

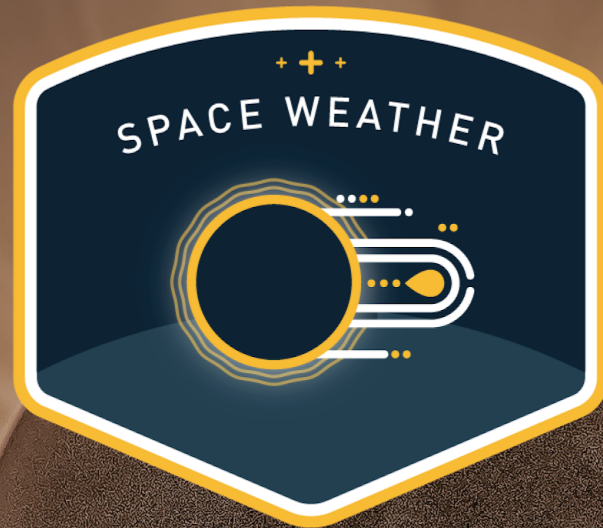
Predicting GNSS Disruptions using Machine Learning

ML in Heliophysics 2019

Laura A. Hayes

Kibrom Ebuy Abraha, Daniel Kumar, Karthik Venkataramani

Asti Bhatt, Red Boumghar, Sylvester Kaczmarek, Ryan McGranaghan, Sean McGregor



Check out Lika's talk on Wednesday!

NASA Frontier Development Lab

- 8 week applied research accelerator
- Public-Private Partnership
- Hosted at SETI Institute and NASA Ames



FDL 2018

NASA Ames Research Center - Silicon Valley - 2018



Machine Learning

THIS IS YOUR MACHINE LEARNING SYSTEM?

YUP! YOU POUR THE DATA INTO THIS BIG PILE OF LINEAR ALGEBRA, THEN COLLECT THE ANSWERS ON THE OTHER SIDE.

WHAT IF THE ANSWERS ARE WRONG?

JUST STIR THE PILE UNTIL THEY START LOOKING RIGHT.

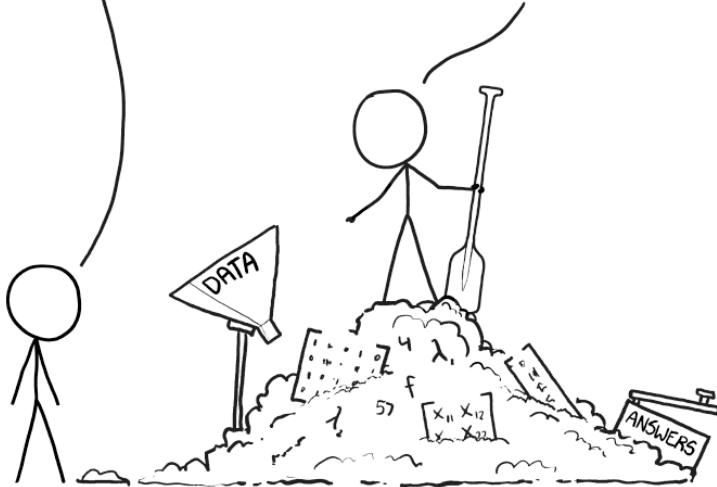


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Machine Learning

- Scientifically informed data-driven approach

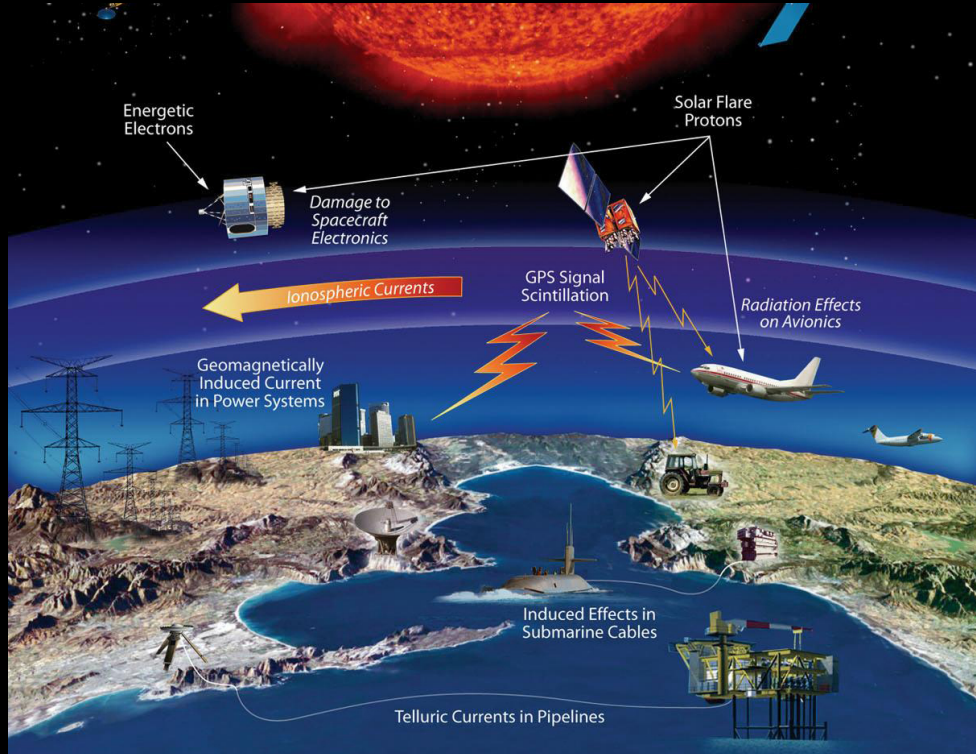
Maximise AI/machine learning techniques to space science challenges

Space Weather

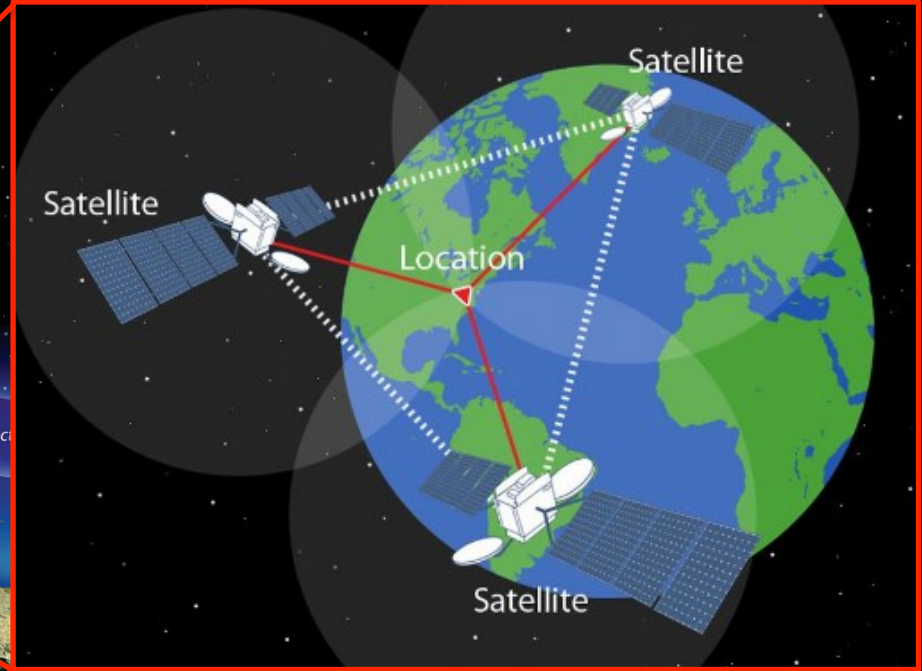
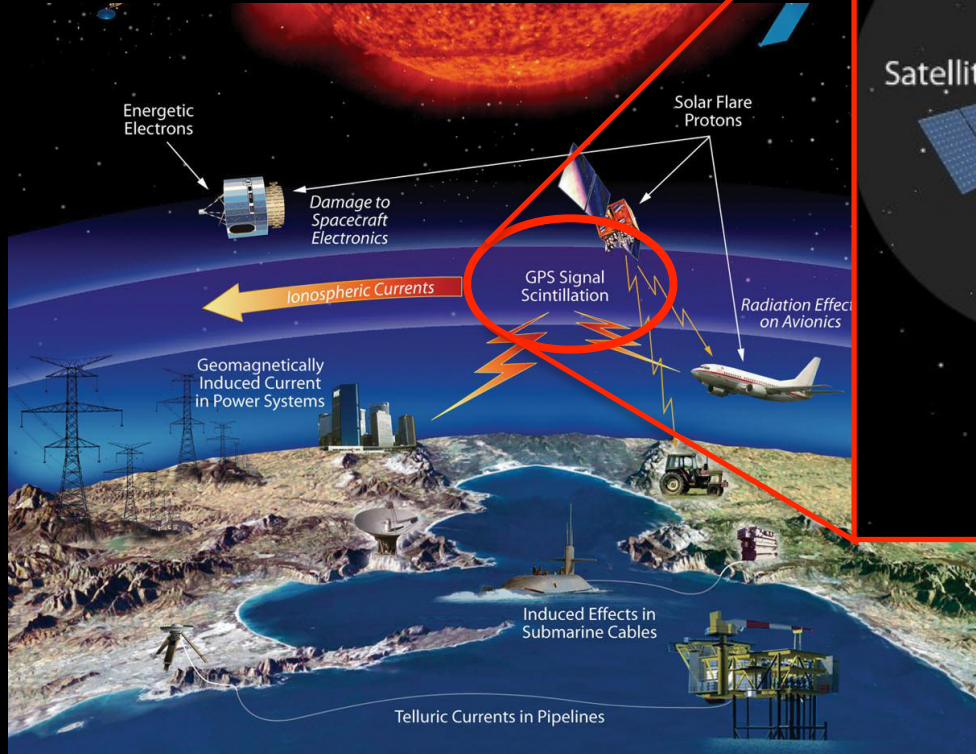


Dynamic solar activity
impacts the Earth's
magnetic field and
terrestrial environment

Space Weather Impacts



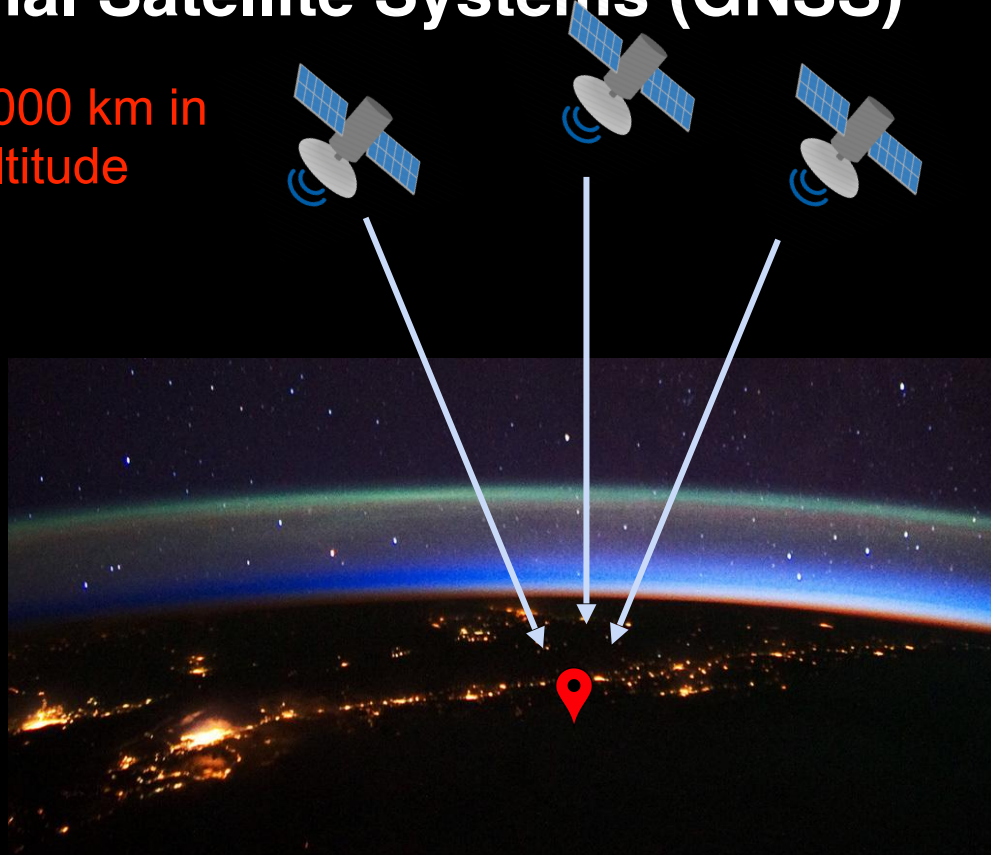
Space Weather Impacts



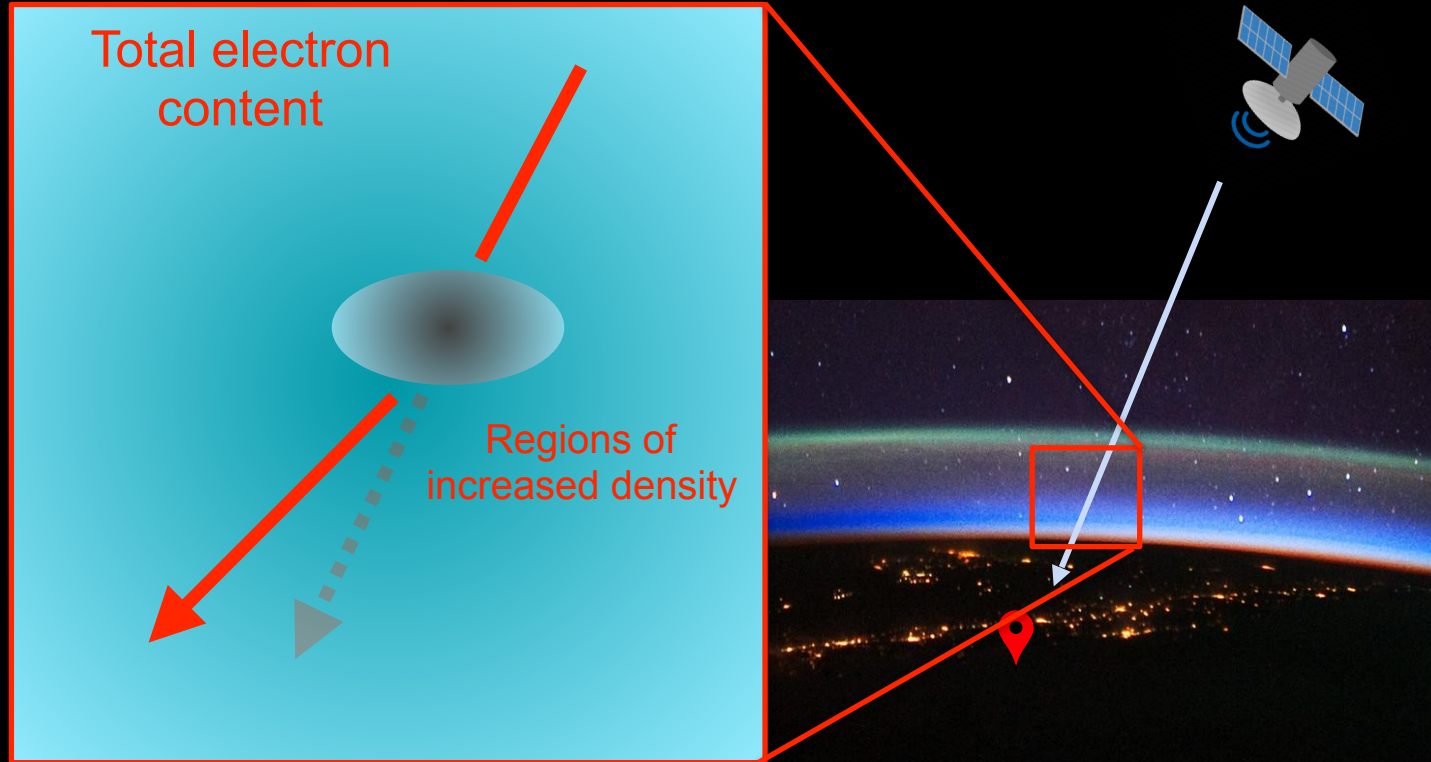
Global Navigational Satellite Systems (GNSS)

~ 20000 km in
altitude

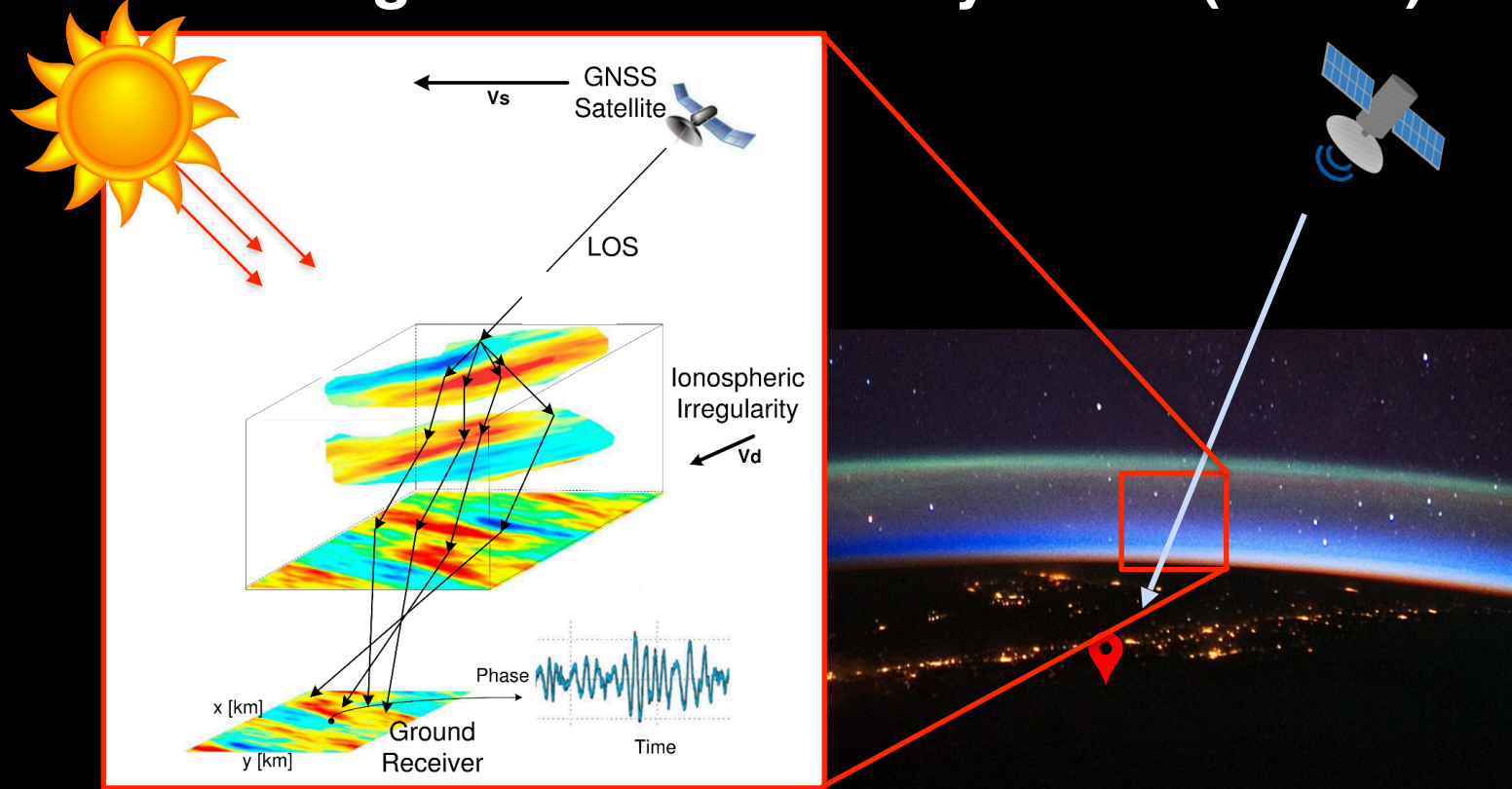
↑
Ionosphere
100-1000 km
↓



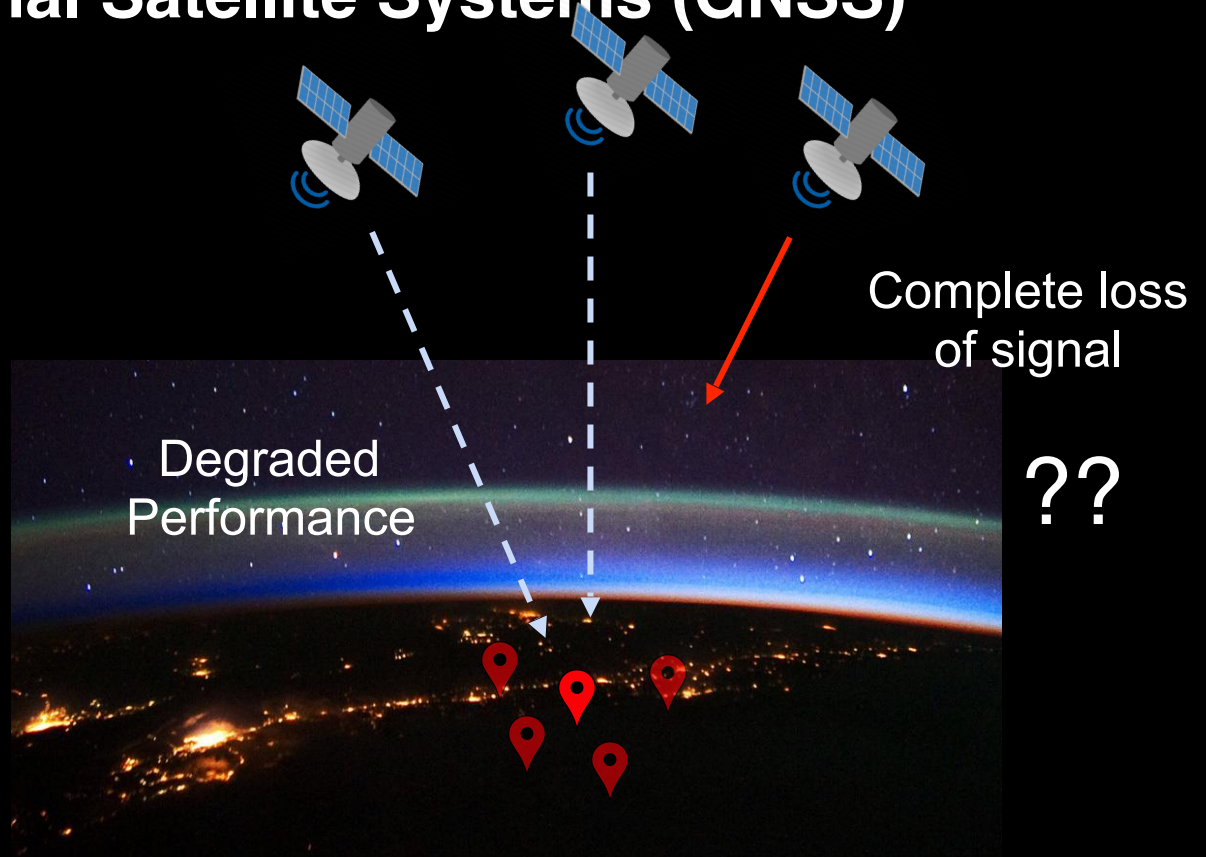
Global Navigational Satellite Systems (GNSS)



Global Navigational Satellite Systems (GNSS)

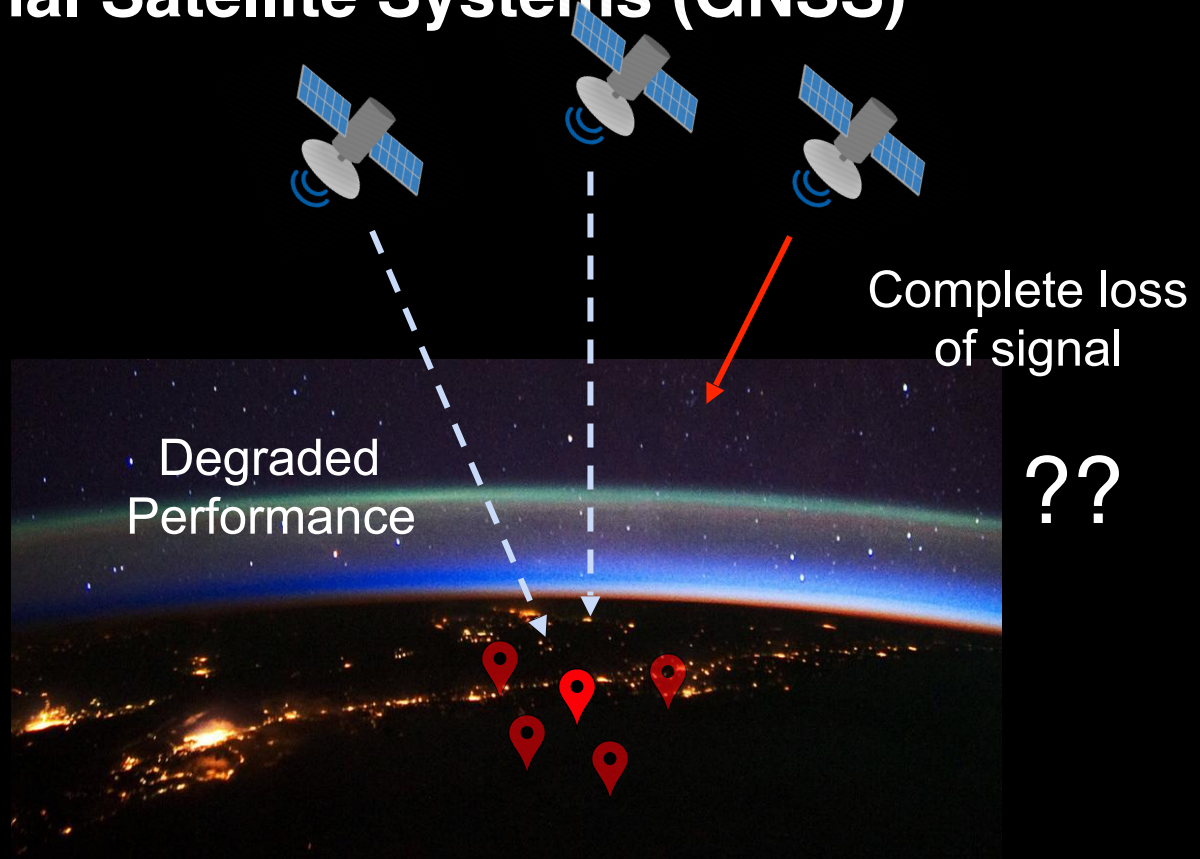


Global Navigational Satellite Systems (GNSS)



Global Navigational Satellite Systems (GNSS)

Need robust
forecasting
methods





Can we use data-driven machine learning techniques to forecast GNSS disruptions?

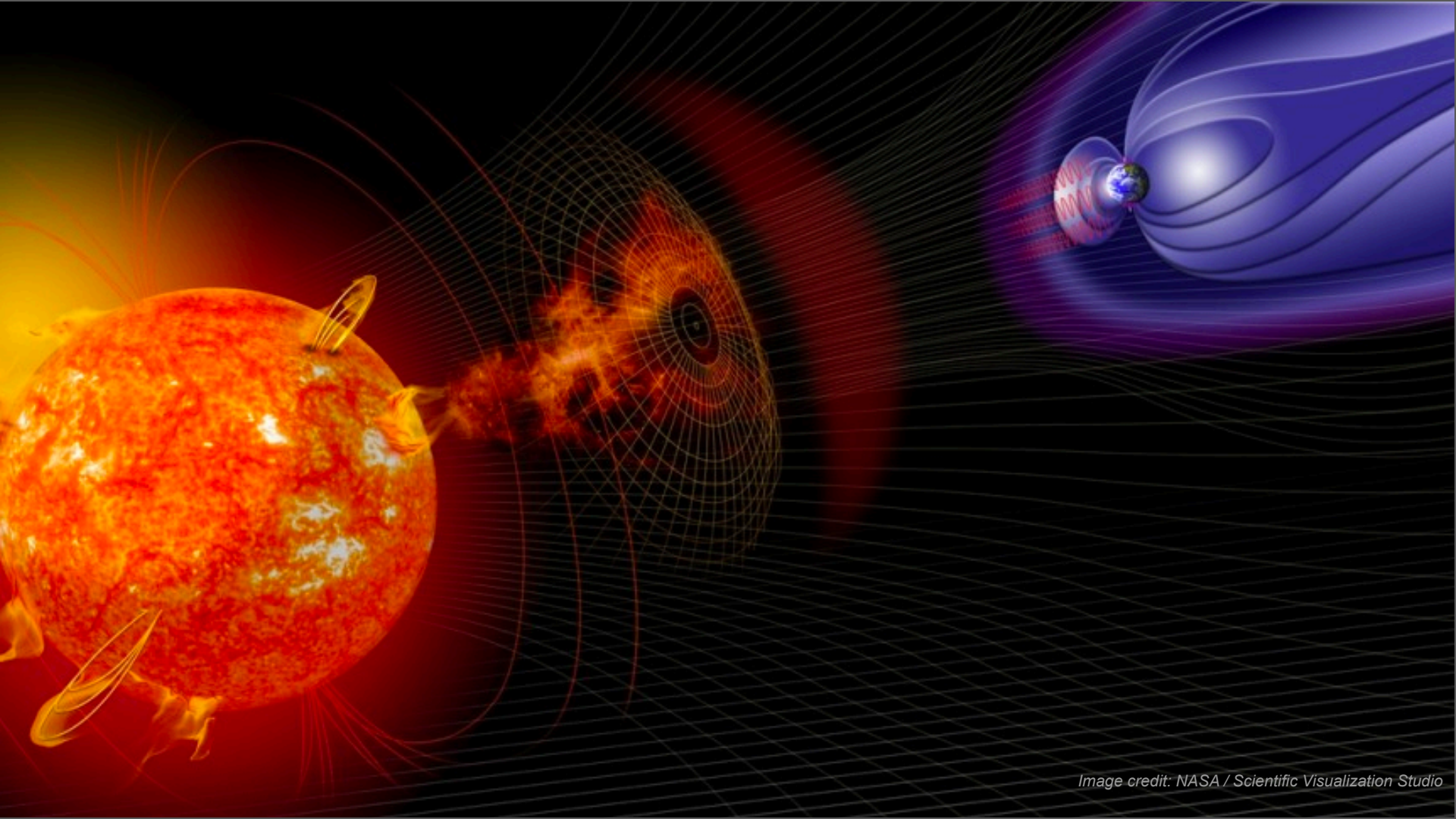
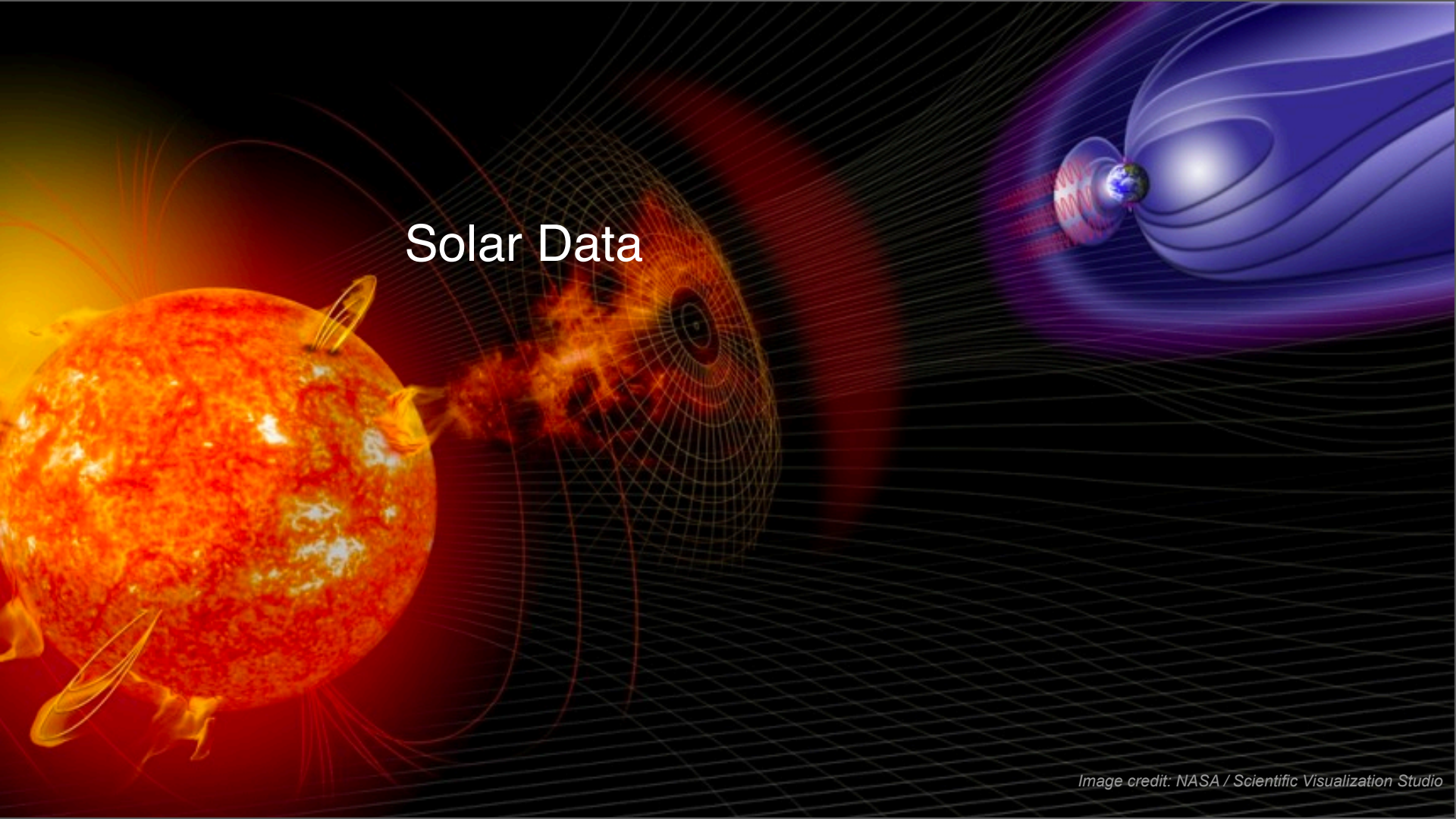


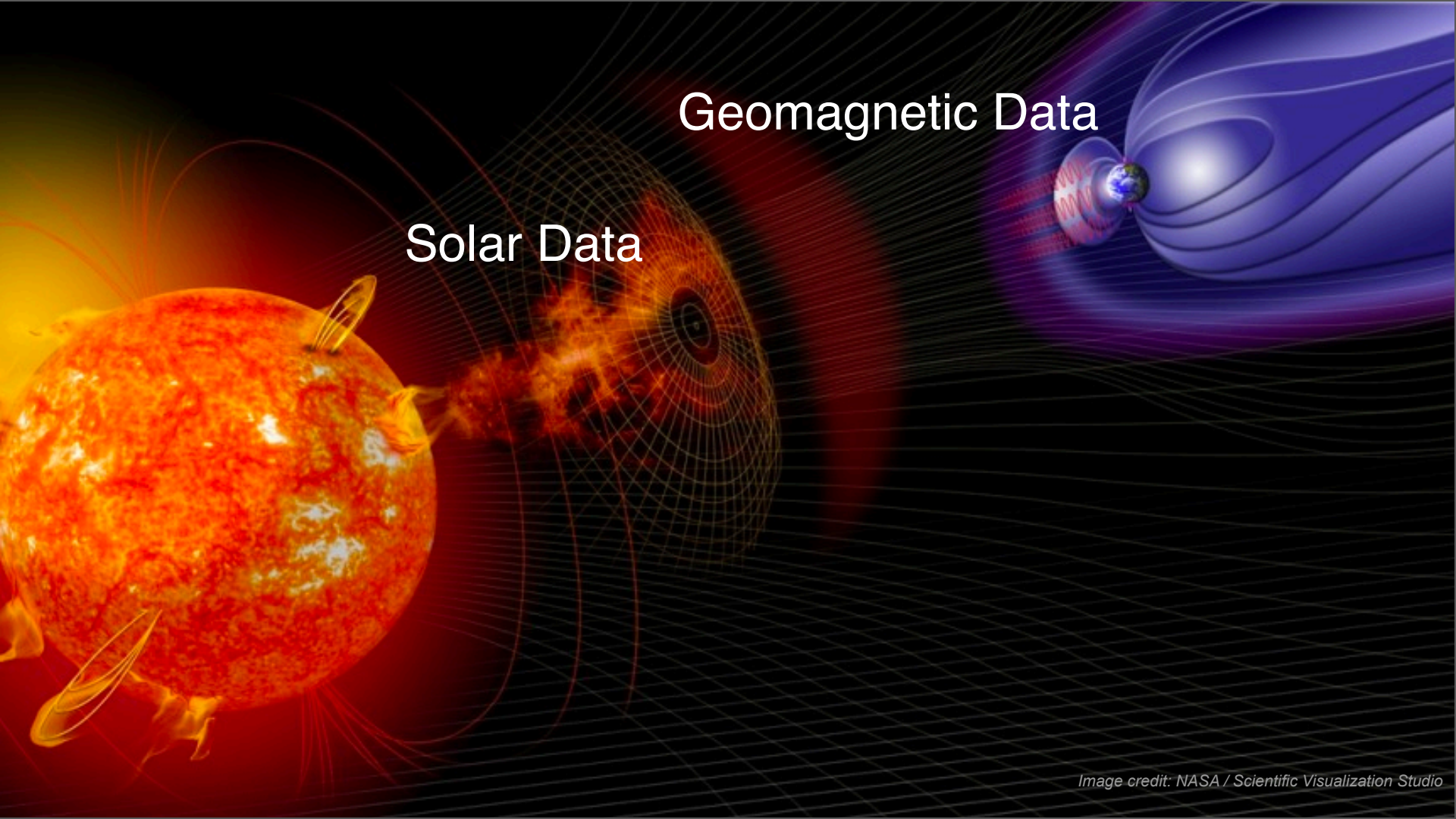
Image credit: NASA / Scientific Visualization Studio

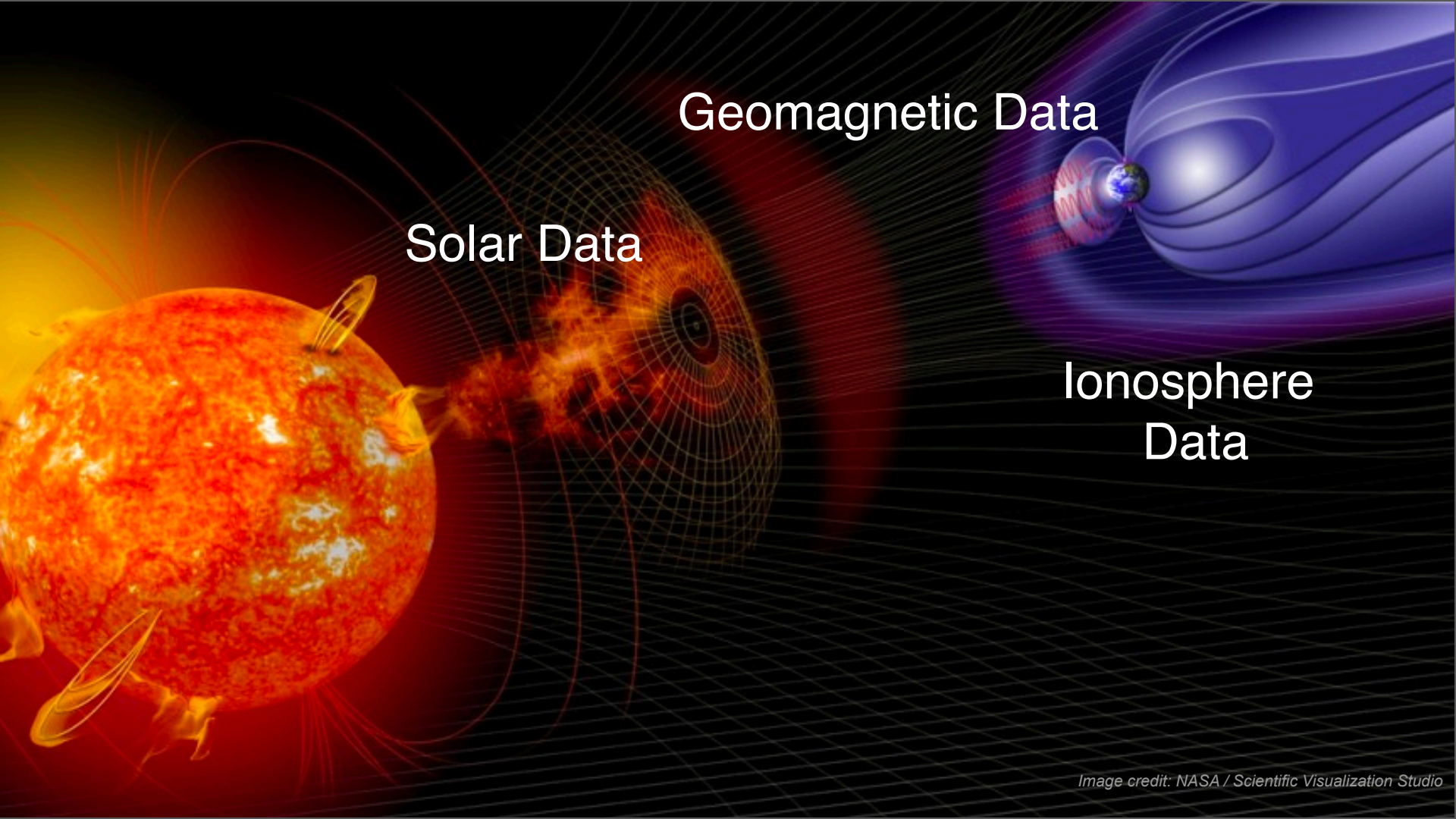
Solar Data



Geomagnetic Data

Solar Data

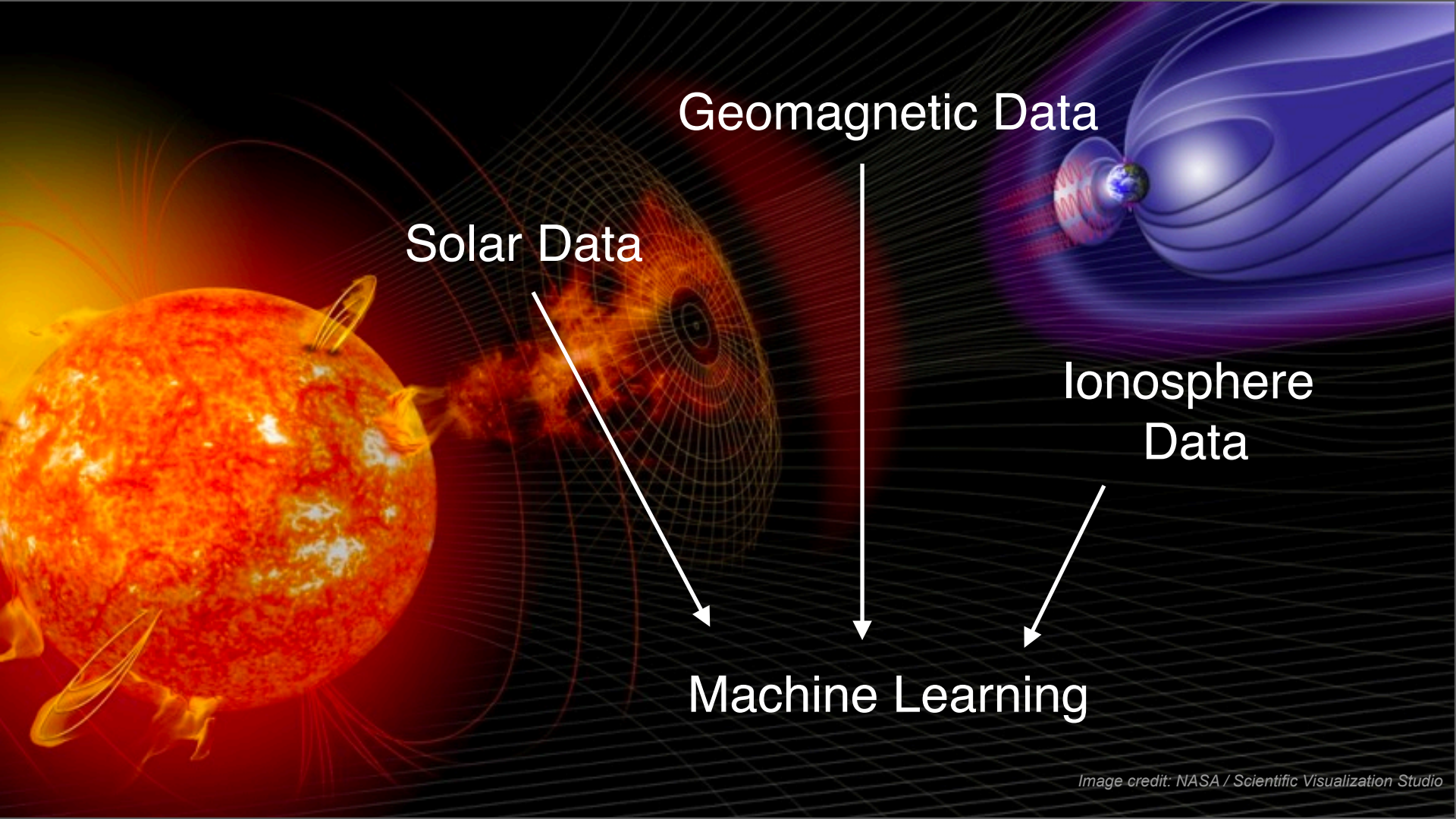




Solar Data

Geomagnetic Data

Ionosphere
Data



Geomagnetic Data

Solar Data

Ionosphere
Data

Machine Learning

High Latitude GNSS Stations

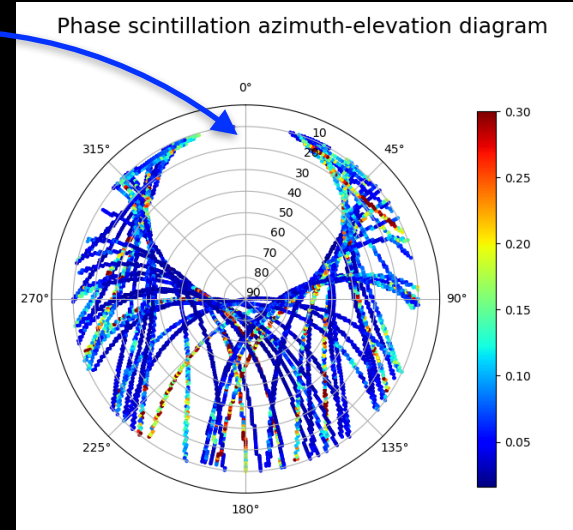


Canadian High Arctic Ionospheric Network
(CHAIN) GPS receivers

High Latitude GNSS Stations



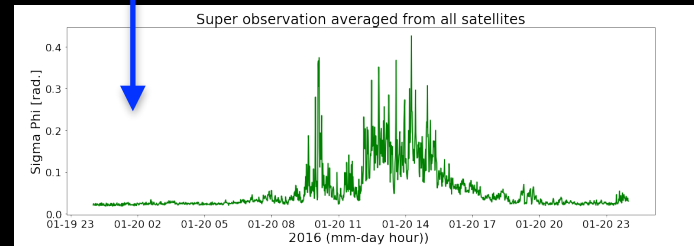
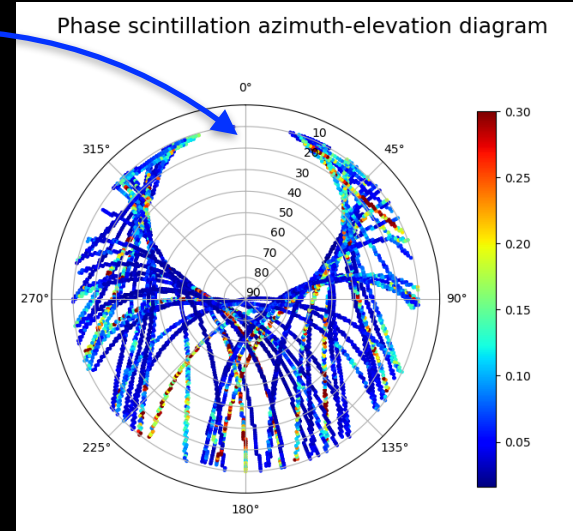
Canadian High Arctic Ionospheric Network
(CHAIN) GPS receivers



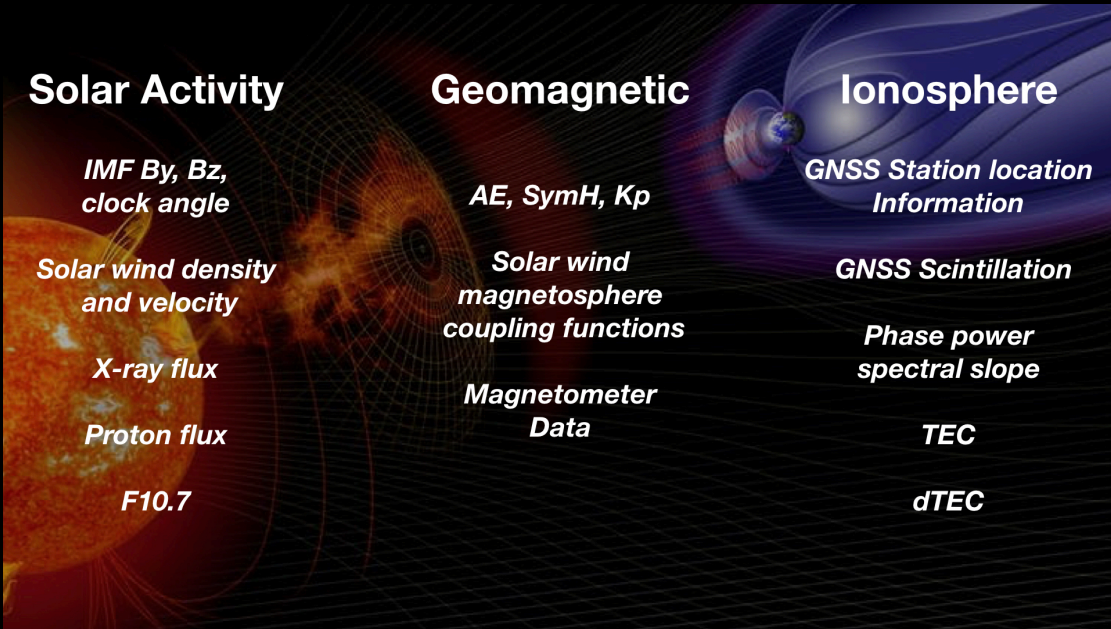
High Latitude GNSS Stations



Canadian High Arctic Ionospheric Network
(CHAIN) GPS receivers



Approach



41 inputs
> 100 feature
engineering

Predicted
GNSS
scintillation

**Build a predictive
model for GNSS
scintillation**

Tools, Compute and Software Environment

Used

- Python open-source tools for data acquisition, wrangling and machine learning
- IBM POWER8 and POWER9 processors
 - Enabled the rapid exploration and testing of ML techniques

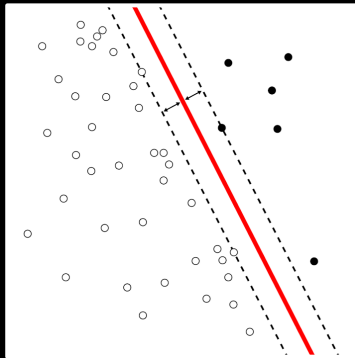
Created

- Python-based tools and cohesive data pipeline generated
- Machine learning framework on data pipeline

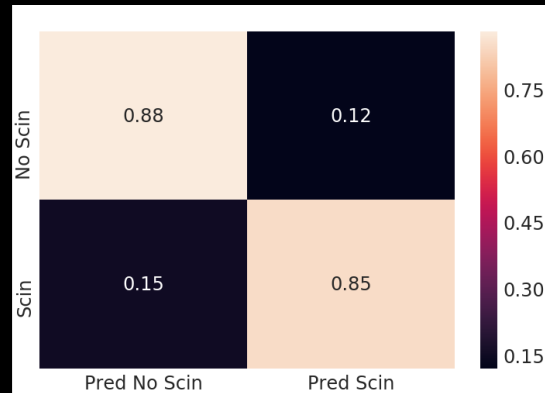


Forecasting GPS disruptions

**Classification
baseline**



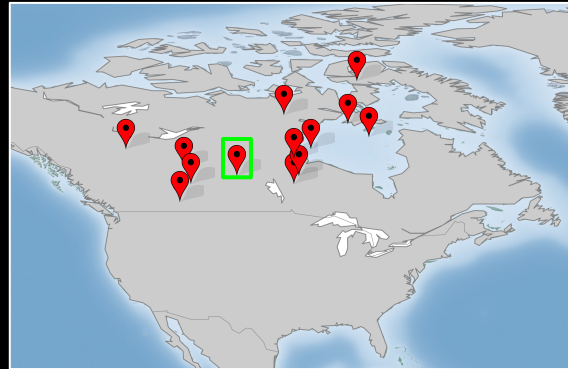
Support Vector
Machines



Forecasting GPS disruptions

**Classification
baseline**

Localized models



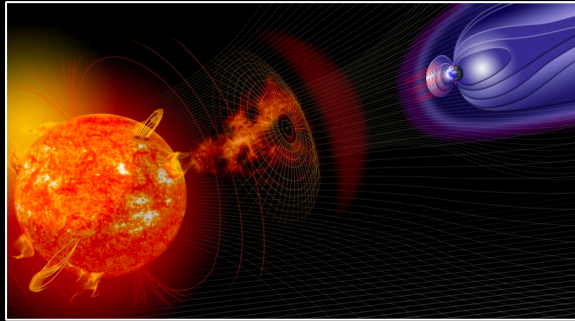
**Localised models improved
performance of model unto 40%**

Forecasting GPS disruptions

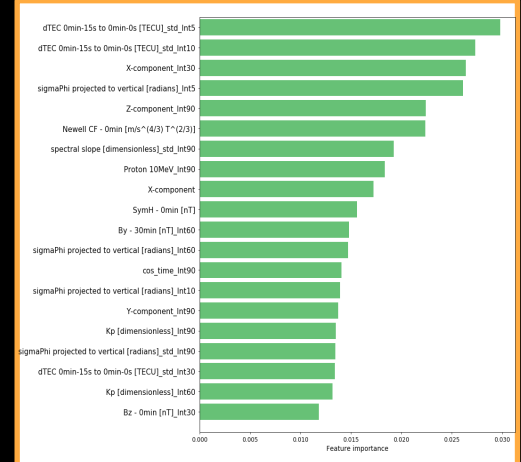
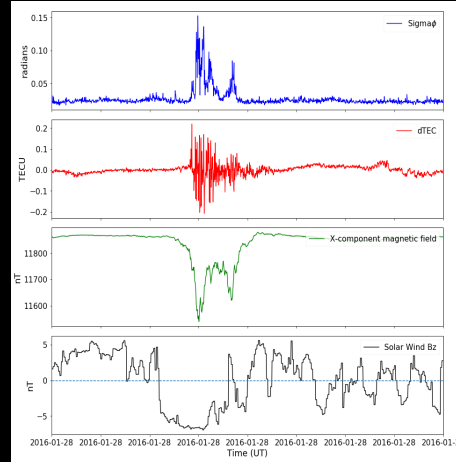
Classification
baseline

Localized models

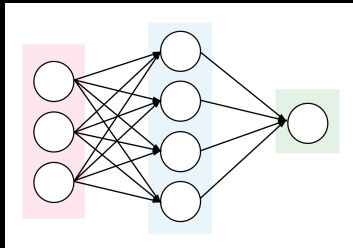
Model input



Feature engineering of inputs
Time history and variance



Forecasting GPS disruptions

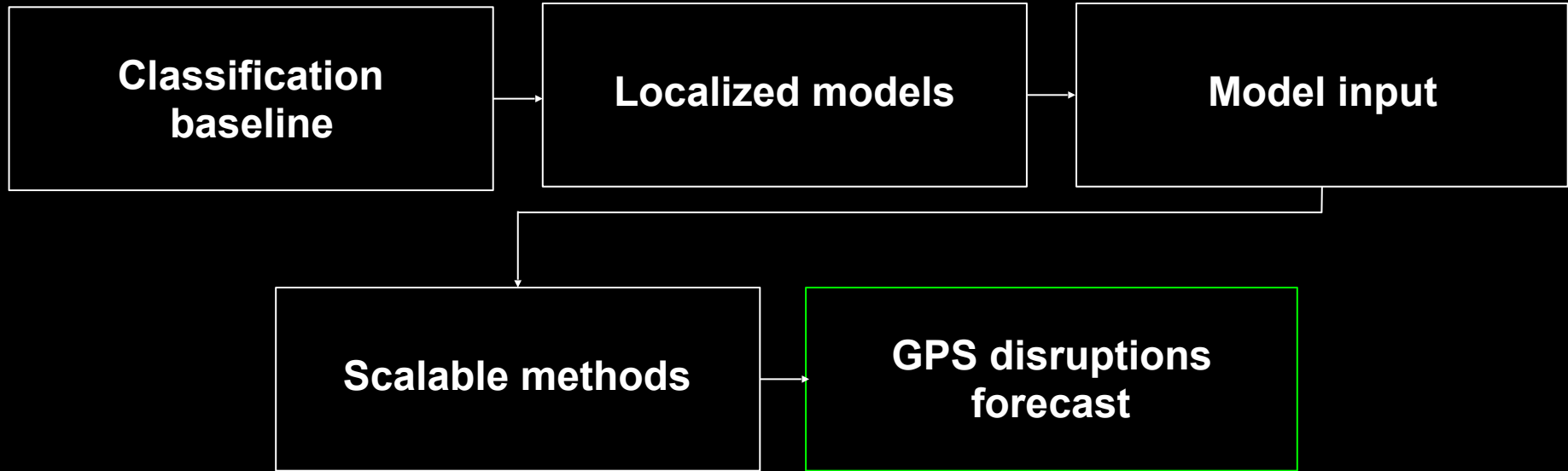


Neural Network



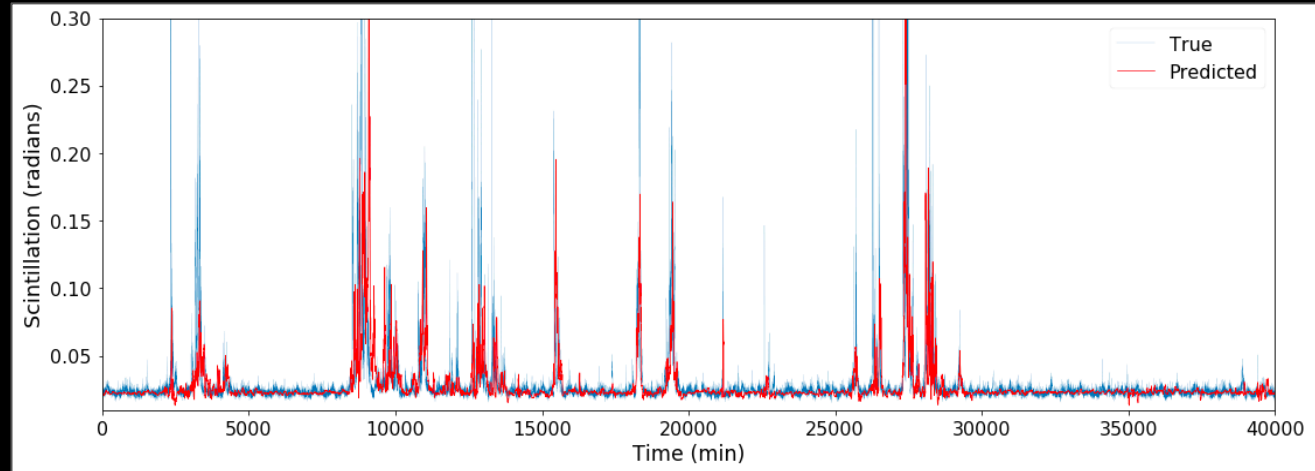
Random Forest, Gradient Boosting

Forecasting GPS disruptions



GNSS forecasts

- GPS disruptions forecasted 1 hr in advance
- Train/validate on 2015, 2016 test on 2017
- Recall of 96%
- **Improve forecasting metrics by 70%**



Model : Neural Network with feature engineered inputs

Outcomes

- **Developed a ML framework for predicting GPS disruptions.**
- **Proof of concept for machine learning applications for forecasting**
- **ML/Data shows localized nature of scintillation.**
- **+70%** on baseline forecasting metrics.

Looking to the Future

- **Pipeline set up for further exploration**
- **FDL 2019 included new parameters - spatially located auroral mapping**
- **See Ryan McGranaghan talk and poster (B session)!**