

Effects of precipitation, heat, and drought on coccidioidomycosis incidence and expansion in California

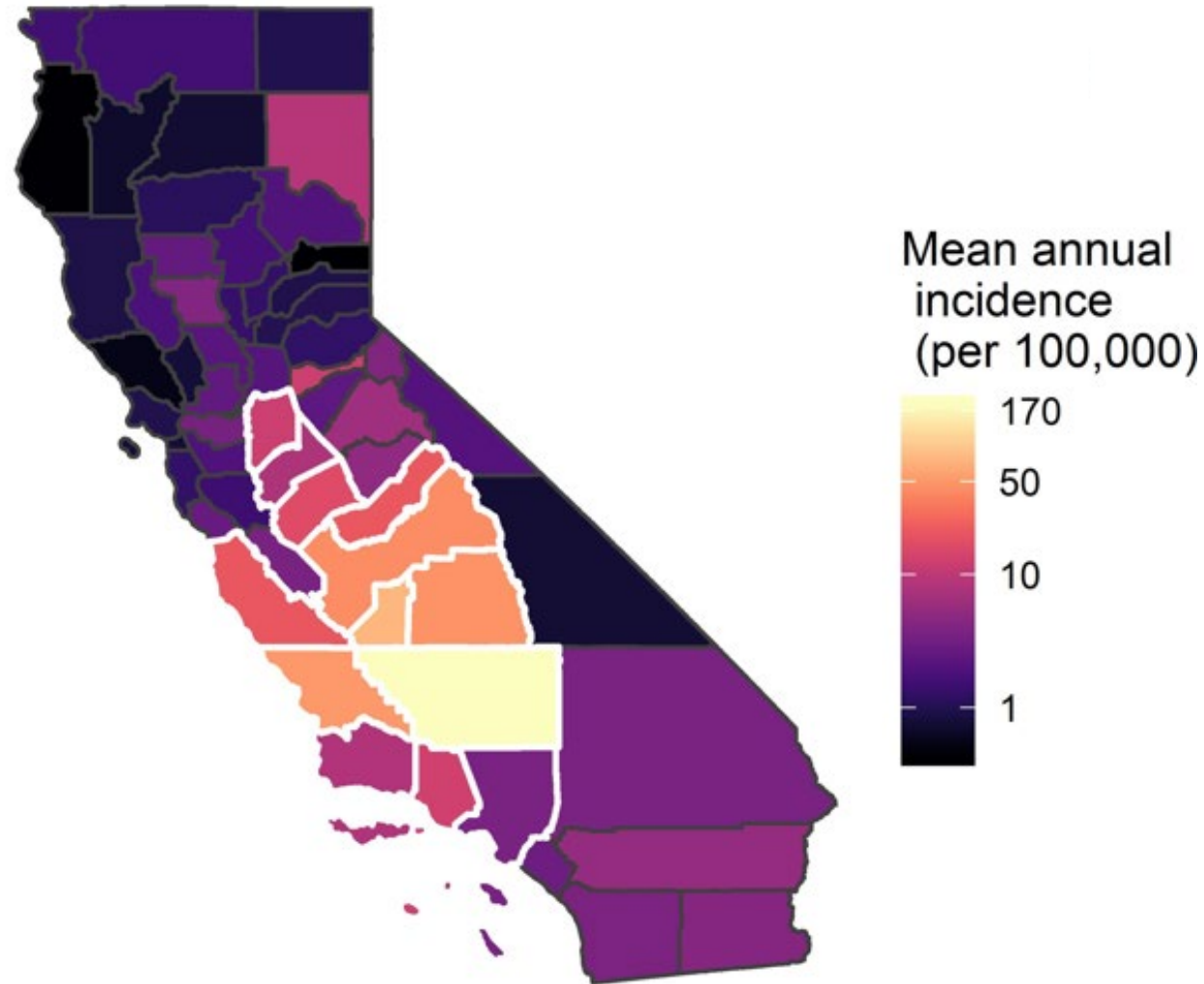
Jennifer R. Head, PhD, MPH
School of Public Health
University of California, Berkeley
November 17, 2022
National Academies of Sciences



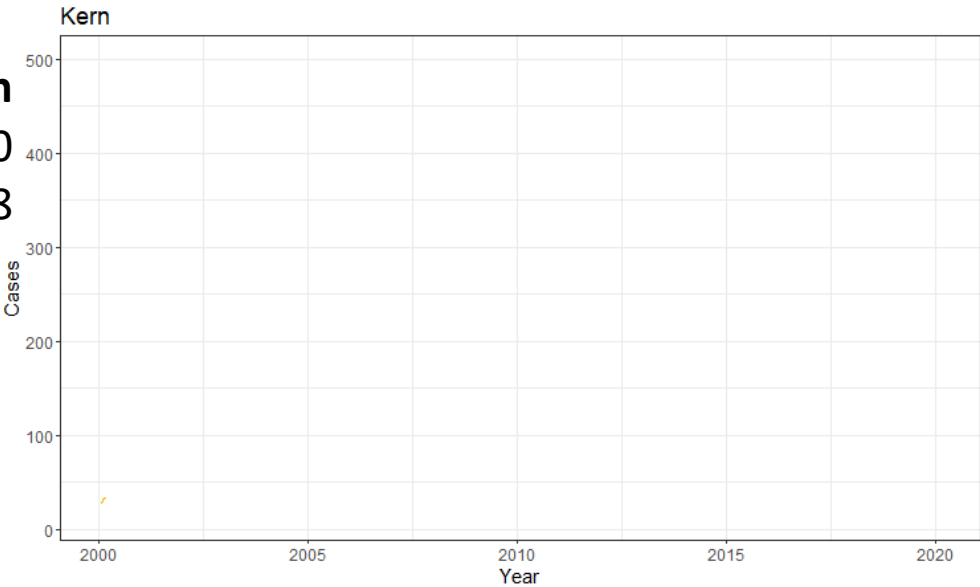
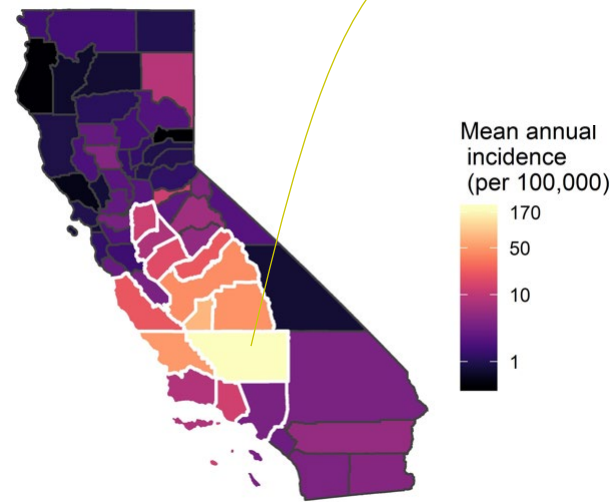
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Photo: Yuan Zhu

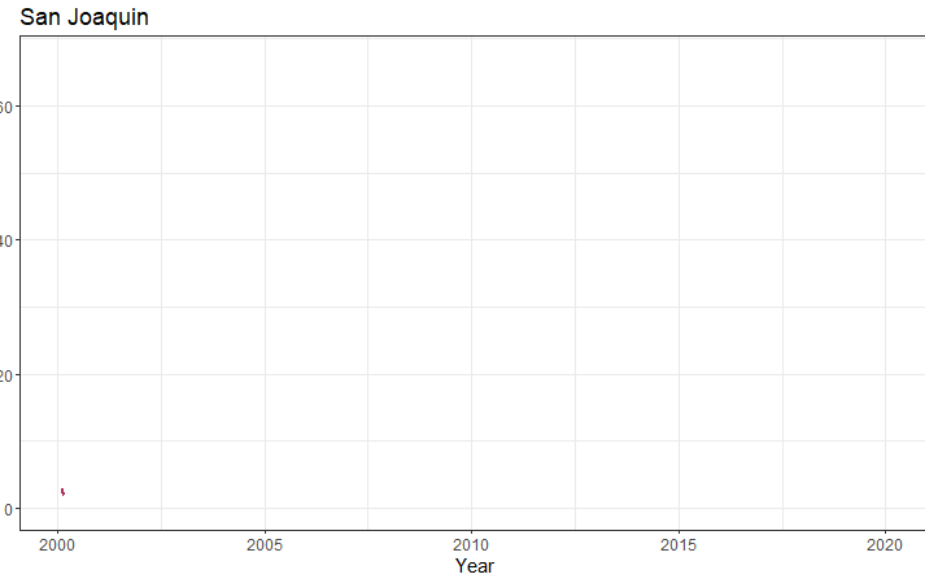
Incidence of infection with *C. immitis* is highest in the lower San Joaquin Valley, which is relatively dry and hot



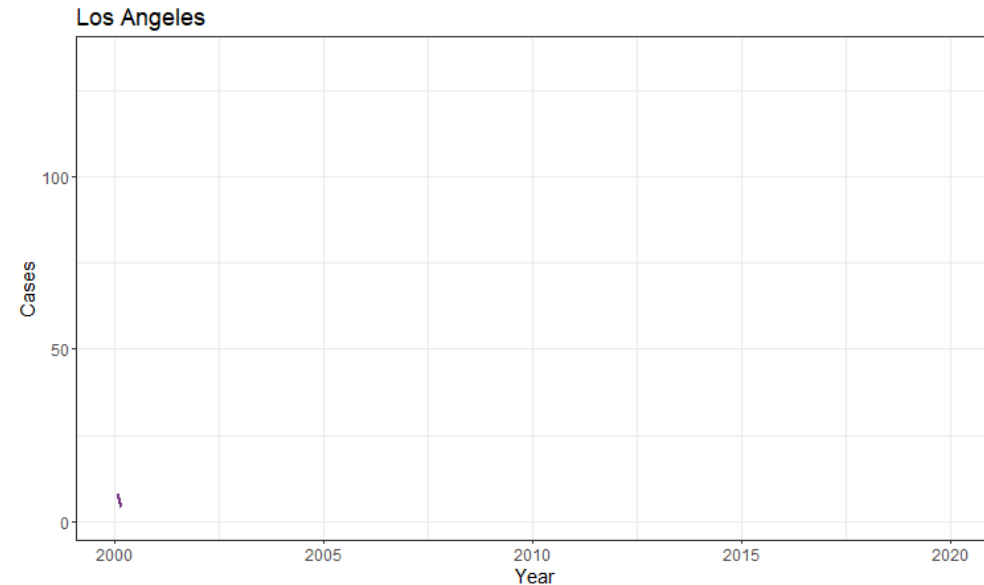
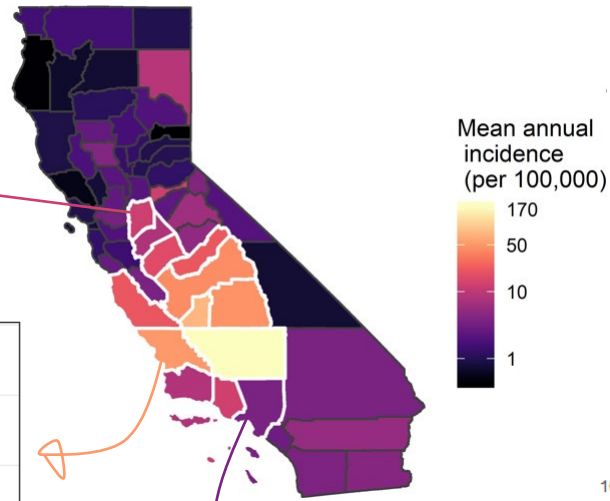
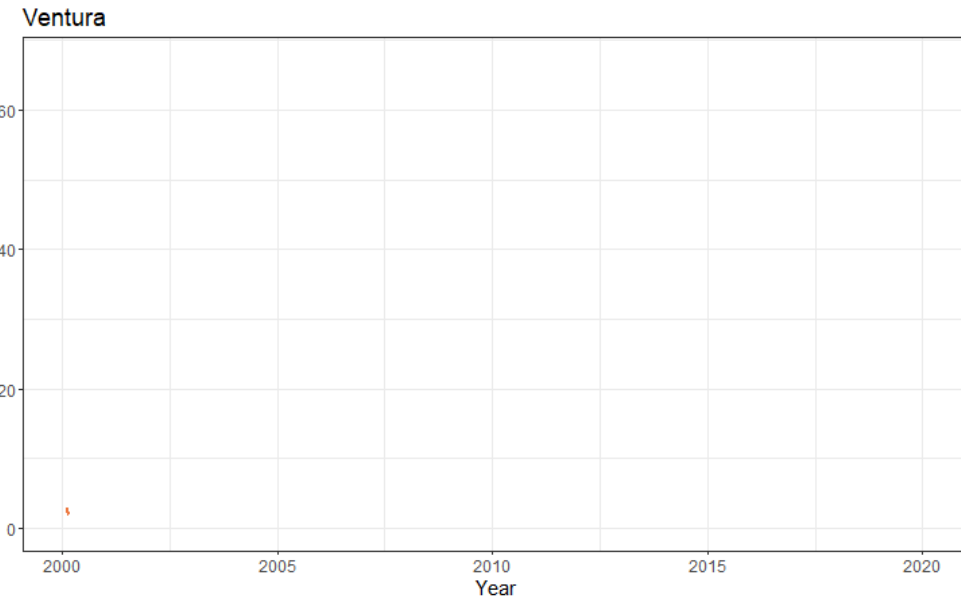
Increases of **>8-fold** in
incidence between 2000
and 2018



Incidence is increasing
fastest in areas that are not
uniformly hot and dry



Increases of **>15-fold** in
incidence between 2000 and 2018



Increasing incidence comes amidst unprecedented megadrought

nature
climate change

BRIEF COMMUNICATION

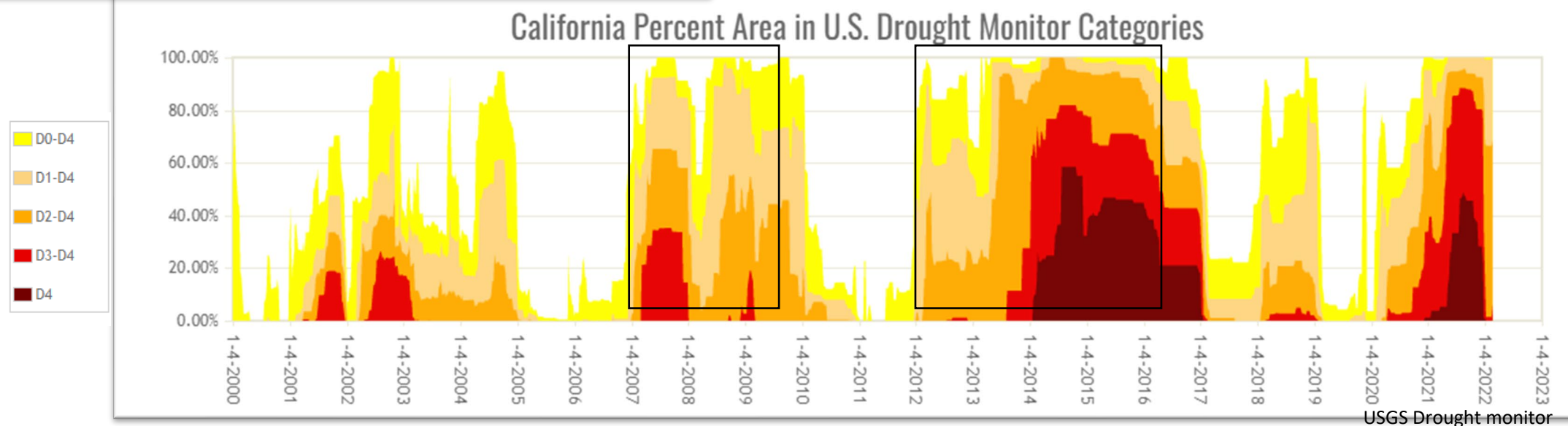
<https://doi.org/10.1038/s41558-022-01290-z>

Check for updates

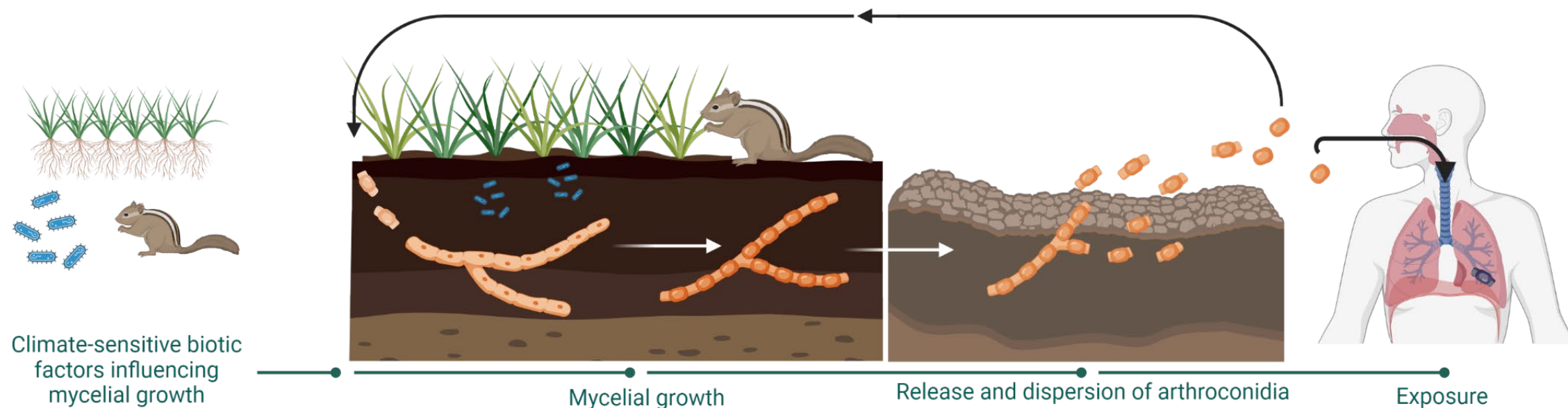
Rapid intensification of the emerging southwestern North American megadrought in 2020–2021

A. Park Williams^{1,2}✉, Benjamin I. Cook^{2,3} and Jason E. Smerdon²

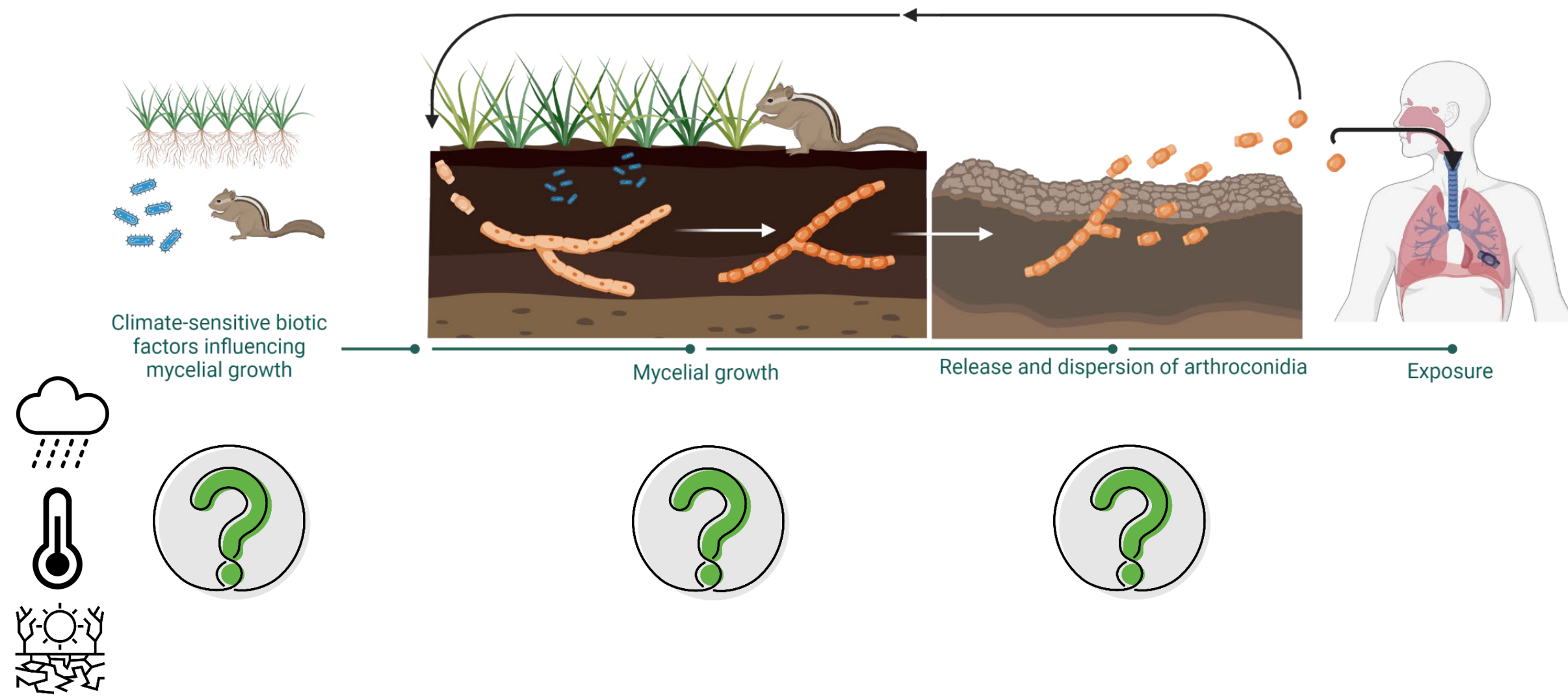
California is experiencing the driest 22-year period since 800 CE



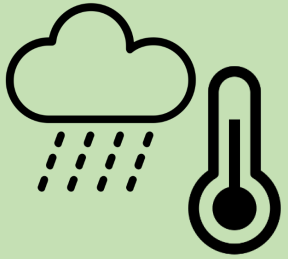
Climate affects *Coccidioides* growth and dispersion differently by time lags



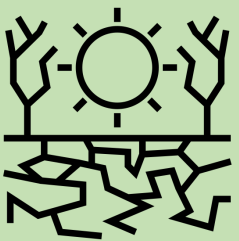
What is the effect of precipitation, temperature, and drought at each stage of the lifecycle?



Objectives:



1. Determine how prior sequences of **temperature** and **precipitation** affect incidence and compare effects across regions



2. Determine **causal effect** of the **2007-2009** and **2012-2015 droughts** on incidence and compare effect across regions

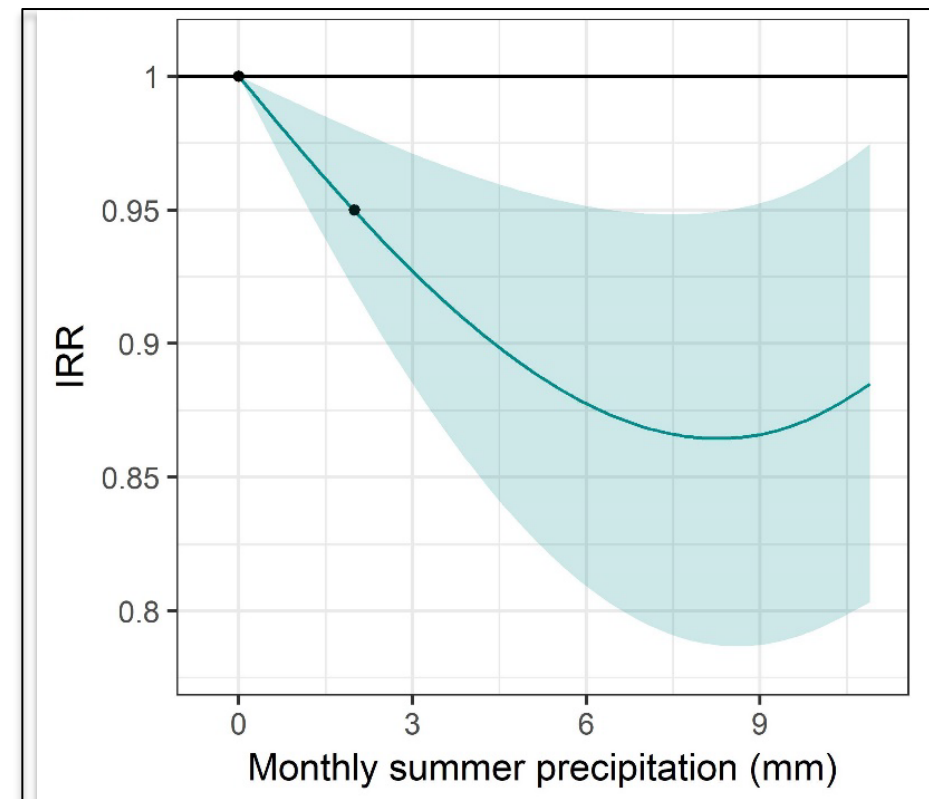
1. Determine how prior sequences of **temperature** and **precipitation** affect incidence and compare effects across regions



Method: Meta-analysis of distributed-lag nonlinear models

Distributed lag nonlinear modeling approach

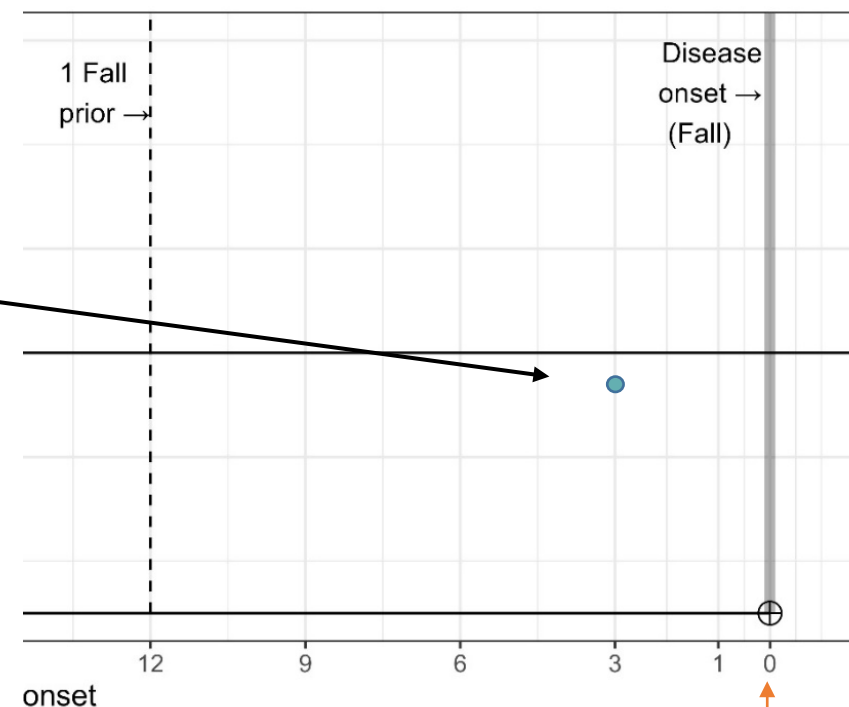
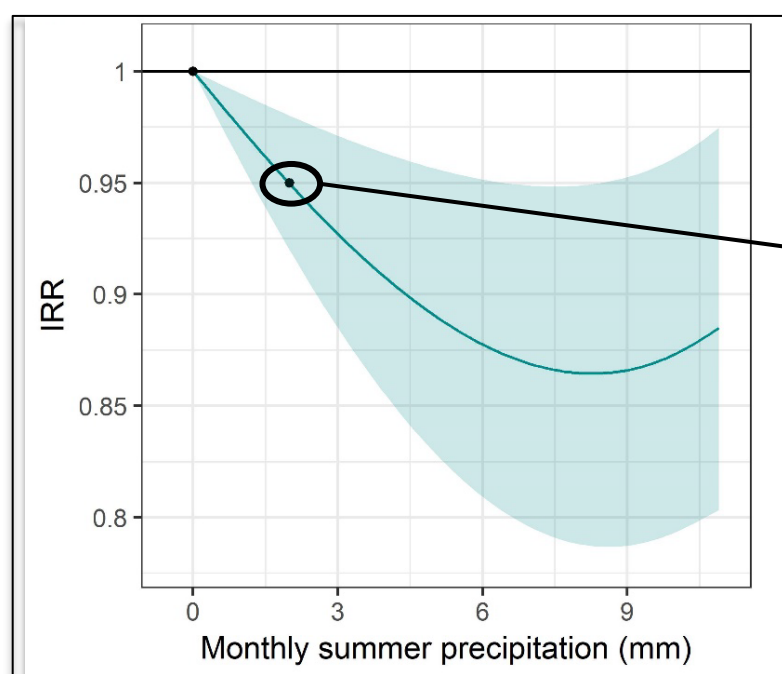
- 1) Linked census tract-level cases between Sept – Nov to environmental data lagged 1-36 mo
- 2) For each county, regressed cases against lagged temperature, precipitation, using splines to capture non-linear relationships



We obtained a non-linear exposure-response relationship for 36 lags

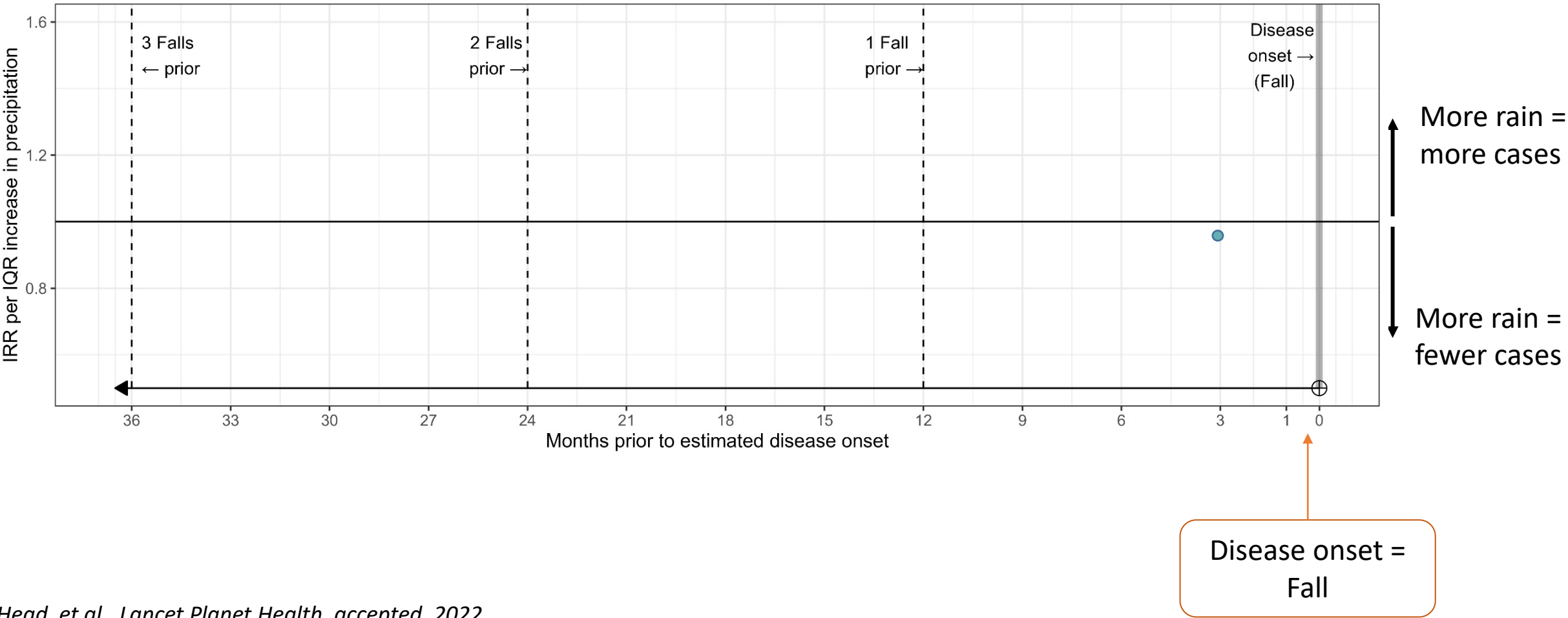
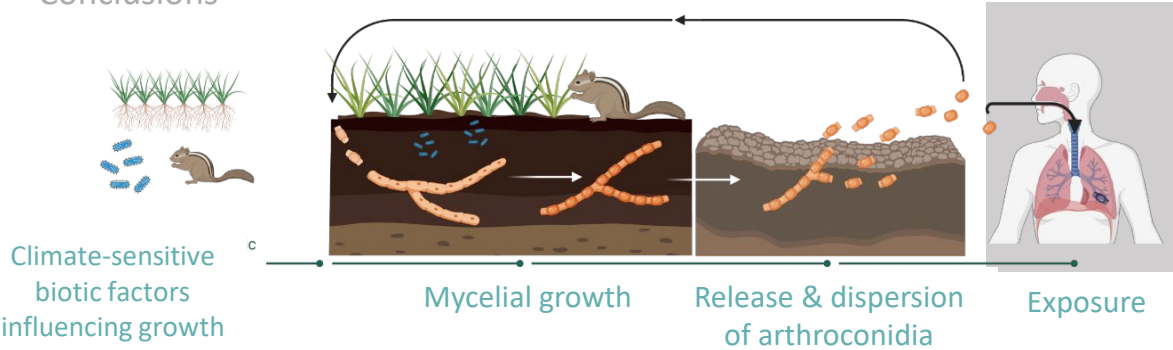
Distributed lag nonlinear modeling approach

3) Computed the incidence rate ratio at the 75th percentile of lagged precipitation or temperature compared to the 25th percentile.

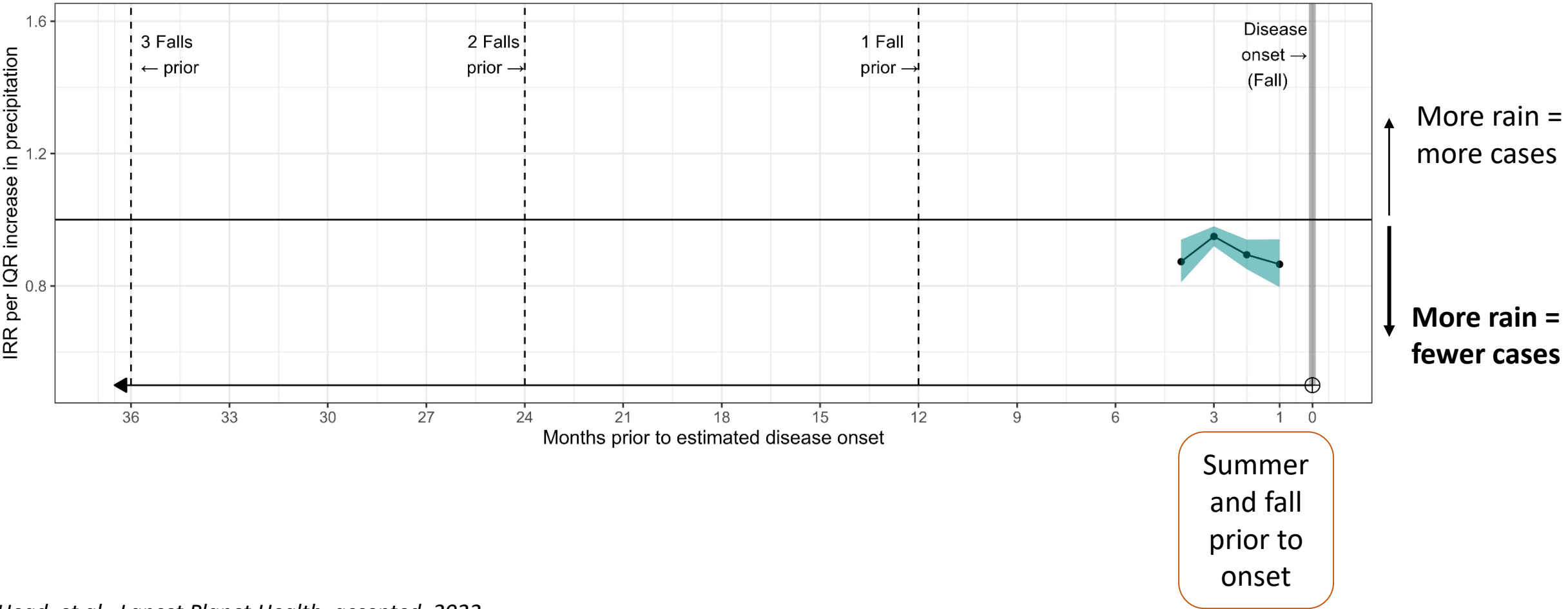
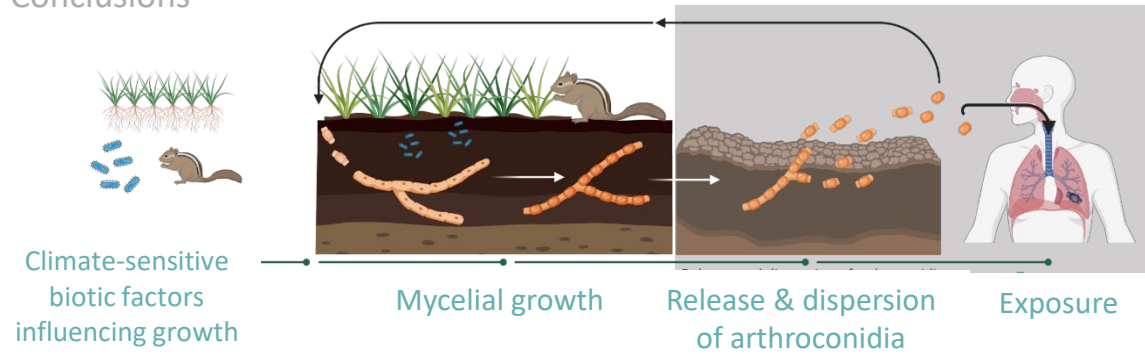


Disease onset =
Fall

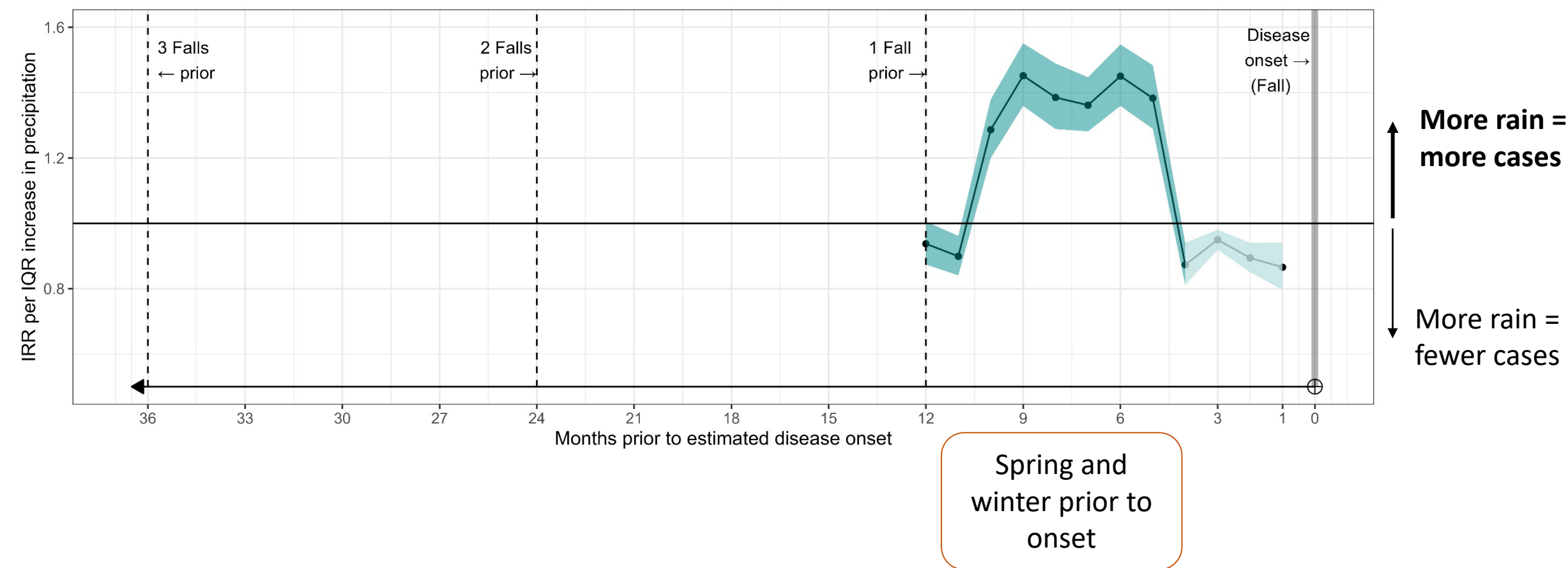
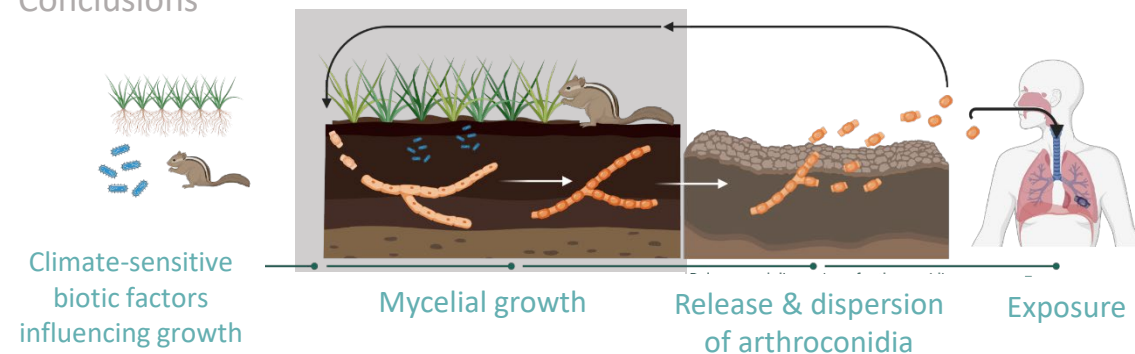
When is precipitation associated with incidence?



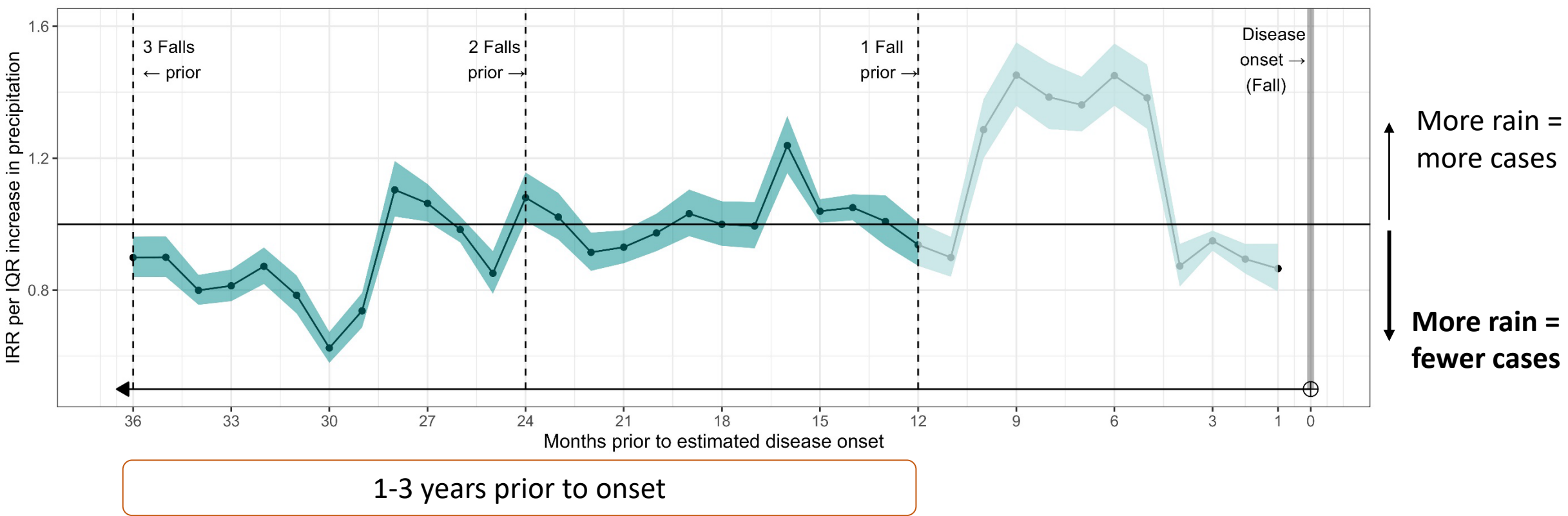
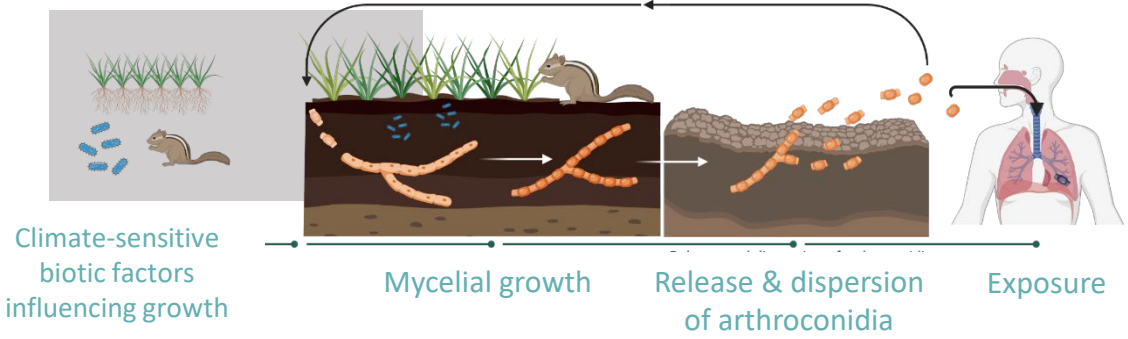
Recent rain → fewer cases



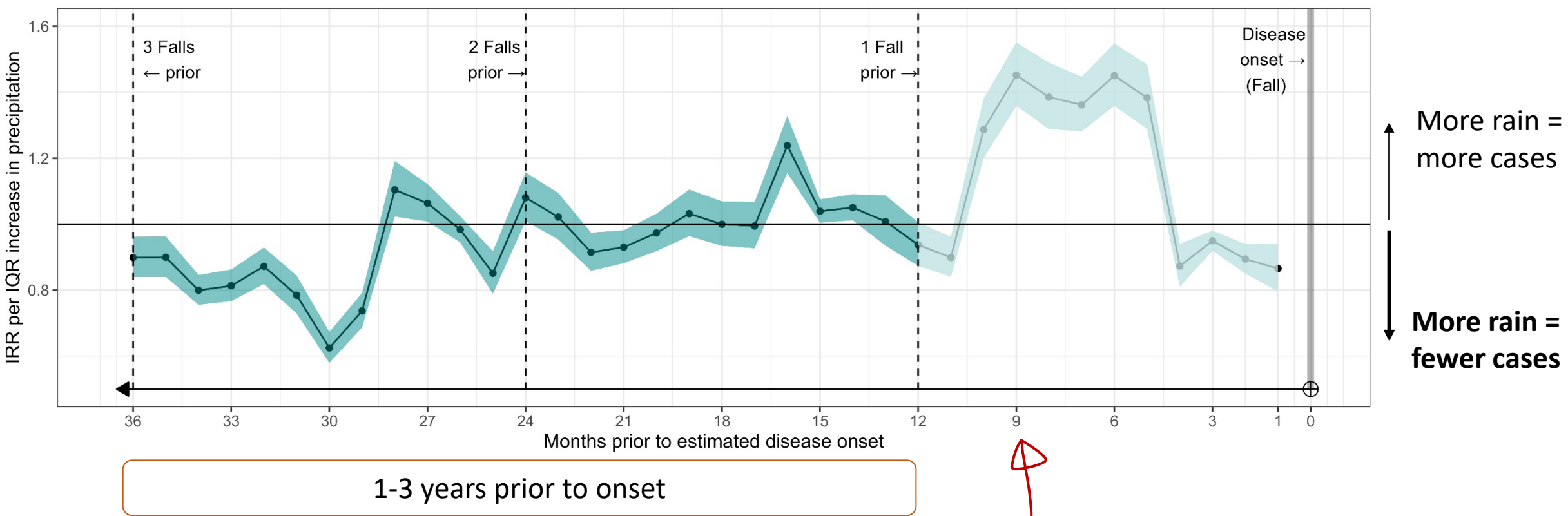
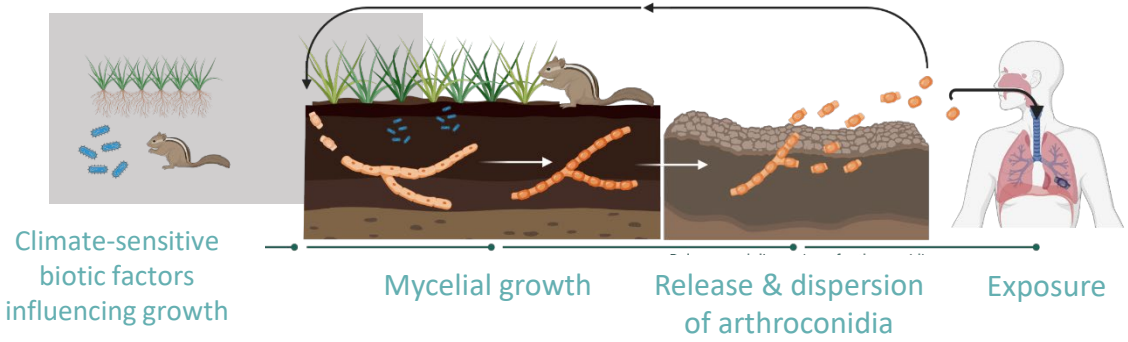
Rainfall lagged 5-10 months → more cases



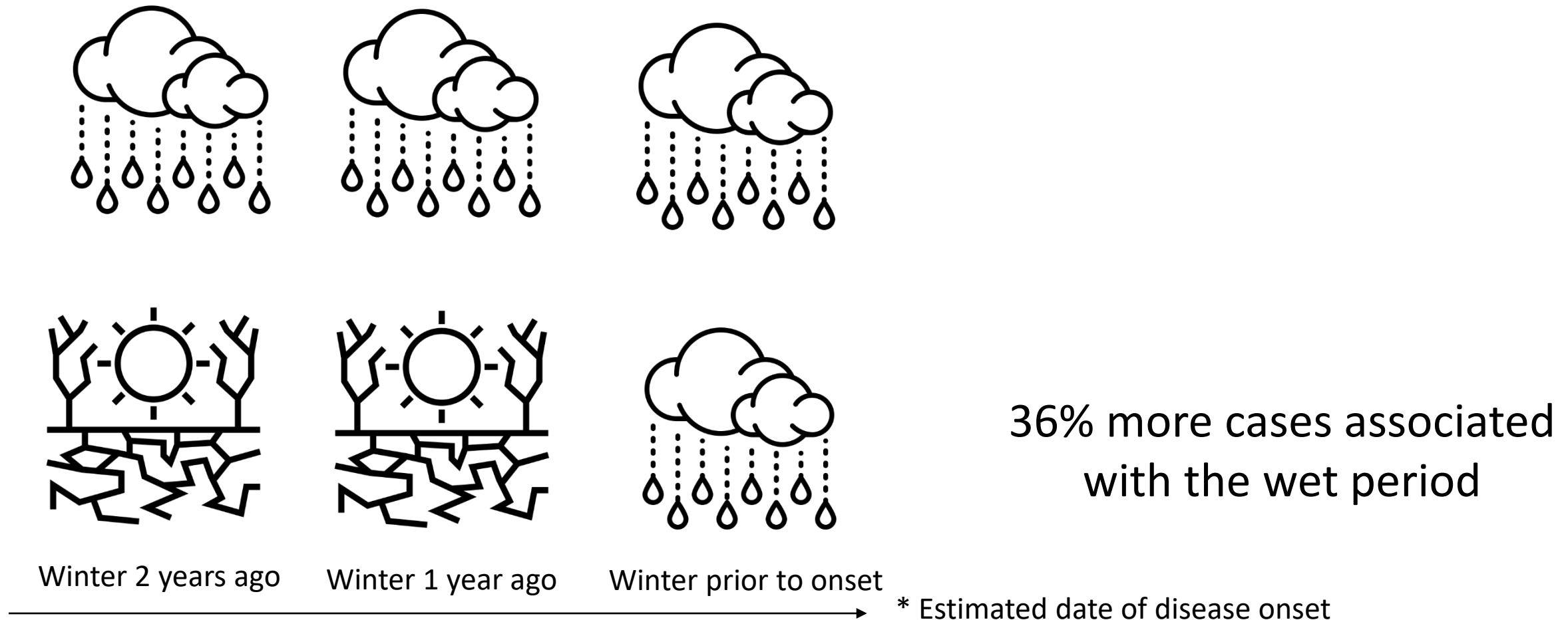
Antecedent rain → variable influence



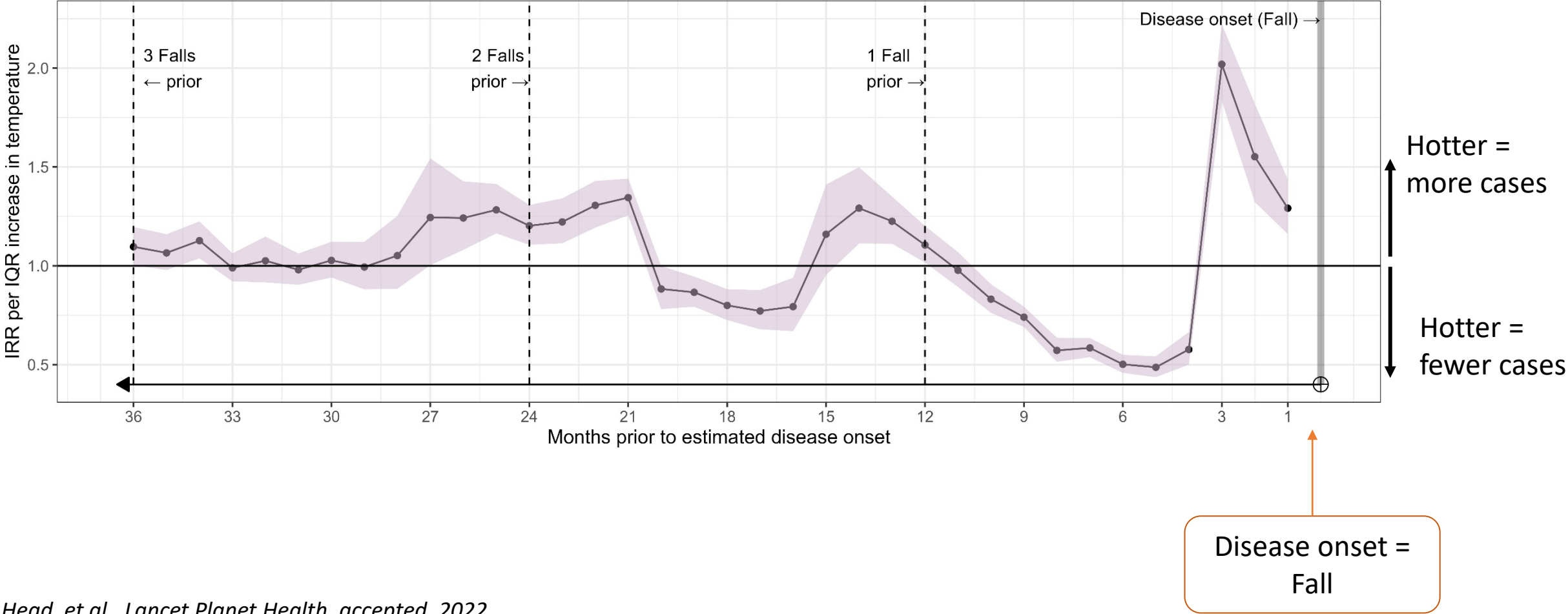
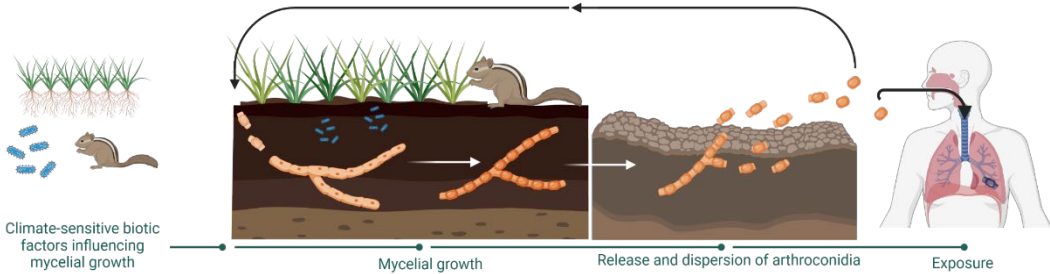
Antecedent rain → variable influence



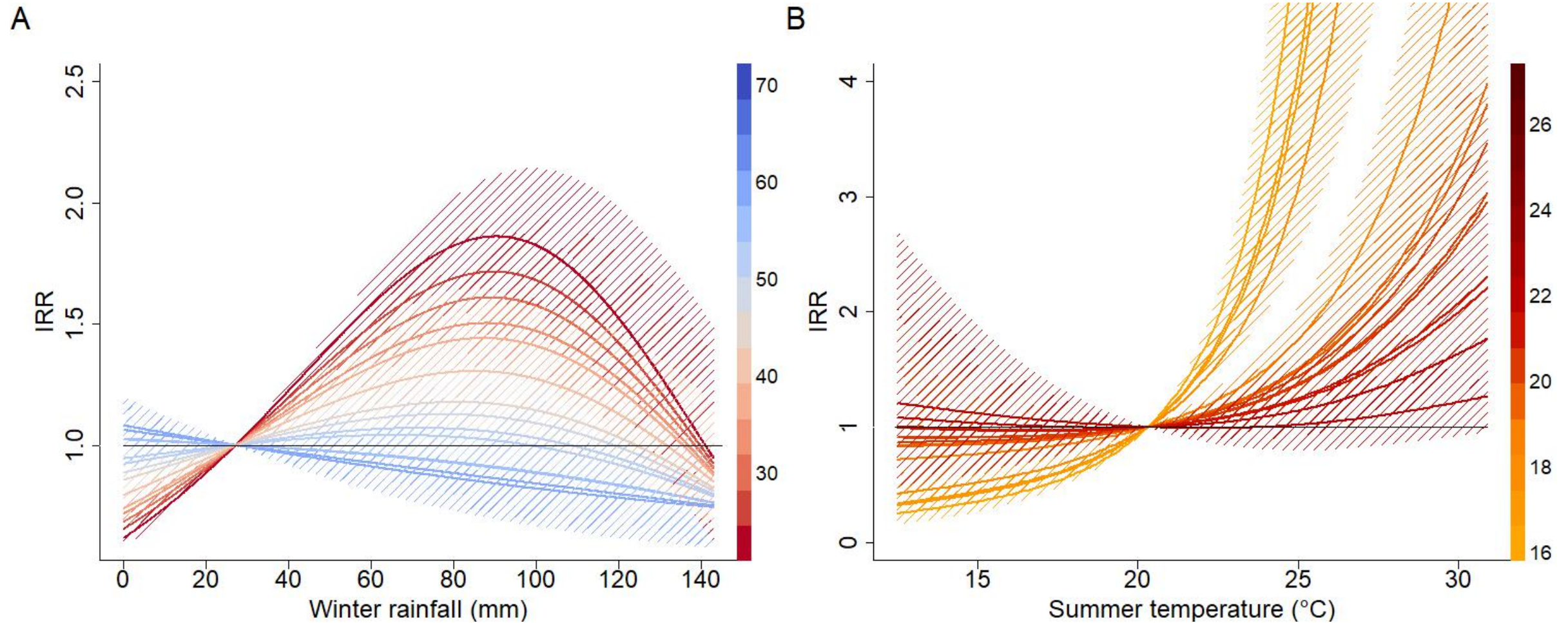
Positive effect of precipitation is *even more* enhanced when a wet winter follows *two* dry winters



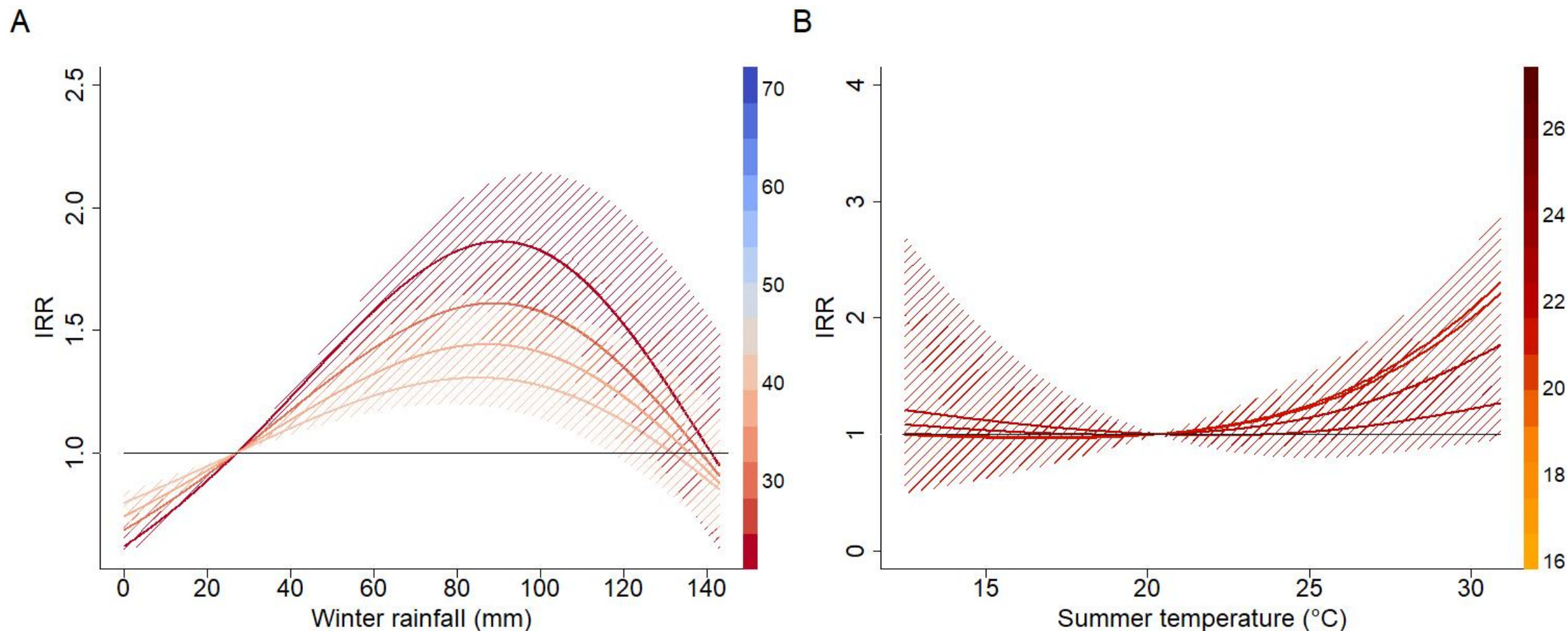
When is temperature associated with incidence?



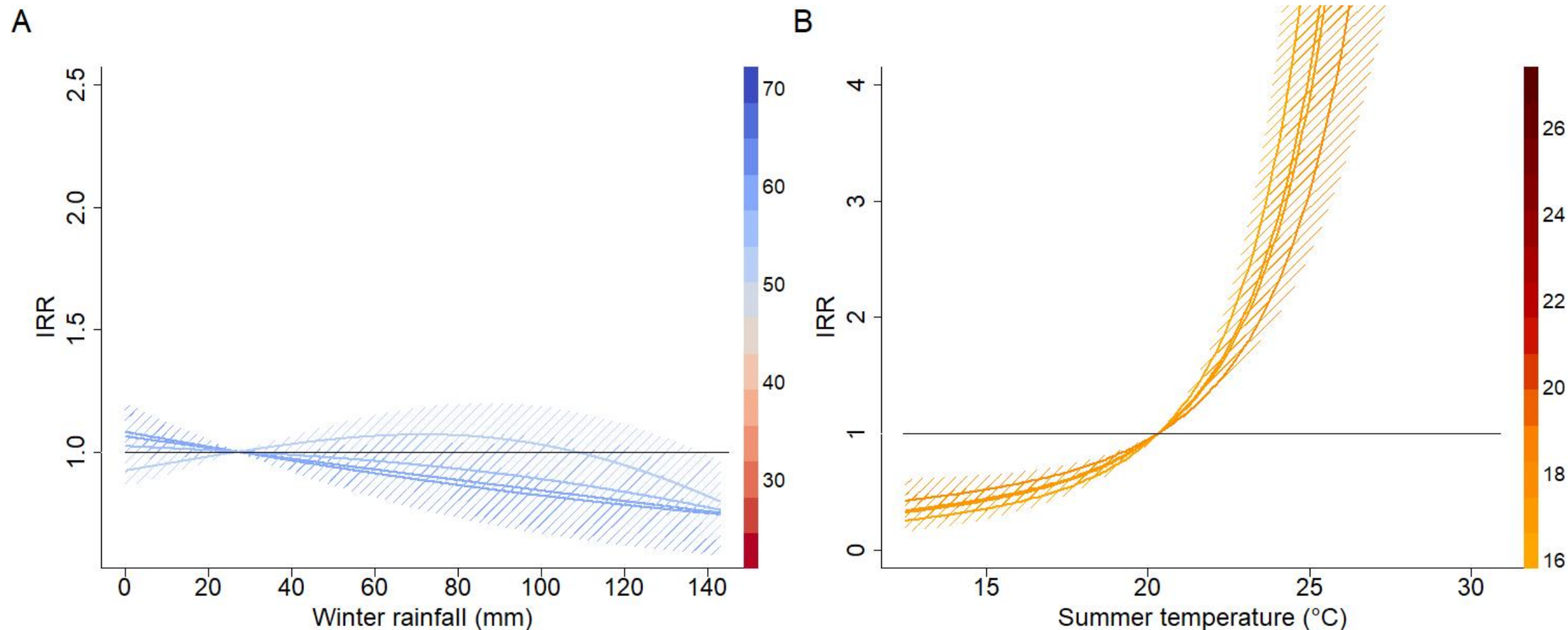
The strength of relationship varies by typical climate



Hotter, drier counties strongly influenced by **rainfall**

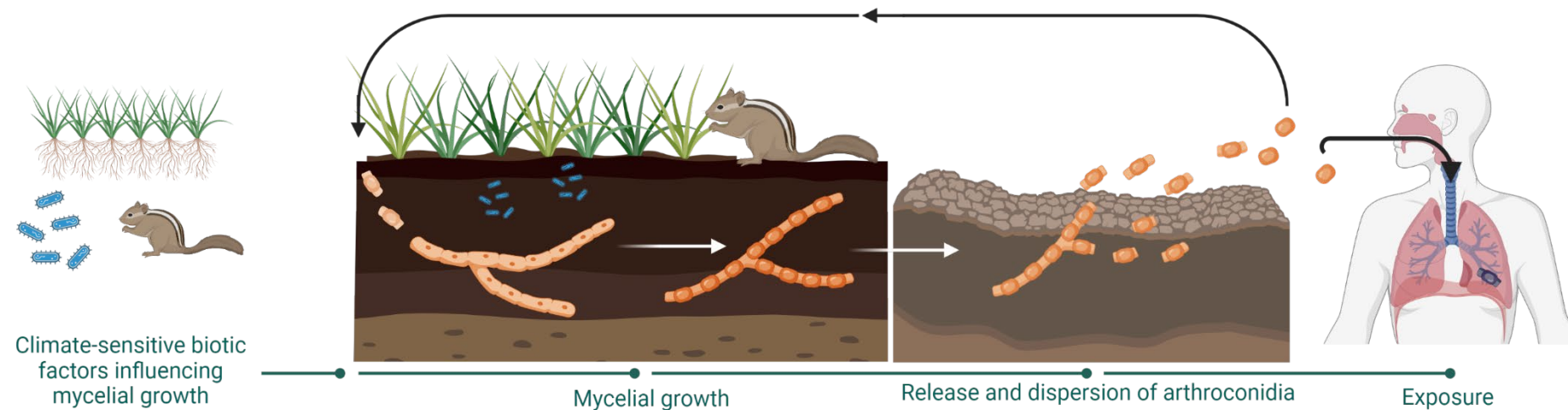


Cooler, wetter counties strongly influenced by **heat**



Summary of findings:

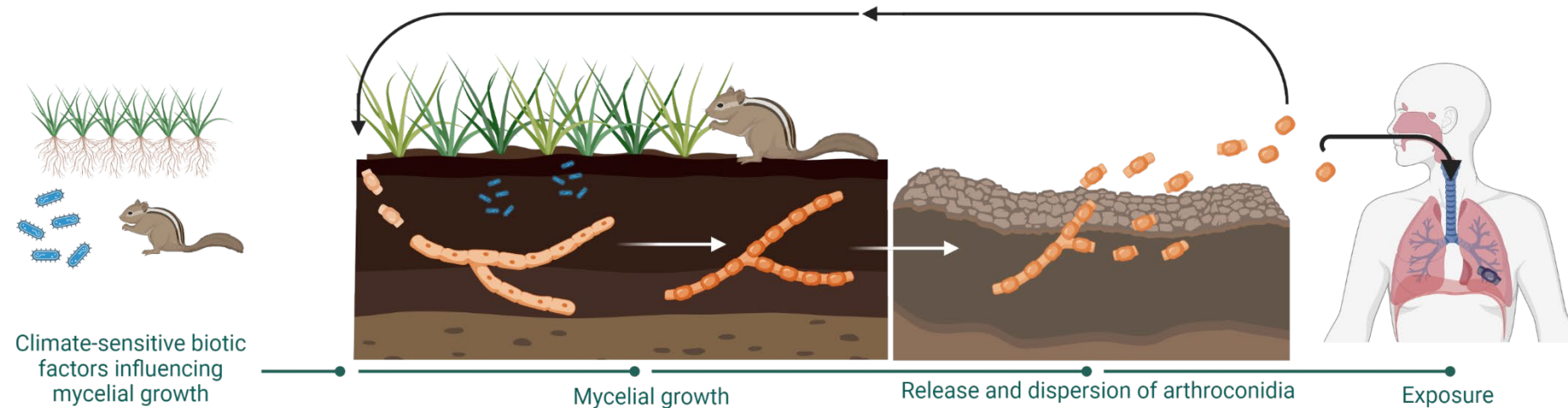
Climate affects *Coccidioides* growth and dispersion differently by time lags



Hot and dry conditions here promote incidence. **Cooler counties are especially sensitive** to heat

Summary of findings:

Climate affects *Coccidioides* growth and dispersion differently by time lags

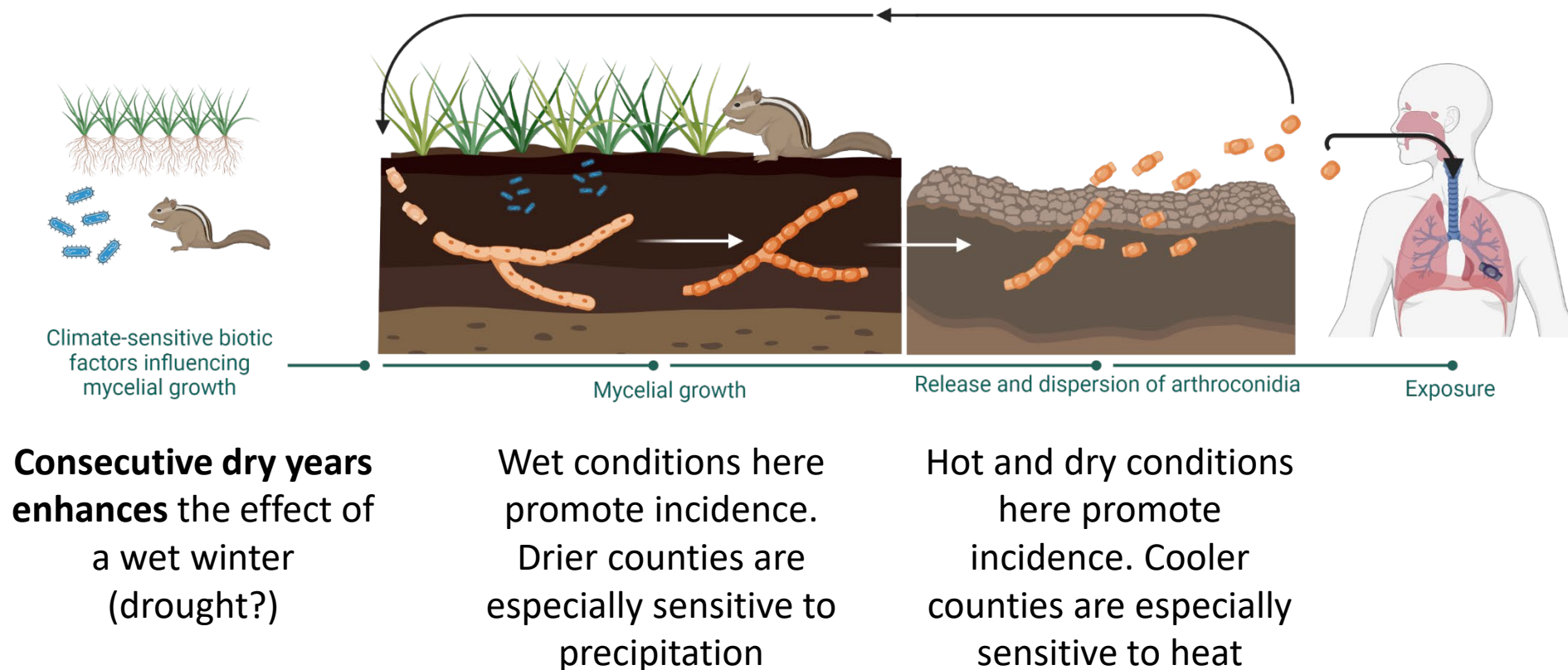


Wet conditions here promote incidence.
Drier counties are especially sensitive to precipitation

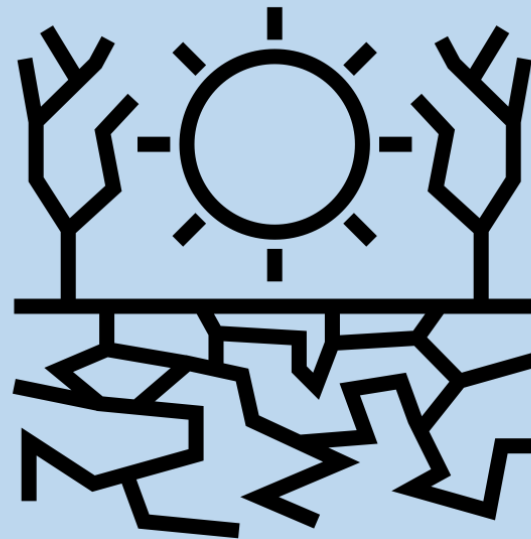
Hot and dry conditions here promote incidence. Cooler counties are especially sensitive to heat

Summary of findings:

Climate affects *Coccidioides* growth and dispersion differently by time lags



2. Determine causal effect of the 2007-2009 and 2012-2015 droughts on incidence and compare effect across regions



Method: Ensemble prediction models with G-computation

First, built ensemble prediction algorithm for each county in the study region

Applied variety of candidate algorithms:

Generalized linear models

Generalized additive models

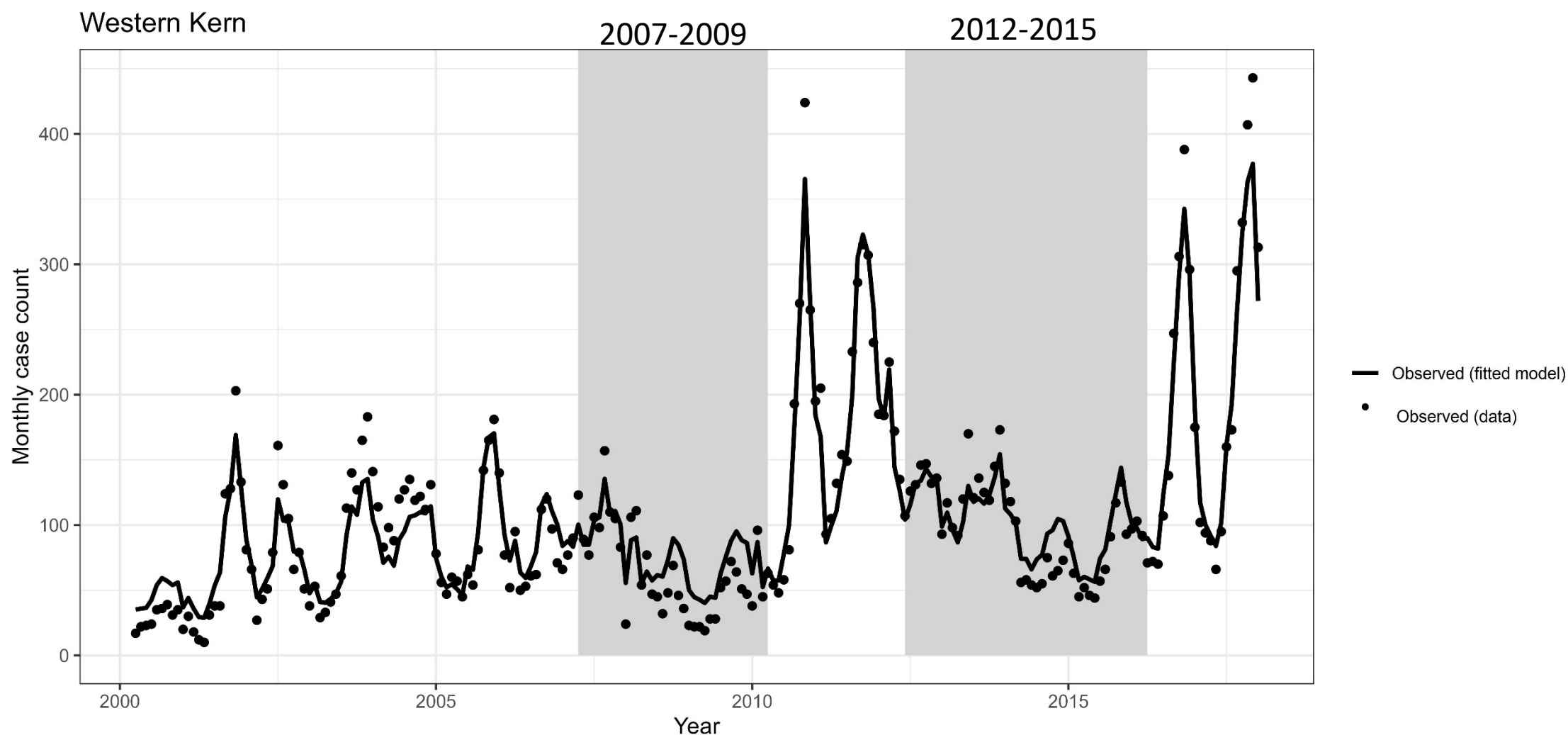
Random forest models

$$\hat{E}(Y_{i,t} | A_{i,t} = a_{i,t}, W_{i,t})$$

Used leave-out-one-year cross validation to determine out of sample prediction error

Computed weighted average of the models so as to minimize out of sample prediction error

First, built ensemble prediction algorithm for each county in the study region



Using G-computation, estimated incidence under a counterfactual “no drought” scenario

$$\hat{E}(Y) = \sum_{t=t_0}^{t=T} \sum_{i=1}^N \hat{E}(Y_{i,t} | A_{i,t} = a_{i,t}, W_{i,t})$$

$A_{i,t} \in (\text{Rainfall}_{it}, \text{Temperature}_{it})$

$W_{i,t} \in (\text{Time}, \text{elevation}, \text{soil texture})$

$\hat{E}(Y)$: expected number of cases under **observed** conditions for $A_{i,t} = a_{i,t}$

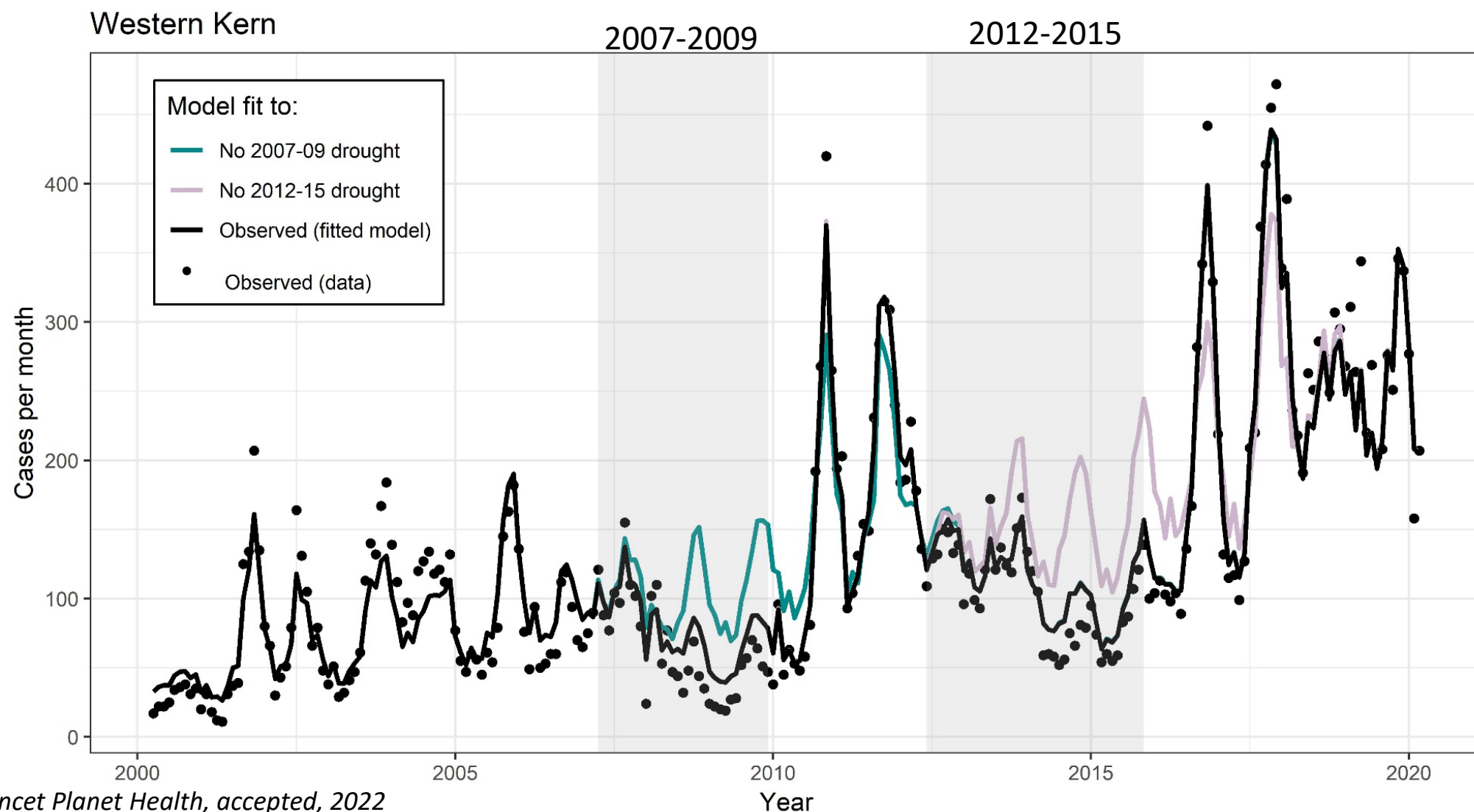
$\hat{E}(Y_0)$: expected number of cases under **counterfactual** conditions for $A_{i,t} = a_{i,t}$

- rainfall below average set to the average

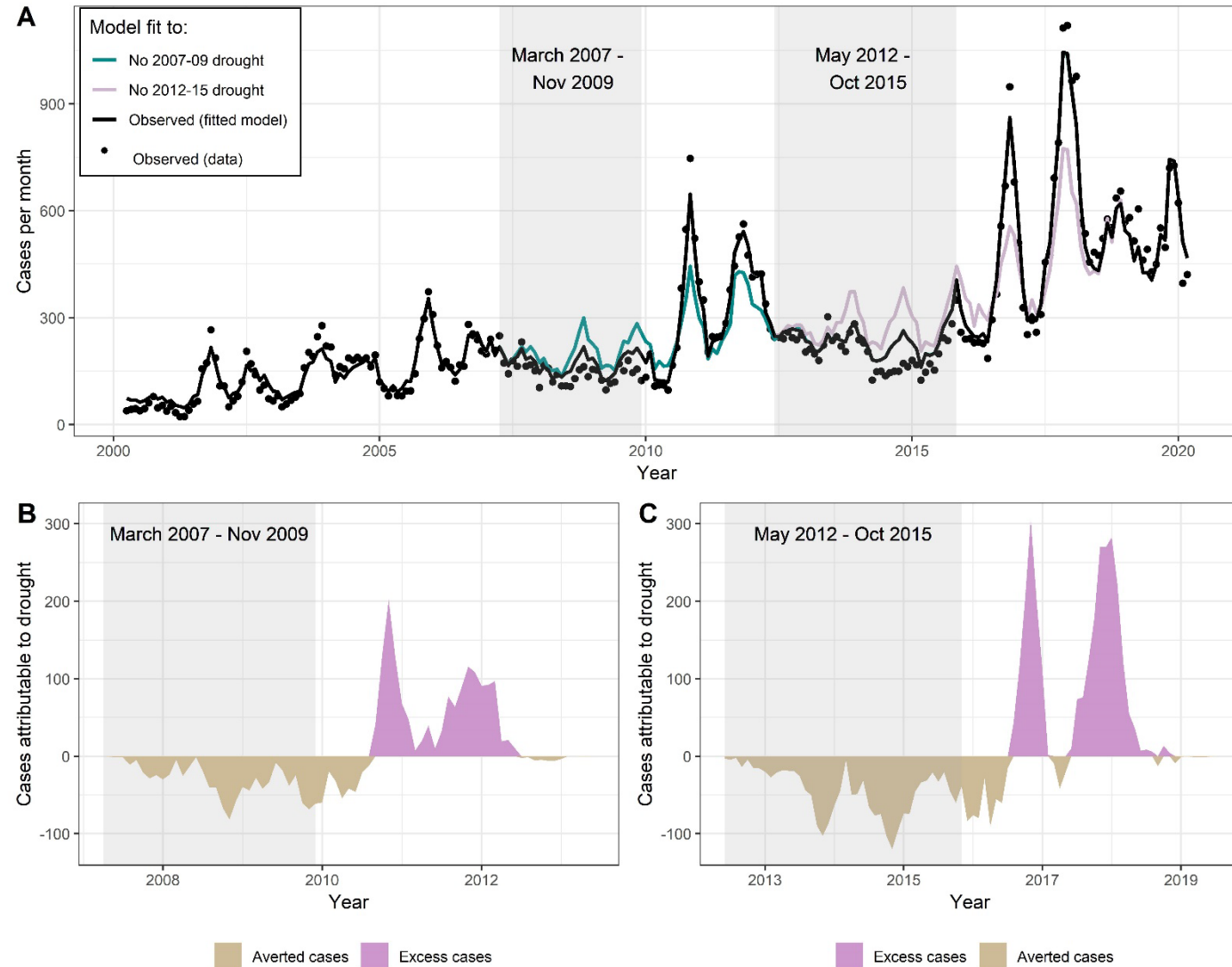
- temperature above average set to the average

$$\hat{\psi} = \hat{E}(Y) - \hat{E}(Y_0)$$

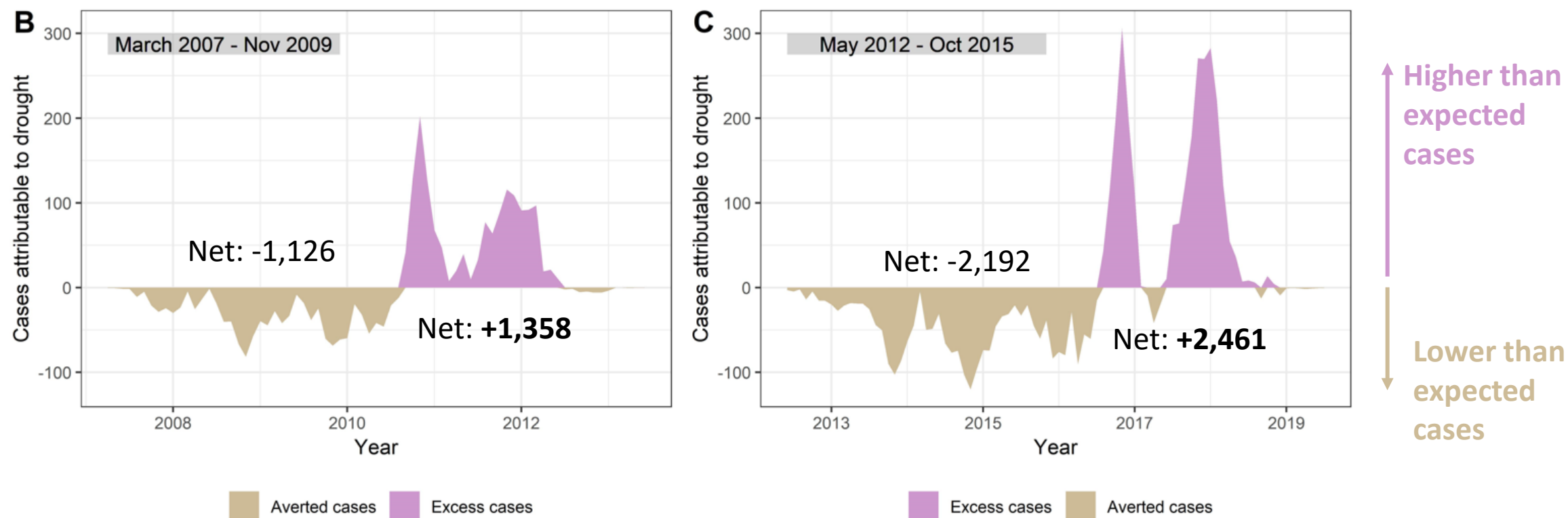
Using G-computation, estimated incidence under a counterfactual “no drought” scenario



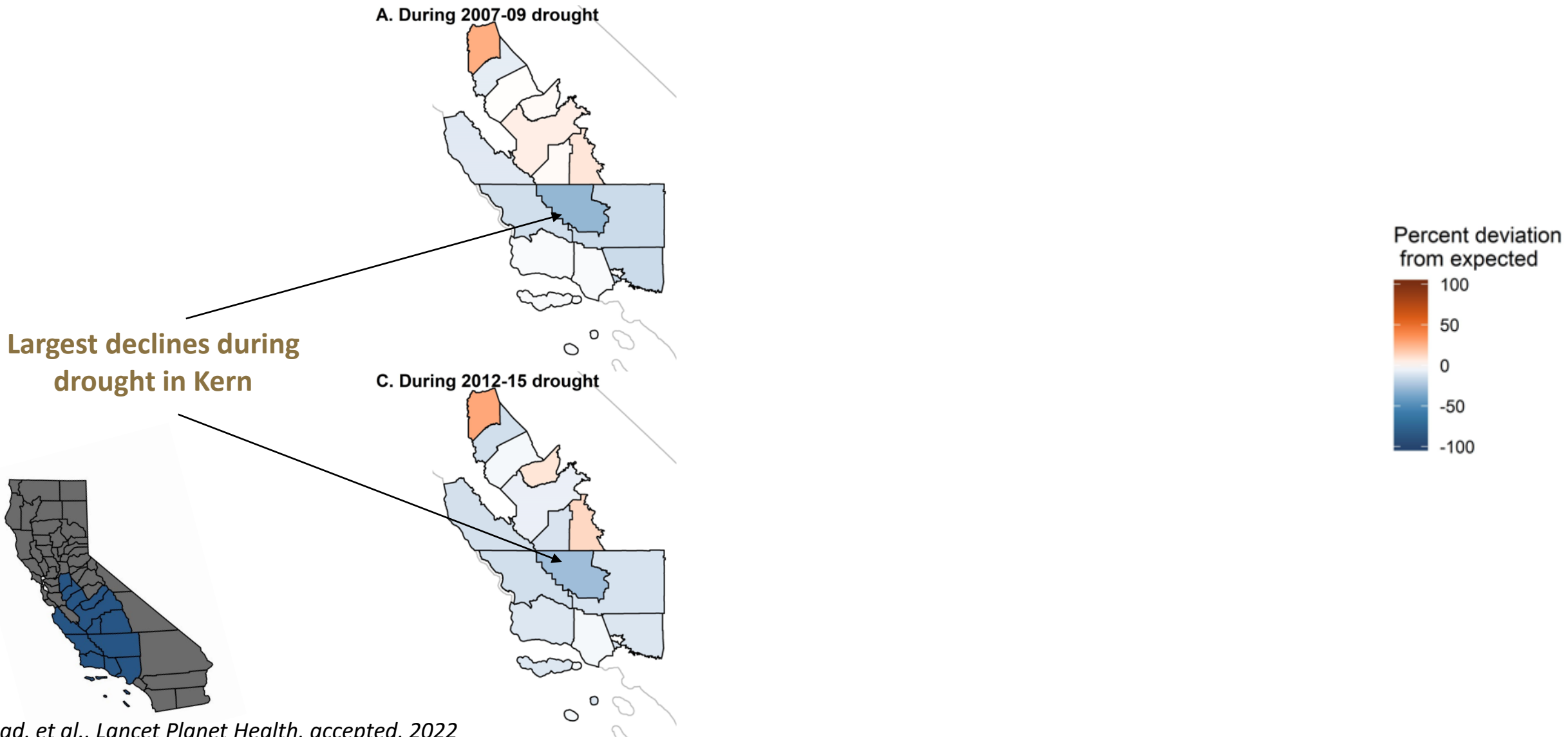
Compared differences across counties and the full study region



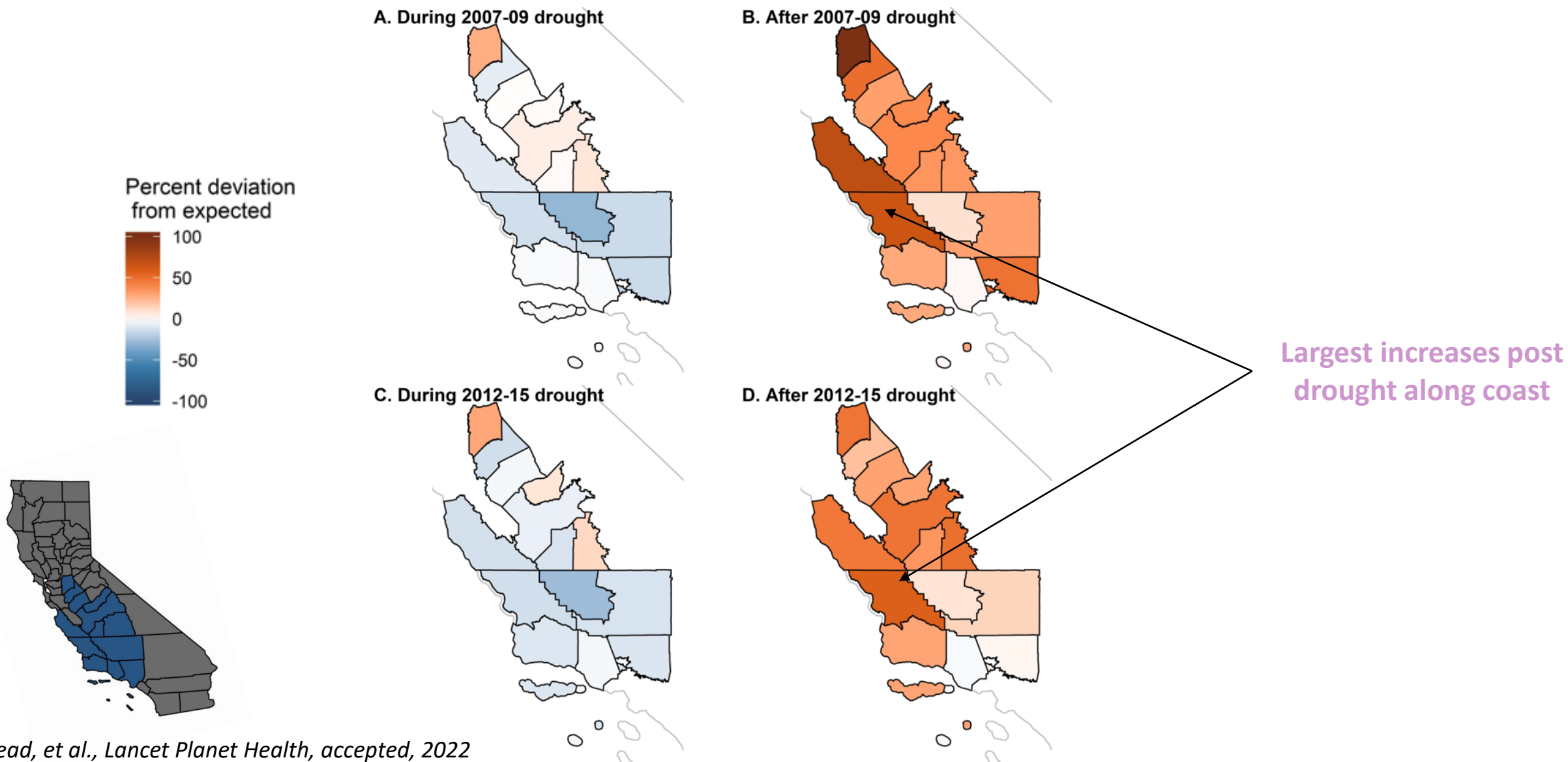
For both droughts, the decline in cases during drought was more than offset by the increase following



Kern County experienced the largest relative declines during droughts

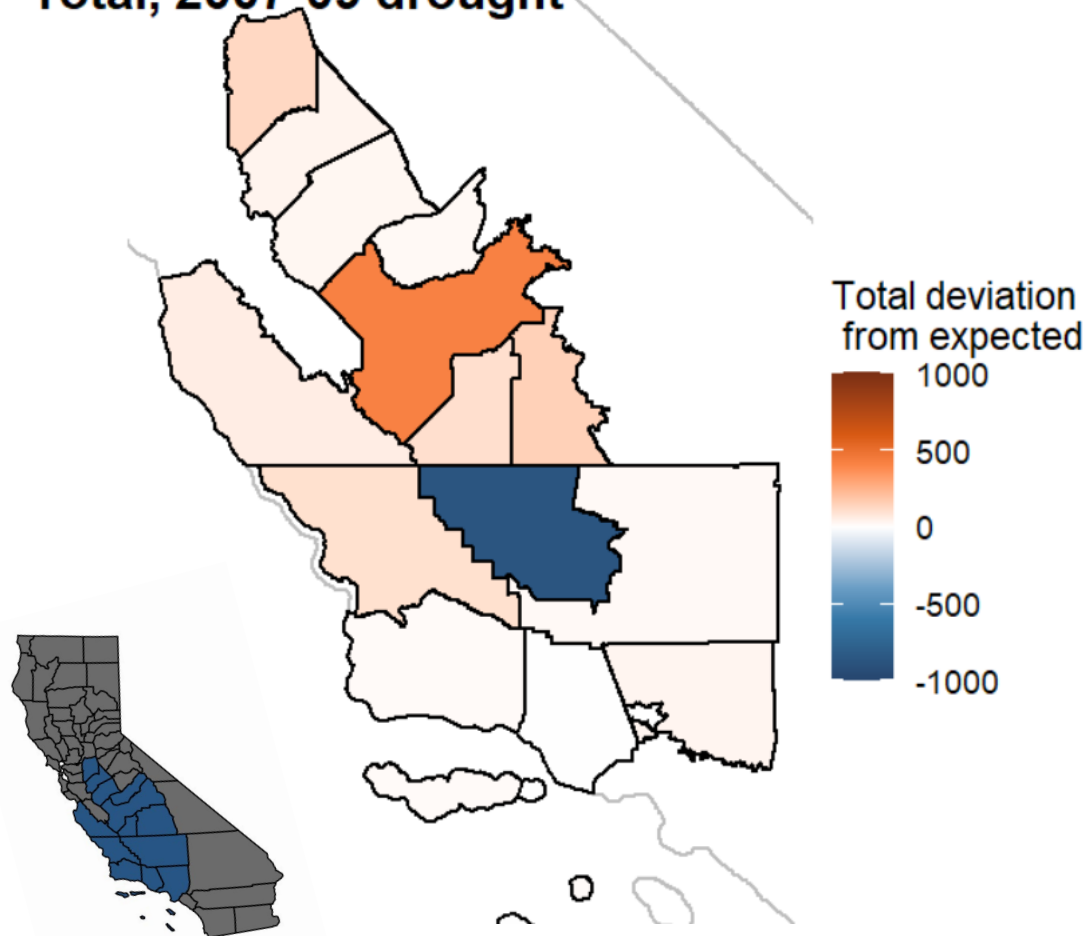


Coastal and northern valley counties experienced largest relative increases post-drought

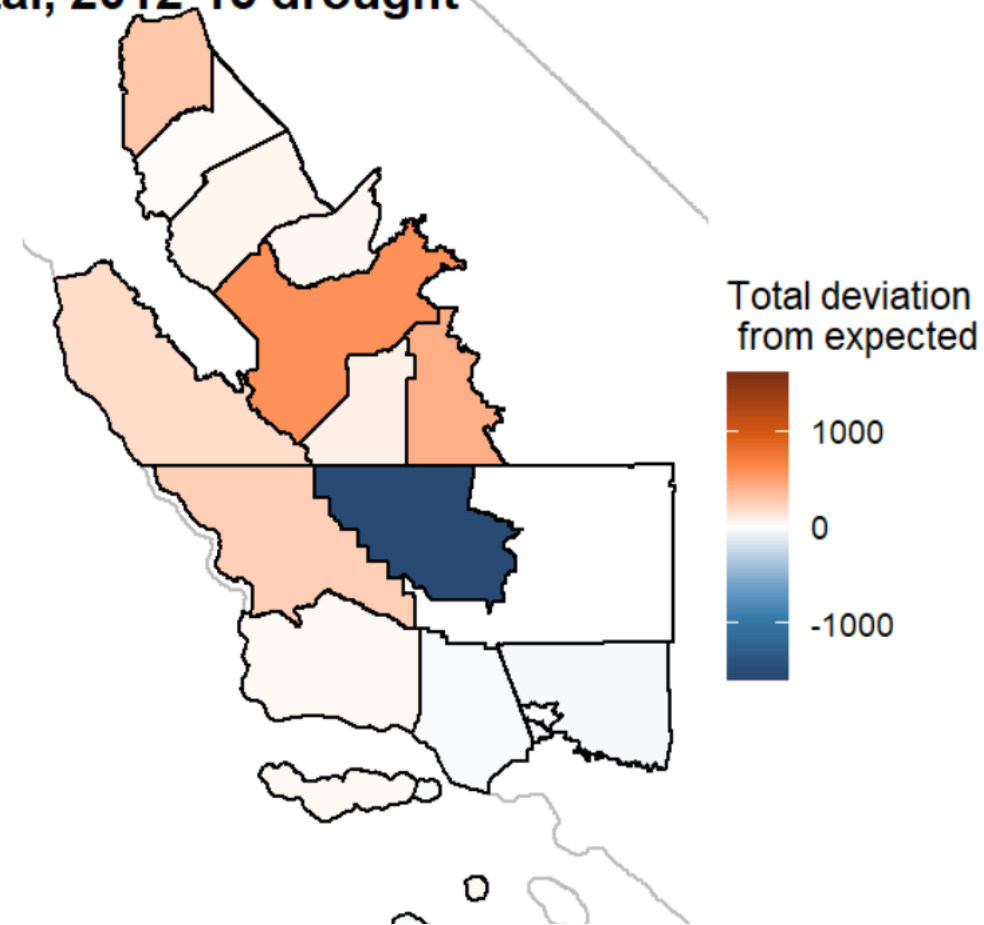


Drought increases were most notable along the coast and northern valley; only western Kern had cases averted by drought

Total, 2007-09 drought



Total, 2012-15 drought



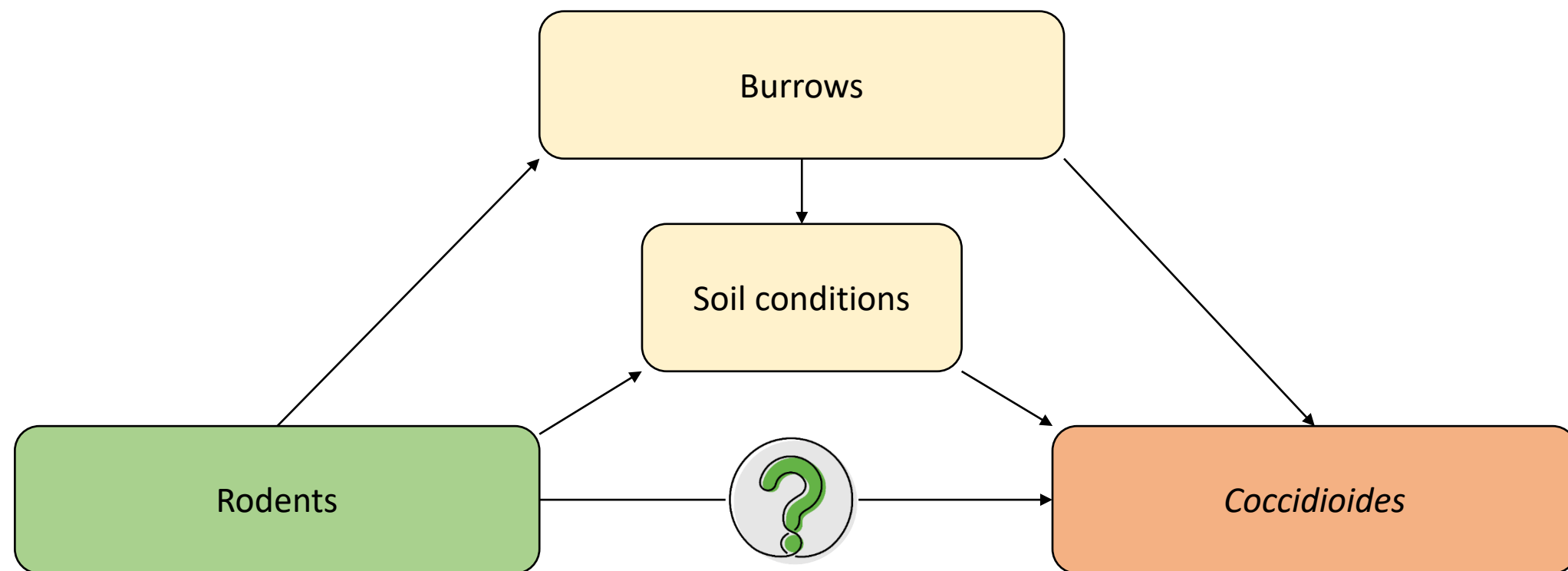
How do our findings relate to the small mammal endozoan hypothesis?

Coccidioidomycosis incidence rebounds sharply after drought

Rodent populations (generally) rebound sharply after drought

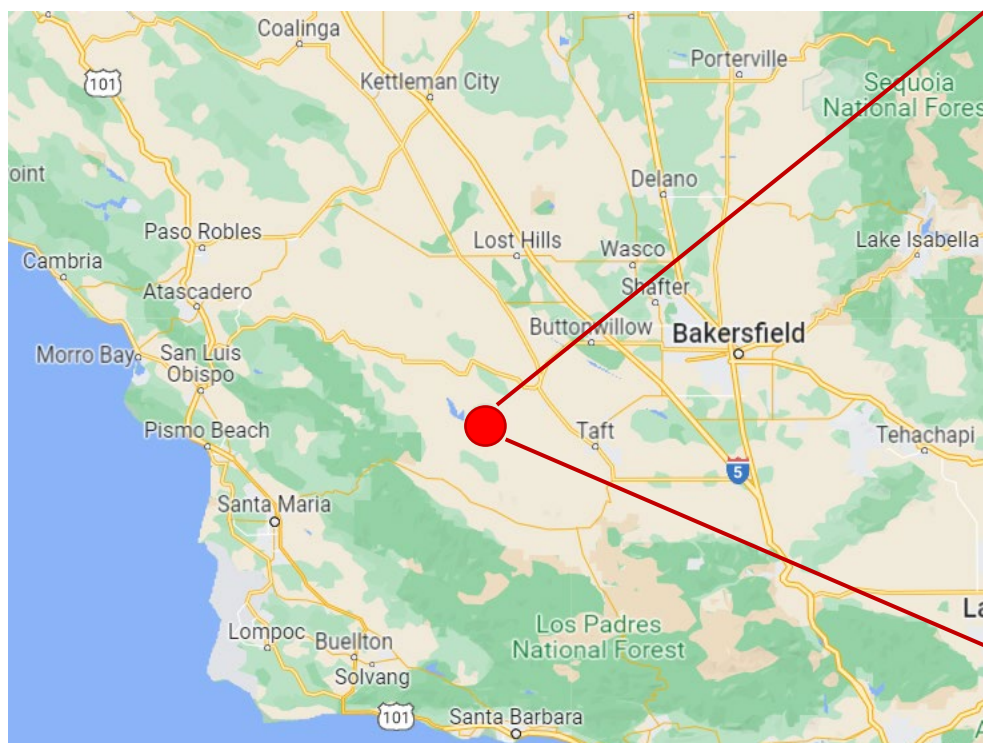


Disentangling the rodent from the burrow has been a challenge



Carrizo Plain Ecology Project

Establishment of rodent exclosures in 2007 permits disentanglement of rodents and burrows



Experimental plot



Rodent precinct

Sample seasonally across a factorial design


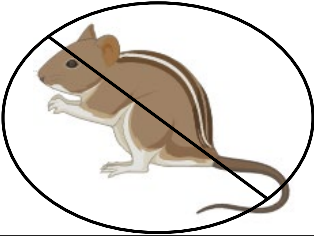

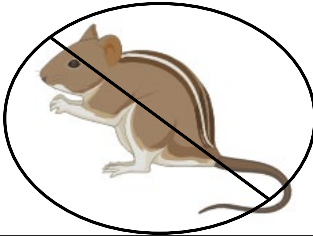




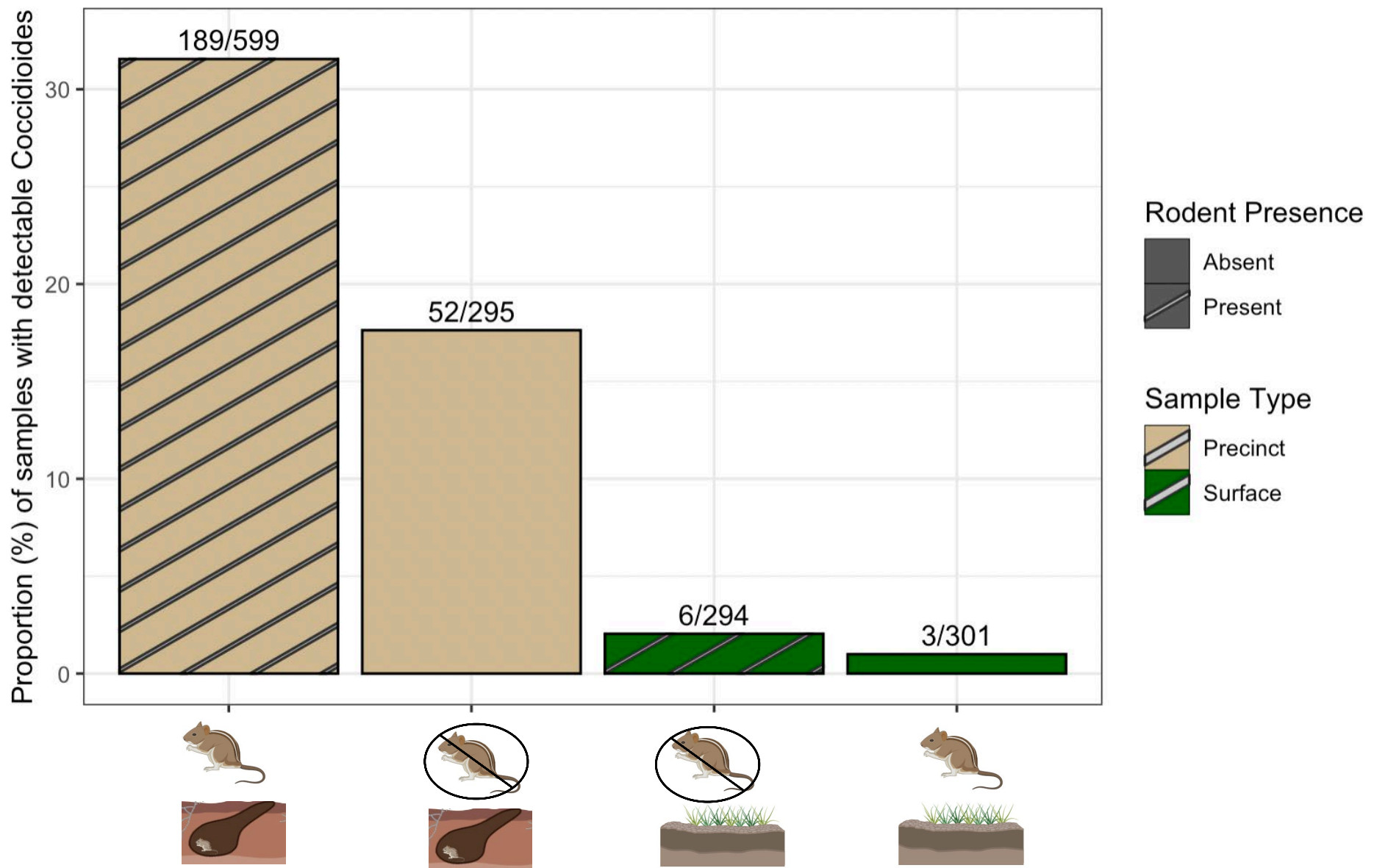

Burrows × Enclosures				
				
Clusters per plot	2	1	1	1
Samples per plot	10	5	5	5
Samples per season	200	100	100	100



Photo: Yuan Zhu

Coccidioides was detected in 250/1,489 (16.8%) of samples



A photograph of a small squirrel, possibly a chipmunk, sitting in a field of dry, yellowish-brown grass. The squirrel is facing right, with its body angled slightly towards the viewer. It has a brown back with a white stripe running down its side. The background is a soft-focus field of similar dry grass, suggesting a dry or arid environment.

Updating of our understanding of the role of climate in coccidioidomycosis epidemiology

- Drought displaces and then amplifies incidence
- Incidence is increasing most rapidly in the wetter and cooler counties. We show evidence that incidence in these counties is very sensitive to temperature variation, and have seen the largest increases following drought.
- Frequency and severity of drought is expected to increase. Increasing aridity and greater swings in climate extremes may lead to expansion in wetter, cooler areas. Increasing aridity may lead to declines among already very dry counties.
- The role of climate and rodent populations on *Coccidioides* may be complex and interrelated

Strengths and limitations of these analyses

- + Applies causal inference techniques to the question of drought and incidence
- + Links incidence to environmental exposures at the census tract level
- + Use flexible model structures to capture interactions and non-linear relationships
- Cases are assigned to the month of estimated date of disease onset
- Cases are assigned to the census tract of residence
- Drought can be defined in multiple manners

University of California, Berkeley

Justin V. Remais, PhD

John Taylor, PhD

Ellen Eisen, ScD

Alan Hubbard, PhD

Alexandra Heaney, PhD

Isabel Jones, PhD

Simon Campo, MPH

Robert Wagner, PhD

Whitney Mgbara, MPH

Sophie Phillips

Nicole Keeney

Amanda Gomez-Weaver, MPH

Erika Lee, MPH

California Department of Public Health

Seema Jain, MD

Duc Vugia, MD

Gail Sondermeyer-Cooksey, MPH

Alexander Yu, MD

University of California, Los Angeles

Gregory Okin, PhD

Dennis Lettenmaier, PhD

Texas Tech University

Abbinash Bhattachan, PhD

California Polytechnic State University

Tim Bean, PhD

Thank you!



Head et al., Effects of precipitation, heat and drought on incidence and expansion of coccidioidomycosis in western USA: a longitudinal surveillance study, *Lancet Planetary Health*, accepted, 2022.



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