

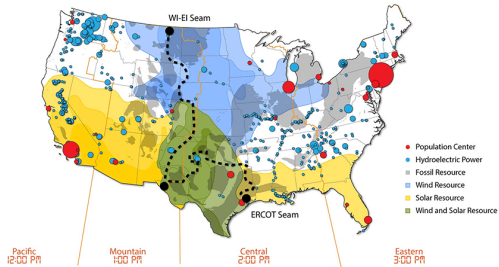


# Weather and Climate Data at NREL

Grant Buster  
June 2024

# Renewable Energy Integration Studies: Envisioning the Future of Energy Systems

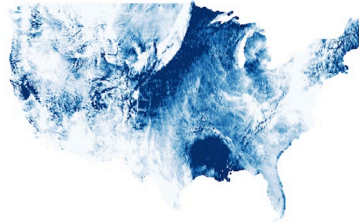
The NREL Seams Study



The North American Renewable  
Integration Study (NARIS)



The Los Angeles 100% Renewable Energy Study



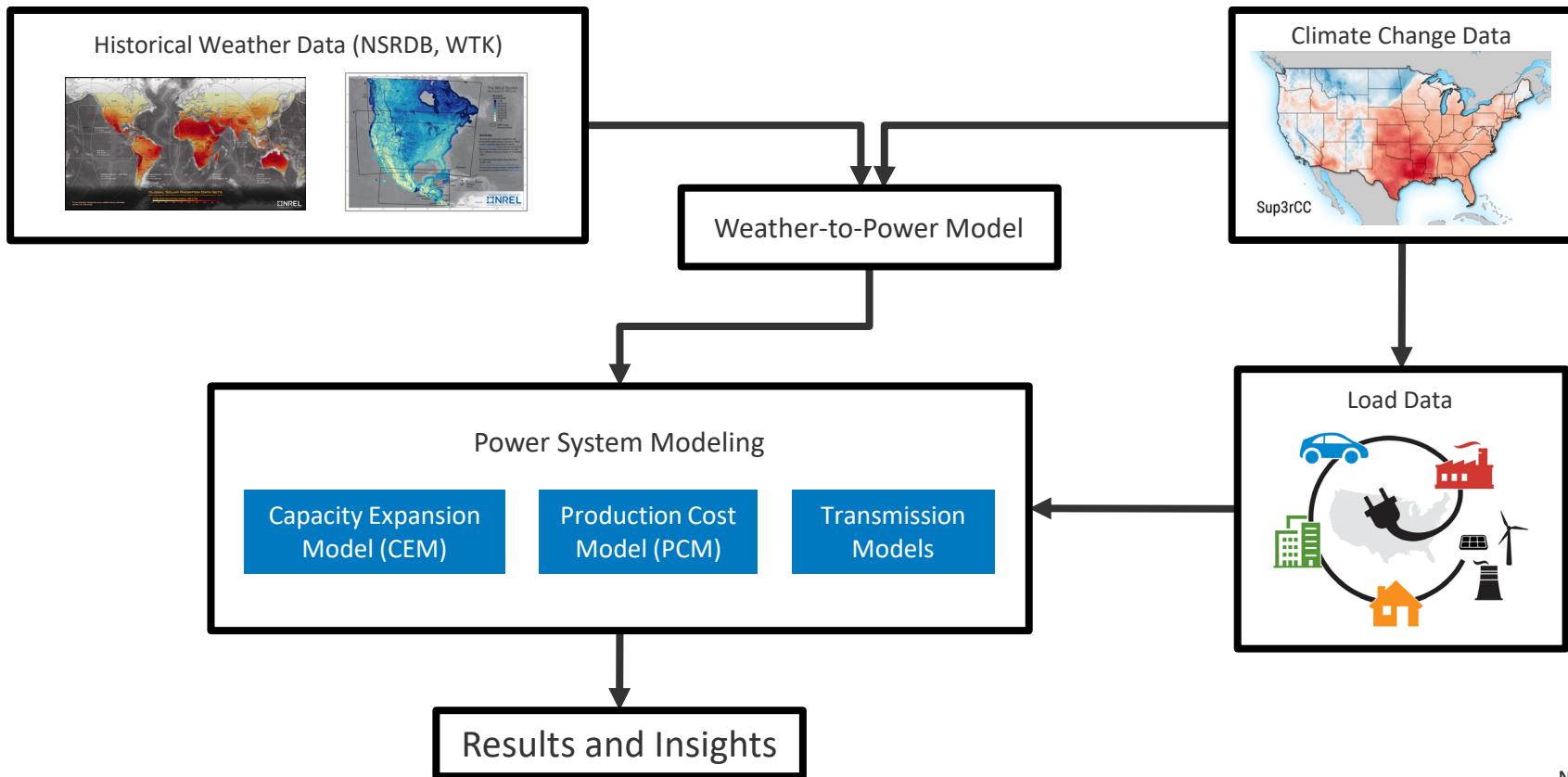
U.S. DEPARTMENT OF ENERGY

**Building a Better Grid**

*National Transmission Planning Study*

“The Evolving Role of Extreme Weather  
Events in the U.S. Power System with High  
Levels of Variable Renewable Energy”

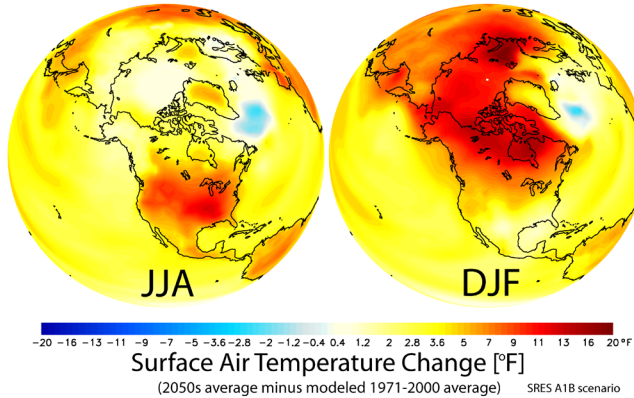
# How Is Weather Data Used In Analysis?



# Integration of Climate Data: Mind the Gap

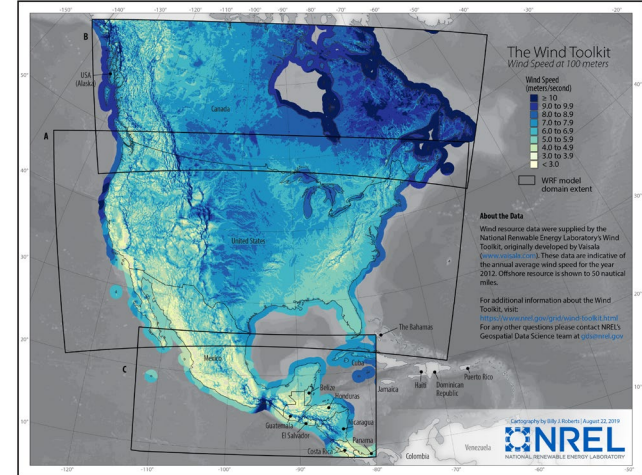
## Global Climate Models (GCMs)

NOAA GFDL CM2.1 Climate Model



<https://www.gfdl.noaa.gov/visualizations-climate-prediction/>

## Mesoscale NREL Datasets (WTK, NSRDB)



~100 km grid resolution  
daily average data  
2000-2100

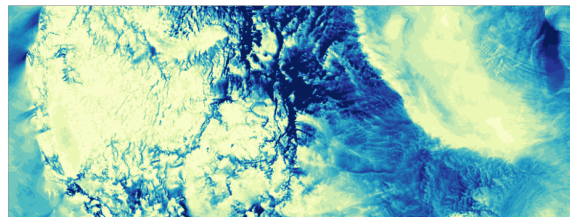
How do we bridge this gap?

~2-4 km grid resolution  
5 min-hourly data  
Historical

Our solution needs to be flexible enough to enable researchers to study any climate model or climate change scenario and to stay current with new climate research.

# Super-Resolution for Renewable Energy Resource Data with Climate Change Impacts (Sup3rCC)

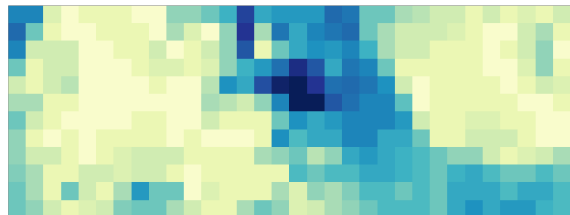
True High Res (WTK or NSRDB)



4km Hourly

Coarsen to  
create  
training data

Low Res (WTK, NSRDB, GCM)

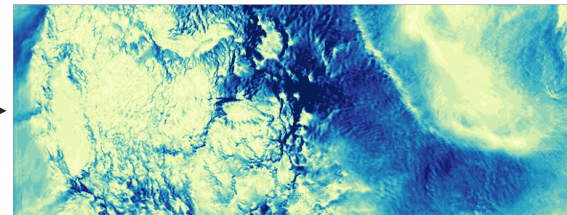


100km Daily

Discriminative  
Model

Generative  
Model

Synthetic High-Res Output



4km Hourly

## Benefits of Downscaling with ML:

1. Computational efficiency (40x-200x faster than WRF)
2. Designed for renewables (wind, solar, temp, humidity)
3. Fully integrated into energy analysis software
4. Open-source: <https://nrel.github.io/sup3r/>

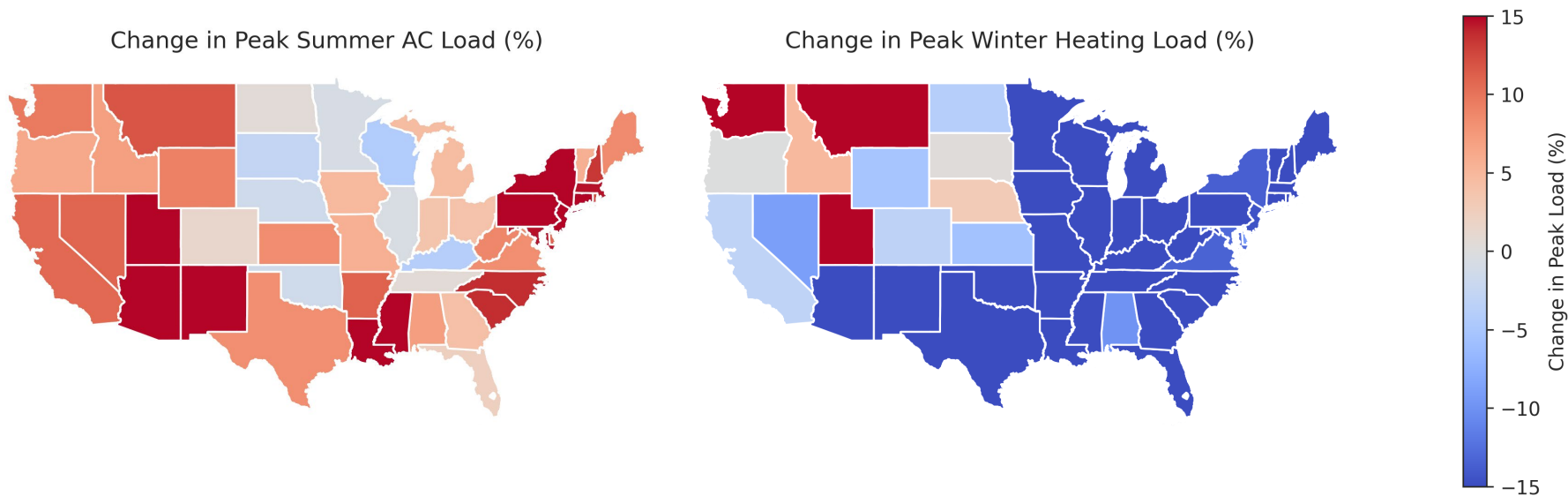
# Sup3rCC

- The Sup3rCC 4km hourly outputs (right) add **high-resolution spatial features and temporal dynamics** conditioned on the low-res GCM input (left)
- Includes **wind speed, solar irradiance, temperature, and humidity**, all spatiotemporally coincident
- Initial data is available for two GCMs, SSP5-8.5, 2015-2059
- Data on NREL-HPC and OEDI:
  - `/datasets/sup3rcc/`
  - [DOI 10.25984/1970814](https://doi.org/10.25984/1970814)



Buster et al., "High-resolution meteorology with climate change impacts from global climate model data using generative machine learning", *Nature Energy* (2024).  
<https://doi.org/10.1038/s41560-024-01507-9>

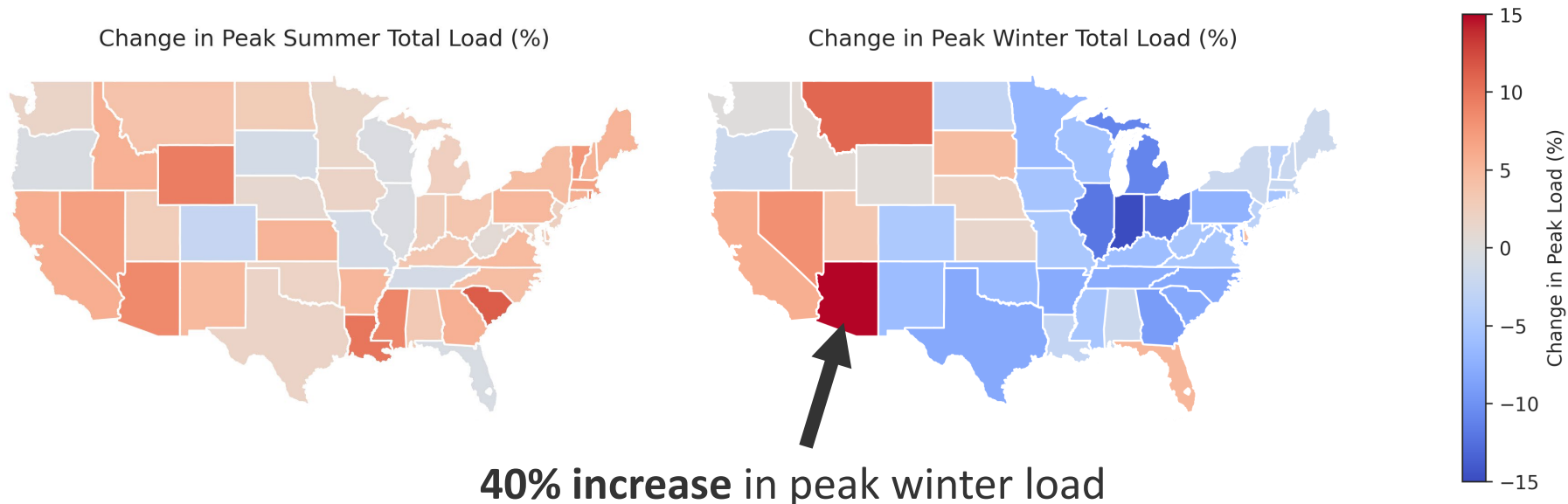
# Impacts of Climate Change on Heat and Cooling Peak Energy Demand



Widespread increases in summer peak AC load and decreases in winter heating load

*this only compares historical load to mid-century weather for a single possible climate scenario*

# Impacts of Climate Change on All-Sector Peak Energy Demand



Less pronounced changes for all-sector peak loads, but some dramatic local changes

*this only compares historical load to mid-century weather for a single possible climate scenario*

# PACES: Power Planning for Alignment of Climate and Energy Systems

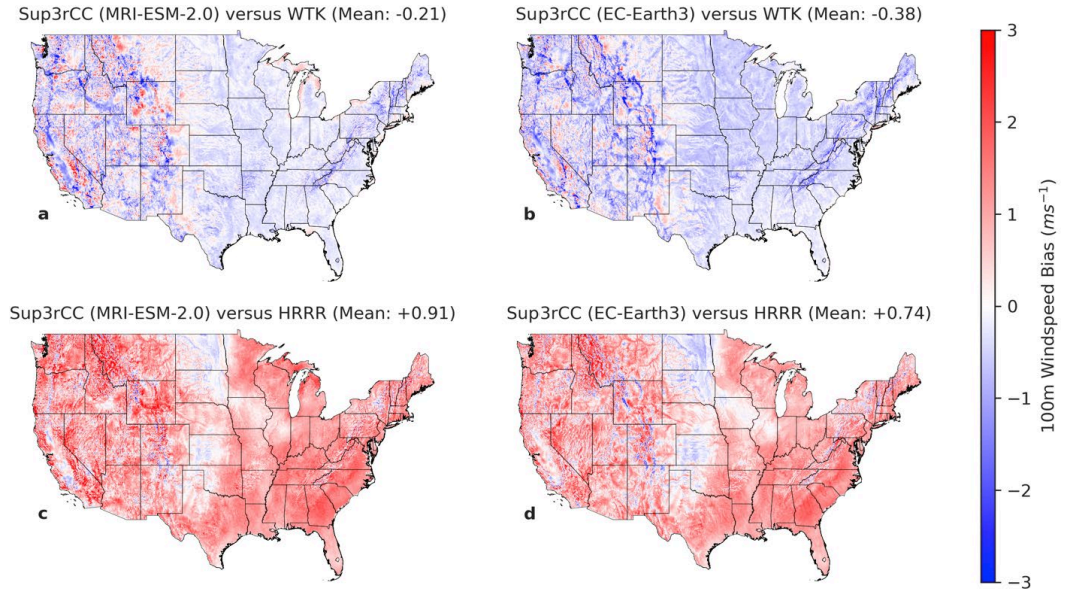


- Downscaling climate projections for power system analysis with multiple methods
  - Generative Machine Learning (Sup3rCC)
  - Numerical Weather Prediction (WRF)
- Exploration of planning strategies under climate change uncertainty
  - Decision making with deep uncertainty (DMDU)
  - Stochastic capacity expansion
- Application to utilities
  - Tennessee Valley Authority
  - Southern Company



# Sup3rCC Known Issues

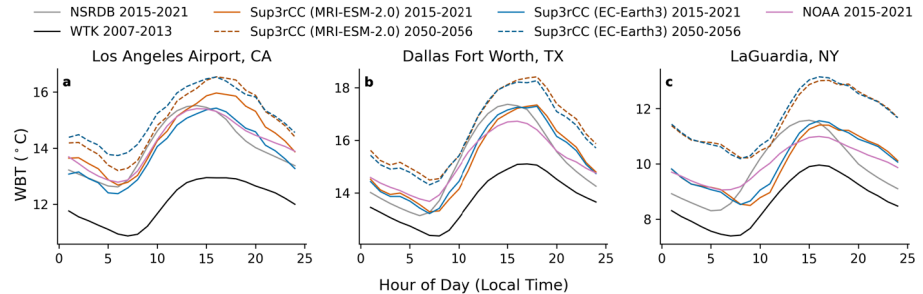
- **Simplistic bias correction**
  - **Linear scalar/adder factors**
  - **7-year comparison**
- Imperfect diurnal cycles
  - Mean wind diurnal cycles are “noisy”
  - Timing of peak temperature can be off



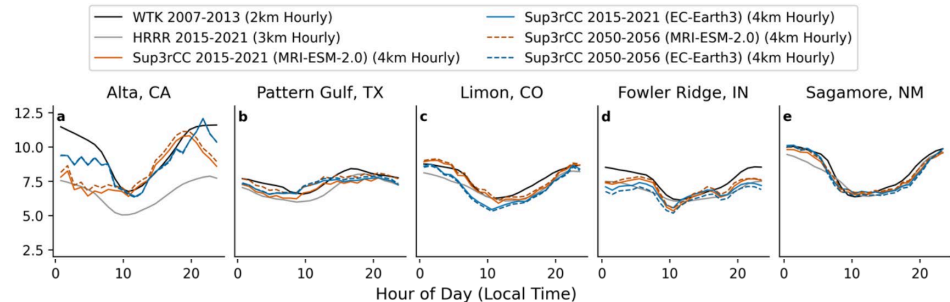
**Supplementary Figure 4. a-d, Maps of mean bias for wind speed.** a, Comparison of Sup3rCC based on MRI-ESM-2.0 to the WTK. b, Comparison of Sup3rCC based on EC-Earth3 to the WTK. c, Comparison of Sup3rCC based on MRI-ESM-2.0 to HRRR. d, Comparison of Sup3rCC based on EC-Earth3 to HRRR. Sup3rCC and HRRR data is from 2015-2021 while WTK is from 2007-2013. Positive bias values indicate Sup3rCC has wind speed greater than the reference dataset.

# Sup3rCC Known Issues

- **Simplistic bias correction**
  - Linear scalar/adder factors
  - 7-year comparison
- **Imperfect diurnal cycles**
  - Mean wind diurnal cycles are “noisy”
  - Timing of peak temperature can be off

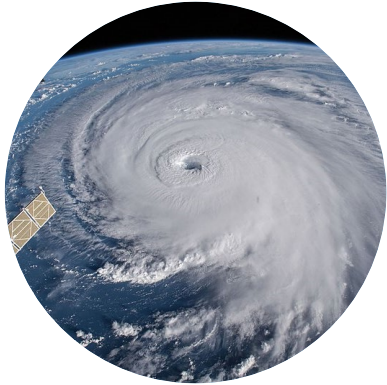


**Supplementary Figure 17. a-c, Mean diurnal cycles for modeled versus observed temperature.** Plots compare gridded datasets (e.g., NSRDB, WTK, and Sup3rCC) to historical observation from NOAA at three large airports.



**Supplementary Figure 14. a-y, Mean diurnal cycles for wind speed.** Data is included for 25 large wind energy facilities in the contiguous United States comparing the WTK, HRRR, and Sup3rCC.

# Gaps!



## Hurricanes

How does climate change affect hurricanes? How can we simulate hurricanes for power system impacts?



## Wildfires

How does climate change affect wildfires? How can we simulate wildfires for power system impacts?



## Migration

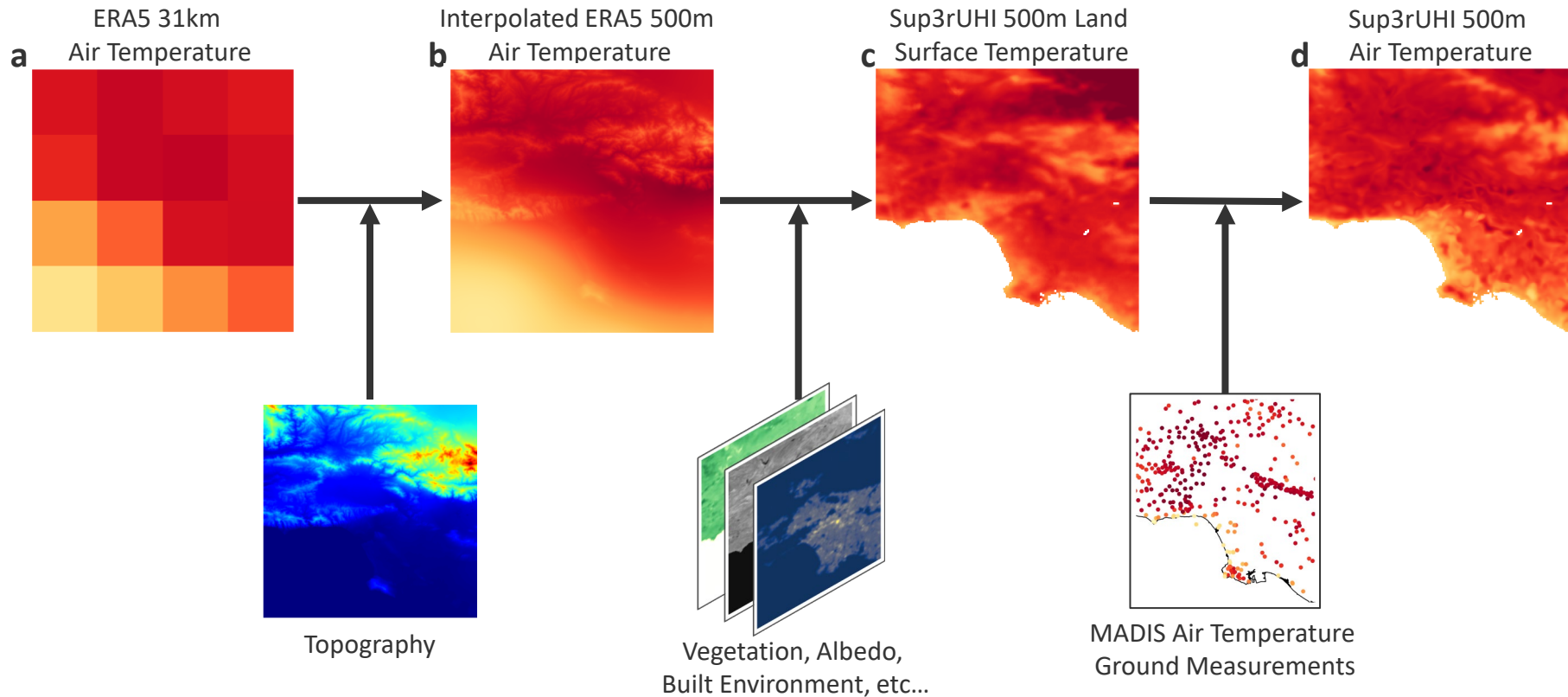
How might climate change drive human migration? How does this stress energy infrastructure and energy demand?



## Urban Effects

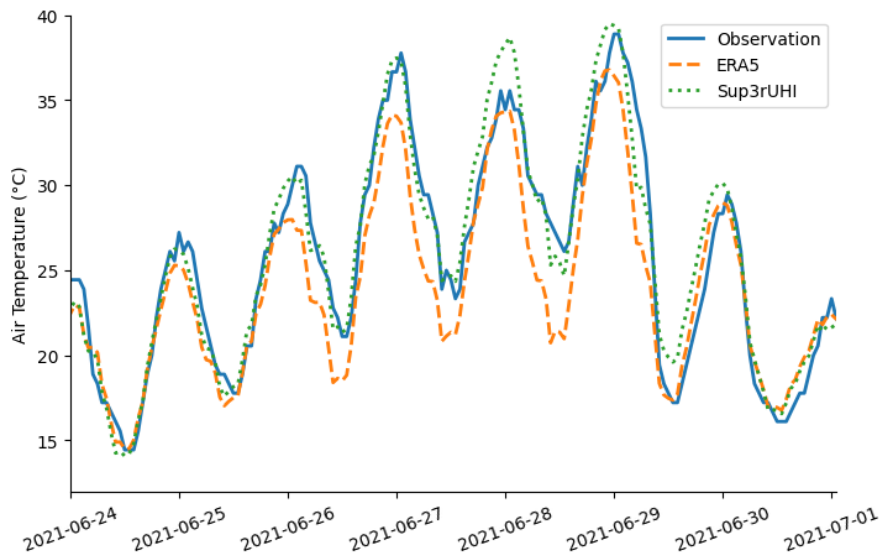
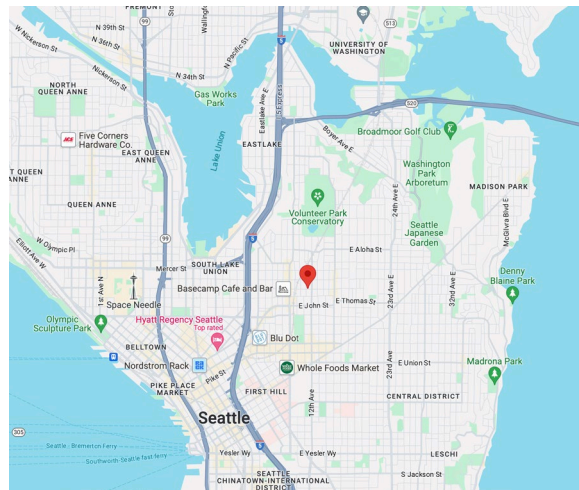
How does the urban microclimate exacerbate impacts from climate change?

# Sup3r Resolution for Urban Heat Island (Sup3rUHI)



# Validation – Seattle

- Sup3rUHI yields much better data on 2021 Seattle heatwave in the urban core than reanalysis, as measured in Capitol Hill
- ERA5 underestimates the heat wave's peak temperature by more than 2°C
- Sup3rUHI overestimates by 0.6°C
- **Supports the use of this model to study unprecedented heat waves in urban microclimates**




# Validation – LA Albedo

- Applying the albedo changes reported in [Taha 2024](#) to Pacoima, CA we see an average change of  $-0.13^{\circ}\text{C}$  with a maximum change of  $-0.75^{\circ}\text{C}$ 
  - This corresponds to a  $T_{\text{air}}$  reduction of  $0.08^{\circ}\text{C}$  per 0.1 increase in albedo
    - Comparable to previous cooling estimates previously reported in the literature
- **Supports the use of this model to estimate effects of albedo manipulation**

**Table 2.** Comparison of reported  $T_{\text{air}}$  reductions per 0.1 increase in albedo from previous modeling studies.

Study	Spatial scale	$T_{\text{air}}$ reduction per 0.1 increase in albedo ( $^{\circ}\text{C}$ )	$T_{\text{air}}$ height (m)
This study	Neighborhood-scale	0.11, 0.10 <sup>a</sup>	1.6, 3 <sup>a</sup>
Taleghani <i>et al</i> [33]	Neighborhood-scale	0.09	1.5
Taha [31, 87]	City-scale	0.17–0.2	2
Mohegh <i>et al</i> [24]	City-scale	0.05–0.22	2
Millstein and Levinson, [88]	Hypothetical hotplate (~200 m)	0.05	2

# Efficacy of Albedo Modification on Residential Cooling Demand in Los Angeles

- Increase in cooling demand through midcentury: +23%
- Road albedo modifications: -2%
- Large-scale neighborhood albedo modifications: -6%
- Cool roof modification: -11%
- Cool roof + large scale neighborhood albedo modifications: -17%
- Based on Sup3rCC data from 2015-2024 versus 2050-2059
- Los Angeles residential building stock represented using  ResStock

# NREL Contributors to Climate Change Data and Climate Impacts Analysis



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Jordan Cox



Ariel Miara



Andrew Glaws



Ryan King



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Guilherme  
Pimenta Castelao



# Thank you

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<https://www.nature.com/articles/s41560-024-01507-9>

<https://data.openei.org/submissions/5839>

This work was authored by the National Renewable Energy Laboratory, operated by Alliance for Sustainable Energy, LLC, for the U.S. Department of Energy (DOE) under Contract No. DE-AC36-08GO28308. Funding provided by the DOE Grid Deployment Office (GDO), the DOE Advanced Scientific Computing Research (ASCR) program, the DOE Solar Energy Technologies Office (SETO), and the Laboratory Directed Research and Development (LDRD) program at the National Renewable Energy Laboratory. The research was performed using computational resources sponsored by the DOE Office of Energy Efficiency and Renewable Energy and located at the National Renewable Energy Laboratory. The views expressed in the article do not necessarily represent the views of the DOE or the U.S. Government. The U.S. Government retains and the publisher, by accepting the article for publication, acknowledges that the U.S. Government retains a nonexclusive, paid-up, irrevocable, worldwide license to publish or reproduce the published form of this work, or allow others to do so, for U.S. Government purposes.

