

# Evaluating the Impact of Different Input Datasets and Model Configuration Uncertainty on Streamflow Simulations by Using SWAT Model

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# Outline

- Introduction
- Aim and Scope
- Study Area
- Model Setup
- Results
- Conclusion

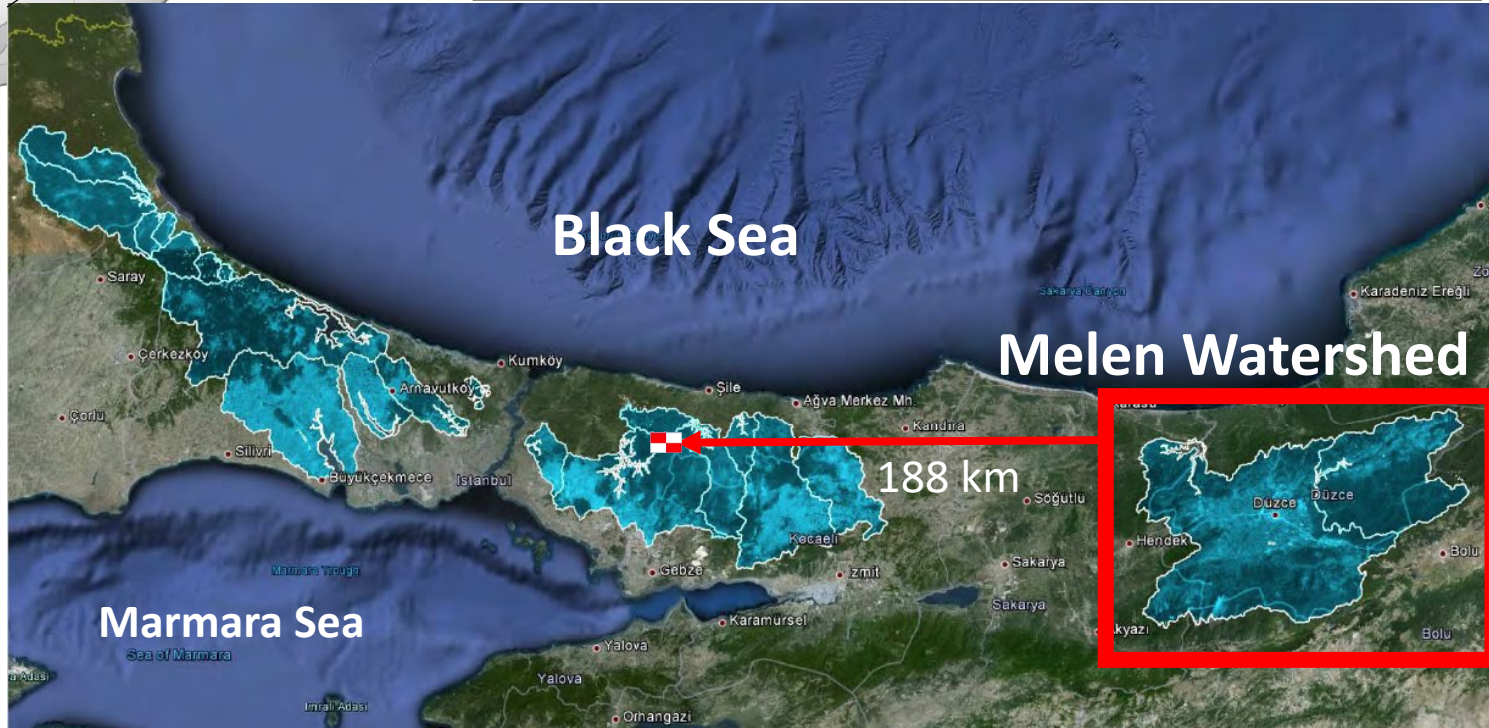
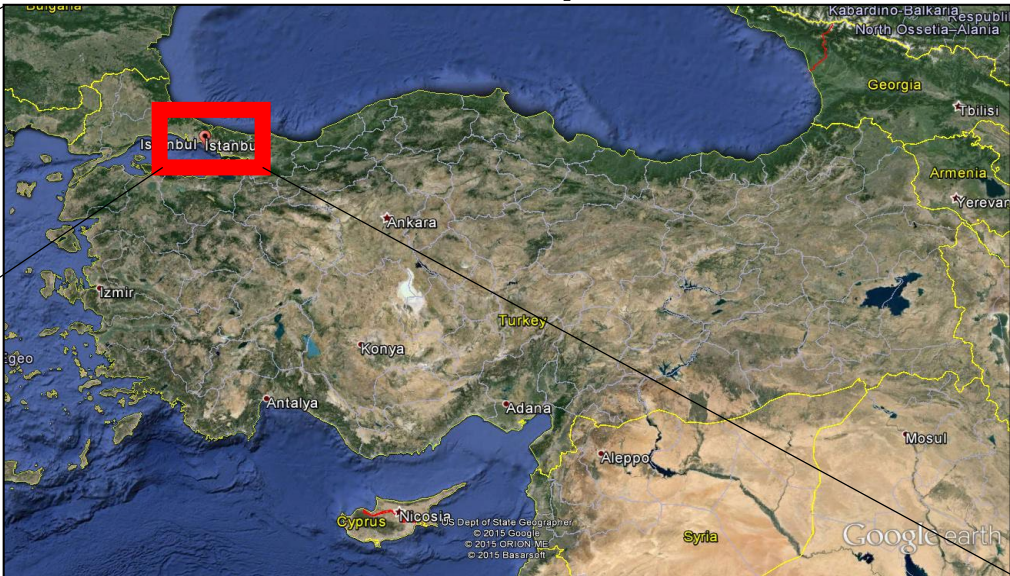
# Introduction

- Performance of the watershed modeling studies is closely associated with the **model input datasets**
- **Numerous databases** are existing and freely available to the modelers in global or regional scale
- These datasets provide required input data to run models, their **resolution** and **accuracies** are needed to be **investigated**.

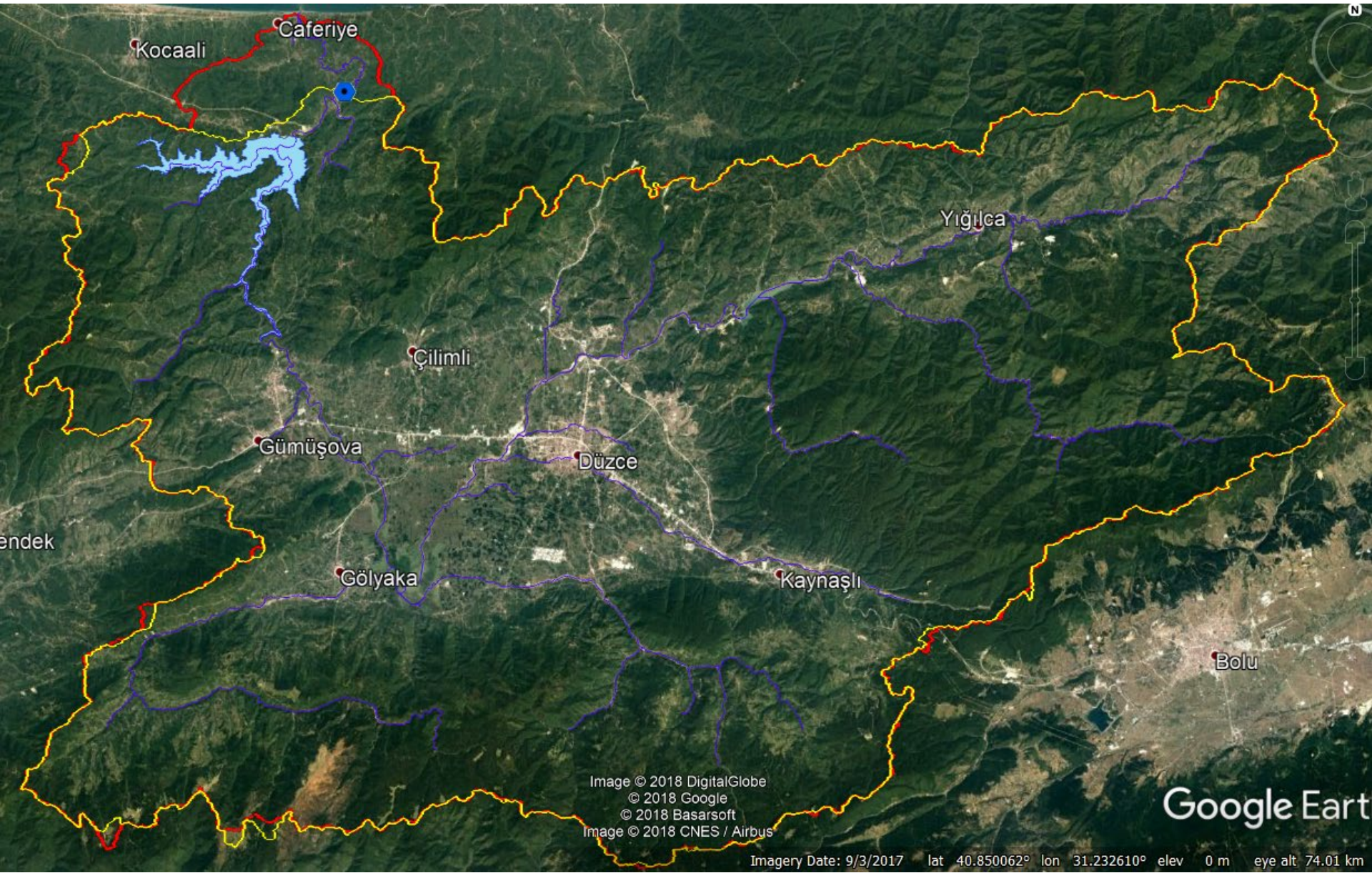
# Aim and Scope

- Investigate the uncertainty caused from
  - Model input data
  - Model configuration (in terms of subbasin number)
  - Model parameterization
- For this purpose we setup a model for Melen Watershed
  - 2 different climate dataset
  - 3 different LULC dataset
  - 3 different subbasin number
- We consider only hydrology
  - Streamflow at outlet of the watershed

# Study Area (Melen Watershed)



# Study Area (Melen Watershed)



# Data Sources

- Digital Elevation Map SRTM 90m
- Soil FAO
- Land Use Data Set
  - Corine Land Cover
  - Landsat Land Use Map
  - STATIP (Turkish Ministry of Agriculture)
- Climate Data Set
  - CFSR (1979-2014)
  - MGM - Local Measurement (1960-2013)
- Discharge Data
  - Turkish State of Hydraulics (1 station - monthly)

Compulsory  
Data

Calibration

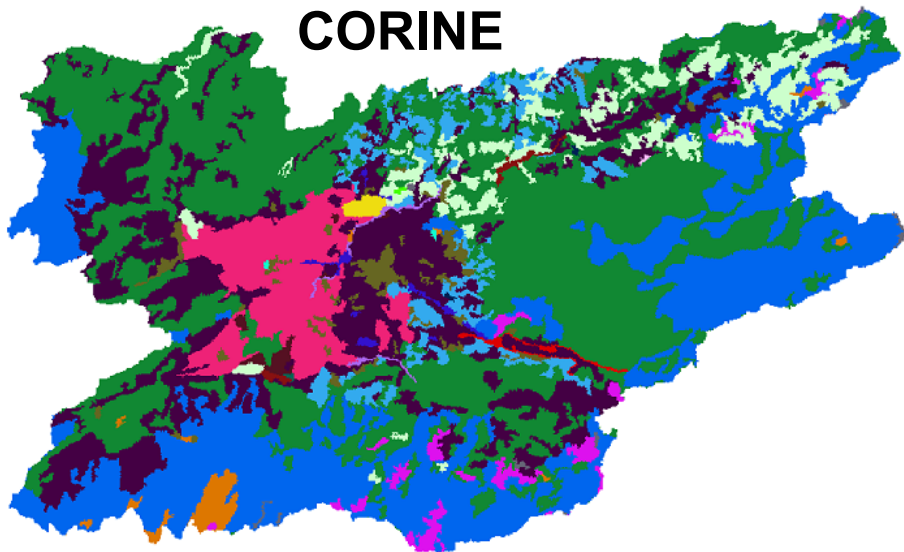
# Model Setup

- Simulation period (1995) 2000 - 2013
- 5 years warmup
- SCS-CN Method
- Hargreaves Method
  
- SWAT-CUP using SUFI-2 algorithm
- 60% of streamflow used for calibration
- NSE as an objective function
- Performance criteria NSE, R2, PBIAS as well as P-factor and R-factor



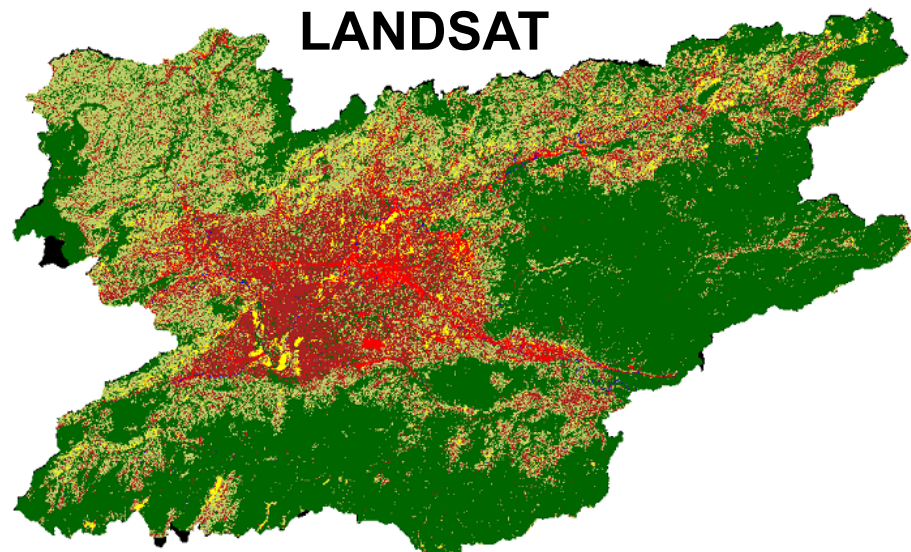
# Landuse Data

**CORINE**



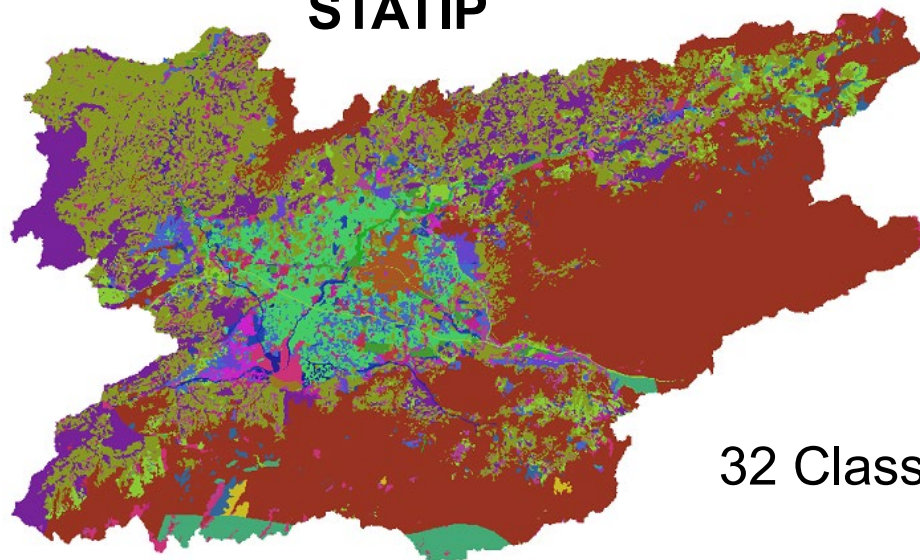
22 Classes, 100m

**LANDSAT**



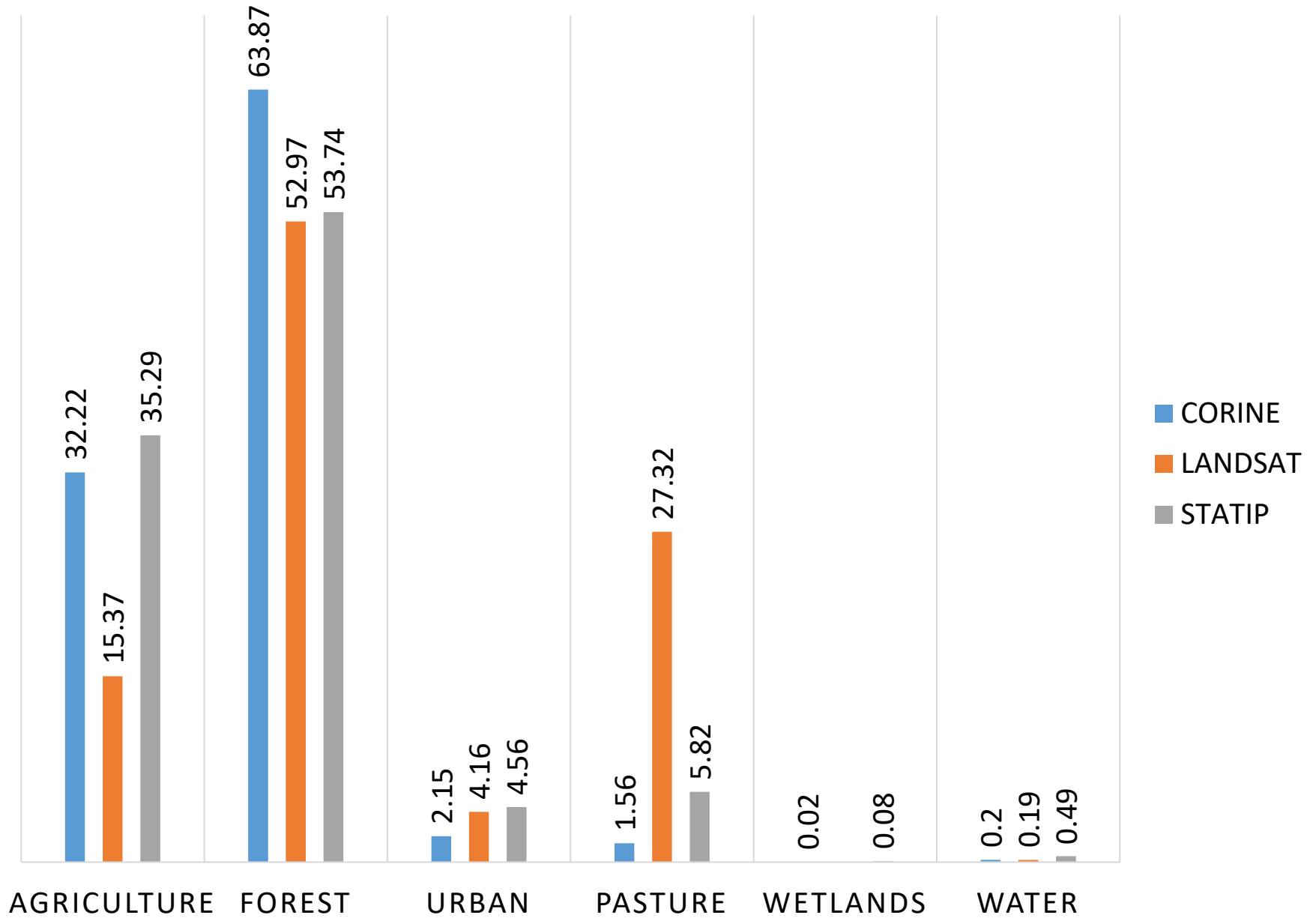
7 Classes, 30m

**STATIP**



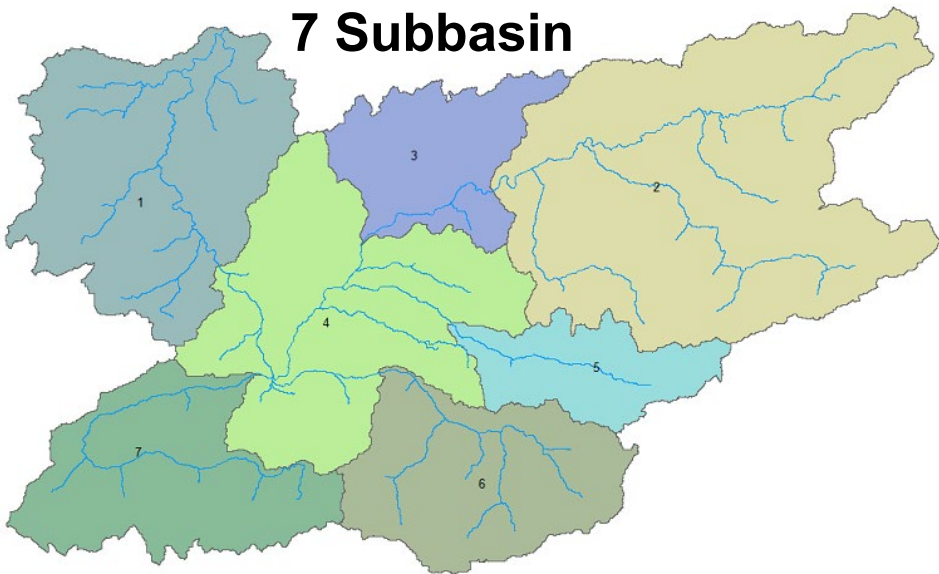
32 Classes, 10m

# Landuse Data

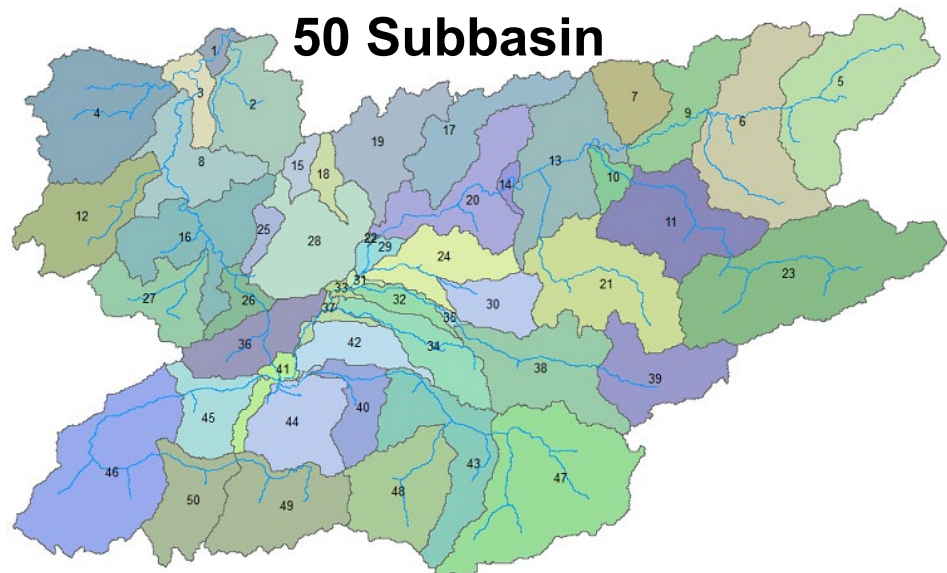


# Subbasin Configuration

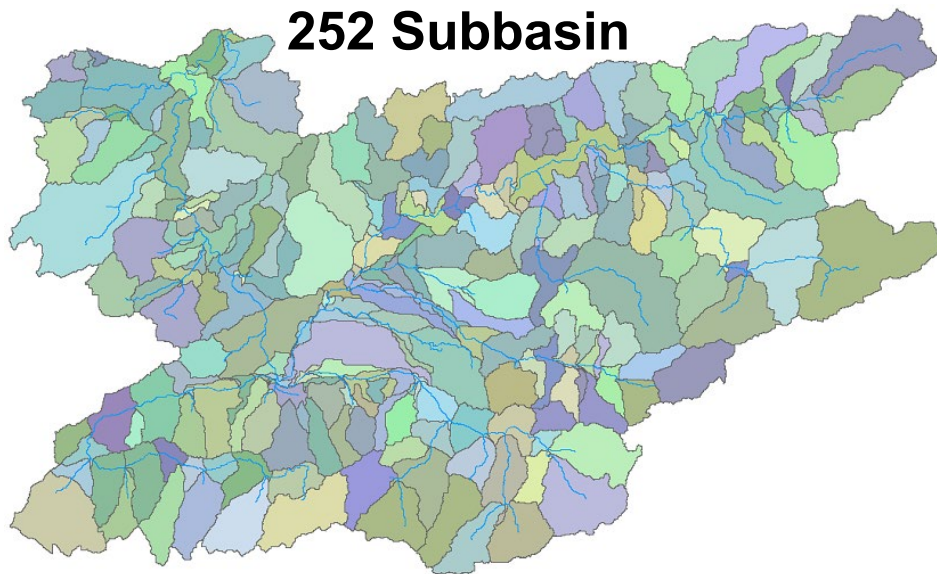
**7 Subbasin**



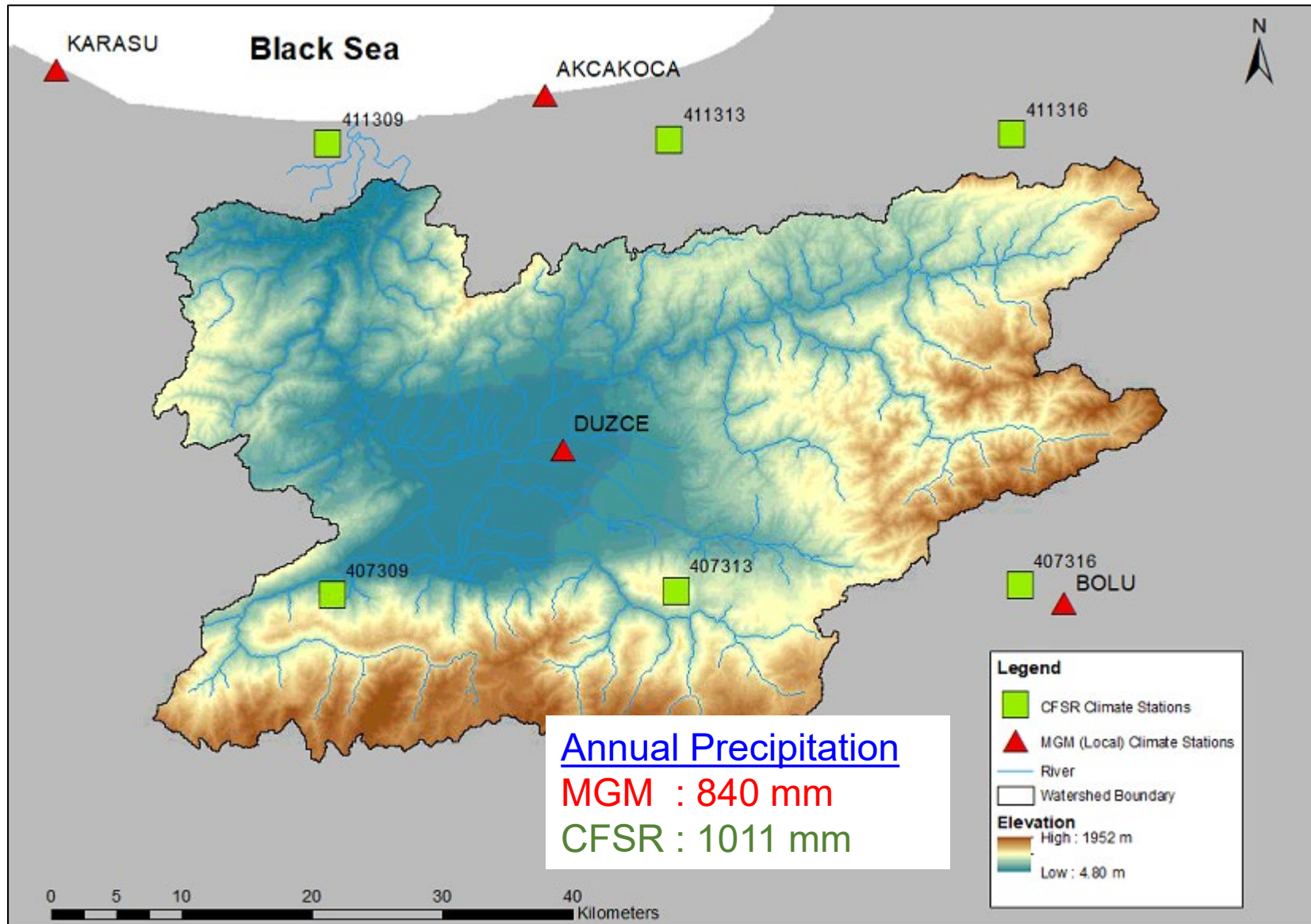
**50 Subbasin**



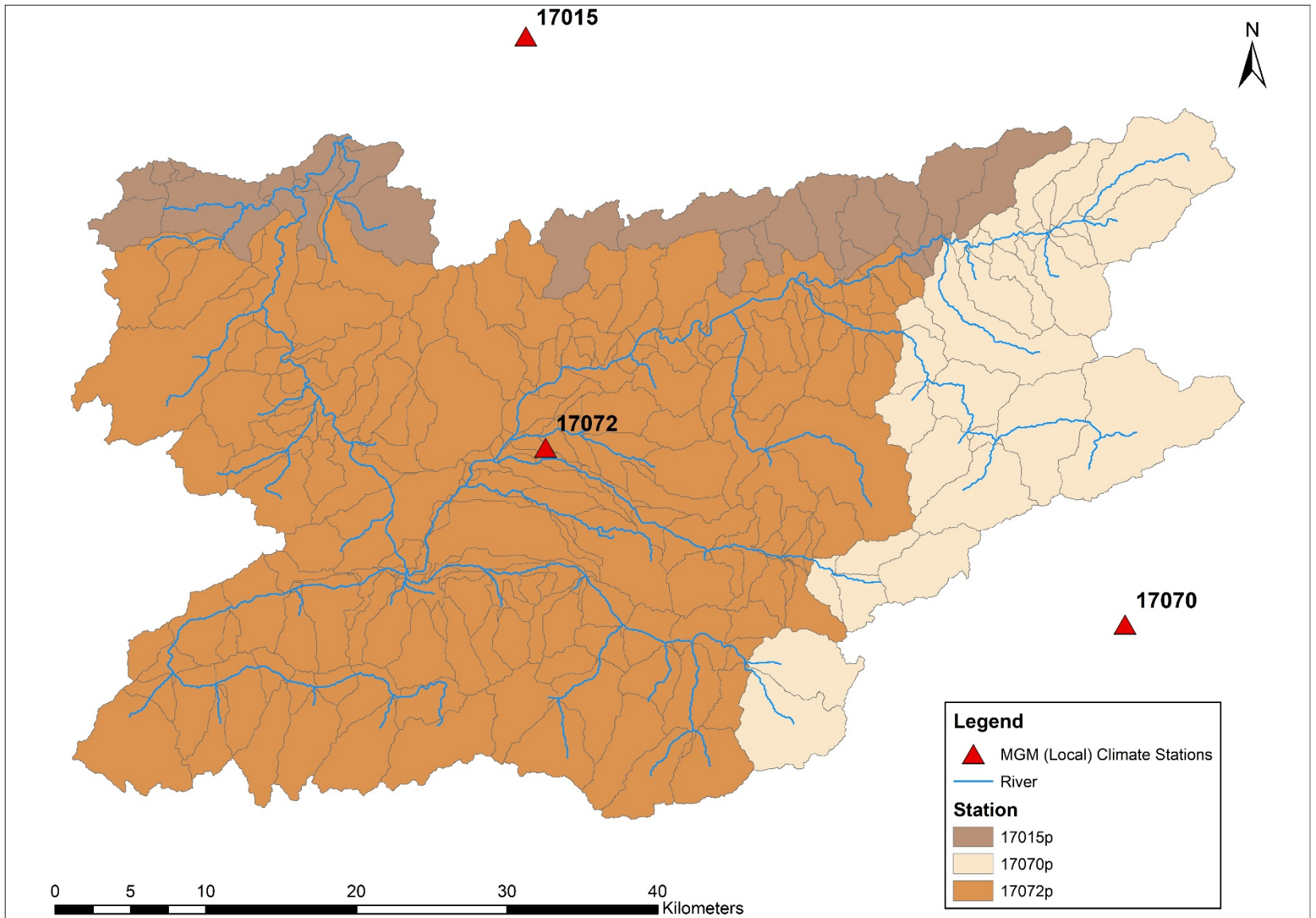
**252 Subbasin**



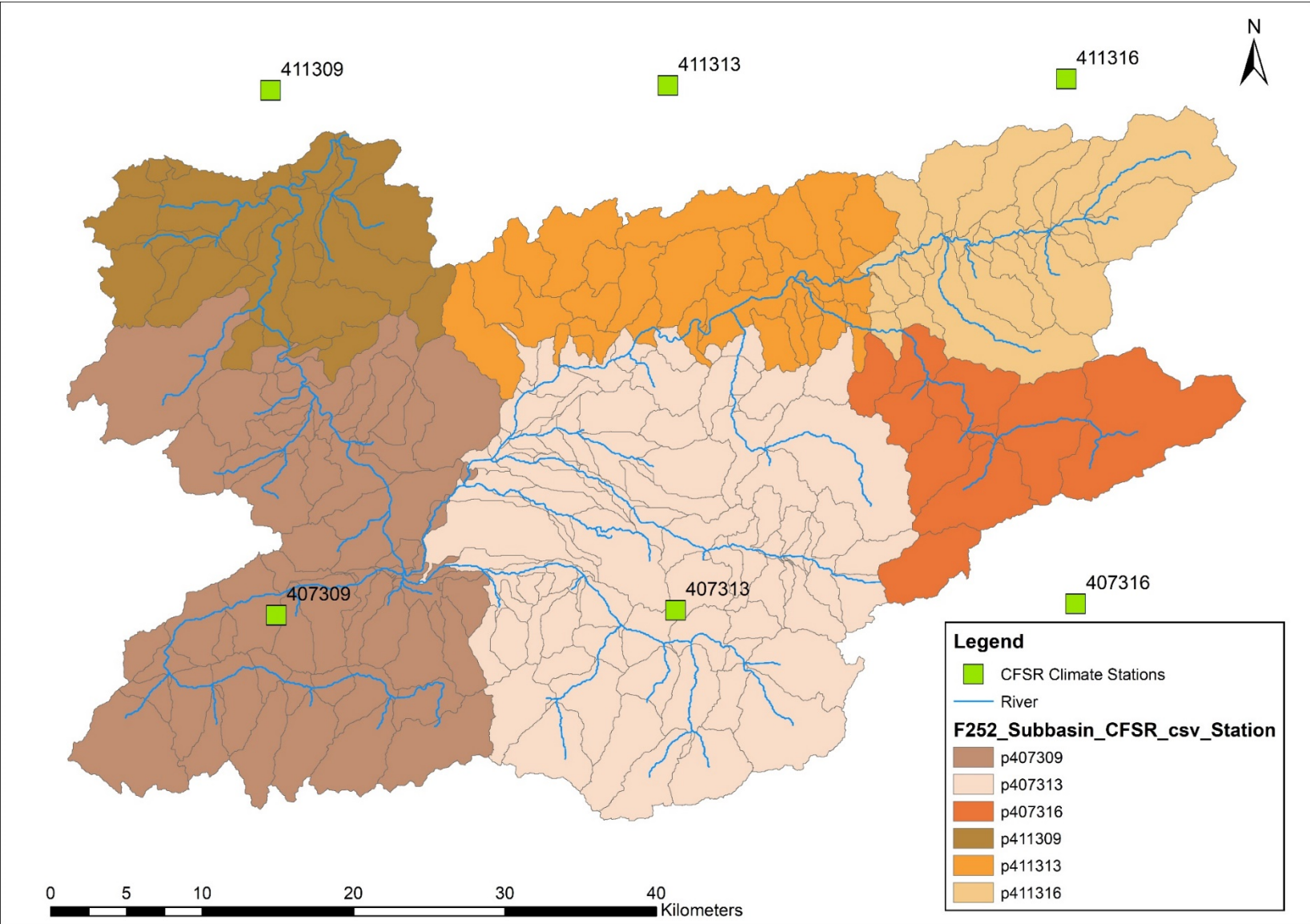
# Climate Data



# Subbasin No vs. Climate Data



# Subbasin No vs. Climate Data

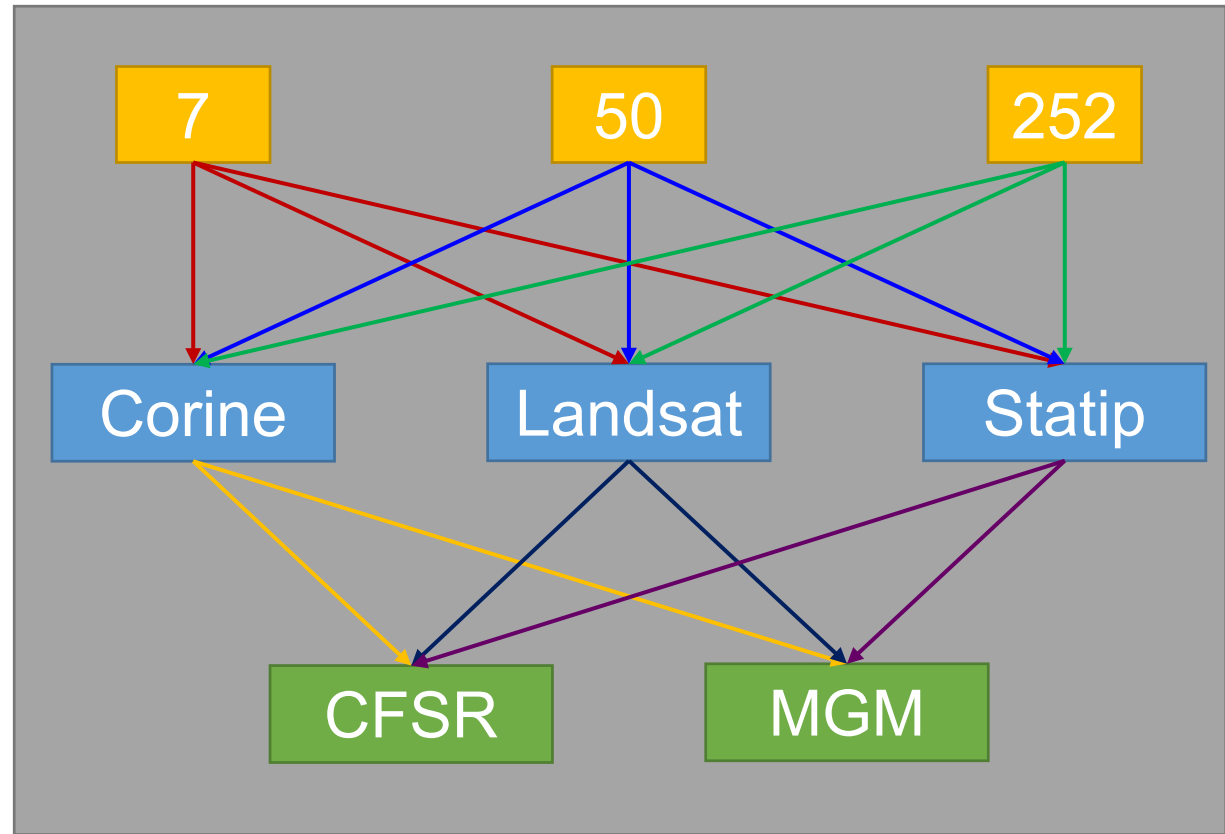


# Model Configuration Setup

Subbasin Number

Landuse Dataset

Climate Dataset



- 18 Different Model Setup

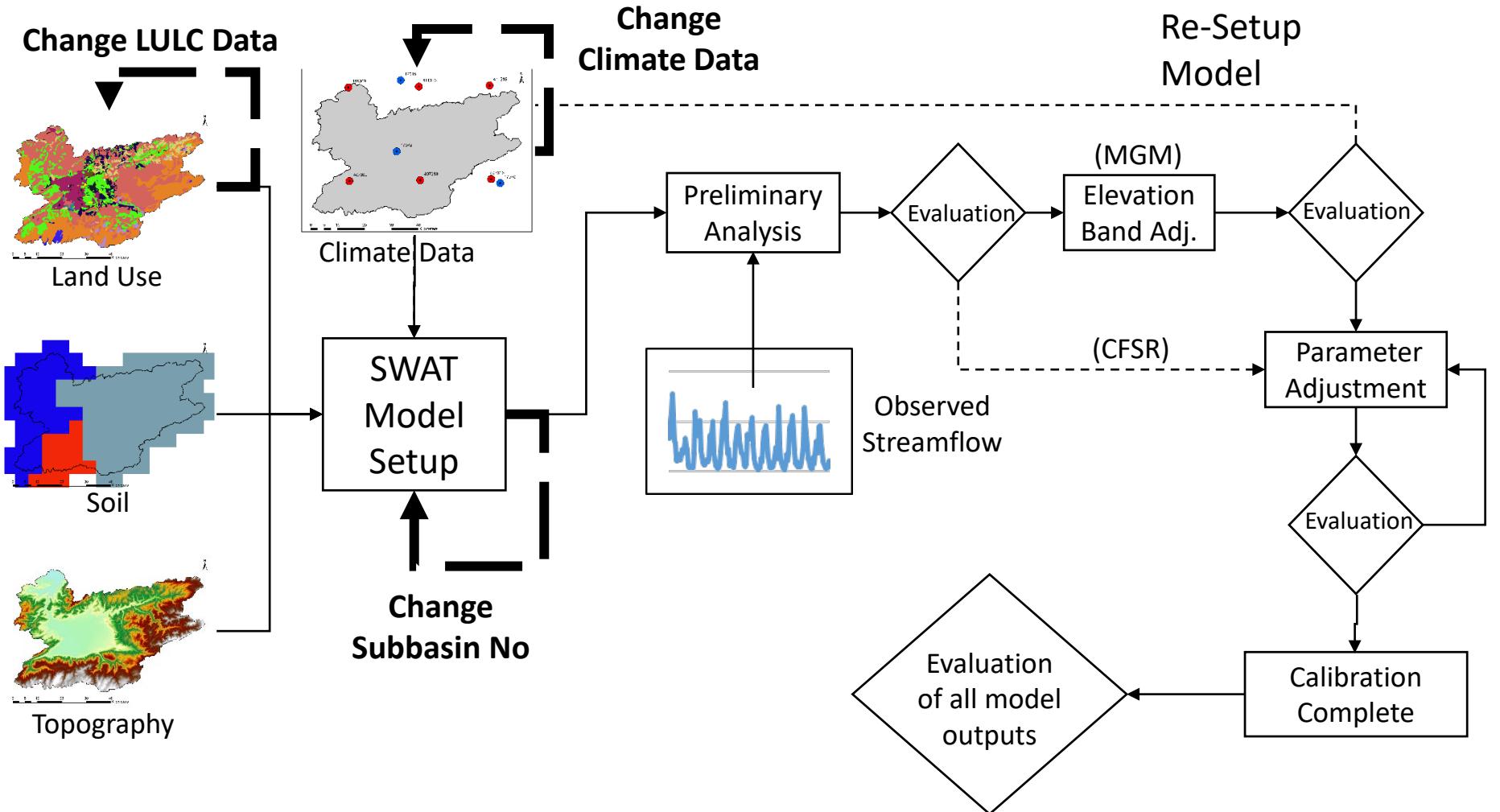
SubbasinNo\_Landuse\_Climate

7\_COR\_CFSR

252\_LAND\_MGM

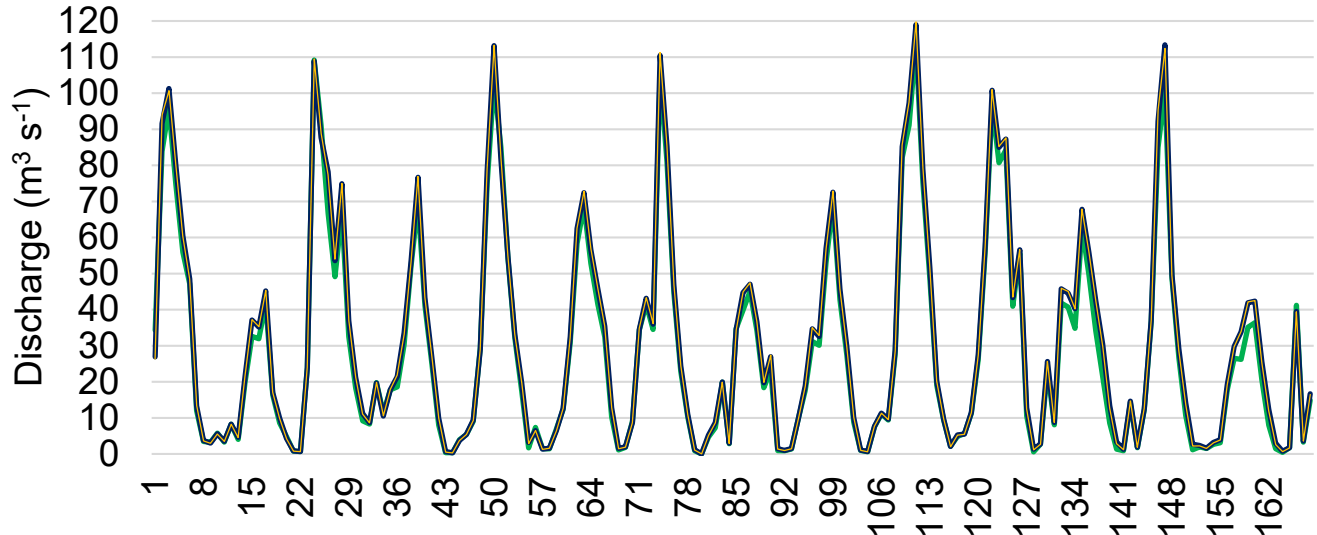
50\_STA\_MGM

# Methodology

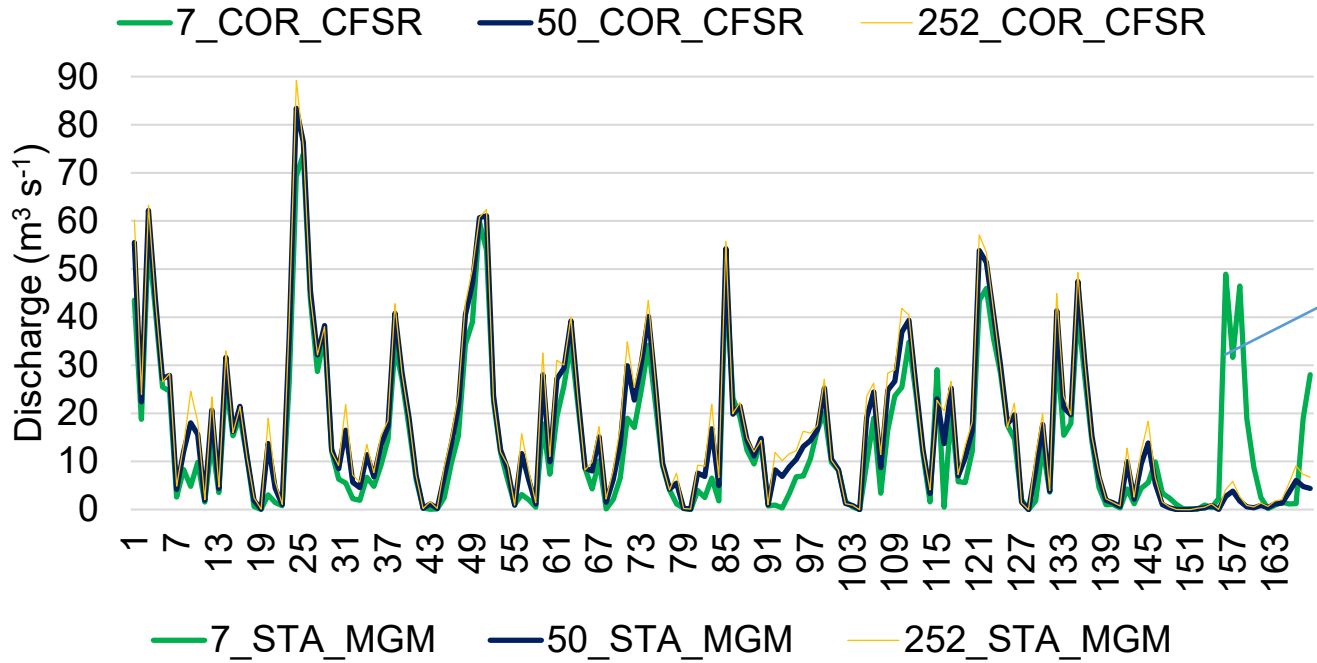




# Comparison of Subbasin No.

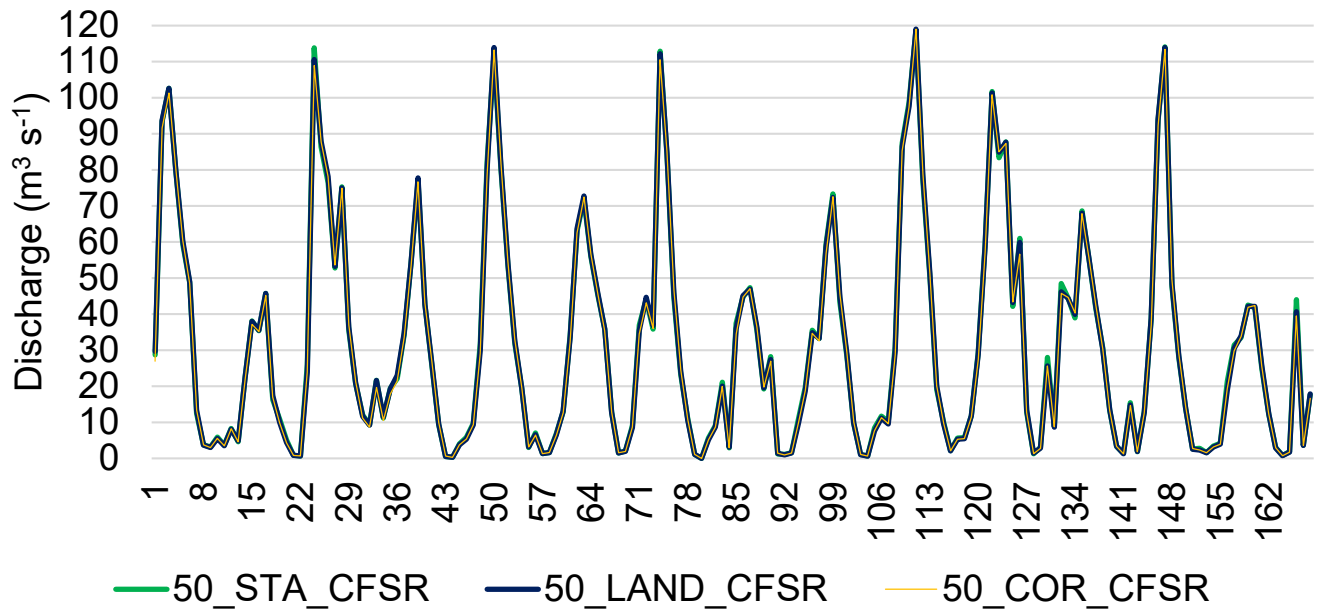


- First Simulations before calibration
- In order to compare effects of **subbasin number**
- No obs. data

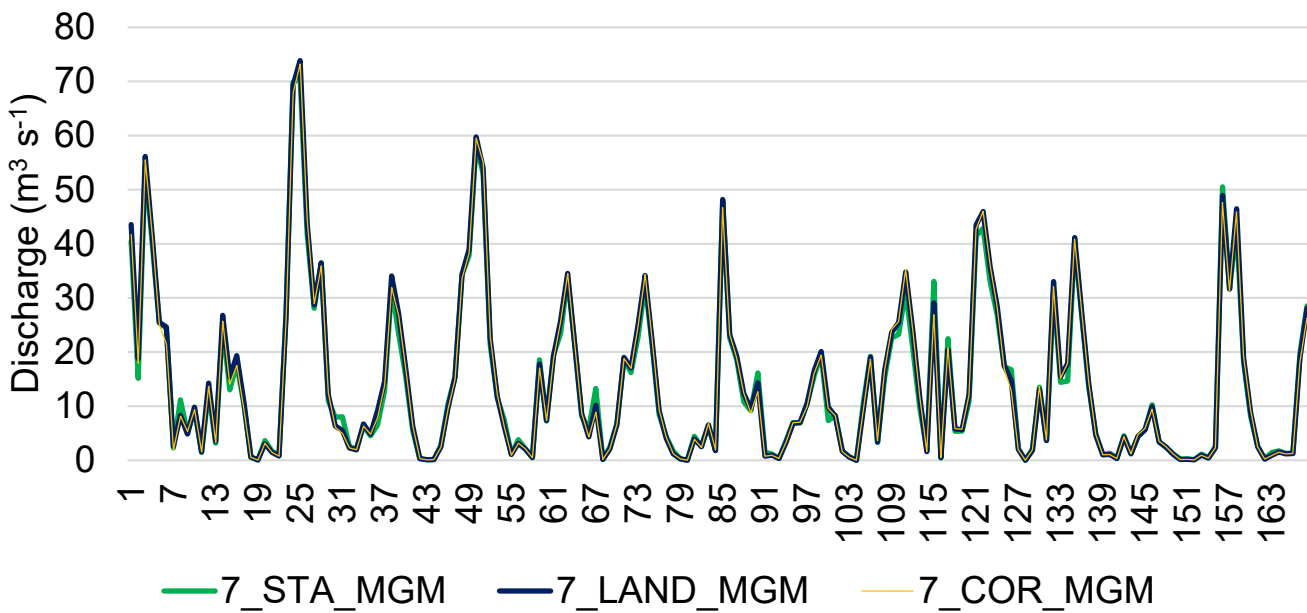


Subbasin Config.

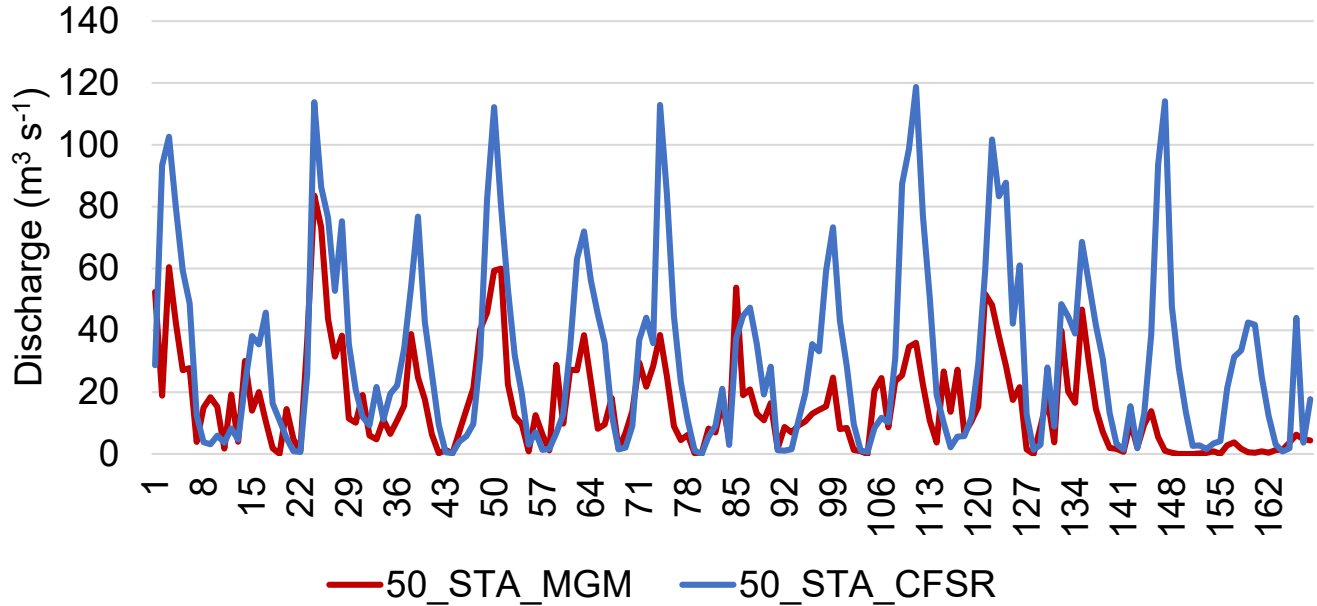
# Comparison of Landuse Data



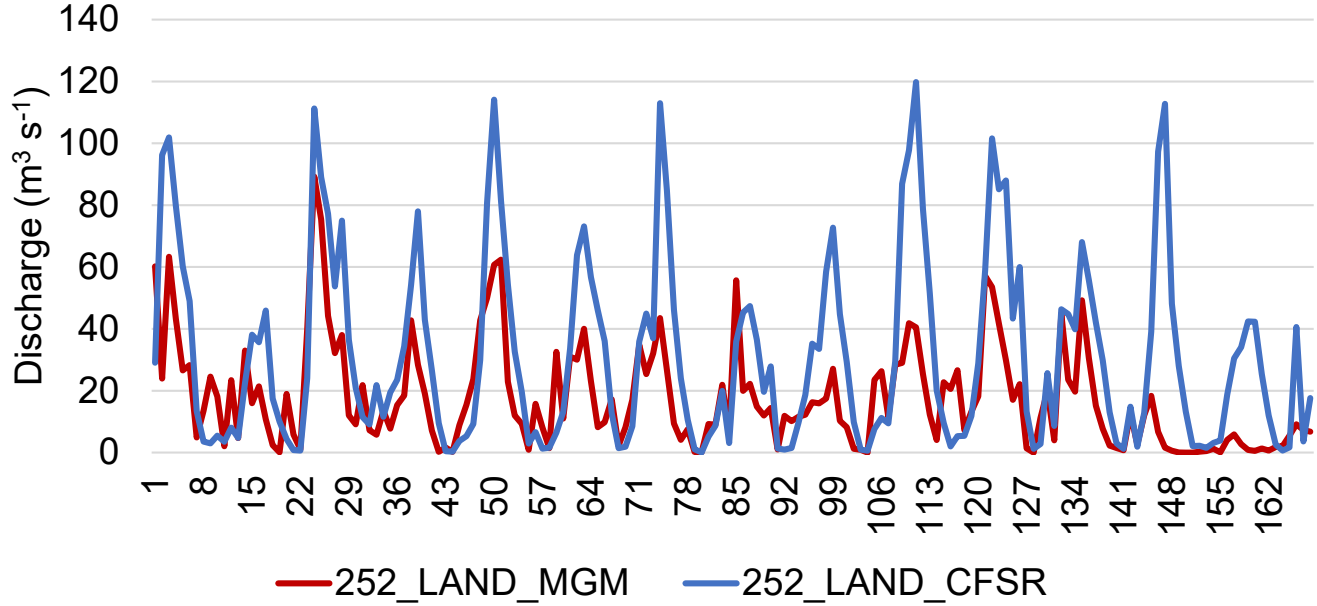
- First Simulations before calibration
- In order to compare effects of **landuse datasets**
- No obs. data



# Comparison of Climate Data



- First Simulations before calibration
- In order to compare effects of climate data
- No obs. data

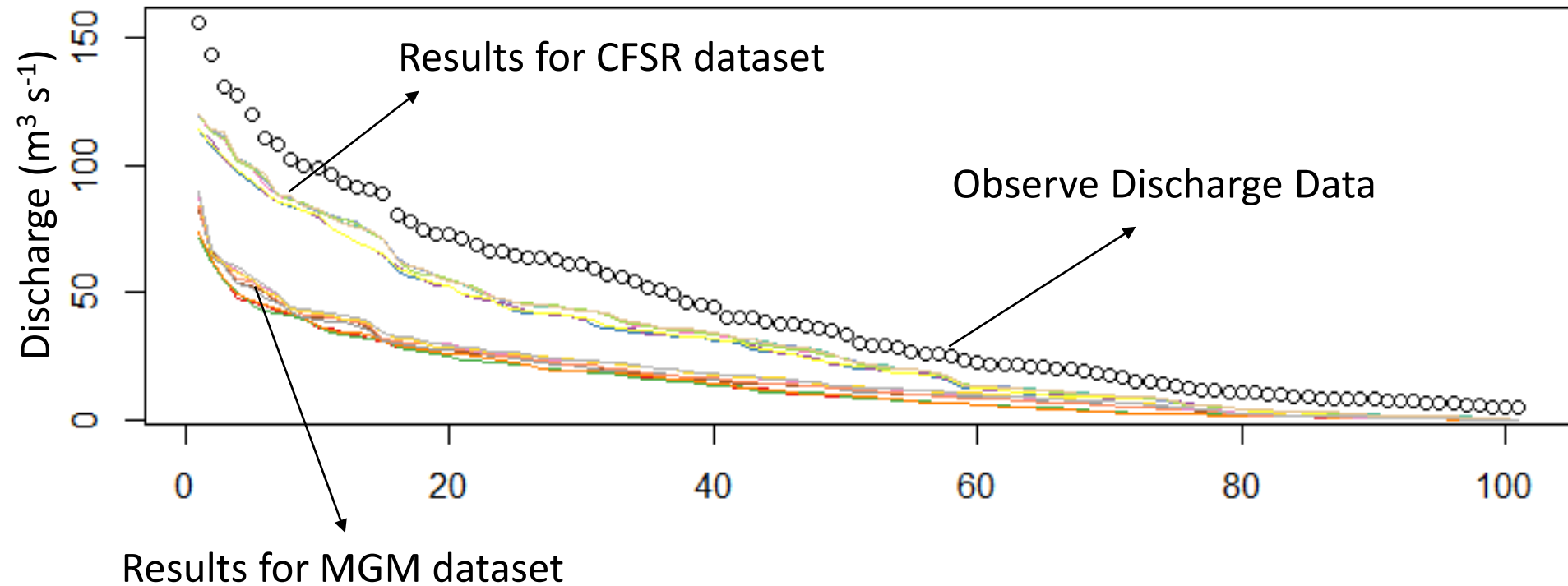


# Performance Criteria for First Sim.

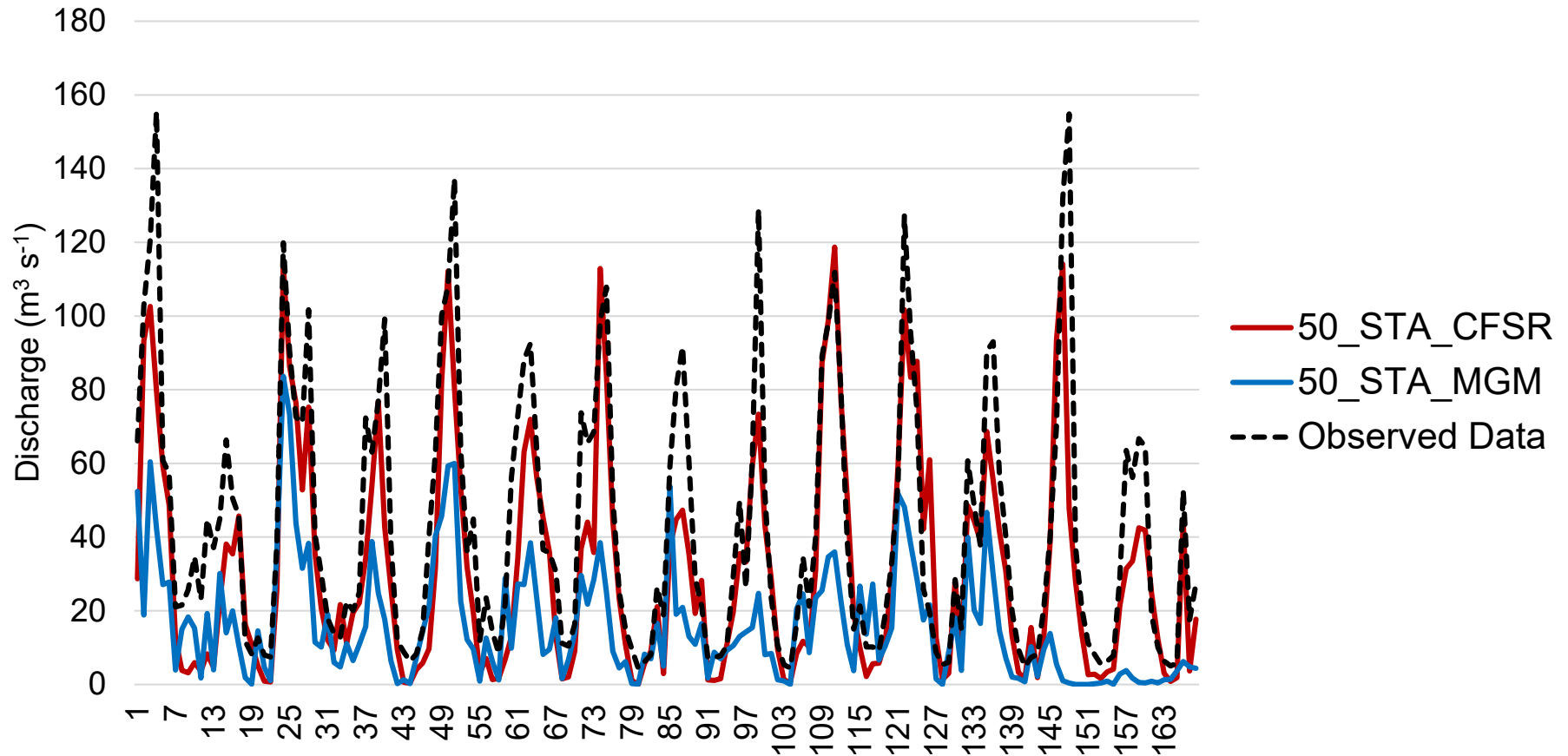
Model Name	R2	bR2	NS	BPIAS
7 COR CFSR	0.79	0.5594	0.63	31.6
7_COR_MGM	0.52	0.1602	-0.22	67.1
7 STA CFSR	0.78	0.5574	0.64	31
7 STA MGM	0.48	0.1404	-0.25	67.4
7 LAND CFSR	0.78	0.5588	0.64	31
7 LAND MGM	0.51	0.1576	-0.21	66.6
50 COR CFSR	0.78	0.5843	0.68	26.9
50 COR MGM	0.48	0.1546	-0.13	62.3
50 STA CFSR	0.78	0.5842	0.68	26.1
50 STA MGM	0.45	0.1375	-0.16	62.5
50 LAND CFSR	0.78	0.5845	0.68	26.3
50 LAND MGM	0.48	0.1533	-0.12	61.7
252 COR CFSR	0.78	0.5853	0.67	26.8
252 COR MGM	0.46	0.1455	-0.09	59.1
252_STA_CFSR	0.78	0.5852	0.68	26
252_STA_MGM	0.43	0.1294	-0.12	59.2
252 LAND CFSR	0.78	0.5857	0.68	26.2
252 LAND MGM	0.45	0.1439	-0.08	58.6

# Flow Duration Curve

Flow Duration curve for the first simulations of all 18 models



# Ex. time series for first simulation



<b>Model Name</b>	<b>R2</b>	<b>bR2</b>	<b>NS</b>	<b>BPIAS</b>
50_STA_CFSR	0.78	0.5842	0.68	26.1
50_STA_MGM	0.45	0.1375	-0.16	62.5

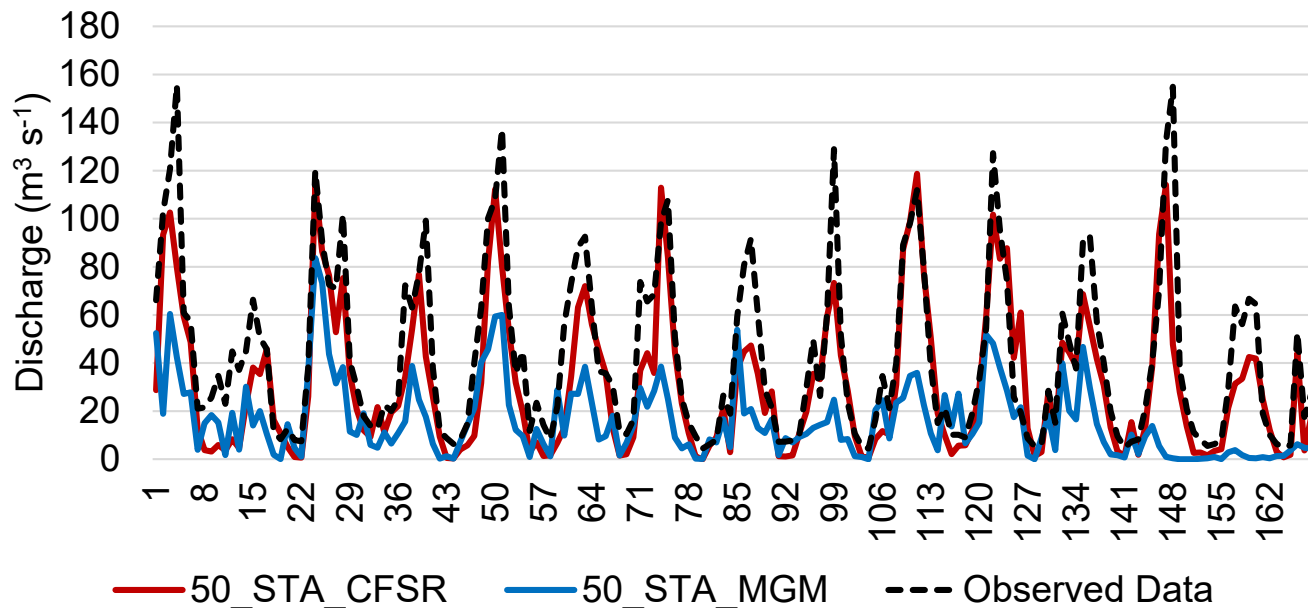
# SWAT Model Parameters

- **SWAT-CUP** used for model calibration/validation

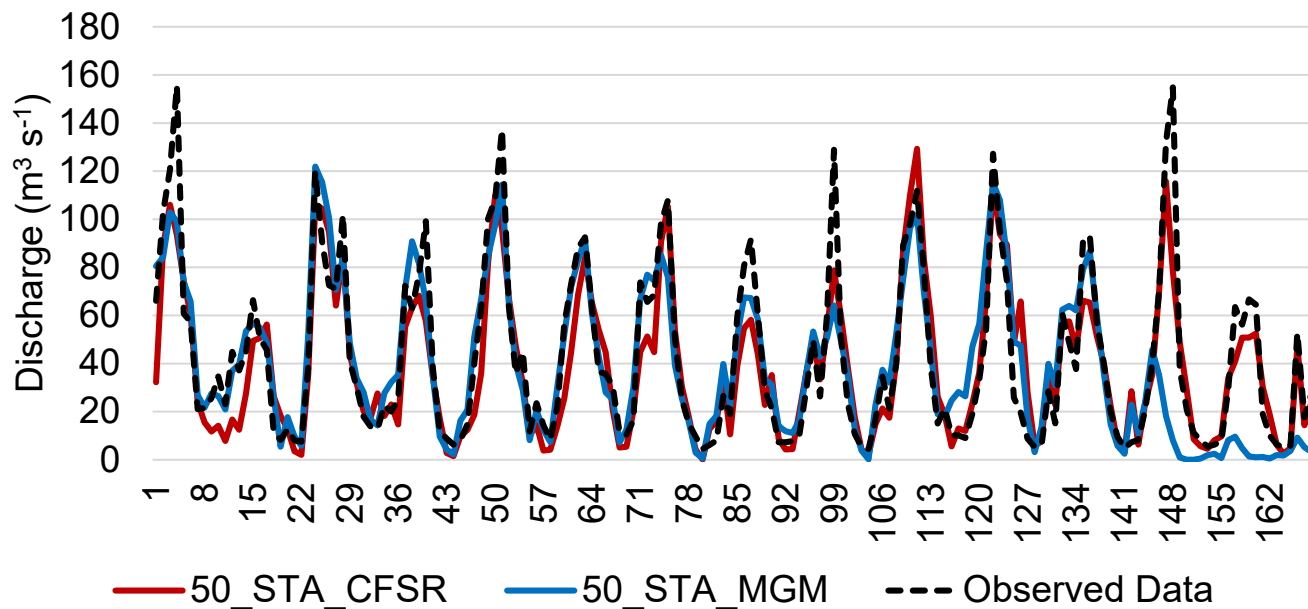
SWAT Parameters	Initial Range
r__CN2.mgt	-0.5 to 0.5
r__GWQMN.gw	-0.5 to 0.5
r__GW_REVAP.gw	-0.5 to 0.5
r__SOL_AWC().sol	-0.5 to 0.5
r__REVAPMN.gw	-0.5 to 0.5
r__ESCO.hru	-0.2 to 0.2
r__ALPHA_BF.gw	-0.5 to 0.5
r__SOL_K().sol	-0.5 to 0.5
r__SOL_BD().sol	-0.5 to 0.5

- **Plapse Rate** is chosen as 600mm/km  
(Cuceloglu and Ozturk, 2017)

480 simulations has been done for each model



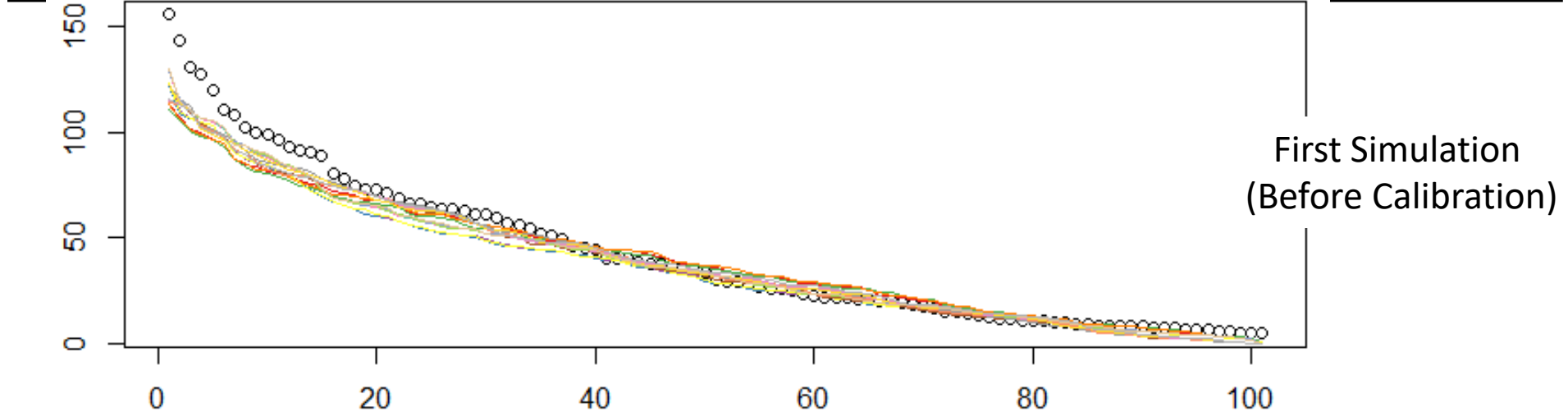
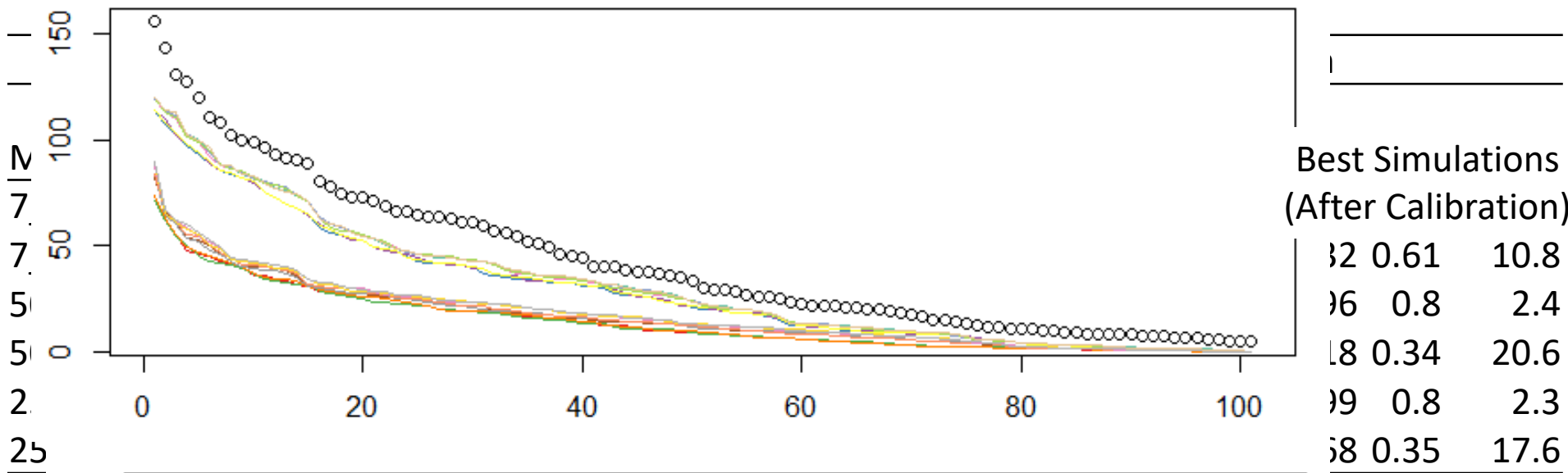
First Simulation  
(Before Calibration)



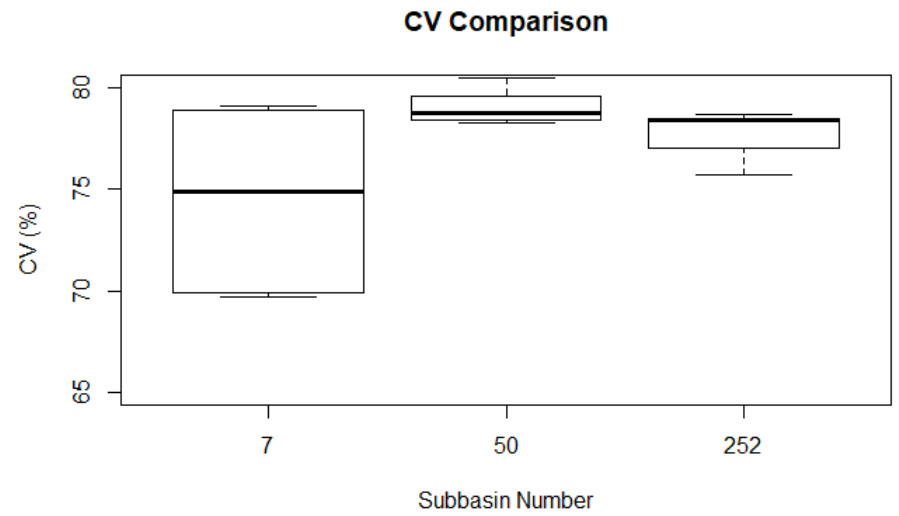
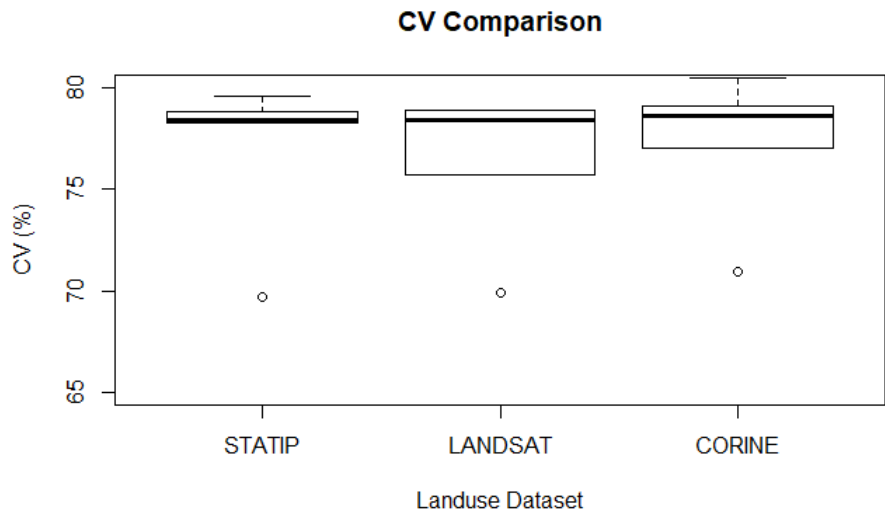
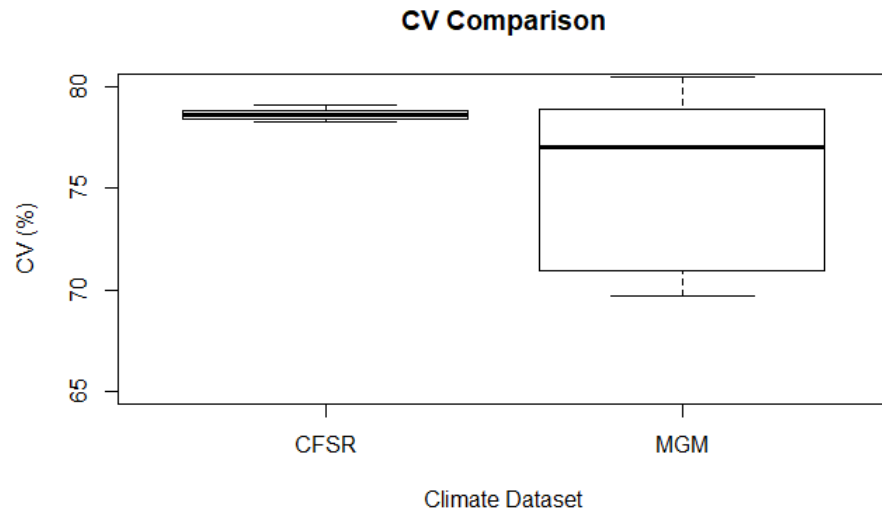
Best Simulations  
(After Calibration)



# Model Performance/Flow Duration Curve



# CV Value for Streamflow



# Results Discussion

- The study shows the climate data set is the most influential dataset and has a larger uncertainty in our study area.
- Global climate dataset (reanalysis data) represents better than the local data.
- The Corine LULC dataset (freely available) yield quite similar results as the local landuse datasets. Detailed local dataset (STATIP) has the smallest uncertainty.
- Subbasin number has no significant effect on the monthly calibrated discharge data but it is important to evaluate relation between the subbasin and climate data.
- Using elevation band is quite efficient (if the available dataset catches the dynamic of the system)

# Conclusion

- Effects of different input data and subbasin configuration are investigated in Melen Watershed, Turkey
- SWAT model has been successfully applied to the region.
- Hydrological simulations could be significantly different for different model setup configurations and could cause large uncertainty.
- Multisite and daily calibration are also important to investigate
- Nowadays it is easy (relatively) to find multiple datasets therefore modeler should take different these datasets into consideration. It is strongly recommend to use different input datasets for modelling studies.

Thanks for your attention...