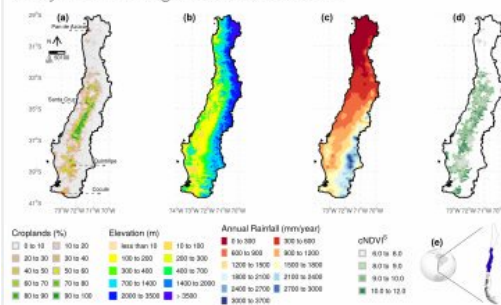


Motivation: Although there is debate whether drought frequency has increased in recent years (Dai, 2012;), climate change is expected to exacerbate the phenomenon and lead to more frequent and intense drought periods, which may even occur in regions where overall precipitation increases are expected (IPCC, 2013). Planning for effective adaptation strategies is thus crucial to mitigate future impacts (Roco et al., 2014). In addition, the ability to anticipate the impact of drought early in the season and take in-season mitigation measures such as more targeted irrigation, or reducing stand density (Bodner et al., 2015) could help to reduce crop losses (Pulwarty and Sivakumar, 2014).

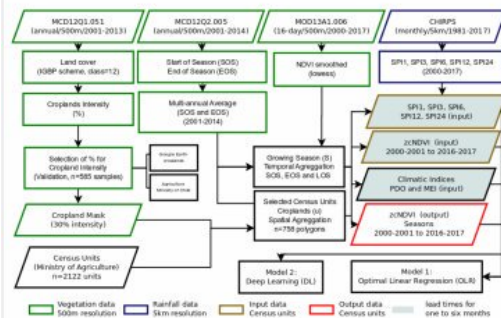
Goal: The main goal of this study is to assess if interannual variability in crop biomass productivity can be accurately predicted using freely-available, near real-time data sources.

Methods

Study area: main agricultural land of Chile



Workflow: for predictions at 1 to 6 months lead times



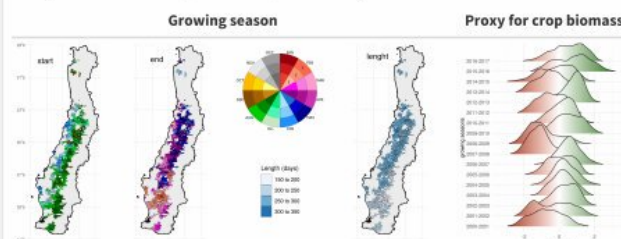
Proxy for crop biomass productivity

$$zcNDVI^S = \frac{cNDVI^S - \overline{cNDVI^S}}{\sigma(cNDVI^S)}$$

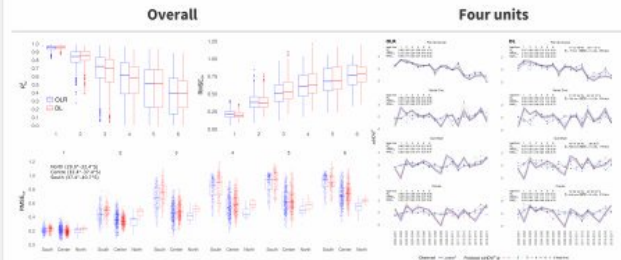
the standardized anomaly of cumulative NDVI over the growing season

Results

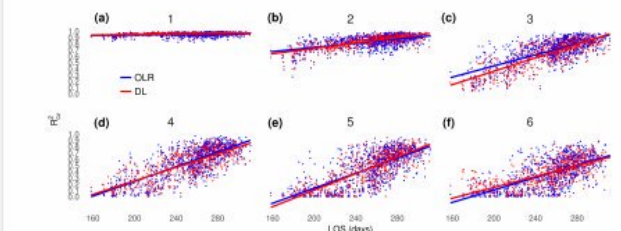
Proxy for seasonal crop biomass productivity



Cross-validated accuracy measures



Accuracies vs. length of growing season for 1 (a) to 6 (f) lead times



Discussion

1. Spatially explicit crop calendar information for Chile would better define this seasonality, allowing for better proxies of agricultural productivity, and more relevant timing of drought prediction
2. Although $zcNDVI$ at lead times was in general the best predictor of seasonal $zcNDVI$, the prediction accuracy strongly depended on the prediction lead time for both OLR and DL with poor predictive power early during the season.
3. Our study shows that complex global models do not always yield better results than simple models that carefully select the best predictors on a case-by-case basis.
4. Our current results are encouraging given that they reach similar or better accuracy compared to existing studies.

Future outlook

1. Additional spatio-temporal predictors such as soil moisture, evapotranspiration, and existing multi-scalar drought indices can be incorporated into future implementation of both OLR and DL approaches.
2. While a few challenges remain, we believe that the models presented in this study can constitute a solid basis for improved operational drought warnings.

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Contact information

francisco.zambrano@umayor.cl
www.researchgate.net/profile/Francisco_Zambrano_Bigiarini
www.linkedin.com/in/frzambra
www.twitter.com/in/frzambra
https://hemera.umayor.cl/

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