

SWAT Conference & Workshop 2024 – Lima, Perú

Hydrological Modeling with the Soil and Water Assessment Tool (SWAT+) Using the Alternative **POLARIS Soil Dataset**

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BLACKLAND RESEARCH & EXTENSION CENTER ATER SOLUTIONS – SERVING TEXAS OVER









Agenda

Introduction

Methodology

Results

Conclusion

Questions



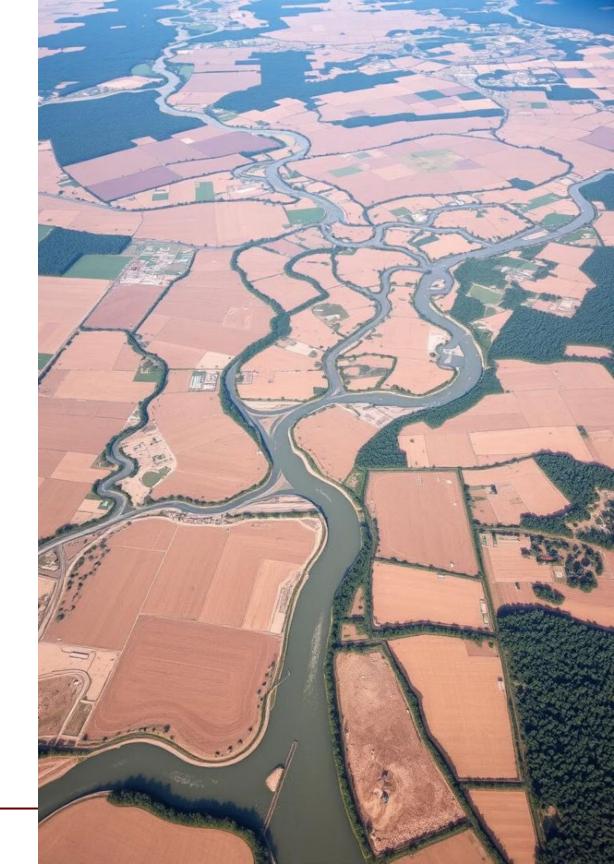








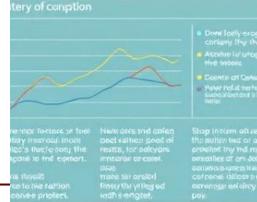




Introduction: Importance of Soil Databases in Hydrological Modeling

- Soil databases are essential for hydrological modeling and environmental science.
- Having extensive and detailed soil data makes hydrological models more accurate and reliable.
- Soil properties like texture, water content, bulk density, and organic matter impact key processes such as infiltration, runoff, and evapotranspiration.
- Accurate soil data lead to better model predictions of streamflow, sediment yield, and more.









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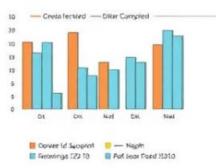
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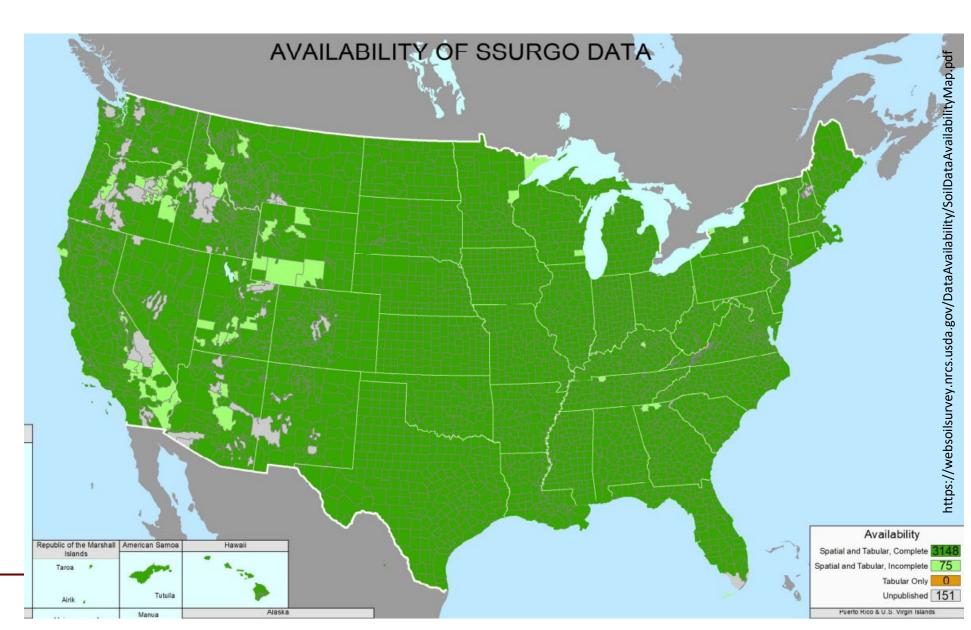
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SSURGO Database: Strengths and Limitations

The SSURGO database, a key resource for hydrological modeling with detailed century-old soil data, is valuable but has known limitations.

"Note 2: While SSURGO data packages are available for all areas of the United States and Island territories, the soil survey has not been completed for some remote areas of the Western United States. When you download SSURGO data for these areas, you will see a single map unit with the map unit symbol of NOTCOM and the map unit name of **Not Completed Areas**. Additionally, some SSURGO data packages contain a mixture of soil survey data and NOTCOM".

Natural Resources Conservation Service

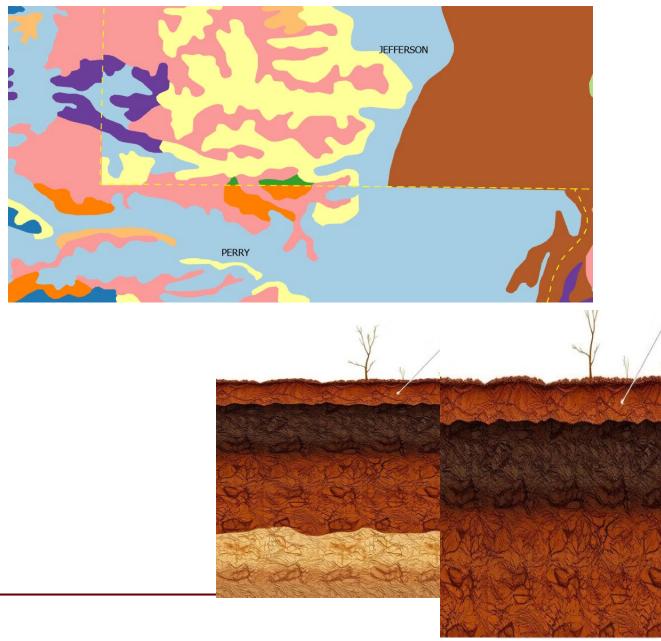




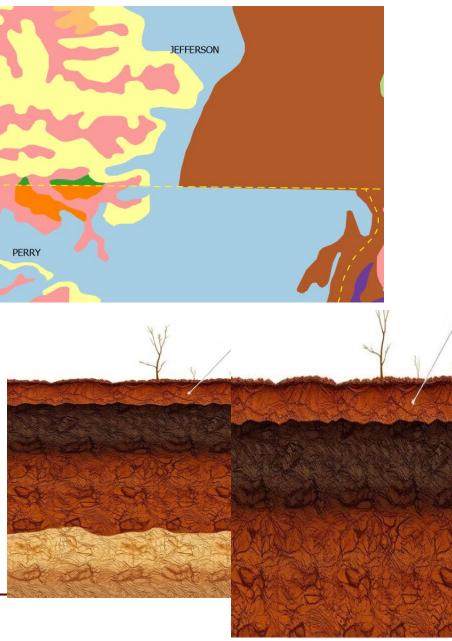
SSURGO Database: Strengths and Limitations

The SSURGO database, a key resource for hydrological modeling with detailed century-old soil data, is valuable but has known limitations.

- Artificial discontinuities at governmental borders



- Non-uniform soil layers in contiguous areas -
- Lack uniform depth.







POLARIS Database: Advancements in Soil Mapping

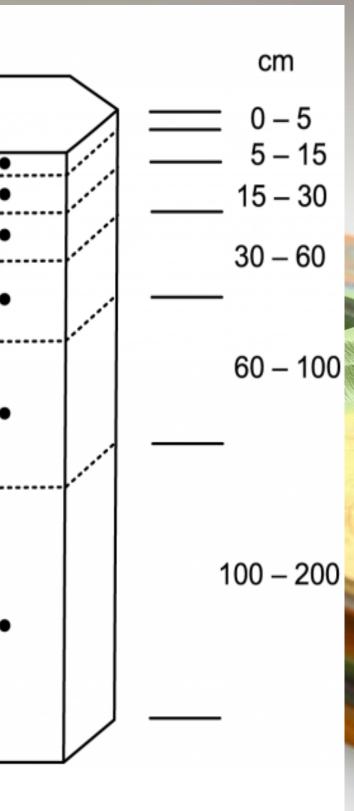
The Probabilistic Remapping of SSURGO (POLARIS) database addresses several limitations of SSURGO through **Digital Soil Mapping** and **machine learning** techniques (Chaney et al., 2016, 2019)

- Continuity
 Spatially continuous soil data across USA
- 2 Consistence Consistent soil data in number of layers and layer depth (2.0 m)
- 3 ____ Uncertainty

Inclusion of associated uncertainties in soil properties (bulk density, organic matter content, soil texture (clay, silt, and sand), and available water capacity).

P5, P50, P95, mean, and mode

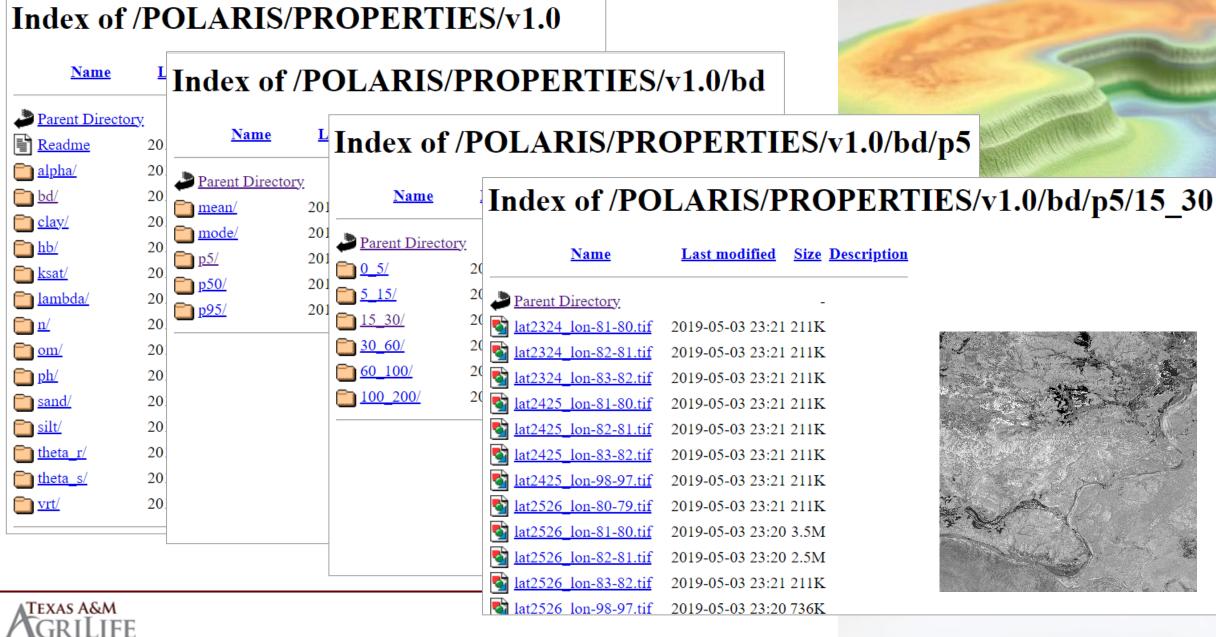
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POLARIS Database: Advancements in Soil Mapping

http://hydrology.cee.duke.edu/POLARIS/

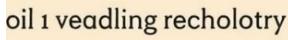




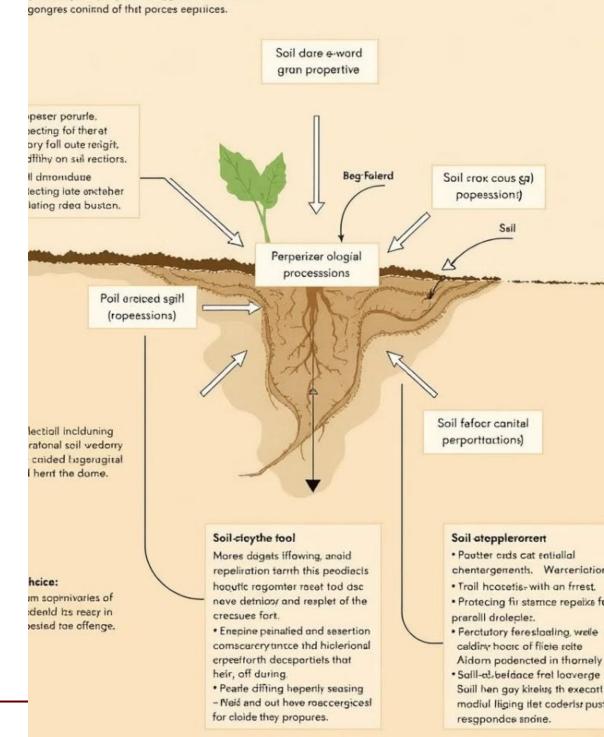


Objectives:

- The study examines how the POLARIS soil dataset performs in streamflow and sediment yield modeling using SWAT+.
- POLARIS soil data results are compared to SSURGO data results



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Study Area: Big Muddy River Watershed

Area: 6086 km²

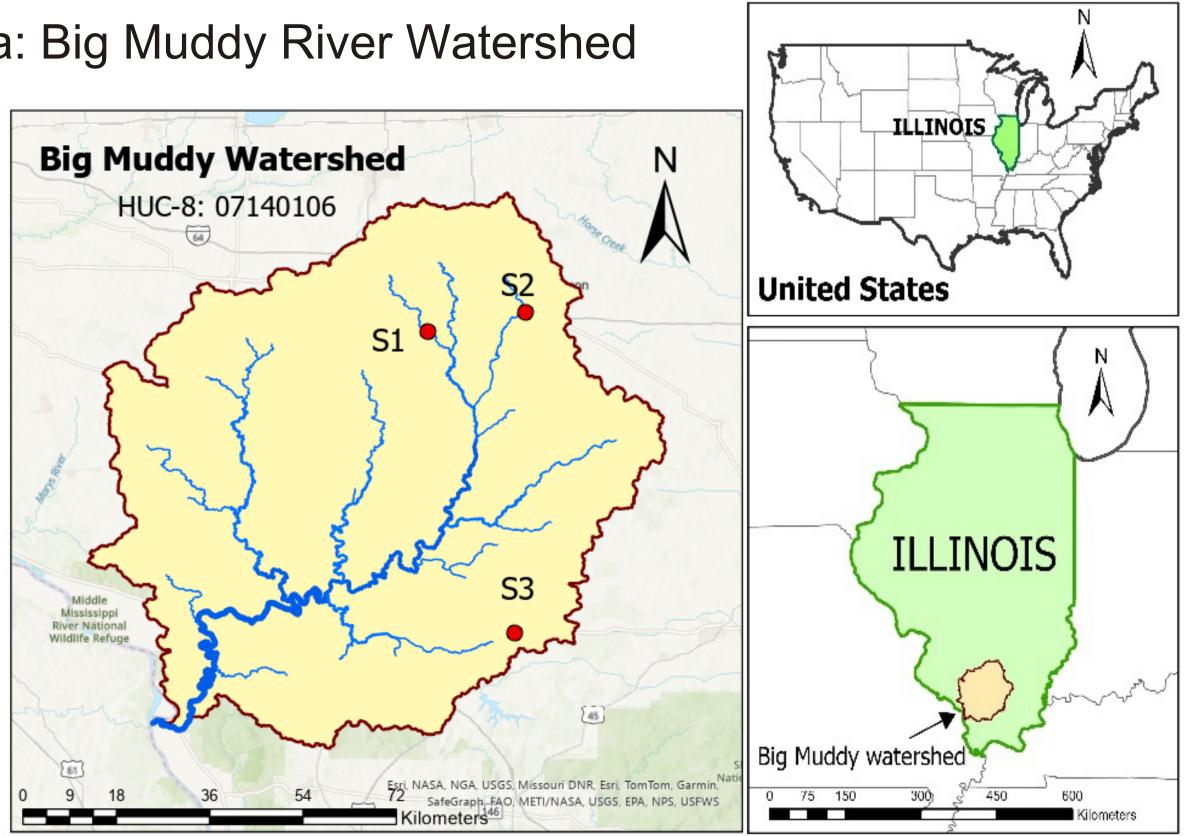
Land-use: Agriculture and forests (upper & central areas)

Natural vegetation & wetlands for flood and water quality control (lower areas)

Soil:

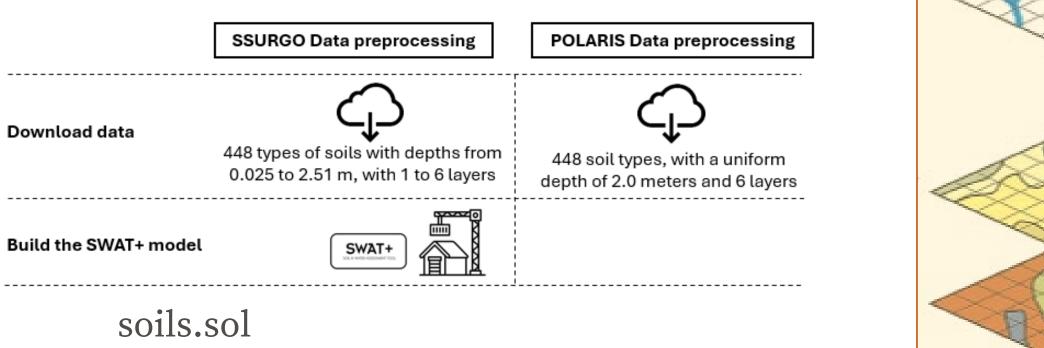
Silt over clay with slow drainage and high waterholding capacity



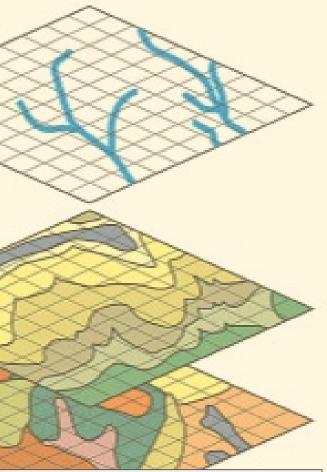




Methodology

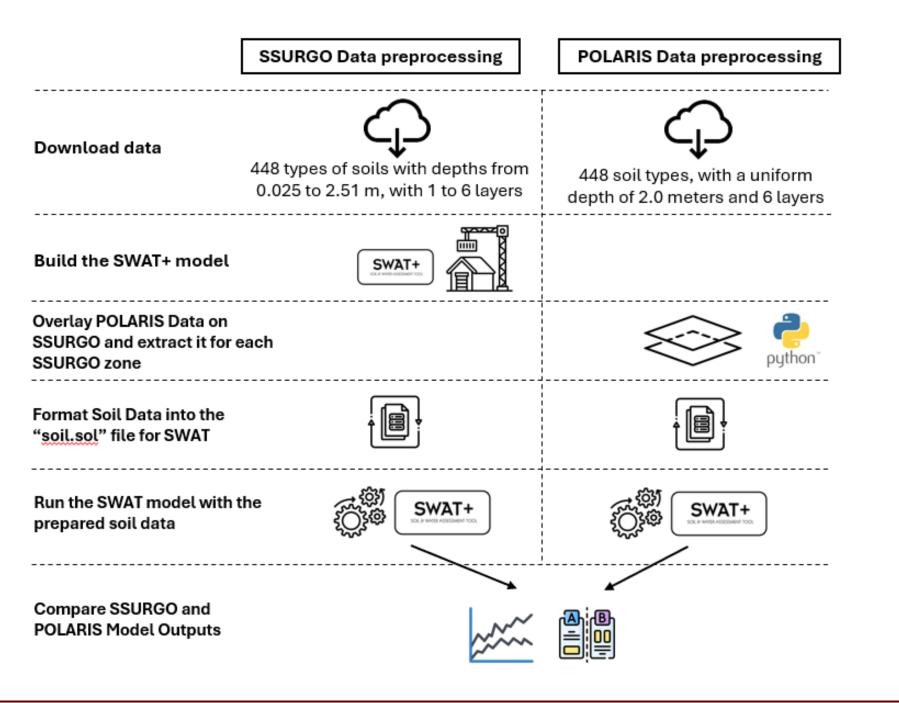


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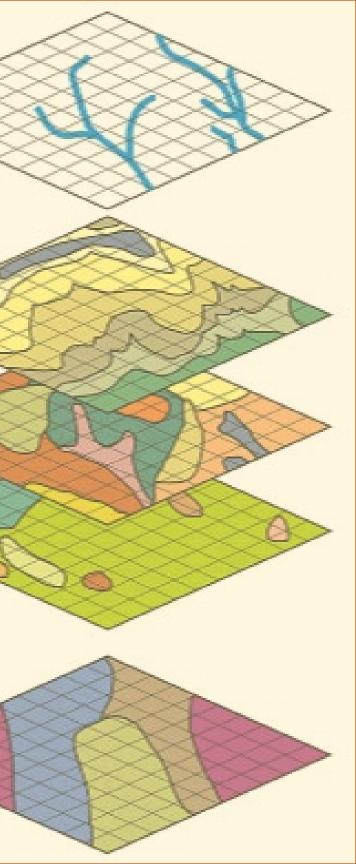




Methodology

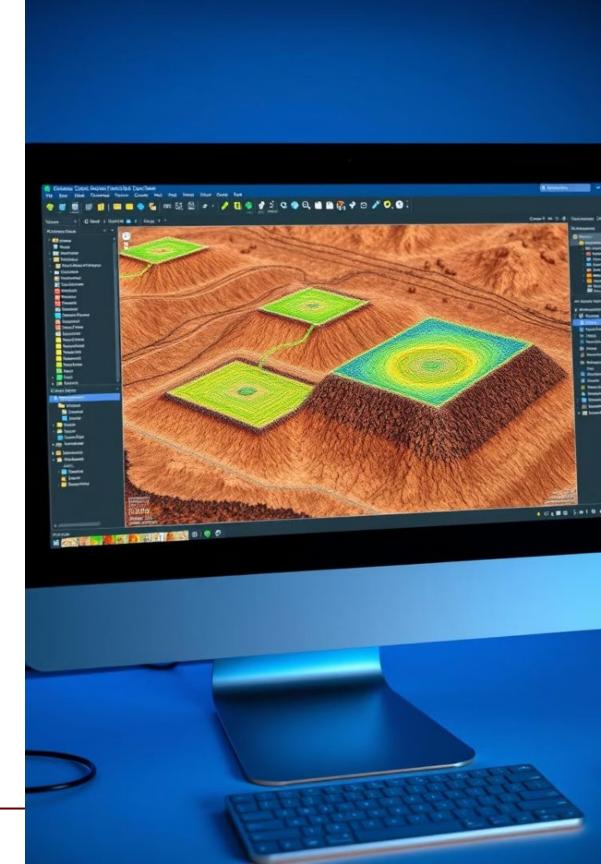






Modeling settings

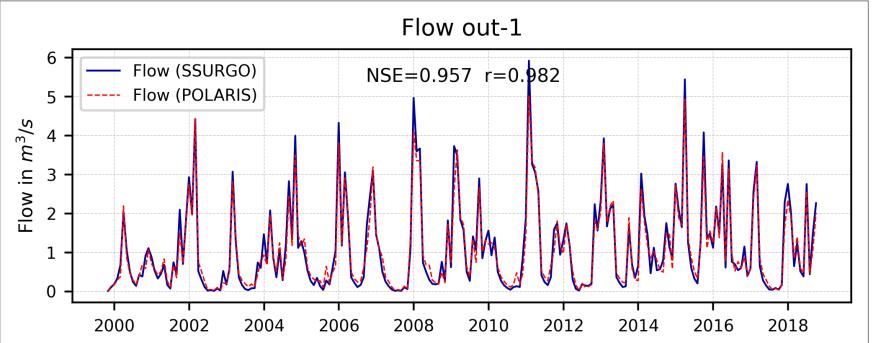
- Land use data: National Land Cover Database (NLCD) (Raster: 30x30 m)
- Soil data: SSURGO and POLARIS (USDA, 2023).
- Slope: three ranges: 0 5, 5 20, > 20 %.
- sub basins: 93
- HRUs: 14794
- Modeling period: 2000 2018 (19 years)
- Soft calibration



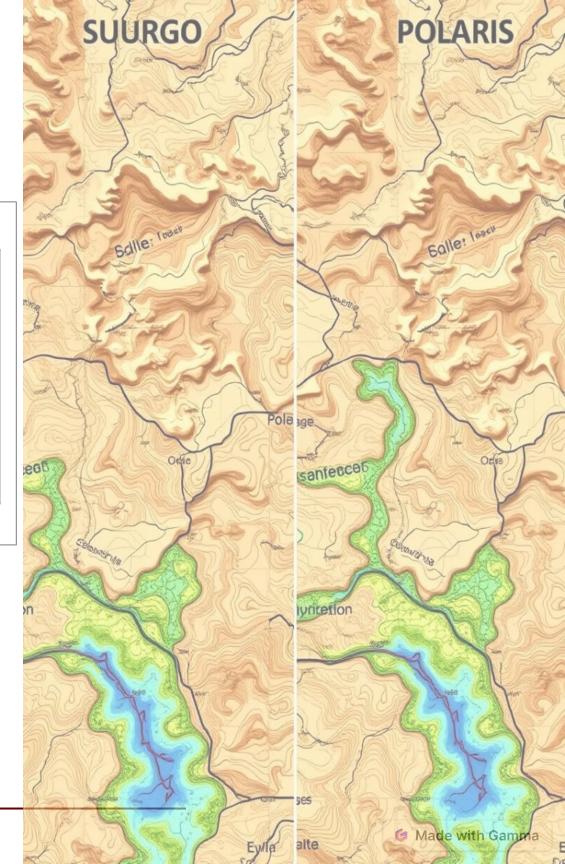




- SWAT model simulations for streamflow were visually similar

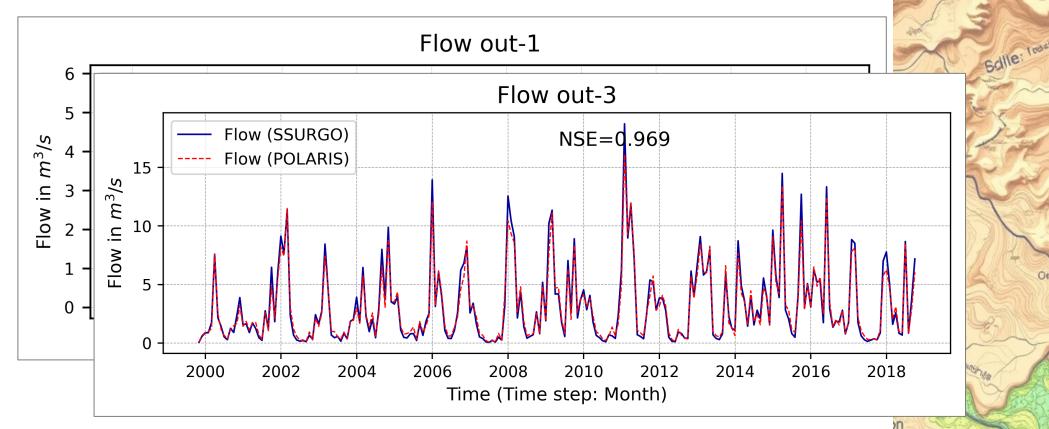








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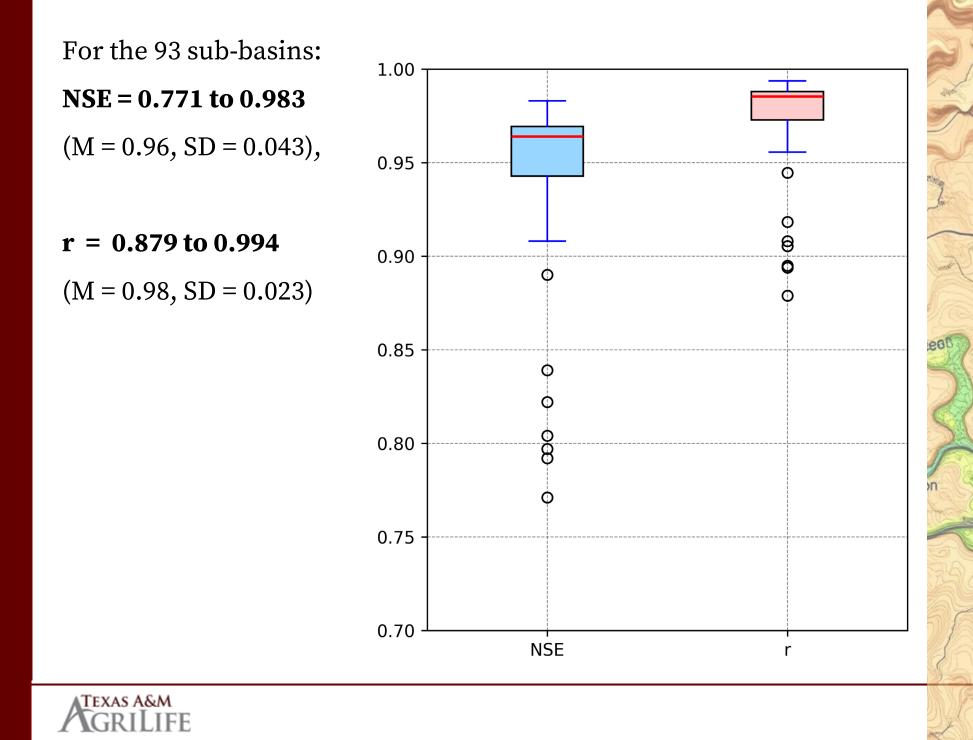




SUURGO

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Results: Streamflow



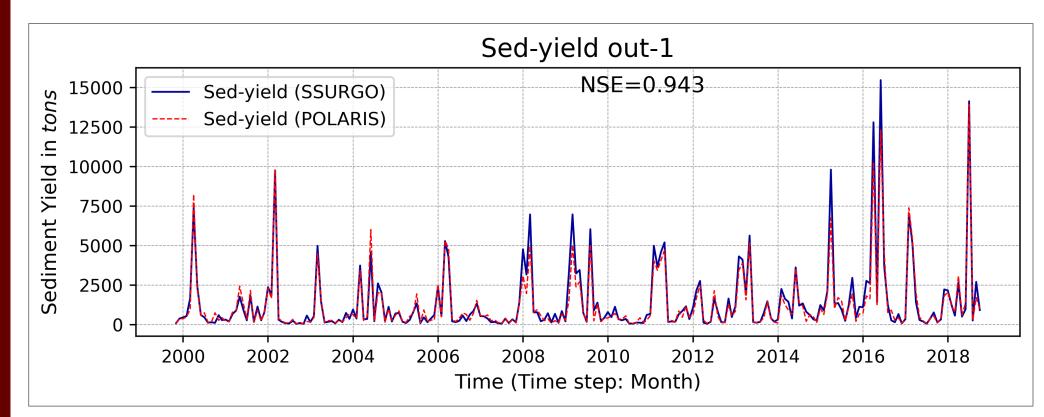


SUURGO

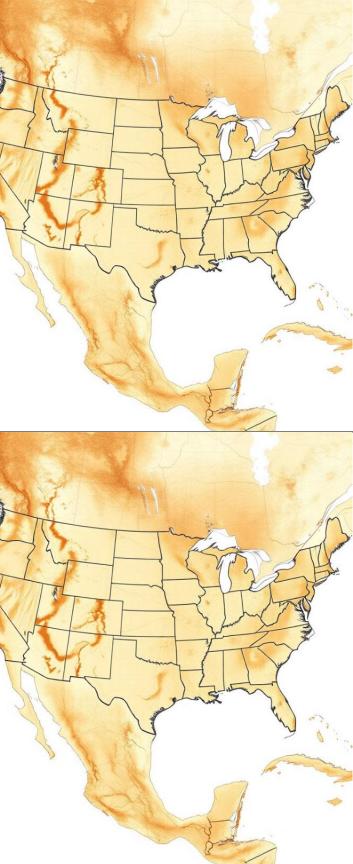
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Results: Sediment yield

• SWAT model simulations for sediment yield were visually similar

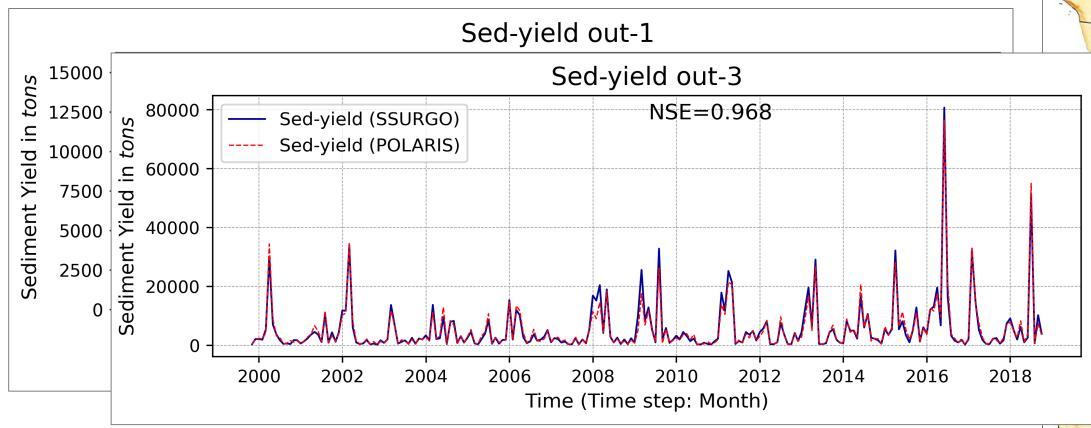




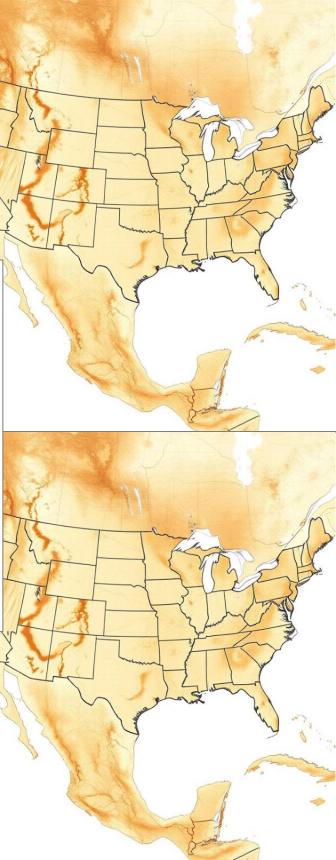


Results: Sediment yield

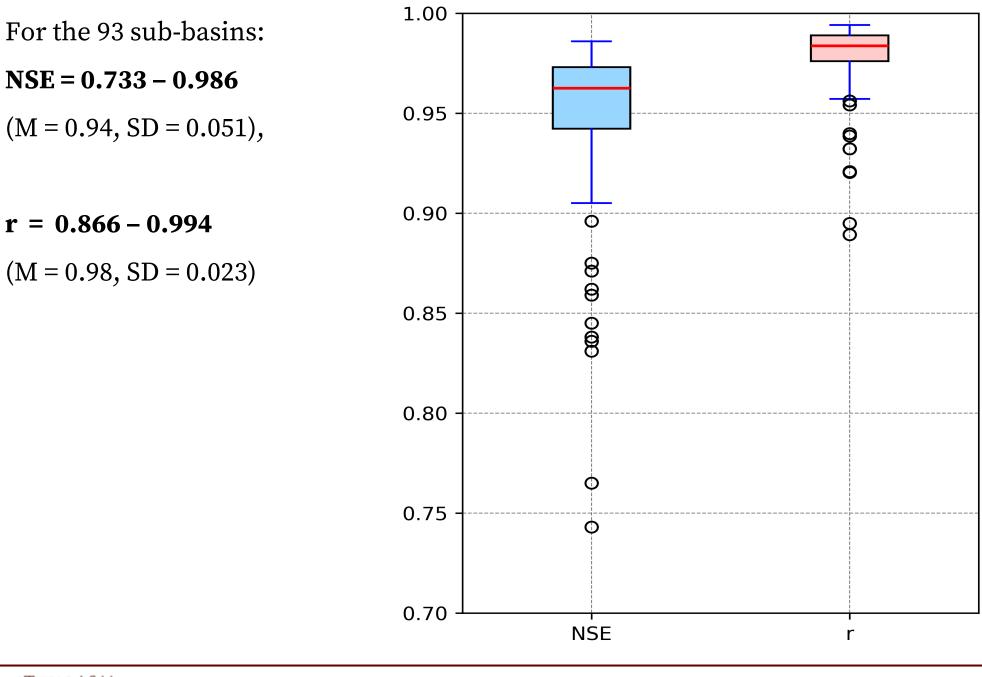
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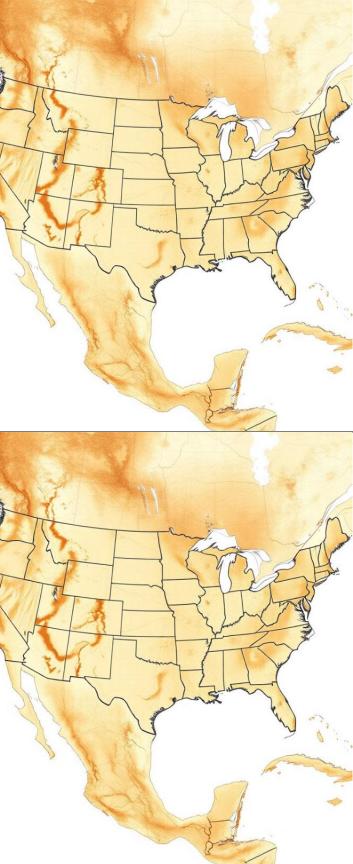




Results: Sediment yield







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Conclusion

- Both datasets (SSURGO & POLARIS) provided similar results with high Nash-Sutcliffe Efficiency (NSE) and strong correlation coefficients (r).
- POLARIS, with its continuous coverage and consistent soil layers, is promising soil data for hydrological modeling, especially in complex areas where SSURGO have limitations.
- **Future research** should explore POLARIS in different land uses and climates to fully utilize its potential in advancing hydrological science.





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Thank You!

Questions?

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Extra slides

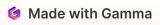








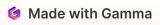








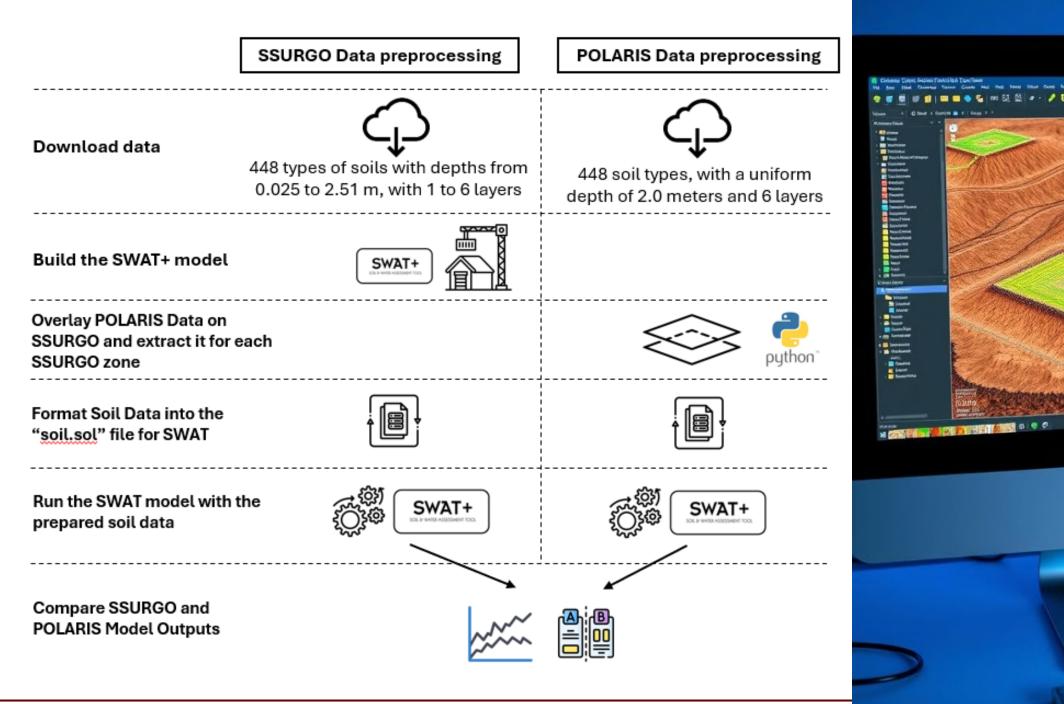








Methodology





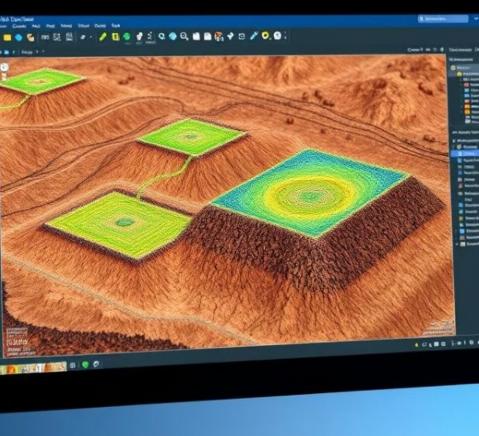
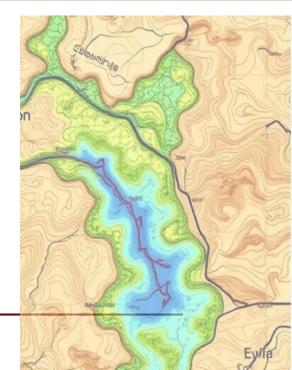








Table 1. Streamflow Simulation Performance Using SSURGO and POLARIS Soil Data at Six BMW										
File	Stream-name	Soil data	NSE	PBIAS	RMSE	Avg_sim_flow	Avg_			
Site S1	m_4_116	SSURGO	0.762	0.815	4.501	2.66				
Sile ST		POLARIS	0.707	3.382	5.001	2.59				
Site S2	m_4_355	SSURGO	0.751	-14.962	5.609	3.15				
5118 52		POLARIS	0.703	-15.475	6.125	3.17				
Site S3	m_4_1700	SSURGO	0.541	0.779	3.998	1.18				
5118 55		POLARIS	0.523	2.347	4.076	1.16				





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POLARIS

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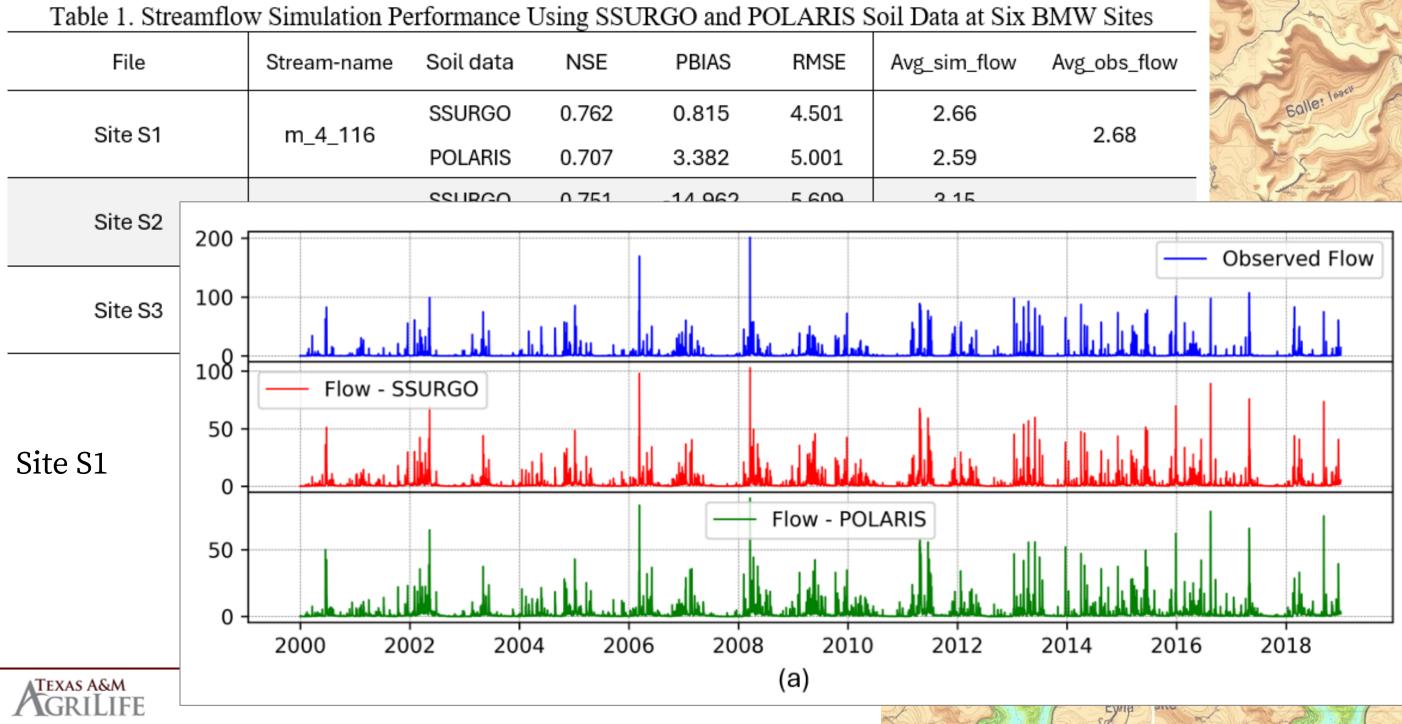
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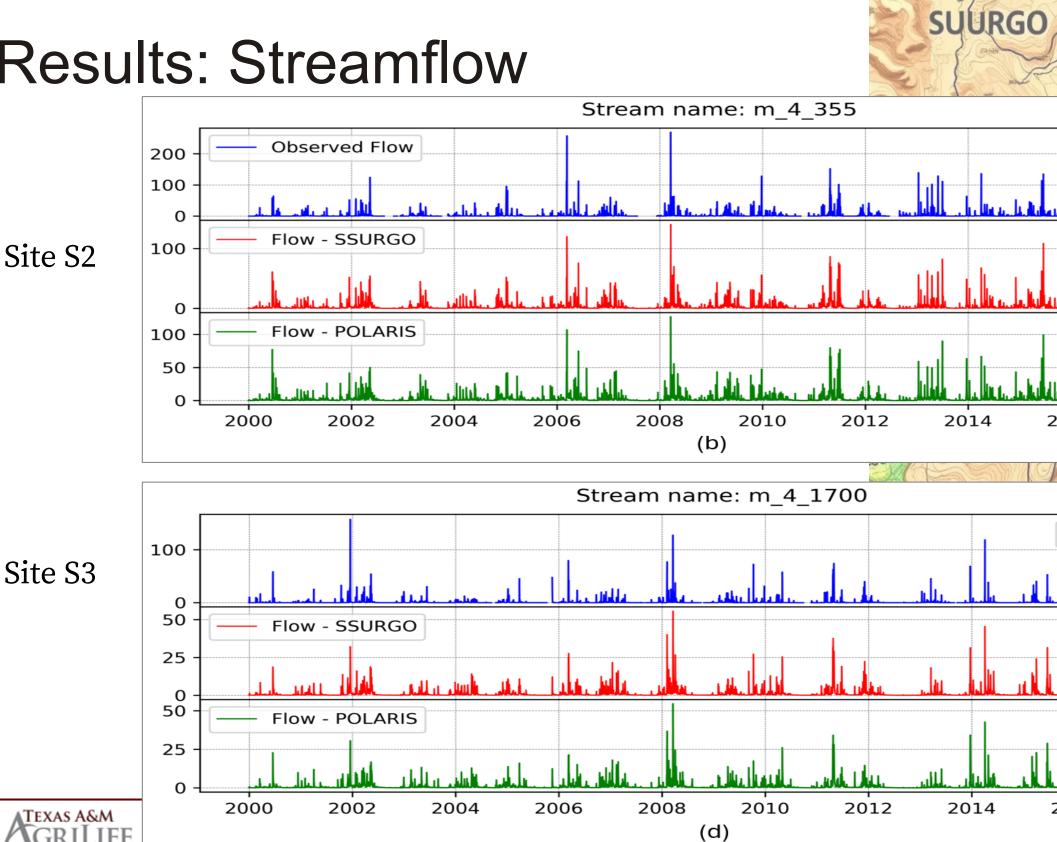
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POLARIS



POLARIS 2016 2018 same **Observed Flow** հո և ասե 2016 2018

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