

Exploring Sea Surface Temperature Data Using Machine Learning

23rd GHRST International Science Meeting
Madolyn Kelm



BACKGROUND

Giving background on the work I built off of.

01

GOALS AND RESULTS

Reporting on the results from the prior proposed goals.

02

TABLE OF CONTENTS

03

THE WRAP UP

The current conclusions of our project.

01

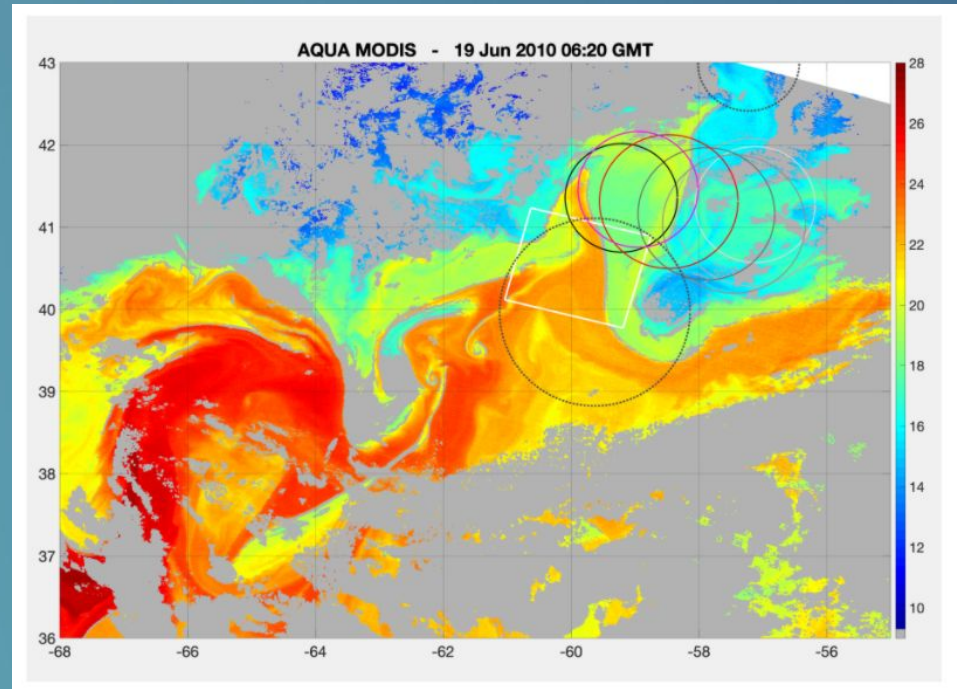
BACKGROUND



Problem

- Sea surface temperature (SST) datasets are massive and show complex patterns
- Current standard oceanographic measures (e.g. power spectrum) fail to describe the full complexity

Possible Solution



THE DATA COLLECTION

	MODIS	VIIRS
Sampling Period		
Spatial Resolution		



1. PRE-PROCESSING

Extract 128x128 km SST regions, known as cutouts. Approx. one million pre-processed cutouts per year.



MODIS

VIIRS

Sample Size

**Cloud Free
Percentage**

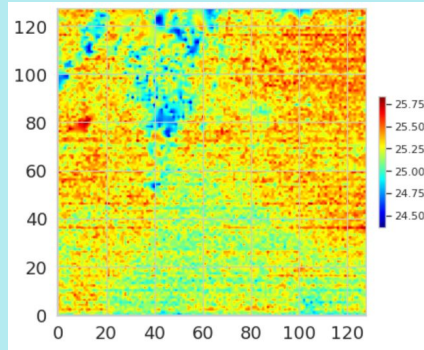
**Final Avg.
Region**

MODIS

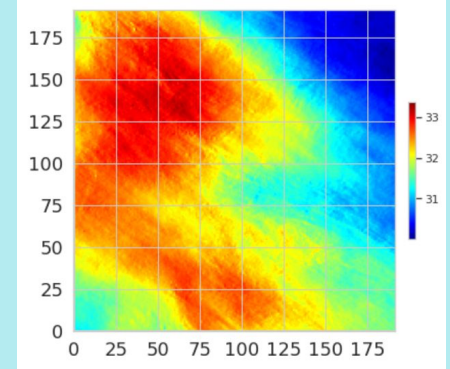
VIIRS

Sample Size

128 x 128 km
128 x 128 pixel



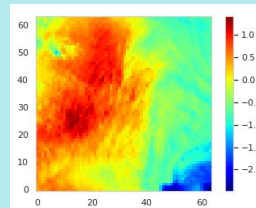
192 x 192 km
128 x 128 pixel



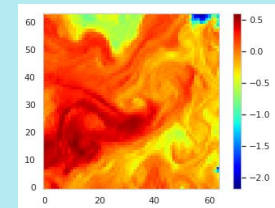
Cloud Free Percentage

Final Avg. Region

64 x 64 pixel



64 x 64 pixel





1. PRE-PROCESSING

Extract 128x128 km SST regions, known as cutouts. Approx. one million pre-processed cutouts per year.



2. TRAINING

Trained on ~150,000 random pre-processed cutouts from 2010.

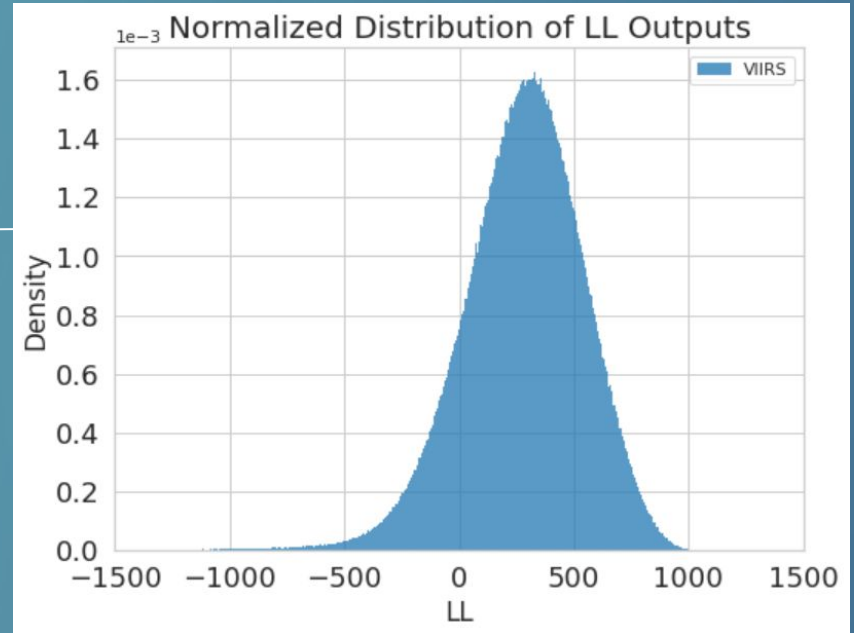
3. APPLY.

Apply the trained model to all pre-processed cutouts.

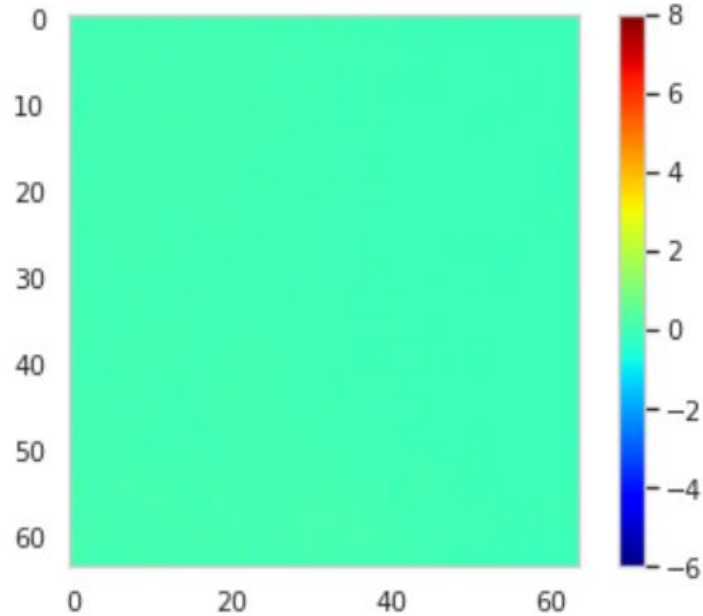
Machine Learning Metric: LL

THE OUTPUT

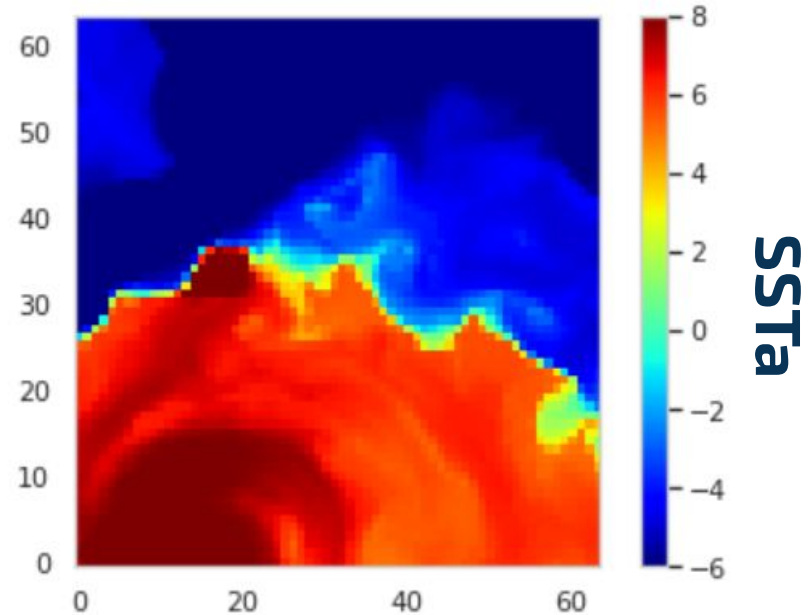
Each cutout is assigned a single number that represents the frequency of images similar to it within the full distribution.



EXAMPLES OF CUTOUTS WITH HIGH and LOW LL VALUES



HIGH
LL Value: 1060.52



LOW
LL Value: -21033.65



02

GOALS AND RESULTS

INITIAL GOALS

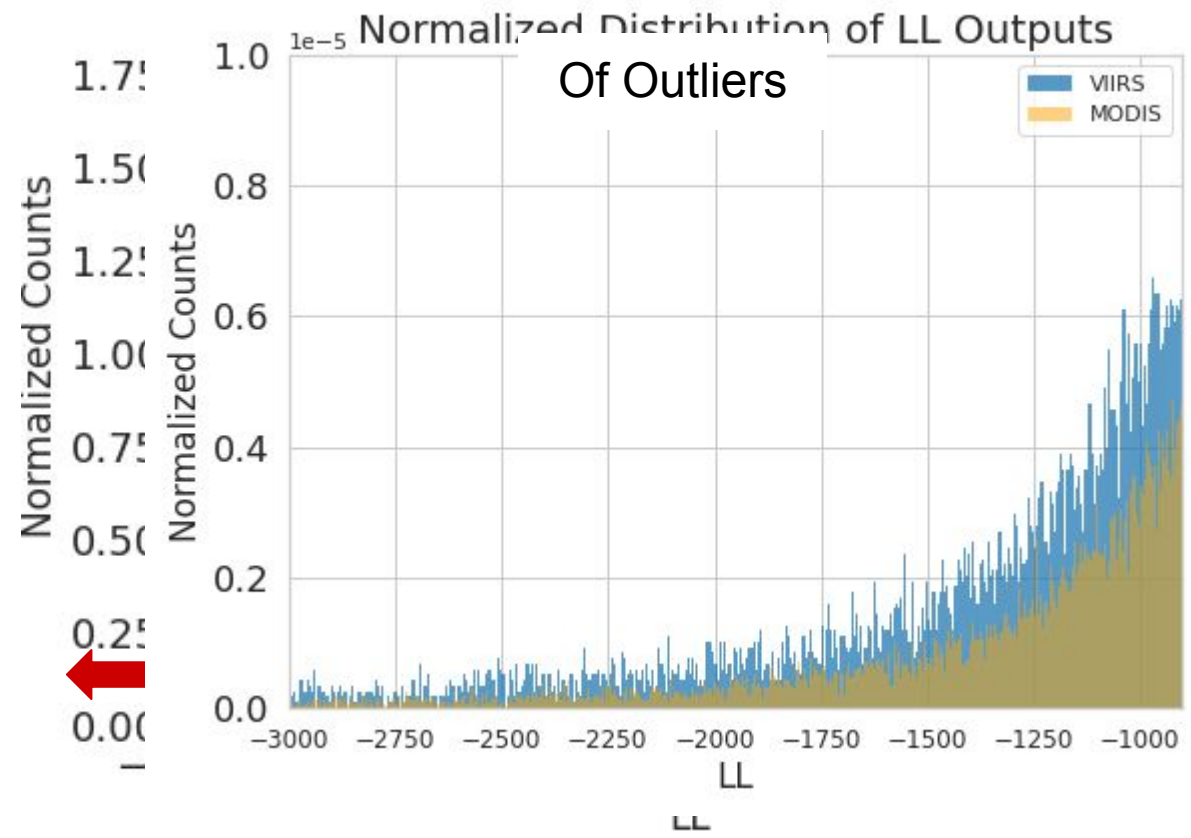


VERIFY ULMO

Use the MODIS
trained ULMO on
VIIRS data

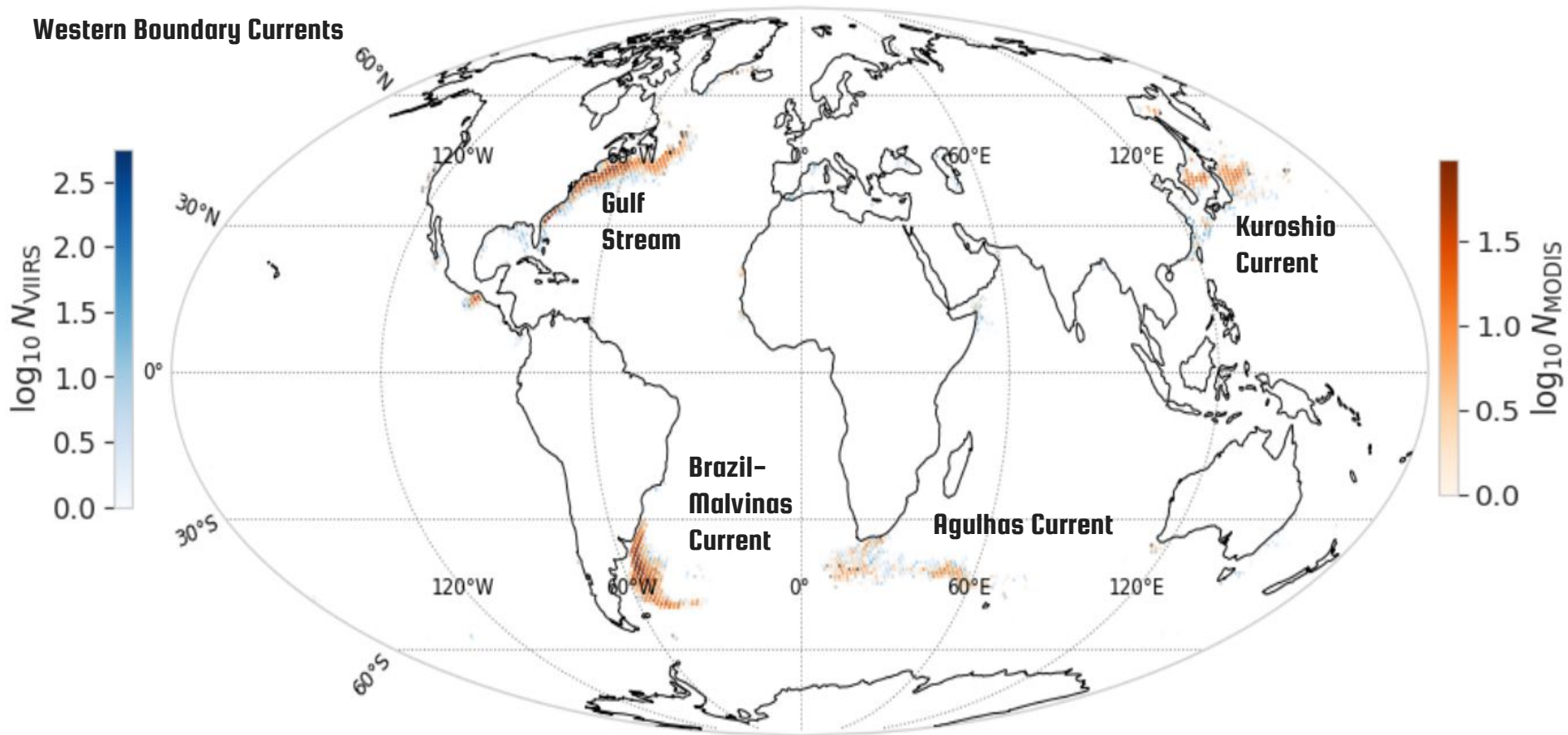


Normalized Distribution of LL Outputs



Cutouts with Log Likelihood Values < -1000 for MODIS and VIIRS

Western Boundary Currents



VERIFY ULMO



HIGHLIGHTS

- The MODIS trained ULMO was able to recognize similar, but not identical, patterns in the VIIRS dataset
- Moving forward with understanding the pattern recognition within VIIRS

INITIAL GOALS

```
graph TD; A[INITIAL GOALS] --> B[VERIFY ULMO]; A --> C[EXPLORE]
```

VERIFY ULMO

Use the MODIS
trained ULMO on
VIIRS data

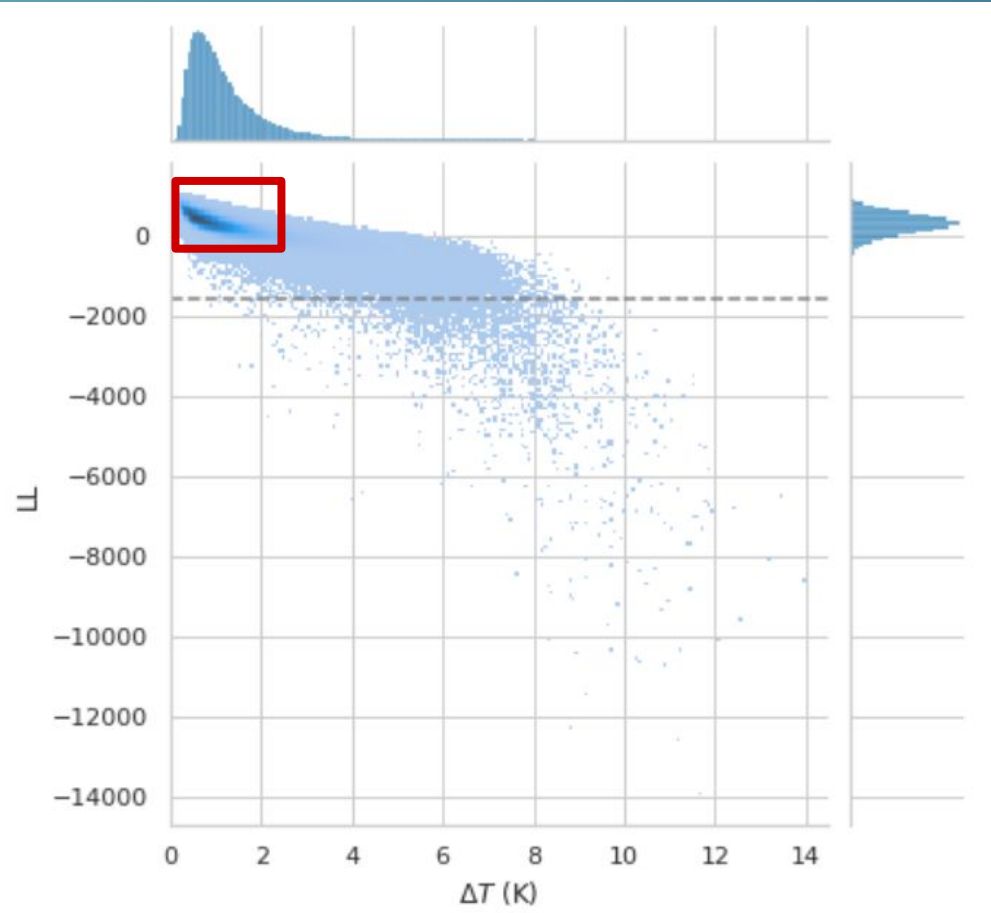
EXPLORE

Explore the patterns
ULMO has
recognized

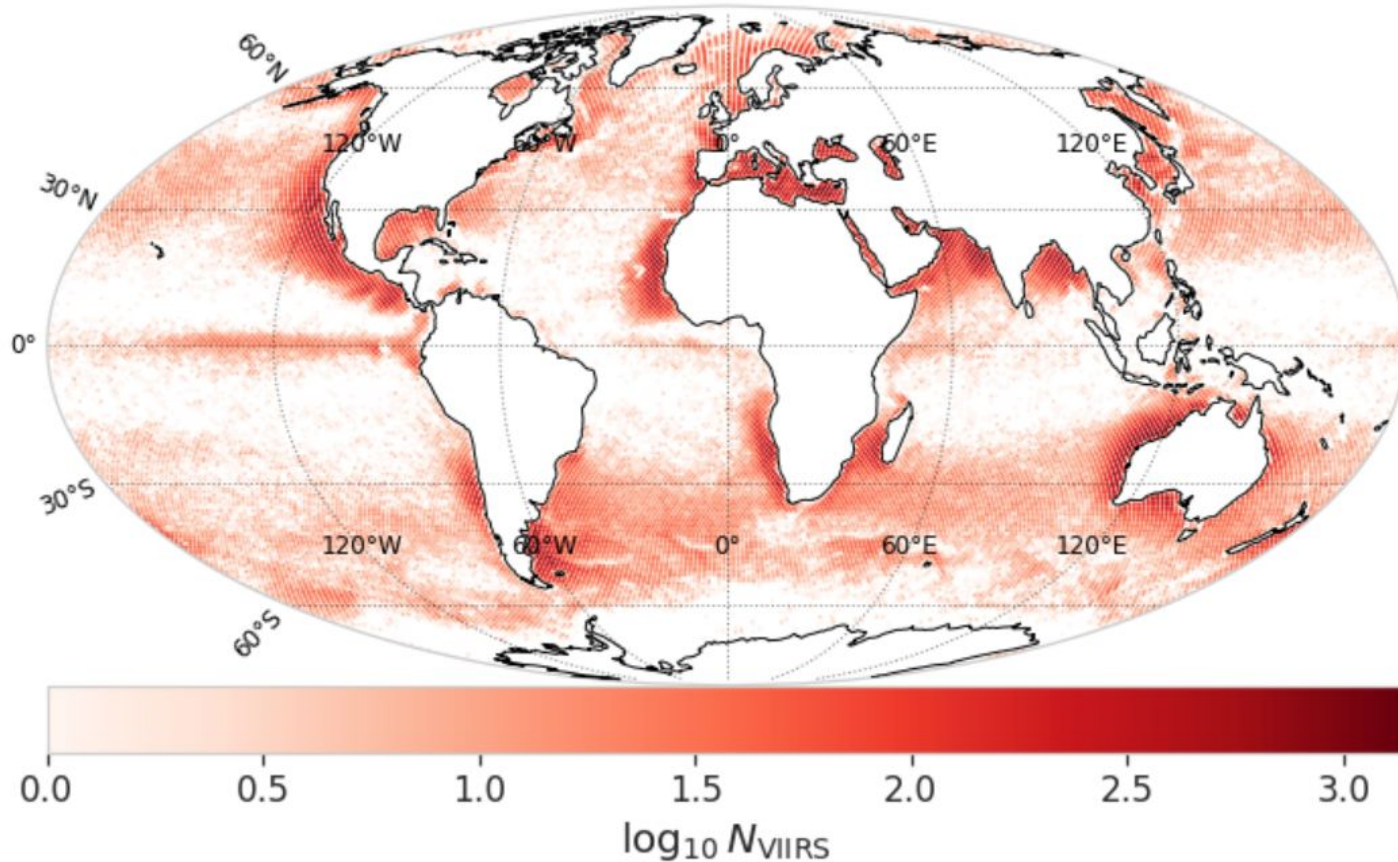
INVESTIGATION OF THE LL VALUES

LL Values

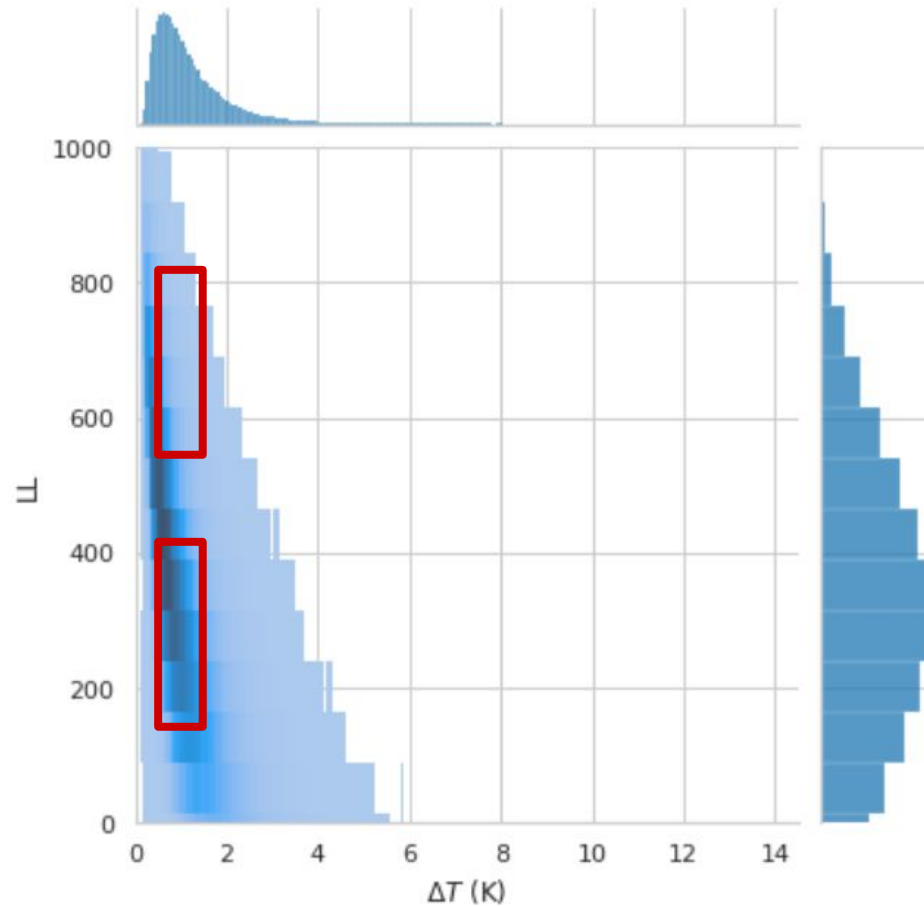
- We have started to see that the outliers ($LL < -1000$) show more structure with some dependence on temperature range.
- Now I will explore the larger portion of the distribution



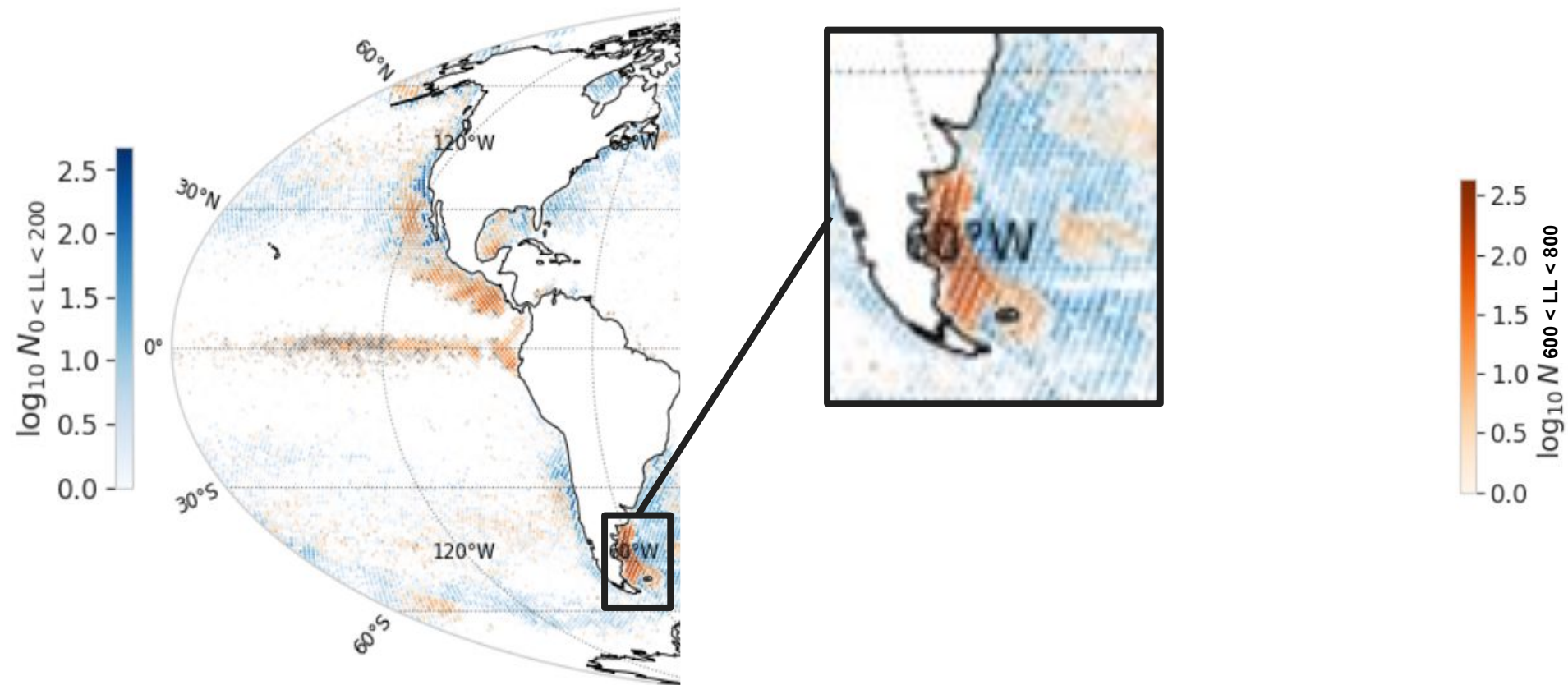
Cutouts with Temperature Range $1 < \Delta T < 1.5$ VIIRS



Cutouts with Temperature Range $1 < \Delta T < 1.5$ VIIRS



Cutouts with $1 < \Delta T < 1.5$ and Specified LL Value Range





03

CONCLUSION NEXT STEPS

CONCLUSIONS

HIGHLIGHTS

- ULMO highlighted differences in SST datasets
- LL is a good discriminator over the entire range of the LL rather than just the tail

FUTURE

- Use this to now explore the LL space
 - What patterns exist in the LL distribution
 - How LL values relate to the basic geophysical properties



TAKE HOME MESSAGE

The ULMO algorithm is successful at identifying complex patterns within extremely large sea surface temperature datasets, even upwards of 100 TB.

Its machine learning metric (LL) highlights geophysical processes in distinct locations.



ACKNOWLEDGEMENTS



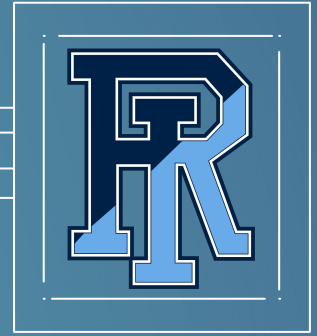
**National Science
Foundation**




Mentors

Peter Cornillon and
J. Xavier Prochaska:

5257-LEARNING THE FUNDAMENTAL
PATTERNS OF SST IMAGERY ON
SUB-MESOSCALES Slides at:
<https://tinyurl.com/x-at-osm2022>



**University of
Rhode island**



**AND THANKS TO
YOU!**
Any Questions?