

1 **Supplementary Information (SI)**

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3 **Divergent urbanization-induced impacts on global surface urban**  
4 **heat island trends since 1980s**

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8 and Shasha Wang

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10 **This file includes:**

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12 **A. Supplementary figures**

13 Figures [S1 to S12](#)

14

15 **B. Supplementary tables**

16 Table [S1](#)

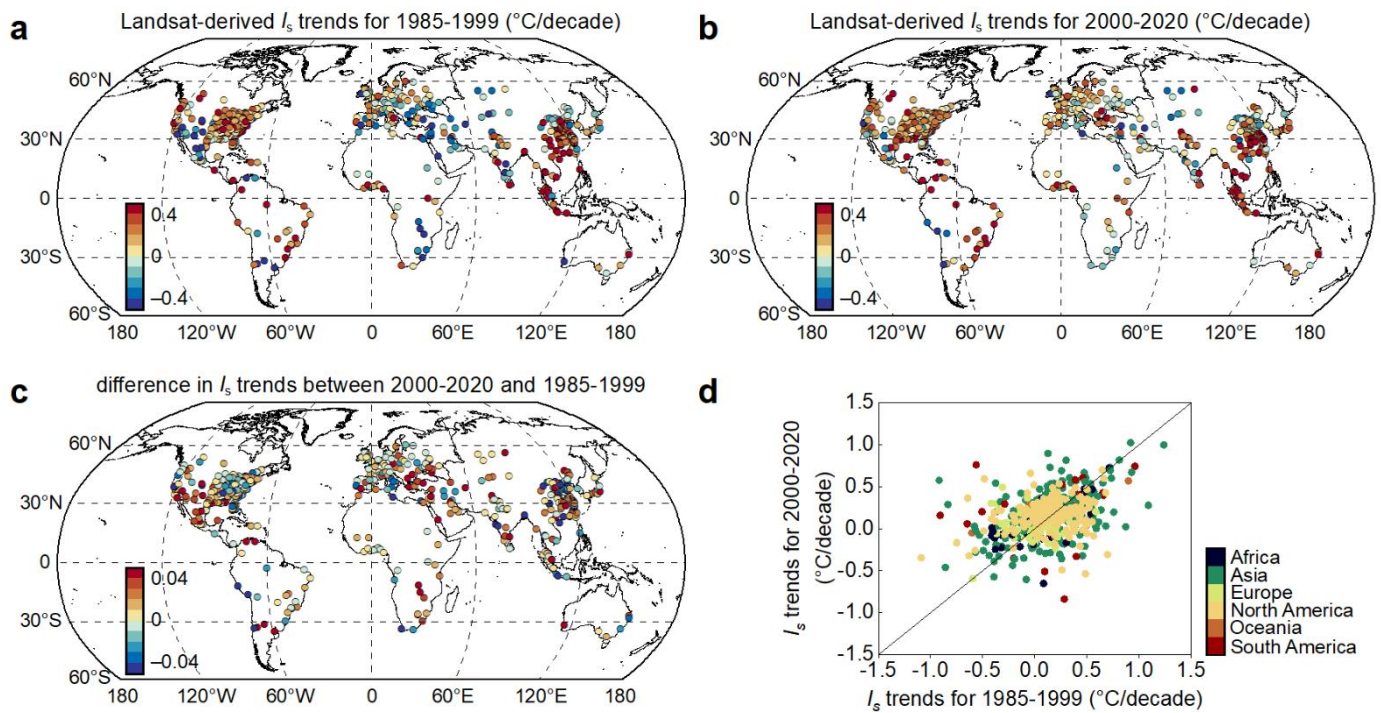
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20 **A. Supplementary figures**

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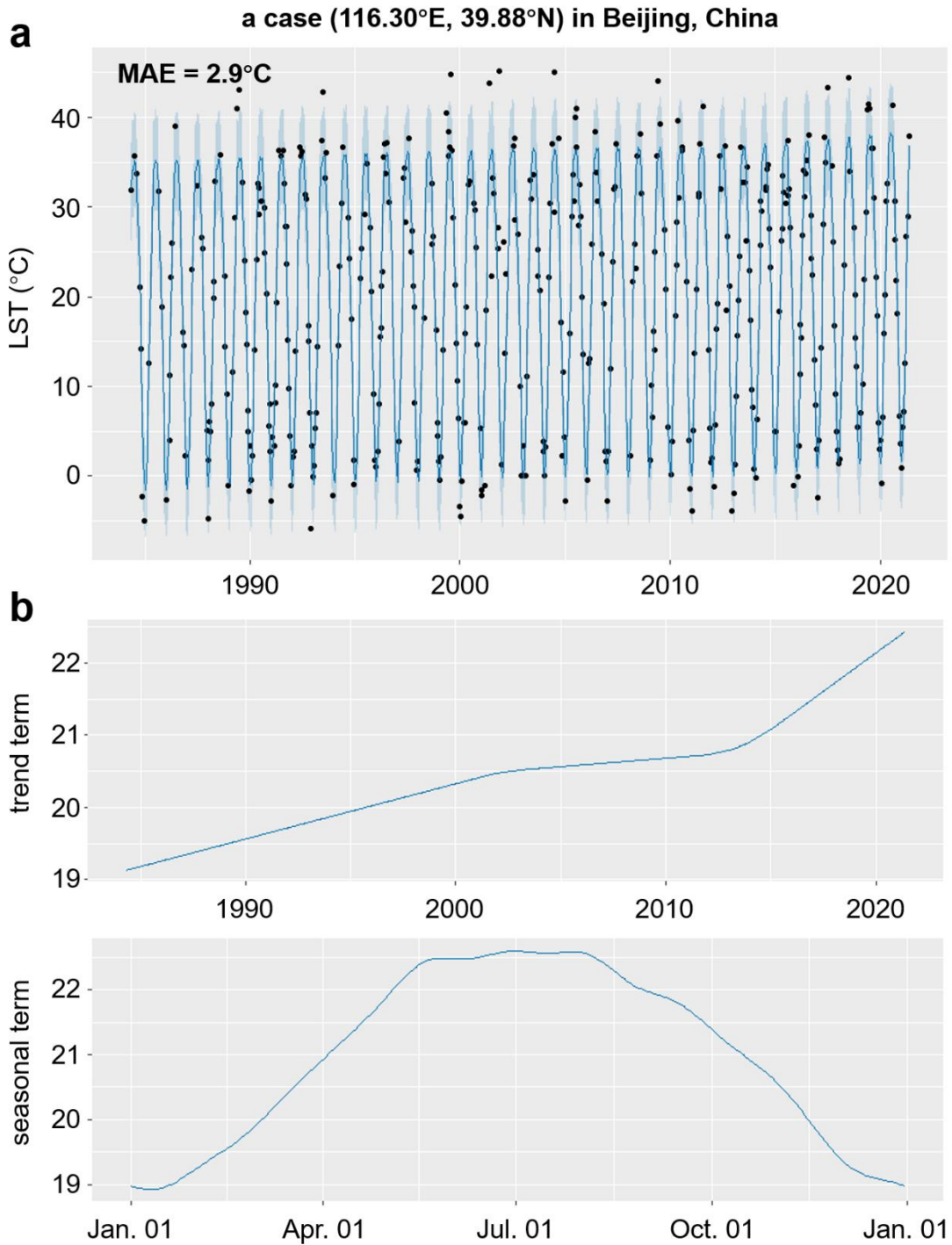
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23 **Figure S1. Comparison of surface urban heat island intensity ( $I_s$ ) trends derived from reconstructed**  
24 **Landsat land surface temperature (LST) data across global major cities. The  $I_s$  trends ( $^{\circ}\text{C}/\text{decade}$ ) for**  
25 **the periods of 1985 – 1999 (a) and 2000 – 2020 (b), difference in  $I_s$  trends between these two periods (c),**  
26 **and the relationship of  $I_s$  trends between the two periods (d).**

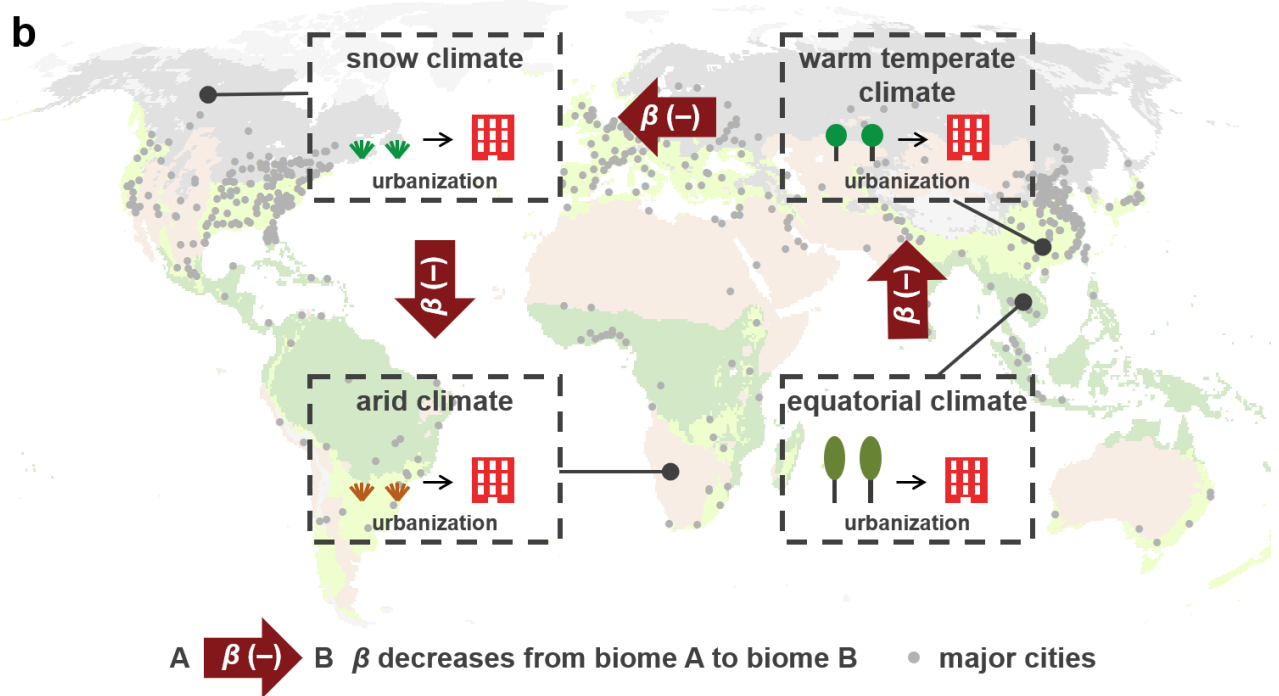
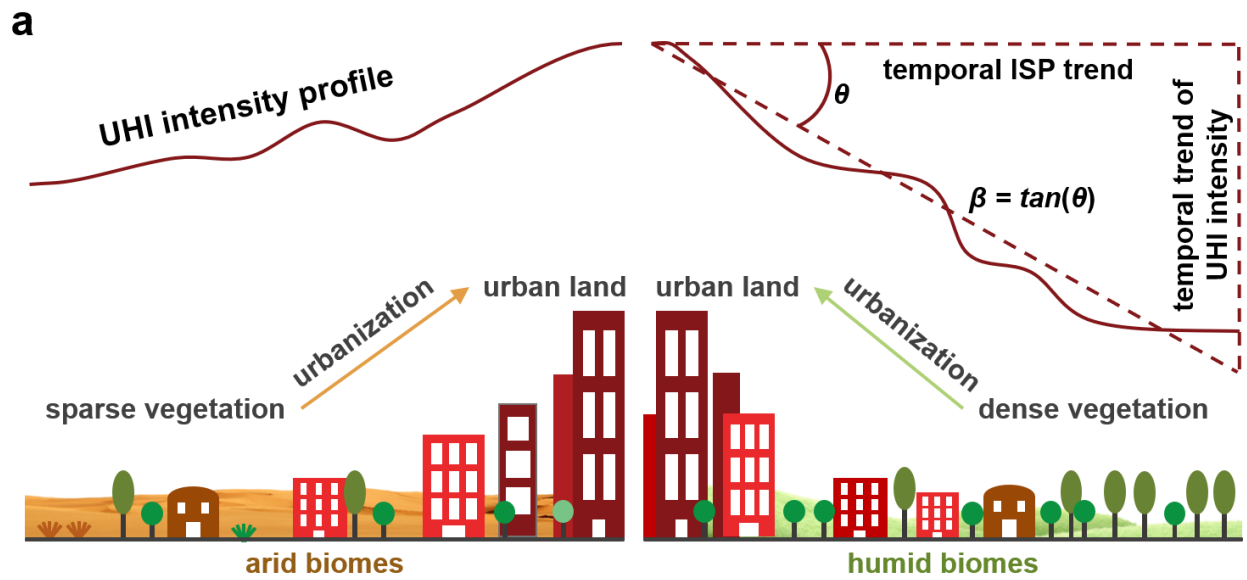
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**Figure S2.** A case (116.30°E, 39.88°N) in Beijing, China to show the procedure of the Prophet model to reconstruct Landsat LST. Observed land surface temperature (LST, i.e., black dots) versus reconstructed LST curve (i.e., blue line) (**a**); the trend and seasonal terms obtained by using the Prophet model showing the interannual and seasonal variations of LST (**b**).



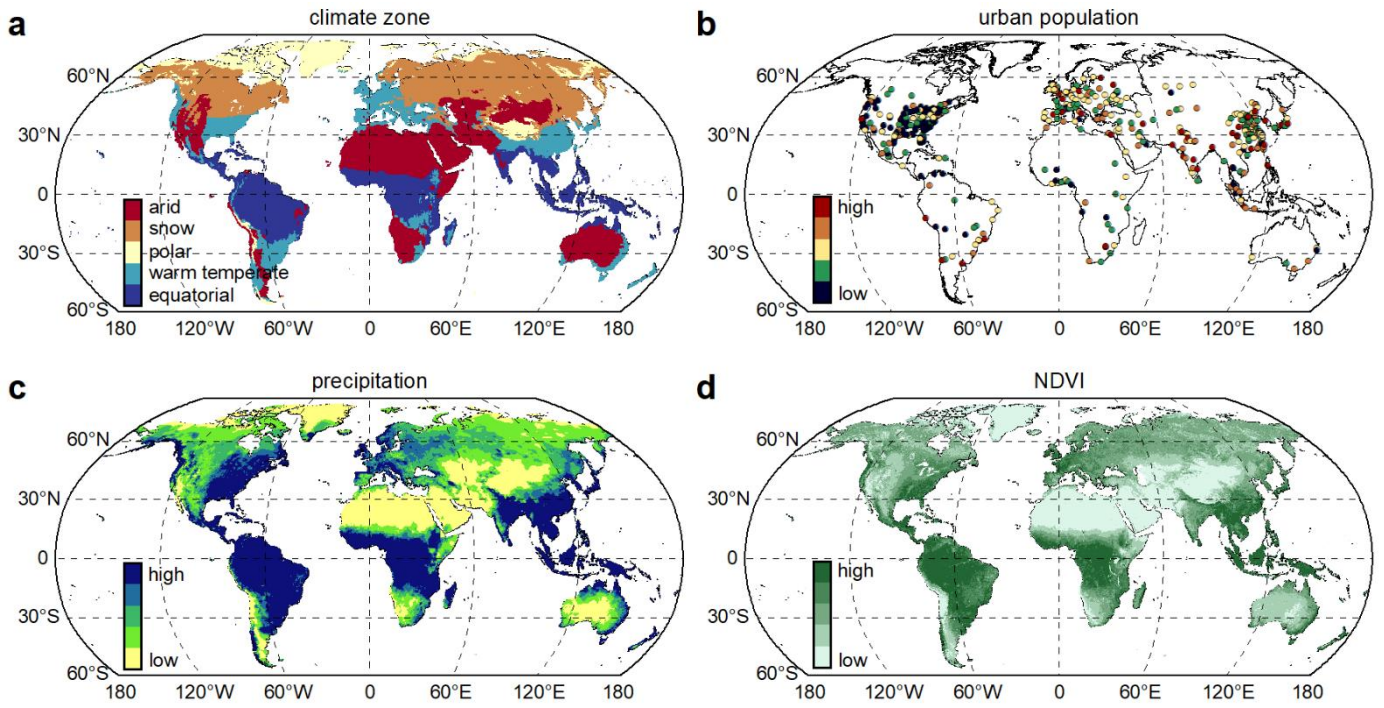
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39 **Figure S3.** Conceptual diagram that displays the contrasts of urbanization-induced impacts on surface  
 40 **urban heat island intensity trends ( $\beta$ ) across biome backgrounds.** The conceptual illustration for the  
 41 calculation of  $\beta$  (a), and the comparison of  $\beta$  across biome backgrounds (b). The  $\beta$  decreases gradually from  
 42 equatorial to warm temperature, snow, and arid climates. The UHI and ISP are the abbreviations for urban  
 43 heat island and impervious surface percentage, respectively.

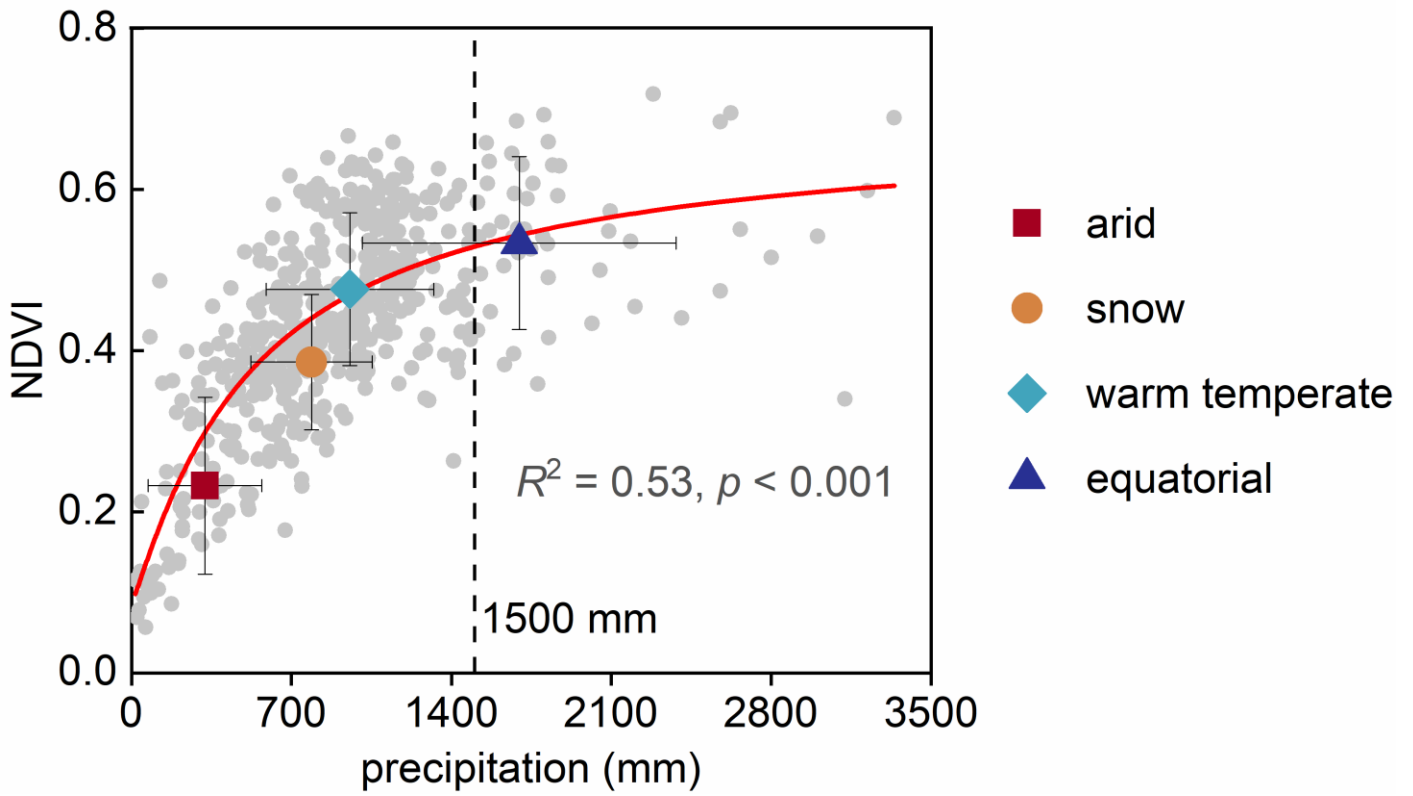
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 48 **Figure S4.** Spatial distribution of global major 511 cities as well as the corresponding climate zone,  
 49 **urban population, precipitation, and normalized difference vegetation index (NDVI).** Climate zone (a),  
 50 **urban population (b), precipitation (c), and NDVI (d).**



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55 **Figure S5. Relationship between rural NDVI and annual total precipitation across global cities.** The

56 rural NDVI generally increases with precipitation when precipitation is < 1500 mm, but there is no

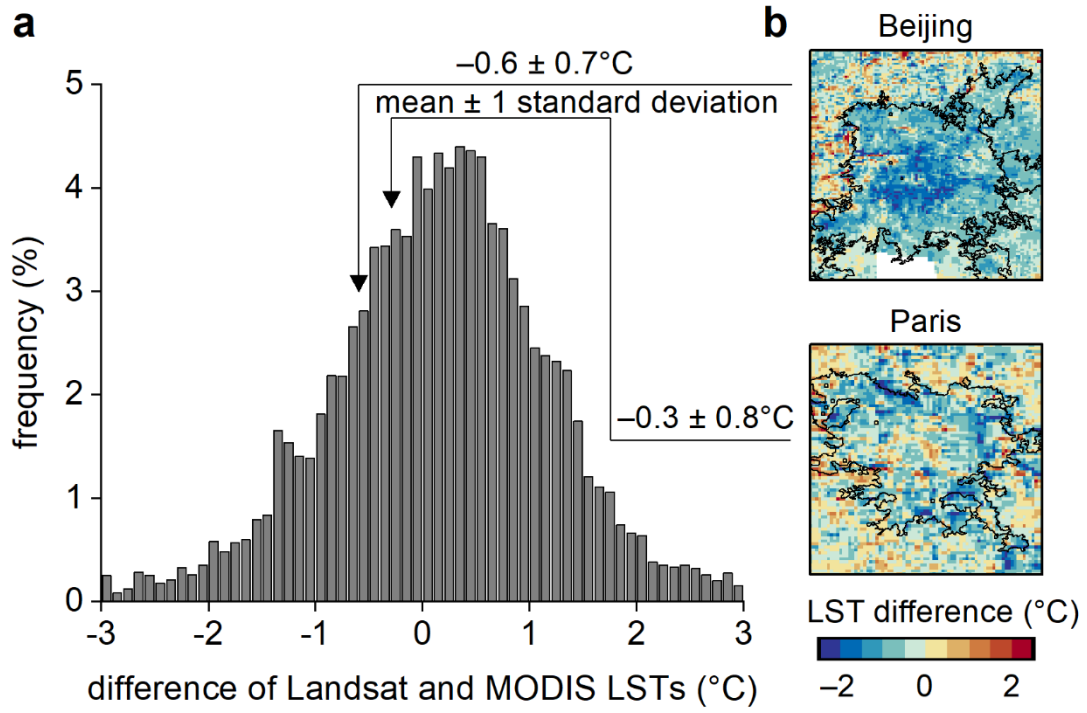
57 significant relationship between rural NDVI and precipitation when precipitation is > 1500 mm.

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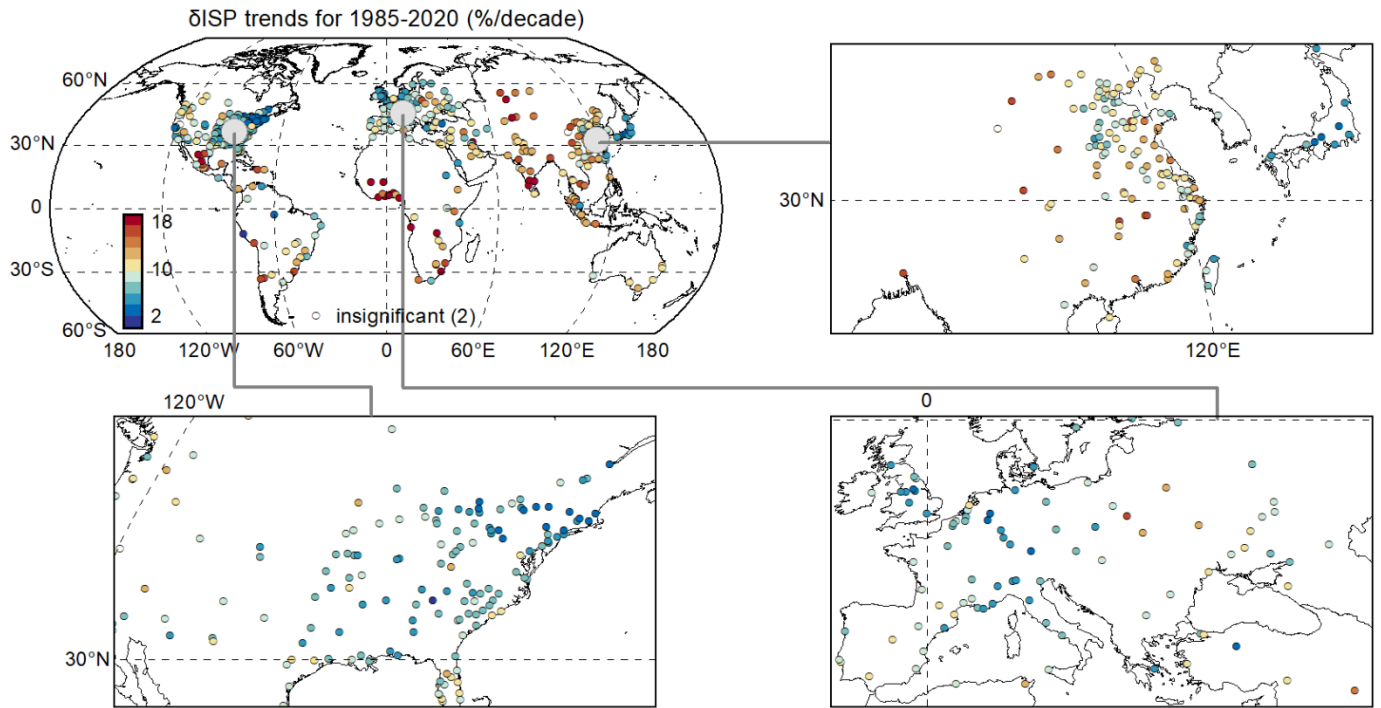
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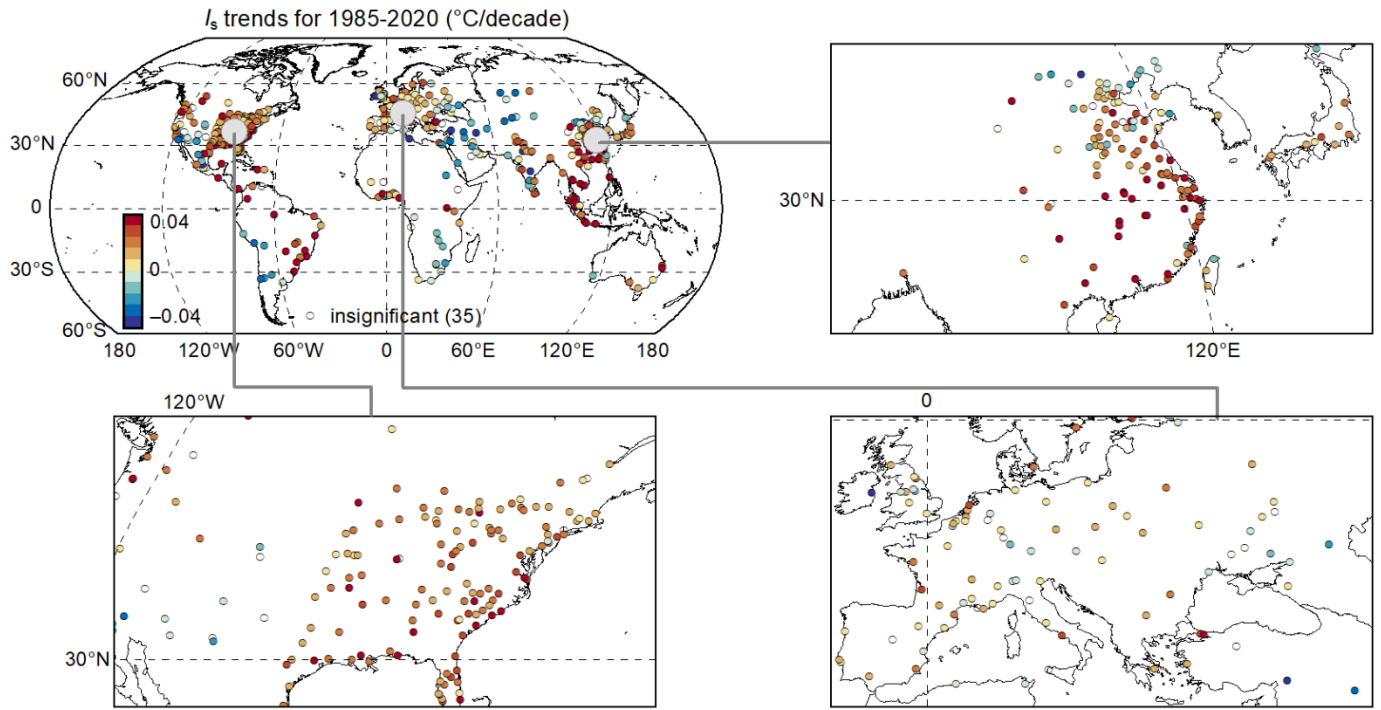


**Figure S6.** The spatial differences between the reconstructed Landsat land surface temperature (LST) and MODIS LST. The frequency distribution of LST difference over all pixels across global cities (a); and the spatial patterns of LST difference in Beijing, China and Paris, France (b).



68  
 69 **Figure S7. Urban-rural contrast in impervious surface percentage ( $\delta$ ISP) trends since 1985 across**  
 70 **global cities, with cities in the regions of North America, Europe, and Asia being enlarged further. The**  
 71  **$\delta$ ISP trends were estimated from Landsat observations.**





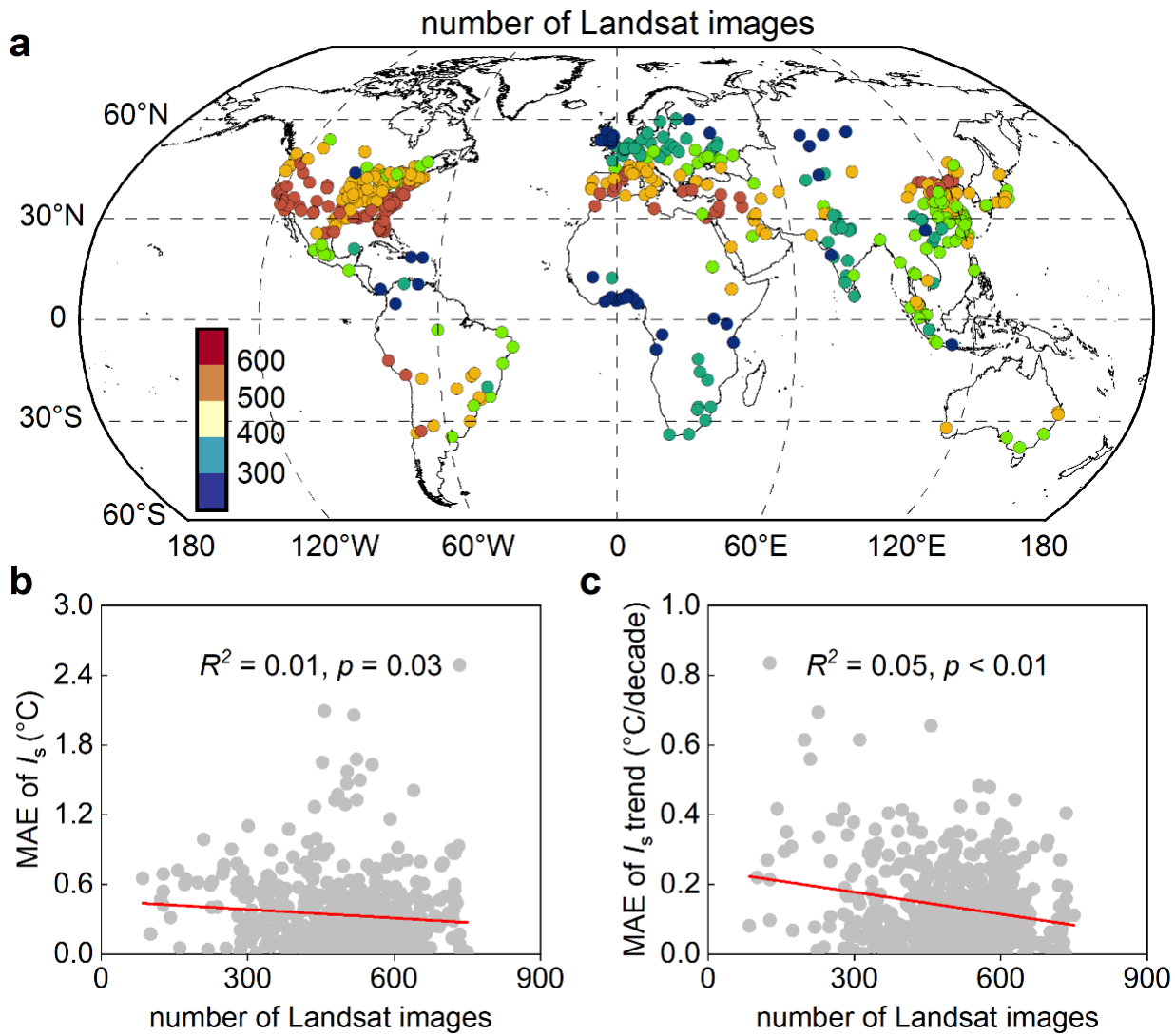
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76 **Figure S8.** Surface urban heat island intensity ( $I_s$ ) trends since 1985 across global cities, with cities in  
 77 the regions of North America, Europe, and Asia being enlarged further. The  $I_s$  trends were estimated  
 78 from Landsat observations.

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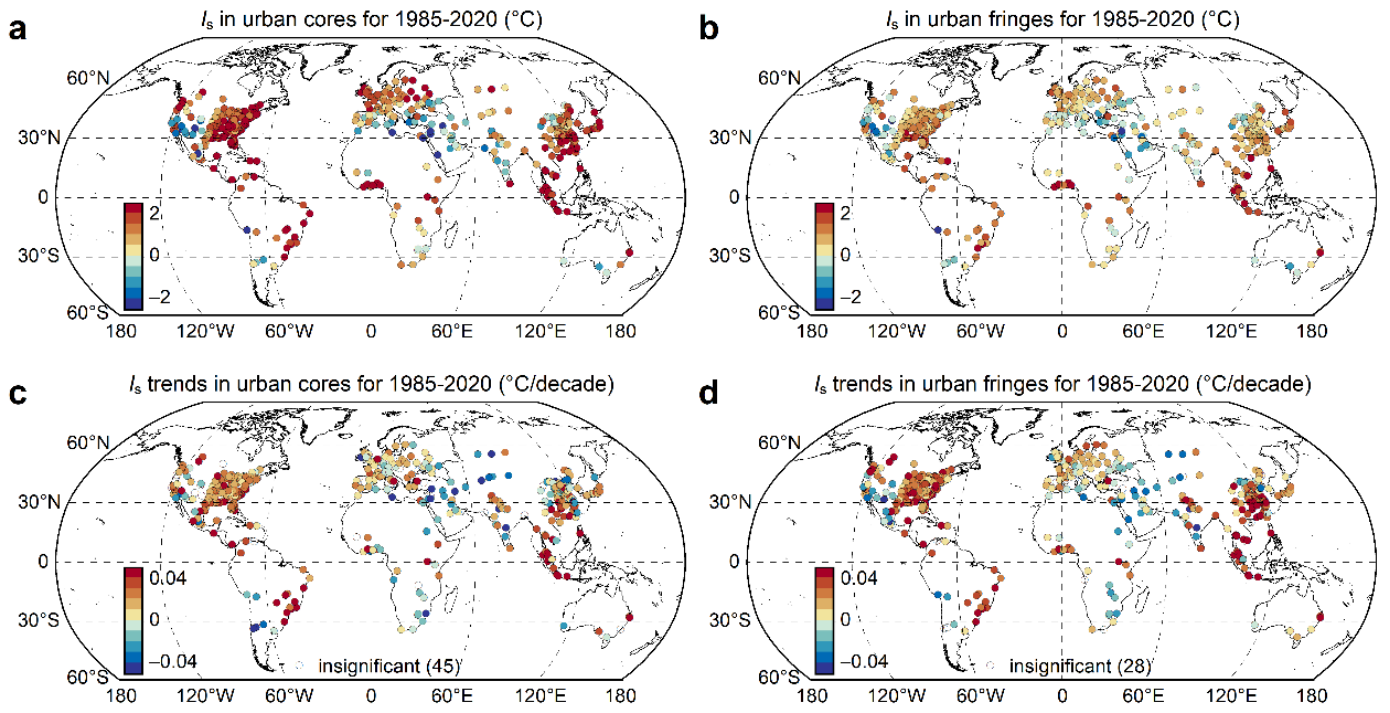
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84 **Figure S9.** The impacts of disparities in the amount of available Landsat land surface temperature  
85 (LST) images on the mean absolute error (MAE) of surface urban heat island intensity ( $I_s$ ) and  $I_s$   
86 trends across global cities. Number of available Landsat LST imageries (a); relationships between the  
87 number of available Landsat LST imageries and MAE of  $I_s$  (b) and  $I_s$  trends (c) across global cities.

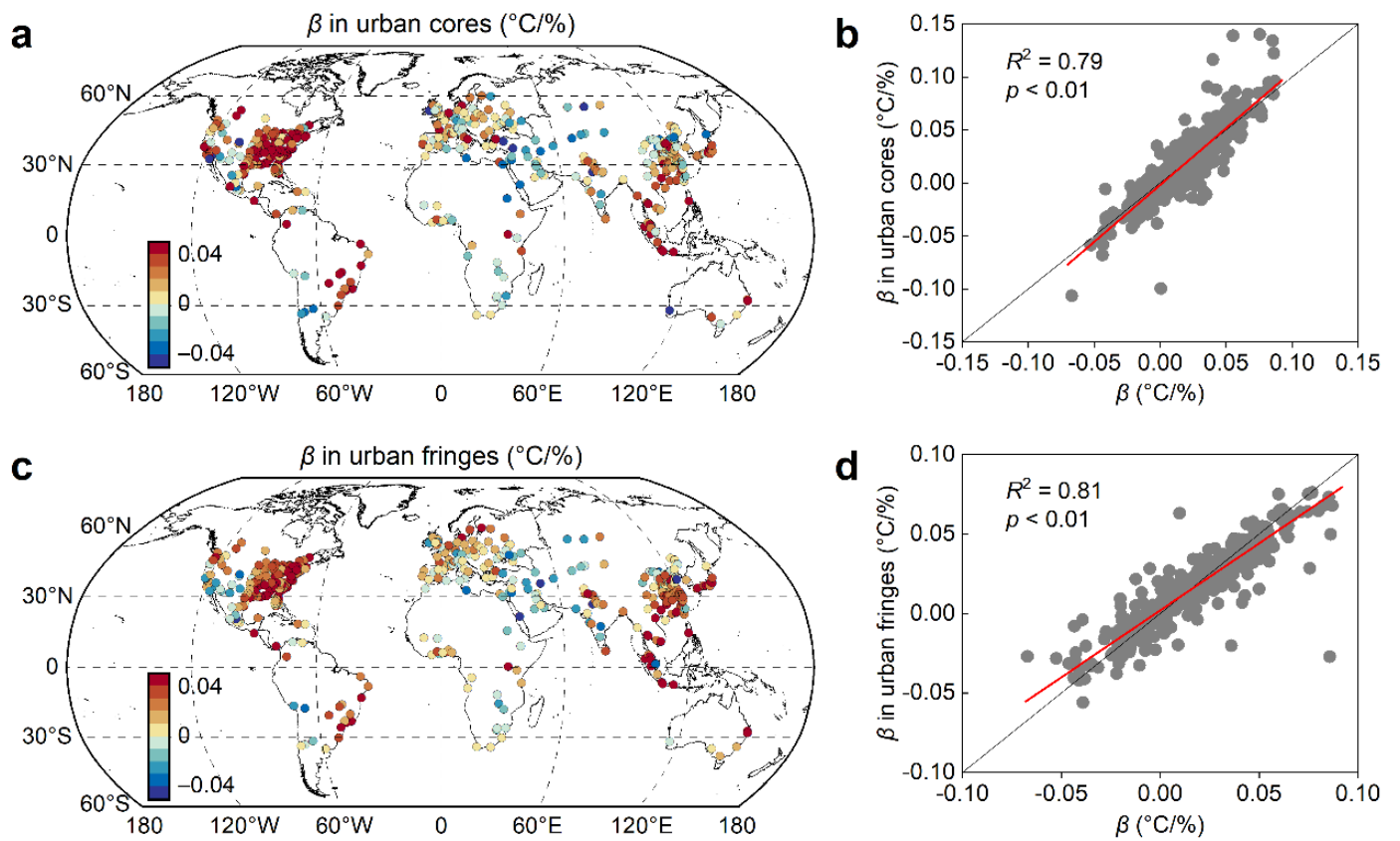
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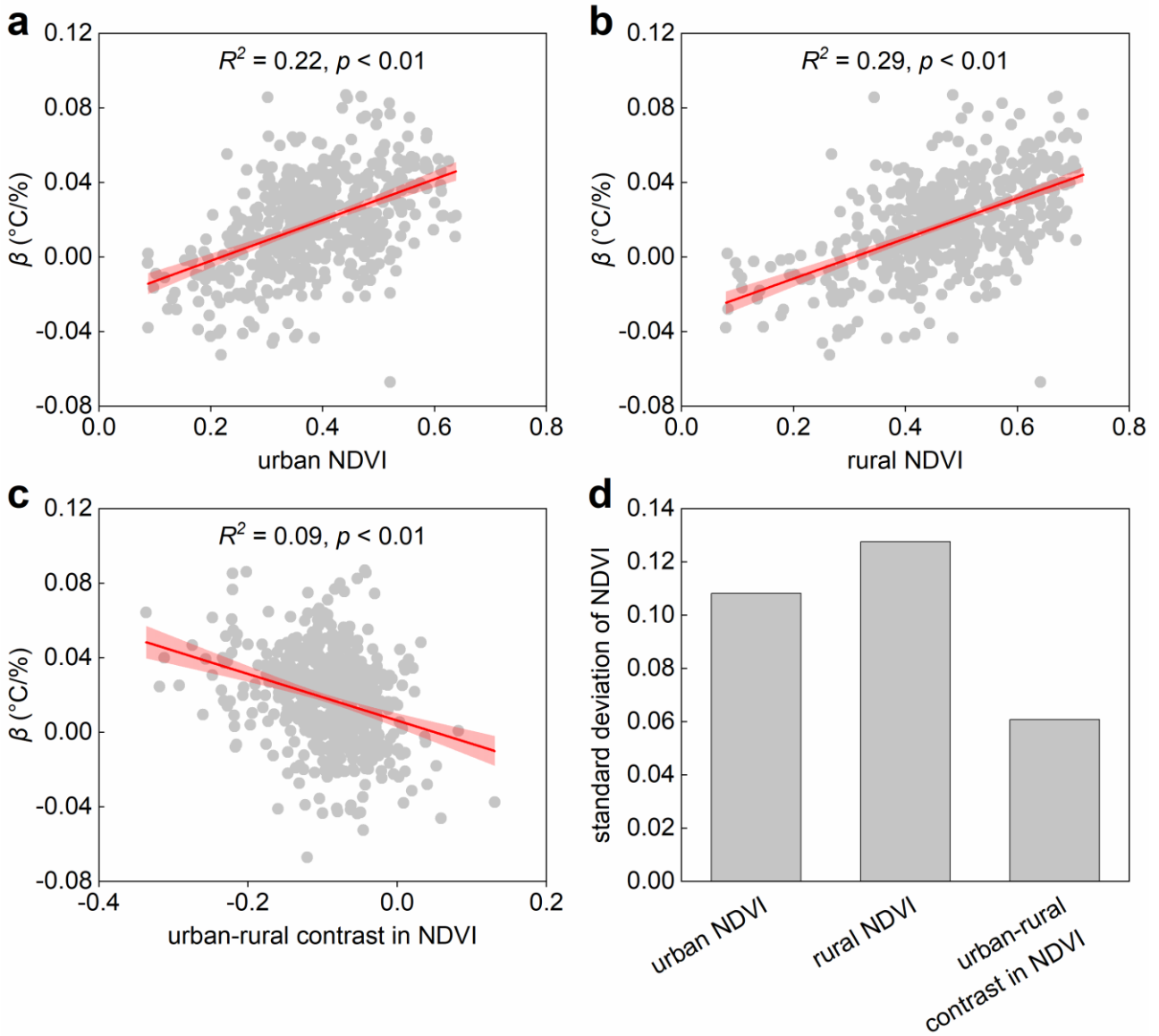
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**Figure S10.** The surface urban heat island intensity ( $I_s$ ) and  $I_s$  trend in different urban categories for 1985 – 2020.  $I_s$  in urban cores (a);  $I_s$  in urban fringes (b);  $I_s$  trend in urban cores (c); and  $I_s$  trend in urban fringes (d).



**Figure S11.** Spatial variations of urbanization-induced impacts on surface urban heat island intensity trends ( $\beta$ ) in different urban categories.  $\beta$  in urban cores (a); correlation of  $\beta$  in urban cores with  $\beta$  (i.e., without distinguishing between urban categories) (b);  $\beta$  in urban fringes (c); and correlation of  $\beta$  in urban fringes with  $\beta$  (i.e., without distinguishing between urban categories) (d).



**Figure S12.** Correlations of urbanization-induced impacts on surface urban heat island intensity trends ( $\beta$ ) with normalized difference vegetation index (NDVI). Relationship of  $\beta$  with urban NDVI (a), rural NDVI (b), and urban-rural contrast in NDVI (c); and standard deviation of three indicators (d).

113 **B. Supplementary tables**

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115 **Table S1. The ranking of the  $\beta$  (i.e., urbanization-induced impacts on surface urban heat island**116 **intensity trends) across countries. The mean and standard deviation (SD) of the  $\beta$  for each country.**

<b>List</b>	<b>Country</b>	<b>Abbreviation of country name</b>	<b>Mean</b>	<b>SD</b>	<b>Number of cities</b>
1	Guatemala	GTM	0.0615	0	1
2	Malaysia	MYS	0.0605	0.0222	4
3	Indonesia	IDN	0.0558	0.0204	6
4	Philippines	PHL	0.0472	0	1
5	Panama	PAN	0.0471	0	1
6	Uganda	UGA	0.0470	0	1
7	Columbia	COL	0.0467	0	1
8	Vietnam	VNM	0.0444	0.0036	2
9	Denmark	DNK	0.0435	0	1
10	Thailand	THA	0.0401	0.0042	2
11	Brazil	BRA	0.0378	0.0129	13
12	Finland	FIN	0.0375	0	1
13	America	USA	0.0323	0.0233	145
14	Sri Lanka	LKA	0.0322	0	1
15	Sweden	SWE	0.0318	0	1
16	Japan	JPN	0.0281	0.0204	10
17	Tanzania	TZA	0.0268	0	1
18	Cambodia	KHM	0.0266	0	1
19	Netherlands	NLD	0.0265	0.0094	2
20	Canada	CAN	0.0255	0.0168	12

21	Dominica	DOM	0.0252	0	1
22	Australia	AUS	0.0212	0.0365	6
23	Romania	ROM	0.0211	0	1
24	Myanmar	MMR	0.0180	0	1
25	Bangladesh	BGD	0.0167	0	1
26	Nigeria	NGA	0.0163	0.0092	5
27	China	CHN	0.0163	0.0187	118
28	Belarus	BLR	0.0157	0	1
29	Pakistan	PAK	0.0154	0.0215	4
30	Bulgaria	BGR	0.0148	0	1
31	Belgium	BEL	0.0145	0.0050	3
32	Poland	POL	0.0139	0.0087	4
33	Ghana	GHA	0.0138	0.0244	2
34	Korea	KOR	0.0137	0	1
35	Portugal	PRT	0.0127	0.0078	3
36	Kenya	KEN	0.0125	0	1
37	France	FRA	0.0115	0.0140	13
38	Benin	BEN	0.0096	0	1
39	Cote d'Ivoire	CIV	0.0095	0	1
40	Puerto Rico	PRI	0.0089	0	1
41	Czech Republic	CZE	0.0070	0	1
42	Italy	ITA	0.0063	0.0164	8
43	Togo	TGO	0.0060	0	1
44	Great Britain	GBR	0.0059	0.0131	9
45	Hungary	HUN	0.0058	0	1
46	Spain	ESP	0.0055	0.0100	6



47	Greece	GRC	0.0041	0	1
48	Angola	AGO	0.0026	0	1
49	Egypt	EGY	0.0021	0.0203	2
50	India	IND	0.0021	0.0222	14
51	Qatar	QAT	0.0019	0	1
52	Morocco	MAR	0.0009	0	1
53	Mali	MLI	0.0007	0	1
54	Ethiopia	ETH	0.0001	0	1
55	Burkina Faso	BFA	-0.0005	0	1
56	Turkey	TUR	-0.0013	0.0382	5
57	Ukraine	UKR	-0.0019	0.0090	7
58	Kazakhstan	KAZ	-0.0024	0.0014	2
59	Germany	DEU	-0.0025	0.0112	8
60	South Africa	ZAF	-0.0028	0.0058	5
61	Russia	RUS	-0.0029	0.0141	11
62	Venezuela	VEN	-0.0030	0.0171	2
63	Tunisia	TUN	-0.0040	0	1
64	Mexico	MEX	-0.0049	0.0217	11
65	Syria	SYR	-0.0052	0	1
66	Congo	COD	-0.0070	0.0007	2
67	Algeria	DZA	-0.0085	0	1
68	Singapore	SGP	-0.0104	0	1
69	Zimbabwe	ZWE	-0.0105	0	1
70	Mozambique	MOZ	-0.0116	0	1
71	Austria	AUT	-0.0116	0	1
72	Chile	CHL	-0.0125	0	1

73	Azerbaijan	AZE	-0.0125	0	1
74	Kuwait	KWT	-0.0165	0	1
75	Saudi Arabia	SAU	-0.0166	0.0153	3
76	Argentina	ARG	-0.0178	0.0085	3
77	Iran	IRN	-0.0178	0.0212	2
78	Zambia	ZMB	-0.0179	0	1
79	Iraq	IRQ	-0.0181	0.0053	3
80	Israel	ISR	-0.0192	0	1
81	Kirghizia	KGZ	-0.0197	0	1
82	Uzbekistan	UZB	-0.0205	0	1
83	Bolivia	BOL	-0.0220	0	1
84	Peru	PER	-0.0245	0.0035	2
85	Turkmenistan	TKM	-0.0246	0	1
86	Sultan	SDN	-0.0282	0	1
87	Jordan	JOR	-0.0313	0	1
88	Libya	LBY	-0.0409	0	1
89	Ireland	IRL	-0.0671	0	1

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