

XVIII World Water Congress

International Water Resources Association(IWRA) Beijing, China | September 11-15, 2023

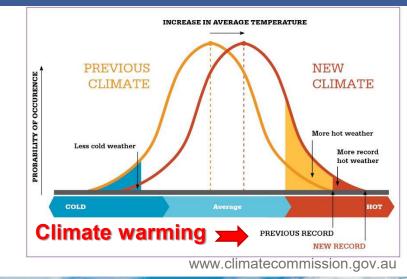


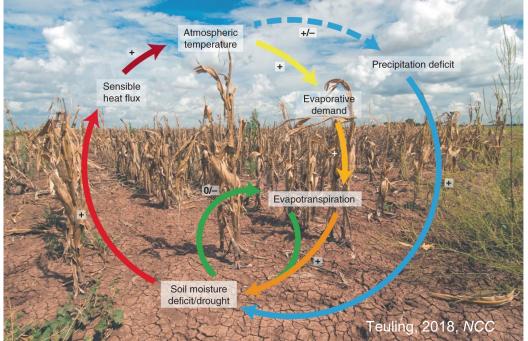
Response of Typical Vegetation Ecosystems to Hotter Droughts under A Changing Climate

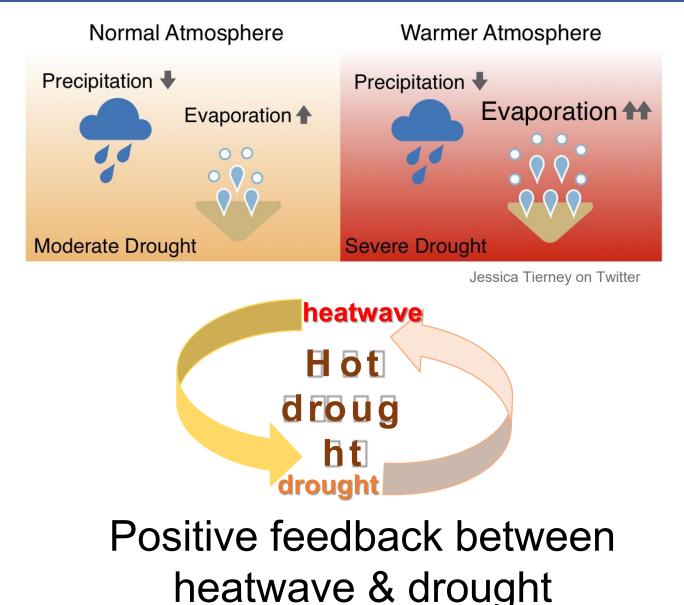
Chunyu Dong, Zhimin Ma, Yu Yan

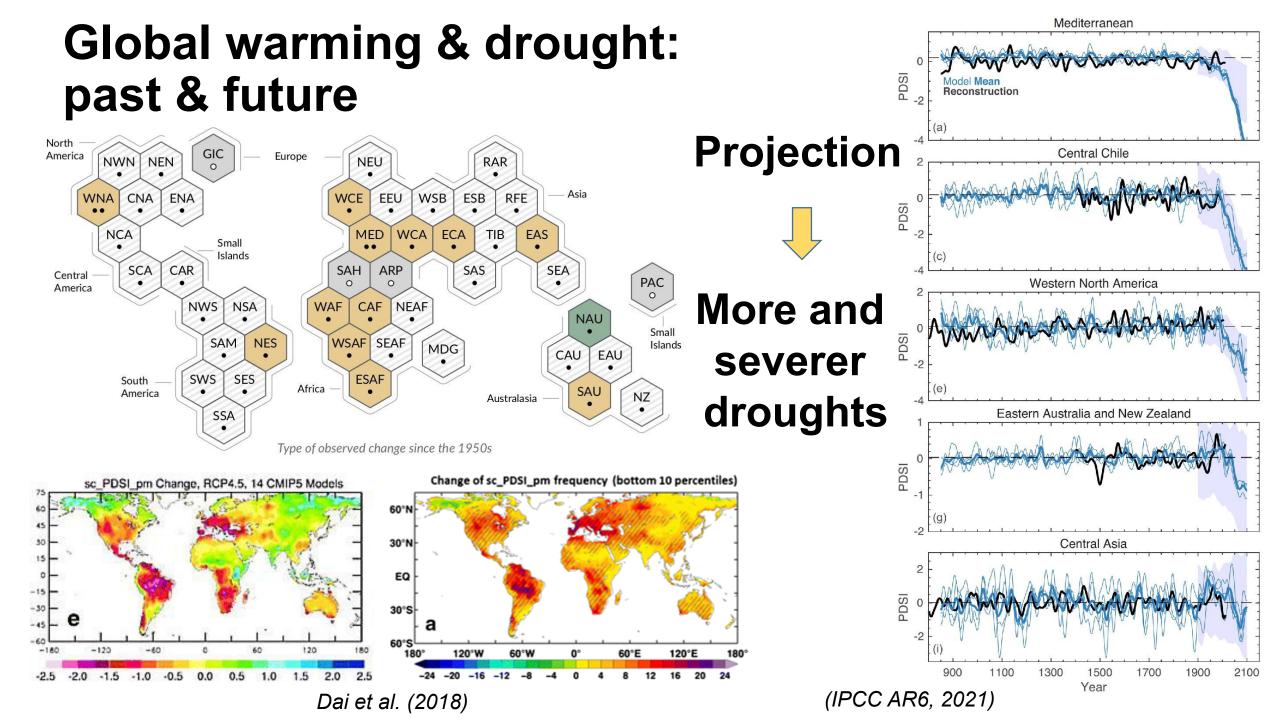
Sun Yat-sen University

Background: climate warming + drought = hotter drought

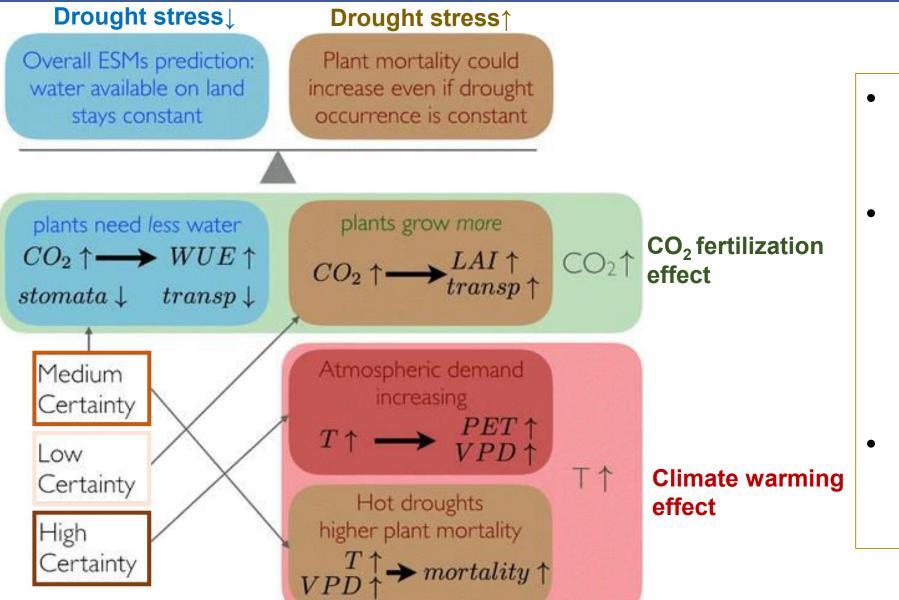








Background: Vegetation's response to future drought

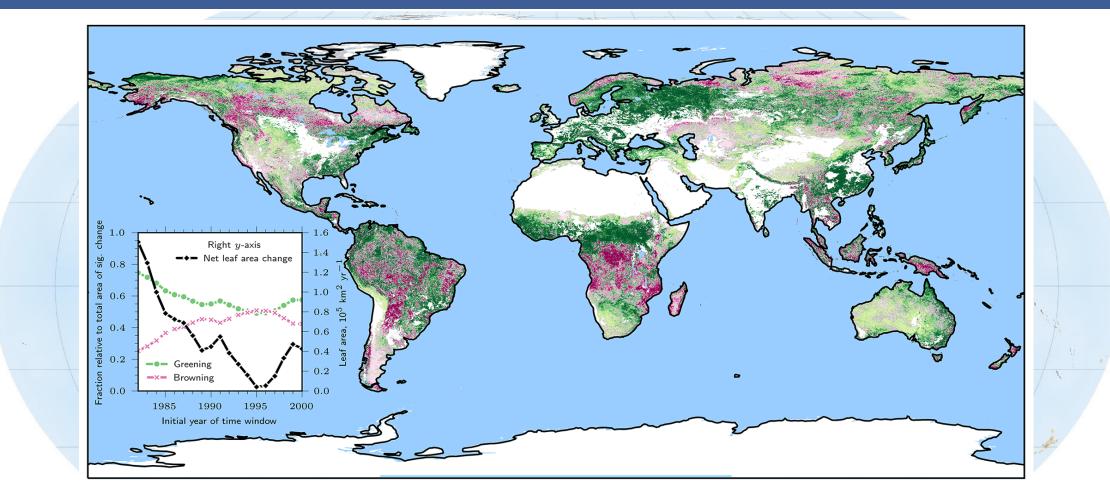


High certainty:

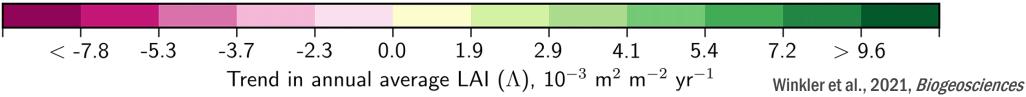
- warming→ PET and VPD increase
- Medium certainty:
 - CO₂ increase→
 WUE increase
 - Warming & VPD increase→ mortality increase
- Low certainty:
 - CO₂ increase →
 LAI increase

Swann, CCCR (2018)

Background: continued global greening?



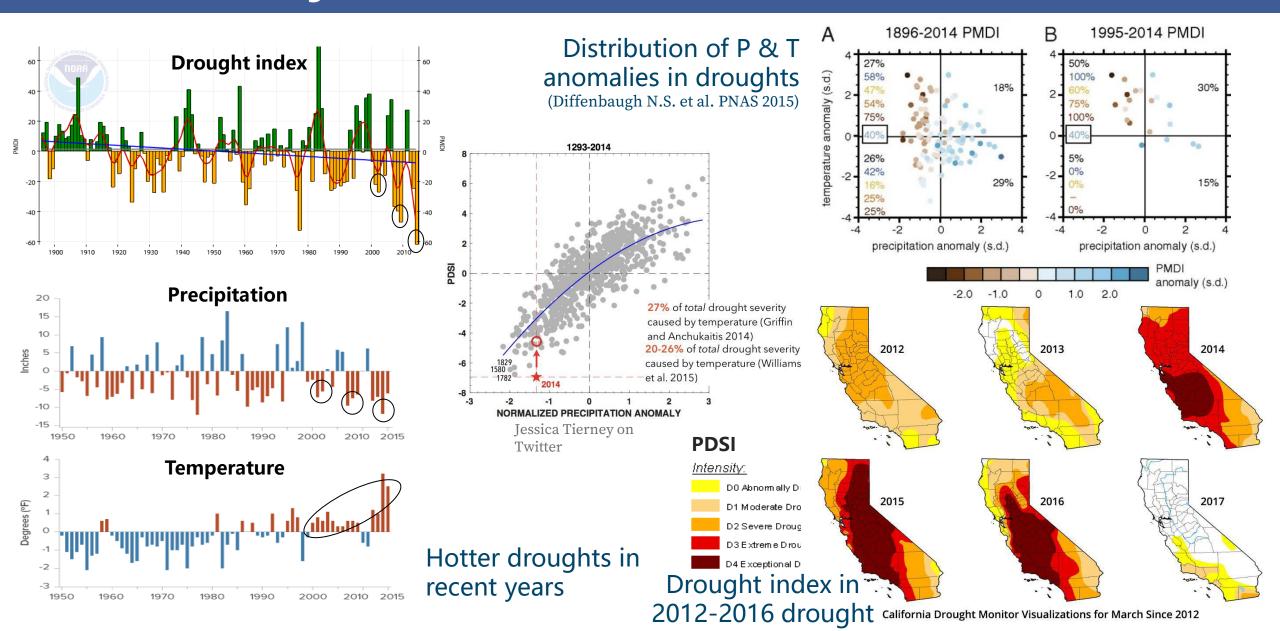
Slow-down of greening + partly browning



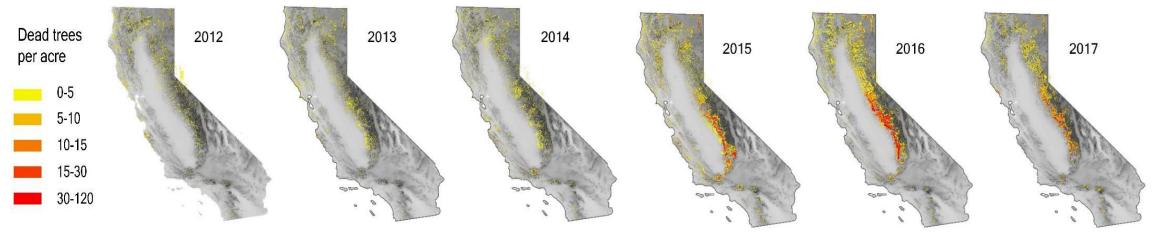
Question

Q1: What are the impacts of hotter droughts ondifferent vegetation ecosystems?Q2: What are the main drivers for different responseof vegetation to drought?

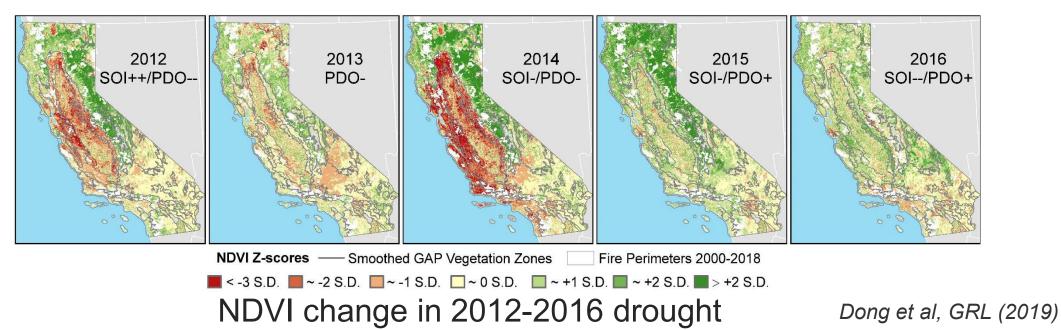
Case study: Mediterranean Climate, California



Latitudinal difference of vegetation's response to drought in CA

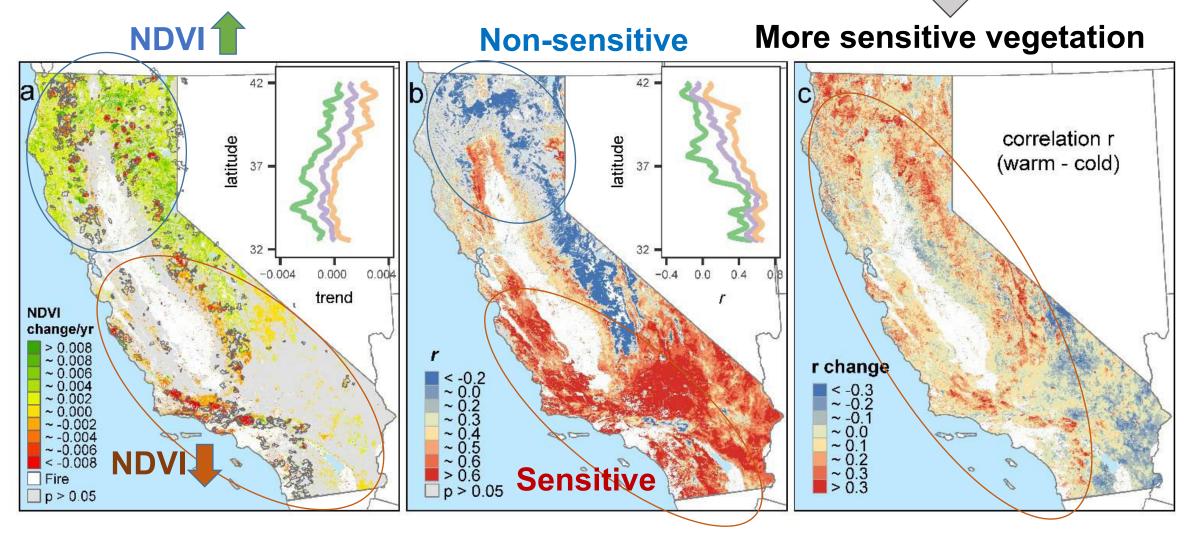


Tree mortality in 2012-2016 CA drought



NDVI trends and sensitivity to drought





NDVI trends

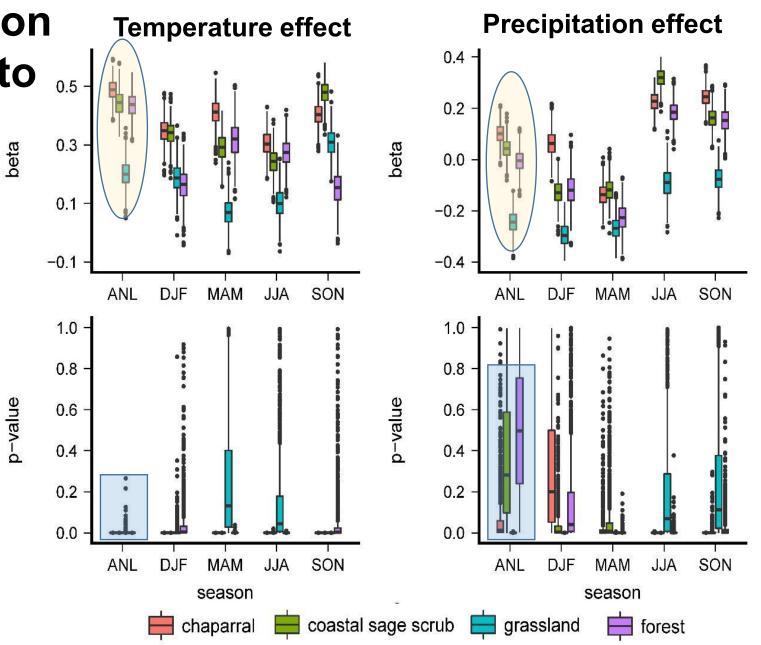
NDVI-PDSI correlation r

 Δ (NDVI-PDSI correlation r) (warm period-cold period)

Dong et al., 2019, *GRL*

Impact of T & P change on vegetation's sensitivity to odd drought

- Warming can increase all vegetation's sensitivity to drought
- Shrubs are most vulnerability
- Grasses will be less affected

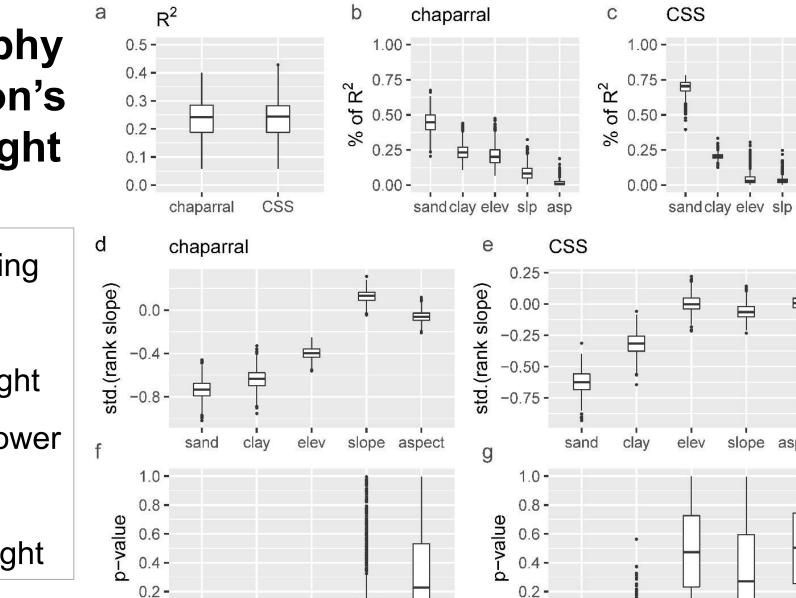


Dong et al., 2019, Remote Sens.

ANL: annual, DJF: winter, MAM: spring, JJA: summer, SON: fall

Impact of topography & soil on vegetation's sensitivity to drought

- Higher soil moisture holding capacity
 - -> less sensitive to drought
- Higher elevation and/or lower slope
 - -> less sensitive to drought



0.0 -

sand

clay

elev

slope

asp

aspect

aspect

Okin & Dong et al., 2018, *JGR: Biogeosciences* Gillespie, Ostermann-Kelm, Dong et al., 2018, Ecological Indicators

0.0-

sand

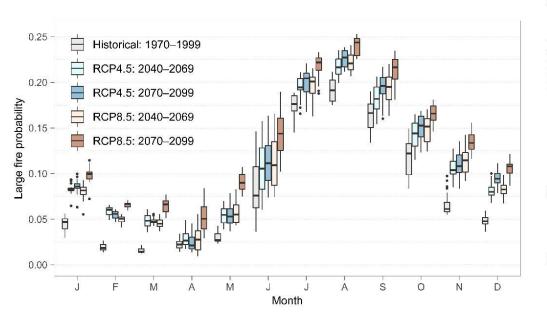
clay

elev

slope

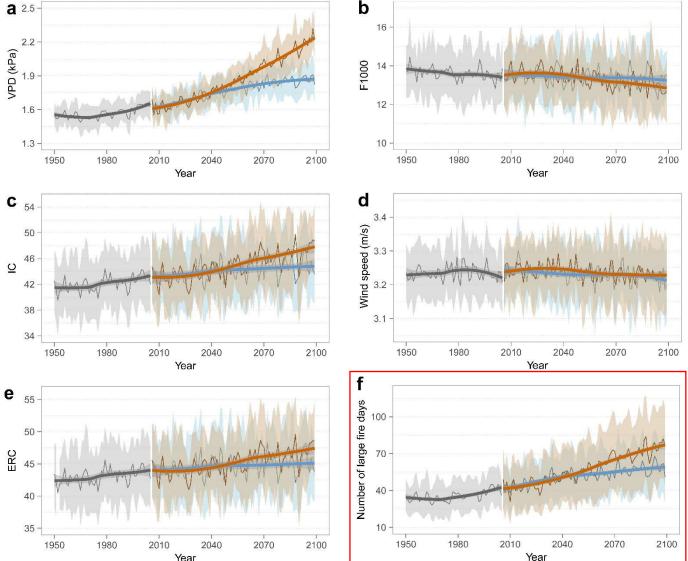
aspect

Climate change increases vegetation degradation and wildfire risk in CA



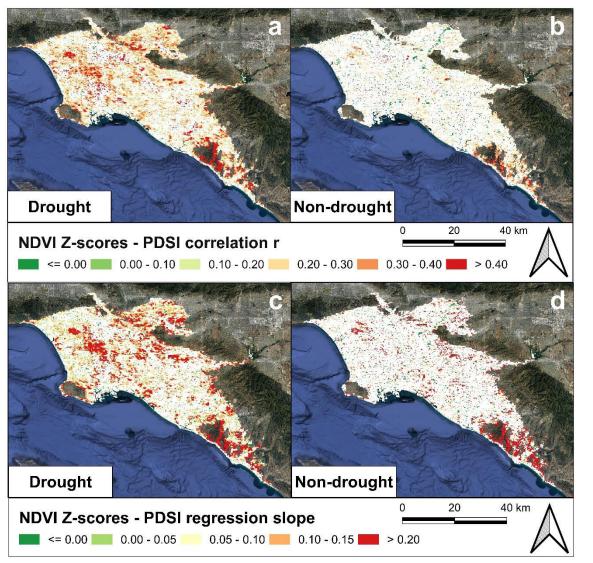


- \rightarrow Increased hotter droughts
- → Prolonged fire season & large-fire days

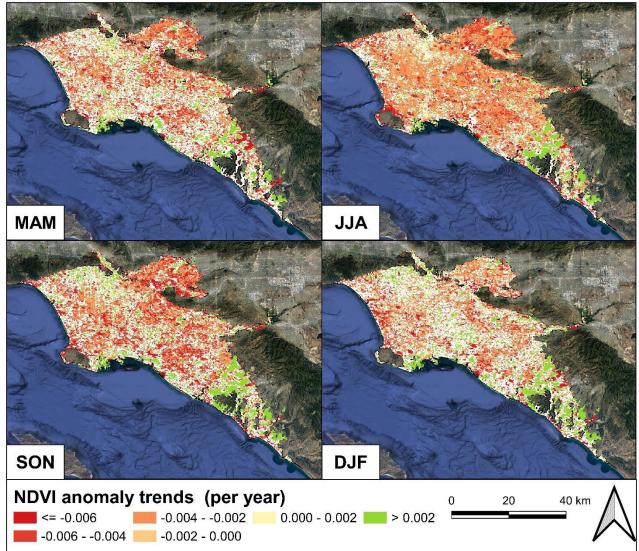


Dong et al., 2022, Communications Earth & Environment

Diverse urban vegetation response in hotter droughts

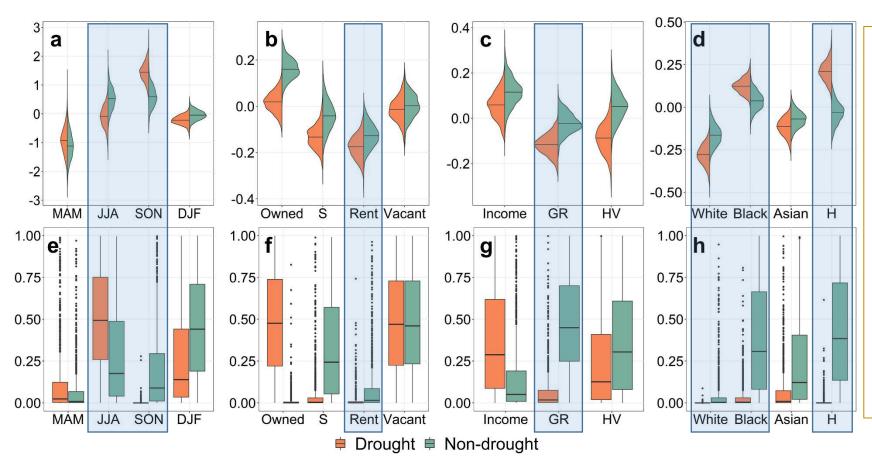


Greenness's sensitivity to drought in LA



Greenness trends in LA

Drivers of different urban vegetation's sensitivity to hotter drought



- Higher dry season T, more sensitive vegetation to drought
- Higher rent, lower sensitivity
- More White population, lower sensitivity
- More Black and/or Hispanic

population, higher sensitivity

MAM: spring T; JJA: summer T; SON: fall T; DJF: winter T; S: seasonal use; GR: gross rent; HV: house value; H: Hispanic population

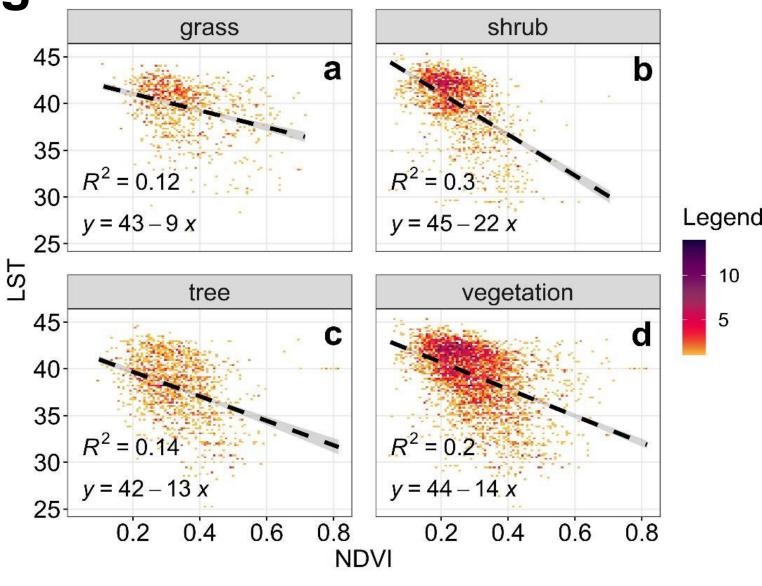
Dong et al., 2023, *Sustainable Cities and Society*

Vegetation's cooling effect change in hotter drought

• Cooling effect:

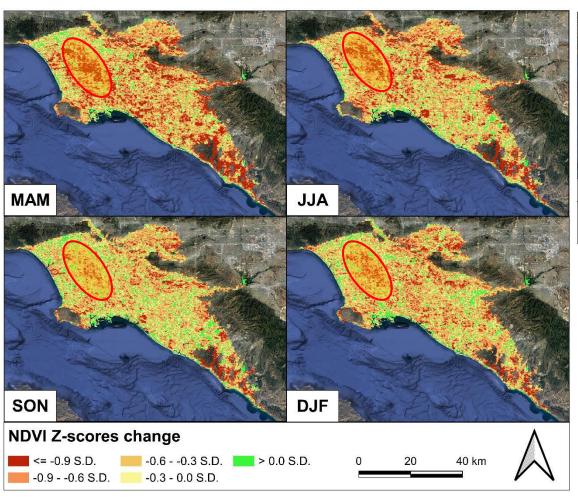
shrub > tree > grass

• NDVI \uparrow , Cooling effect \uparrow

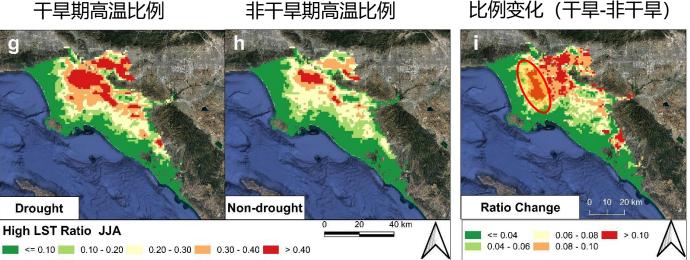


LST~NDVI regressions

Comparisons between NDVI & heatwave change in drought



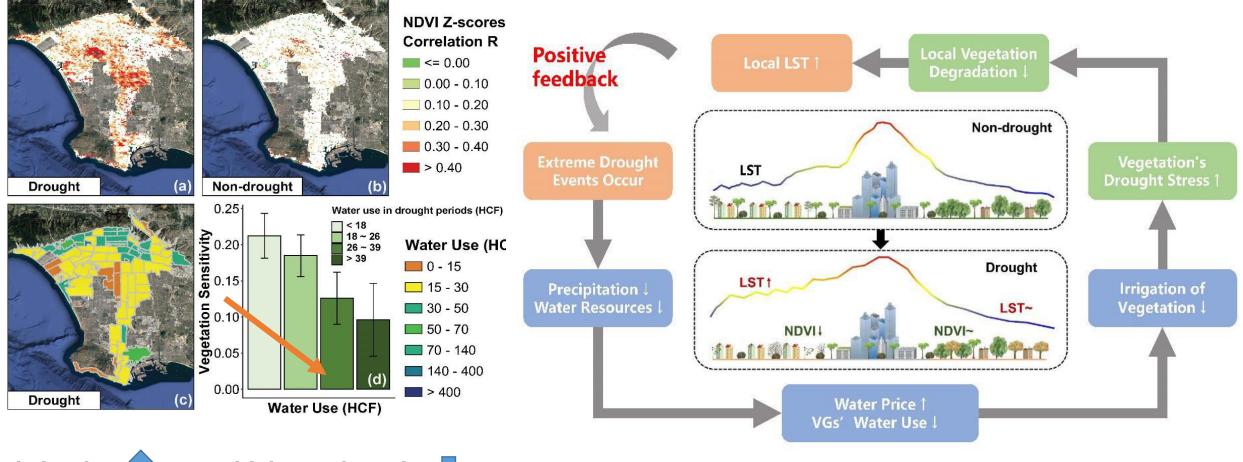
NDVI change



- NDVI declines —> heatwave increase
- —> More NDVI declines
- Mostly occurred in disadvantaged

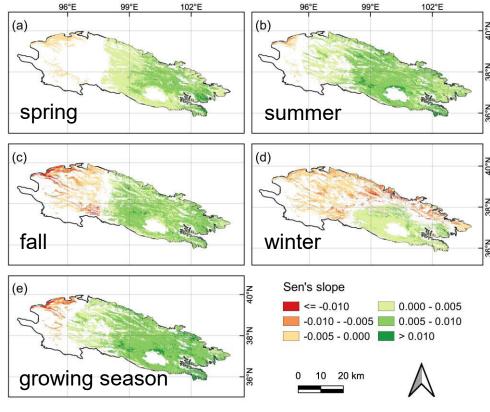
communities

Positive feedback of vegetation-waterheatwaves in cities



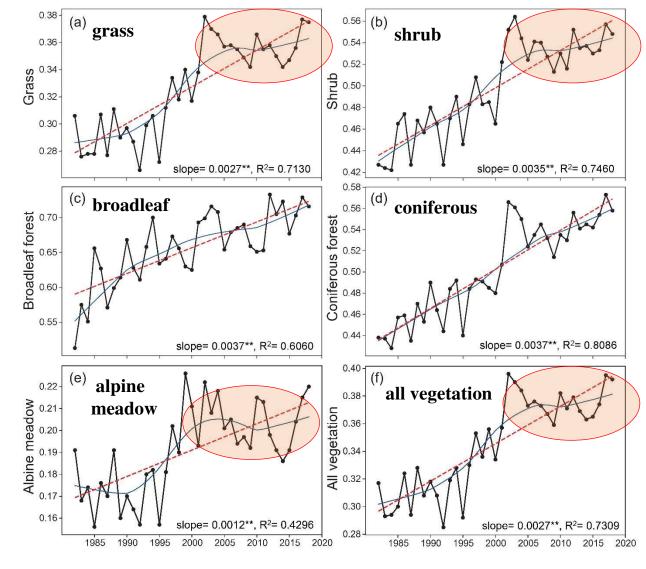
irrigation 1 sensitivity to drought

Case study: Qilian Mountains (QLMs), NW China



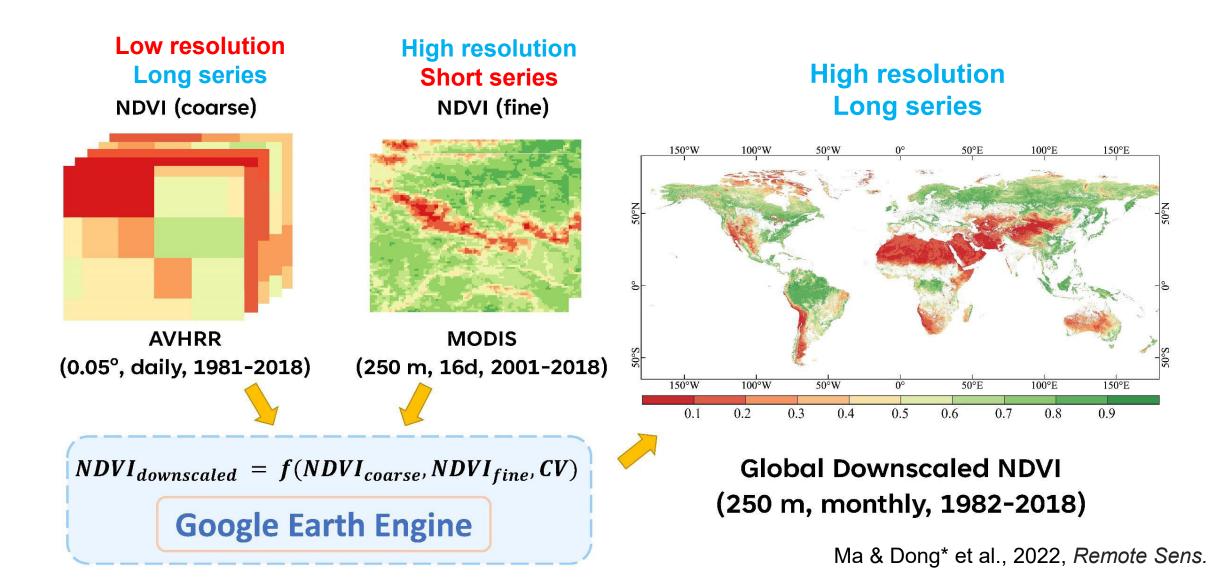
Seasonal NDVI trends

- Greening in east & browning in west
- Fast greening during 1995~2005
- Slow-down of greening since 2005

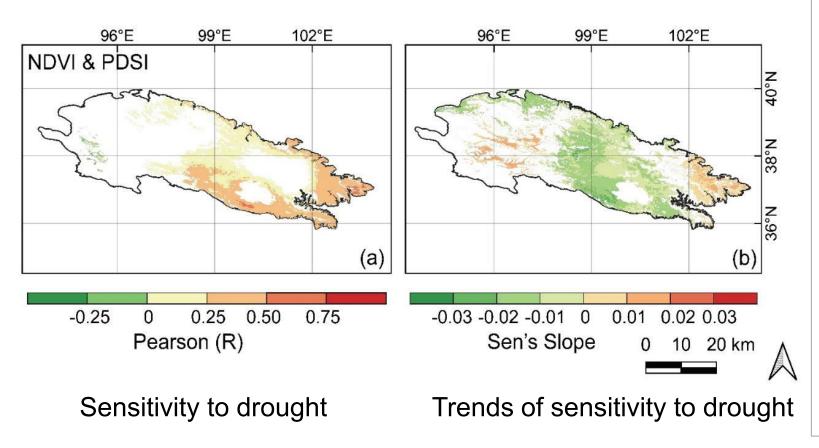


Ma & Dong et al., 2023, *J. Hydrol.*, under review

Development of an NDVI downscaling algorithm



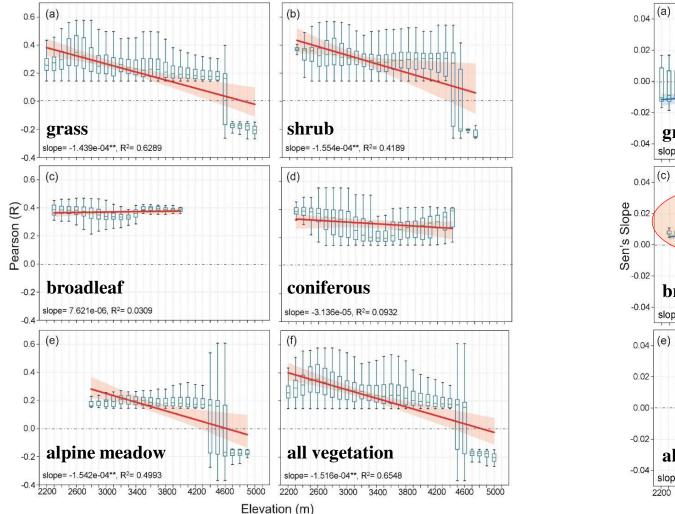
Spatial variability of vegetation's sensitivity to drought



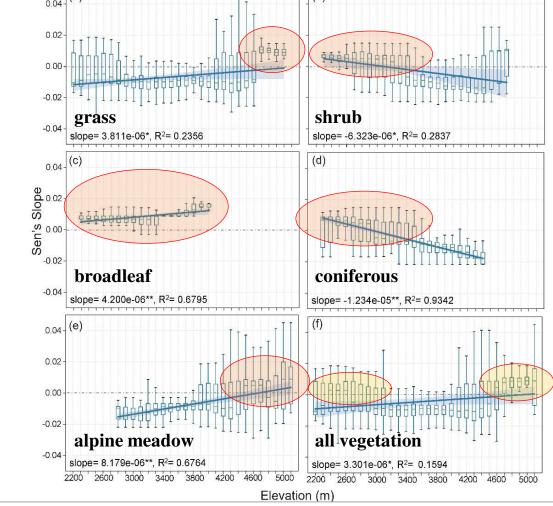
- High sensitivity to drought for vegetation in the east, and low sensitivity in the west;
- High sensitivity for the low and medium elevations;
- Decreasing sensitivity for the central areas, and increased sensitivity for the eastern and western parts.

Varied vegetation's response to drought along elevations

Sensitivity to drought

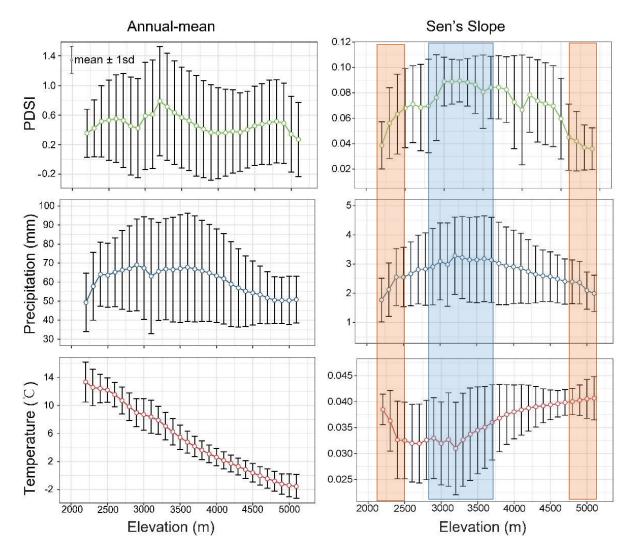


Trends of sensitivity to drought



Increased sensitivity for the low and high elevations, decreased sensitivity for the middle elevations;
Highest and increased sensitivity for broadleaf forests.

Reasons for the varied response of vegetation at different elevations



• High and low elevations:

- Fastest warming & slowest wetting
 -> slowest PDSI increase
- Middle elevations:
- Slowest warming & fastest wetting
 -> fastest PDSI increase

Climate change trends at different elevations

Summary

- The ecosystems in lower-elevated and warmer areas dominated by moisture-limited vegetation were relatively more sensitive to hotter drought;
- The ecosystems in higher-elevated and colder areas dominated by heatlimited vegetation were relatively insensitive to hotter drought;
- The warming climate has led to an increased sensitivity of many vegetation communities to drought;
- The response of **urban vegetation** to hotter drought is largely driven by differences in irrigation water use and the underlying socioeconomic factors, leading to **increased environmental injustice**.

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Thanks!

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Downscaling algorithm

Key equations

2001-2018:

$$NDVI_{H,x,y,t} = NDVI_{H,x,y,bl} \times (1 + K_{x,y,t} \times R_{CVx,y,m}) + \varepsilon_{x,y,t}$$

1982-2000:

$$NDVI_{H,x,y,t} = NDVI_{H,x,y,bl} \times (1 + K_{x,y,t} \times R_{CVx,y,m} \times R_{CVx,y,n}) + \varepsilon_{x,y,t}$$

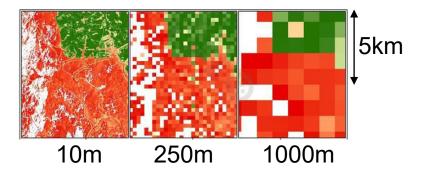
in which, $NDVI_{H,x,y,t}$ is the downscaled high-resolution NDVI, $NDVI_{H,x,y,bl}$ is the monthly median MODIS NDVI during the baseline period of 2001~2018, $\varepsilon_{x,y,t}$ is random error.

Extraction of temporal information

$$K_{x,y,t} = (NDVI_{L,x,y,t} - NDVI_{L,x,y,bl}) / NDVI_{L,x,y,bl}$$

in which, $NDVI_{L,x,y,t}$ is the monthly AVHRR NDVI (1982~2018), $NDVI_{L,x,y,bl}$ is the monthly median AVHRR NDVI during the baseline period (2001~2018).

Coefficient of variation $CV = \sigma/\mu$



Extraction of spatial information

 $R_{CVx,y,m} = MODIS_CV/AVHRR_CV_{post}$ in which, $MODIS_CV$ and $AVHRR_CV_{post}$ refer to pixel-wise coefficients of variation (CVs) of MODIS and AVHRR NDVI.

 $R_{CVx,y,n} = AVHRR_{CV_{post}}/AVHRR_{CV_{pre}}$ in which, $AVHRR_{CV_{pre}}$ and $AVHRR_{CV_{post}}$ refer to pixel-wise coefficients of variation (CVs) of AVHRR NDVI for the two periods of 1982~2000 and 2001~2018.

Ma & Dong* et al., 2022, Remote Sens.