Supplementary Tables

Supplementary Table 1. Summary of previous research on temperature and outdoor activities.

Litanatuna	Data	Extrama aald	Extrama hat
Literature	Data	Extreme cold	Extreme hot
		temperature	temperature
(Chan and Wichman	Daily bicycle trips (bikeshare	< 30 °F (~ -1 °C):	> 80 °F (~
2017)	usage)	-75%	27 °C):
			no sig. effect
(Dundas and von	Daily marine recreational fishing	< 30 °F (~ -1 °C):	80-95 °F: no sig.
Haefen 2020)	(survey)	-22%	effect
			>95 °F (~35 °C): -
			2%
(Graff Zivin and	Daily outdoor time	< 30 °F (~ -1 °C):	>95 °F (~35 °C):
Neidell 2014)	(survey)	- 40 min	-8 min
(Obradovich and	Monthly physical activity	< 0 °C:	high temperature:
Fowler 2017)	(survey)	-8-9%	no sig. effect
(Heaney et al., 2019)	Daily hours and distance of	Max T -10-0 °C:	Max T 30-35 °C:
	public bike (bikeshare usage)	~ -20% km	~ -4% km

Supplementary Table 2. Summary statistics.

			Standard		25%	75%	
	N	Mean	Deviation	Min	pctl	pctl	Max
Statistic	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Park visitation	44,425,309	260.258	591.559	0	86	290	99,925
Temperature (°C)	44,526,965	17.36	10.593	-41.7	10	25.7	48.5
Air Quality Index	44,913,818	72.818	49.647	1	42	88	500
Precipitation (mm)	44,054,119	0.128	1.146	0	0	0	133.4
Wind Speed (m/s)	44,529,555	2.891	1.639	0	1.7	3.7	34.3
Relative Humidity (%)	44,523,233	65.086	22.99	0	49	84	100
Atmos. water vapor (mm)	44,521,631	15.292	9.785	0	7	23.1	53.7
Atmos. pressure (hPa)	44,519,306	989.558	48.769	635	991	1,015	1,057
Cloud coverage (%)	45,749,465	49.365	36.717	0	9	87	100

Note: All variables except cloud coverage are summarized on hourly basis. An observation is a park-hour. Observations between Earlier than 6 AM or later than 11 PM are excluded.

Supplementary Table 3. Effect of temperature on park visitation on hourly and daily resolutions.

	Hourly			Daily	
			Minimum	Average	Maximum
Temperature (°C)	(1)	Temperature (°C)	(2)	(3)	(4)
(-10,-5]	-0.2451***	(-10,-5]	-0.0848***	-0.2602***	-0.1978***
	(0.0237)		(0.0198)	(0.0371)	(0.0294)
(-5,0]	-0.2105***	(-5,0]	-0.0295	-0.2390***	-0.1771***
	(0.0169)		(0.0152)	(0.0290)	(0.0198)
(0,5]	-0.1776***	(0,5]	0.0193	-0.1868***	-0.1689***
	(0.0111)		(0.0169)	(0.0168)	(0.0175)
(5,10]	-0.1279***	(5,10]	0.0443***	-0.1197***	-0.127***
	(0.0079)		(0.0121)	(0.0144)	(0.0124)
(10,15]	-0.0770***	(10,15]	0.0228***	-0.0481***	-0.0589***
	(0.0059)		(0.0063)	(0.0088)	(0.0064)
(15,20]	-0.0334***	(20,25]	-0.0679***	0.0466***	0.0565***
	(0.0041)		(0.0081)	(0.0088)	(0.0083)
(25,30]	0.0039	(25,30]	-0.1633***	0.0743***	0.1264***
	(0.0034)		(0.0179)	(0.0089)	(0.0135)
(30,35]	-0.0548***	(30,35]	-0.3065***	0.1091***	0.1871***
	(0.0101)		(0.0371)	(0.0135)	(0.0112)
(35,40]	-0.1379***				
	(0.0171)				
Hour FE	Yes		No	No	No
DOW FE	Yes		Yes	Yes	Yes
Park FE	Yes		Yes	Yes	Yes
Month-City FE	Yes		Yes	Yes	Yes
Weather Controls	Yes		Yes	Yes	Yes
Observations	43,067,963		2,623,934	2,653,213	2,546,273
Adjusted R ²	0.8358		0.8128	0.8124	0.8125

Note: *p<0.01; **p<0.005; ***p<0.001. Dependent variable is ln(Park visitation +1). Control variables include air quality, precipitation, wind speed, relative humidity, atmospheric water vapor, atmospheric pressure, cloud coverage. Hourly regression has 20-25 °C as reference while daily temperature has 15-20 °C as reference. Standard errors clustered within city. In Fig.1, the regression coefficients are adjusted into percentage change by $e^{\beta} - 1$.

Supplementary Table 4. Effects of non-temperature variables.

	Main Regression		Interactions
	(1)		(2)
Precipitation (mm)	-0.0073***	(30,35] °C × Precipitation	0.0003
	(0.0005)		(0.0018)
Wind Speed (m/s)	-0.0061***	(35,40] °C × Precipitation	0.0020
	(0.0006)		(0.0031)
Relative Humidity (%)	-0.0013***	(30,35] °C × Wind Speed	-0.0019
	(0.0002)		(0.0020)
Atmos. water vapor (mm)	0.0019*	(35,40] °C × Wind Speed	-0.0027
	(0.0007)		(0.0037)
Air Pressure (hPa)	0.0006**	(30,35] °C × Humidity	0.0006
	(0.0002)		(0.0003)
Cloud Coverage (%)	-0.0001	(35,40] °C × Humidity	-0.0008
	(0.00003)		(0.0005)
Air Quality Index	0.0001		
	(0.00004)		
Temperature Bins	Yes		Yes
Hour FE	Yes		Yes
DOW FE	Yes		Yes
Park FE	Yes		Yes
Month-City FE	Yes		Yes
Observations	43,067,963		43,067,963
Adjusted R ²	0.8358		0.8361

Note: *p<0.01; **p<0.005; ***p<0.001. Dependent variable is ln(Park visitation +1). Column 1 shows the regression coefficients for weather controls, with temperature bins and fixed effects included in the regression. Column 2 shows results when further adding the interaction between temperature bins and weather variables (Only interaction results for temperature bin (30,35] °C and (35,40] °C are displayed in the table, yet all the temperature bins are included for the interaction terms). Standard errors clustered within city.

Supplementary Table 5. Heterogeneous effect of extreme temperature on park visitation.

					Subgroup
		< 0 °C	>30 °C	Observations	Z-tests
Group	Sub-sample	(1)	(2)	(3)	(4)
All	all	-0.2233***	-0.0595***	43,067,963	
		(0.0157)	(0.0096)		
By park types	city parks/ plazas	-0.2129***	-0.0858***	21,469,912	< 0 °C:
		(0.0112)	(0.0094)		P=0.6677
	tourism attractions	-0.2240***	-0.0291**	19,990,861	>30 °C:
		(0.0233)	(0.0091)		P=0***
By time	weekdays	-0.1752***	-0.0517***	28,650,887	< 0 °C:
		(0.0096)	(0.0062)		P=0***
	weekends/ holidays	-0.2780***	-0.0880***	11,294,291	>30 °C:
		(0.0179)	(0.0088)		P=0.0007***
By region	north	-0.2759***	-0.0326	16,213,628	< 0 °C:
		(0.0190)	(0.0145)		P=0.0010***
	south	-0.1947***	-0.0805***	26,645,175	>30 °C:
		(0.0156)	(0.0073)		P=0.0032***
By average	higher than median	-0.1967***	-0.0816***	27,223,458	< 0 °C:
temperature		(0.0177)	(0.0071)		P=0.0057***
	lower than median	-0.2709***	-0.0287	15,844,514	>30 °C:
		(0.0202)	(0.0138)		P=0.0007***
By park	Rural	-0.2488***	-0.0222	14,548,542	< 0 °C:
location		(0.0242)	(0.0094)		P=0.0625*
	Urban	-0.1992***	-0.0765***	28,519,430	>30 °C:
		(0.0111)	(0.0098)		P=0***
By income	higher than median	-0.2189***	-0.0584***	34,144,936	< 0 °C:
		(0.0186)	(0.0113)		P=0.6744
	lower than median	-0.2087***	-0.0794***	7,540,220	>30 °C:
		(0.0156)	(0.0069)		P=0.1127

Note: *p<0.01; **p<0.005; ***p<0.001. Dependent variable is $\ln(\text{Park visitation } +1)$. Control variables include air quality, precipitation, wind speed, relative humidity, atmospheric water vapor, atmospheric pressure, cloud coverage. Hour, day-of-week, city-by-month and park fixed effects are included in the regressions. 20-25 °C bin is set as reference. Standard errors clustered within city. In Fig.2, the regression coefficients are adjusted into percentage change by $e^{\beta} - 1$. Column 4 shows the z-test (reference) results across groups within each category to test whether there are statistically significant differences between coefficients.

Supplementary Table 6. Daily average temperature and activity in different times of the day.

	All times	Morning/ Night	Noon/ Evening
Daily Average Temperature (°C)	(1)	(2)	(3)
(-10,-5]	-0.1934***	-0.1844***	-0.1999***
	(0.0290)	(0.0272)	(0.0309)
(-5,0]	-0.1676***	-0.1657***	-0.1692***
	(0.0238)	(0.0238)	(0.0242)
(0,5]	-0.1166***	-0.1184***	-0.1172***
	(0.0091)	(0.0100)	(0.0087)
(5,10]	-0.0669***	-0.0684***	-0.0677***
	(0.0050)	(0.0059)	(0.0048)
(10,15]	-0.0286***	-0.0333***	-0.0262***
	(0.0038)	(0.0031)	(0.0045)
(20,25]	0.0336***	0.0526***	0.0198***
	(0.0053)	(0.0056)	(0.0052)
(25,30]	0.0324***	0.0796***	-0.0026
	(0.0069)	(0.0071)	(0.0075)
(30,35]	0.0128	0.0927***	-0.0468***
	(0.0102)	(0.0098)	(0.0118)
(35,40]	-0.0488***	0.0511***	-0.1225***
	(0.0135)	(0.0124)	(0.0166)
Hour FE	Yes	Yes	Yes
DOW FE	Yes	Yes	Yes
Park FE	Yes	Yes	Yes
Month-City FE	Yes	Yes	Yes
Observations	43,147,335	20,324,544	22,822,791
Adjusted R ²	0.8348	0.8389	0.8652

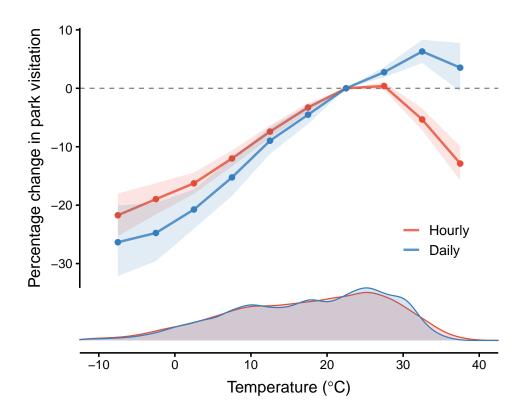
Note: *p<0.01; **p<0.005; ***p<0.001. Dependent variable is ln(Park visitation +1). Sub-samples are restricted by time periods and regressions are run separately for each sub-sample. Control variables include air quality, precipitation, wind speed, relative humidity, atmospheric water vapor, atmospheric pressure, cloud coverage. 15-20 °C bin is set as reference. Standard errors clustered within city. In Fig.3a, the regression coefficients are adjusted into percentage change by $e^{\beta} - 1$.

Supplementary Table 7. Within-day hourly activity distribution patterns.

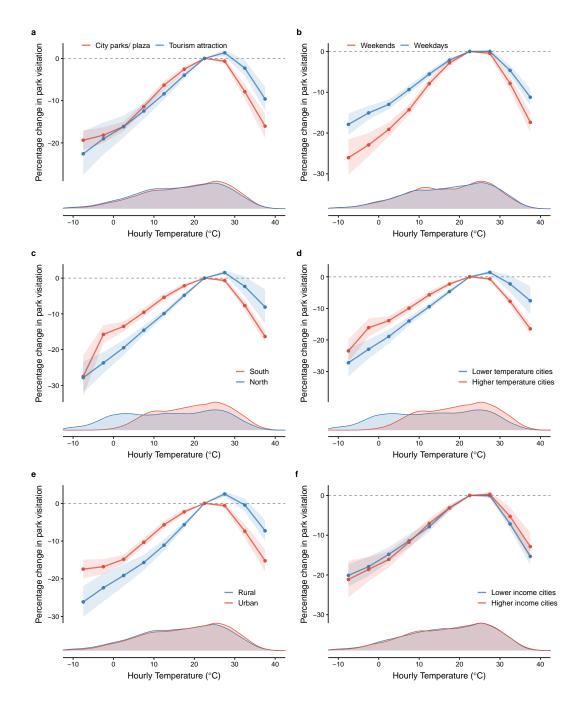
11		<i>J</i> J		
	>30 °C	25-30 °C	20-25 °C	15-20 °C
Time	(1)	(2)	(3)	(4)
6 AM	-0.6277***	-0.6542***	-0.6951***	-0.7900***
	(0.0154)	(0.0185)	(0.0210)	(0.0238)
7 AM	-0.2471***	-0.2699***	-0.2860***	-0.3317***
	(0.0134)	(0.0150)	(0.0137)	(0.0138)
8 AM	-0.0430***	-0.0692***	-0.0862***	-0.1219***
	(0.0071)	(0.0088)	(0.0088)	(0.0086)
9 AM	0.0360***	0.013	-0.0062	-0.0371***
	(0.0055)	(0.0053)	(0.0061)	(0.0059)
10 AM	0.0762***	0.0644***	0.0477***	0.0249***
	(0.0072)	(0.0051)	(0.0055)	(0.0050)
11 AM	0.0774***	0.0684***	0.0555***	0.0413***
	(0.0061)	(0.0045)	(0.0046)	(0.0038)
12 PM	0.0420***	0.0360***	0.0279***	0.0208***
	(0.0034)	(0.0024)	(0.0028)	(0.0023)
2 PM	0.0175***	0.0220***	0.0226***	0.0277***
	(0.0029)	(0.0025)	(0.0023)	(0.0028)
3 PM	0.0387***	0.0417***	0.0389***	0.0432***
	(0.0047)	(0.0043)	(0.0035)	(0.0041)
4 PM	0.0746***	0.0696***	0.0559***	0.0496***
	(0.0061)	(0.0059)	(0.0054)	(0.0059)
5 PM	0.1127***	0.0850***	0.0495***	0.0231*
	(0.0083)	(0.0091)	(0.0082)	(0.0084)
6 PM	0.1208***	0.0623***	0.0003	-0.0476***
	(0.0118)	(0.0131)	(0.0107)	(0.0113)
7 PM	0.1172***	0.0342	-0.0440***	-0.1049***
	(0.0164)	(0.0158)	(0.0122)	(0.0125)
8 PM	0.0818***	-0.0048	-0.0883***	-0.1536***
	(0.0240)	(0.0216)	(0.0166)	(0.0167)
9 PM	-0.0443	-0.1230***	-0.2035***	-0.2603***
	(0.0275)	(0.0249)	(0.0190)	(0.0196)
10 PM	-0.2953***	-0.3598***	-0.4283***	-0.4698***
	(0.0280)	(0.0275)	(0.0222)	(0.0228)
Hour FE	No	No	No	No
DOW FE	Yes	Yes	Yes	Yes
Park FE	Yes	Yes	Yes	Yes
Month-City FE	Yes	Yes	Yes	Yes
Observations	3,744,994	8,320,998	8,017,164	6,826,441
Adjusted R ²	0.8809	0.8693	0.8599	0.8499
Note: *n<0.01: **n<0) 005· ***n<0 001 Der	andant variable is In	(Park visitation +1)	Subsamples are base

Note: *p<0.01; **p<0.005; ***p<0.001. Dependent variable is ln (Park visitation +1). Subsamples are based on average daily temperature and regressions are run for each sample separately. Control variables include air quality, precipitation, wind speed, relative humidity, atmospheric water vapor, atmospheric pressure, cloud coverage. 1 PM is set as the reference. Each time point represents the following hour (e.g., 6 AM represents 6 AM to 7 AM). Standard errors clustered within city. In Fig.4a, the regression coefficients are adjusted into percentage change by $e^{\beta} - 1$.

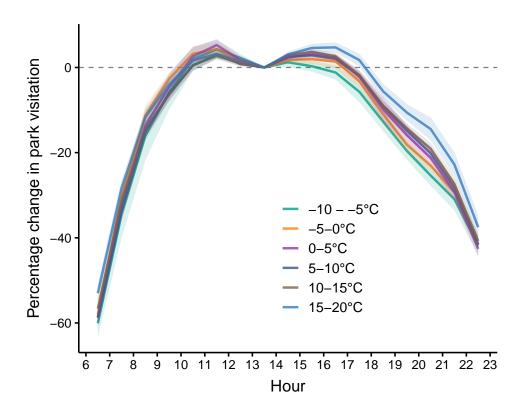
Supplementary Figures



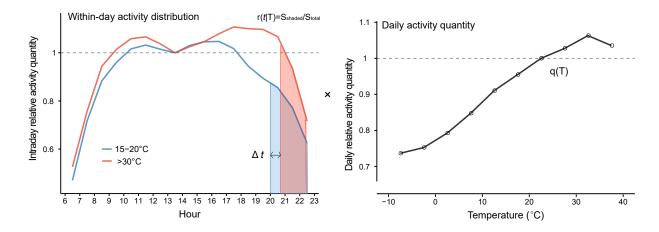
Supplementary Figure 1. Effects of daily temperature from different time granularity. Estimation outputs from daily and hourly observations are presented respectively. Shaded areas are 95% confidence intervals. The density plots at the bottom of each figure show the share of parks exposed to different temperatures in the sample used to estimate each curve.



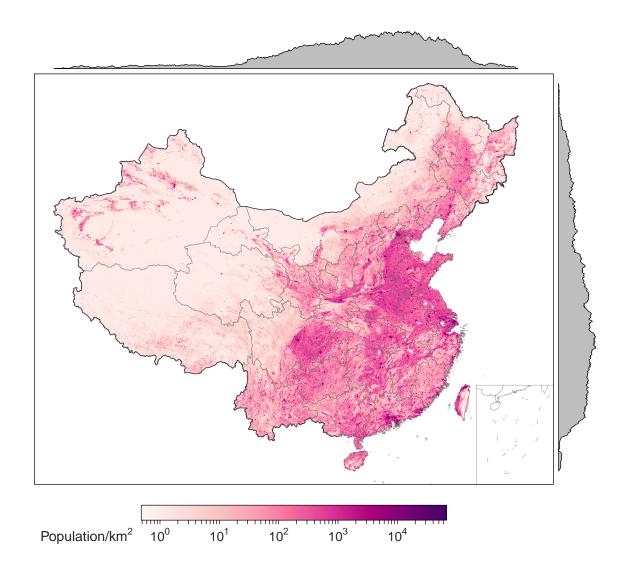
Supplementary Figure 2. Heterogeneity results for each temperature bin. **a**, by park types. **b**, by time. **c**, by geographical regions. **d**, by city custom temperature level. **e**, by park location (urban/rural). **f**, by city income. Shaded areas are 95% confidence intervals. The density plots at the bottom of each figure show the share of the park exposed to different temperatures in the sample used to estimate each curve.



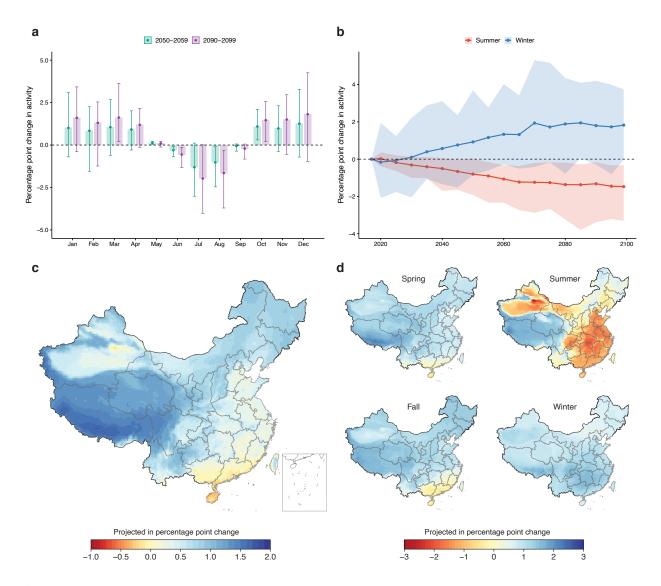
Supplementary Figure 3. Effects of cold temperature on the within-day activity distribution pattern. Each line represents the within-day activity pattern for days lying in each daytime average temperature range. Shaded areas are 95% confidence intervals.



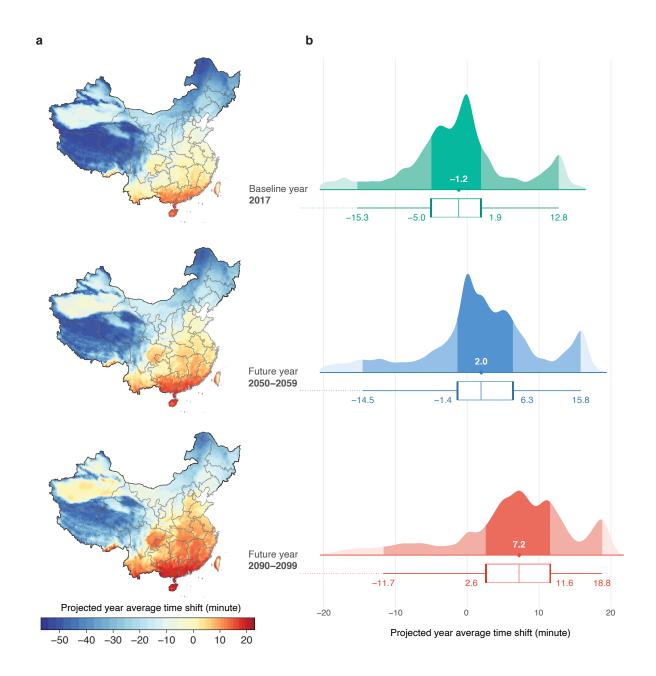
Supplementary Figure 4. Schematic illustration of the definition of time shift and its calculation procedure. Within-day distribution daily quantity results are from the regression results of Fig. 4a and Supplementary Figure 1. Equalizing activity share times activity quantity under different temperatures produces the time shift function in Fig. 4b.



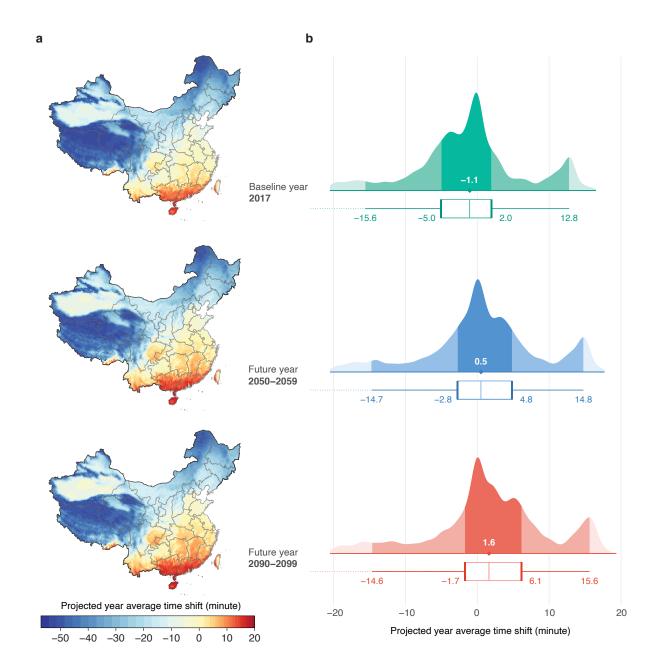
Supplementary Figure 5. Distribution of population density of China in 2017. The Qinghai-Tibet Plateau in west China is the only region where climate change has positive human activity impacts, yet with a very limited population. The grid population data is from LandScan: https://landscan.ornl.gov/.



Supplementary Figure 6. Projected change in park visitation quantity under RCP4.5 scenario. **a**, Additional percentage point change in park visitation quantity by month in the year 2050 and 2099 compared with 2017 due to climate change. **b**, The percentage change in activity quantity for summer and winter respective for all years in the future. Geographic estimates of the projected change in park visitation quantity in all areas of China in 2090-2099. Error bars and shaded areas represent 95% confidence intervals calculated through Monte Carlo sampling of econometric parameter estimates and climate models. **c**, Geographic dispersion of the predicted additional yearly average effects of climate change-induced activity change in 2090-2099. **d**, Prediction of additional climate change-induced activity change in 2090-2099 by seasons.

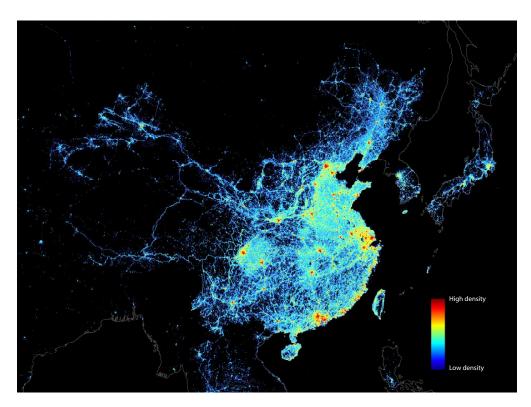


Supplementary Figure 7. Projected time shift in intraday timing of activity under RCP8.5. **a**, Distribution of average yearly activity time shift due to temperature in summer of 2017, at midcentury (2050-2059) and end of the century (2090-2099) under RCP8.5 scenario. **b**, Densityplot of the size of populations subjecting to each yearly average time shift levels.

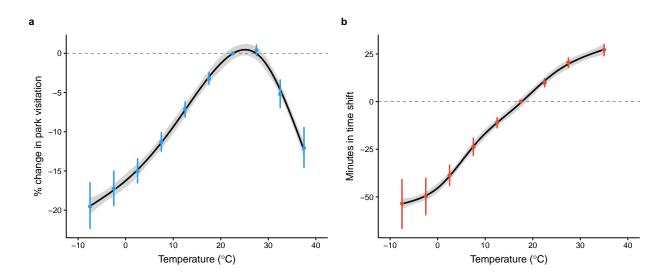


Supplementary Figure 8. Projected time shift in intraday timing of activity under RCP4.5.

a, Distribution of average yearly activity time shift due to temperature in 2017, at mid-century (2050-2059) and end of the century (2090-2099) under RCP4.5 scenario.
b, Densityplot of the size of populations subjecting to each yearly average time shift levels.



Supplementary Figure 9. Main datasets. Cell-phone location data heatmap.



Supplementary Figure 10. Natural cubic spline fitted models. **a**, Hourly temperature and percentage change in activity quantity; **b**, Daily average temperature and activity time shift. Dots and error bars show point estimates and 95% confidence intervals for each temperature bin. The dark lines connecting the dots are smoothed through the Loess algorithm.