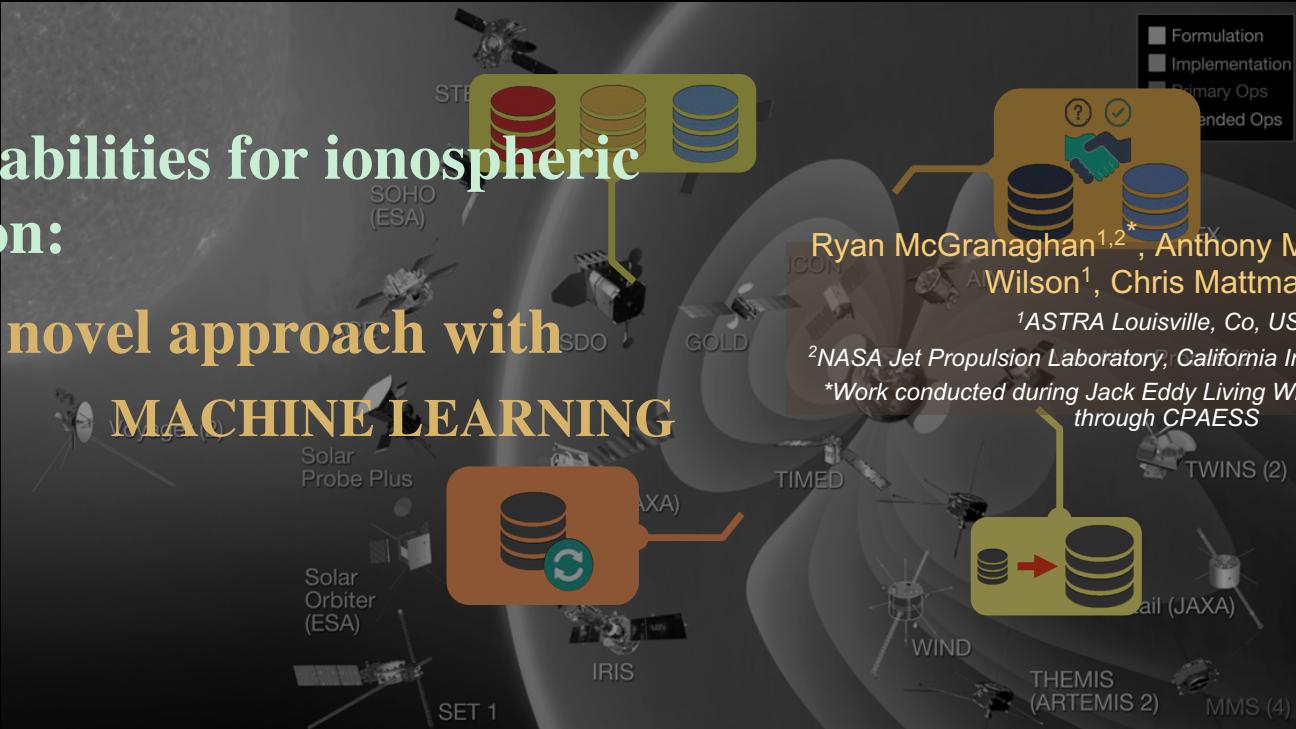


New capabilities for ionospheric prediction: A novel approach with **MACHINE LEARNING**



Ryan McGranaghan^{1,2*}, Anthony Mannucci¹, Brian Wilson¹, Chris Mattmann¹

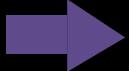
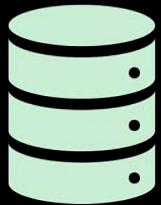
¹ASTRA Louisville, Co, USA

²NASA Jet Propulsion Laboratory, California Institute of Technology

*Work conducted during Jack Eddy Living With a Star Fellowship through CPAESS



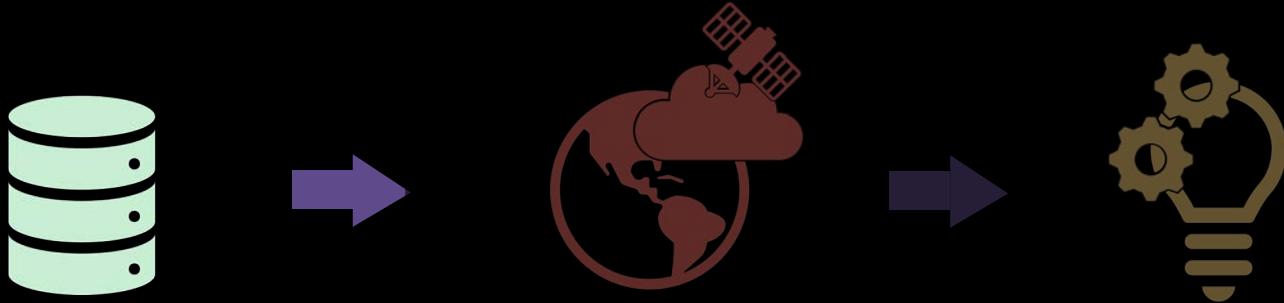
Agenda



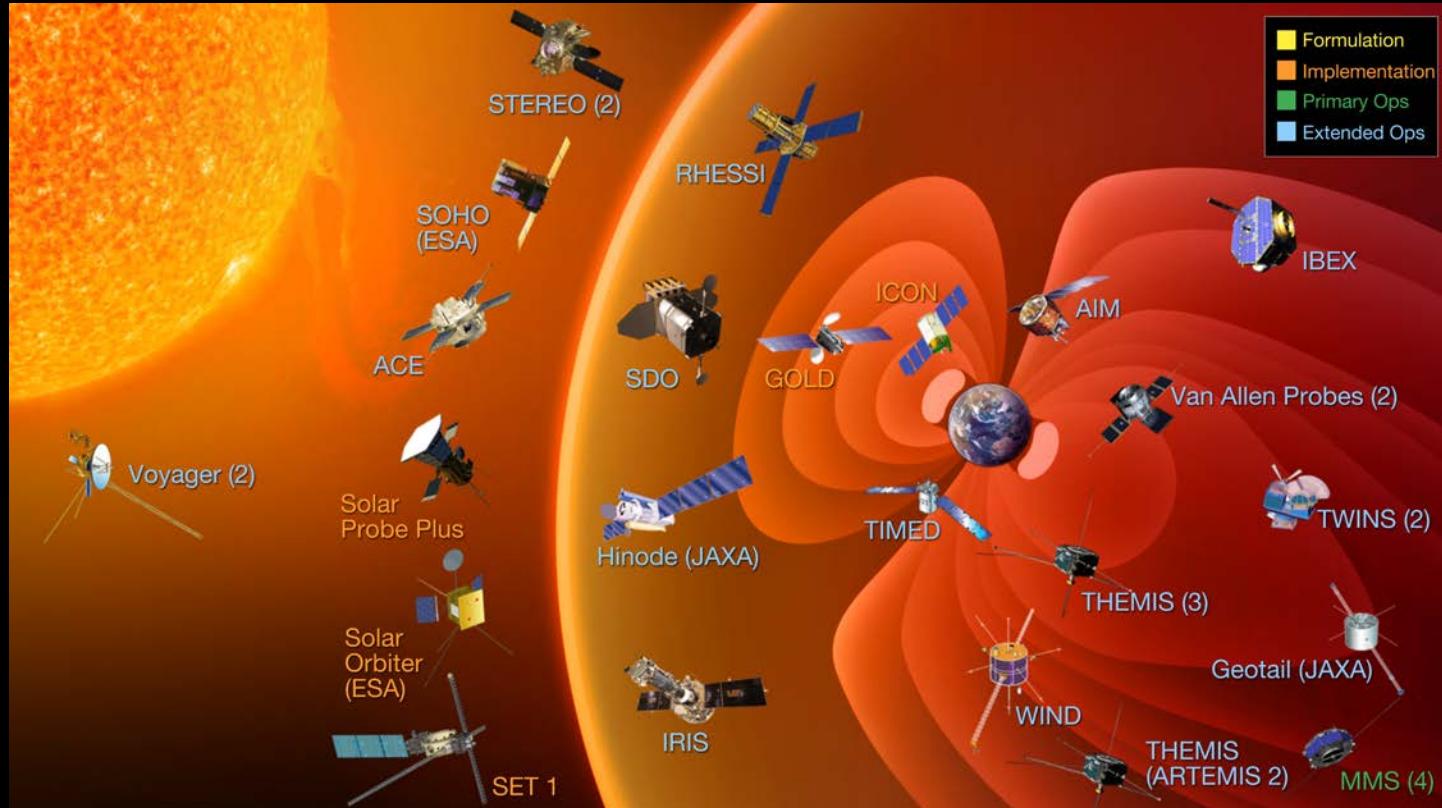
**How is CEDAR evolving
and why do we need
data science?**

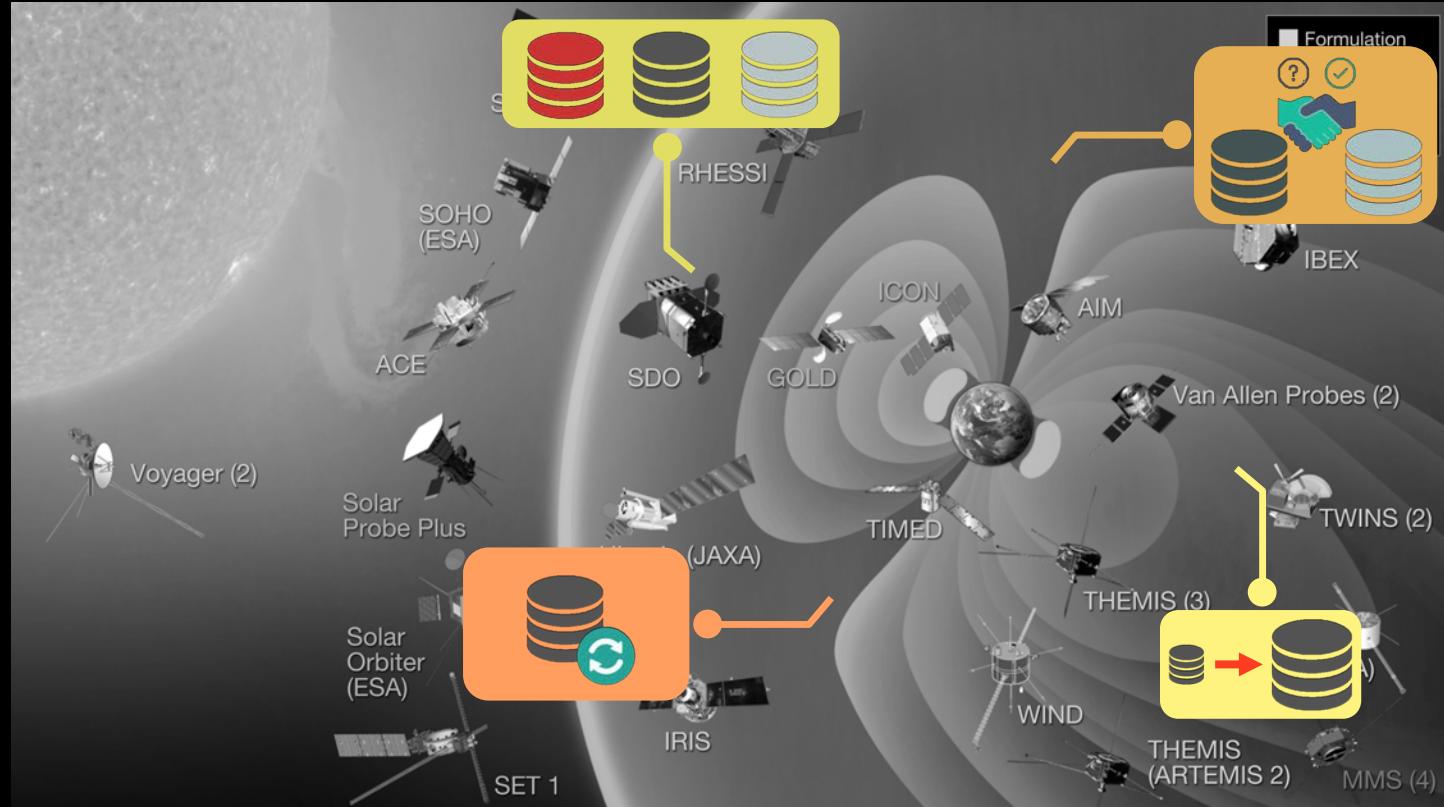
**Why is ionospheric
scintillation a
fantastic use case
and what progress
have we made?**

**What *trends* does
this reveal?**



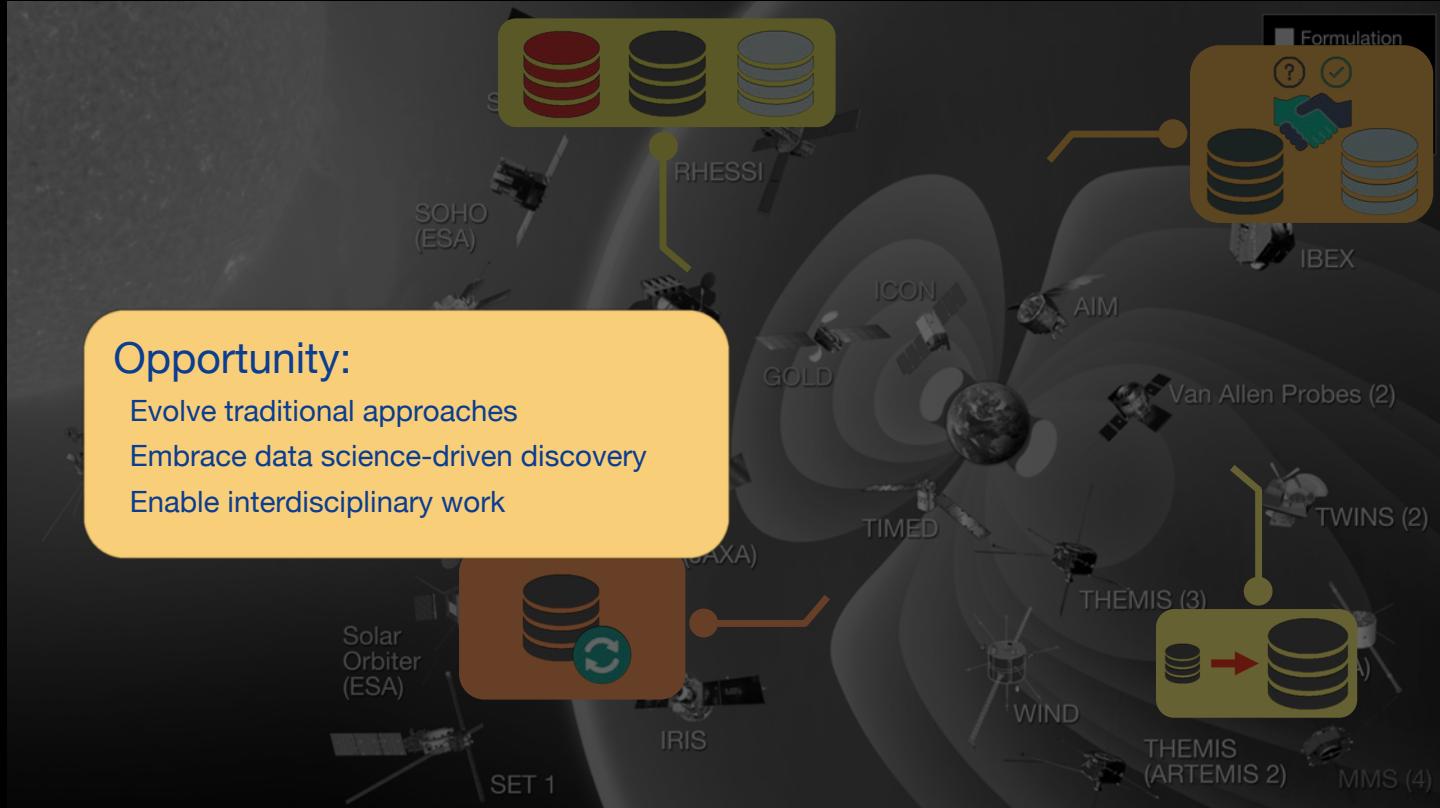
**How is CEDAR evolving
and why do we need
data science?**





Opportunity:

- Evolve traditional approaches
- Embrace data science-driven discovery
- Enable interdisciplinary work





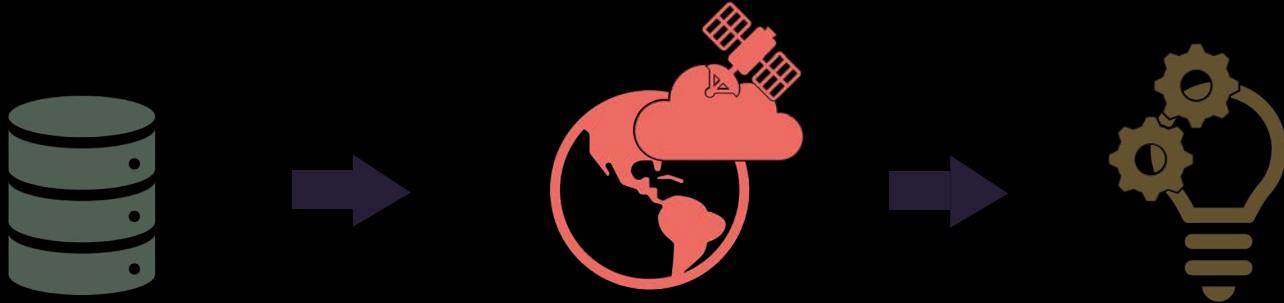
Scalable architectural approaches, techniques, software and algorithms which alter the paradigm by which data are collected, managed and analyzed.

Dan Crichton, JPL

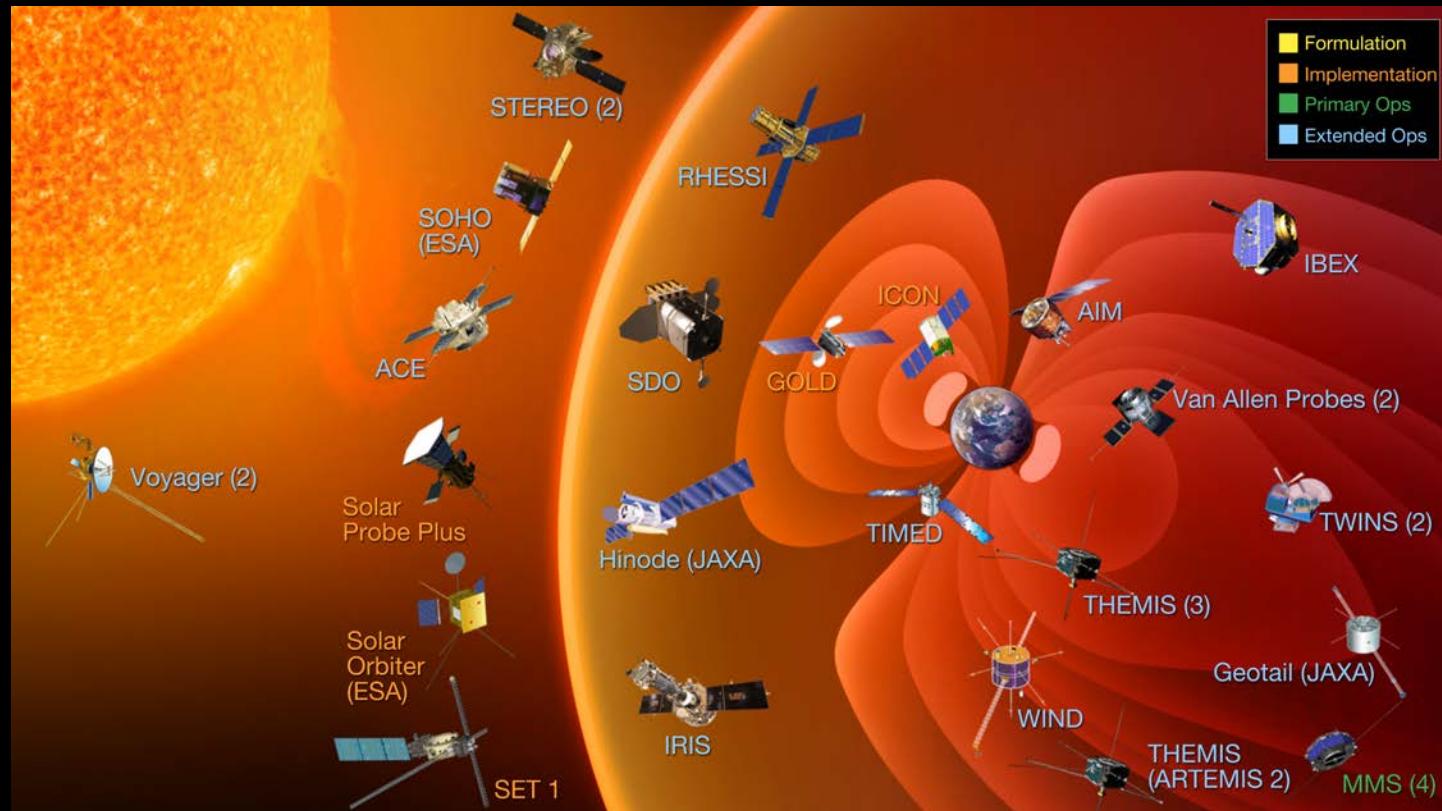


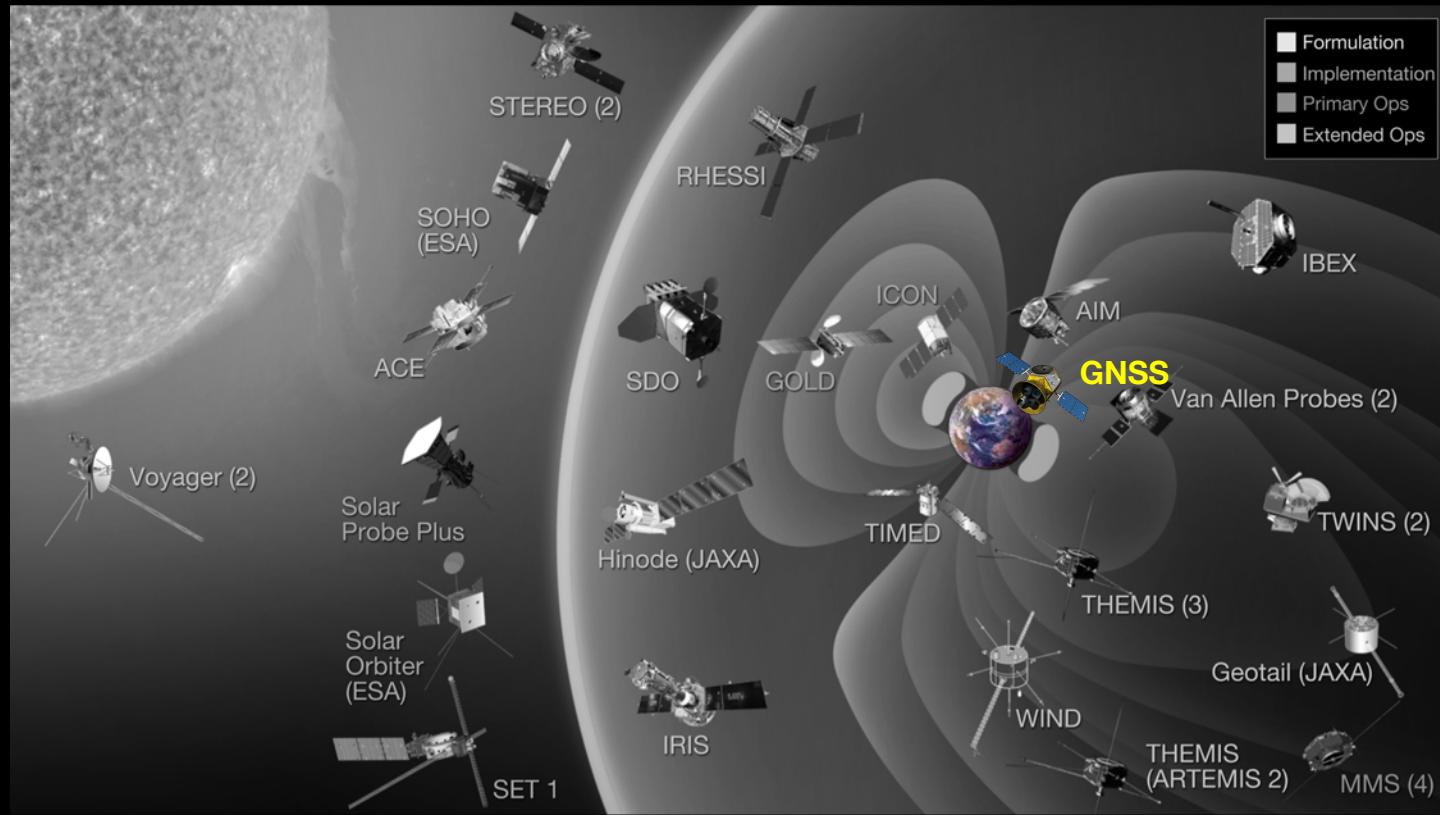
Someone or something that doesn't fit within traditional academic discipline-a field of study with its own particular words, frameworks, and methods

Joi Ito, MIT Media Lab, "Antidisciplinary"



**Why is ionospheric
scintillation a
fantastic use case
and what progress
have we made?**

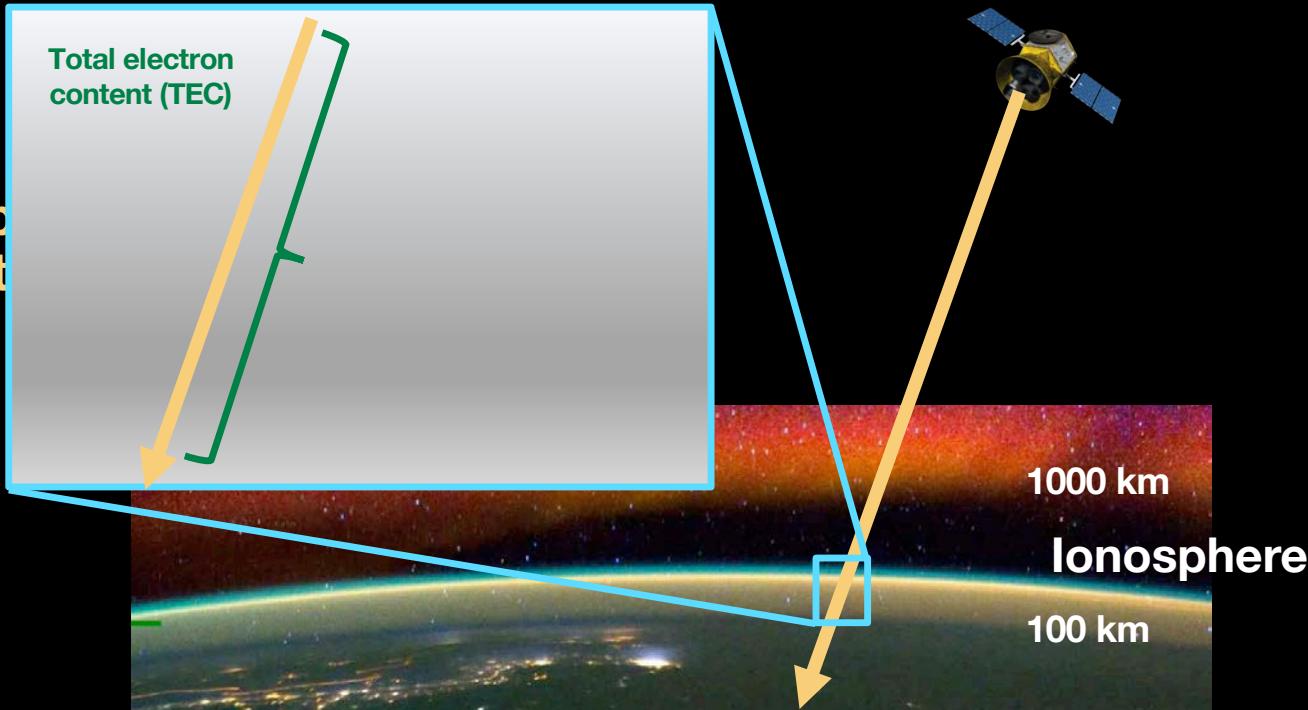




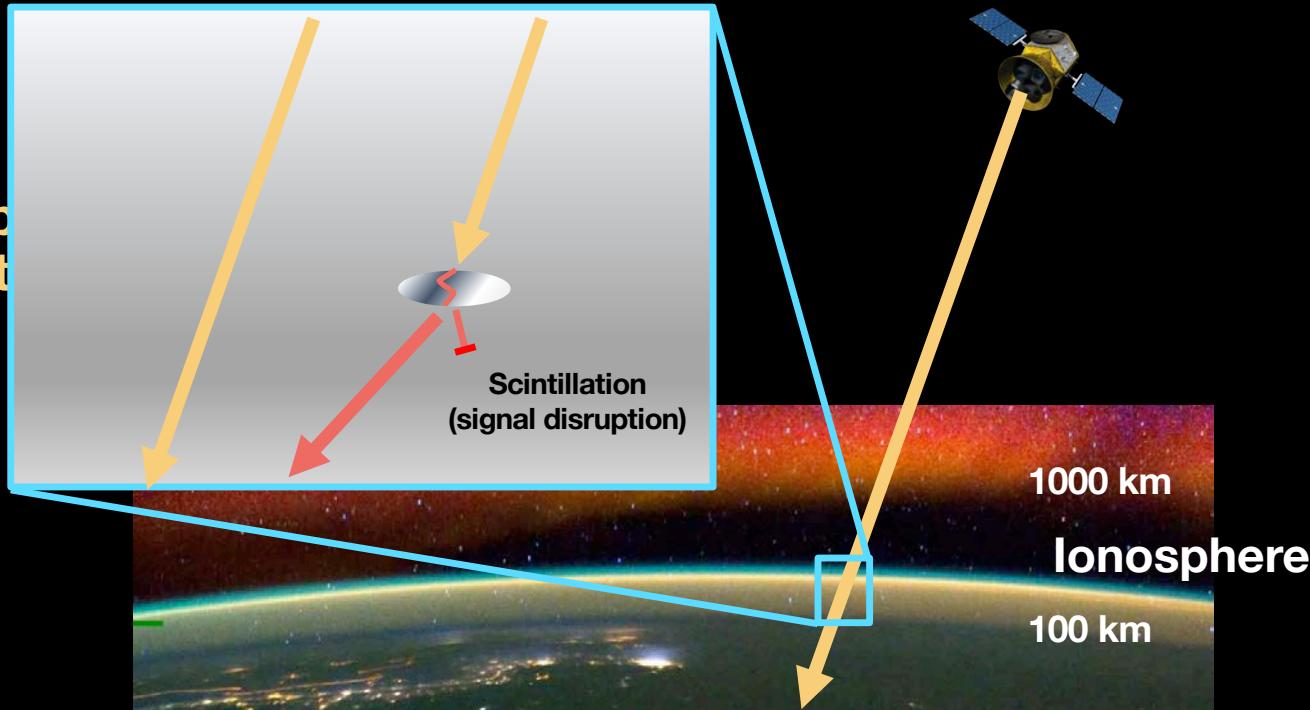
Global Navigation Satellite System (GNSS) signals for Space Science

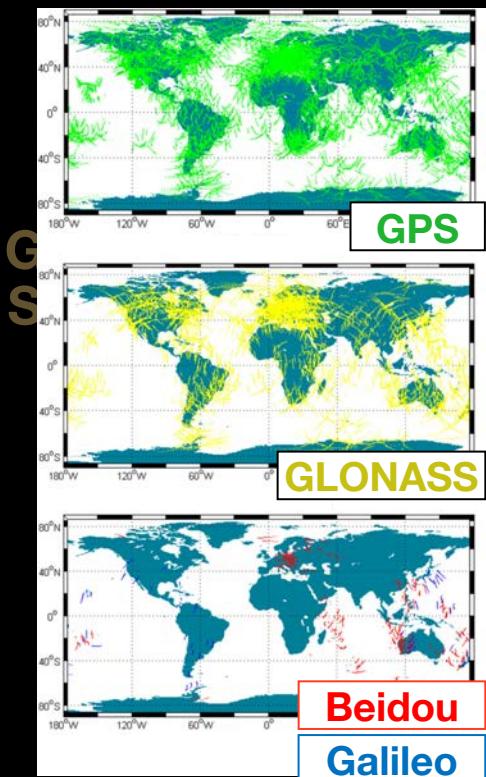


Global System



Global System





Satellite
signals for
navigation



Support Vector
Machine (SVM)

Decision Trees

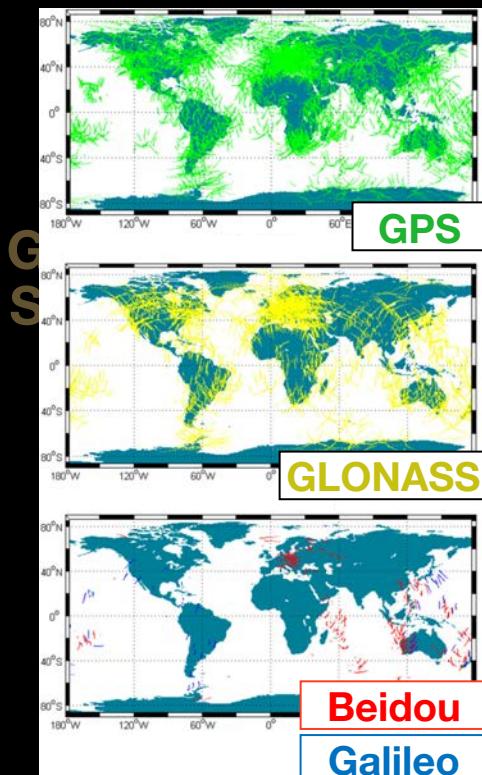
Random
Forests

Neural
Networks

Easily explainable

Difficult to explain

Create a narrative of new scientific understanding
across spectrum of machine learning approaches



Satellite
signals for
navigation



1000 km
Ionosphere
100 km

Support Vector Machine (SVM)

Decision Trees

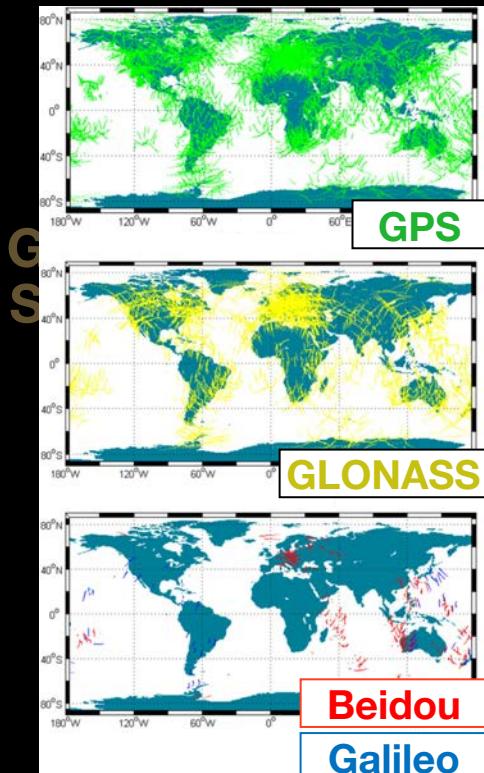
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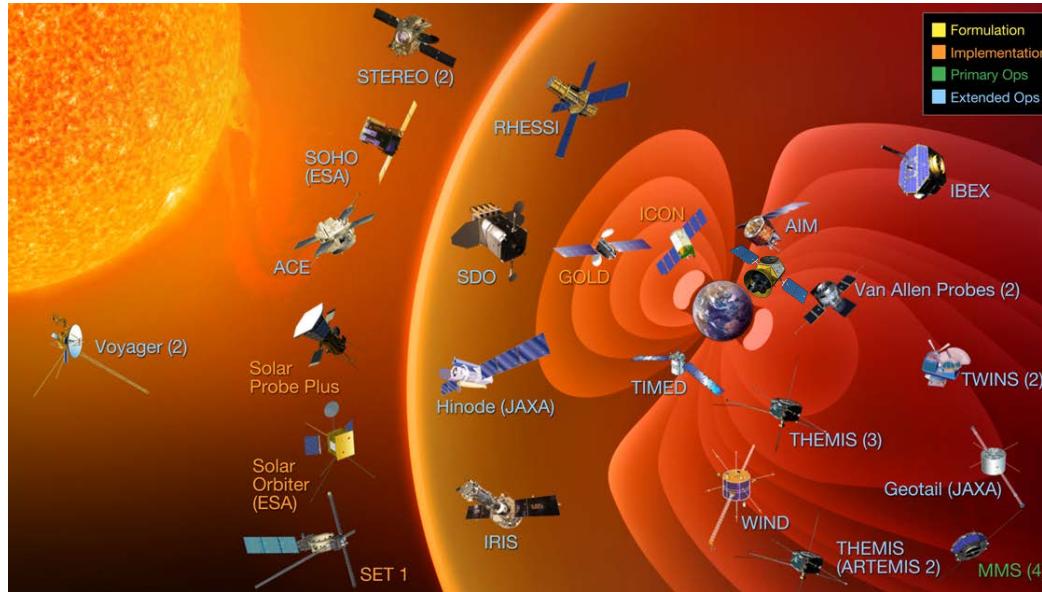


Satellite
signals for
navigation



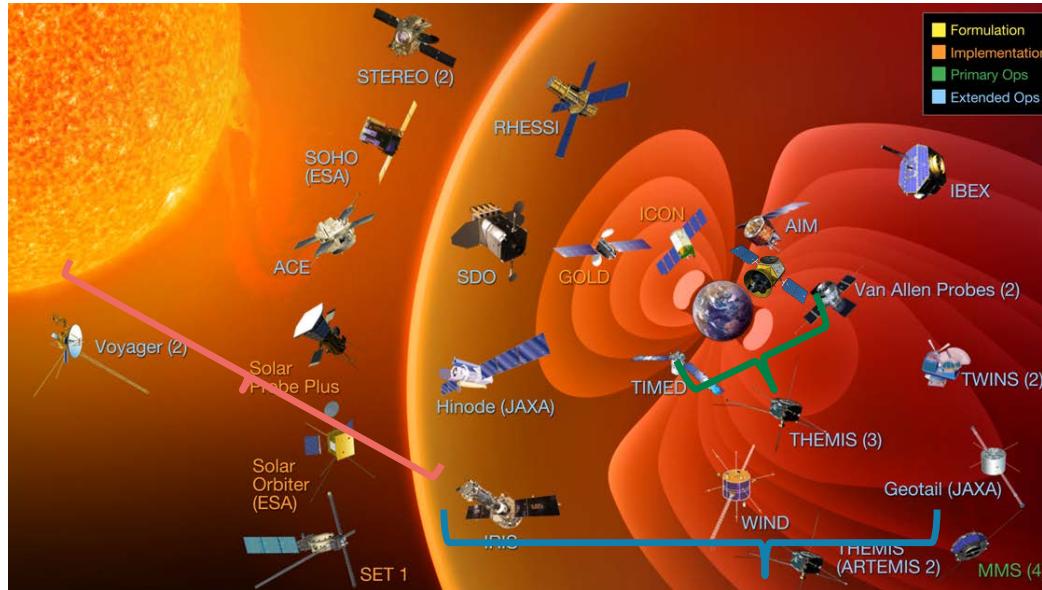
Step 1:

Obtain solar, geomagnetic, and ionospheric data



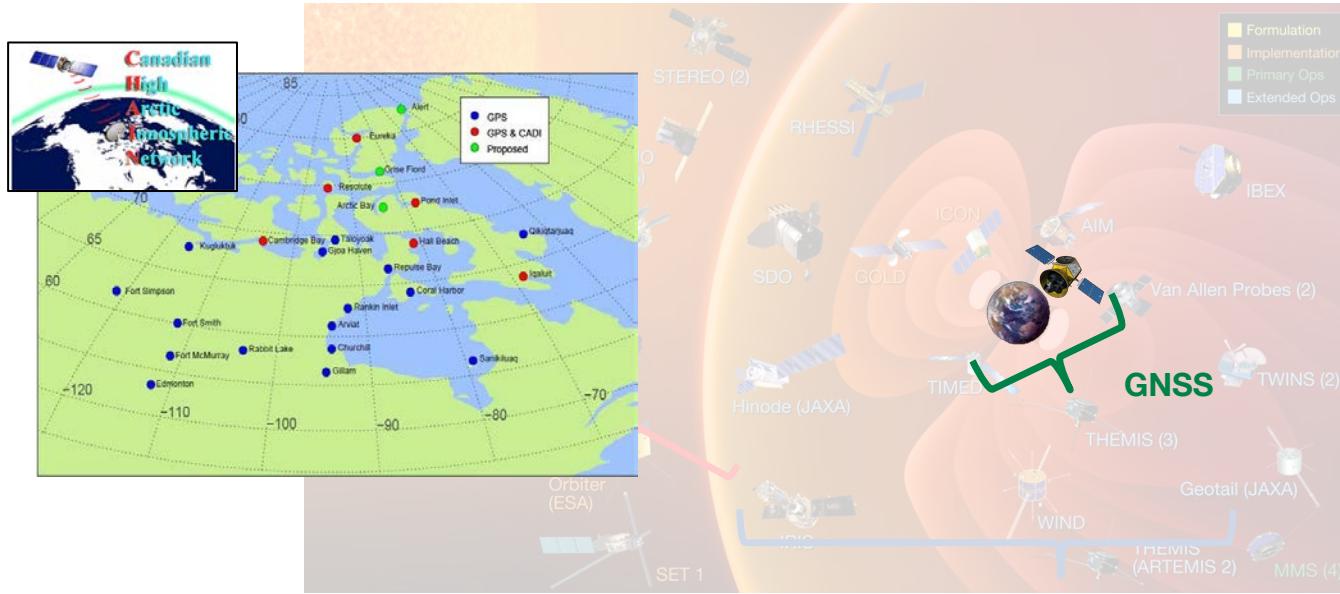
Step 1:

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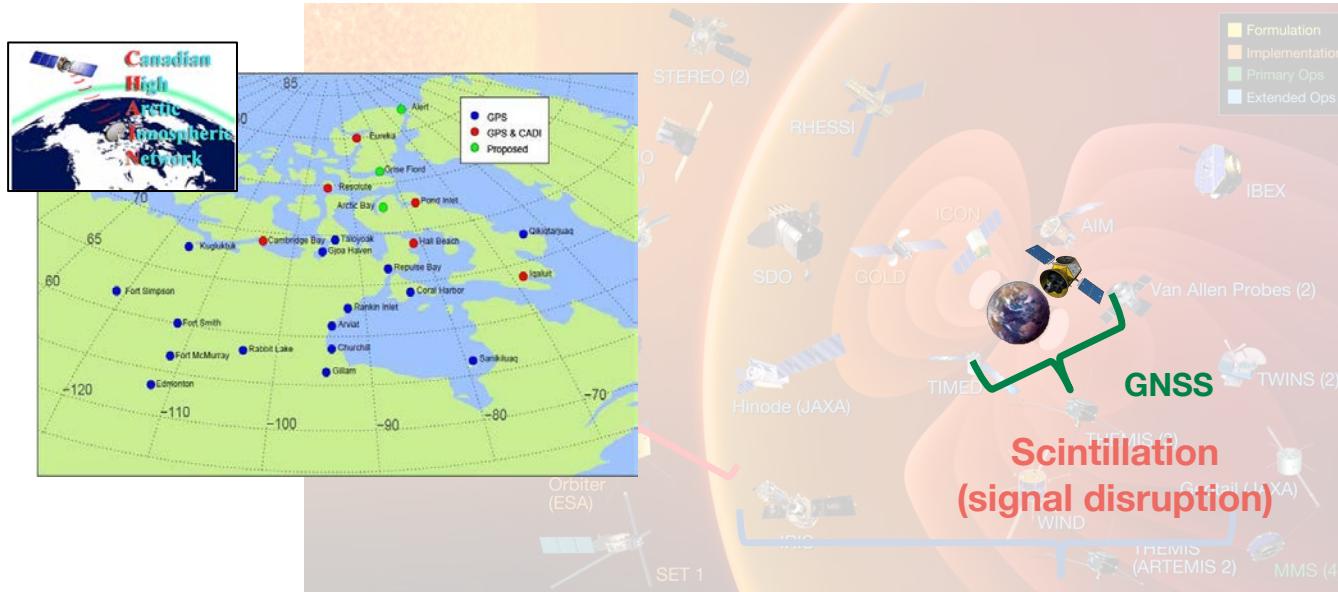
Step 1:

Obtain solar, geomagnetic, and ionospheric data



Step 2:

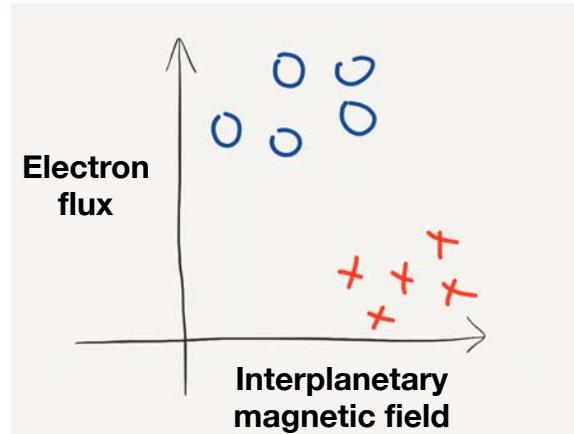
Define the predictive task



Step 3:

Machine learning algorithm for prediction

Support Vector Machine

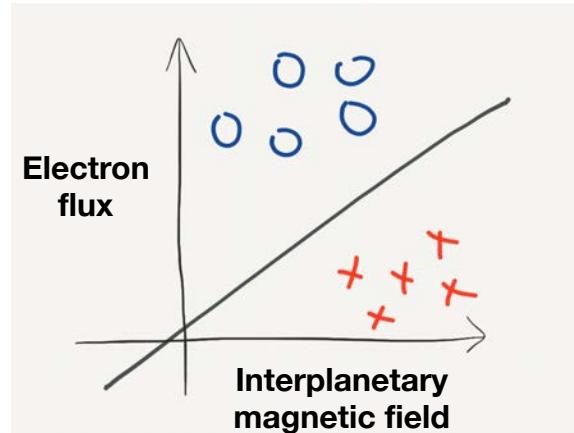


Cortes and Vapnik (1995)

Step 3:

Machine learning algorithm for prediction

Support Vector Machine

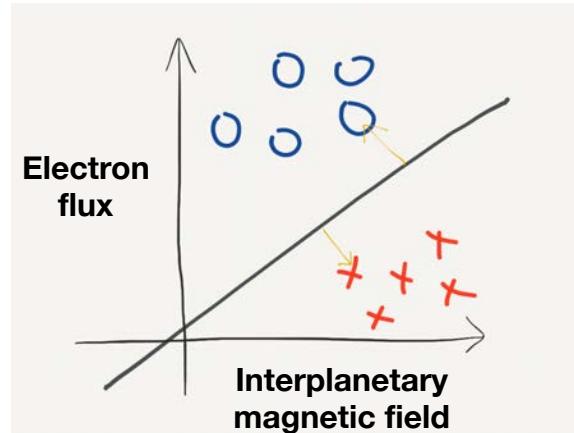


Cortes and Vapnik (1995)

Step 3:

Machine learning algorithm for prediction

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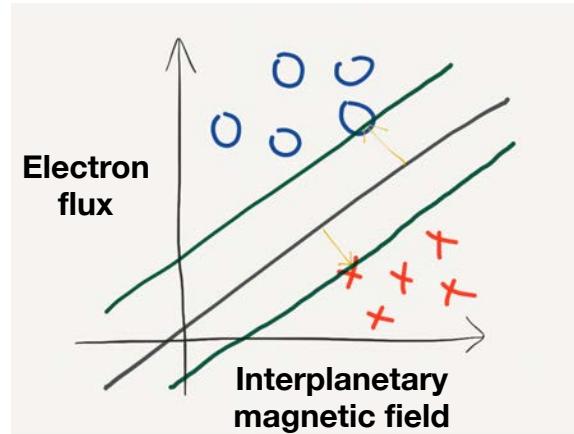


Cortes and Vapnik (1995)

Step 3:

Machine learning algorithm for prediction

Support Vector Machine



Cortes and Vapnik (1995)

Step 3:

Machine learning algorithm for prediction

no scintillation

True label

scintillation

True
negative

False
positive

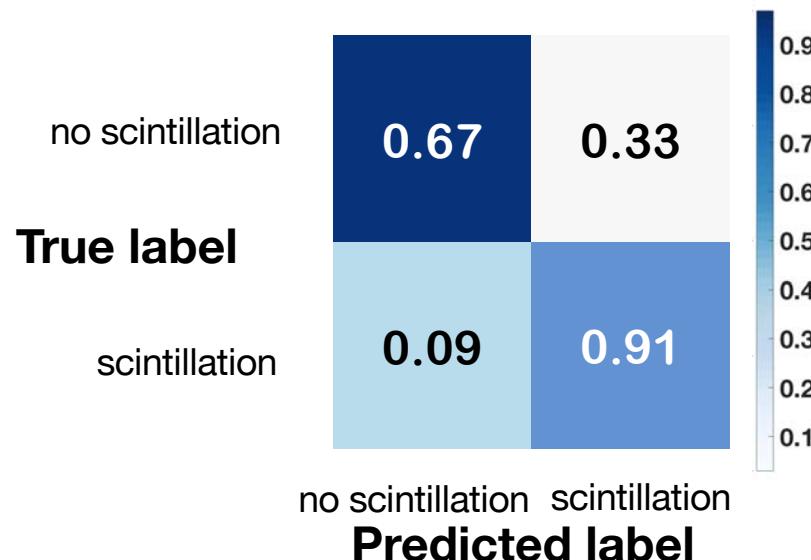
False
negative

True
positive

no scintillation scintillation
Predicted label

Step 3:

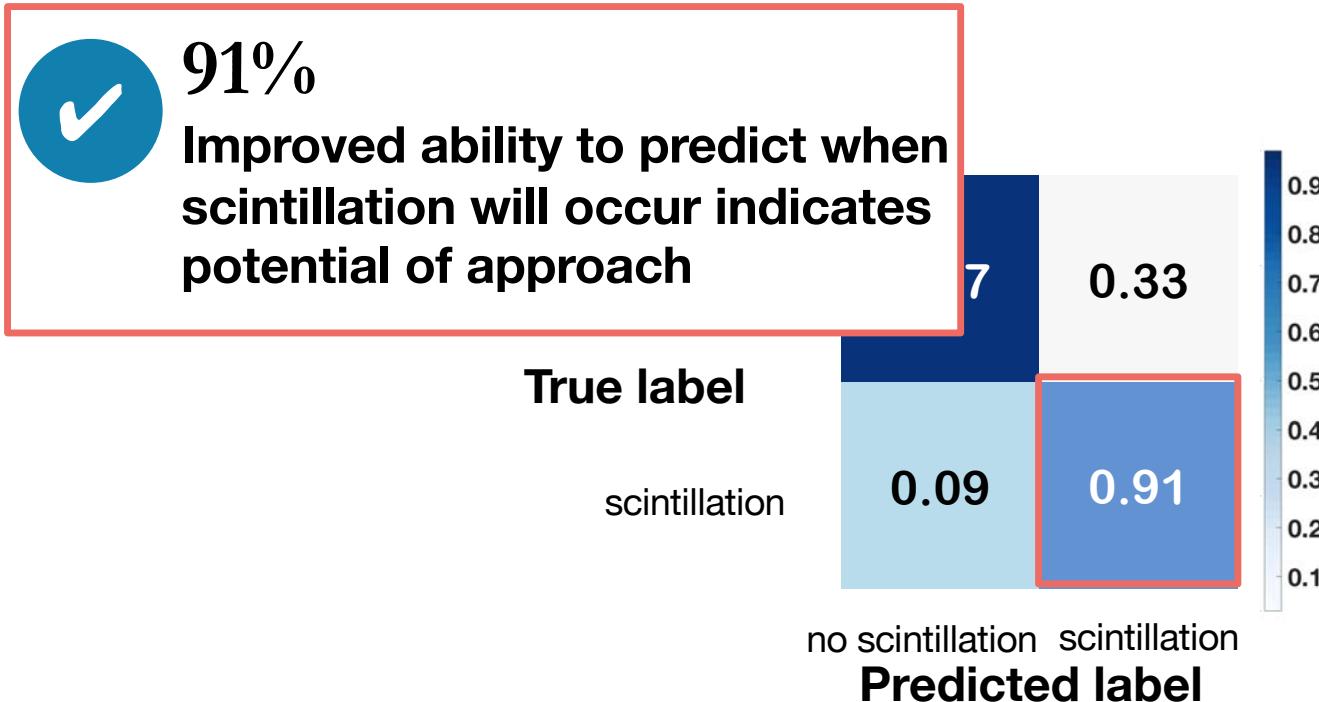
Machine learning algorithm for prediction



McGranaghan et al., (2018)

Step 3:

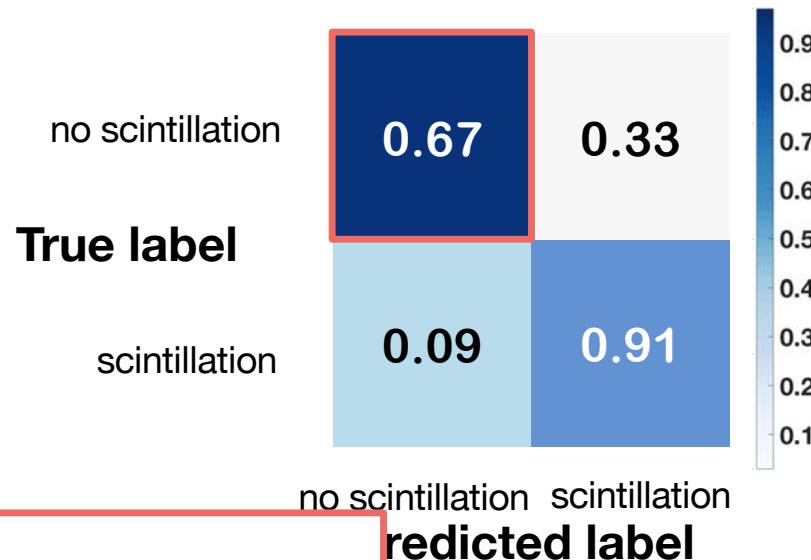
Machine learning algorithm for prediction



McGranaghan et al., (2018)

Step 3:

Machine learning algorithm for prediction



67%

High accuracy predicting when
scintillation would not occur

Step 4:

Interrogate the model

Step 4:

Interrogate the model

Evaluation

Step 4: Interrogate the model

Evaluation

True Skill Statistic (TSS)

$$TSS = \frac{TP}{TP + FN} - \frac{FP}{FP + TN}$$

no scintillation

True label

scintillation

True negative	False positive
False negative	True positive

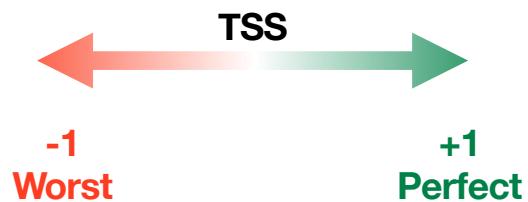
no scintillation scintillation
Predicted label

Step 4: Interrogate the model

Evaluation

True Skill Statistic (TSS)

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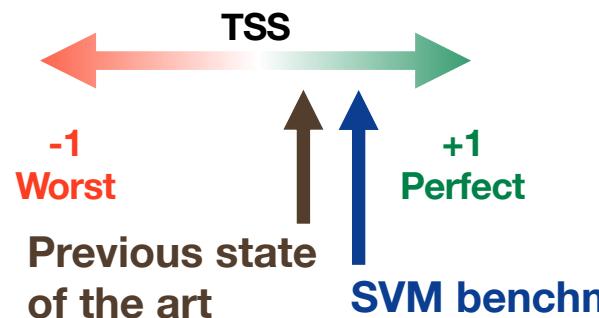
no scintillation scintillation
Predicted label

Step 4: Interrogate the model

Evaluation

True Skill Statistic (TSS)

$$TSS = \frac{TP}{TP + FN} - \frac{FP}{FP + TN}$$



no scintillation

True label

scintillation

True negative	False positive
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no scintillation scintillation
Predicted label

Step 4: Interrogate the model

Evaluation

Explanation

Step 4: Interrogate the model

Evaluation

Explanation

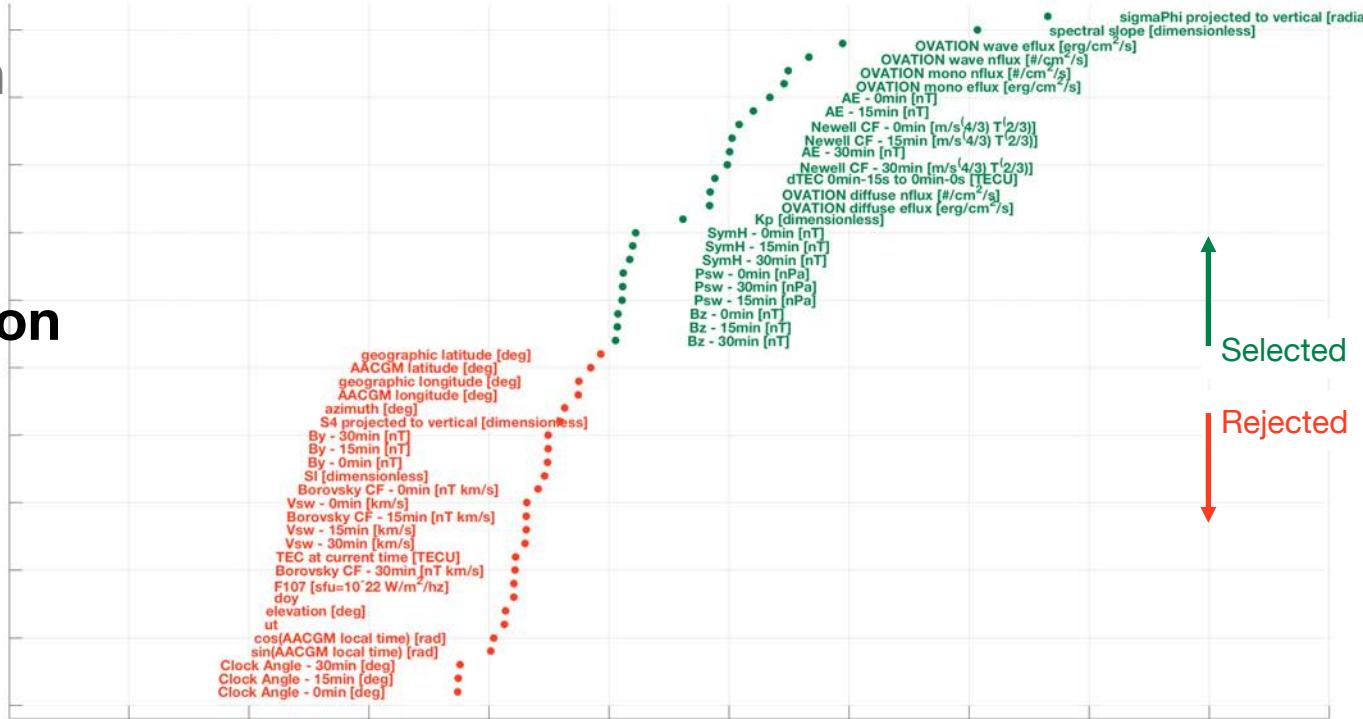
Step 4: Interrogate the mode

Less important

More important

Evaluation

Explanation



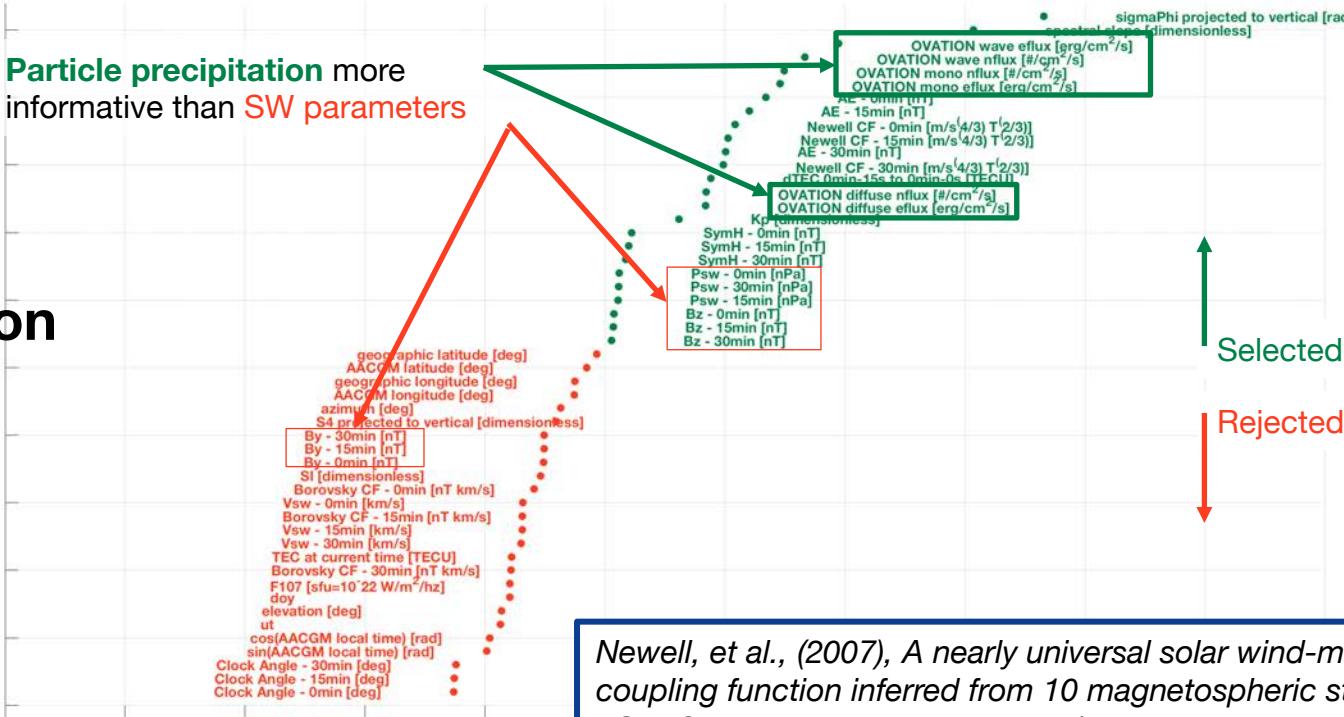
F-Score (i.e., importance)

Step 4: Interrogate the mode

Evaluation

Particle precipitation more informative than **SW parameters**

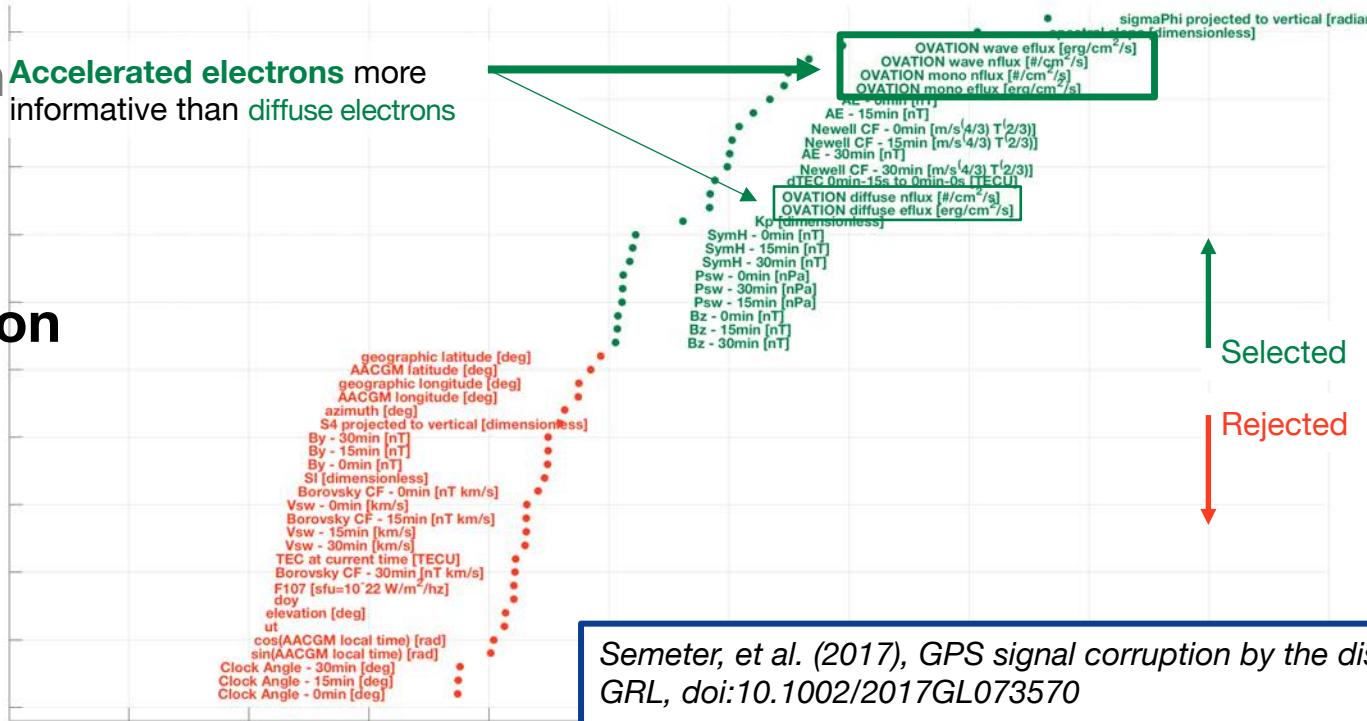
F-Score (i.e., importance)



Newell, et al., (2007), *A nearly universal solar wind-magnetosphere coupling function inferred from 10 magnetospheric state variables*, JGR: Space Physics, doi:10.1029/2006JA012015.

Step 4: Interrogate the mode

Evaluation **Accelerated electrons** more informative than **diffuse electrons**



F-Scor

Semeter, et al. (2017), GPS signal corruption by the discrete aurora. GRL, doi:10.1002/2017GL073570

Mrak, et al., (2017), Field-aligned GPS scintillation: Multisensor data fusion, JGR: Space Physics, doi:10.1002/2017JA024557.

Step 4: Interrogate the mode

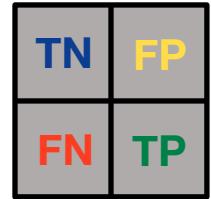
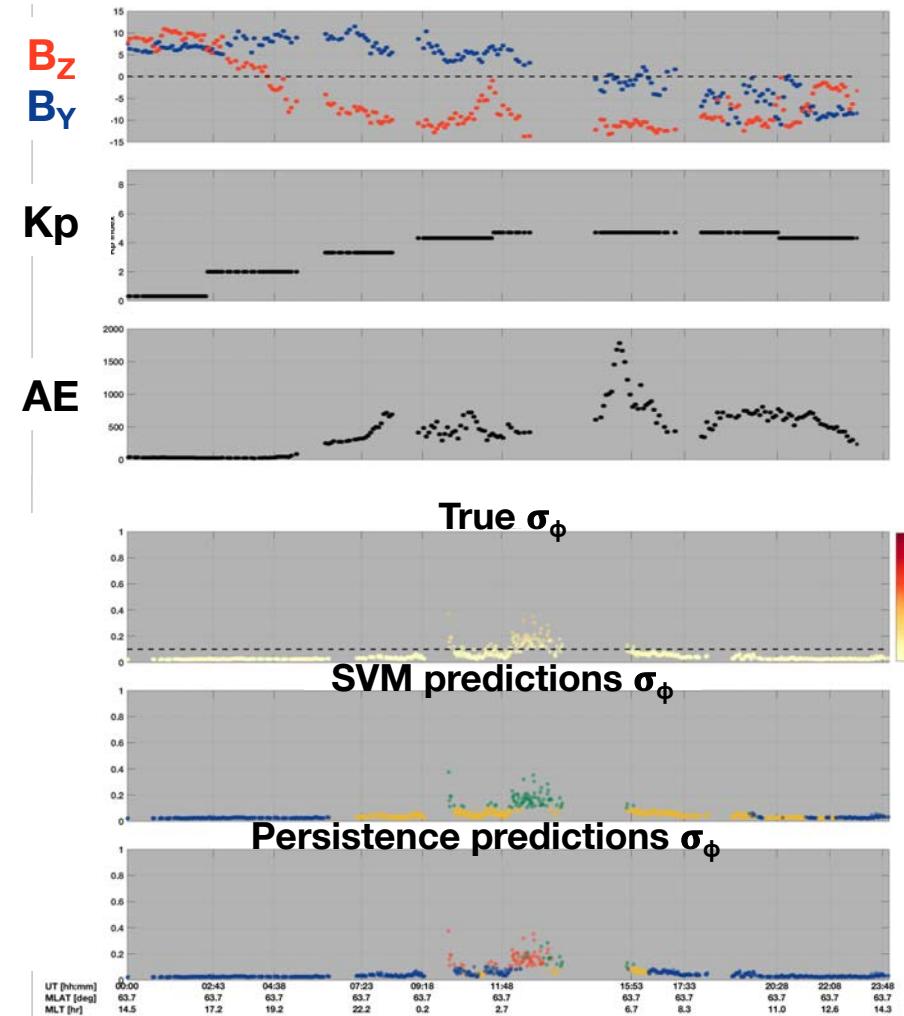
Evaluation

Explanation

January 20, 2016

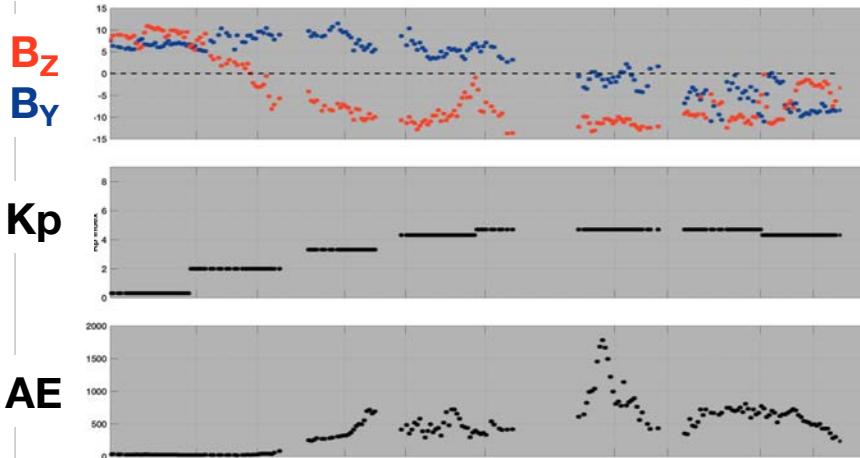
Evaluation

Explanation

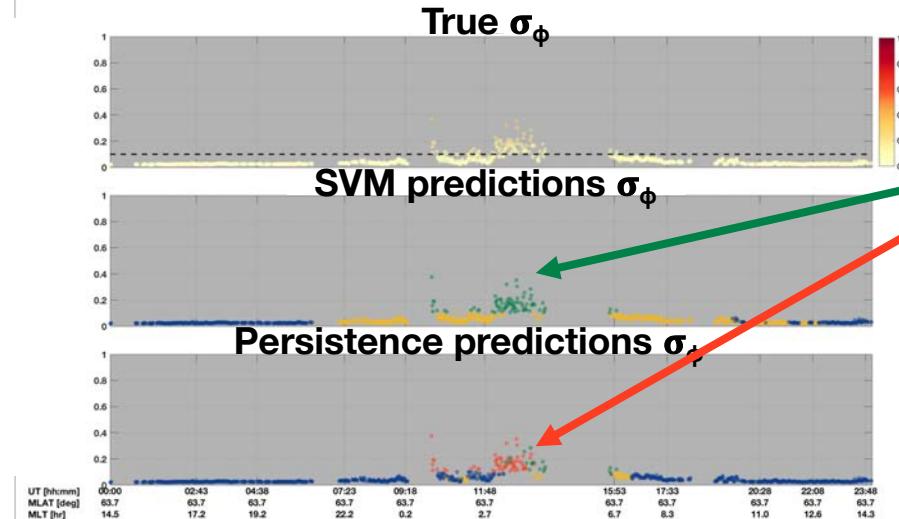


January 20, 2016

Evaluation



Explanation

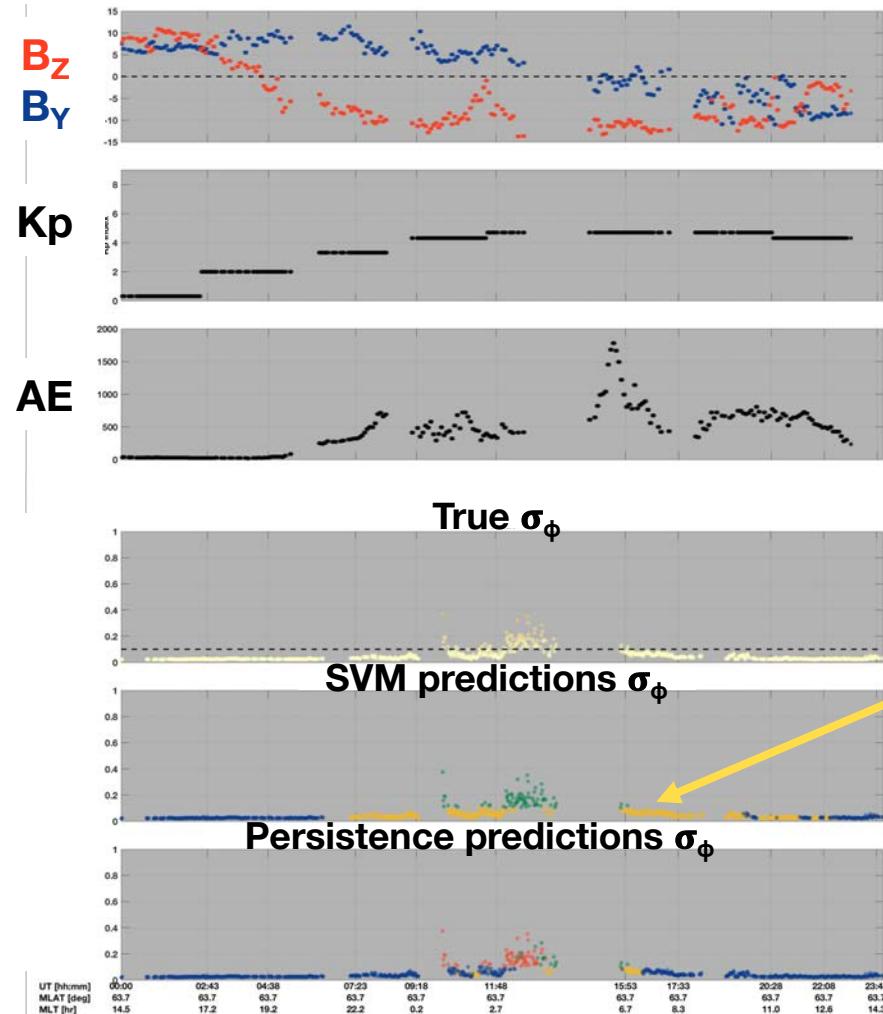


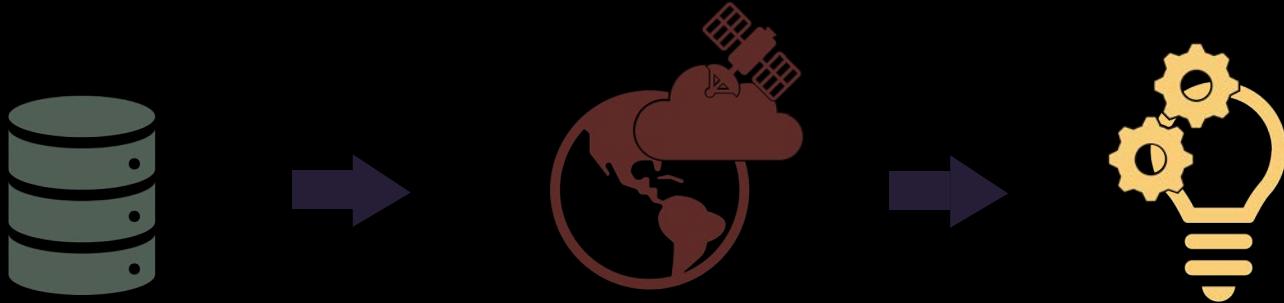
SVM identifies strong
scintillation,
persistence does not

January 20, 2016

Evaluation

Explanation





**What *trends* does
this reveal?**

Be antidisciplinary



What *trends* does
this reveal?

Be antidisciplinary

Be open by default



What *trends* does
this reveal?

Be antidisciplinary

Be open by default

Understand the models



What *trends* does
this reveal?

Trends

Be *antidisciplinary*

Be open by default

Understand the models



@AeroSciengineer



ryan.mcgranaghan@gmail.com



RyanMcGranaghan.com

McGranaghan, R. M., Bhatt, A., Matsuo, T., Mannucci, A. J., Semeter, J. L., & Datta-Barua, S. (2017). Ushering in a new frontier in geospace through data science. *Journal of Geophysical Research: Space Physics*, 122, 12,586–12,590.

<https://doi.org/10.1002/2017JA024835>

McGranaghan, R. M., A.J. Mannucci, B.D Wilson, C.A. Mattmann, and R. Chadwick. (2018), New capabilities for prediction of high-latitude ionospheric scintillation: A novel approach with machine learning, *Space Weather*, 16. <https://doi.org/10.1029/2018SW002018>

Curated Sources of Data Science Learning Resources

Ryan McGranaghan running list of resources (Github repository)

- https://github.com/rmcgranaghan/data_science_tools_and_resources

HelioAnalytics website and list of resources

- <https://sites.google.com/view/heliodata/resources?authuser=0>