Attribution of Urban Heat Island for Cities in China

YNCenter Video Conference Chang Cao 3/26/2015

Outline

- Background
- Motivation
- Method
- Results
- Conclusion
- On-going work

Background

- More than half of the world's population reside in urban area which consists of 2% of global land area.
- Urbanization gives rise to Urban Heat Island (UHI) which has impact on local climate and human health.
- Community Earth System Model (CESM) climate model has incorporated urban canopy model to study urban climatic response under atmospheric condition in the future.
- Study on UHI for cities in North American showed that convection efficiency is the main driver of UHI variation across climatic zones.

Motivation

- What is the UHI spatial pattern across different climatic zones in China?
- What is the driver of UHI variation for cities in China?
- What is the contribution of each biophysical process to UHI?

Method

- 39 cities in China
- Exclude elevation and latitude difference between urban and rural area more than 100 meters and 1°
- Exclude water pixels
- Avoid 'oasis effect' for semi-arid cities (Hotan)

Other data

- Precipitation data: National Climate Center
- Air temperature data: National Climate Center
- Normalized Difference Vegetation Index data: MODIS MYD13Q1 250m 16-day
- White Sky Albedo data: MODIS MCD43A3
 500m 16-day
- Population data: China Population and Employment Statistics Yearbook, 2013

Observational land surface temperature

- 2003-2013
- MODIS Land Surface Temperature (LST) MYD11A2 1km 8-day daytime 13:30 nighttime 1:30

Model simulation

- Community Land Model 4.0 CLMU
- 1972-2004, 0.23° longitude × 0.31° latitude
- Offline, Qian et al forcing data
- Extract data at 1:00 and 13:00 (local time) and for clear-sky condition



Figure 1 | Schematic overview of urban land unit in CLM

Source: Oleson, Bonan, Feddema et al. (2008) *J Appl Meteor Climatol* 47: 1038-1060

Result

Spatial variation



Figure 2 | Vegetation of China Source: Zhou, Zhao, Liu et al. (2014) *Sci of Tol Env* 488: 136-145



Figure 3 | Spatial variation of MODIS-derived annual-mean daytime (a) and night-time (b) UHI.

Cropland pixel fraction of rural area

(MODIS land cover)

	Humid	Semi-humid	Semi-arid	Averaged
China	0.53	0.80	0.29	0.51
North American	0.20	0.44	0.23	0.30



a

С

Figure 4 | Precipitation influence on annual daytime UHI. a, All 39 cites in. b, Without cities whose rural area has cropland pixel fraction bigger than 0.5. c and d are the same as a and b but for North American

cities

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13

Relationship between UHI and variables



Figure 5 | The square of linear correlation coefficients (R²) (a) and stepwise multiple linear regression derived explanation (b) of each variable for cities in China.



Figure 5 | The square of linear correlation coefficients (R²) (c) and stepwise multiple linear regression derived explanation (d) of each variable for cities in North American.



С

Figure 6 | Modelled albedo influence on modelled annual mean (a), winter (b) and summer (c) night-time UHI 16 intensity

Attribution to model-predicted UHI $(1-\alpha)K_{\parallel}+L_{\parallel}-\sigma T_{s}^{4}=H+LE+Q_{s}-Q_{AH}$ (1) $\sigma T_s^4 = \sigma T_a^4 + 4\sigma T_a^3 (T_s - T_a)$ (2) $\Delta T \approx \frac{\lambda_0}{1+f} \Delta R_n^* + \frac{-\lambda_0}{(1+f)^2} (R_n^* - Q_s + Q_{AH}) \Delta f_1$ $R_n^* = (1 - \alpha) K_{\perp} + L_{\perp} - (1 - \varepsilon) L_{\perp} - \varepsilon \sigma T_a^4$ $+\frac{-\lambda_0}{(1+f)^2}(R_n^*-Q_s+Q_{AH})\Delta f_2+\frac{-\lambda_0}{1+f}\Delta Q_s \qquad \qquad \Delta f_1=\frac{-\lambda_0\rho C_p}{r_a}(1+\frac{1}{\beta})\frac{\Delta r_a}{r_a}$ $\Delta f_2 = \frac{-\lambda_0 \rho C_p}{r} \frac{\Delta \beta}{\beta^2}$ $+\frac{\lambda_0}{1+f}\Delta Q_{AH}$ (3)



Validation



Figure 7 | Model-predicted ΔT vs calculated ΔT . a, Annual-mean daytime. b, Annual-mean night-time. c, Summer mean daytime. d, Summer mean night-time. e, Winter mean daytime. f, Winter mean night-time. 19

Tuning emissivity

North American: 0.88 (urban area)
 0.96 (rural area)

$$T_{s} = \sqrt[4]{\frac{L_{\downarrow} + (L_net)/\varepsilon}{\sigma}}$$

 Chinese cities (combined with CLM and MODIS emissivity, H/W ratio)

	Humid	Semi-humid	Semi-arid
Urban	0.914	0.960	0.969
Rural	0.927	0.980	0.973



Figure 8 | Attribution of UHI intensity in three climate zones in annual-mean (a), summer (b) and winter (c) daytime.²¹



MODIS CLM



Figure 9 | Attribution of UHI intensity in three climate zones in annual-mean (a), summer (b) and winter (c) night-time.

Covariance analysis

$\Delta T = C_R + C_H + C_{LE} + C_s + C_{AH} + e \quad (6)$

$Cov(\Delta T, P) = Cov(C_R, P) + Cov(C_H, P) + Cov(C_{LE}, P) + Cov(C_s, P) + Cov(C_{AH}, P) + Cov(C_{R}, P) + Cov(e, P)$ ⁽⁷⁾

Daytime

Convection

Evaporation

Anthropogenic heat

Storage

Figure 10 | Relationship between model-predicted daytime ΔT and precipitation among the cities in China. a (up), Correlation of ΔT and the biophysical components with annual-mean precipitation. a (bottom), ΔT -precipitation covariance explained by different biophysical factors. b and c are the same as a but for summer and winter.

Temporal sensitivity

Figure 11 | Temporal sensitivity of UHI intensity to precipitation. a, c, Map of the temporal sensitivities according to MODIS (a) and climate model (c). b, d, Dependence of MODIS (b) and model-predicted (d) temporal sensitivity on annual mean precipitation.

Conclusion

- UHI pattern for cities in China are mainly depend on annual precipitation but high cropland fraction in rural area can weaken this relationship.
- Convection efficiency and storage heat is the dominant contributor to UHI at day time and night time respectively. Evaporation can be a significant factor for cities in drier region in summer daytime.

Future work

- Finding solutions to the underestimation of nighttime UHI
- Can soil moisture be an important factor?
- Investigating relationship between UHI and biochemical factors (aerosol..)

Thank you