



Automated Calibration of Watershed Models

David Radcliffe, Zhulu Lin, Mark Risse, and Rhett Jackson
University of Georgia

CSREES 2004 Integrated Programs Grant
“A Framework for Trading Phosphorus Credits in the Lake Allatoona Watershed”

Watershed-Scale Models

- Used in TMDL analysis
 - Estimate current pollutant loads (especially non-point source loads)
 - Estimate effect of BMP's in reducing loads
- Dynamic vs. static
 - *Static* models predict annual loads
 - *Dynamic* models predict daily or hourly stream flow and pollutant concentration

Watershed-Scale Models

- Lumped models
 - Group land-uses and soils into large units that flow directly into the watershed outlet
 - Parameters averaged over watershed
- Semi-distributed models
 - Divide watershed into sub-basins
 - Model flow from sub-basin to sub-basin
 - Within each sub-basin land-uses and soils lumped
 - Parameters can vary among sub-basins

Watershed-Scale Models

- Dynamic semi-distributed watershed-scale models
 - Soil Water Assessment Tool (SWAT)
 - Hydrological Simulation Program Fortran (HSPF)
 - Many parameters
- Models require calibration
 - Some parameter values determined by fitting model predictions to observed flow and pollutant concentration (usually at outlet)

Calibration

- Most dynamic watershed-scale models calibrated manually
- Software exists for automated calibration
- Parameter Estimator (PEST) software developed by John Doherty
 - Can be used with any Fortran code model
 - Reads input and output files
 - Finds optimum values and does sensitivity analysis
 - Requires starting values and may not find global “best-fit”
 - Free software
 - Can be run with HSPF (but not SWAT) in BASINS 3.1?

Objectives

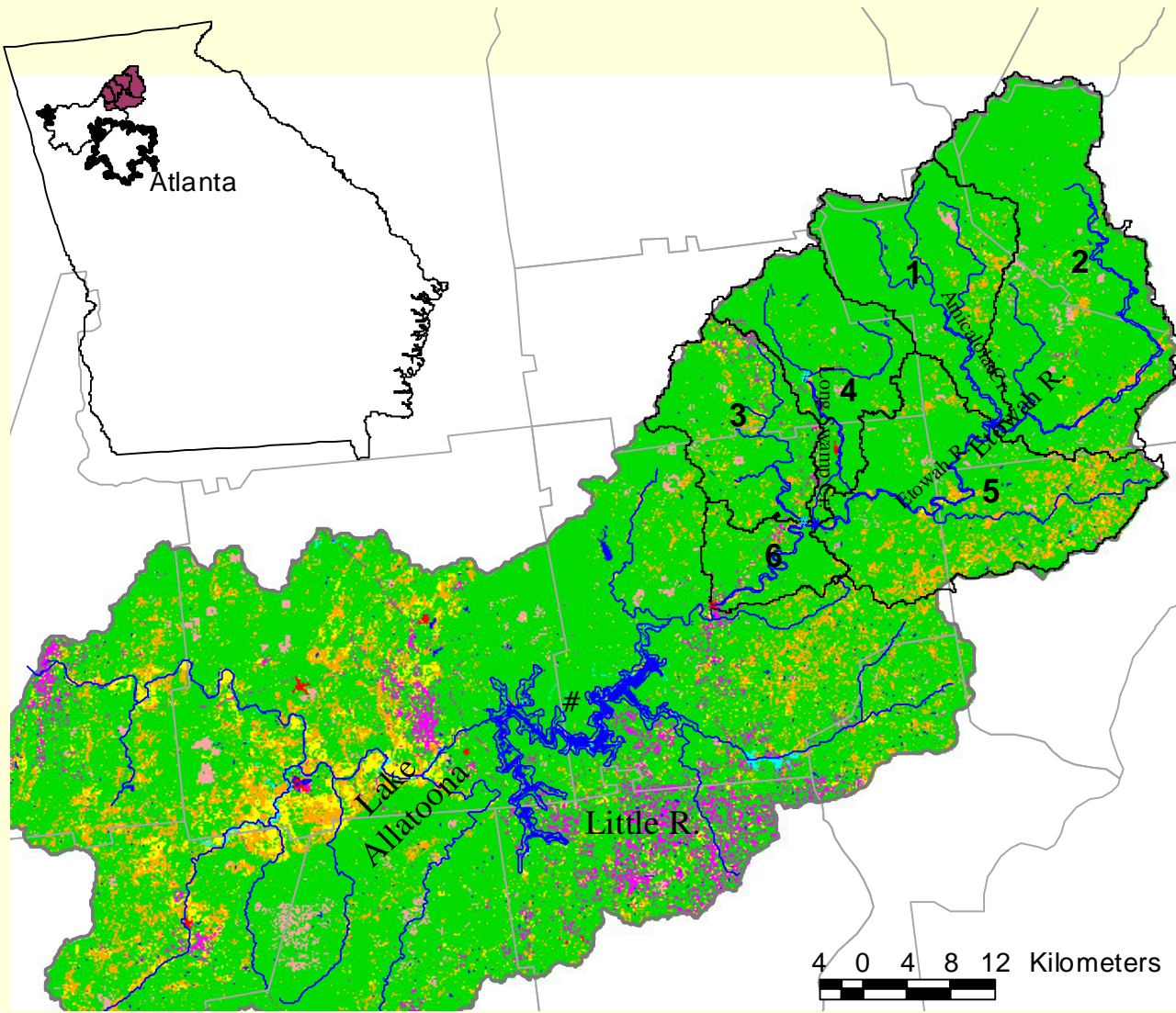
- Short-term: perform automated calibration
 - Using SWAT and PEST
 - Upper Etowah River watershed
 - Daily water flow available 1983-1993
- Short-term: determine uncertainty of model predictions
 - Automated calibration is first step in process
 - Put confidence intervals (CI's) on model predictions of annual and daily flow
- Long-term: use CI's to determine trading ratios between point and non-point sources

Outline

- No calibration
- Manual calibration
- Automated calibration
- Model uncertainty

SWAT Set-up

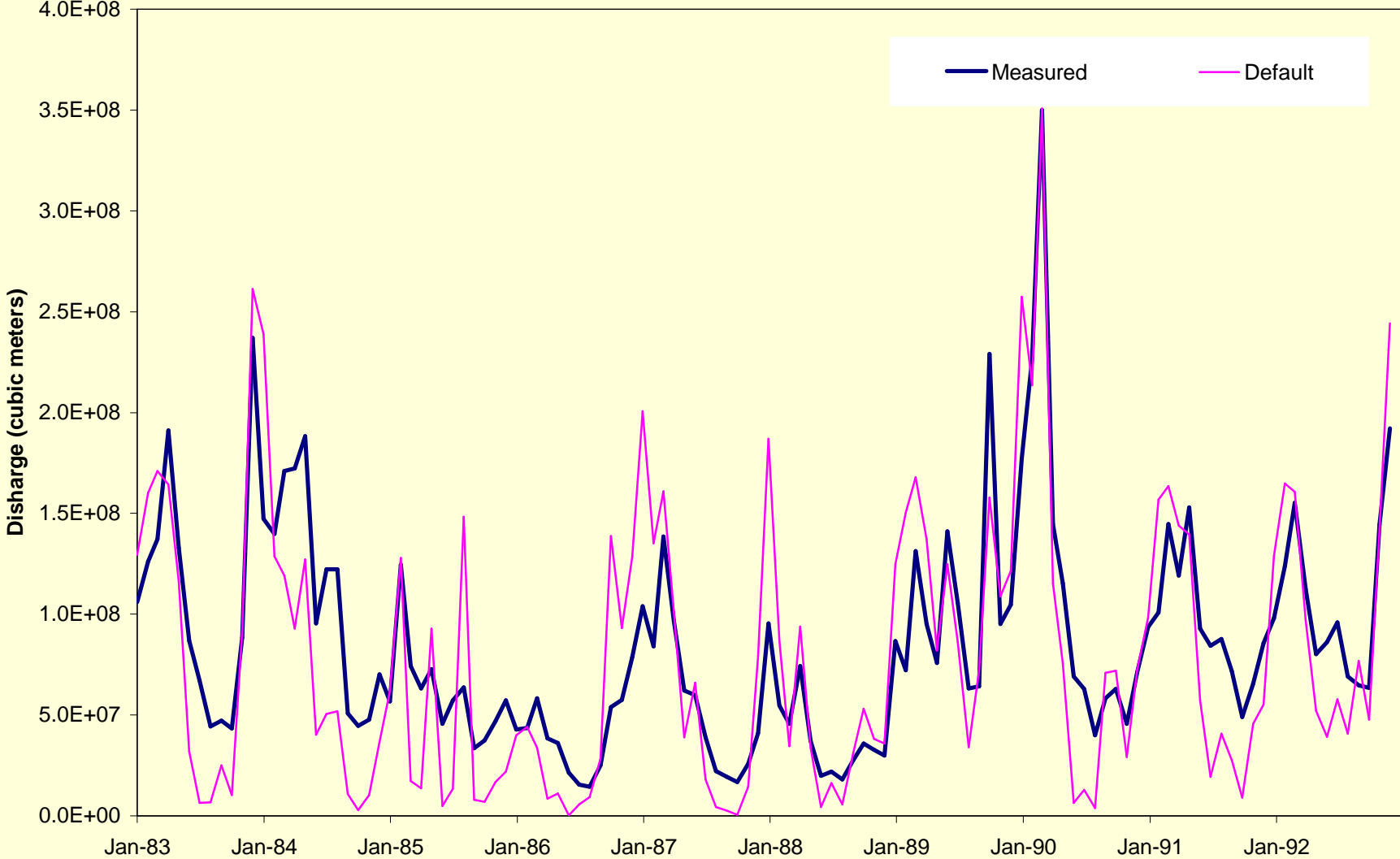
- Automated delineation of watershed from USGS gauging station at Canton
 - Watershed area 1,580 km²
 - 89.7% forest, 7.9% grass, 1.9% row crop, <0.5% urban
 - STATSGO soil layer
 - Precipitation from 4 gages in watershed
- 6 sub-basins
- Within each sub-basin each combination of land-use and soil is a hydrologic response unit (HRU)
 - 72 potential combinations of sub-basin, soil, and land-use
 - 48 actual HRU's (not all soils and land-uses appear in each sub-basin)



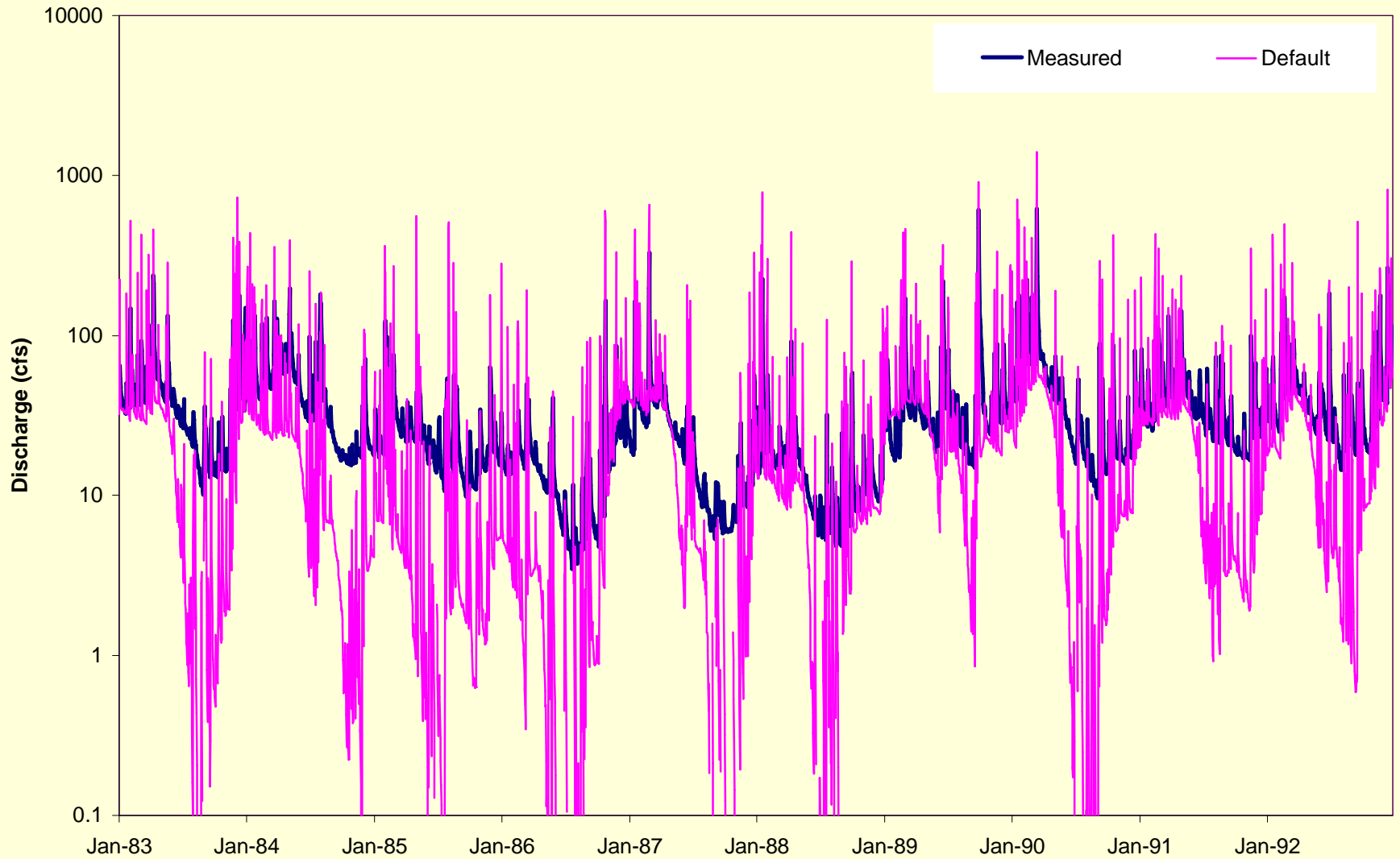
Outline

- No calibration (default parameter values)
- Manual calibration
- Automated calibration
- Model uncertainty

Monthly Discharge 1983-1993



Daily Discharge 1983-1993



Parameter	Description	Sensitivity	Class
ESCO	Soil evaporation depth coefficient	1.917	I
GW_DELAY	Time for groundwater to reach shallow aquifer	0.670	II
CN2F	Curve number forestry	0.666	II
SOL_AWC	Soil available water content	0.275	II
GW_REVAP	Threshold depth for water transfer from shallow aquifer to root zone	0.254	II
CN2P	Curve number pasture	0.119	II
CH_K(1)	Ephemeral channel bottom conductivity	0.042	III
GWQMN	Threshold depth for groundwater flow to stream	0.033	III
SURLAG	Surface lag coefficient delays runoff	0.027	III
ALPHA_BF	Base flow recession constant	0.014	III
RCHRGDP	Deep aquifer percolation fraction	0.011	III
CN2U	Curve number urban	0.009	IV
CH_N	Manning's "n" for tributary channels	0.009	IV
CN2A	Curve number row crop agriculture	0.008	IV
OV_N	Manning's "n" for overland flow	0.002	IV
REVAPMN	Threshold depth for water transfer from shallow aquifer to the unsaturated zone or deep groundwater	0.001	IV

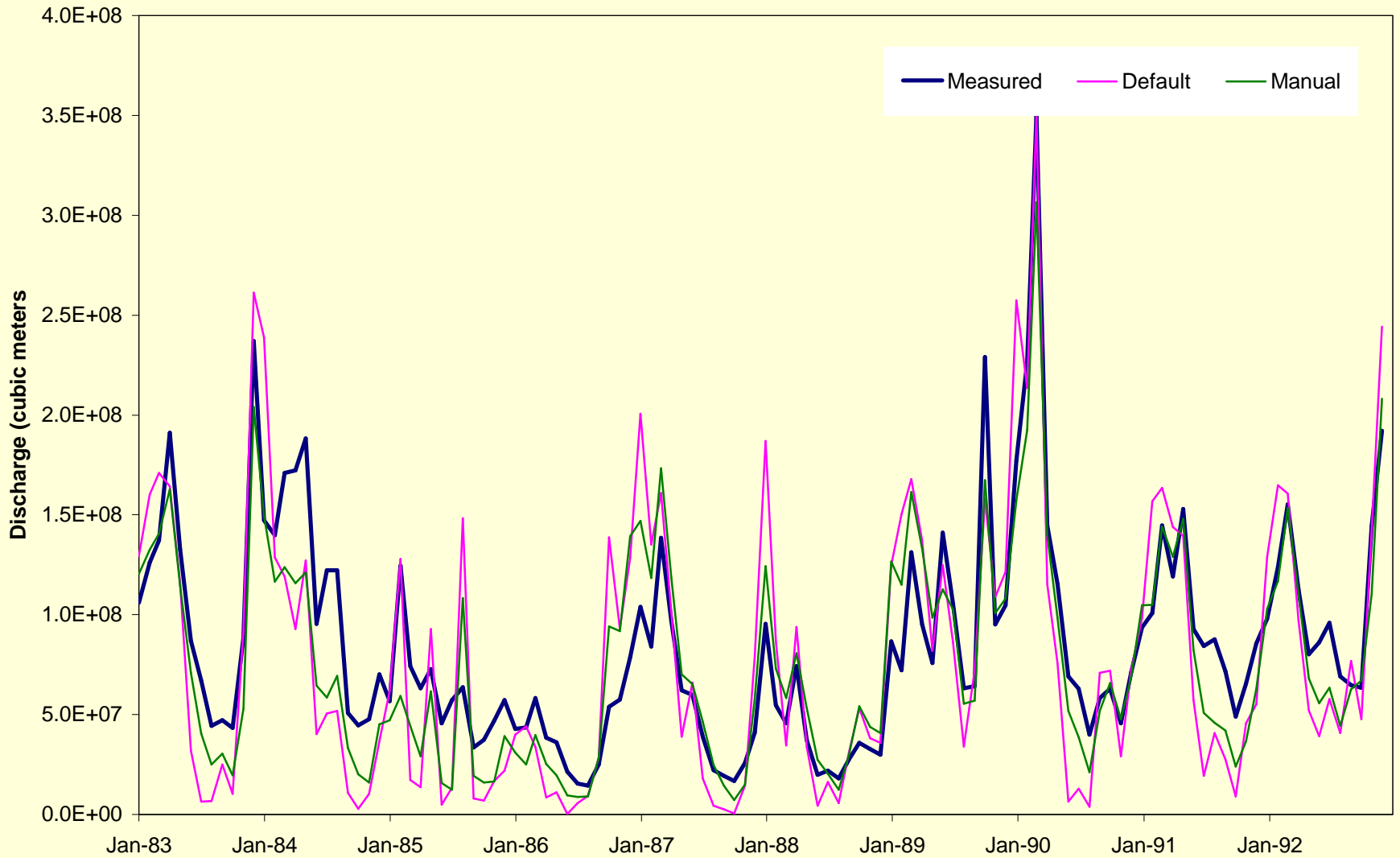
Outline

- No calibration
- Manual calibration
- Automated calibration
- Model uncertainty

Manual Calibration

- Manually adjusted 11 most sensitive parameters
- Class I–III sensitivity
- All parameters had the same values across sub-basins
 - Lumped fitted parameter values

Monthly Discharge



Nash-Sutcliffe

- Nash-Sutcliffe efficiency

$$E = 1 - \frac{\mathbf{S}_{error}^2}{\mathbf{S}_{observed}^2}$$

σ_{error} = *error variance* or root mean squared error of observed – predicted

$\sigma_{observed}$ = variance of the observations

- $E = 1$: perfect fit (zero error)
- $E = 0$: no better than a model that uses the mean of observed as prediction
- $E < 0$: worse than mean of observed

Nash-Sutcliffe

Calibration	Monthly Discharge	Daily Discharge
Default	0.50	-1.39
Manual	0.77	0.62

Outline

- No calibration
- Manual calibration
- Automated calibration
- Model uncertainty

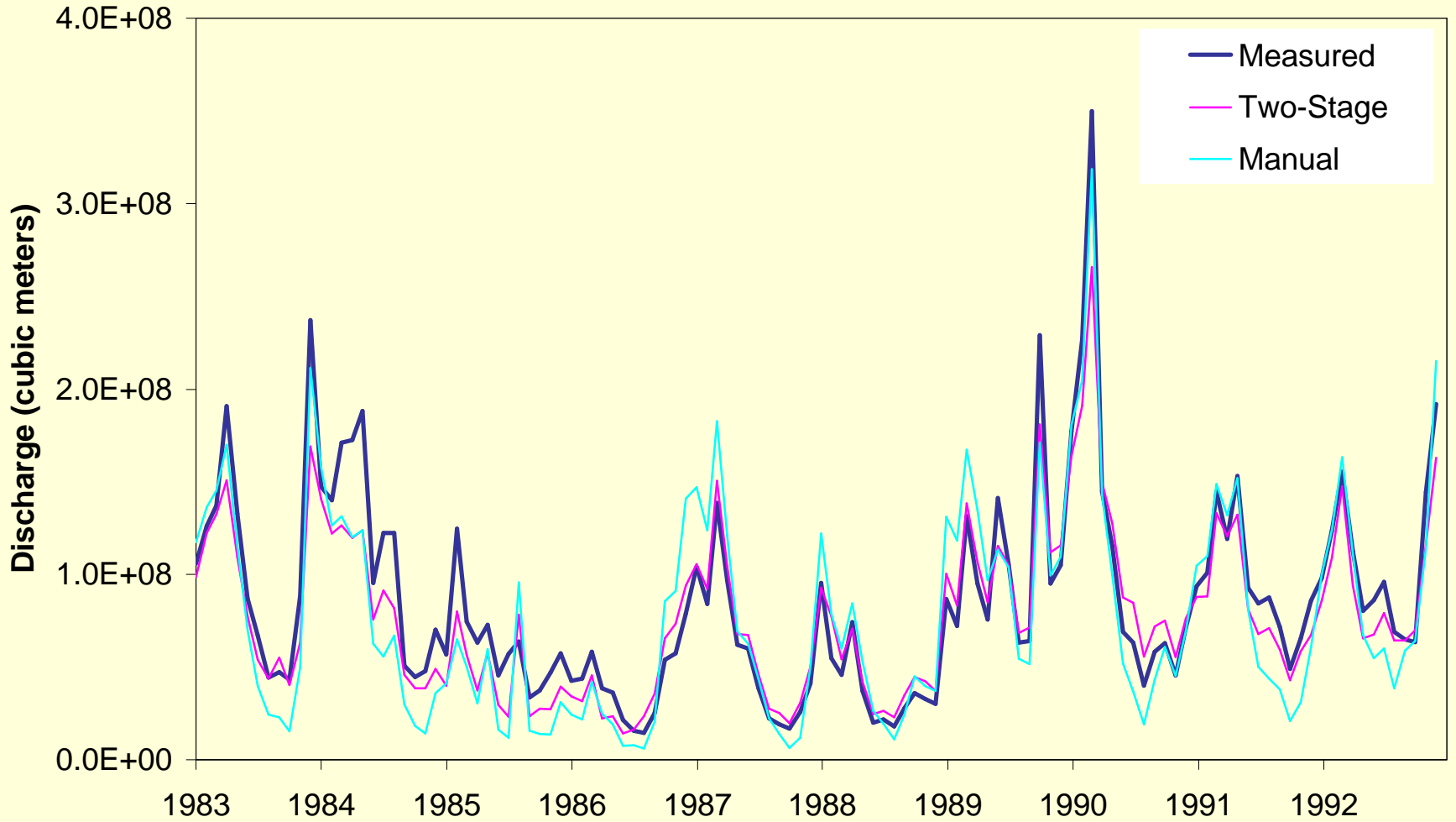
Automated Calibration

- Two-stage automated calibration using PEST
- First stage
 - Used Shuffled Complex Evolution (SCE) option in PEST to find approximate global minimum
 - Accelerated process by
 - using lumped parameters (11 parameters)
 - relaxing requirements for fit

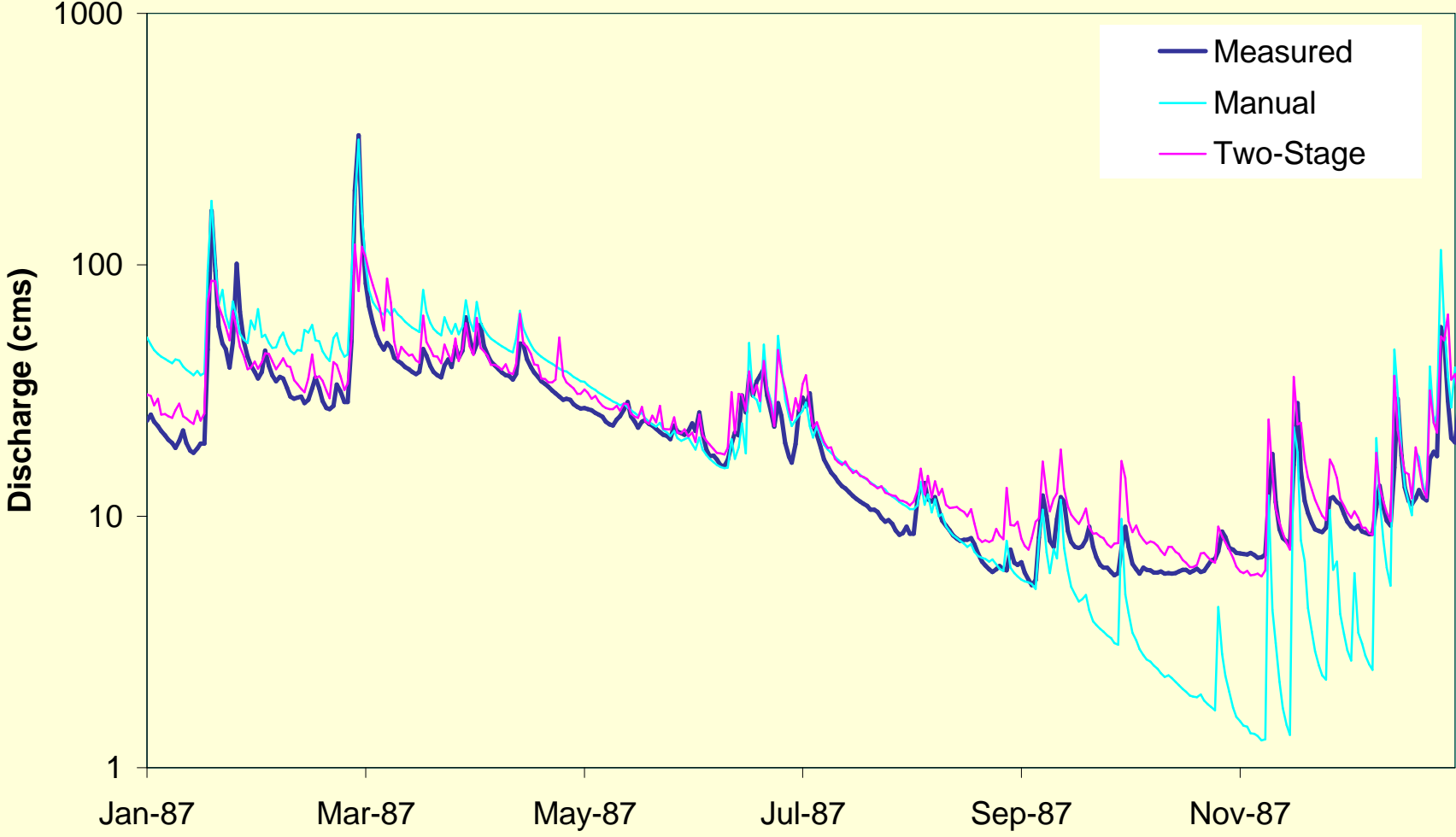
Automated Calibration

- Second stage
 - Used local method to refine global minimum
 - Used parameter values found in first stage as starting values
 - Allowed parameters to vary among HRU
 - Increased parameter number from 11 to 69
 - Used “regularization” to restrain variation among HRU

Monthly Discharge (cubic meters)



Daily Discharge 1987



Nash-Sutcliffe

Calibration	Monthly Discharge	Daily Discharge
Default	0.50	-1.39
Manual	0.77	0.62
Two-stage PEST	0.86	0.61

Number of Model Runs

Calibration Method	Approximate Number of Model Runs	Nash-Sutcliffe Coefficient for Monthly Flow
Two-stage PEST	3000	0.86
Conventional Global Method (SCE)	8000	0.83

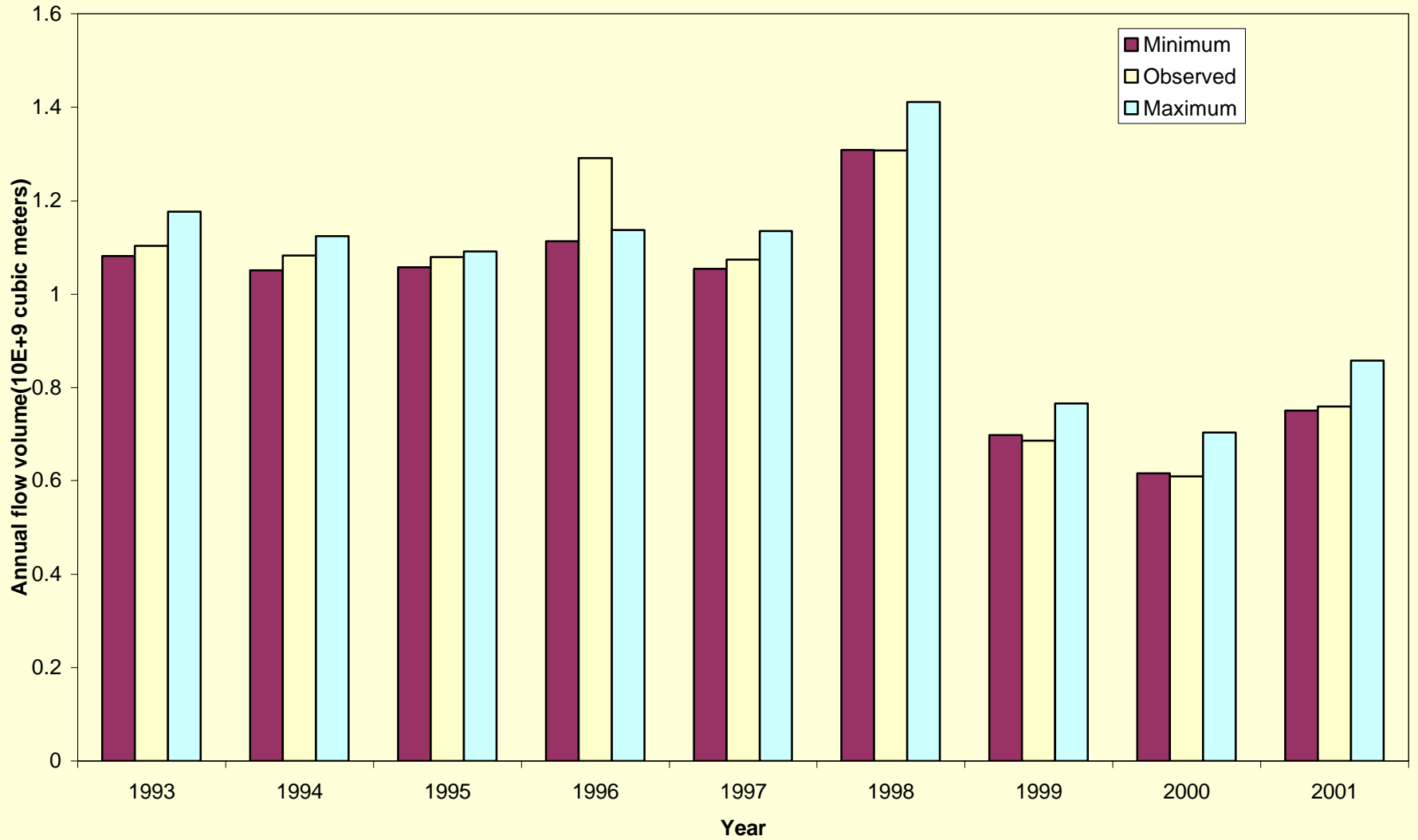
Outline

- No calibration
- Manual calibration
- Automated calibration
- Model uncertainty

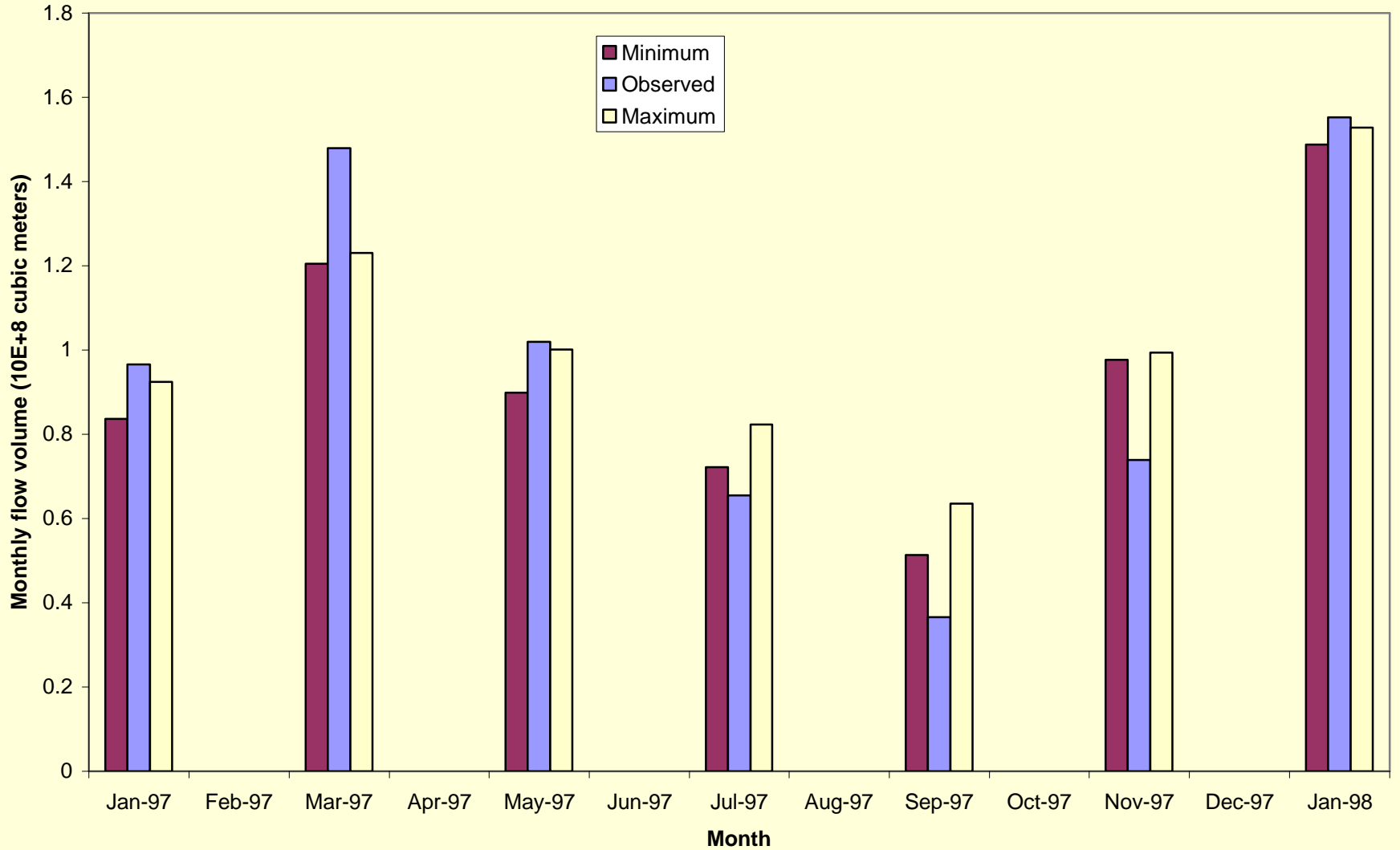
Model Uncertainty

- No widely accepted method has been developed for setting confidence limits on predictions of dynamic watershed-scale models
- PEST includes a “non-linear calibration constrained optimization” method for setting 95% confidence intervals (CI) on predictions
- We used this to set CI on annual, monthly, and weekly predictions for calibration period

Uncertainty in Annual Predictions



Uncertainty in Monthly Predictions 1997



Conclusions

- Automated calibration of SWAT
 - Can be done with PEST (free software)
 - Allows a large number of distributed parameters to be fitted
 - First step in analyzing model uncertainty
- Model uncertainty
 - PEST has a method for setting CI on predictions
 - For annual flow CI were narrow and encompassed observed flow CI for monthly and weekly flow did not encompass observed flow
 - Time scale for trading program is annual