

# WATER CHEMISTRY OF THE GULF COAST AQUIFER

Kelsey Calvez, Environmental Scientist, Freese & Nichols, Inc.

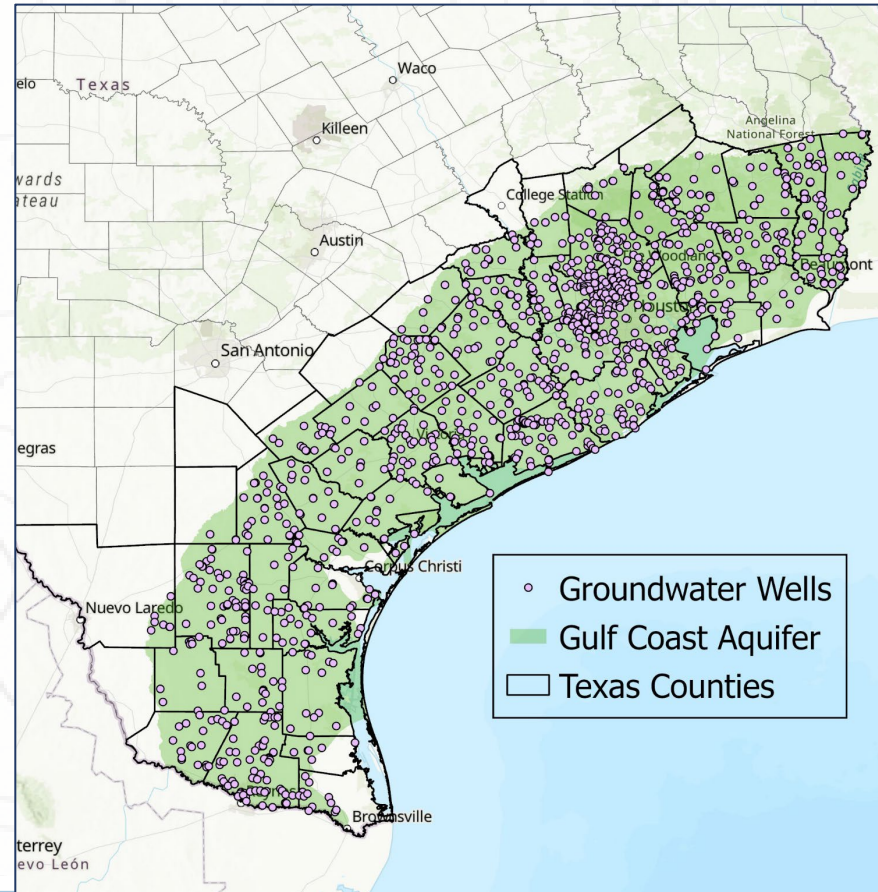
*Penn State University, MGIS Graduate Studies Peer-Review Report Presentation*



# Texas Water Development Board

## GROUNDWATER DATA VIEWER

- Database of **private and public groundwater wells** across the entire state
- Water quality and water level monitoring data dating back to early 1900s
- **3,000 groundwater wells** in the Gulf Coast Aquifer



# STUDY OBJECTIVES

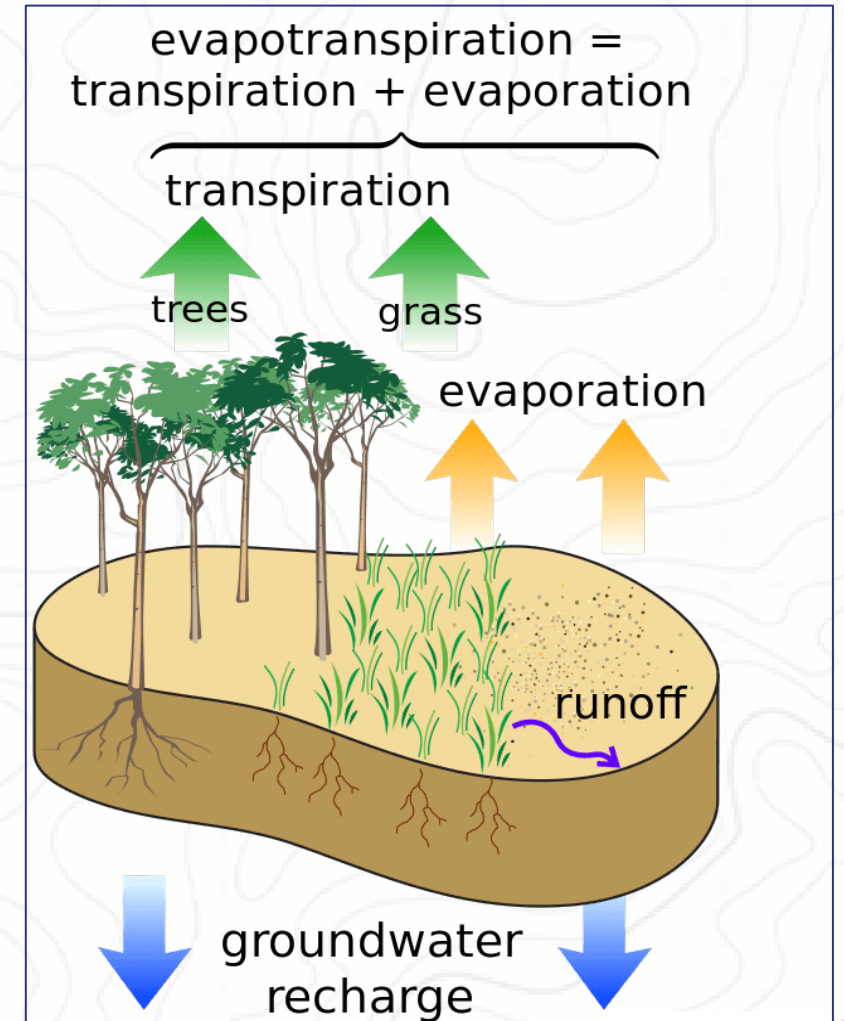
- Evaluate the full suite of chemical compounds (all 33) against land use and climate data for the **Gulf Coast Aquifer**.
- **Total Dissolved Solids** (TDS) used a dependent variable (the variable that's being predicted).
- TDS has been used as an **indicator of groundwater contamination** in previous studies.
- **Fluoride**
  - Natural and anthropogenic sources (coal burning, oil refining, fertilizer plants)
  - Beneficial and harmful impacts above MCL
  - Texas known as a state to have high fluoride concentrations





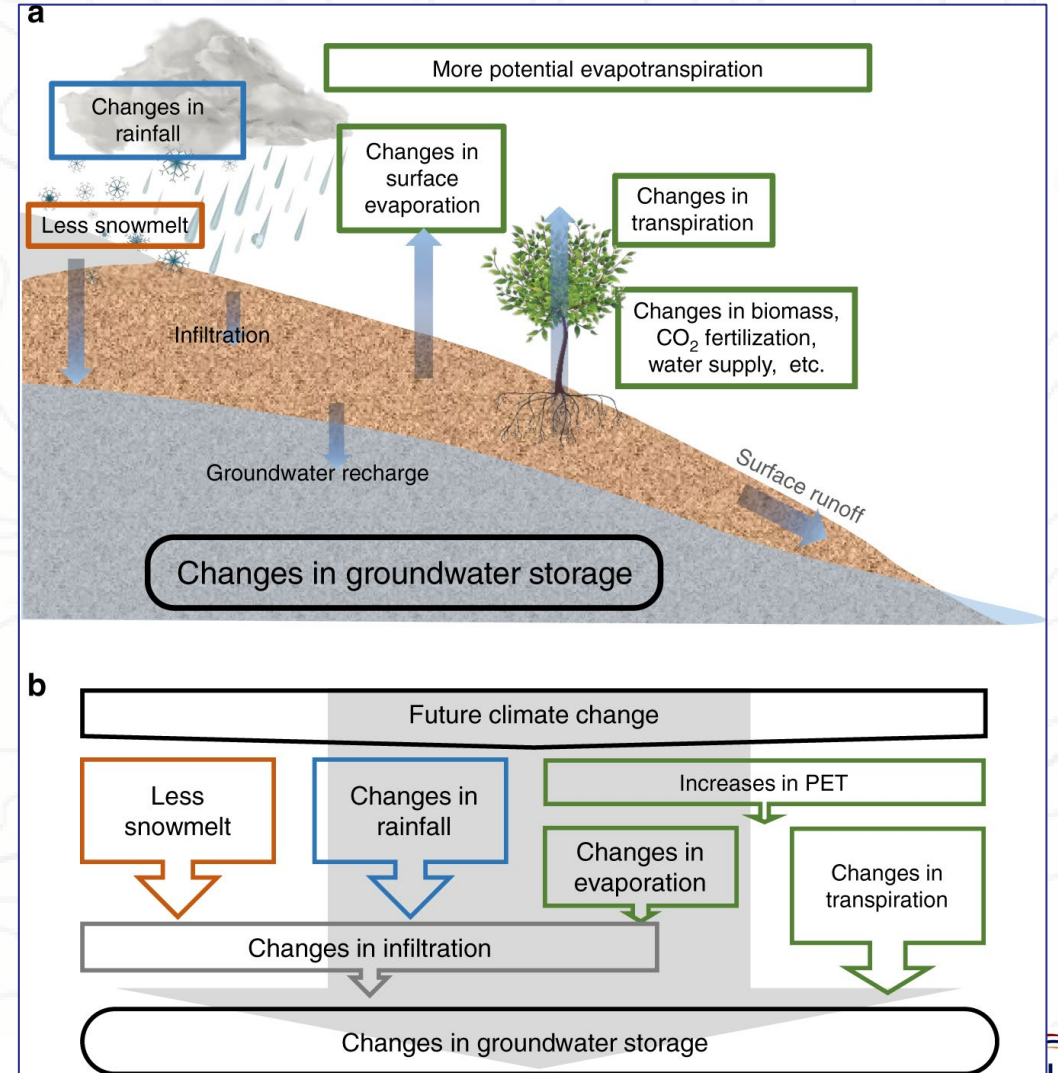
# LAND USE IMPACTS ON GROUNDWATER

- ↑ land development ↓ supply of groundwater
- Impervious surfaces (roads, driveways, roofs) replace forests and wetlands, rainfall/precipitation no longer replenishes aquifers
- Texas has been in a drought for awhile and already has depleted groundwater resources



# CLIMATE INFLUENCE ON GROUNDWATER

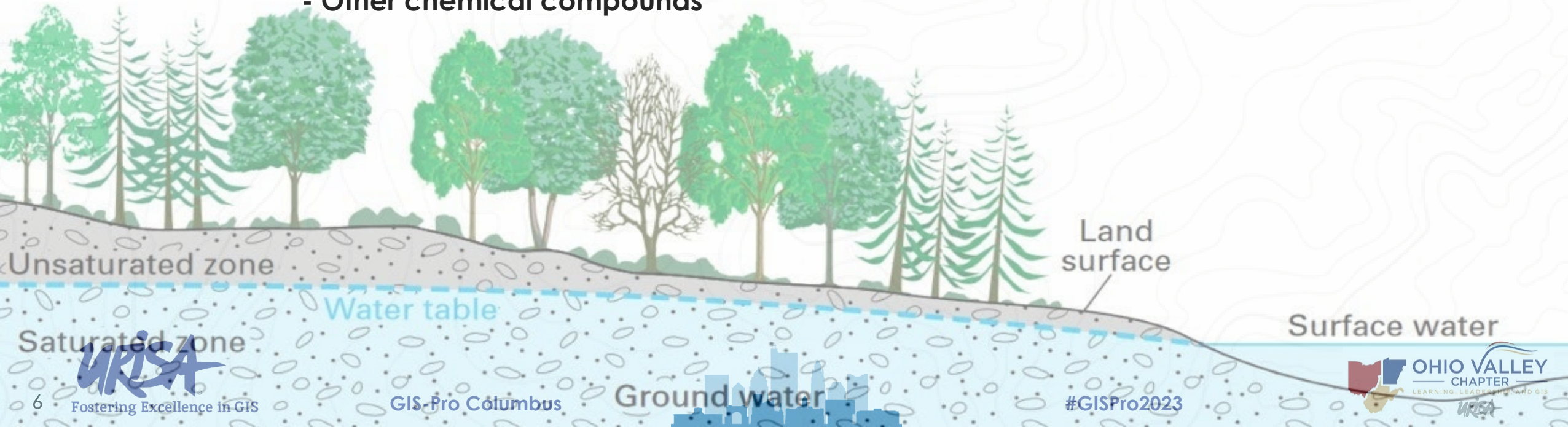
- Less precipitation and less recharge back into the groundwater
- Increased pumping due to less surface water resources
- Decreased soil moisture, increased runoff, and decreased soil infiltration



# STUDY OBJECTIVE (CONT.)

**Question:** What primary or secondary drivers are affecting groundwater quality and thus groundwater contamination?

- Independent variable – **Total Dissolved Solids (TDS)**
- Dependent variables:
  - **Land use (primary)**
  - **Climate (secondary)**
  - **Other chemical compounds**





# LAND USE MODEL VARIABLES

Land Cover

Soil Moisture

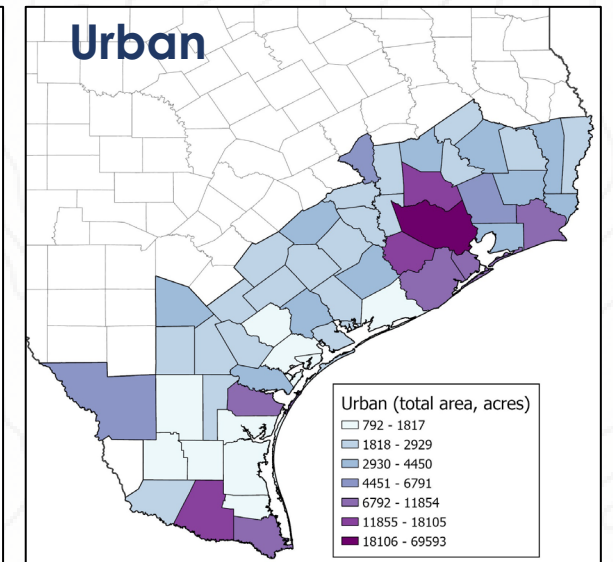
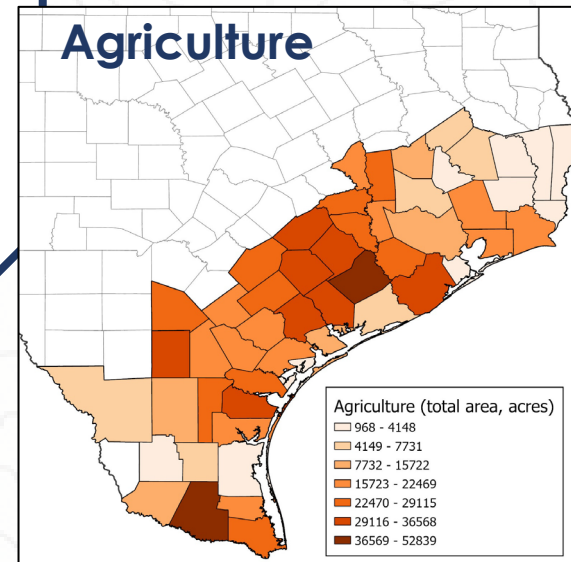
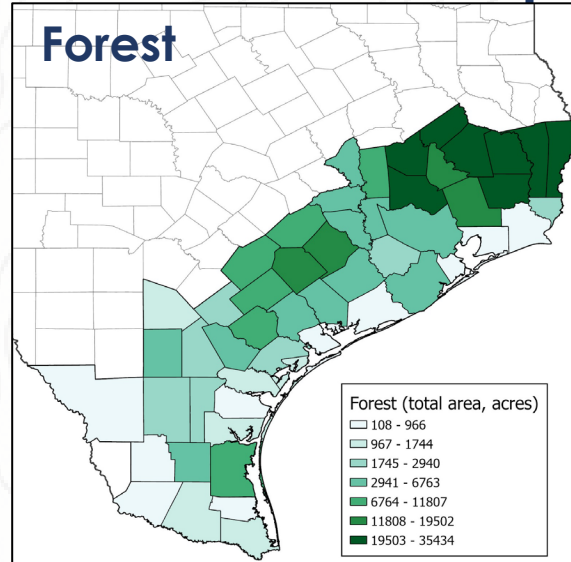
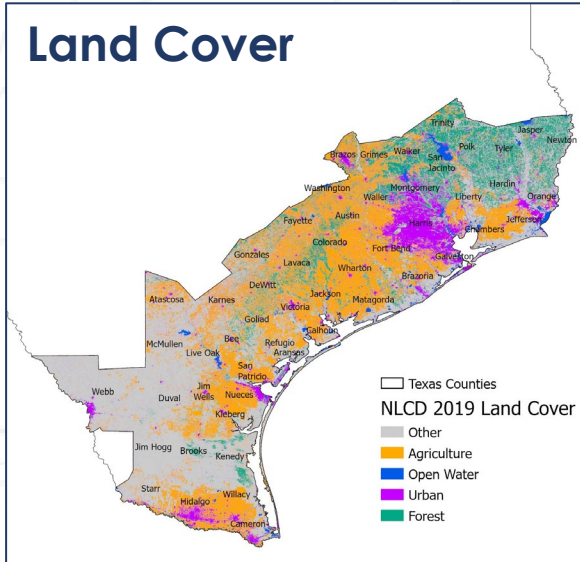
Drought Severity

Streamflow

Groundwater Recharge

Precipitation

Groundwater Quality



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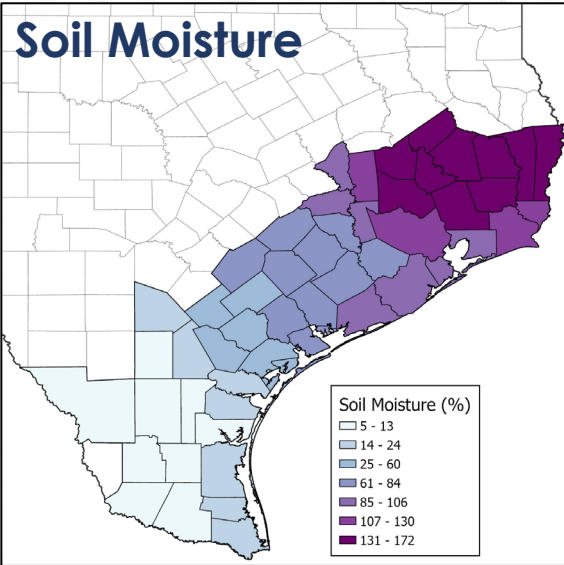
Streamflow

Groundwater Recharge

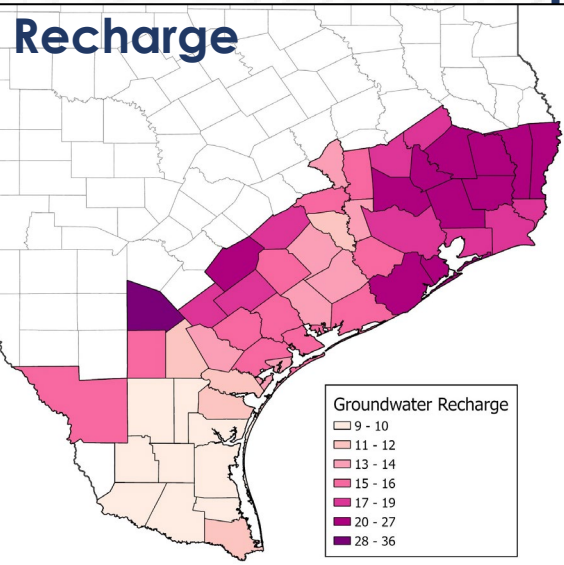
Precipitation

Groundwater Quality

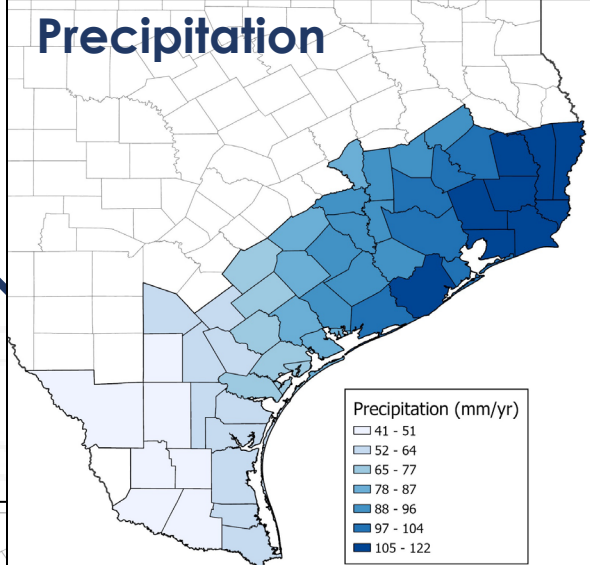
Soil Moisture



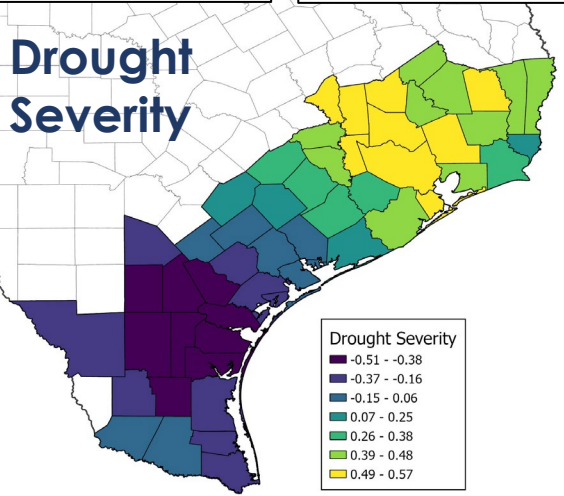
Recharge



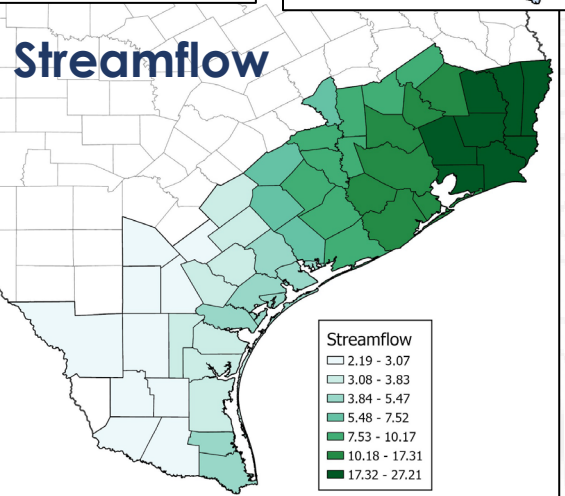
Precipitation



Drought Severity



Streamflow



Over 33 chemical compounds analyzed, including:

- Antimony
- Arsenic
- Barium
- Beryllium
- Boron
- Bromide
- Cadmium
- Chromium
- Fluoride
- Lithium
- Mercury
- Nitrate
- Nitrite
- Selenium
- Thallium



# GLOBAL & LOCAL REGRESSION MODELS

## Global Models

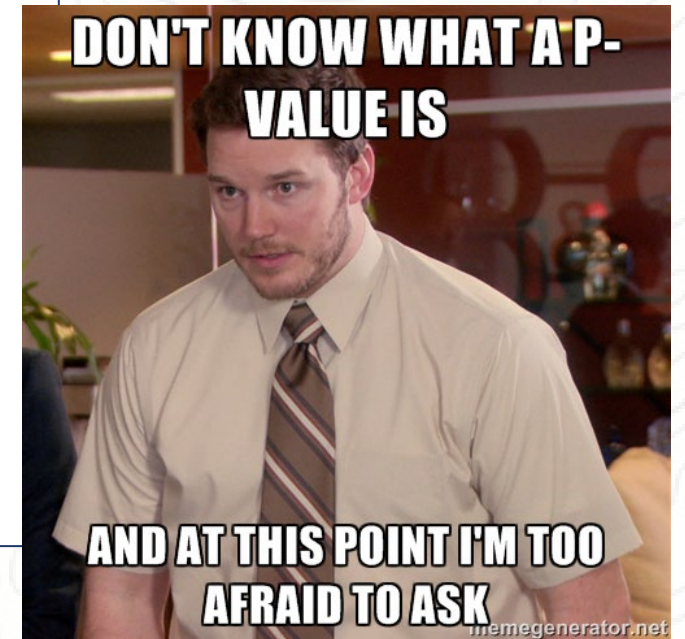
- Exploratory Regression
- Ordinary Least Squares (OLS)
- Spatial Autocorrelation (Global Moran's I)

## Local Model

- Multiscale Geographically Weighted Regression (MGWR)



Photo Credit: ESRI, 2023. Exploratory Regression. ArcGIS Pro Documentation.



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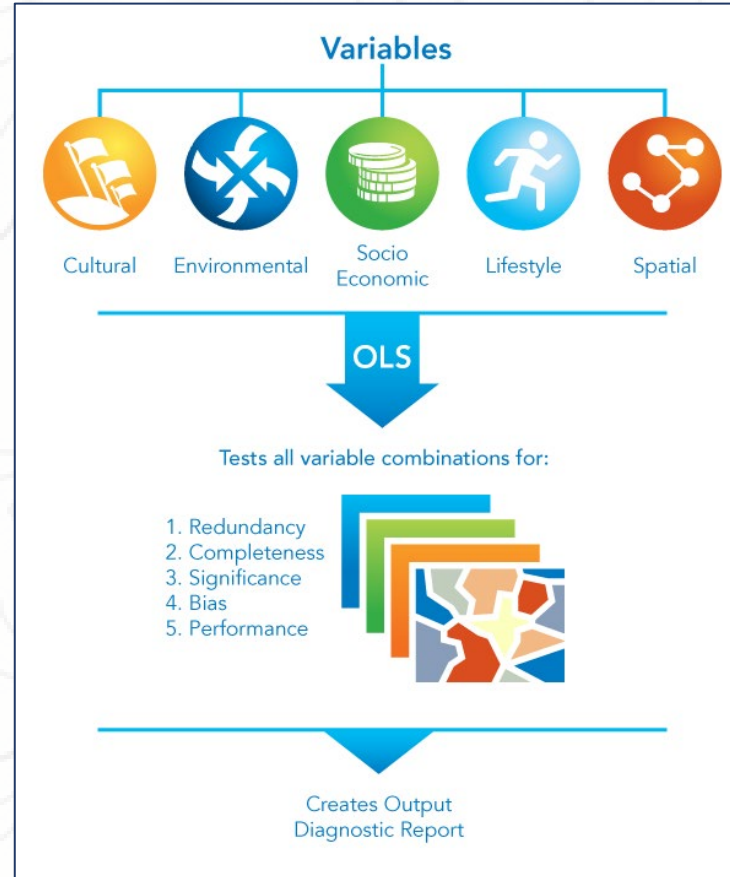
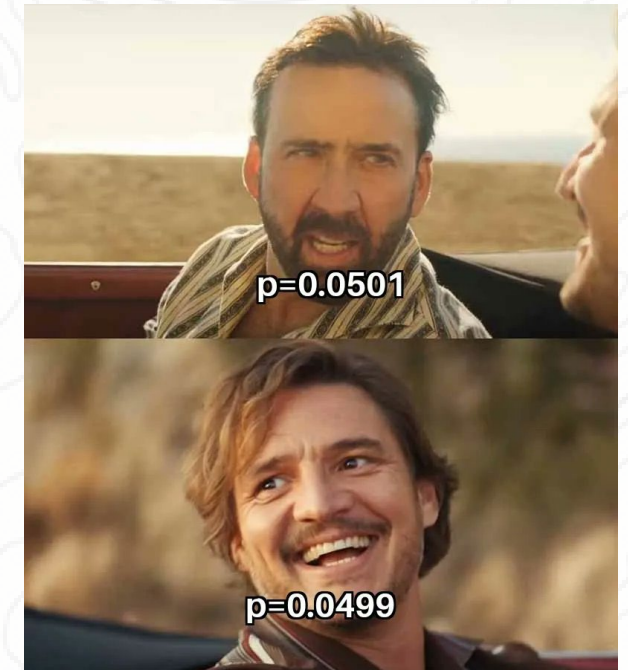


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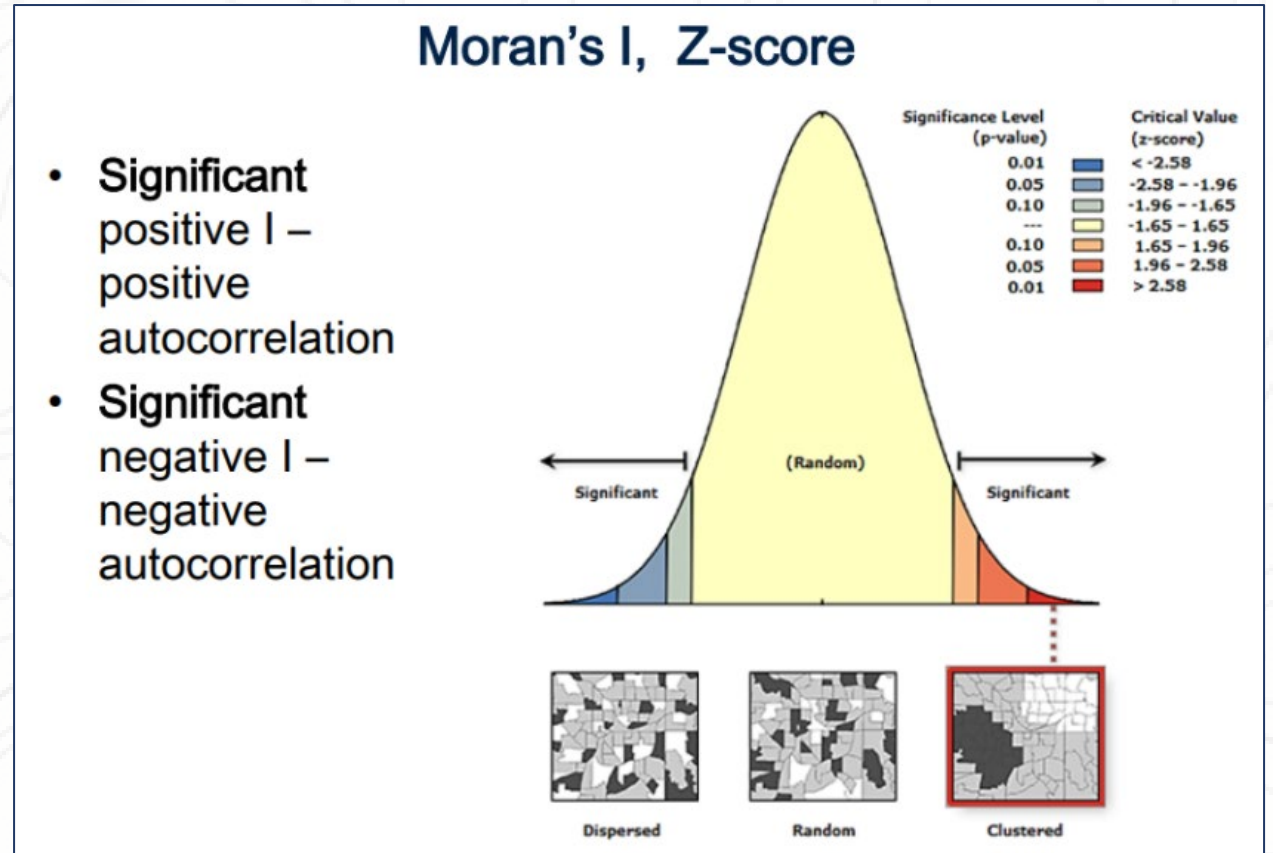


Photo Credit: ESRI, 2023. Spatial Autocorrelation (Global Moran's I). ArcGIS Pro Documentation.



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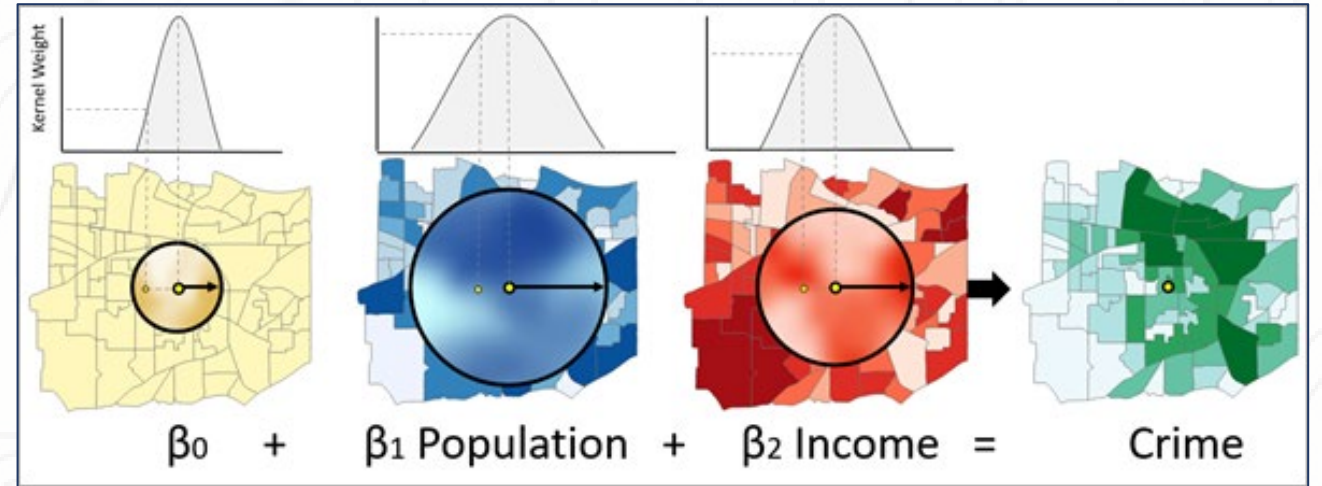


Photo Credit: ESRI, 2023. Multiscale Geographically Weighted Regression (mGWR). ArcGIS Pro Documentation.

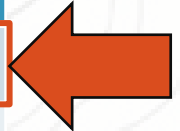
## Local Model

- Multiscale Geographically Weighted Regression (MGWR)

Local models allow you to visualize your data in a **geographic framework**

# CLIMATE MODEL EXPLORATORY REGRESSION

| Adjusted R-Squared | K(BP) <sup>1</sup> | Passing Models <sup>2</sup>  |
|--------------------|--------------------|--|
| 0.81               | 0.04               | +fluoride*** -streamflow***  |
| 0.82               | 0.04               | +fluoride*** -recharge*** -streamflow***                                 |
| 0.91               | 0.01               | +barium*** +boron*** +thallium** +streamflow**                           |
| 0.92               | 0.02               | +barium*** +boron*** +cadmium*** -soil moisture** +streamflow*** +PDSI** |
| 0.91               | 0.02               | +barium*** +boron*** +phosphorus** +thallium** +streamflow**             |

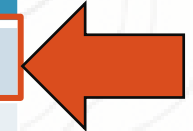


<sup>1</sup>Koenker (BP) Statistic p-value; <sup>2</sup>Model Variable significance (\* = 0.10; \*\* = 0.05; \*\*\* = 0.01).



# LAND USE MODEL EXPLORATORY REGRESSION

| Adjusted R-Squared | K(BP) <sup>1</sup> | Passing Models <sup>2</sup>  |
|--------------------|--------------------|--|
| 0.80               | 0.03               | +fluoride*** -forest**   |
| 0.95               | 0.02               | +boron*** +bromide*** -nitrate*** +agriculture***                    |
| 0.86               | 0.01               | -barium** +cadmium*** +fluoride*** -forest***                        |
| 0.93               | 0.04               | +barium** + boron *** +cadmium** -nitrite*** -fluoride***            |
| 0.93               | 0.04               | + boron *** +cadmium** -chromium** +fluoride*** -forest***           |
| 0.87               | 0.01               | -barium** +fluoride*** +thallium** +agriculture** -forest** -urban** |



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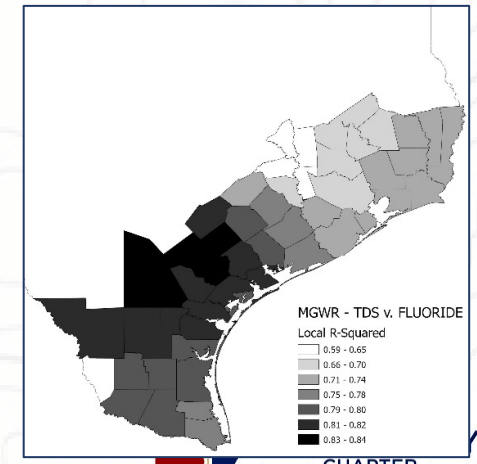
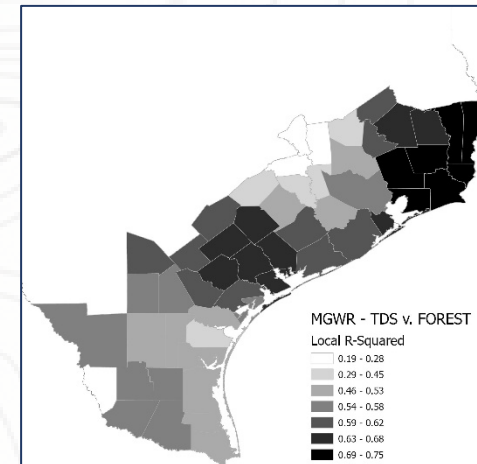
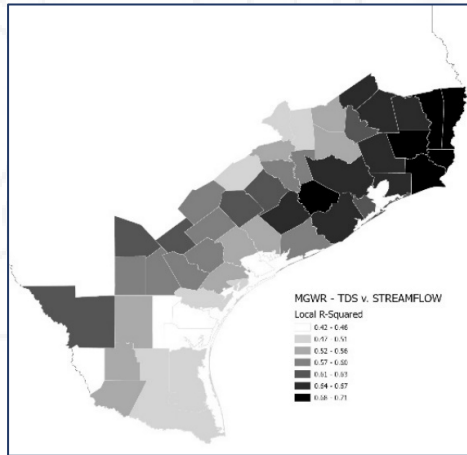
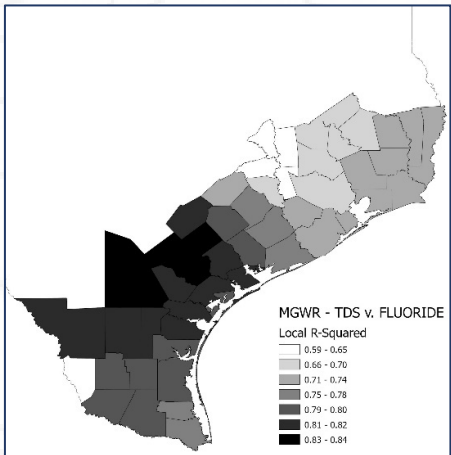
Exploratory Regression Model

Passing Climate Model:  
+fluoride, -streamflow

Passing models then run through OLS to check for:  
redundancy, completeness, significance, bias, performance

Passing Land Use Model:  
+fluoride, -forest

mGWR



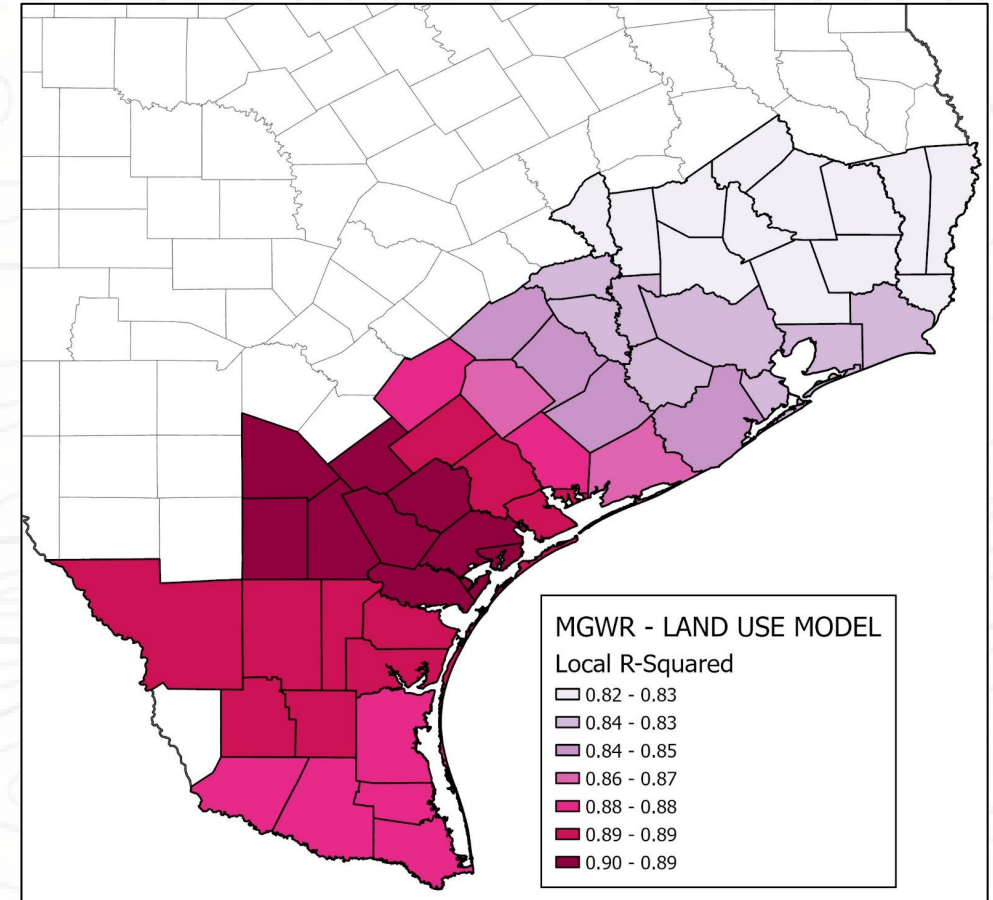
# LAND USE MODEL – MGWR RESULTS

## TDS, +FLUORIDE, -FOREST

**Question:** What primary drivers are impacting groundwater quality?

**mGWR:**

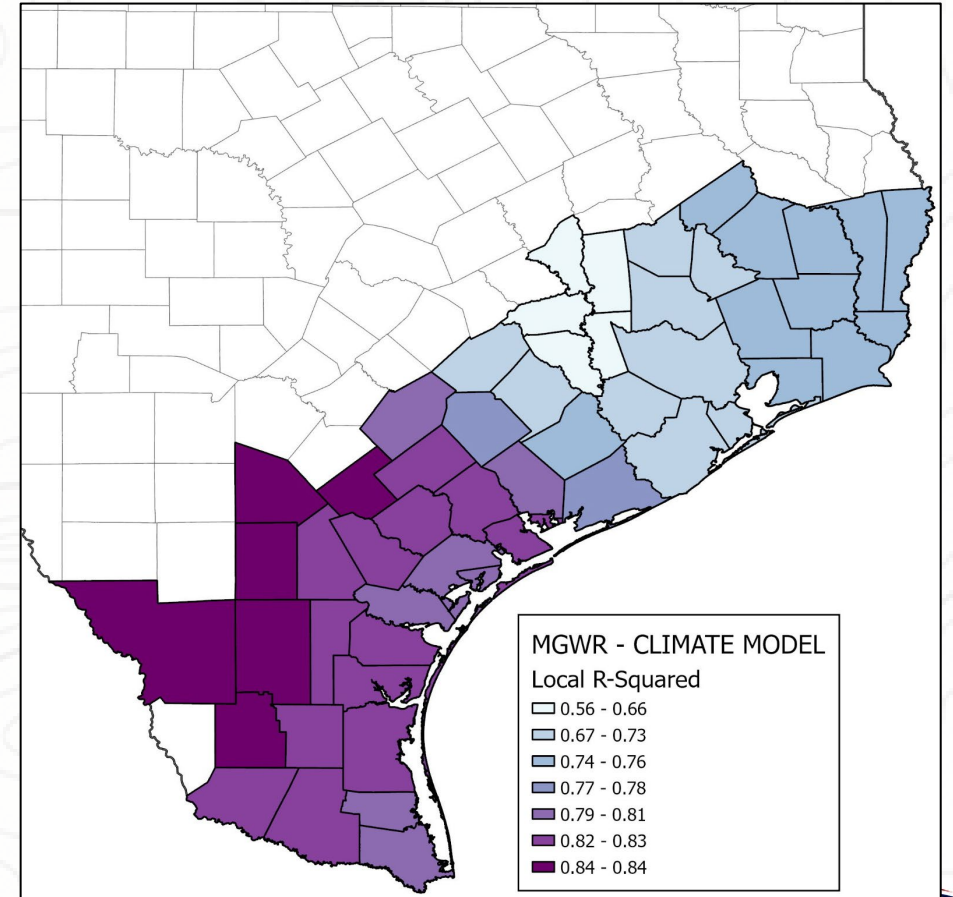
- **TDS and fluoride are highly correlated** (high  $r^2$ ) in the south-west counties
- There is a **lack of forest** in these counties, both due to the natural landscape of TX and due to increased development and agricultural practices
- Forest and other herbaceous land covers (i.e., wetlands) **filter** **contaminates** before they reach the aquifers



# CLIMATE MODEL – MGWR RESULTS TDS, +FLUORIDE, -STREAMFLOW

**Question:** What secondary drivers are impacting groundwater quality?

- Indicative of **overall climate change** that is occurring
- Less precipitation, less streamflow, **less dilution** of contaminants in groundwater
- Intensification of the water cycle which is propagating a drought = **less infiltration** of surface water into the aquifers





# THANK YOU!

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