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COMMUNITY EARTH
SYSTEM MODEL



The Grainger College of Engineering
Civil & Environmental Engineering

Advancing urban representation in CESM to understand urban climate at large scales

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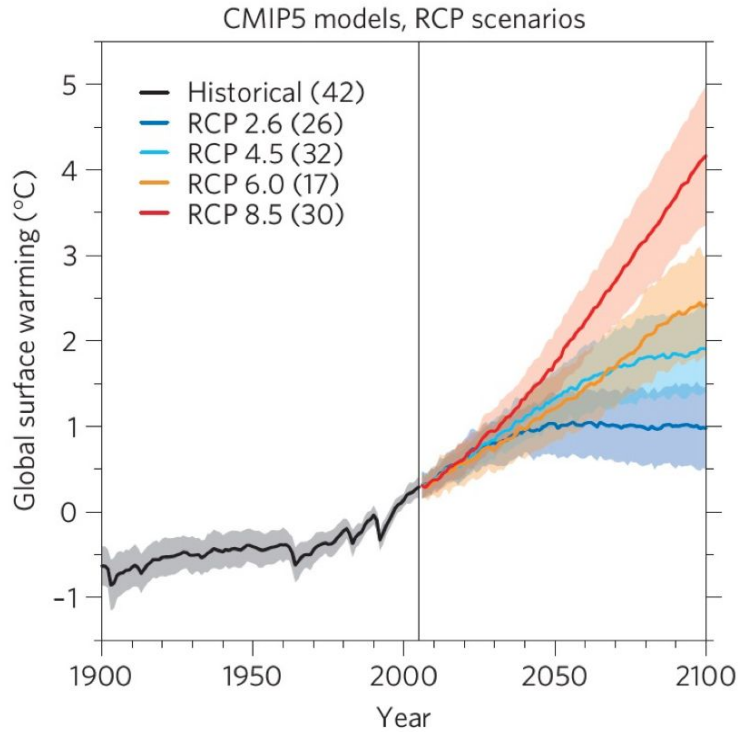
work with/led by:

Group members: Yifan Cheng, Bowen Fang, Laura Gray, Cathy Li, Joyce Yang, Yiwen Zhang, Zhonghua Zheng,
&

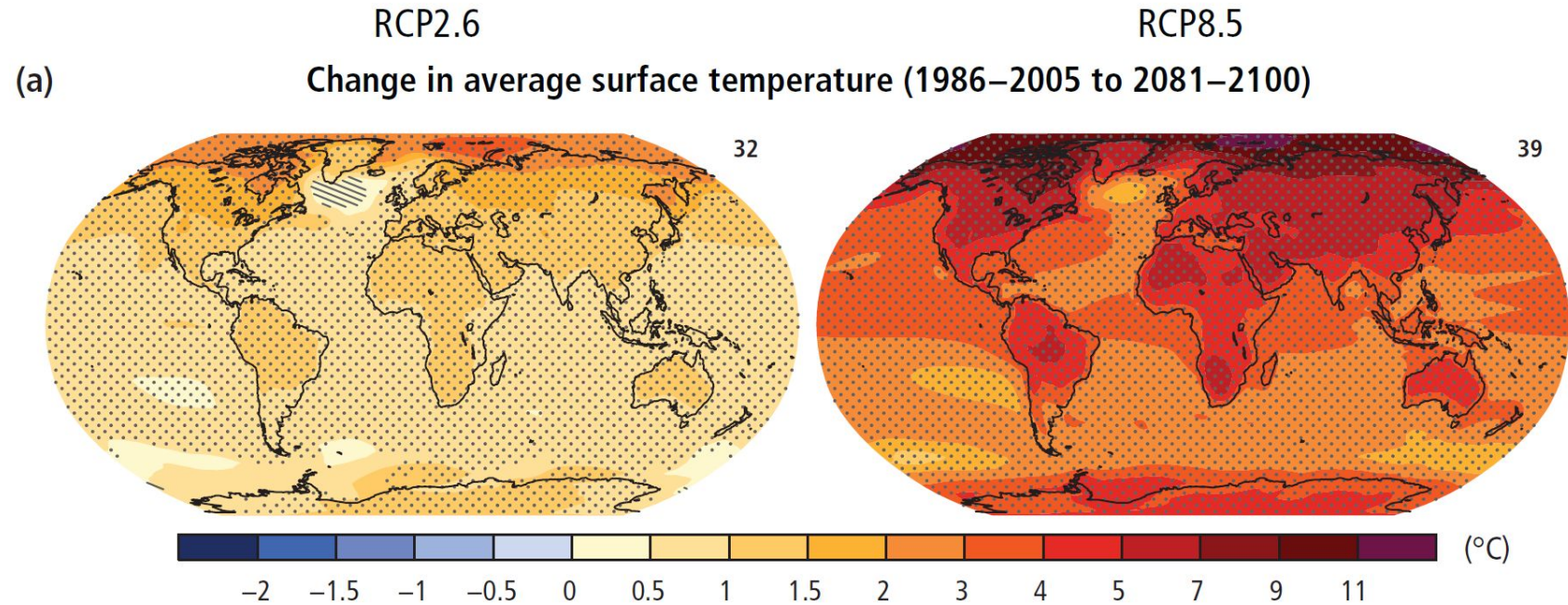
Collaborators: Keith Oleson (NCAR), Keer Zhang (Yale), Elie Bou-Zeid (Princeton), Bill Sacks (NCAR), Peter Lawrence (NCAR),
Dave Lawrence (NCAR), Scott Krayenhoff (U. Guelph), Pierre Gentine (Columbia), and many more

CESM Land Model Working Group Winter Meeting 2024

Global “surface” climate projections



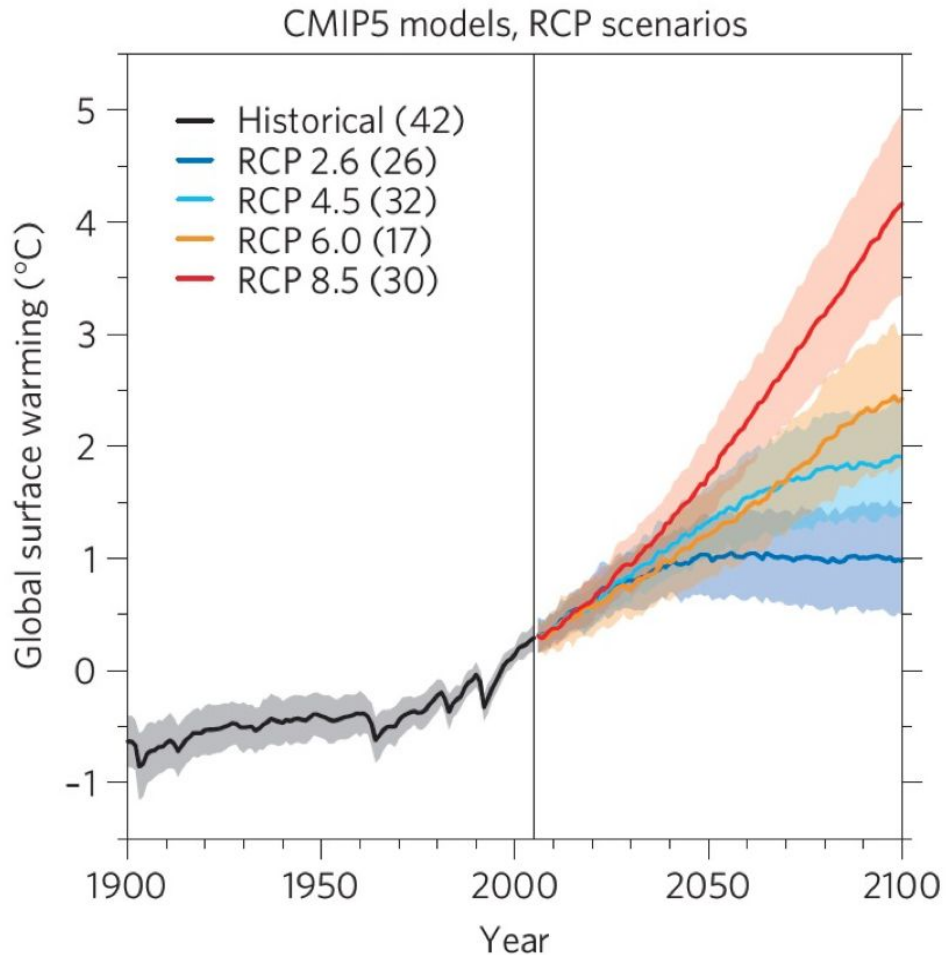
Knutti & Sedlacek, *Nat. Clim. Change* 2013



IPCC AR5

Are these modeled results “good” for urban areas?

Nearly all large-scale Earth system models lack an urban representation!

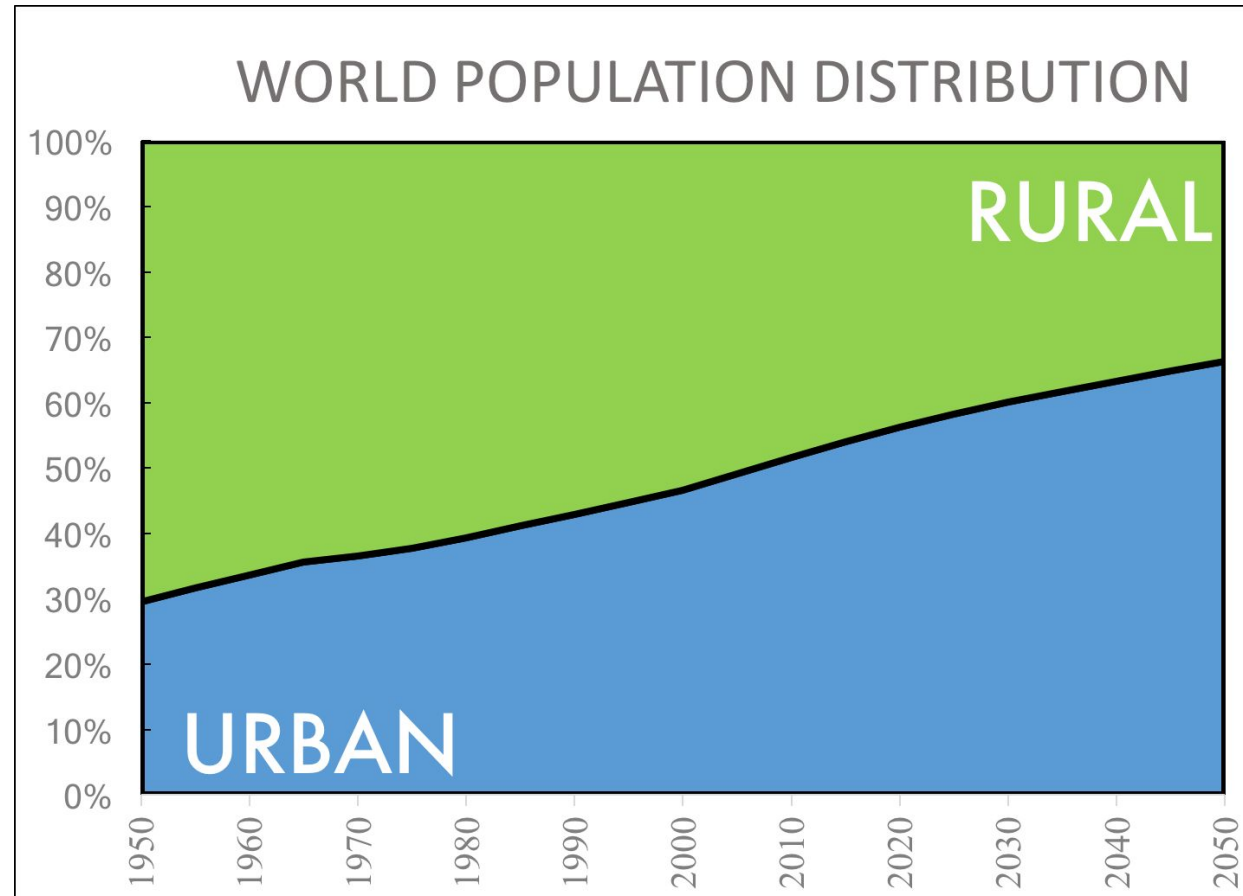


Traditional projections of “global surface climate” are essentially global “non-urban” surface climate.

Human-perceived impacts

2-3 %
of Earth's land surface

>50 %
of world population





Climate-driven urban risks will worsen in the future because of ...

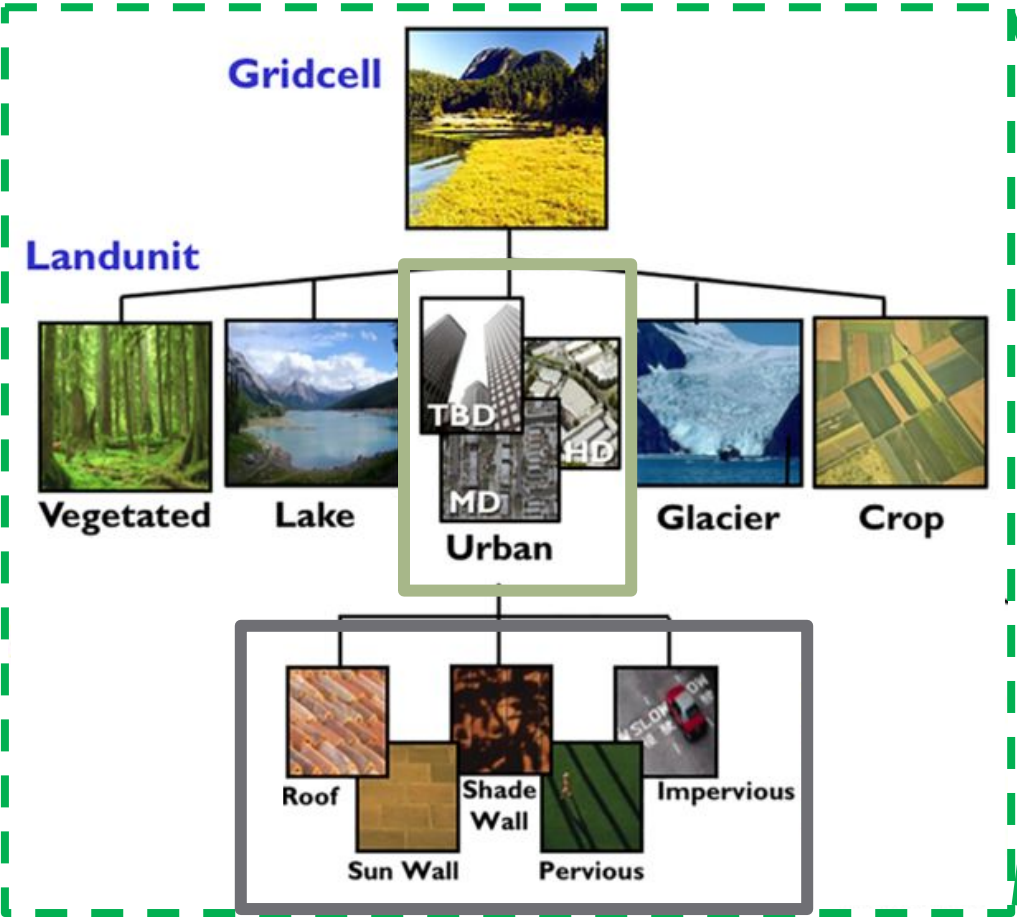


Urban development

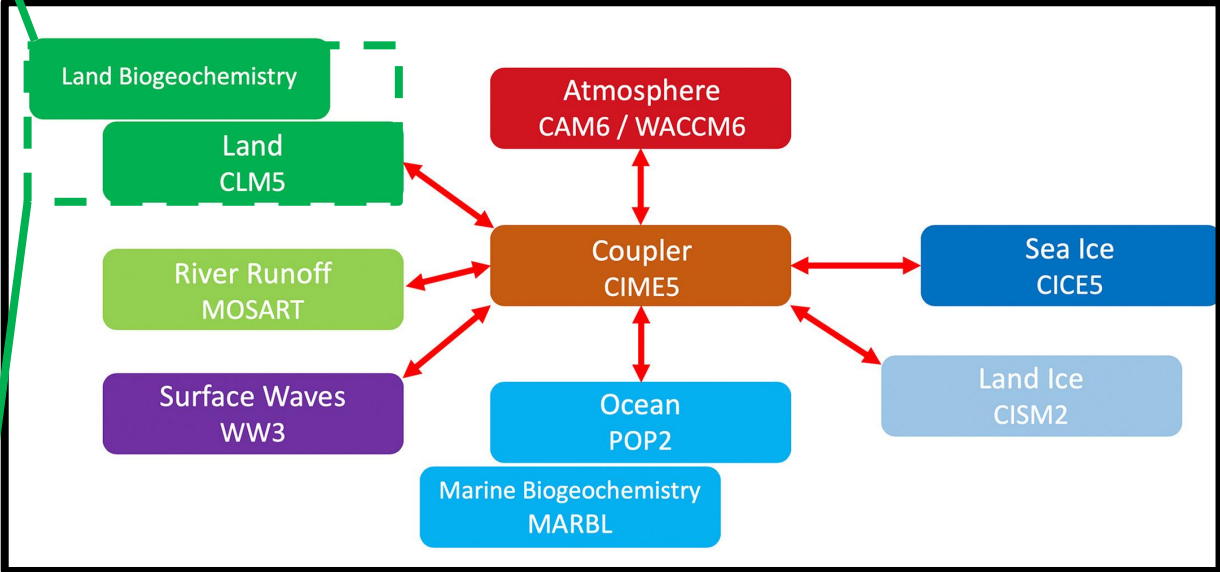


Climate change

Urban representation in CESM

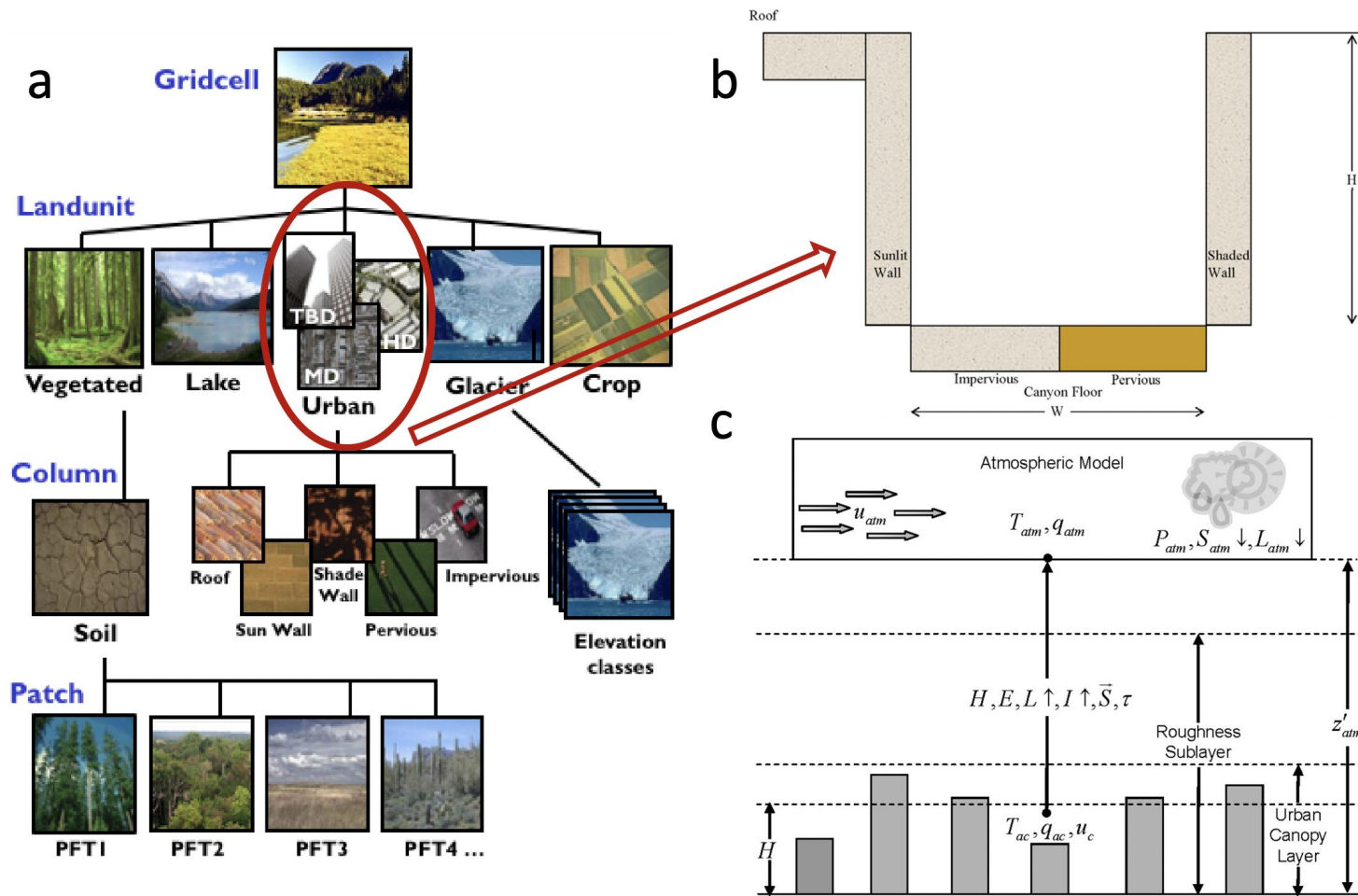


Community Earth System Model (CESM)



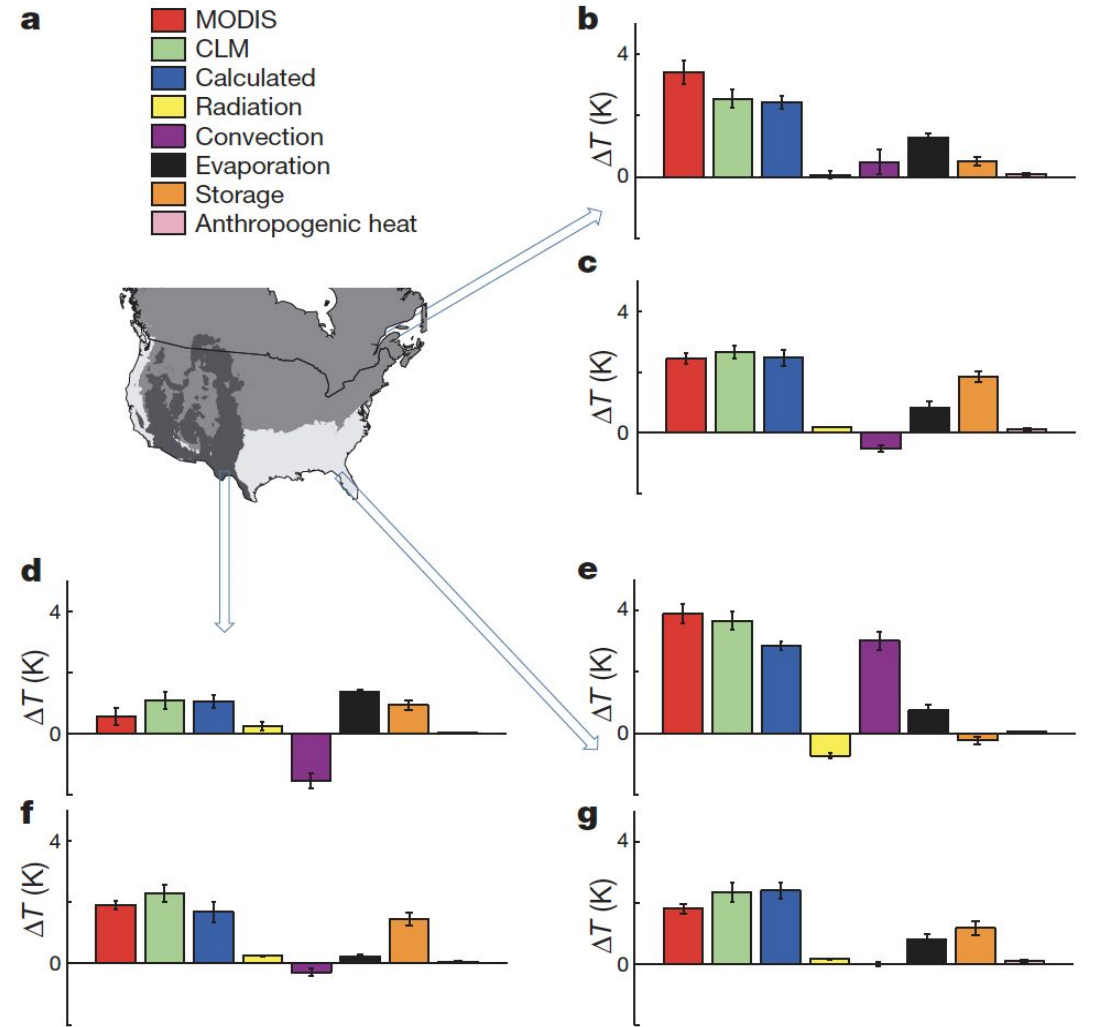
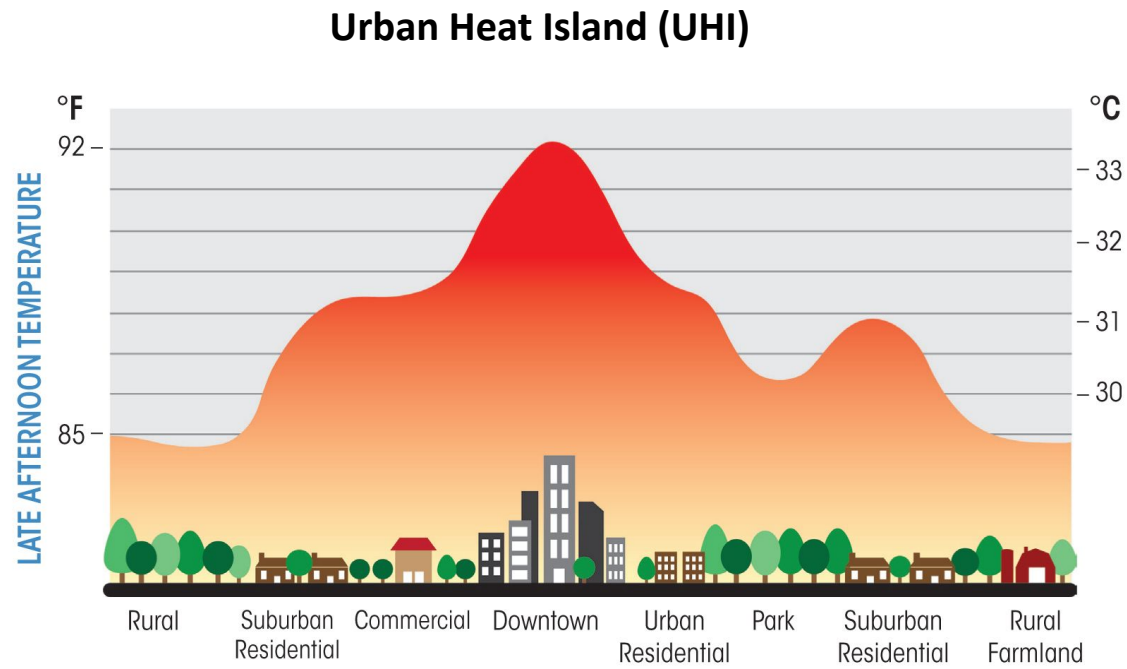
Lawrence et al. (2019), Danabasoglu et al. (2020)

Urban representation in CESM

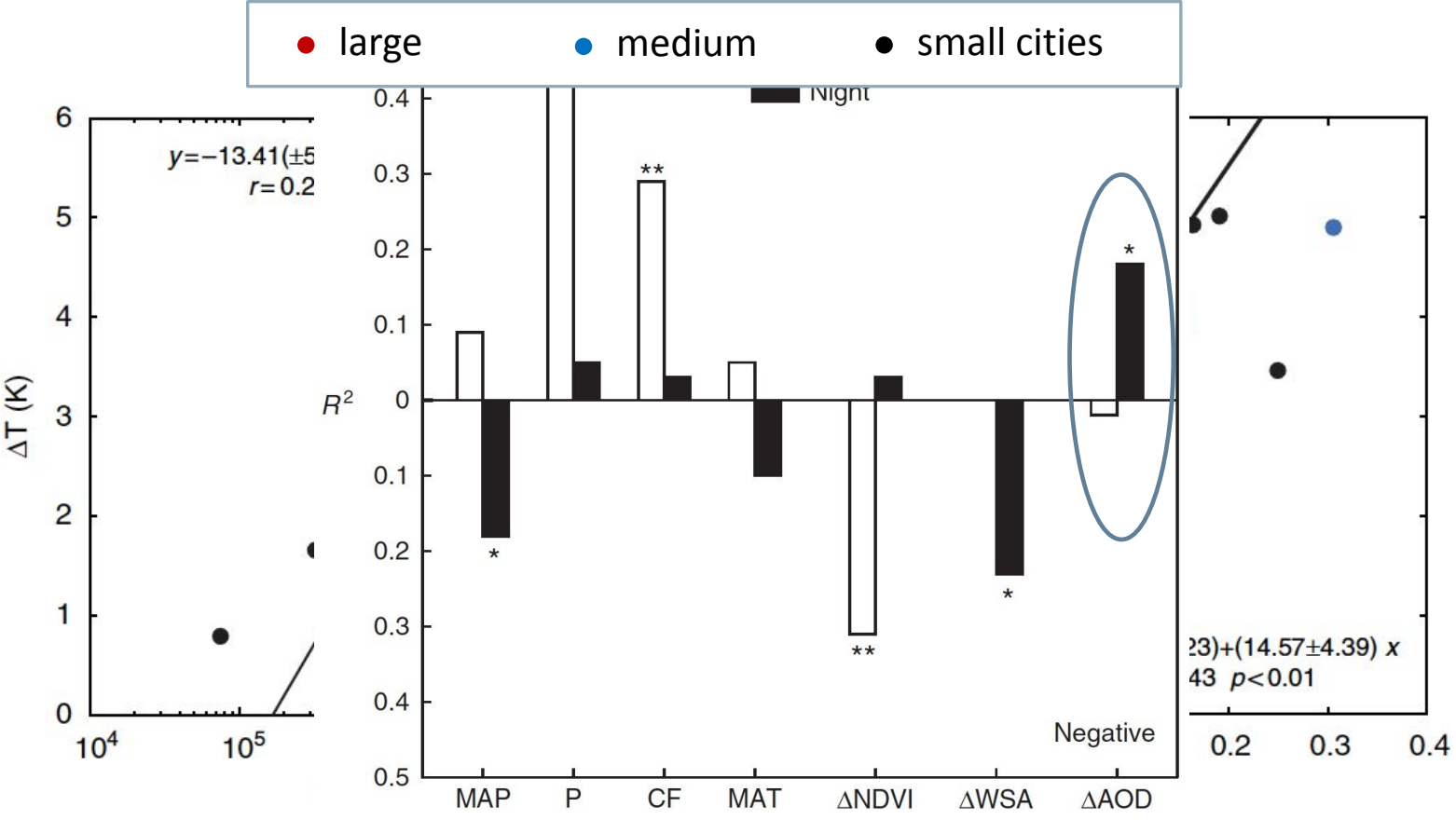


Understanding urban heat at large scales

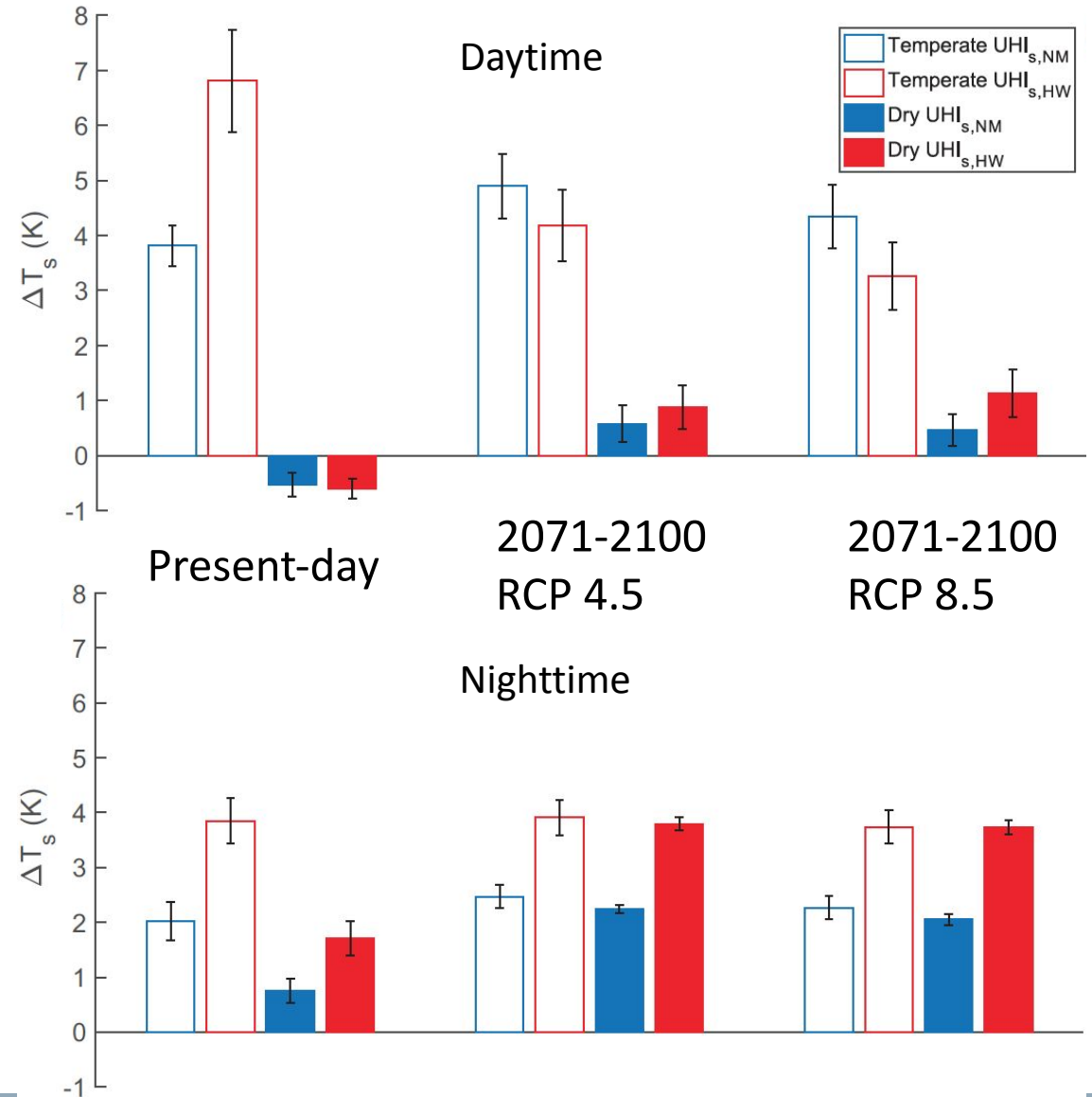
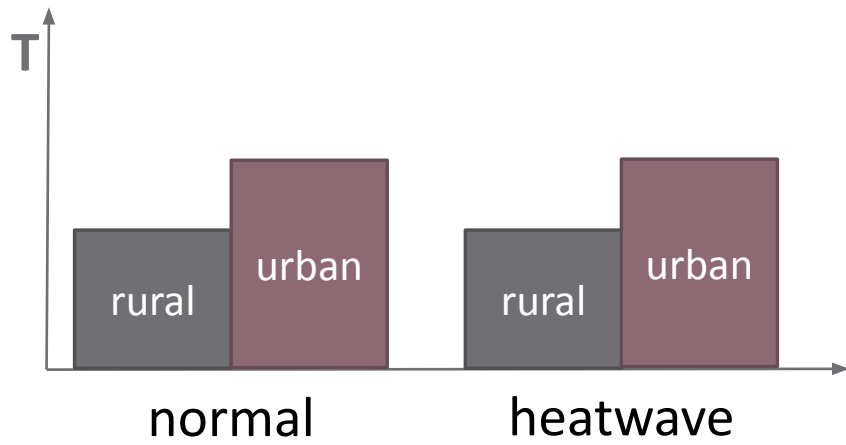
Urban heat island



UHI interacts with air pollution



UHI interacts with extremes



Can we provide robust multi-model *urban* projections?

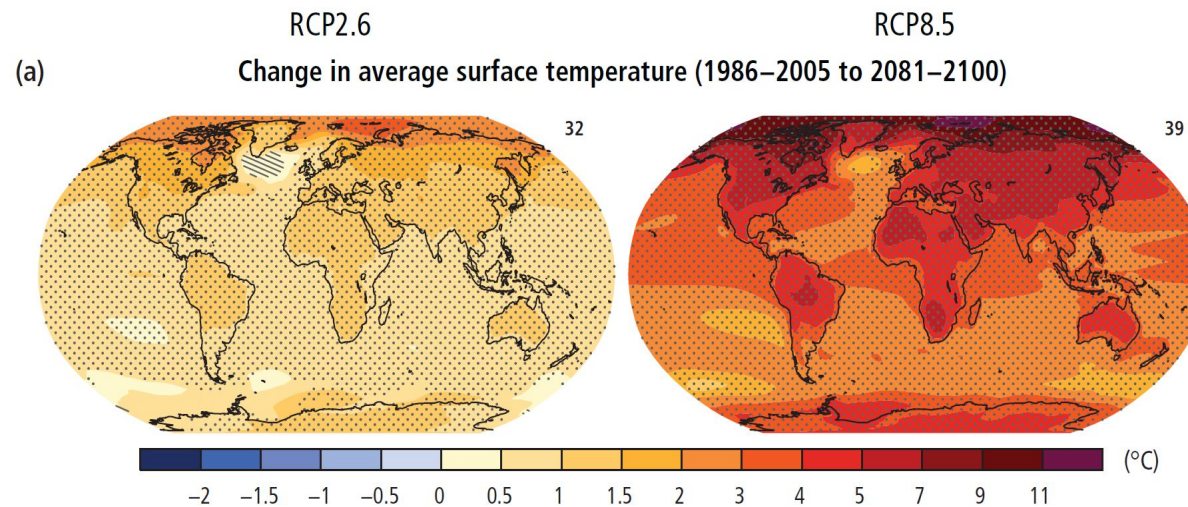
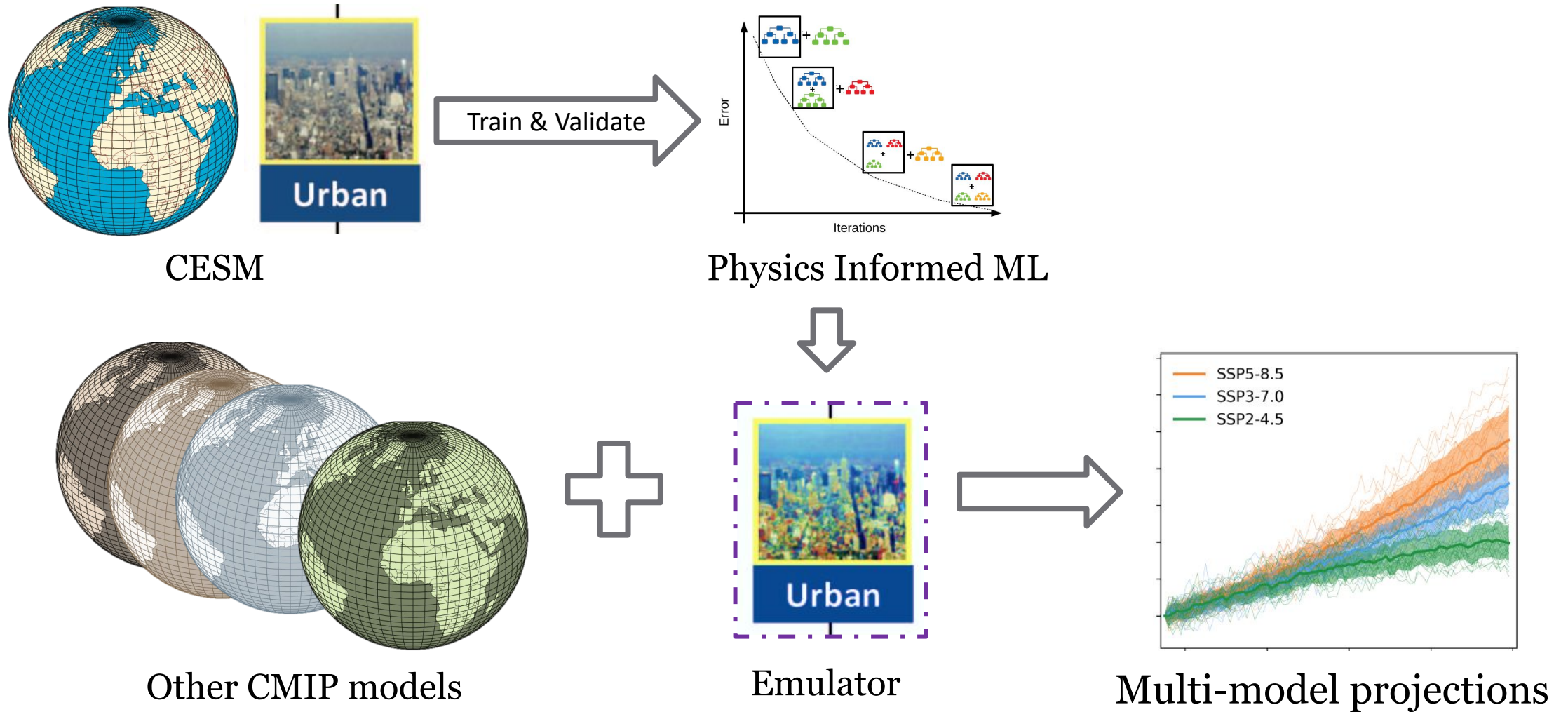


Figure SPM.7 | Change in average surface temperature (a) and change in average precipitation (b) based on multi-model mean projections for 2081–2100 relative to 1986–2005 under the RCP2.6 (left) and RCP8.5 (right) scenarios. The number of models used to calculate the multi-model mean is indicated in the upper right corner of each panel. Stippling (i.e., dots) shows regions where the projected change is large compared to natural internal variability and where at least 90% of models agree on the sign of change. Hatching (i.e., diagonal lines) shows regions where the projected change is less than one standard deviation of the natural internal variability. [2.2, Figure 2.2]

An urban climate emulator framework



Robust projections of urban warming

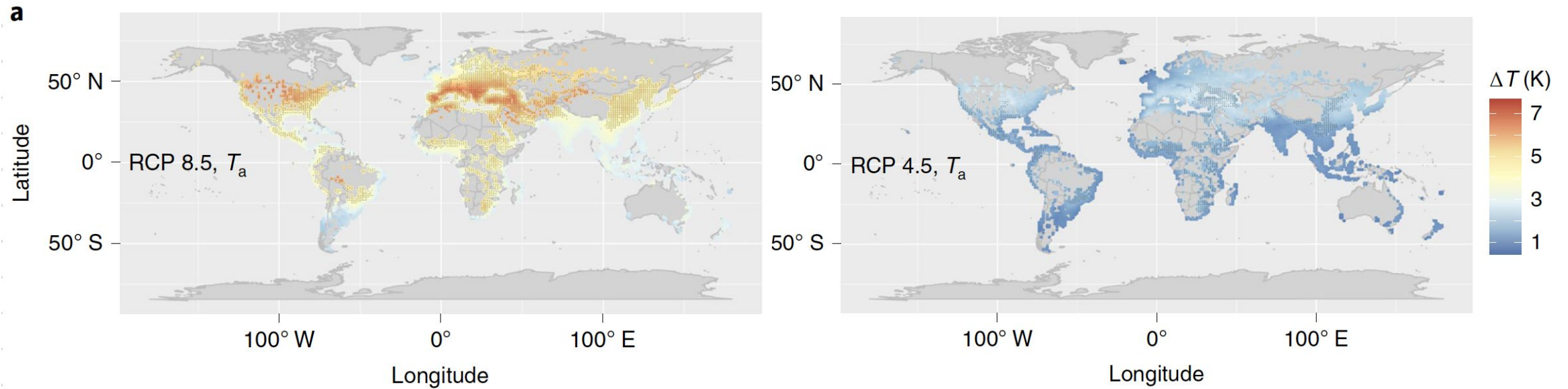


Figure: Multi-model ensemble mean urban warming for JJA under RCP 8.5 and RCP 4.5. Stippling indicates significant change ($\Delta T \geq 4$ K under RCP8.5 and $\Delta T \geq 1.5$ K under RCP45) with high inter-model robustness ($\text{SNR} > 2.5$).

Future urban heatwaves

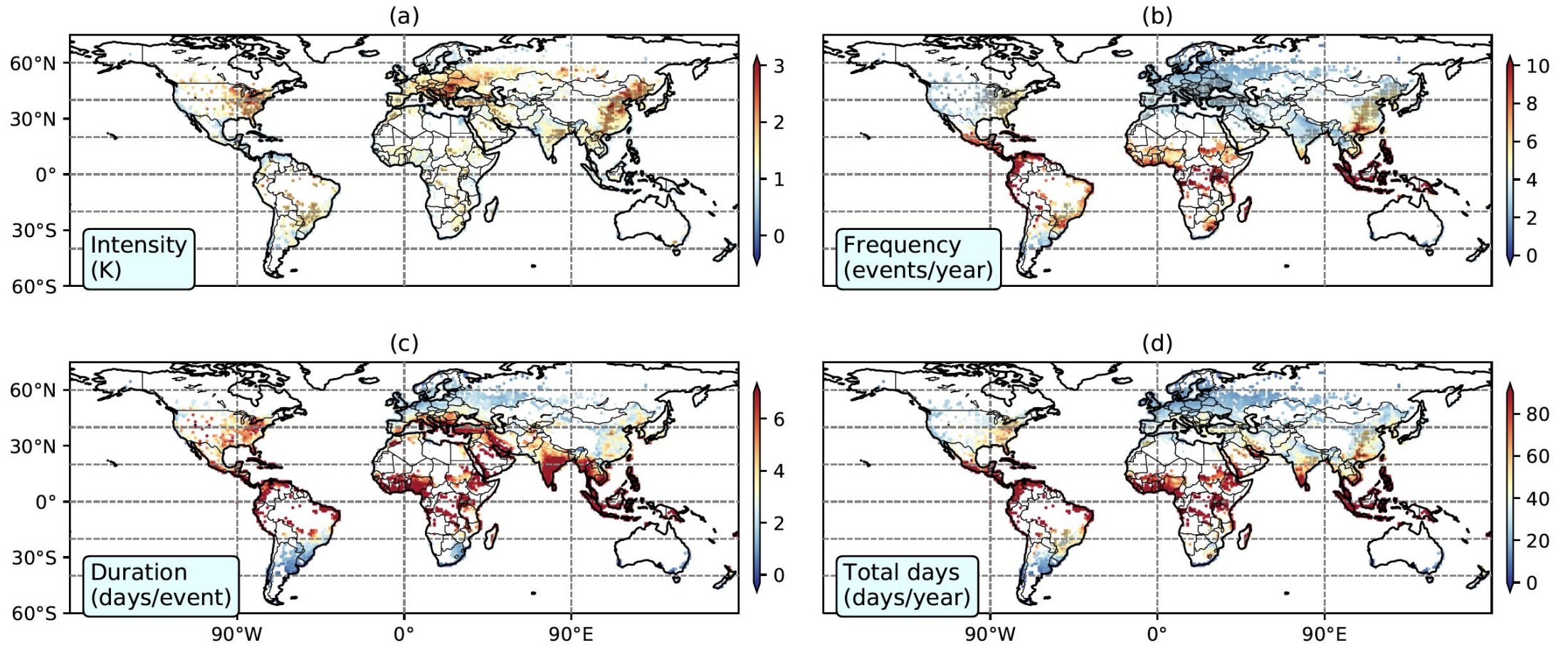


Figure: Multi-model ensemble mean change in average UHW intensity (K), frequency (events per year), duration (days per event), and total days (days per year) in 2061–2070 relative to 2006–2015. Stippling indicates substantial change (intensity > 1.5 K) with high inter-model robustness (SNR > 2.0).

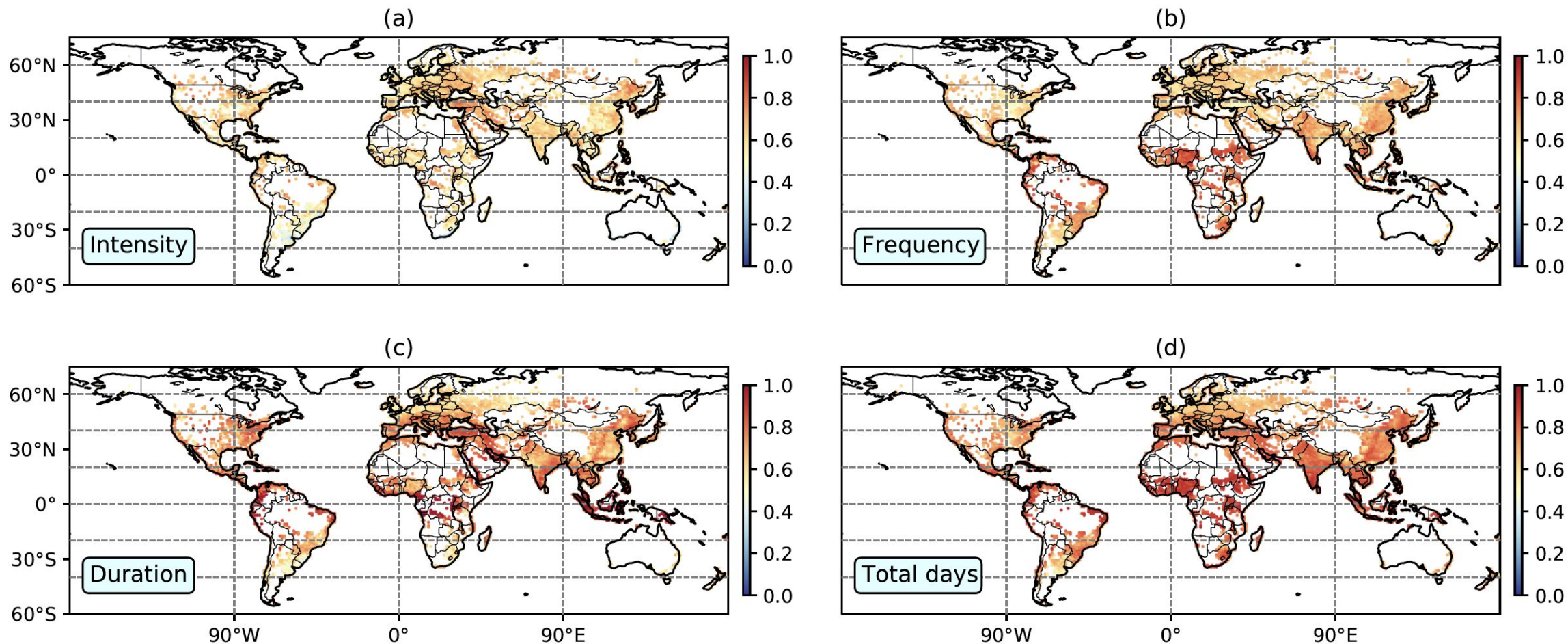
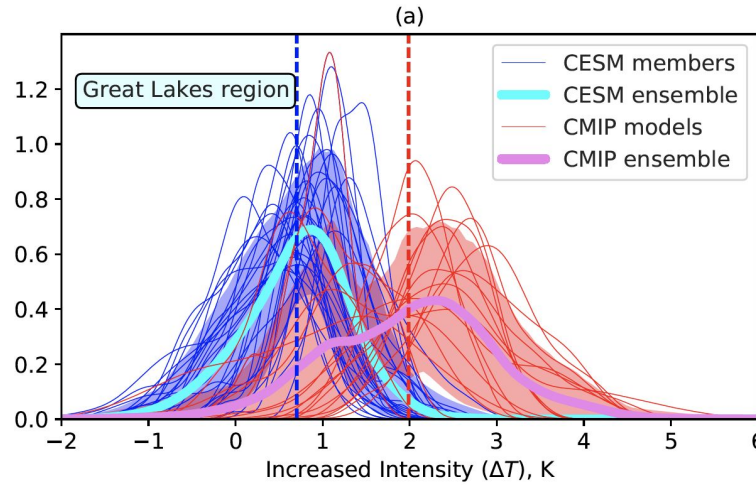


Figure 3. Relative contribution of the model structural variability in UHW projections. Each colored point represents a decadal mean structural uncertainty fraction (SUF) defined as $\frac{\sigma_{CMIP}}{\sigma_{CMIP} + \sigma_{CESM}}$

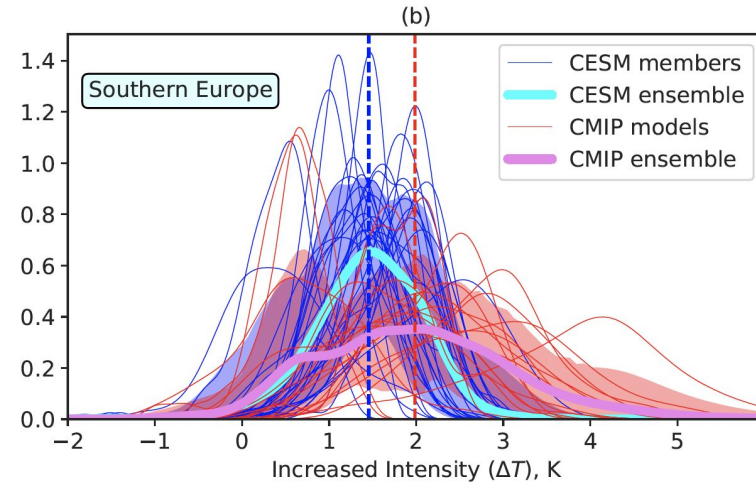
Grey-swan urban heatwaves

Seemingly unlikely
0.01% UHW

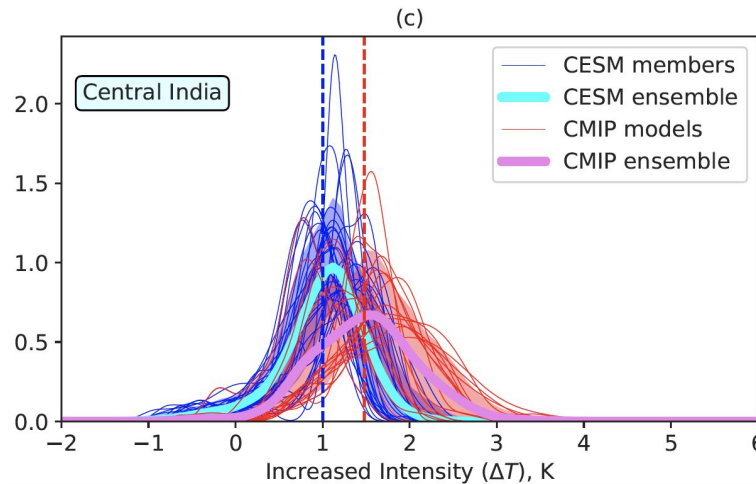
23.7%



4.2%



1.6%



14.8%

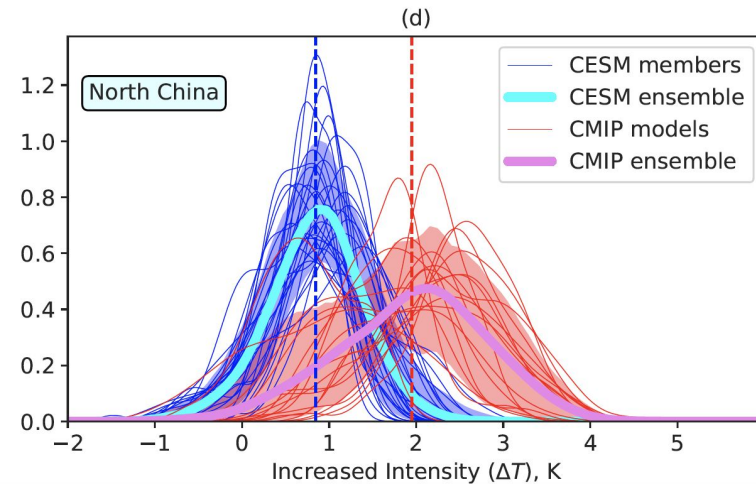


Figure: Spatial distribution of changes in urban heat waves intensity by 2061 – 2070. The thick lines mark the CMIP5 multi-model mean (thick violet lines) and CESM-LE multi-member mean (thick cyan lines).

Urban humid heat

Urban warming and “drying”

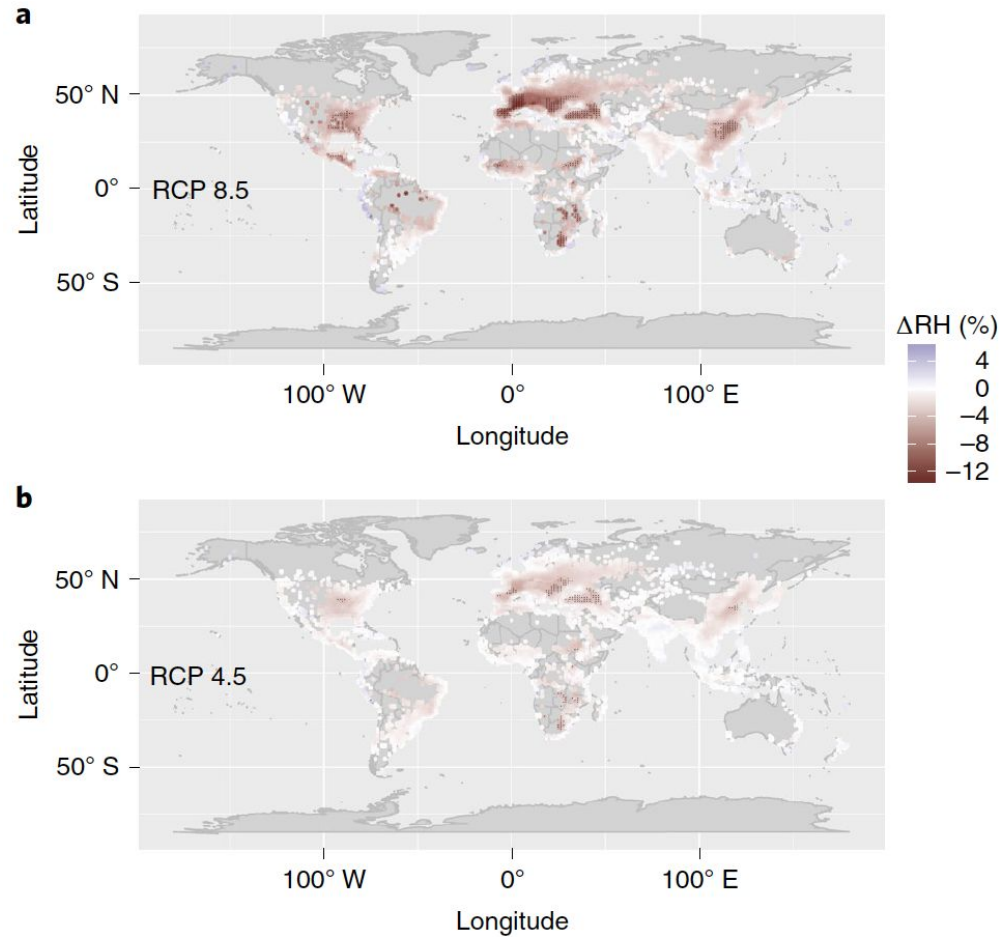
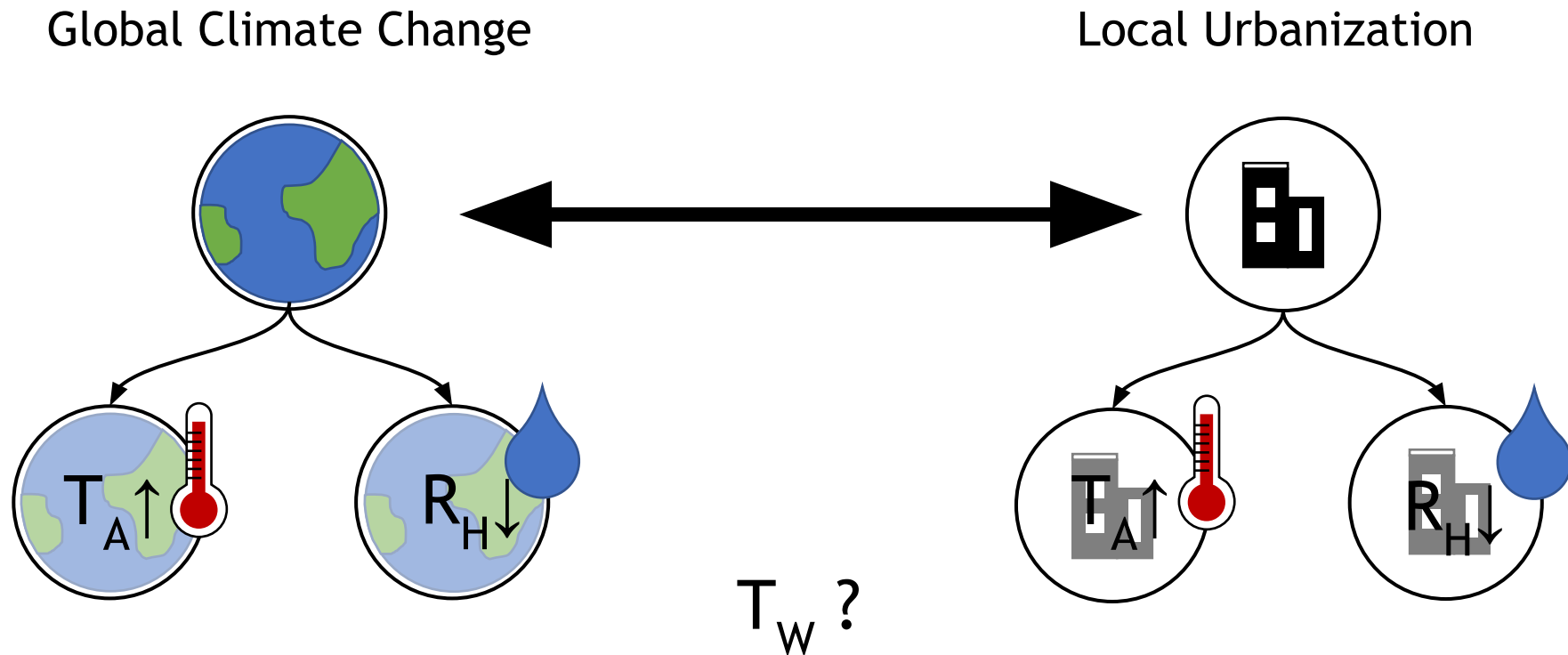


Fig. 4 | Multi-model mean urban relative humidity change for JJA. a,b, Seasonal mean urban RH changes between 2006–2015 and 2091–2100 under RCP 8.5 (**a**) and RCP 4.5 (**b**). Stippling indicates substantial change ($\text{abs}(\Delta\text{RH}) > 5\%$ under RCP 8.5 or $\text{abs}(\Delta\text{RH}) > 2.5\%$ under RCP 4.5) with high inter-model robustness ($\text{SNR} > 1$).

Zhao et al. *Nature Climate Change* (2021)

Urban warming and “drying”

Both climate change and urbanization tend to result in **higher temperatures and lower humidity**.



Urban wet-bulb heat island

Energy balance of web-bulb temperature $T_w + \frac{e_s^*(T_w)}{\gamma} = T_a + \frac{e_a}{\gamma} \longrightarrow \Delta T_w = w_1 \Delta T_a + w_2 \frac{\Delta e_a}{\gamma}$

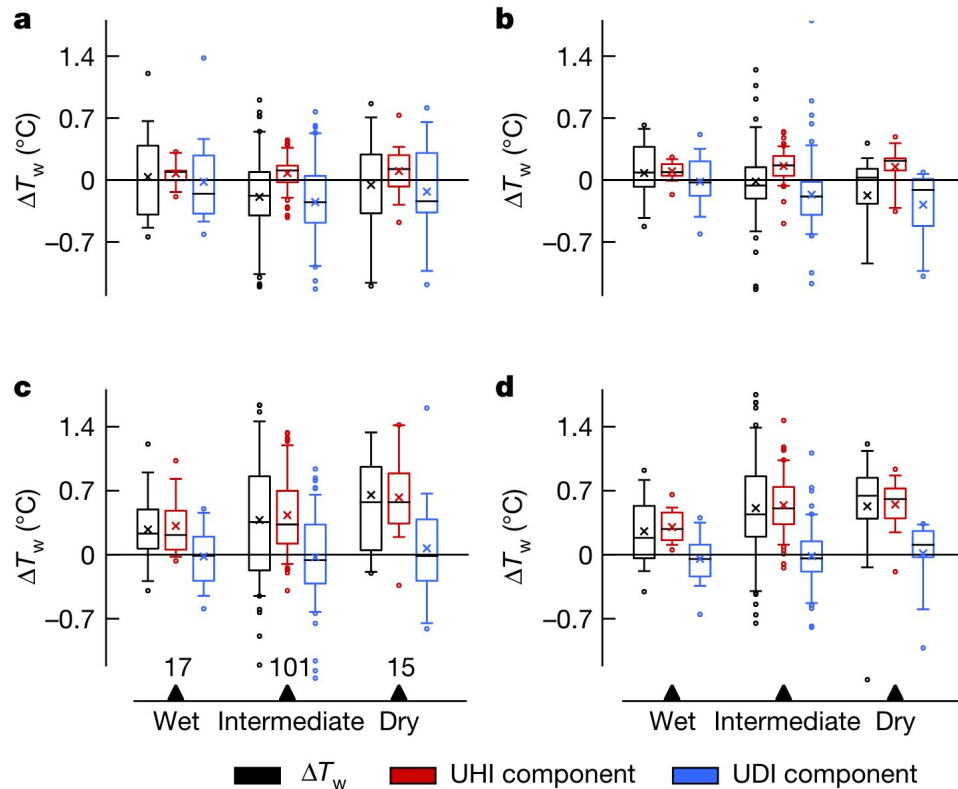
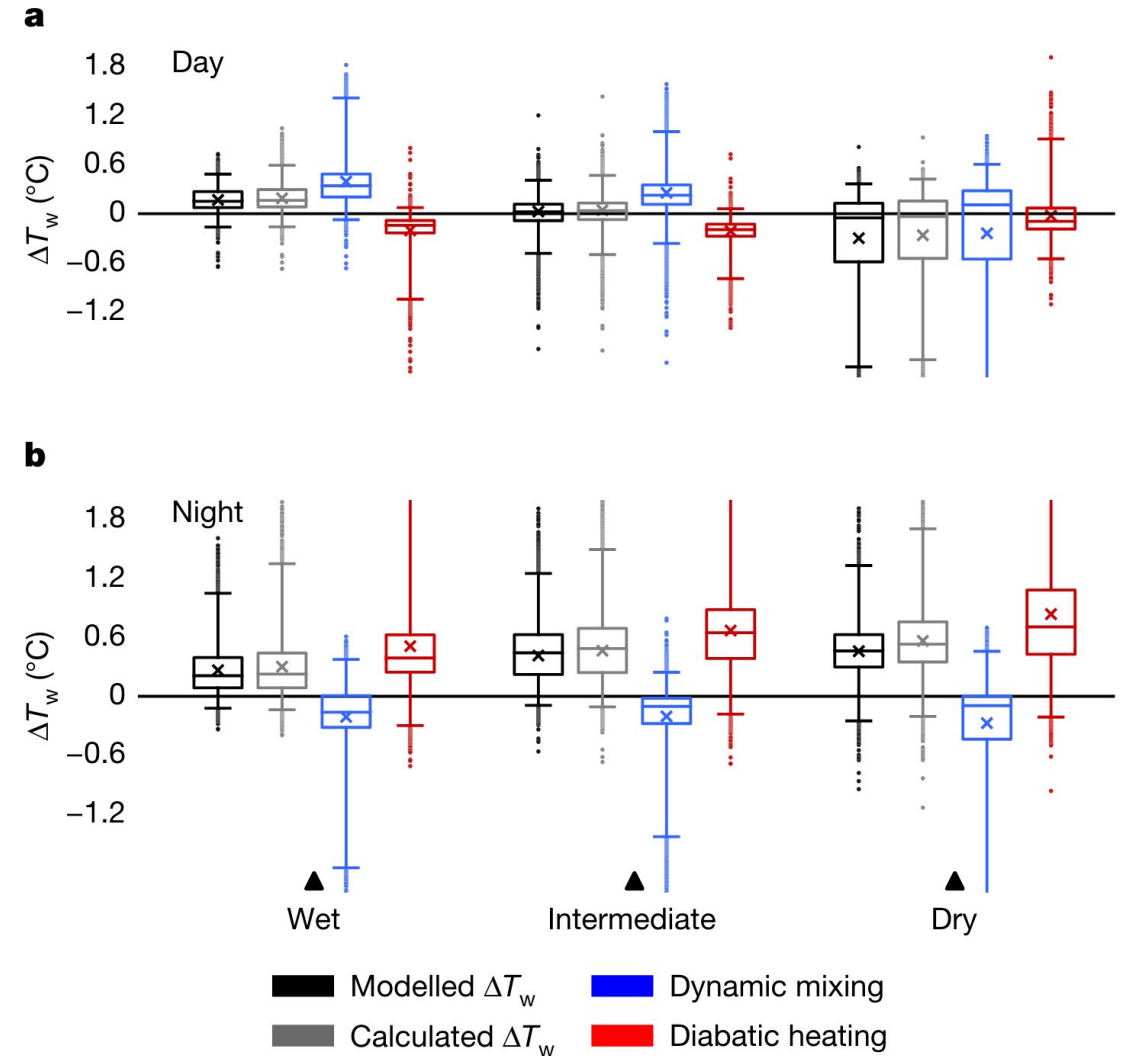


Fig. 1 | The urban wet-bulb island depends on time of the day and on climate wetness. a,c, Observed daytime (a) and nighttime (c) ΔT_w and its components. **b,d**, Modelled daytime (b) and nighttime (d) ΔT_w and its components. The model results are for grids corresponding to the urban–rural station pairs and from the same time periods. The box plots show the median (line), 25–75% range (box), 5–95% range (whiskers) and the mean value (cross). Station pair numbers are indicated in c.

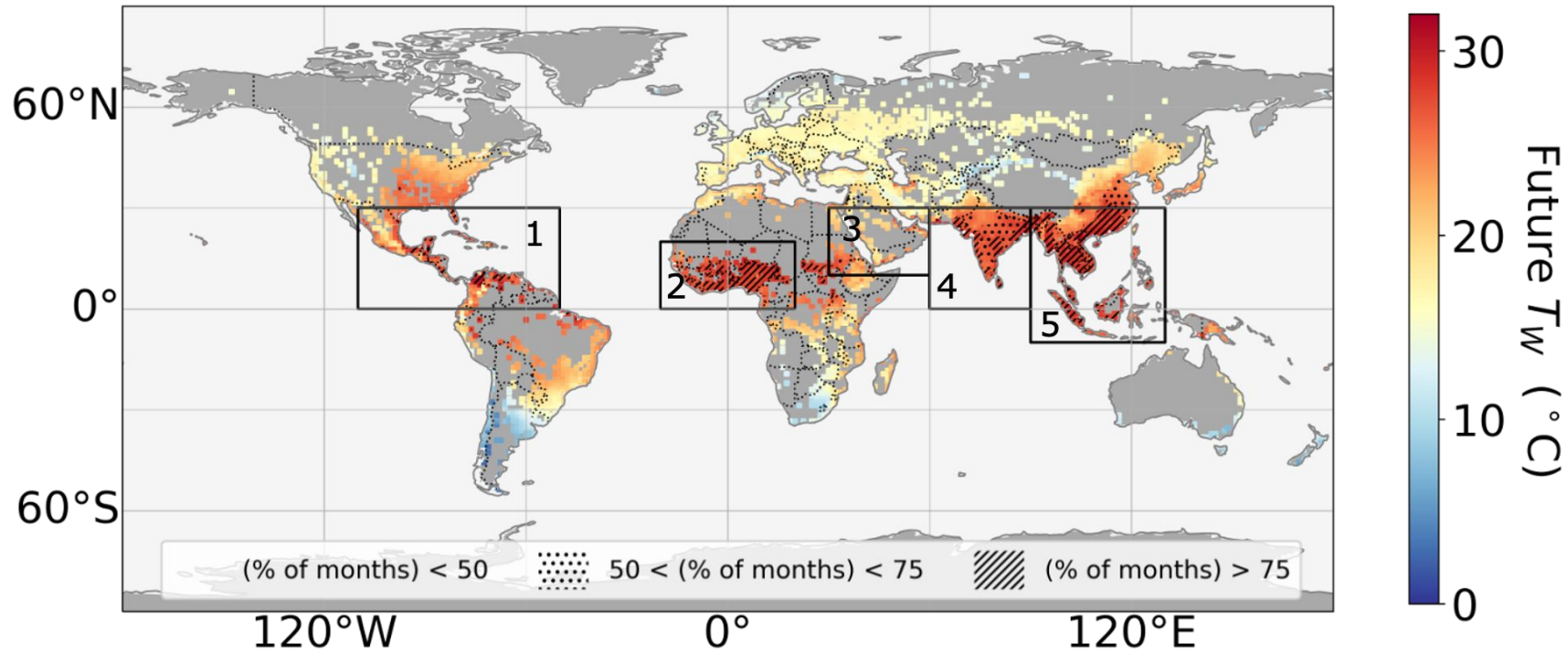
Urban wet-bulb heat island

$$\Delta T_w = \frac{1}{\rho c_p (1 + \Delta_w / \gamma)} [\Delta r_a (R_n + Q_A - G) + r_a \Delta (R_n + Q_A - G)]$$

Dynamic mixing
Diabatic heating difference

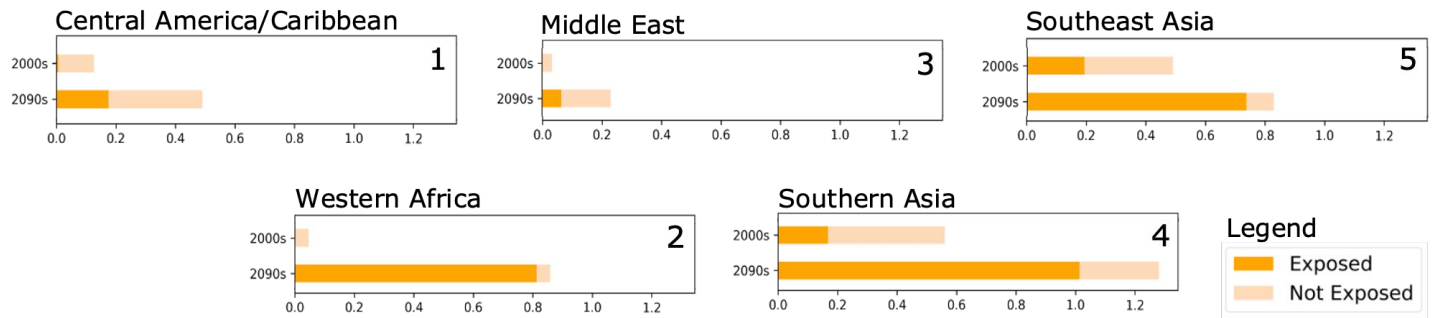
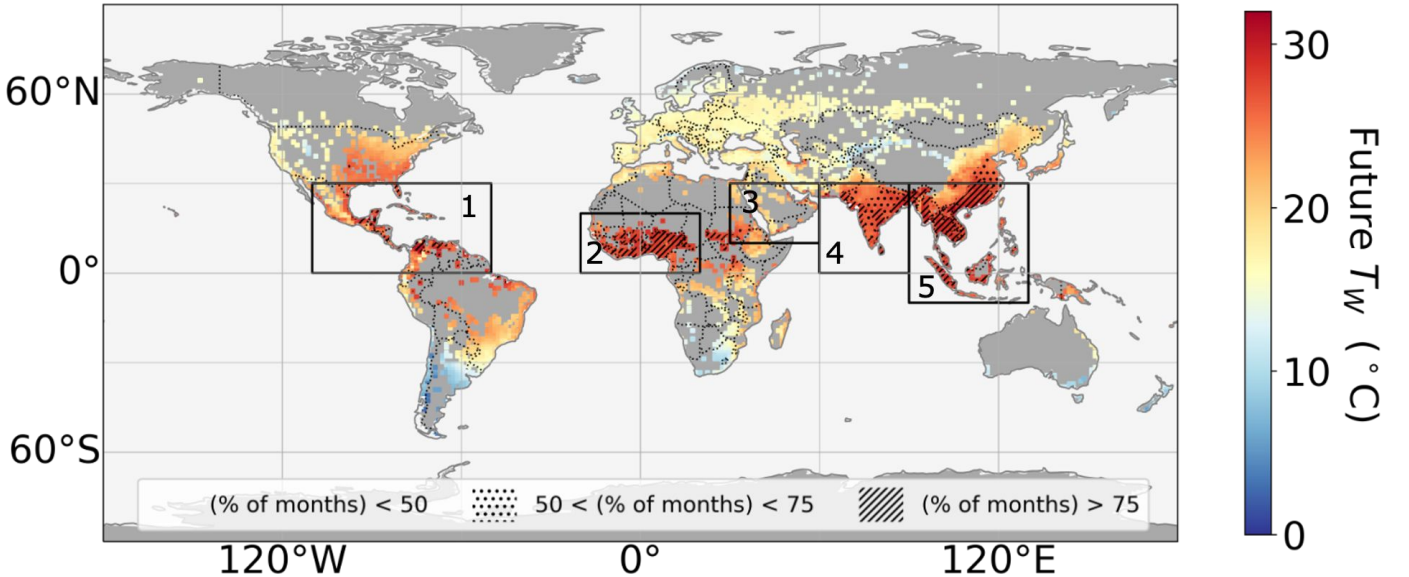


Urban humid heat increases globally, despite projected urban “drying”. There is a tendency towards high urban humid heat in coastal, tropical urban areas – with potential for chronic heat stress.



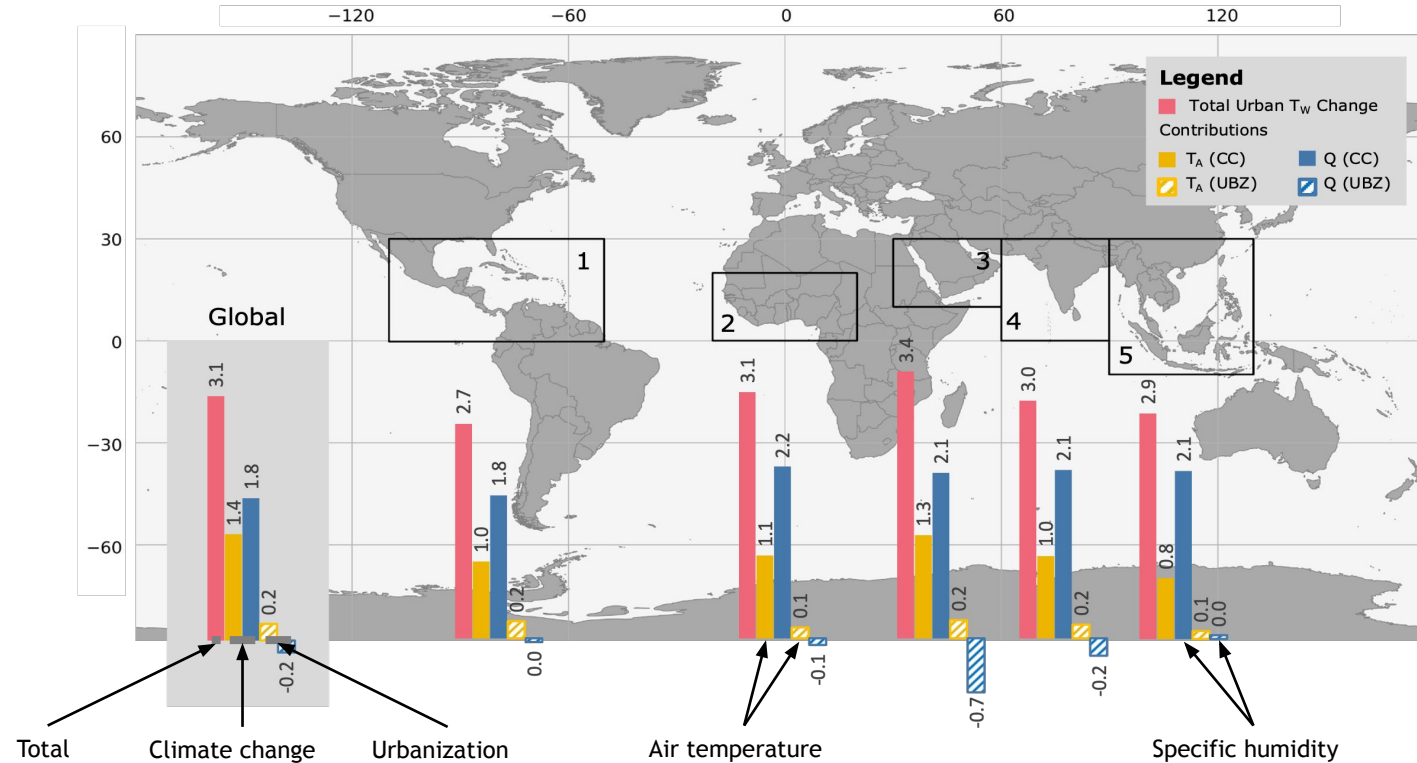
Urban humid heat (June-July-August, JJA) in the 2090s under SSP3-RCP7

In some areas of the world, growth in urban population potentially exposed to chronic heat stress is faster than growth of the total urban population.



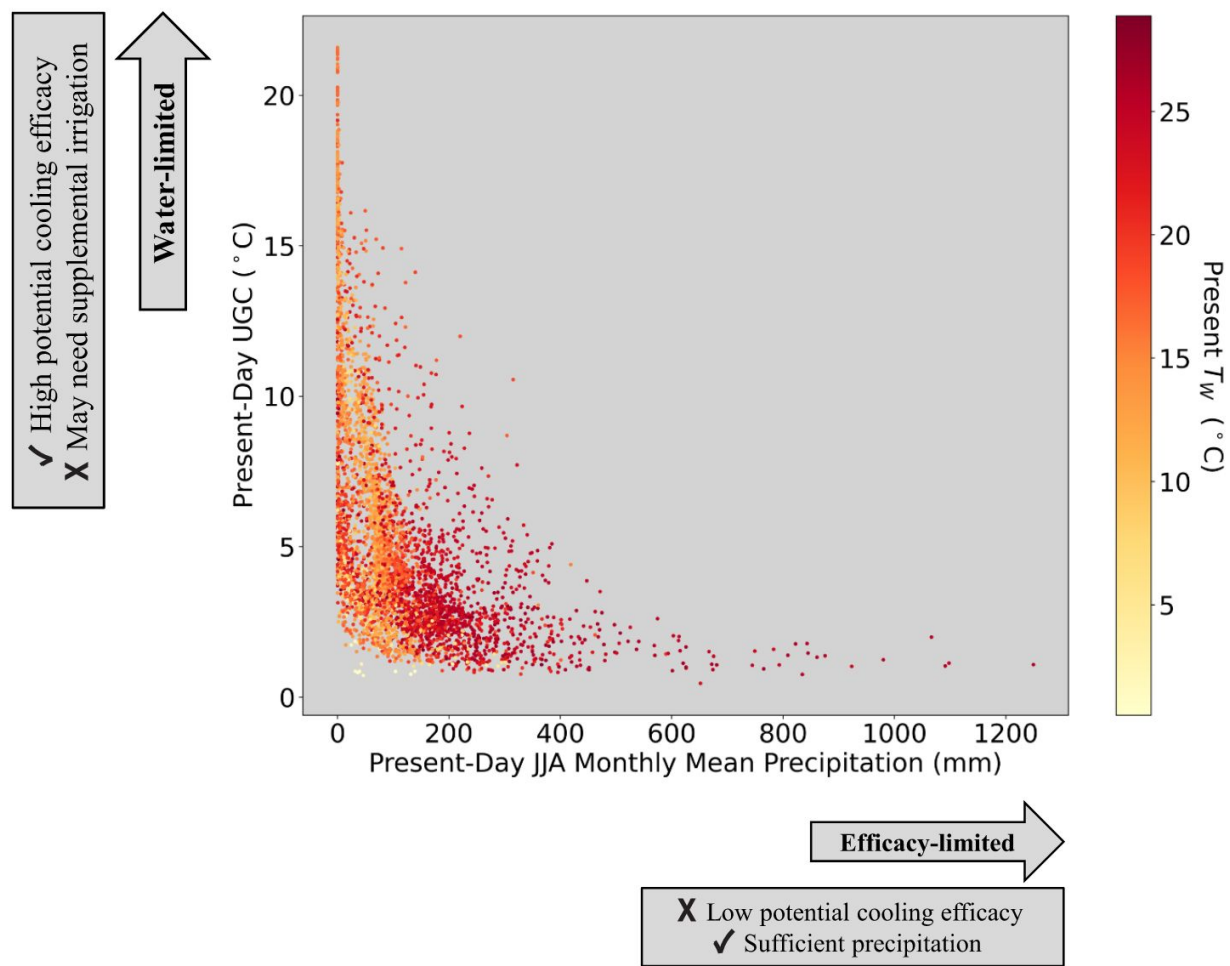
Growth in urban humid heat exposure under SSP3-RCP7

The near-universal elevated urban humid heat stress is largely driven by climate change and the substantial resulting specific humidity increases.



Total and decomposed change in urban humid heat

Globally-consistent, climate-driven dilemma to cooling through *urban greenery*: cities with greater (lower) atmospheric water availability tend to have lower (greater) cooling efficacy.



T_w : moist static energy

$$\Delta T_w = w_1 \Delta T_a + w_2 \frac{\Delta e_a}{\gamma}$$

Is urban green strategy really *evaporative "cooling"*?

Figure 5. Irrigation needs and cooling efficacy of UGI at the start of the century (2000–2009). Irrigation needs and cooling efficacy are reflected through the proxies of precipitation and UGC, respectively. Precipitation and UGC are decadal mean urban JJA values. Colors indicate the decadal mean JJA urban T_w .

Advancing urban representation in CESM

Key gaps and challenges

- Large urban underrepresentation in SOTA ESMs
- Globally consistent, reliable, and realistic urban surface property data
- Understanding and representing urban-scale processes and dynamics

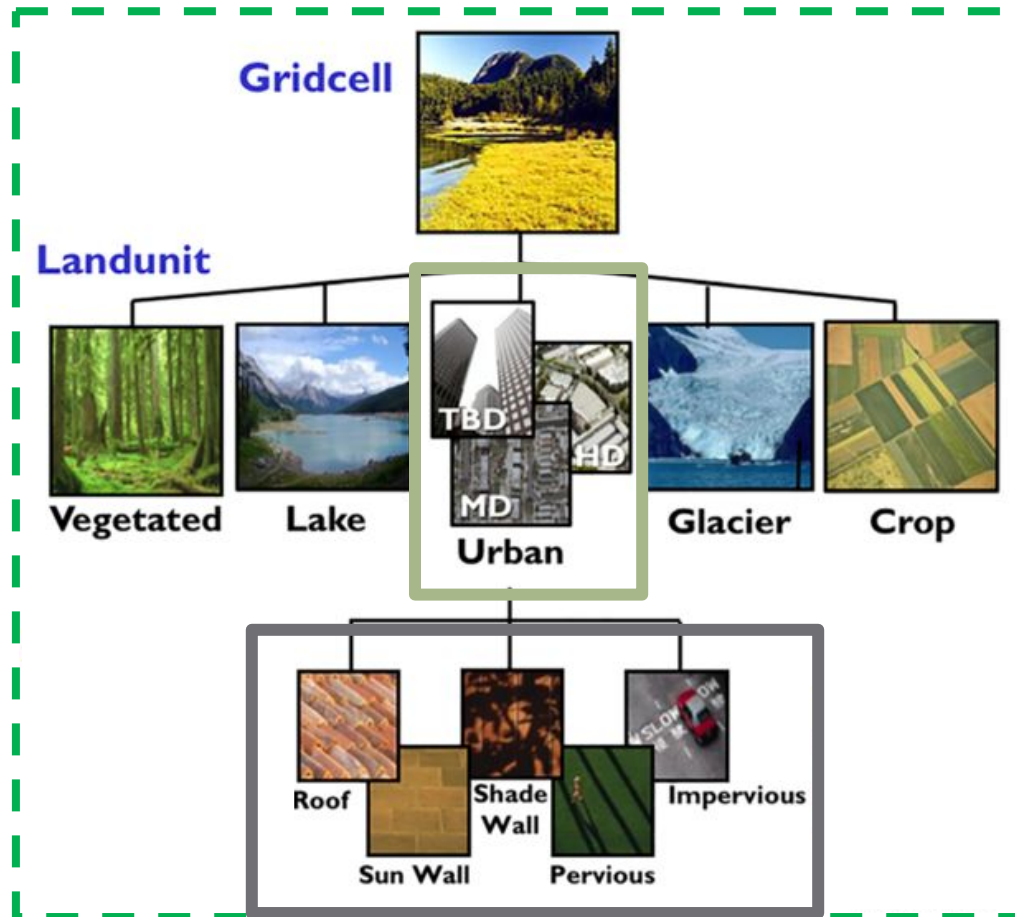
Data

Science

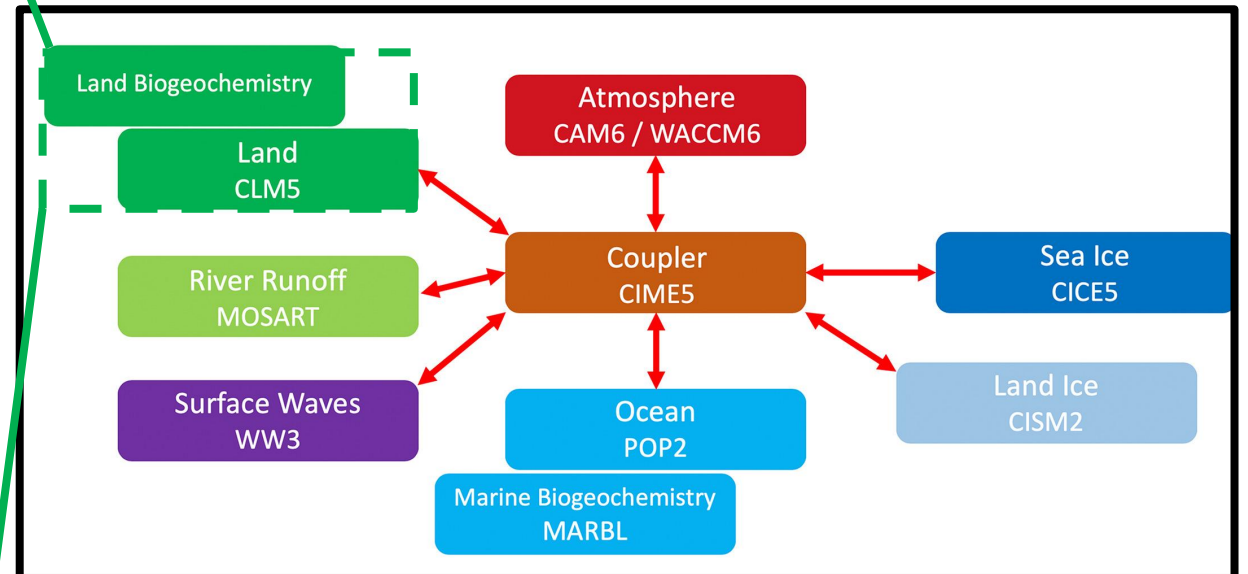
Computing

1. Dynamic-urban CESM

A critical limitation in current ESMs: static urban



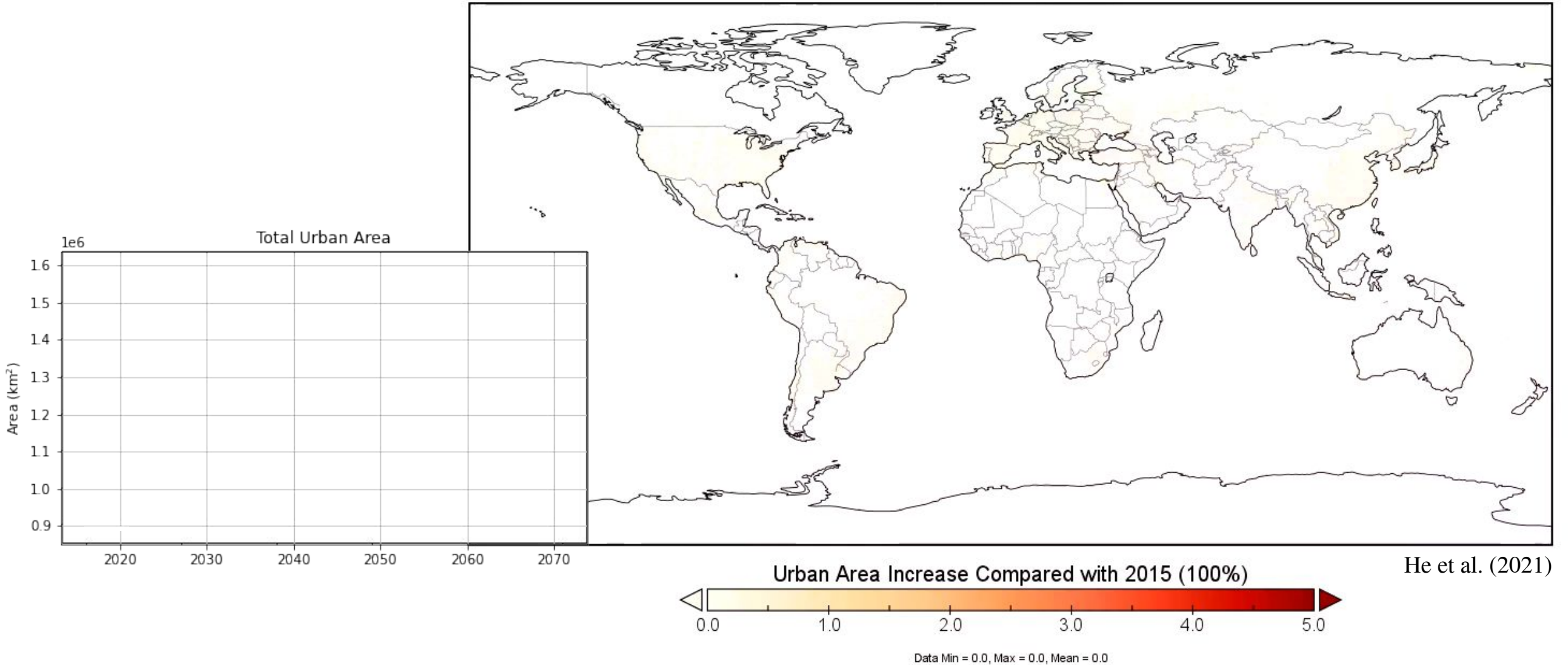
Community Earth System Model (CESM)



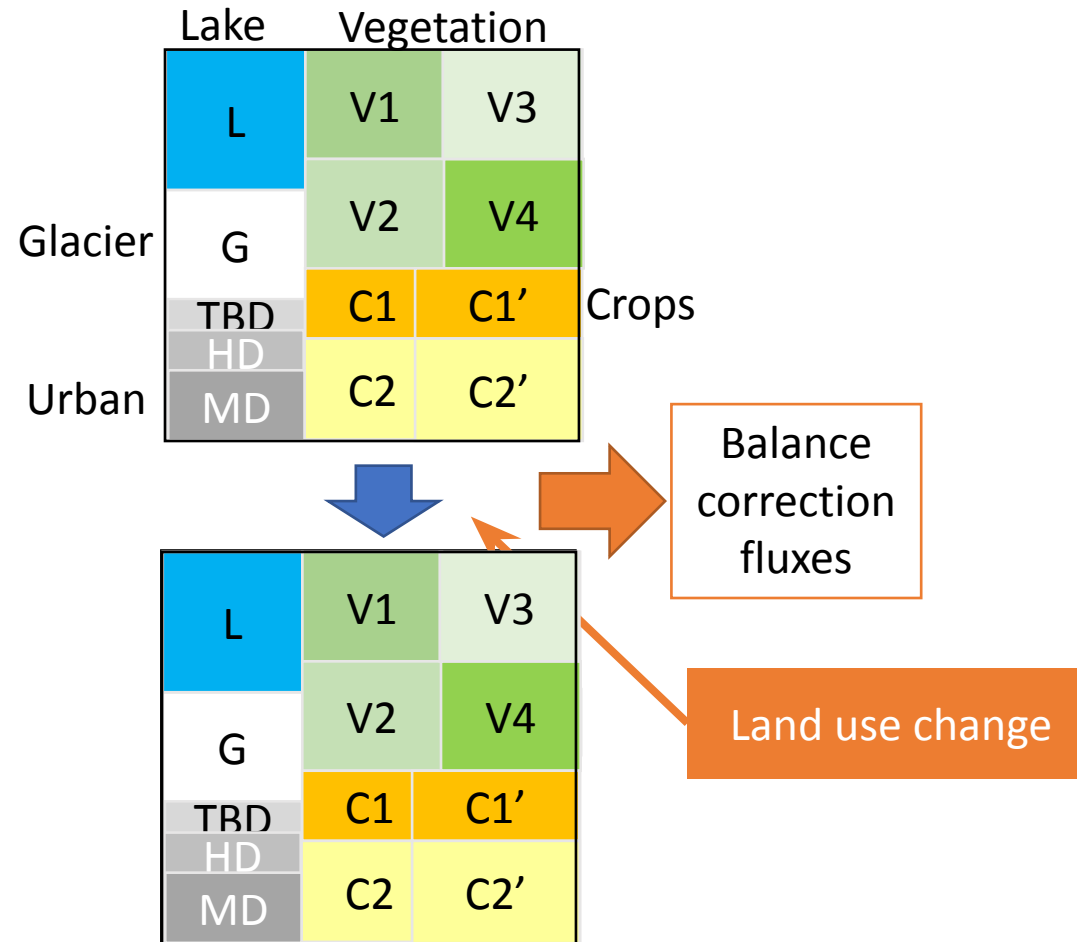
Rapid urbanization could have climate impacts at local scale and beyond

Projected Urbanization under SSP5-8.5 Scenario

Time: 2015-01-01 00:00

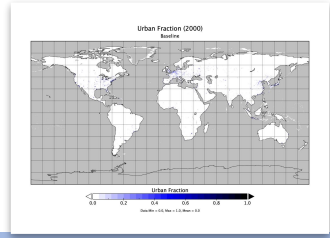


A new dynamic urban scheme in CESM



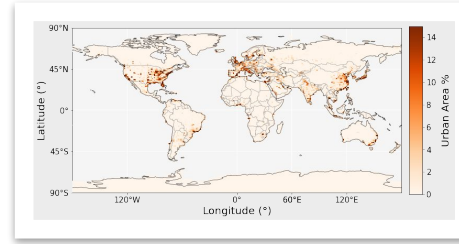
Dynamic urban scheme in CESM

A Original (static) urban representation

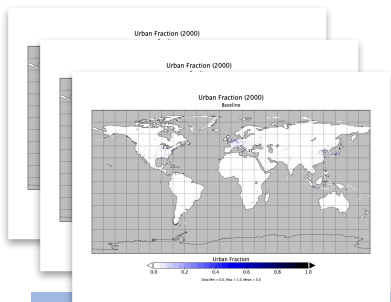


CESM
Surface dataset

Model driver:
Static initialization



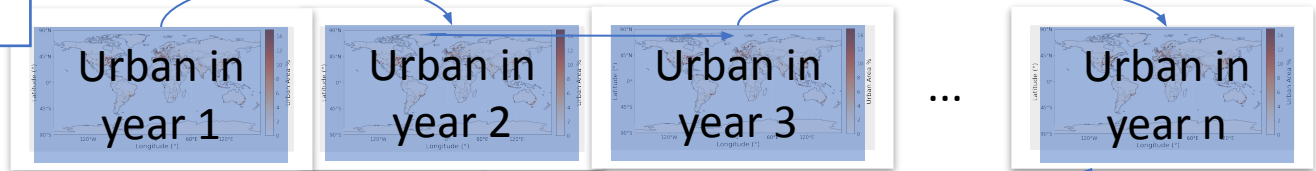
Same urban extents and properties throughout simulation



CESM land use
time series files

Model driver:
Selective initialization

Dynamic scheme: Conservation check



B Dynamic urban representation

Dynamic scheme:
Annual update

The new dynamic urban scheme opens up new opportunities

master CTSM / doc / ChangeSum

Go to file

bill sacks Update ChangeLog Latest commit 563ce93 24 days ago

13 contributors

753 lines (753 sloc) 71.4 KB

1	Tag	Who	Date	Summary
2	-----			
3	ctsm5.1.dev098	swensosc	05/20/2022	Correct perched water table calculation
4	ctsm5.1.dev097	sacks	05/20/2022	Update ccs_config to fix issue on izumi and maybe elsewhere
5	ctsm5.1.dev096	erik	05/18/2022	Fix a few glitches from the last tag
6	ctsm5.1.dev095	erik	05/09/2022	Turn soil BGC off for FATES-SP mode, externals updates, FatesSP user-mode directory and compset
7	ctsm5.1.dev094	negins	05/06/2022	subset_data allows zeroing out nonveg landunits without any dompft selected
8	ctsm5.1.dev093	slevis	05/03/2022	Modifications for FATES-MIMICS to work
9	ctsm5.1.dev092	sacks	04/29/2022	Refactor NutrientCompetition / CNAallocation to provide hooks for AgSys
10	ctsm5.1.dev091	rgknox	04/22/2022	clm decomp method is now passed to fates to enable mimics coupling
11	ctsm5.1.dev090	samrabin	03/31/2022	Fix misleading name of "gddplant"
12	ctsm5.1.dev089	sacks	03/31/2022	For CLM45 apply peakl1 to aleef in grainfill
13	ctsm5.1.dev088	samrabin	03/28/2022	Add outputs for annual crop sowing and harvest dates
14	ctsm5.1.dev087	sacks	03/24/2022	Reduce aux_clm testing cost
15	ctsm5.1.dev086	negins	03/22/2022	Updates necessary for NEON v2 files and server updates
16	ctsm5.1.dev085	sacks	03/16/2022	Expand crop reproductive pools; remove some unused options
17	ctsm5.1.dev084	glemieux	03/15/2022	FATES parameter file updated to align with clm pft optical parameters
18	ctsm5.1.dev083	multiple	03/08/2022	Implement PCT_URBAN_MAX to minimize dynamic urban memory
19	ctsm5.1.dev082	slevis	02/28/2022	Replace dom_nat_pft with dom_plant to enable crop in fsurdnt_modifier tool
20	ctsm5.1.dev081	swensosc	02/24/2022	Do not subtract irrigation from QRUNOFF diagnostic
21	ctsm5.1.dev080	sacks	02/24/2022	Use avg days per year when converting param units
22	ctsm5.1.dev079	sacks	02/24/2022	Changes to CropPhenology timing
23	ctsm5.1.dev078	sacks	02/24/2022	Rework single-point testing
24	ctsm5.1.dev077	rgknox	02/22/2022	Updates to FATES API, including removal of patch dimensions from fates history and using soil instead of ground layers for fates history
25	ctsm5.1.dev076	negins	02/18/2022	updating subset_data.py script and move to the Python package.
26	ctsm5.1.dev075	erik	02/16/2022	Small answer changes: urban ventilation, fire-emission, irrigate off when not crop, fix two SSP ndep files
27	ctsm5.1.dev074	slevis	02/02/2022	Introduce vert. resolved MIMICS as new method to solve below ground decomp.
28	ctsm5.1.dev073	sacks	01/25/2022	Some fixes for Gregorian calendar
29	ctsm5.1.dev072	negins	01/17/2022	mksurfdnt toolchain part 1: gen_mksurf_namelist
30	ctsm5.1.dev071	glemieux	01/16/2022	Small changes to enable new fates dimension and update fates tag
31	ctsm5.1.dev070	multiple	01/10/2022	Update externals, remove need for LND_DOMAIN_FILE and LND_DOMAIN_PATH, etc.
32	ctsm5.1.dev069	multiple	12/15/2021	Implement dynamic (transient) urban capability
33	ctsm5.1.dev068	multiple	12/13/2021	Adding fsurdnt_modifier tool
34	ctsm5.1.dev067	jedwards	12/13/2021	NEON UI update, externals updates, small miscellaneous fixes

Atmospheric Fluxes

- **Dynamic Urban**, new **urban datasets**, & building property parameterization [Keith Oleson]
- **Surface roughness modifications** [Ronnie Myer]
- **Dust emissions and more** [Longlei Li, Natalie Mahowald, Danny Lueng, Jasper Kok]
- **Updated PFT optical properties** [Keith Oleson]
- **Biomass heat storage** [Sean Swenson]
- **Snow burial of vegetation** [Danica Lombardozzi]
- **Multilayer canopy** [Gordon Bonan & Ned Patton]

Hydrology

- **Representative hillslope hydrology** [Sean Swenson & others]
- **Water isotopes** [Bill Sacks]
- **Excess Ice** [Lei Cai, Matvey Debolskiy]
- **Irrigation** [Bill Sacks & Sean Swenson]
- **mizuRoute** [Naoki Mizukami & Erik Kluzek]
- **Reservoir operations** and dynamic lakes [Inne Vanderkelen]

Ecosystems and Biogeochemistry

- **FATESsp**, **no competition**, and **other simplified configurations**. [Rosie Fisher, Charlie Koven, Jackie Shuman, Adrianna Foster & many more]
- FATES point and regional configurations [Jackie, Adrianna, Polly, others]
- **MIMICS-CN soil bgc model** [Will W. & Sam Levis]
- **Ozone impacts on Jmax** in LUNA [Stefanie Falk]
- **Fire Model**: **bug fixes, improvements, & tuning** [Fang Li]
- **Arctic/Boreal phenology & allocation** [Leah Birch]
- **CN-Matrix** for biogeochemistry [Chris Lu & Yiqi Luo]
- **Aerosols: FAN** (NH3 emissions) [Julius Vira & Peter Hess]
- **Terrestrial DOM fluxes** and transport in river model [Dev N.]

Crop Model

- **Clean up crop phenology, planting & harvest dates** [Sam Rabin]
- **Bioenergy crop** [Yanyan Cheng & Maoyi Huang]
- **Winter wheat** [Yaqiong Lu], **implemented but not currently used**
- **Shifting cultivation** [Peter Lawrence]
- APSIM crop phenology [Bin Peng]
- Tillage and crop residue management [Mike Graham & Danica L]
- Soil degradation with Land Use [Johann Feddema, Pei-Ling]

Features

- **Switch to NUOPC** [Erik, Mariana]
 - Single point & regional workflow including **NEON simulations, supported towers & generic grid cells** [Danica L, Negin Sobhani, Will Wieder, Adrianna Foster, Erik]
 - **Parameter Perturbation Ensemble** [Daniel, Katie, Dave]
 - **Moving hard coded parameters to parameter file** [Keith Oleson]
 - **Simple Land Model**, SLIM [Marysa Lague]
 - Tool chain modifications **part 1** [Sam, Bill, Negin]
 - **WRF-CTSM beta release** [Dave Lawrence & more]
 - **No anthro compsets**: turn off irrigation, crop, urban, LULCC, fire
 - **Prescribed soil moisture** [Sean Swenson]
 - **Soil and snow layer flexibility + trimming land units & PFTs** [Sam Levis & Bill Sacks]
 - **SSP-RCP anomaly forcing compsets** for land only simulations
-
- **Bold items** have been merged into CTSM5.1_main
 - **Linked items** point to github pull requests.
 - **Highlighted items** are outstanding CTSM5.1 milestones
 - **Highlighted items** will require new datasets that will come in for CTSM5.2.

2. Improving the BEM in CESM

New explicit-AC-adoption scheme

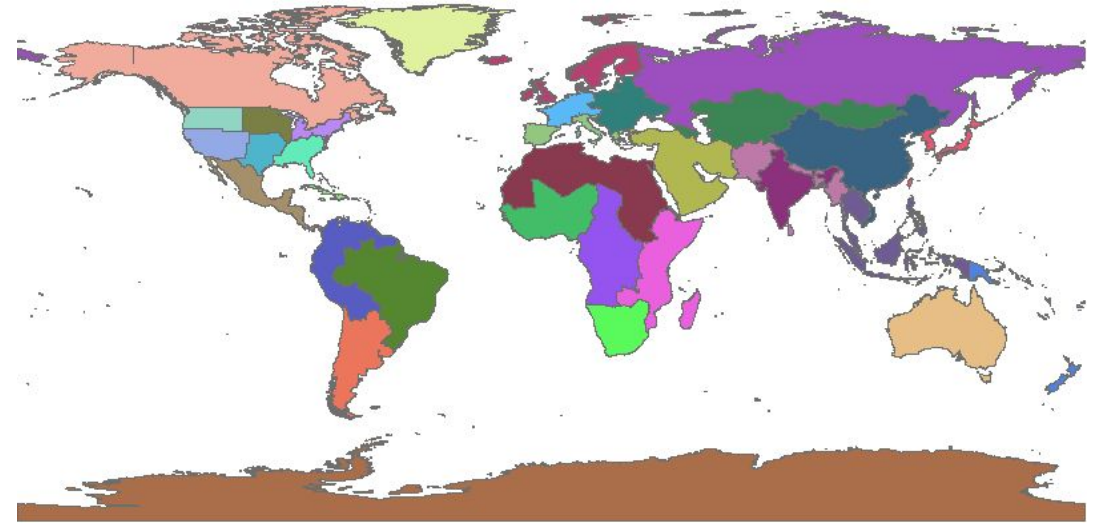
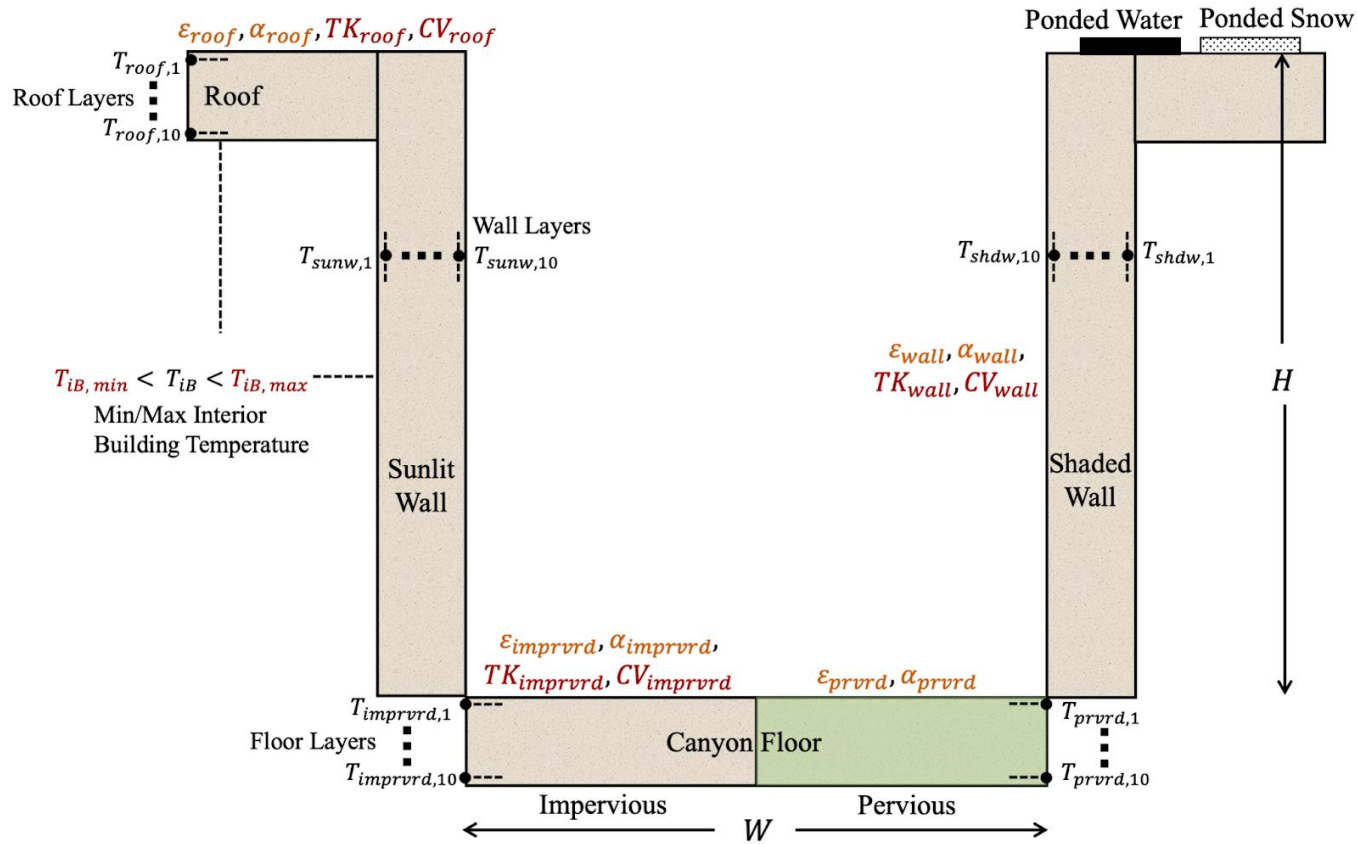
Table 1. Comparison of the original and new AC adoption modeling scheme.

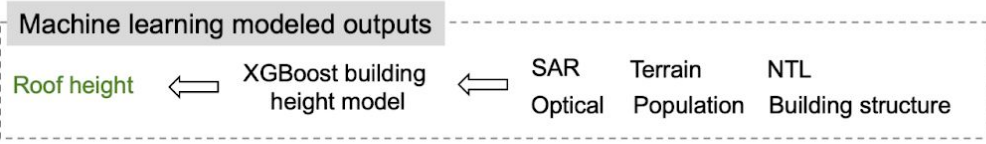
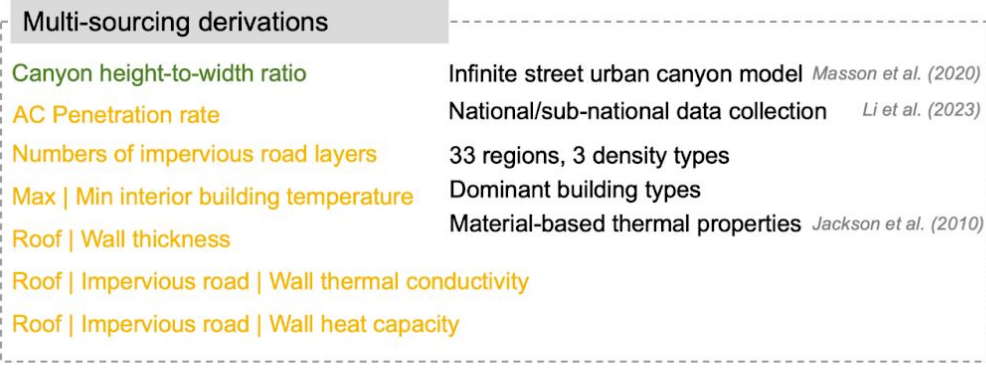
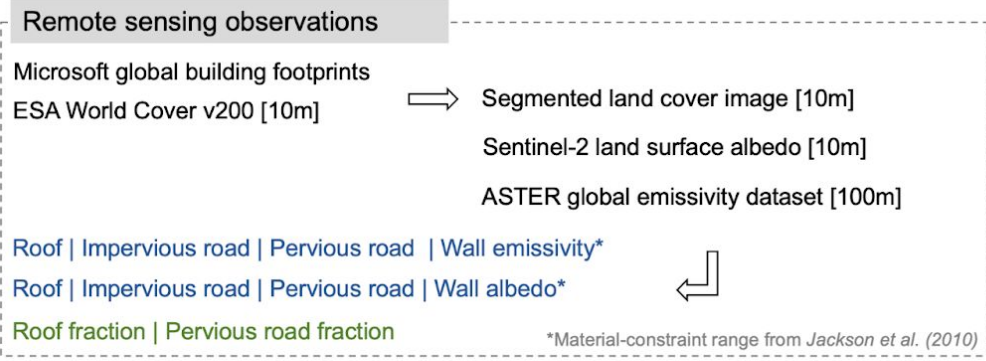
	Original	New
AC adoption	Implicitly modeled by proxy setpoint T_{max}	Explicit AC adoption rate, p_{AC} + Saturation setpoint at 100% adoption, $T_{sat,max}$
AC trigger	Indoor temperature $T_{i_B}^{t+1} > T_{max}$	$T_{i_B}^{t+1} > T_{sat,max}$
AC flux	$F_{AC} = \frac{H\rho C_p}{\Delta t} (T_{i_B}^{t+1} - T_{max})$	$F_{AC} = p_{AC} \cdot F_{sat,AC},$ where $F_{sat,AC} = \frac{H\rho C_p}{\Delta t} (T_{i_B}^{t+1} - T_{sat,max})$
Updating $T_{i_B}^{t+1}$	$T_{i_B}^{t+1} = T_{max}$	$T_{i_B}^{t+1} = \frac{(1 - p_{AC})F_{sat,AC} \cdot \Delta t}{H\rho C_p} + T_{sat,max}$

H : building height [m], ρ : air density [kg/m^3], C_p : specific heat of dry air [$\text{J}/\text{kg}\cdot\text{K}$], Δt : timestep of simulation [s]; t+1 denotes the next time step.

3. Global high-res urban representation for km-level Earth system modeling

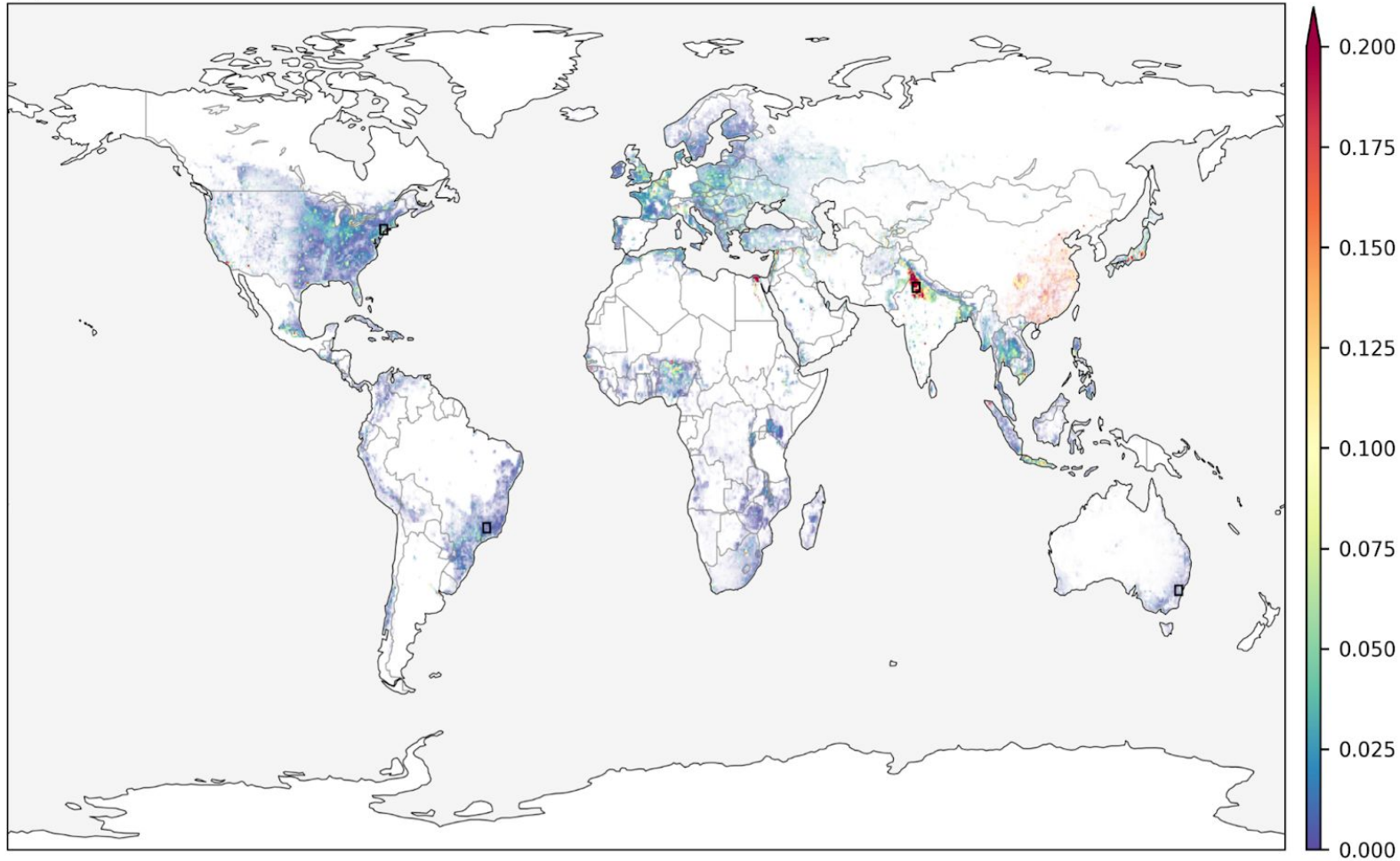
Urban surface parameters





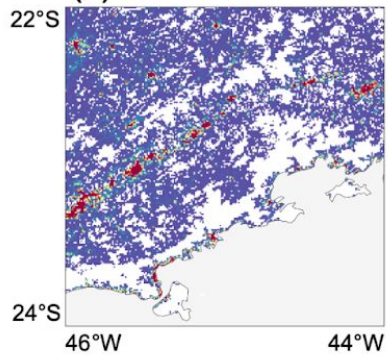
Category	Urban Parameters	This Study
Radiative	Roof Emissivity	<ul style="list-style-type: none"> Source: 100m ASTER v3 satellite emissivity product Time span: 2000-2008 Resolution: 1km
	Impervious Road Emissivity	
	Pervious Road Emissivity	
	Wall Emissivity	
	Roof Albedo	<ul style="list-style-type: none"> Source: 10m Sentinel2 satellite albedo product and NTB algorithm Time span: 2019-2021, 2021-2022 Resolution: 1km
	Impervious Road Albedo	
	Pervious Road Albedo	
Wall Albedo		
Morphological	Roof Height	<ul style="list-style-type: none"> Source: Time span: Resolution: 1km
	Canyon Height-to-width Ratio	<ul style="list-style-type: none"> Source: infinite canyon street model (Masson et al., 2020) Time span: Resolution: 1km
	Roof Fraction	<ul style="list-style-type: none"> Source: Microsoft global building footprints Time span: 2014-2021 Resolution: 1km
	Pervious Road Fraction	<ul style="list-style-type: none"> Source: 10m ESA Worldcover v200 Time span: 2021-2022 Resolution: 1km
	Urban Percentage	<ul style="list-style-type: none"> Source: 1km global urban extent (Gao and O'Neill, 2020) Time span: 2000 Resolution: 1km
Thermal	Air Conditioning Penetration Rate	<ul style="list-style-type: none"> Source: global AC penetration rate (Li et al., 2024) Time span: Resolution: national-level
	Number of Impervious Road Layers	<ul style="list-style-type: none"> Source: local building codes, municipal documentation Time span: Resolution: Regional-level, density-class-specific
	Roof Thickness	
	Wall Thickness	
	Minimum Interior Building Temperature	
	Maximum Interior Building Temperature	
	Roof Thermal Conductivity	
	Impervious Road Thermal Conductivity	
	Wall Thermal Conductivity	
	Roof Volumetric Heat Capacity	
Impervious Road Volumetric Heat Capacity		
Wall Volumetric Heat Capacity		

(a)

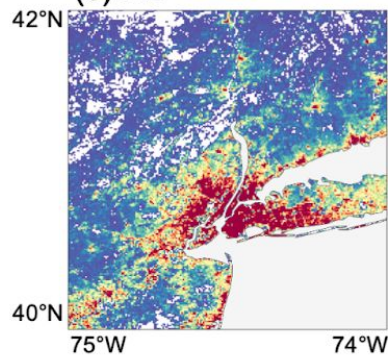


Spatial distribution of roof fraction

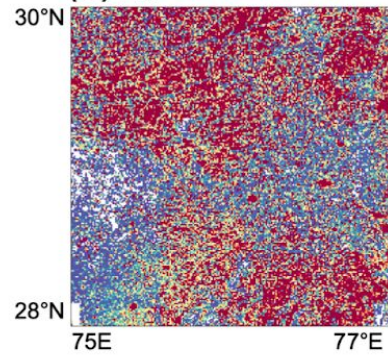
(b) R1



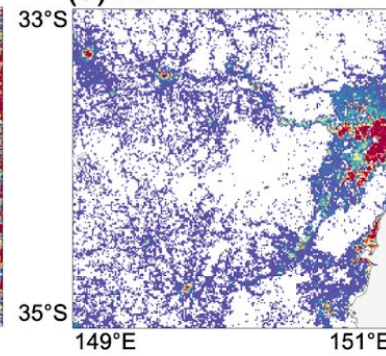
(c) R2

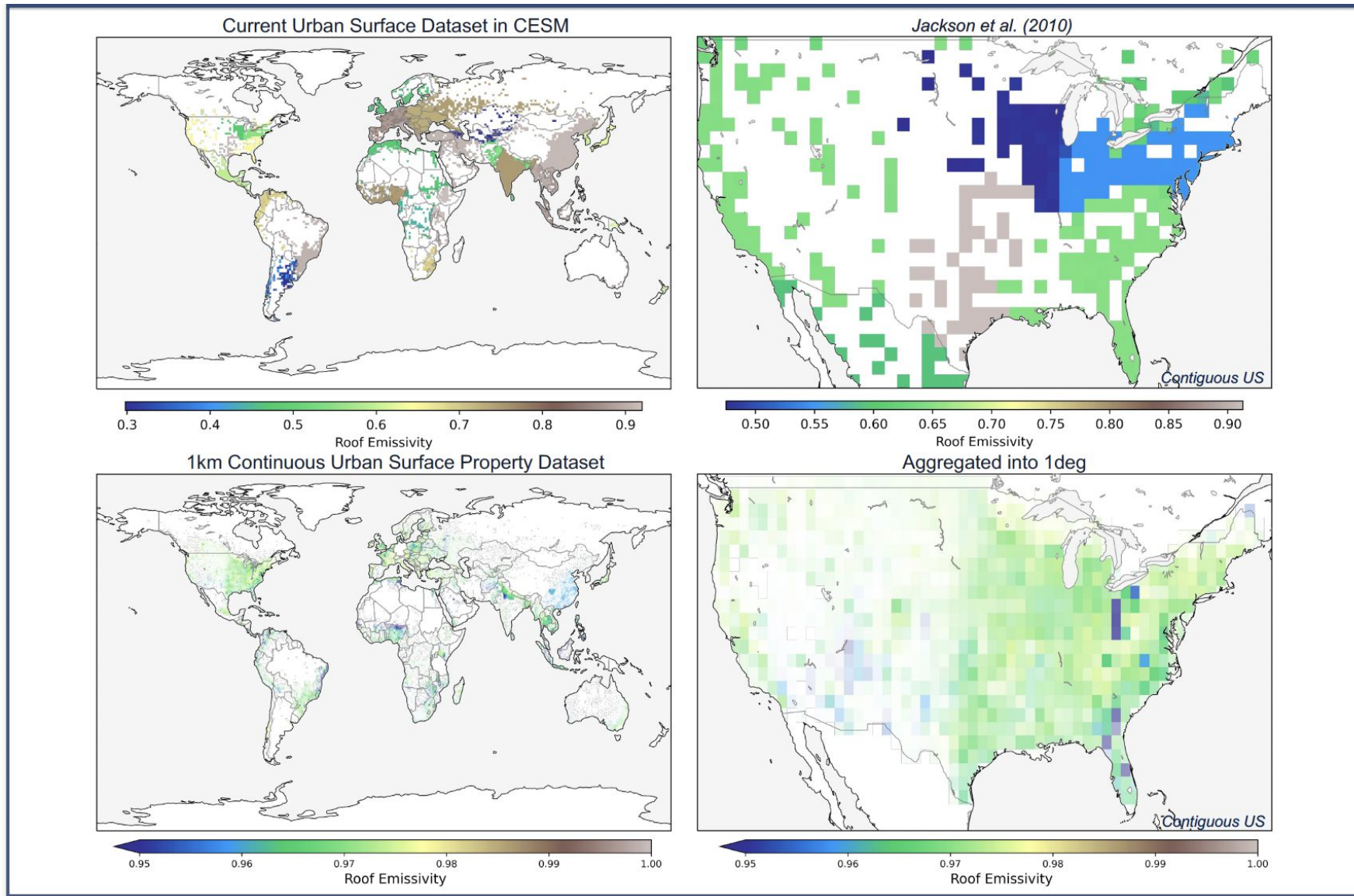


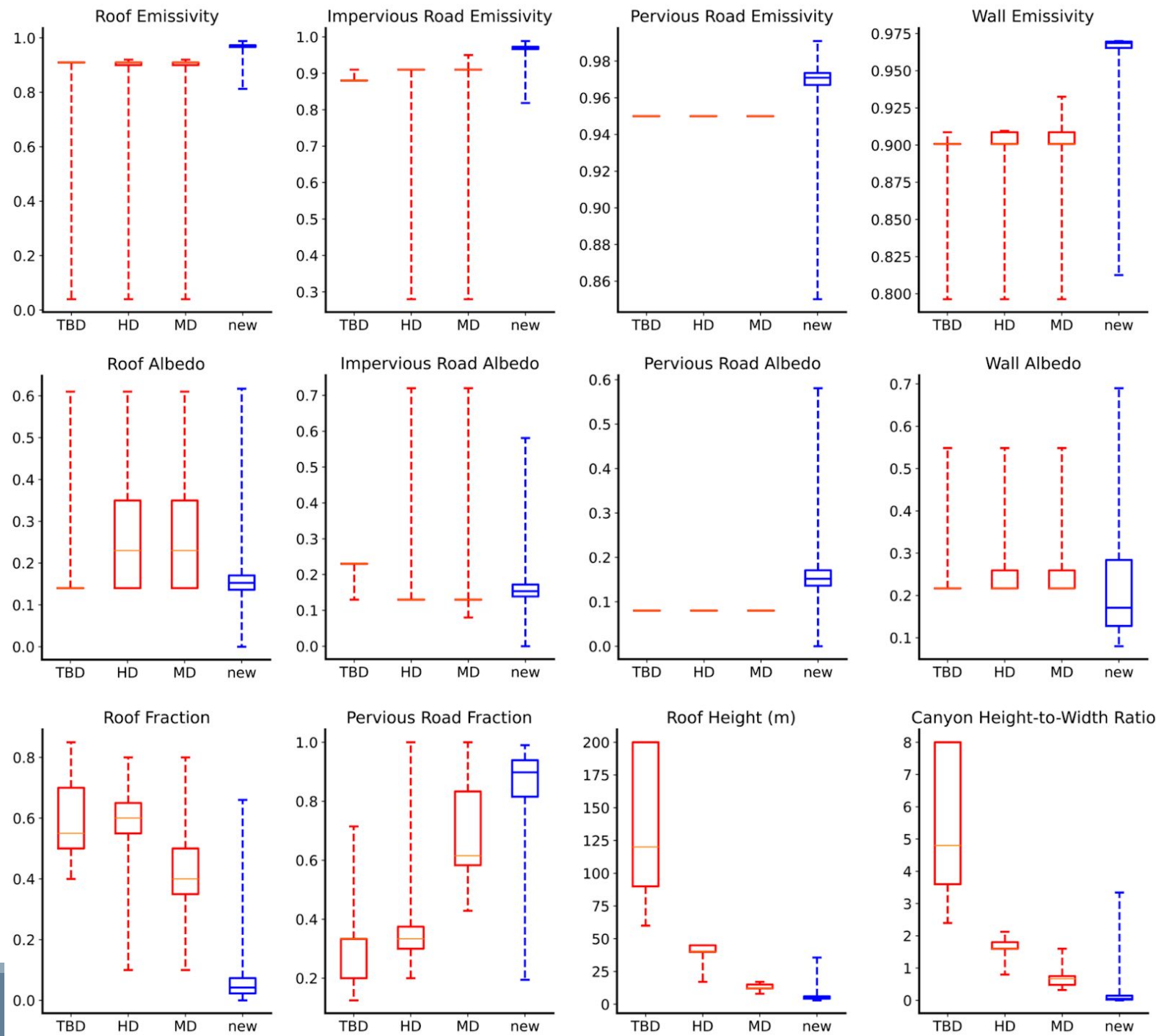
(d) R3



(e) R4



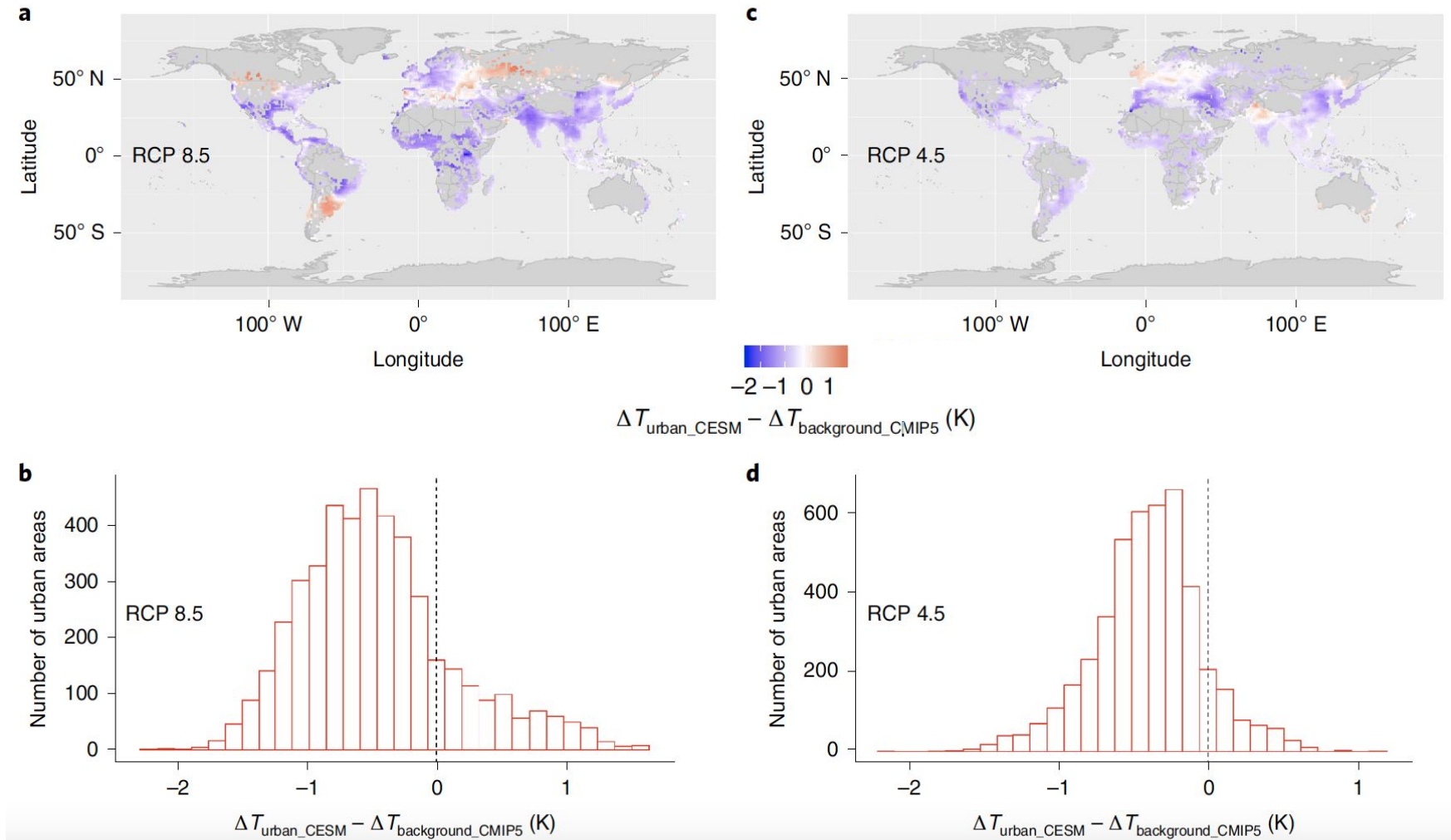




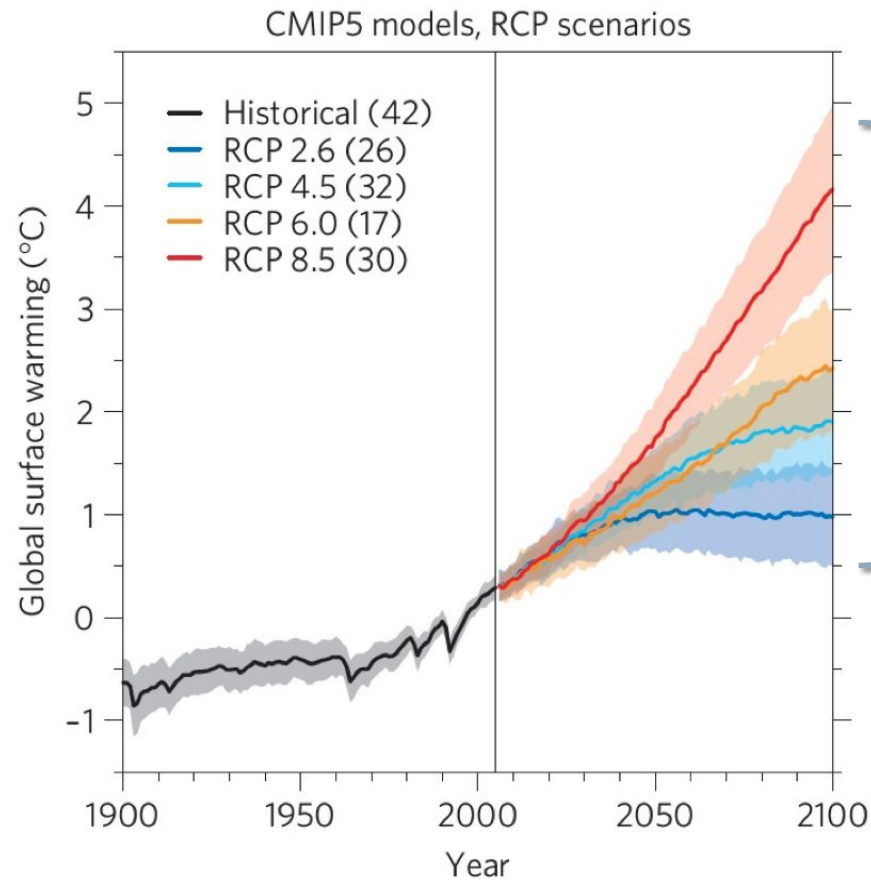
Thank you!

Comments & questions?

Traditional CMIP multi-model projections fail to capture urban warming signals



Understanding the uncertainties



Internal variability

+

Structural uncertainty

+

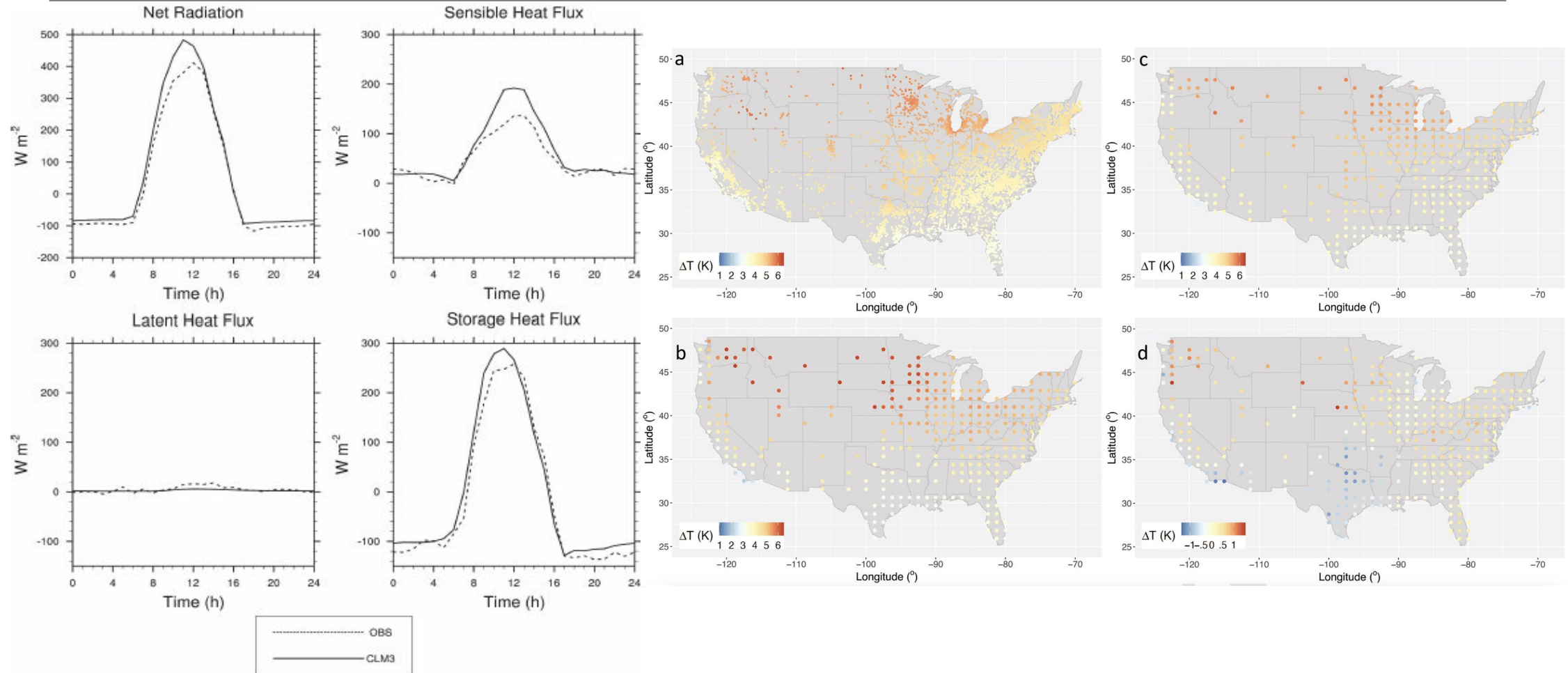
Parameter uncertainty

+

Scenario uncertainty

Dominant source of uncertainty at the time horizons of multiple decades or longer

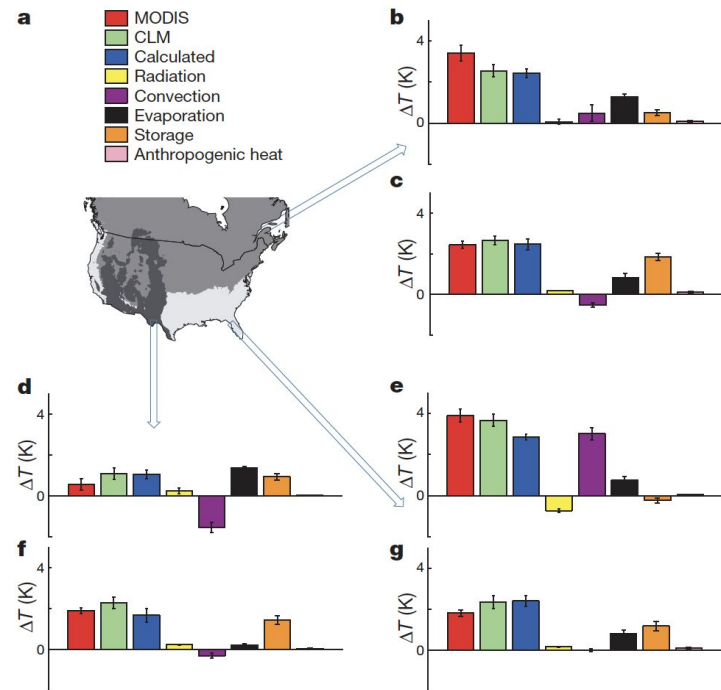
Evaluation of the CESM urban modeling



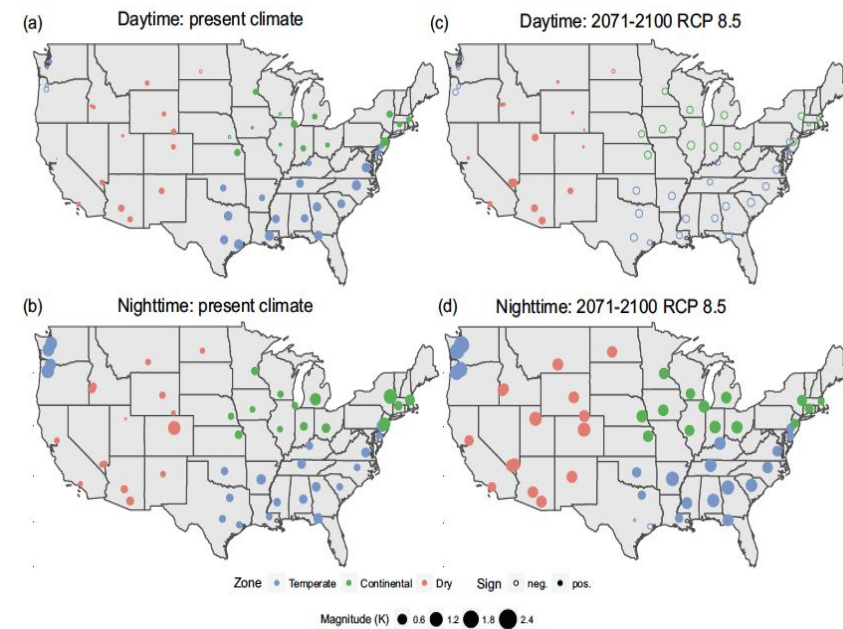
Source: Oleson et al. 2008

Source: Zhao et al. *Nature Climate Change* (2021)

Urban warmth

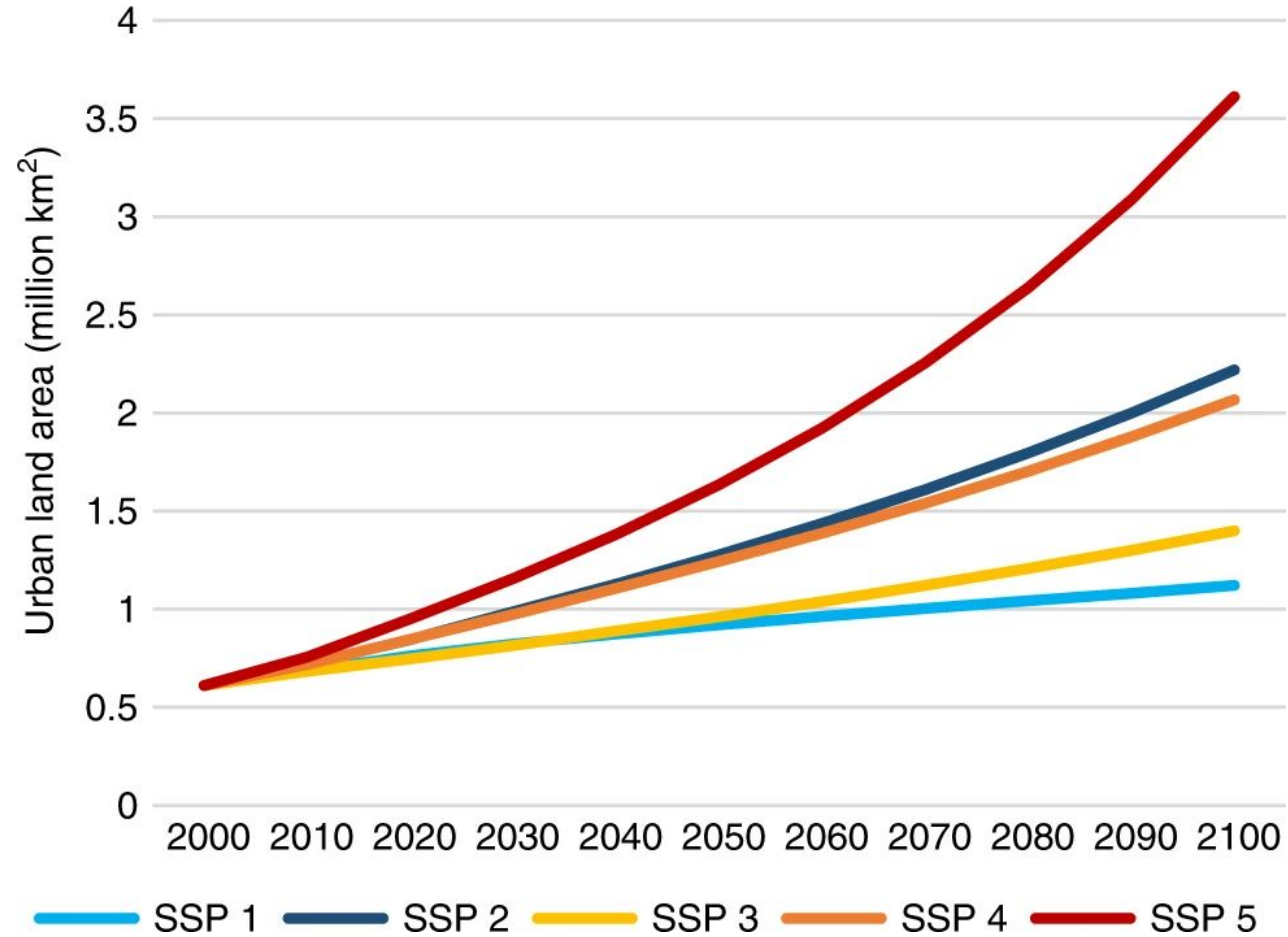


Ref: Zhao et al. *Nature* 2014



Ref: Zhao et al. *Env. Res. Lett* 2018

Urban growth



Gao, Jing, and Brian C. O'Neill, 2020