

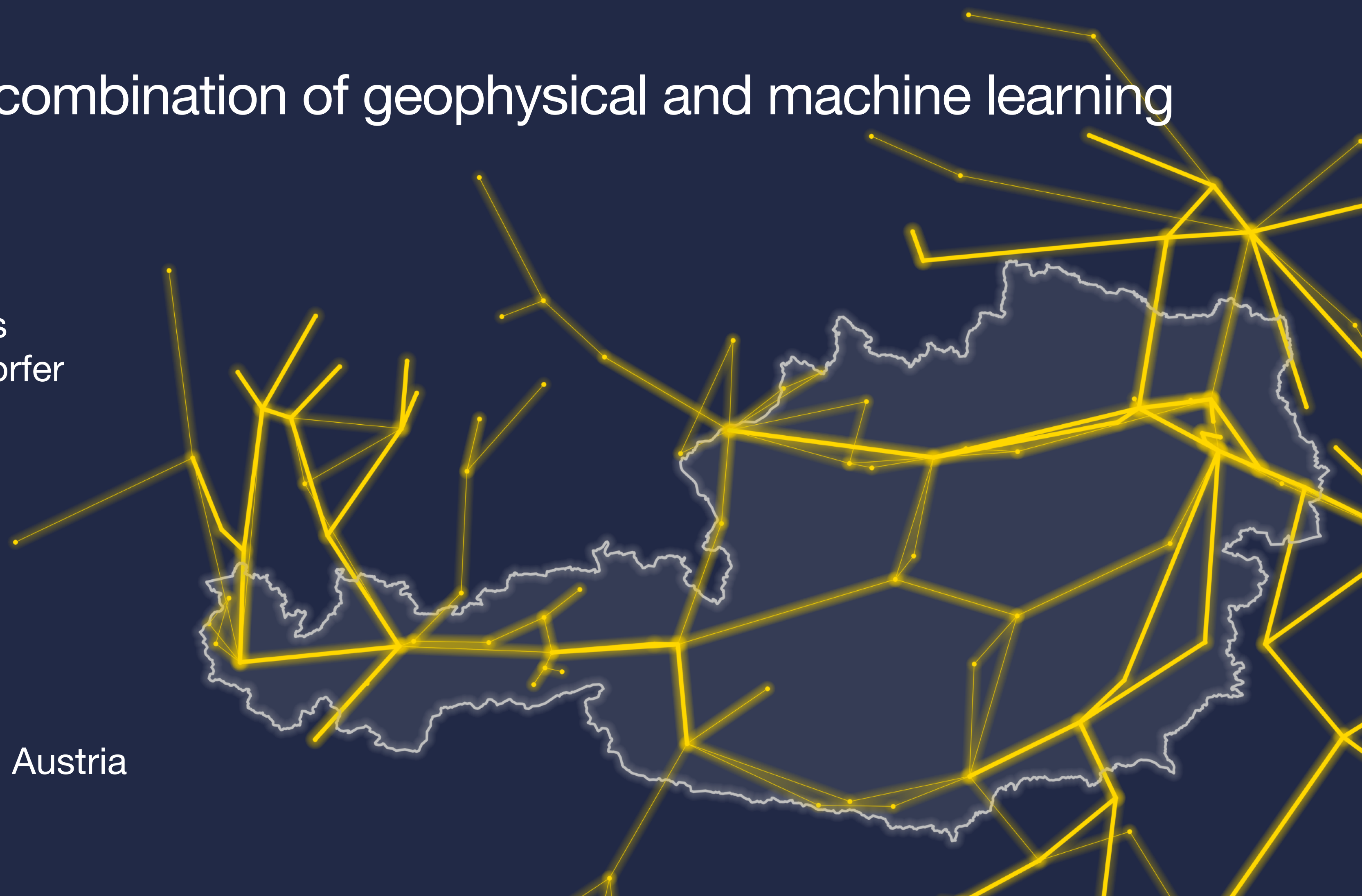
# Building a GLC forecasting tool based on geomagnetic and solar wind data: challenges and future avenues

A study in Austria using a combination of geophysical and machine learning methods

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# GICs are a modern-day phenomenon

March 13, 1989

March 9th: coronal mass ejection leaves  
Sun

March 12th: mild geomagnetic  
disturbances

March 13th:

01:00: Hydro-Quebec power grid  
operation problems

02:44: geomagnetic storm starts

02:45: sequence of HV transformers and lines trip,  
within 60s collapse of entire network,  
**power blackout**

...

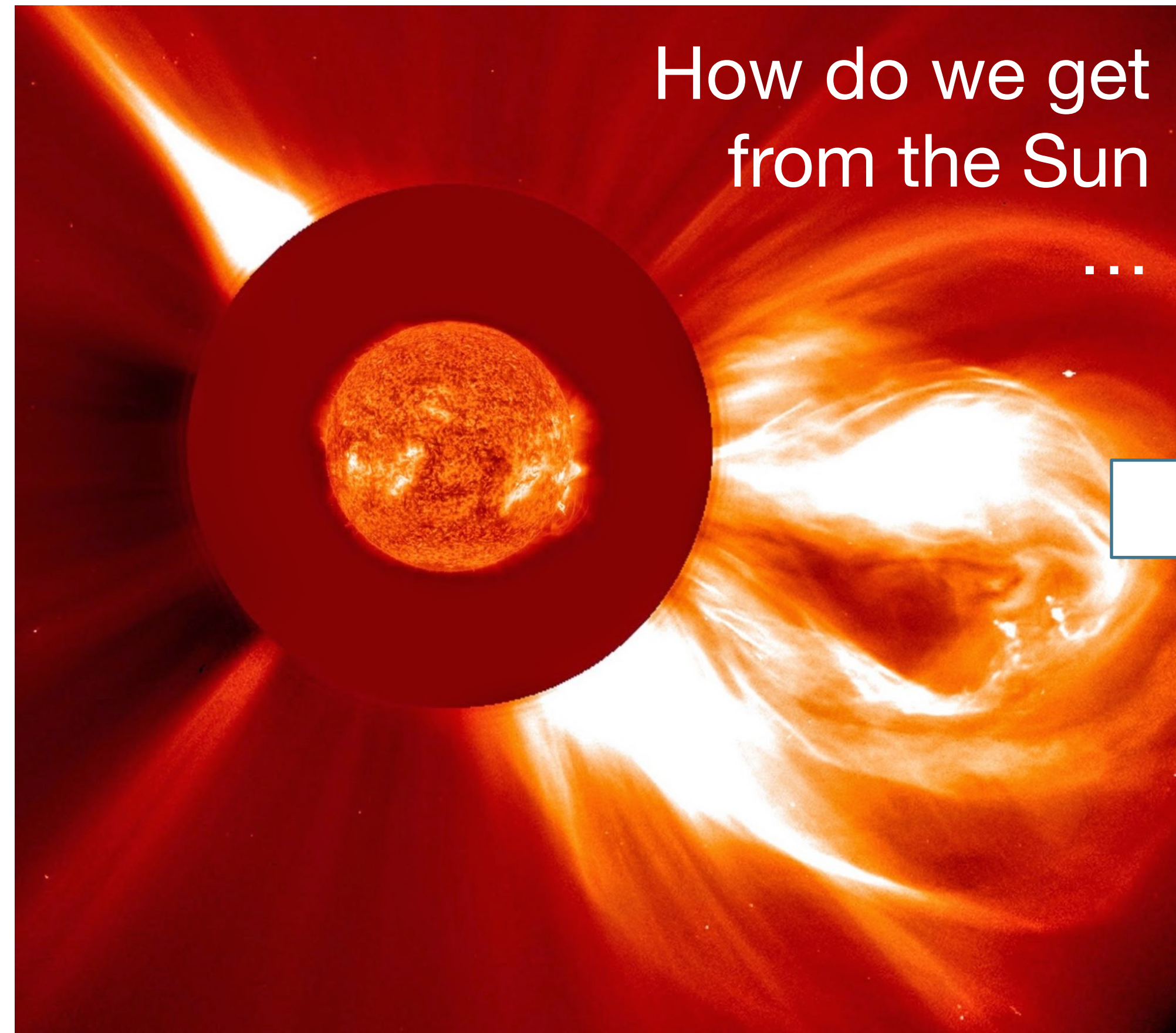
12:00: 83% of power restored



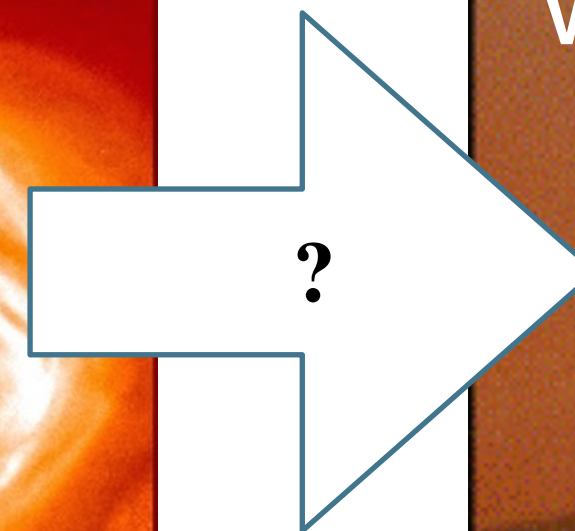
Credit: NASA



# GICs are a modern-day phenomenon



Credit: SOHO, 2003-12-02



Credit: PSE&G



# Rapid geomagnetic variations cause GICs

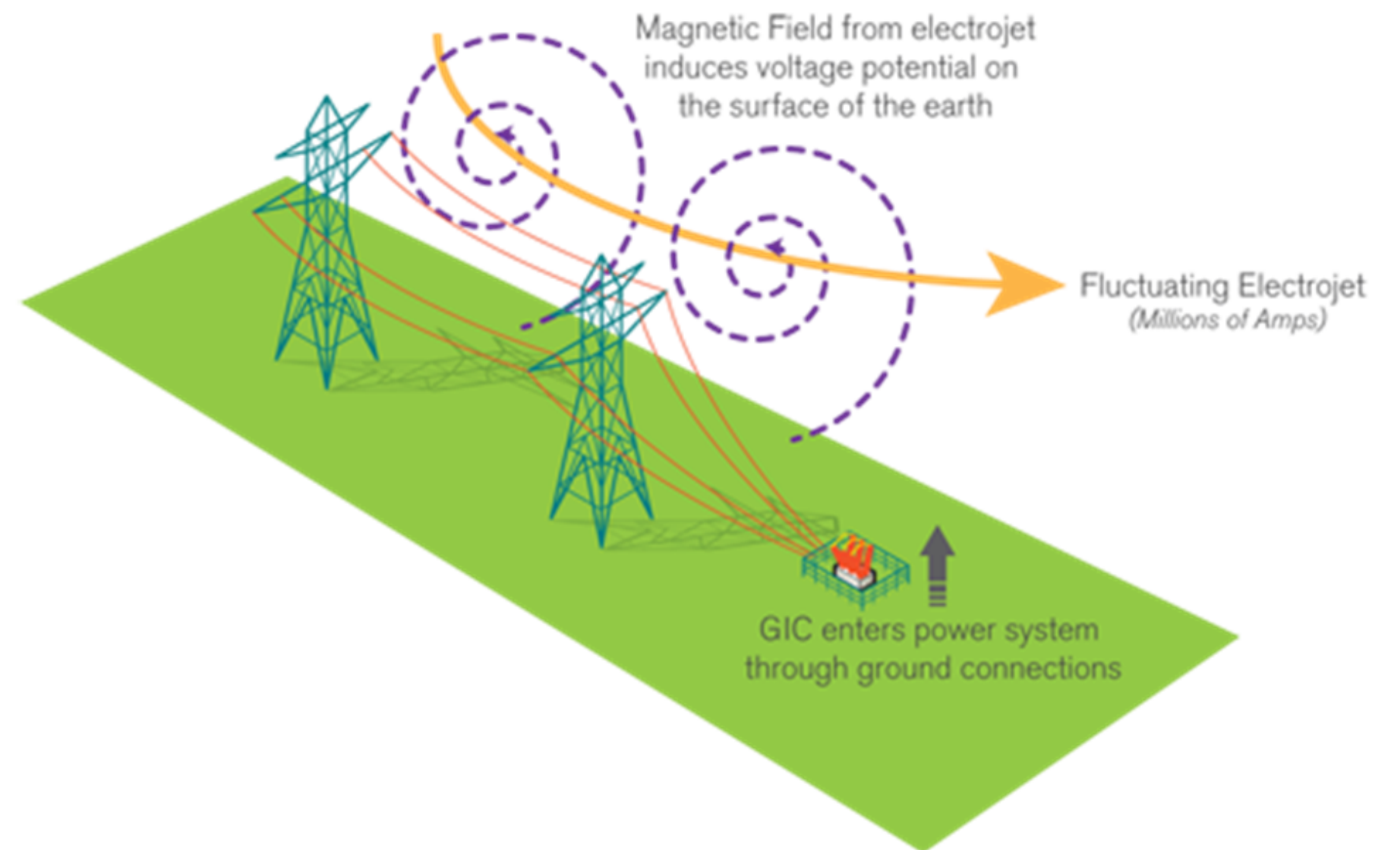
Geomagnetic storms lead to **rapid magnetic field variations**

Varying magnetic field → induced electric field

$$\frac{E_x}{B_y} = Z_{xy} = \sqrt{i\rho\omega\mu}$$

( $\rho$  = ground resistivity,  $\omega$  = angular freq.,  $\mu$  = magnetic permeability)

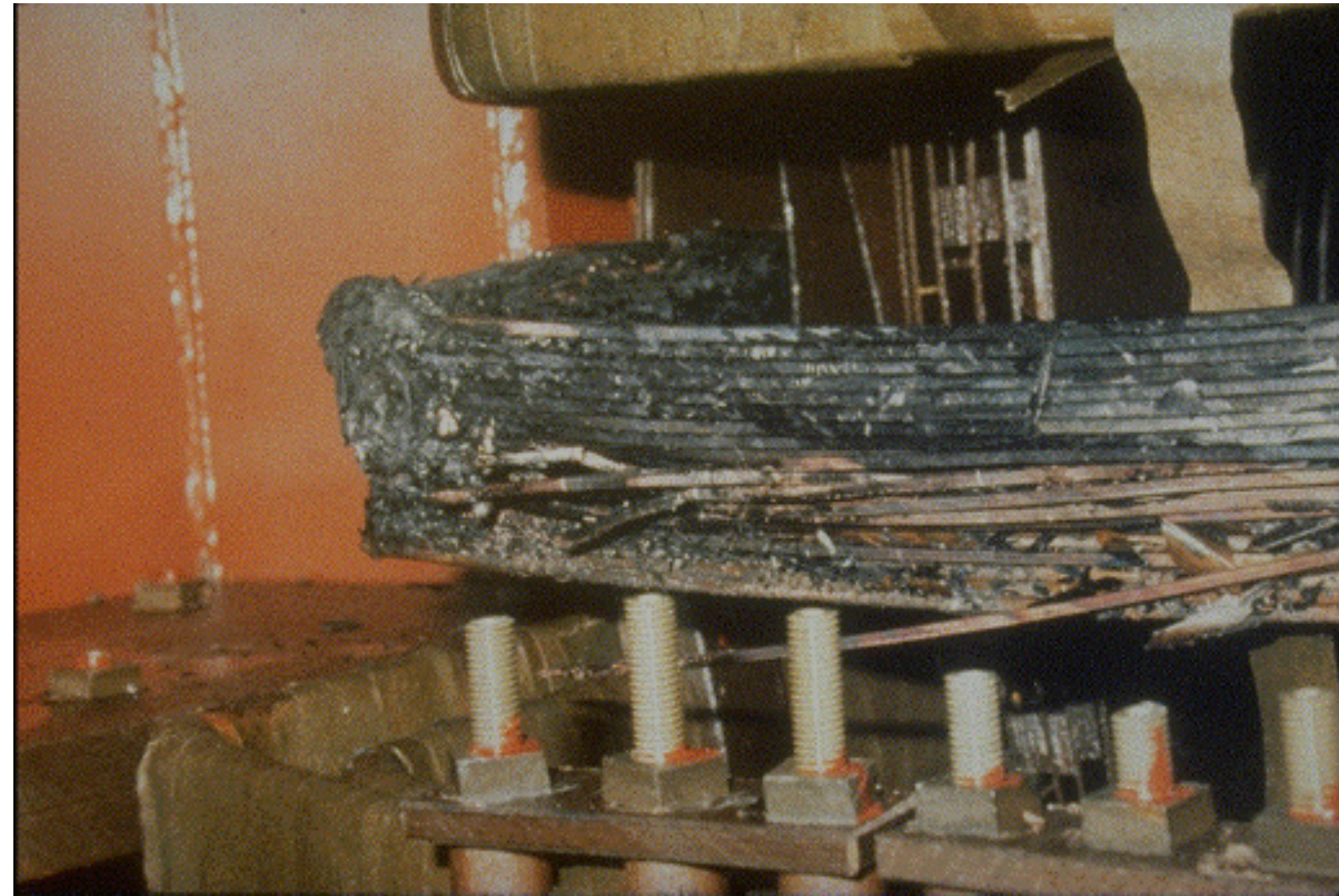
**Geoelectric field** leads to electro-motive force in grounded conductive network  
→ currents through lines / transformers





# GICs flow through transformers

- GICs **flow from Earth to power grid** and vice versa —> DC is flowing through transformers
- Transformers are not built to handle large amounts of DC
- GICs can be **damaging** when flowing through transformers
- Large GICs cause overheating —> burnout (e.g. Quebec)



Credit: PSE&G



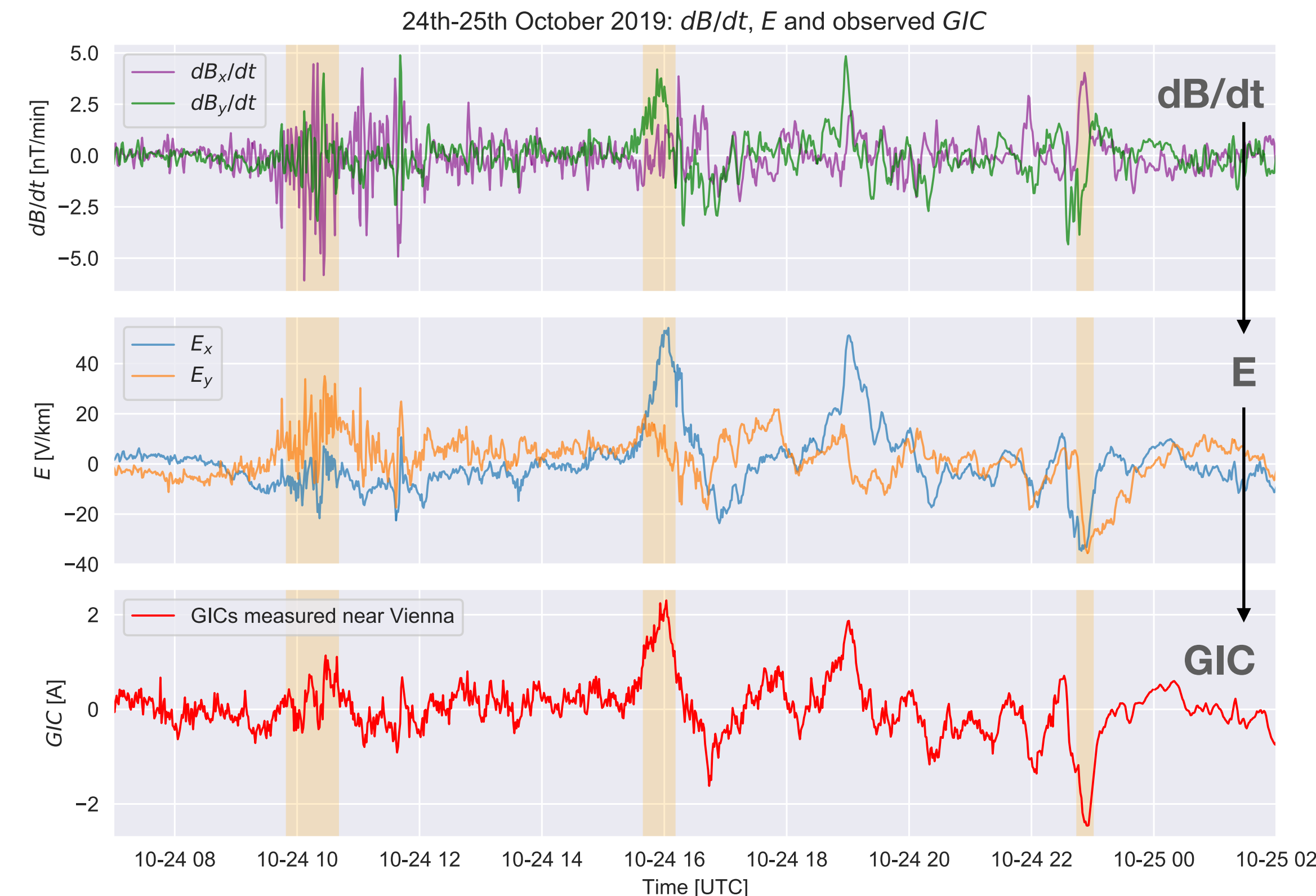
# Forecasting GICs

## The history

- Solar wind plasma and B-field data measured at L1 gives advance warning (30-60 minutes) of incoming geoeffective events.
- Many have tackled this problem before using solar wind data to **forecast  $dB/dt$  as a proxy**<sup>1</sup>.
- **$dB/dt$  is not ideal** - it does not contain the same frequency content as the geoelectric field<sup>2</sup>, the primary driver behind GICs.
- For a local grid forecast, the geoelectric field or the GICs themselves will be more accurate.

<sup>1)</sup>  
[Wintoft+\(2005\)](#)  
[Lotz+\(2015\)](#)  
[Wintoft+\(2015\)](#)  
[Camporeale+\(2019\)](#)  
[Keese+ \(2020\)](#)

<sup>2)</sup>  
[Pulkkinen+\(2006\)](#)  
[Pulkkinen+\(2013\)](#)

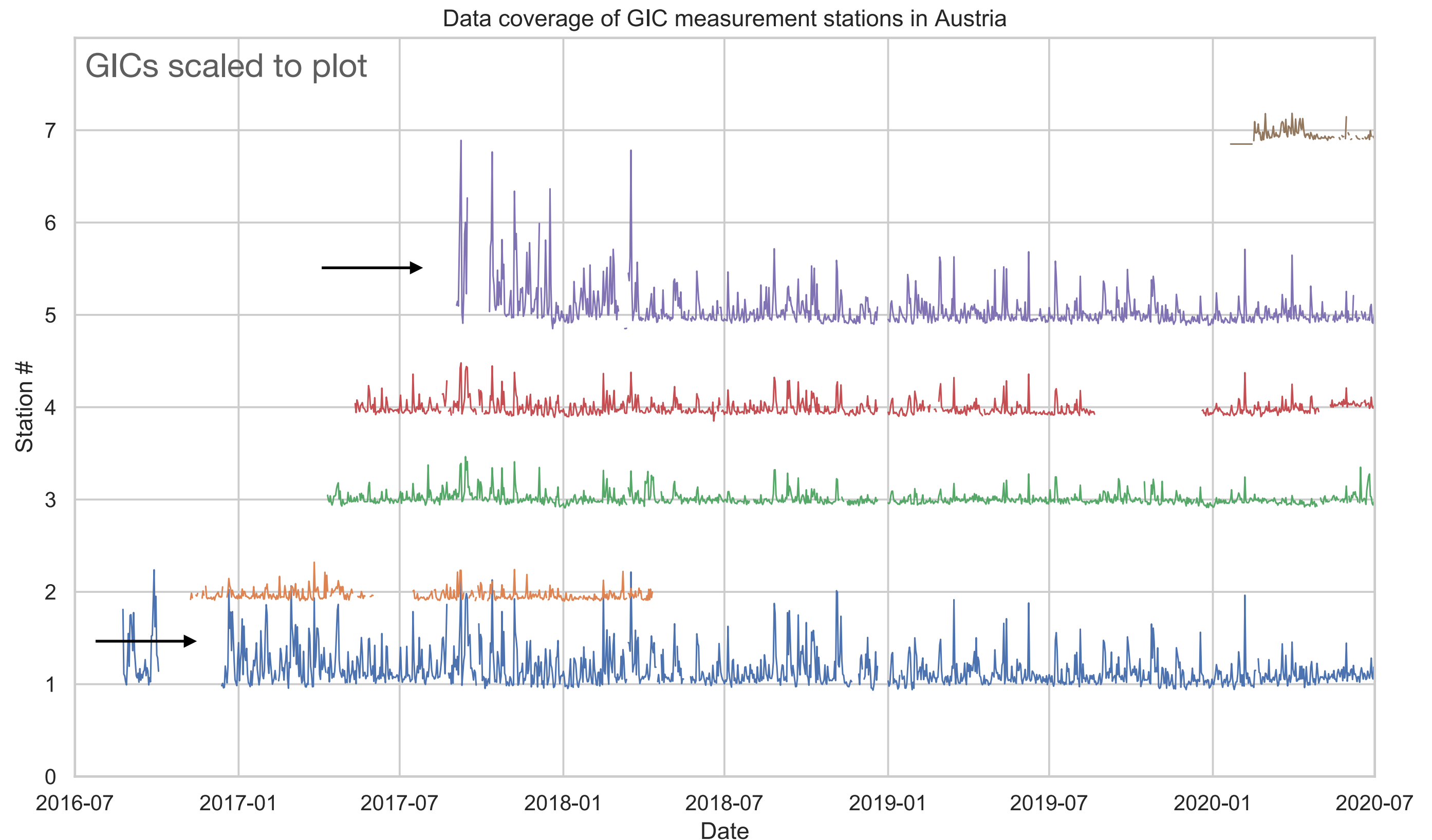


An example from 24th October 2019 of  $dB/dt$  (top), modelled geoelectric field  $E$  (middle) and measured GICs (bottom). The areas where  $dB/dt$  would either over- or underestimate the risk of high GICs are highlighted in orange.

# GIC Measurements in Austria

## An overview

- Measurements since Aug. 2016 with some small breaks at individual stations
- Up to **five stations** running in parallel
- Two stations (#s **1** and **5**) experience larger ( $> 10$  A) GICs during geomagnetic storms —> Can we make **forecasts of max(GIC)** for these stations?



# Which variable do we want to forecast?

GIC from raw station data?

... OR ...

Local modelled geoelectric field

$E_x$  &  $E_y$ ?

- + Real measurements (~4 years) define the ground truth
- No extreme events and few events with large values
- Very noisy data and often a cap on max. value

- No measurements as ground truth
- + Can model extreme events with good accuracy
- + LOTS of magnetic field data (>20 years) for modelling

$$GIC_j = a_j \cdot E_x + b_j \cdot E_y$$



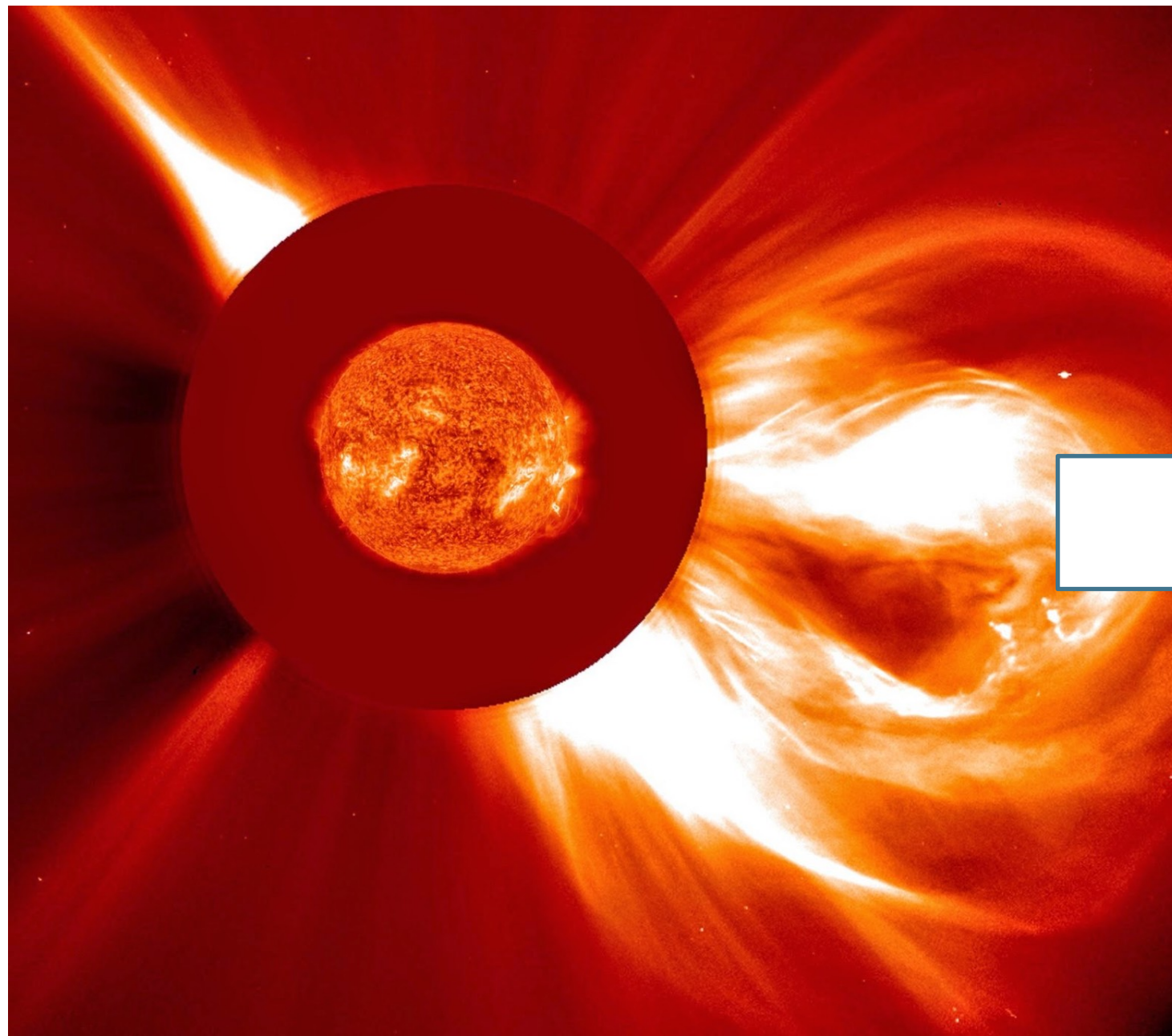
So... try both, compare and optimise!



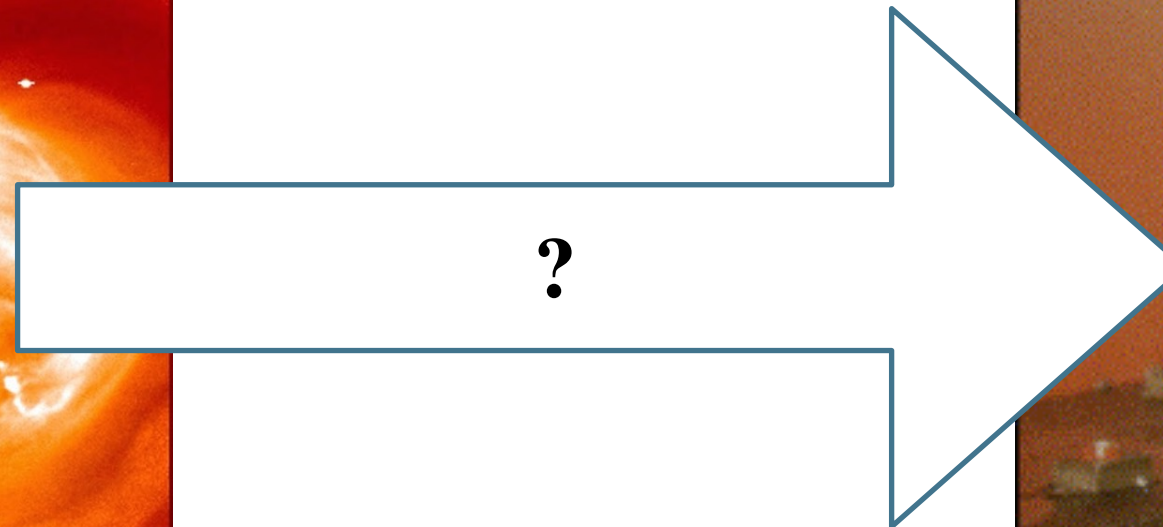
# Forecasting GLCs

The end result should essentially be a transfer function between solar wind and the local geoelectric fields or GLCs. The question is:

—> **How well can it work?**



Credit: SOHO, 2003-12-02



Credit: PSE&G



# LSTM Neural Net for predicting GLCs

- Develop a neural network (LSTM) to map from solar wind variations to a measure of GLC at the Earth
- LSTMs have memory of past data - they are a good tool for this problem and have been used in similar studies
- Use past **120 mins of solar wind data** to **predict max(GLCs)**
- In data preparation: balance quiet events with active events to not overtrain on quiet times

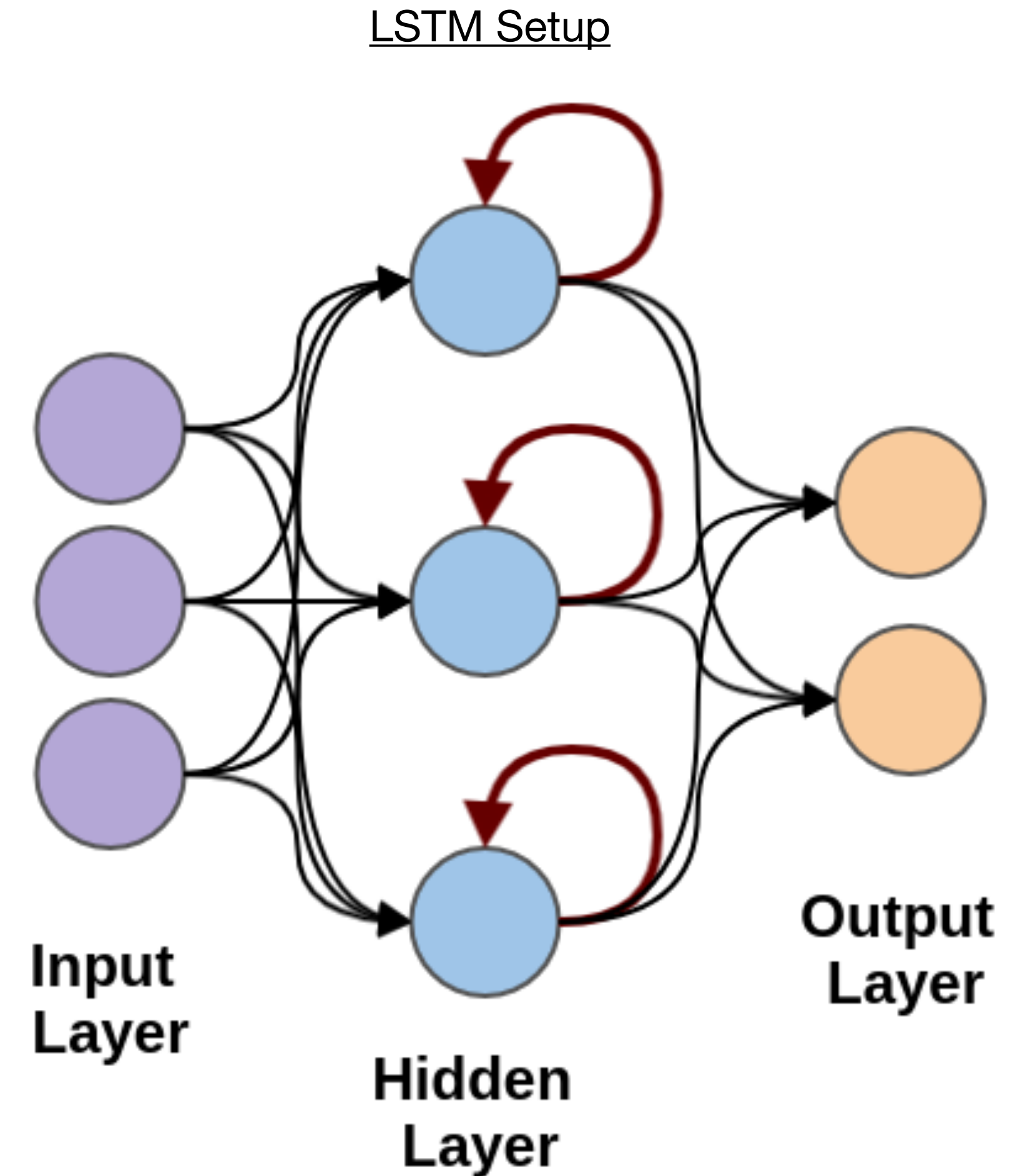
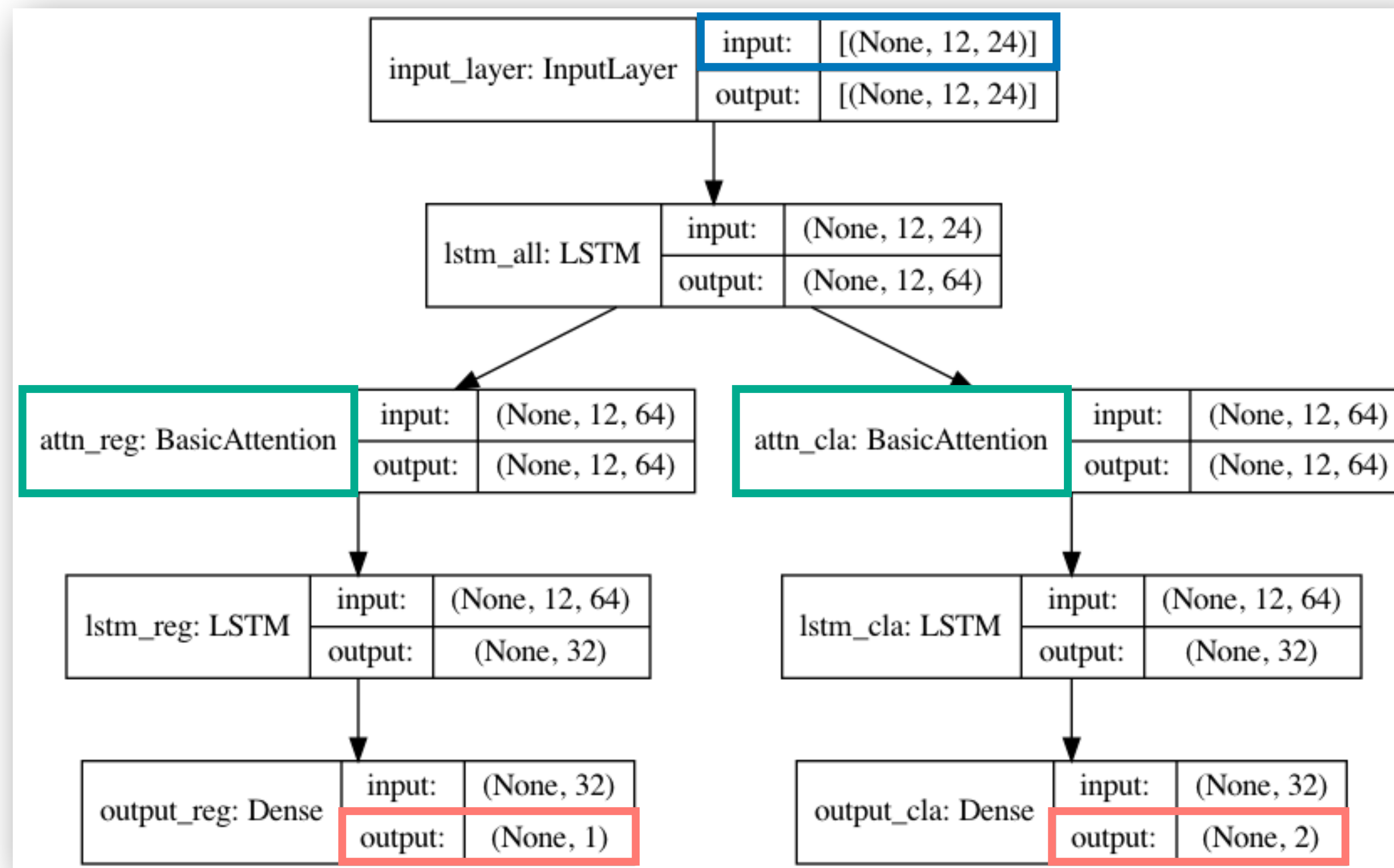


Image courtesy of [austingwalters.com](https://austingwalters.com)



# LSTM Neural Net Architecture



|E|

+ or -

- For geoelectric field: predict **magnitude E** (main target, regression), and **direction of E-field** (as secondary target, classification) from **input solar wind variables** (speed, density, B-field)
- Include an **attention** mechanism (neural net that mimics human attention)
- *Training:*  
21 years of data  
*Testing:*  
5 years of data



# Comparing LSTM Output

- Train three *LSTMs* to predict three targets from the solar wind data:

$$\left. \begin{array}{l} 1. \text{ Modelled max}(E_x) \\ 2. \text{ Modelled max}(E_y) \end{array} \right\} (GIC = a \cdot E_x + b \cdot E_y) = \mathbf{GIC_E}$$

$$3. \text{ Measured max}(GICs) = \mathbf{GIC_{OBS}}$$

- Substation #1** is better predicted using LSTM-E trained on E-field data.

**Substation #5** is better predicted using LSTM-GIC trained on raw GIC data.

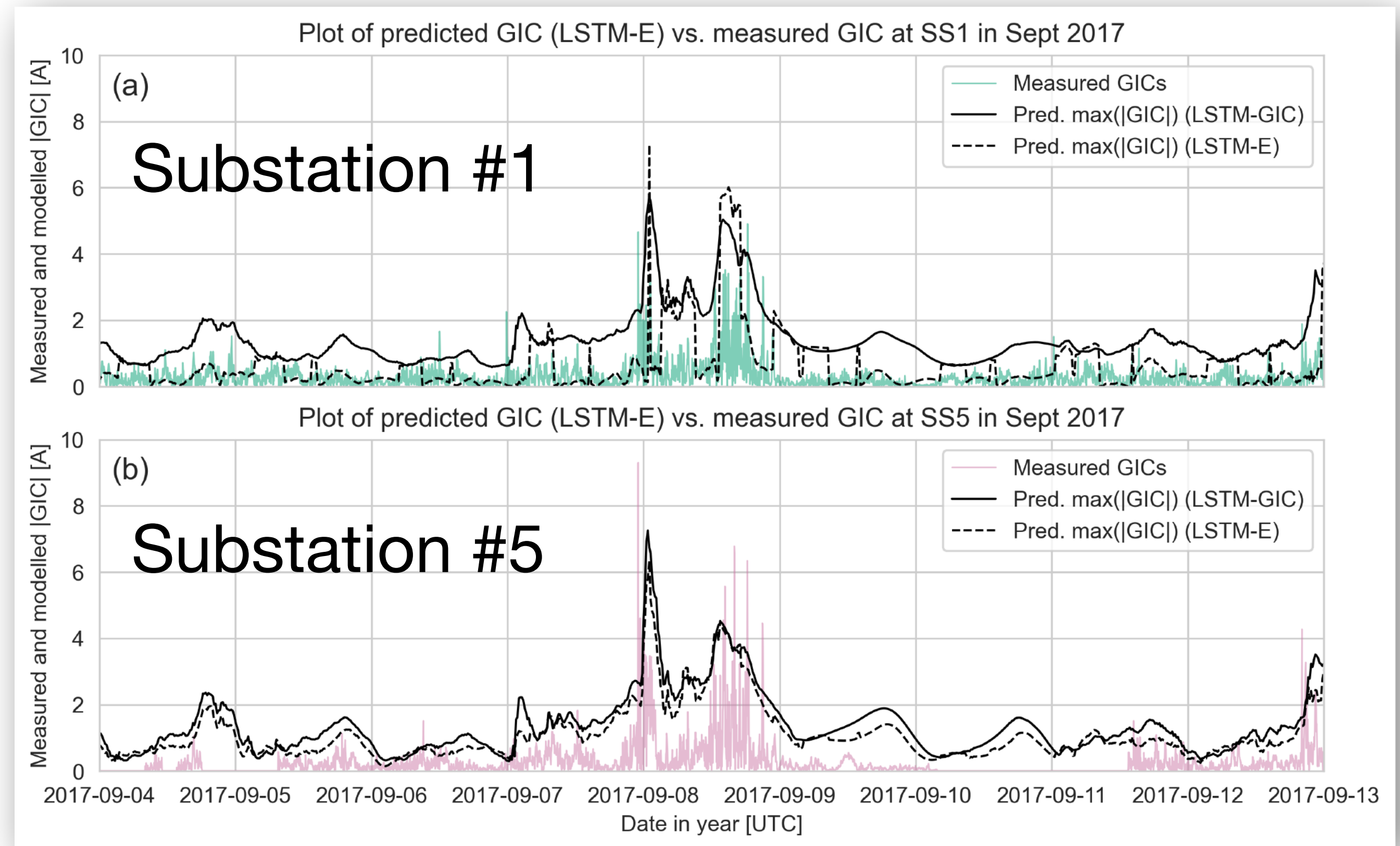
—> Different E-field directions dominate GICs in each station, likely cause of discrepancy.

Error measures: LSTM vs. Meas	<b>GIC<sub>E</sub></b>	<b>GIC<sub>OBS</sub></b>
<b>Pearson's R, #1</b>	0.35	<b>0.56</b>
<b>Pearson's R, #5</b>	<b>0.67</b>	0.64
<b>RMSE, #1</b>	<b>0.49 A</b>	0.67 A
<b>RMSE, #5</b>	<b>0.59 A</b>	0.78 A



# Results

- The model can predict times of heightened activity, and predicts a daily variation otherwise (see Figure for Sept. 2017 storm —>).
- **Probability of detecting events:** 50% for moderate activity ( $> 2$  A), 15% for strong activity ( $> 4$  A).
- Would need more information on status of magnetosphere and current systems to improve predictions.

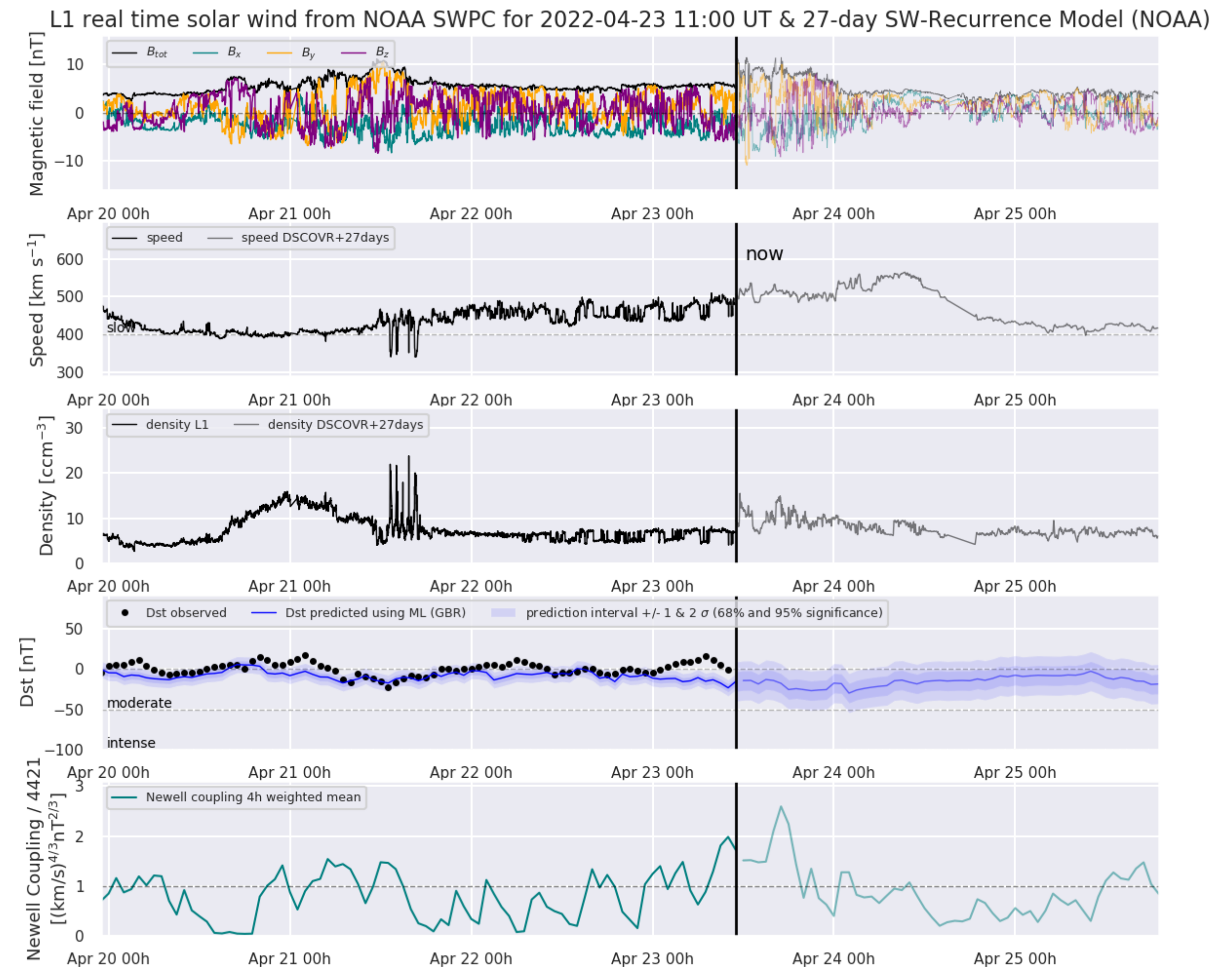


**Figure: LSTMs applied in virtual “real-time mode.”** Output of the forecasting models (black lines) vs. the observed GICs (coloured lines) at two stations (above and below). The solid black line shows results from LSTM-GIC trained on modelled GIC values, while the dashed line shows results from LSTM-E trained on modelled geoelectric field values.



# Where are we going with this?

- **PREDSTORM** (Fig. —>) provides forecasts of the in situ solar wind from various methods (STEREO, 27-day recurrence, flux rope modelling, ...) <https://helioforecast.space/solarwind>
- **Incorporate local GIC forecasts into the model** (for nowcast and expected future values)
- See the full study here: <https://doi.org/10.1029/2021SW002907>
- Code: <https://github.com/bairaelyn/SOLARWIND2GIC>



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