Climate-Change Projections of Hazardous Convective Weather Using an Environment-Informed, Convection-Permitting Dynamical Downscaling Methodology (96)

S. Wang, R. Trapp, University of Illinois; **J. Allen, D. Gopalakrishnan**, Central Michigan University; **E. Robinson**, Aon, Inc.



MOTIVATION

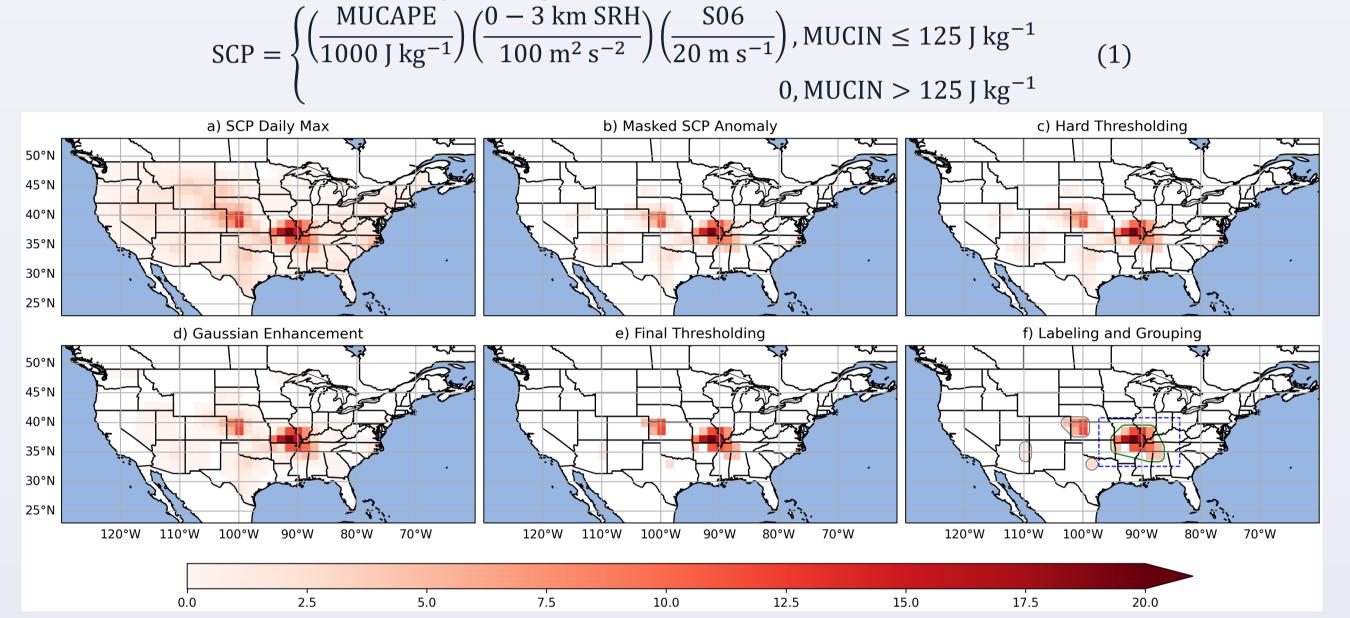
• Studies of hydro-climate extremes benefit from application of convectionpermitting dynamical downscaling (CPDD)

ILLINOIS

- e.g., storm-scale generators of locally heavy precipitation, tornadoes, hail, and severe winds (HCW) are explicitly represented with few-kilometer grid spacings
- The scope of CPDD studies, however, tend to be more computational—resource limited than GCM studies, which imposes limitations on:
- o integration lengths, computational domain size, number of different GCM drivers, etc.
- Such studies have have tended to downscale a single GCM (or GCM composite), and thus are absent of uncertainty measures otherwise determined with an ensemble of simulations via an ensemble of GCMs
- o exceptions include Kendon et al. (2020) and Coppola et al. (2018)
- With our "environment-informed" approach, only a subset of days/geographical domains are downscaled, therefore eliminating unnecessary model integrations: This allows us to downscale multiple GCMs, providing HCW projections with more confidence.

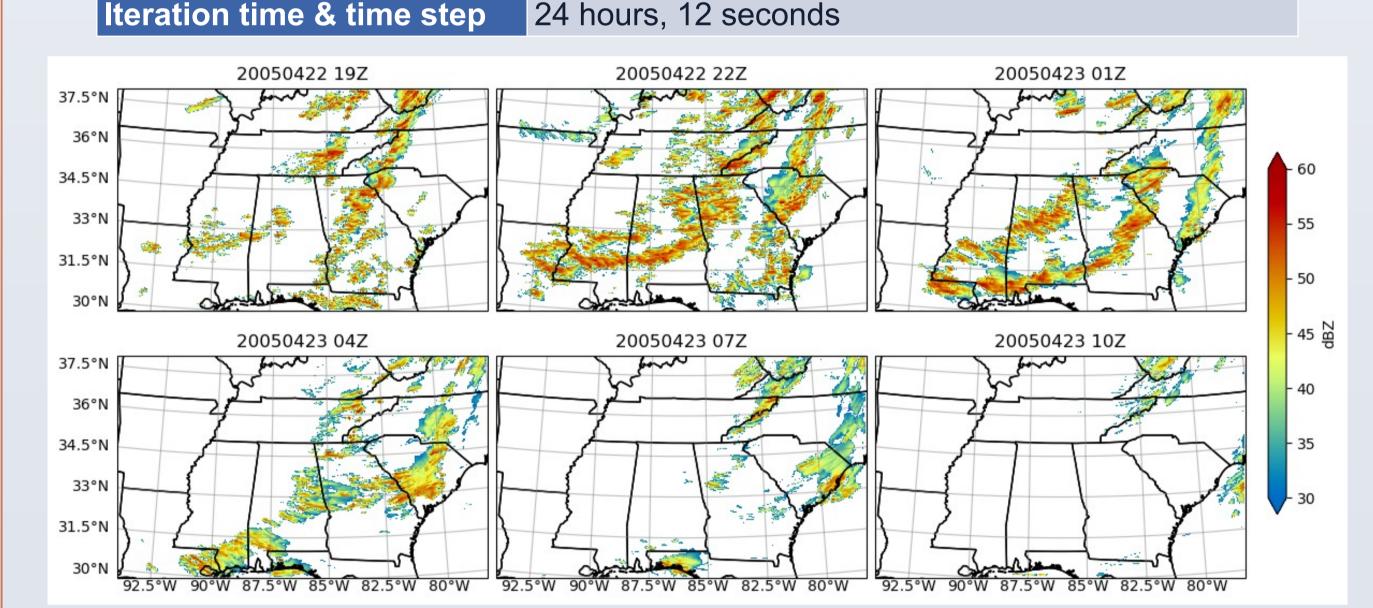
METHODOLOGY

1. Environment screening using supercell composite parameter (SCP):

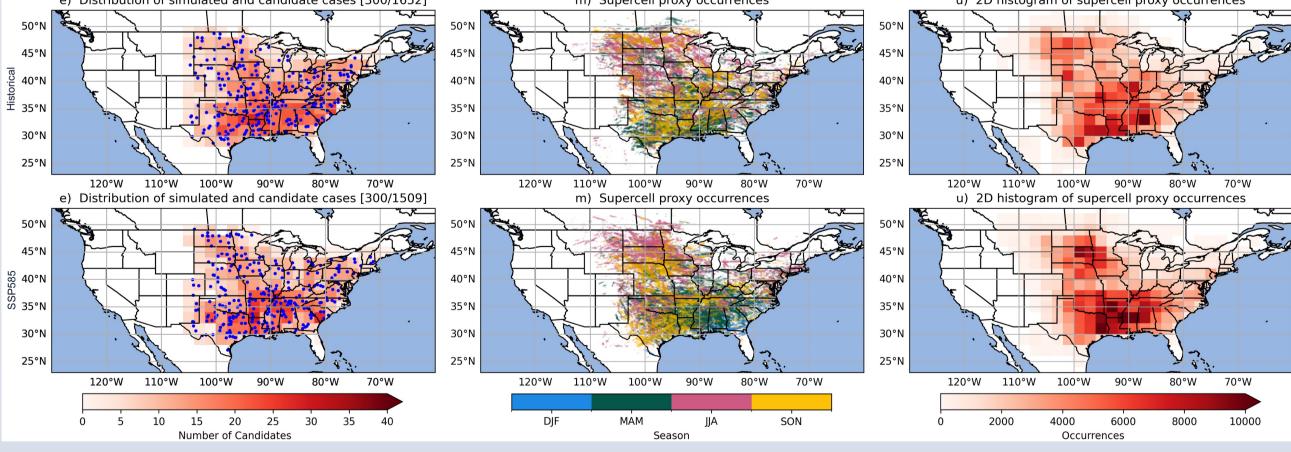


2. WRF simulation for sampled cases:

Domain size & resolution500×300 grid points, 3 km gridpoint spacingVertical level & model top50 levels, 50 hPa



3. Collect candidate cases and sample 300 cases for dynamical downscaling:



- 4. Repeat for 8 GCMs from CMIP6:
 - BCC-CSM2-MR, CanESM5, CESM2, CNRM-ESM2, MIROC6, MPI-ESM1-2-HR, MRI-ESM2-0, NorESM2-MM.
- Historical (1995-2014), SSP585 (2040-2059):

NOTES ON APPROACH

This "environment-informed" (EI) approach, which is used to generate a CPDD ensemble:

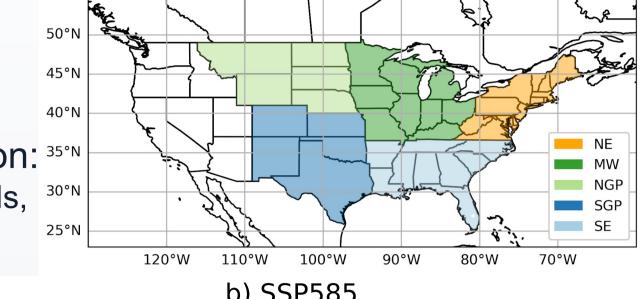
- offers flexibility: one could repeat a given set of runs with different physics, refined resolution, different downscaling models/versions;
- + can be done using fairly modest HPC resources;
- + is adaptable for different applications;
- + is also adaptable for different scenarios, time periods, etc.;
- is not suitable for problems involving feedbacks (e.g., land-atmosphere)
 on multi-day or longer periods;
- is not suitable to study the water cycle

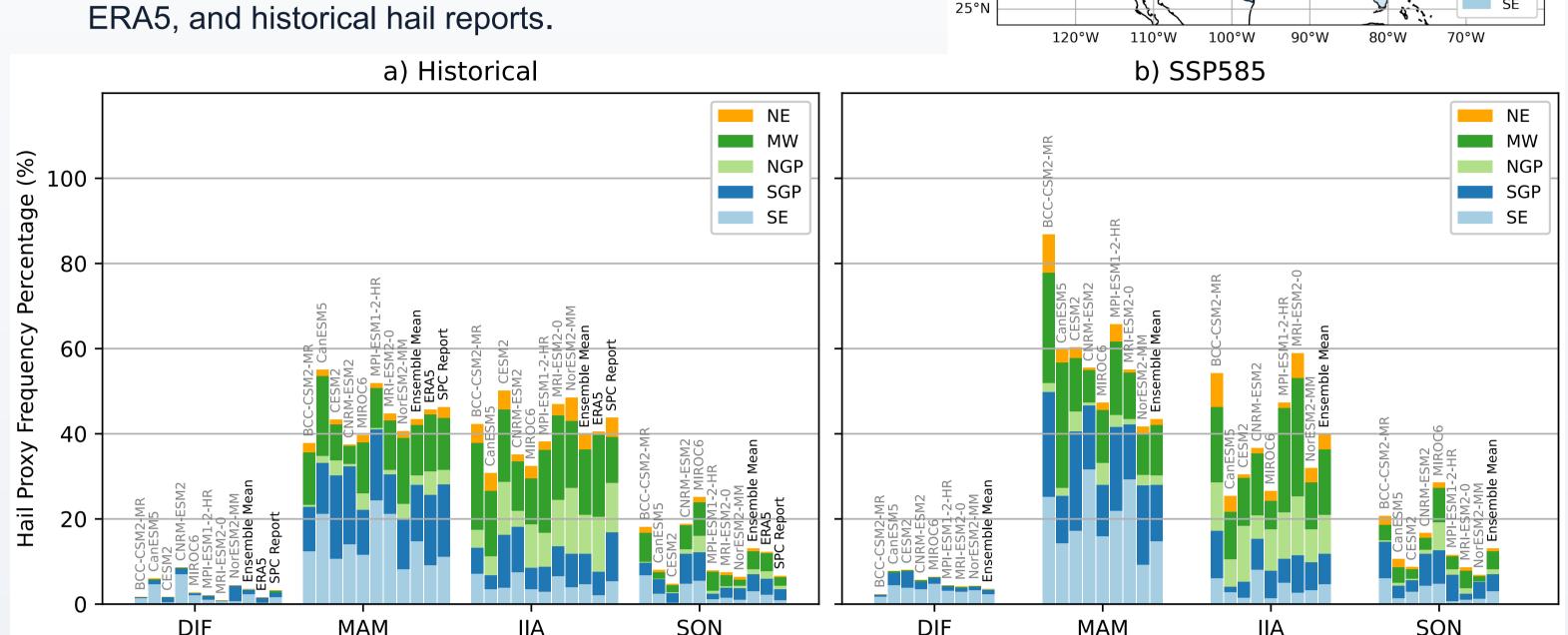
RESULTS FROM CPDD ENSEMBLE

UNIVERSITY

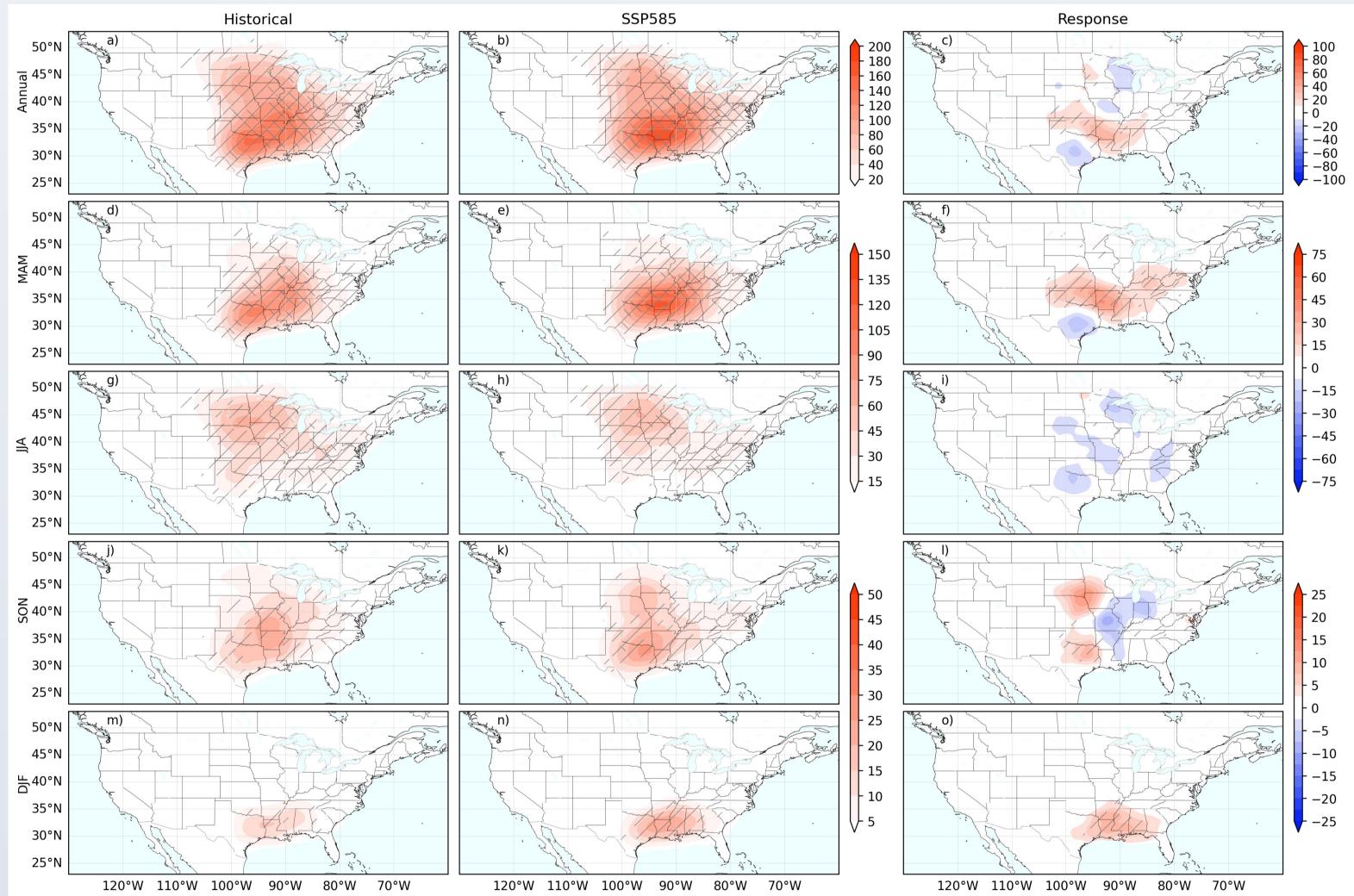
Hail proxy: HAIL_MAXK1>1.8 cm & WUP>20 m/s & REFD MAX>55 dBZ

1. Hail proxy occurrence stratified by region and season: 35°N
 Similar seasonal and regional distribution among GCMs, 30°N

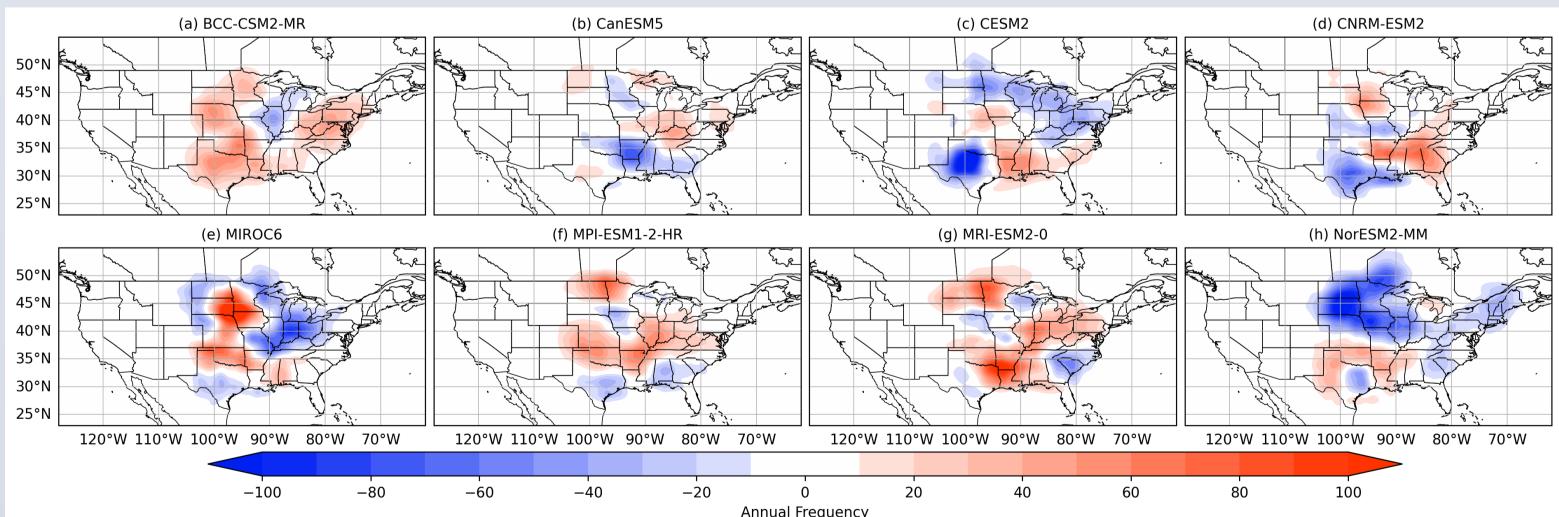




- 2. 2D hail proxy occurrence with robustness/significance:
 - Consistent with observed climatology, the center of hail occurrence moves northwards during summer, while in winter is confined to the coast of the Gulf of Mexico.
- Hail increases in the future in the region extending from Oklahoma to Tennessee-Alabama border.
- Spring is the season that contributes the most to hail occurrence in the historical and future; the climate change response is largest in spring as well.
- The historical and future occurrence is robust among the ensemble (Mean>Std).
- The response of hail occurrence is significant only in limited regions. (T-test: P<0.10)



- 3. Ensemble spread and GCM biases:
 - GCM biases, etc. contribute to different responses in future hail occurrence across the ensemble.
- The spread in this response shows the value of a multi-GCM ensemble.

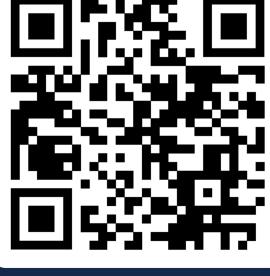


4. Hail size distribution across the ensemble:

- Hail between 1–1.75 cm tends to increase in the future; hail larger/smaller than this decreases.
- The future distribution of hail size has larger mean and smaller variance.

ONGOING WORK

- Future projection of tornado and severe wind activity.
- Upscaling: machine learning methods to build the relationship between large-scale environment and convective storm activity.



This research was sponsored by Aon, Inc