

# **Planning Observations for Intelligent Science Experimentation**

Principal Investigator: Yuliya Marchetti (398); Co-Investigators: Andrew Branch (397), James Montgomery (398), Steve Chien (397), Margaret Johnson (398), Longtao Wu (398), James Mason (397), Jason Swope (397), Peyman Tavallali (JPMorgan Chase), Hui Su (Hong Kong University of Science and Technology)

> Program: FY22 R&TD Strategic Initiative Strategic Focus Area: Intelligent Adaptive Observing System Strategic Initiative Leader: Benjamin Smith

## **Objectives**

Planning Observations for Intelligent Science Experimentation (POISE) develops autonomous technologies that enable coordinated, targeted, adaptive observations across multiple observing systems guided by the estimated improvement to our model-based understanding, predictive skills, and scientific understanding of highest priority Earth processes in the Earth Science Decadal Survey.



- with machine learning.
- **Coordination** across observing assets to schedule targeted measurements.
- Assimilation of obtained observations back into the forecast model.
- Hurricane intensification as a case study.

### Approach

New technologies include 1) variational ensemble generation for many hurricane cases and dates, 2) machine learning model to accelerate forecast, 3) observational utility metric and machine learning (ML) utility model, 4) utility-based observation scheduling:



# Planning

- Select observation requests to maximize utility while obeying operational constraints of multiple heterogeneous assets.
- Utilizes combinatorial optimization (i.e. simulated annealing) to allocate requests to individual assets.

#### A new automated federated observation planning strategy developed, motivated by the adaptive observing system:



#### **Planned Observations**



 Combines observations from independent heterogenous assets to maximize utility

• Decouples utility from low-level planning, easier integration with existing asset operational systems.

#### Variance of the forecast ensembles



Provides	Fast		Sub-selects
ensembles of	emulation of	from ensembles	observations to
forecasts into	the physical	and provides a	request from
the machine	model	set of impactful	multiple
learning	forecast	locations and	heterogeneous
models	ensembles	types of	assets
		observations	··

### Significance/Benefits to JPL and NASA

- Maximizes science return of observations from multiple assets
- Improve science understanding and predictive capabilities
- Applicability to systems with strong assimilative models
- Strong support for mission proposal competitiveness

#### **National Aeronautics and Space Administration**

**Jet Propulsion Laboratory** California Institute of Technology Pasadena, California

www.nasa.gov

Clearance Number: CL# Poster Number: RPC#R20037 Copyright 2022. All rights reserved.

### **Publications:**

Branch, A.; Chien, S.; Marchetti, Y.; Su, H.; Wu, L.; Montgomery, J.; Johnson, M.; Smith, B.; Mandrake, L.; and Tavallali, P., "Federated Scheduling of Model-Driven Observations for Earth Science", In International Workshop on Planning & Scheduling for Space (IWPSS), July 2021.

Tavallali, P., Chien, S.; Mandrake, L.; Marchetti, Y.; Su, H.; Wu, L.; Smith, B.; Branch, A.; Mason, J.; and Swope, J., "Adaptive, Model-Driven Observations for Earth Science", In Proceedings of the International Symposium on Artificial Intelligence, Robotics and Automation in Space (i-SAIRAS 2020), January 2020.

### **PI/Task Mgr. Contact Information:**

Email: Yuliya.Marchetti@jpl.nasa.gov

### Results

- Demonstrate a ~25% reduction in sea level pressure forecast ensemble variance in OSSE-type simulation experiments over "blind" observations.
- Tropical cyclone intensity projections closer to the true state •
- Improved numerical model stability