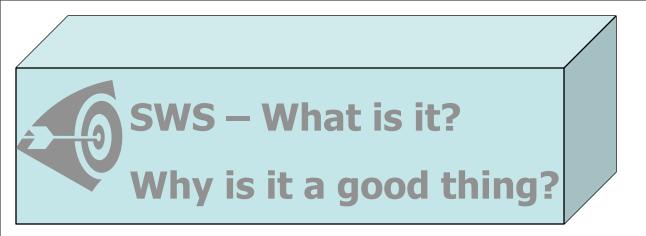
Statistical Methods/ Programs to Support Water Supply Forecasting

**CBRFC October 2006** 

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## The current software package of inter-related programs was initiated in 1992 It is called SWS or Statistical Water Supply

Many renditions later...



 SWS – Statistical Water Supply - a package of inter-related programs to support water supply forecasting

 Monthly data – reap the benefits of the relational database (library of functions as well as standard SQL methods)

 Ancillary programs – take advantage of many programs to report and manipulate monthly data

 Companion to ESP – "Super Ensemble" – one or more models to forecast the same thing (model diversity)

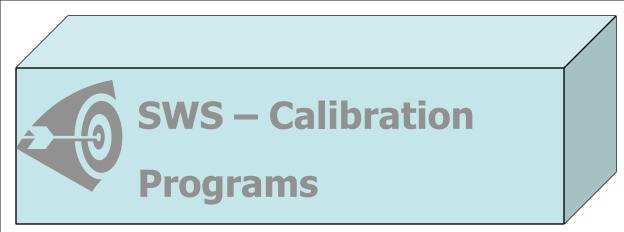
 Ease of use has been (and will be) a continuous priority during software development

The often used phrase: "wouldn't it be nice if..." – features are more easily accommodated/incorporated as the software development environment and working environment are the same



## I. CalibrationII. Operation





- Regcand find candidate variables, calculate correlation matrix
- Regdata prepare input file for regcomb
- Regcomb find optimal equations (top 20)
- Epal Evaluate, Pick and Load equations to the postgres db structure
- Cstats Calculate and store period statistics

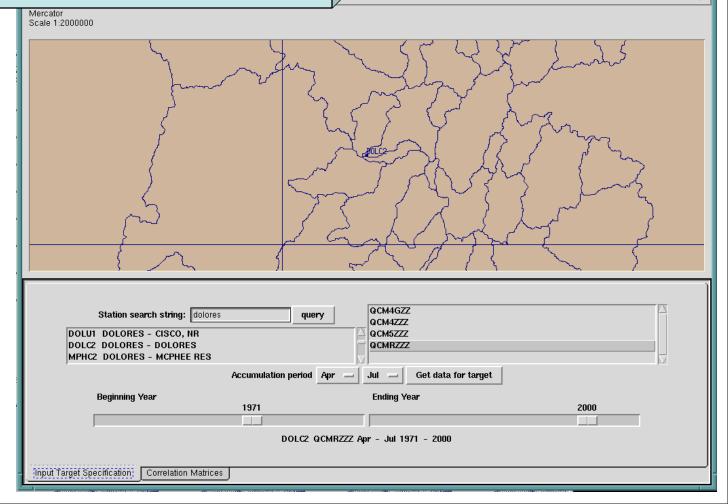
- Equation # 3
  - Y 1 DOLC2/QCMRZZZ, Ap-J1, DOLORES DOLORES
    - = -5.203
- + 14.859 X4 HSPC2/PPMRZZZ,No-Ja,FORT LEWIS
- + 57.175 X5 DRRC2/QCMRZZZ,Ja,DOLORES RICO, BLO
- + 10.161 X6 LNCC2/SWIRMZZ, Fe, LONE CONE
- Number of observations used = 30
- Number of principal components used = 1
- CORRELATION COEFFICIENT (R) =0.709STANDARD ERROR =80.055 (rank = 4)JACKKNIFE CORRELATION COEFFICIENT =0.674JACKKNIFE STANDARD ERROR =84.035JACKKNIFE BIAS: above average flow =-56.063 (14 obs.)
  - below average flow = 48.137 (16 obs.)

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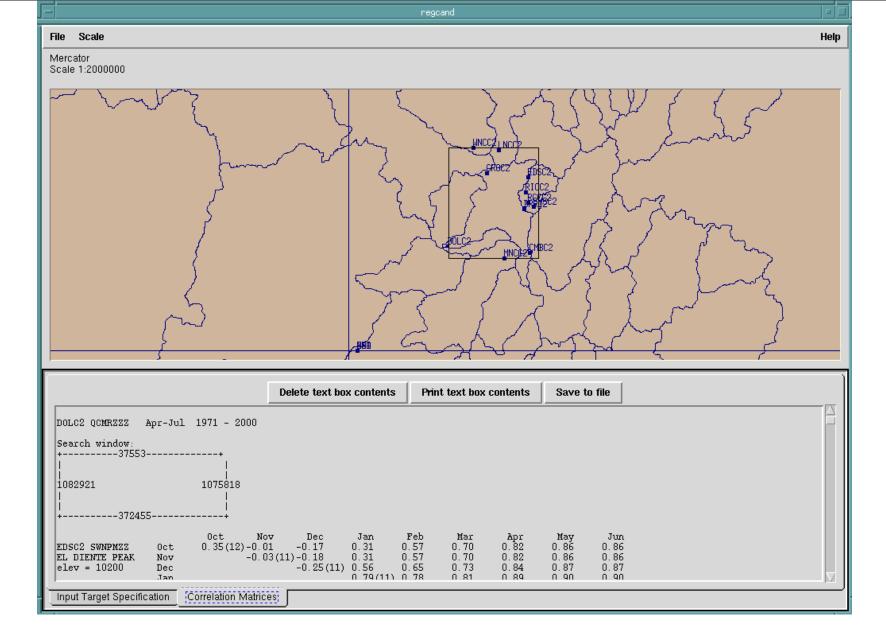
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ERROR	COMPUTED	ERROR	COMPUTED	OBSERVED	YEAR
-56.13	130.37	-51.29	135.21	186.50	61
-8.74	247.56	-7.40	248.90	256.30	62
51.63	178.03	48.72	175.12	126.40	63
-28.66	123.94	-26.15	126.45	152.60	64
32.20	368.00	30.38	366.18	335.80	65
98.37	304.57	93.46	299.66	206.20	66
129.02	252.52	124.87	248.37	123.50	67



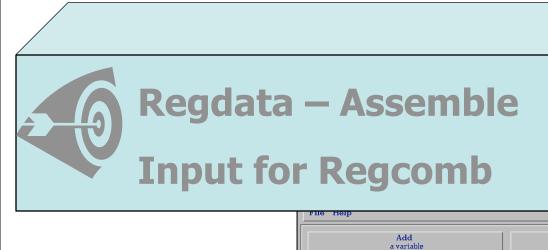
Lets user visualize spatial relationship of potential predictor variables. Calculates accumulation of Y var. Calculates correlation matrix for each independent variable; accumulations as well as discrete months.



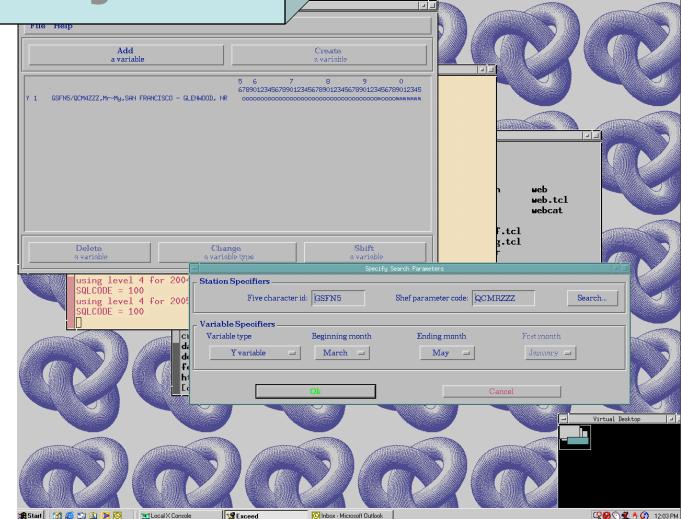
Help

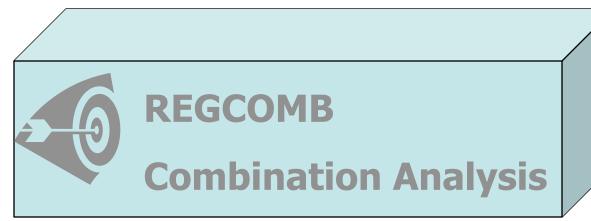


Future software development will include GIS where it is useful

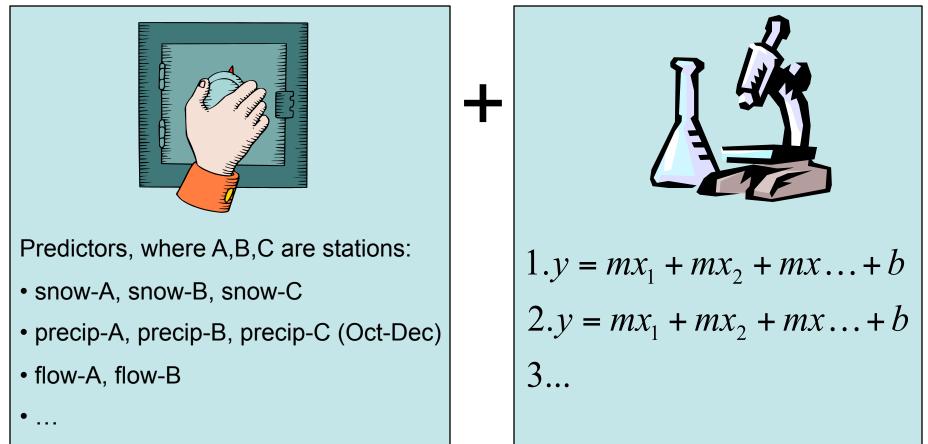


Extracts data from postgresql, accumulates as needed, builds formatted input file for Regcomb



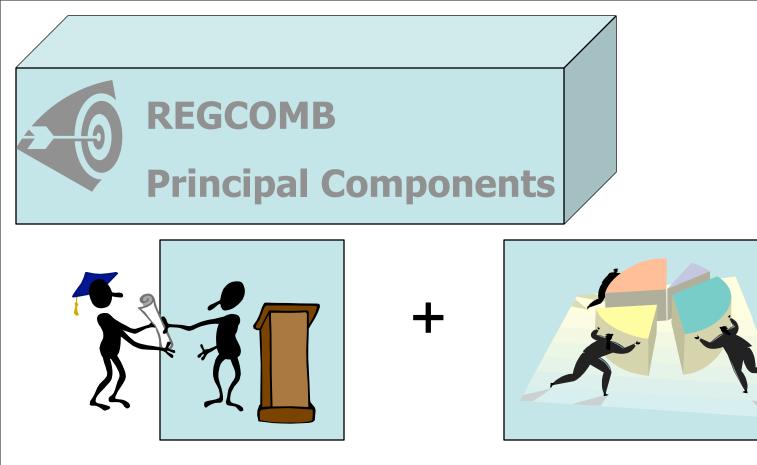


Why? ...there are over 500 million unique combinations of just 30 variables.

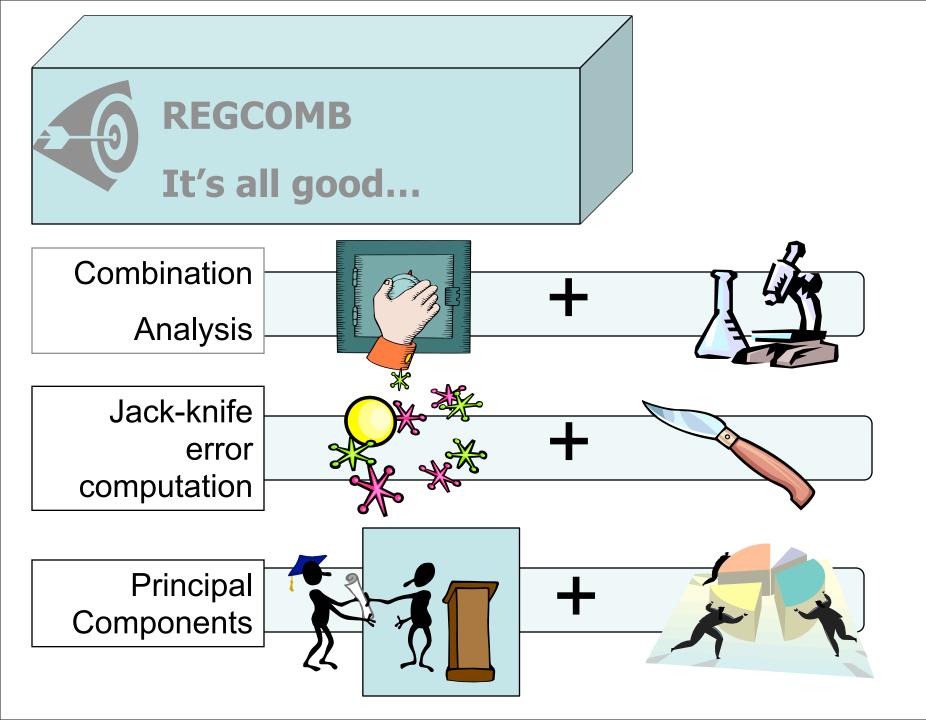


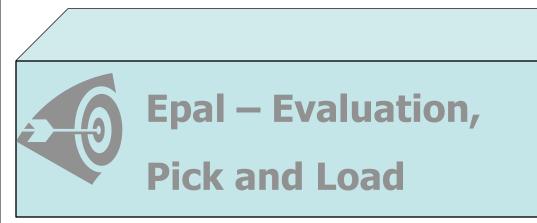


For a given set of predictors, one observation (one year) is deleted from the data set. Optimal coefficients are determined. The equation is then measured as to how well it predicted the selected year. The idea here is to simulate how well the equation will perform in an operational environment where the predictand is not known at the time of equation execution.

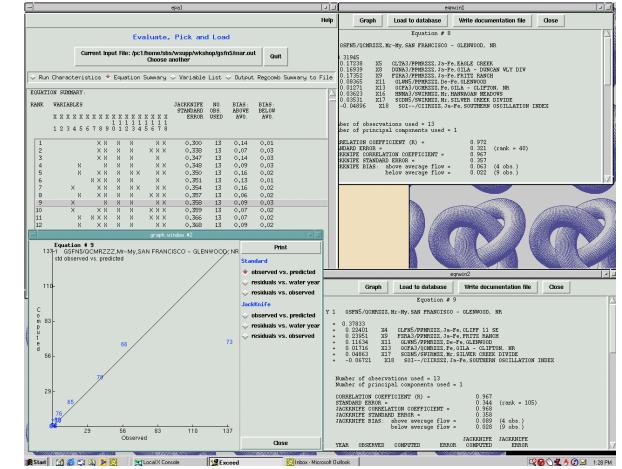


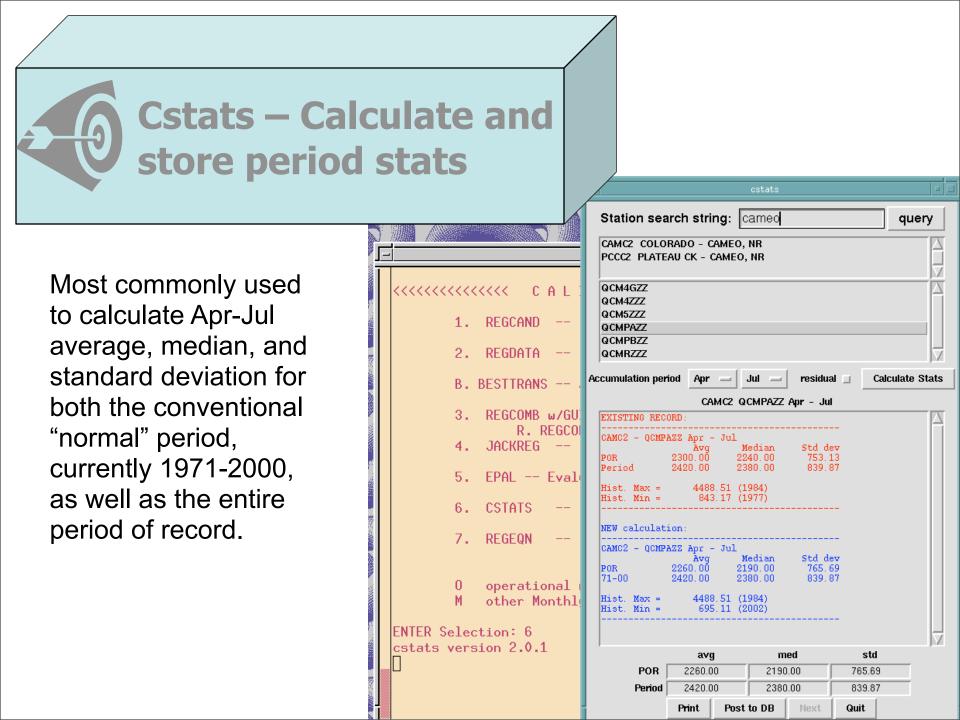
Variables in a water-supply equation tend to have high correlation with each other. This causes problems when trying to determine optimal coefficients via traditional regression techniques. Principal components analysis is a way to determine optimal coefficients while recognizing and addressing the intercorrelation problems.





This program allows the user to examine the output from REGCOMB and ultimately choose and store an equation to the postgresql database for operational use.

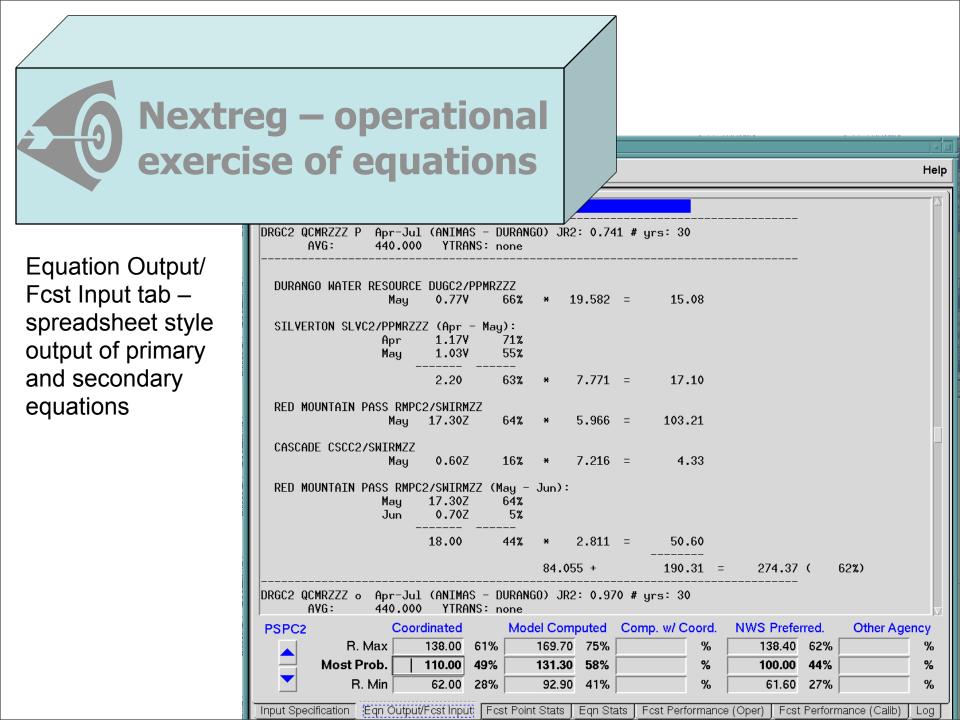








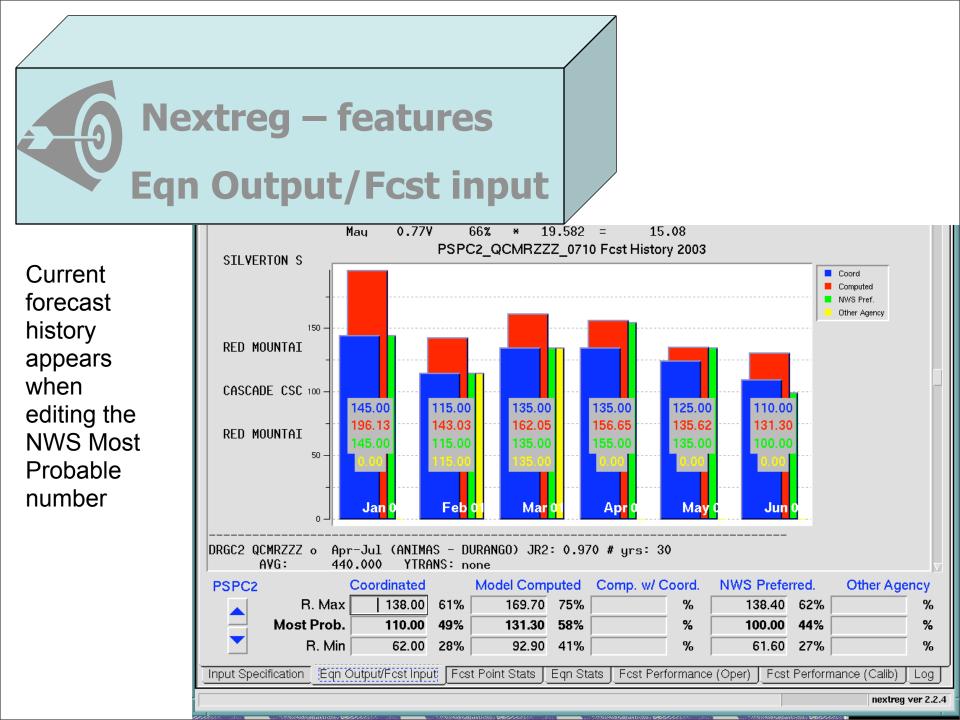
- Nextreg exercises equation with operational data and allows user to store forecasts in the database
- Nextpub extracts forecasts from database and outputs several different formats for various purposes

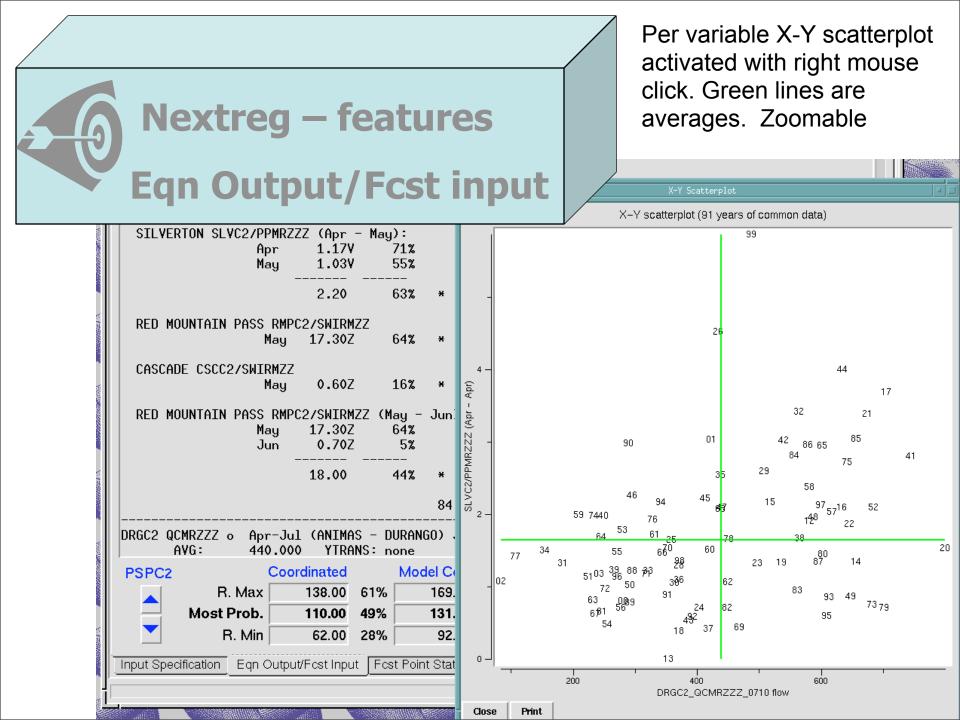


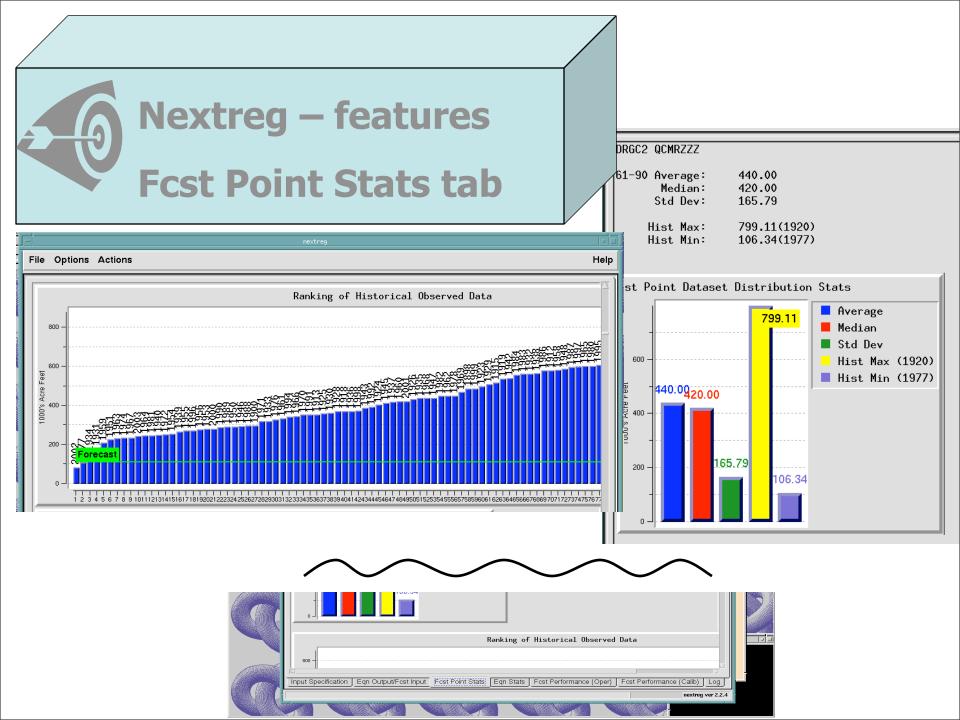


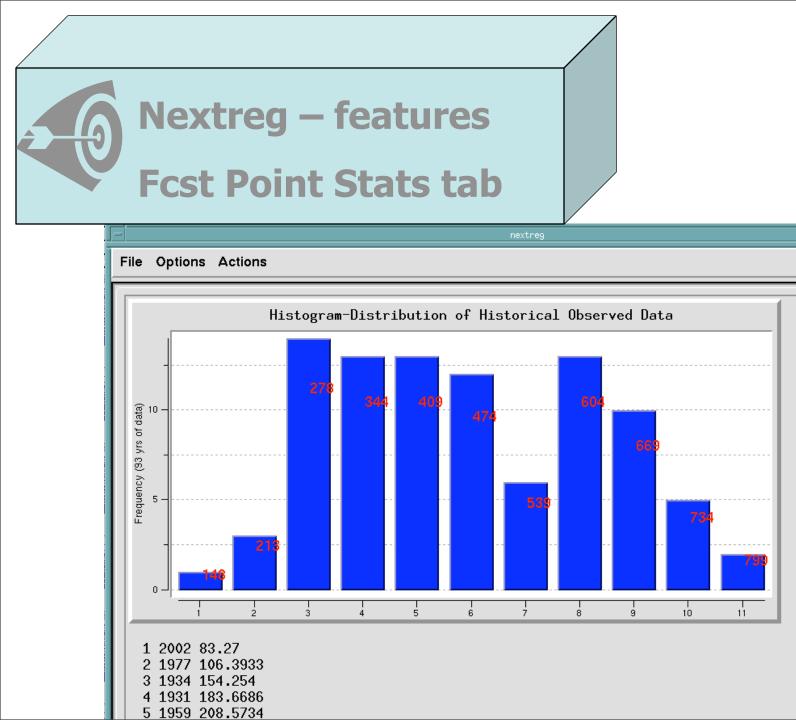
Historical Max/Min appears when entering the edit box for Rmax and Rmin

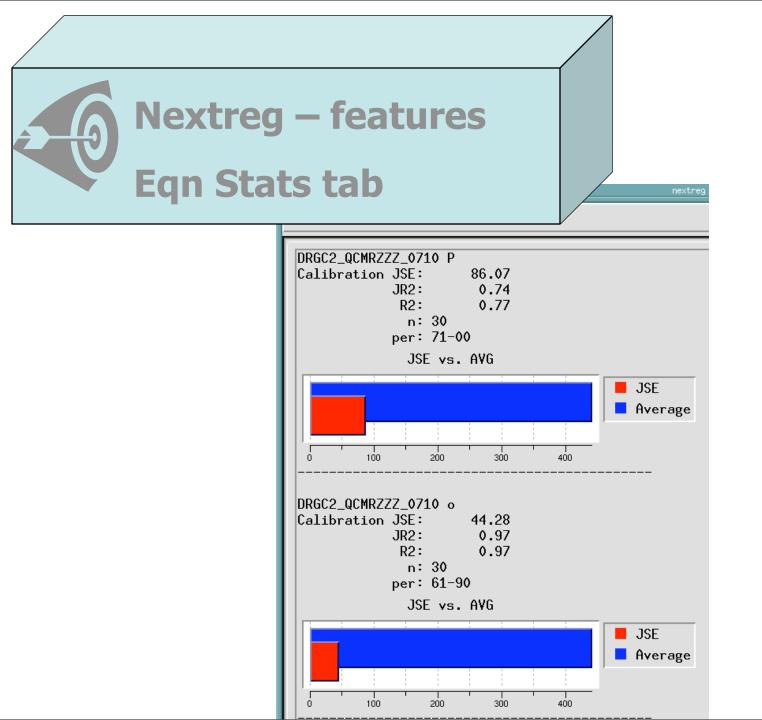
	RED MOUNTAIN PASS Ma Ju	iy 17.30Z	64	1%							
く語		18.00	44	- 1% * 2	2.811	=	50.60				
HISTORY .	84.055 + 190.31 = 274.37 ( 62%)										
	DRGC2         QCMRZZZ         39.54 (1977)         - DURANGO)         JR2:         0.970         # yrs:         30           AVG:         455.62 (1941)         S: none         S:         None         None										
	PSPC2	Coordinated	- N	Aodel Comp	outed	Comp	. w/ Coord.	NWS Prefer	rred.	Other Age	ncy
	R. Max	138.00	61%	169.70	75%		%	138.40	62%		%
1	Most Prob.	110.00	49%	131.30	58%		%	100.00	44%		%
会場	R. Min	62.00	28%	92.90	41%		%	61.60	27%		%
	Input Specification Eqn C	output/Fest Input	Fcst F	Point Stats	Eqn Sta	ts Fc	st Performanc	e (Oper) _ Fcst	Perforr	mance (Calib)	Log
										nextreg	ver 2.2.4

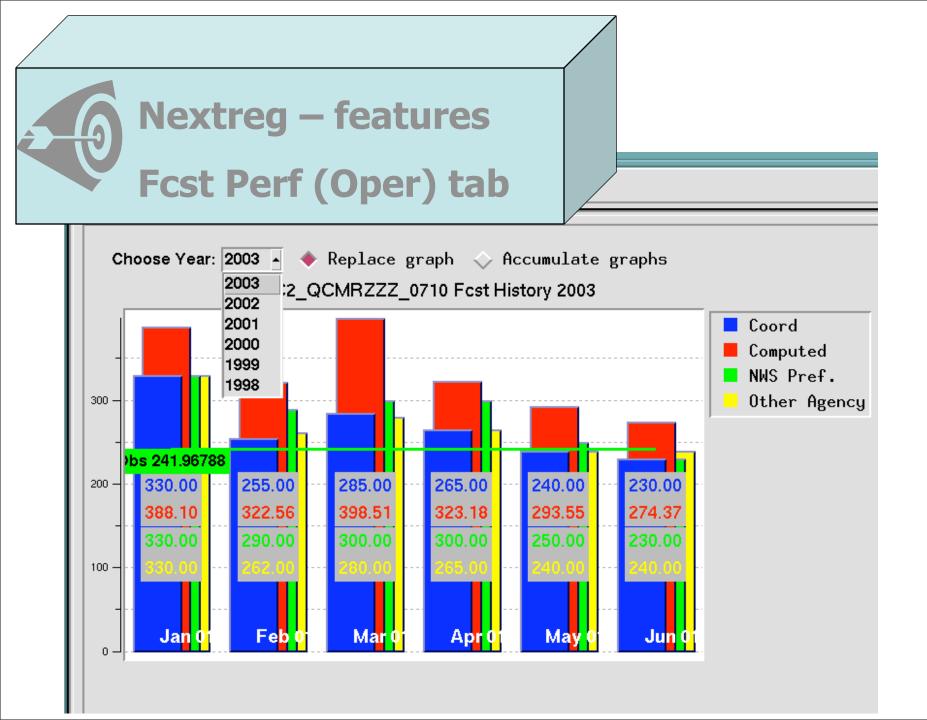




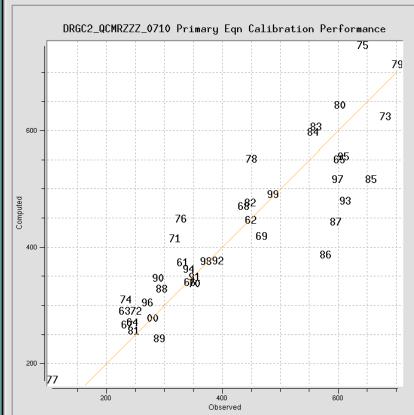


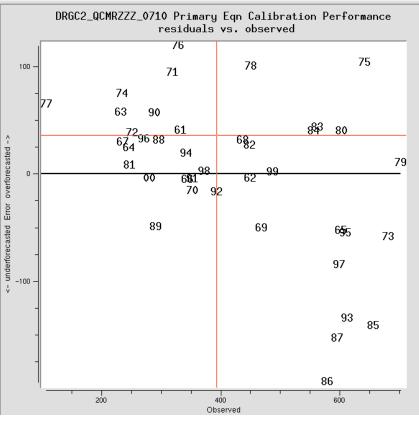


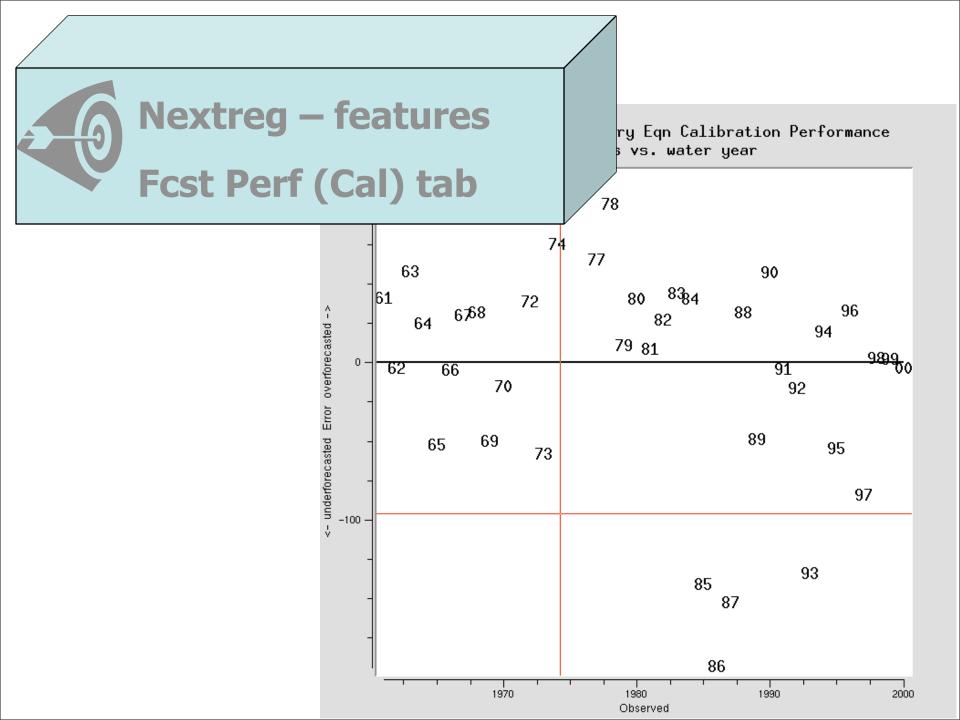










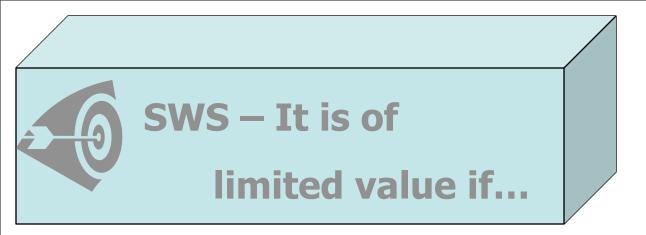




- Continue trend of reducing number of programs
- GIS ground has been broken with Regcand; continue this approach
- Revise database table structures
- Get others involved in programming/support
- Investigate/incorporate newer statistical techniques and new ways of looking at verification



- A package of beginning-to-end integrated programs for water supply forecasting, or really, any kind of statistical forecasting
- Monthly data stored in relational database
- Other programs that deal with data of a monthly time step
- Another way to forecast volume, in addition to ESP
  Ease of use
- Software has been polished by a lot of "wouldn't it be nice if..."'s



- There is no dominate driving force (like snowmelt)
- There is not a substantial period of record of data i.e. 30+ years
- The predictand data set does not closely approximate natural flow
- The predictors used in the equations are not recorded early enough in the month
- The predictors used in the equations are not recorded reliably month to month and year to year
- The ability to "time distribute" the forecast volume is required

Climate Change Forecasts as Input to CBRFC's Water Supply Models

- Historical MAP and MAT time series are adjusted relative to current climate outlooks before they are used as input into ESP.
  - The probability anomaly shift described by the CPC forecasts must be translated into real physical temperature or precipitation.
  - This is done through the simple method of computing the shift in the medians of the distributions of the temperature and precipitation based on the CPC forecasts and then using this shift to adjust all values of the time series.
    - temperature uses an additive adjustment
    - precipitation uses a multiplicative adjustment

## Multi-model Multi-agency Forecast Process

