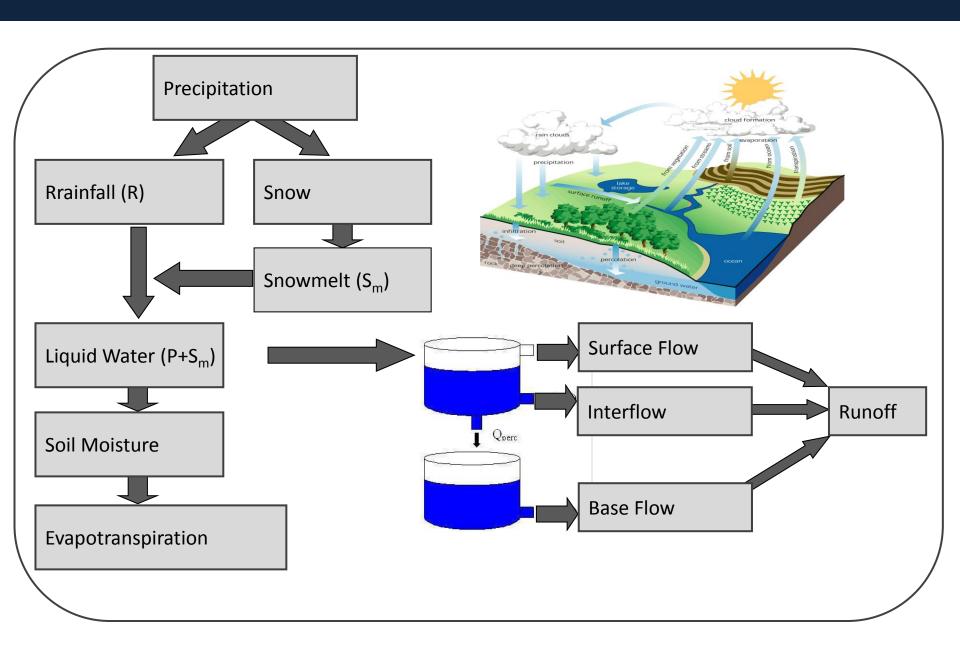


2nd Workshop on Water Resources in Developing Countries: Planning and Management in a Climate Change Scenario, 6-17 May 2013, Trieste, Italy.

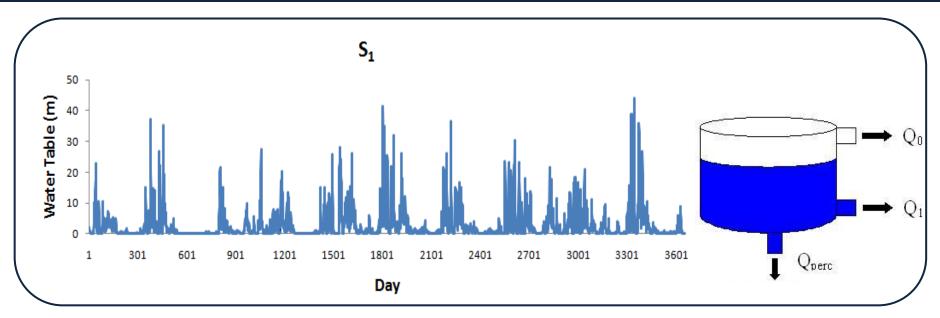


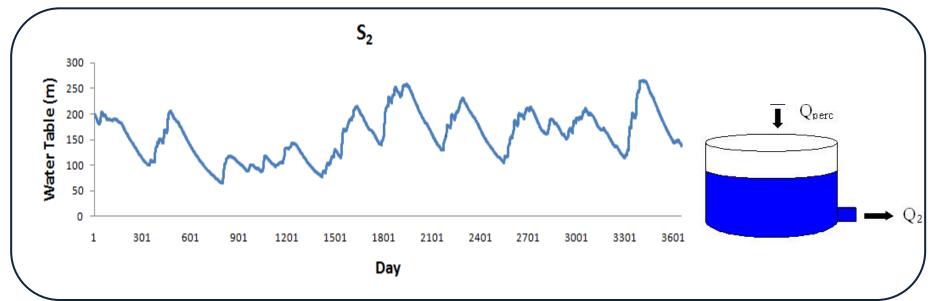


Model Structure (Review)

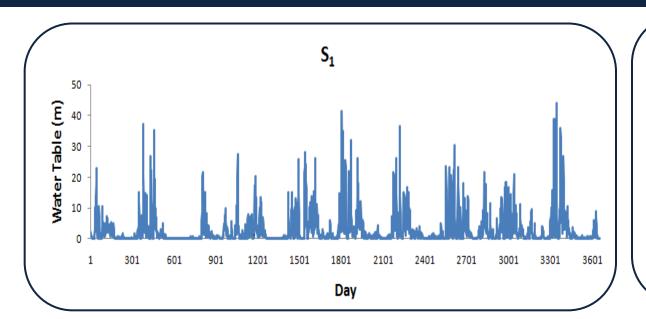


Reservoir Concept



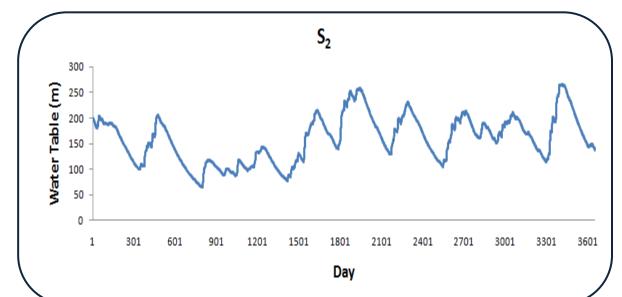


Reservoir Concept



$$Q_0 = K_0 .(S_1-L). A$$

 $Q_1 = K_1 . S_1 . A$
 $Q_2 = K_2 . S_2 . A$



$$K_0 > K_1 > K_2$$

Runoff Response

$$Q = Q_0 + Q_1 + Q_2$$

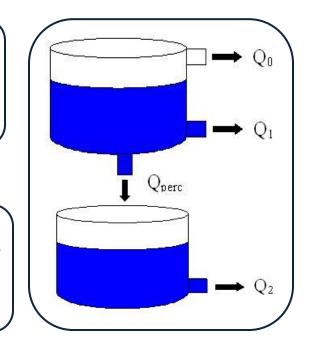


=MAX(0,K31-\$F\$9)*\$F\$8+K31*\$F\$10+L31*\$F\$11

 $K_0 (S_1-L)$

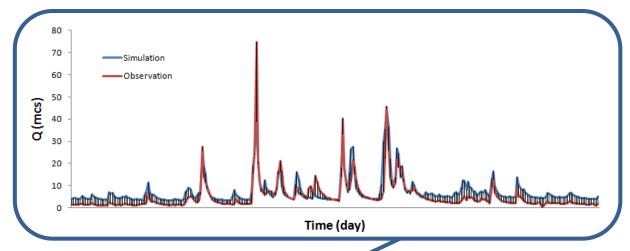
 $K_1(S_1)$

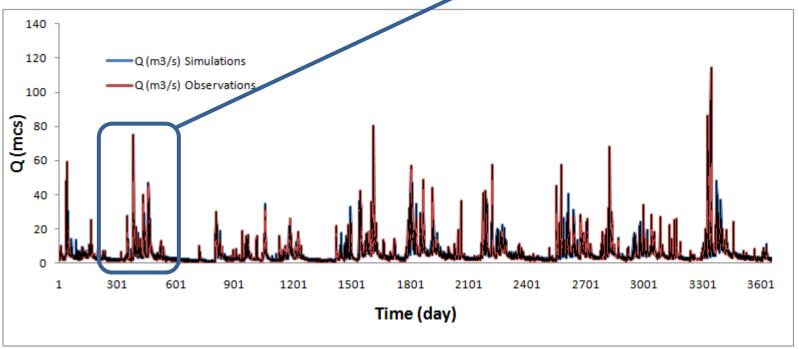
 $K_2(S_2)$



Temp.	Preci.	Snow	Liquid Water	Soil Moisture	DQ (mm/day)	Potential	Ea	\mathbf{S}_1	S ₂	Total Q (Q _t)		$Q(m^3/s)$	Q (m ³ /s)
(C)	(mm)	(mm)		OR P _{ef}		E. (PE _a)	(mm/day)			(mm/dax)			Observations
		25		100.0				2.000	200.000		1.065		
-1.5	0.4	25.4	0	99.8	0.000	0.161	0.153	1.291	199.644		0.969	4.600	4.5
-0.8	10.5	35.9	0	99.7	0.000	0.164	0.156	0.833	199.133		0.907	4.303	11
-2.8	0.9	36.8	0	99.5	0.000	0.155	0.147	0.538	198.521		0.865	4.106	6.6
-3.7	4.4	41.2	0	99.4	0.000	0.150	0.142	0.347	197.847		0.837	3.973	5
-6.1	0.6	41.8	0	99.3	0.000	0.139	0.131	0.224	197.133		0.818	3.883	4.1
-3	0	41.8	0	99.1	0.000	0.154	0.145	0.145	196.394		0.805	3.819	3.5
-0.7	4.4	46.2	0	99.0	0.000	0.165	0.155	0.093	195.640		0.795	3.772	3.2
1.8	3.1	40.8	8.5	107.0	0.336	0.177	0.167	0.396	194.879		0.832	3.948	3.2
0.6	1.7	39	3.5	110.1	0.211	0.171	0.171	0.467	194.187		0.838	3.979	5
1.8	3.6	33.6	9	118.3	0.633	0.177	0.177	0.934	193.514		0.898	4.259	7.9

Runoff Response





Validation Criteria

Root Mean Square Error

$$RMSE = \sqrt{\frac{\sum_{i=1}^{n} (Q_s - Q_o)^2}{n}}$$

Ideal RMSE= 0

Q_s=simulated discharge

Q_o=observed discharge

Bias

$$Bias = \frac{\sum_{i=1}^{n} Q_s}{\sum_{i=1}^{n} Q_o}$$

Ideal Bias= 1

Q_s=simulated discharge

Q_o=observed discharge

Validation Criteria

Correlation Coefficient

Ideal
$$R_p = 1$$

R_p= -1 Negatively correlated

R_p= 0 Not correlated

 R_p = 1 Correlated

$$R_{p} = \frac{\sum_{i=1}^{n} (Q_{o}^{i} - \overline{Q}_{o}) \cdot \sum_{i=1}^{n} (Q_{s}^{i} - \overline{Q}_{s})}{\sqrt{\sum_{i=1}^{n} (Q_{o}^{i} - \overline{Q}_{o})^{2} \cdot \sqrt{\sum_{i=1}^{n} (Q_{s}^{i} - \overline{Q}_{s})^{2}}}}$$

where:

 R_P Pearson correlation coefficient [-]

 \bar{O}_s mean simulated discharge [L³T⁻¹]

Nash-Sutcliff Coefficient

Ideal $R_{NS} = 1$

Negative R_{NS} means that the mean of observations is a better predictor than the model.

$$R_{NS} = 1 - \frac{\sum_{t=1}^{n} (Q_{s}^{t} - Q_{o}^{t})^{2}}{\sum_{t=1}^{n} (Q_{o}^{t} - \overline{Q}_{o}^{t})^{2}}$$

where:

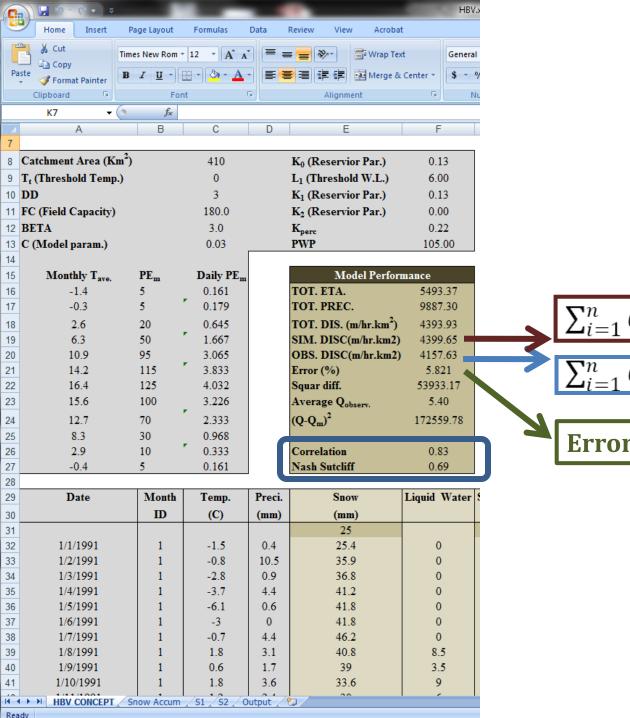
 R_{NS} Nash-Sutcliffe coefficient [-]

 Q_s simulated discharge [L³T⁻¹]

 Q_o observed discharge [L³T⁻¹]

 \bar{O}_o mean observed discharge [L³T⁻¹]

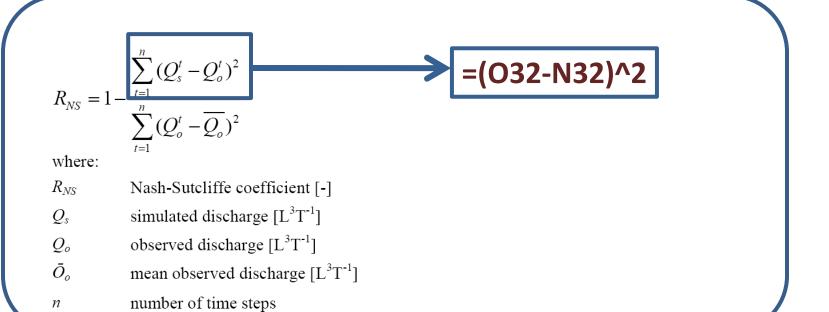
n number of time steps



$$Bias = \frac{\sum_{i=1}^{n} Q_{s}}{\sum_{i=1}^{n} Q_{o}}$$

Error (%) of total runoff

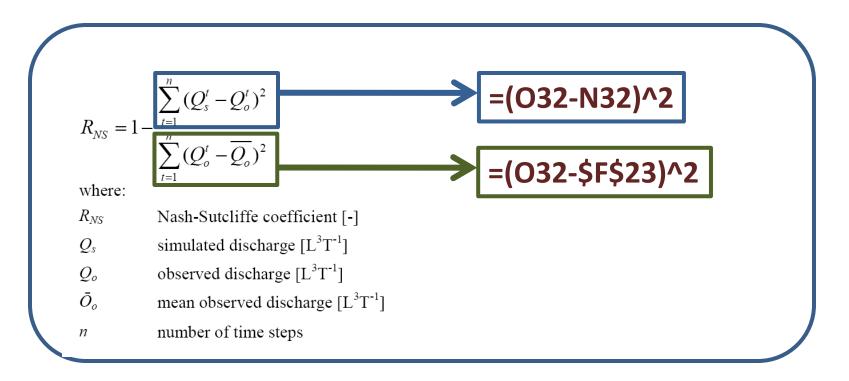
Nash-Sutcliff Coefficient



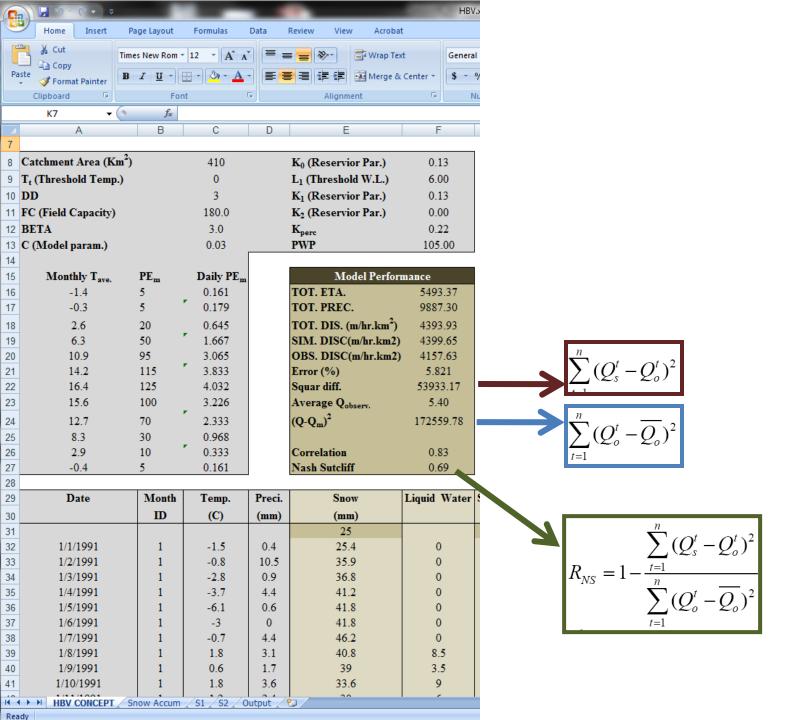
29	Month	Temp.	Preci.	Snow	Liquid Water	Soil Moisture	DQ (mm/day)	Potential	Ea	S ₁	S_2	Total Q (Q _t)	$Q(m^3/s)$	$Q(m^3/s)$	(Q-QT) ²	(Q-Qm) ²
30	ID	(C)	(mm)	(mm)			OR P _{eff}	E. (PE _a)	(mm/day)			(mm/day)	Simulations	Observations		
31				25		100.0				2.000	200.000	1.065				
32	1	-1.5	0.4	25.4	0	99.8	0.000	0.161	0.153	1.291	199.644	0.969	4.600	4.5	0.010	0.817
33	1	-0.8	10.5	35.9	0	99.7	0.000	0.164	0.156	0.833	199.133	0.907	4.303	11	44.850	31.317
34	1	-2.8	0.9	36.8	0	99.5	0.000	0.155	0.147	0.538	198.521	0.865	4.106	6.6	6.221	1.431
35	1	-3.7	4.4	41.2	0	99.4	0.000	0.150	0.142	0.347	197.847	0.837	3.973	5	1.054	0.163
36	1	-6.1	0.6	41.8	0	99.3	0.000	0.139	0.131	0.224	197.133	0.818	3.883	4.1	0.047	1.700
37	1	-3	0	41.8	0	99.1	0.000	0.154	0.145	0.145	196.394	0.805	3.819	3.5	0.102	3.625
38	1	-0.7	4.4	46.2	0	99.0	0.000	0.165	0.155	0.093	195.640	0.795	3.772	3.2	0.327	4.857
39	1	1.8	3.1	40.8	8.5	105.9	1.413	0.177	0.167	1.473	194.879	0.974	4.624	3.2	2.028	4.857
40	1	0.6	1.7	39	3.5	108.5	0.713	0.171	0.171	1.663	194.426	0.998	4.735	5	0.070	0.163
41	1	1.8	3.6	33.6	9	115.4	1.971	0.177	0.177	3.045	194.018	1.179	5.595	7.9	5.314	6.231

n

Nash-Sutcliff Coefficient



29	Month	Temp.	Preci.	Snow	Liquid Water	Soil Moisture	DQ (mm/day)	Potential	Ea	$\mathbf{S_1}$	S_2	Total Q (Q _t)	Q (m ³ /s)	Q (m ³ /s)	$(Q-QT)^2$	(Q-Qm) ²
30	ID	(C)	(mm)	(mm)			OR P _{eff}	E. (PE _a)	(mm/day)			(mm/day)	Simulations	Observations		
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32	1	-1.5	0.4	25.4	0	99.8	0.000	0.161	0.153	1.291	199.644	0.969	4.600	4.5	0.010	0.817
33	1	-0.8	10.5	35.9	0	99.7	0.000	0.164	0.156	0.833	199.133	0.907	4.303	11	44.850	31.317
34	1	-2.8	0.9	36.8	0	99.5	0.000	0.155	0.147	0.538	198.521	0.865	4.106	6.6	6.221	1.431
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38	1	-0.7	4.4	46.2	0	99.0	0.000	0.165	0.155	0.093	195.640	0.795	3.772	3.2	0.327	4.857
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41	1	1.8	3.6	33.6	9	115.4	1.971	0.177	0.177	3.045	194.018	1.179	5.595	7.9	5.314	6.231



Model Parameters

Conceptual

- BETA (β)
- (
- |
- K_C
- K₁
- K₂
- K_{perc}

Conceptual & Measurable

- FC
- DD
- PWP
- T_t

Initial Conditions

- Snow
- Soil Moisture
- S₁
- S₂

Error Sources

Error in Initial Conditions

 Error in the initial values of soil moisture, snow, field capacity, permanent wilting point

Error in Model Processes

- Unrealistic model assumptions
- Unrepresentative conceptual description of the system

Error in Observations

- Error in input data (e.g., precipitation, temperature, etc.)
- Error in observed discharge

Error in Model Parameterization

- Inability to obtain the optimal set of parameters.
- Deficiencies in parameter estimation scheme

Error Sources

Error in Initial

 Error in the initial values of soil moisture, snow, field capacity, permanent wilting point

Error in Model Processes

Unrealistic mode assumptions

Unrepresentative

Conceptual

- BETA (β)
- C
- |
- K₀
- K₁
- K₂
- K_{pero}

Error in Observations

e, etc.)

served

• Error in input data (e.g., precipitation,

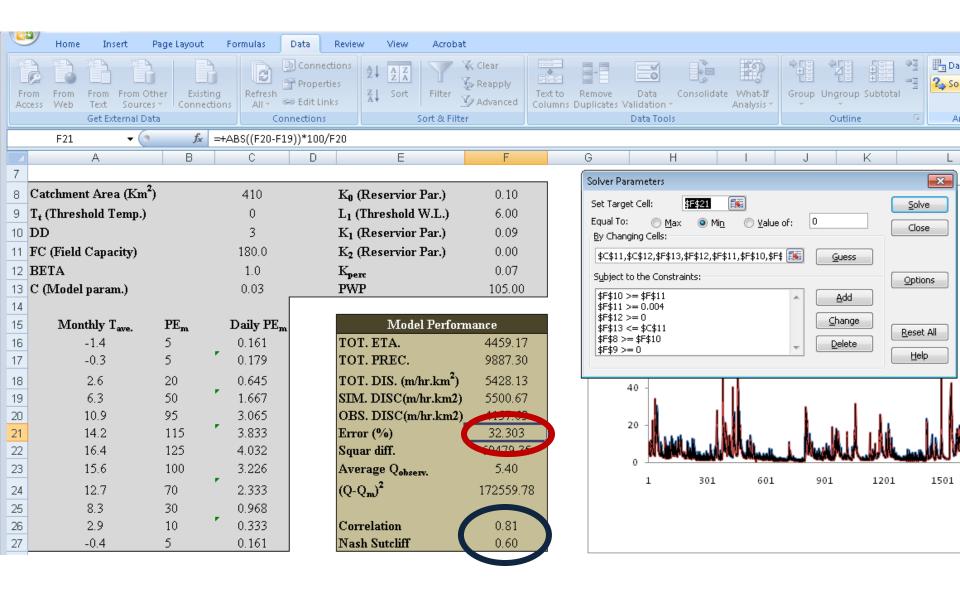
Conceptual & Measurable

- FC
- DD
- PWP
- T_t

Error in Model Parameterization

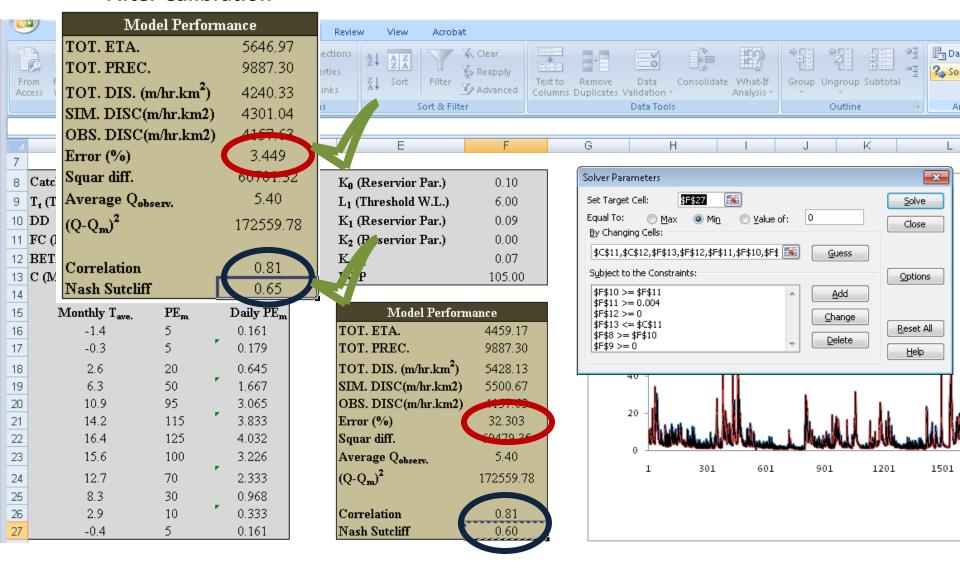
- Inability to obtain the optimal set of parameters.
- Deficiencies in parameter estimation scheme

Parameter Estimation

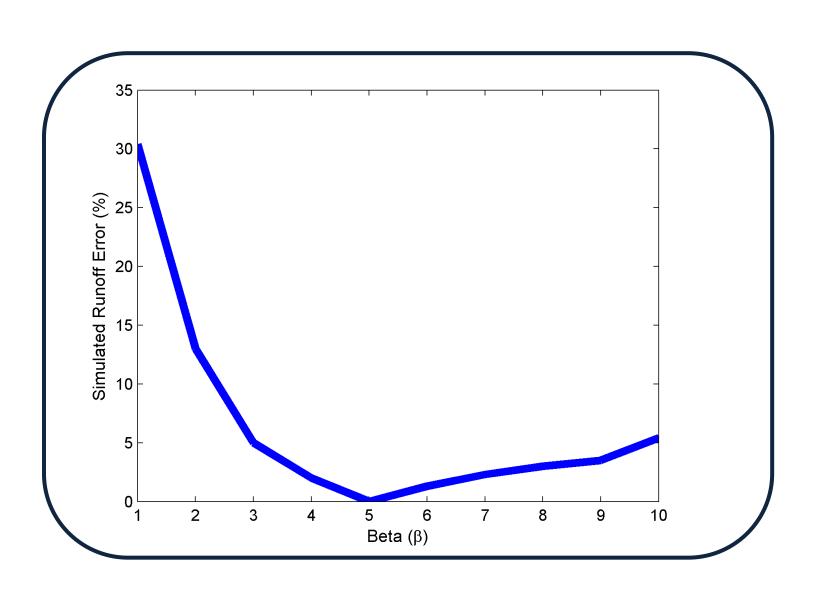


Parameter Estimation

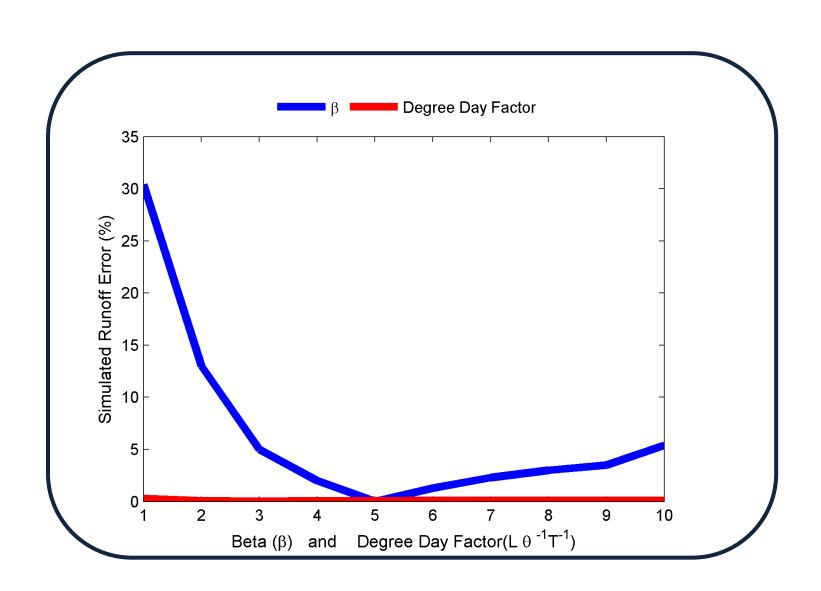
After Calibration



Parameter Sensitivity



Parameter Sensitivity



Error Sources

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Error in Observations

- Error in input data (e.g., precipitation, temperature, etc.)
- Error in shserved discharge

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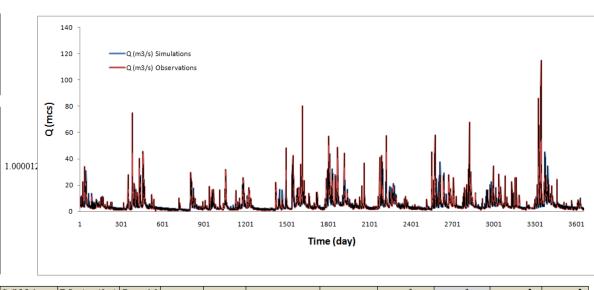
Initial Conditions

- Snow
- Soil Moisture
- S₁
- S₂

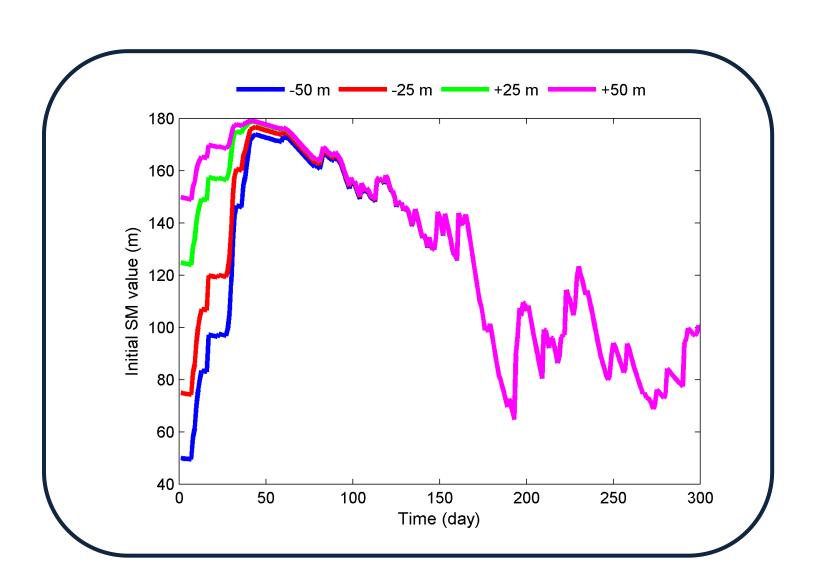
Catchment Area (Km²)	410	K ₀ (Reservior Par.)	0.13
T _t (Threshold Temp.)	0	L ₁ (Threshold W.L.)	6.00
DD	3	K1 (Reservior Par.)	0.13
FC (Field Capacity)	180.0	K ₂ (Reservior Par.)	0.00
BETA	5.0	K_{perc}	0.22
C (Model param.)	0.03	PWP	105.00

Monthly Tave.	PE_m	Daily PE _m
-1.4	5	0.161
-0.3	5	0.179
2.6	20	0.645
6.3	50	1.667
10.9	95	3.065
14.2	115	3.833
16.4	125	4.032
15.6	100	3.226
12.7	70	2.333
8.3	30	0.968
2.9	10	0.333
-0.4	5	0.161

Model Perform	ARTICO.
TOT. ETA.	5736.08
TOT. PREC.	9887.30
TOT. DIS. (m/hr.km ²)	4151.22
SIM. DISC(m/hr.km2)	4157.68
OBS. DISC(m/hr.km2)	4157.63
Error (%)	0.001
Squar diff.	52400.87
Average Q _{observ} .	5.40
$(Q-Q_m)^2$	172559.78
0 12	0.04
Correlation	0.84
Nash Sutcliff	0.70



Date	Month	Тетр.	Preci.	Snow	Liquid Water	Soil Moisture	DQ (mm/day)	Potential	Ea	\mathbf{S}_1	S ₂	Total Q (Q _t)	$Q(m^3/s)$	$Q(m^3/s)$	$(Q-QT)^2$	$(Q-Qm)^2$
	ID	(C)	(mm)	(mm)			OR P _{eff}	E. (PE _a)	(mm/day)			(mm/day)	Simulations	Observations		
				25		100.0				2.000	200.000	1.065				
1/1/1991	1	-1.5	0.4	25.4	0	99.8	0.000	0.161	0.153	1.291	199.644	0.969	4.600	4.5	0.010	0.817
1/2/1991	1	-0.8	10.5	35.9	0	99.7	0.000	0.164	0.156	0.833	199.133	0.907	4.303	11	44.850	31.317
1/3/1991	1	-2.8	0.9	36.8	0	99.5	0.000	0.155	0.147	0.538	198.521	0.865	4.106	6.6	6.221	1.431
1/4/1991	1	-3.7	4.4	41.2	0	99.4	0.000	0.150	0.142	0.347	197.847	0.837	3.973	5	1.054	0.163
1/5/1991	1	-6.1	0.6	41.8	0	99.3	0.000	0.139	0.131	0.224	197.133	0.818	3.883	4.1	0.047	1.700



Butterfly Effect

Sensitive dependence on initial conditions

small variations in the initial condition of a dynamic model may lead to large differences in the behavior of the system.

Butterfly Effect

Edward N. Lorenz 1917-2008

Meteorologist
Massachusetts Institute of Technology



Error Sources

Error in Initial Conditions

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Error in Model Processes

- Unrealistic model assumptions
- Unrepresentative conceptual description of the system

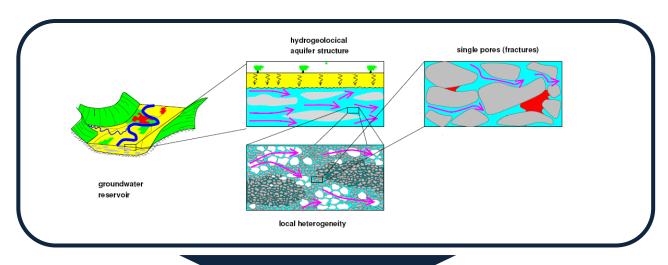
Error in Observations

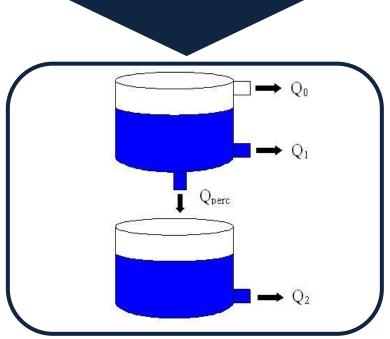
- Error in input data (e.g., precipitation, temperature, etc.)
- Error in observed discharge

Error in Model Parameterization

- Inability to obtain the optimal set of parameters.
- Deficiencies in parameter estimation scheme

Error in Model Processes





Error Sources

Error in Initial Conditions

 Error in the initial values of soil moisture, snow, field capacity, permanent wilting point

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- Error in input data (e.g., precipitation, temperature, etc.)
- Error in observed discharge

Error in Model Parameterization

- Inability to obtain the optimal set of parameters.
- Deficiencies in parameter estimation scheme

Error in Observations

Temp.	Preci.	$Q(m^3/s)$	$Q(m^3/s)$
(C)	(mm)	Simulati	Observat
-1.5	0.4	4.6	4.5
-0.8	10.5	4.3	11.0
-2.8	0.9	4.1	6.6
-3.7	4.4	4.0	5.0
-6.1	0.6	3.9	4.1
-3.0	0.0	3.8	3.5
-0.7	4.4	3.8	3.2
1.8	3.1	4.0	3.2
0.6	1.7	4.0	5.0
1.8	3.6	4.4	7.9
1.2	2.4	4.6	11.9
1.5	0.0	4.7	10.4
1.1	0.0	4.7	10.4
-0.5	0.0	4.3	8.5
-3.2	1.3	4.1	6.8
-0.9	0.6	3.9	6.1
3.2	5.0	5.5	11.6
-1.5	20.7	4.8	22.5
-2.8	8.4	4.4	12.3
-5.1	1.5	4.1	9.2
-2.9	2.4	3.9	7.3
-0.6	0.9	3.8	6.1
0.1	0.9	3.9	5.2
-0.8	1.2	3.8	4.5

Model Perform	ance
TOT. ETA.	5736.08
TOT. PREC.	9887.30
TOT. DIS. (m/hr.km ²)	4151.22
SIM. DISC(m/hr.km2)	4157.68
OBS. DISC(m/hr.km2)	4157.63
Error (%)	0.001
Squar diff.	52400.87
Average Q _{observ.}	5.40
$(Q-Q_m)^2$	172559.78
Correlation	0.84
Nash Sutcliff	0.70

Example Applications of the Model

Water Science and Technology Library

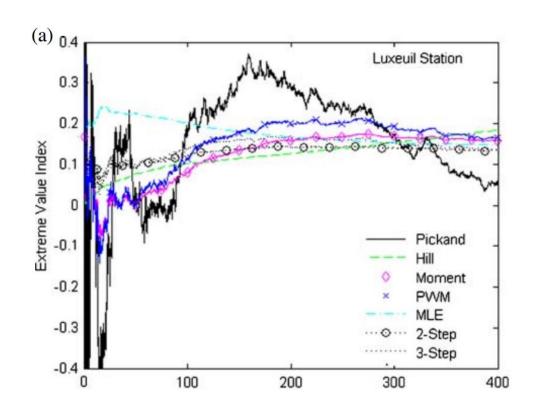
Amir AghaKouchak · David Easterling Kuolin Hsu · Siegfried Schubert Soroosh Sorooshian *Editors*

Extremes in a Changing Climate

Detection, Analysis and Uncertainty

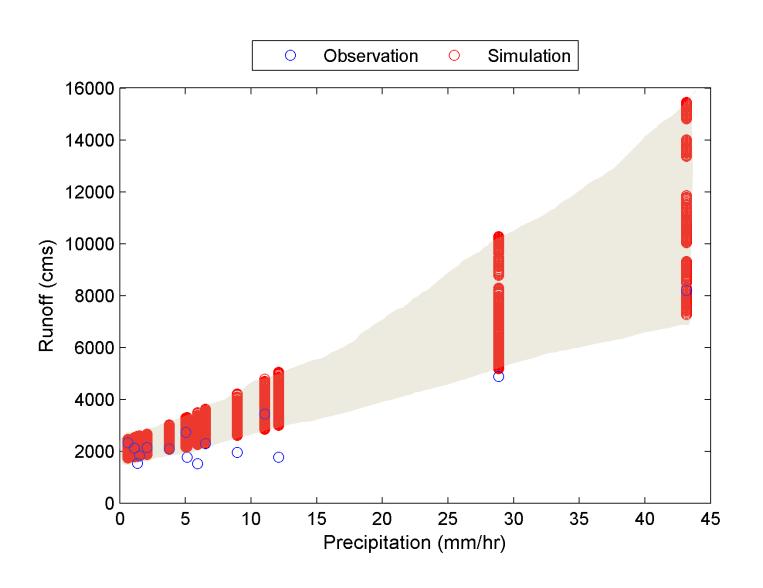
1- Run the model with different GCM outputs

2- Analyze data using the Extreme Value Theory

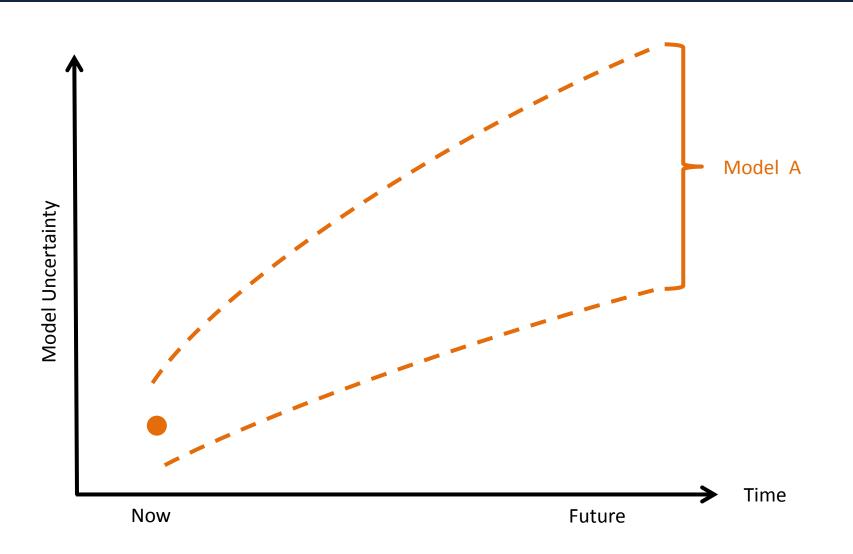




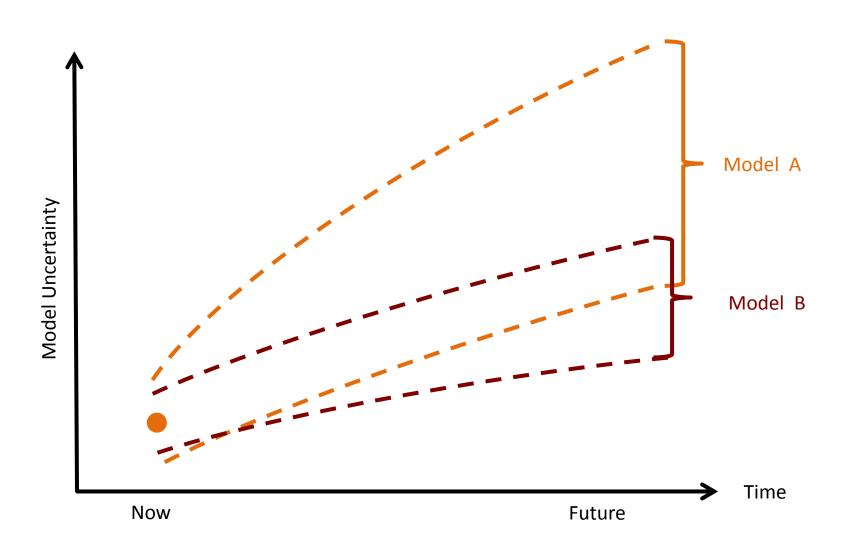
Model Uncertainty - Ensemble Simulation



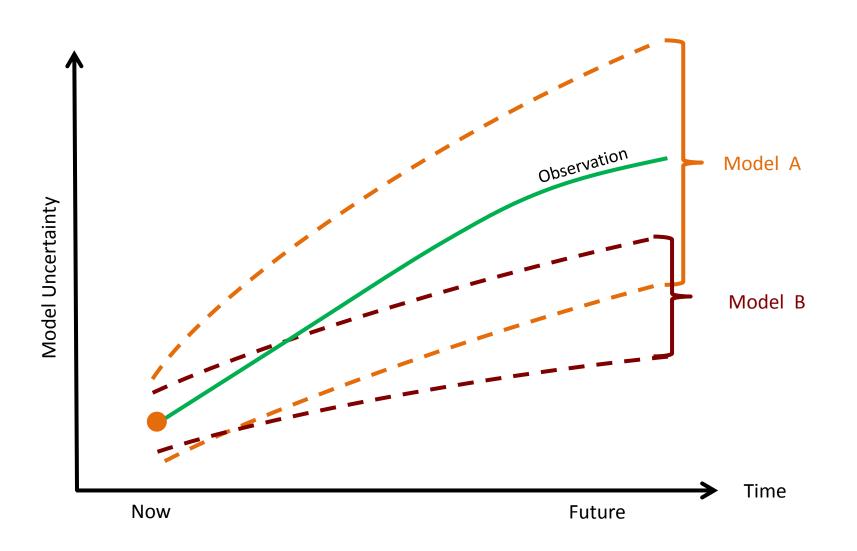
Model Uncertainty



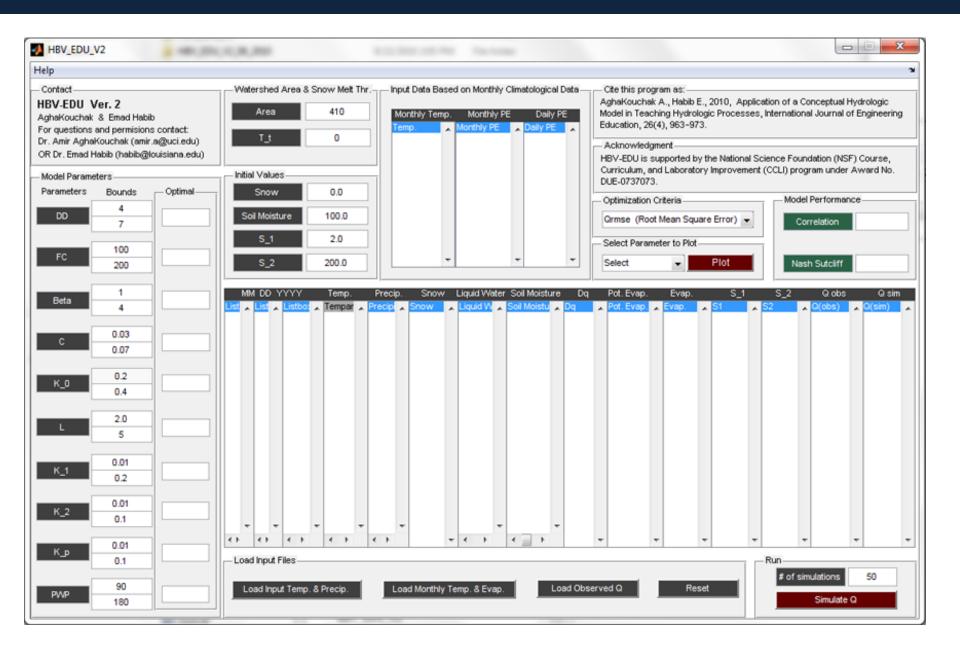
Model Uncertainty



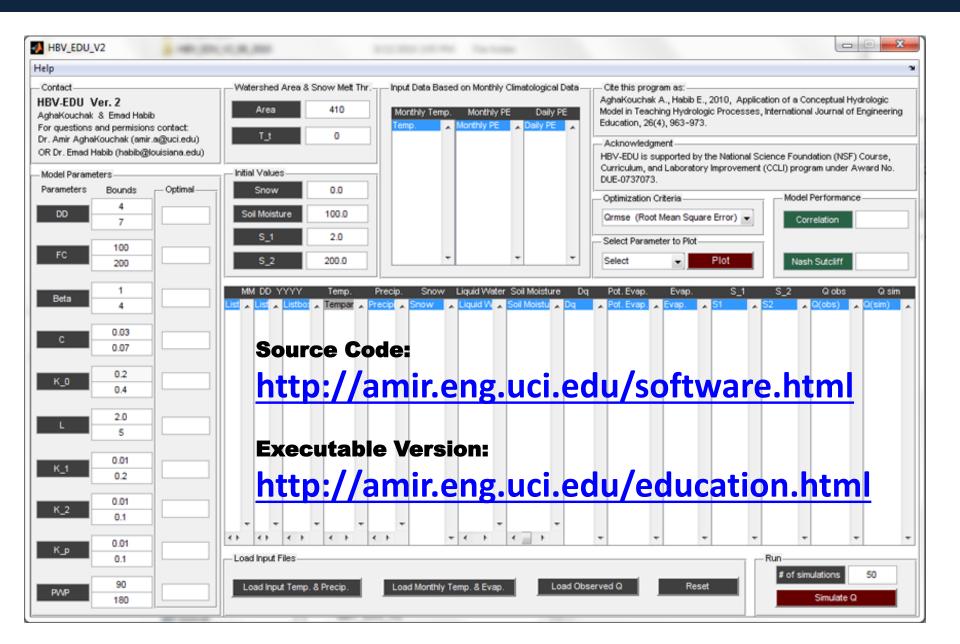
Model Uncertainty



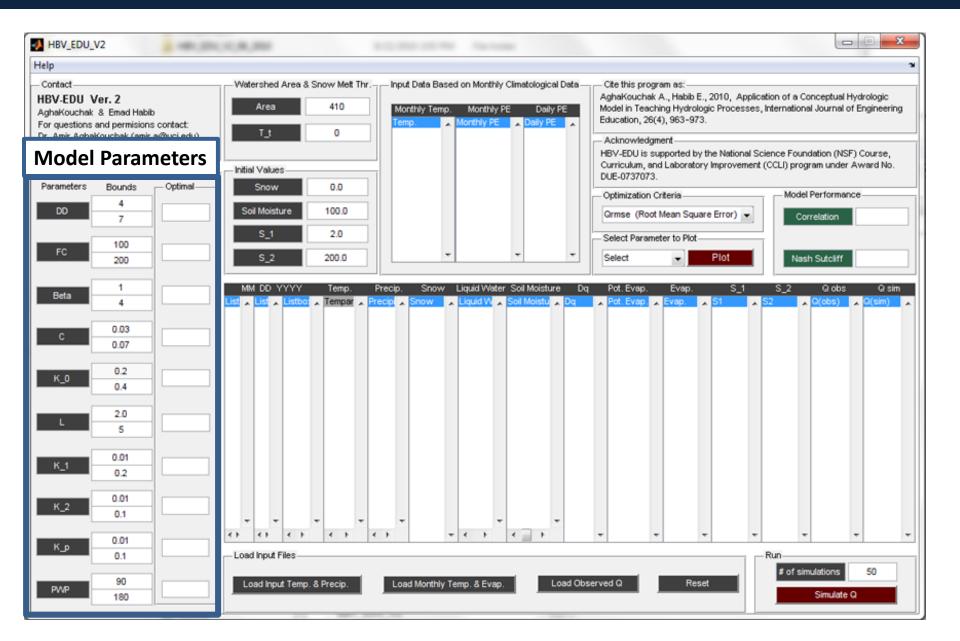
HBV-EDU

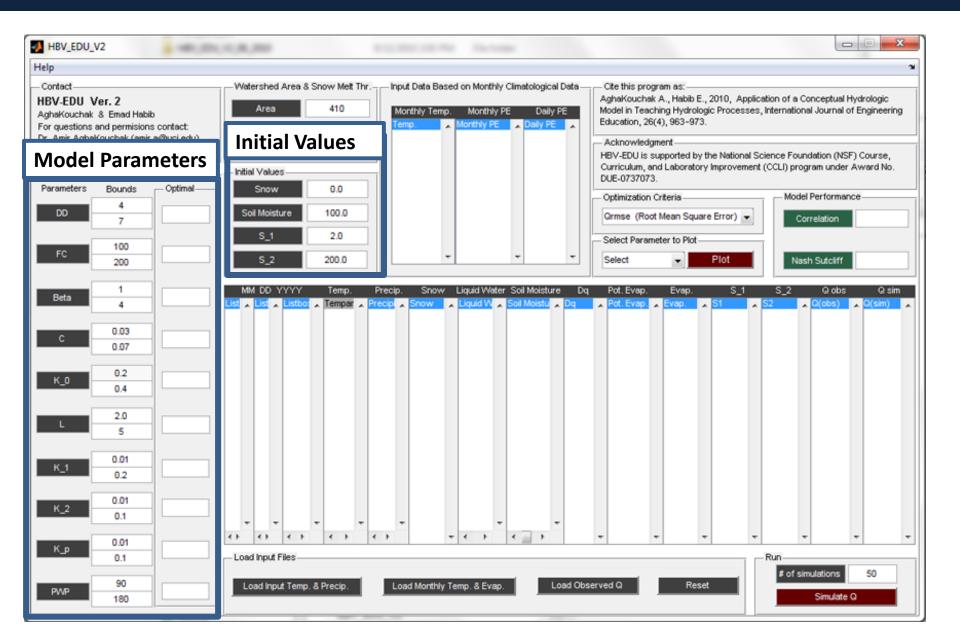


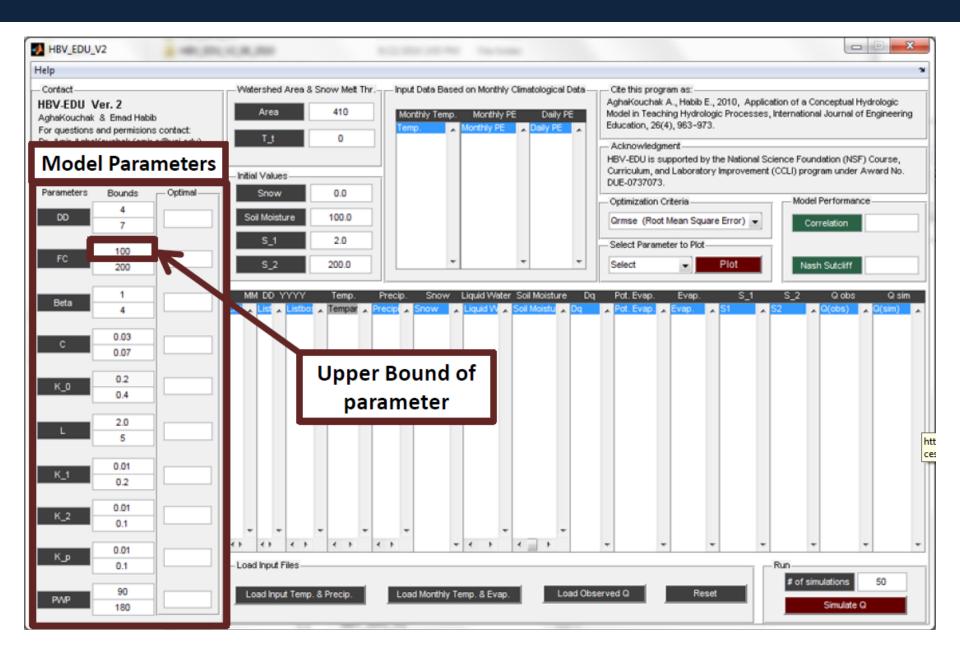
HBV-EDU

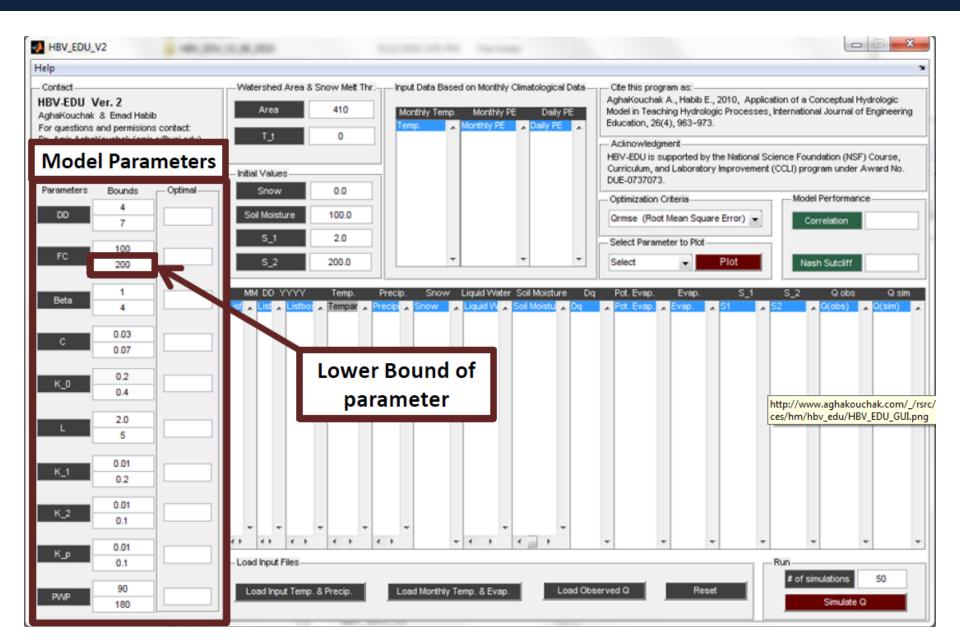


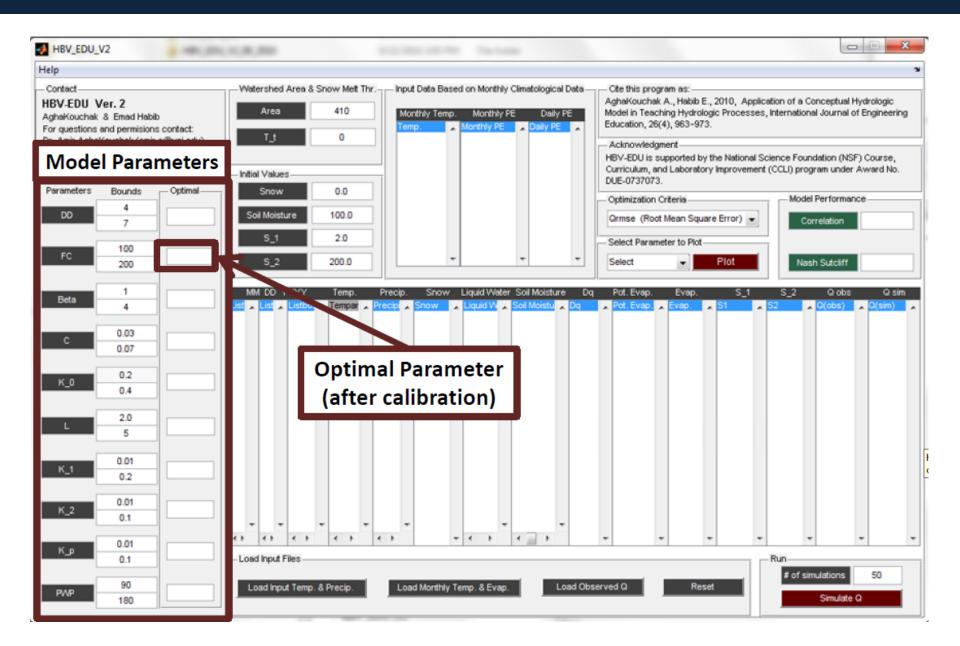
HBV-EDU

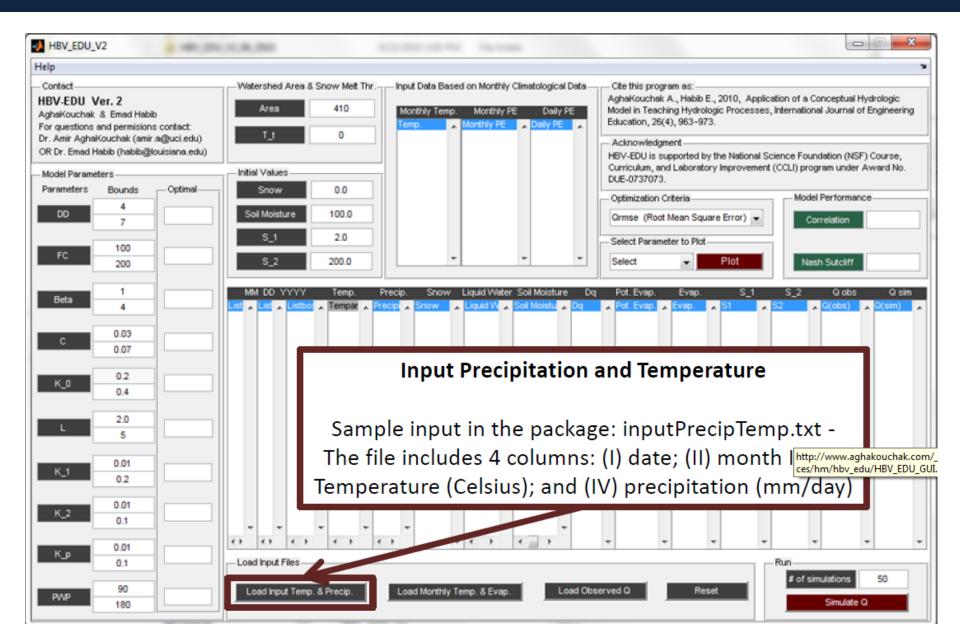


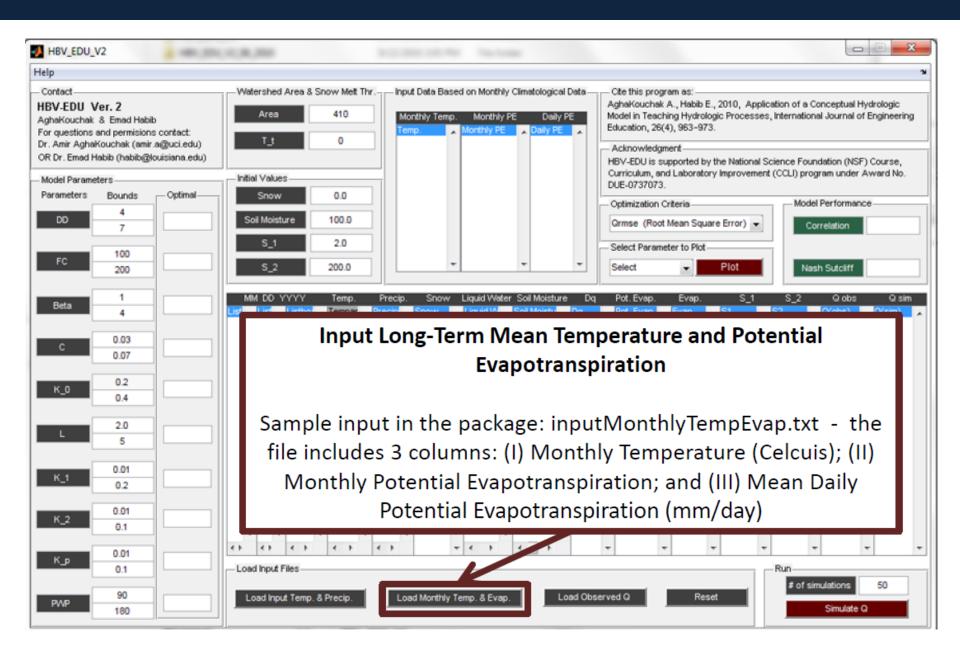


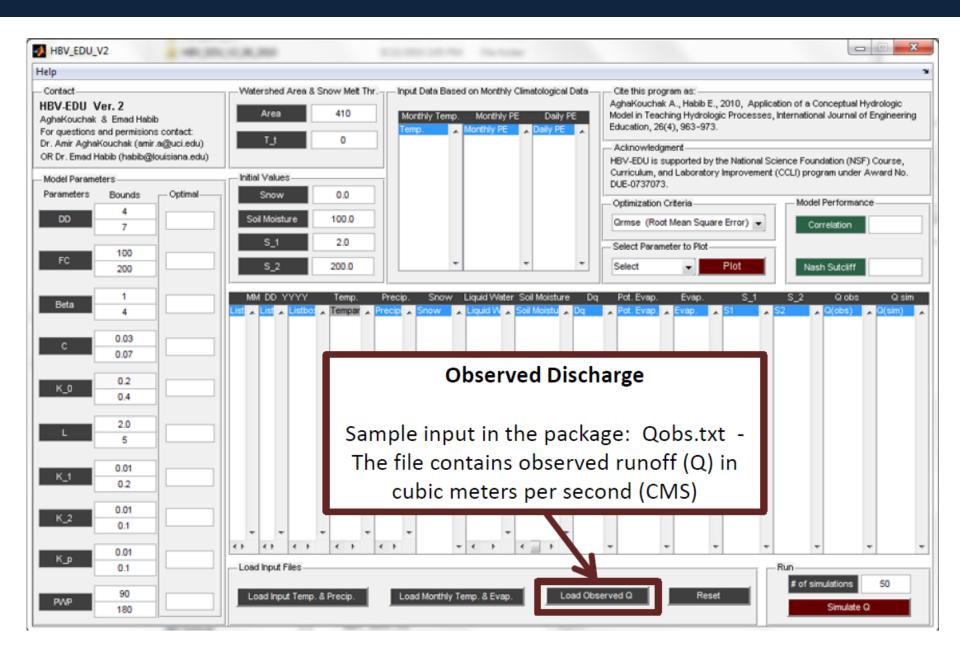


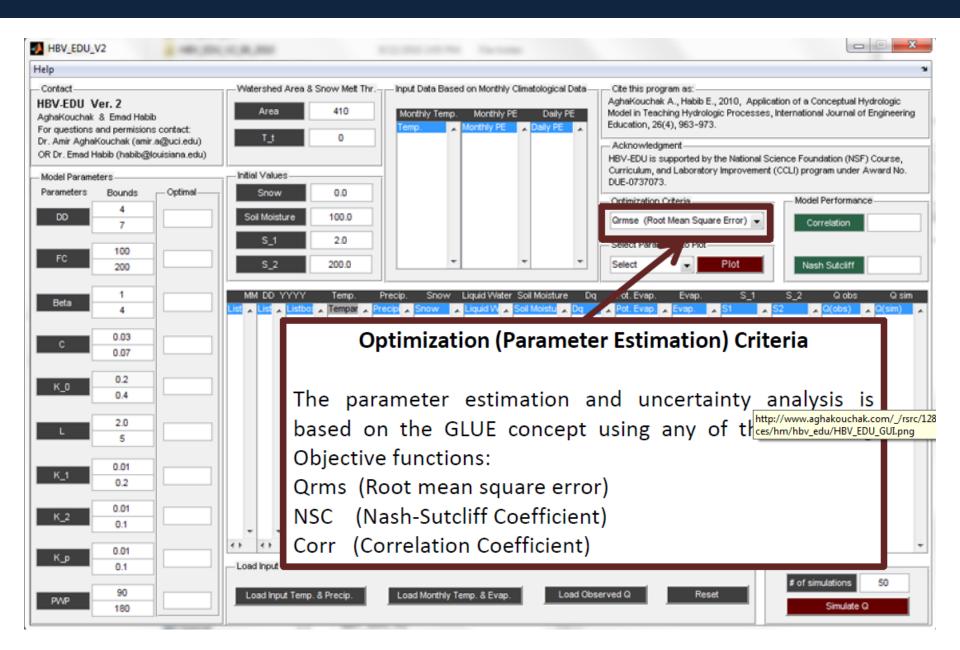


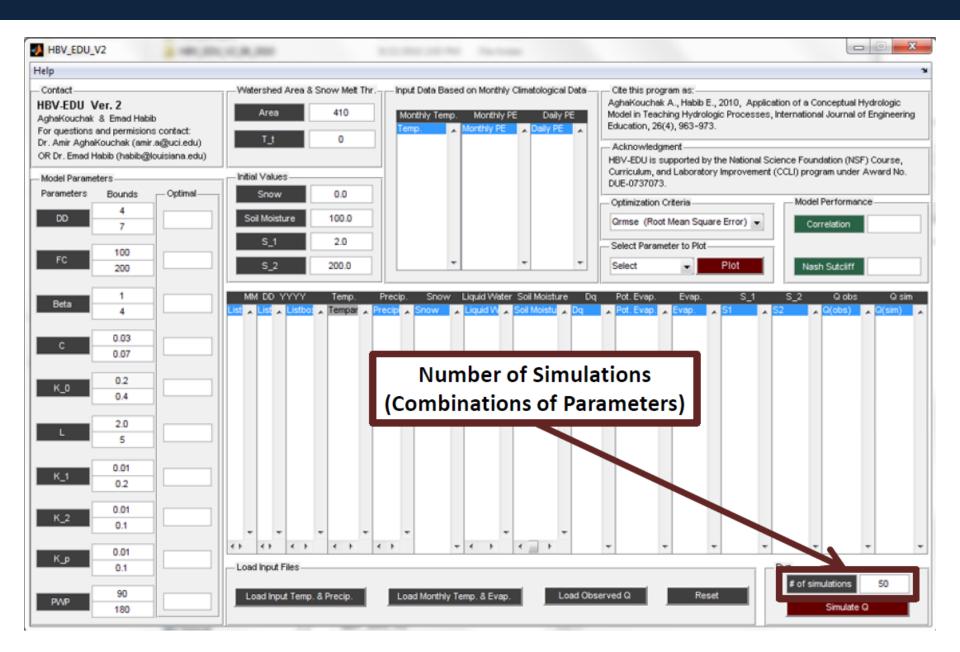


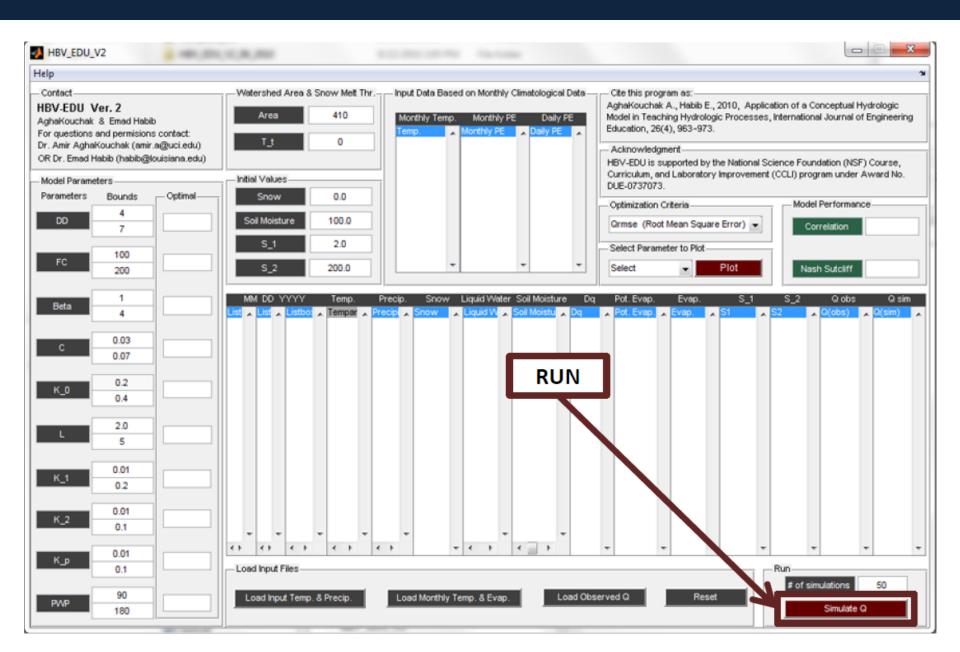


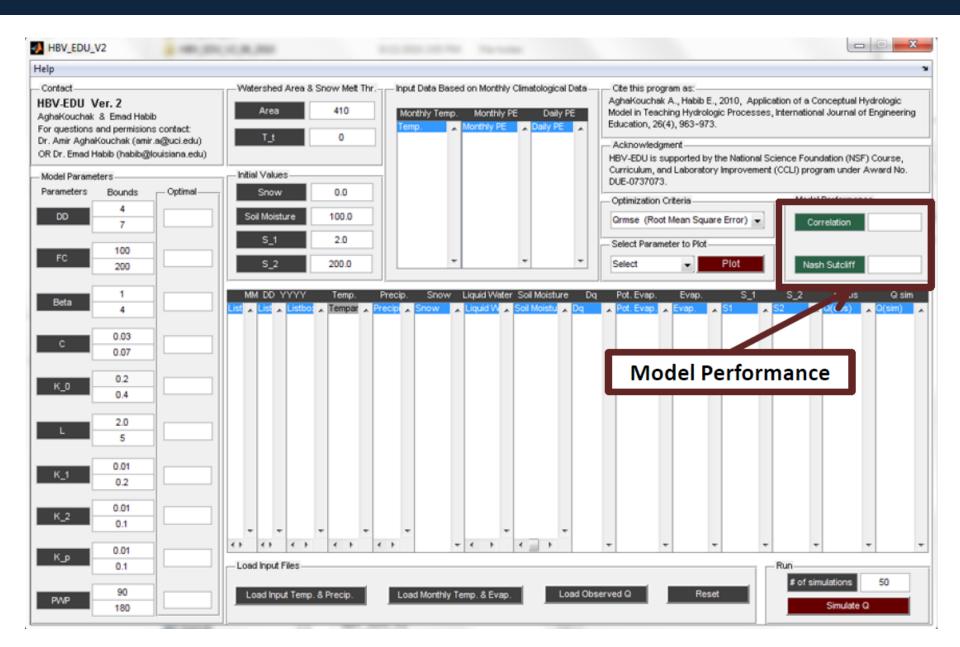


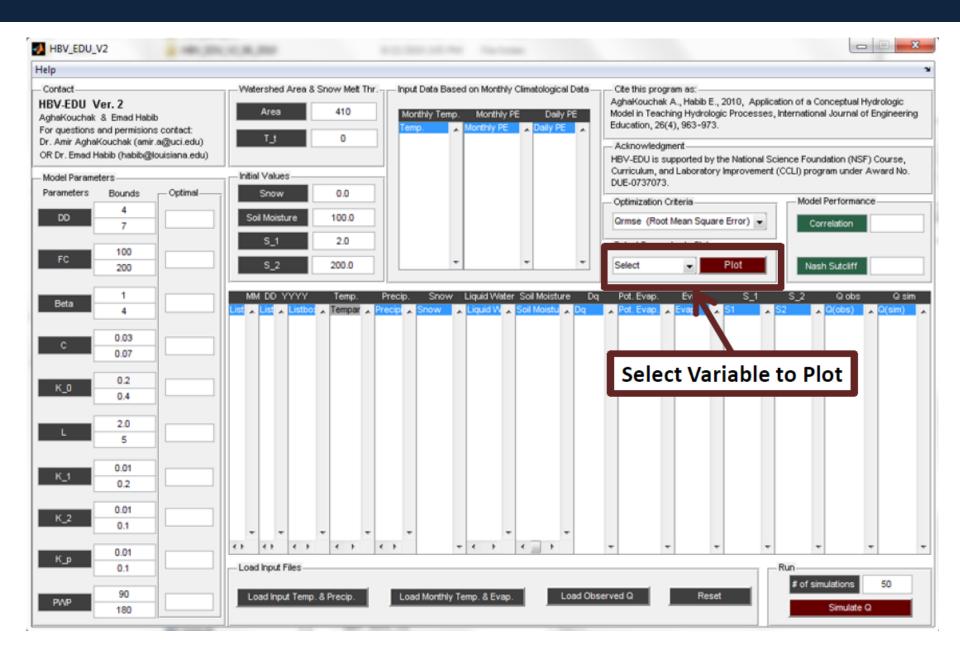




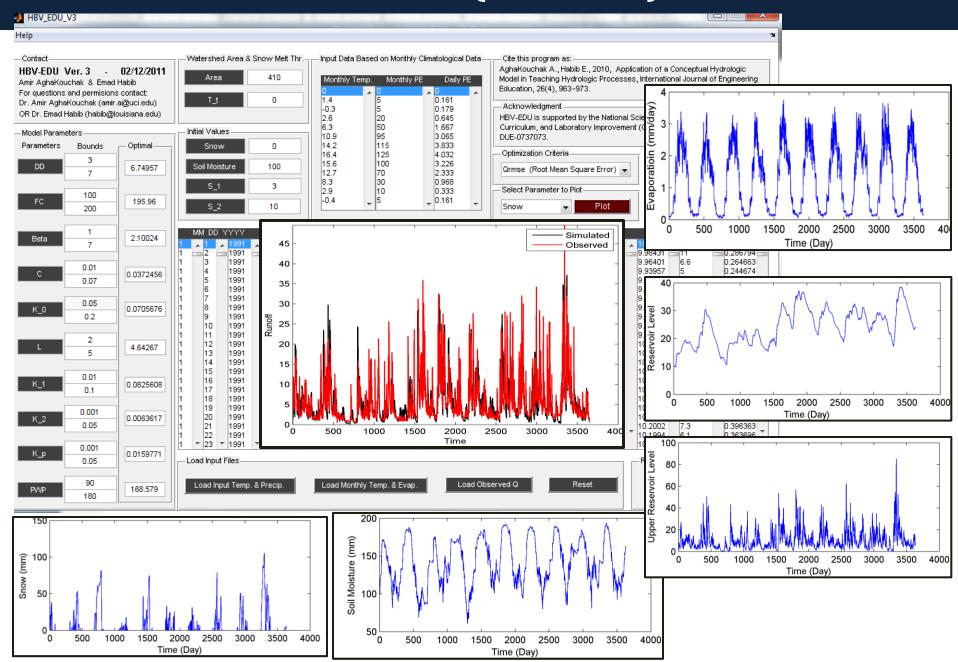








HBV-EDU (Version 3)



Alfred Wegener (1880 – 1930)

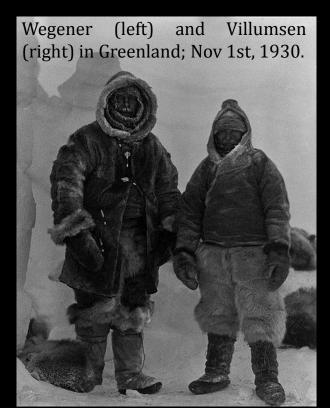


Famous hydrologist, meteorologist and interdisciplinary scientist Alfred Wegener was born in Berlin, Germany in November of 1880 He created the first balloons that were used to track weather and air masses. In order to better study the circulation of polar air, Wegener was part of several expeditions that went to Greenland. He and a companion went missing in November of 1930 on a Greenland expedition. Wegener's body was not found until May of 1931 (Sources: wikipedia.org & about.com).

"...Wegener and Villumsen took two dog sleds and made for West camp. They took no food for the dogs and killed them to feed the rest until they could only run one sled. They never reached the camp."









Multi-Index Drought Monitoring





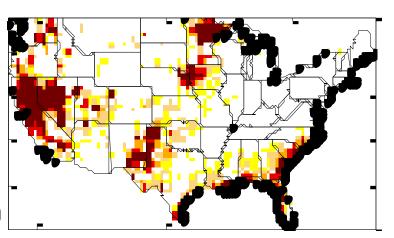
Different drought indices based on different climate variables (e.g., Precipitation, soil moisture):

- Standardized Precipitation Index (SPI)
- Standardized Soil Moisture Index (SSI)
- Standardized runoff Index (SRI)
- Palmer Drought Severity Index (PDSI)

SPI 2012-1

Intensity: D0 Abnormally Dry D1 Drought - Moderate D2 Drought - Extreme D3 Drought - Extreme D4 Drought - Exceptiona

SSI 2012-1

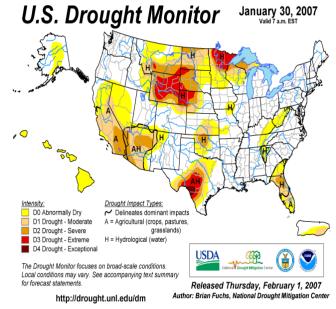


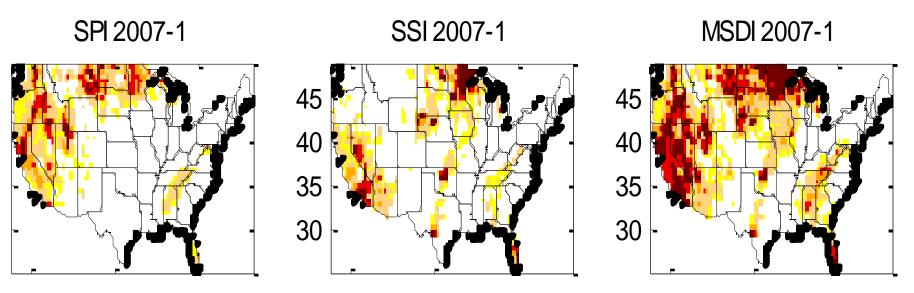
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Multi-Index Drought Monitoring

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1-Month SPI and SSI Derived Using NASA MERRA-LAND Precipitation and soil moisture Data.



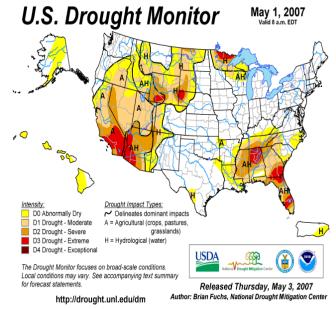


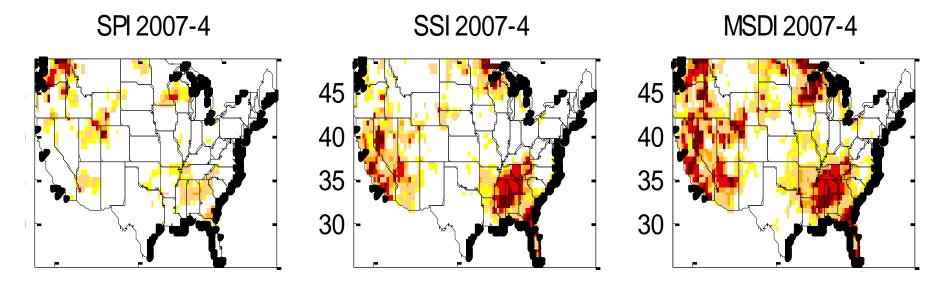
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Multi-Index Drought Monitoring

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1-Month SPI and SSI Derived Using NASA MERRA-LAND Precipitation and soil moisture Data.



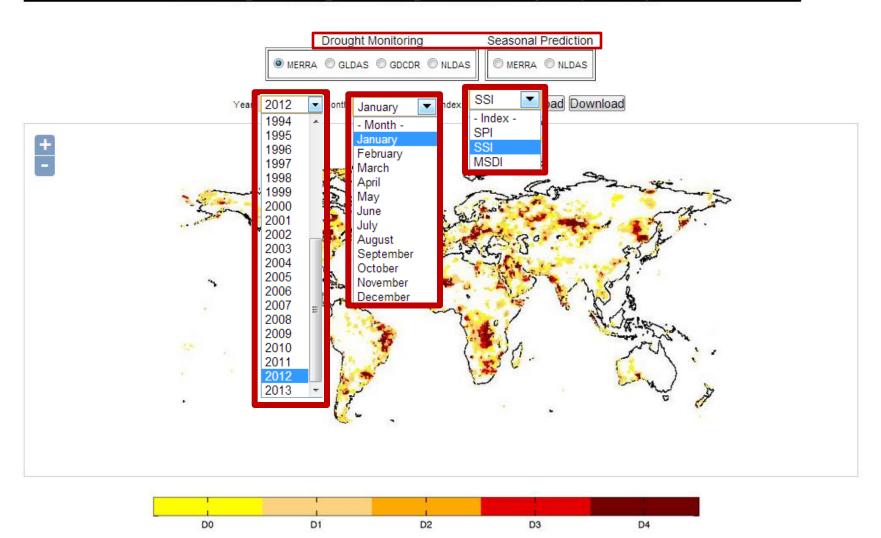






http://drought.eng.uci.edu/

Global Integrated Drought Monitoring and Prediction System (GIDMaPS)







Input Data Set	ID	Source	Resolution
NASA Modern-Era Retrospective analysis for Research and Applications – Reichle et al., 2011 - Precipitation and Soil Moisture	MERRA	NASA	2/3°x 1/2°
North American Land Data Assimilation System - Kumar et al., 2006 - Precipitation and Soil Moisture	NLDAS	NASA	0.125°
Global Drought Climate Data Record - AghaKouchak and Nakhjiri, 2012 – Precipitation – combines real-time PERSIANN satellite data (Sorooshian et al., 2000; Hsu et al., 1997) with long-term GPCP (Adler et al., 2001) observations.	GDCDR	UCI	0.5°
Global Land Data Assimilation System (GLDAS) - Peters-Lidard et al., 2007 - Precipitation and Soil Moisture	GLDAS	NASA	1°

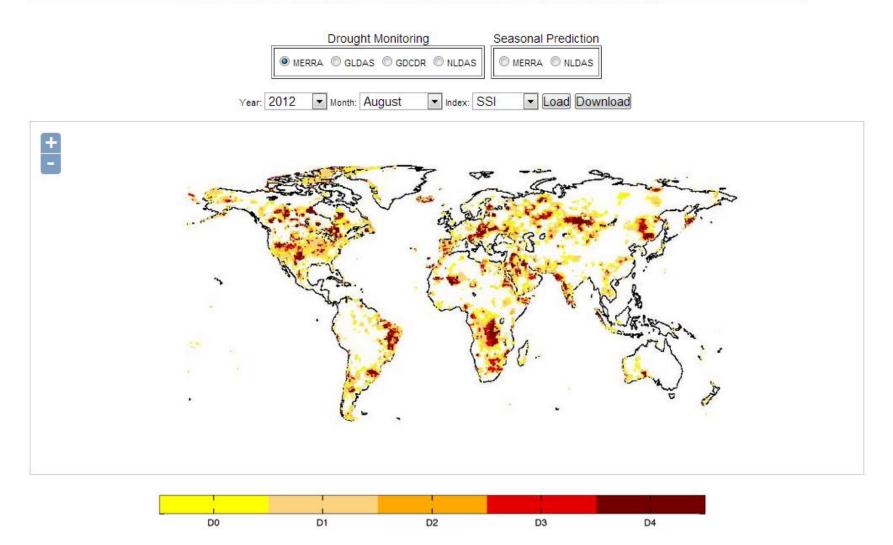
Drought Indicator	ID	Reference
Standardized Precipitation Index	SPI	McKee et al., 1993
Standardized Soil Moisture Index	SSI	Hao and AghaKouchak, 2013a
Multivariate Standardized Drought Index	MSDI	Hao and AghaKouchak, 2013a,b





http://drought.eng.uci.edu/index.html

Global Integrated Drought Monitoring and Prediction System (GIDMaPS)

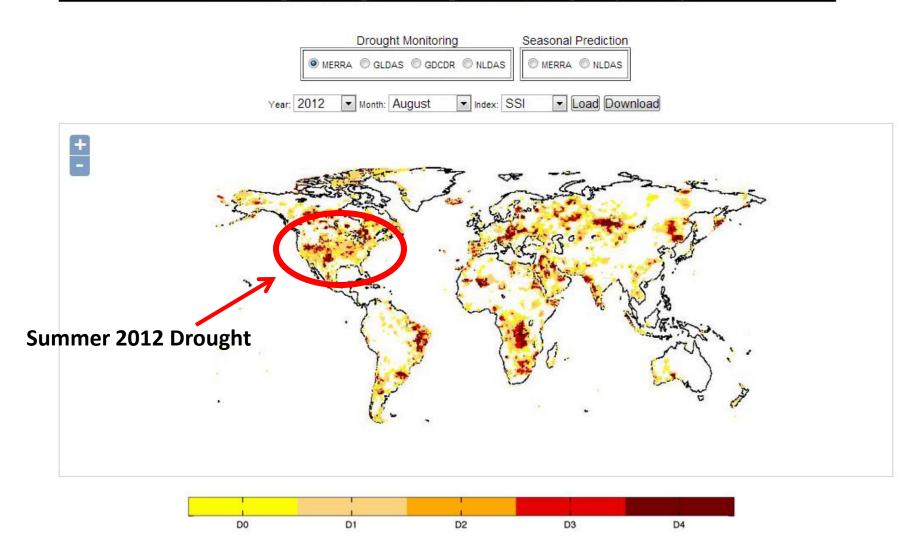






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Global Integrated Drought Monitoring and Prediction System (GIDMaPS)

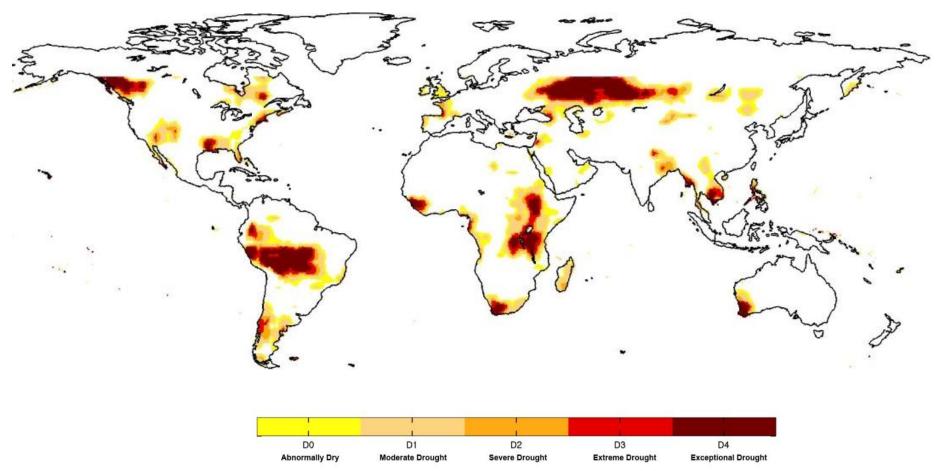






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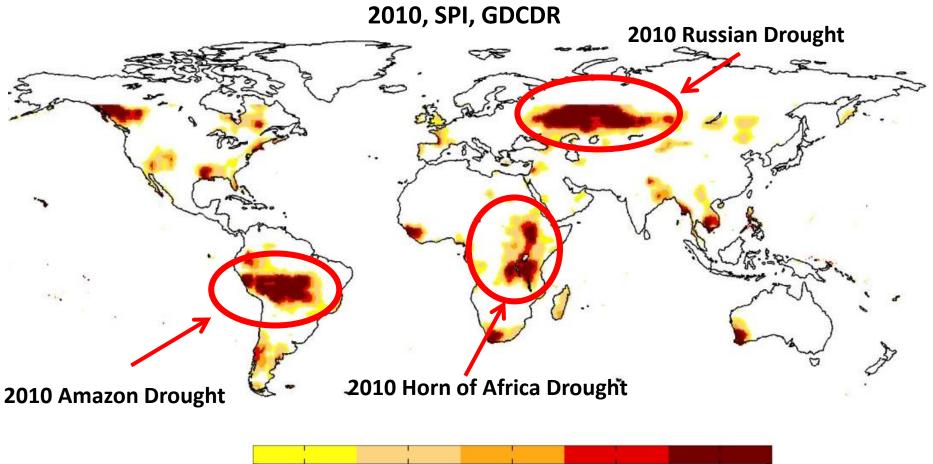
2010, SPI, GDCDR (PERSIANN combined with GPCP)

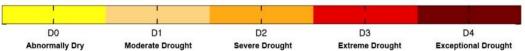






http://drought.eng.uci.edu/







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18 Months Real-Time

Prediction component is based on a drought persistence model which requires historical observations. The seasonal drought prediction component is based on two input data sets (MERRA and NLDAS) and three drought indicators (SPI, SSI and MSDI).

Ai+1(1)= Si-4+ Si-3+ Si-2 +Si-1+ Si +S(1)i+1 Ai+1(2)= Si-4+ Si-3+ Si-2 +Si-1+ Si +S(2)i+1

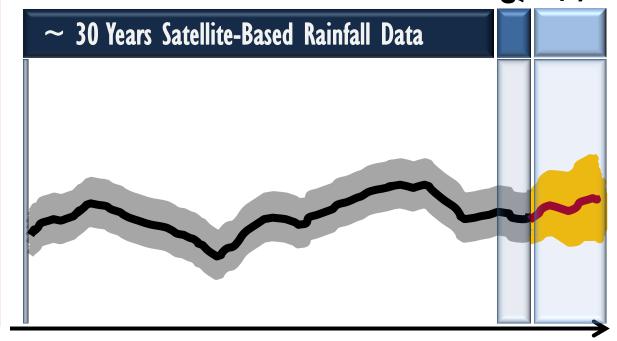
Ai+1(m)= Si-4+ Si-3+ Si-2 +Si-1+ Si +S(m)i+1 IOP PUBLISHING
Environ. Res. Lett. 7 (2012) 044037 (8pp)

A near real-time satellite-based global drought climate data record

Amir AghaKouchak and Navid Nakhjiri

University of California Irvine, E4130 Engineering Gateway Irvine, CA 92697, USA

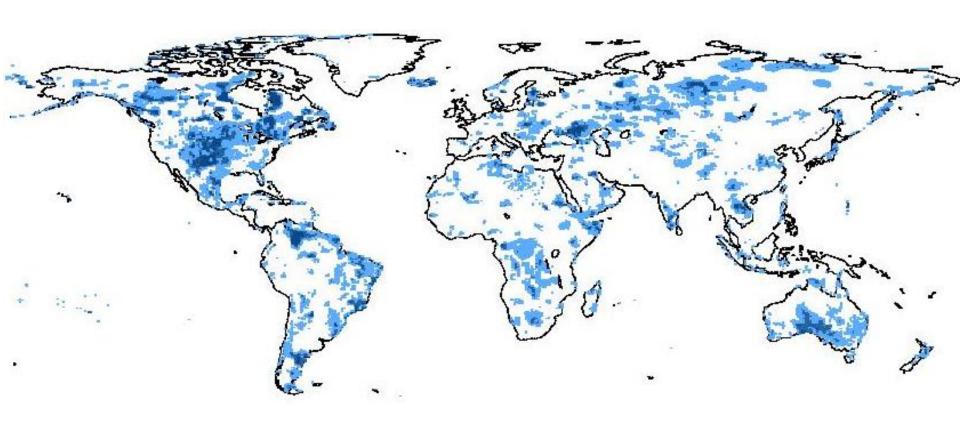
E-mail: amir.a@uci.edu and nnakhjir@uci.edu







March 2013

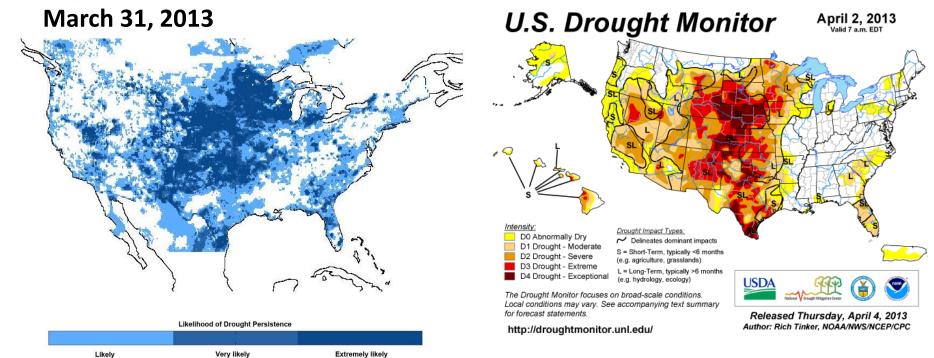


Likelihood of Drought Persistence

Likely Very likely Extremely likely







The probability values of the drought prediction component are converted to a 3-catergory drought likelihood measure:

- (a) drought likely to persist (≥70% probability)
- (b) drought very likely to persist (≥90% probability)
- (a) drought extremely likely to persist (≥95% probability)



Validation Toolbox: Evaluation of Climate and Remote Sensing Data

Validation Toolbox



http://amir.eng.uci.edu/downloads/ValidationToolbox.zip



Performance Metrics **Evaluation** of Remote Sensing Observations and Climate Model Simulations: A simple and easy to use Validation Toolbox (MATLAB source code) that can be for validation used gridded data including satellite observations. reanalysis data, and weather and climate model simulations. In addition to the commonly used categorical indices, the includes toolbox the Volumetric Hit Index (VHI), Volumetric False Alarm Ration (VFAR), Volumetric Missed Index (VMI), and Volumetric Critical Success Index (VCSI).

