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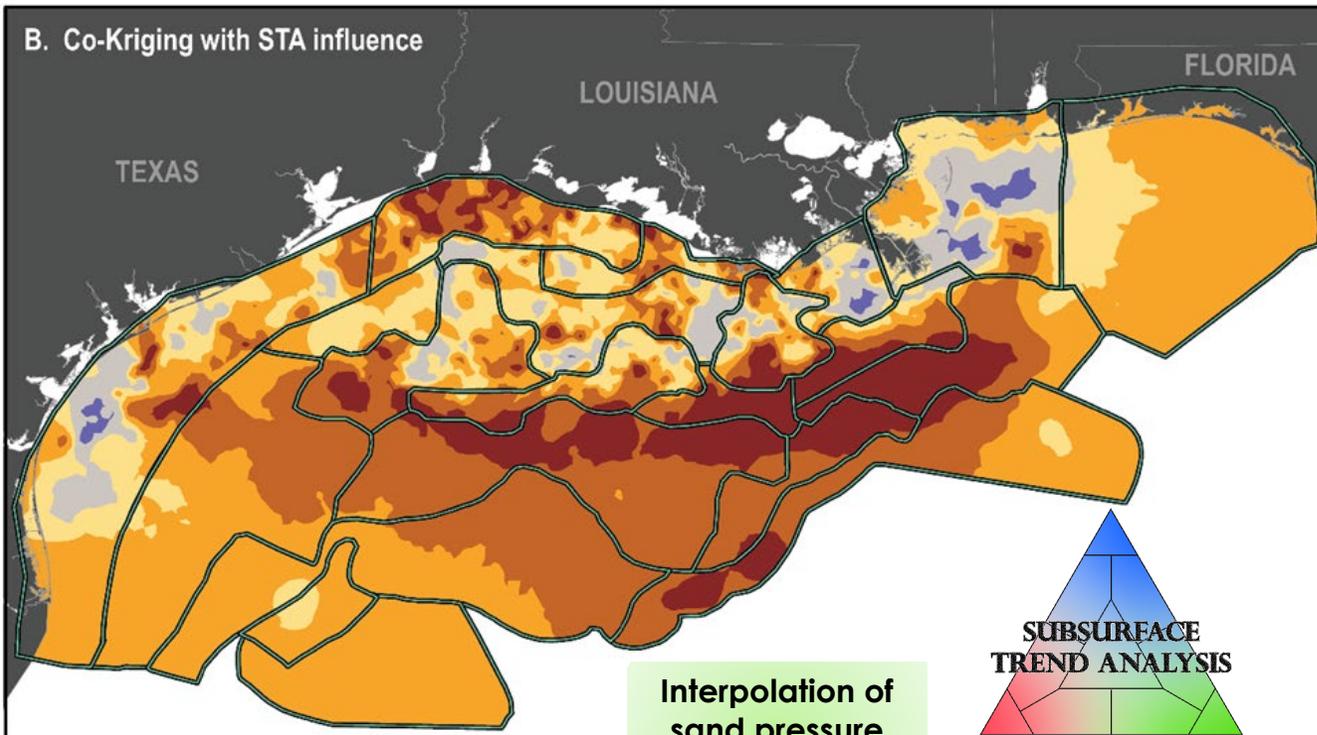
Integrating Artificial Intelligence/Machine Learning (AI/ML) to Improve Prediction of Subsurface Conditions

Issue/R&D Need

- Subsurface introduces hazards that are difficult to constrain and predict
- **Ongoing operations increase risk** in offshore regions
- There is **a need for rapid, accurate, and efficient tools that predict subsurface conditions**, even in areas with little to no data

Task Objectives

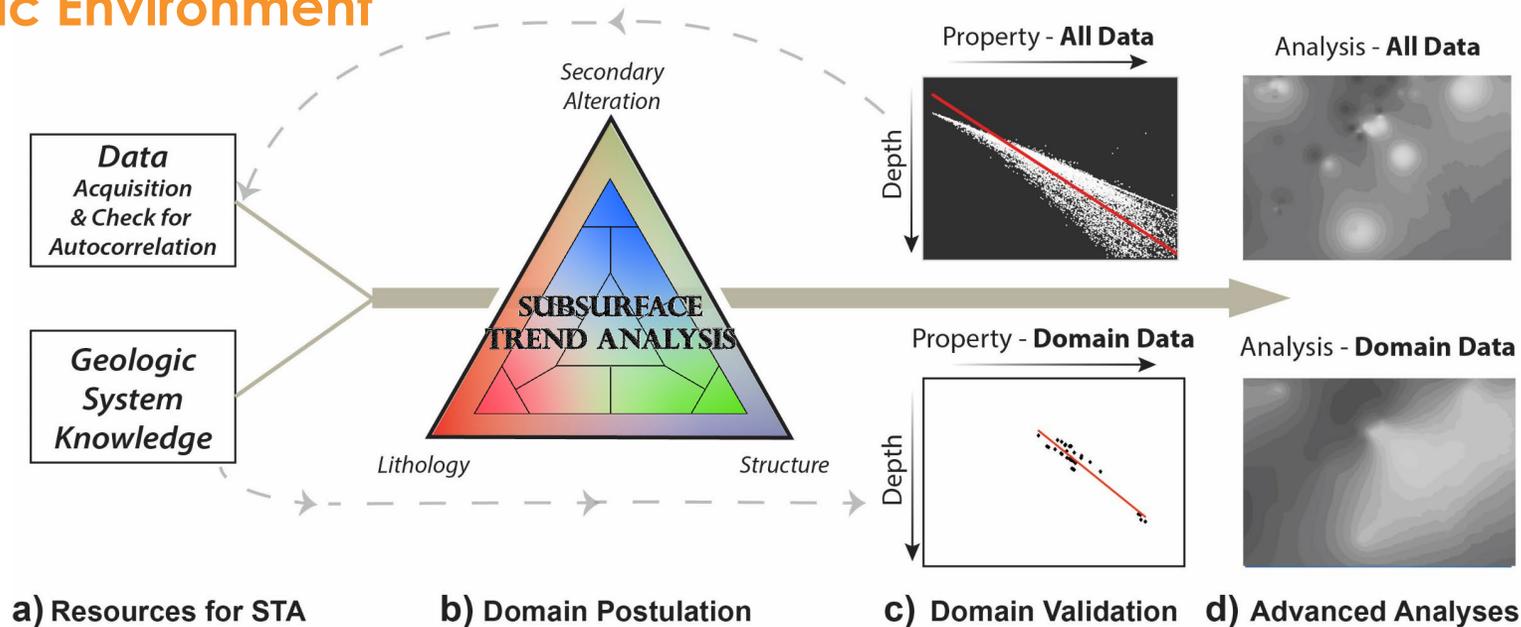
- Develop a 2D/3D, adaptive smart tool using the Subsurface Trend Analysis (STA) method framework
- Test and validate the STA smart tool in the offshore Gulf of Mexico (GOM)
- Release the tool for public use



Rose et al., 2020

Subsurface Trend Analysis

An AI/ML-Informed Methodical Framework to Predict Subsurface Properties and the Geologic Environment



Initial method published: Rose, K., Bauer, J.R., and Mark-Moser, M. (2020) Subsurface Trend Analysis, a multi-variate geospatial approach for subsurface evaluation and uncertainty reduction, *Interpretation*

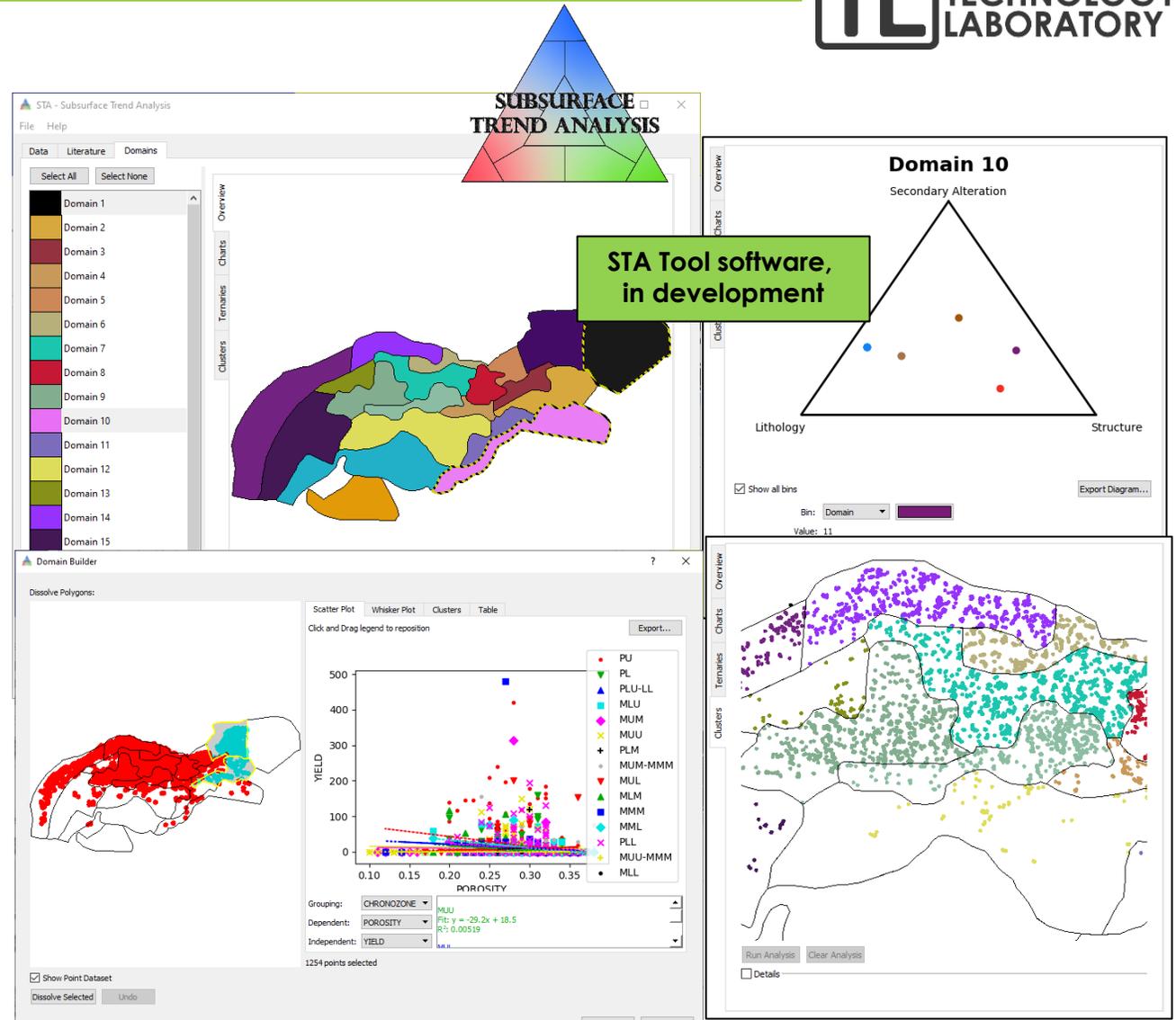
Initially developed to address spill risks following Deepwater Horizon

Validated method is expanding to an AI/ML, 2D/3D adaptive Smart Tool: the STA Tool

Subsurface Trend Analysis Tool

STA Tool – a virtual research assistant designed to:

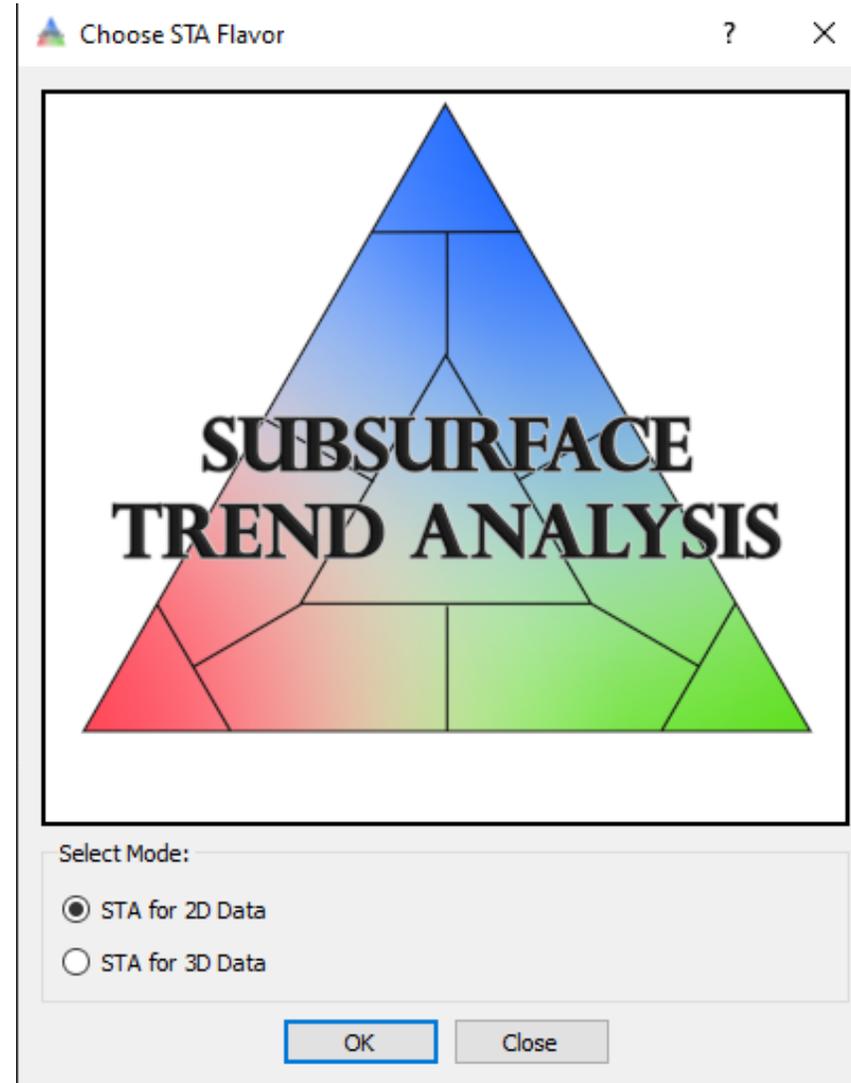
- Organize and visualize disparate big data and knowledge resources
- Simplify and automate geologic domain formation
- Provide and execute statistical analyses and validation
- Utilize AI/ML to accelerate knowledge gathering and subsurface property analysis



STA Tool Analytical Components

2D STA Components

- Autocorrelation
- Literature management
 - Natural Language Processing (NLP)
 - CNN embedded image analysis
- Domain Builder – GIS file import, data and polygon viewer, analytical charts
 - Statistical charts
 - AI/ML statistical analyses
- Wireline analysis – ingests .las files to update data



3D STA Components (in dev)

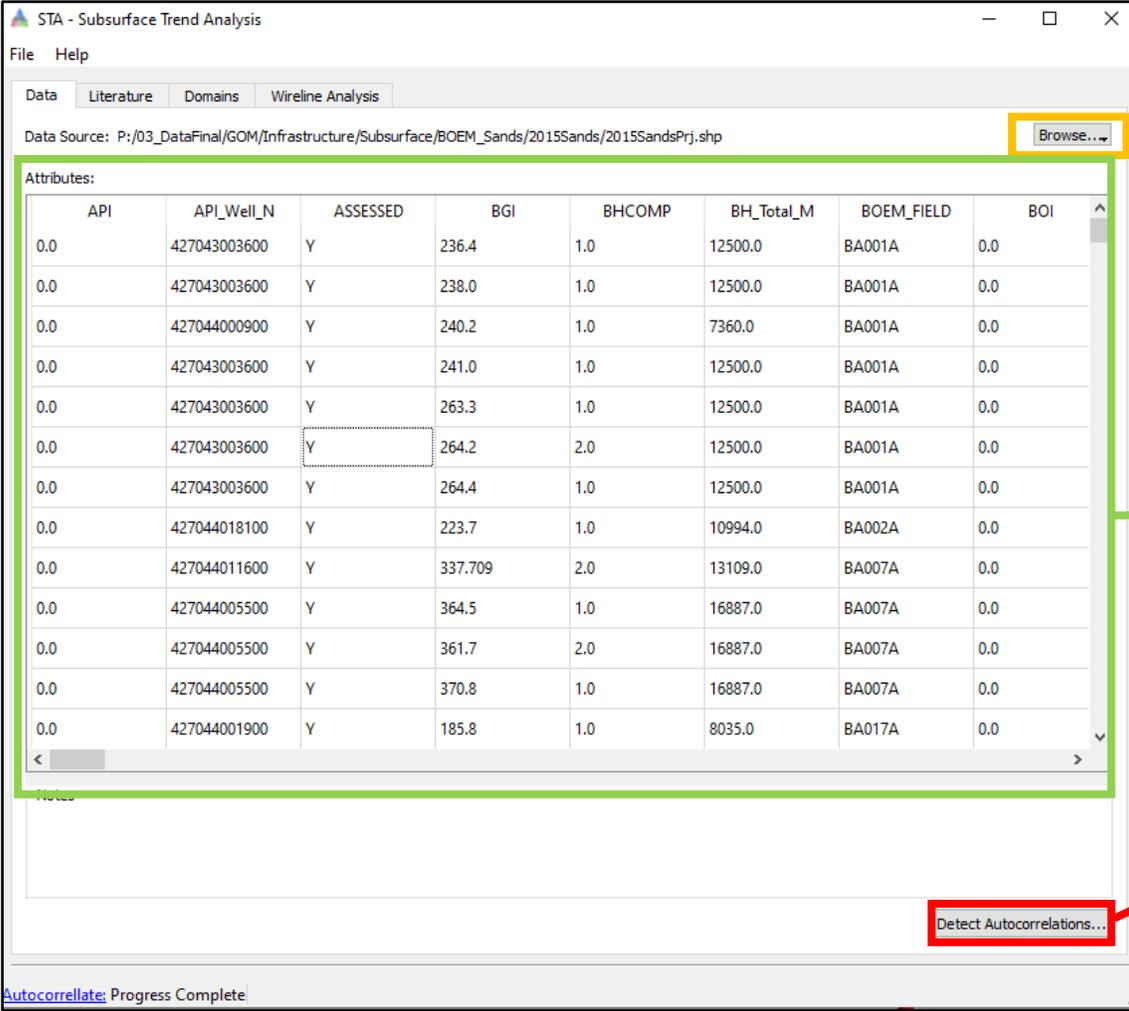
- 3D Domain Builder
- 3D Interpolation
- Wireline analysis conversion to 3D in progress

[Download the 2D STA Tool](https://edx.netl.doe.gov/datas-et/subsurface-trend-analysis-2d-tool)

<https://edx.netl.doe.gov/datas-et/subsurface-trend-analysis-2d-tool>

3D Tool to be released Dec. '22

Knowledge and Data Management in the STA Tool



STA - Subsurface Trend Analysis

File Help

Data Literature Domains Wireline Analysis

Data Source: P:\03_DataFinal\GOM\Infrastructure\Subsurface\BOEM_Sands\2015Sands\2015SandsPrj.shp

Browse...

Attributes:

API	API_Well_N	ASSESSED	BGI	BHCOMP	BH_Total_M	BOEM_FIELD	BOI
0.0	427043003600	Y	236.4	1.0	12500.0	BA001A	0.0
0.0	427043003600	Y	238.0	1.0	12500.0	BA001A	0.0
0.0	427044000900	Y	240.2	1.0	7360.0	BA001A	0.0
0.0	427043003600	Y	241.0	1.0	12500.0	BA001A	0.0
0.0	427043003600	Y	263.3	1.0	12500.0	BA001A	0.0
0.0	427043003600	Y	264.2	2.0	12500.0	BA001A	0.0
0.0	427043003600	Y	264.4	1.0	12500.0	BA001A	0.0
0.0	427044018100	Y	223.7	1.0	10994.0	BA002A	0.0
0.0	427044011600	Y	337.709	2.0	13109.0	BA007A	0.0
0.0	427044005500	Y	364.5	1.0	16887.0	BA007A	0.0
0.0	427044005500	Y	361.7	2.0	16887.0	BA007A	0.0
0.0	427044005500	Y	370.8	1.0	16887.0	BA007A	0.0
0.0	427044001900	Y	185.8	1.0	8035.0	BA017A	0.0

Detect Autocorrelations...

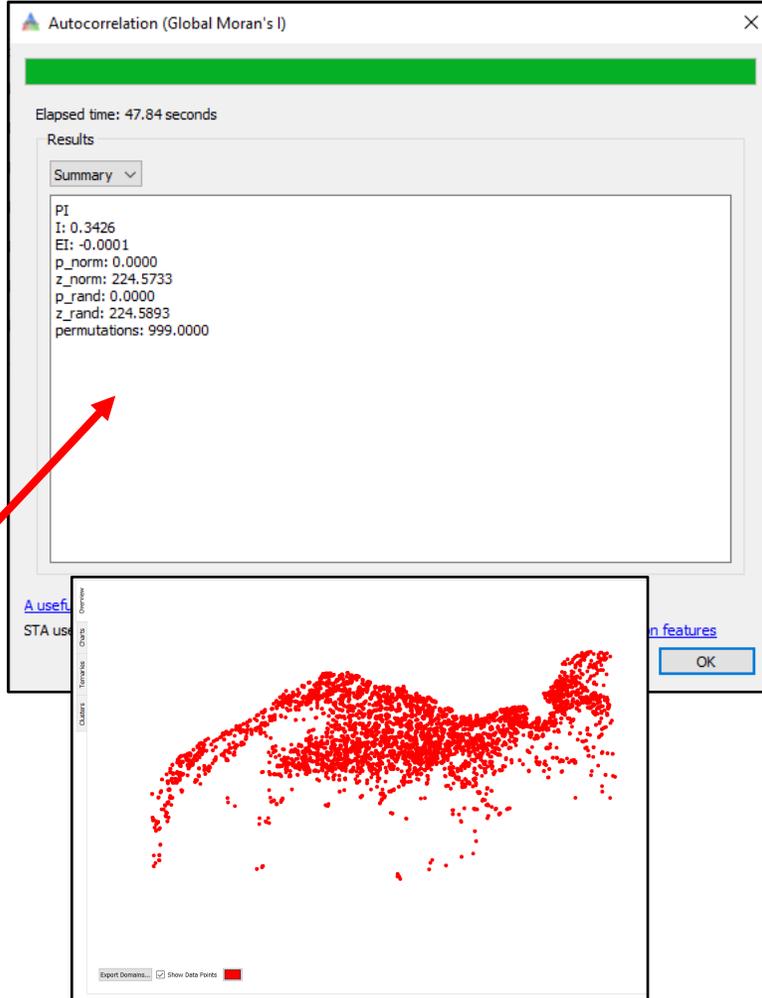
Autocorrelate: Progress Complete

Spatial Data upload and autocorrelation check

User upload of point dataset

Data attribute viewer

Autocorrelation detection



Autocorrelation (Global Moran's I)

Elapsed time: 47.84 seconds

Results

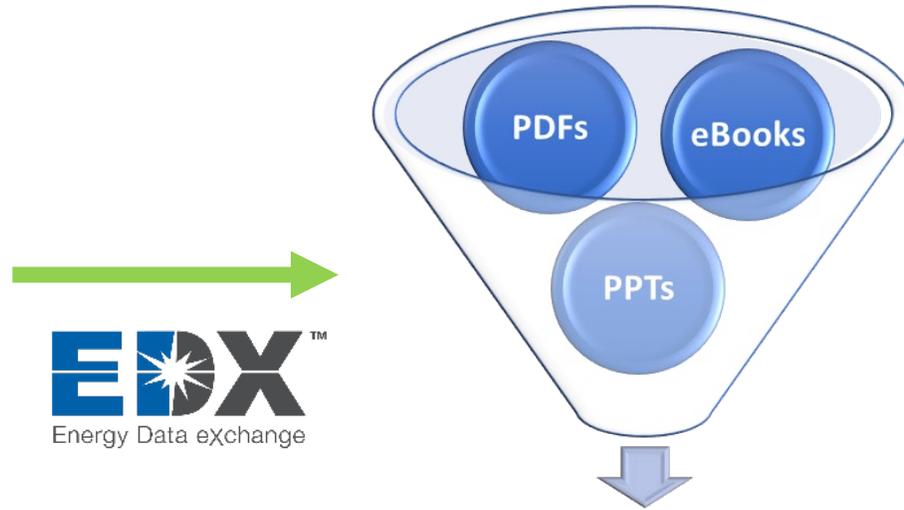
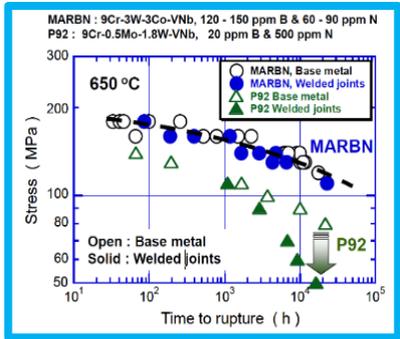
Summary

PI
I: 0.3426
EI: -0.0001
p_norm: 0.0000
z_norm: 224.5733
p_rand: 0.0000
z_rand: 224.5893
permutations: 999.0000

Export Domains... Show Data Points

Natural Language Processing for Unstructured Data

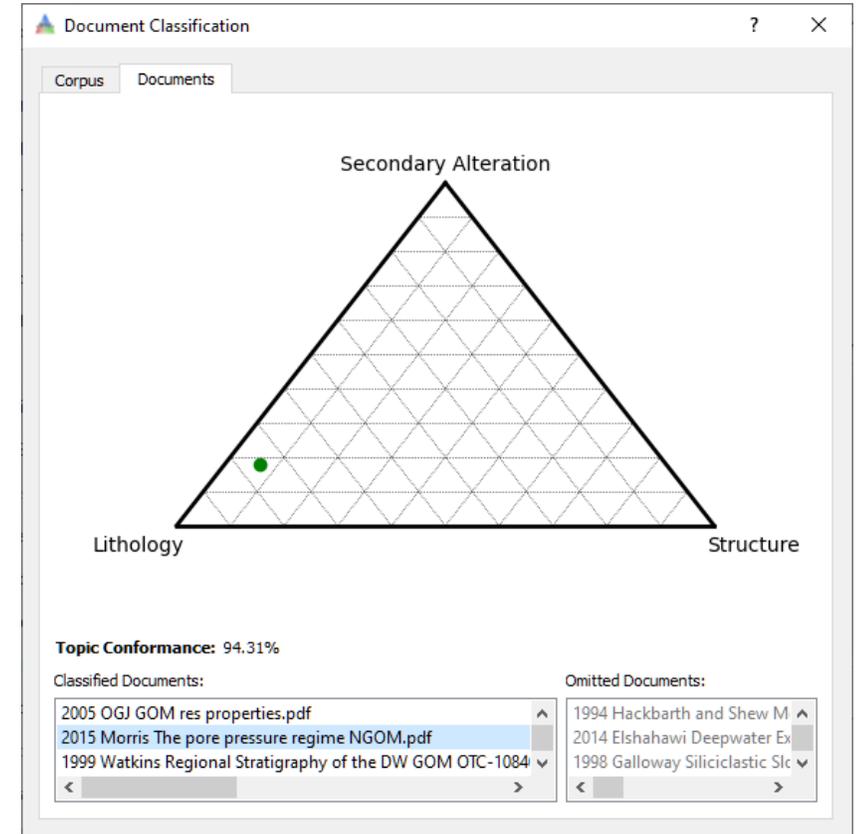
Extracting Knowledge



Structured digital data

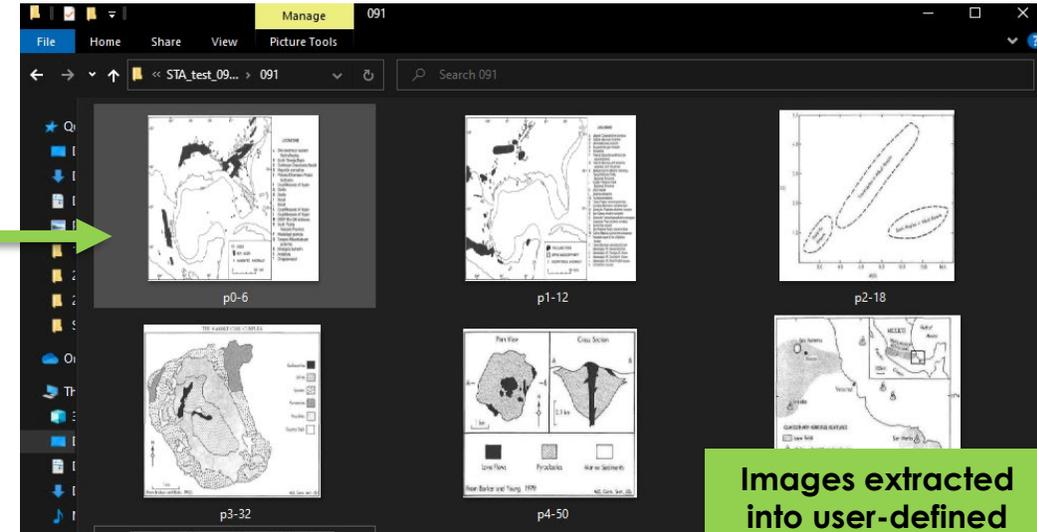
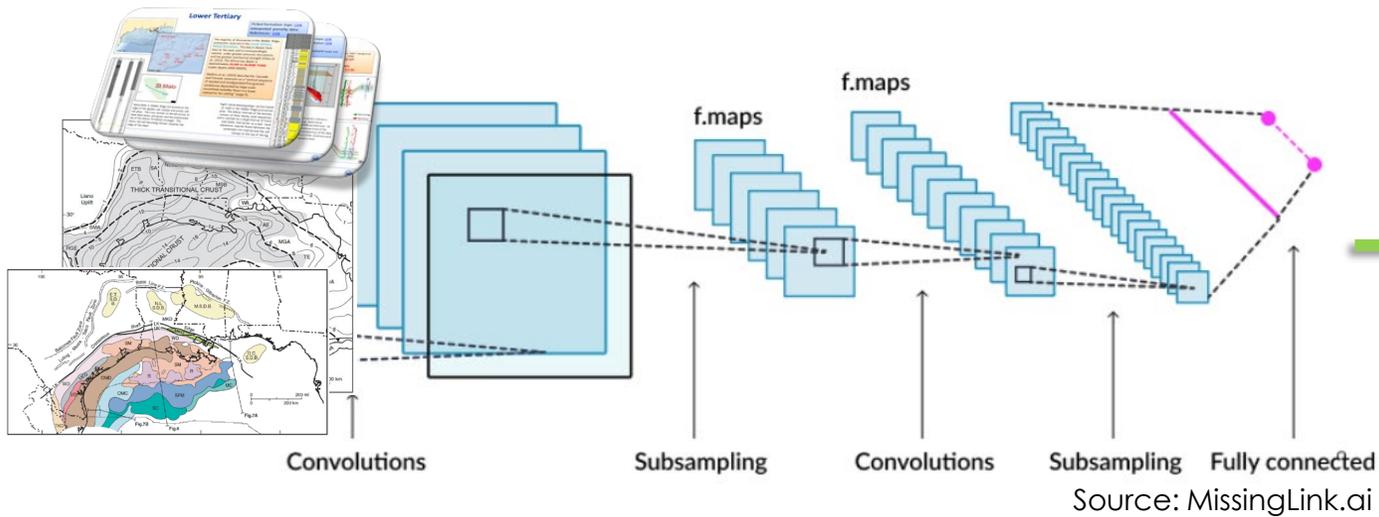
Latent Dirichlet Allocation
topic model
Jaccard similarity-based
categorization

Mining data from documents, R&D products, presentations, etc., using NLP



Computer Vision for Image Extraction

Extracting Knowledge

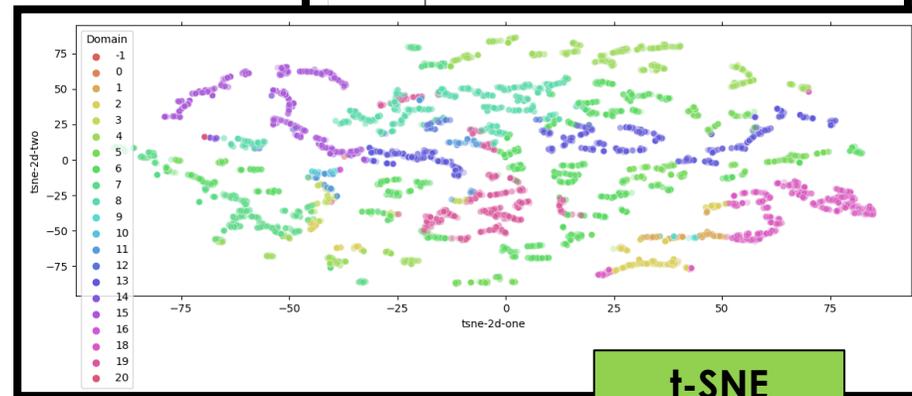
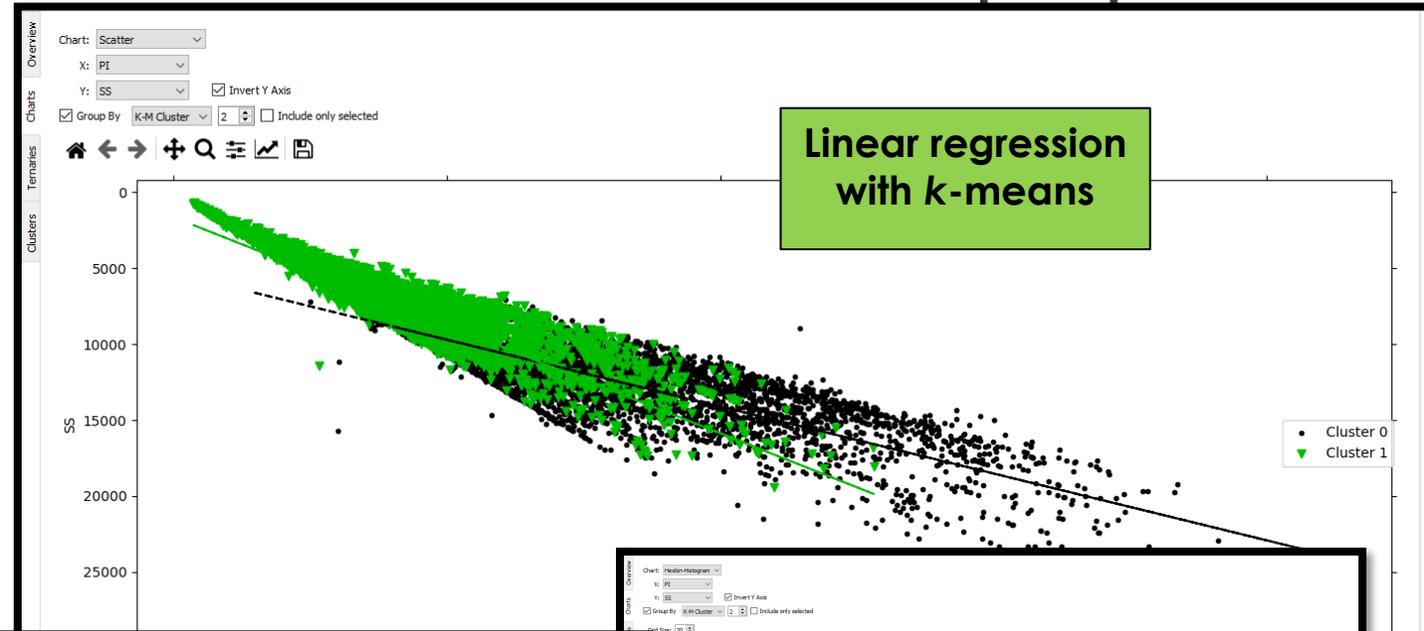


- User feeds unstructured data to Convolutional Neural Network model that returns all relevant images, **saving hours of work and reducing error**
- Transfer learning via VGGNet, tested on >1300+ images
- Multiclass image identification accuracy ~77%

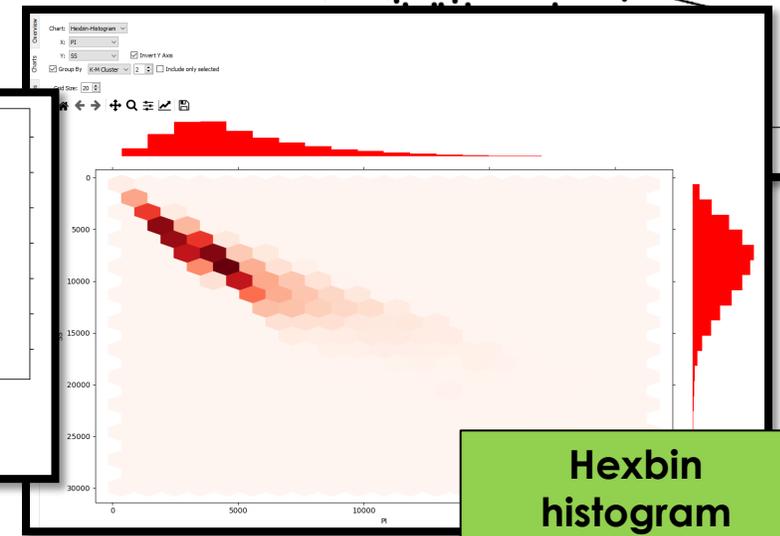
Breadth of Statistical Analytics in STA Tool

STA Tool provides both classic and AI/ML statistical analyses for broad insight into subsurface property data:

- Scatter (linear regression included)
- Box-and-whisker
- Hexbin-histogram
- DBSCAN
- t-SNE
- k-means



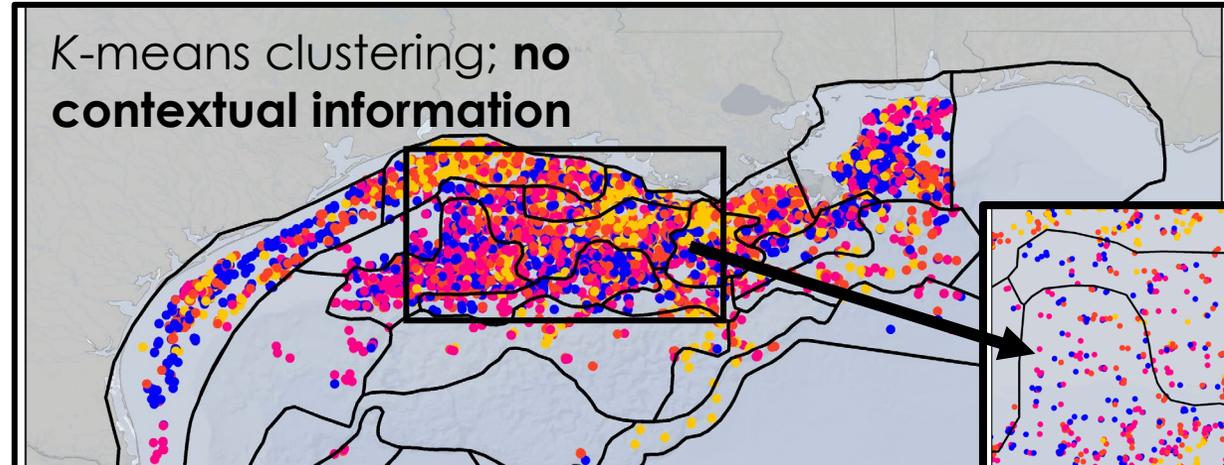
t-SNE



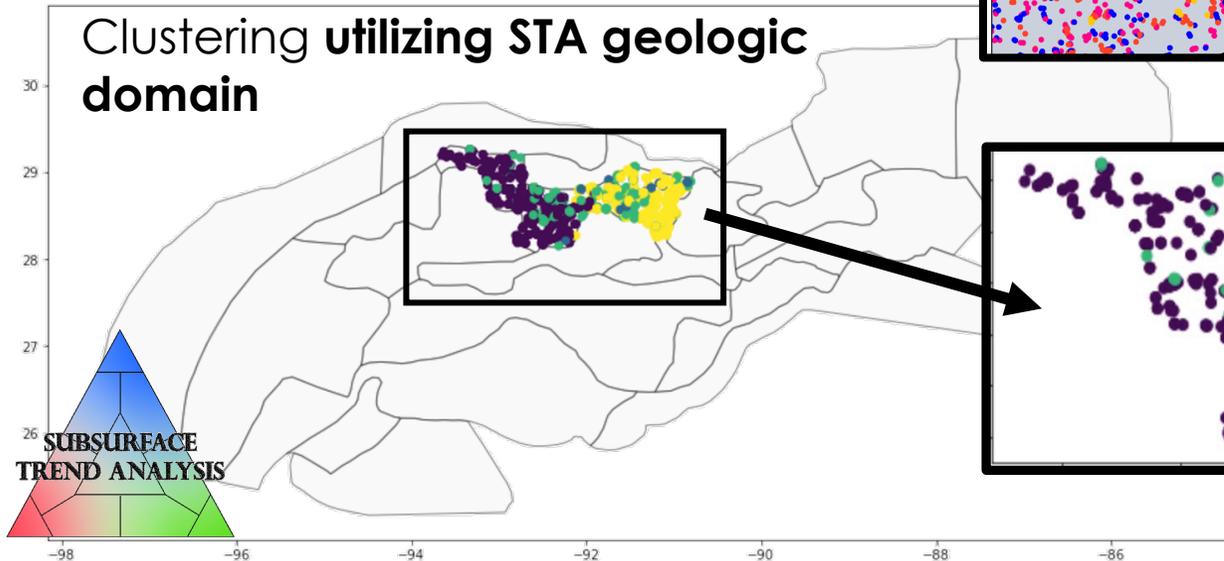
Hexbin histogram

Domain Validation and Clustering Analysis

Gulf of Mexico application using reservoir properties



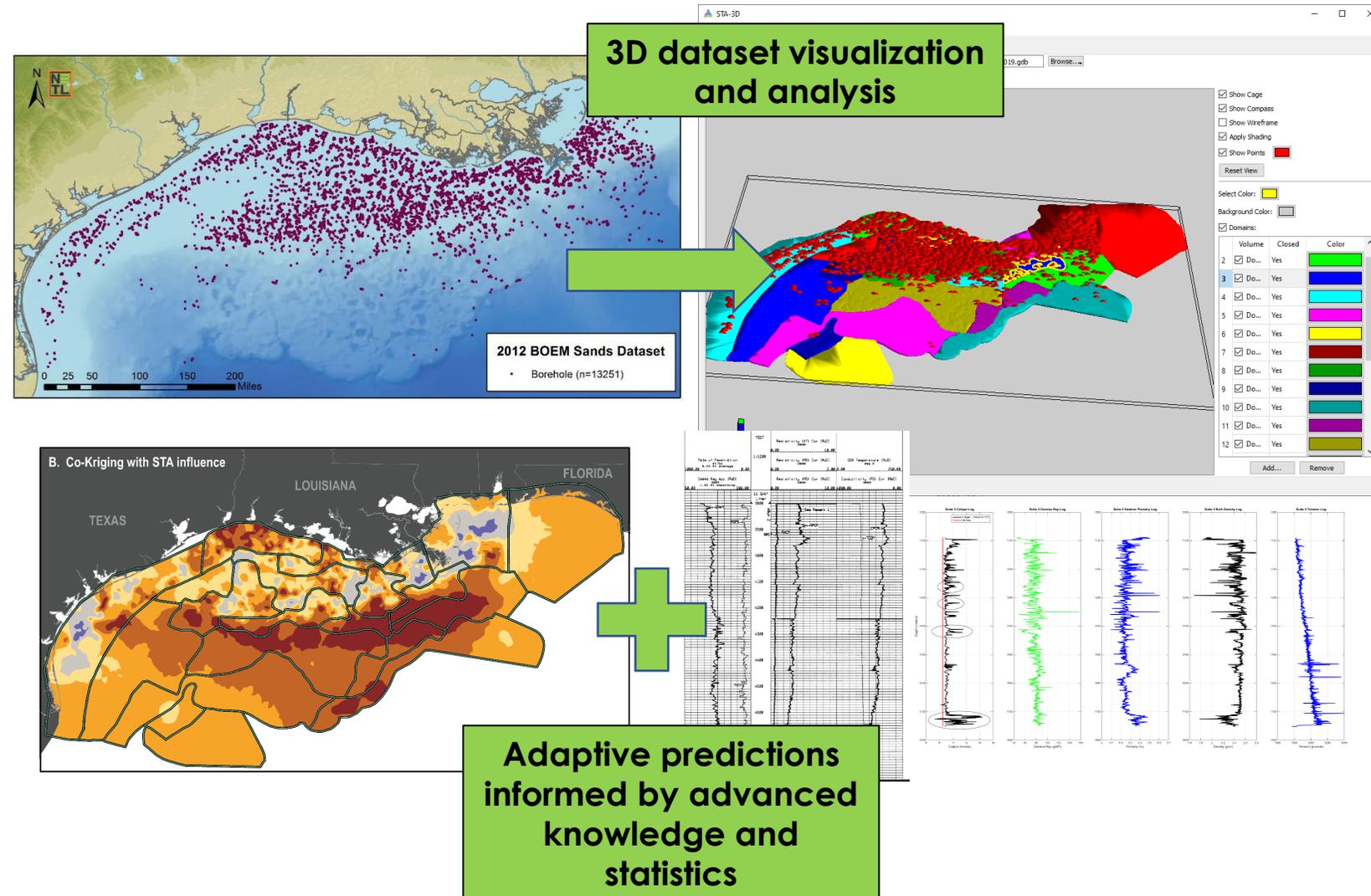
- **Four** clusters
- Poor continuity among clusters
- Cohesion score = ~10



- **Three** clusters
- Improved continuity among clusters
- Cohesion score: ~3

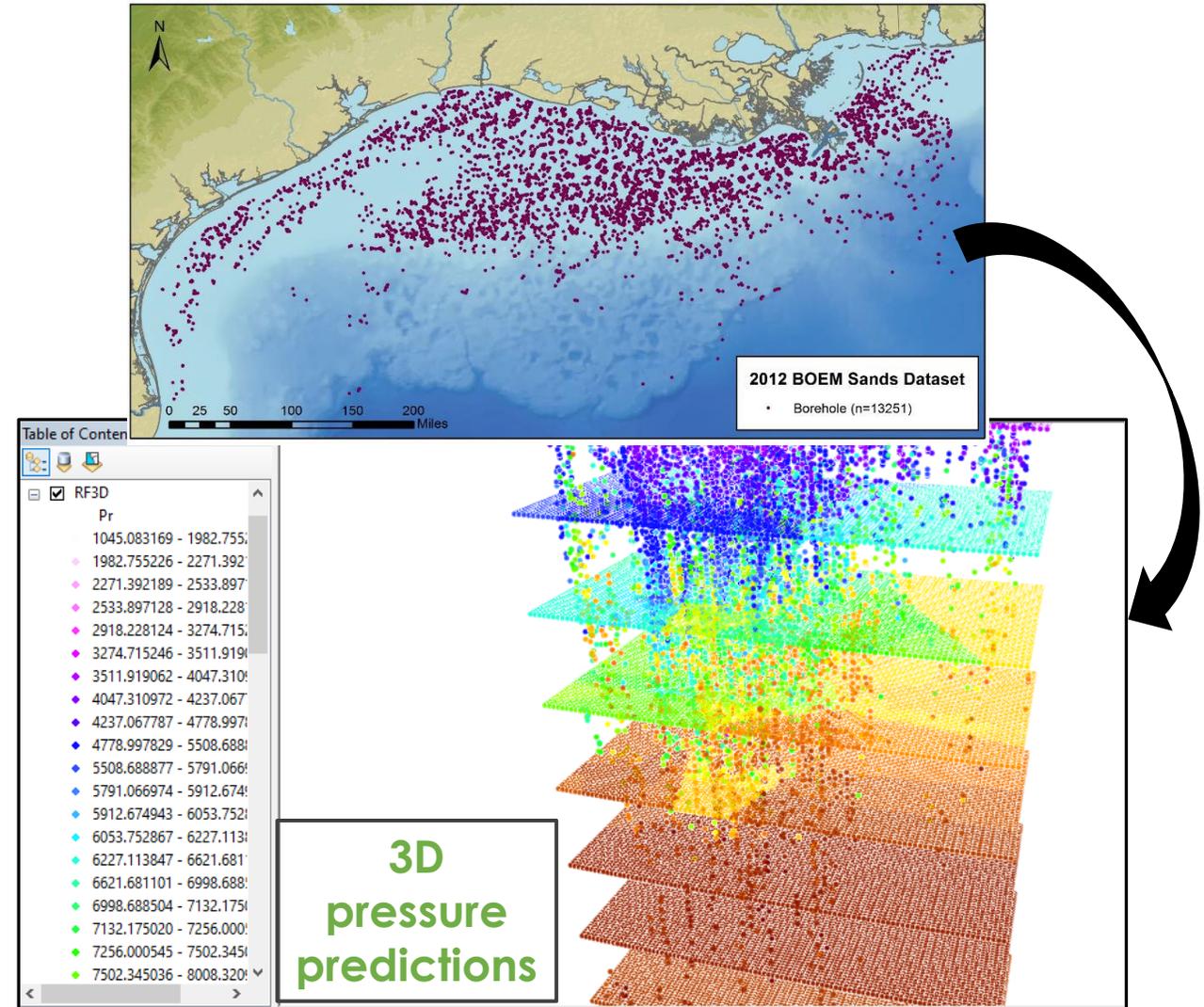
3D Enhancements and Adaptive Prediction

- 3D visualizations to gain perspective on data and subsurface predictions
- Integrating geologic knowledge and new field data to update prediction of subsurface conditions

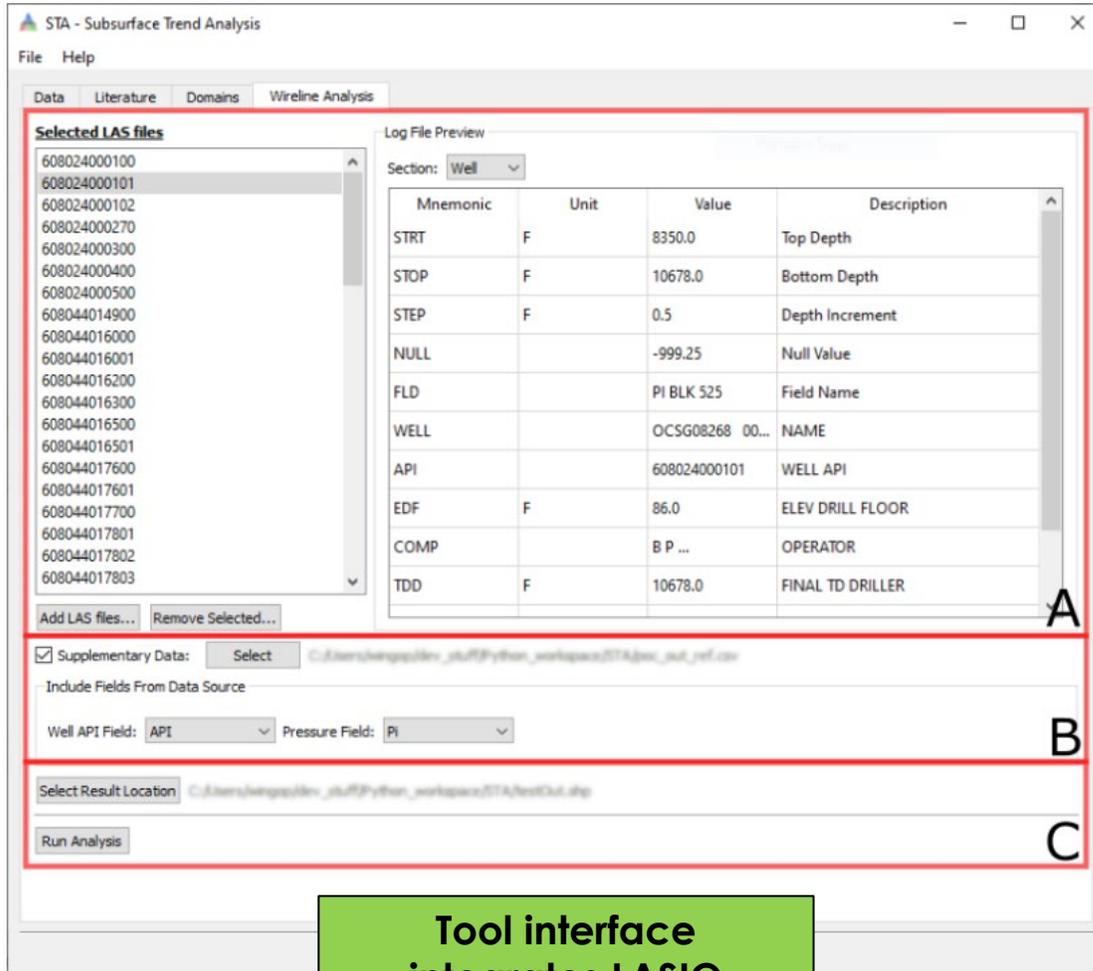


Expanding Property Prediction to 3D

- 3D interpolations of property predictions can constrain subsurface conditions and forecast hazards
- **This method expands variography from 2D to 3D**
- **Integrates multiple variables**, including quantitative and qualitative information, for 3D co-kriging
- Ultimately, model produces 3D subsurface property predictions



2D/3D Wireline Analysis



Tool interface
integrates LASIO
Python package

Automated Wireline Analysis performs the following:

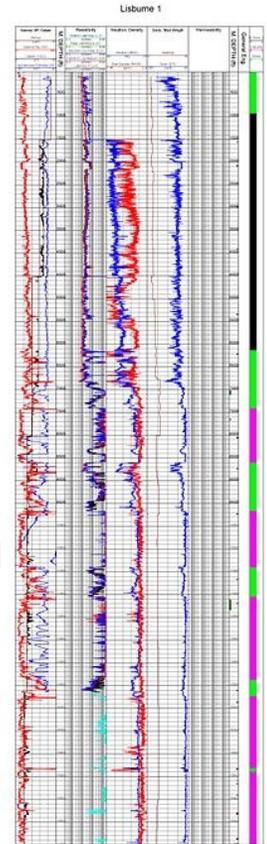
A -- LAS File management

B -- Additional data inputs

C -- Analysis run and output

Updates previously existing point data with newly extracted properties from well logs

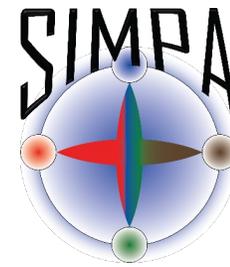
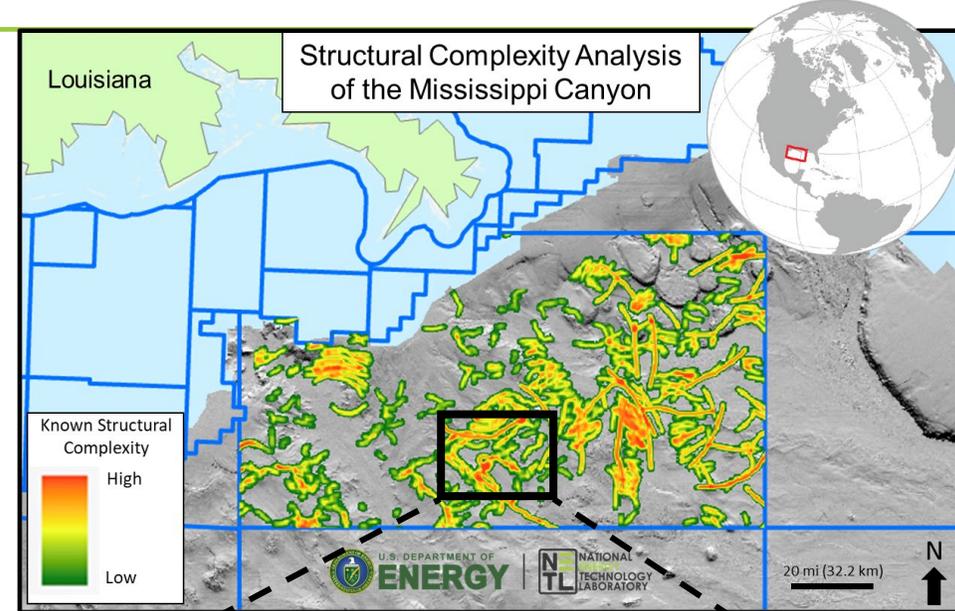
3D version in dev calculates overburden stress to indicate pressure along the wellbore



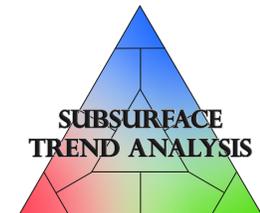
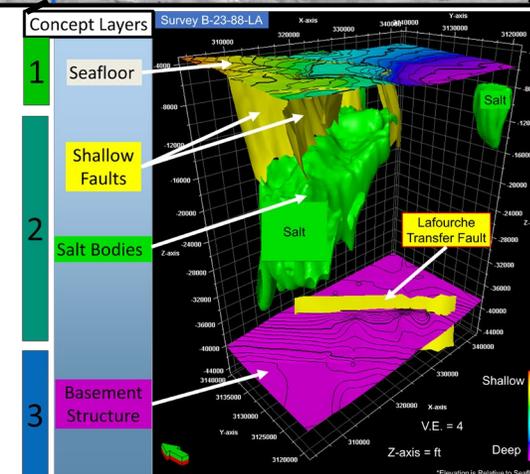
Source: USGS

Generating Datasets to Validate Geohazard Analysis in the STA Tool

- Use case: Mississippi Canyon in the Gulf of Mexico
- Multiple datasets analyzed
 - Seafloor fault mapping, seismic, basement structure
 - Subsurface data analyzed using **SIMPA** to produce STA geologic domains
 - Combined with spatial wellbore data to test tool capabilities



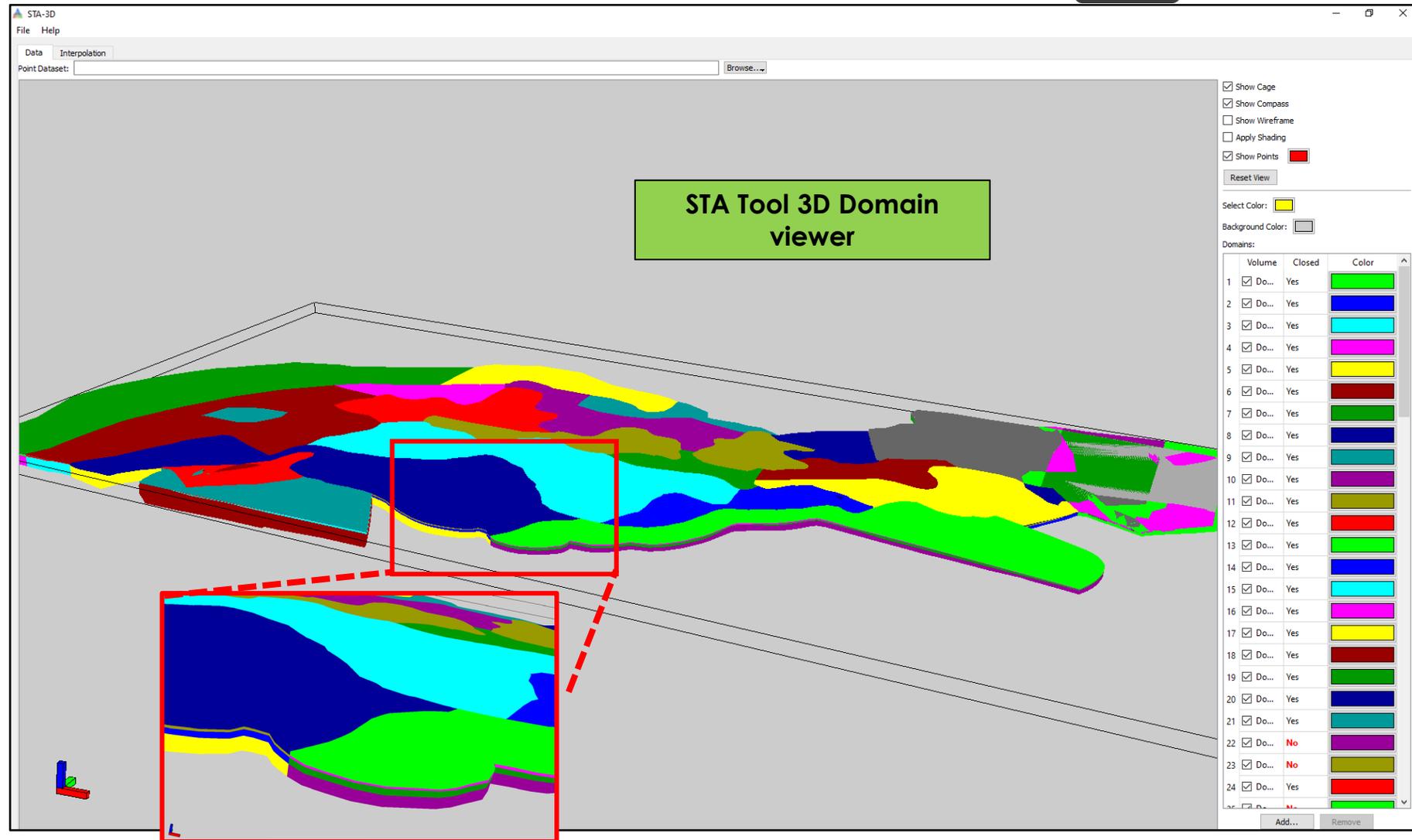
Justman et al., 2020



3D STA Domains for 3D Tool visualization and testing

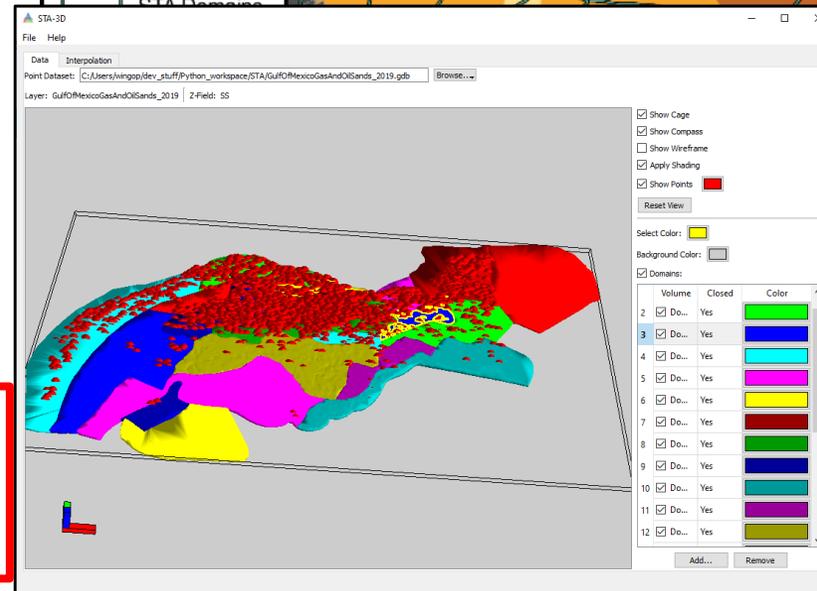
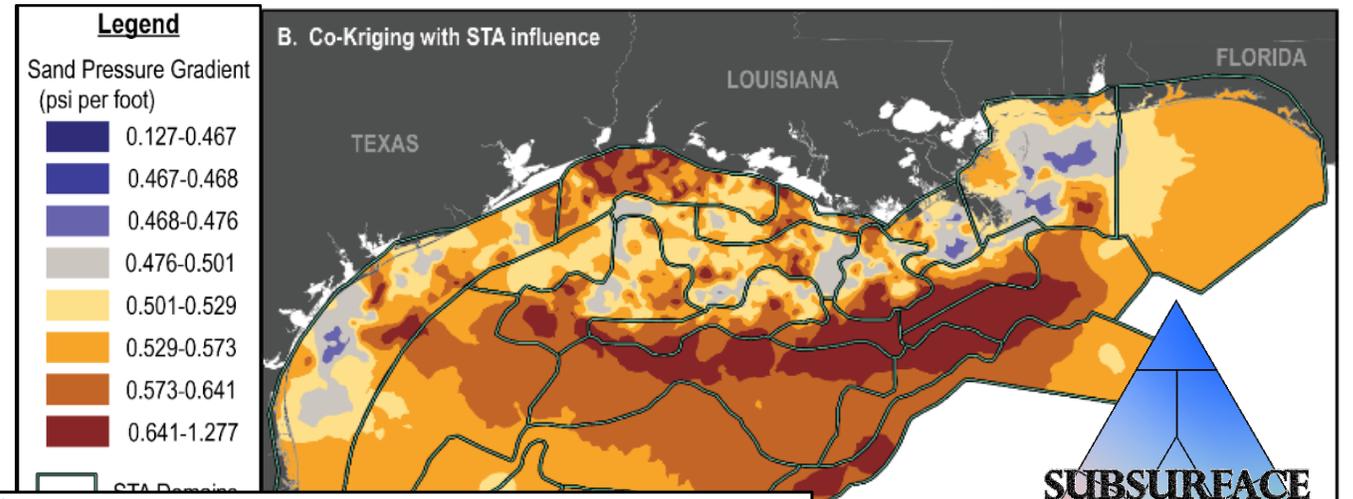
- Basin-scale 3D STA Domains based on chronozone analysis from original STA Gulf of Mexico use case
- To be utilized with 3D co-kriging interpolation

EON	ERA	PERIOD	EPOCH	Ma	
Phanerozoic	Cenozoic	Quaternary	Holocene	0.011	
			Pleistocene	Late 0.8 Early 2.4	
		Tertiary	Pliocene	Late	3.6
				Early	5.3
				Miocene	Late 11.2 Middle 16.4 Early 23.0
			Oligocene	Late	28.5
				Early	34.0
				Eocene	Middle 41.3 Early 49.0 Late 55.8
		Mesozoic	Cretaceous	Late	61.0
				Early	65.5
			Jurassic	Late	99.6
				Early	145
	Middle			161	
	Triassic		Late	176	
			Early	200	
	Paleozoic		Permian	Late	228
				Middle	245
				Early	251
			Pennsylvanian	Late	260
				Early	271
		Middle		299	
	Paleozoic	Mississippian	Late	306	
			Early	311	
			Middle	318	
Devonian		Late	326		
		Early	345		
		Middle	359		
Silurian	Late	385			
	Early	397			
	Middle	416			
Ordovician	Late	419			
	Early	423			
Cambrian	Late	428			
	Early	444			
			488		
			501		
			513		
			542		

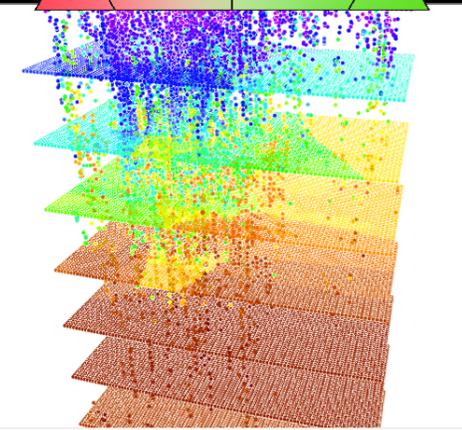


Summary

- The STA Method and Tool integrate data, knowledge and science-based analytics to accelerate data collection and knowledge integration, and improve predictions of subsurface properties
- **STA Method and Tool:**
 - Adapt to different regions and scales of analysis
 - Integrate newly acquired data and update predictions of subsurface properties
 - Seeks to provides insights where gaps in data and analytical tools exist



SUBSURFACE TREND ANALYSIS



Download the 2D STA Tool!

<https://edx.netl.doe.gov/dataset/subsurface-trend-analysis-2d-tool>

NETL RESOURCES

VISIT US AT: www.NETL.DOE.gov



Contacts

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Offshore information available at
<https://edx.netl.doe.gov/offshore/>



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ENERGY

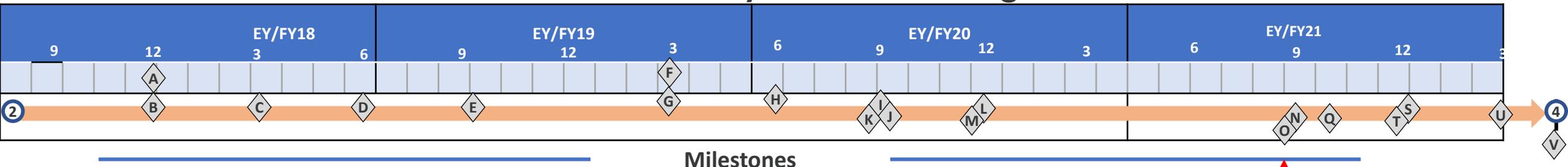


Offshore Unconventional FWP

Key Team Members: PI – Kelly Rose - CO-PI – Mackenzie Mark-Moser



Task 5: Geohazards & Subsurface Uncertainty Smart Modeling



Milestones

Number	Date	Description
EY20.5.H	06/2020	Initiate STA analysis of GOM subregion to refine use of the STA at multiple scales (wellbore-to-field scale) and provide a validation/test of the enhanced STA tool.
EY20.5.I	09/2020	Continue analysis of advanced subsurface properties in the central GOM and use new 3-D approach and visualization techniques as available. These may include CO ₂ and/or H ₂ S occurrence, fracture and fault distributions, and reservoir thickness. The team will pick one of these variables, based on data availability, and initiate an STA analysis for this parameter.
EY20.5.J	09/2020	Release of the ML-NLP enhanced STA beta tool for 2-D analytics.
EY20.5.K	09/2020	Initiate development of 3-D analytical and visualization logic into the ML-NLP-STA tool.
EY20.5.L	12/2020	Finalize development of ML driven neural network analysis of statistical dimensions and image embedding.
EY20.5.M	12/2020	Initiate 3-D analysis use-case using STA analysis with a separate fuzzy logic-driven NETL tool of a GOM subregion for subsurface structural complexity or other advanced subsurface property. This effort will use high-resolution bathymetry to map seafloor features (this has ties to Offshore Task 6.0 goals) in combination with geophysical datasets to constrain the GOM subsurface and basement boundary to understand and forecast subsurface structural complexity (areas of more likely faults and fractures) and uncertainty.
EY20.5.N	09/2021	Continue and finalize development on the component to handle real-time ingestion of subsurface geologic property data from LWD/SWD and wireline data streams. Real-time subsurface prediction and uncertainty reduction by combining STA with LWD/SWD data streams in the 2-D ML/AI STA tool.
EY21.5.O	09/2021	Evaluate and implement beta augmented reality/virtual reality (AR/VR) capabilities that may be paired with outputs from the tool for 3-D visualization of STA properties, end-user benefit.
EY21.5.P	–	If appropriate, seek and engage a partner for a DOE TCF proposal to commercialize ML-NLP enhanced STA tool during summer of 2021.
EY21.5.Q	09/2021	Complete journal manuscript describing and documenting the 2-D ML/AI enhanced STA model and validation use-case from EY/FY20.
EY21.5.R	10/2021	Complete application of the 3-D ML/AI enhanced STA model, test case, initiated in EY/FY20 for structural complexity and bathymetric analytics. Evaluate for any changes or enhancements required to finalize and validate the tool.
EY21.5.S	12/2021	Complete a 2-D validation test case for 2-D ML NLP enhanced STA tool. The STA hybrid GOM analysis (see previous bullet) and/or the LWD/SWD real-time property capability can be used for this validation.
EY21.5.T	12/2021	Improve and finalize development of 3-D analytical and visualization logic into the ML-NLP enhanced STA tool.
EY21.5.U	03/2022	Enhance handling of real-time ingestion of subsurface geologic property data from LWD/SWD and wireline data streams. Real-time subsurface prediction and uncertainty reduction by combining STA with LWD/SWD data streams in the 3-D ML/AI STA tool.
EY22.5.V	06/2022	Complete development, testing, and validation of the 3-D ML NLP enhanced STA tool.

Chart Key

- # TRL Score
- Go / No-Go Timeframe
- Project Completion
- Milestone

Publications, Presentations, External Interest



Upcoming Publications

- Mark-Moser, M., Rose, K., Suhag, A., Wingo, P., Hoover, B., Bean, A., Pantaleone, S., and Bauer, J., Exploring integrated data science techniques for subsurface property trend analysis and prediction. In prep, Computers & Geosciences
- Hoover, B., Zaengle, D., Mark-Moser, M. Enhancing knowledge discovery of unstructured data to support context in subsurface-modeling predictions. In prep, Environmental Modeling and Software

Presentations

- Pantaleone, S., Mark-Moser, M., Bean, A., Walker, S., Rose, K. Accepted. Forecasting 3D Structural Complexity with AI/ML Methods: Mississippi Canyon, Gulf of Mexico. **AAPG/SEG IMAGE conference 2021**, Sept. 26th-Oct. 1, Denver, CO/Virtual.
- Rose, K., Mark-Moser, M., Suhag, S., Bauer, J. Submitted, invited talk. Improving prediction of subsurface properties using a geoscience informed, multi-technique, artificial intelligence approach. **AGU Fall Meeting 2021**, Dec. 13-17, New Orleans, LA/Virtual. Session: H071 – Machine Learning Applications in Geosciences Modeling and Measurement.
- Mark-Moser, M., Wingo, P., Duran, R., Dyer, A., Zaengle, D., Suhag, A., Hoover, B., Pantaleone, S., Shay, J., Bauer, J., Rose, K. Submitted. AI/ML integration for accelerated analysis and forecast of offshore hazards. **AGU Fall Meeting 2021**, Dec. 13-17, New Orleans, LA/Virtual. Session: EP027 - Proven AI/ML applications in the Earth Sciences.

Past publications*

- Rose, K., Bauer, J.R., and Mark-Moser, M., 2020, A systematic, science-driven approach for predicting subsurface properties, *Interpretation*, 8:1, 167-181 <https://doi.org/10.1190/INT-2019-0019.1>
- Mark-Moser, M.; Miller, R.; Rose, K.; Bauer, J.; Disenhof, C. [Detailed Analysis of Geospatial Trends of Hydrocarbon Accumulations, Offshore Gulf of Mexico](#); NETL-TRS-13-2018; NETL Technical Report Series; U.S. Department of Energy, National Energy Technology Laboratory: Albany, OR, 2018; p 108. DOI: 10.18141/1461471.

Past Presentations*

- Machine Learning for Oil and Gas, November 2020
- SMART Webinar December 2019
- AGU Fall Meeting 2019
- Machine Learning for Unconventional Resources 2019
- AAPG Special Topic Forum Invited Talk 2018
- Geological Society of America 2017

Datasets

- Mark-Moser, M. Subsurface Trend Analysis domains for the northern Gulf of Mexico, 3/25/2020, <https://edx.netl.doe.gov/dataset/subsurface-trend-analysis-domains-for-the-northern-gulf-of-mexico>, DOI: 10.18141/1606228

*Previous project ended in 2016. Some of these are subsequent products from that relate to this ongoing AI/ML offshore geohazard research effort

External
Interest



Advanced Offshore Research Task 5

Research Problem:

- Offshore sedimentary systems are complicated, and heterogeneous subsurface introduces hazards and risks that are hard to constrain and predict pre-drill leading to deleterious impacts, such as the Macondo blowout and Deepwater Horizon spill in 2010
- There is a need for rapid, accurate, and efficient tools that effectively predict pre-drill subsurface conditions, even in areas with little to no data

Proposed Research:

- Develop a 2D, and eventually 3D, real-time “smart” tool using the Subsurface Trend Analysis method framework
- Integrate machine learning and artificial intelligence (AI/ML) to improve efficacy and robustness of analyses
- Test and validate the AI/ML-enhanced STA Tool utilizing LWD/SWD datasets and analyses of structural complexity in the Gulf of Mexico (GOM)

Benefit:

- Reduction of subsurface hazards and risks by utilizing these predictions to assist in efficient geohazard prevention and resource estimation, e.g., prevention of oil spills, CO₂ storage estimation, rare earth element enrichment, and geothermal prospects

