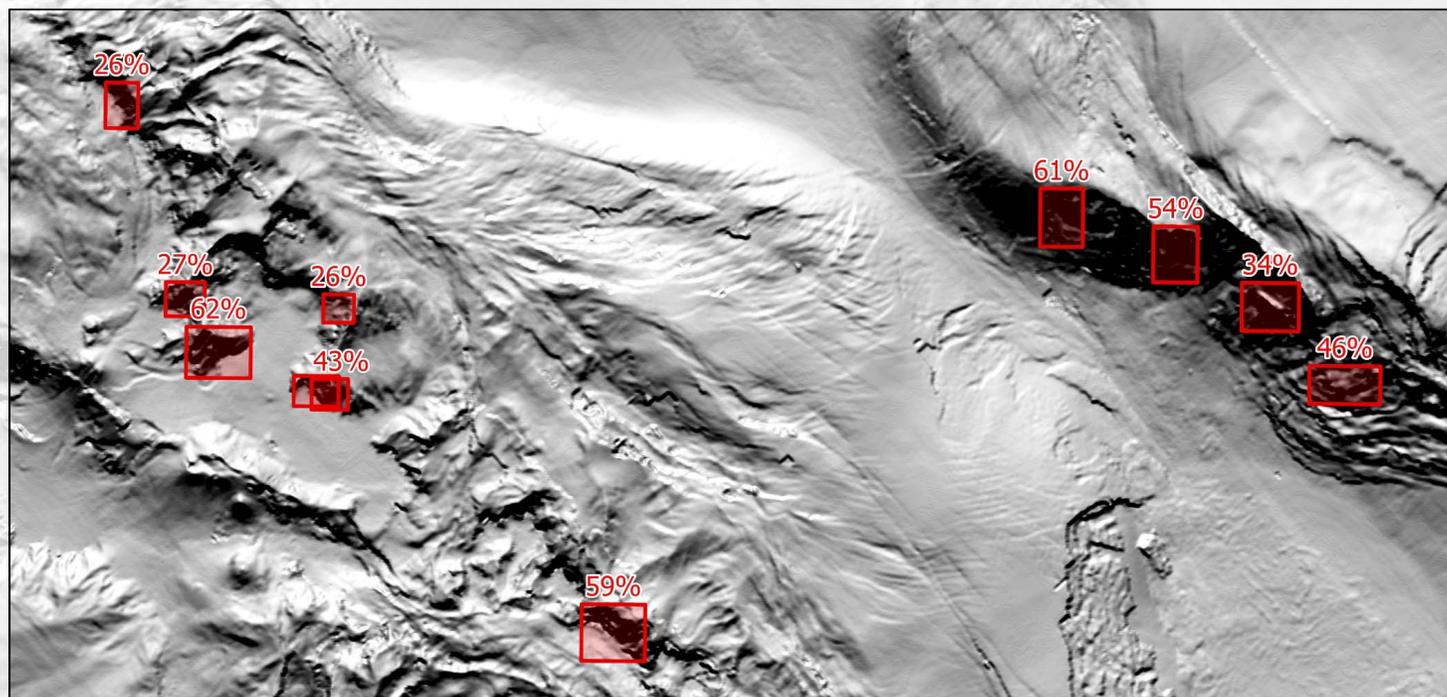


Infrastructure and Metocean Technology: The Ocean and Geohazard Analysis Tool



Submarine Landslide Detection Using YOLO Algorithm



50% — Confidence Level (%)
[Red Box] — Landslide Prediction



0 1 Miles



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Disclaimer



This project was funded by the United States Department of Energy, National Energy Technology Laboratory, in part, through a site support contract. Neither the United States Government nor any agency thereof, nor any of their employees, nor the support contractor, nor any of their employees, makes any warranty, express or implied, or assumes any legal liability or responsibility for the accuracy, completeness, or usefulness of any information, apparatus, product, or process disclosed, or represents that its use would not infringe privately owned rights. Reference herein to any specific commercial product, process, or service by trade name, trademark, manufacturer, or otherwise does not necessarily constitute or imply its endorsement, recommendation, or favoring by the United States Government or any agency thereof. The views and opinions of authors expressed herein do not necessarily state or reflect those of the United States Government or any agency thereof.

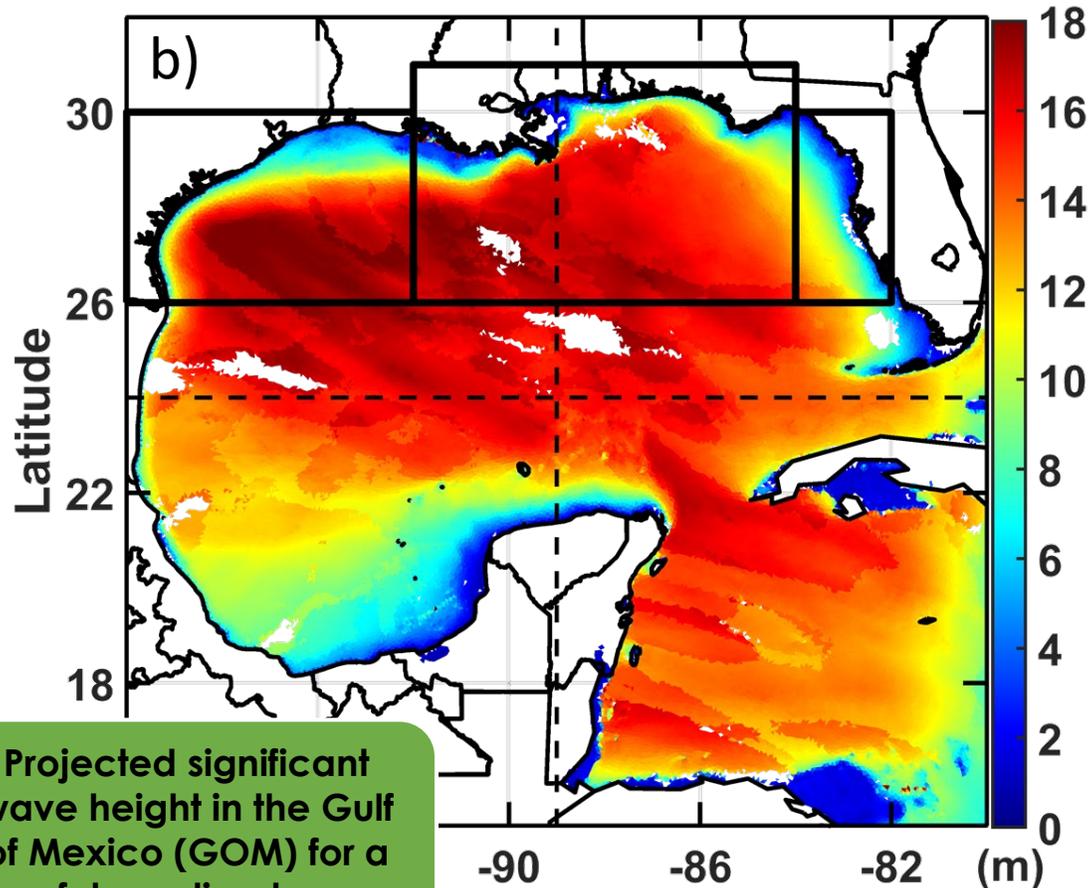
PIs: MacKenzie Mark-Moser^{1,2}, Rodrigo Duran^{1,2,3}, Jennifer Bauer¹

Researchers: Patrick Wingo^{1,2}, Dakota Zaengle^{1,2}, Izzy Pfander^{1,2}, Cat Schooley^{1,2}, Chukwuemeka Okoli^{1,2}, Michael C. Gao^{1,2}, Kelly Rose¹

¹National Energy Technology Laboratory, 1450 Queen Avenue SW, Albany, OR 97321, USA

²NETL Support Contractor, 1450 Queen Avenue SW, Albany, OR 97321, USA

³Theiss Research, 7411 Eads Avenue, La Jolla, CA 92037, USA



Why is this work important?

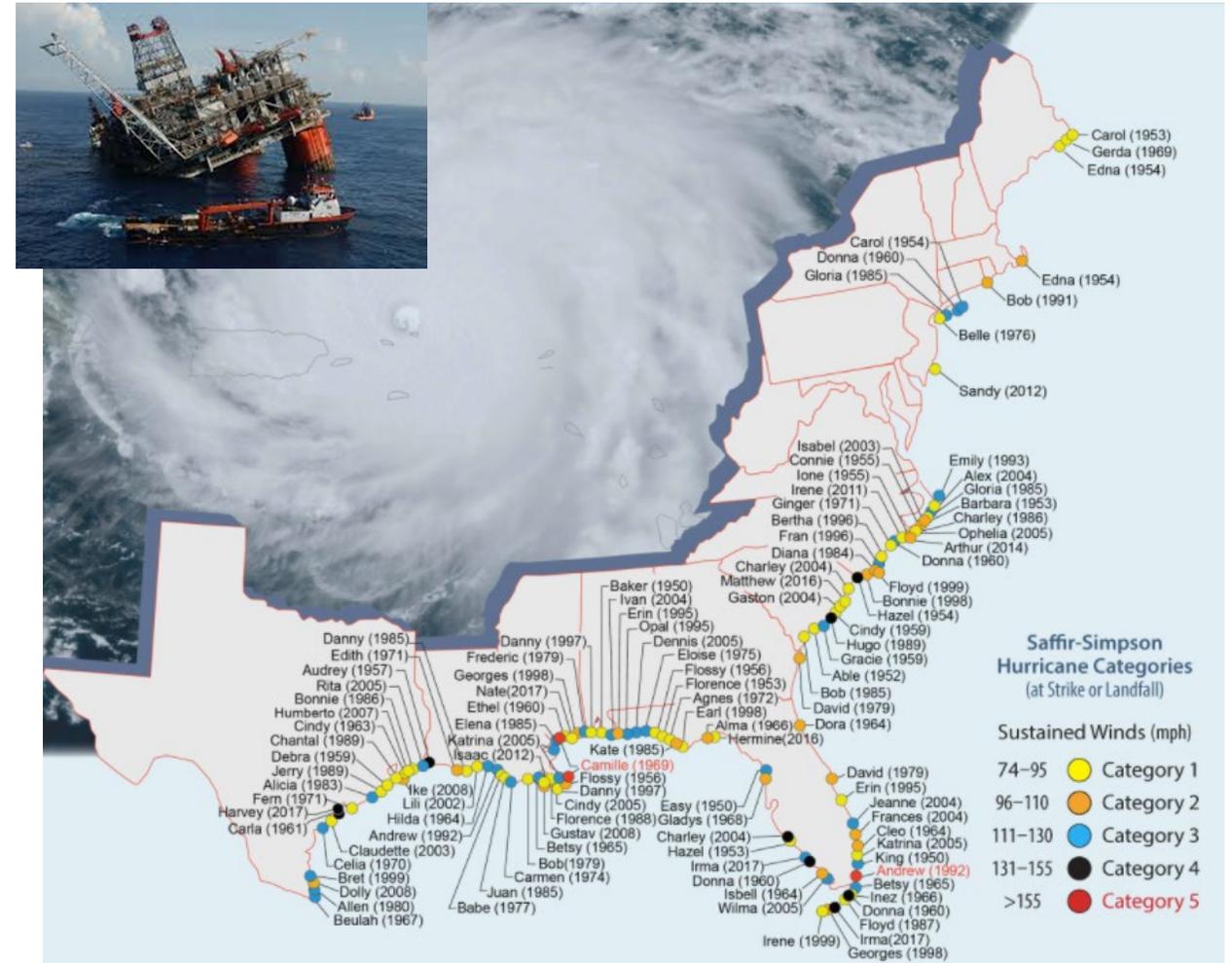
Limiting environmental and community impact and improving safety of offshore energy operations and legacy infrastructure depends on **forecasting and avoiding hazards**

Issue/R&D Need

- **Technology that integrates big data and science-based analytics for offshore hazards does not exist**
- Advanced analytics can offer near real-time assessment of risks, integrate different hazard types, and also forecast vulnerabilities

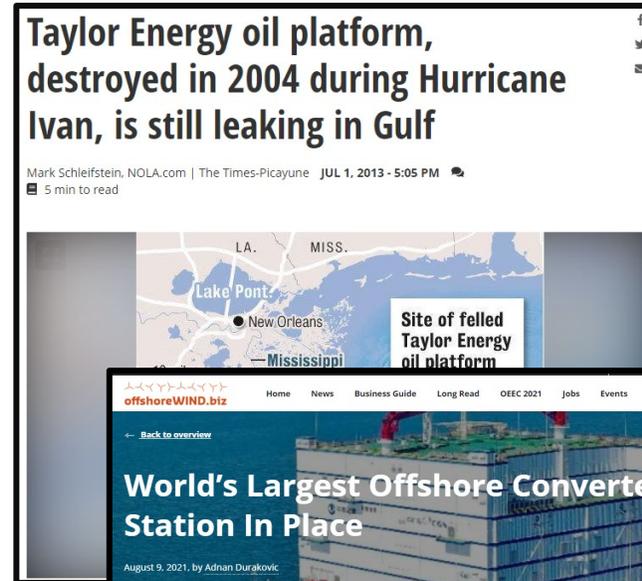
Motivation for Artificial Intelligence/Machine Learning (AI/ML), Data-Driven Offshore Hazard Tools

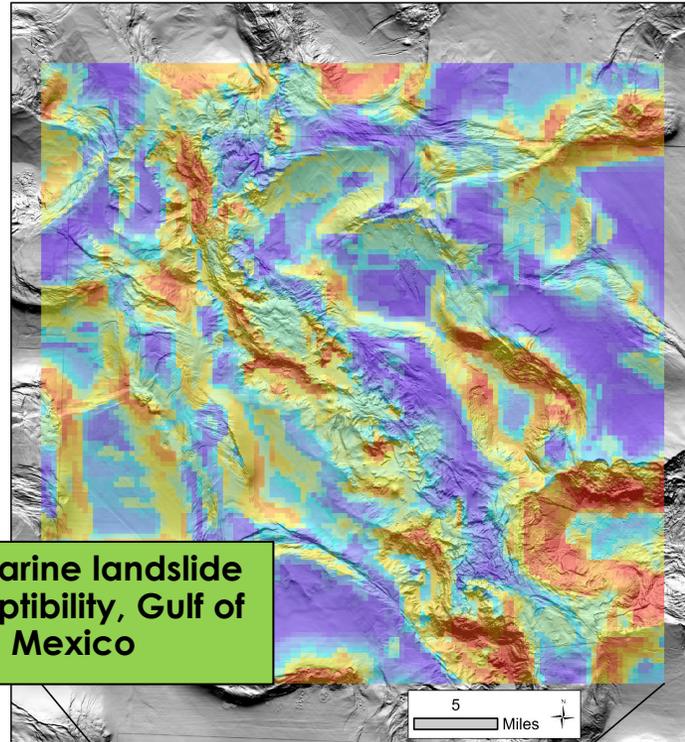
- Demand on offshore Exclusive Economic Zone (EEZ) in the U.S. and around the world is increasing, **with offshore infrastructure expected to increase 50–70% by 2028**
- **Between 2004–2008, 181 structures and 1,673 wells in the Gulf of Mexico were destroyed by five hurricanes**
- Climate change is projected to intensify extreme events, increasing the frequency of major tropical cyclones



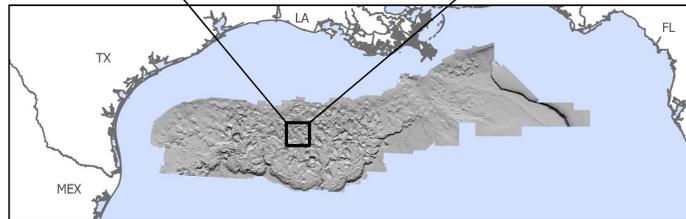
Motivation for AI/ML, Data-Driven Offshore Hazard Tools

- Hazards related to the metocean, seafloor, and subsurface environments include seabed instability, extreme wind/wave/current events, earthquakes, and hazardous material spills.
- **Hazards are often interrelated.** E.g., hurricanes and submarine landslides





Predicted Seafloor Landslide Potential
Low High



- Assessing offshore hazards often requires massive amounts of data and length of time to assess the entire system
- Diverse offshore hazards require various approaches for analytics
- Packaging analytics in a flexible smart software tool improves accessibility and forecasting at multiple scales

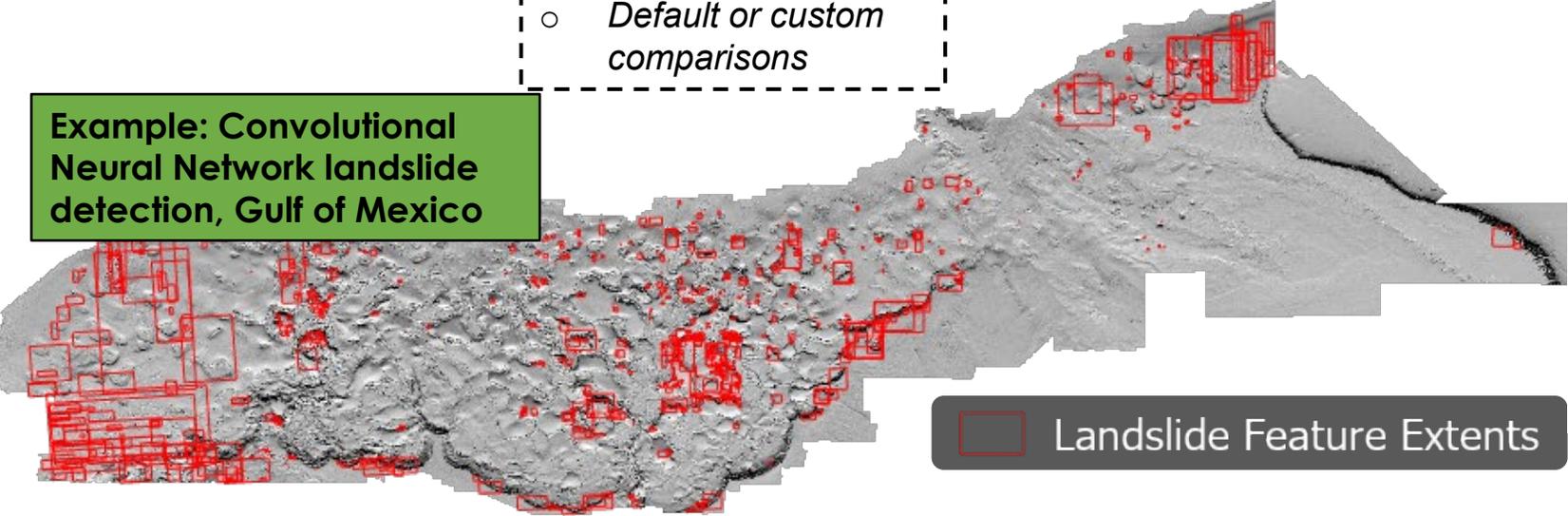
Enabling efficient research for offshore metocean and seafloor hazard assessments

Ocean and Geohazard Analysis Smart Tool Workflow



- Determine hazards and hazard triggers to be analyzed
- Default or custom comparisons

Example: Convolutional Neural Network landslide detection, Gulf of Mexico



□ Landslide Feature Extents

Selecting, training, testing AI/ML is key to an effective workflow

- Identify datasets for diverse hazard analyses
- Develop analytical framework for smart modeling
- Train and validate AI/ML models
- Integrate metocean statistical and probabilistic analyses

OGA Tool Hazard Components

- Analyses are **selected for suitability** of predicting a given hazard or condition
- Each analysis is developed, validated, and prepped for **integration into OGA Tool**

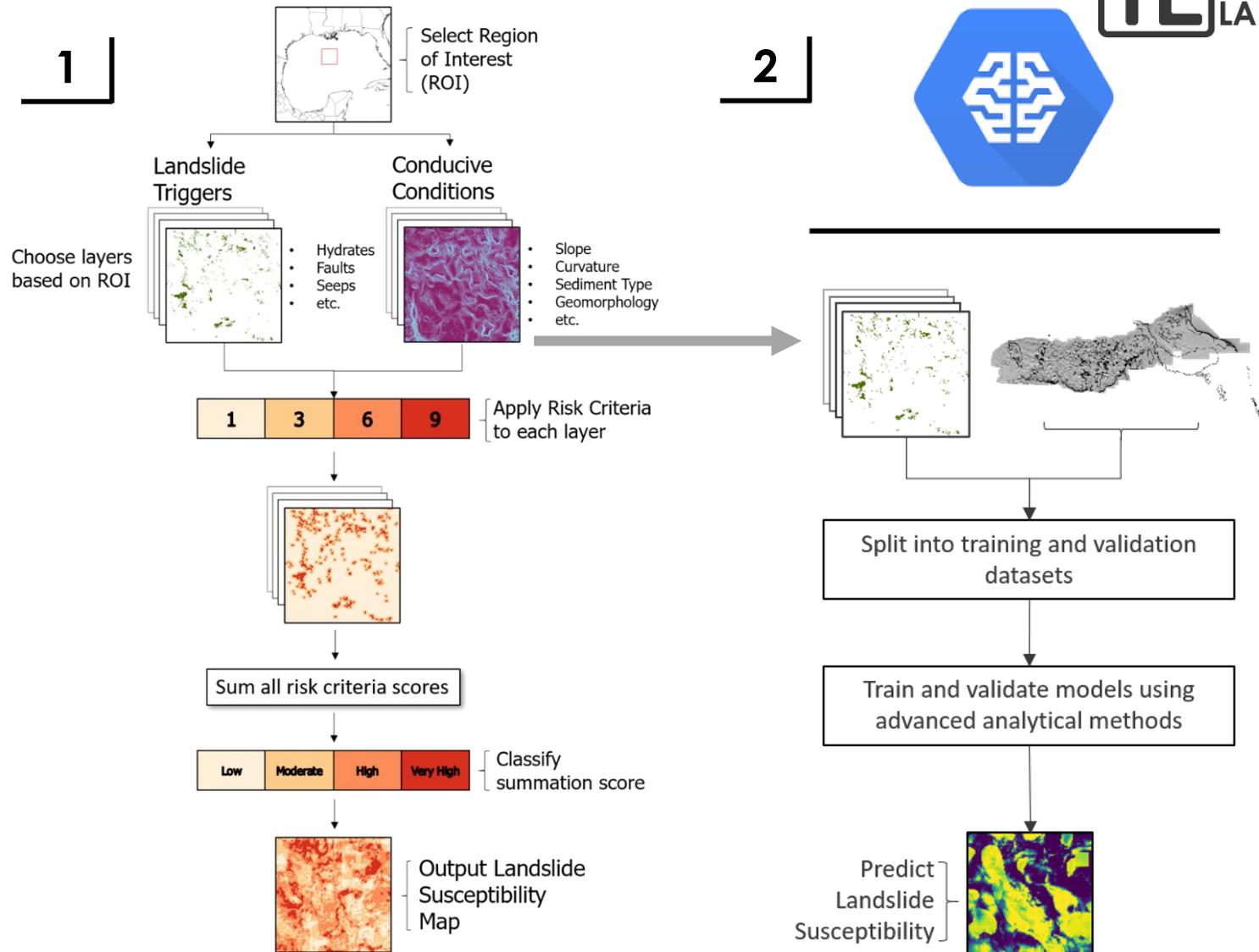
Hazard and/or Process	Analysis Approach(s)
Landslide susceptibility	GIS (risk-based) Machine learning
Landslide detection	Convolutional neural network
Turbidity current susceptibility	GIS interpolation AI/ML spatial analysis
Wave height	Synthetic storm events simulate future extreme events under climate change Generalized extreme value
Wind speed	Generalized extreme value
Current speed	Generalized extreme value
Metocean	Lagrangian Coherent Structures (CIAM)
Loop current eddy shedding	Self-organizing maps

Submarine Landslide Susceptibility Mapping

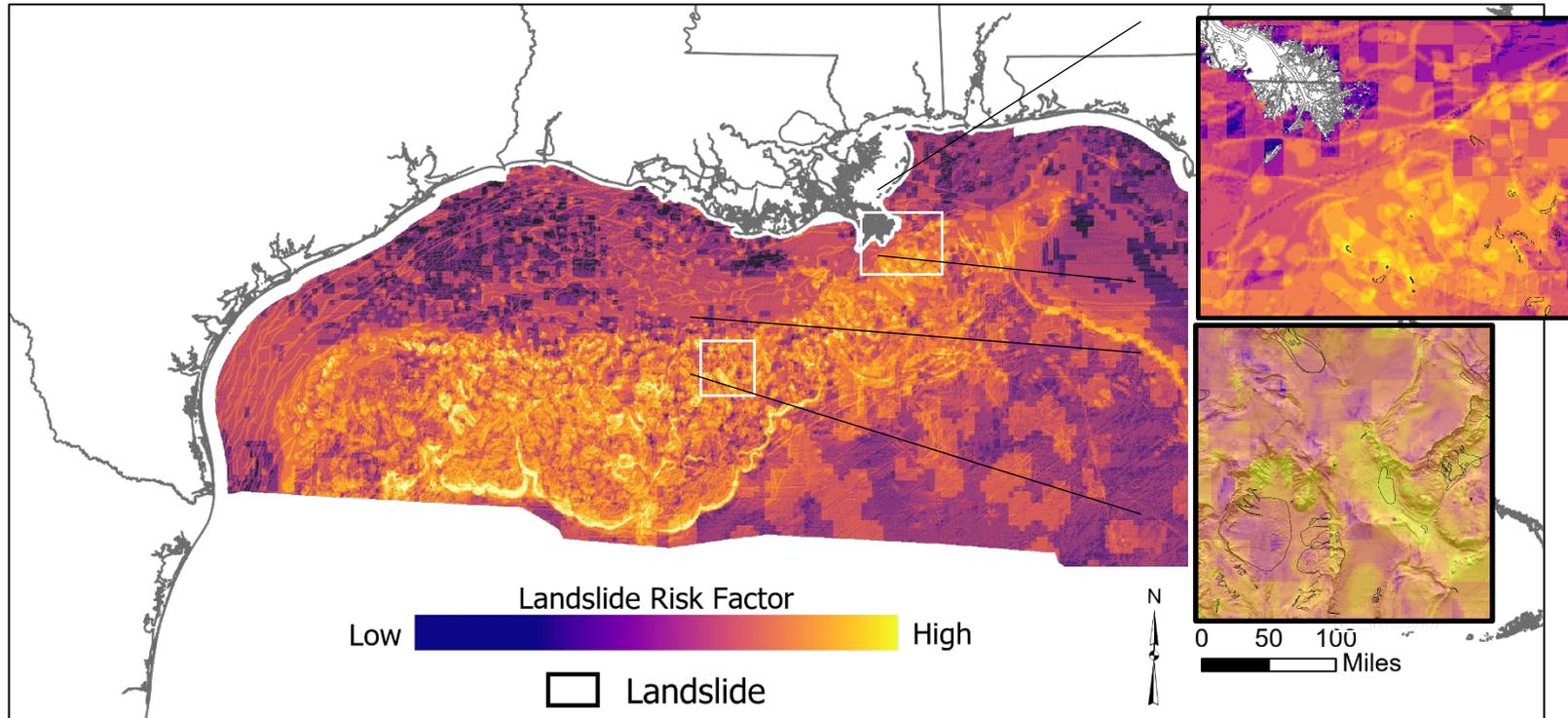
Two approaches for analyzing seafloor landslide potential in the GOM

1. Risk-Based Approach

2. Machine Learning (ML) Approach

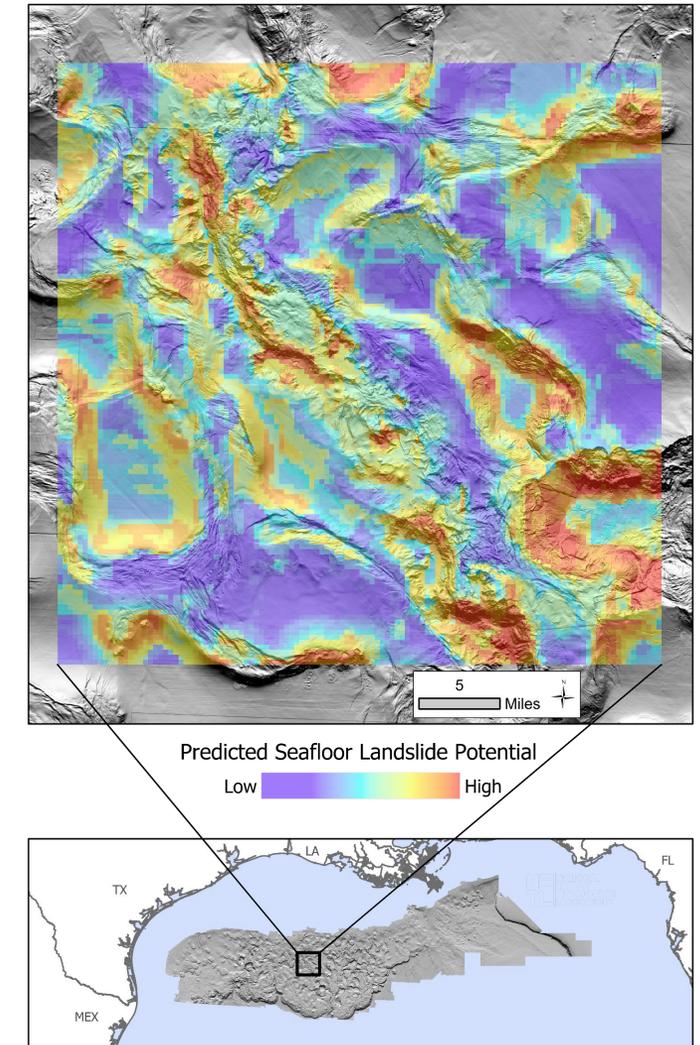


Submarine Landslide Susceptibility Mapping



Two approaches for analyzing seafloor landslide susceptibility in the GOM

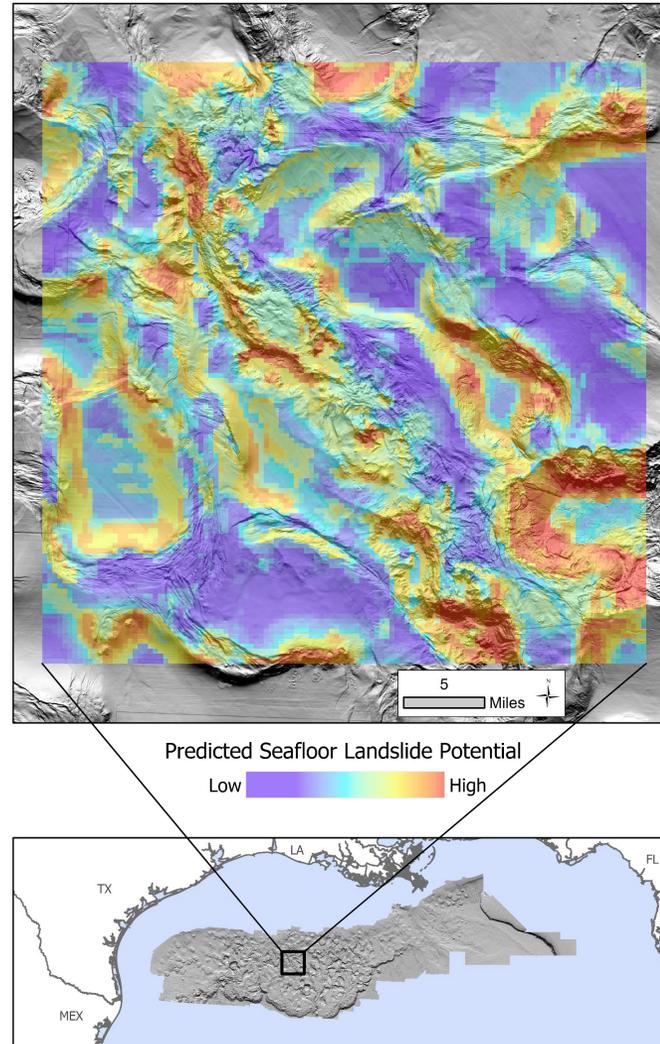
1. *Risk-based GIS Approach (above)*
2. *Machine Learning (ML) Approach (at right)*



Submarine Landslide Susceptibility Mapping

- Utilizing the same input criteria along with **robust ML models** to predict landslide potential
 - **Gradient Boosting Classifier (GBC)**
 - **Artificial Neural Network (ANN)**
- Improved accuracy using **tuning methods**
 - **Hyperparameter random search**
 - **Dimensionality reduction (SVD)**

GBC Model Output



Accuracy evaluated against
validation dataset

GBC: 70.0%

ANN: 65.3%

Submitted to
Natural
Hazards for
publication

Dyer, A.S., Mark-Moser, M., Duran, R., Bauer, J.R. (submitted) Submarine Landslide Susceptibility in the Northern Gulf of Mexico. Natural Hazards, Springer.

Nearshore Adaptation for Submarine Landslide Susceptibility Mapping

- Landslides in the Mississippi River delta front have been recognized to threaten offshore infrastructure since the 1950s
- In 2004, Hurricane Ivan caused a landslide that resulted in the longest lasting spill in U.S. history, with heavy oil sheens still observed as late as 2019
- Our effort leverages big data and ML approaches to assess risk in the region after developing a ML model in deeper waters where the quality of data is favorable
- Nearshore submarine LSM considers shallow waters and effects of waves

Taylor Energy oil platform, destroyed in 2004 during Hurricane Ivan, is still leaking in Gulf
Mark Schleifstein, NOLA.com | The Times-Picayune JUL 1, 2013 - 5:05 PM
5 min to read

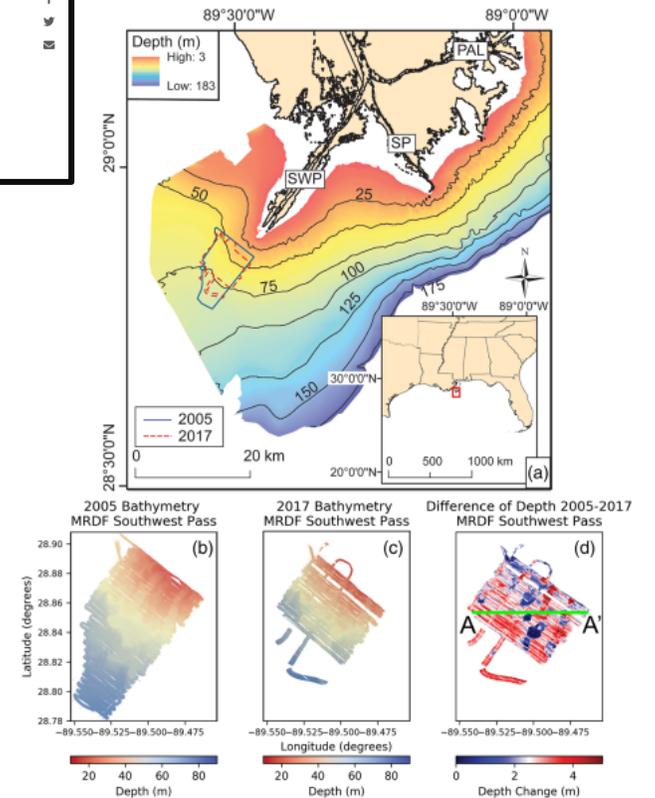
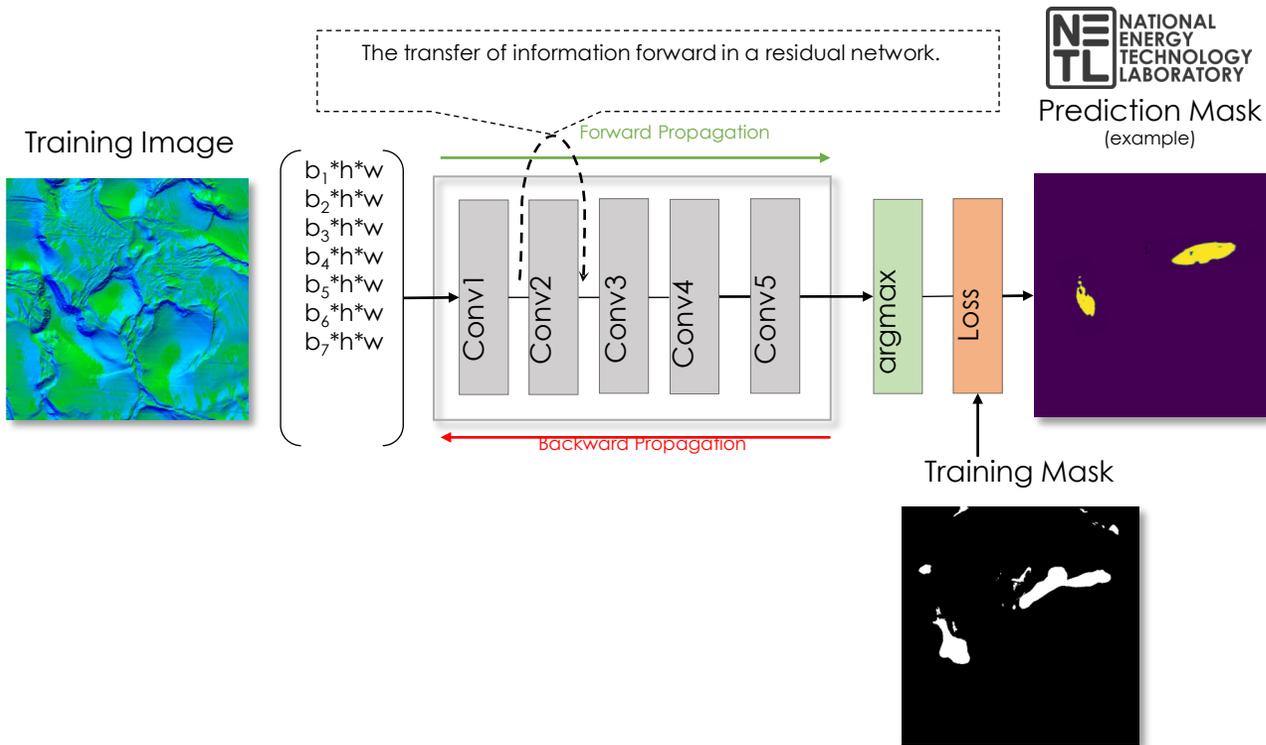


Figure 1. (a) The Mississippi River Delta Front region. Color shaded bathymetry derived from last full-coverage survey of the MRDF region in late 1970s. Blue and red polygons show locations of 2005 and 2017 multibeam surveys, respectively. Acronyms: PAL = Pass a Loutré; SP = South Pass; SWP = Southwest Pass. (b) 2005 25 m² resolution Southwest Pass multibeam bathymetric survey. (c) 2017 100 m² resolution multibeam Southwest Pass bathymetric survey. (d) Difference of depth between 2005 and 2017 bathymetric surveys; the entire area deepened by an average of ~2.6 m in 12 years, with more dynamic depth changes in mudflow zones. The green line represents the 1-D transect in Figure 5.

Obelcz et al., 2020

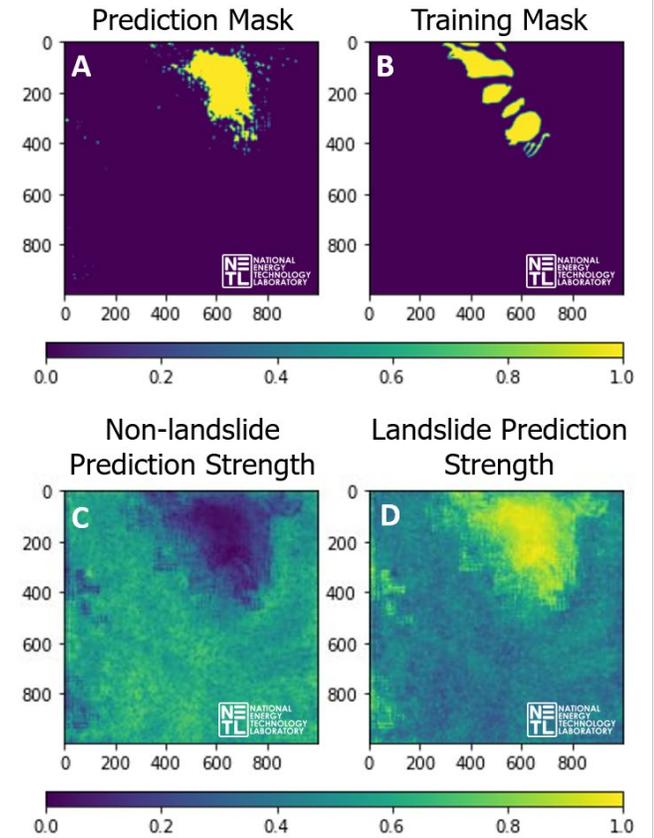
Submarine Landslide Detection

Using high-resolution seafloor images, a data-driven neural network model is used to identify the locations of submarine landslides



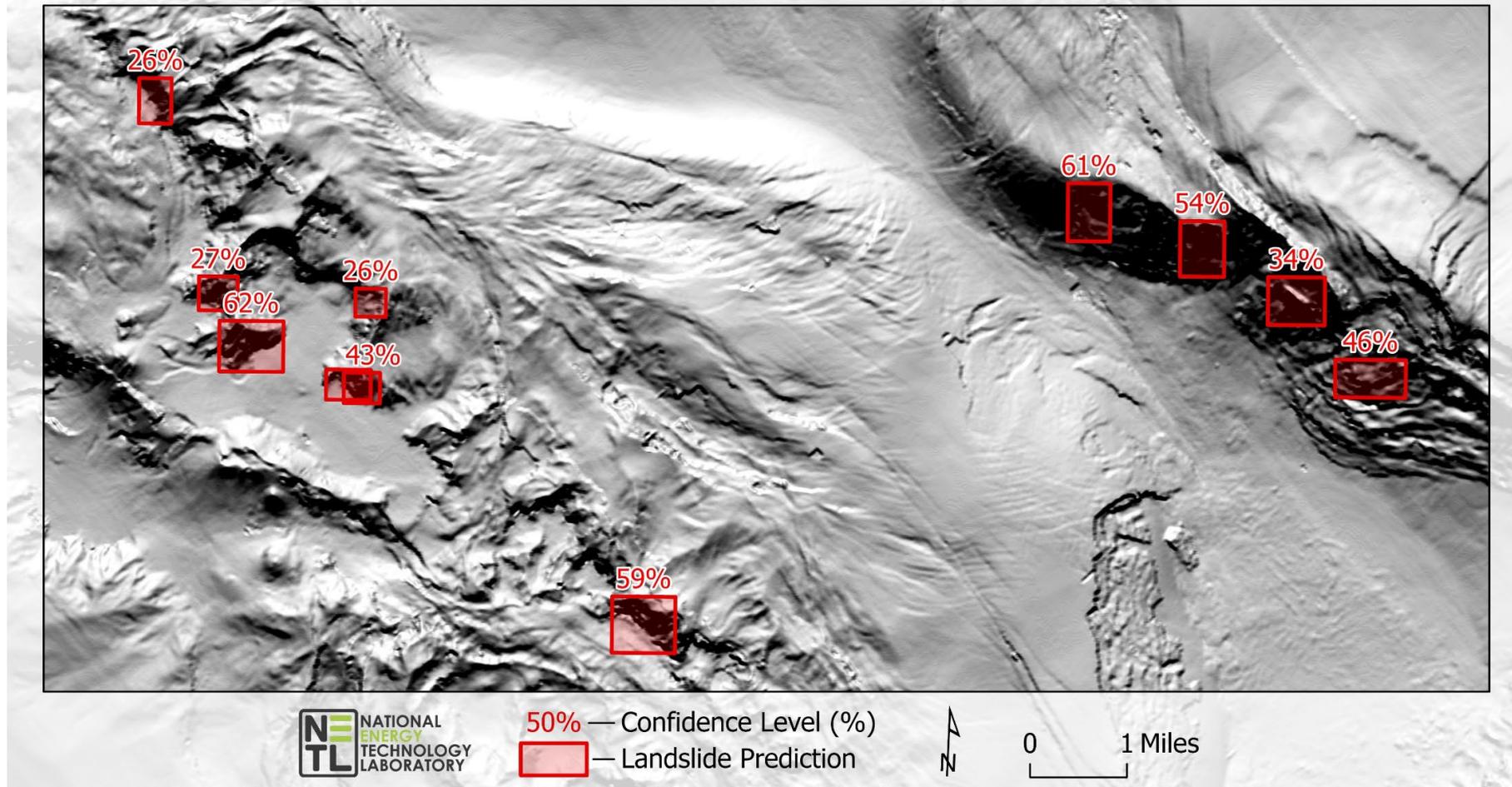
Model Design

- The Fully Convolutional ResNet model was used, a prebuilt network available with the PyTorch framework.
- The model performs semantic segmentation to create an output mask highlighting landslides given an input image.



Landslide Detection Results

Submarine Landslide Detection Using YOLO Algorithm



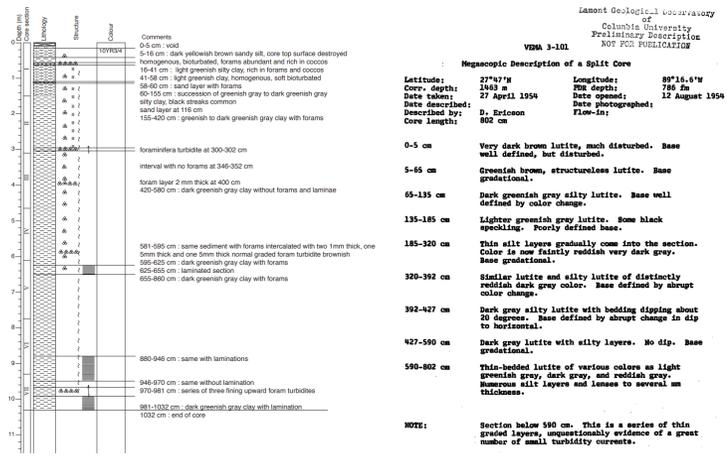
Turbidity Current Hazard Modeling

- Turbidity currents are significant and powerful offshore hazards that are similar but distinct from submarine landslides
- Core analysis is accelerated using automated text extraction and can assist in locating potential turbidity currents
- Locations can be used to inform ML for turbidity current susceptibility mapping and forecasting

“turbidite”

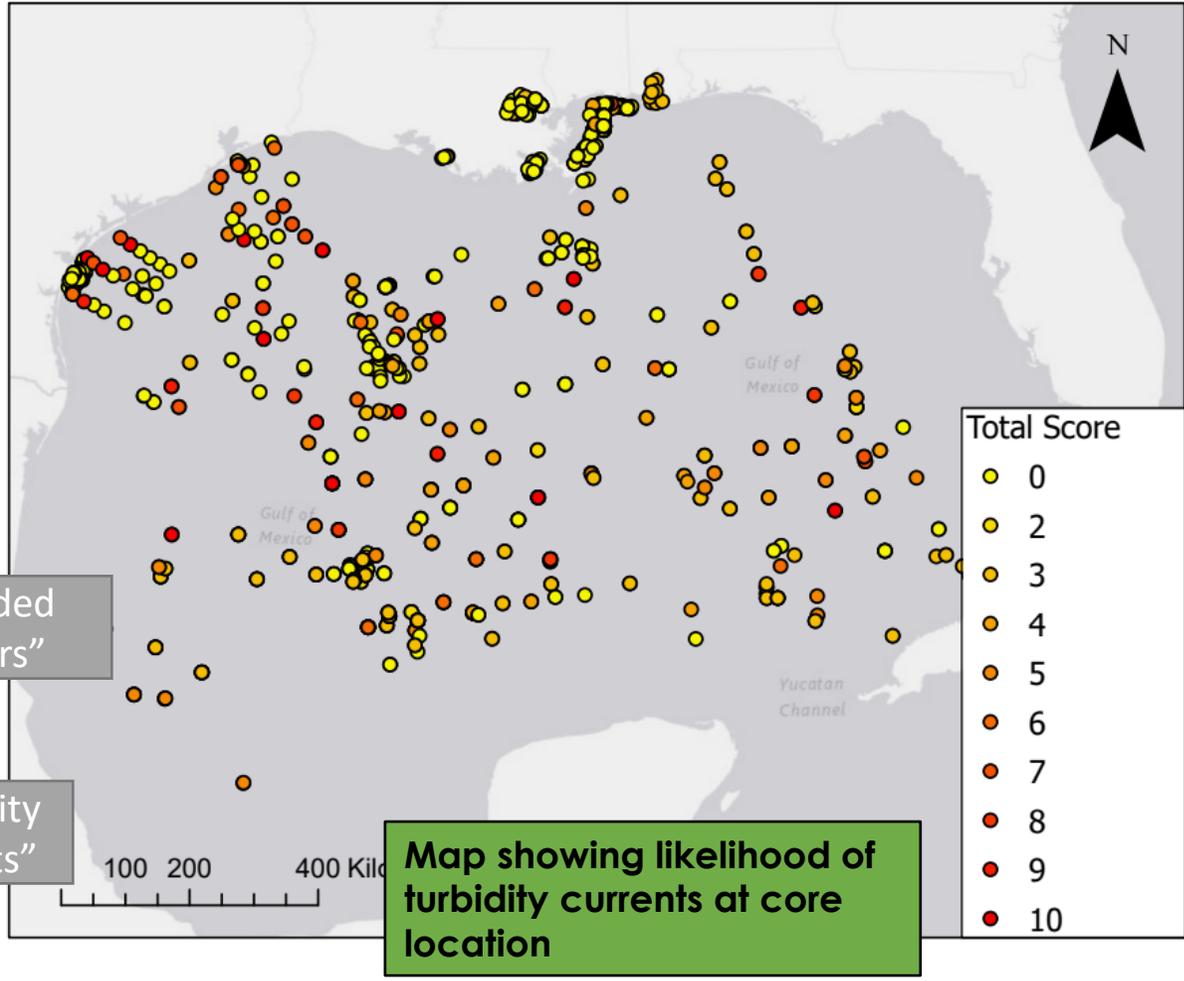
“fining upward”

“foram turbidites”



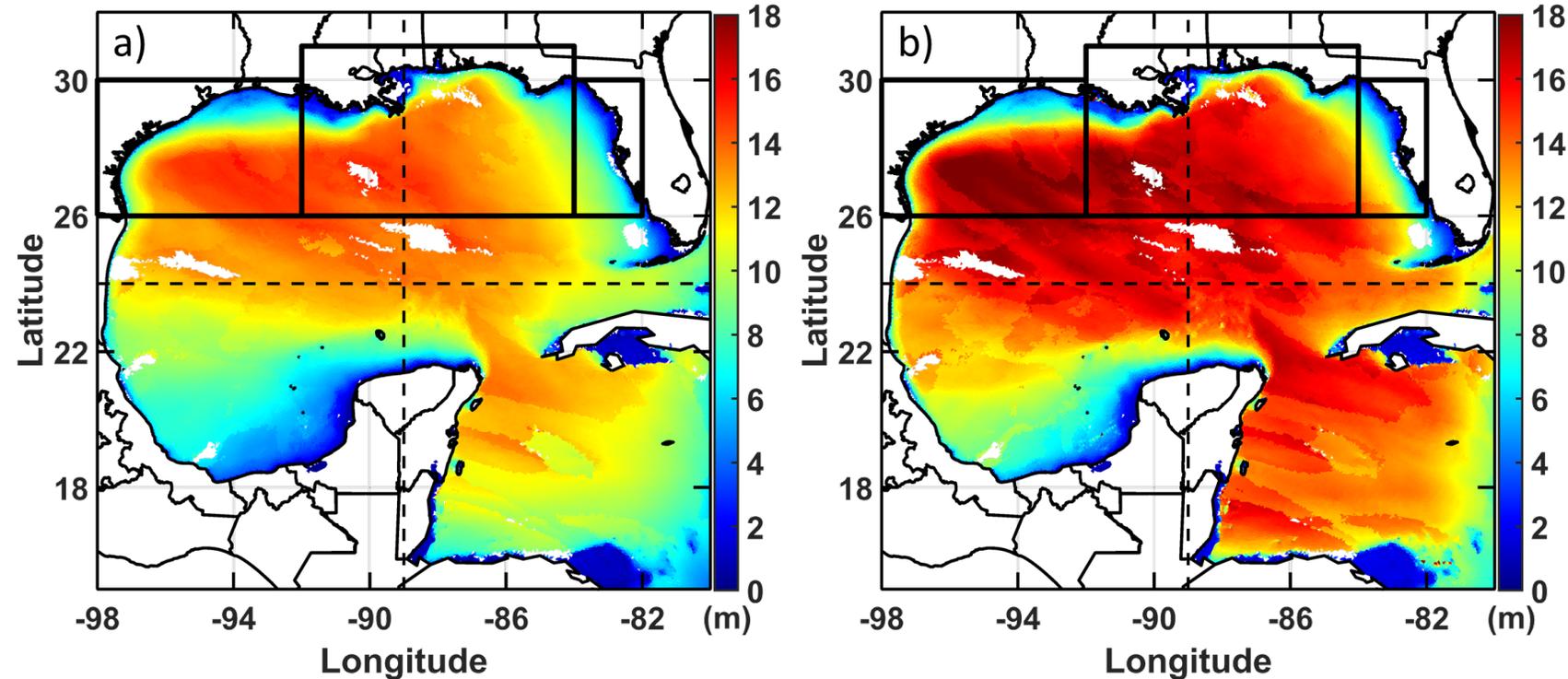
“graded layers”

“turbidity currents”



Wave Modeling Development

- API and industry have needed to revise platform design criteria due to unforeseen extreme waves
- We are creating wave data from synthetic physics-based tropical cyclones using Joule supercomputer
- Critical for risk projections in a changing climate

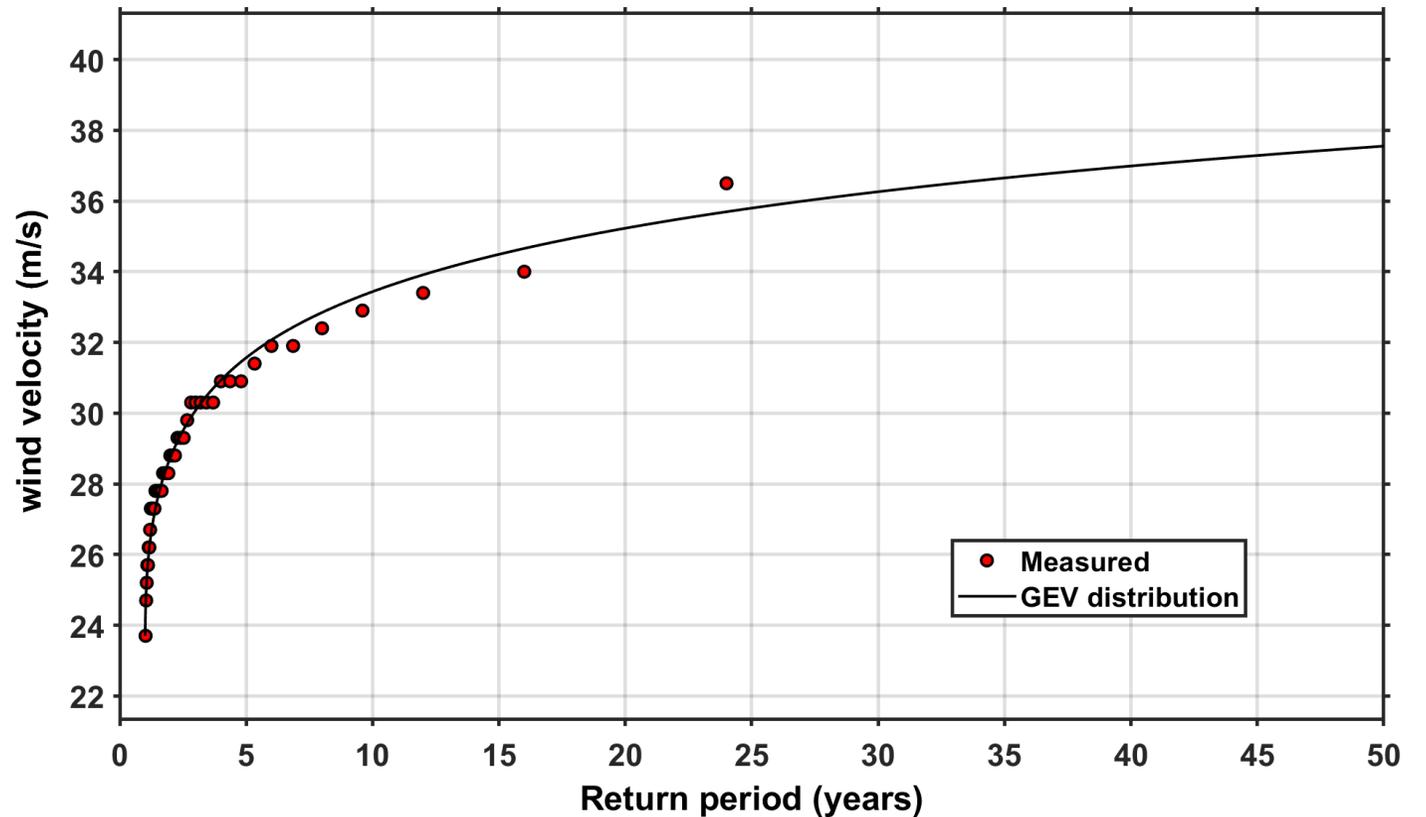


Significant wave height for the 100-years return period obtained from the general circulation model-derived events ensemble for the (a) present and (b) future wave climates. Blank areas denote regions where less than 4 models show the same trend.

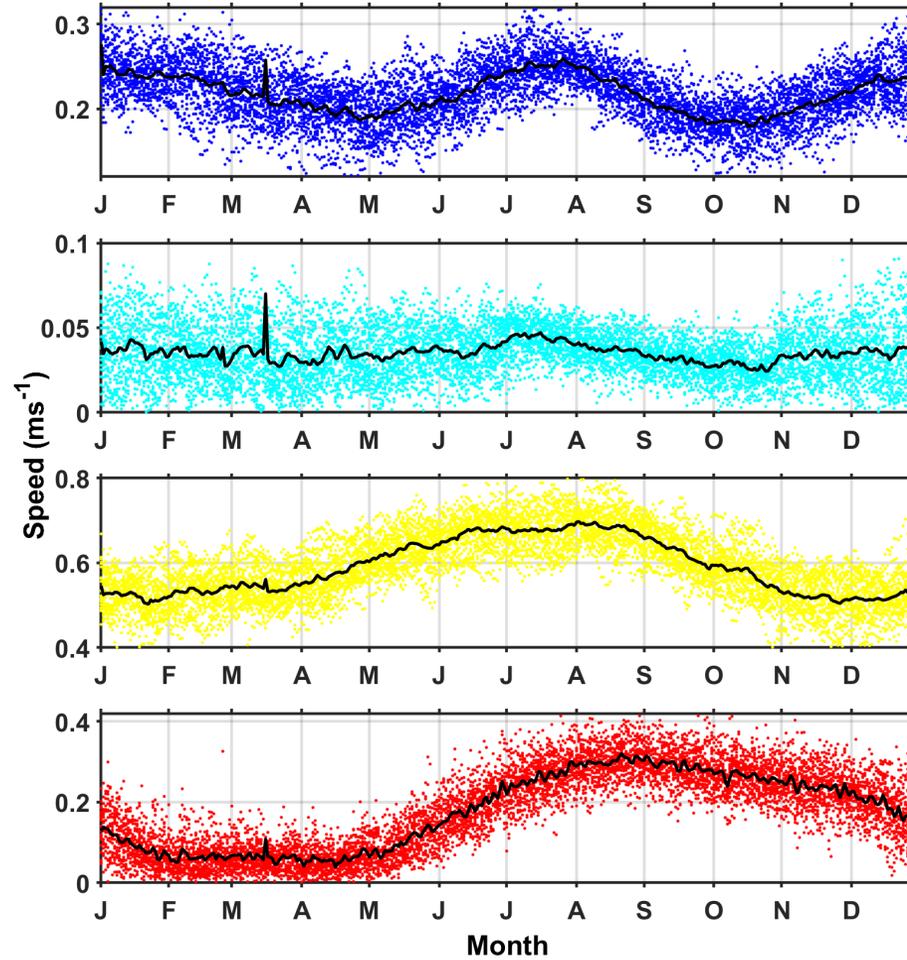
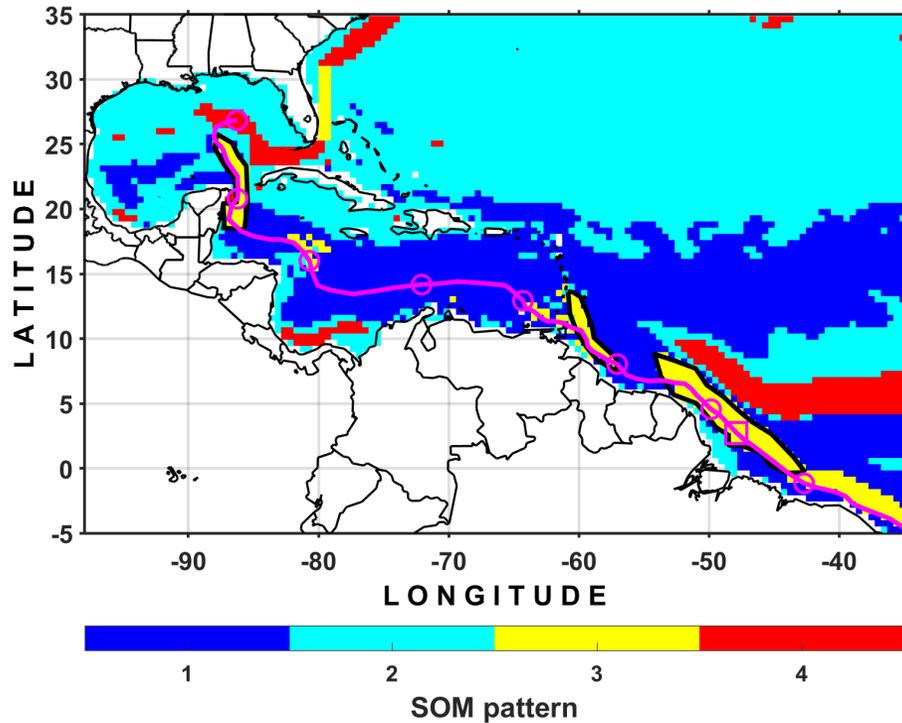
**Generalized Extreme Value
(GEV) distributions for wind
velocity**

**Indicates likelihood of future
extreme events**

$$G(z) = \exp \left\{ - \left[1 + \xi \left(\frac{z - \mu}{\sigma} \right) \right]^{-1/\xi} \right\}$$

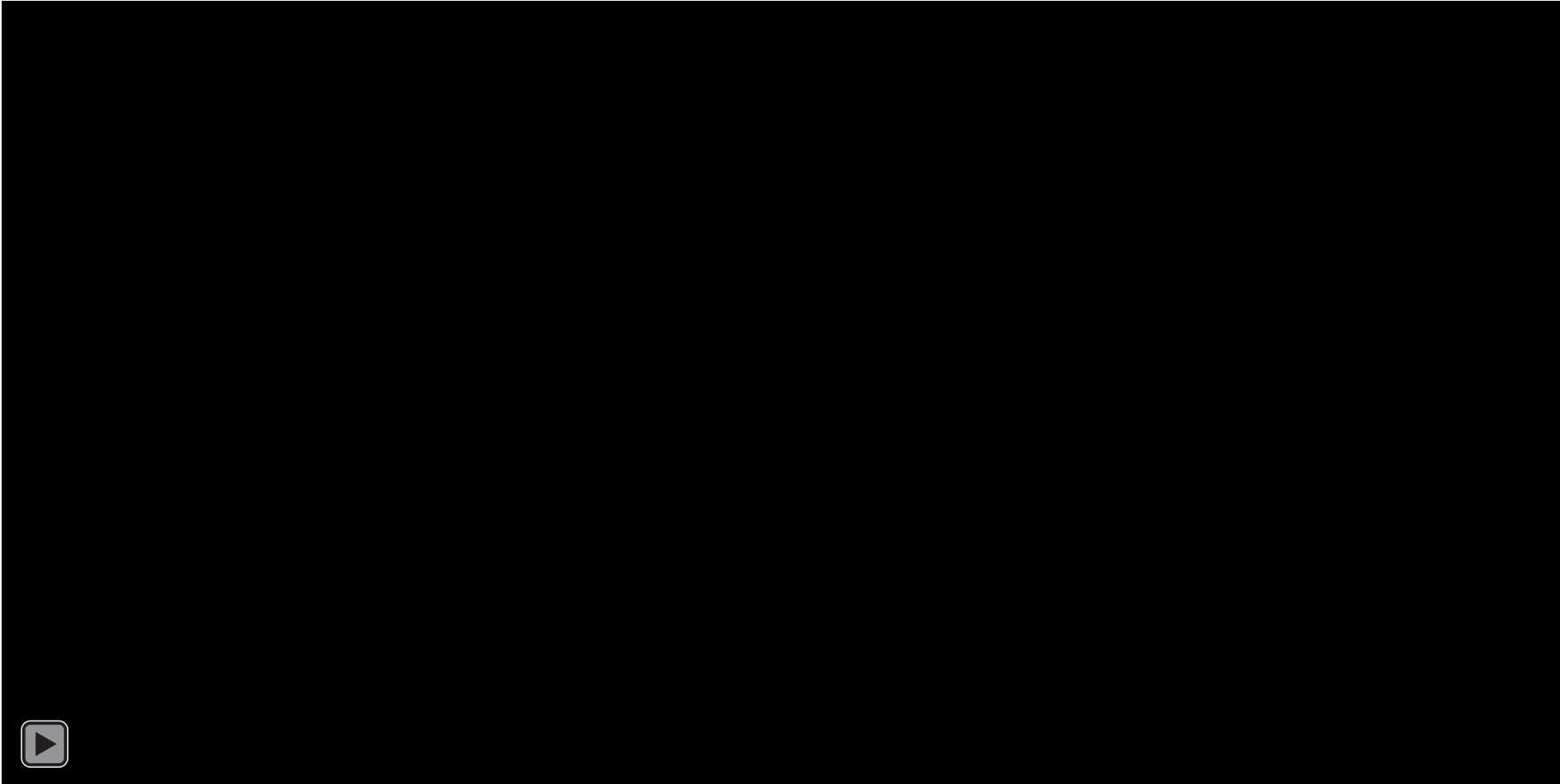


Self-Organizing Maps – An Unsupervised Neural Network



- Temporal patterns from self-organizing maps identifying predictable patterns in sea-surface velocity
- These insights, in combination with advanced analyses of energy and information transfers in the ocean are expected to improve Loop Current predictability

Loop Current Eddy Shedding

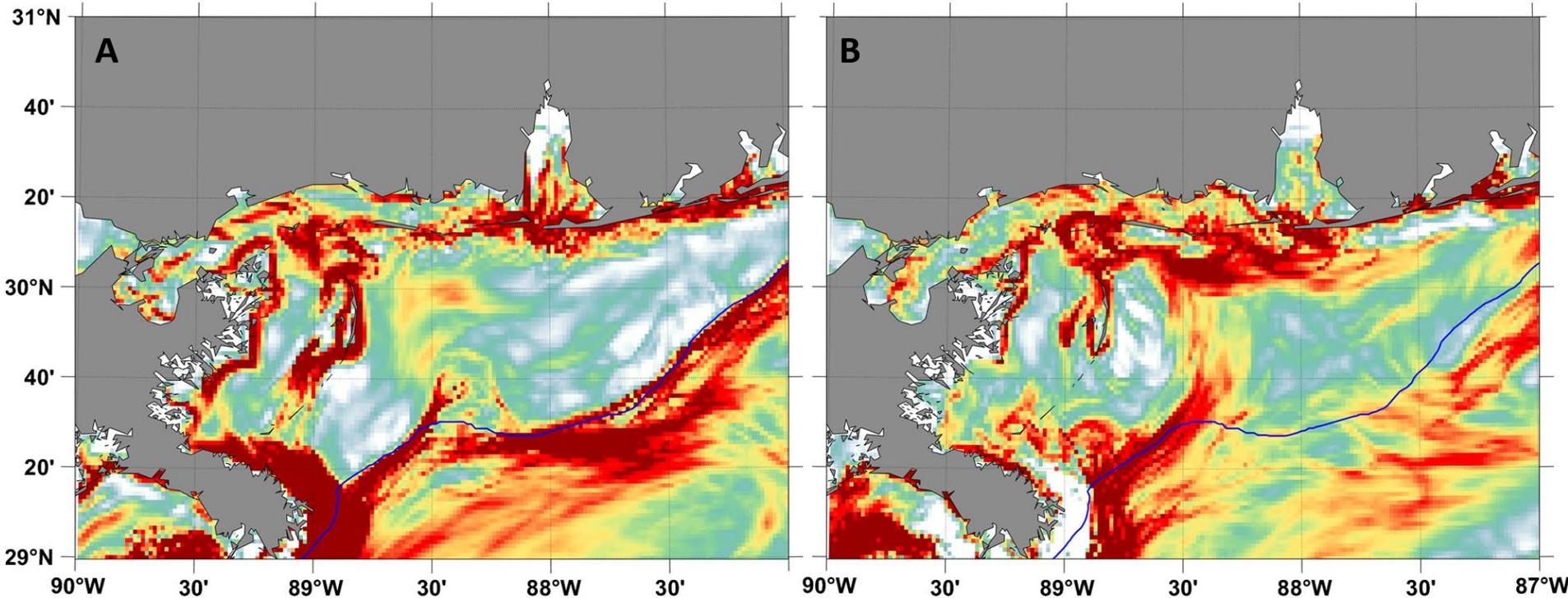


- The Loop Current (LC) and associated eddies are among the most intense currents in the world and are a major concern for offshore infrastructure
- Predicting LC eddy shedding has been intractable so far
- Insights from self-organizing maps are leading to novel analyses of Loop Current eddy shedding events using oceanic energy transfers

CIAM Model Updates

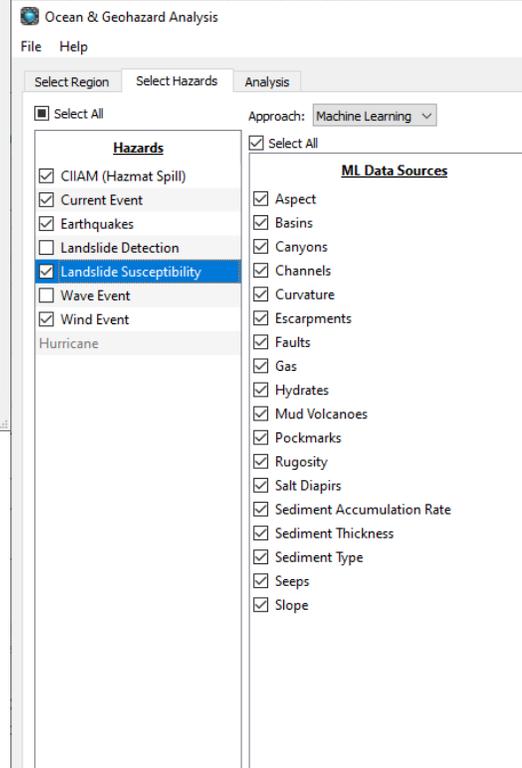
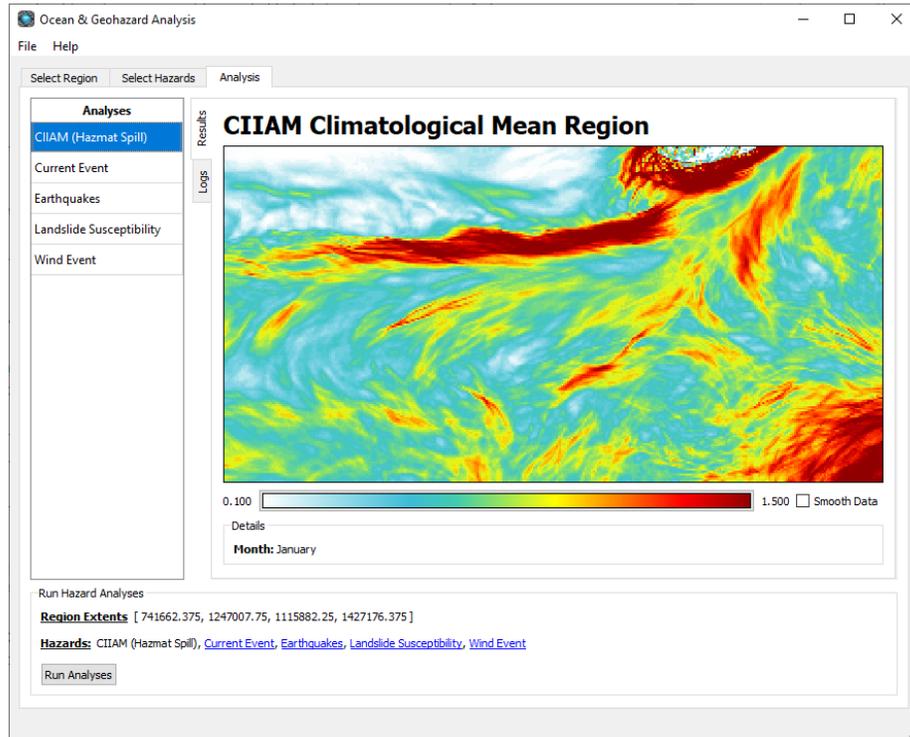


Metocean Pathways:
Red=attracting
White=isolated



CIAM outputs for the ocean near southeastern Louisiana in (A) winter and (B) spring, showing transport barrier as open allowing particulate to reach coastline

OGA Smart Tool Interface



Smart Tool allows users to interact with their data and select or integrate appropriate models

Produces forecasts of areas more susceptible to metocean and seafloor hazards

Collaboration and External Interest



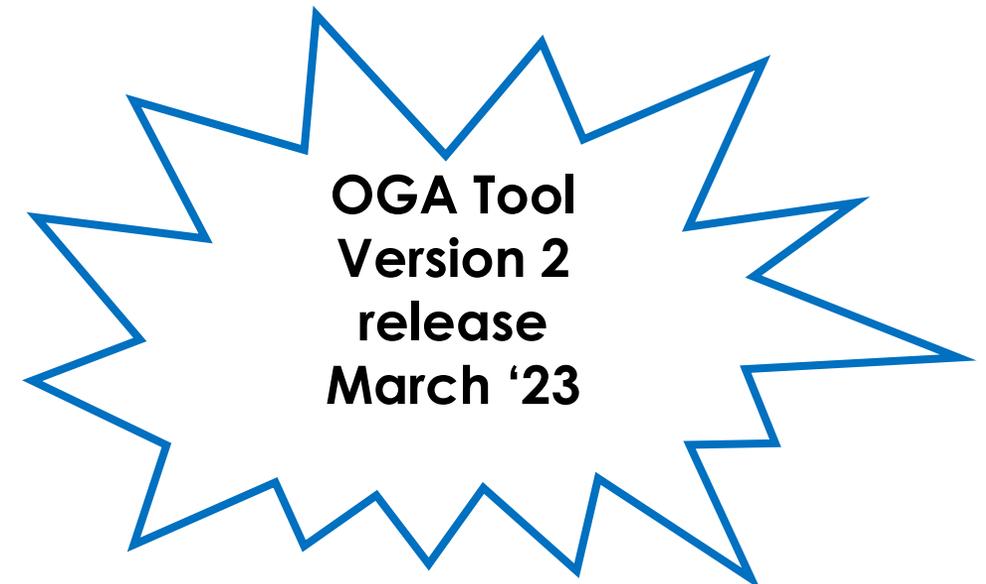
External CIAM Users			
Country	Research Institute.	Study region	Status
Spain	ICM Marine Science Institute Spain.	Mediterranean	Publication in progress
India	National Institute of Oceanography India	Gulf of Bengal	Publication in progress
Mexico	Engineering & Coastal Processes UNAM Mexico	Caribbean & Loop Current	Publication in progress
Brazil	National Institute for Space Research Brazil	Tropical Atlantic	Gouveia et al (2021). https://www.nature.com/articles/s41598-020-79386-9
Mexico	CICESE Ensenada Center for Scientific Research and Higher Education, Mexico	Deep GOM	Maslo, A., et al. (2020). https://doi.org/10.1016/j.jmarsys.2019.103267
Mexico	CICESE Ensenada Center for Scientific Research and Higher Education, Mexico	NW GOM	Gough, M. K., et al. (2019). https://doi.org/10.1175/JPO-D-17-0207.1
United Kingdom	National Oceanography Centre Marine Systems Modelling Group	North Sea and Caribbean	Preliminary results obtained
Saudi Arabia	Red Sea Modeling and Prediction Group KAUST	Red Sea	Preliminary results obtained
Mexico	Consortium for Sargassum forecasts (CICESE, UNAM, ECOSUR)	Caribbean and GOM	Preliminary results obtained

MOU for Collaboration OASIS:
BOEM and NETL
MOU AGMT-1082.AMD1



Next Steps for OGA Tool

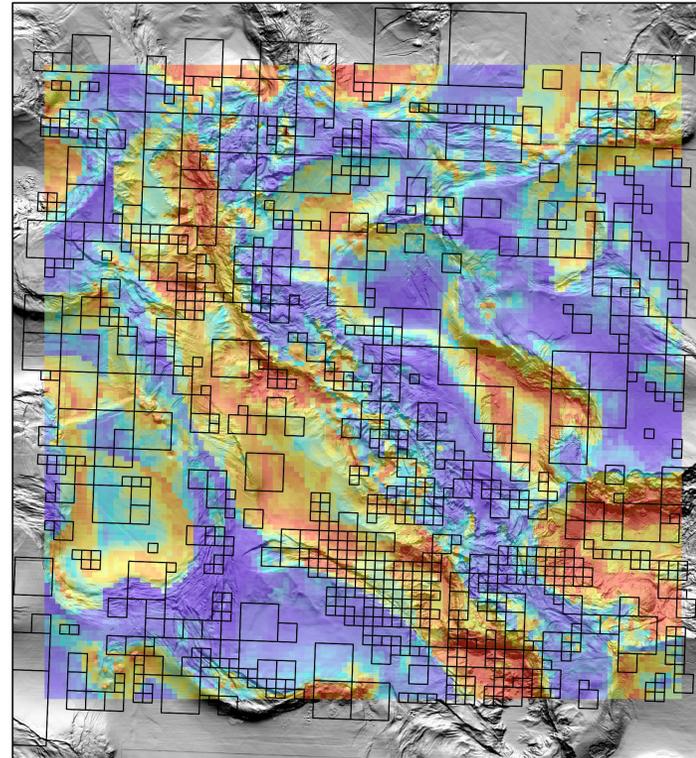
- Integrating analyses for turbidity currents, nearshore submarine landslide susceptibility, extreme waves and wind in a changing climate, Loop Current predictability
- Assembly of database containing metocean and seabed datasets that feed OGA analyses
- Strategize conversion of OGA Tool to online-accessible web application



Email: MacKenzie Mark-Moser
Mackenzie.mark-moser@netl.doe.gov

Key Takeaways

- Technology that integrates big data and science-based analytics for offshore hazards
- Advanced analytics can offer near-real time assessment of risks and also forecast vulnerabilities
- Smart Tool:
 - adapts to data availability/quality
 - adapts to different regions
 - flexible to integrate NETL tools and user tools for advanced predictive and spatial analysis
- Next steps are to integrate additional hazard analyses, validate tool, and strategize conversion to online tool

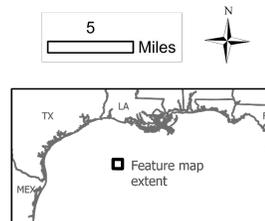


Predicted Seafloor Landslide Potential
Low  High

Variable Grid Method

Uncertainty determined using radial mean accuracy

-  Grid cells with accuracy > 0.65
-  Grid cells with accuracy > 0.45
-  Grid cells with accuracy > 0.28



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

Values Delivered

Advancing the current state of knowledge, supporting offshore activities, forecasting hazards to maintain environmental integrity that may evolve with a changing climate

Improved characterization of metocean and seabed related hazards will help to prevent catastrophic incidents as human and engineered systems integrate with natural systems in the offshore environment

NETL RESOURCES

VISIT US AT: www.NETL.DOE.gov



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Jennifer.bauer@netl.doe.gov



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



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Advanced Offshore Research Task 6 Timeline

Research Problem:

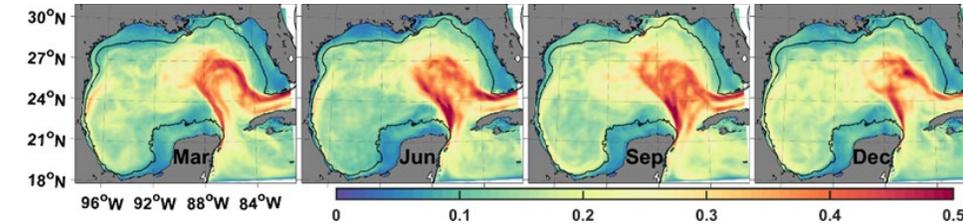
- Changes in the ocean environment (e.g., mudslides or burial from subsea currents, strong weather events or natural fluctuations) have been linked to billions of dollars of impacts. Climate change is expected to intensify many of these problems.
- These events can have a significant effect on the success and longevity of offshore infrastructure, as well as affect safety and cost during exploration, production, and storage activities.

Research Approach:

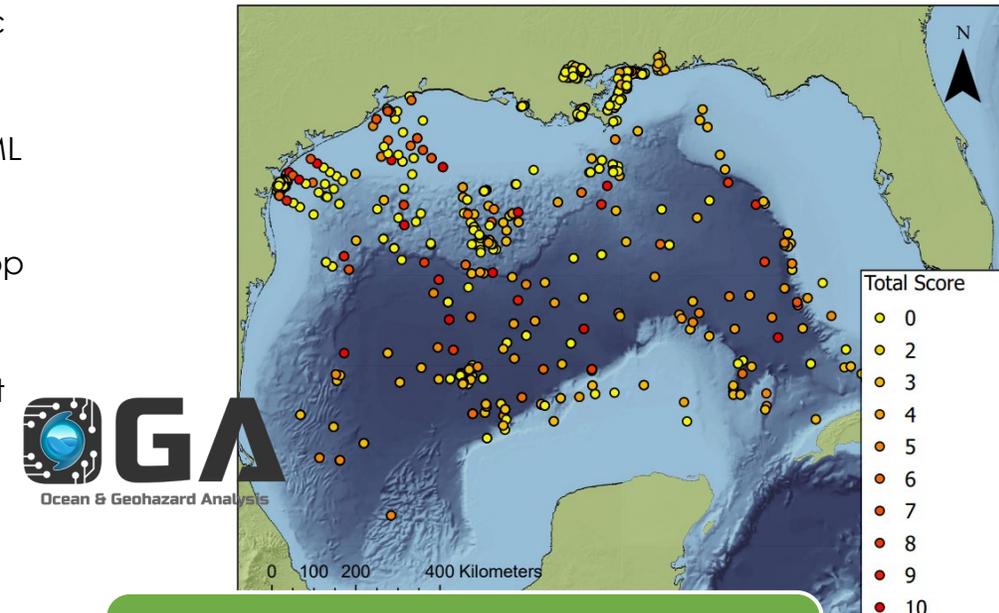
- Determine current state of knowledge regarding hazardous metocean and bathymetric conditions, and data availability regarding these conditions and historic events.
- **EY19-EY21:** Evaluated if AI/ML models can be developed to better identify current hazardous metocean and bathymetric conditions. Developed, trained, and tested AI/ML models to identify conditions and forecast changes and vulnerabilities to offshore infrastructure. Refined Smart Tool to host AI/ML models and develop user interface. Developed forecasting and integrated selected hazard types into tool. Released desktop version at end of EY.
- **EY22:** Refine analytical logic and smart tool functionalities through user testing and development. Build metocean and seabed hazard database for release on EDX. Report research in technical report or publication.
- **EY23+:** Strategize conversion of OGA Tool to online platform. Submit integrated seabed hazard database for release to EDX. Continue to produce technical publication(s).

Benefit:

- Improved characterization of metocean and seabed related hazards in the offshore can help prevent catastrophic incidents that impact the environment, coastal communities, and their economies while supporting offshore energy and carbon storage efforts.



Average bottom current velocity (12 yr.)



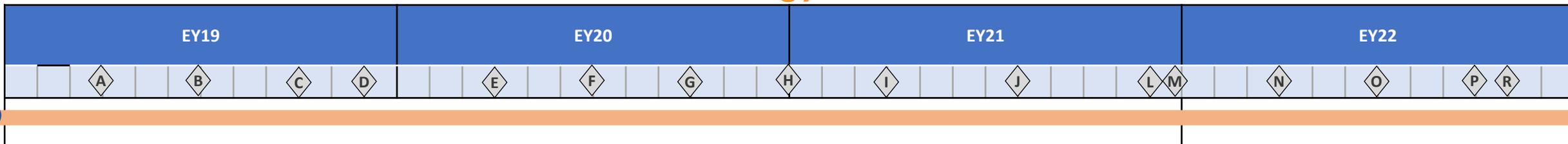
Potential locations of turbidity currents throughout the Gulf of Mexico based on text extraction core analysis

Offshore Unconventional FWP

Key Team Members: PI – Jennifer Bauer - CO-PI – Mackenzie Mark-Moser, Rodrigo Duran



Task 6: Infrastructure and Metocean Technology



Milestones

Number	Date	Description
EY21.6.I	06/2021	List summarizing identified improvements and enhancements for analytical logic and smart tool.
EY21.6.L	02/2022	Internal release of the Ocean & Geohazard Analysis tool, desktop version, to EDX.
EY21.6.M	03/2022	Evaluate TRL for smart tool and determine if additional development or enhancements are needed to obtain target TRL.
EY22.6.N	06/2022	List summarizing tool enhancements priorities identified by user testing on OGA Version 1.
EY22.6.O	09/2022	Draft manuscript(s) of individual smart tool model(s) or algorithm(s) completed
EY22.6.P	12/2022	List optimizations made to the Ocean & Geohazard Analysis tool.
EY22.6.Q	01/2023	Assemble metocean and seafloor database to support smart tool analysis.
EY22.6.R	06/2023	Strategize conversion of Ocean & Geohazard Analysis tool to online platform.
EY22.6.S	10/2023	Update integrated metocean and seabed hazard database for management review and approvals to release on EDX.
EY22.6.T	12/2023	Outline a technical report or additional publications.

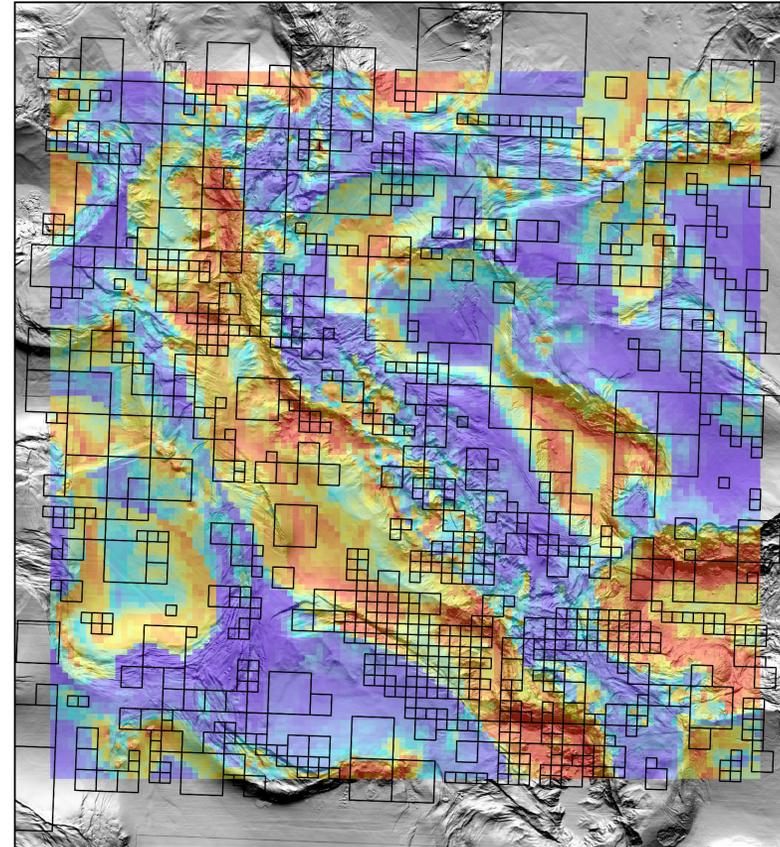
Chart Key

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

Landslide Susceptibility Results

ML Approach with Variable Grid Method

- The **Variable Grid Method (VGM)** (Bauer & Rose, 2015) utilized to visualize spatial uncertainty.
- **Smaller grid sizes** indicate a **higher certainty** of model predictions for that region while **larger grid sizes** indicate **lower certainty**.



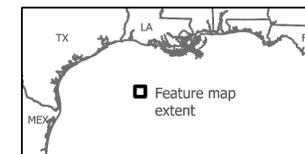
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-  Grid cells with accuracy > 0.45
-  Grid cells with accuracy > 0.28



Publications, Datasets & Presentations



Publications

- Dyer, A.S., Mark-Moser, M., Duran, R., Bauer, J.R. (submitted) Submarine Landslide Susceptibility in the Northern Gulf of Mexico. Natural Hazards, Springer.
- Alec Dyer, Scott Pantaleone, MacKenzie Mark-Moser, Andrew Bean, Paige Morkner, Samuel Walker, Jennifer Bauer, Historic Submarine Landslides in the Northern Gulf of Mexico, 8/8/2022, <https://edx.netl.doe.gov/dataset/historic-submarine-landslides-in-the-northern-gulf-of-mexico>, DOI: 10.18141/1879673
- Duran, R., T. Nordam, M. Serra and C. Barker (2021). Horizontal transport in oil-spill modeling. Book chapter in Marine Hydrocarbon Spill Assessments, Elsevier. <https://arxiv.org/abs/2009.12954>
- Nordam T., J. Skancke, R. Duran and C. Barker (2021). Vertical transport in oil spill modeling. Book chapter in Marine Hydrocarbon Spill Assessments, Elsevier. <https://arxiv.org/abs/2010.11890>
- Nordam, T. & R. Duran (2020). Numerical integrators for Lagrangian oceanography. Geoscientific Model Development. <https://gmd.copernicus.org/preprints/gmd-2020-154/>.
- Gouveia, M. B., R. Duran, J. A. Lorenzetti, A. T. Assireu, R. Toste, L. P. de F. Assad and D. F. M. Gherardi (submitted, revision in progress, 2020). Persistent meanders and eddies lead to quasi-steady Lagrangian transport patterns in a weak western boundary current. <https://arxiv.org/abs/2008.07620>
- Zhang, R., P. Wingo, R. Duran, K. Rose, J. Bauer, R. Ghanem (2020). Environmental Economics and Uncertainty: Review and a Machine Learning Outlook. Oxford Encyclopedia of Environmental Economics. <https://doi.org/10.1093/acrefore/9780199389414.013.572>.
- Gough M. K., F. J. Beron-Vera, M. J. Olascoaga, J. Sheinbaum, J. Jouenno, R. Duran (2019). Persistent Lagrangian transport patterns in the northwestern Gulf of Mexico. J. Phys. Oceanogr., 49, 353–367, <https://doi.org/10.1175/JPO-D-17-0207.1>
- Duran, R., F. J. Beron-Vera, M. J. Olascoaga (2018). Extracting quasi-steady Lagrangian transport patterns from the ocean circulation: An application to the Gulf of Mexico. Scientific Reports, 8(1), 5218. <https://www.nature.com/articles/s41598-018-23121-y>
- Appendini C. M., P. Ruiz-Salcines and R. Duran (in preparation). Tropical cyclone waves under climate change in the Gulf of Mexico
- Kurczyn, J. A., R. Duran, E. Beier, and A. J. Souza (2021). On the advection of upwelled water on the western Yucatan Shelf. Frontiers in Marine Science. <https://doi.org/10.22541/au.162126717.71153804/v1>

Presentations

- 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
- Zaengle, D., Dyer, A., Duran, R., Mark-Moser, M., Rose, K., Bauer, J., Wingo, P. **Accepted**. Seafloor Landslide Detection in the Gulf of Mexico Using Computer Vision. IMAGE 2021
- Dyer, A., Duran, R., Mark-Moser, M., Rose, K., Bauer, J., Zaengle, D., Wingo, P. **2021**. Geohazard Analysis of Seafloor Landslide Potential for Infrastructure Protection. Esri User Conference. July 12-16, 2021.
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