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PREDICTING SHOALING ON THE MISSISSIPPI RIVER USING CROWD- SOURCED AND USACE OPERATIONAL DATA

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SmartPort

*WEDA Gulf Coast
Chapter 2022*

BACKGROUND & INCEPTION

PHASE 1



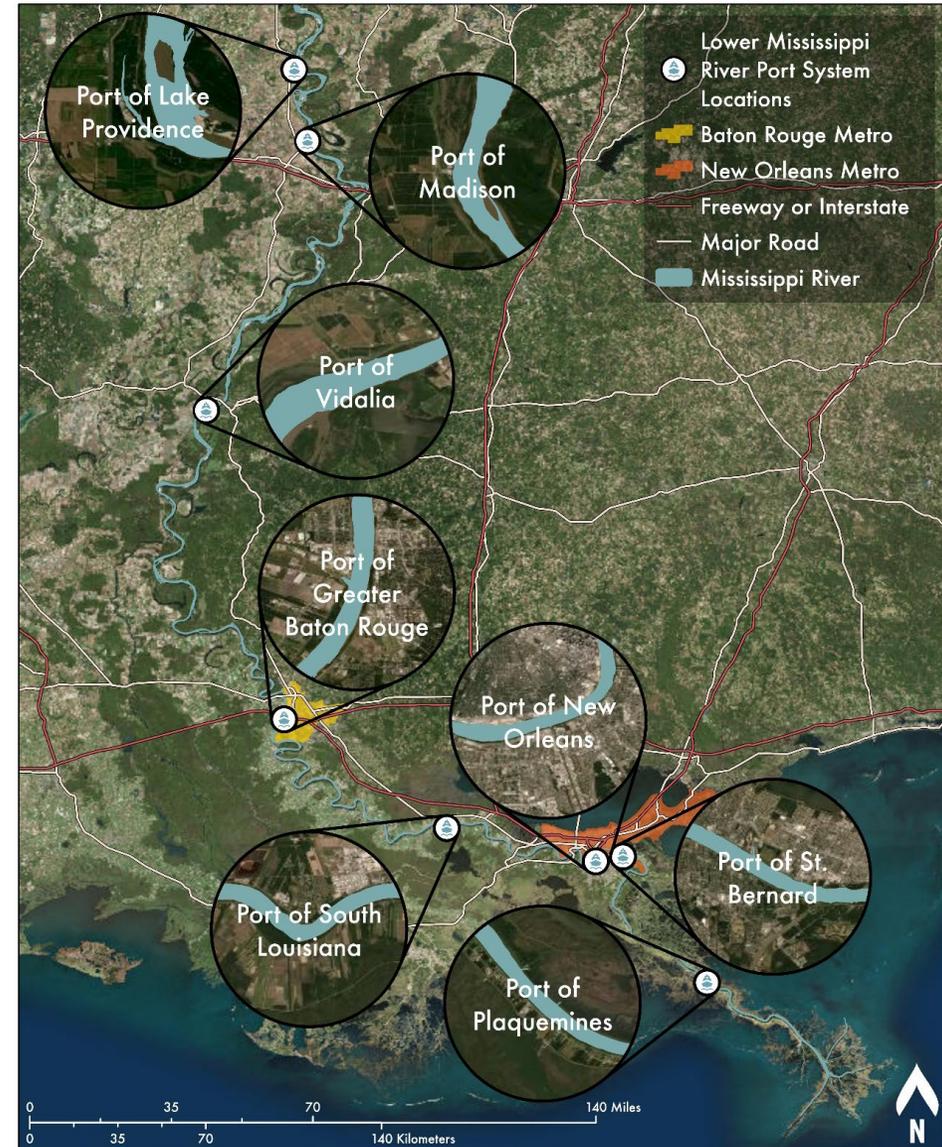
PHASE 2



SMARTPORT OVERVIEW

Project objectives include developing:

- A shoaling forecast tool
 - Days-months
- Resiliency strategy
- Customized dashboards
- Physical center for emergency operations



Source: Earthstar Geographics, U.S. Department of Transportation, National Transportation Atlas Database



SHOALING TOOL OBJECTIVES

Questions

- Do we have draft to berth a ship this afternoon?
- Can we anticipate future dredging needs (weeks to months)?

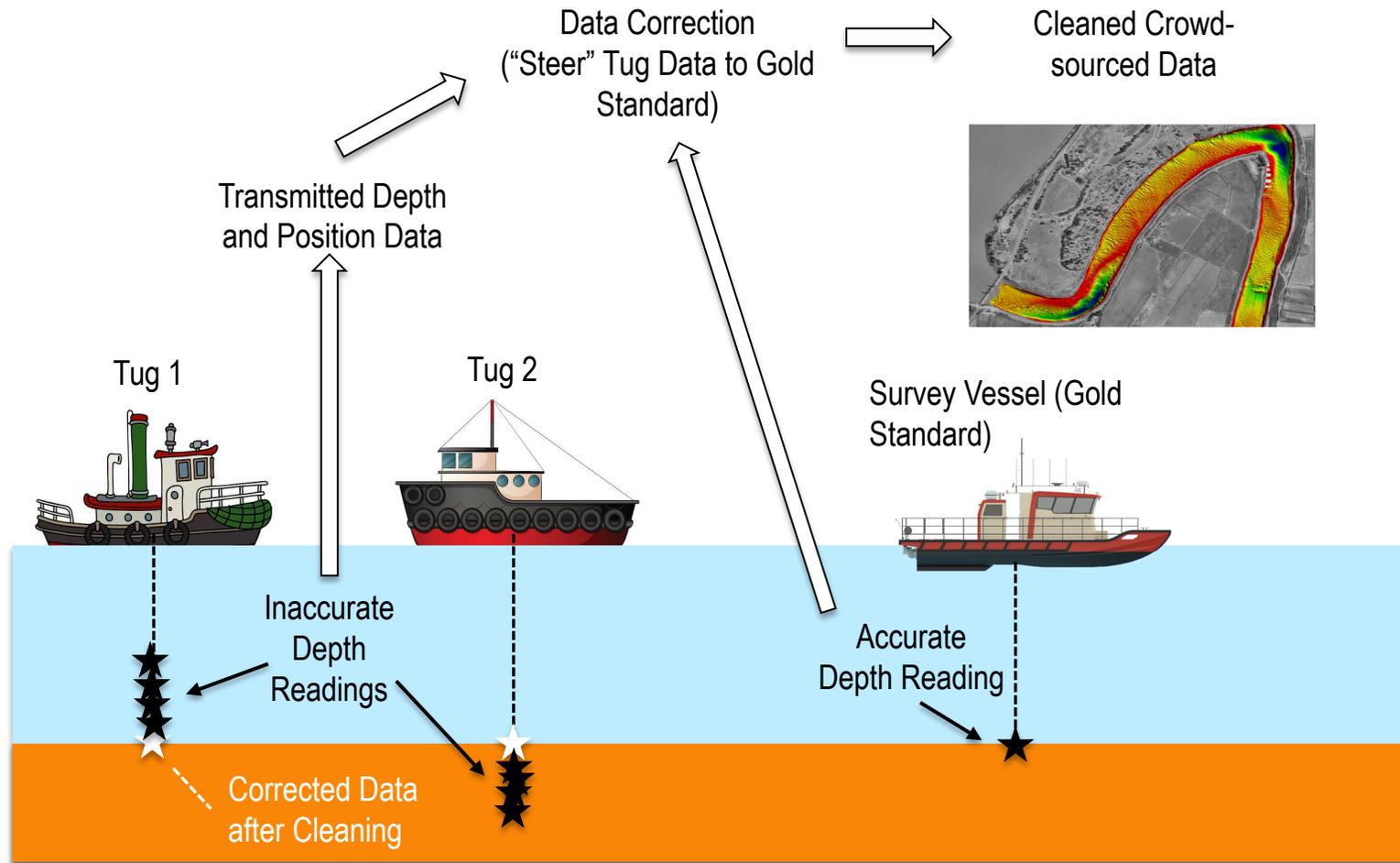
Solution

- Crowd-sourced data; increases data volume and frequency and cost effective.
- Implement a machine learning (ML) approach to forecast shoaling.



Port of New Orleans

CROWD-SOURCED DATA



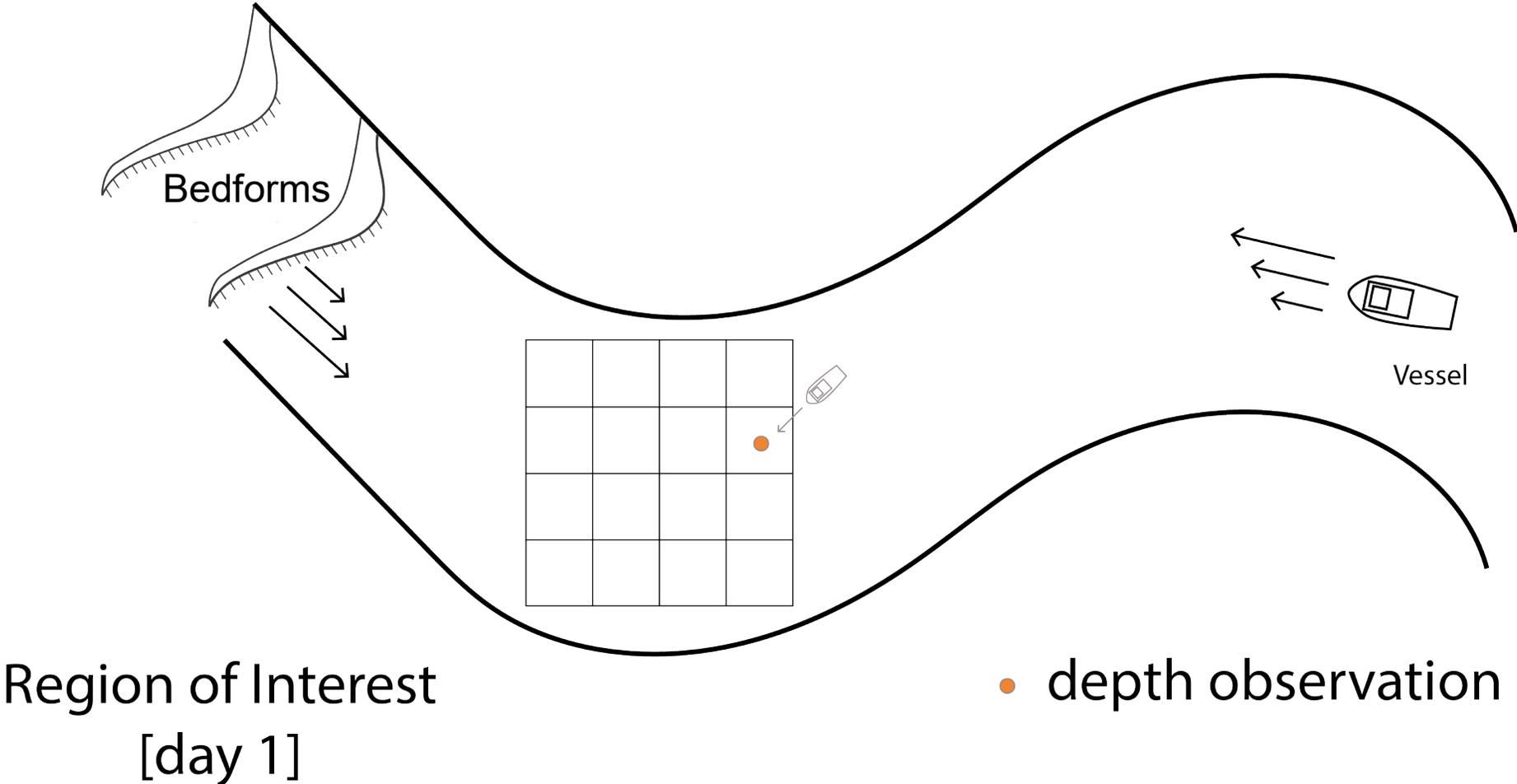
- Utilize existing sensors on vessel and networks
- Software captures available data streams on vessel:
 - GPS
 - AIS
 - Depth
- Transmit data to cloud storage



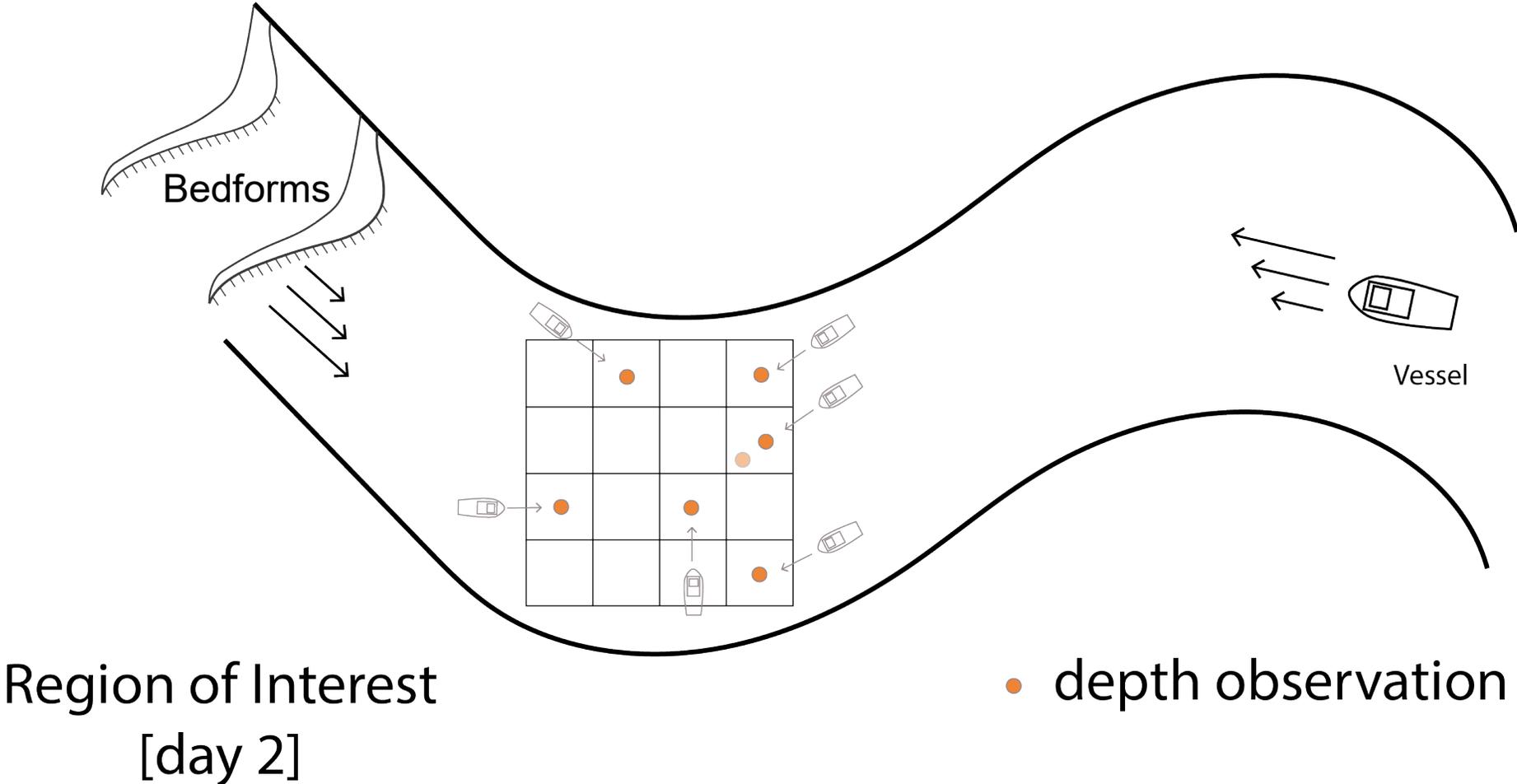
PARTNERS ENGAGED IN CROWD-SOURCE EFFORT



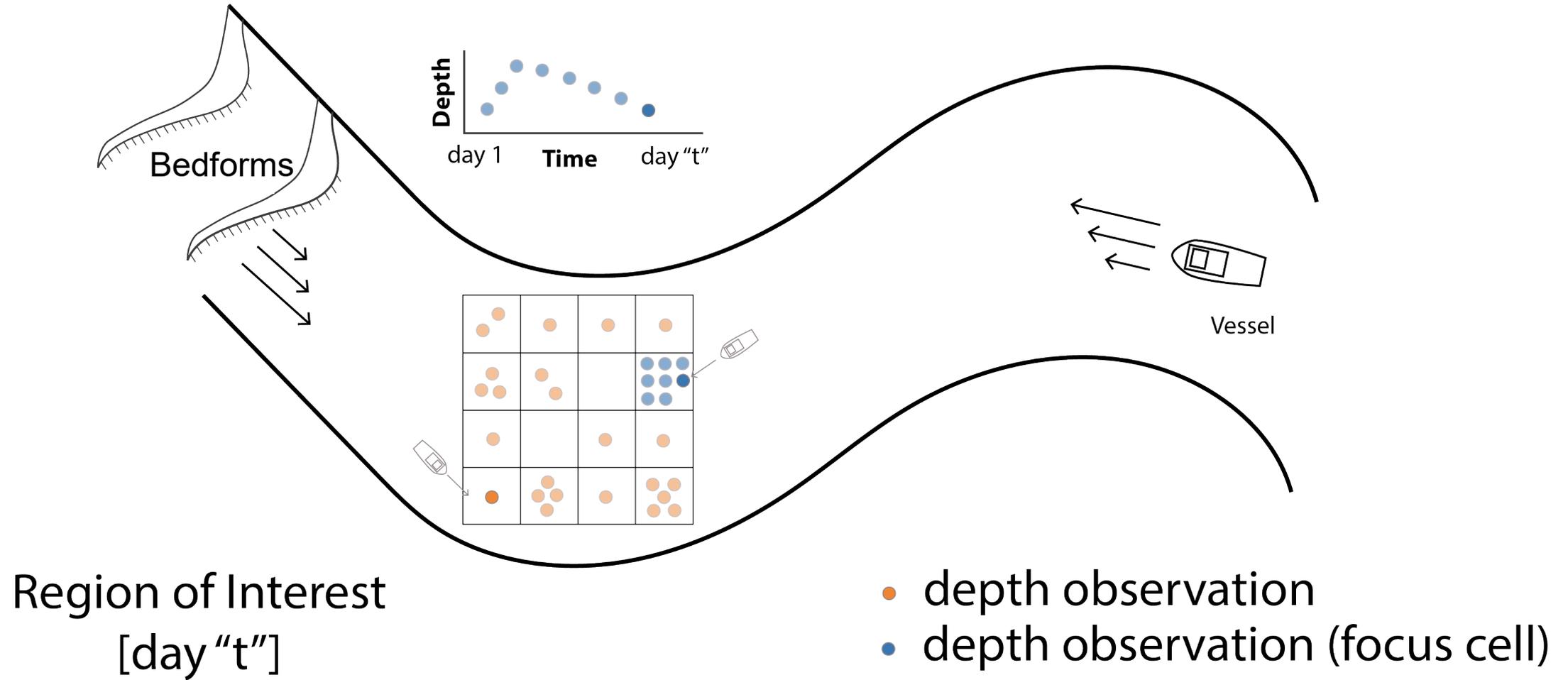
SMARTPORT ILLUSTRATION



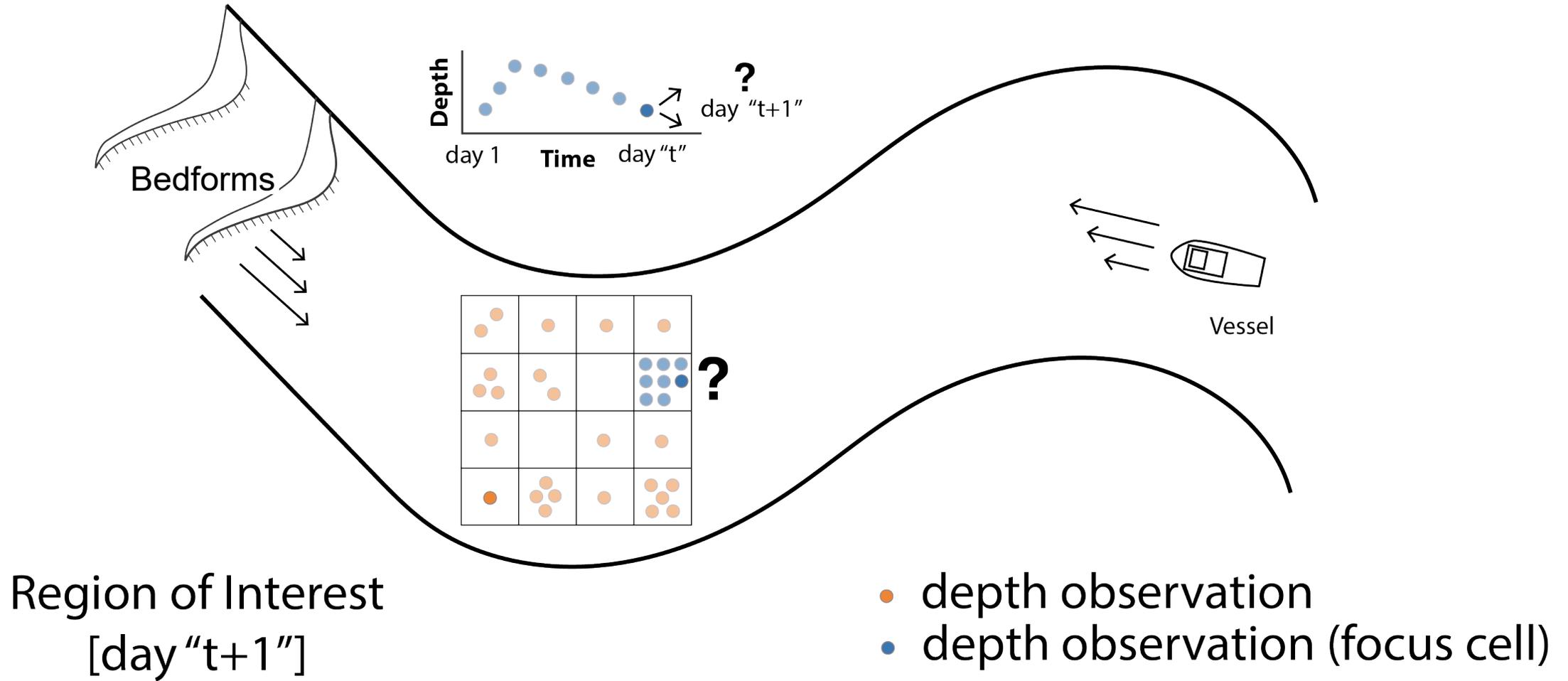
SMARTPORT ILLUSTRATION



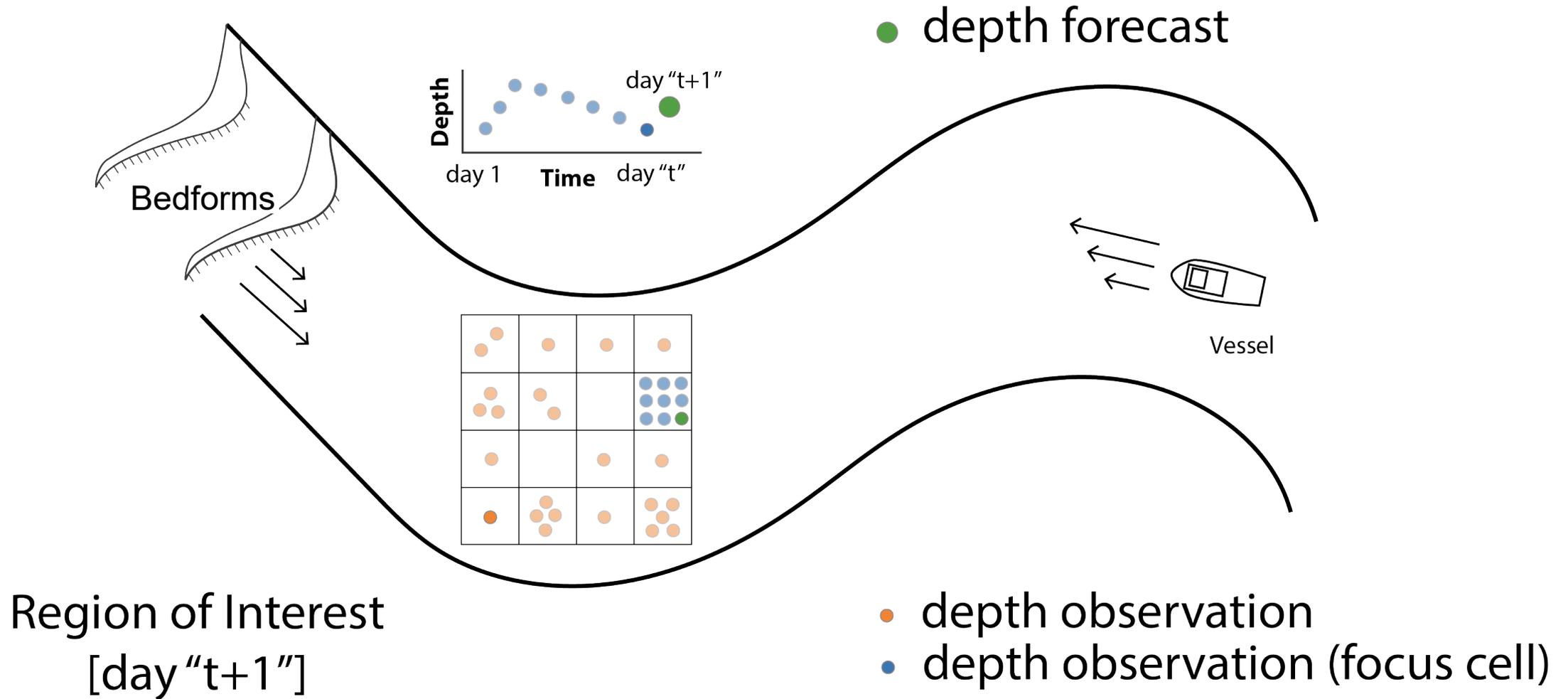
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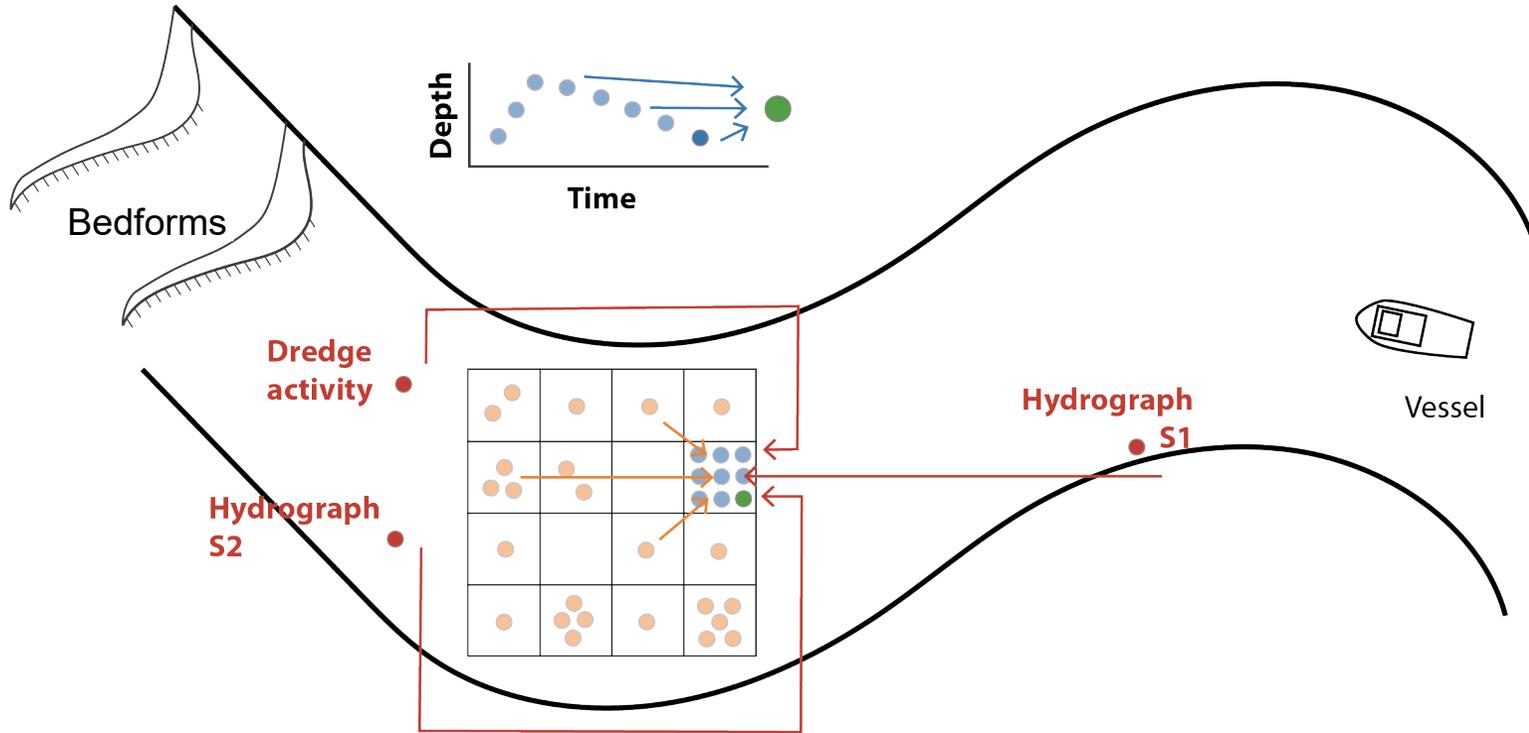
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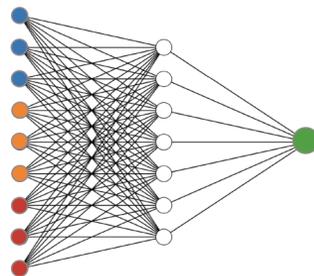
SMARTPORT ILLUSTRATION



SMARTPORT ILLUSTRATION



depth (time)
depth (space)
context (exogenous)

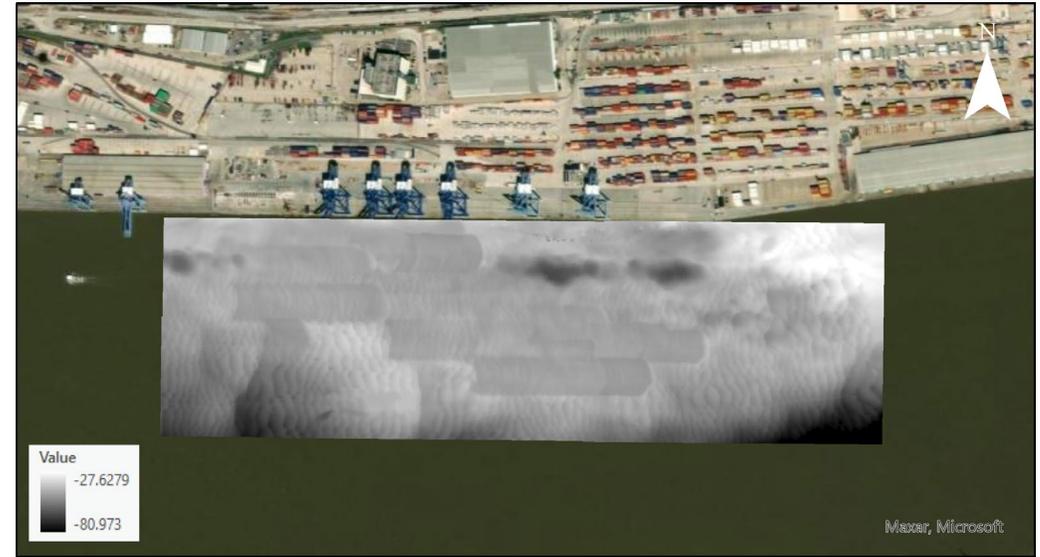


depth forecast (target)



DATA GATHERING

- Crowd-sourced vessel data
- USACE survey data (eHydro)
- USACE & USGS river gages
- Port survey data
- NWS river forecast



Port of NOLA Multibeam



USACE, 2022



Bisso Towboat Company, 2022



USACE, 2022



DATA CLEANSING FOR 'XY' AND 'Z' OUTLIERS USING SUPERVISED MACHINE LEARNING

Issue

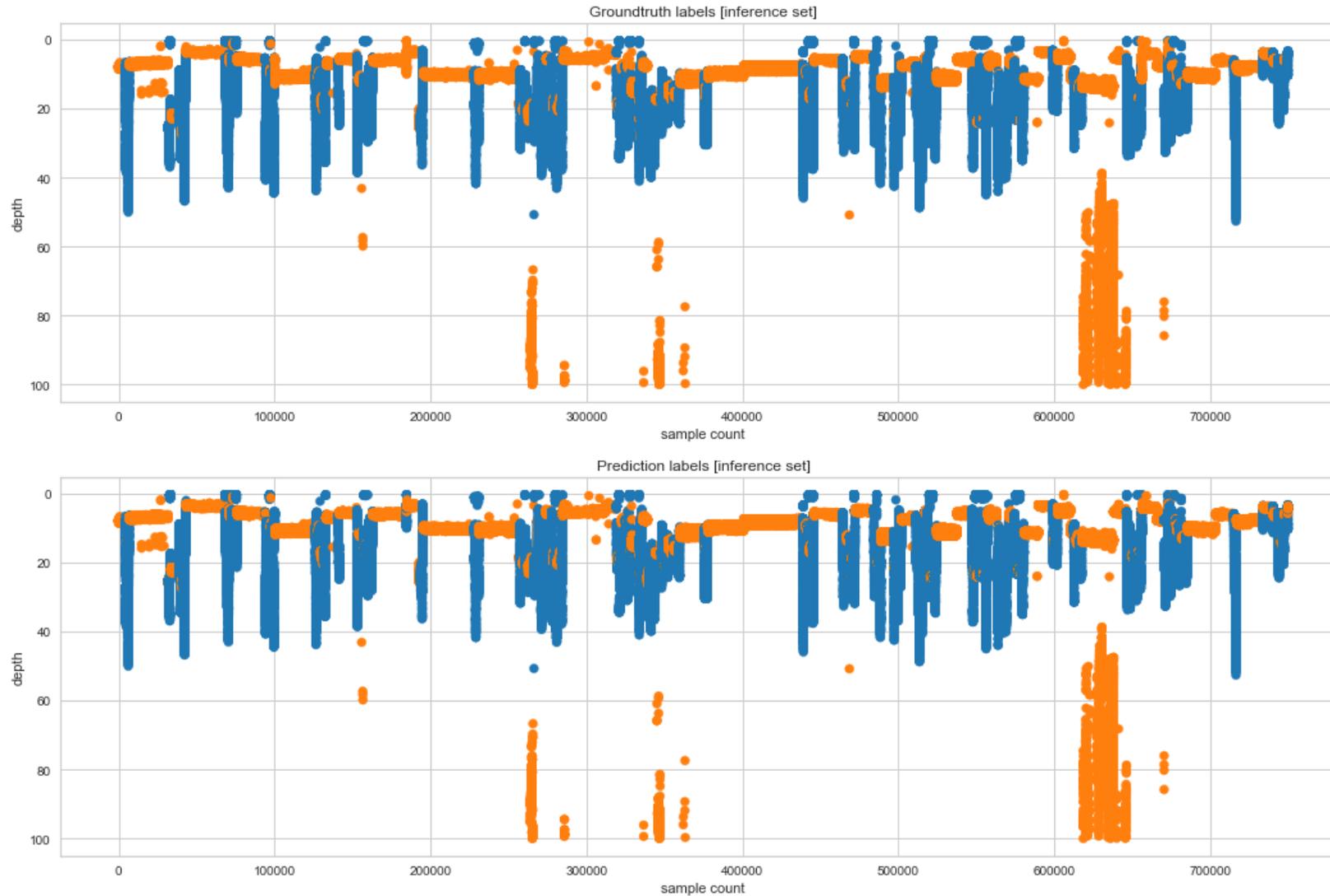
- Both GPS and depth sounder measurements are prone to error for a variety of reasons

Solution

- Train a ML model to classify 'xy' and 'z' outliers based on user labelling (supervised learning)



DWELL/'XY' DETECTION



- MVP Dwell
 - Accuracy: 97.0%
 - Precision: 97.4%
 - Recall: 98.9%
 - F1: 98.1%

Test set: "base-20"
20 mmsi-days

Legend

- : non-dwell
- : dwell

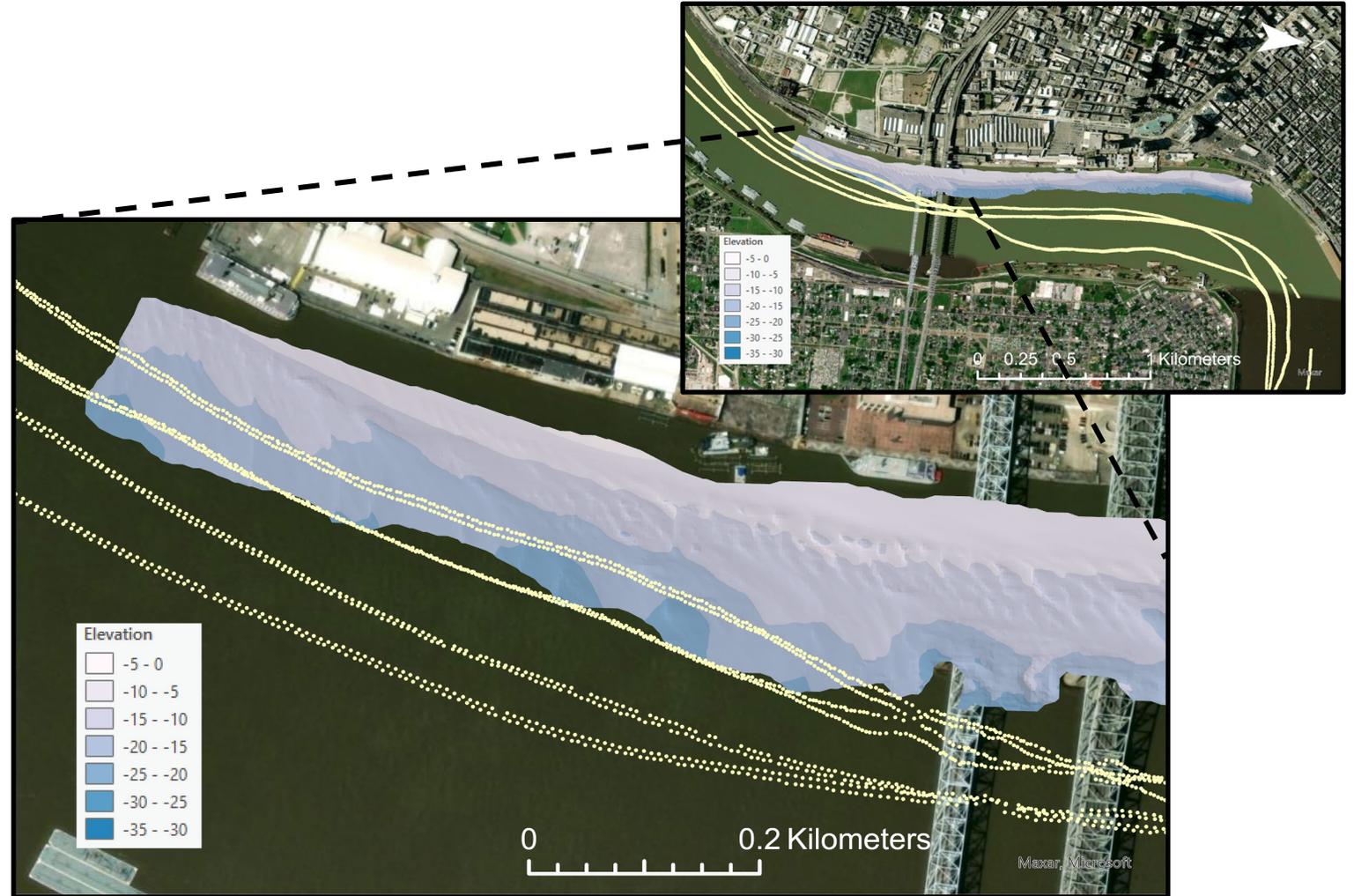


TUG DEPTHS TO ELEVATIONS

Data shown:

- USACE Multibeam;
MR_56_NO3
- Vessel track data

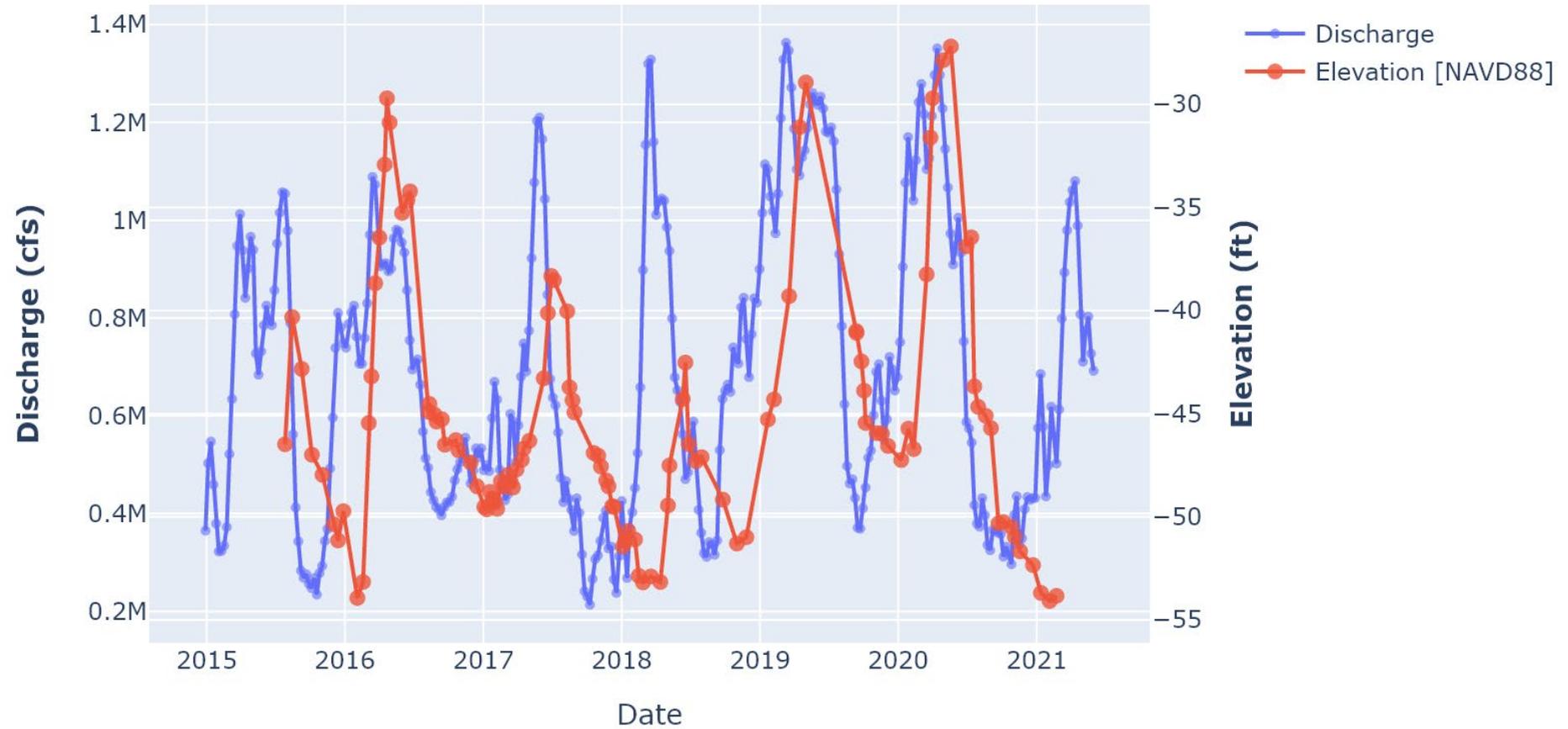
A known bed elevation, water surface elevation, and vessel data provide necessary information to 'convert' tug depths to elevations.



REDEYE CROSSING

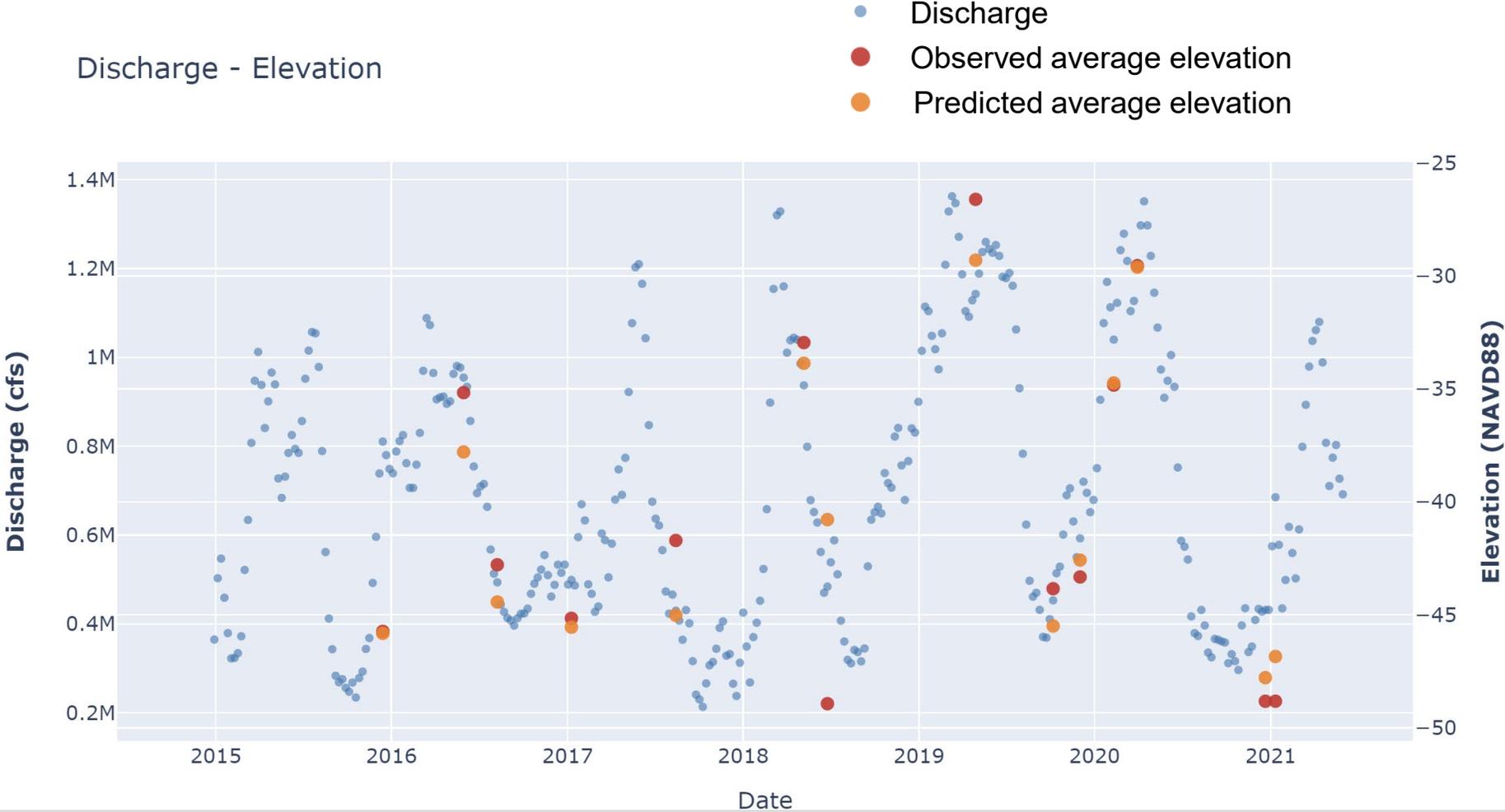


"Discharge - Elevation": RED



OBSERVATIONAL ML PRELIMINARY RESULTS

- Redeye Crossing
- Predict bed elevation using multiple river parameters

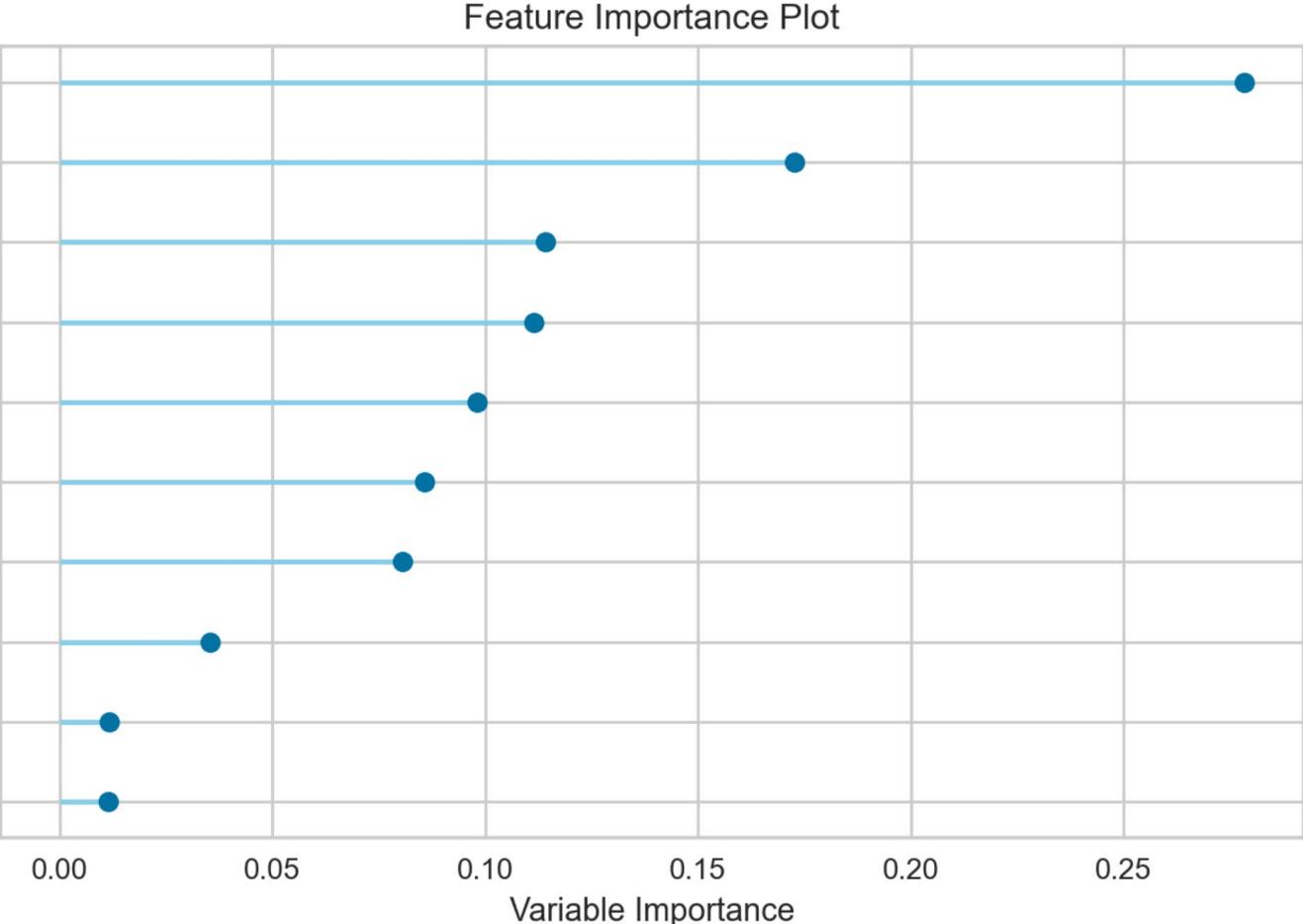


OBSERVATIONAL ML PRELIMINARY RESULTS



OBSERVATIONAL ML PRELIMINARY RESULTS

Water Surface Slope (RR-Don), t=2 weeks
Red River Landing WSE, t=4 weeks
Baton Rouge Discharge, t=4 weeks
Red River Landing WSE, t=2 weeks
Baton Rouge Discharge, t=2 weeks
Water Surface Slope (RR-Don), t
Water Surface Slope (RR-Don), t=4 weeks
Donaldsonville WSE, t
Red River Landing WSE, t
Baton Rouge Turbidity, t



CONCLUSIONS

- Established process to crowd-source and ‘correct’ vessel data
- ML model to predict bed elevation based on observational data

NEXT STEPS

- Integrate crowd-sourced and dredging data into ML model
- Expand ML model to other areas of the river
- Version 1 of Shoaling Forecast Tool expected Q1 2023
- Customized dashboards for real-time decision management





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THANK YOU

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