



FEMA

Predicting the “Unpredictable”

ACSE/EDSML MINI-PROJECT

- Team Barry
- Last updated: 2025.01.31

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Who are we?

Team Barry

- Founded at Imperial College London in the Ada Lovelace Academy



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Today's Discussion

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Problem

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The Problem:

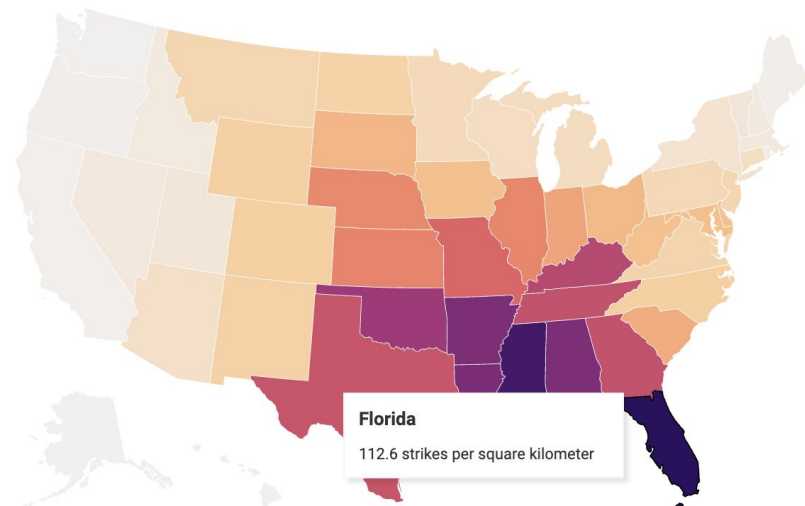
⚡ Lightning ⚡ *Where will it strike next?*

- Economic, natural, and human impact
- Increased frequency

Lightning strike density by state in 2023

While Texas had more total lightning strikes in 2023, Florida consistently leads the nation in lightning strike density, measured in strikes per square kilometer.

Strikes per square kilometer





Our Solution:

flash.io

- Harnessing ML to Predict Storms & Lightning Strikes
 - Predicting radar images 1-hour into the future
 - Predicting radar images from other satellite images in the same time period
 - Predicting Lightning Strikes



Feature 1

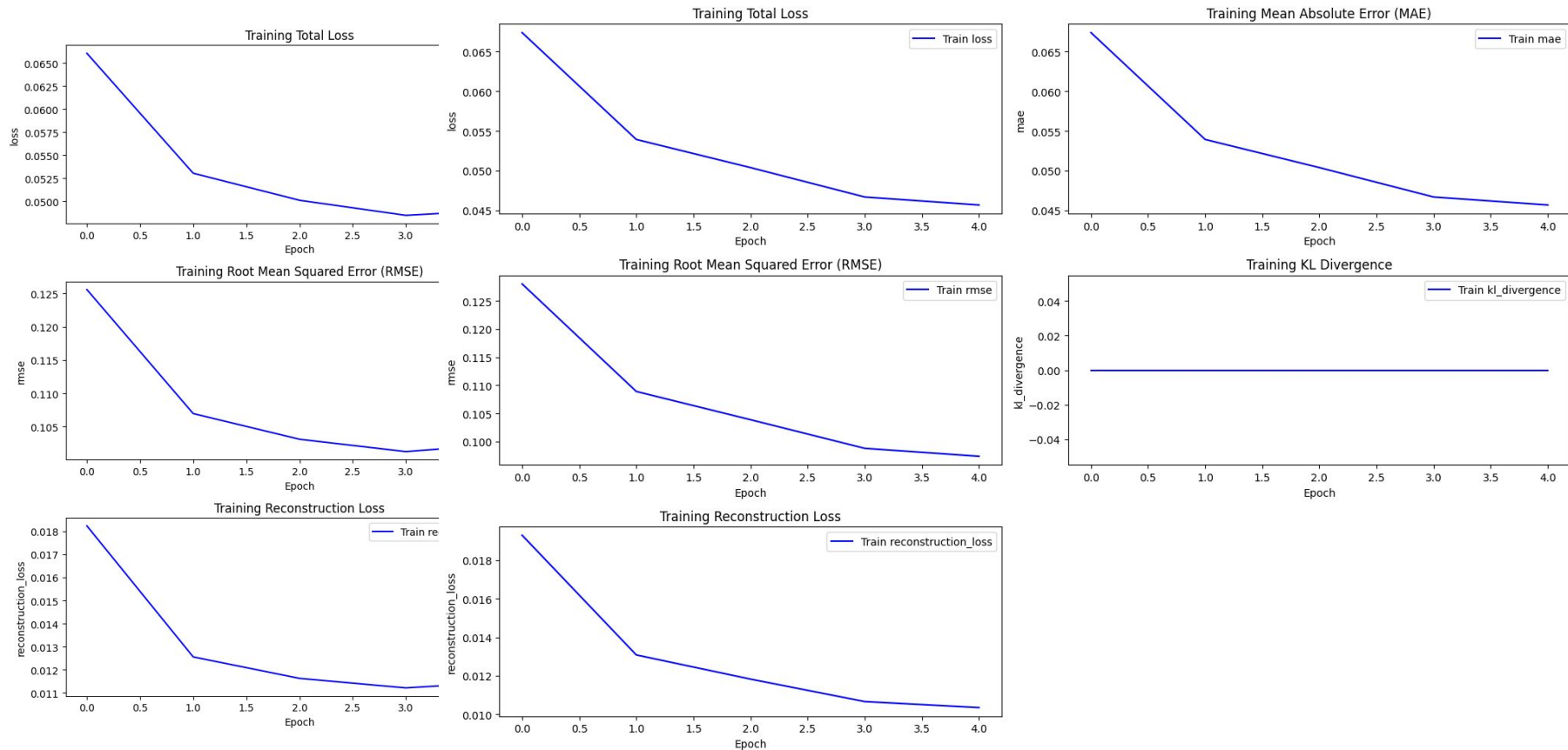
Goal: Given one hour, predicting the next hour
of vertically integrated liquid (radar)

Model Architecture and Training Workflow



- **ConvLSTM2D**
- Input layer (height, width, time steps, channels)
- **3 ConvLSTM2D** Layers
- Output Layer: **Conv3D**
- For compare, we use the **same model** in A&B
- **7 slide windows**
- **800 * 7** data per epoch
- Train: 70%
- Test: 10%
- Validation: 20%
- Optimizer: adam
- Loss : mae
- Saved as **pretrained model**
- Augmentation
- Normalization
- Predicts **12 future vil frames**
- Metrics: total_loss, mae, rmse, kl_divergence, reconstruction_loss
- **used for** storm tracking, weather forecasting
- **Visualization**: the input, predicted, and actual frames

Model Training Monitoring

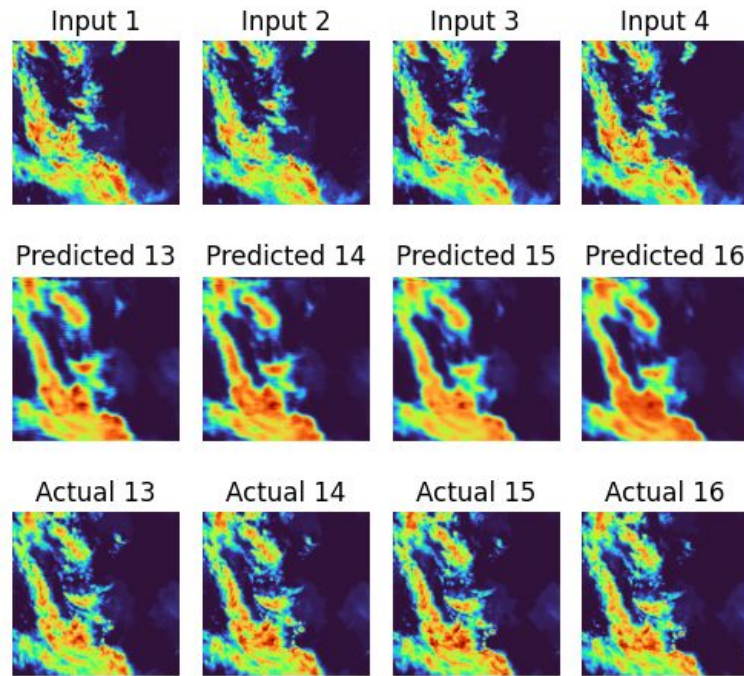
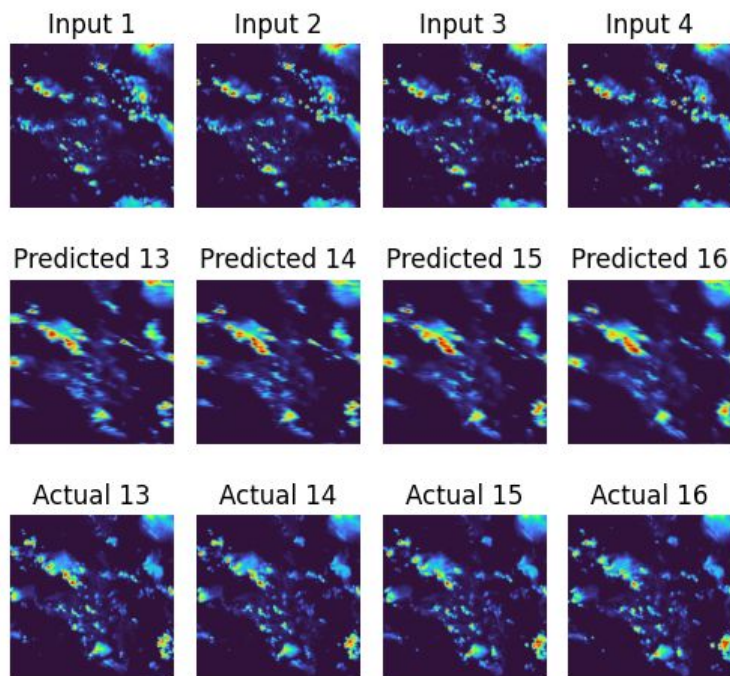


Model Evaluation

Index	Model A	Model B
RMSE	0.09183	0.09582
MAE	0.04197	0.04592
KL	0.06719	0.06502
Reconstruction Loss	0.00929	0.01021
Total Loss	0.07648	0.07522

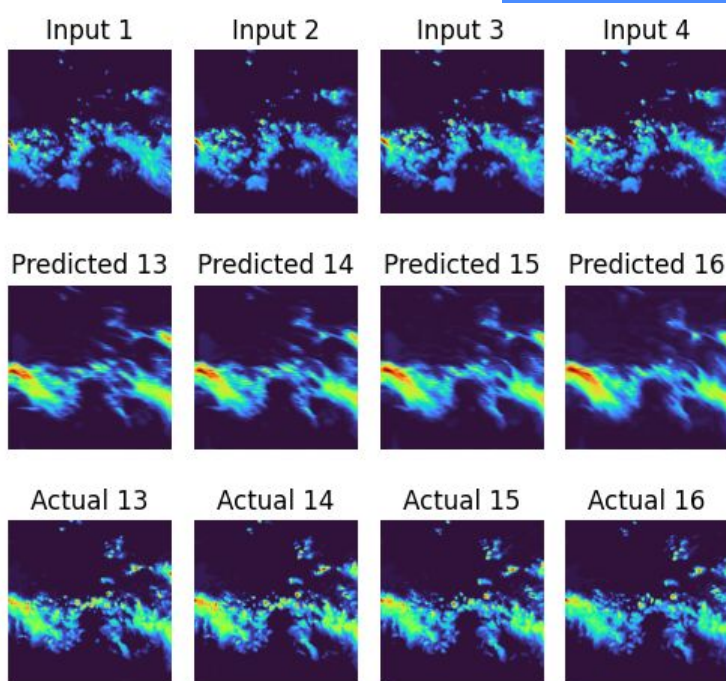
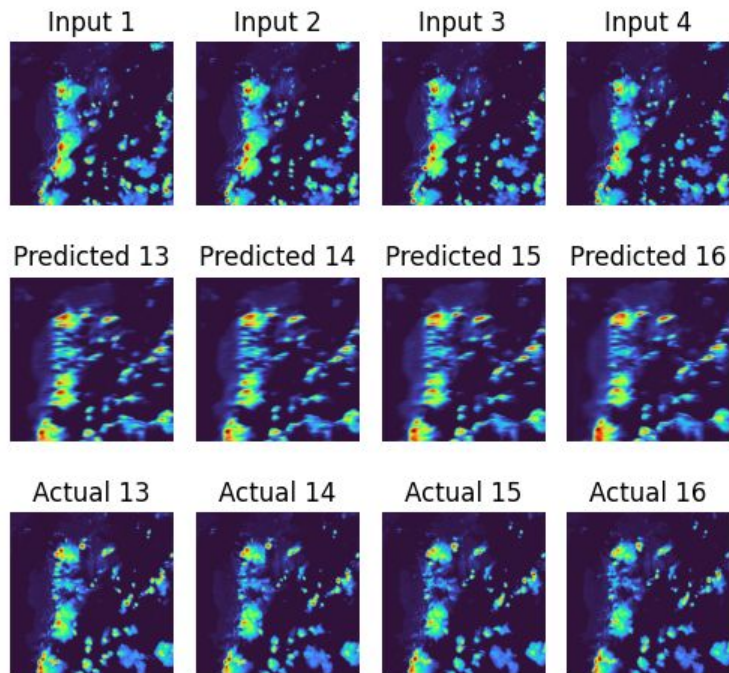
B is more concerned with the overall structure.

Model A output - Training



Good ! But...

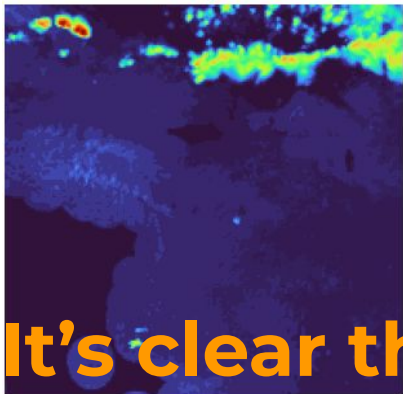
Model B output - Training



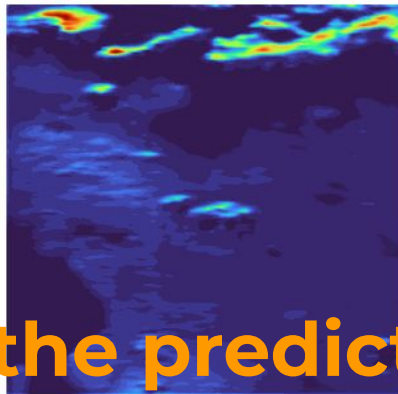
More reasonable !!

Model output - surprise storms

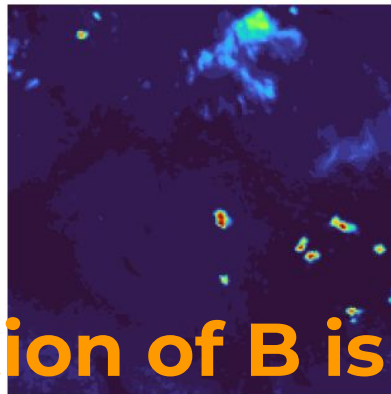
S844398 - Input Frame 1



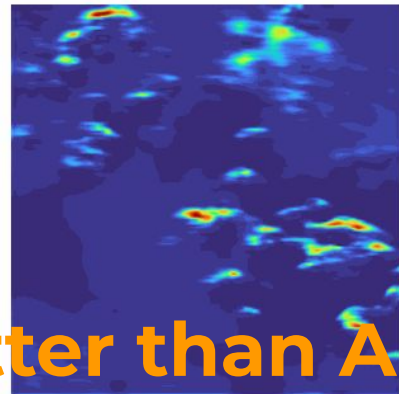
S844398 - Task 1A Prediction Frame 1



S837416 - Input Frame 1

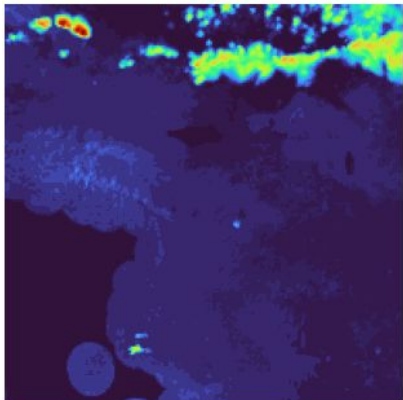


S837416 - Task 1A Prediction Frame 1

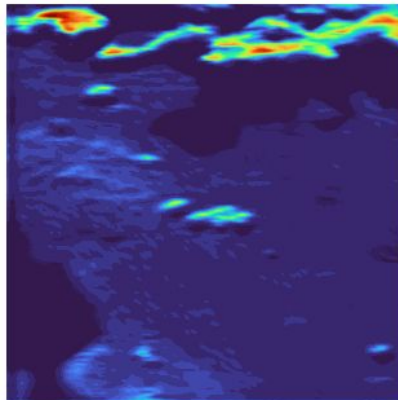


It's clear that the prediction of B is better than A

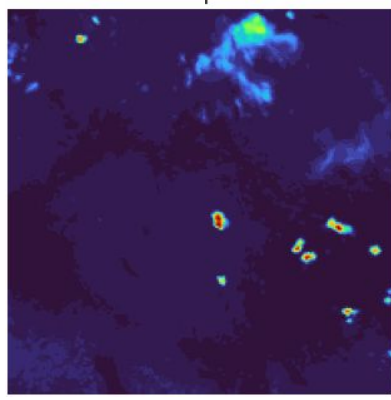
S844398 - Input Frame 1



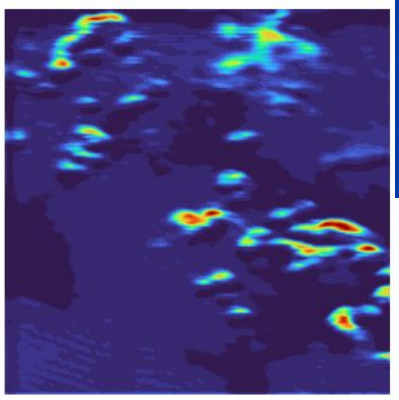
S844398 - Task 1B Prediction Frame 1



S837416 - Input Frame 1



S837416 - Task 1B Prediction Frame 1

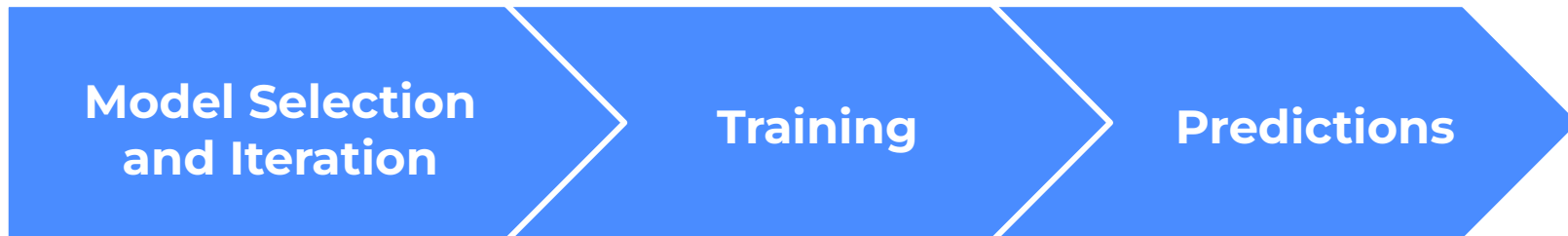




Feature 2

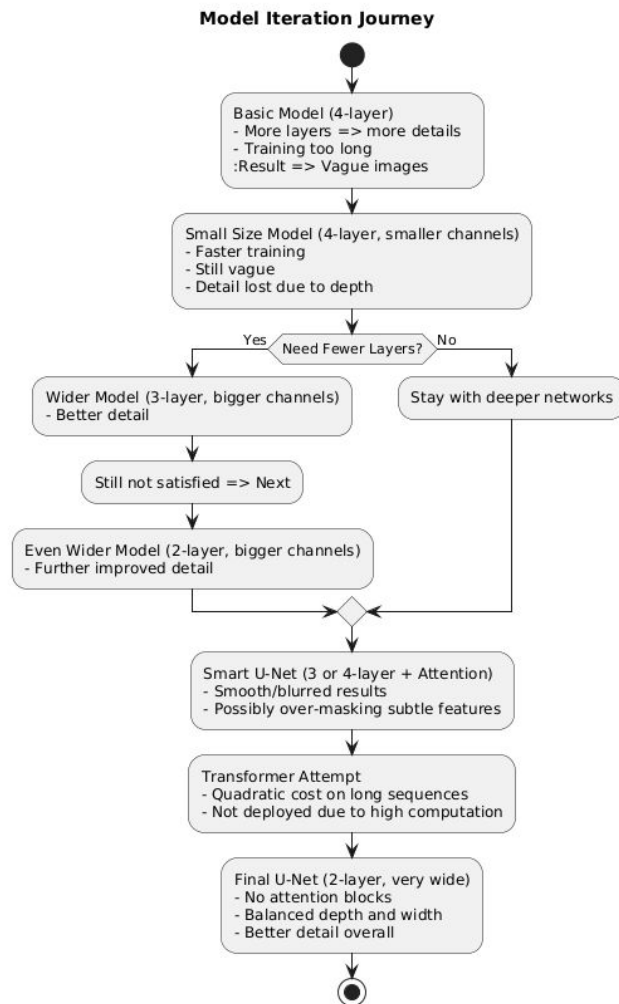
Goal: predicting vertically integrated liquid
(radar) images when missing

Iteratively develop our task

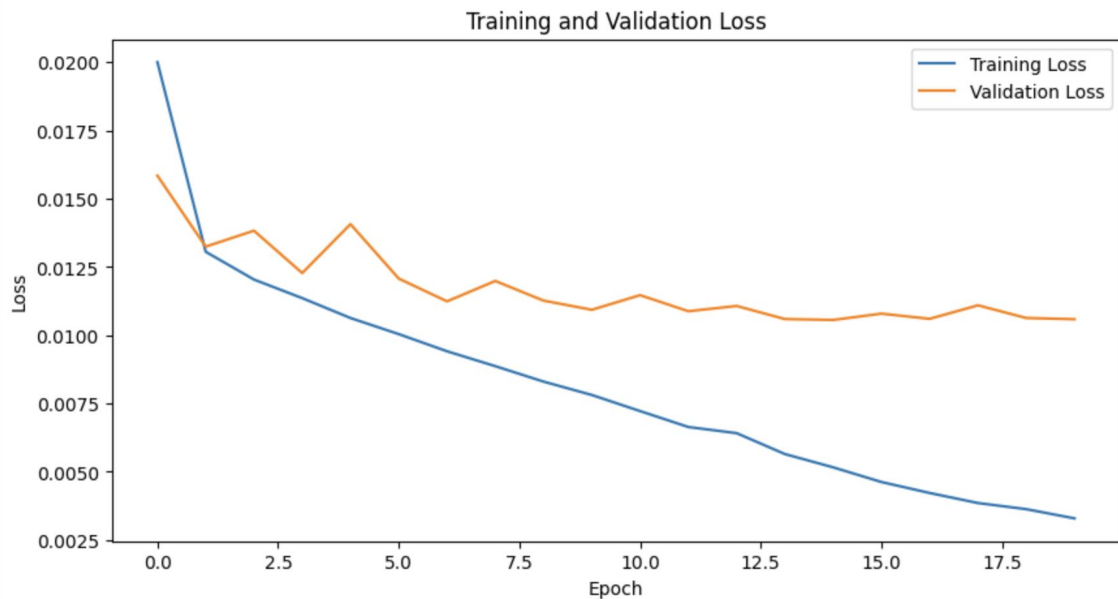


- Determine the main objective
 - Design model that solve the task well
 - Iteratively update the structure of the model
- Training the model based on different loss, parameter and epoch.
 - Iteratively feedback to the model selection and design
- Using the trained model to predict the data.

Model Selection and iteration

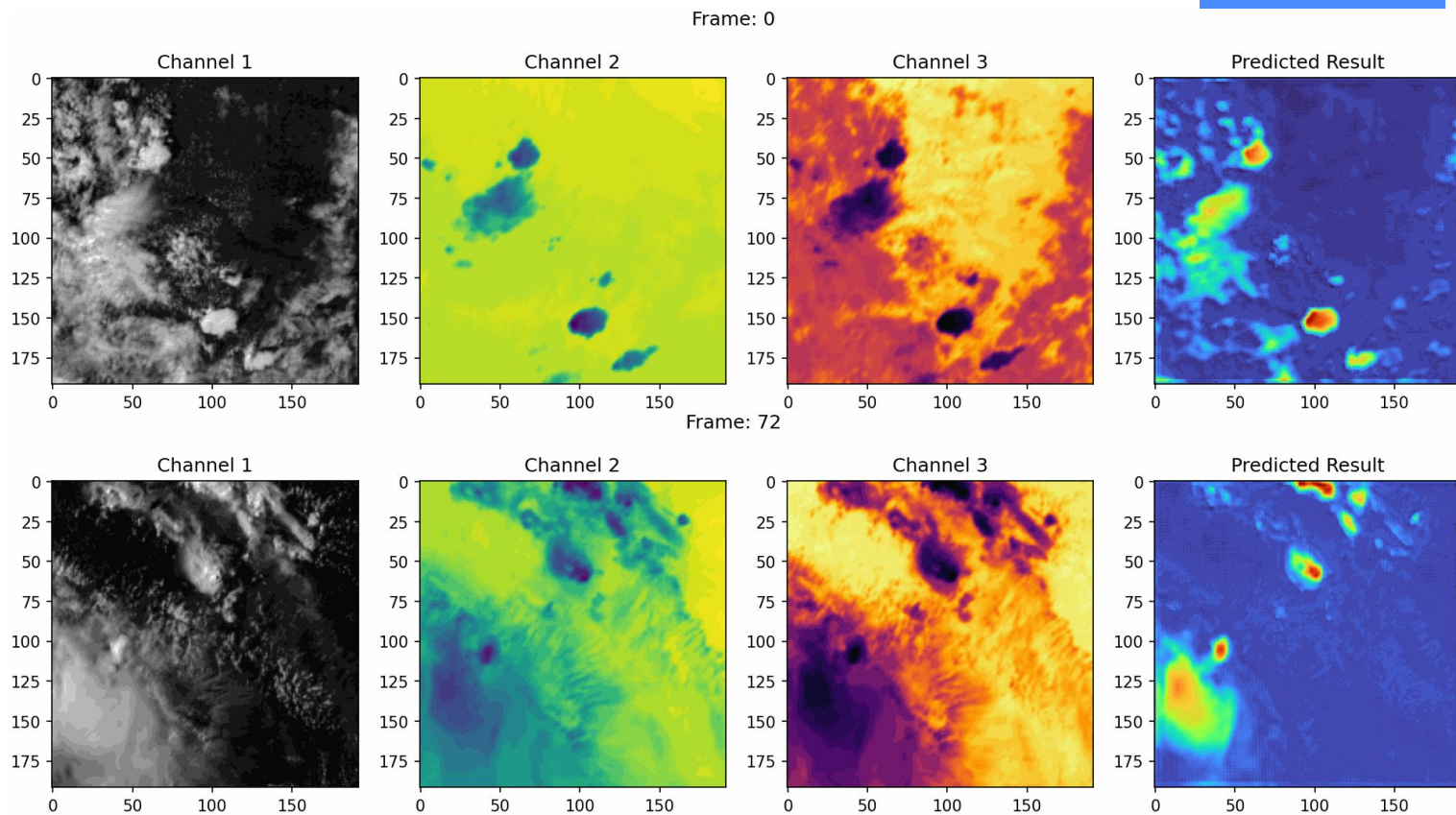


Model Training



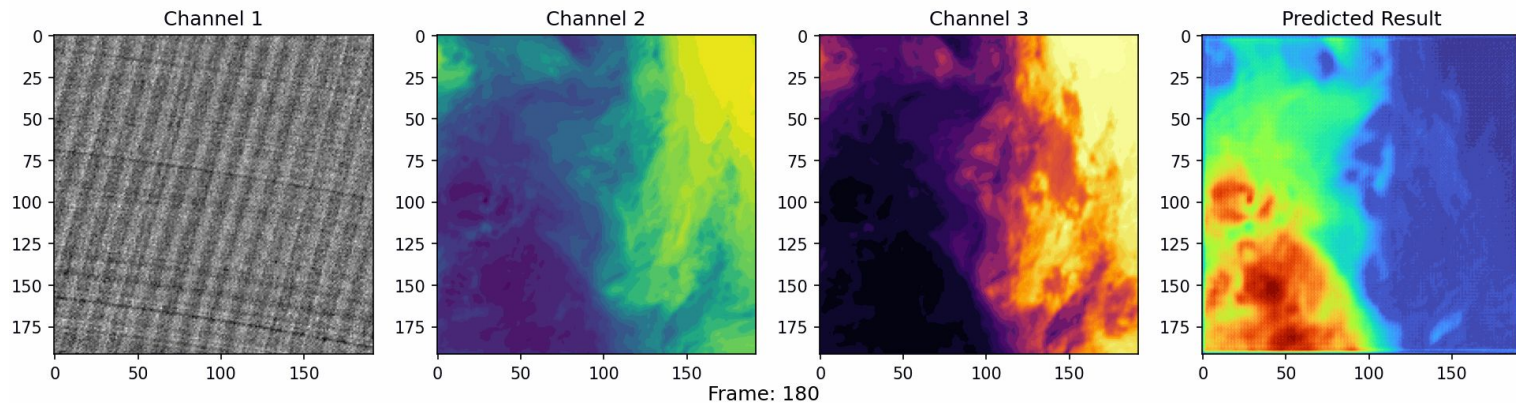
Epochs = 30
Loss: MSE
Parameters:
Optimised: Adam

Model output

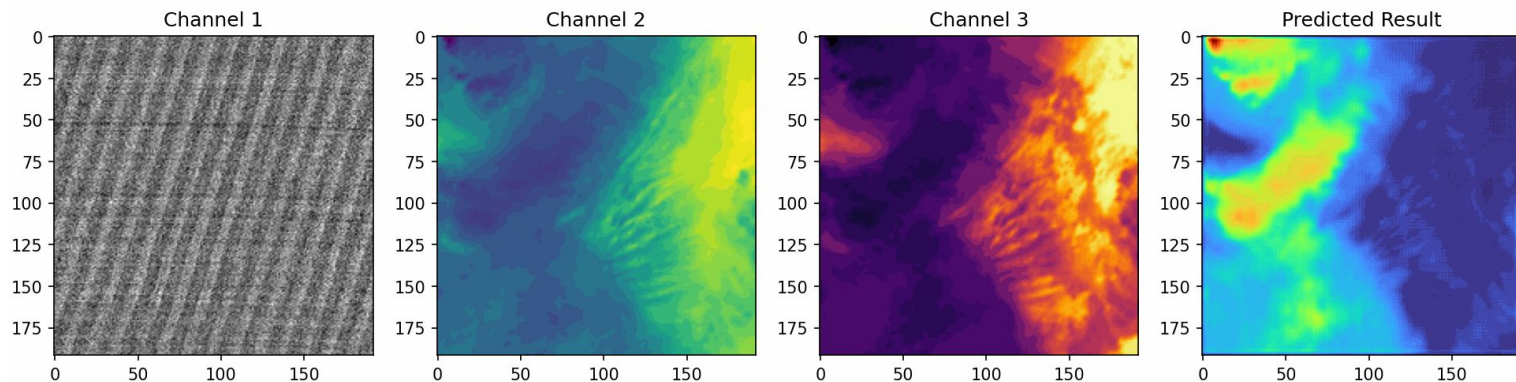


Model output

Frame: 144



Frame: 180

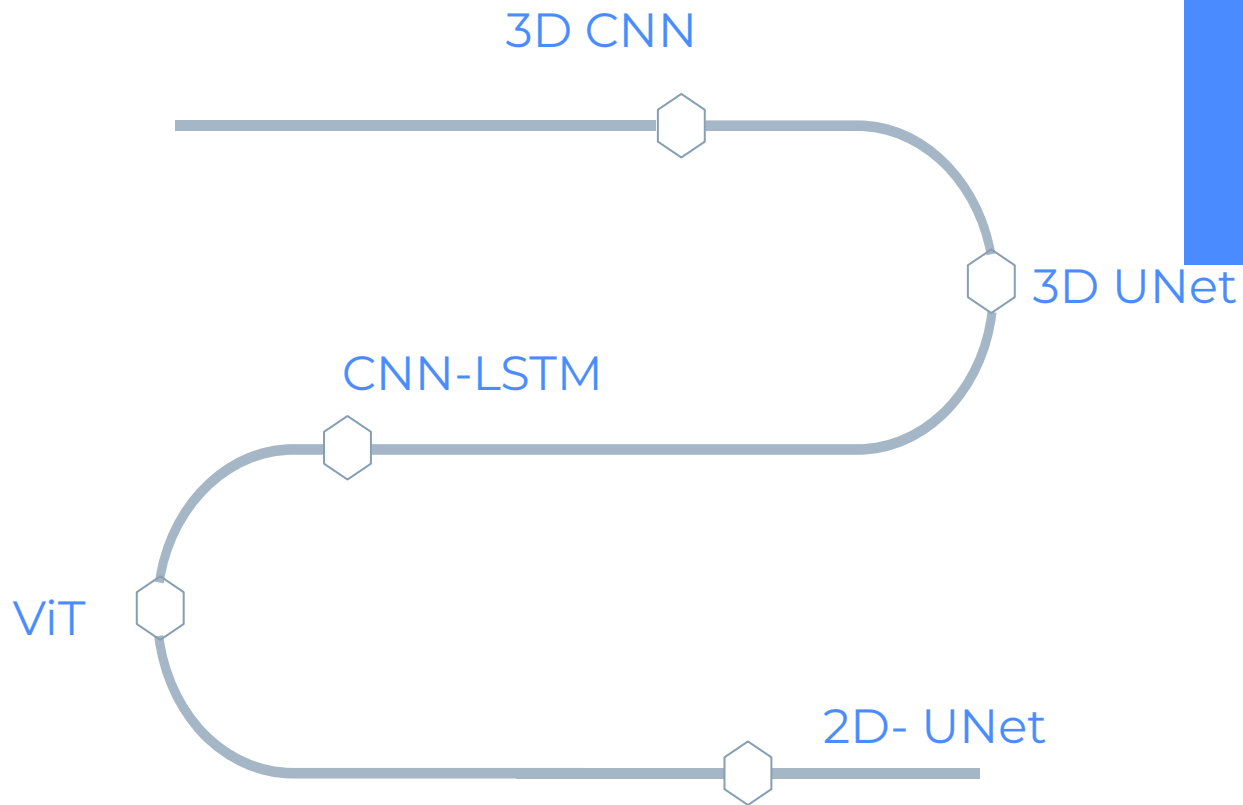




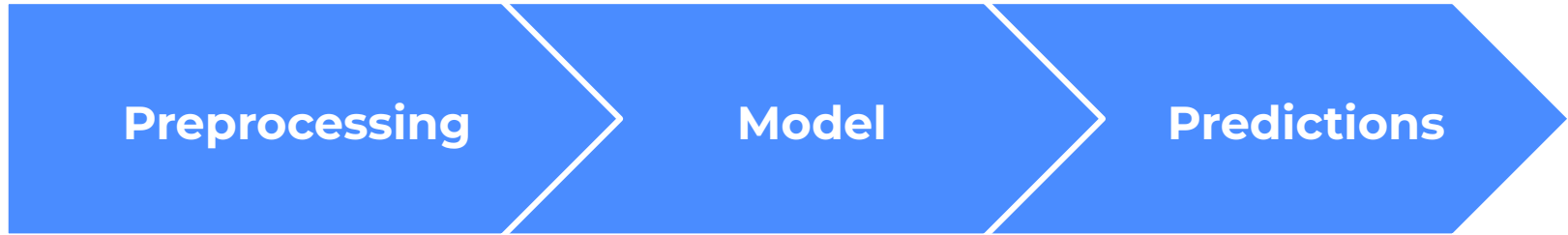
Feature 3

Given a set of 4 different bands of satellite images, predict lightning locations.

Model Selection and iteration



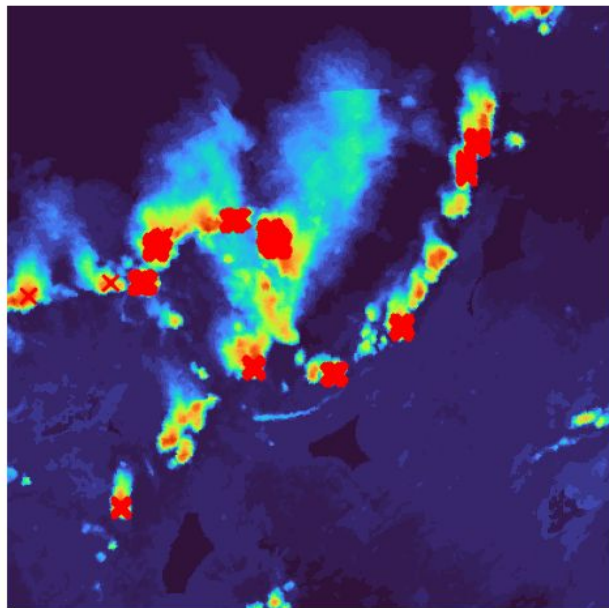
Data Gathering Survey



- Normalize images using local Min Max
 - Upsized the IR images
 - Grouped lightnings in frames
- UNet Architecture
 - Skip Connections
 - Dice Loss
 - Relu
- Outputs x,y grid of pixels locations.
 - One specific frame
 - Time of lightning in 5 min frames

Predicted Outputs

Frame 0



Lightning events predictions of Storm Surprise over the 36 frames of VIL band.

- Train Loss: 0.9968
- Val Loss: 0.9989

Potential Concerns

1. Data Quality and Availability
2. Real Time Performance
3. Integration with Existing Systems
4. Dependency on Weather Data Sources



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References

- <https://www.fastcompany.com/91041966/us-yearly-lightning-ground-strikes-damage-prevention>
- <https://www.livescience.com/planet-earth/weather/electrifying-time-lapse-image-captures-100-lightning-bolts-torching-the-sky-over-turkey>
- GITHUB:
<https://github.com/ese-ada-lovelace-2024/acds-storm-prediction-barry>



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Conclusion

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Thanks!

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