
Electrical activity prediction in the WRF-ELEC model using ML



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Summary

1 Why do we study electrical activity?

2 WRF and WRF-ELEC models

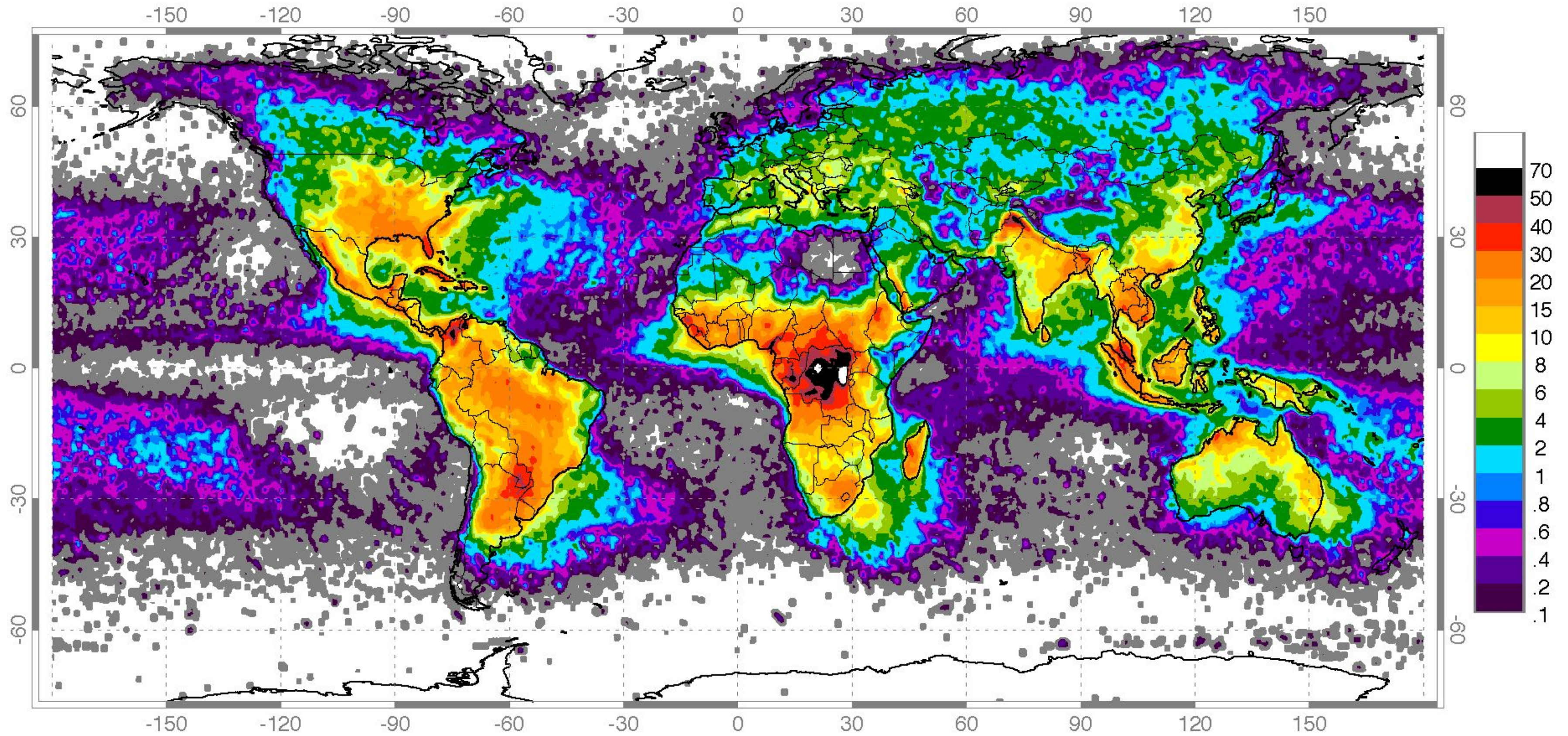
3 ML techniques

4 Results

5 Conclusions

6 Future studies

GLOBAL MAP OF LIGHTNING FREQUENCY [FLASHES/KM2/YR].

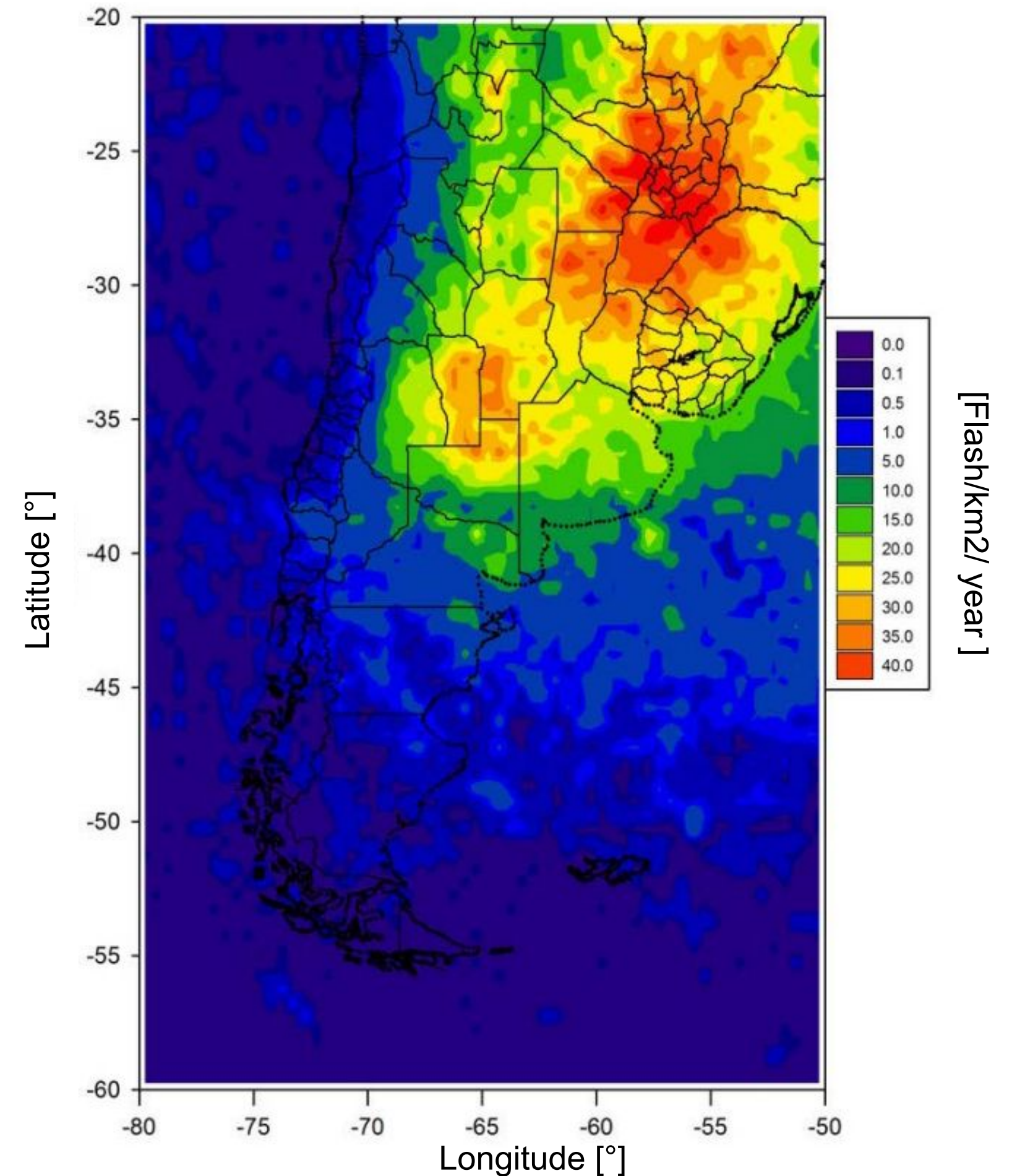


SOUTH - SOUTH AMERICA

- Is the region between 20° & 60°S latitudes and between 50° & 80° W longitudes.
- Due to local climate, precipitations in Center and North of Argentina are mainly generated by thunderstorms (Liu et al., 2010).
- The annual map of Thunderstoms Days (Tds) shows an absolute maximum located in northwestern Argentina with values greater than 100 Td/year

(Serrano et al., 2022).

- As figure shows, central region have relevant electrical activity with 25 flashes per km² per year, while northwest area shows more than 40 flashes per km² per year (Nicora et al., 2014)

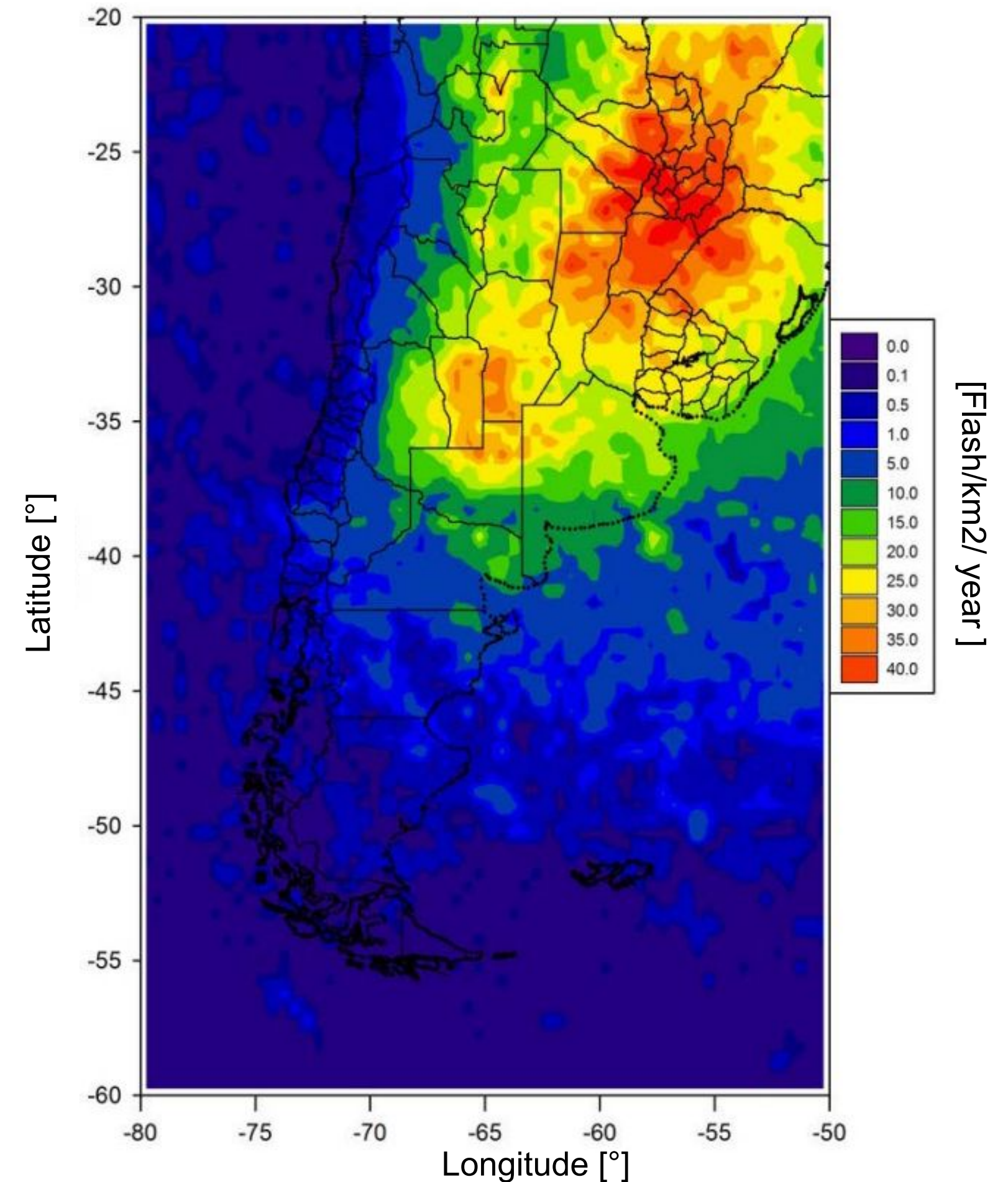


ELECTRICAL ACTIVITY AS A PROXY

Research indicates a correlation between electrical discharges and the microphysics and dynamics of thunderstorms. Correlations with severe weather events have also been identified.

- Updraft maximum (Deierling and Petersen, 2008).
- Precipitation ice mass (Deierling and Petersen, 2008).
- Updraft volumes (Schultz, et al., 2017)
- Graupel mass (Carey et al., 2019)
- CTT and hail precipitation (Borque et al., 2020)

Then, it could be possible to predict severe events from electrical activity.



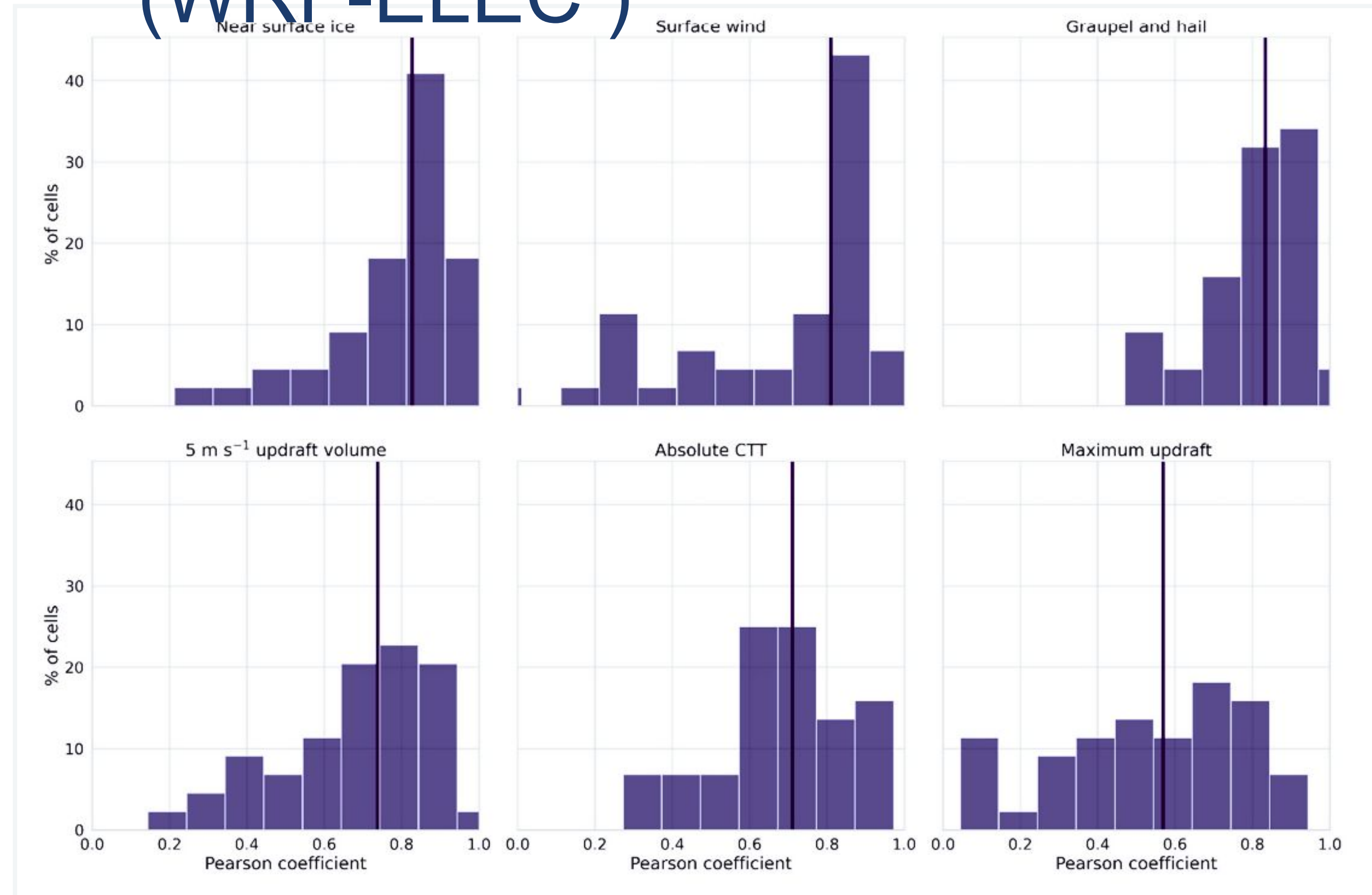
WRF / WRF-ELEC models

- WRF is a mesoscale forecast model and assimilation system developed by numerous institutes and centers in the U.S.
- WRF-ELEC is a storm electrification package for the WRF model which explicitly represents the charge and discharge processes of the clouds (Fierro et al., 2013).

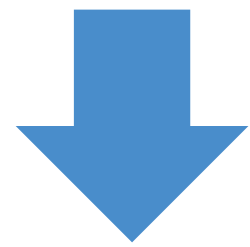


Correlations with electrical activity (WRF-ELEC)

Histograms for maximum Pearson correlation coefficients between the LIGHT variable and: near-surface ice, surface wind, graupel + hail mixing ratios, 5 m/s updraft volume, CTT and maximum updraft. Vertical lines indicate median value at each case.



WRF-ELEC



ML MODEL

Only works with WRF 3.9.1.
Last WRF is 4.5.1

Uses WRF to create dataset.
ML with Python.



Models



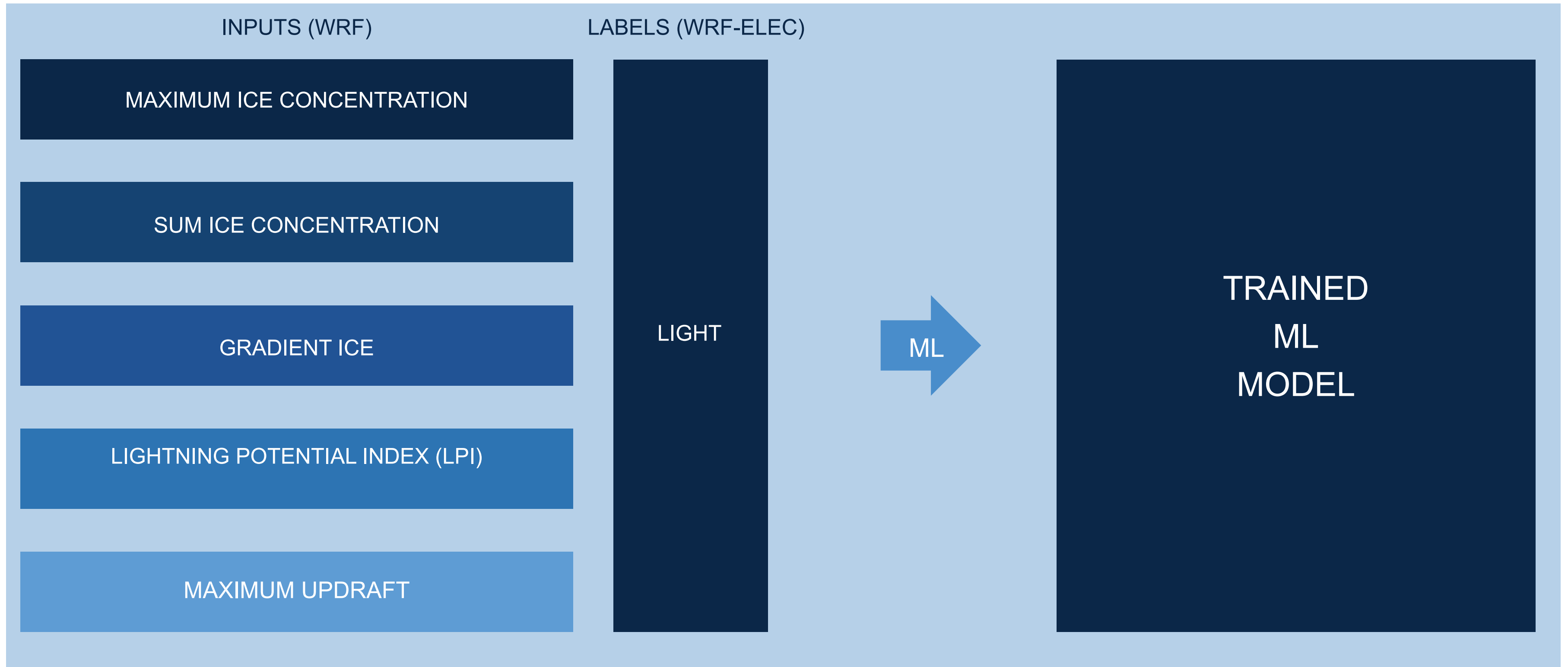
RANDOM FOREST
is an ensemble learning
method for classification,
regression and other tasks
that operates by
constructing a multitude of
decision trees at training
time.

For regression tasks, the
mean or average
prediction of the individual
trees is returned.



NEURAL NETWORKS
are a branch of machine
learning models that are
built using principles of
neuronal organization.
Artificial neural networks
can be described as
models with at least two
layers: one input and one
output, as well as, in most
cases, other intermediate
layers (hidden layers).

Dataset and training



ML implementation

NEW INPUTS (WRF)

MAXIMUM ICE CONCENTRATION

SUM ICE CONCENTRATION

GRADIENT ICE

LIGHTNING POTENTIAL INDEX (LPI)

MAXIMUM UPDRAFT

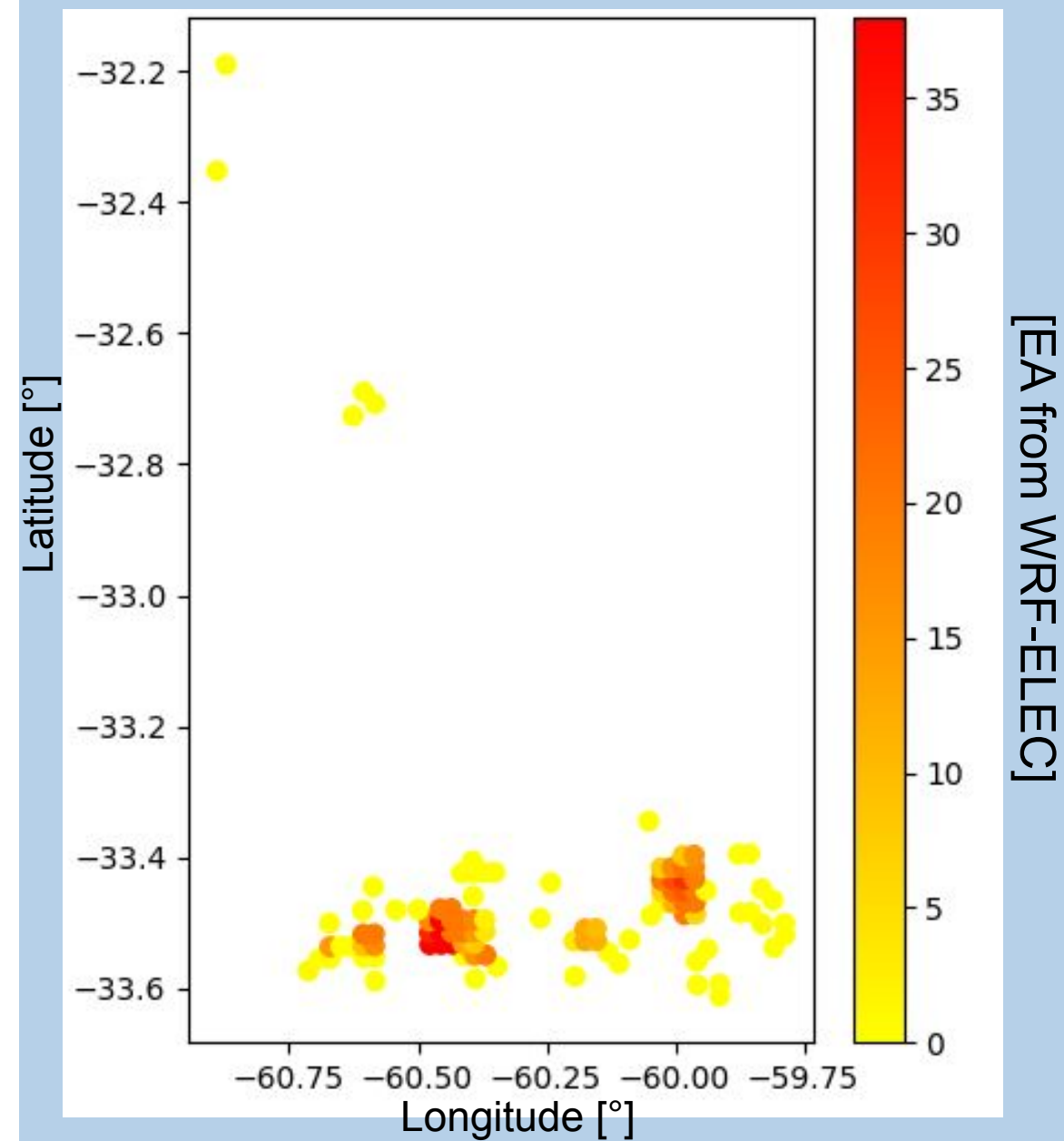
TRAINED
ML MODEL

PREDICTED
ELECTRICAL
ACTIVITY

Results



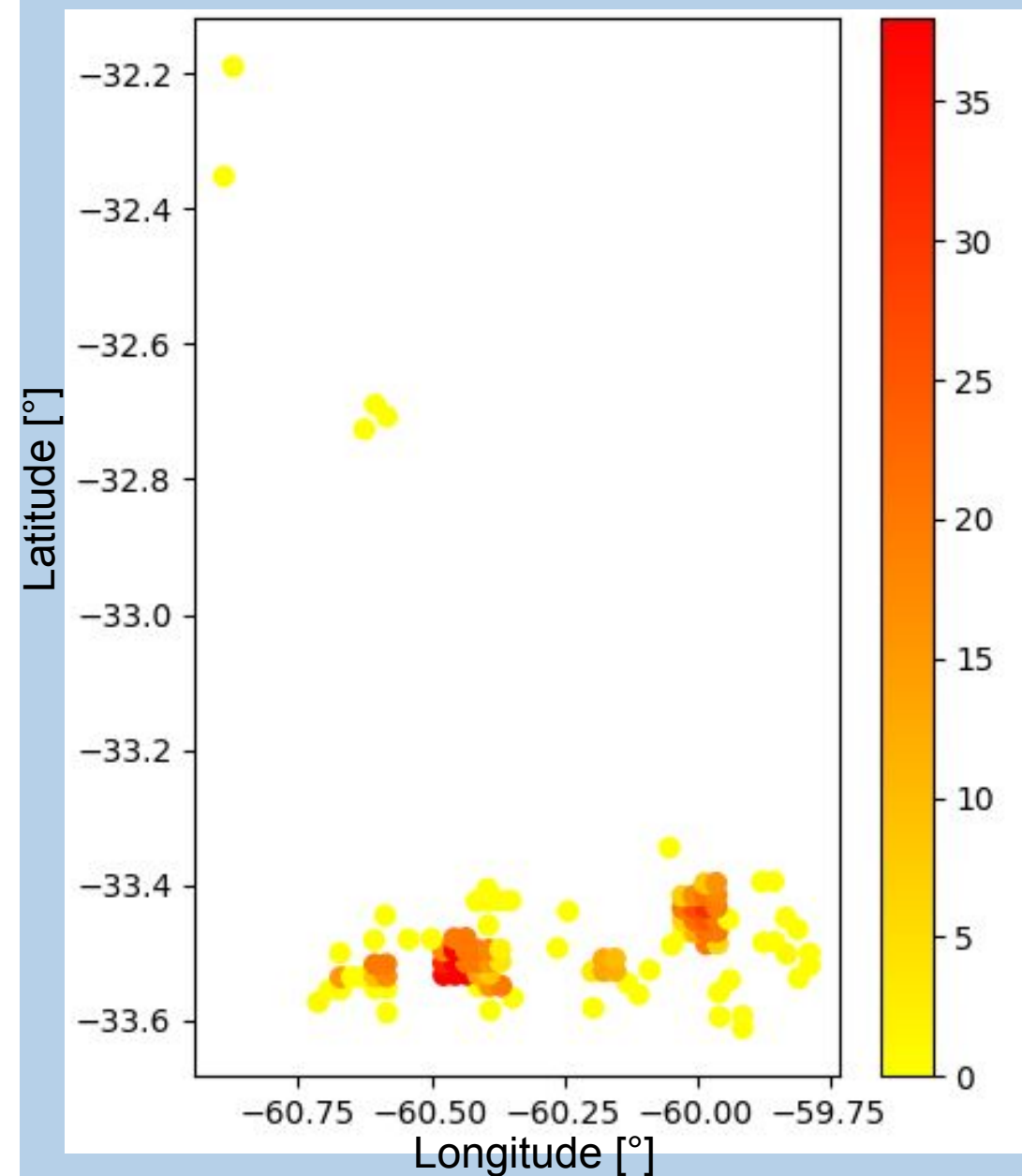
WRF-ELEC SIMULATED
ELECTRICAL
ACTIVITY



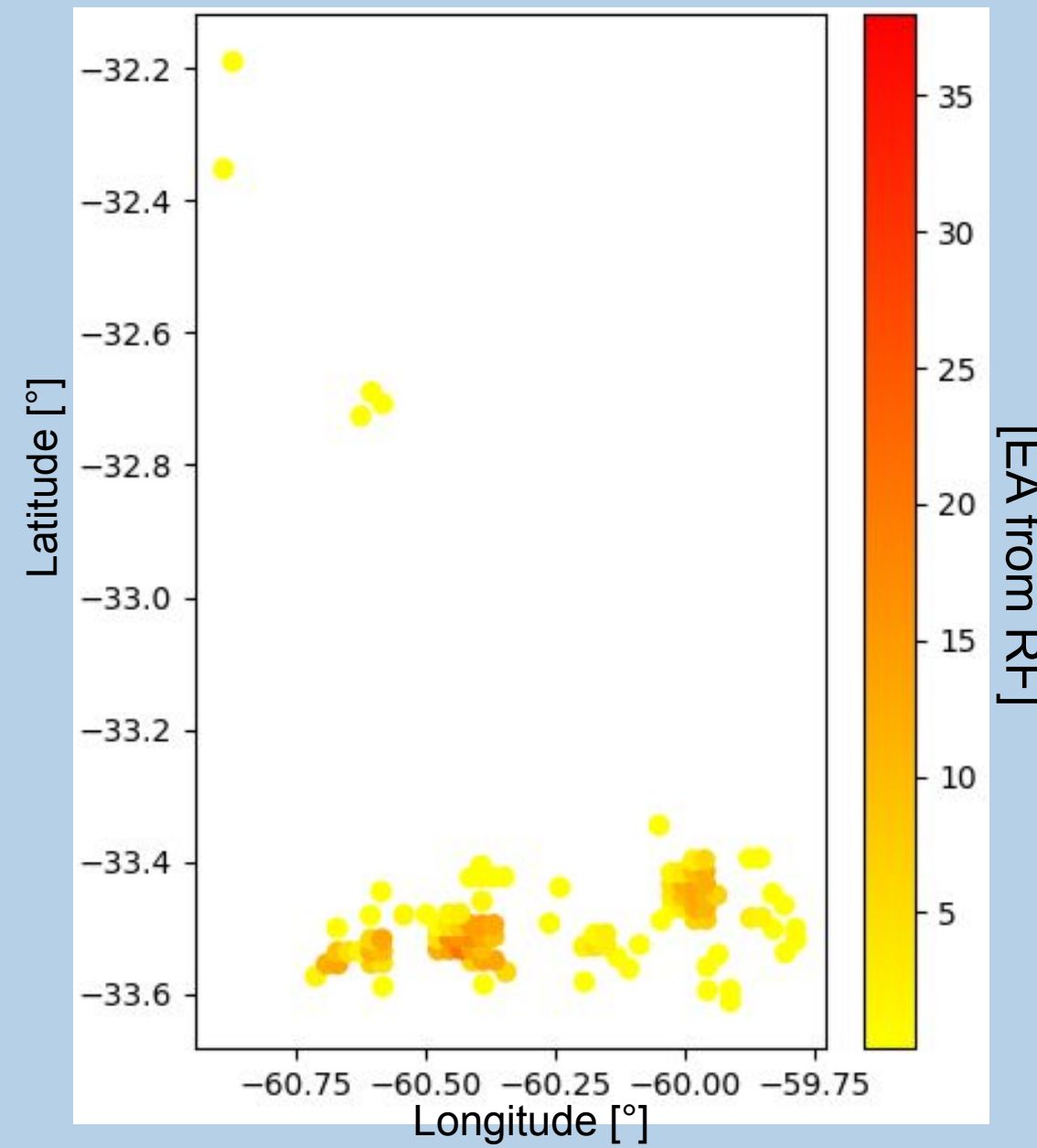
Results



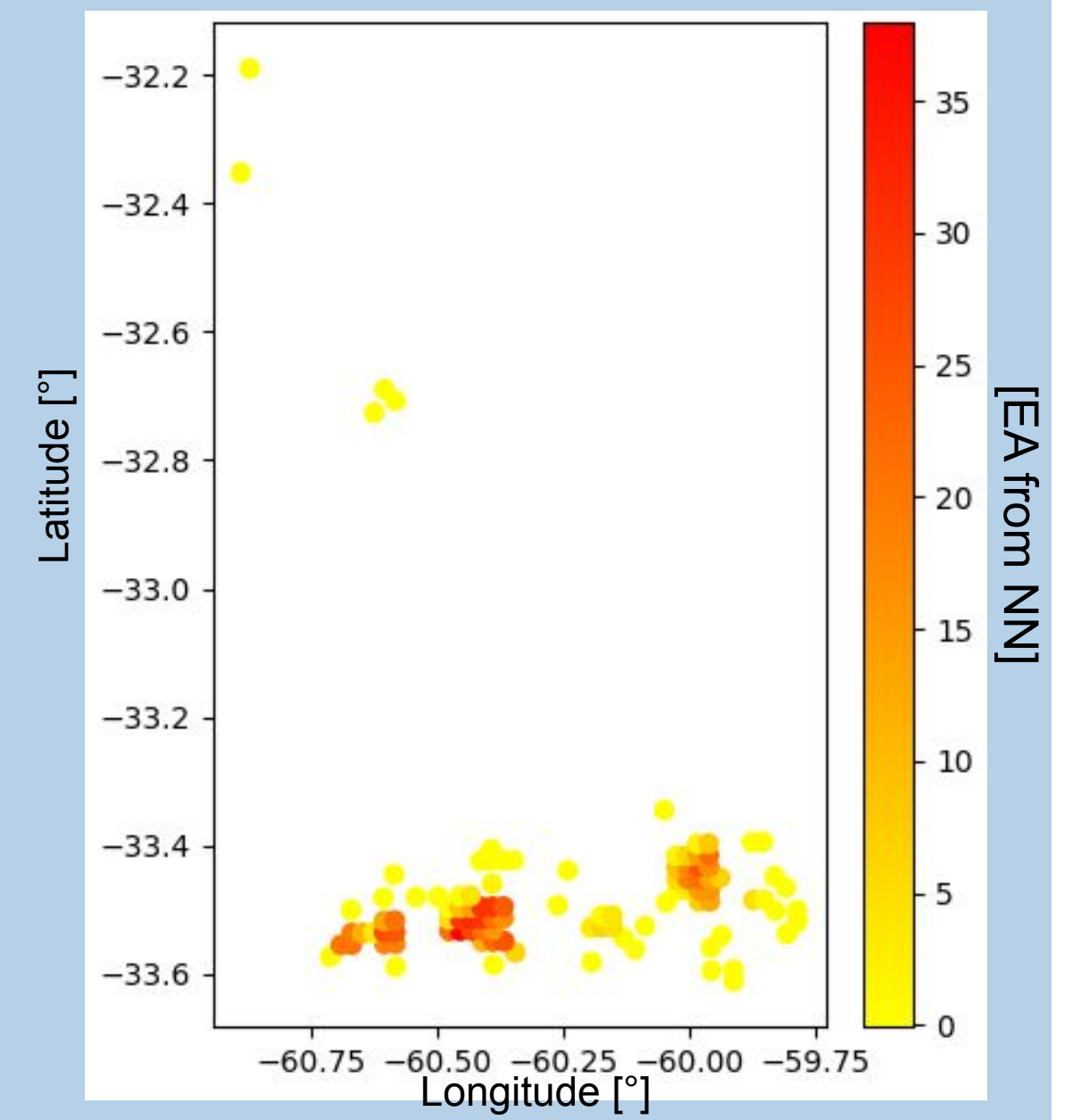
REAL ELECTRICAL
ACTIVITY



RANDOM FOREST



NEURAL NETWORKS





Conclusions

- BOTH ML MODELS REPRODUCE SUCCESSFULLY THE LOCATION OF DISCHARGES SIMULATED BY WRF-ELEC.
- NN SHOWS BETTER PERFORMANCE THAN RF MODEL WITH VARIANCES OF 0.67 AND 0.6, RESPECTIVELY.
- TAKING INTO ACCOUNT THE PRESENT RESULTS, ML MODELS DEVELOPED HERE COULD BE USED AS AN ALTERNATIVE TO WRF-ELEC MODEL.



Future studies

- IMPLEMENT MORE COMPLEX MODELS SUCH AS CONVOLUTIONAL NEURAL NETWORKS.
- EXTEND STUDIES TO OTHER SOURCES OF LIGHTNING INFORMATION SUCH AS WWLLN INCLUDING NOT ONLY WRF DATA BUT ALSO RADAR DATA.



**Thank
you!**