



Analysis of Flow Signatures and Catchment Similarity Indices for Catchment Classification in Yesilirmak Basin

Batuhan SOYUGUR, Koray K. YILMAZ

Department of Geological Engineering, Middle East Technical University, Ankara, Turkey

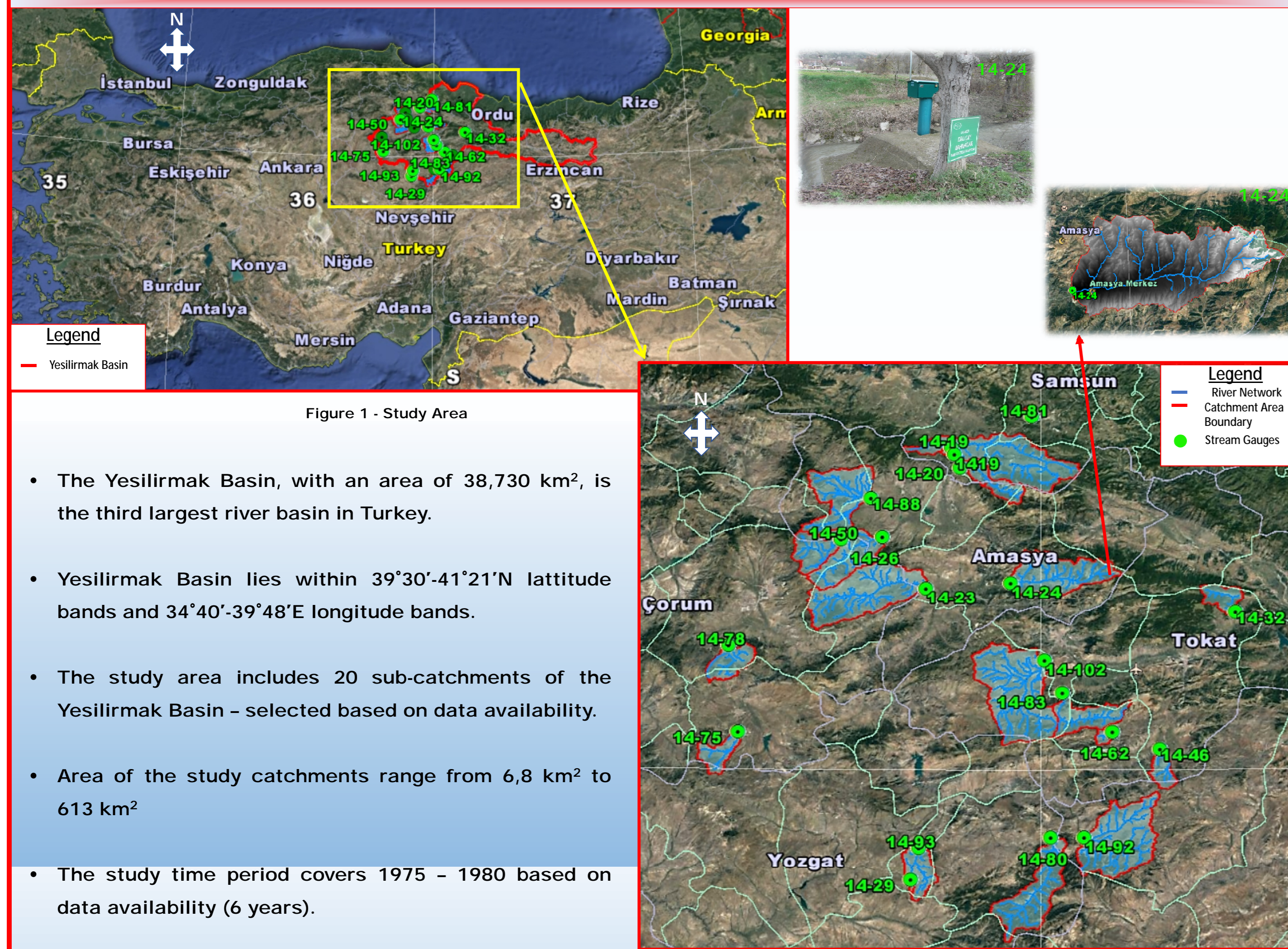
batuhan.soyugur@metu.edu.tr, yilmazk@metu.edu.tr



1) STUDY OBJECTIVES

- Catchment classification schemes aim to identify groups of hydrologically similar catchments to enable a mapping between catchment physical characteristics and hydro-climatic conditions with the catchment functioning.
- This mapping, together with the quantified uncertainties, potentially facilitates improved process understanding, transfer of this understanding to ungauged catchments, model parameter regionalization and hence improve operational applications and watershed management.
- The aim of this study is:
 - To carefully derive hydrologically relevant similarity metrics from catchment physical (elevation, area, aspect, slope, geology, soils, land use), climatic and hydrologic response characteristics (flow signatures).
 - To utilize Affinity Propagation clustering algorithm to determine the optimal number of groupings based on individual as well as a combination of these similarity metrics.

2) STUDY AREA



3) DATA SETS

- STREAMFLOW DATASET:**
 - Daily streamflow data is provided by the General Directorate of State Hydraulic Works (DSI).
- RAINFALL DATASET:**
 - There are 75 rain gauge data (monthly precipitation) provided from Turkish State Meteorological Service.
- SOIL MAP:**
 - National Soil Map of Turkey with 1:25000 resolution
- DIGITAL ELEVATION MODEL:**
 - A digital elevation model having a resolution of 30 m x 30 m was used.
- GEOLOGICAL MAPS:**
 - 1 / 100 000 scale geological maps were provided by General Directorate Of Mineral Research And Exploration (MTA)
- LAND COVER:**
 - Corine 2000 Land Cover Map

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4) METHODS

SELECTED VARIABLES	
AREA	AREA
CLIMATE	Annual Total Precipitation
FLOW INDICES	Q5: High Flow Conditions
	Q50: Medium Flow Conditions
TOPOGRAPHIC	Q85: Low Flow Conditions
	Slope of FDC: Catchment Response to Precipitation
GEOLOGY	Elevation
	Aspect
SOIL	Aluminum (A)
	Volcanic Rocks (VR)
CORINE LAND COVER	Plutonic Rocks (PR)
	Metamorphic Rocks (MT)
SOIL	Clayrich (C)
	Limestone (LS)
SOIL	Schists (SCH)
	Reddish Brown Soils (RBS)
SOIL	Gray Brown Podzolic Soils (GBPD)
	Brown Forest Soils (BFS)
SOIL	Brown Forest Soils Without Litter (BFSWL)
	Hydromorphic Soils (HS)
SOIL	Chestnut (macon) Color Soils (CCS)
	Cultivated Soils (CS)
SOIL	Reddish Marsh Soils (RMS)
	Settlement Areas (SA)
SOIL	Brown Soils (BS)
	Bare Rocks and Rubble Areas (BRRA)
SOIL	Brown Soils Without Litter (BSWL)
	Mixed Data (MD)
CORINE LAND COVER	Dam - Small Dam (DS)
	Water Bodies (WB)
CORINE LAND COVER	Mineral Extraction Sites (MES)
	Vineyards (VY)
CORINE LAND COVER	Discontinuous Built-Up Areas with Family Houses with Gardens (DVAFHG)
	Arable Land with Large Fields (ALWLF)
CORINE LAND COVER	Land Principally Occupied by Agriculture (LPOCA)
	Permanently Irrigated Arable Lands (PIAL)
CORINE LAND COVER	Broad-Leaved Forest (BLF)
	Orchards (OR)
CORINE LAND COVER	Coniferous Forest (CF)
	Complex Cultivation Patterns without Scattered Houses (CCPWSH)
CORINE LAND COVER	Mixed Forests (MF)
	Complex Cultivation Patterns with Scattered Houses (CCPSH)
CORINE LAND COVER	Natural Grassland (NG)
	Bare Rocks (BR)
CORINE LAND COVER	Transitional Woodland Shrub (TWS)
	Beaches, Dunes and Sand Plains (BDSP)
CORINE LAND COVER	Sparsely Vegetated Areas (SVA)

- All the variables were normalized (Z-score) prior to the analysis to match the scale.
- Flow duration curves were prepared for all gauging stations and high flows (Q₅), medium flows (Q₅₀), and low flows (Q₈₅) were calculated for the study time period (1975-1980).

CLUSTERING ALGORITHM

- AFFINITY PROPAGATION** (Frey and Dueck, 2007)
 - Three advantages of the algorithm:
 - The algorithm does not require an initial parameterization (i.e. exemplars, # of clusters)
 - An exemplar is identified which is the most representative of each cluster
 - Highly effective algorithm with lower clustering error compared to majority of the existing methods

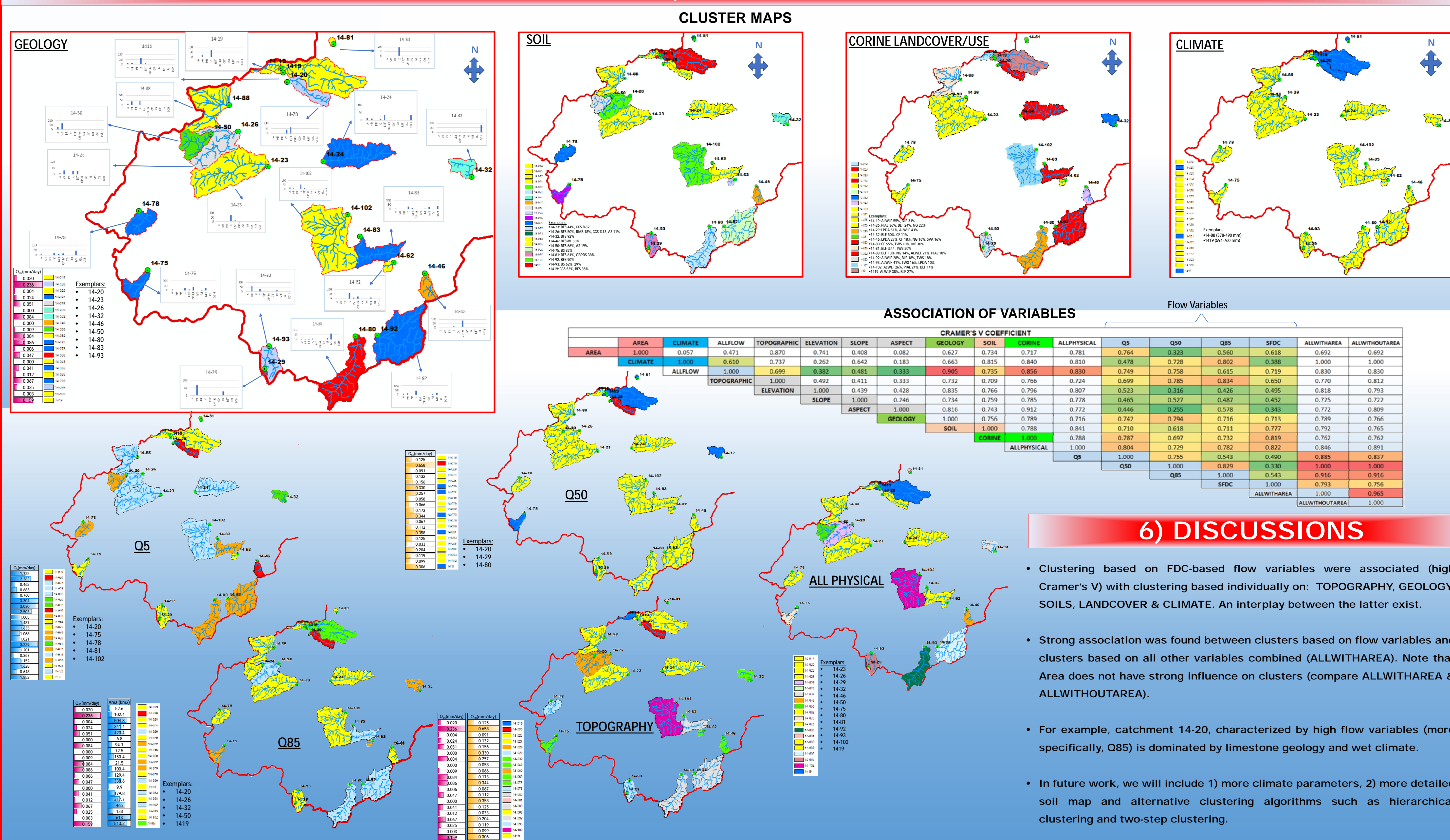
ASSOCIATION BETWEEN CLUSTERS OF DIFFERENT VARIABLES

- CRAMER'S V COEFFICIENT**
 - Cramer's V Coefficient is a way of estimating the extent of relationship between two variables, in our case clusters based on different variables. Cramer's V Coefficient varies between 0 and 1.
 - Groups which have a larger value for Cramer's V can be considered to have a strong relationship between the variables, with a smaller value for V indicating a weaker relationship.

$$V = \sqrt{\frac{\chi^2}{\min(k-1, r-1)}} = \sqrt{\frac{\chi^2/n}{\min(k-1, r-1)}}$$

χ^2 = Pearson's chi-squared test
 n = Grand total of observations
 k = Number of rows
 r = Number of columns

5) RESULTS



6) DISCUSSIONS

- Clustering based on FDC-based flow variables were associated (high Cramer's V) with clustering based individually on: TOPOGRAPHY, GEOLOGY, SOILS, LANDCOVER & CLIMATE. An interplay between the latter exist.
- Strong association was found between clusters based on flow variables and clusters based on all other variables combined (ALLWITHAREA). Note that Area does not have strong influence on clusters (compare ALLWITHAREA & ALLWITHOUTAREA).
- For example, catchment 14-20, characterized by high flow variables (more specifically, Q85) is dominated by limestone geology and wet climate.
- In future work, we will include 1) more climate parameters, 2) more detailed soil map and alternative clustering algorithms such as hierarchical clustering and two-step clustering.