

LIGHTNING AND EXTREME WEATHER



Google Cloud

USGS
science for a changing world

MAYO CLINIC

ESA
LUXEMBOURG SPACE AGENCY

SPACE RESOURCES LU

LOCKHEED MARTIN

MIT Portugal

intel

IBM

NVIDIA

planet.

POSTER TITLE AND RESEARCHERS

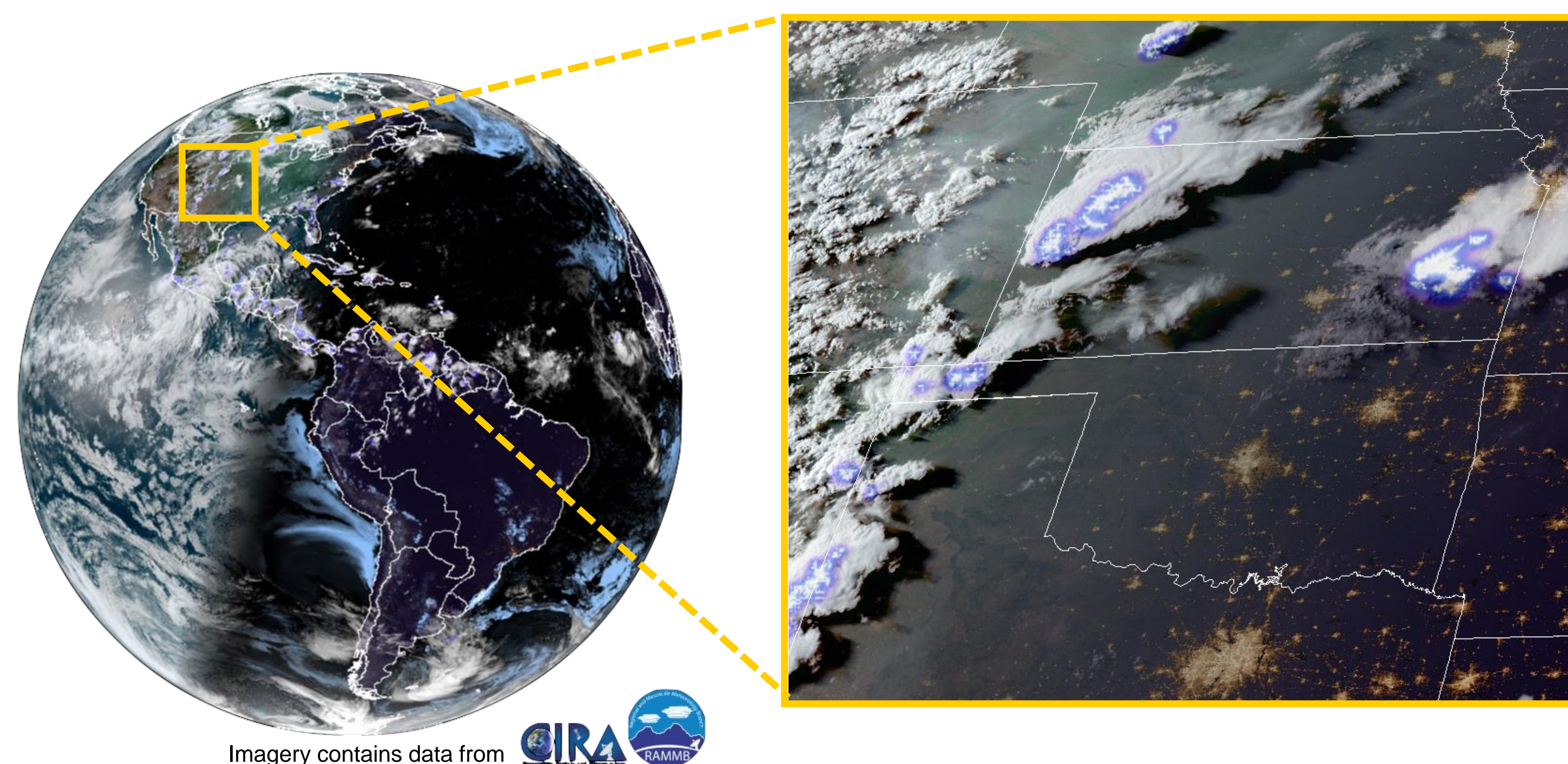
SEVERE WEATHER PREDICTION USING LIGHTNING DATA

Iván Venzor-Cárdenas, Nadia Ahmed, Maria J. Molina, Marek Slipski



CHALLENGE & OPPORTUNITY

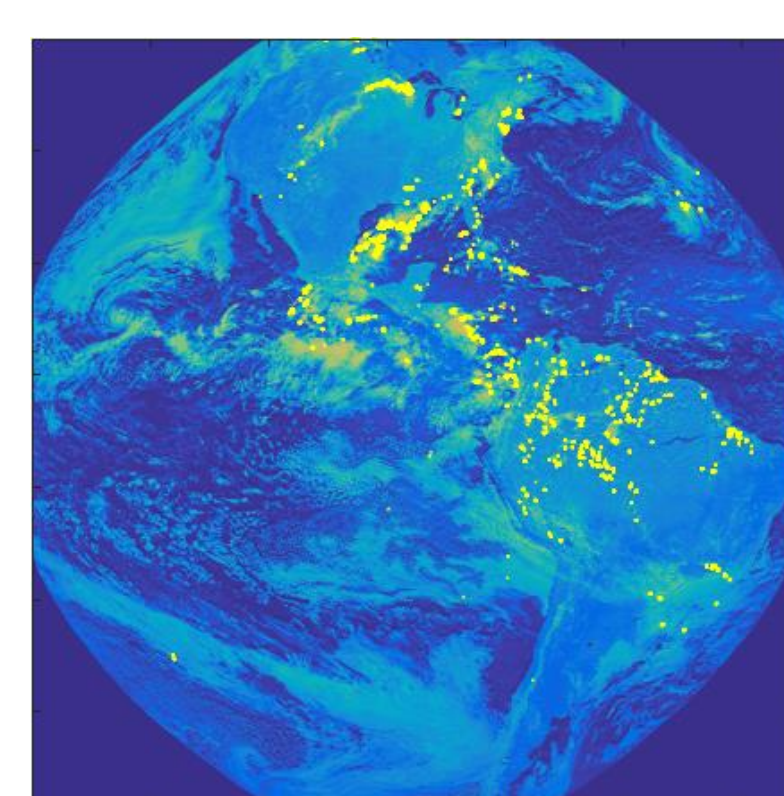
PROBLEM: Can we use lightning observations from GOES satellite to improve predictions of severe thunderstorms and help forecasters keep the public safe?



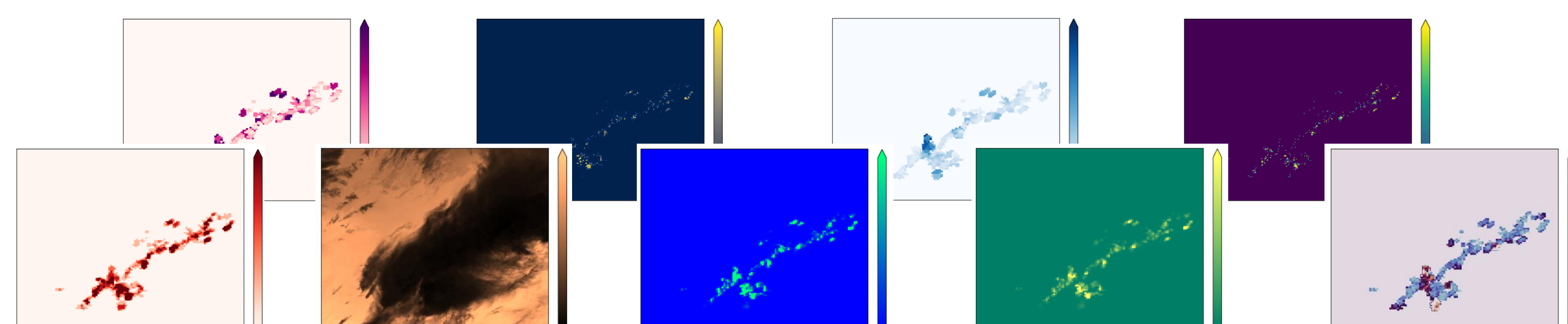
Imagery contains data from CIRA RAMMB

Data from the Geostationary Lightning Mapper (GLM) and Advanced Baseline Imager (ABI) instruments aboard GOES can be used to create proxies for convection intensity that can then be used to identify possible severe weather.

OPPORTUNITY: Use high temporal resolution and large spatial coverage of GLM and ABI data aboard GOES to improve predictions of severe thunderstorms.



8 variables from GLM (e.g., flash extent density, group centroid density) and 1 variable from ABI (channel 13).



Requires **significant** computational resources.



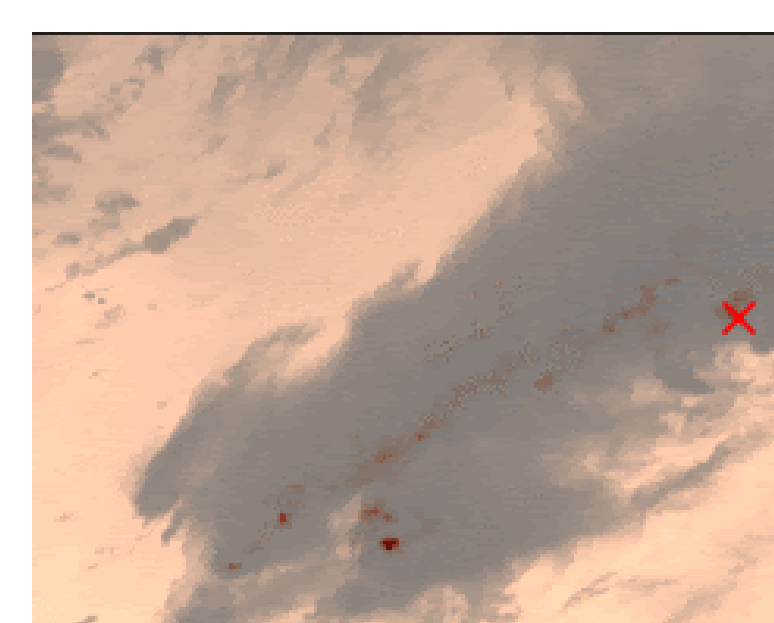
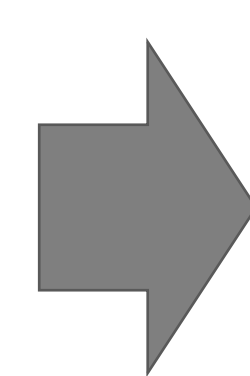
Google Cloud

240 CPUs

9 input channels, 1 min cadence

10 GPUs

3,000,000+ GLM files processed



OUTCOME Our time series convolutional-kernels model suggests GLM + AI can improve forecasting of severe weather events with a 15 min lead time.



15 min nowcasting lead time



8 GLM lightning time series derived quantities

1/2

Reduced false alarms for warned thunderstorms

8/10

Correctly classified tornadoes and severe hail reports



Train model using 8 GLM quantities 60 minutes prior to events to classify as severe or non-severe.

Non-severe: Warning with no confirmed report in a 24-hour period.

Severe: Confirmed tornado or severe hail report.



ROCKET
RandOm Convolution
Kernel Transform
(Dempster et al. 2020)

10,000 convolutions

Classifier



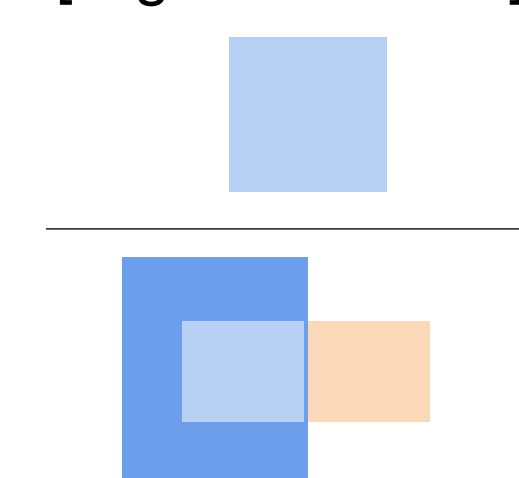
15 min lead time

Severe

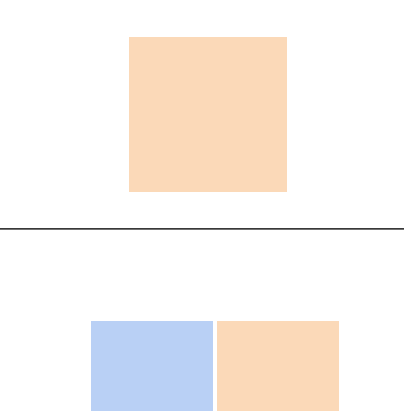
Non-severe

Miss	Correct Rejection	83 Hit	26 Miss
Hit	False Alarm	59 False alarms	53 Correct rejection

Critical Success Index (CSI)
[Higher is better]



False Alarm Ratio (FAR)
[Lower is better]



Project Pikajoule

State of the Art

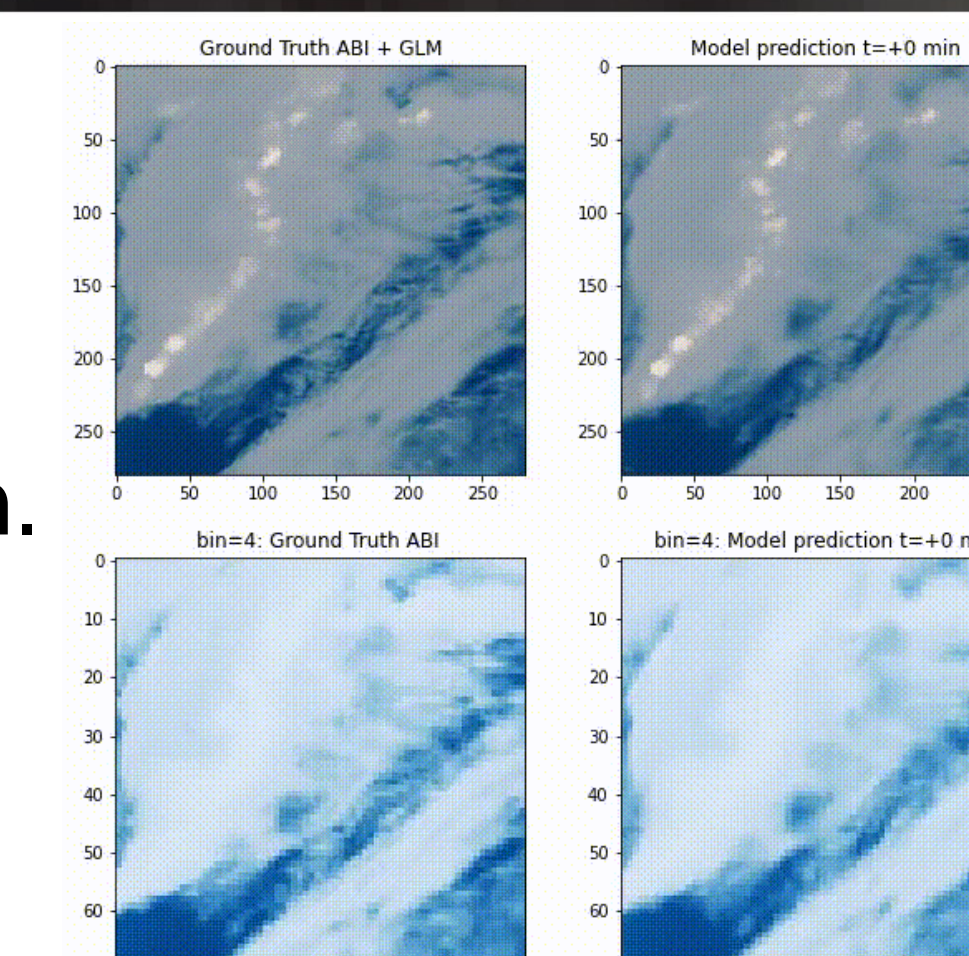
<i>Lead time</i>	15 min	> 15 min
<i>Coverage</i>	Central US (1,000 x 800 km)	CONUS
<i>Period</i>	Mar-Jun 2019	May-Jul 2014, Mar-Dec 2016
CSI	0.49*	~0.35
FAR	0.41*	~0.55

* Mean of 100 trained models in test dataset.

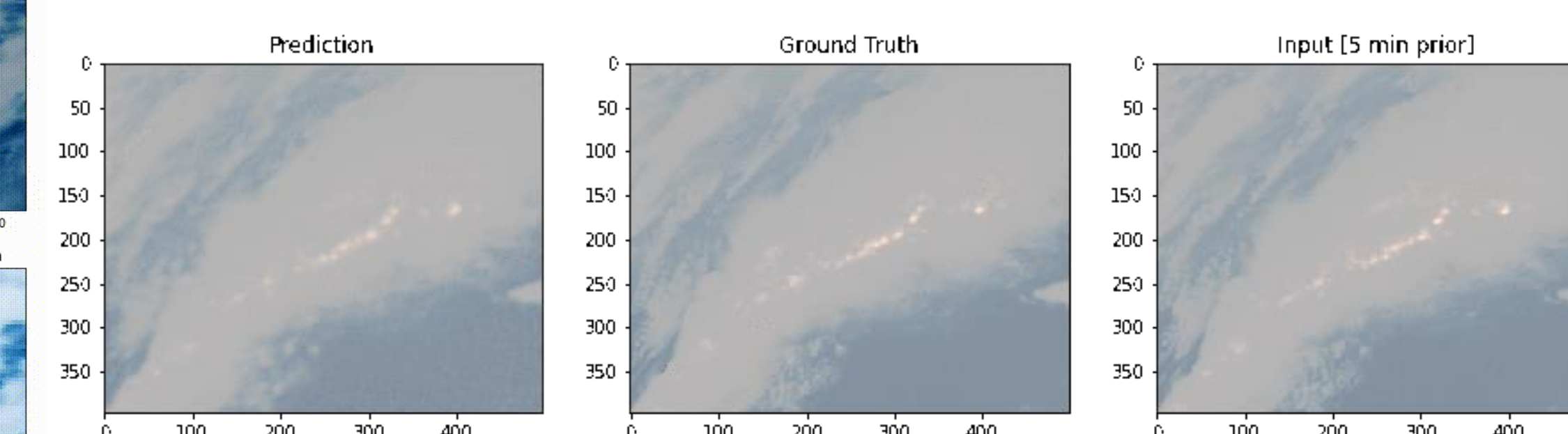
National Weather Service, (Cintineo et al. 2018)

NEXT STEPS

- ML explainability.
- Video frame prediction.
- Geographic and seasonal extension.
- Increased lead time.



U-Net model (GLM, ABI).



Recurrent model (GLM, ABI, severe mask).