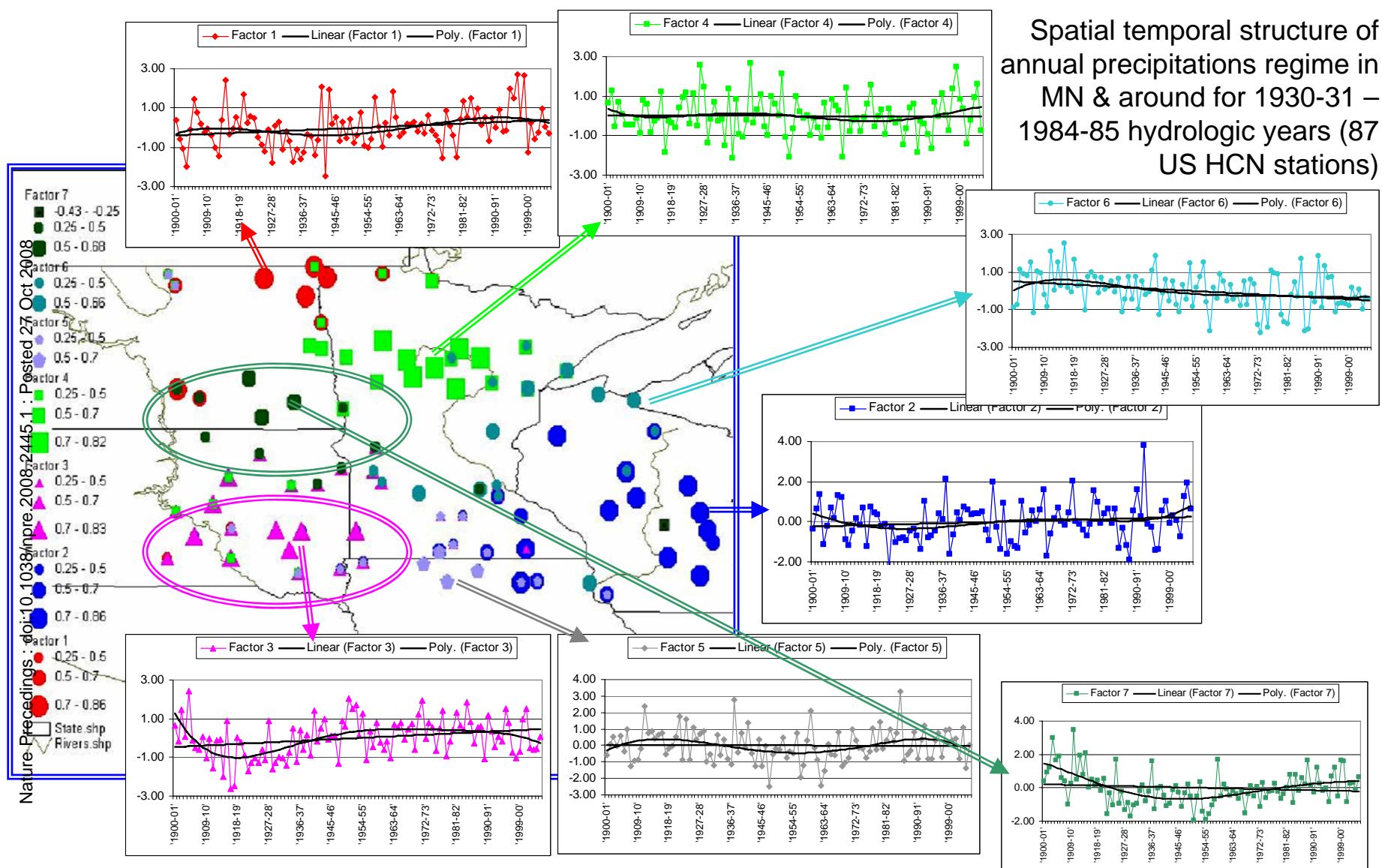
where are: t = number of years  
(67- 1931-1998) & n = number of meteorological stations (25),



# Regime & snowfall distribution



# Spatial temporal structure of annual precipitations regime in MN & around for 1930-31 – 1984-85 hydrologic years (87 US HCN stations)

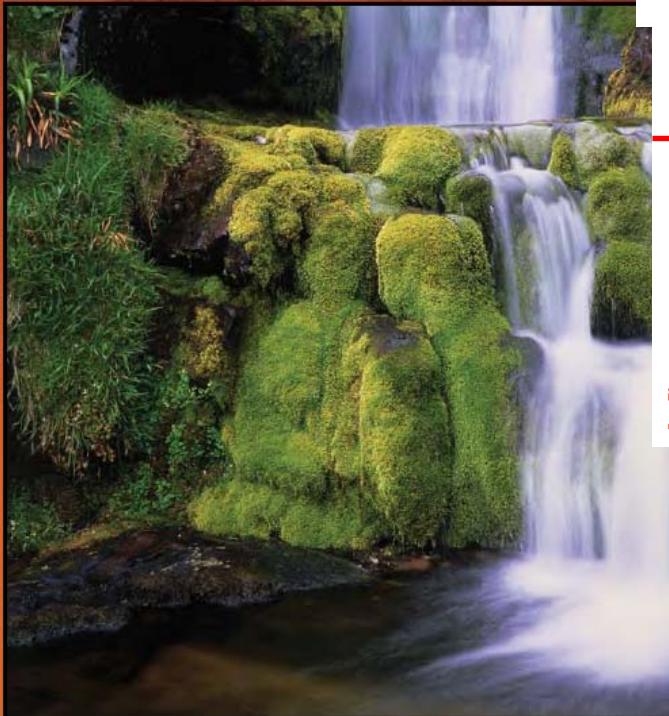


# Precipitation regime around SD

# Improving the knowledge & creating maps of natural resources of SD

# CLIMATE CHANGE AND WATER

IPCC Technical Paper



Observed warming over several decades has been linked to changes in the large-scale hydrological cycle such as: increasing atmospheric water vapour content; changing Current water management practices may not be robust enough to cope with the impacts of climate change on water supply reliability, flood risk, health, agriculture, energy and aquatic ecosystems. In many locations, water management Several gaps in knowledge exist in terms of observations and research needs related to climate change and water. Observational data and data access are prerequisites for adaptive management, yet many observational networks are shrinking. There is a need to improve understanding and modelling of climate changes related to the hydrological cycle at scales relevant to decision making. Information about the water-

From  
Reports'  
conclusions



Intergovernmental Panel on Climate Change



# "... gaps in knowledge ..." - from Reports' conclusions

NOAA Satellites and Information  
National Environmental Satellite, Data, and Information Service

10/27/2008 : Posted 27 Oct 2008

10/23/2008 /npre.2008.2445.1 : doi:10.3138/npre.2008.2445.1

North Sno

SSD Snow and Ice Detection

Satellite Services Division  
Snow and Ice Products

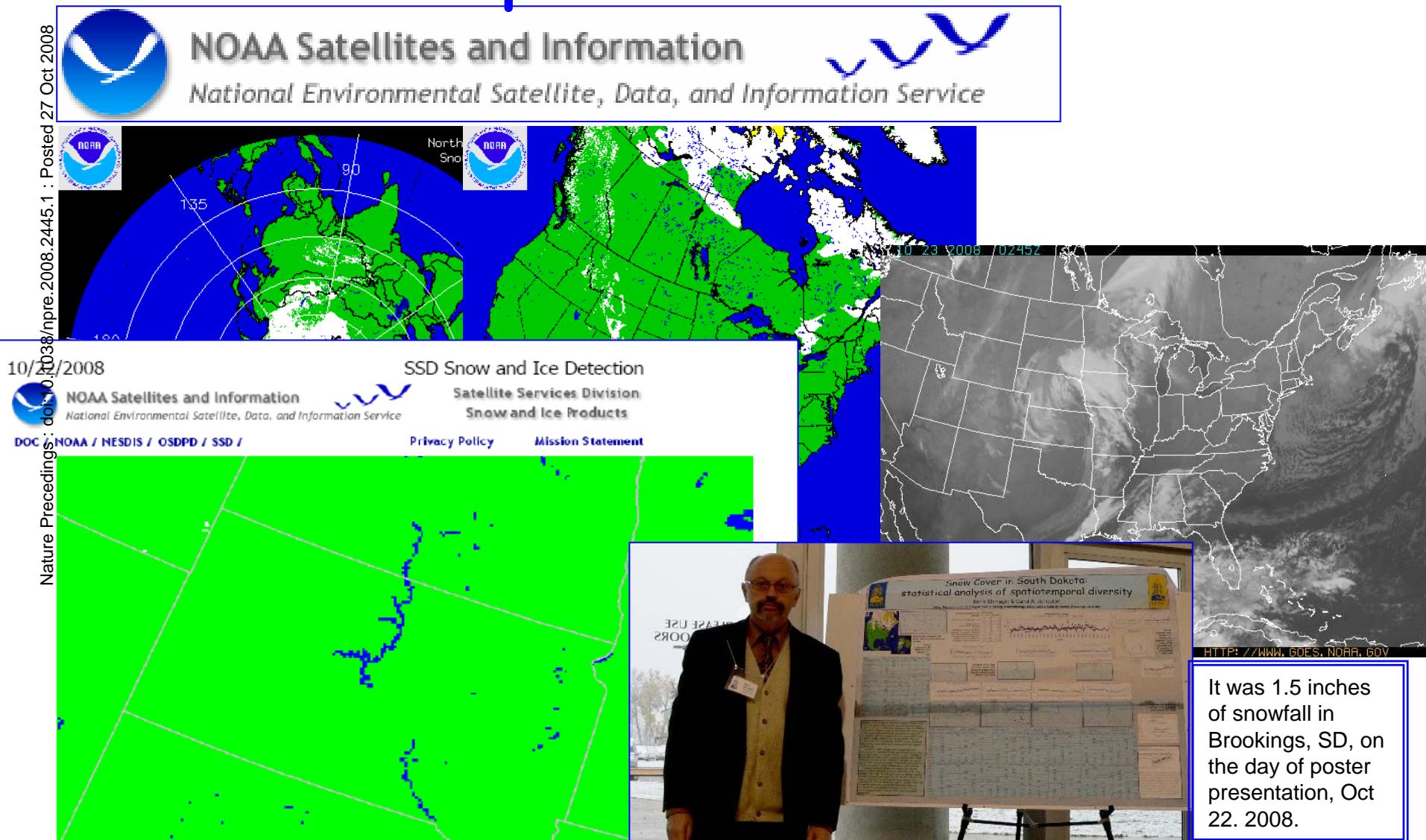
Privacy Policy   Mission Statement

DOC : NOAA / NESDIS / OSDPD / SSD /

Nature Precedings

It was 1.5 inches of snowfall in Brookings, SD, on the day of poster presentation, Oct 22. 2008.

HTTP://WWW.GOES.NOAA.GOV



# "... at scales relevant to decision making.."- from Reports' conclusions

The screenshot shows a Microsoft Internet Explorer window with the following details:

- Title Bar:** South Dakota Diversity of Temperature: Pictures from Statistical Analysis : Nature Precedings - Windows Internet Explorer
- Address Bar:** http://precedings.nature.com/documents/1082/version/1
- Toolbar:** Includes standard buttons for Back, Forward, Stop, Refresh, and Home, along with links for Google search and Page Tools.
- Navigation:** Jump to main content, Jump to navigation.
- Header:** natureprecedings logo, Pre-publication research and preliminary findings, a group of people illustration, Log in, Register.
- Menu Bar:** Home, Browse by subject, Advanced search, Search: All documents, GO.
- Content Area:**
  - Document Information:** hdl:10101/npre.2007.1082.1
  - Article Summary:** South Dakota Diversity of Temperature: Pictures from Statistical Analysis by Boris Shmagin<sup>1</sup> & Dennis Todey<sup>1</sup>. Correspondence: (Login to view email address).  
1. South Dakota State University  
[PDF \(4.3 MB\)](#)
  - Document Type:** Manuscript
  - Date:** Received 22 September 2007 15:18 UTC; Posted 24 September 2007
  - Subjects:** Earth and Environment
  - Tags:** multidimensional statistics, regional temperature, temperature seasonality
  - Abstract:** The regional diversity of monthly temperature was analyzed based on long-term data obtained for South Dakota (SD) from the High Plains Regional Climate Center. Multidimensional statistical methods were used and the principal results presented as a
- Right Sidebar:** Participate (Submit a document, My recent activity, Invite a colleague, Feedback), Frequently Asked Questions (What is Nature Precedings?, What is voting and who can vote?).

# Not from the Reports' conclusions

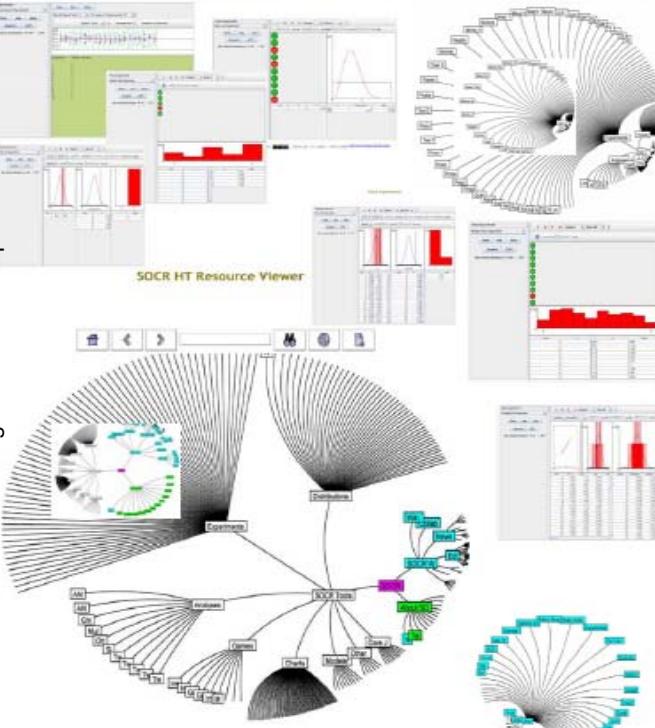
27 Oct 2008

Nature Precedings : doi:10.1038/npre.2008.2445.1 : Posted

## Academic Course as Research Seminar in Data Analysis on South Dakota Climate & Water Resources (for undergrad & graduate students)

Boris Shmagin & Din Chen

Department of Agricultural & Biosystems Engineering & Department of Mathematics and Statistics , South Dakota State University, Brookings, SD 57007



SDSC Resource Visitor number: 16,553, since Jan. 01, 2002 | 1997-2008 dinov@stat.ucla.edu



### Abstract:

Data analysis allows quantitative description of the behavior of natural and human dynamic systems & opens a way to understand their structure. Seminar style academic courses are fundamental tools for creating the craft atmosphere & research culture in the university. The seminar course "Climate & Water Resources" based on the use of special prepared reasons (small initial matrix ( $X_{0,0}$ ) with special constrained properties). This matrix is used to introduce all main statistical conceptions & techniques for students (mean, variability, correlation – as descriptive statistics, & also clustering and factorization – as multidimensional and multivariate exploratory techniques).

The second novelty are the illustrations based on the webpage (<http://www.socr.ucla.edu/>) to visualize the associated statistical principles of probability & distribution along with examples of real data that is a more appropriate way for students with non-mathematical majors grasp the concept presented. The designed course introduces models & tools for analysis of empirical climatic, hydrologic or environmental (economic, medical & others) data for students & then provides the hands on experience involving techniques of data analysis for research.

The goal for students will be to produce a publication, oral or poster conference presentation or other presentation in accordance with the level of completed research (undergraduate, master or PhD). The student's reports will be the basis for evaluating success in learning. Analyses completed by students may be used in civil and ag engineering projects in SD for improving the normative base in the state, & also in research that require descriptions of SD's natural conditions like agriculture & wildlife.

### SD Climate & Water Resources: Analysis of Existing Empirical Data

- PREFACE -- Course as a seminar for undergraduate & graduate students
- INTRODUCTION: Basic Data & Statistical analysis
- DATA PREPARATION ANALYSIS

  - Table of data (matrix of the initial model)
  - Parameters of sampling from five definitions.
  - Table of the parameters of variables.
  - Variables.
  - Histograms of standardized data.
  - The graphs of variables connected from the number.
  - The method of factor analysis according to Anderson & Fisher tests.
  - PATRIOT REGRESSIONS & CORRELATION WITH MATRIX

- Simple linear model.
- Variables of the dependent variable and the coefficient of paired correlation.
- Table of the independent-dependent variable ratio.
- Regression & correlation for four pairs of variables.
- Regression & correlation for the entire set of variables.
- Correlation matrix.
- Correlation matrix for samples from 15 different states.

### MULTIPLE REGRESSION & CORRELATION

- Multiple regression model.
- Regression analysis for XSDR from 7 remaining variables.
- The results of the step-by-step regression for the variables XSDR & X1DR from the rest.

### VI. CLUSTER ANALYSIS

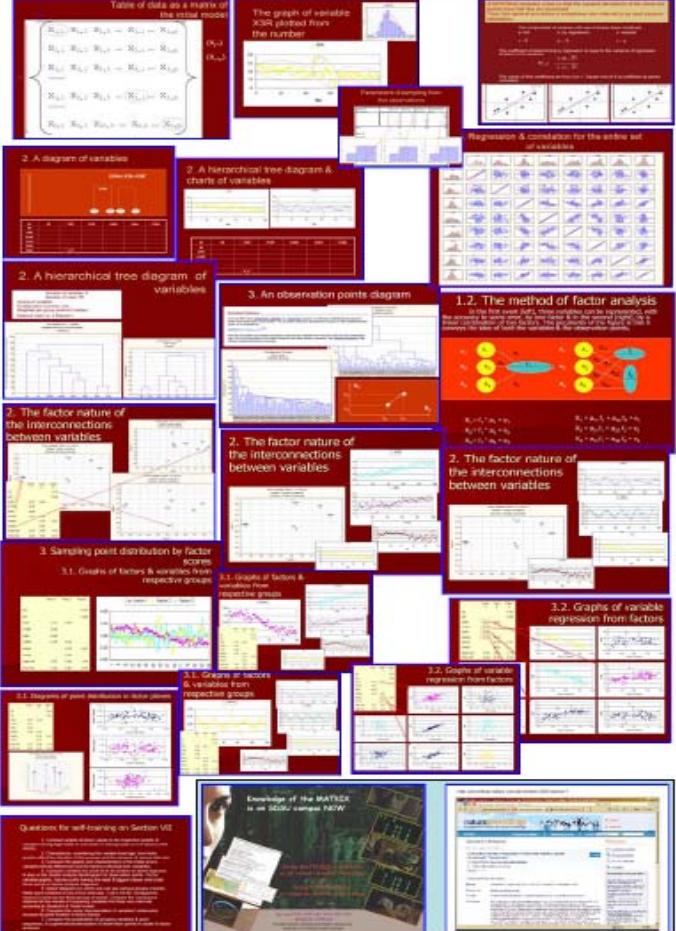
- The models & procedures of cluster analysis.

### VII. FACTOR ANALYSIS

- Analysis of factor analysis.

### VIII. CONCLUSION AND PRESENTATION OF EVERY STUDENT RESULTS ON SEMINAR

- Dr. Boris Shmagin teaches courses for SDSU as bases of similar courses: "Systems analysis of hydrological data" he was teaching for graduate students in Hydrology for eleven years (1985-1996) Department of Hydrology Faculty of Geography Moscow State University, Russia. For courses developed & published textbook, & for one time (Ged 5281 Water Quarter 1998 with Dr. S. K. Tolosa) he was invited to teach at LUMS, Lahore, Pakistan.
- Dr. Dennis Tolley is the state climatologist and regularly deals with practical issues of climatic data, including quality control & analysis of trends in such data. He teaches a 300-level class in climate to team-teachers & graduate level class in microclimatology.
- Dr. Boris Shmagin is a Professor in Geosciences in Department of Mathematics & Statistics, South Dakota State University.



# For discussion

Regional analysis of empirical hydro & climate data has to be the direction of SDSU research.

The goal is Atlas of climate & natural resources of South Dakota.

A photograph of a sunset over a body of water. In the foreground, the dark silhouettes of palm fronds frame the scene. The sky is a vibrant mix of orange, yellow, and pink hues, with the sun low on the horizon. The water reflects these colors. A small, bright white object, possibly a boat or a buoy, is visible in the distance.

# Questions

## Acknowledgement

Research project is privately funded.

Additional support & funding for presented research provided by  
Prof. Carol Peterson, SD EPSCoR &  
Water Resource Institute of SDSU