

Role of Climate Variability in Modulating Surface water and Groundwater Interaction over the Southeast US

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and

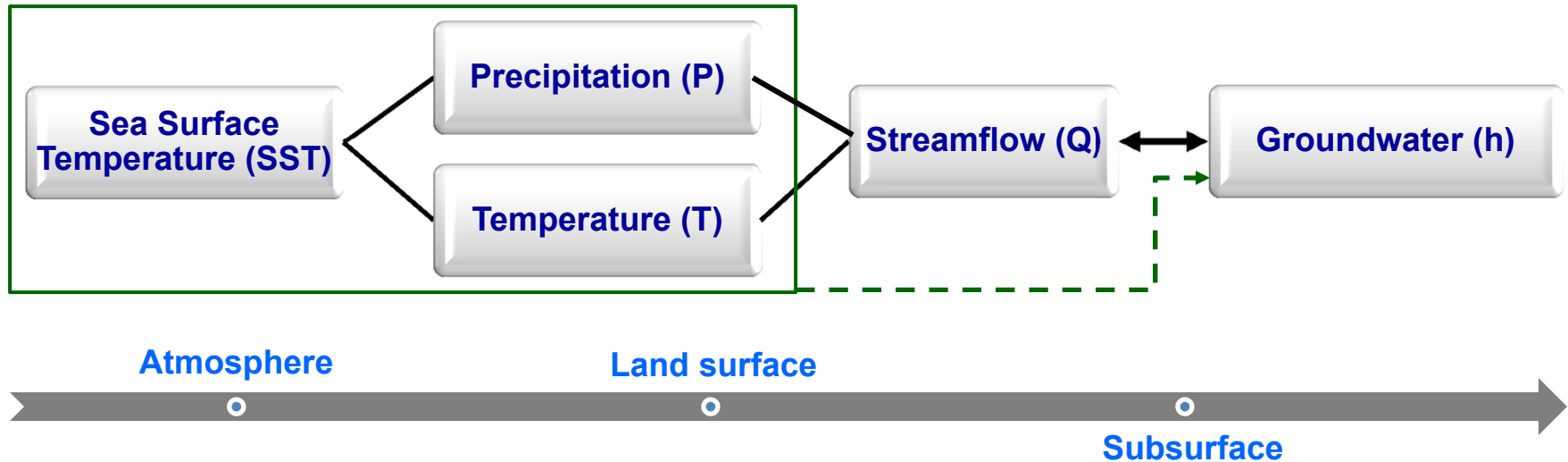
Jerad Bales

United States Geologic Survey, Reston

NOAA in the Carolinas, March 15, 2012, Charleston, SC



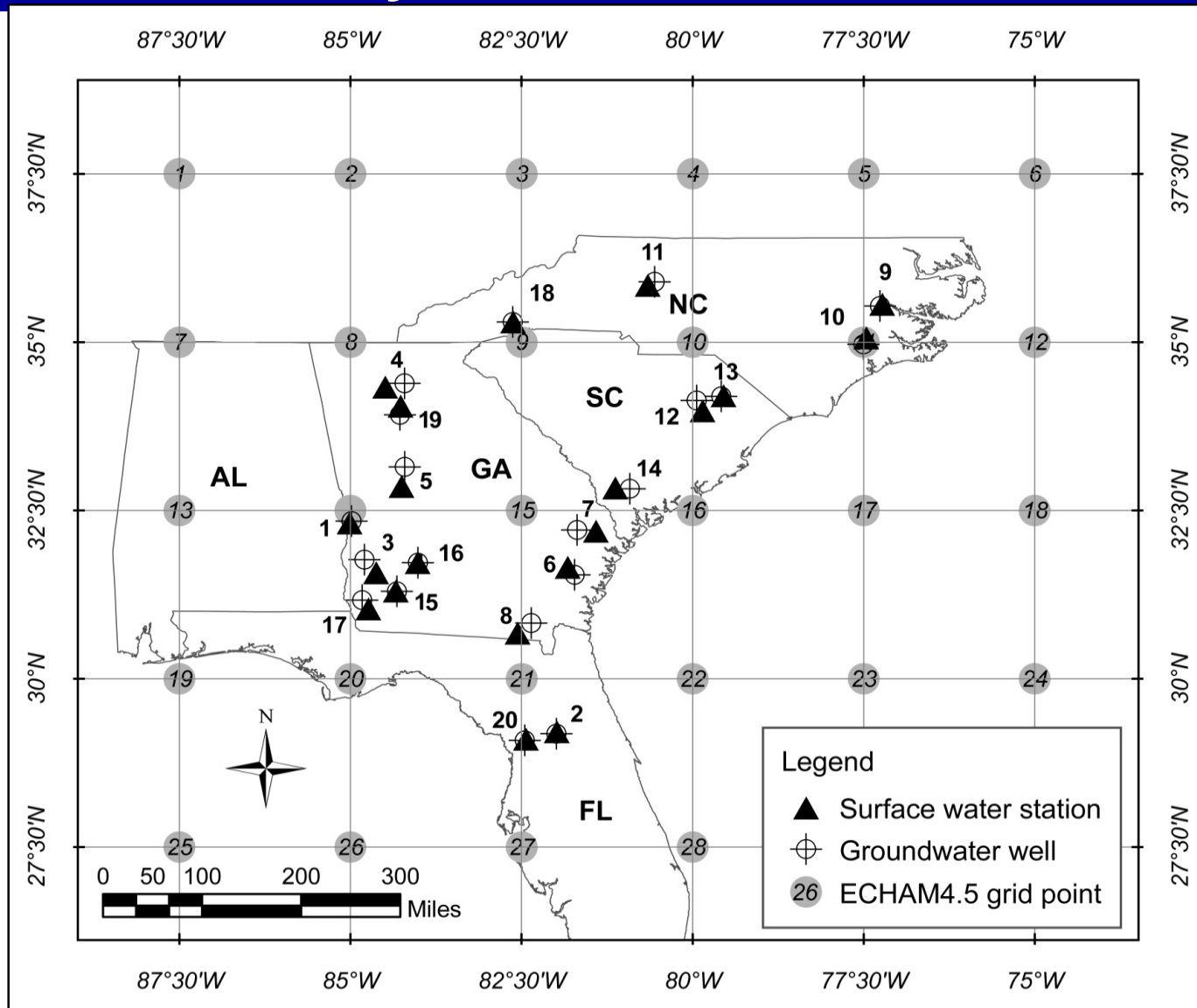
Introduction: Motivation



❑ Motivation:

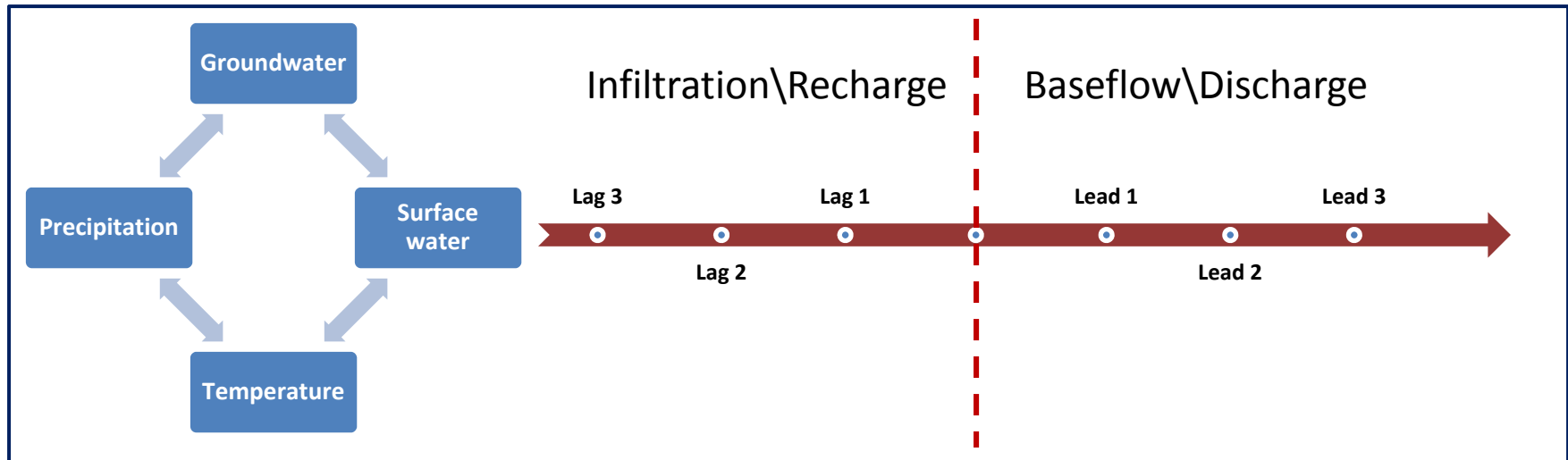
- 1) **Climate variability influences precipitation, temperature and streamflow**
- 2) **Goal is to understand how climate influences basin hydroclimatology including groundwater availability.**

Study Area and Database



Basins: HCDN (Streamflow); Groundwater (Climate and Groundwater response network - USGS)

Dependency Analysis: Recharge and Discharge Months



18 variables:

- Q, h, P and T at various months

2 Seasons

- JFM, and JAS

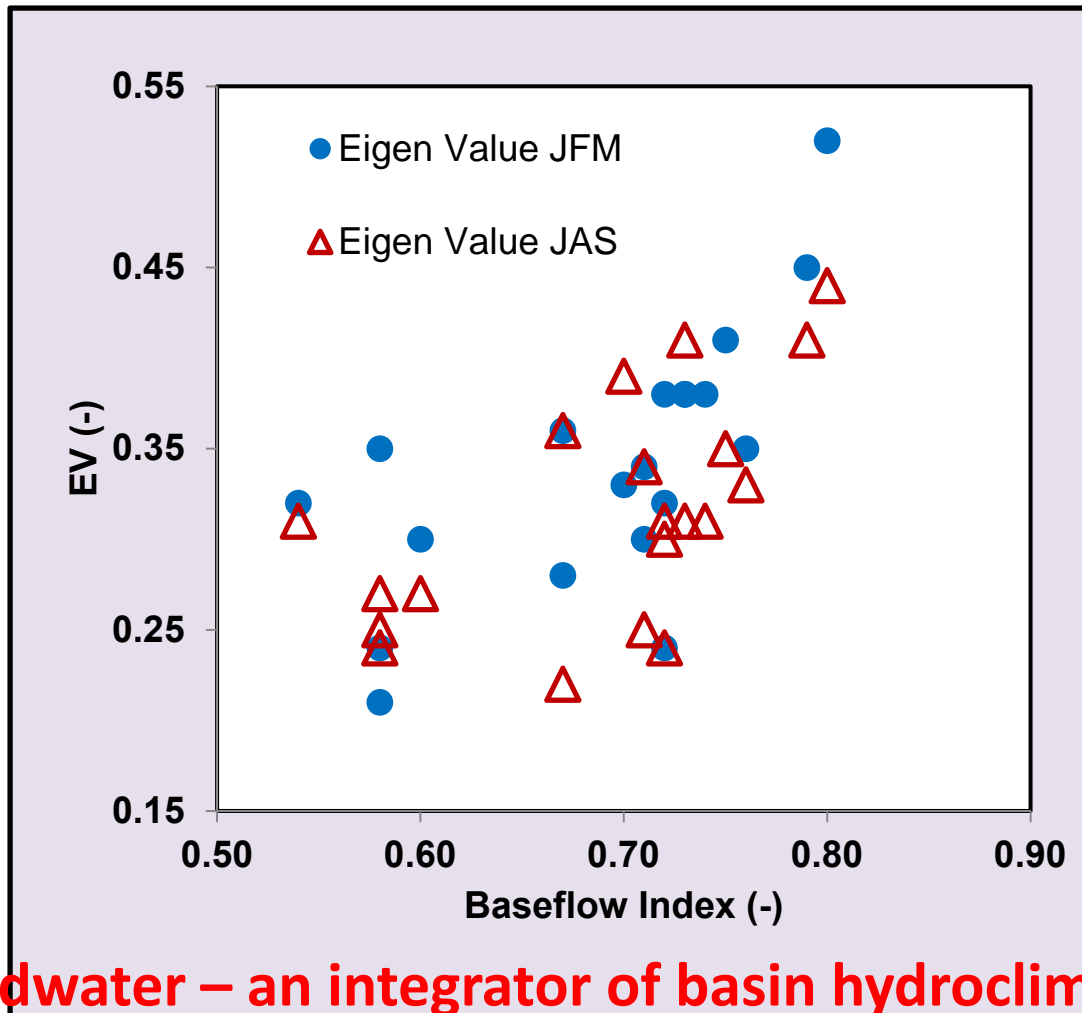
Eigen Values

- Variance Proportion

Eigen Vectors

- Source of variability

Principal Component Analysis - Eigen Values



Groundwater – an integrator of basin hydroclimatology

Rank Correlation:

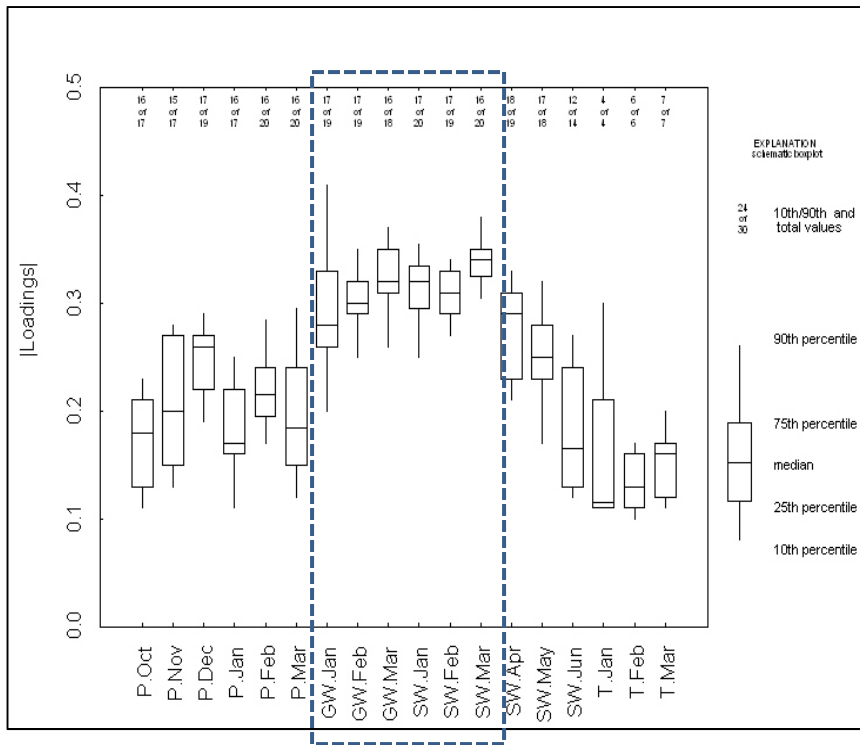
$$\rho (EV_{JFM}, BFI_{JFM}) = 0.72$$

$$\rho (EV_{JAS}, BFI_{JAS}) = 0.66$$

Principal Component Analysis - Eigen Vectors

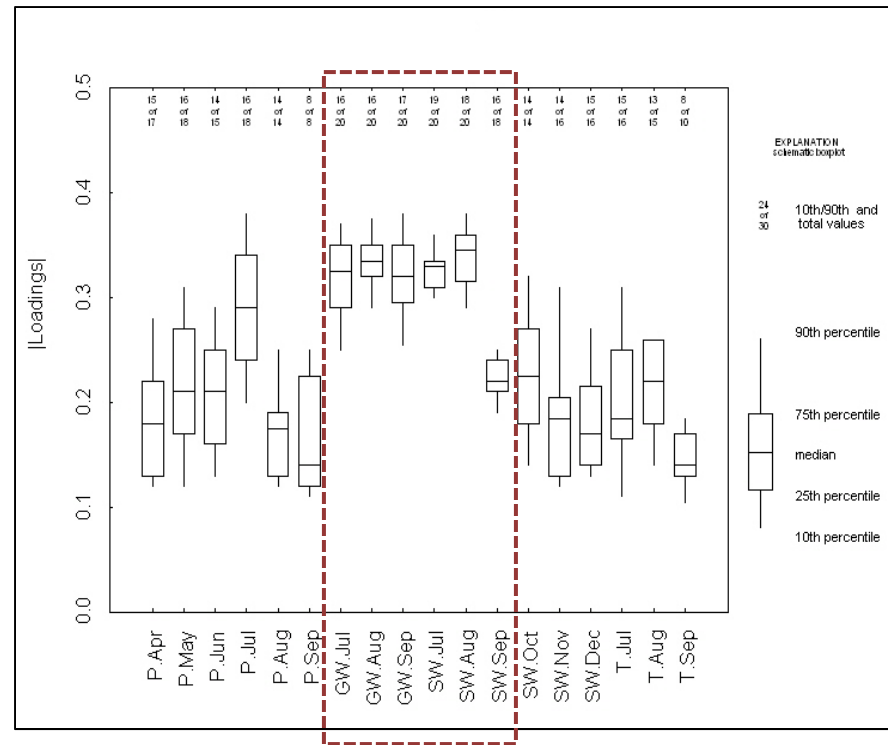
Primary Source of Variability – Surface water and Groundwater

JFM Loadings



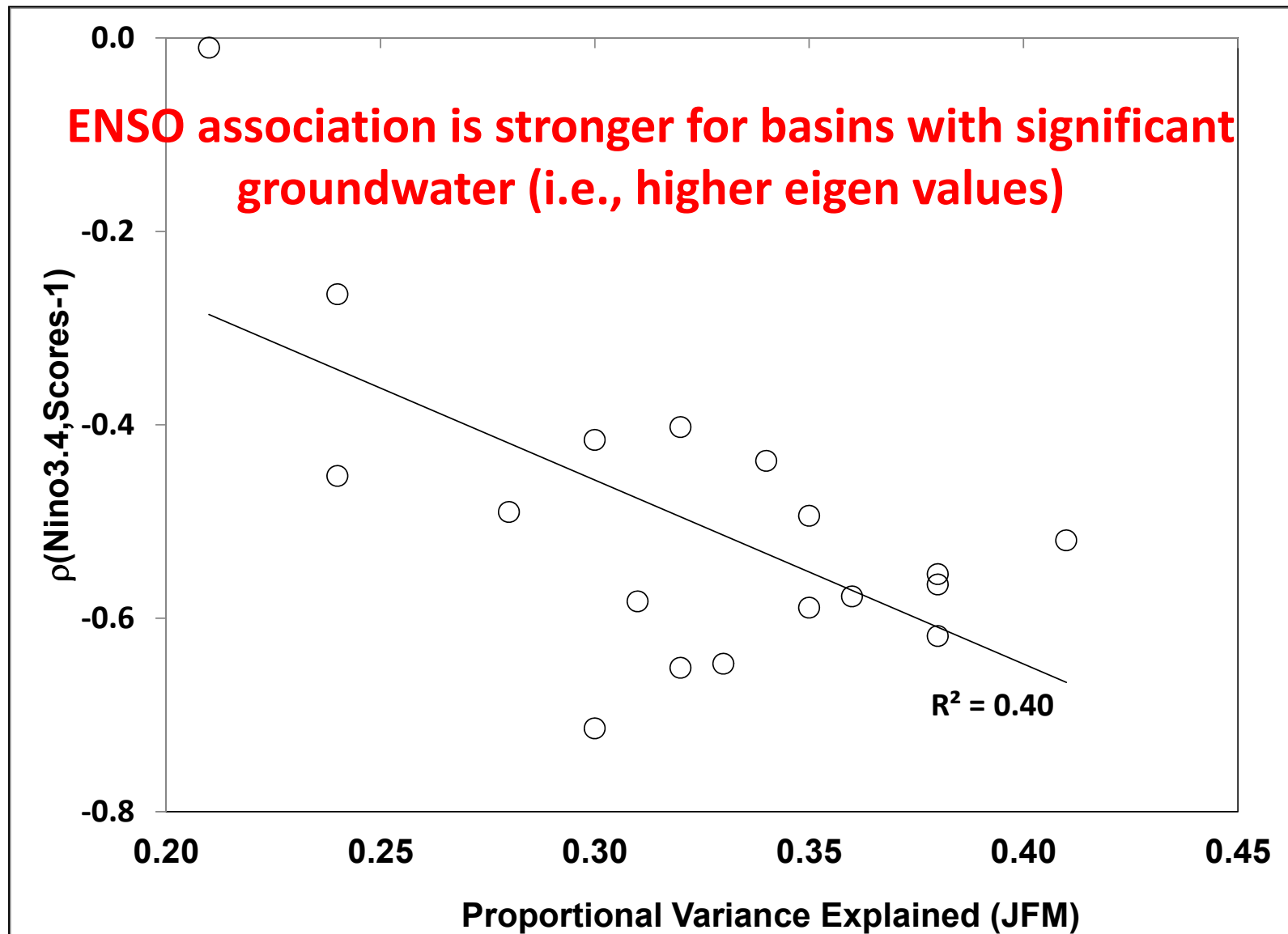
P	6	O	N	D	J	F	M			
T	3				J	F	M			
Q	6				J	F	M	A	M	J
h	3				J	F	M			

JAS Loadings



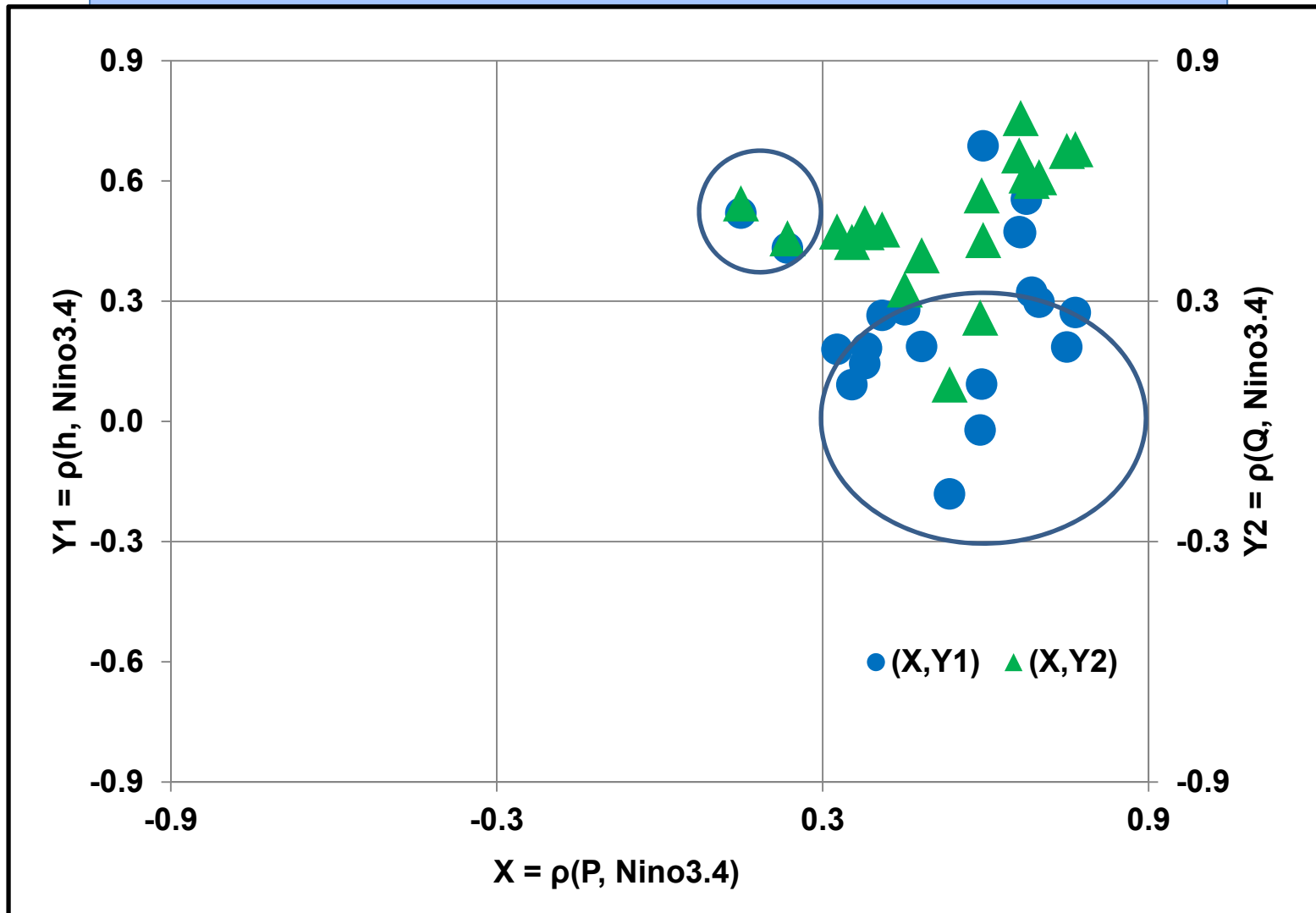
P	6	A	M	J	J	A	S			
T	3				J	A	S			
Q	6				J	A	S	O	N	D
h	3				J	A	S			

PCA – Eigen values vs Nino3.4 (ENSO)

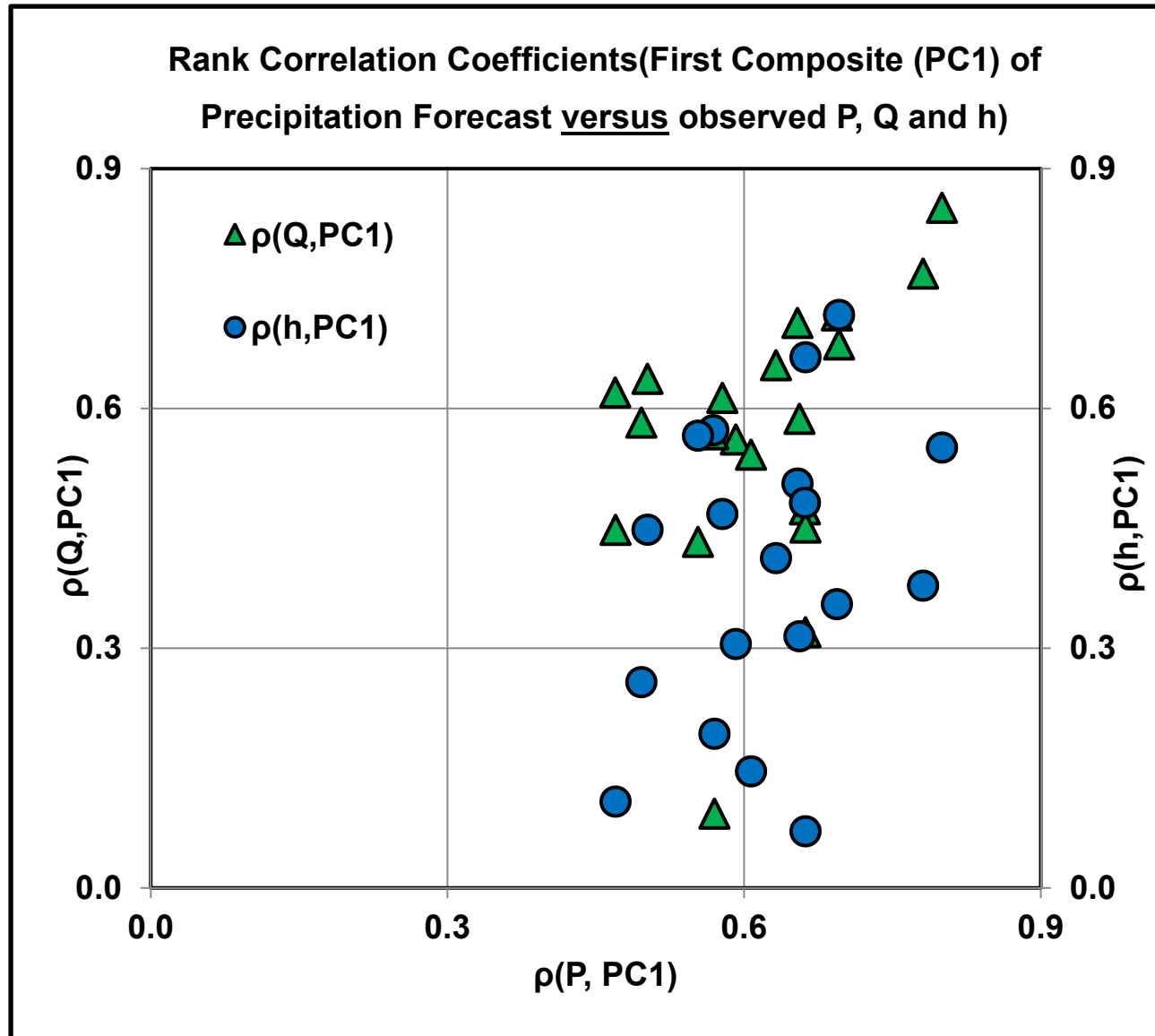


Role of ENSO Winter (JFM)

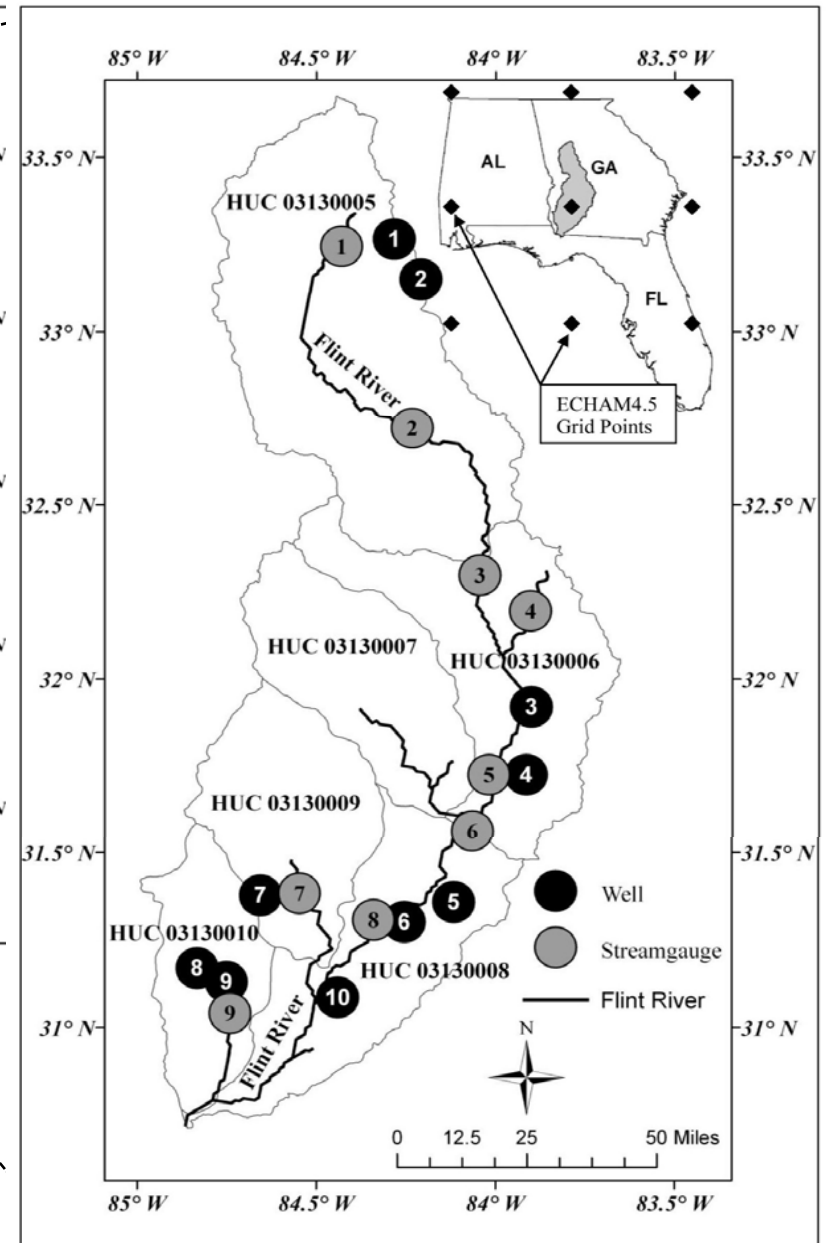
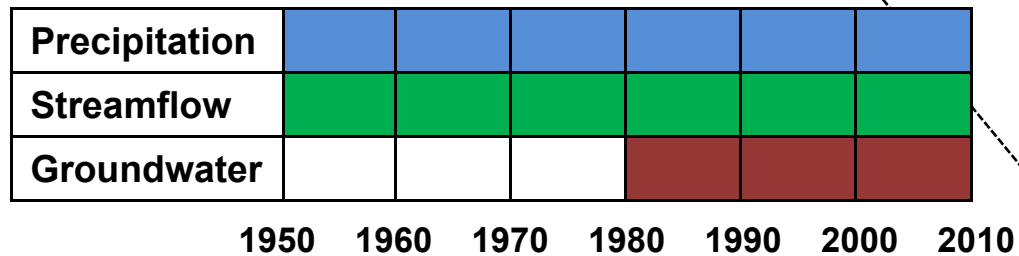
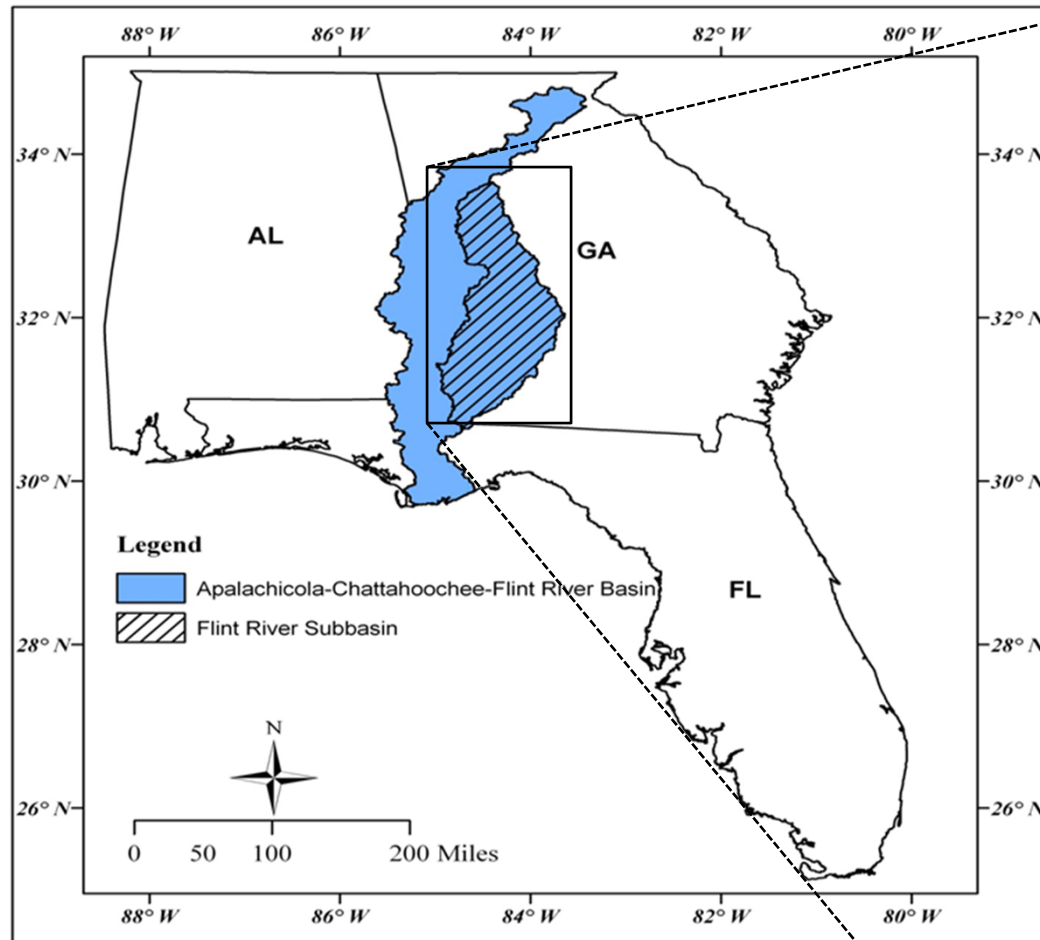
Rank correlation: JFM Nino3.4 versus JFM P, Q and h



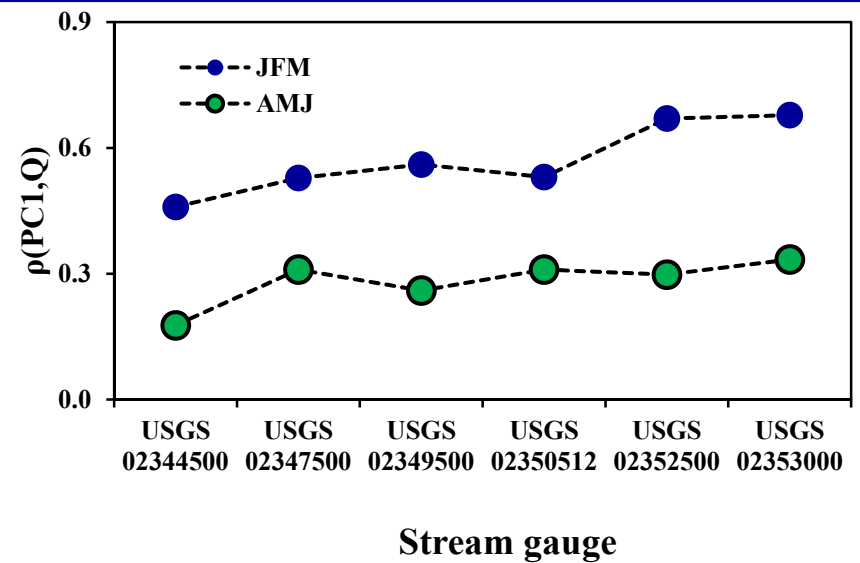
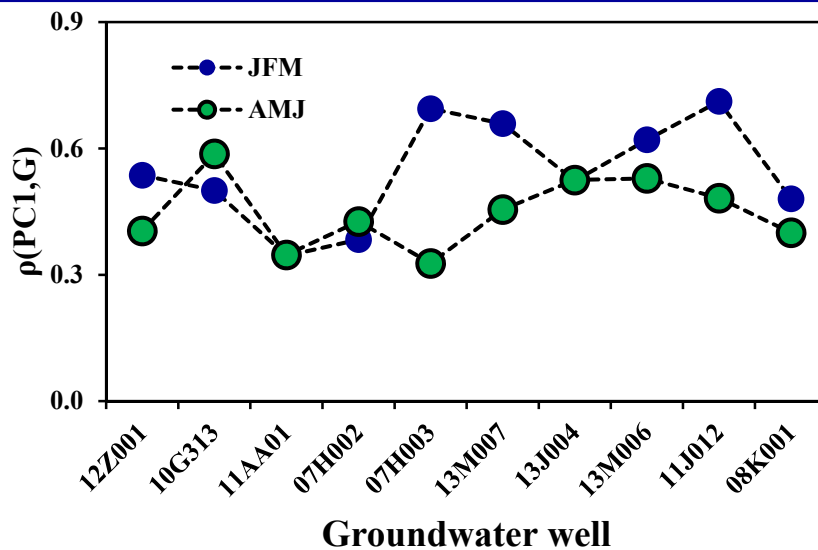
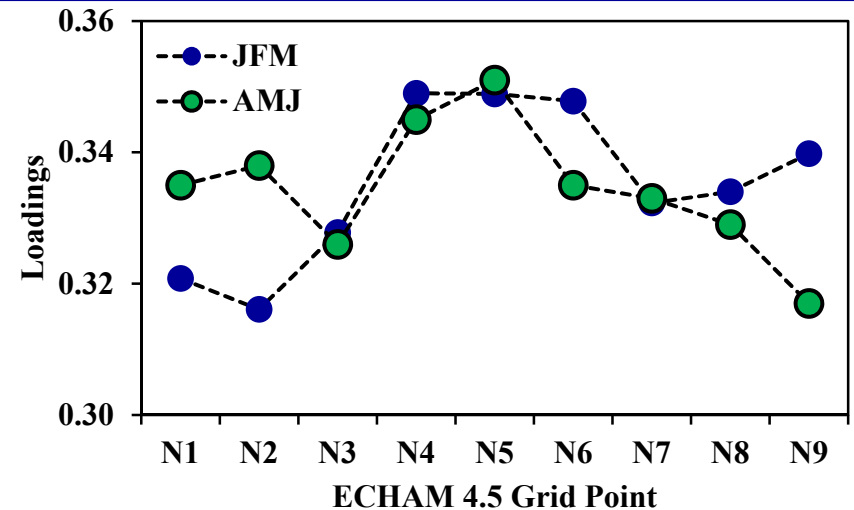
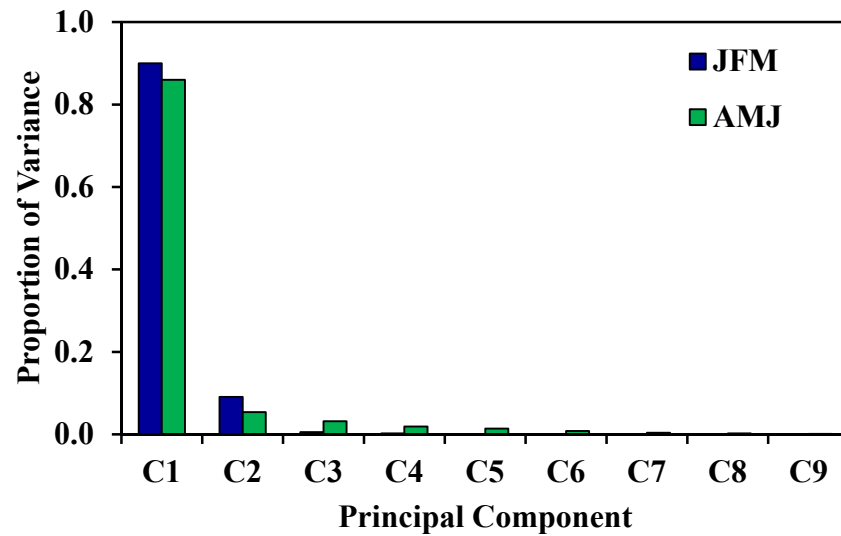
Precipitation Forecasts versus observed P, Q and h



Study Area and Data Description



Correlation between Precipitation Forecasts and Q, P and h

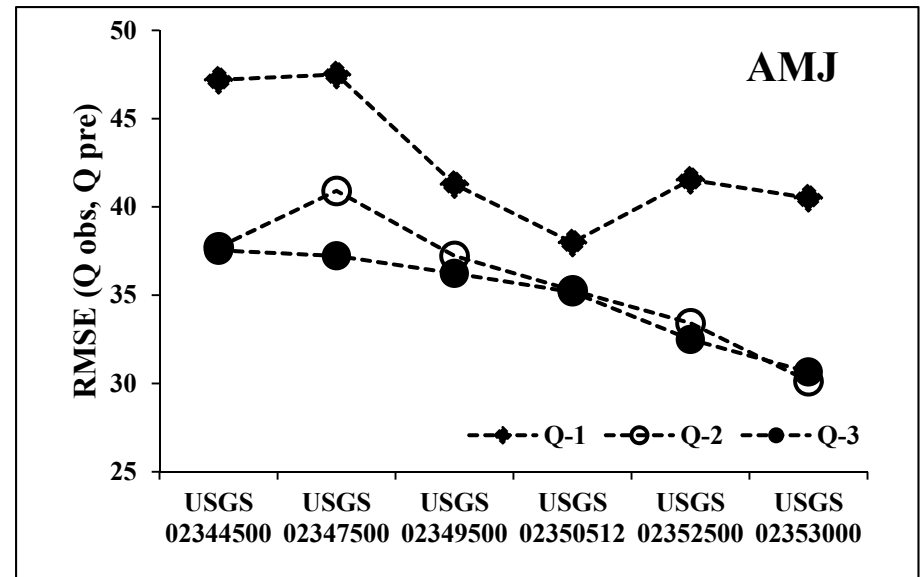
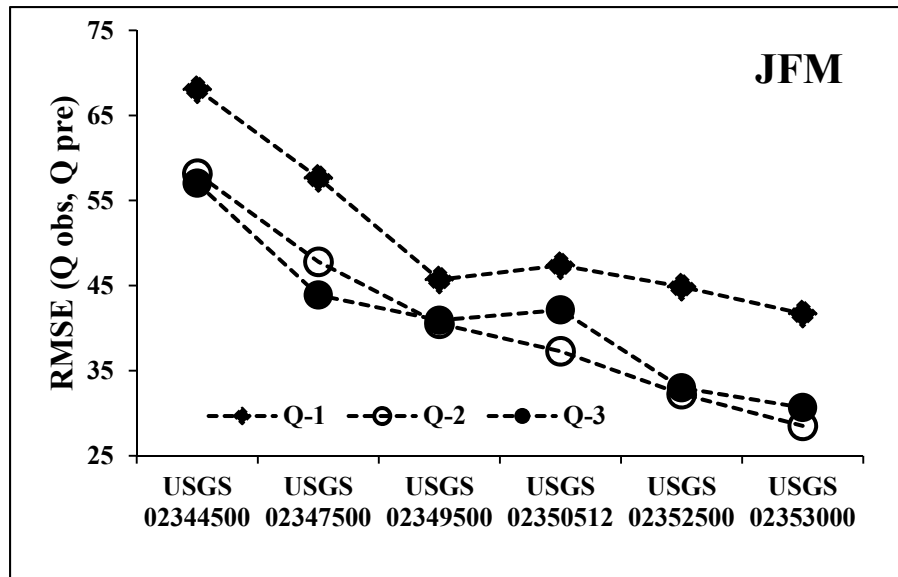
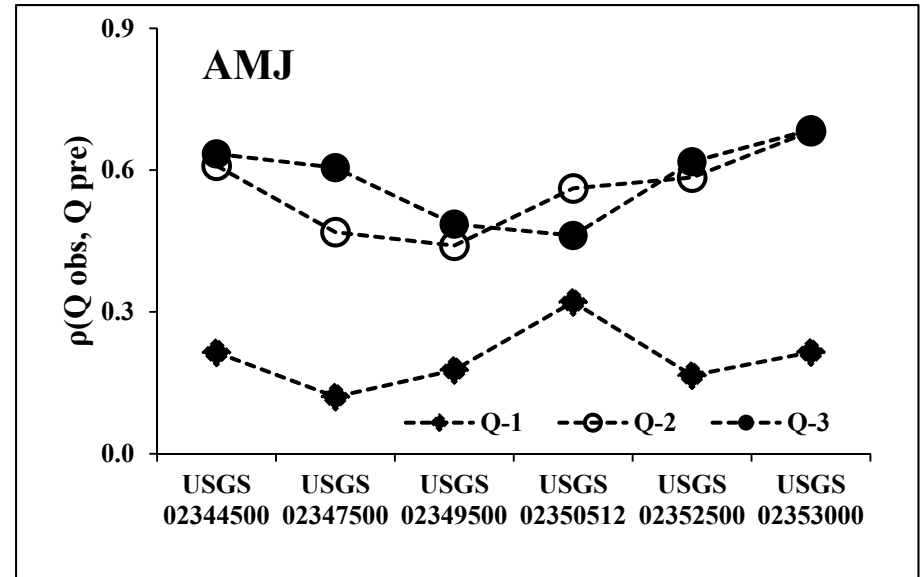
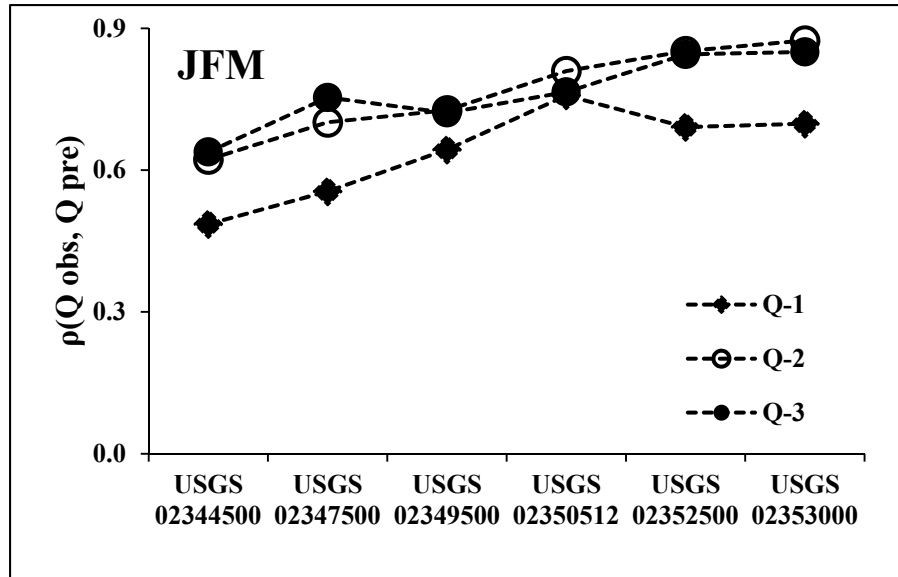


Statistical Models: Predictors and Predictands

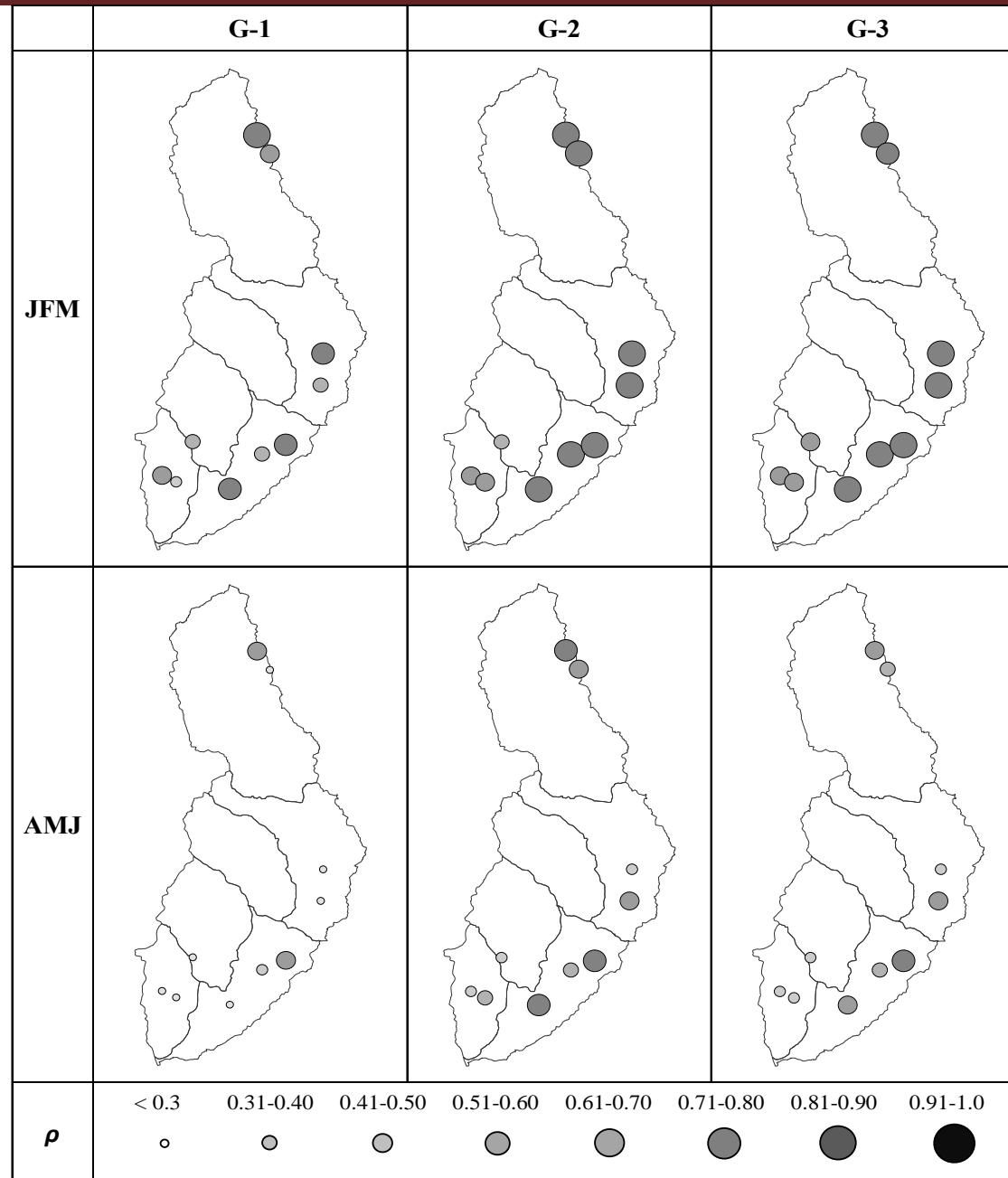
- **Principal Component regression (PCR)**
- **Canonical Correlation Analysis (CCA)**
- **Leave-5 Out Cross-Validation 1980 – 2010**

Variable	Model	Data Period	Method	Predictors			Predictand	
				G	Q	P forecasts	G	Q
G	G-1	1980 - 2010	PCA	OND	-	-	JFM AMJ Jan - Jun	-
	G-2			OND	-	JFM AMJ		
	G-3		CCA	OND	-			
Q	Q-1	1980 - 2010	PCA	-	OND	-	-	JFM AMJ Jan - Jun
	Q-2			-	OND	JFM AMJ		
	Q-3		CCA	-	OND			

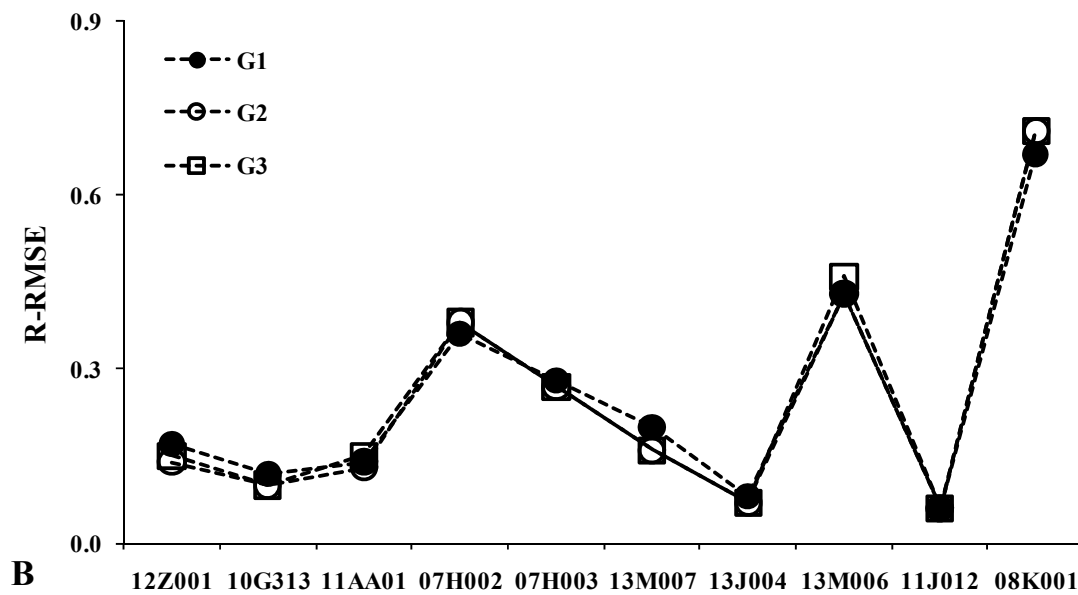
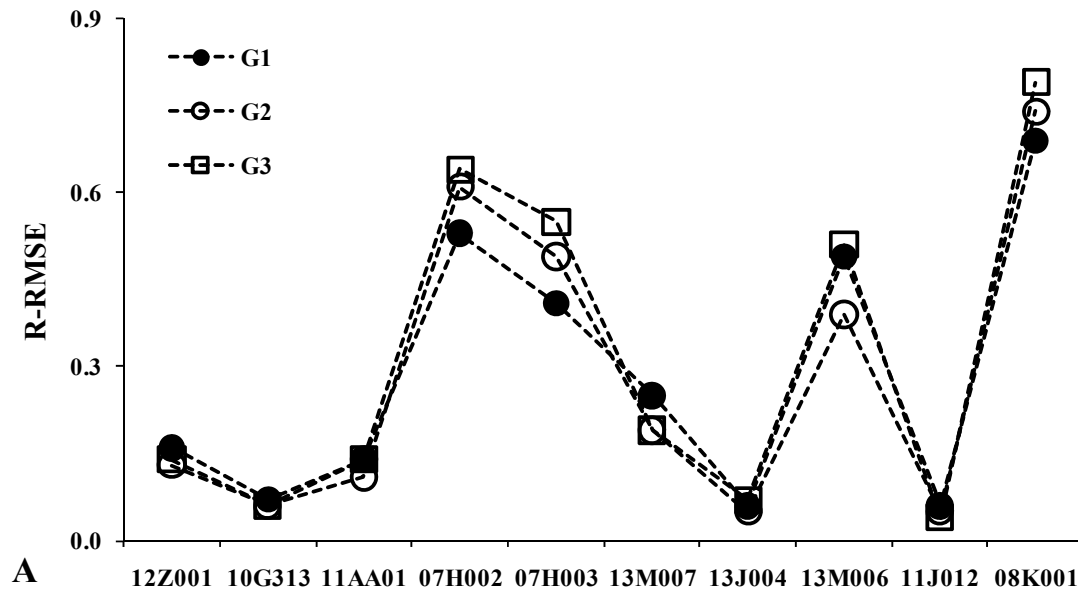
Statistical Models for Seasonal Streamflow



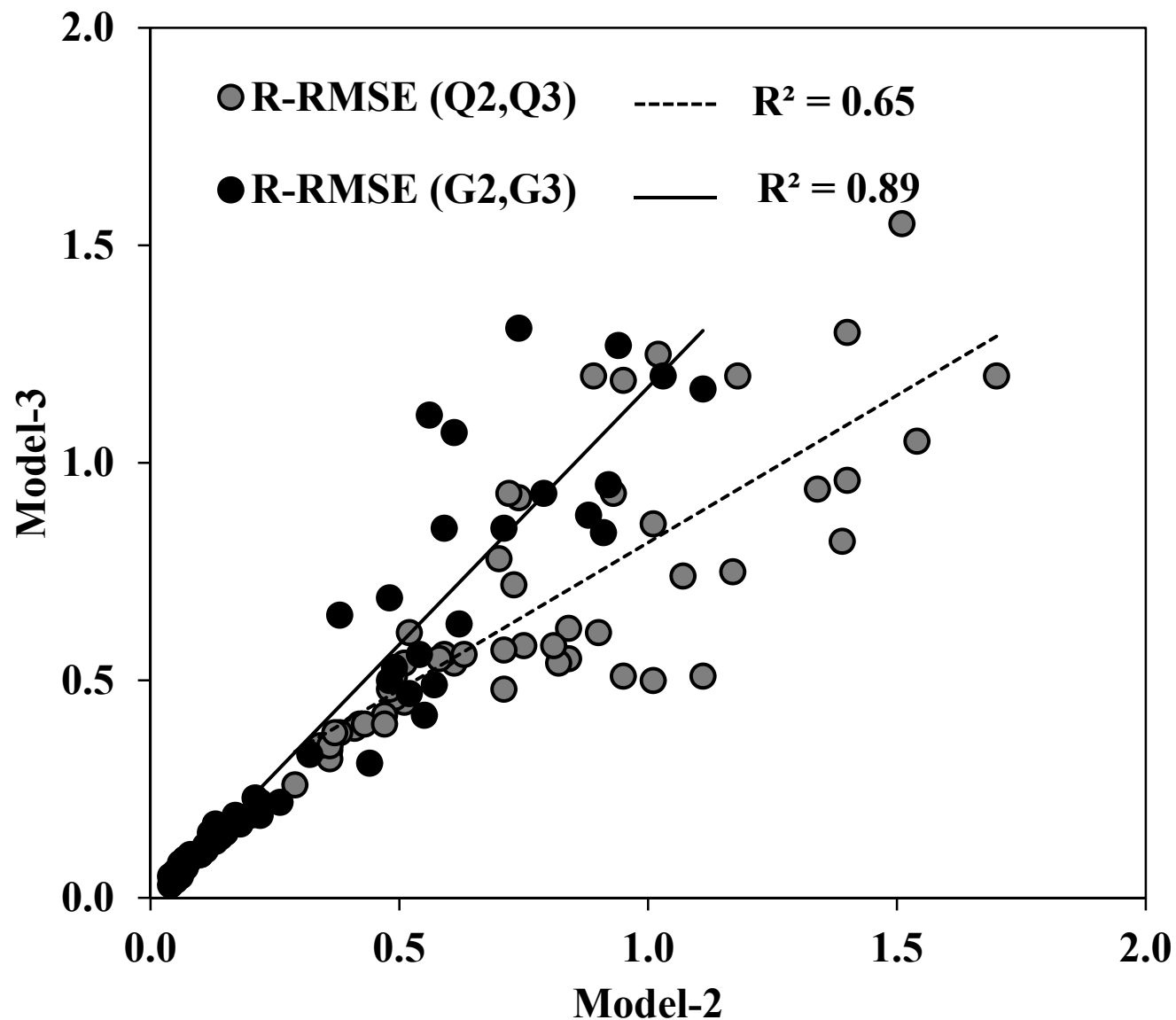
Statistical Model: Seasonal Groundwater



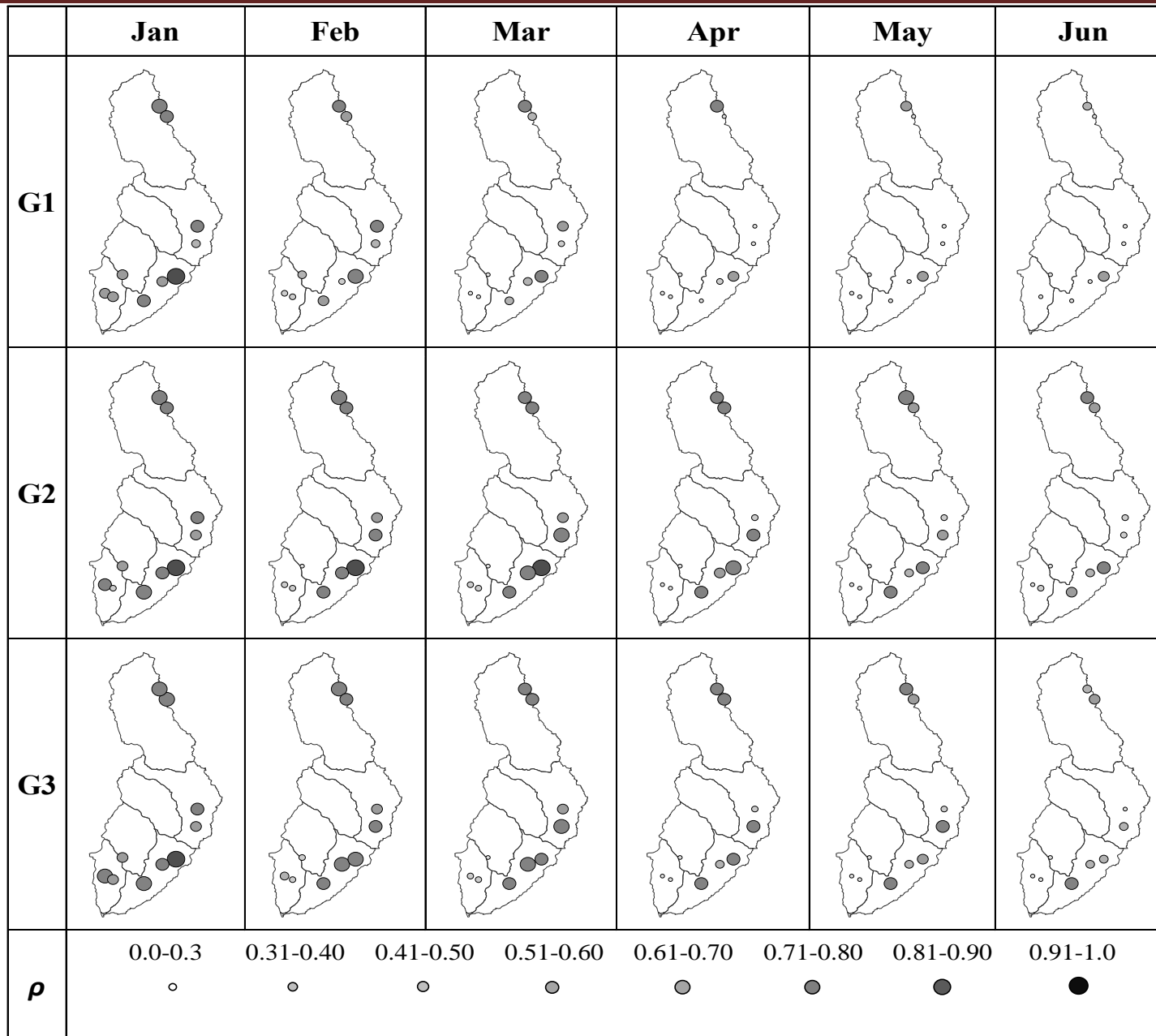
Statistical Model: Seasonal Groundwater



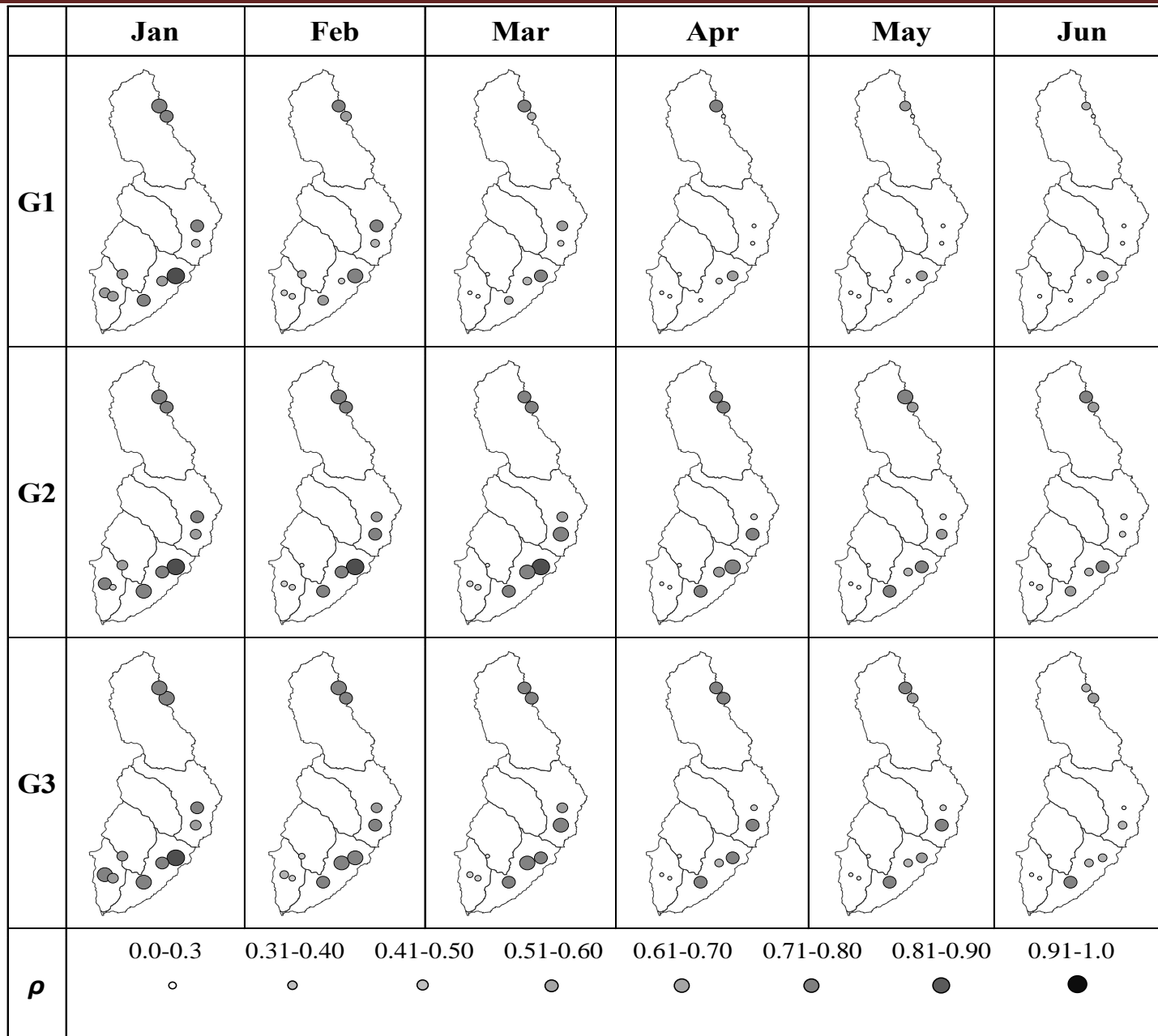
RMSE: Seasonal Streamflow and Groundwater Forecasts



Statistical Model: Monthly Groundwater Forecasts



Statistical Model: Monthly Groundwater Forecasts



Summary and Conclusions

- 1) **Climate variability signature is prevalent over the entire basin**
- 2) **Groundwater integrates basin hydroclimatology**
- 3) **Initial results show potential in predicting surface water up to three months and groundwater up to a lead time of six months**
- 4) **Skill in predicting both surface water and groundwater improves as we approach downstream**
- 5) **Canonical Correlation Analyses, in general, performs better than Principal Component Regression since it uses the spatial correlation in both predictors and predictand**

Acknowledgements

- NC Sea Grant and NC WRRRI
 - Early Career grant supported Naser Almanaseer's PhD dissertation
- NSF - CAREER grant
 - Partially supported Naser's PhD.