## A Planetary-Scale Data–Model Integration Framework to Resolve Urban Impacts Across Scales

TC (Tirthankar Chakraborty)

Earth Scientist,

Earth System Modeling Group,

Atmospheric, Climate & Earth Sciences Division,

Pacific Northwest National Lab



## A Planetary-Scale Data–Model Integration Framework to Resolve Urban Impacts Across Scales and Examine Weather Extremes over Coastal U.S. Cities

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### **Presentation Overview**

# BACKGROUND

# MOTIVATION

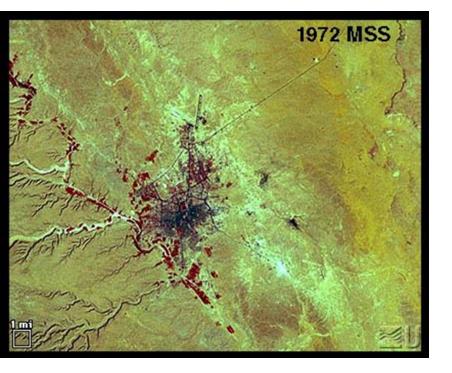
# **RESEARCH PLANS**

# BACKGROUND

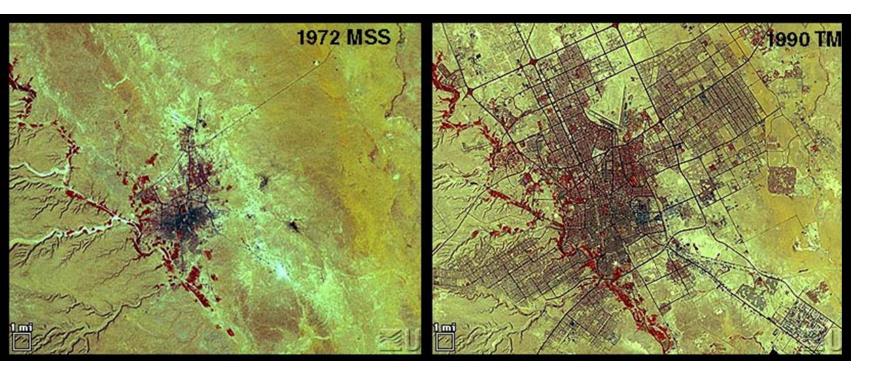
# MOTIVATION

## **RESEARCH PLANS**

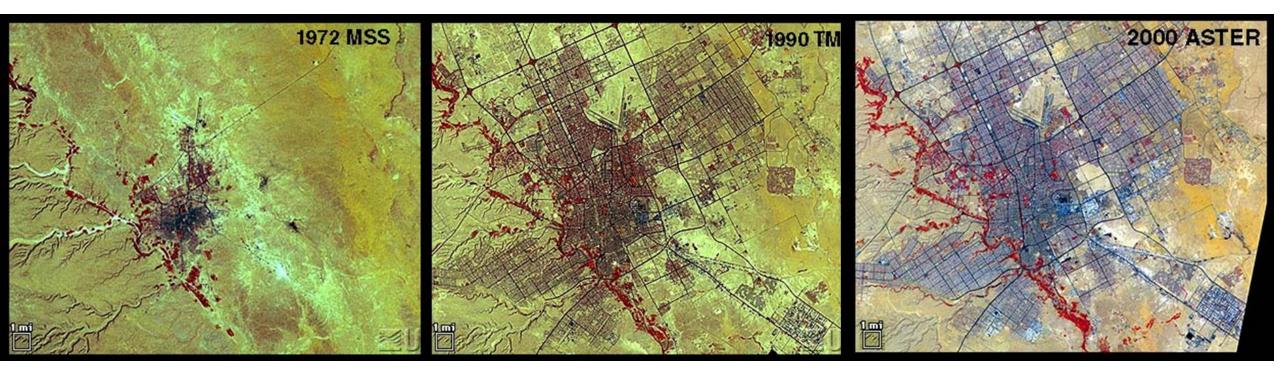
Replacement of natural land surfaces with built-up structures



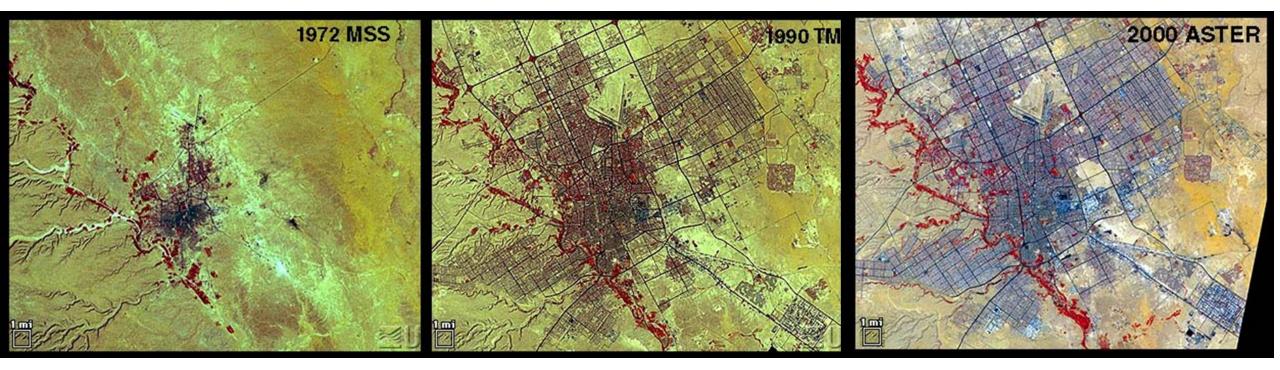
Replacement of natural land surfaces with built-up structures



Replacement of natural land surfaces with built-up structures



- Replacement of natural land surfaces with built-up structures
- Broadly, any change in natural land cover due to urban planning (e.g. addition of urban green spaces and transition from native to non-native vegetation), constitutes urbanization



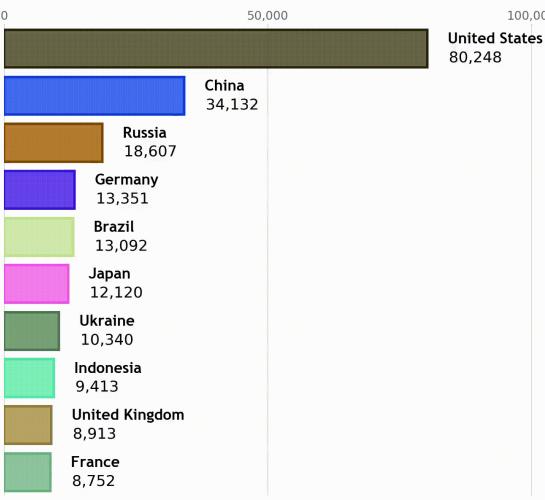
### **Rapid historical urban expansion**

100,000

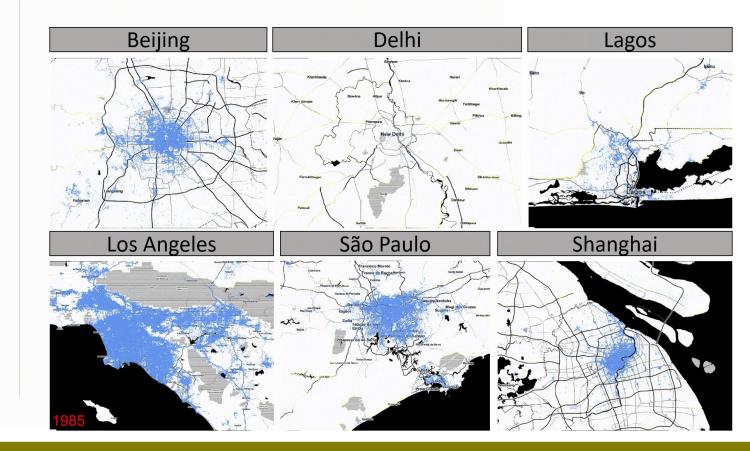
#### Countries with the highest built-up area

1992

Area (in sq. km)



**Urban areas have grown** tremendously in the last few decades



### **Urban impacts on weather and climate**

Increase in urban extent + changes in surface properties + anthropogenic activities

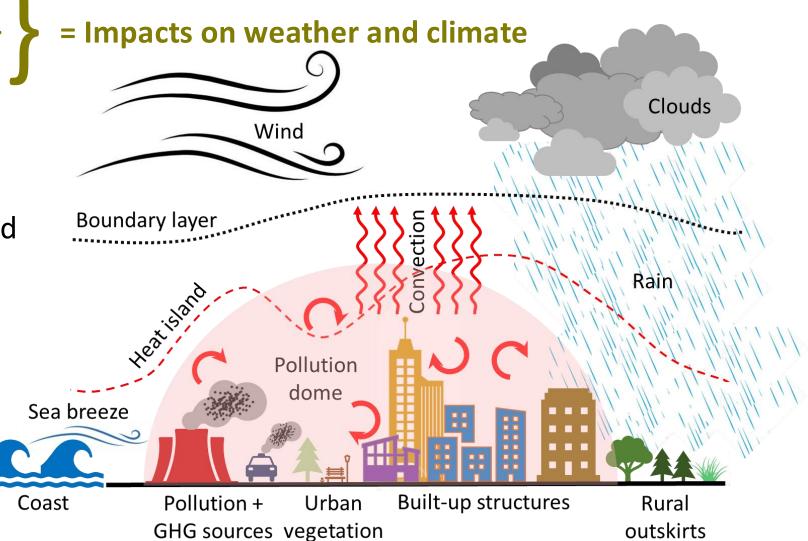
= Impacts on weather and climate

#### Modified from Qian, Chakraborty et al., 2022

### Urban impacts on weather and climate

Increase in urban extent + changes in surface properties + anthropogenic activities

- Urbanization causes localized warming (urban heat island), drying (urban dry island), cloud formation and precipitation over and downwind of cities, enhanced air pollution, etc.
- Urban growth and densification may lead to further feedbacks to the atmosphere



#### Modified from Qian, Chakraborty *et al.*, 2022

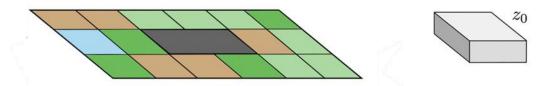
# BACKGROUND

# MOTIVATION

# **RESEARCH PLANS**

### **Urban representation in Earth system models**

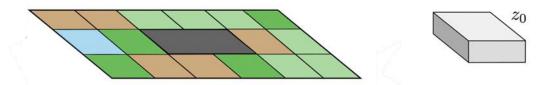
#### Mostly slabs or no urban



- Urban areas are rarely represented in global models
- We assume that urban land fraction is small and won't impact broader climate
- The few global models with urban representation are too simple

### **Urban representation in E3SM**

#### Mostly slabs or no urban

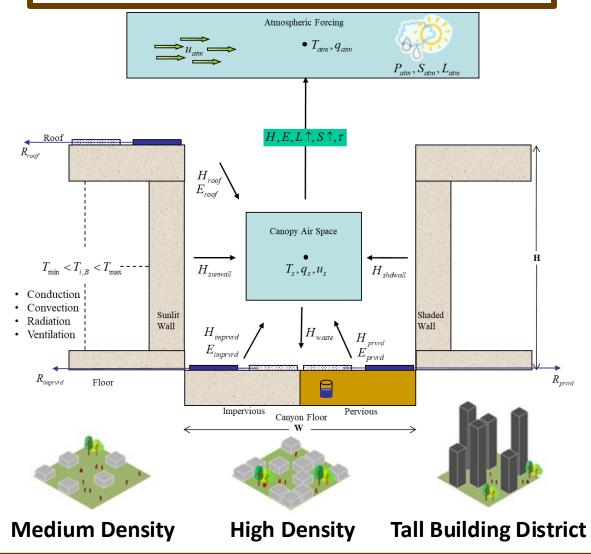


- Urban areas are rarely represented in global models
- We assume that urban land fraction is small and won't impact broader climate
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Hertwig et al., 2020

E3SM: DOE's Energy Exascale Earth System Model

#### **Urban canopy structure in E3SM**

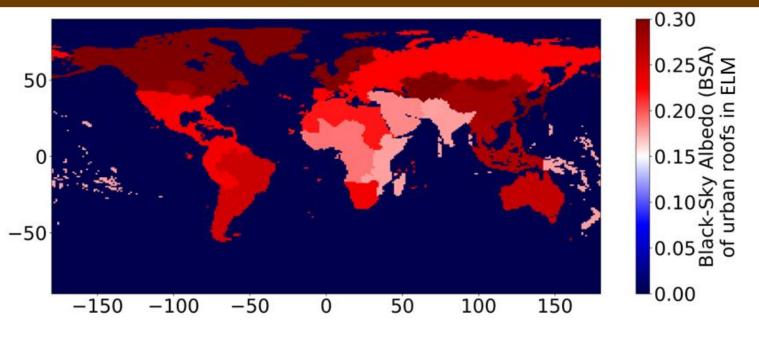


#### Modified from Ching *et al.*, 2014

### Poor surface constraints for urban areas

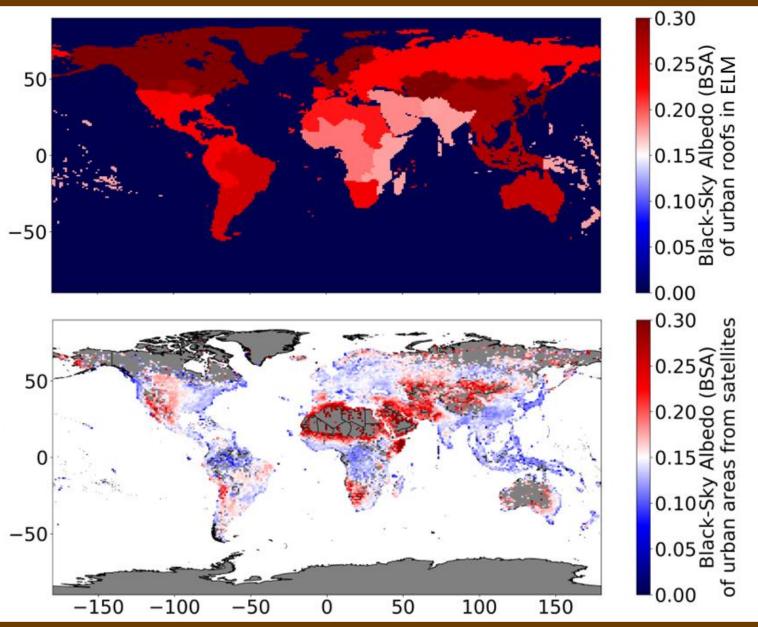
oof

urban



In E3SM, the world is divided into 33 regions, each with unique values for urban radiative, thermal, and morphological parameters

### Poor surface constraints for urban areas

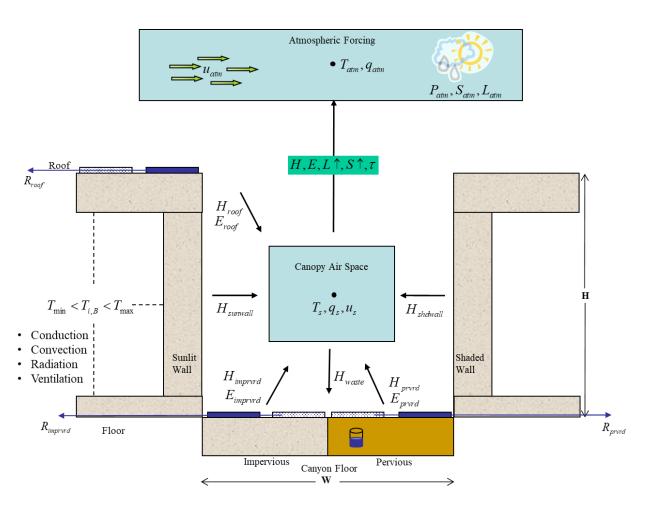


In E3SM, the world is divided into 33 regions, each with unique values for urban radiative, thermal, and morphological parameters

- Actual variability, as seen from satellites, is much higher than what is represented in the model.
- Would strongly impact simulations of cross-sample variability in urban climate

### No vegetation within urban areas

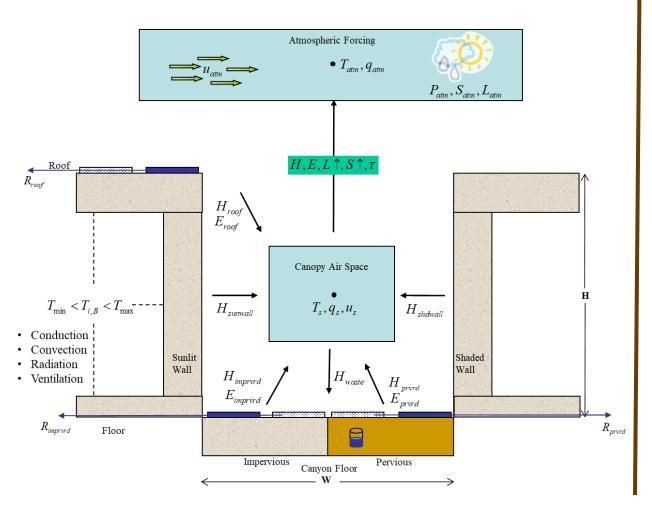
#### **Urban canopy structure in E3SM**



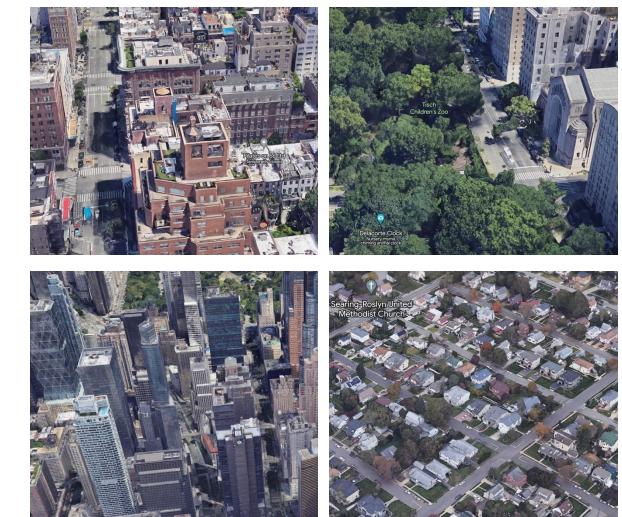
#### Modified from Ching et al., 2014

### No vegetation within urban areas

#### **Urban canopy structure in E3SM**



#### Real cities?

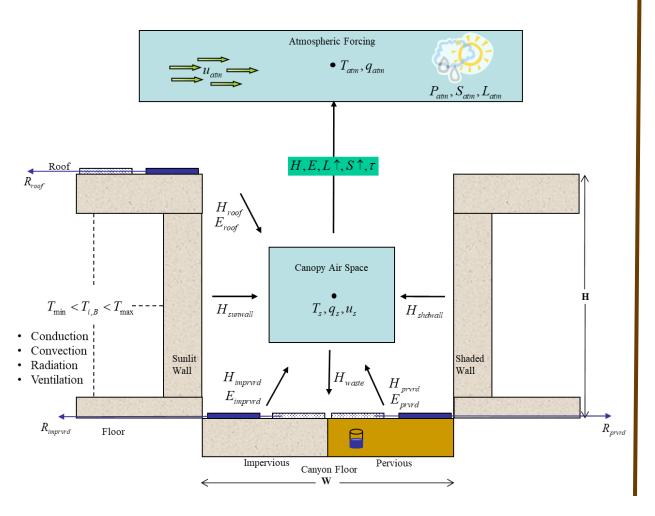


#### Modified from Ching et al., 2014

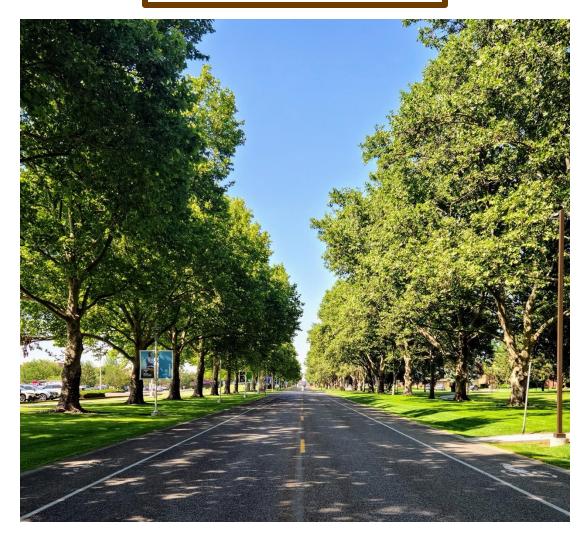
#### Source: Google

### No vegetation within urban areas

#### **Urban canopy structure in E3SM**



#### Real cities?

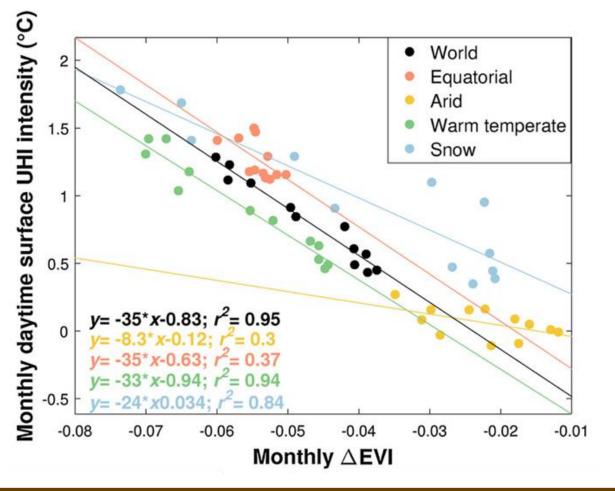


#### Modified from *Ching et al.*, 2014

#### Source: My Instagram

### Importance of urban vegetation

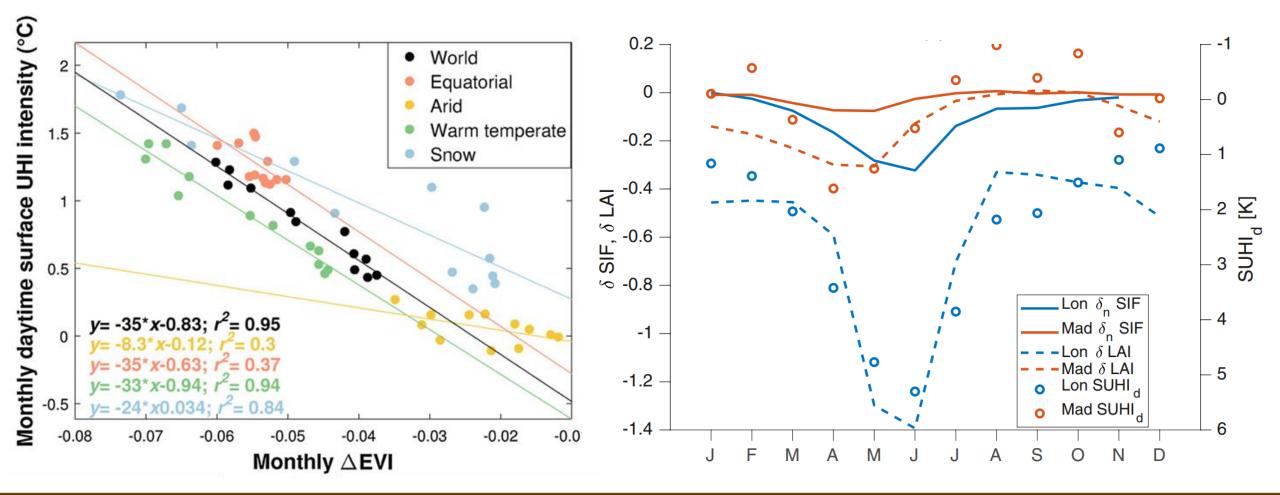
- Urban vegetation modifies weather and climate
- Urban and rural vegetation are not identical, both in amount and properties



Chakraborty and Lee, 2019

### Importance of urban vegetation

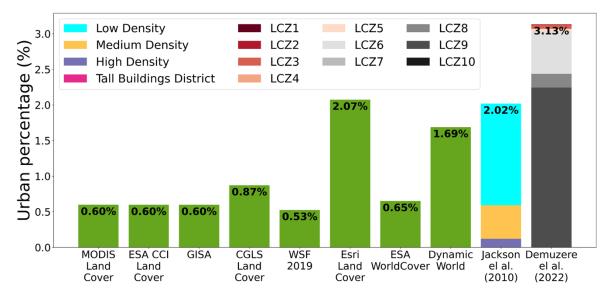
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Chakraborty and Lee, 2019

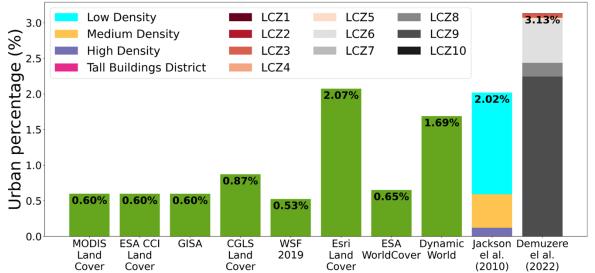
Paschalis et al., 2021

### Static urban representation in E3SM



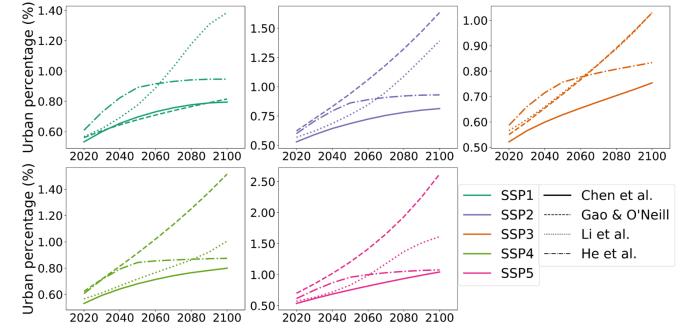
- Current urban extent in E3SM Land Model (ELM) is based on a circa 2001 dataset
- Uses population-based urban definitions
- Low density class data are not used
- Static for historical and future simulations

### Static urban representation in E3SM



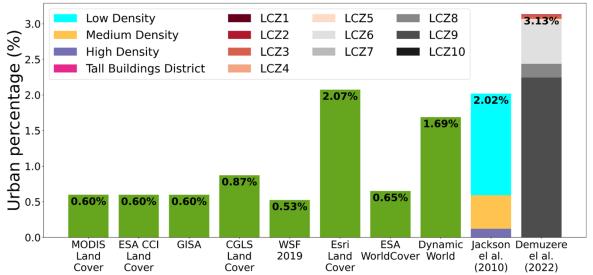
- Significant future urban expansion
- Large uncertainties across datasets

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#### Chakraborty et al., Under Review

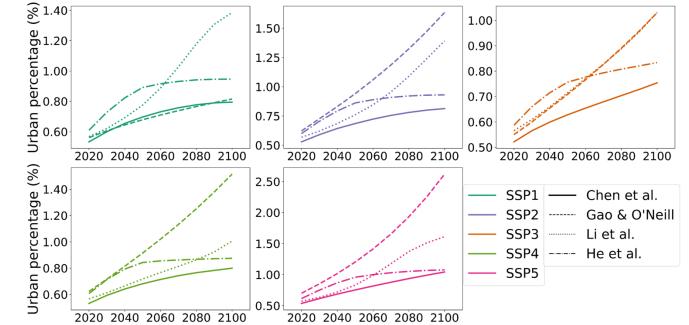
### Static urban representation in E3SM



- Significant future urban expansion
- Large uncertainties across datasets
- Urban properties will also change

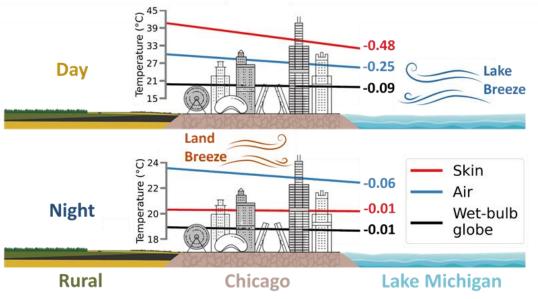
Urban Evolution = Urban expansion + change in urban properties over time

- Current urban extent in E3SM Land Model (ELM) is based on a circa 2001 dataset
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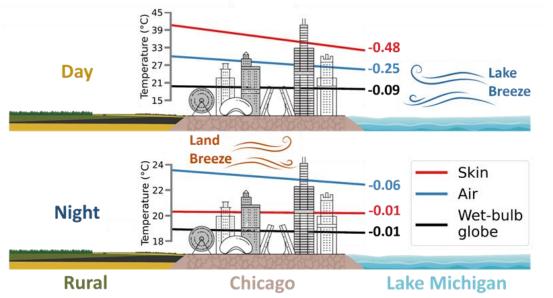
Chakraborty & Qian, Accepted; Chakraborty et al., Under Review

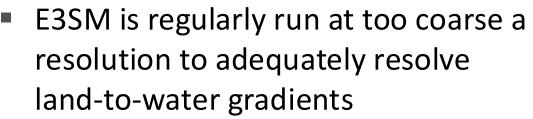
## **Resolving coastal processes?**



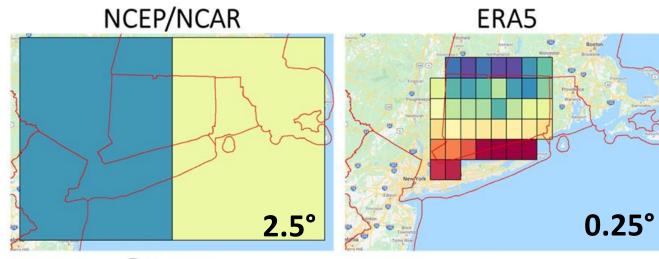
- Majority of global urban population lives in coastal cities
- Urban-coastal interactions are complex
- Large spatial gradients expected along coastlines, as well as hard-to-simulate extreme weather phenomena

## **Resolving coastal processes?**





- For coastal cities, spatial variability of surface properties not represented
- Model does not care where urban area is within grid



0.008°

#### Daymet



#### Mean air temperature for Connecticut: NCEP/NCAR = -21.1 °C ERA5 = -19.1 °C Daymet = -19.5 °C

-14

Chakraborty et al., 2023

# BACKGROUND

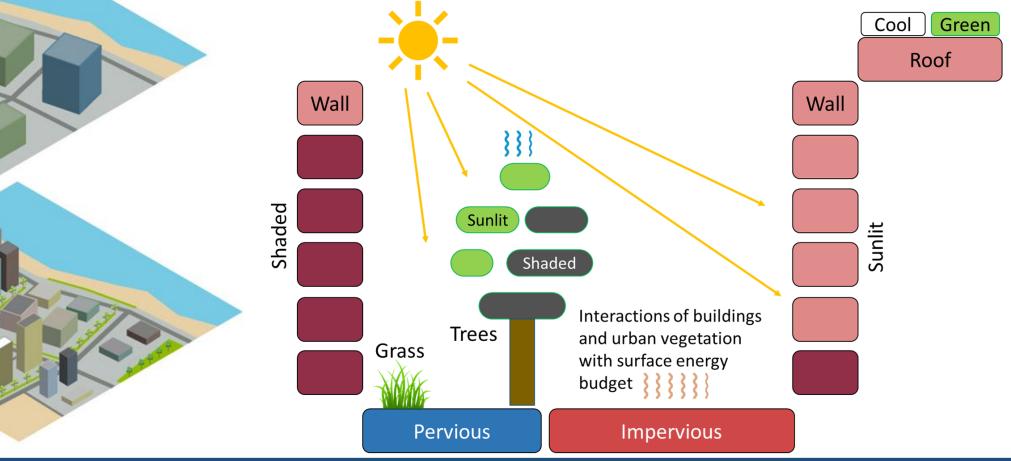
# MOTIVATION

# **RESEARCH PLANS**

### New urban parameterization for E3SM

Objective 1: Develop a new urban parameterization for E3SM

- Explicitly representing urban vegetation and its interactions with climate
- Global spatially continuous urban surface dataset

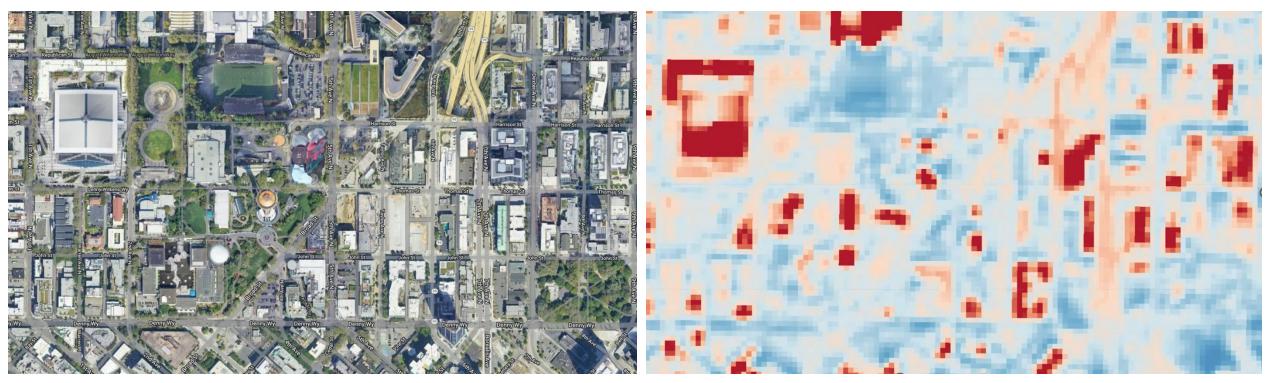


### Generating a global 1 km urban dataset



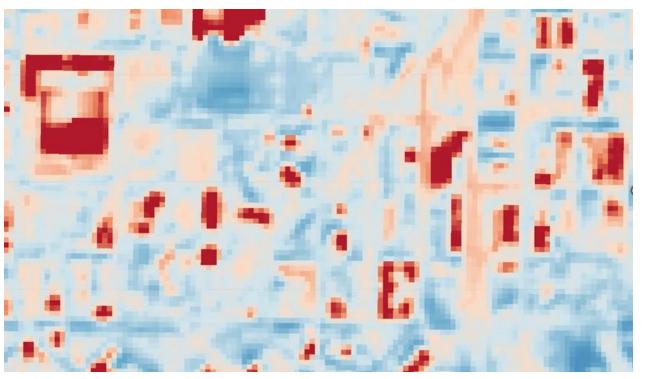
 Leverage and develop high resolution satellitederived global products

### Generating a global 1 km urban dataset



 Leverage and develop high resolution satellitederived global products

### Generating a global 1 km urban dataset



- Leverage and develop high resolution satellitederived global products
- Combine products with building and road datasets
- Extract roof, wall, and road parameters for all 1 km urban pixel in the world



#### Wu et al., In Prep

#### Source: Microsoft

### Some initial results

#### **E3SM urban surface constraints**

Radiative	Roof   Impervious Road   Pervious Road   Wall Emissivity
	Roof   Impervious Road   Pervious Road   Wall Albedo
Morphological	Roof Height
	Canyon Height-to-width Ratio
	Roof Fraction   Pervious Road Fraction
	Urban Percentage
Thermal	Numbers of impervious road layers
	Roof   Wall Thickness
	Minimum   Maximum Interior Building Temperature
	Roof   Impervious Road   Wall Thermal Conductivity
	Roof   Impervious Road   Wall Volumetric Heat Capacity

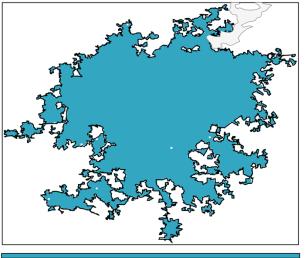
#### Cheng et al., In Prep

### Some initial results

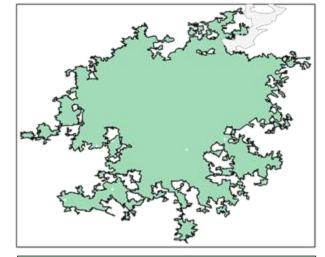
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#### **Default** roof fraction



#### **Default** roof albedo



0.55

0.14

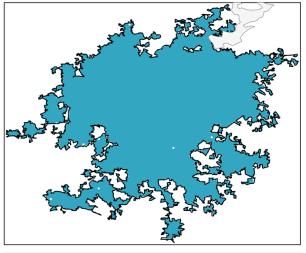
#### Cheng et al., In Prep

### Some initial results

#### E3SM urban surface constraints

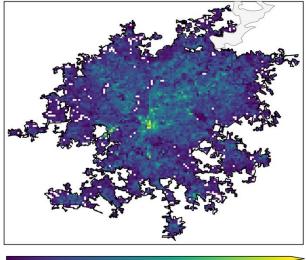
Radiative	Roof   Impervious Road   Pervious Road   Wall Emissivity
	Roof   Impervious Road   Pervious Road   Wall Albedo
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#### **Default** roof fraction



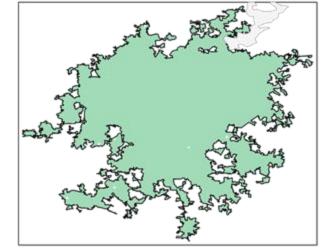
Updated roof fraction

0.55



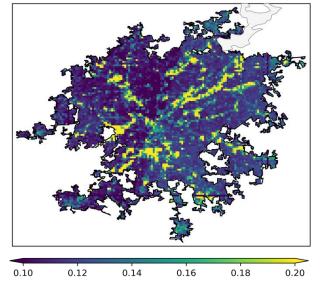
0.00 0.05 0.10 0.15 0.20 0.25 0.30 0.35 0.40

#### Default roof albedo



Updated roof albedo

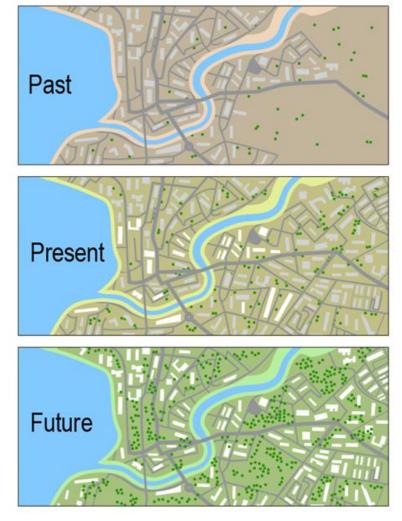
0.14



#### Cheng et al., In Prep

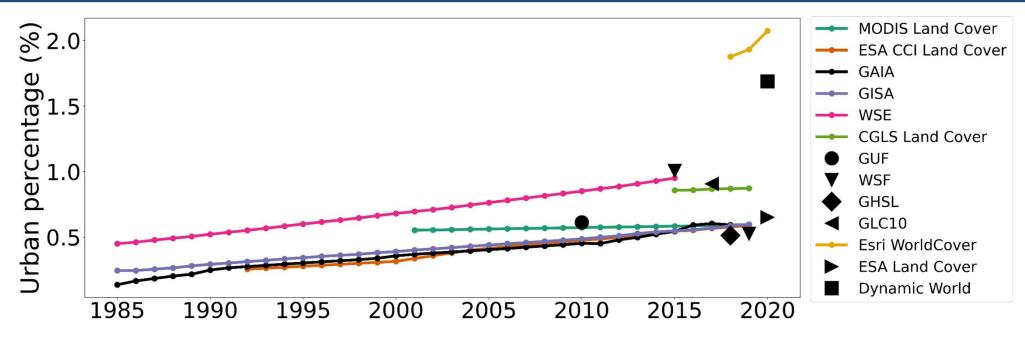
### **Examining urban evolution and its impacts**

Objective 2: Isolate the role of urban evolution on surface climate from continental to coastal scales



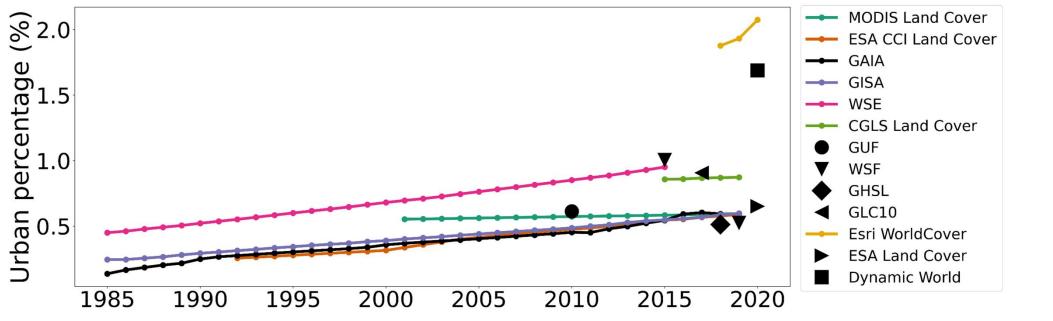
- Generate spatiotemporally varying estimates of urban evolution
- Develop high resolution forcing over coastal areas across scenarios (historical to future climate projections)
- Land only simulations to examine urban impacts on surface climate across scales

## Developing temporally varying urban parameters

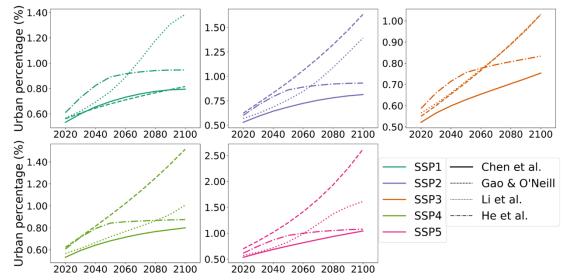


 Time series forecasting models as a function of historical properties

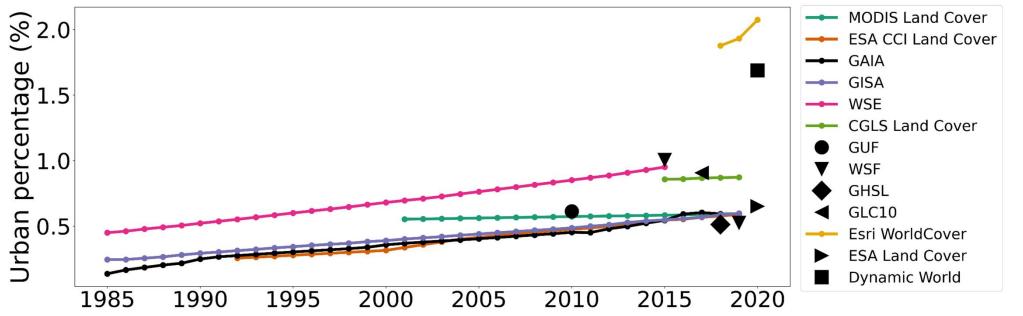
## Developing temporally varying urban parameters



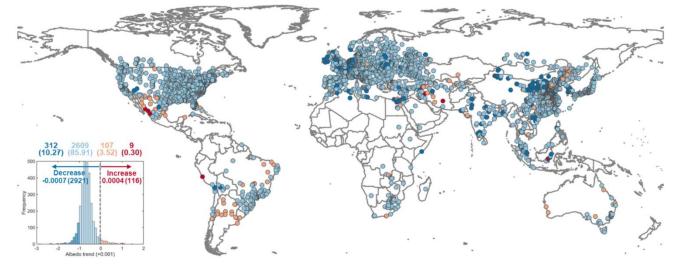
- Time series forecasting models as a function of historical properties
- Constrained by future projections by scenarios



## Developing temporally varying urban parameters



- Time series forecasting models as a function of historical properties
- Constrained by future projections by scenarios
- Change in urban properties and extent over space and time

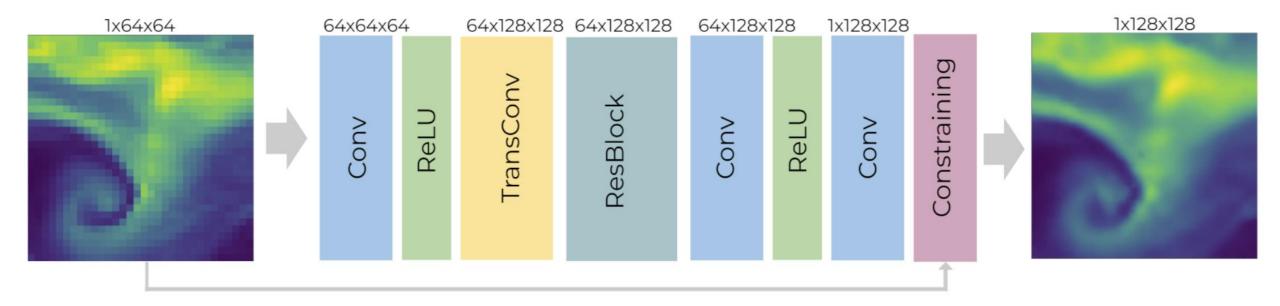


Albedo trend (× 0.001) • < -1

Wu et al., 2024

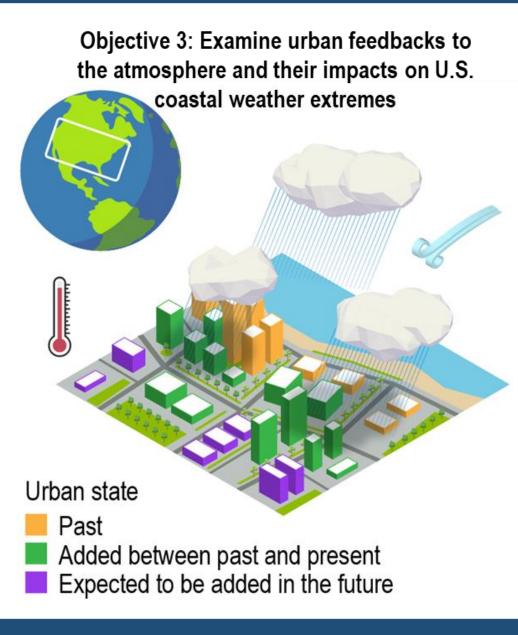
#### Chakraborty et al., Under Review

## Downscaled forcing datasets along U.S. coastline



- Physics-constrained convolutional neural networks to capture coastal gradients
- Constrained by mass and energy budget of the entire grid (based on coarser future climate projections)

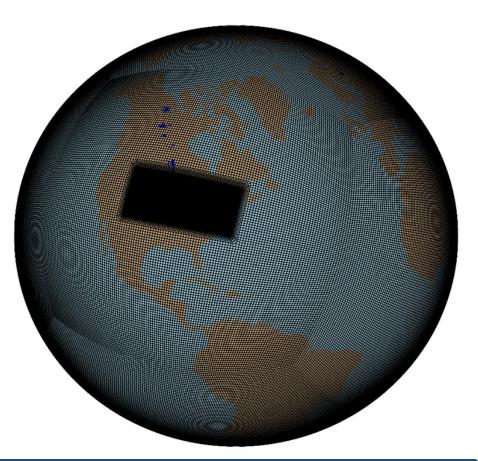
## Examining feedbacks and studying extreme events



 Examining feedback from CONUS to coastal scales

 Urban impacts on coastal weather extremes

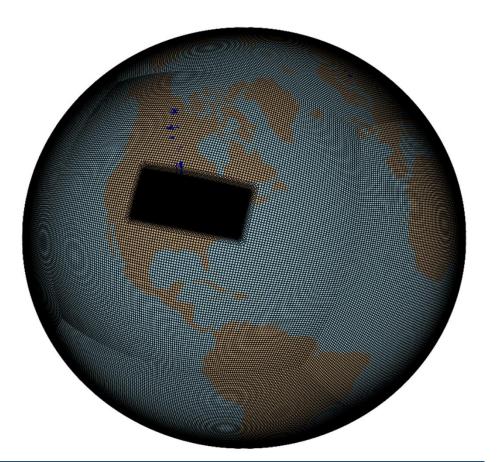
## Running E3SM with regionally refined grids at 1 km



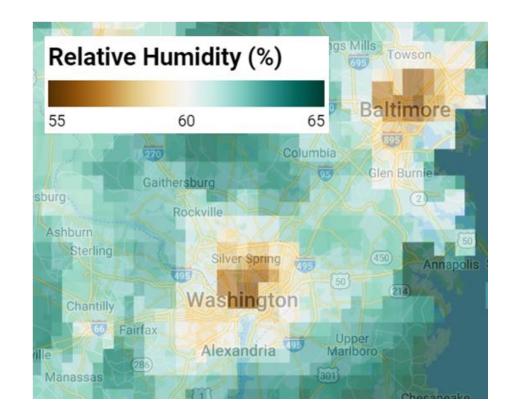
Regionally refined Simple Cloud-Resolving E3SM Atmosphere Model (SCREAM) at ~3.25 km resolution

Liu *et al.*, 2023

## Running E3SM with regionally refined grids at 1 km



Regionally refined Simple Cloud-Resolving E3SM Atmosphere Model (SCREAM) at ~3.25 km resolution



- Isolate specific extreme events from coarser coupled simulations (drought, heatwave, storms)
- Run perturbation simulations with different urban scenarios

#### Chakraborty et al., In Prep

#### Liu *et al.*, 2023

**Multiple critical** collaborations across labs, other institutions, and countries

> **Overall schematic of** proposed research

Objective 1: Develop a new urban

Objective 2: Isolate the role of urban evolution on

Objective 3: Examine urban feedbacks to surface climate from continental to coastal scales parameterization for E3SM the atmosphere and their impacts on U.S. coastal weather extremes Past Present Urban state Future Past Added between past and present Expected to be added in the future Data sharing & model **Co-developing** Calibrating model benchmarking future urban improvements to isolate evolution coastal-urban interactions CROCUS Community Research on Climate & Urban Science scenarios GREAT LAKES MODELING Z Across Systems and Scales Thank you! 

Integration with other **DOE-funded projects**