

## **Agricultural Livelihoods in a Changing World**

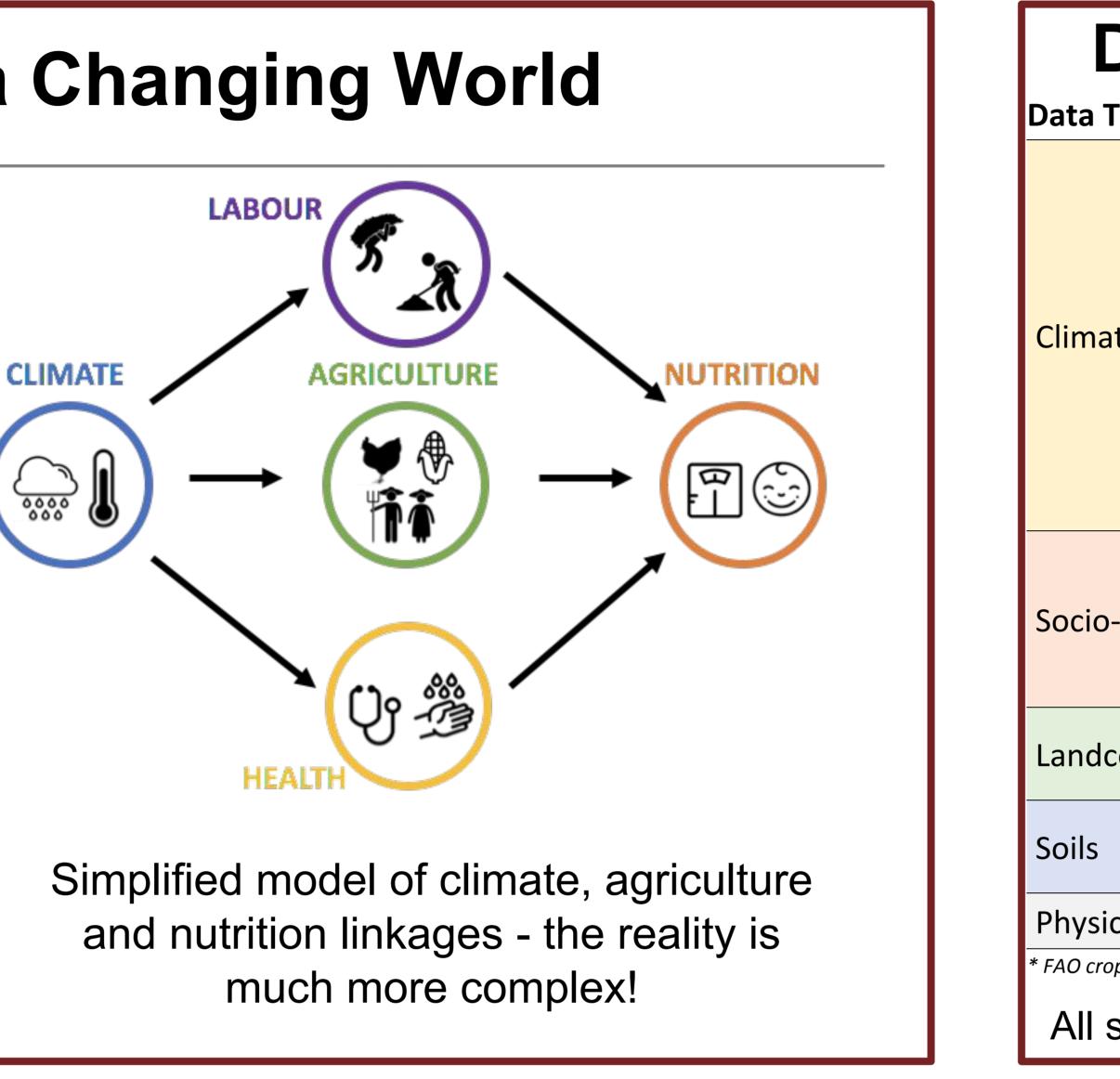
Multiple pathways have been proposed to link agriculture to nutrition outcomes, including direct consumption of farm products, generation of income, labour, and gender-based decision making. However, climate change can directly impact agriculture, nutrition, and indirectly impact the pathways that link them as well. While RHoMIS captures much of the complexity of agricultural livelihoods, it cannot necessarily capture climate realities experienced by farmers. Here we developed tools to link RHoMIS observations with climate, biophysical, and socioeconomic datasets to analyze links between climate, agriculture and nutrition.

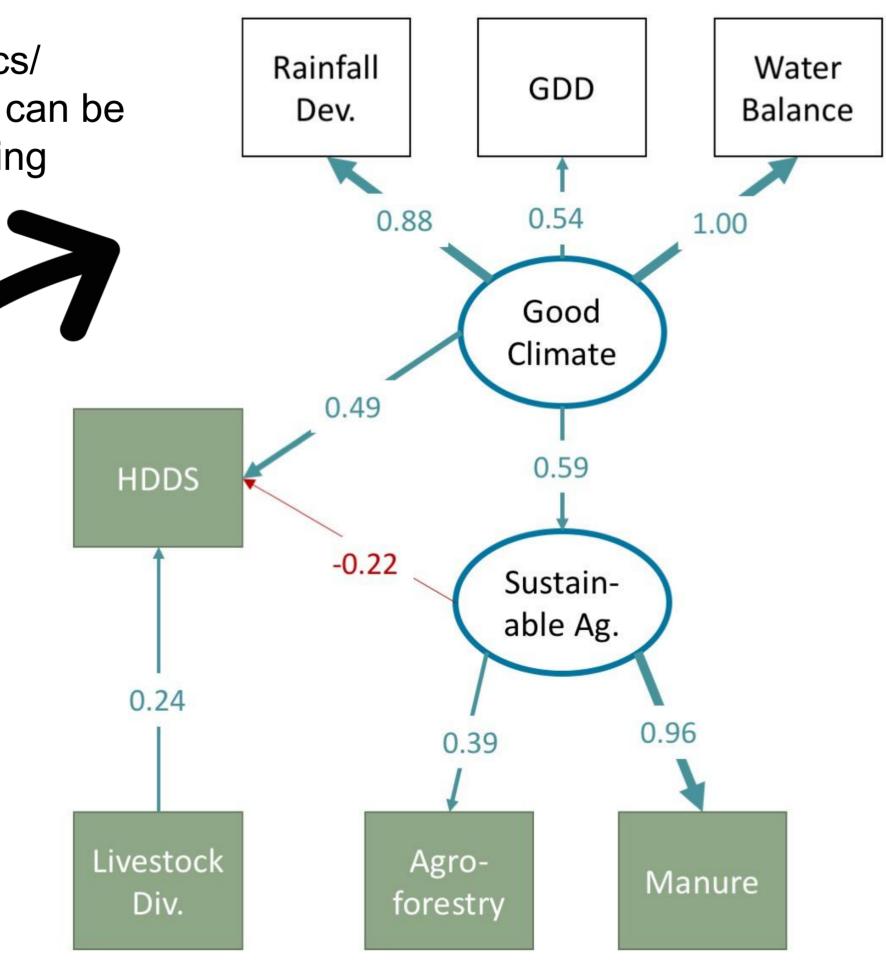
#### **Do Climate and Agriculture Impact Nutrition in RHoMIS Households?** Through **Structural** Bengamis n = 135 Create statistics/ Rainfall Water **Equation Modeling**, we GDD indicators that can be Balance Dev. found evidence that a used in modeling favorable climate had a Legend **RHOMIS SCAN Site** strong positive influence Kenya on dietary diversity, as O Tanzania Good Zambia Climate

**Buffer HH locations** to extract spatial data

# Linking Climate, Agriculture and Nutrition: Leveraging external data to unlock livelihood pathways in RHoMIS

#### Peter Steward, Kristal Jones, Christine Lamanna, Todd Rosenstock, Suneetha Kadiyala





did livestock diversity. However, we found a negative linkage between sustainable agriculutral practices and dietary diversity! At the same time, favorable climate conditions mean households are more likely to practice sustainable agriculture.

Datasets used & statistic added to				
Туре	Dataset	Variables	Resolution	Sta
ate	<u>POWER</u>	_ Daily temp, solar rad., windspeed, humidity, pressure	55.5 km 1983-Present	1) 2) 3) _ Th _ I
	<u>AgMERRA</u>		28-111 km 1980-2010	
	<u>CHIRPS</u>	- Daily precipitation	5.5 km 1981-Present	
	<u>TARCAT</u>		4 km 1983-Present	
-Economic	<u>Harvest</u> <u>Choice</u>	13 variables relating to health, wealth, markets and production	9.25 km 2005-2012	FA cr tir %
cover	ESA CCI-LC	38 landcover classes calculated annually	300 m 1992-2015	%
	<u>SoilGrids</u>	24 soil parameters, 3-6 depths	250 m 2015-16	Te ca
cal	SRTM DEM	Elevation	30 m, 2014	Ele
op calendars used to were used estimate growing seasons for different crops. Statistics were generated for the entire season a				

All spatial data we can currently extract <u>https://goo.gl/SpPgeV</u>; Other spatial datasets <u>https://goo.gl/CEruuB</u>

- We have the tools to extract spatial data for resilience and adaptation!
- "highly context specific outcomes"
- SEM shows strong climate, agriculture and nutrition linkages, but some are households
- data to compare across sites.

innana nnovative Methods and Metrics for Agriculture and Nutrition Actions





### **RHOMIS** observations

tatistics

- Long-term average (LTA)
- Seasonal value\* across multiple years
- Seasonal deviance from LTA
- The above were calculated for variables including: Rainfall (total, rain days, dry spells)
- Temperature, GDD (low, optimum, high, >max)
- PET, water balance
- Planting date

AO farming system, livestock unit density, cropland area, maize & pulse production, ime to market (5 sizes), DHS wealth index, TPHR, 6 stunting and wasting

6 cropland and woodland within 2 km

exture, chemistry, structure, taxonomy, water capacity

Elevation, aspect, slope

and for a 30 day window after planting

RHoMIS and explore climate stress, shocks,

• Free and rich data can help avoid unexplained

counter-intuitive, highlighting the complexity of

## • Next Steps: expanding SEM to include more variables, and potentially adding more external