

# Summary of Remote Sensing Hydrology Lab's Research in Parowan Valley, Utah

# The Remote Sensing Hydrology Lab - vision

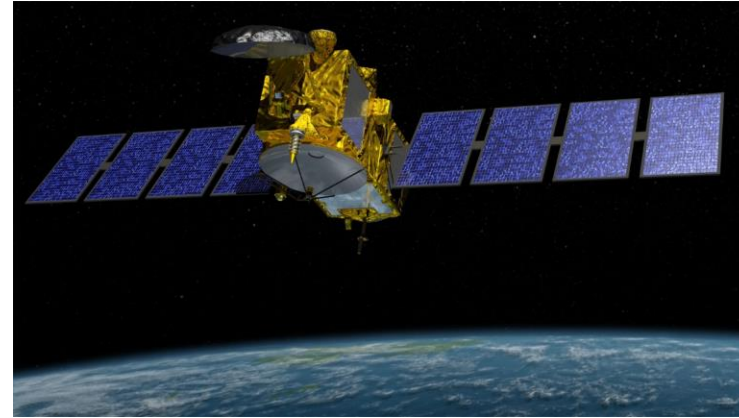
**Develop groundwater methods that integrate in-situ data with the wealth of satellite and ground-based geophysical datasets, improving predictive capabilities and enabling groundwater evaluation in data-sparse regions.**

# The Remote Sensing Hydrology Lab - datasets

In-situ



Satellite



Ground-based



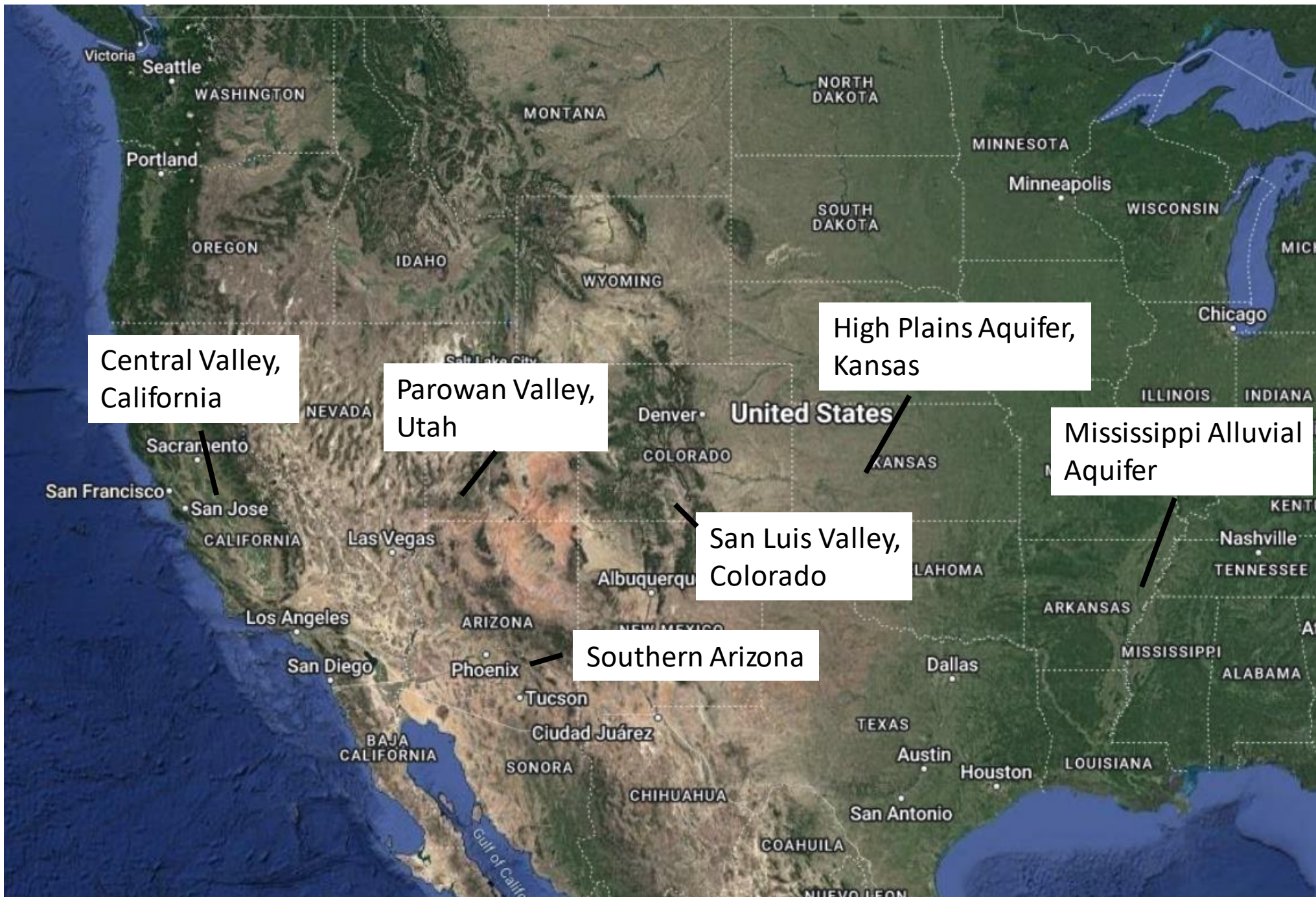
Airborne



# My background

- BS in Geology at Brigham Young University (2014)
- PhD in Geophysics at Stanford University (2018)
- Assistant Professor at Missouri University of Science and Technology (2018-2022)
- Assistant Professor at Colorado State University (2022-present)





Central Valley,  
California

Parowan Valley,  
Utah

San Luis Valley,  
Colorado

Southern Arizona

High Plains Aquifer,  
Kansas

Mississippi Alluvial  
Aquifer

# Why Parowan Valley?

- I'm motivated to study this area because there is an intersection of
  - High-quality data availability
  - Groundwater management priority
  - Compelling science questions
  - Local partnerships

# Authors on projects shown here (in addition to myself)



Jiawei Li, PhD student



Katherine Grote, Associate Professor,  
Missouri University of Science and Technology



Jim Butler, Senior Scientist, Kansas Geological Survey

# Outline of work that has been done to date – contact me for a copy of the papers

- Analysis of key drivers of subsidence in the valley (published)
- Modeling ground deformation with satellite and groundwater level data (published)
- Water budget analysis using satellite, groundwater level, and pumping data (currently under review)
- Geophysical survey of the top ~150 ft of the valley (analysis in progress)

[Ryan.G.Smith@colostate.edu](mailto:Ryan.G.Smith@colostate.edu)

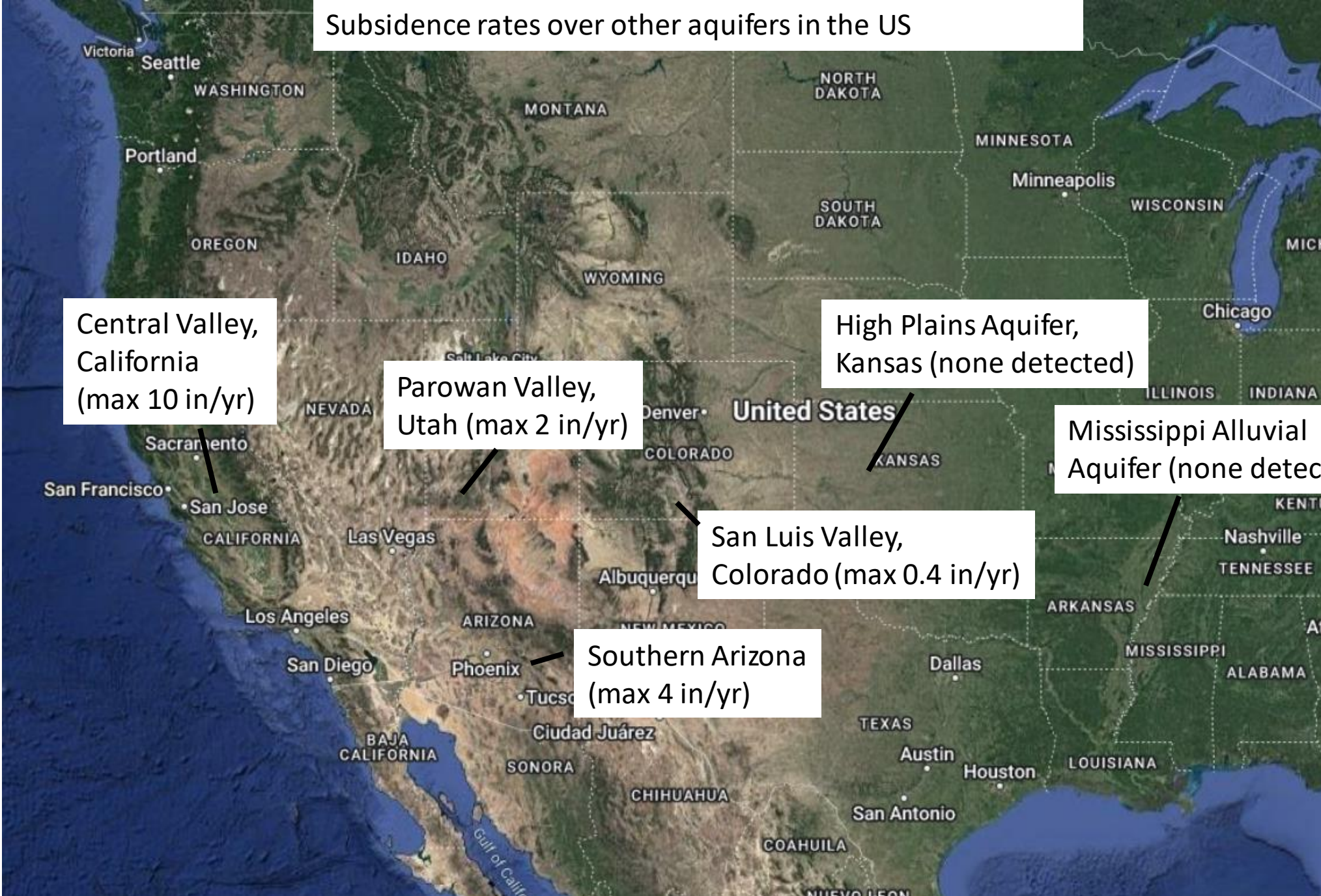


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# Subsidence rates over other aquifers in the US



Central Valley,  
California  
(max 10 in/yr)

Parowan Valley,  
Utah (max 2 in/yr)

High Plains Aquifer,  
Kansas (none detected)

Mississippi Alluvial  
Aquifer (none detected)

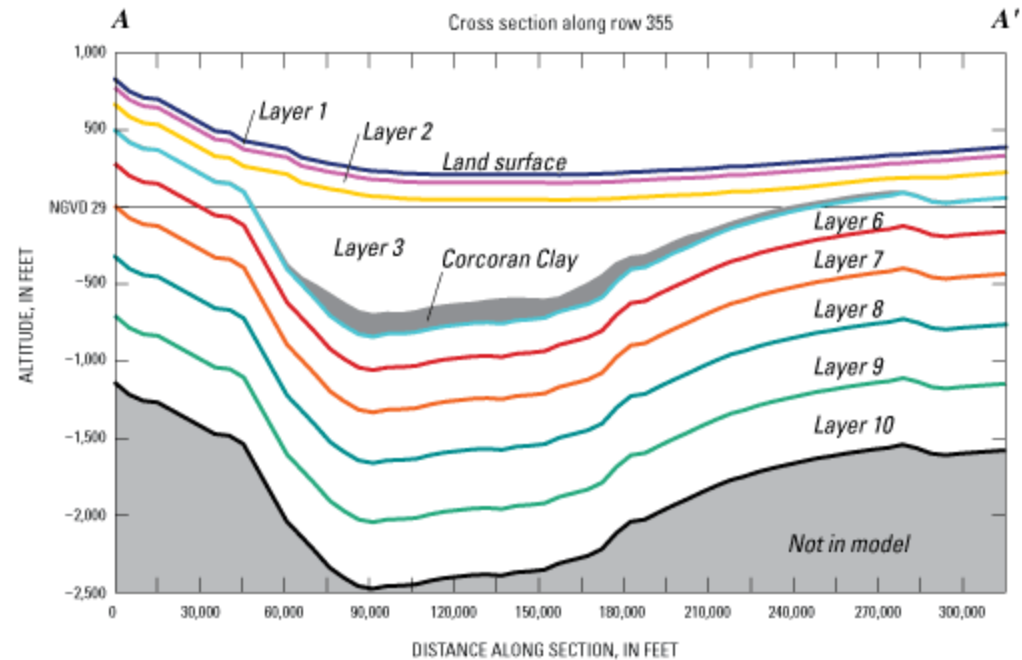
San Luis Valley,  
Colorado (max 0.4 in/yr)

Southern Arizona  
(max 4 in/yr)

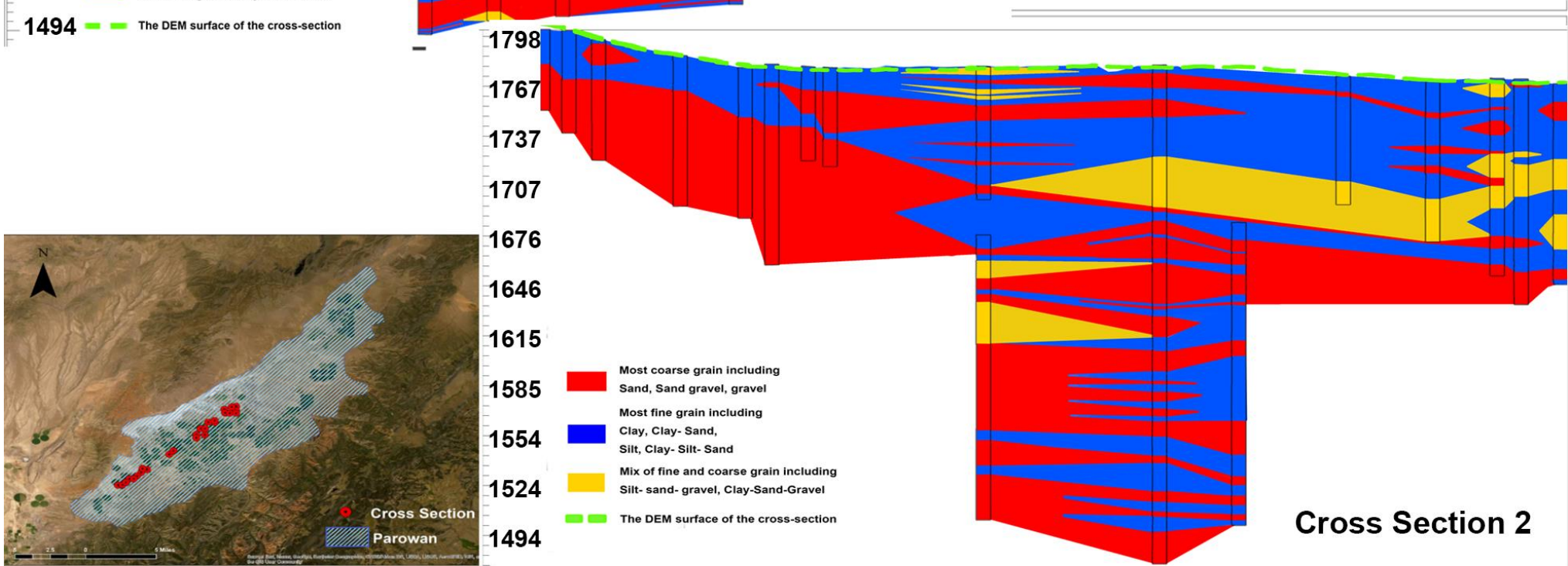
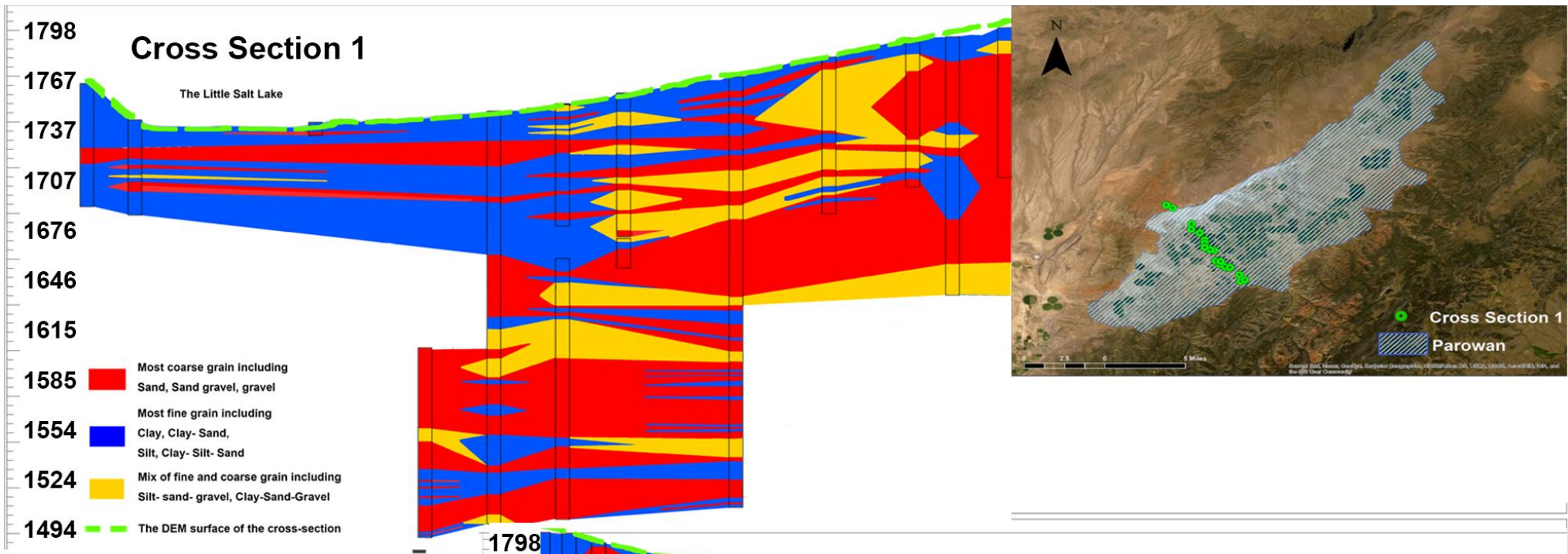
# What causes subsidence?

- Three things are needed:
  - Groundwater pumping
  - Significant clay in the aquifer system being pumped
  - Confining unit

Example in California: the Corcoran Clay confines the deep aquifers, causing pressurization

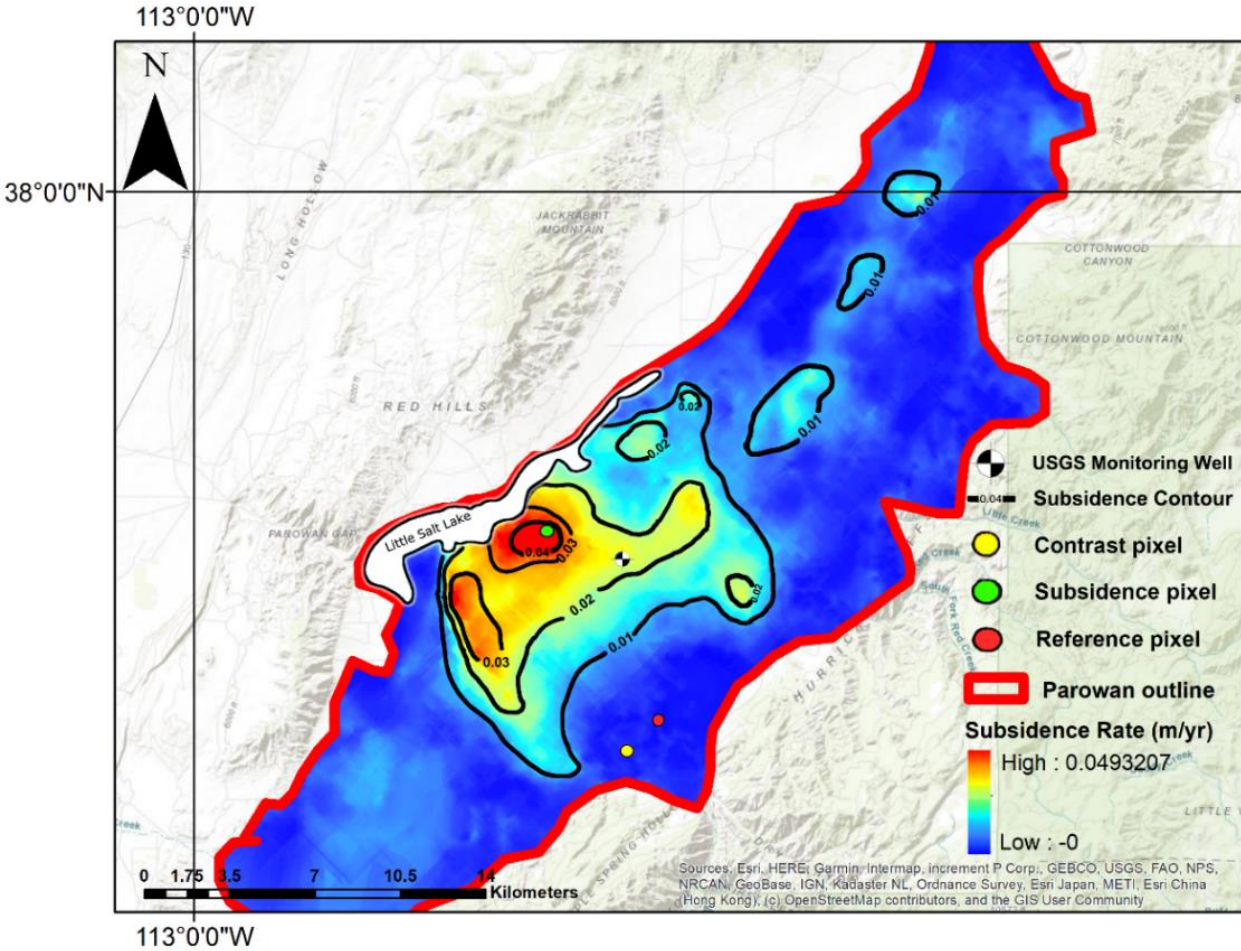




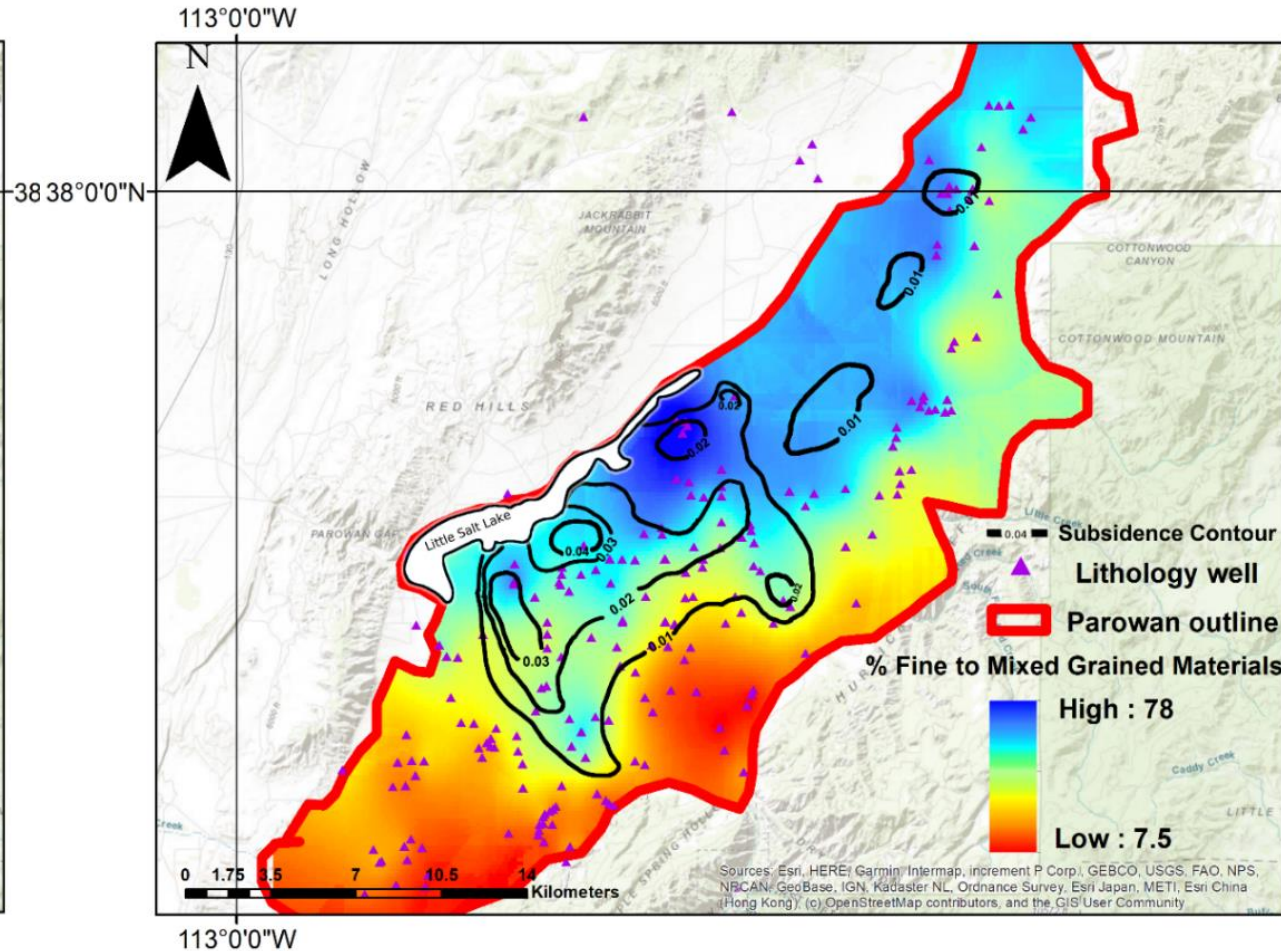




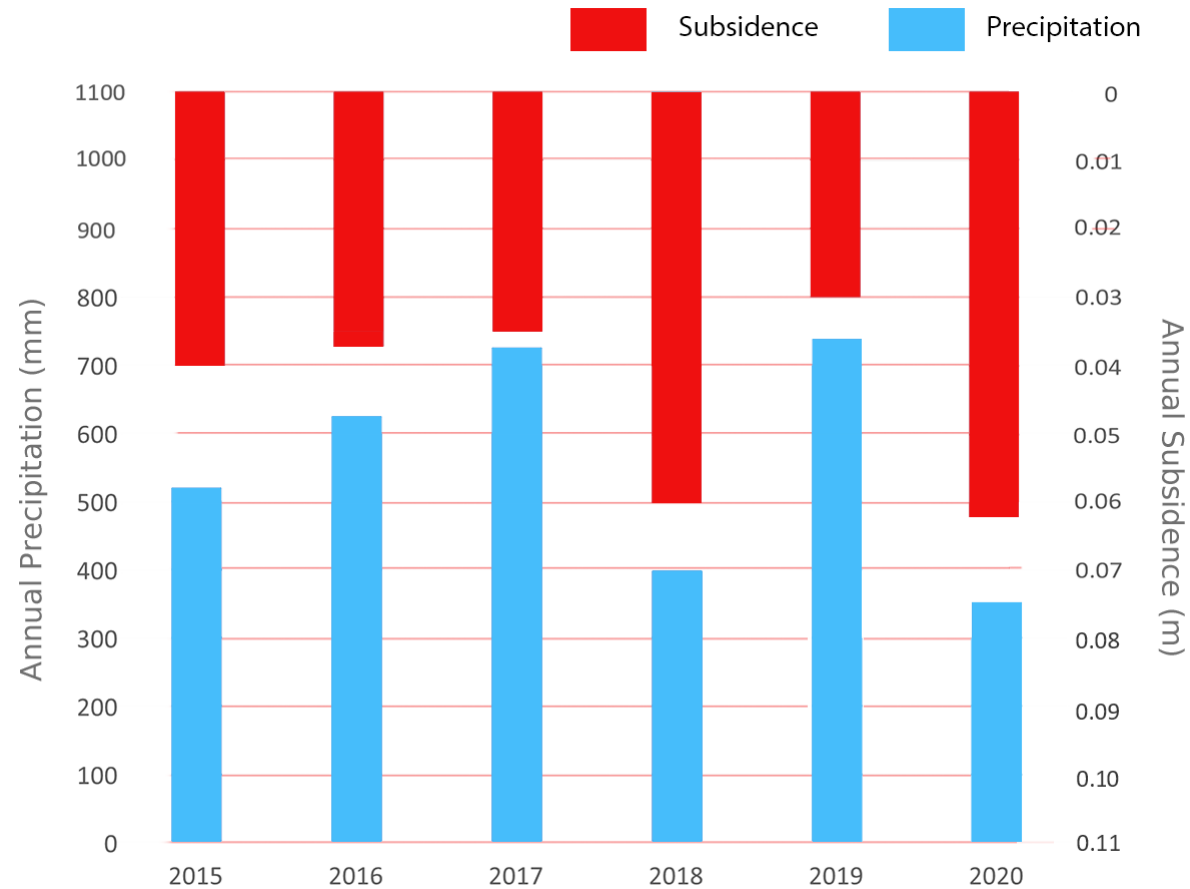
Subsidence rate mapped by satellites



% clay or other fines with subsidence overlain



# Yearly subsidence compared with precipitation



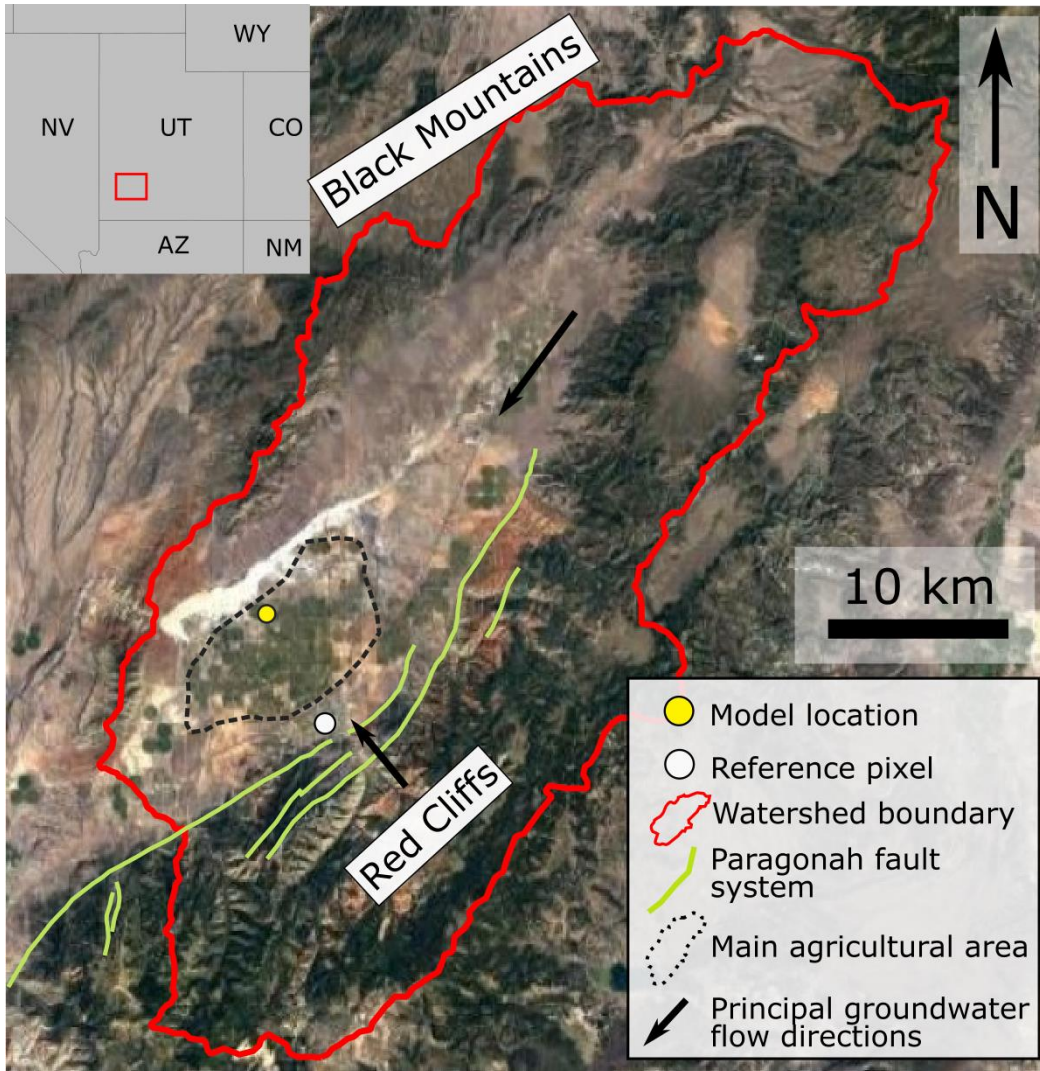
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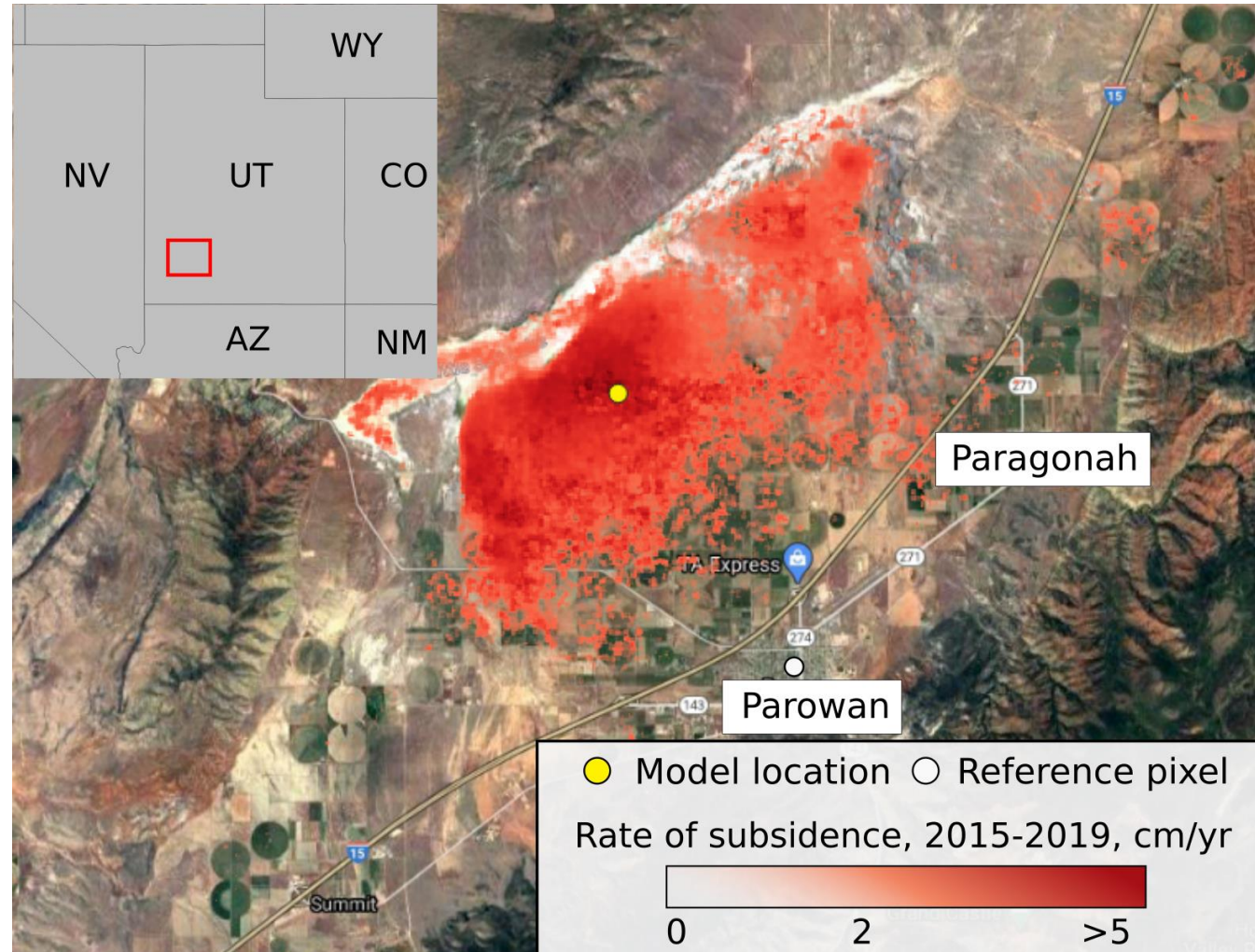


# Modeling ground deformation with satellite and groundwater level data (published)

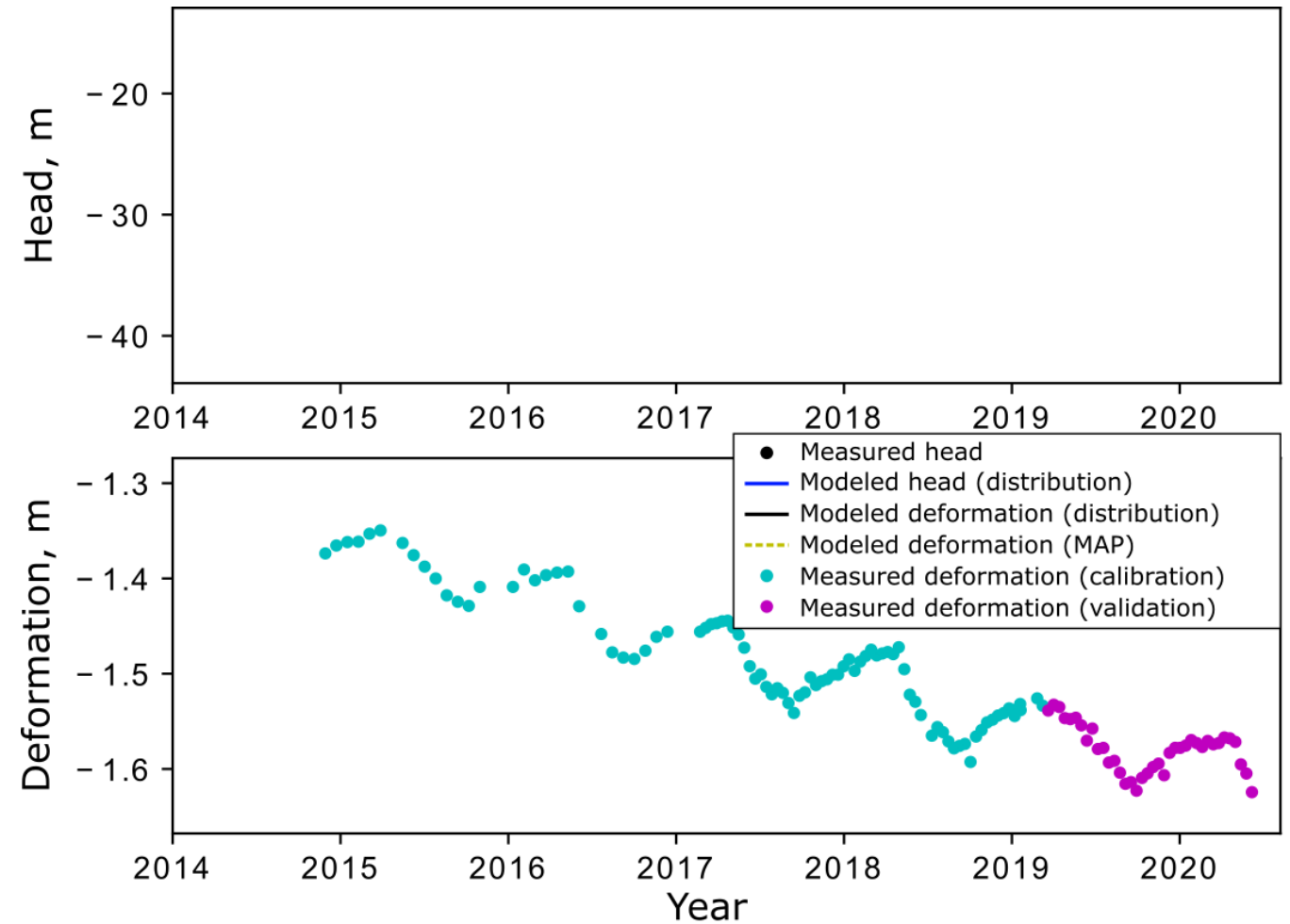




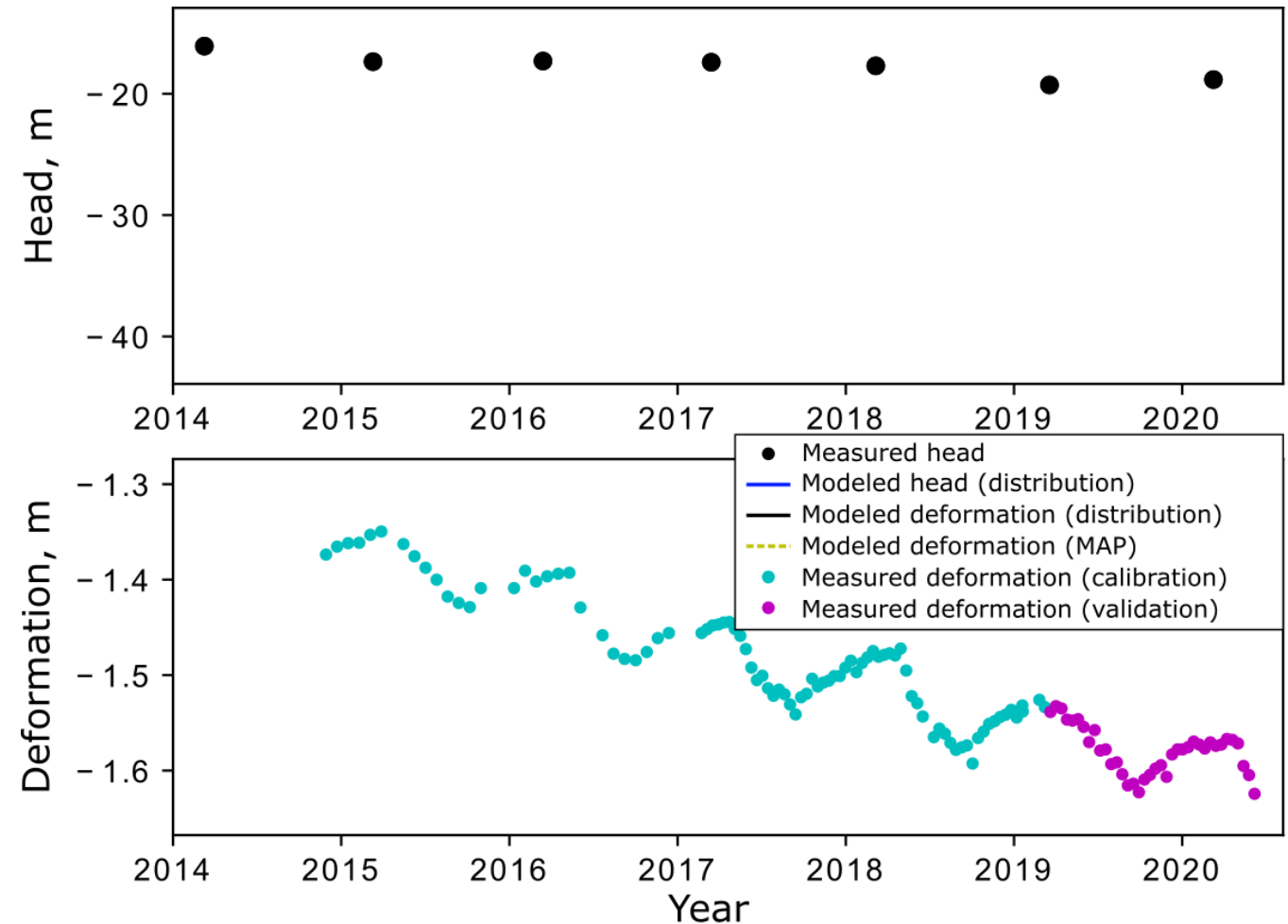
# Long-term subsidence rate in the southern part of Parowan Valley



# Subsidence and uplift over time at one location

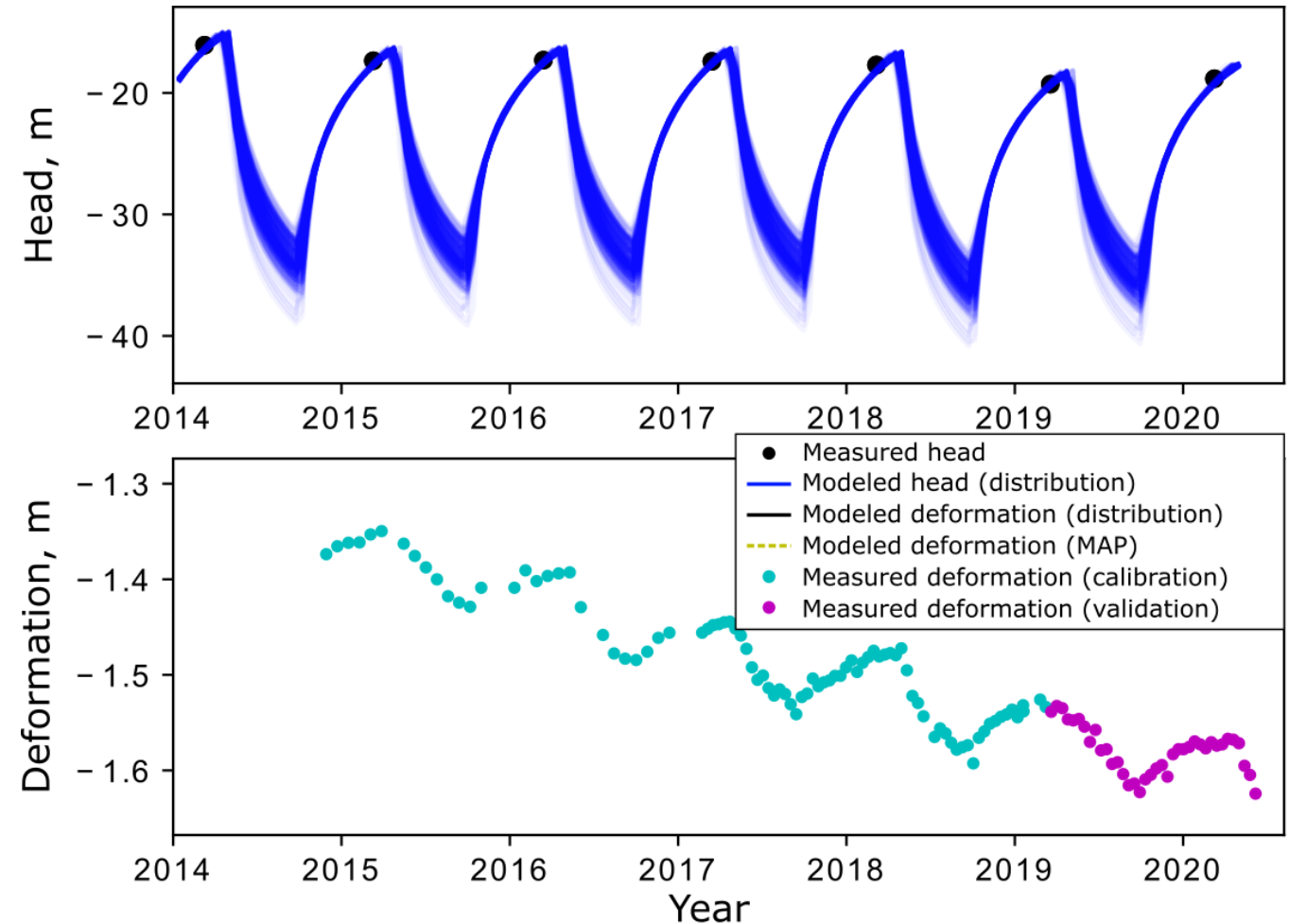


# Subsidence and uplift over time at one location is controlled by groundwater levels



# We simulated groundwater levels to model deformation

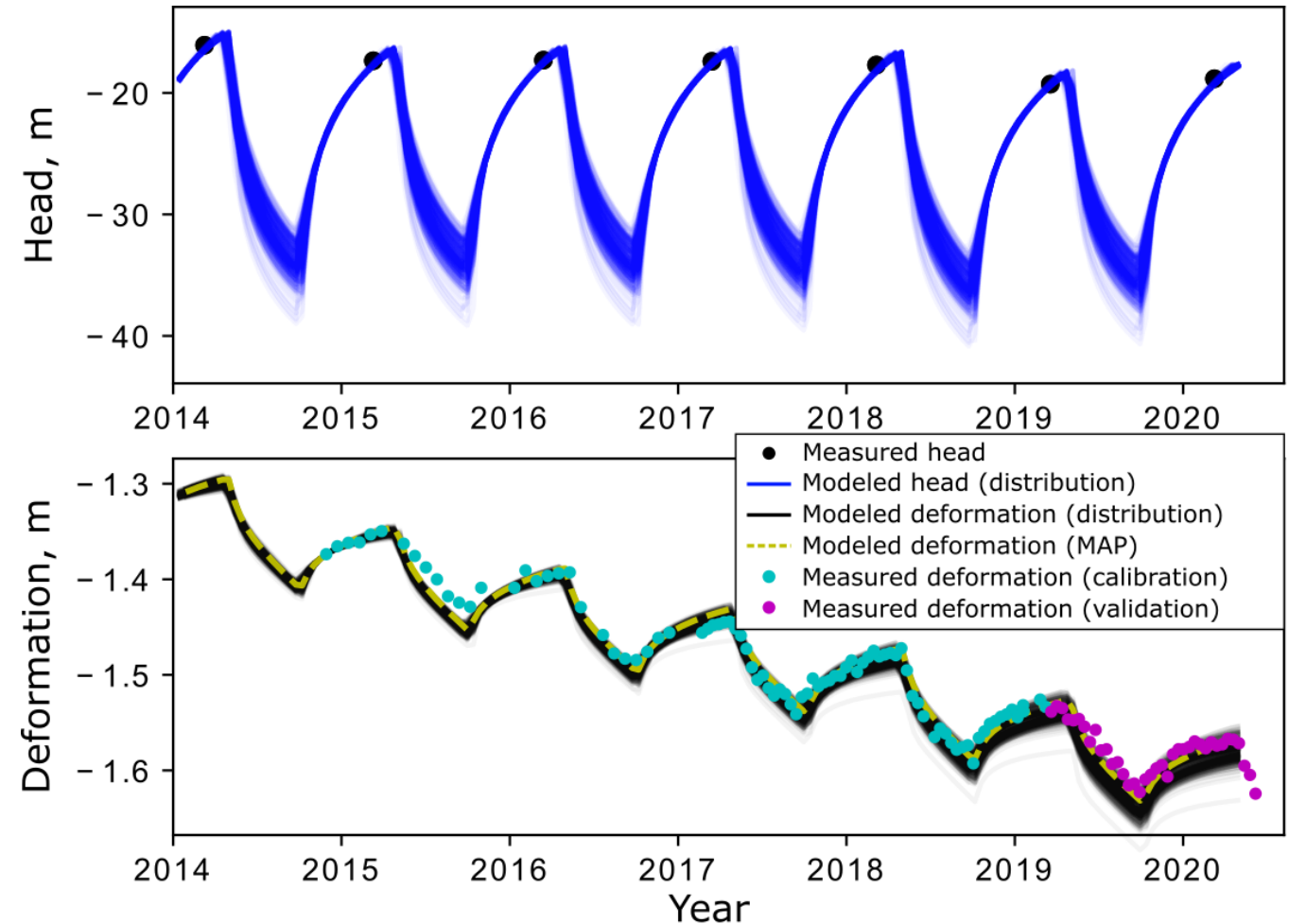
- This model can predict subsidence based on changes in groundwater level (head)
- The model can also estimate the elastic (recoverable) and inelastic (permanent) portions of subsidence



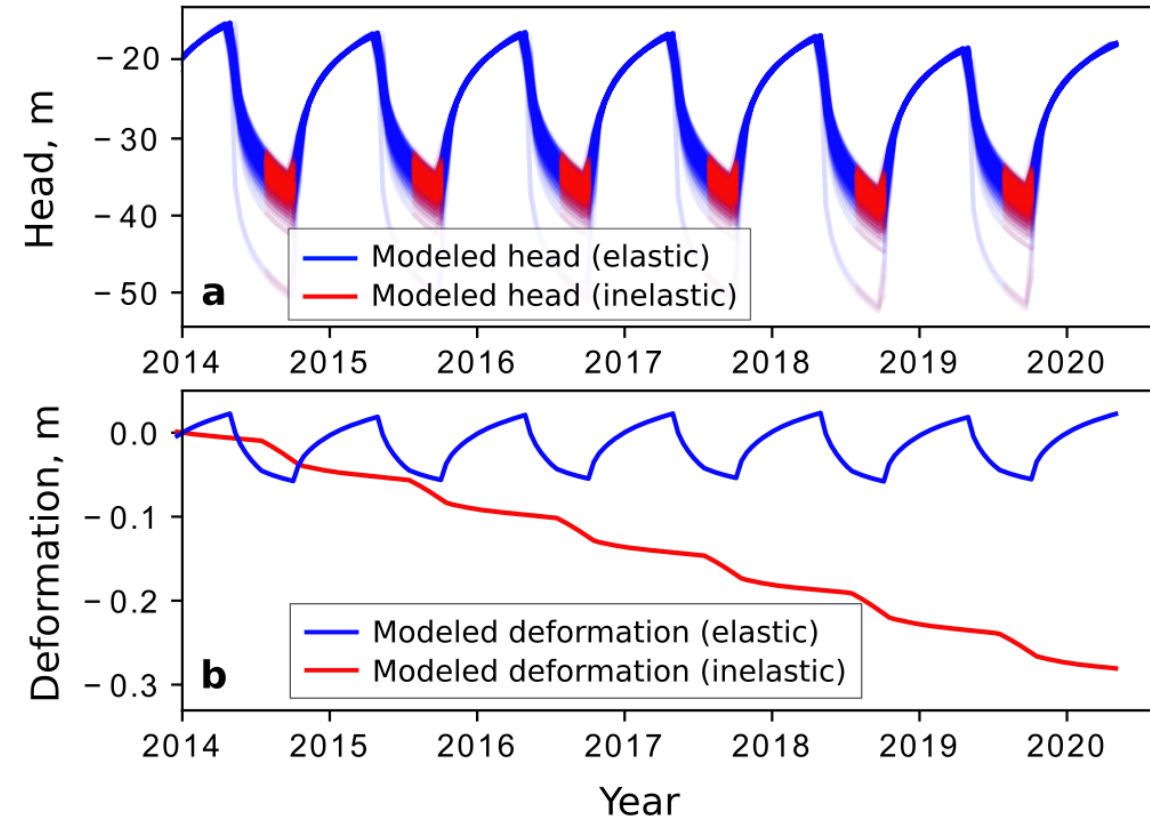
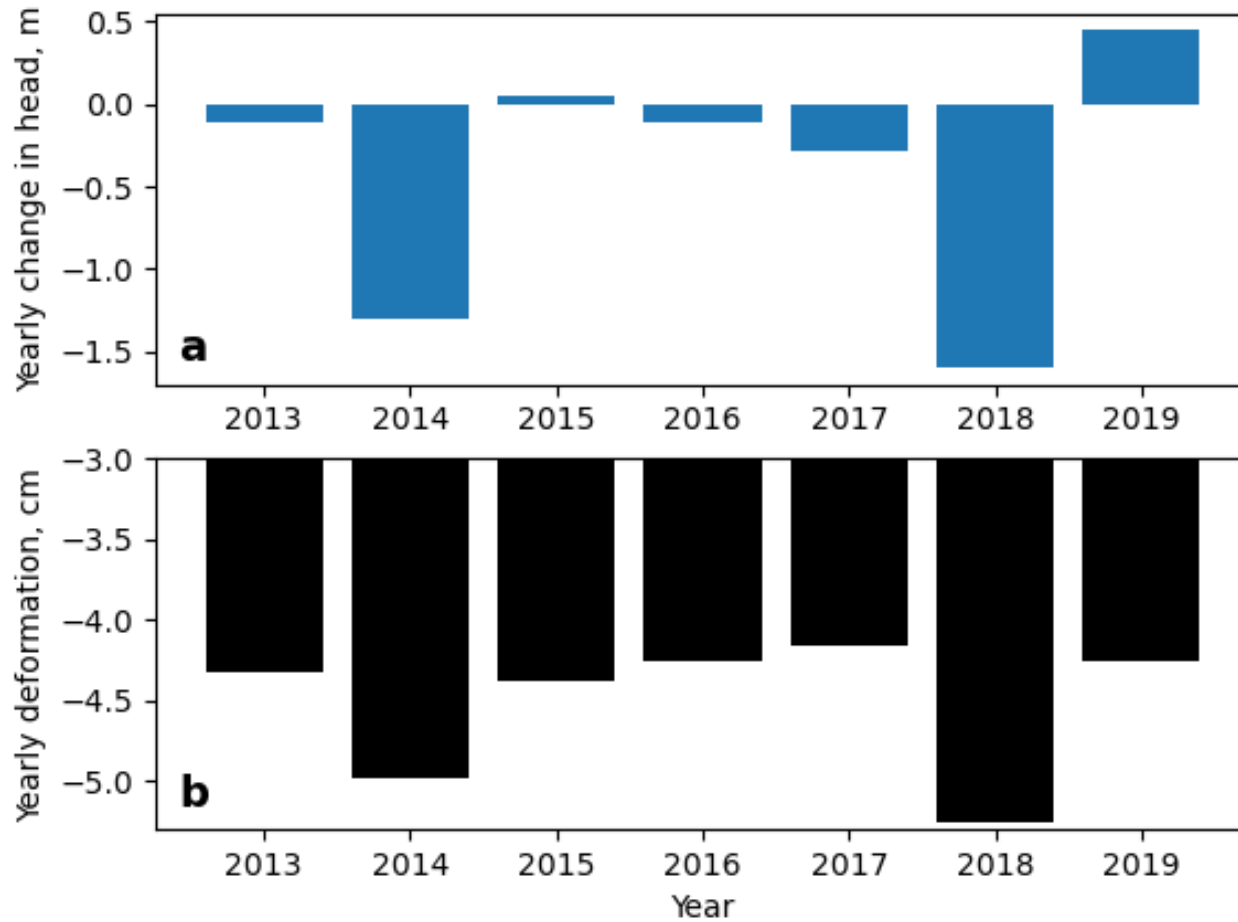


# Our model accurately predicts subsidence and rebound

- This model can predict subsidence based on changes in groundwater level (head)
- The model can also estimate the elastic (recoverable) and inelastic (permanent) portions of subsidence



# Most of the long-term subsidence is permanent (inelastic)

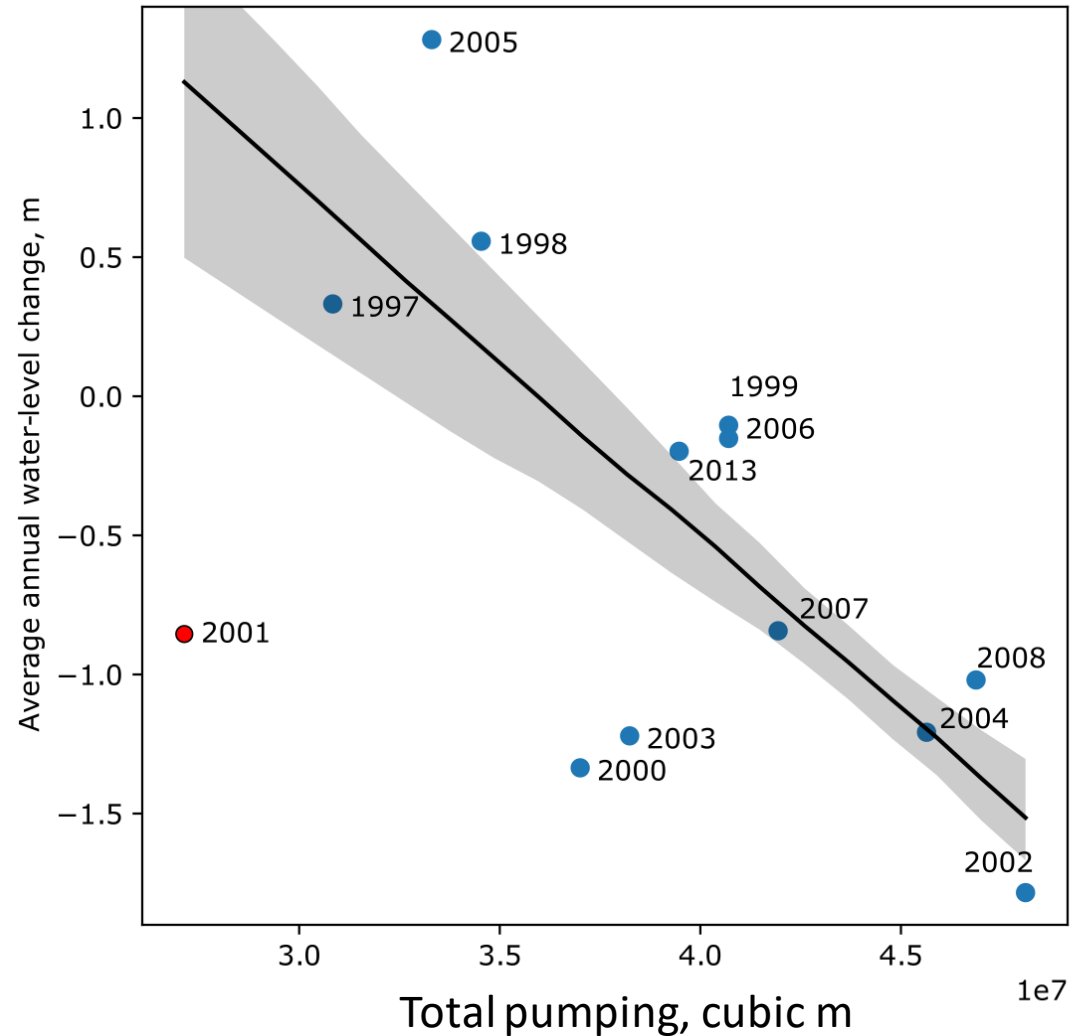


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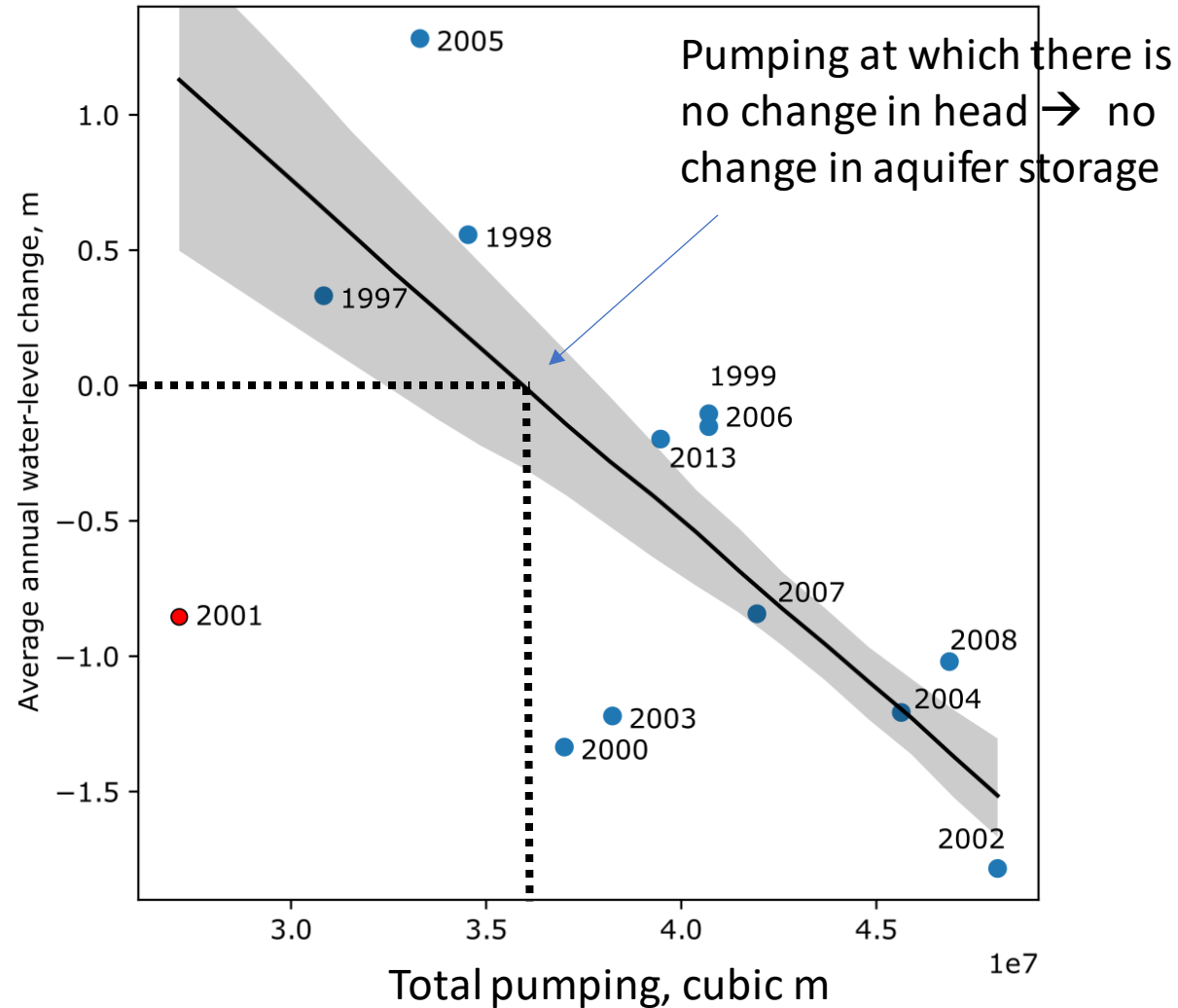
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# Water budget approach: compare total annual pumping with average annual change in head

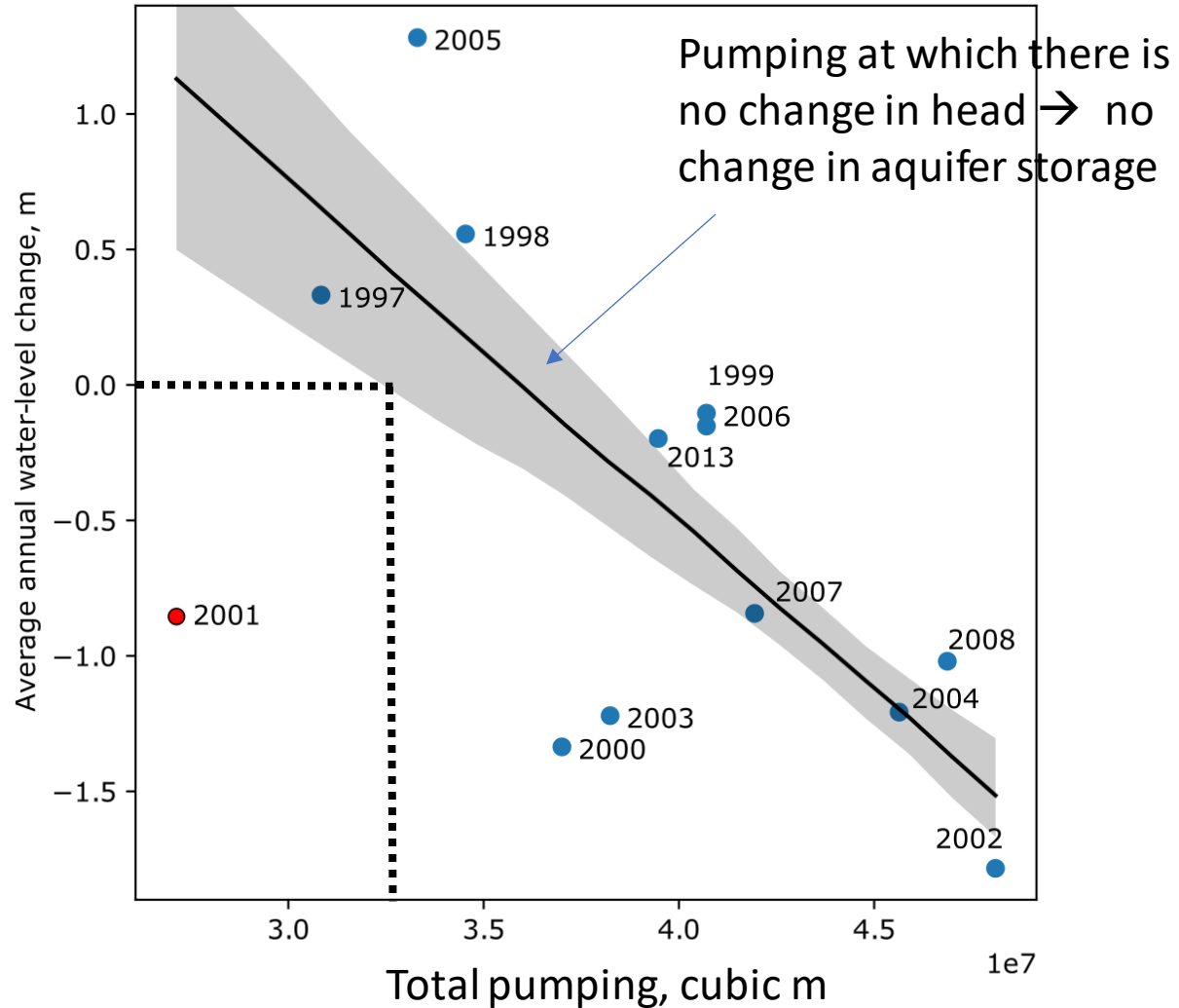




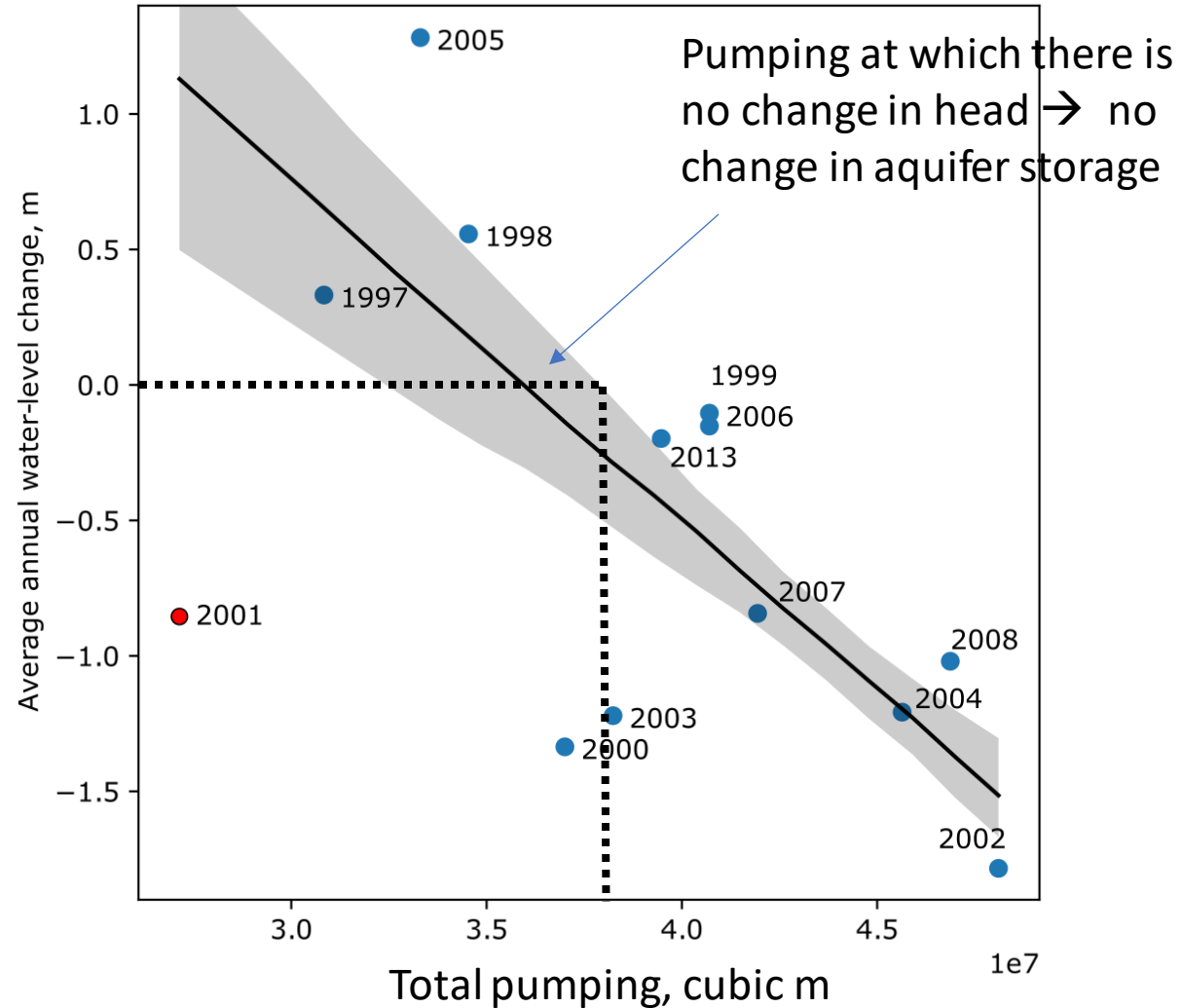
# Water budget approach: compare total annual pumping with average annual change in head



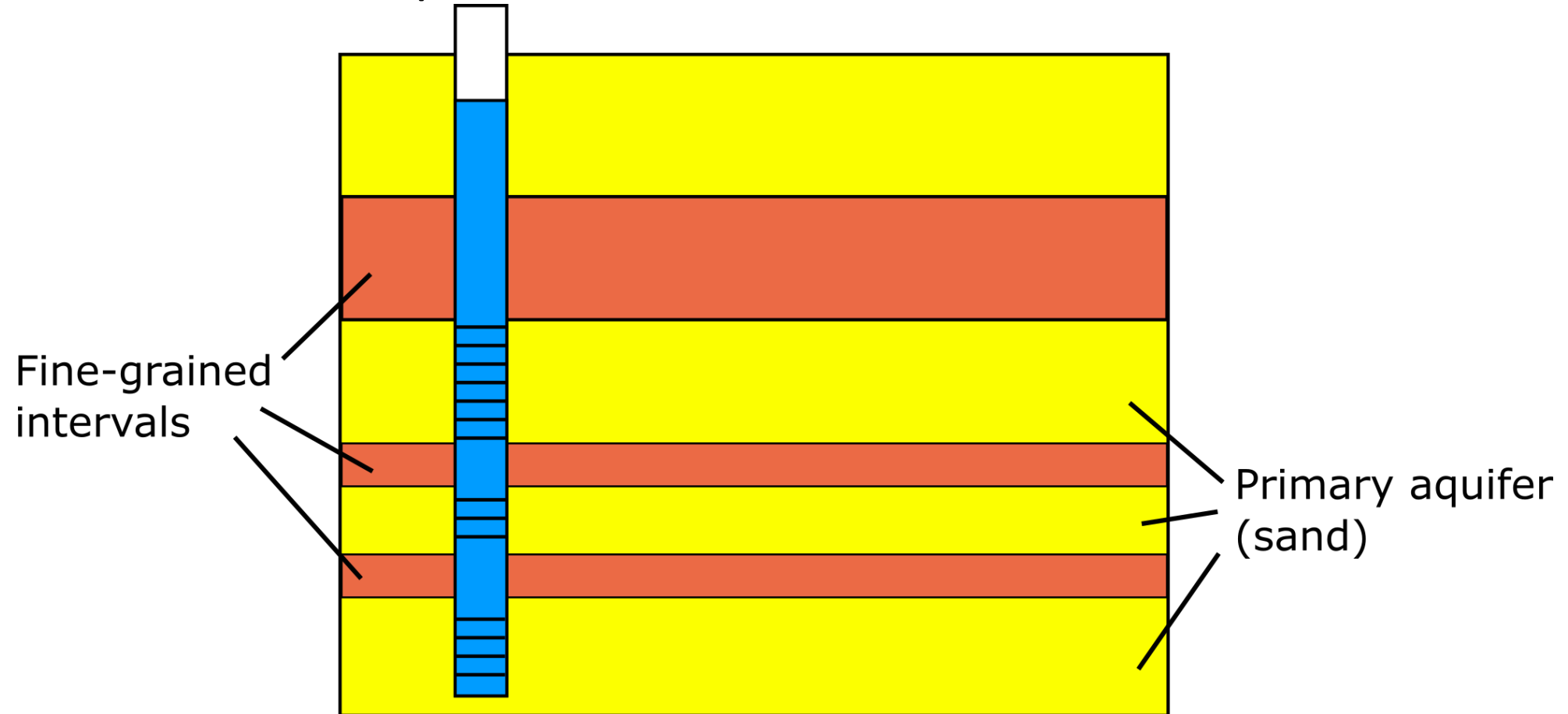
# There's some uncertainty in this value!



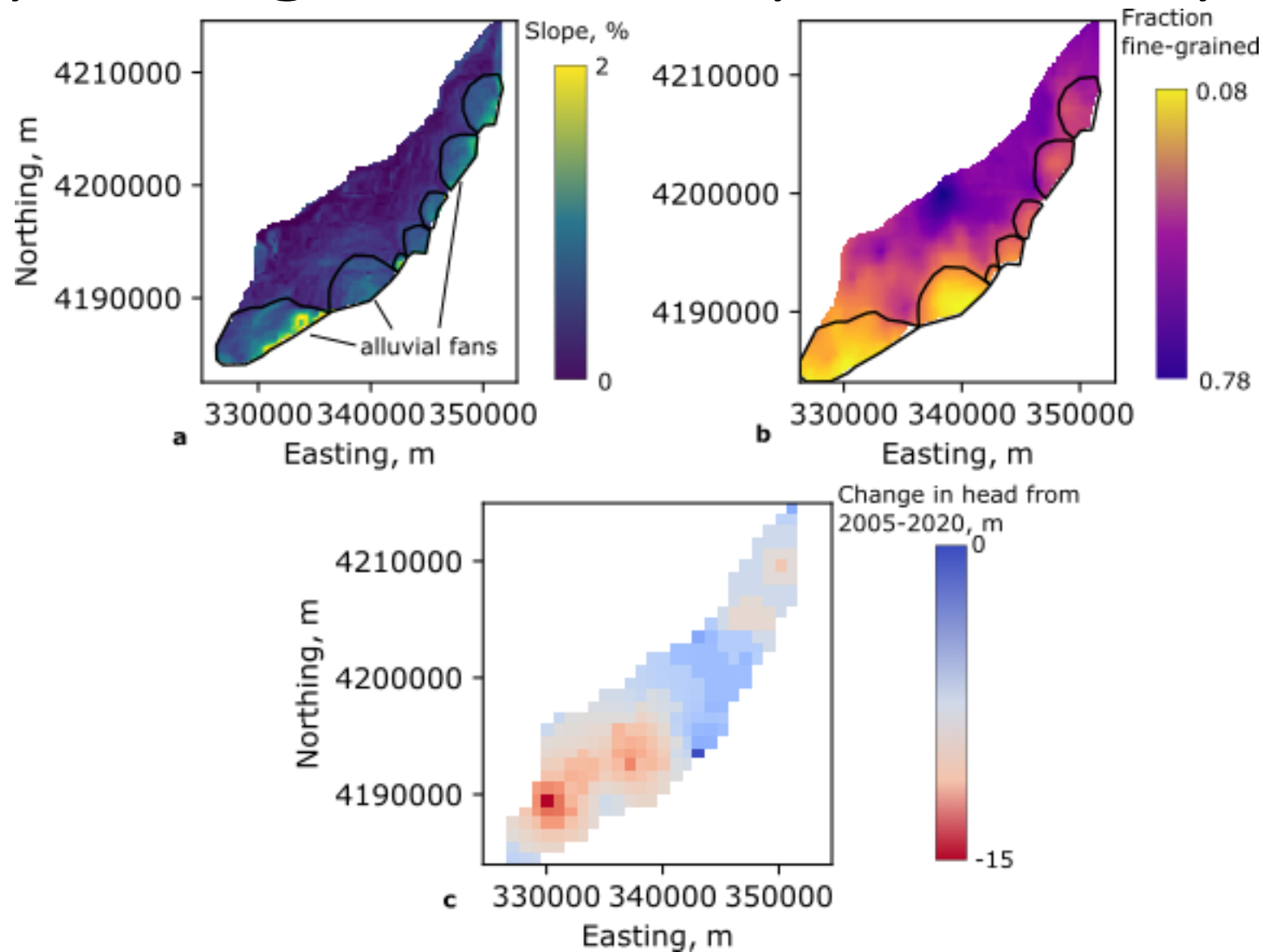
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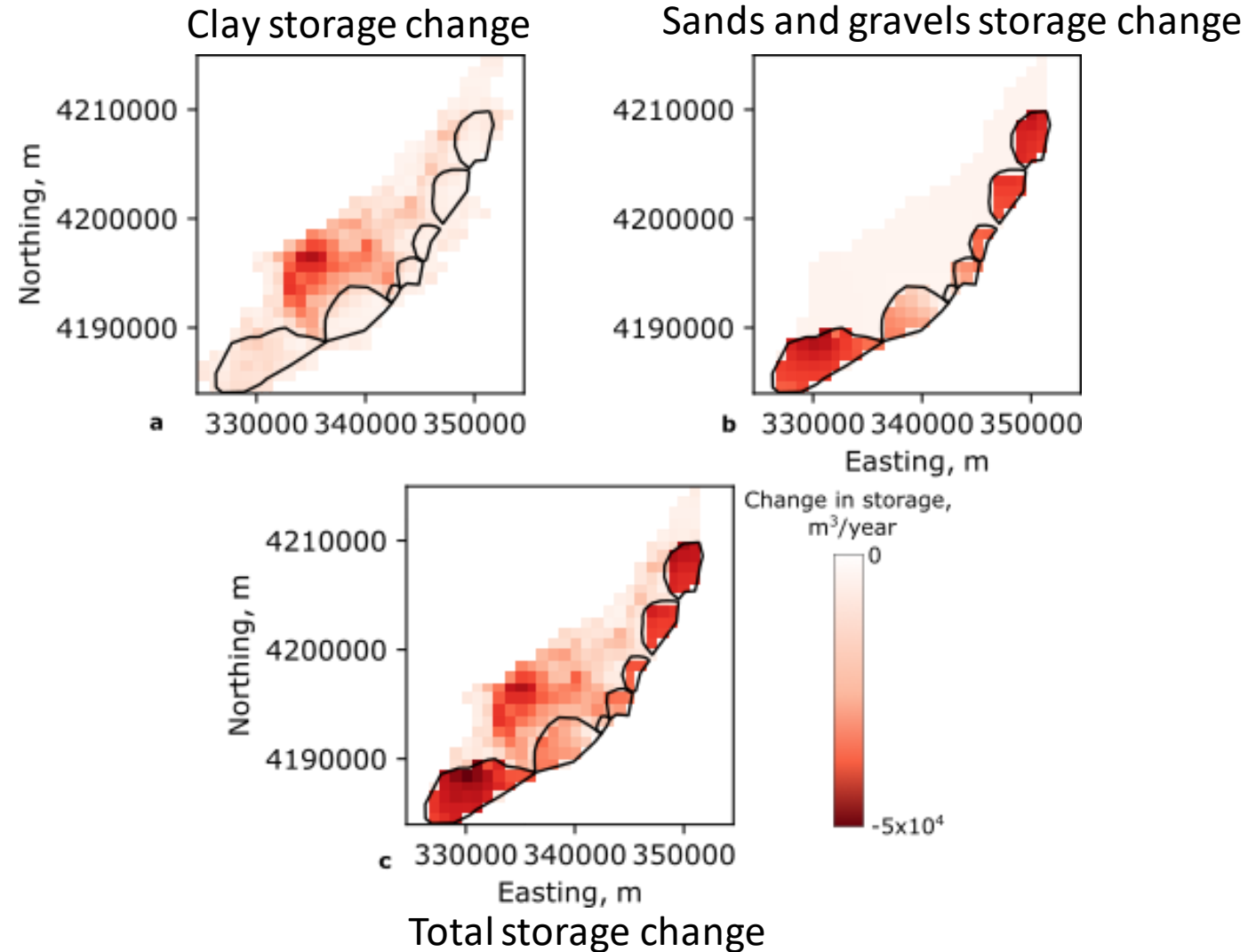
We also know that there is some storage coming into the aquifer from clays that will go away once drawdown stops



Another approach for storage change:  
multiply change in head by storativity

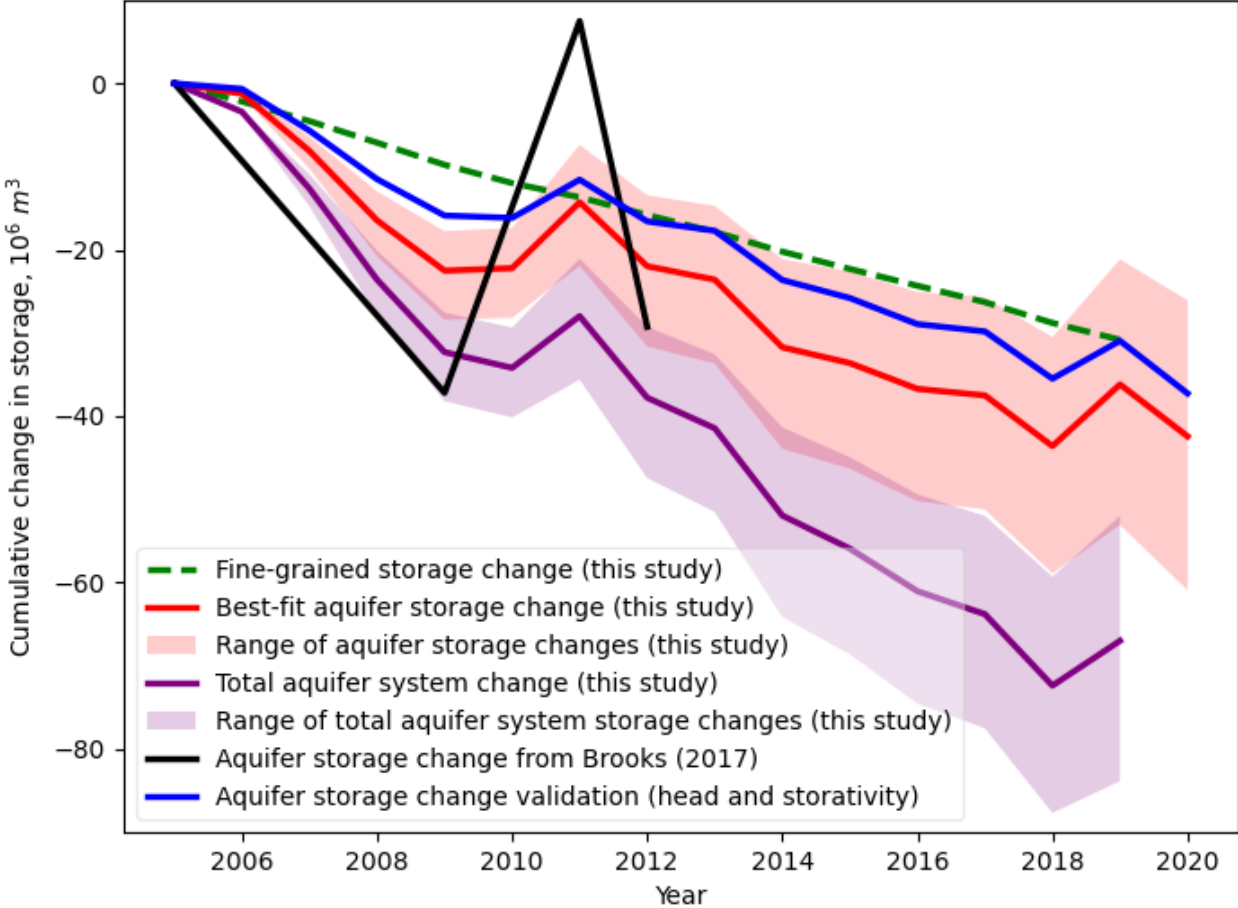


We can then estimate storage change in both the aquifer (sands and gravels) and the clays



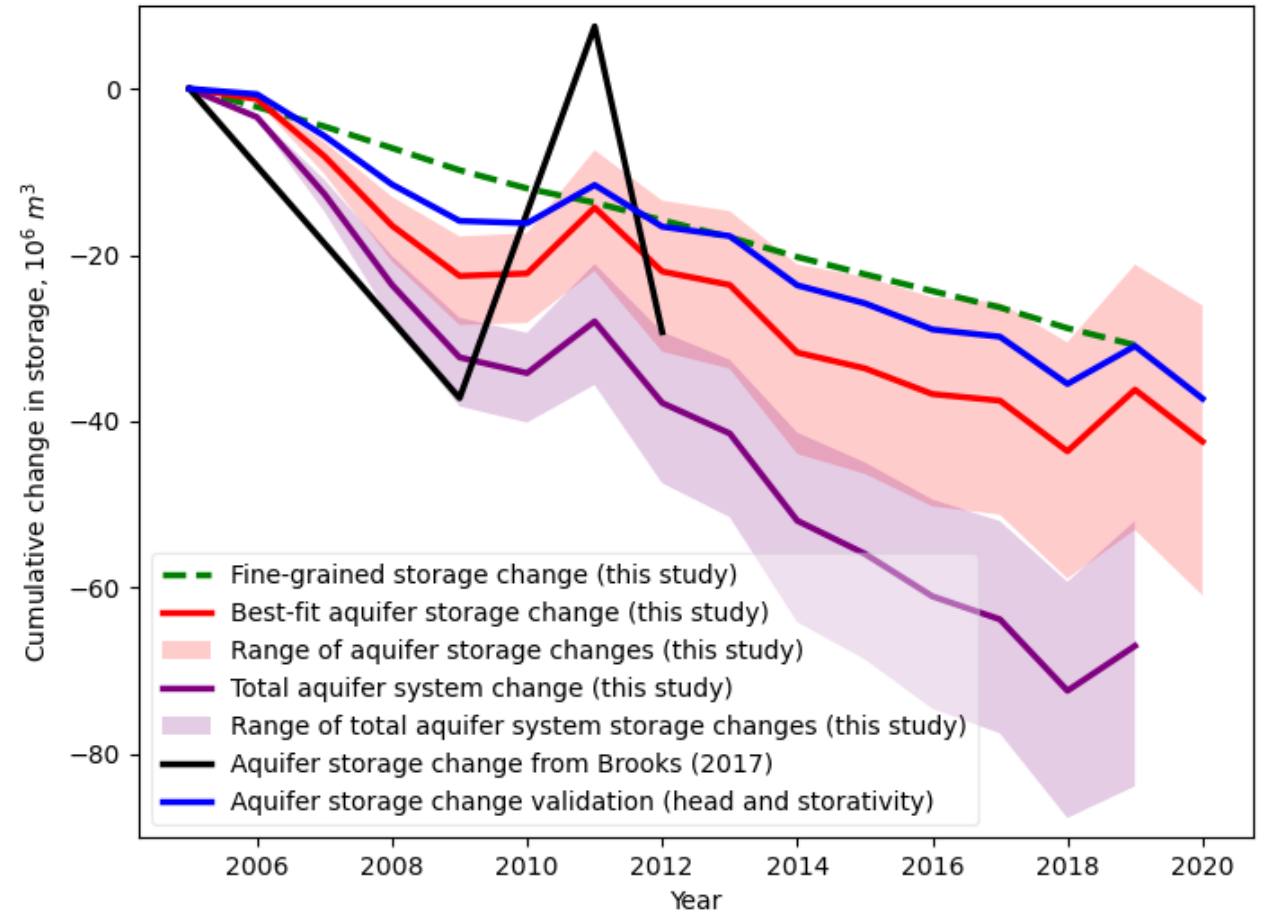


# Storage change estimates from this study, compared with USGS groundwater model (Brooks 2017, black line)



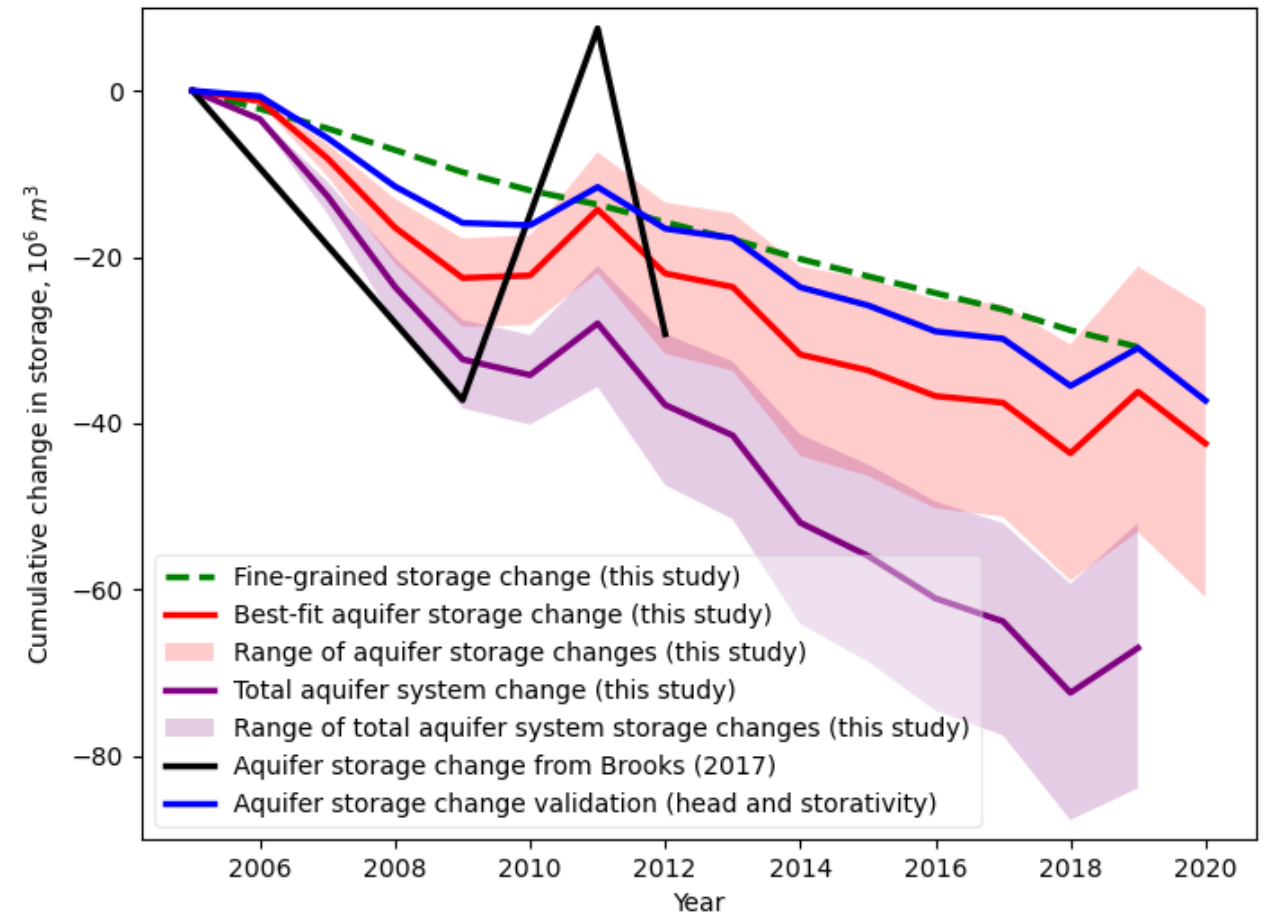
# Lots of uncertainty in storage change estimates!

Average annual storage loss, 2005-2012 (acre-feet)				Average annual pumping, 2005-2012 (acre-feet)
Brooks (2017)	Marston (2017)	This study, range ( <b>best estimate</b> )	Validation, this study	
3,388	10,863	5,493 -3,386 ( <b>4,376</b> )	1,921	34,125



# Lots of uncertainty in storage change estimates!

- Each of these storage change estimates is much lower than that of Marston (2017), who used a water budget approach
- In water budgets, recharge is very difficult to estimate
- Possible explanations for discrepancy:
  - Marston (2017) under-estimated recharge
  - Mountain aquifers are being depleted and considered a source of 'inflow' in our water balance approach



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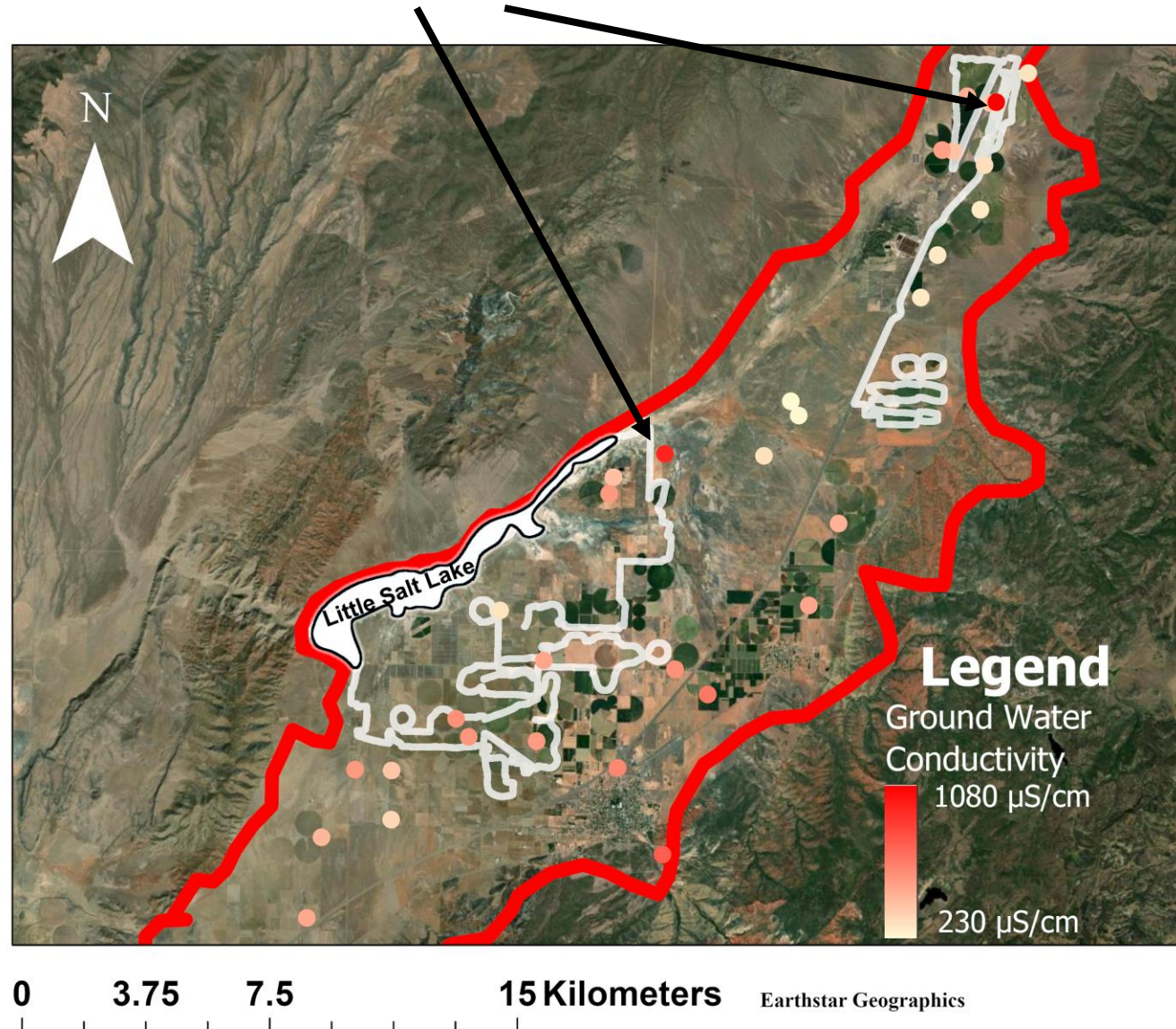
# Geophysical survey – thanks so much to the community for helping us do this!

- November 2021
- Conducted with a towed Time-domain ElectroMagnetic system (tTEM)
- Towed behind an ATV at ~8 mph
- Images resistivity from the surface to a depth of ~150 ft
- This can be used to identify aquifers (sands/gravels) and aquitards (clays and other fines) in the subsurface
- Resistivity of common materials
  - Sand: 40-200 ohm m
  - Clay: 5-20 ohm m
  - Freshwater: >6 ohm m
  - Brackish water: 0.6-10 ohm m



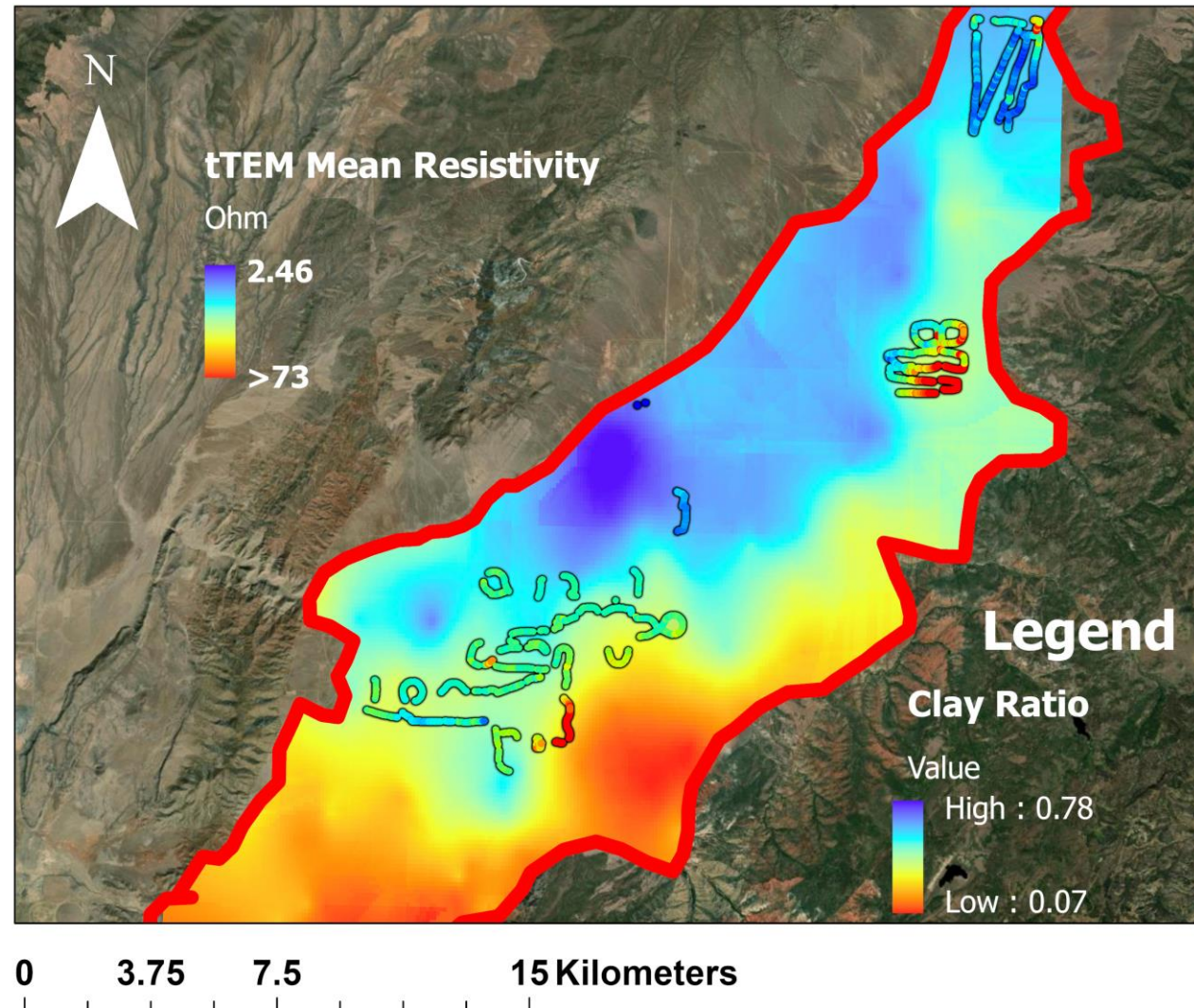


It appears that the water conductivity is fairly consistent except for one spot in the north, and one near Little Salt Lake





# Mean resistivity from tTEM compared with % clay from drillers' logs



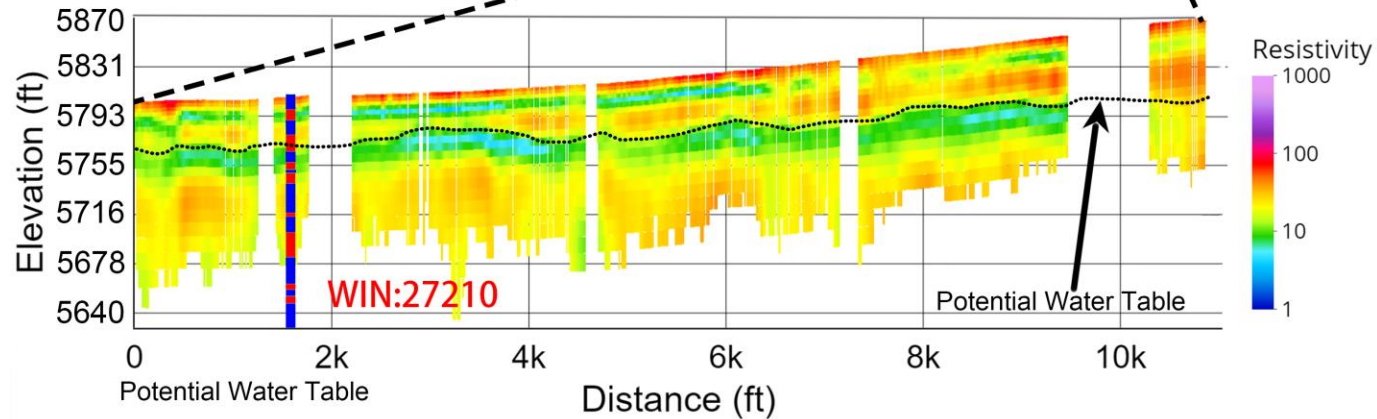
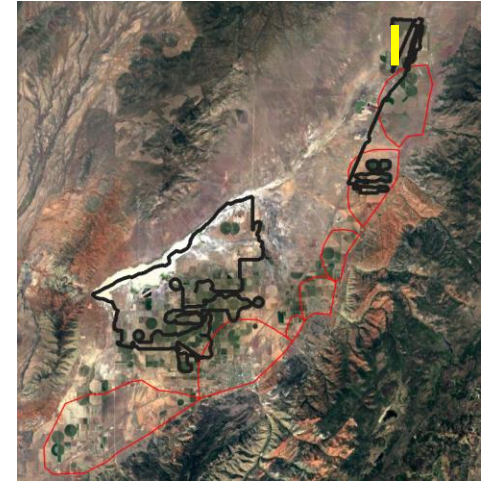
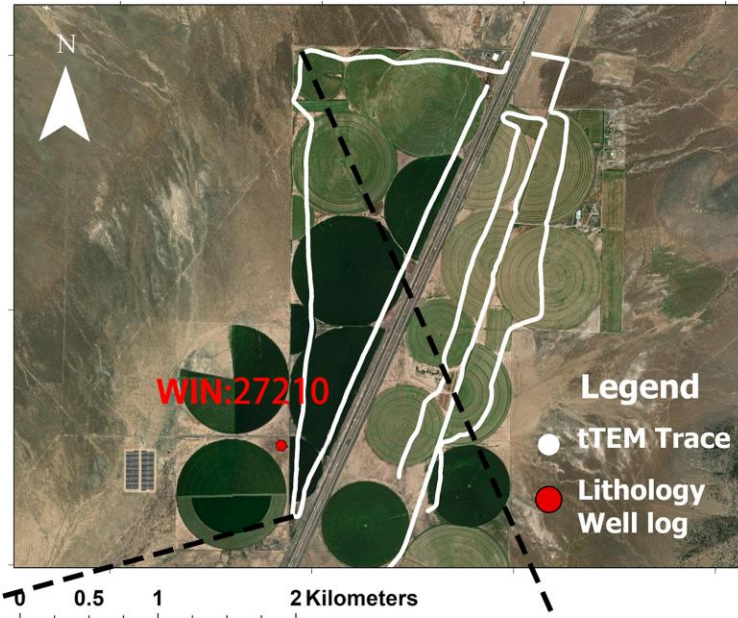
# tTEM Trace 110 Northern Parowan Valley Nov 2021

## Lithology Classification for Well Log

**Red** Most coarse grain including Sand, Sand-Gravel, Gravel

**Blue** Most fine grain including Clay, Clay-Sand, Silt, Clay-Silt-Sand

**Yellow** Mix of fine and coarse grain including Silt-Sand-Gravel, Clay-Sand-Gravel

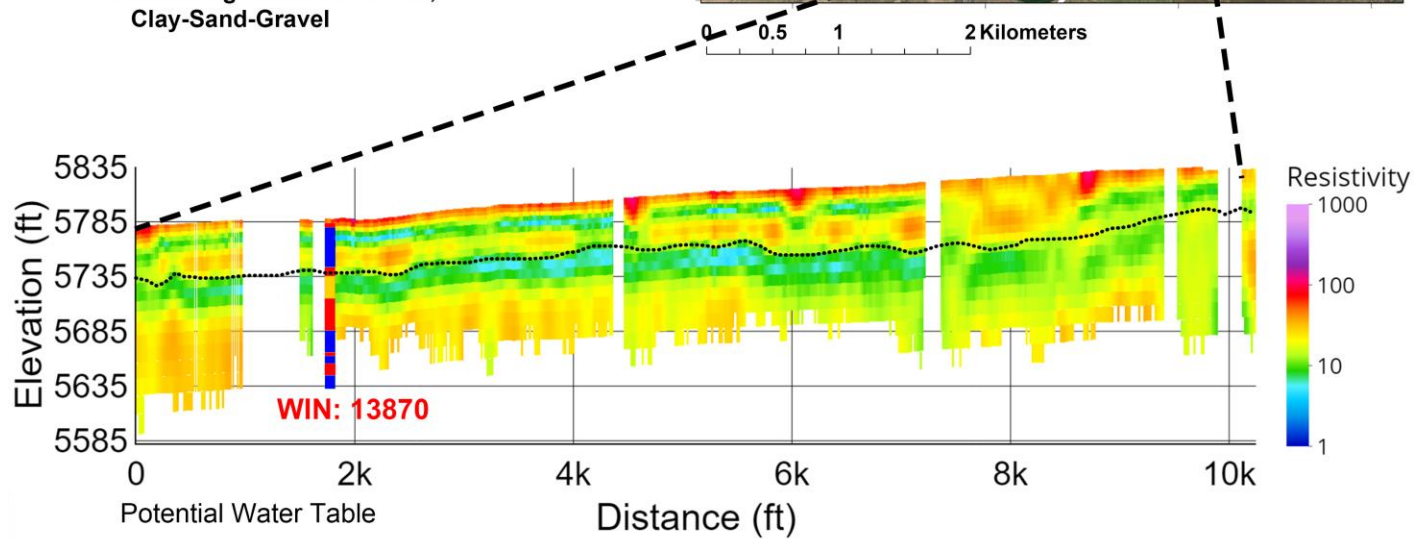
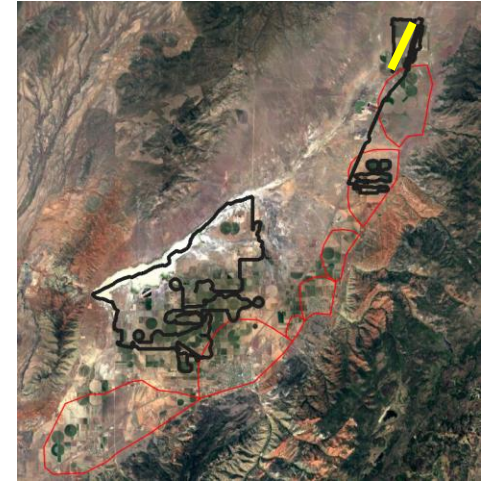
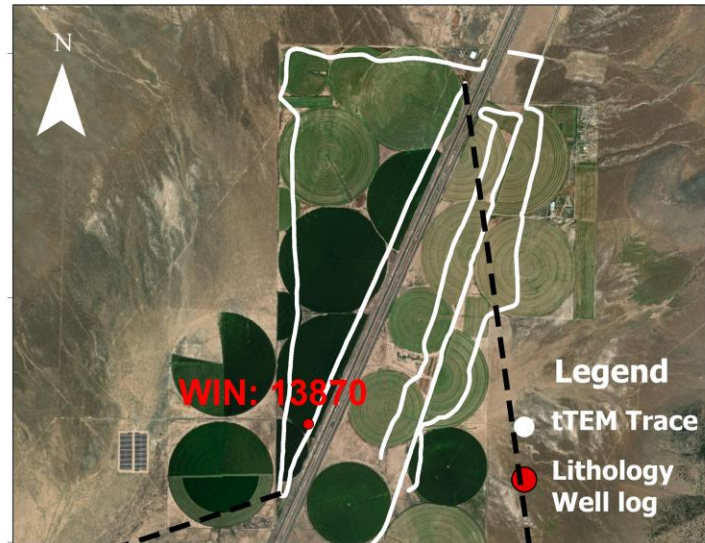




# tTEM Trace 100 Northern Parowan Valley Nov 2021

## Lithology Classification for Well Log

- Most coarse grain including Sand, Sand-Gravel, Gravel
- Most fine grain including Clay, Clay-Sand, Silt, Clay-Silt-Sand
- Mix of fine and coarse grain including Silt-Sand-Gravel, Clay-Sand-Gravel



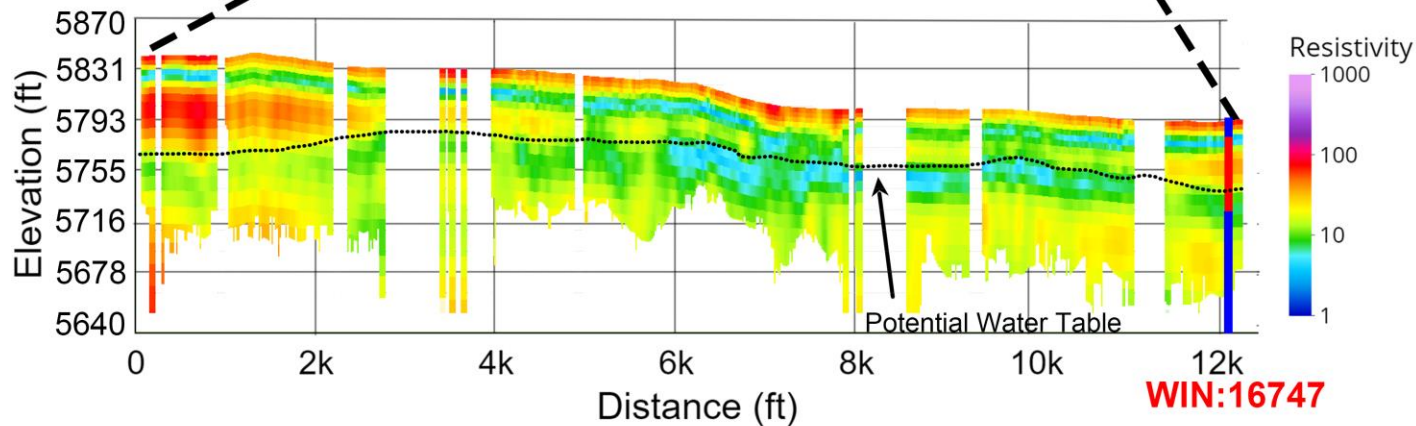
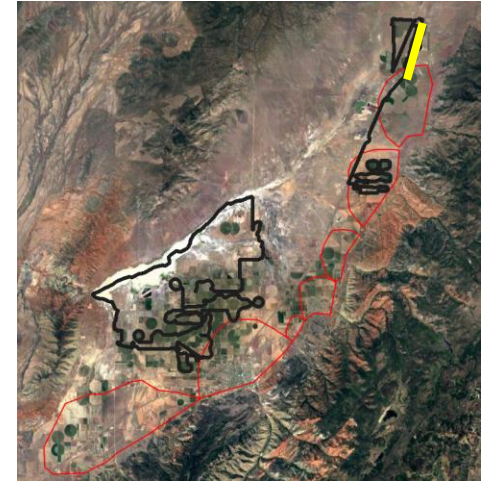
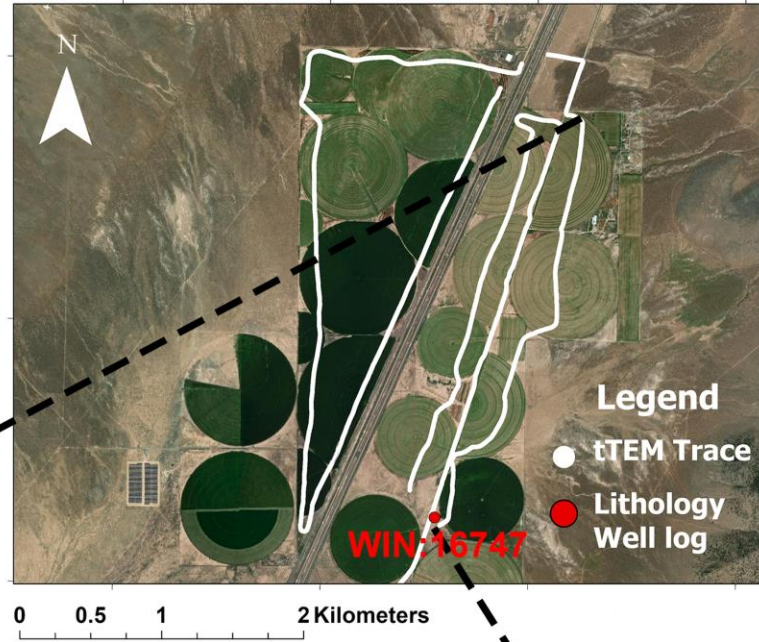
# tTEM Trace 120 Northern Parowan Valley Nov 2021

## Lithology Classification for Well Log

**Red** Most coarse grain including Sand, Sand-Gravel, Gravel

**Blue** Most fine grain including Clay, Clay-Sand, Silt, Clay-Silt-Sand

**Yellow** Mix of fine and coarse grain including Silt-Sand-Gravel, Clay-Sand-Gravel





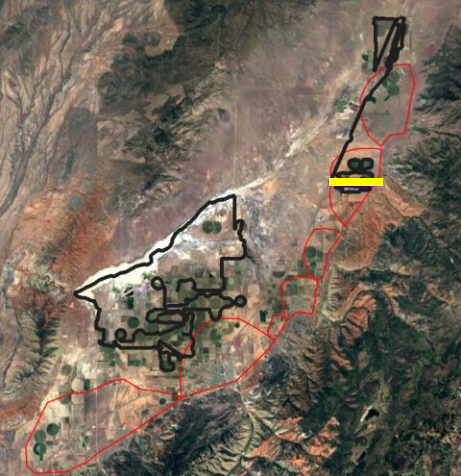
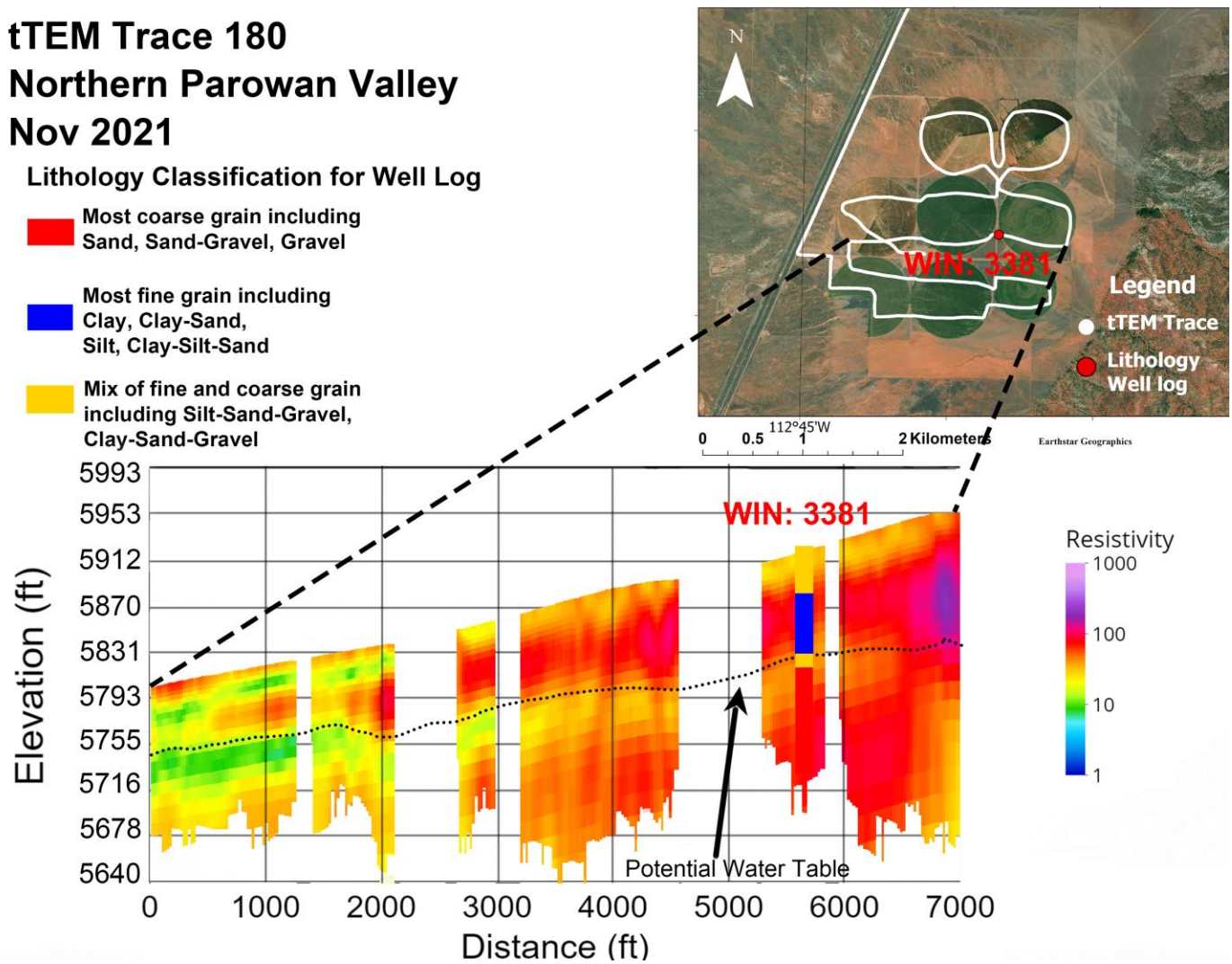
# tTEM Trace 180 Northern Parowan Valley Nov 2021

## Lithology Classification for Well Log

**Red** Most coarse grain including Sand, Sand-Gravel, Gravel

**Blue** Most fine grain including Clay, Clay-Sand, Silt, Clay-Silt-Sand

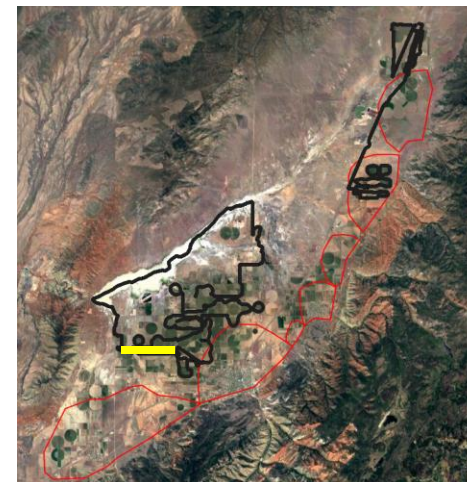
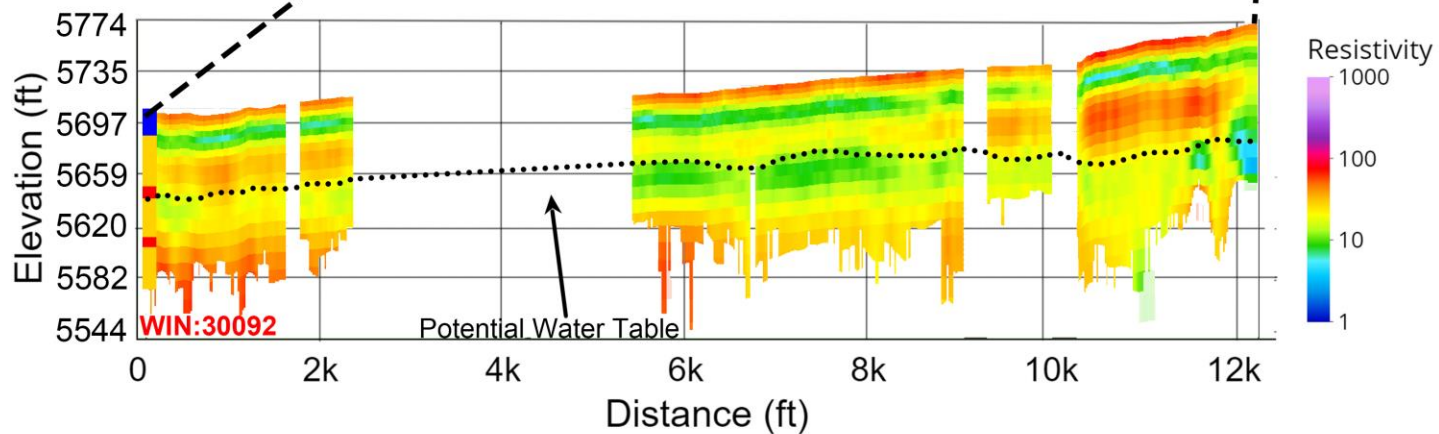
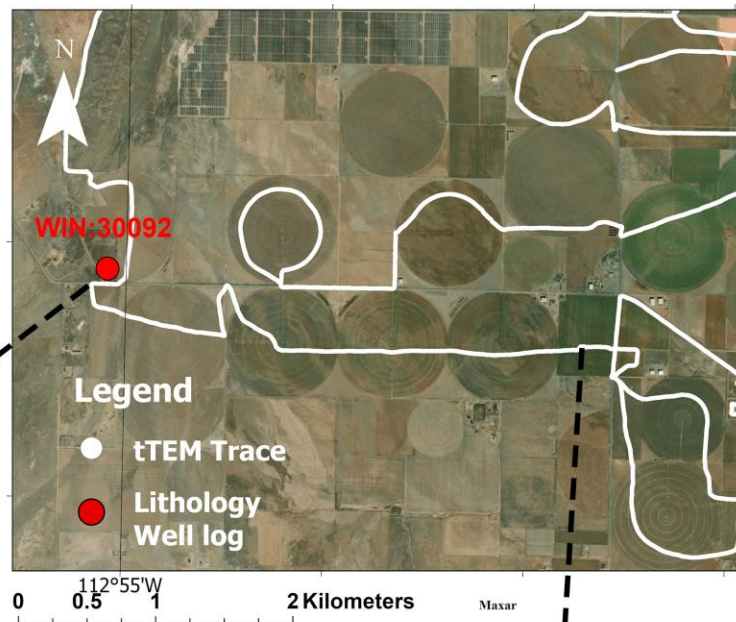
**Yellow** Mix of fine and coarse grain including Silt-Sand-Gravel, Clay-Sand-Gravel



# tTEM Trace 230 Central Parowan Valley Nov 2021

## Lithology Classification for Well Log

- Most coarse grain including Sand, Sand-Gravel, Gravel
- Most fine grain including Clay, Clay-Sand, Silt, Clay-Silt-Sand
- Mix of fine and coarse grain including Silt-Sand-Gravel, Clay-Sand-Gravel

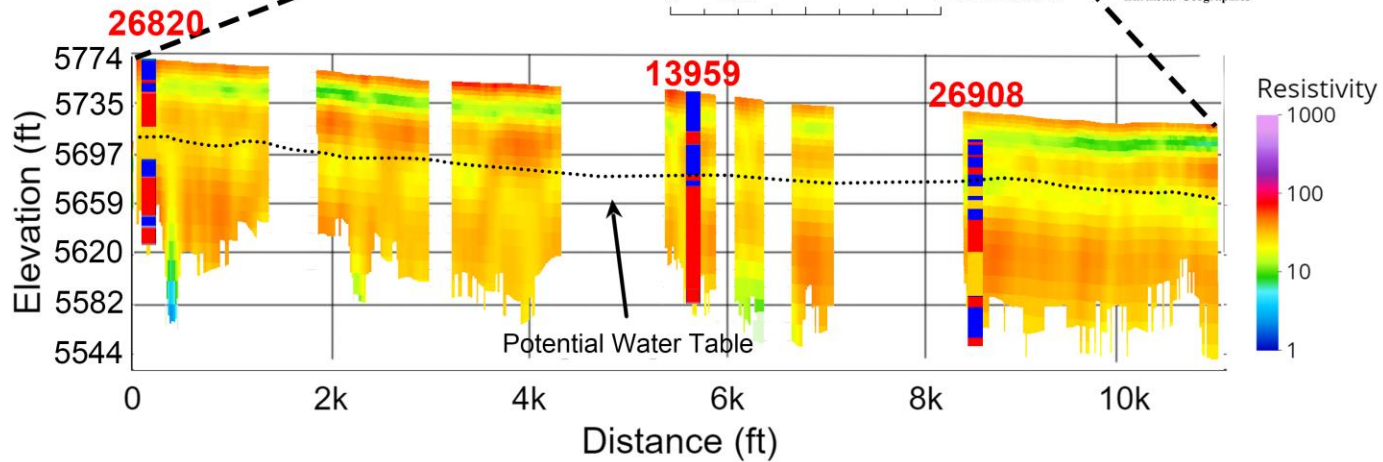
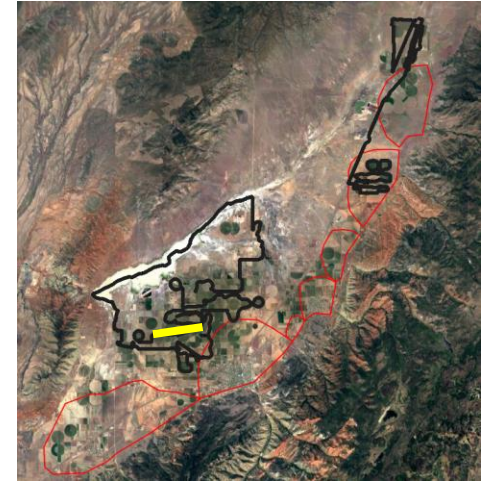
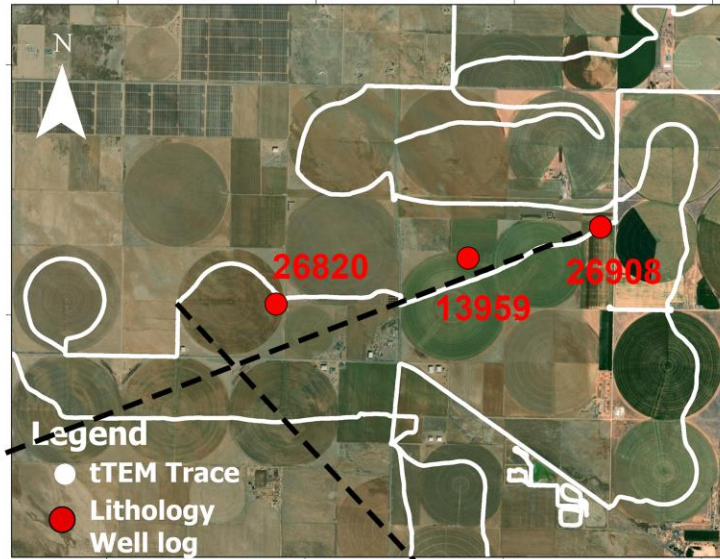




# tTEM Trace 300 Central Parowan Valley Nov 2021

## Lithology Classification for Well Log

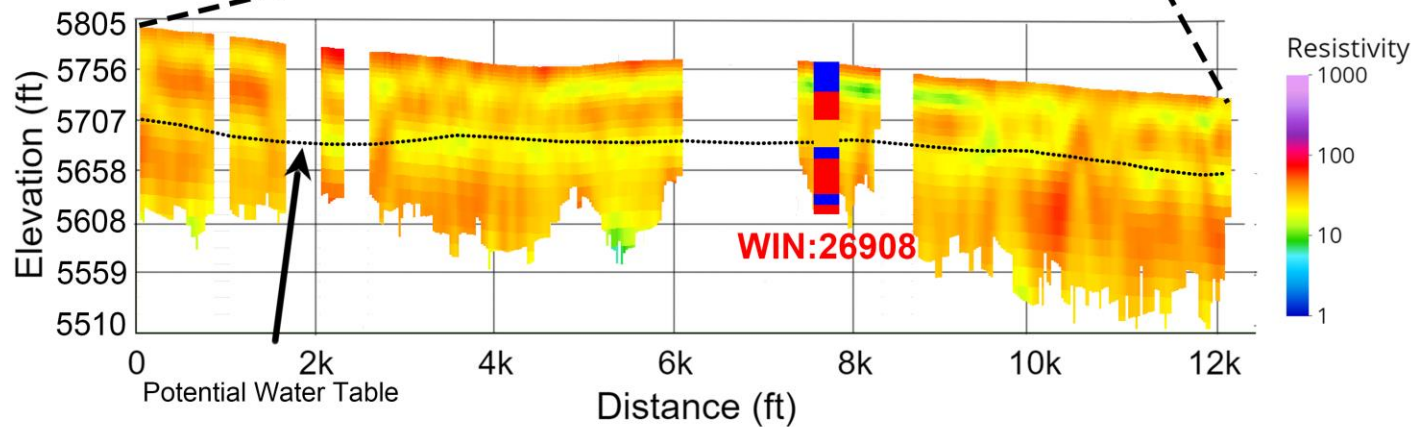
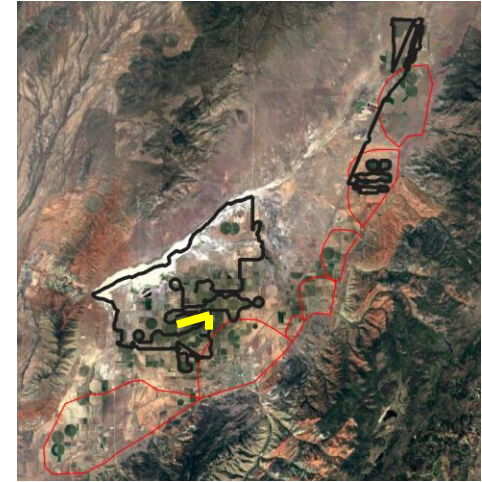
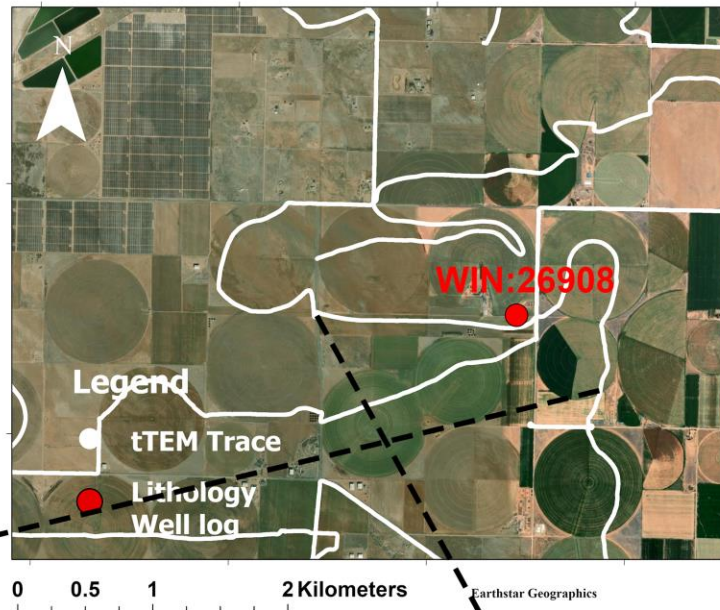
- Most coarse grain including Sand, Sand-Gravel, Gravel
- Most fine grain including Clay, Clay-Sand, Silt, Clay-Silt-Sand
- Mix of fine and coarse grain including Silt-Sand-Gravel, Clay-Sand-Gravel



# tTEM Trace 270 Central Parowan Valley Nov 2021

## Lithology Classification for Well Log

- Most coarse grain including Sand, Sand-Gravel, Gravel
- Most fine grain including Clay, Clay-Sand, Silt, Clay-Silt-Sand
- Mix of fine and coarse grain including Silt-Sand-Gravel, Clay-Sand-Gravel

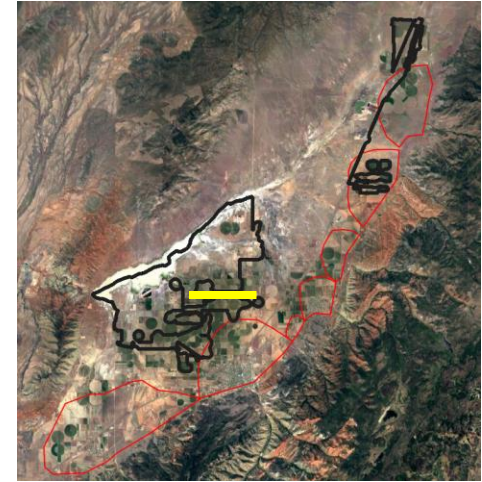
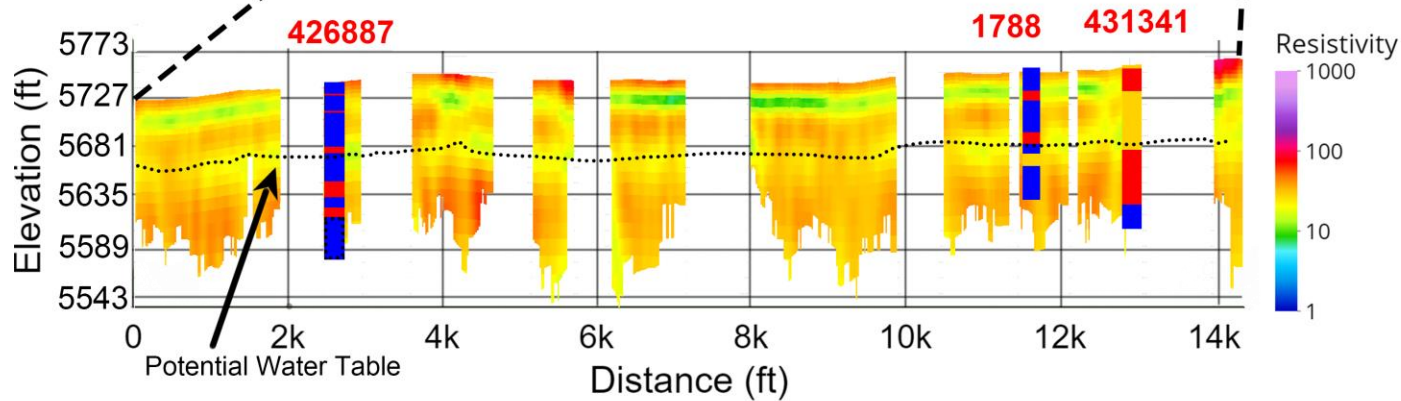
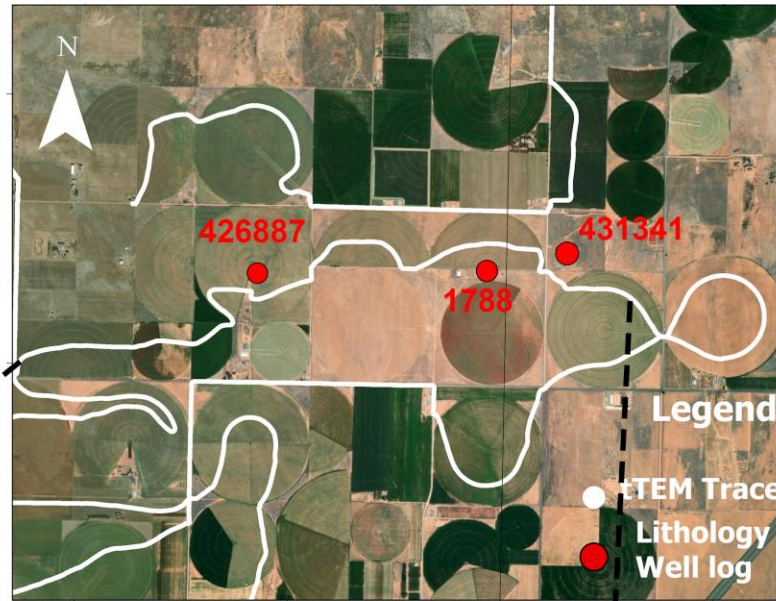




# tTEM Trace 290 Central Parowan Valley Nov 2021

## Lithology Classification for Well Lo

- Most coarse grain including Sand, Sand-Gravel, Gravel
- Most fine grain including Clay, Clay-Sand, Silt, Clay-Silt-Sand
- Mix of fine and coarse grain including Silt-Sand-Gravel, Clay-Sand-Gravel



# Future tTEM work

- Could be used to identify best locations for managed aquifer recharge
- Could better inform texture (relative amount of sands/clays) of the valley for modeling

# Questions?

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[www.remote-sensing-hydrology.com](http://www.remote-sensing-hydrology.com)