

# Evaluation of SWAT Hydrological Model for Streamflow Simulation In Yasu River, Japan

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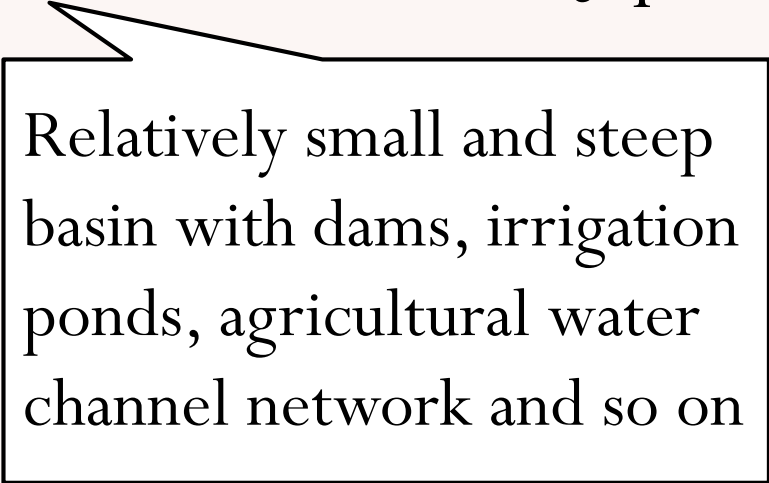
# Introduction

- Water is fundamental resources for all lives on earth including human beings.
- High threats such as increasing population, pollution, and climate change lie in its future availability.
- Establishment of sustainable water use is urgent need.
- For suitable water management, grasping local hydrological environment and local climate is inevitable.
- Hydrological modeling is one of the most popular method.
- SWAT (Soil Water Assessment Tool) is employed to model Japanese basin.

# Objectives

## Main Objective

- To simulate streamflow in Yasu River Basin, Japan

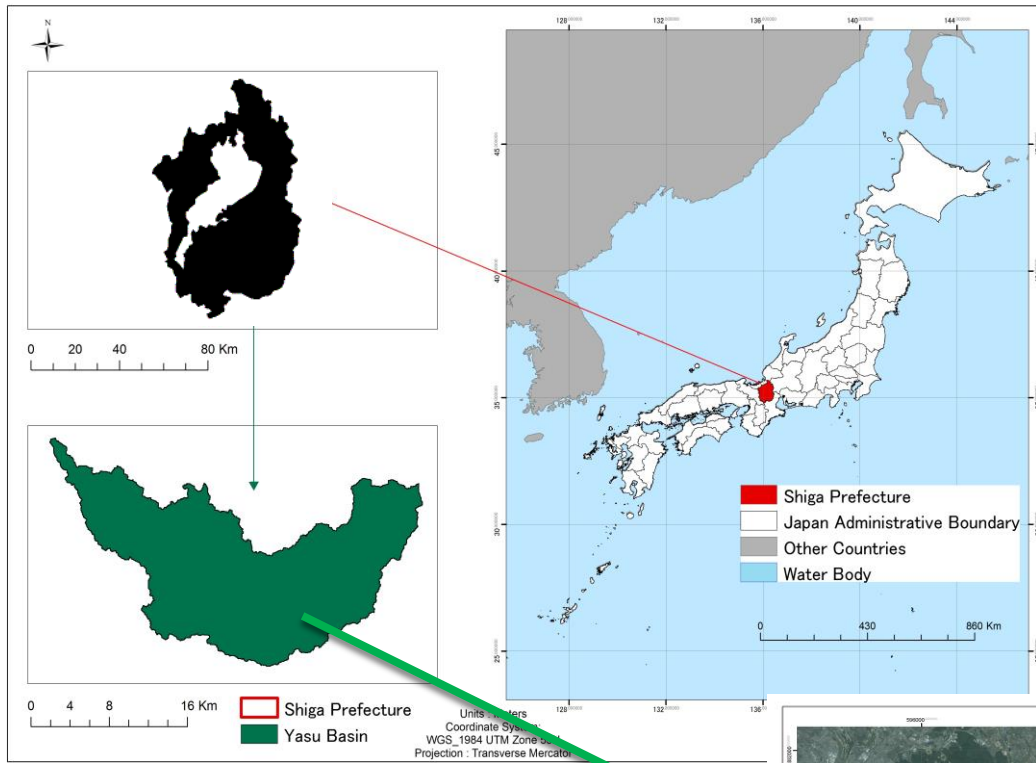


Relatively small and steep basin with dams, irrigation ponds, agricultural water channel network and so on

## Specific objectives

- To check availability of SWAT to Japanese basin through **sensitivity analysis, calibration, validation, and uncertainty analysis.**

# Study area: Yasu River basin



## Elevation

- 97masl to 1235masl

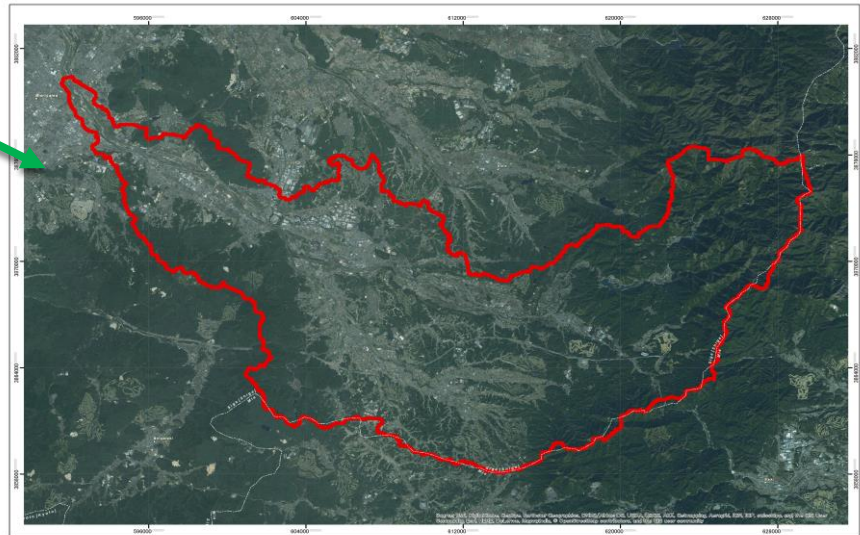
## Land use

- Forest, Paddy field, Settlement

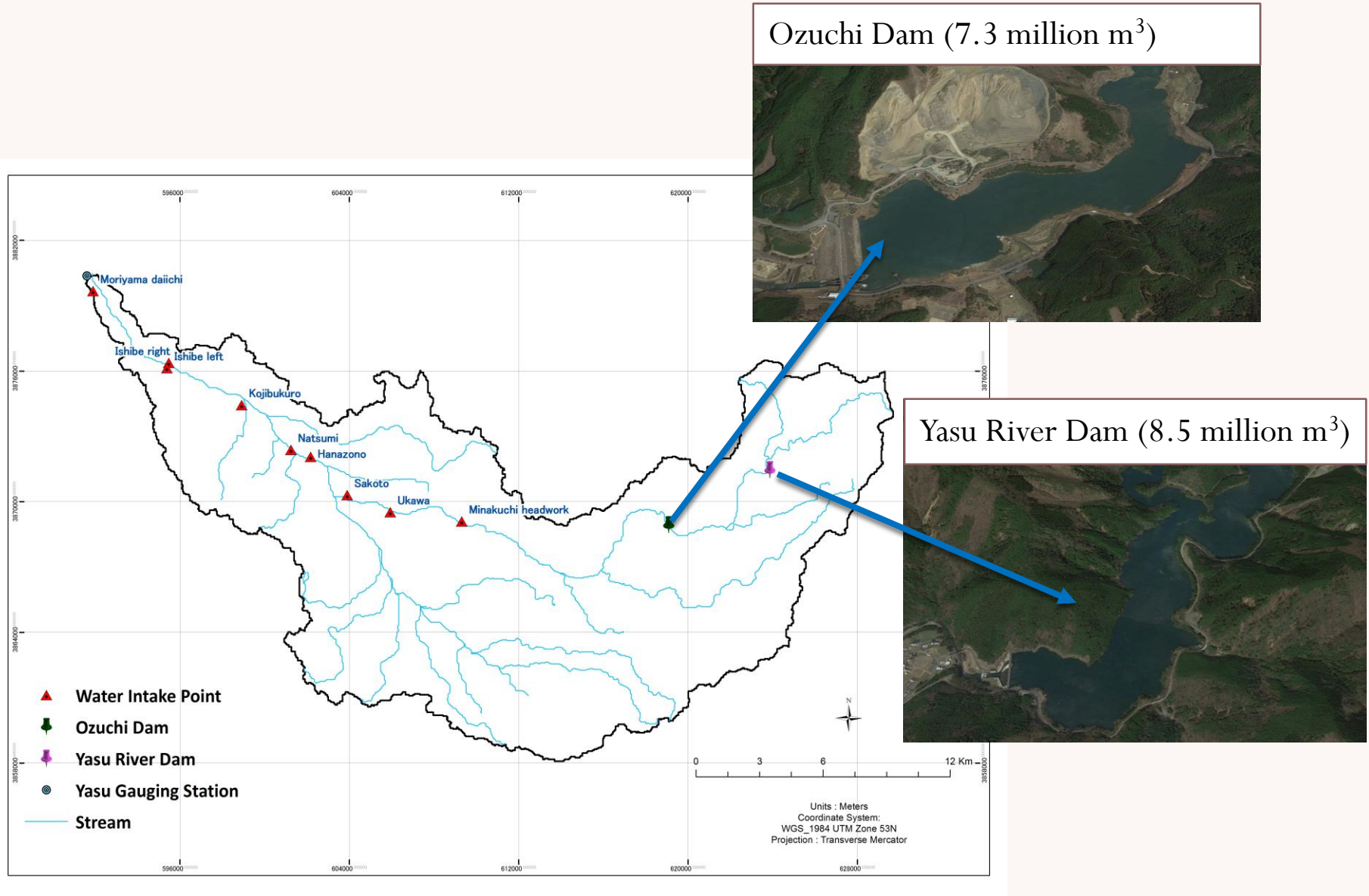
## Soil

- Brown forest soil

Rain: 1,587 mm  
Area: 377 km<sup>2</sup>  
Length: 65 Km



# Study area: Yasu River basin



# SWAT model

Uses water balance equation (Arnold et al 1998)

$$SW_t = SW_0 + \sum_{i=1}^t (R_{day} - Q_{surf} - E_a - W_{seep} - Q_{gw})_i$$

$SW_t$  = final soil water content (mm)

$SW_0$  = soil water content on day  $i$  (mm)

$t$  = time in days

$R_{day}$  = of precipitation on day  $i$  (mm)

$Q_{surf}$  = amount of surface runoff on day  $i$  (mm)

$E_a$  = amount of evaporation on day  $i$  (mm)

$W_{seep}$  = amount of water entering the vadose zone from soil profile on day  $i$  (mm)

$Q_{gw}$  = amount of return flow on day  $i$  (mm)

# SWAT CUP (Calibration and Uncertainty Programs)

- Application that conducts calibration, validation, and sensitivity and uncertainty analyses for SWAT model.
- Optimization programs
  - SUFI-2 (Sequential Uncertainty Fitting)
  - GLUE (Generalized Likelihood Uncertainty Estimation)

## Sensitivity analysis

- Sensitivity is given as the average change of objective function as a result of change in each parameter.

$$g = \alpha + \sum_{i=1}^m \beta_i b_i$$

$$\alpha = \frac{\sigma_s}{\sigma_m}$$

$$\beta = \frac{\mu_s}{\mu_m}$$

$b_i$  = parameter,

$\sigma_s$  and  $\sigma_m$  = standard deviation of simulated and measured data

$\mu_s$  and  $\mu_m$  = are means of simulated and measured data.



# Model Efficiency

- By  $R^2$  (coefficient of determination) and  $NS$  (Nash-Sutcliffe), the model efficiency is evaluated.

$$R^2 = \frac{[\sum_{i=1}^n (Q_{obs,i} - \bar{Q}_{obs})(Q_{sim,i} - \bar{Q}_{sim})]^2}{\sum_{i=1}^n (Q_{obs,i} - \bar{Q}_{obs})^2 \sum_{i=1}^n (Q_{sim,i} - \bar{Q}_{sim})^2}$$

$$NS = 1 - \left[ \frac{\sum_{i=1}^n (Q_{obs,i} - Q_{sim,i})^2}{\sum_{i=1}^n (Q_{obs,i} - \bar{Q}_{obs})^2} \right]$$

$Q_{obs,i}$  = observed streamflow

$Q_{sim,i}$  = simulated streamflow

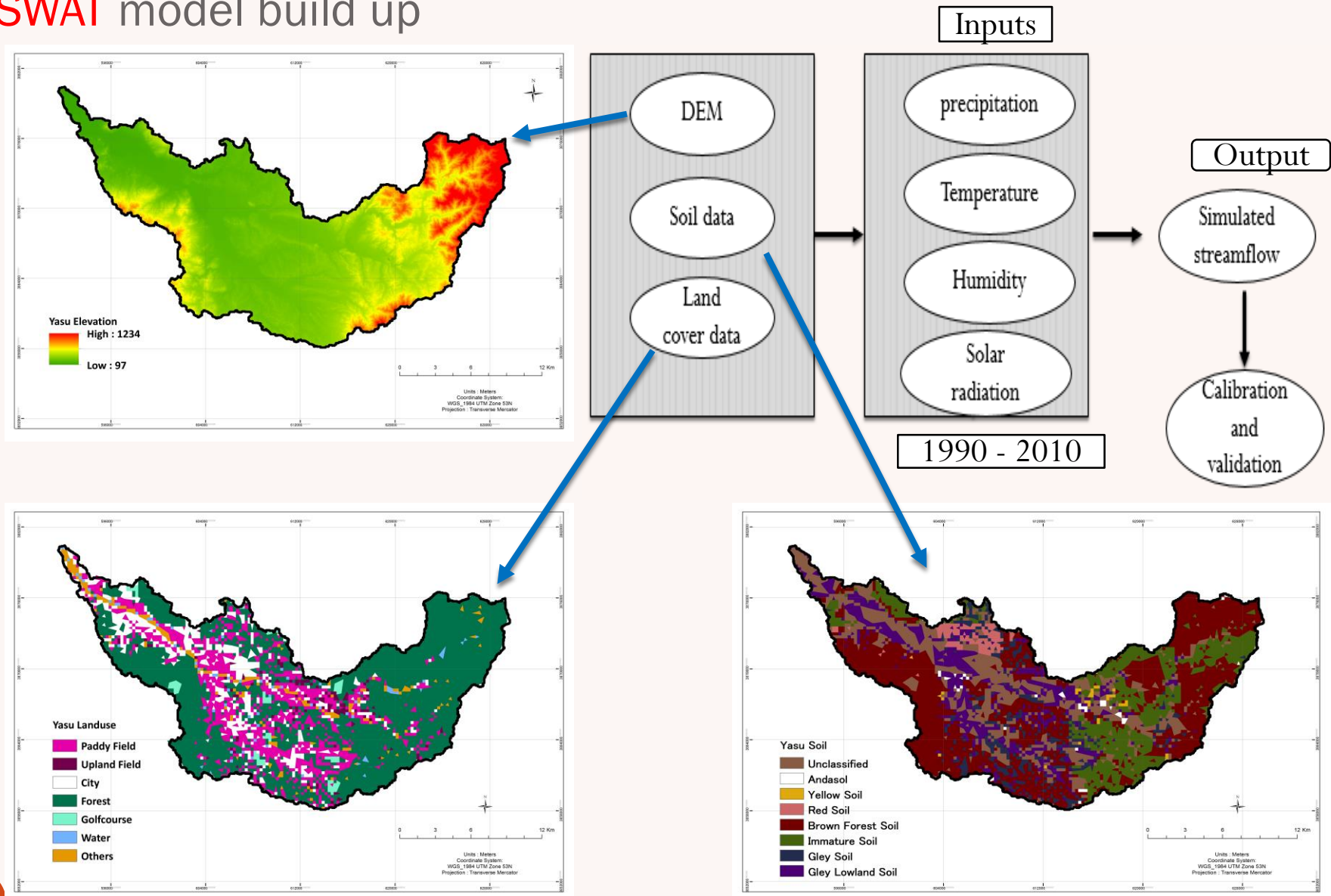
$\bar{Q}_{obs}$  = mean of observed streamflow

$\bar{Q}_{sim}$  = mean of simulated stream flow

$n$  = total number of observation

# Application

## SWAT model build up



# Sensitivity analysis results of Monthly Streamflow

Parameter	SUFI-2			GLUE		
	t-stat	p-value	Rank	t-stat	p-value	Rank
R_CN2.mgt	37.947	7.60E-195	1	41.551	9.94E-219	1
V_ALPHA_BF.gw	18.953	1.50E-150	2	17.982	9.94E-63	2
V_ALPHA_BF.gw	16.9	1.50E-150	3	16.9	6.33E-33	3
V_ALPHA_BF.gw	15.1	1.50E-150	4	15.1	1.89E-32	4
V_ALPHA_BF.gw	13.79	1.50E-150	5	13.79	6.01E-11	5
V_ALPHA_BF.gw	12.49	1.50E-150	6	12.49	1.49E-06	6
V_ALPHA_BF.gw	11.22	1.50E-150	7	11.22	3.93E-06	7
R_SOL_K(..).sol	2.3836	0.0173	8	4.1435	3.71E-05	8
V_GWQMN.gw	-1.966	0.0498	9	-3.488	0.000508	9
V_GW_REVAP.gw	-1.479	0.139	10	-1.866	0.0622	10
V_GW_DELAY.gw	-0.588	0.556	11	-1.836	0.0667	11
R_SOL_AWC(..).sol	-0.553	0.580	12	-0.886	0.375	12
V_REVAPMN.gw	-0.382	0.702	13	-0.543	0.587	14
A_ESCO.hru	-0.371	0.710	14	-0.471	0.637	15
V_SURLAG.bsn	-0.302	0.762	15	-0.722	0.470	13

A coefficient of a parameter divided by its standard error

P-value is given from Student's t distribution table

# Sensitivity analysis results of Daily Streamflow

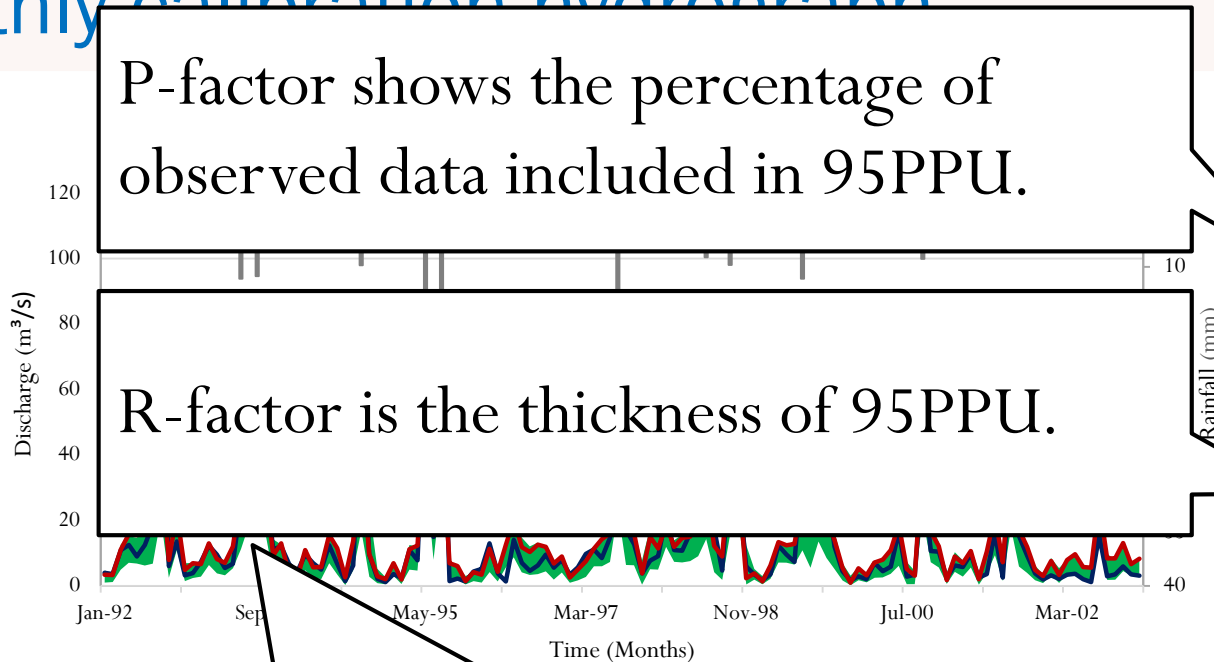
Parameter	SUFI-2			GLUE		
	t-stat	p-value	Rank	t-stat	p-value	Rank
R__CN2.mgt	13.47	4.04E-38	2	17.97	1.15E-62	4
V__ALPHA_BF.gw	18.97	1.13E-68	1	19.70	4.30E-73	3
V__SLSUBBSN.hru	-17.35	4.75E-59	4	-20.56	1.90E-78	2
V__HRU_SLP.hru	13.21	8.03E-37	5	16.60	8.47E-55	6
V__CH_K2.rte	-17.42	1.99E-59	3	-21.35	2.03E-83	1
V__CH_N2.rte	-5.158	3.02E-07	7	-6.646	4.99E-11	7
V__OV_N.hru	-12.46	3.28E-33	6	-17.96	1.36E-62	5
R__SOL_K(..).sol	0.715	0.474	10	1.552	0.120	9
V__GWQMN.gw	0.782	0.433	9	-1.364	0.172	10
V__GW_REVAP.gw	0.493	0.621	13	0.656	0.511	14
V__GW_DELAY.gw	0.628	0.529	12	-0.863	0.388	12
R__SOL_AWC(..).sol	-0.383	0.701	15	-0.412	0.680	8
V__REVAPMN.gw	0.436	0.662	14	-0.881	0.378	13
A__ESCO.hru	-0.711	0.477	11	-0.462	0.644	15
V__SURLAG.bsn	-1.278	0.201	8	-1.068	0.285	11

# Fitted value

Parameter Name	Min_value	Max_value	Fitted Value			
			SUFI-2 Daily	GLUE Daily	SUFI-2 Monthly	GLUE Monthly
R__CN2.mgt	-0.025	0.5	0.404	0.432	0.493	0.342
V__ALPHA_BF.gw	0	1	0.870	0.885	0.325	0.377
V__SLSUBBSN.hru	10	150	33.310	24.384	28.270	16.827
V__HRU_SLP.hru	0	1	0.725	0.404	0.821	0.616
V__CH_K2.rte	0	150	88.725	67.53	64.275	54.644
V__CH_N2.rte	0.001	0.3	0.284	0.191	0.122	0.211
V__OV_N.hru	0.01	30	11.331	11.719	4.433	15.861
R__SOL_K(..).sol	-0.25	0.25	-0.067	-0.055	-0.036	0.127
V__GWQMN.gw	0	5000	4857.5	413.377	4127.5	990.442
V__GW_REVAP.gw	0.02	0.2	0.096	0.030	0.056	0.167
V__GW_DELAY.gw	0	500	62.750	275.576	5.750	341.898
R__SOL_AWC(..).sol	-0.25	0.1	-0.161	-0.064	0.034	0.038
V__REVAPMN.gw	0	500	20.75	208.58	173.25	375.791
A__ESCO.hru	0	1	0.733	0.732	0.742	0.413
V__SURLAG.bsn	0.5	5	1.208	0.710	0.583	3.278

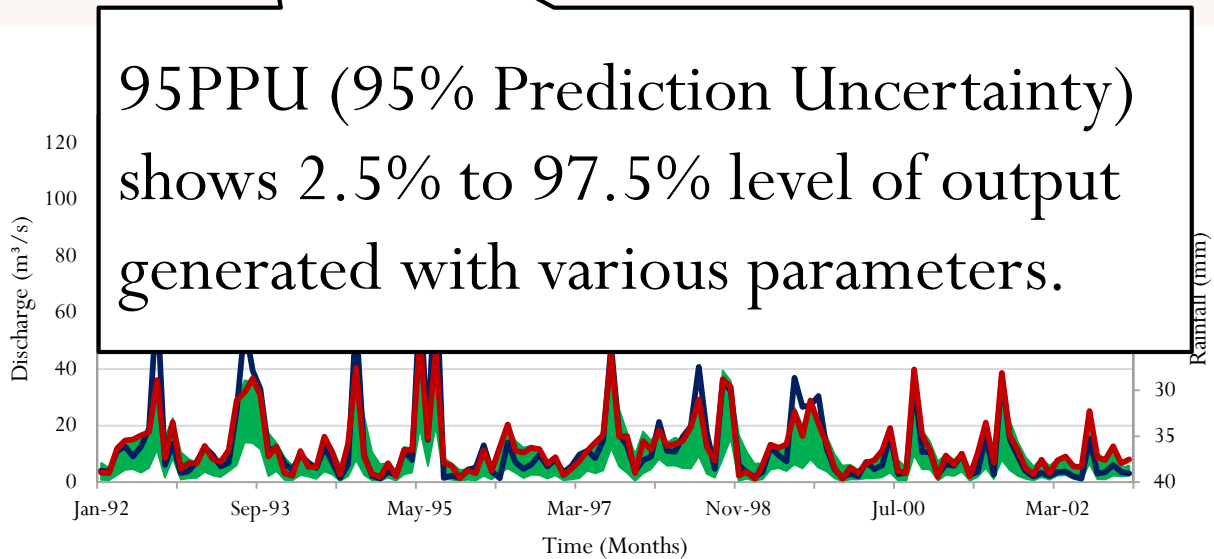
# Monthly calibration hydrograph

## GLUE



$R^2=0.85$   
NS=0.82  
P-factor=0.74  
R-factor=0.58

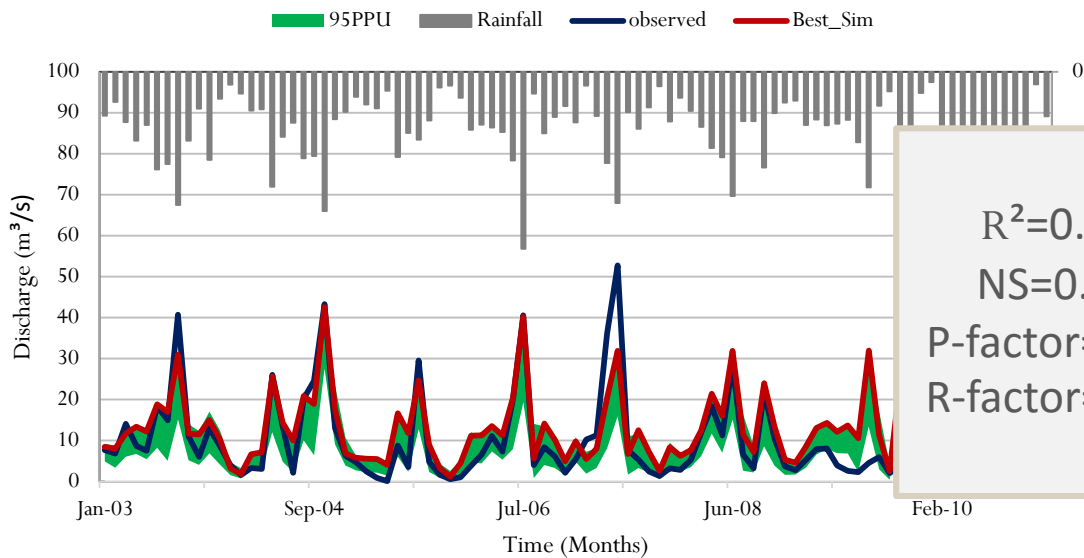
## SUFI-2



$R^2=0.85$   
NS=0.82  
P-factor=0.86  
R-factor=0.77

# Monthly validation hydrograph

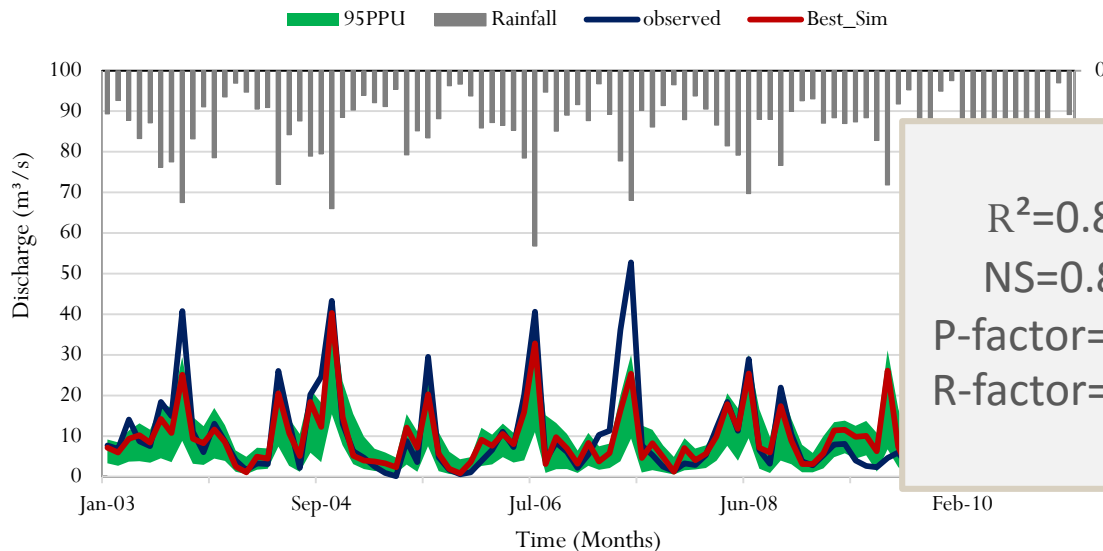
**GLUE**



$R^2=0.85$   
 $NS=0.82$   
 $P\text{-factor}=0.74$   
 $R\text{-factor}=0.58$

$R^2=0.64$   
 $NS=0.63$   
 $P\text{-factor}=0.66$   
 $R\text{-factor}=0.78$

**SUFI-2**

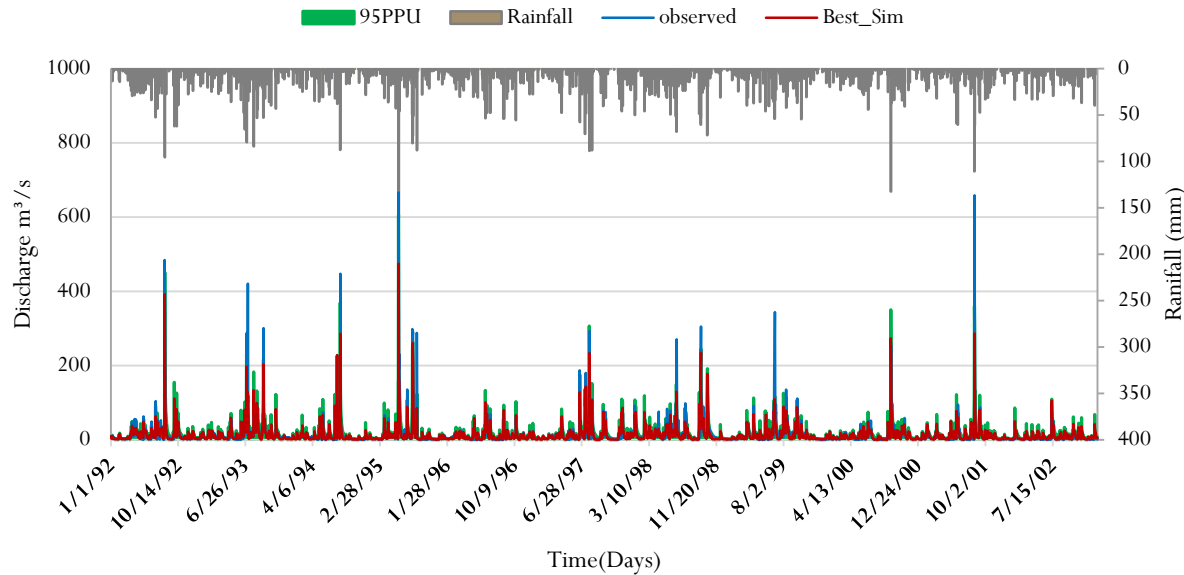


$R^2=0.85$   
 $NS=0.82$   
 $P\text{-factor}=0.86$   
 $R\text{-factor}=0.77$

$R^2=0.63$   
 $NS=0.62$   
 $P\text{-factor}=0.82$   
 $R\text{-factor}=1.03$

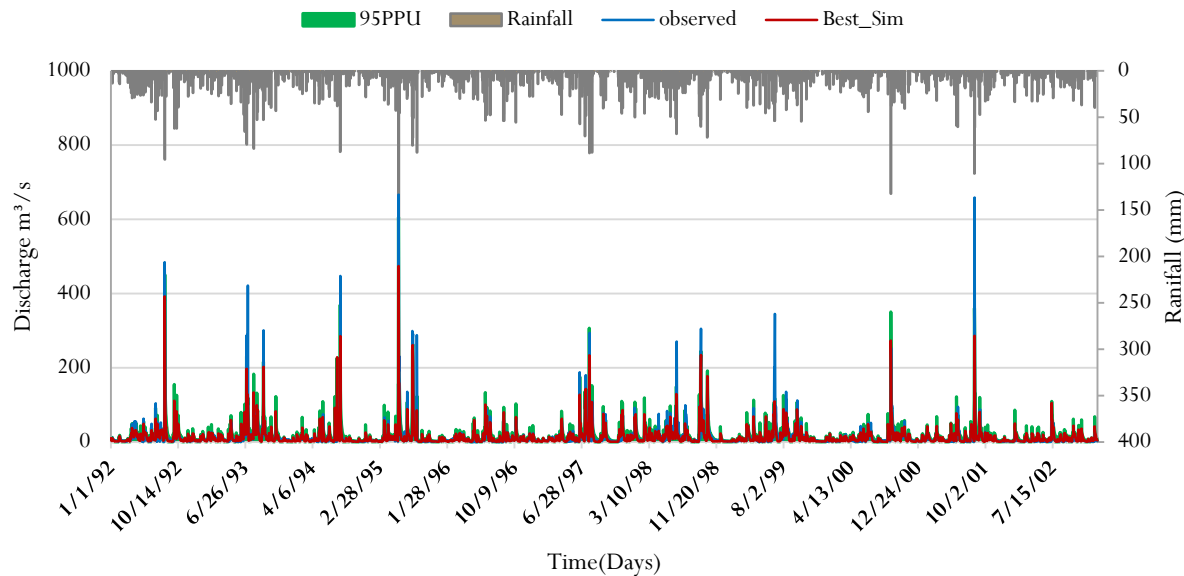
# Daily calibration hydrograph

GLUE



$R^2= 0.71$   
NS= 0.70  
P-factor=0.80  
R-factor=0.45

SUFI-2

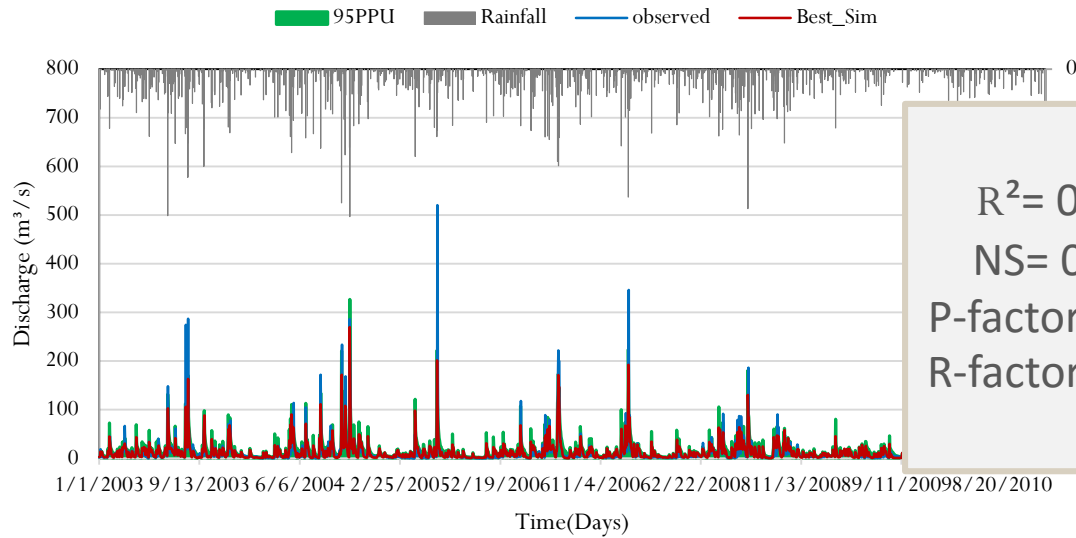


$R^2=0.71$   
NS=0.70  
P-factor=0.90  
R-factor=0.56



# Daily validation hydrograph

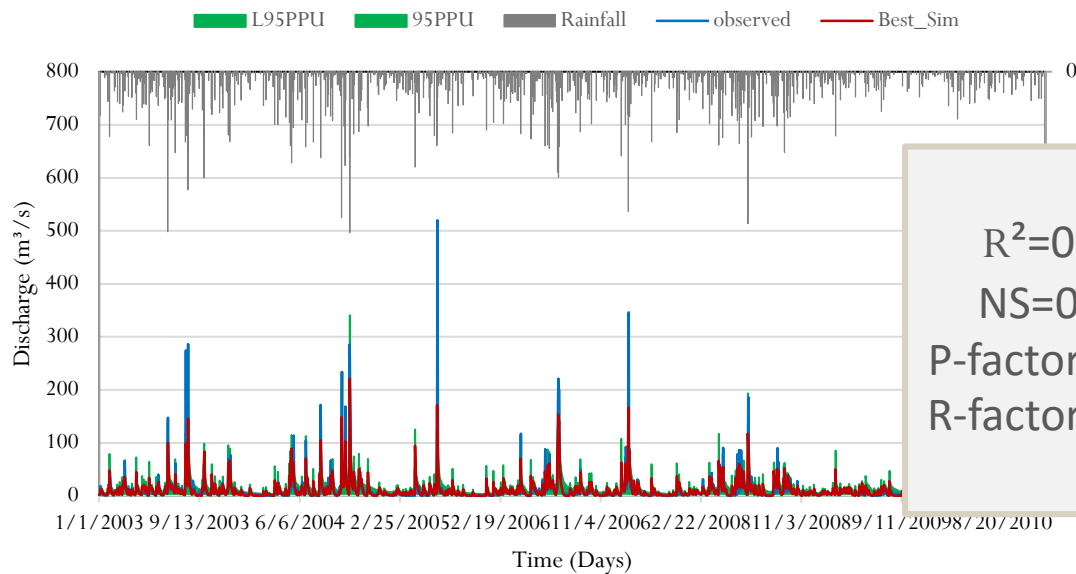
**GLUE**



$R^2= 0.71$   
 $NS= 0.70$   
 $P\text{-factor}=0.80$   
 $R\text{-factor}=0.45$

$R^2=0.60$   
 $NS=0.59$   
 $P\text{-factor}=0.88$   
 $R\text{-factor}=0.58$

**SUFI-2**



$R^2=0.71$   
 $NS=0.70$   
 $P\text{-factor}=0.90$   
 $R\text{-factor}=0.56$

$R^2=0.62$   
 $NS=0.61$   
 $P\text{-factor}=0.94$   
 $R\text{-factor}=0.71$

# Conclusion

- This study provided an approach for streamflow simulation based on SWAT model in a Japanese relatively small and steep basin.
- Sensitivity analysis, calibration, validation and uncertainty analysis were carried out.
- Results revealed parameters' rank on sensitivity based on optimization techniques (SUFI-2 and GLUE) in daily and monthly stream flows.
- Monthly period yielded high NS and  $R^2$  as compared to daily period.
- Uncertainty range was larger in SUFI-2 as compared to GLUE for both daily and monthly period during calibration and validation.

# Future Plan

- To introduce artificial structures such as dams, irrigation ponds, channel network, head works, and pumping stations into SWAT model.