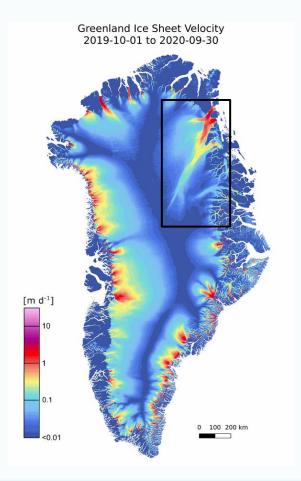
## Predicting Seasonal Velocity Values at Zachariæ Isstrøm, Northeast Greenland

Claire Jensen ESS-569 Final



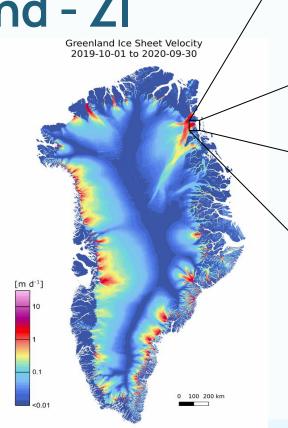
## Background

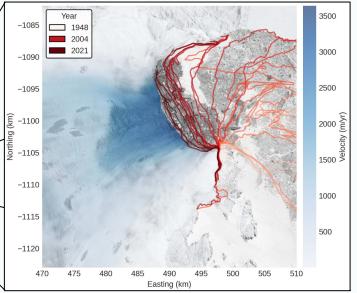
- NEGIS **drains ~12%** of the ice sheet
- 3 glaciers drain the NEGIS:
   (Zachariæ Isstrom (ZI),
   Nioghalvfjerdsfjorden (79N) and
   Storstrømmen (Store))



Background - ZI

lce tongue
 collapsed in
 2013, leading
 to dramatic
 speedup



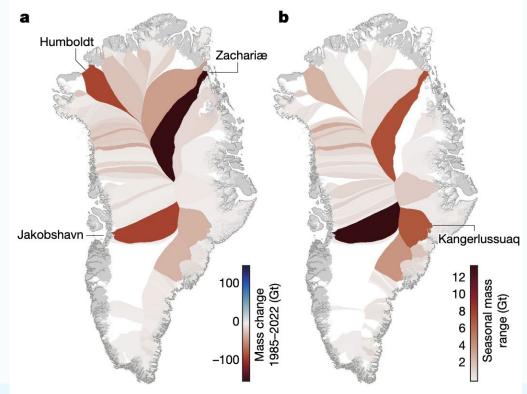


Zl's velocity (blue) and terminus over time (red)

Copernicus Climate Change Service (2020)

## Seasonality at NEGIS

- Higher seasonality correlates with larger long-term mass loss
- NEGIS has lost the most mass and has the second most seasonal mass change



Greene et al. (2024)

a. Cumulative mass change by catchment b. Seasonal mass change by catchment

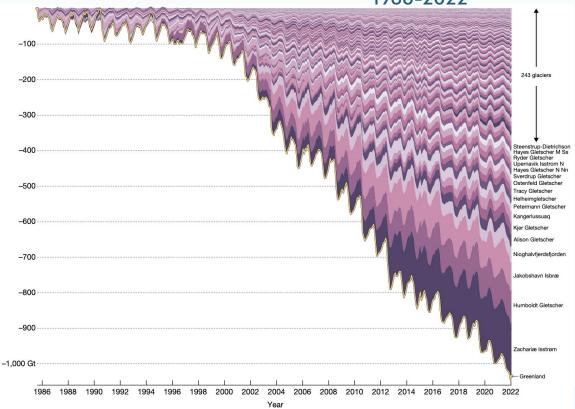
Seasonality at ZI

- ZI has the **highest** 

cumulative mass loss

- Summer meltwater discharge and mélange (sea ice) buildup might contribute to seasonality

Cumulative mass change 1986–2022



Greene et al. (2024)

## Goals

Tier 1

Predict velocity values at 3 points on ZI

Tier 2

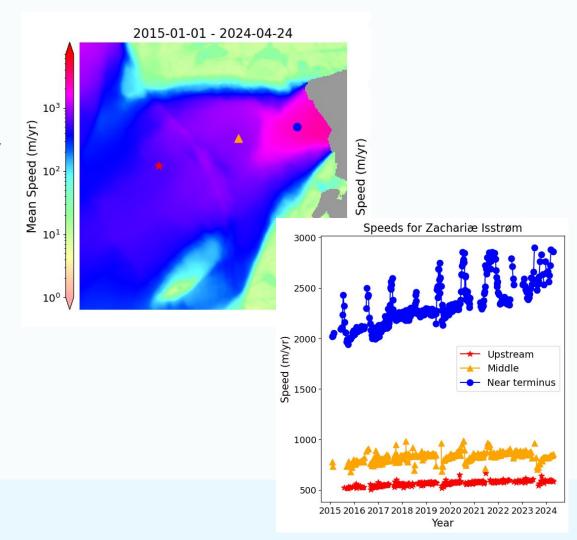
Predict velocity along a flowline

Tier 3

Predict velocity at entire glacier

#### **Data**

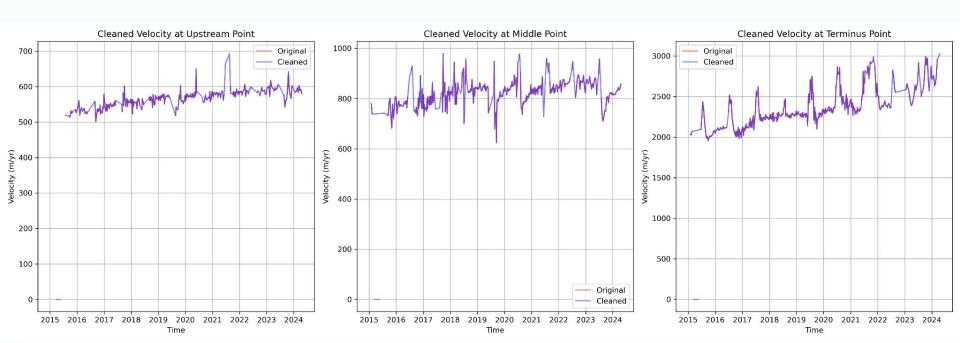
- Satellite-derived velocity estimates from GrIMP at 6 or 12 day frequency
- Picked 3 points: upstream, middle, terminus
- GrIMP promotes open access!



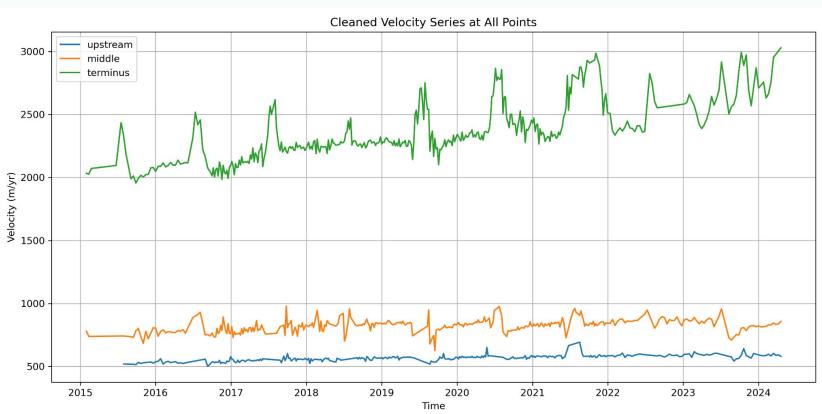
## Preprocessing

Look at the scale bars!!

NaNs and zeroes were removed from each point

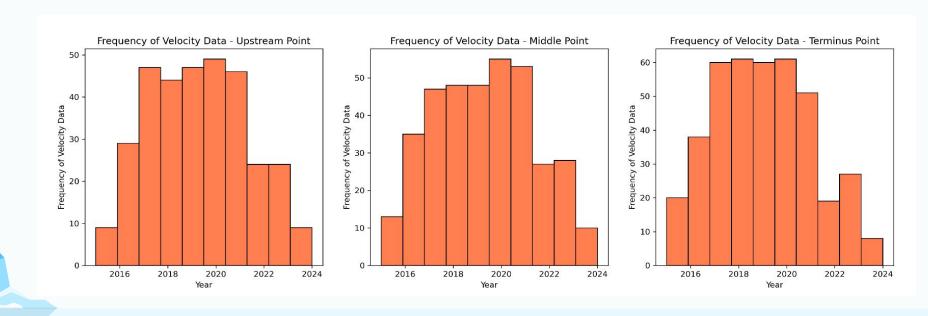


## Preprocessing



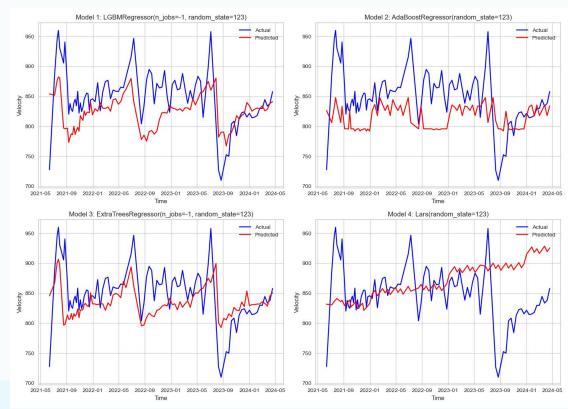
## **Data**

- Most data where Sentinel-IA and Sentinel-IB overlap (~2016-2022)



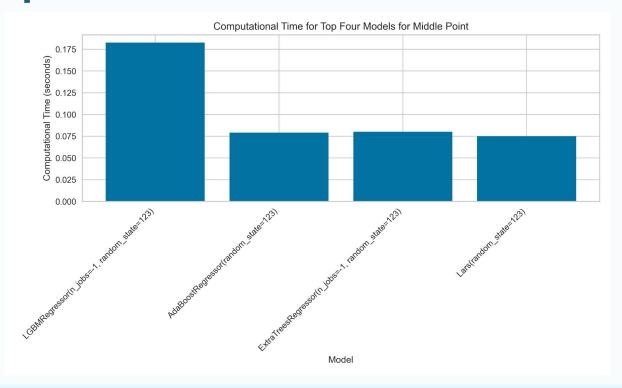
## CML - Hyperparameter Tuning

- Used PyCaret library to select optimal model for each point
- LightGBM and
  ExtraTrees capture
  seasonal trends
- Tuned and chose
  ExtraTrees (splits randomly rather than optimally)



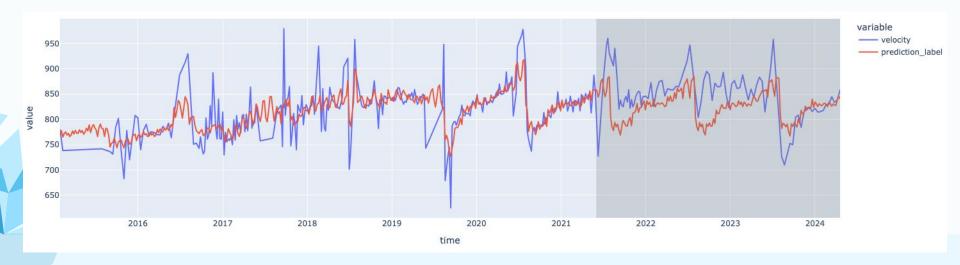
## **CML - Computational Time**

- LightGBM was slowest
- Others are relatively similar
- Not much of a concern right now...



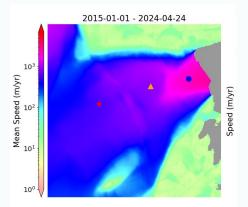
#### CML - Train/Test

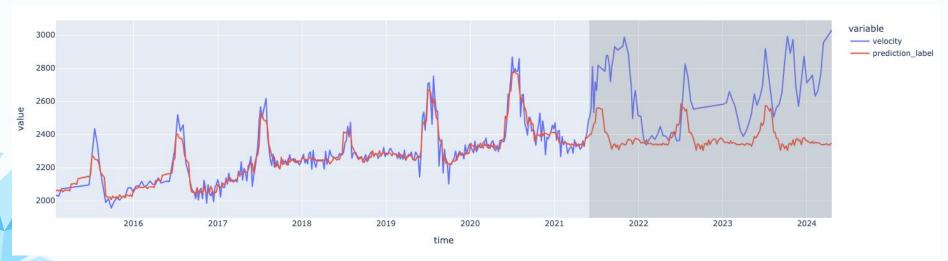
- Train (red) and test (red, highlighted gray) vs. ground truth (blue)
- Paying attention to seasonality



## CML - Train/Test (terminus)

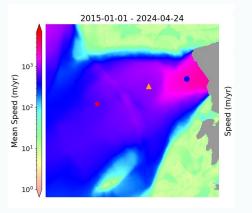
 Train (red) and test (red, highlighted gray) vs. ground truth (blue)

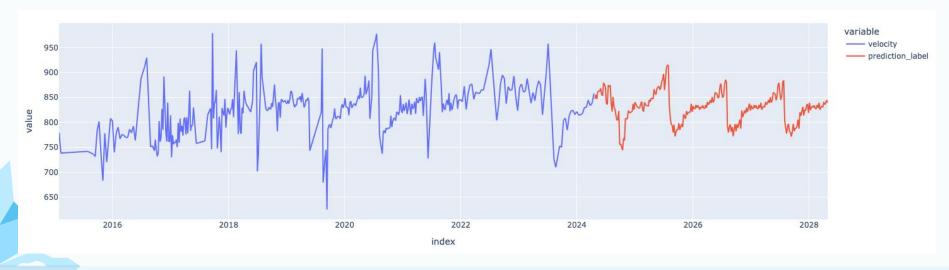




## CML - Predictions (middle)

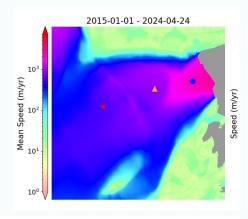
- Used ExtraTrees to predict values 2 years in the future
- Prediction (red) vs. ground truth (blue)





## CML - Predictions (terminus)

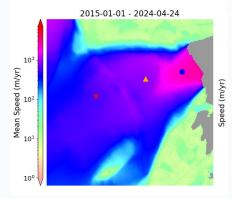
- Used **LightGBM** to predict values 2 years in the future
- Prediction (red) vs. ground truth (blue)

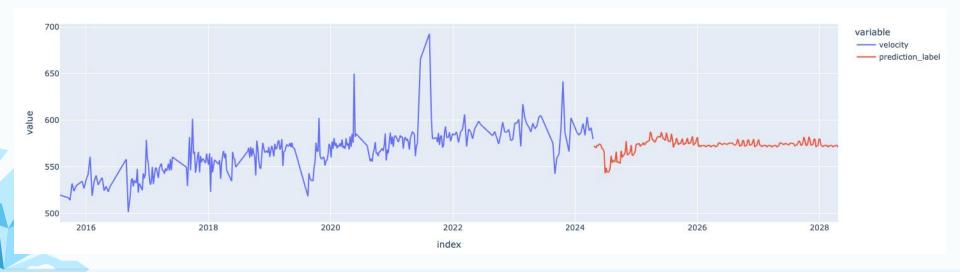




# CML - Predictions (upstream) (My) pad (upstream) (hy) pad (upstream) (upstrea

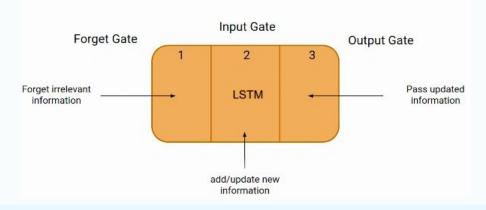
- Used RandomForest to predict values 2 years in the future
- Prediction (red) vs. ground truth (blue)





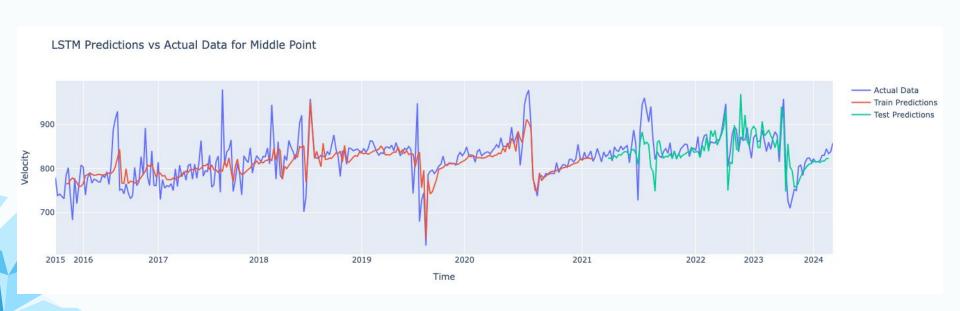
#### DL-LSTM

- LSTM uses input, forget, output gates to store memory
- Must choose optimal lookback (memory length) for predictions
- Trained 500 epochs with 0.2 dropout, lookback of 5



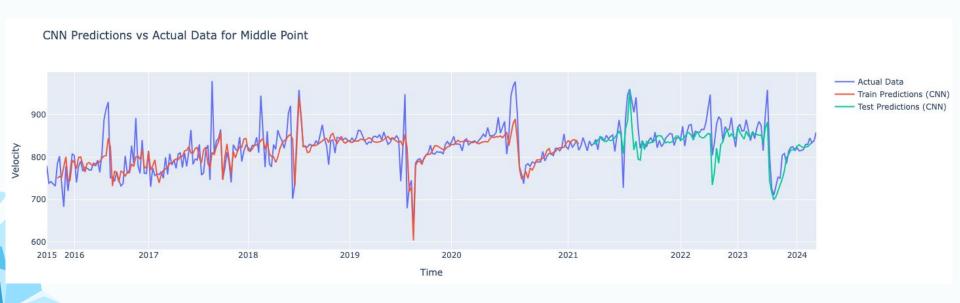
## DL-LSTM

- Trained (red) and tested (green) on ground truth (blue)



#### DL - CNN

- Trained (red) and tested (green) on ground truth (blue)

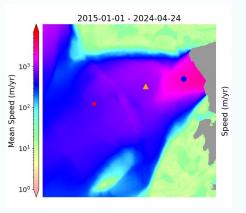


LSTM Train Score: 36.88 RMSE LSTM Test Score: 41.10 RMSE CNN Train Score: 821.66 RMSE CNN Test Score: 843.57 RMSE Horrible RMSE scores but do predictions capture seasonality?

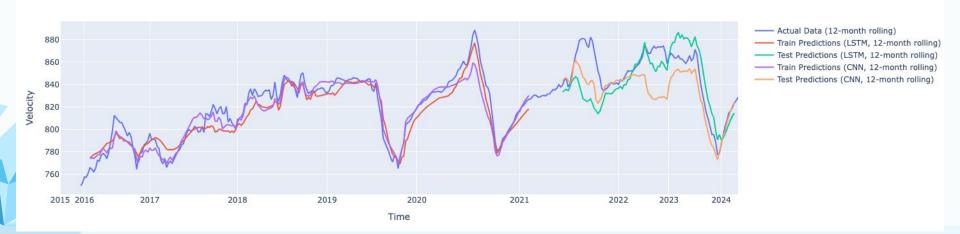
LSTM vs CNN Predictions vs Actual Data for Middle Point



LSTM Train Score: 36.88 RMSE LSTM Test Score: 41.10 RMSE CNN Train Score: 821.66 RMSE CNN Test Score: 843.57 RMSE

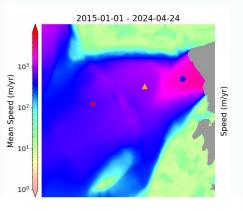


12-Month Rolling Average of Predictions vs Actual Data for Middle Point

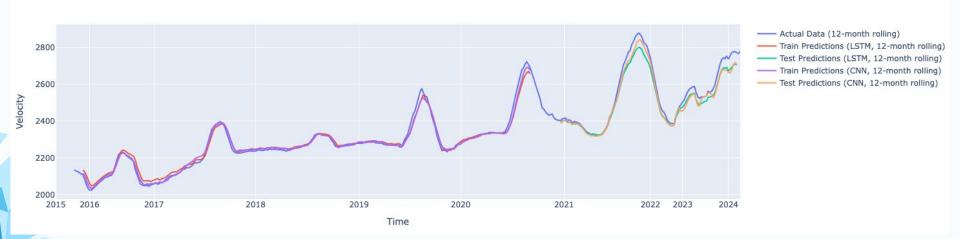


For the terminus point, yes!

... but is it overfitting??

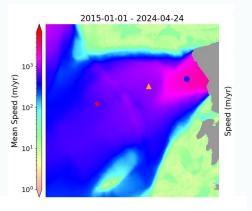


12-Month Rolling Average of Predictions vs Actual Data for Terminus Point

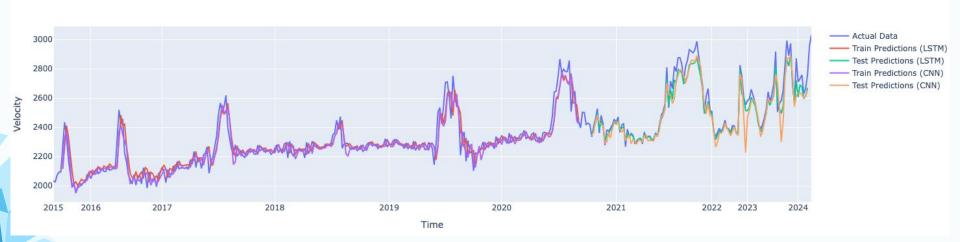


For the terminus point, yes!

... but is it overfitting??



LSTM vs CNN Predictions vs Actual Data for Terminus Point



## Scaling Up

# **Exogenous** Variables

Incorporate temperature, meltwater runoff, etc.

#### Frequency

Test monthly predictions

#### **Spatial Coverage**

Incorporate many glaciers or many points along flowline

If different models perform better on different parts of the glacier, can one accurately describe the entire glacier??

Python library to automate point extraction, training, model dev?

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