

Climate Change Impacts on Rice Yields in Lao PDR

Drew Behnke, Sam Heft-Neal, and David Roland-Holst
University of California

Workshop on the Regional Rice Initiative (RRI) - Strategic Objective 2
Food and Agriculture Organization of the United Nations

Plaza Athénée Bangkok, 61 Wireless (Witthayu) Road, Bangkok, Thailand
25-26 November 2013



Climate Change and Food Security

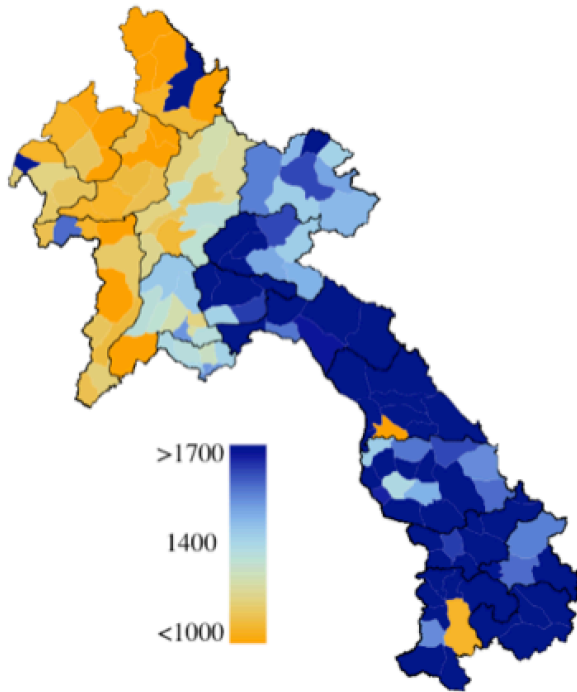
Climate Risk: About Mean and Variance

- Like sea level, the mean and variance of temperature are both rising:
 - Trend movement in climate variables will have the most lasting effects on patterns of food production, but
 - Stochastic variation is the greater and more immediate threat to local and national food security
- Adaptation needs to be learned, but we already have a lot of experience with variance.
- By improving early warning capacity, we can begin now to design and target appropriate interventions.



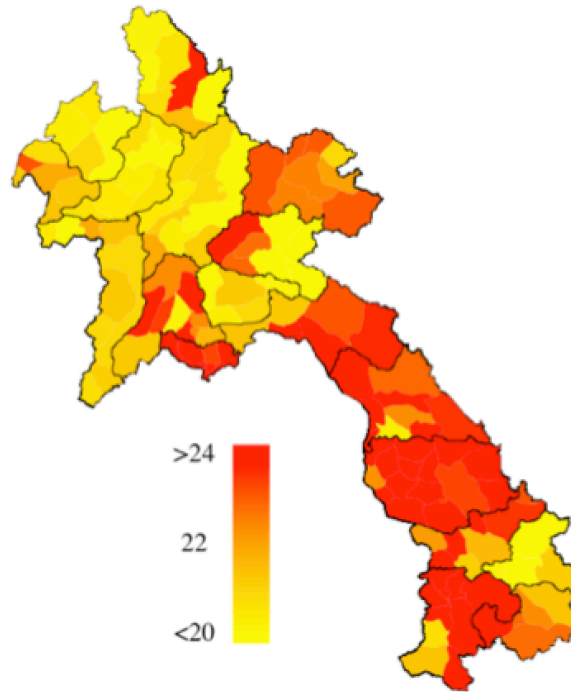
Mean Wet-Season Weather Conditions

Rainfall



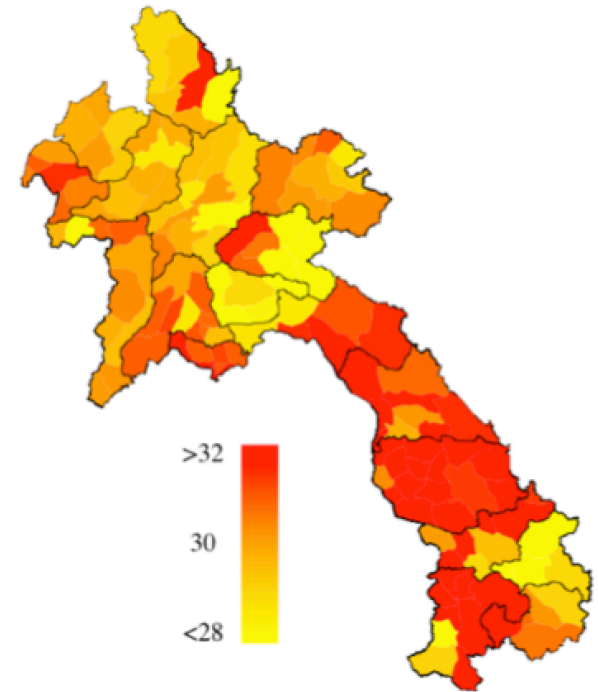
(Units = mm)

Minimum Temp



(Units = Degrees C)

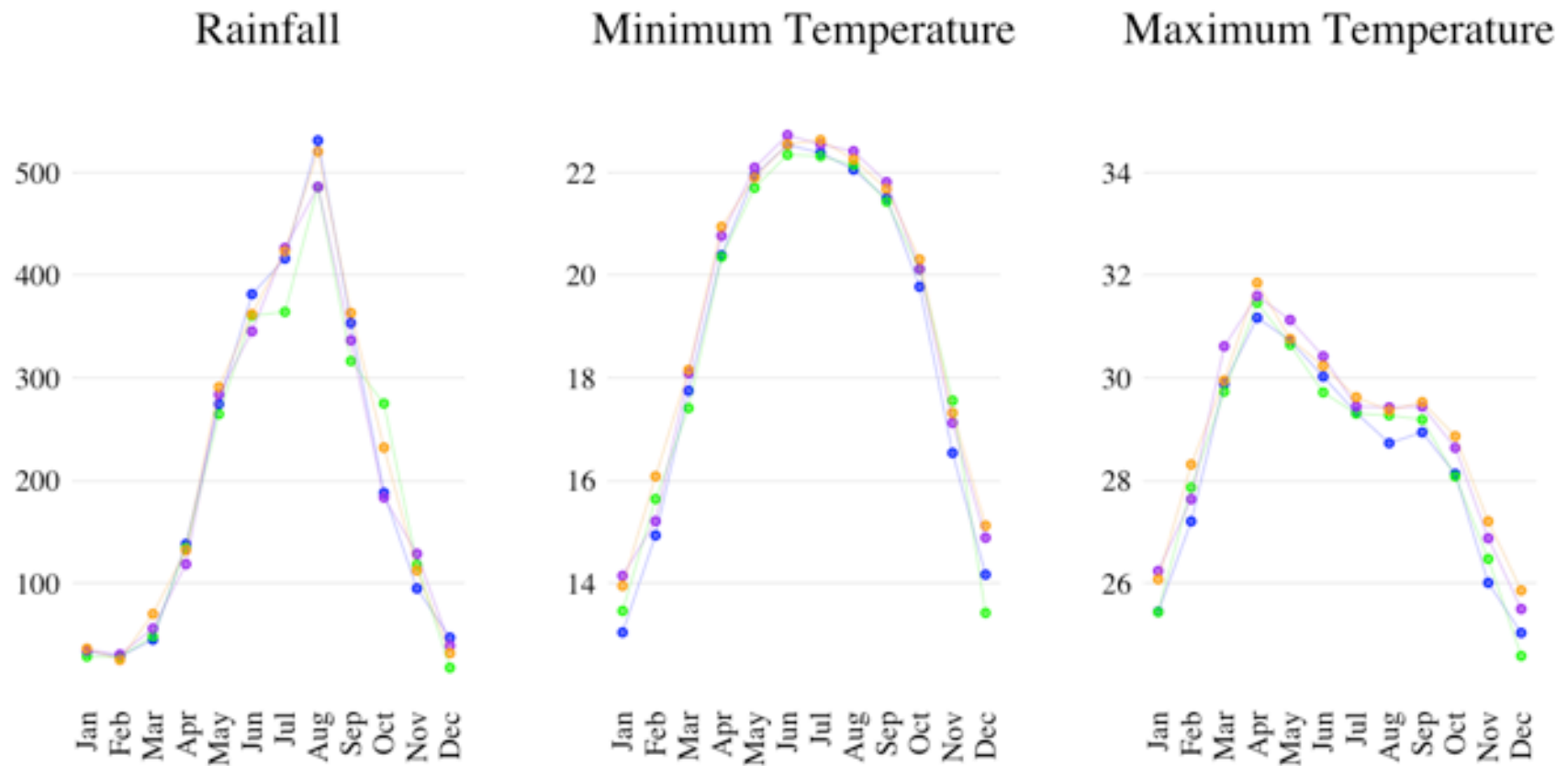
Maximum Temp



(Units = Degrees C)



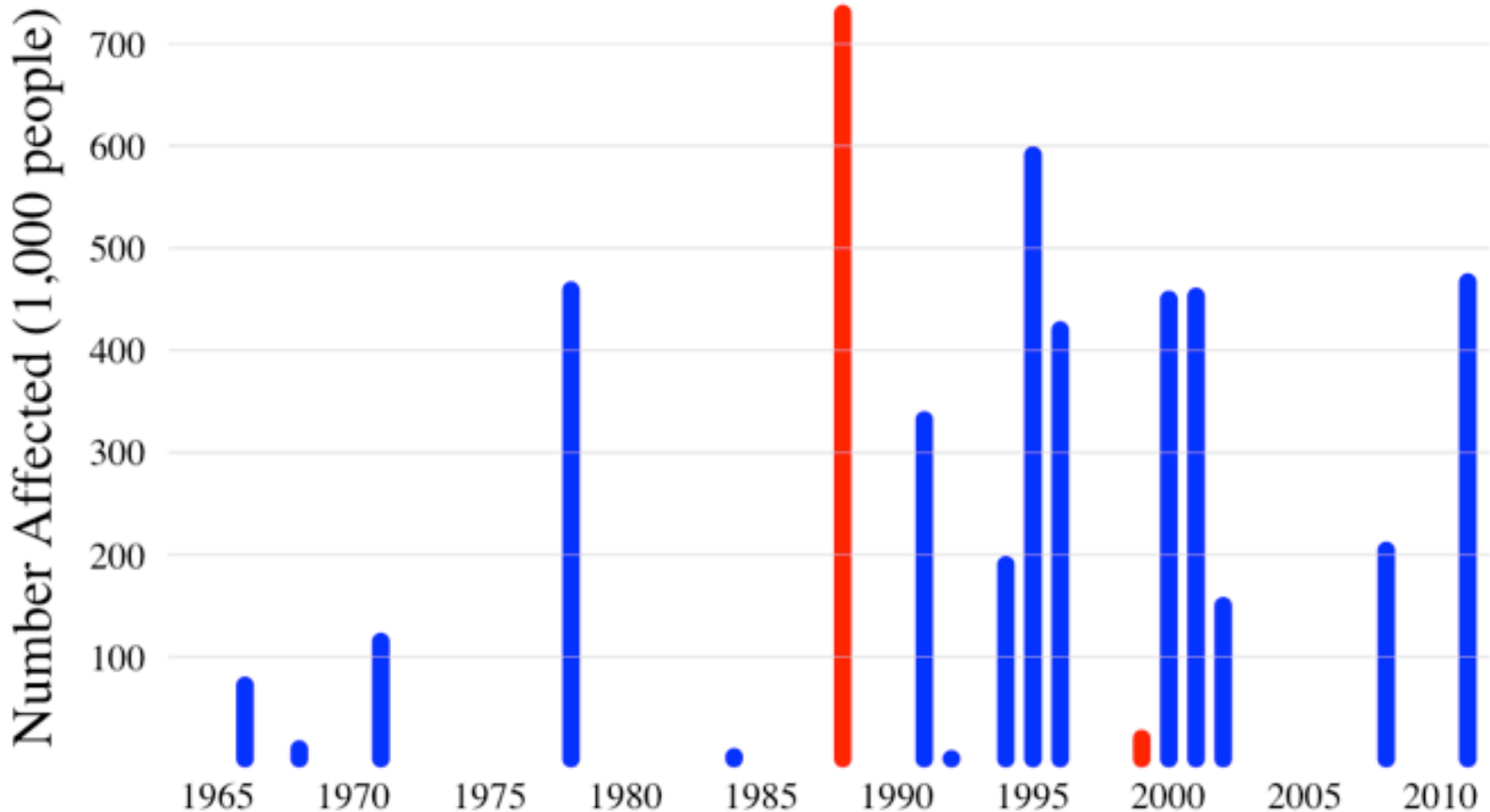
Mean Decadal Changes in Seasonal Weather Conditions



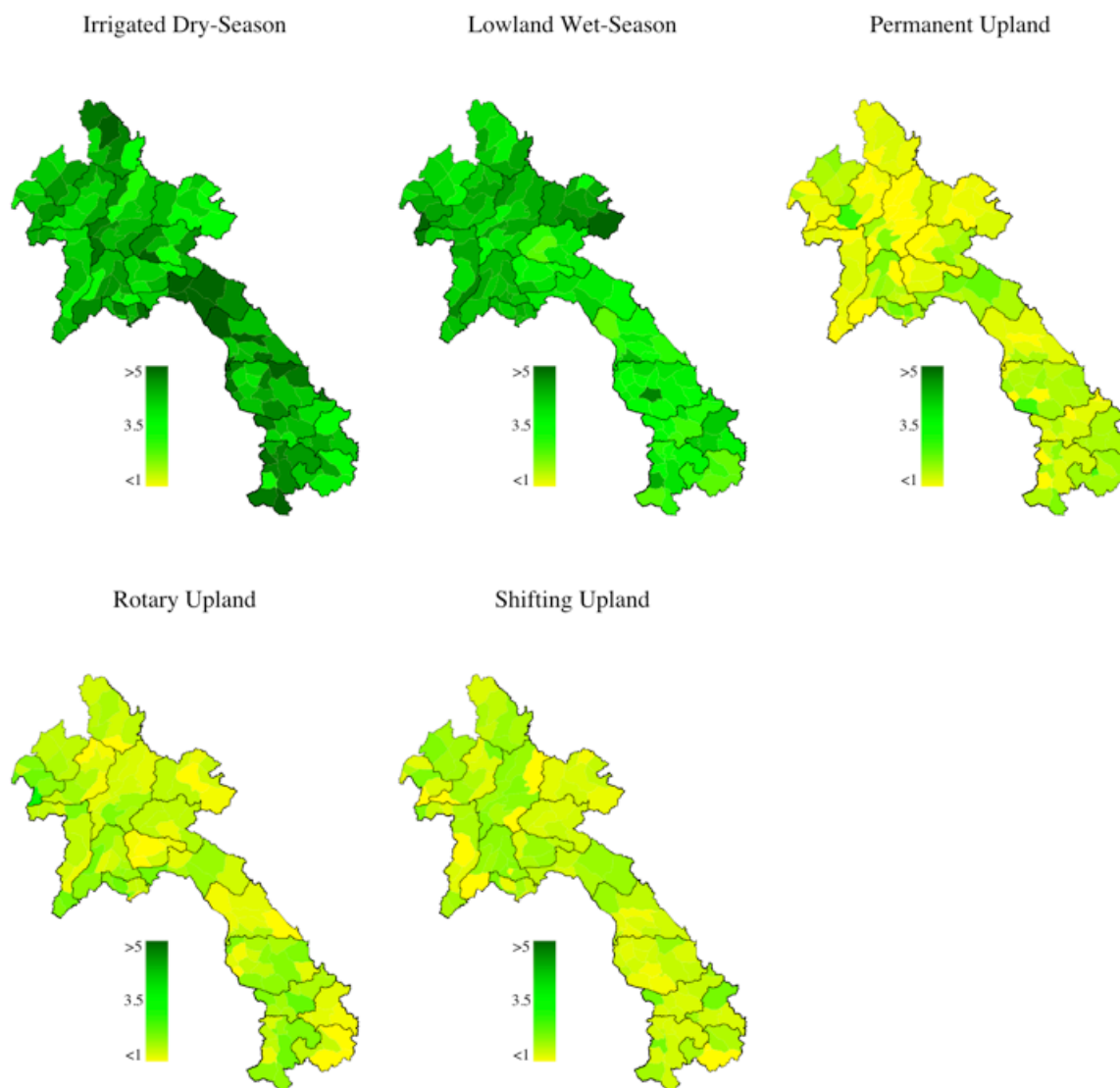
blue = 1970s, green = 1980s, purple = 1990s, orange = 2000s



Population Affected by Major Flood and Drought Events in Lao PDR



Average Rice Yields (t/ha), 2006-2012

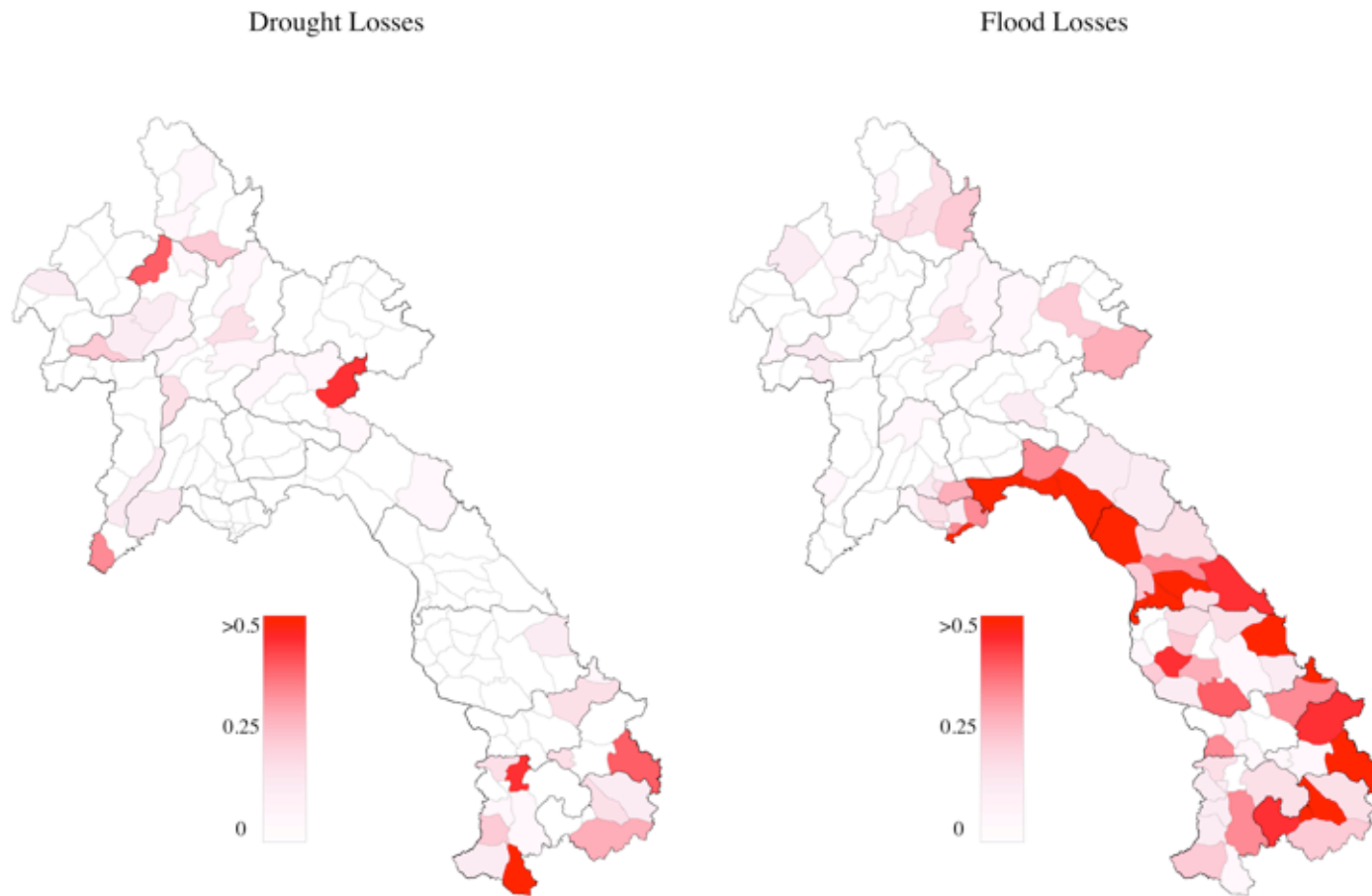


Source: Crop Statistics Yearbook (DOA, Lao PDR)

Slide 6



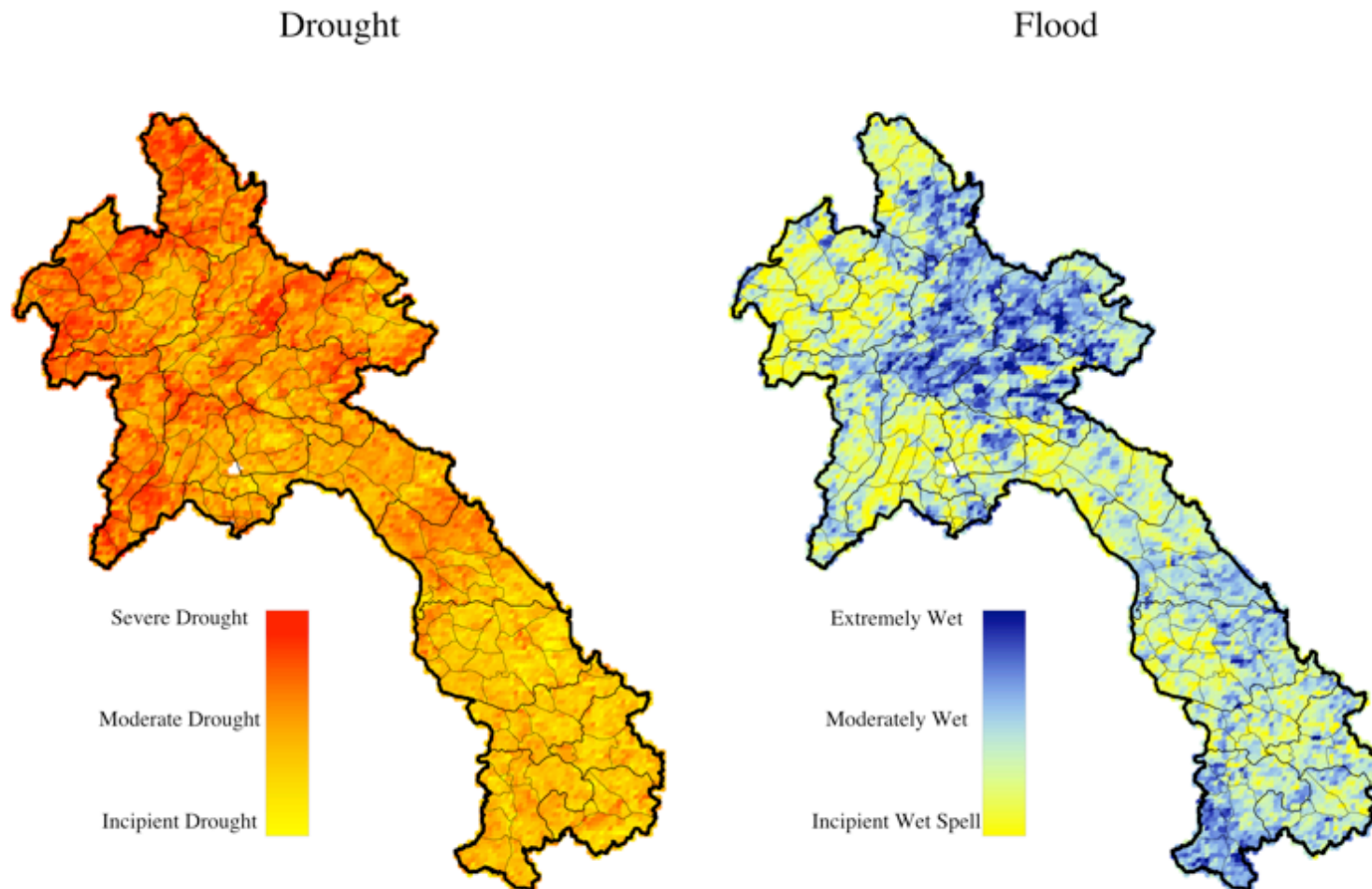
Largest Rice Area Losses by Cause, 2006-2012



Source: Crop Statistics Yearbook (DOA, Lao PDR)



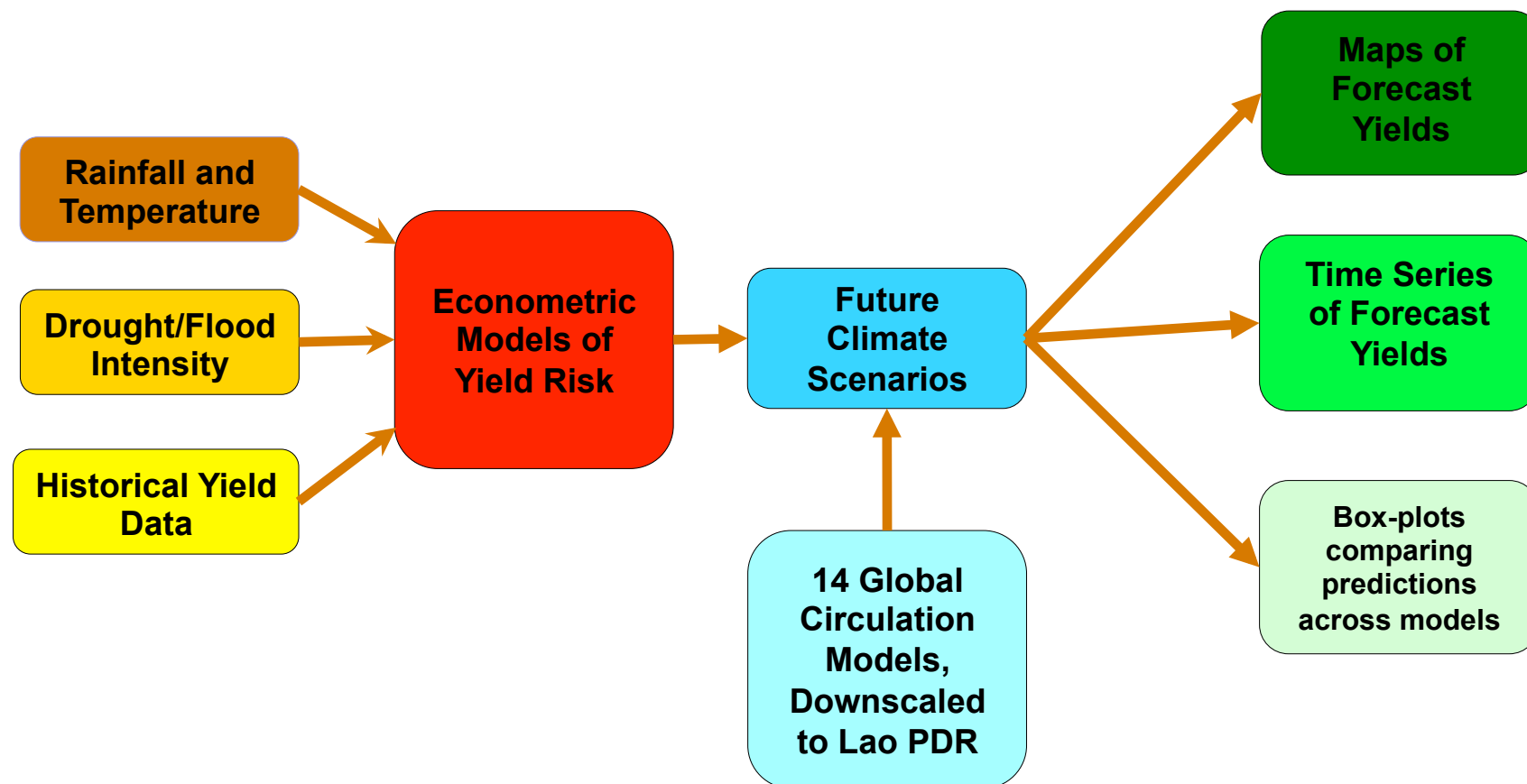
Most Extreme Growing-Season Weather Conditions, 2006-2012



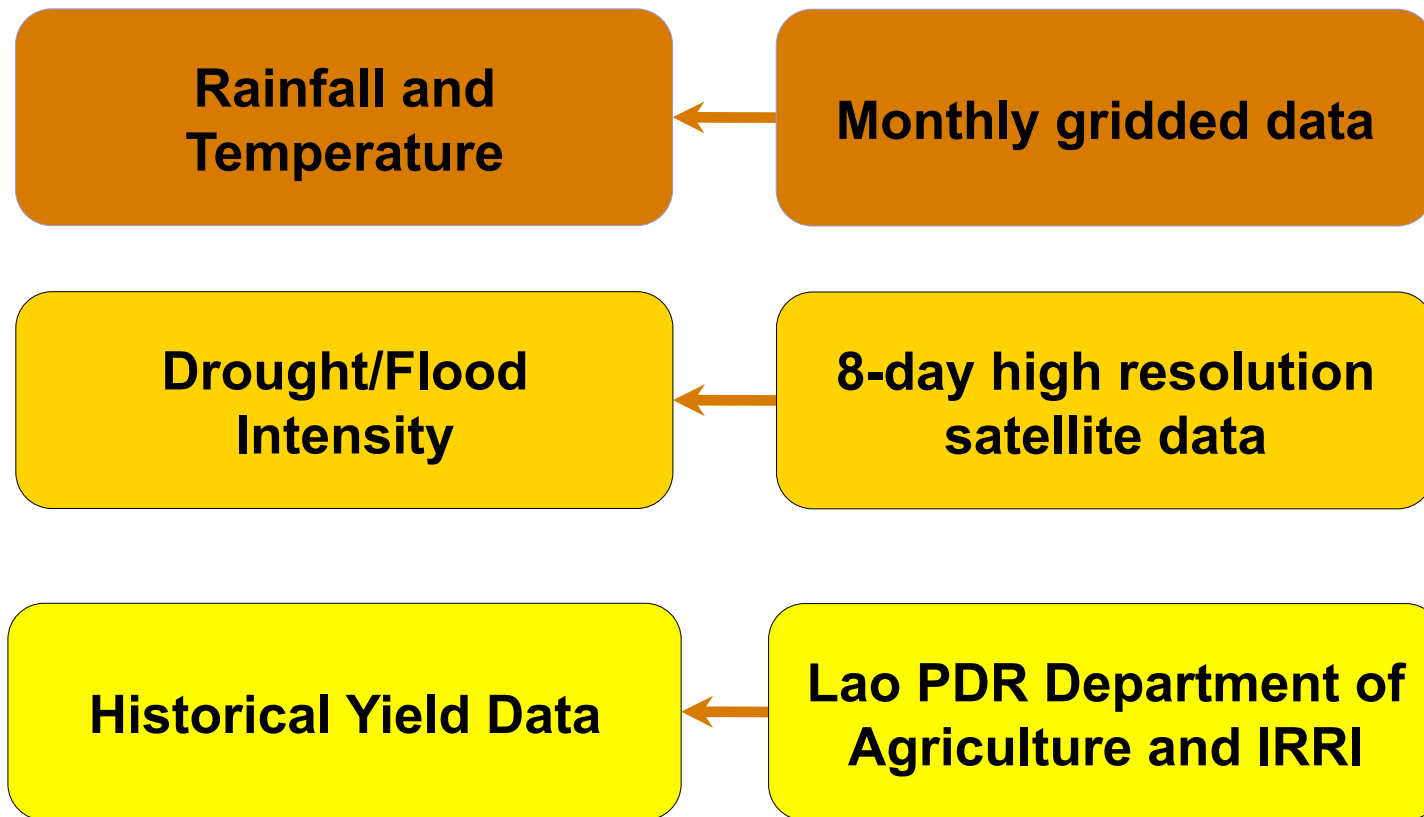
Source: Drought Severity Index (DSI) described in Mu et al (2013)



Modeling Climate Risk to Rice Yields



Data Resources



Panel Models of Average Weather Effects and Shocks

Equation 1: Trend Risk

$$\log(Y_{dt}) = \gamma_d + \theta_t + \beta_1 \text{Min}T_{dt} + \beta_2 \text{Max}T_{dt} + \beta_3 P + \varepsilon_{dt}$$

Y_{dt} is yield for district d in year t .

The model includes province fixed effects γ_d and year fixed effects θ_t .

β_{1-3} represent the coefficients on our weather variables

Equation 2: Shock Risk

$$\text{Log}(Y_{dt}) = \gamma_d + \theta_t + \beta_1 \text{Dr}_{dt} + \beta_2 \mathbf{X}_{dt} + \varepsilon_{dt}$$

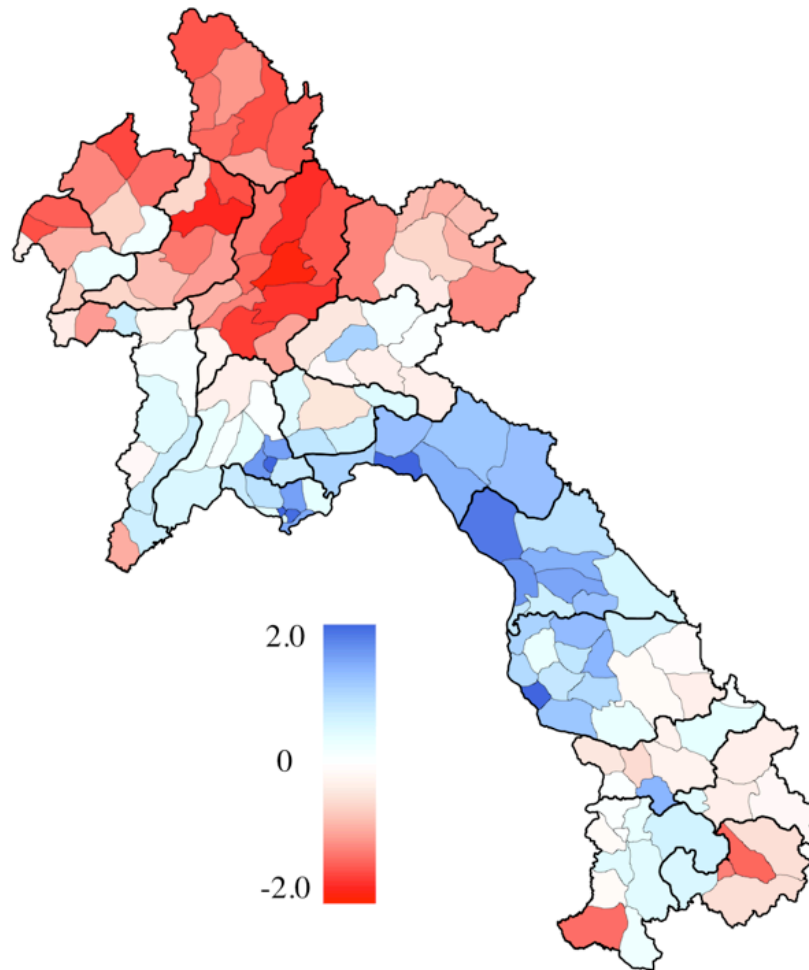
β_1 represents the coefficients on our drought measure.

\mathbf{X}_{dt} are other controls.



Most Extreme Growing-Season Weather Conditions, 2006-2012

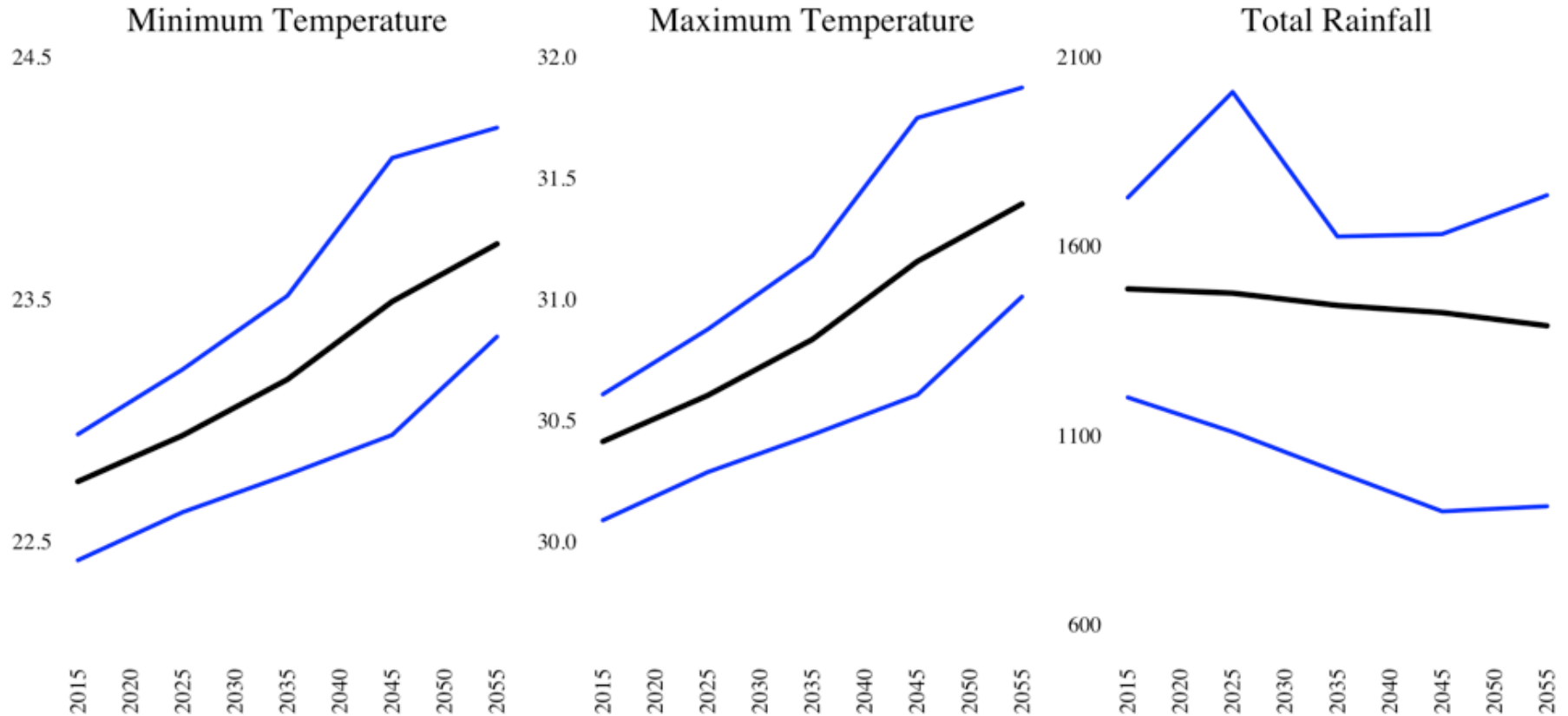
Drought Severity Index (DSI)



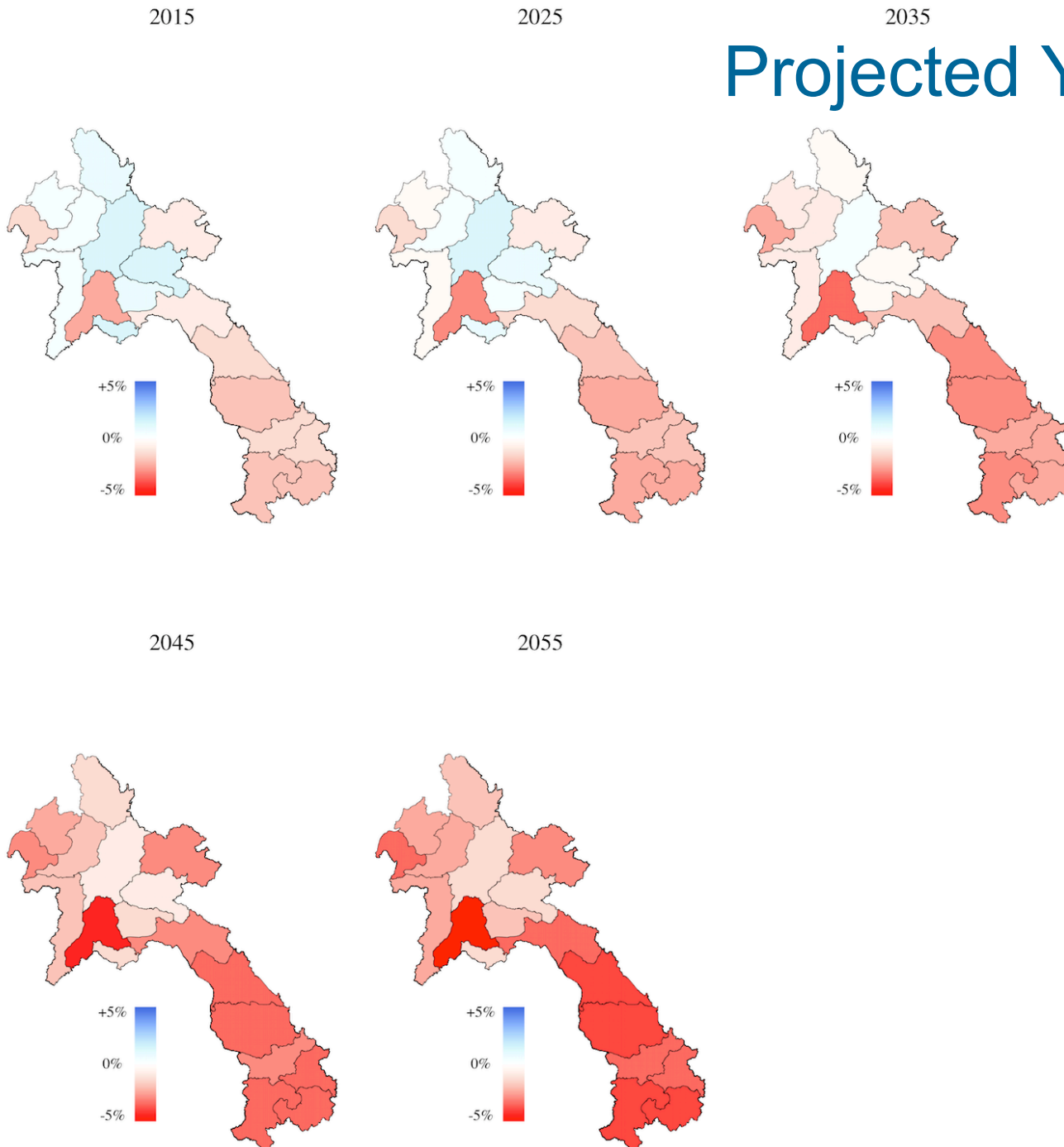
The figure shows average area-weighted DSI values for Lao PDR districts. Blue represents greater than normal and red represents less than normal water levels.



Forecast Lao PDR Climate Conditions: Averaged across 14 Global Circulation Models



Projected Yield Changes



Yields may actually increase in some upland (lower temperature) areas for two decades, then decline with national yields.

Other provinces should be targeted for adaptation support according to emergent yield risk (red).

Estimates are conservative (linear model).

2020

2030

2040

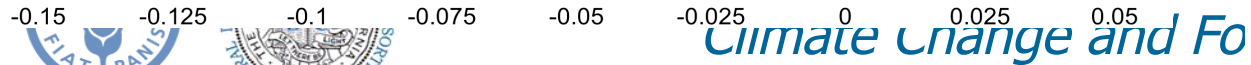
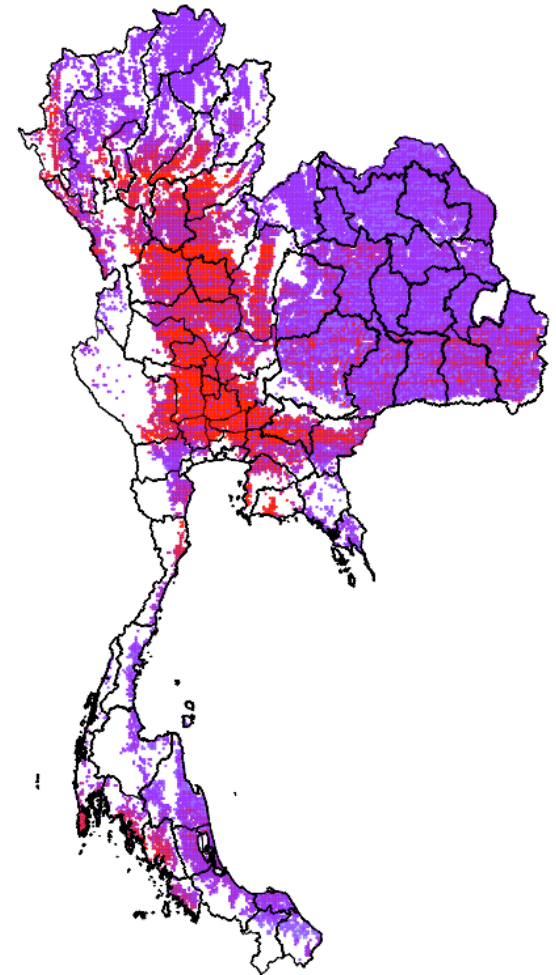
Yield Changes: A1B Scenario

Thailand:
Results of a non-linear model
2070

2050

2060

2070



Preliminary Conclusions

- Climate risk to the Lao PDR rice sector in will vary significantly over space and time.
- This heterogeneity presents an opportunity, for policymakers and private stakeholders to learn adaptation.
- To support this, we propose an early warning mechanism that can identify and target risks as they emerge.
- Two generic risk categories need to be monitored:
 - Direct – risks to domestic food production
 - Indirect – risks to food security transmitted through markets

