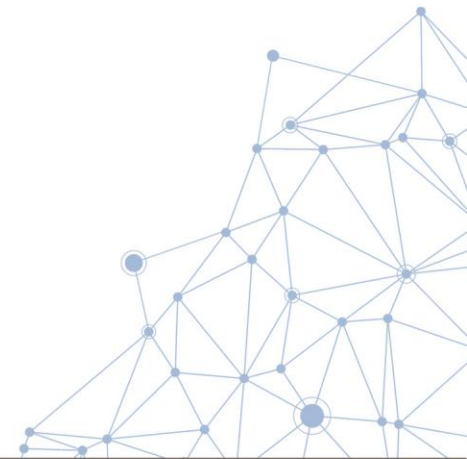


# Assessing drought and food security data for use in trigger design

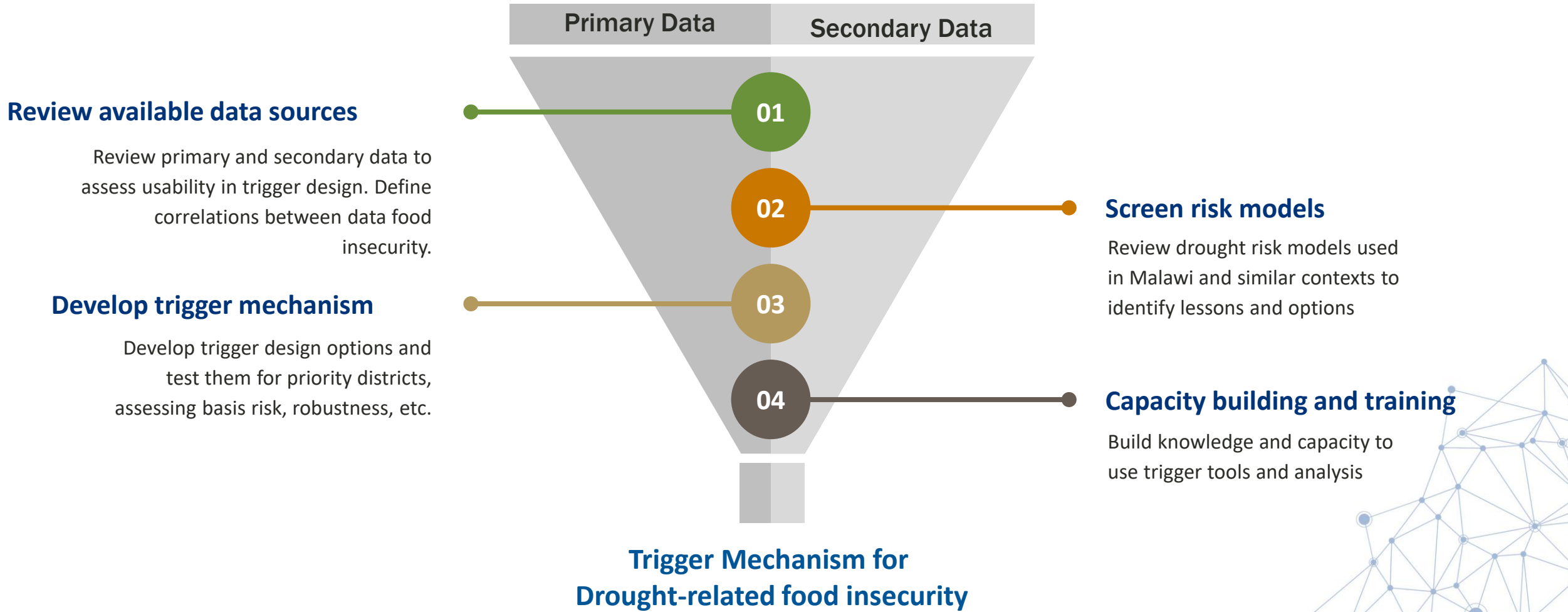
Richard Choularton  
Director of Agriculture and Economic Growth  
[Richard.Choularton@TetraTech.com](mailto:Richard.Choularton@TetraTech.com)

# Objectives

- 1. Review the use of data in the design of trigger mechanisms for shock responsive social protection**
- 2. Review available data for Malawi**
- 3. Present preliminary analysis on what the data tells us in Malawi**



# Overview of Trigger Mechanisms Design Process



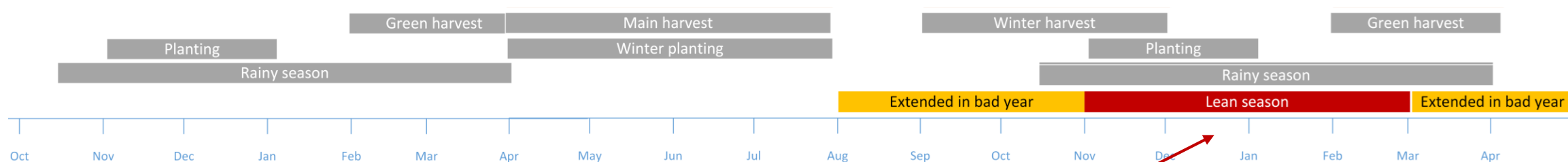
# Assessing data for use in trigger design

## Part 1: Using data to define trigger indicators

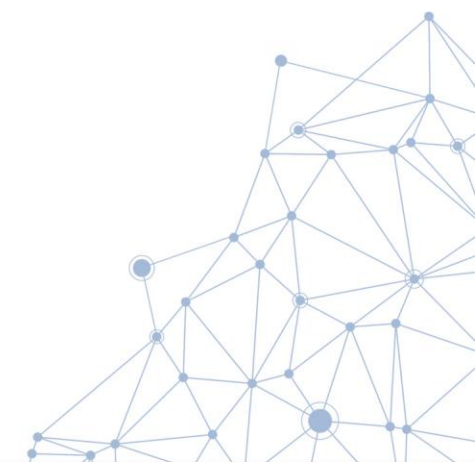


# Linking data to trigger design

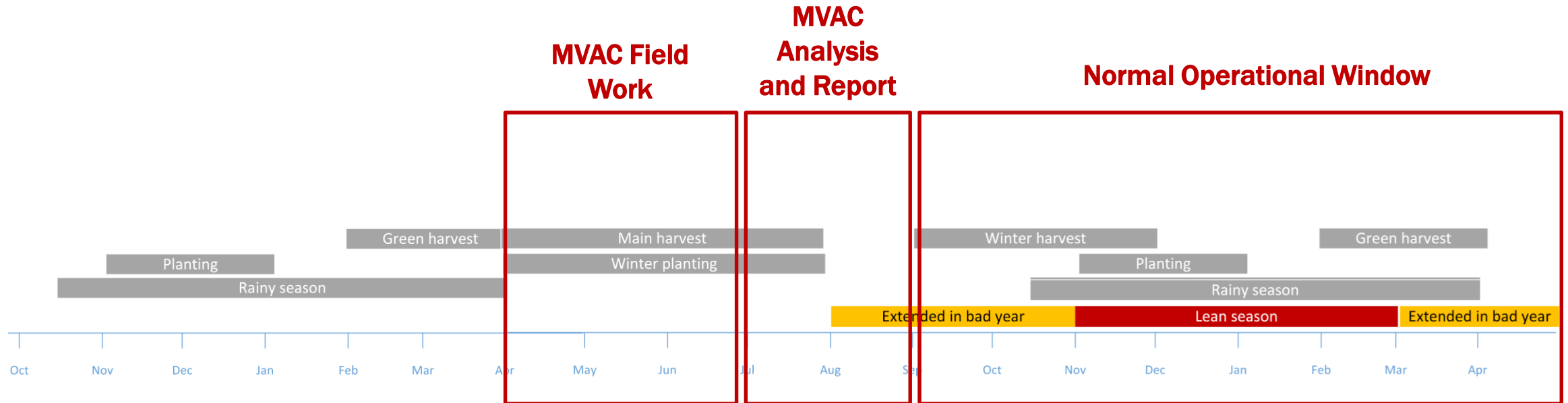
**Key principle: Understanding the links between livelihoods, food security, and shocks is the foundation of selecting indicators and designing triggers**



**Food insecurity peaks seasonally during the lean season**

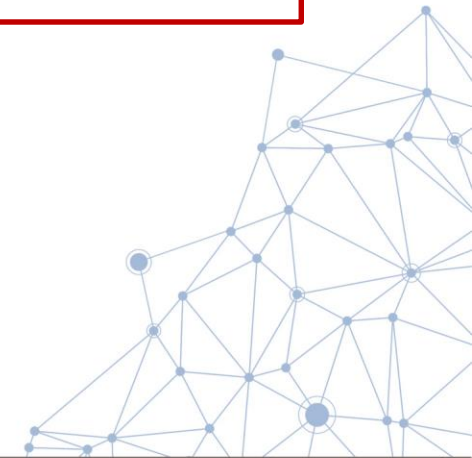


# Current MVAC Assessment Process

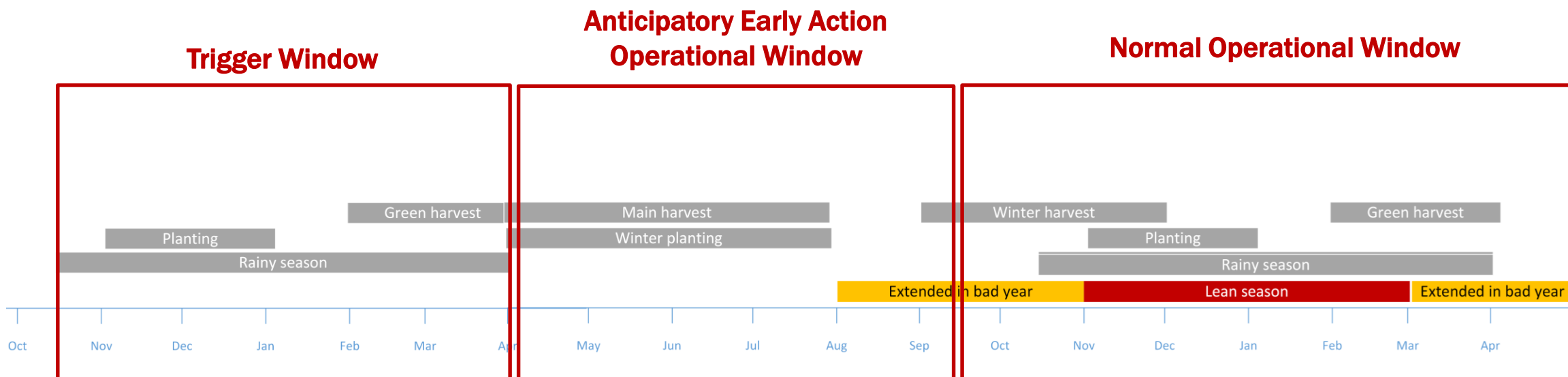


## Key Points?

- The MVAC provides a comprehensive national assessment each year
- Can other indicators complement the MVAC results to trigger earlier anticipatory action?

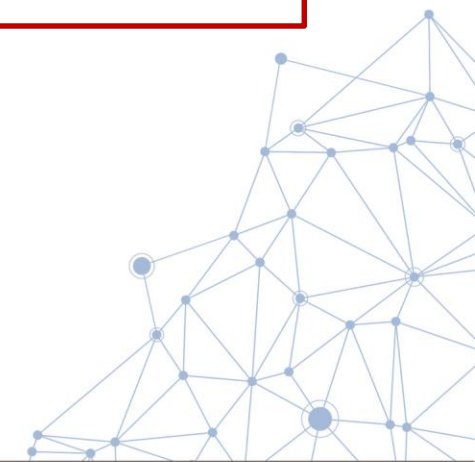


# Linking data to trigger design – Trigger and Operational Windows



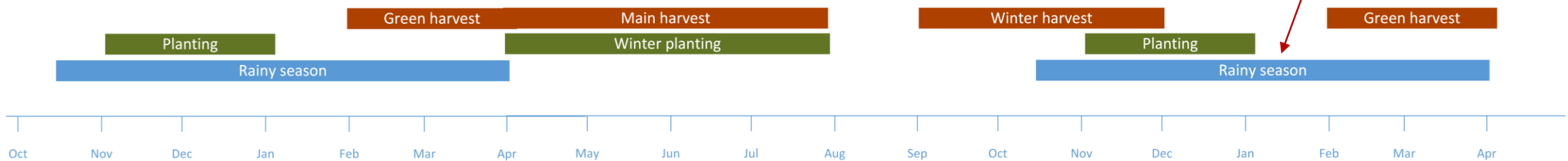
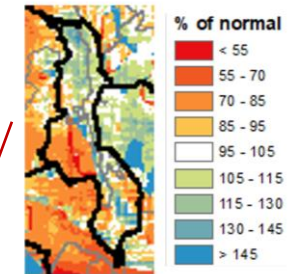
## Key Questions

- What indicators can we reliably collect during the trigger window?
- Do they have a high correlation to negative food security outcomes during the lean season?



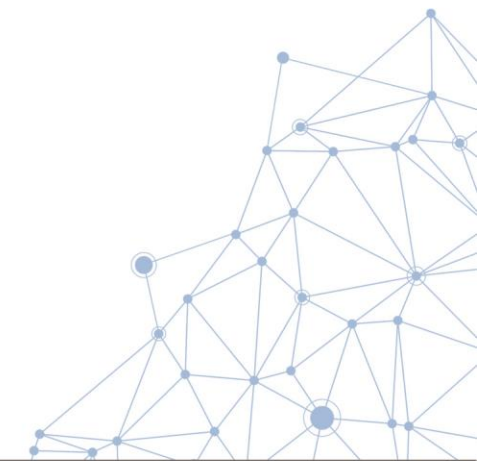
# Linking data to trigger design: Key data types

Seasonal Rainfall Accumulation Percent of Normal by pentad  
 2020-2021 season Oct - May  
 (Oct pentad 1 2020 thru Oct pentad 1 2020) / Avg (1981-2010) \* 100



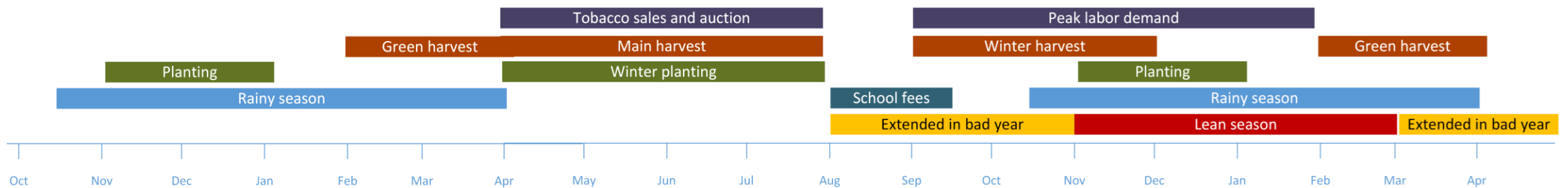
## Key Data: Agricultural Season Monitoring

- Remote sensing
- Weather data
- Crop cuts
- Food prices
- Input availability
- Pest and disease



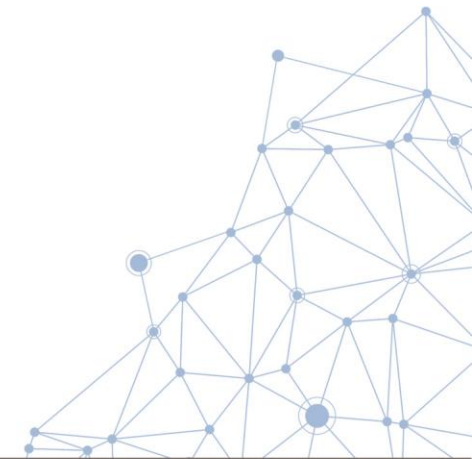


# Linking data to trigger design: Key data types



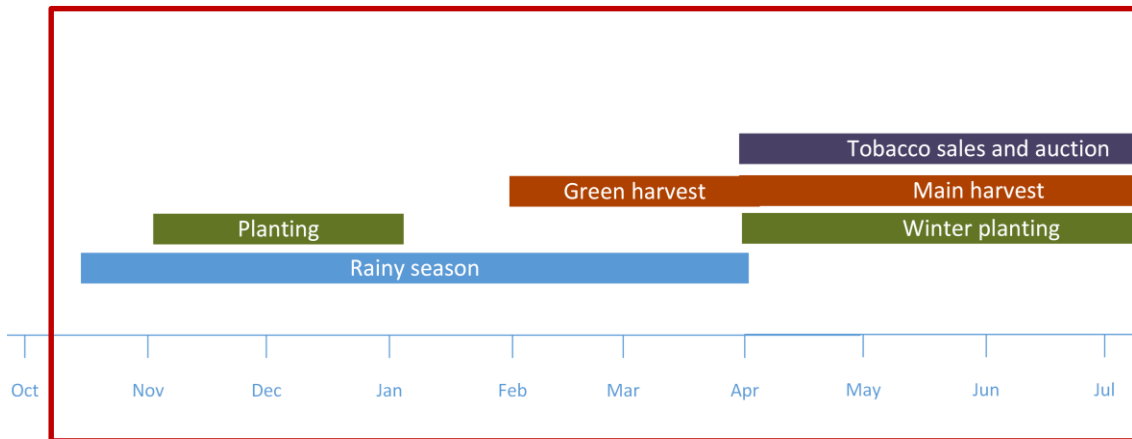
## Key Data: Food Security Monitoring

- Food prices
- Food consumption
- Malnutrition
- Coping Strategies
- Labor opportunities

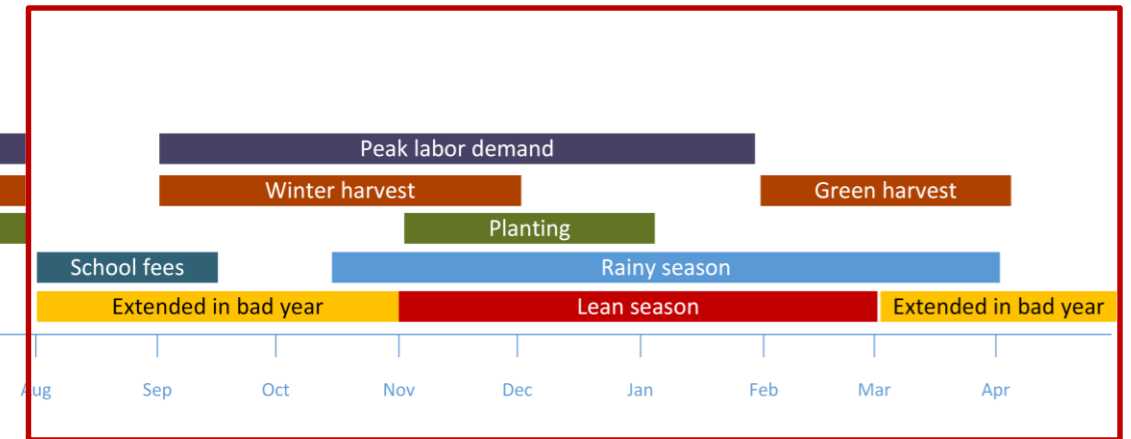


# Linking data to trigger design: Leading and Trailing Indicators

## Leading Indicators



## Trailing Indicators



**Early indicators of shocks, stressors, and early signs of food insecurity stress**

**Food security outcome indicators that indicate a food crisis is occurring, e.g. increased wasting**



# Assessing data for use in trigger design

## Part 2: Review of drought and food security indicators for Malawi



**Quick question: What criteria would you use to evaluate data for triggering response?**




# Trigger Data Review Criteria

## First Level Criteria

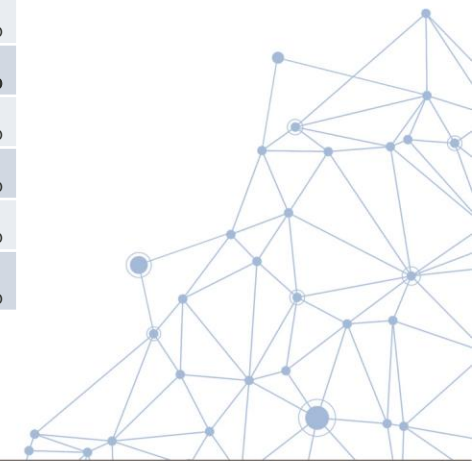
- Historical availability
- Temporal resolution
- Data latency
- Cost
- Spatial resolution and coverage
- Experience in index insurance
- Continuity
- Rigor

## Second Level Criteria

- Statistical analysis
  - Ease of understanding
  - Transparency
  - Ease of access
  - Consistency
  - Correlation to drought, agricultural loss, food insecurity, farmer feedback
  - Exposure to microclimate conditions
- 

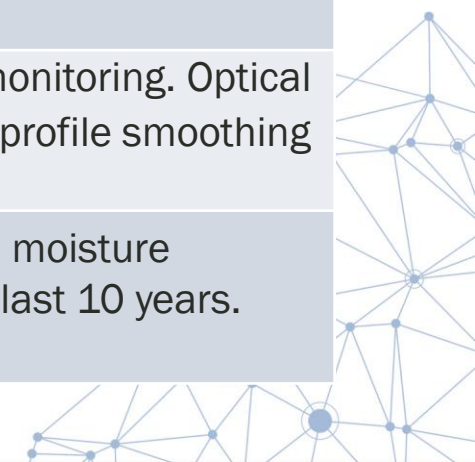
# Primary Trigger Remote Sensed Data Sources & Ratings

No.	Product	Indicator	Rating %
1	TAMSAT	Rainfall	100%
2	CHIRPS	Rainfall	100%
3	ARC2	Rainfall	95%
4	Landsat NDVI	Vegetation Index	89%
5	MOD13 NDVI	Vegetation Index	88%
6	MOD16 ET	Evapotranspiration	87%
7	RFE	Rainfall	84%
8	ECOSTRESS	Evapotranspiration	76%
9	SPOT-VGT	Vegetation Index	76%
10	SMAP	Soil Moisture	72%
11	SMOS	Soil Moisture	72%
12	TRMM/GPM	Rainfall	69%
13	AVHRR NDVI	Vegetation Index	64%
14	OCO-2 SIF	Chlorophyll Fluorescence	52%
15	GRACE-FO	Ground-water	33%



# Recommended primary trigger data sources

Data Source	Indicator	Overview
<b>TAMSAT</b>	Rainfall Index	TAMSAT produces daily rainfall estimates for all of Africa at 4km resolution. The TAMSAT archive spans 1983 to the delayed present. The longevity of the dataset makes it especially suitable for risk assessment.
<b>CHIRPS</b>	Rainfall Index	Climate Hazards Group InfraRed Precipitation with Station data (CHIRPS) is a 35+ year global rainfall data set. CHIRPS incorporates multiple inputs to create a gridded rainfall time series for trend analysis and seasonal drought monitoring.
<b>ARC2</b>	Rainfall Index	Developed by NOAA, ARC2 combines satellite and quality-controlled ground weather station data to create an improved rainfall data set. ARC2 data slightly underestimate rainfall levels
<b>MOD13 NDVI</b>	Vegetation Index	Maintained by NASA, NDVI measures vegetation and is commonly used for crop monitoring. Optical satellite images can be hampered by cloud cover, but though techniques such as profile smoothing this can be addressed.
<b>SMOS</b>	Soil Moisture Index	Maintained by ESA. Previous studies have shown good correlation with SMOS soil moisture drought intensity, especially in arid and semi-arid areas. Data is available for the last 10 years.



# Potential Secondary Trigger Data Sources

- **FEWS NET Food Security Outlook (Current and Projection)\***
- **Food prices\***
- MVAC IPC Assessment (Medium Term Projection)
- CRW Global IPC Trigger
- Household Economy Analysis (Scenario and post-harvest)
- Crop production
- Wasting (GAM, SAM, MUAC)
- Labor rates/Availability
- SAFEX Futures Prices
- Food Consumption Score (FCS)
- Coping Strategies Index (CSI and rCSI)
- Household Hunger Scale
- Household Dietary Diversity Score (HHDD)
- Mortality (CDR, U5DR)
- Pest and disease surveillance





# Assessing drought and food security data for use in trigger design

## Part 3: Preliminary Analysis Overview



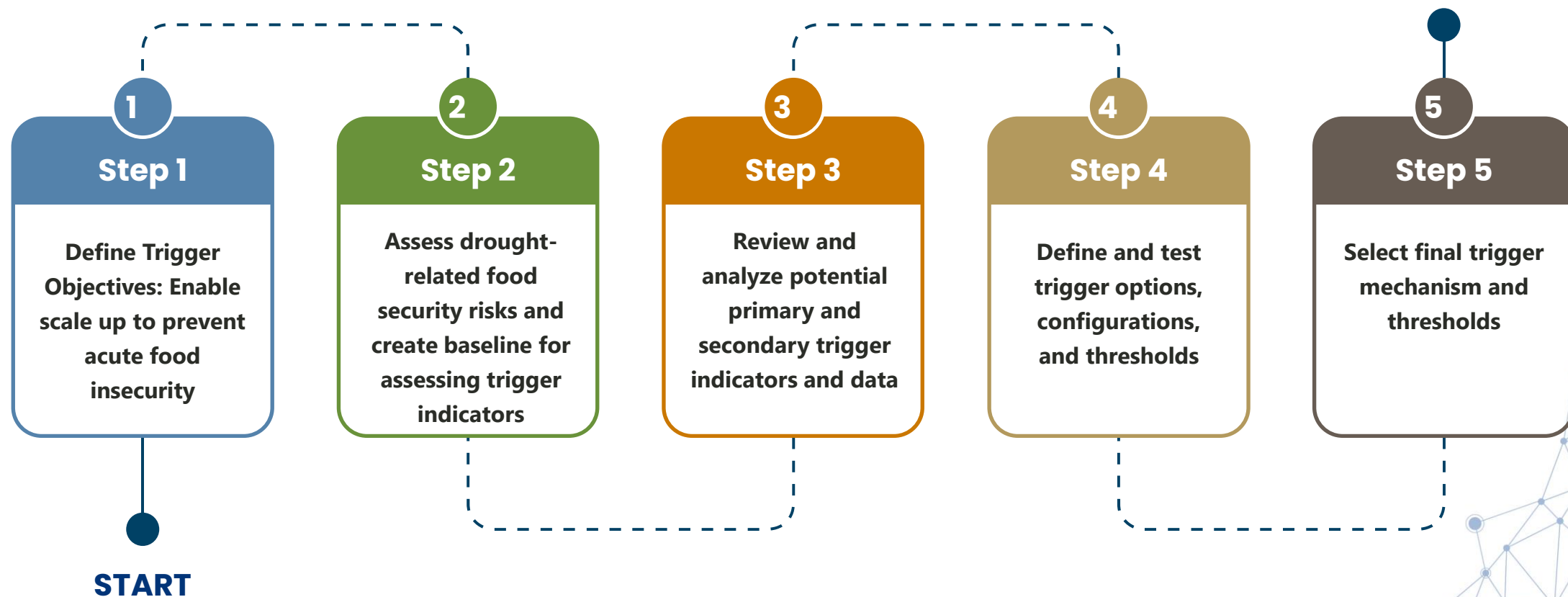
## **Interactive Poll:**

**What districts are most at risk from drought and food insecurity?**



# Trigger design process

## Operational Scale up Mechanism



Assess food security risks and create baseline for assessing trigger indicators

## Key Points

- Most districts have faced a food crisis in the last 10 years.
- Food crises are more common in the south.

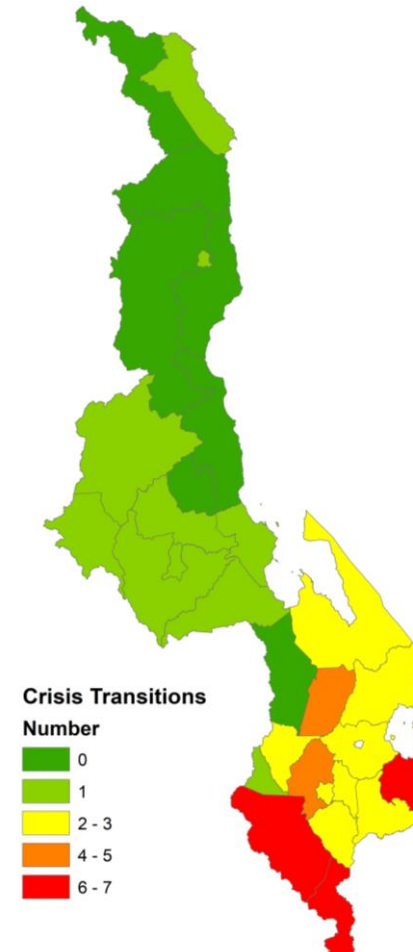
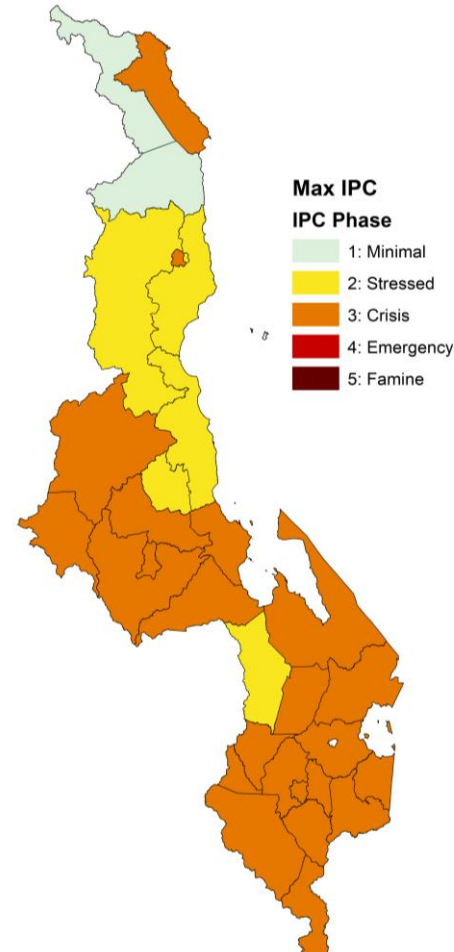
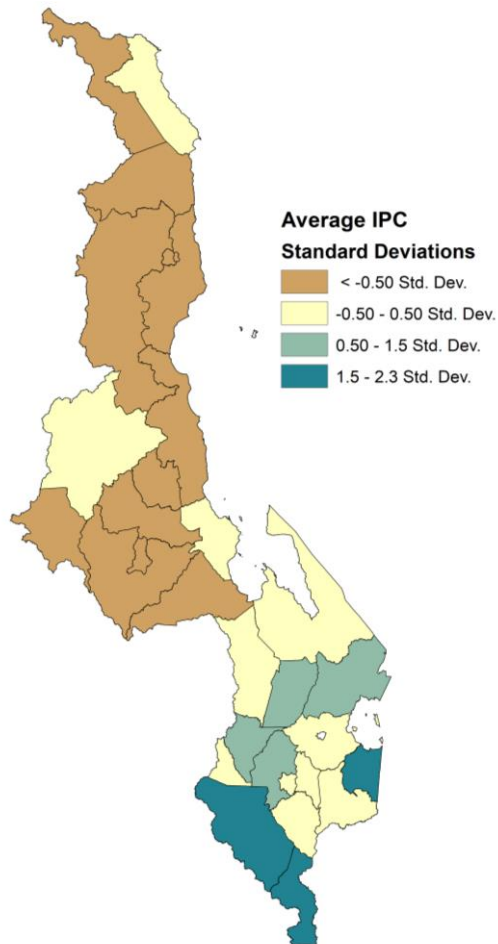
*Note: This analysis used FEWS NET Food Security Outlook data's CS layer. CS is the best available evaluation of current food security conditions based on evidence available and interpreted using IPC compatible analysis.*

# Measuring food insecurity and food crises

FEWS NET Average IPC

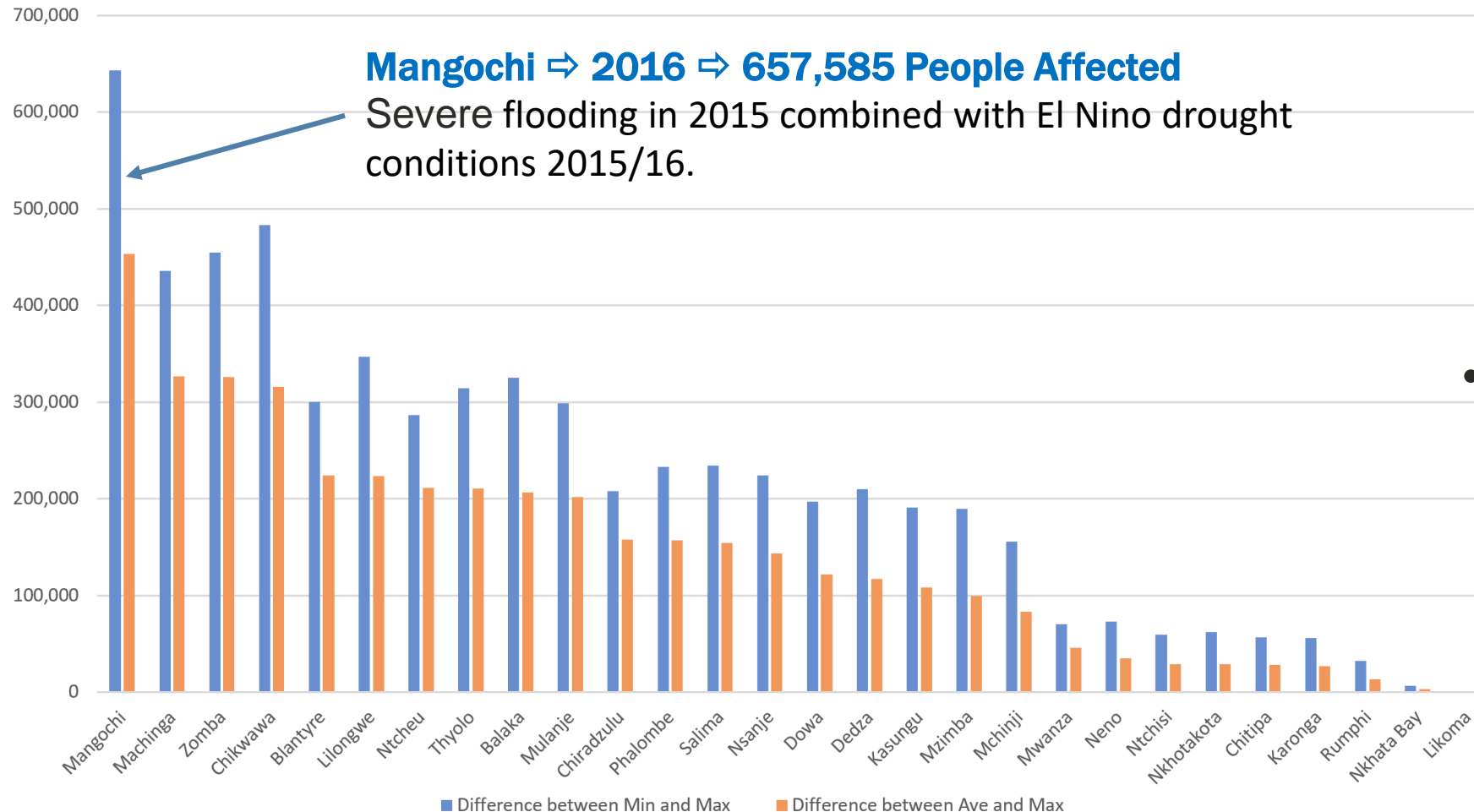
FEWS NET Maximum IPC

No. of Crisis Transitions



Assess food security risks and create baseline for assessing trigger indicators

# MVAC Population Affected – Difference between Minimum, Maximum and Average by District (2011 – 2018)



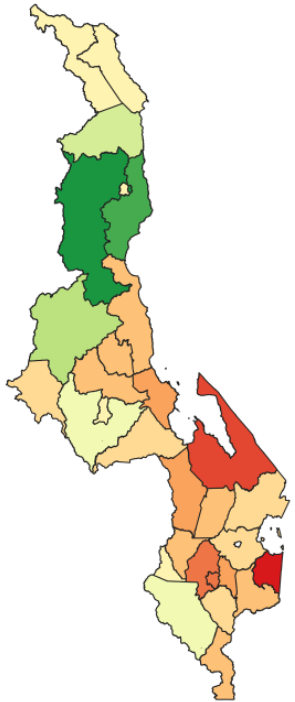
## Key Point

- Many districts have faced significant scale up events over the last 10 years

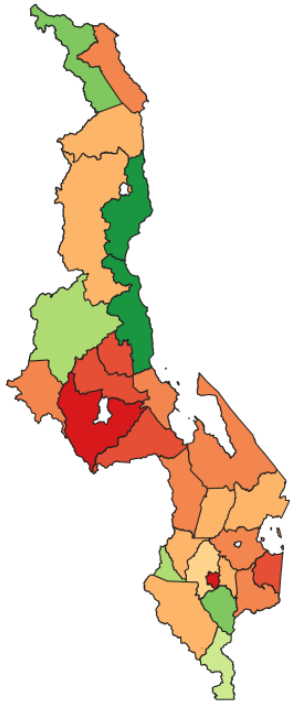
Review and analyze potential primary and secondary trigger indicators and data

# Drought frequency

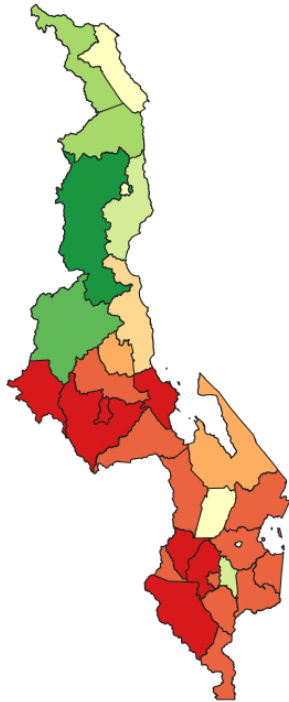
SMOS – soil moisture



MODIS - NDVI



MODIS - ET



Frequency of metrics below lowest quartile



These maps show the frequency of each district having a value in the lowest quartile of each data set over the last 10 years.

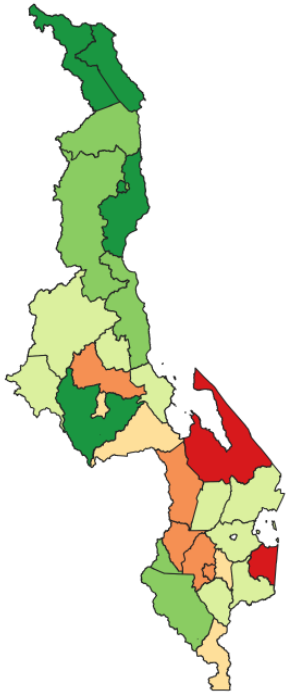
## Key Points

- Different data sources show different aspects of drought
- South and Center more frequently experience drought

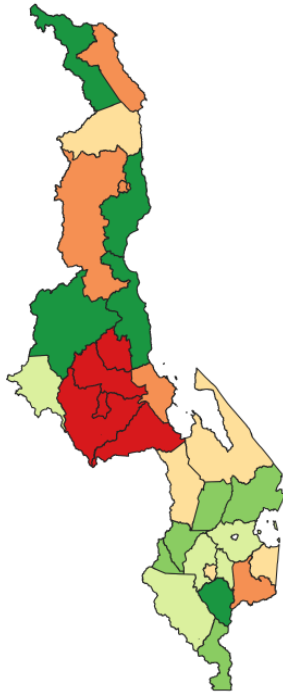
Review and analyze potential primary and secondary trigger indicators and data

# Drought anomalies

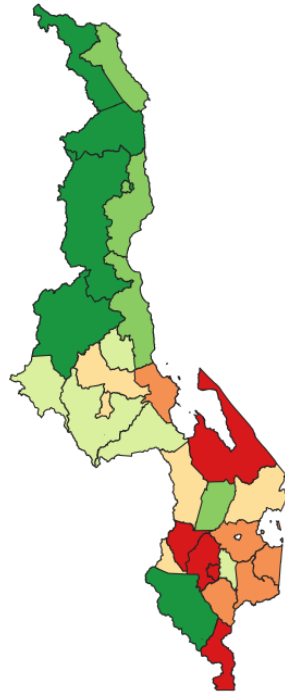
SMOS – soil moisture



MODIS - NDVI



MODIS - ET



Drought as measured by long-term anomaly



## Key Points

- Anomalies are often more related to shocks.
- Assessing multiple options helps identify approaches which are fit for purpose.

These maps show how often a district experienced a value below the long-term average for each indicator in the last 10 years.

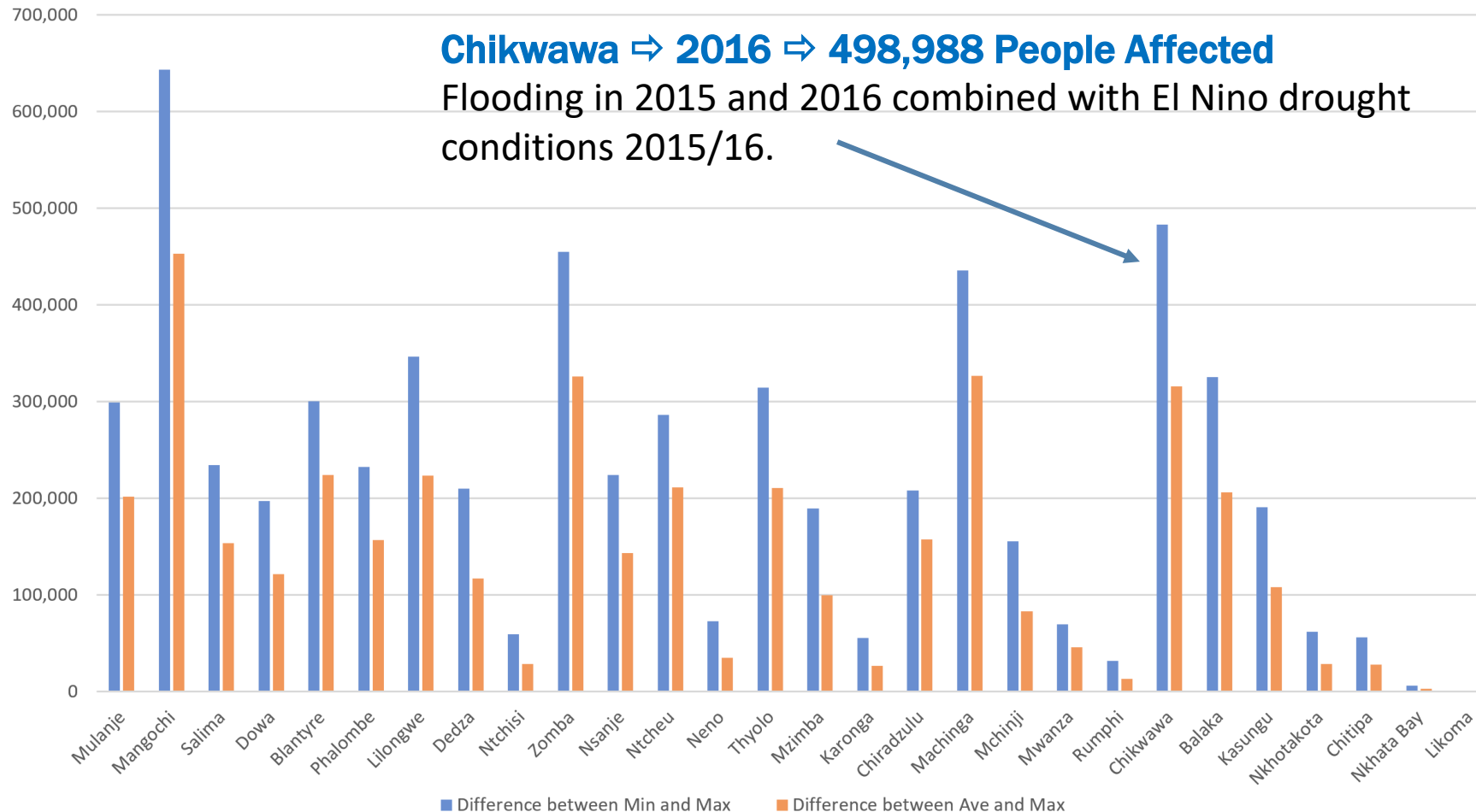
Review and analyze potential primary and secondary trigger indicators and data

### Key Point

- Major droughts can have major impacts in areas that are less drought prone

Note: Districts have been reordered so that starting from left to right the districts are shown with the highest frequency of drought measured by an index of two drought measures.

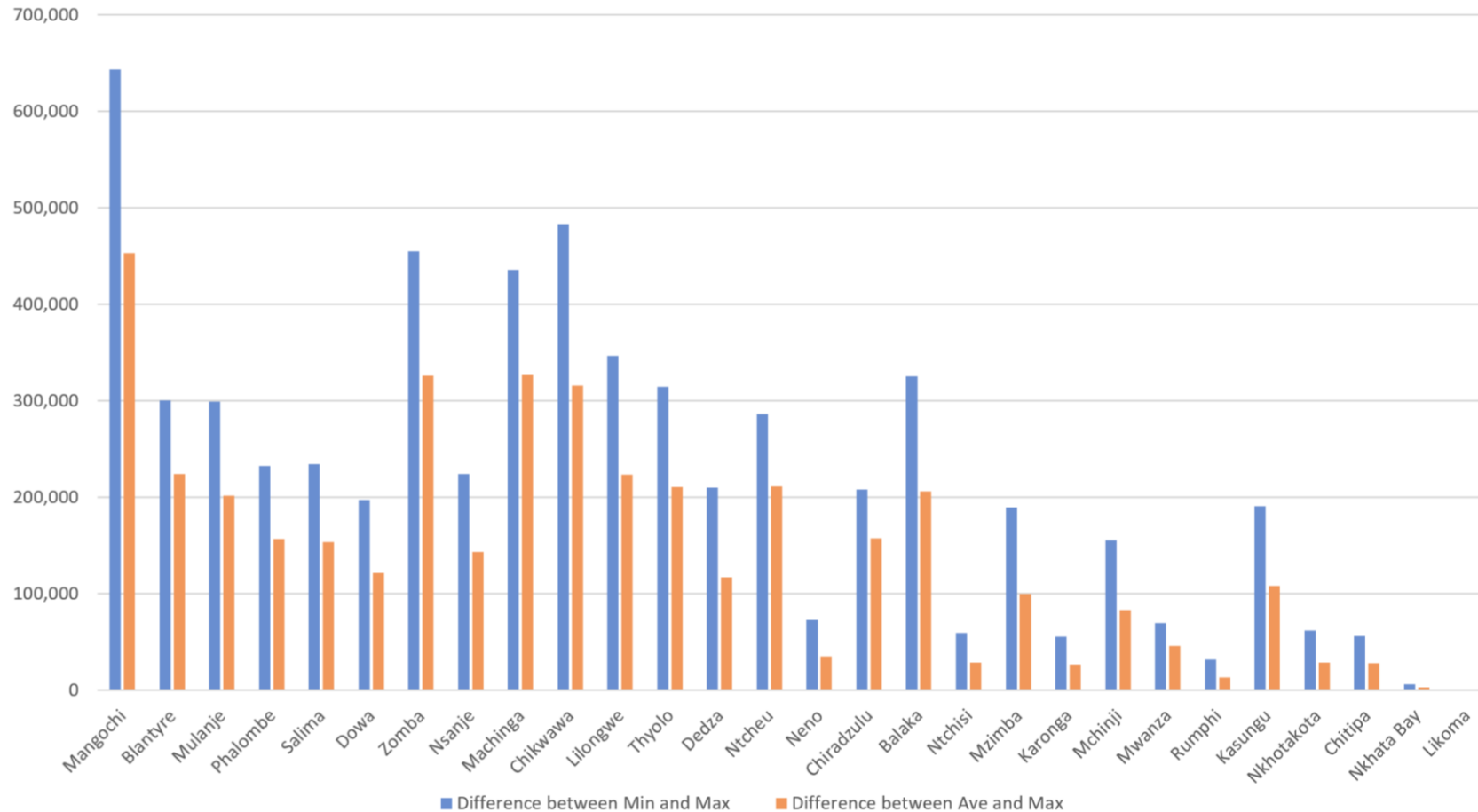
# MVAC Population Affected – Difference between Minimum, Maximum and Average by District (2011 – 2018) - Reordered by drought anomalies





Review and analyze potential primary and secondary trigger indicators and data

# MVAC Population Affected – Difference between Minimum, Maximum and Average by District (2011 – 2018) - Reordered by food security and drought risk



- Top 10 Districts Combined Risk
- Mangochi
  - Blantyre
  - Phalombe
  - Mulanje
  - Zomba
  - Dowa
  - Nsanje
  - Salima
  - Chikwawa
  - Machinga

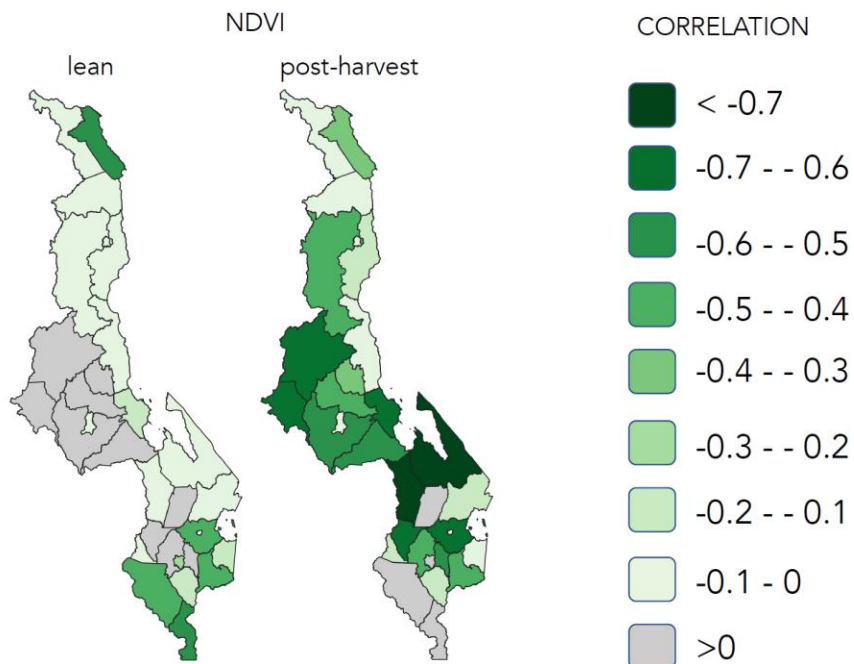
Note: Districts have been reordered based on a ranking index of food security risk (MVAC data and FEWS NET data) and two drought risk measures.

Review and analyze potential primary and secondary trigger indicators and data

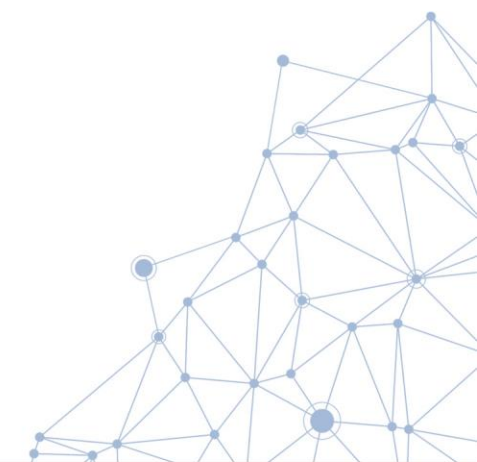
# Remote Sensing Data Correlations to Food Insecurity

## Key Point

- High correlation between remote sensing indicators and food security data is a solid starting point for a trigger design.



This maps show the correlations between a remote sensing indicator of drought and food security severity, comparing food security during the post harvest and lean season periods with the previous 3-month average values.



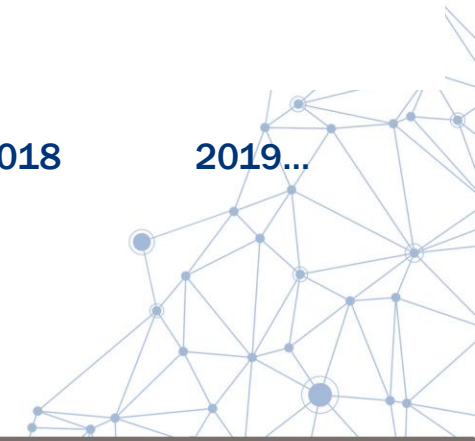
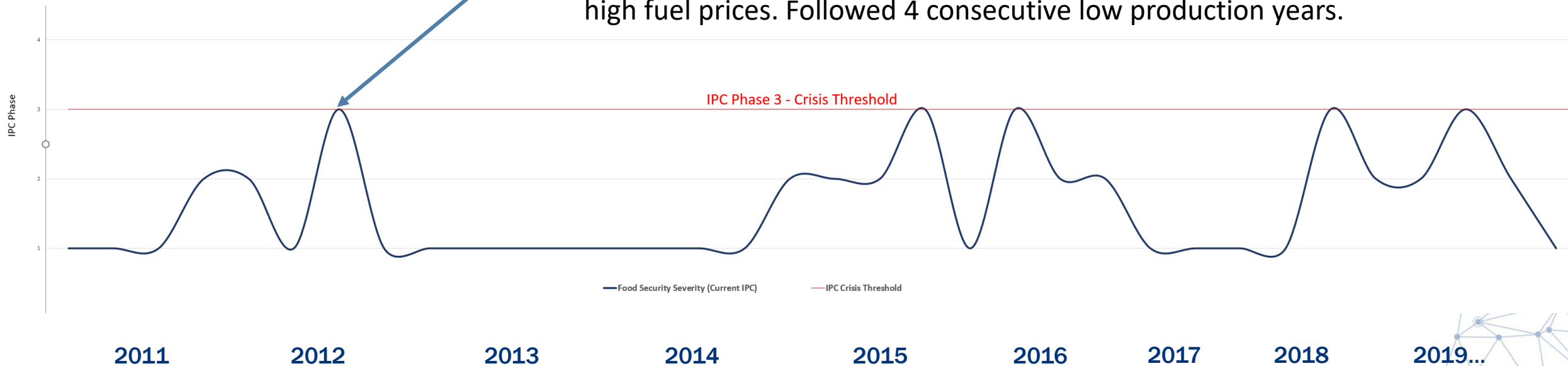
# Assessing drought and food security data for use in trigger design

## Part 3: Preliminary Analysis Overview – Balaka Case Study

# Food Security Trends - Assessing whether triggers hit the mark -Balaka

**2012 ⇒ 208,501 People Affected**

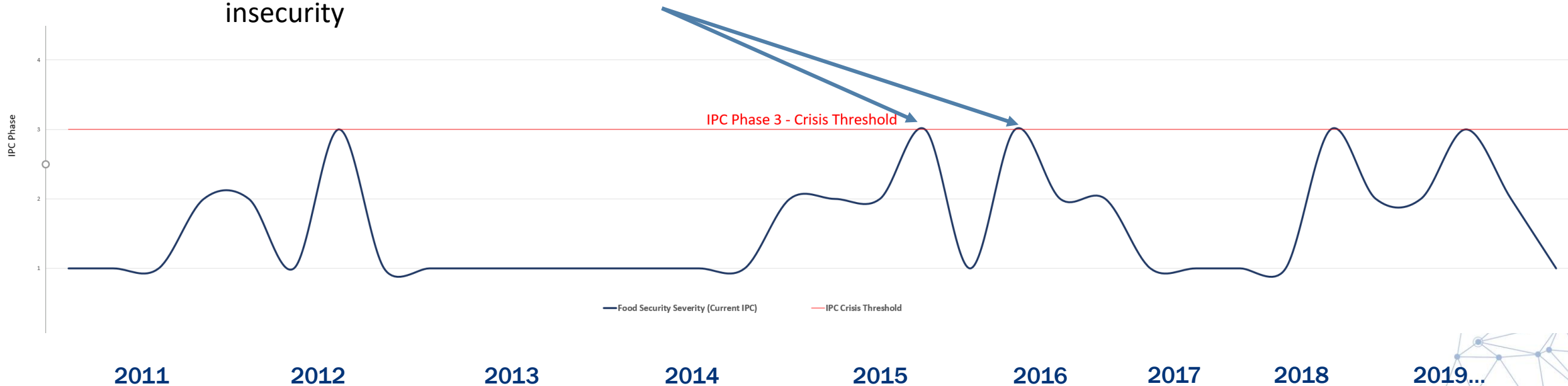
Below average maize production. Cyclone Funso caused erratic and heavy rains. Depressed *ganyu* availability. Maize prices higher than average due to high fuel prices. Followed 4 consecutive low production years.



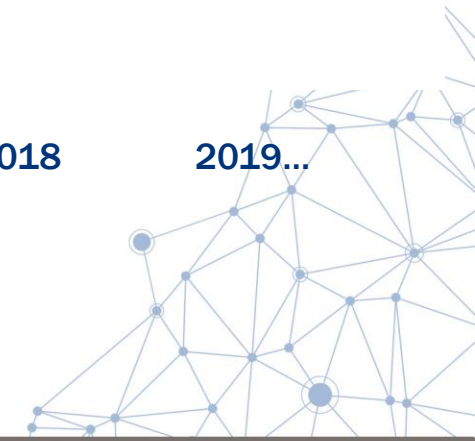
# Food Security Trends - Assessing whether triggers hit the mark -Balaka

**2015/2016 ⇒ 333,943 People Affected**

El Nino drought conditions led to stressed conditions and acute food insecurity



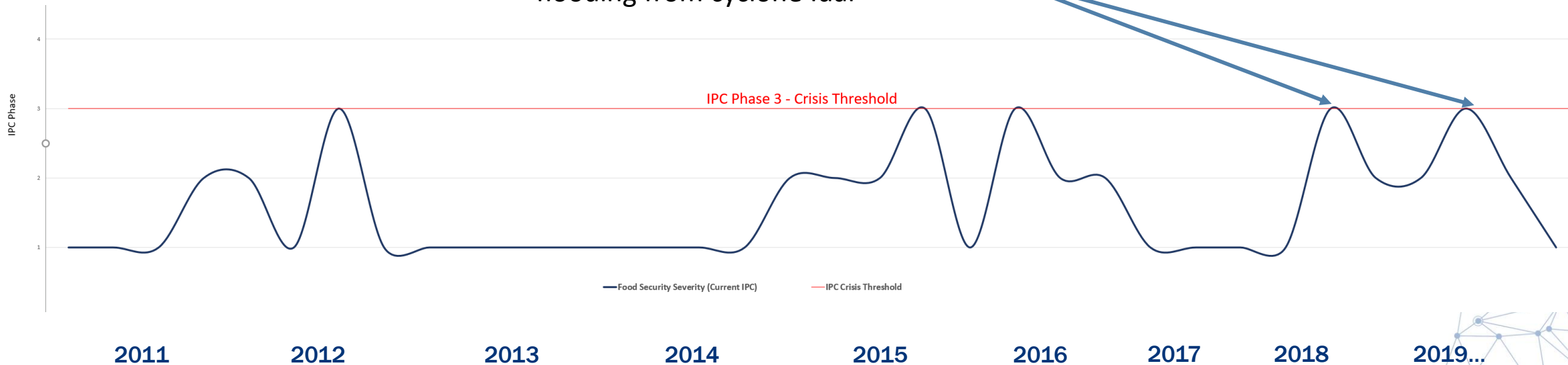
- **2015/16** - drought conditions led to stressed conditions and acute food insecurity.
- **2018/19** - prolonged dry spells and pest infestations followed by heavy rainfall and flooding from cyclone Idai



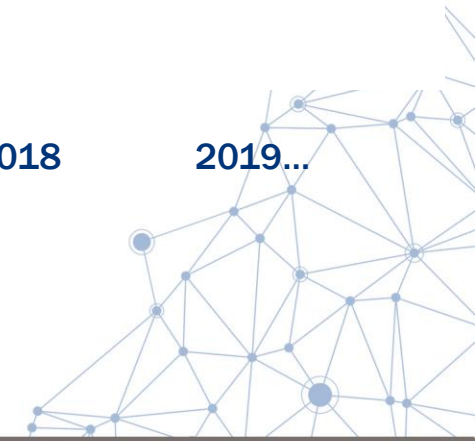
# Food Security Trends - Assessing whether triggers hit the mark -Balaka

**2018 ⇒ 166,036 People Affected**

Prolonged dry spells and pest infestations followed by heavy rainfall and flooding from cyclone Idai

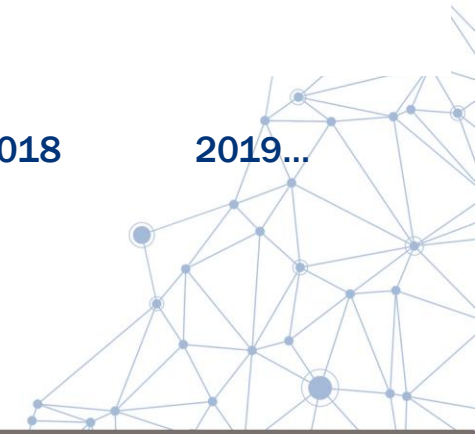
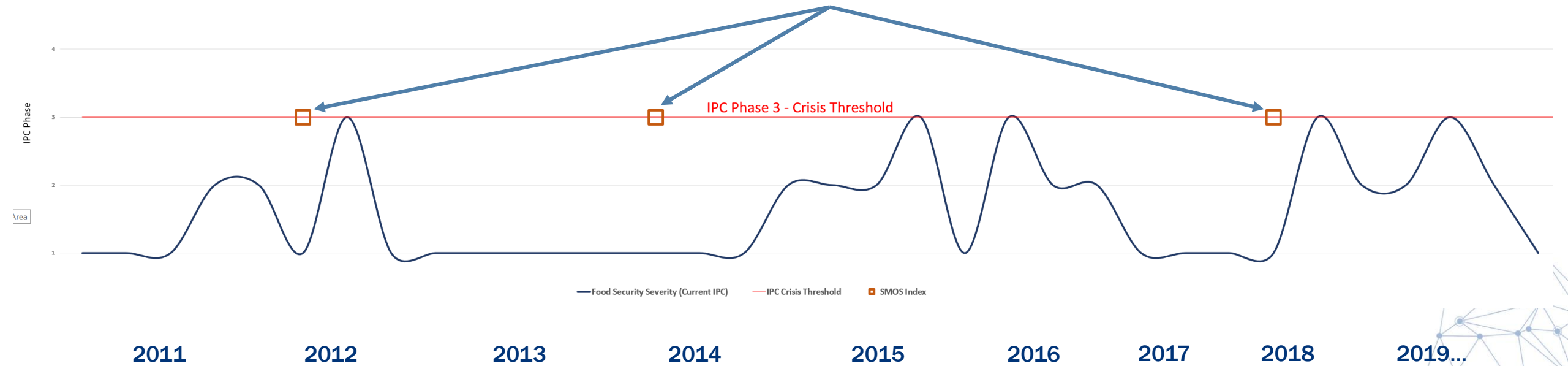


- **2015/16** - drought conditions led to stressed conditions and acute food insecurity.
- **2018/19** - prolonged dry spells and pest infestations followed by heavy rainfall and flooding from cyclone Idai



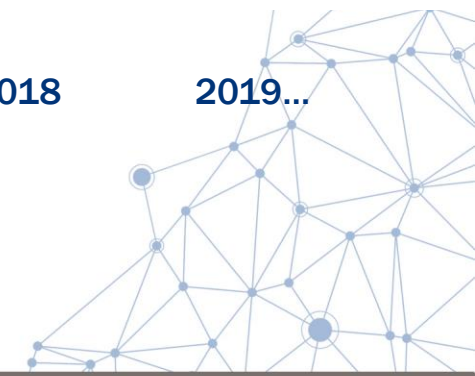
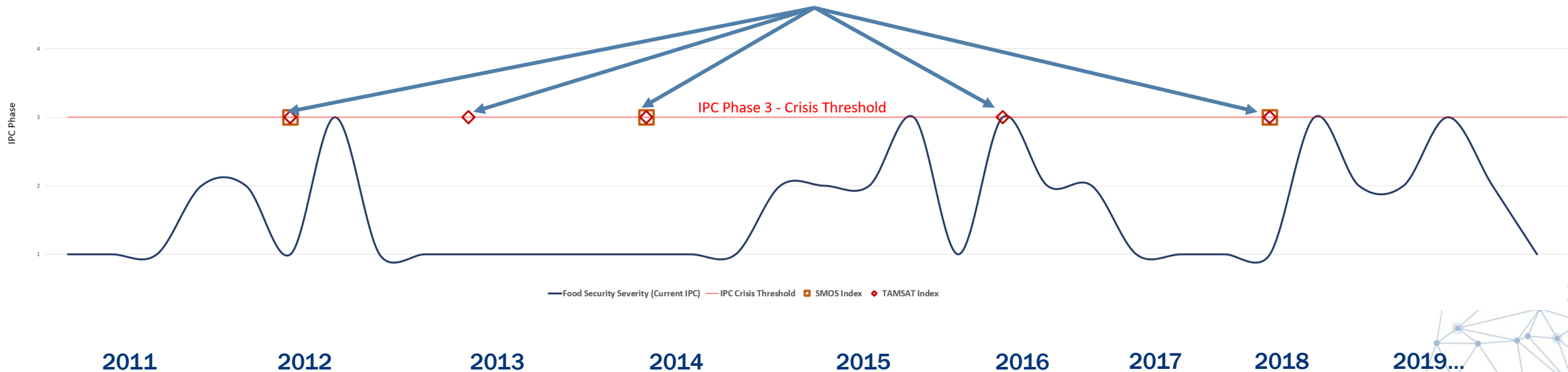
# Assessing whether triggers hit the mark – Balaka – Prototype Soil Moisture Index

**Soil Moisture Index using SMOS Data - Poor Start to the Season**  
**Triggers: 2011/12, 2014/15, 2017/18**



# Assessing whether triggers hit the mark – Balaka – Prototype Rainfall Index

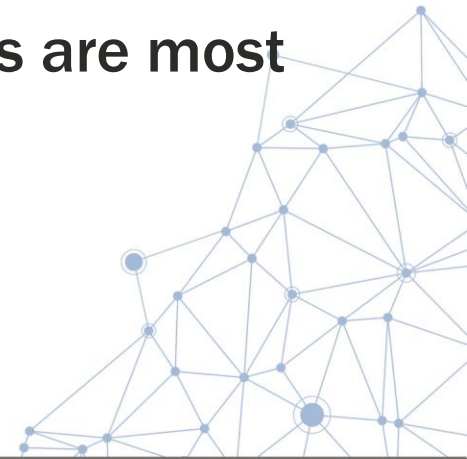
**Rainfall Index using TAMSAT Data - Poor Start to the Season**  
**Triggers: 2011/12, 2013/14, 2014/15, 2016/15/16, 2017/18**





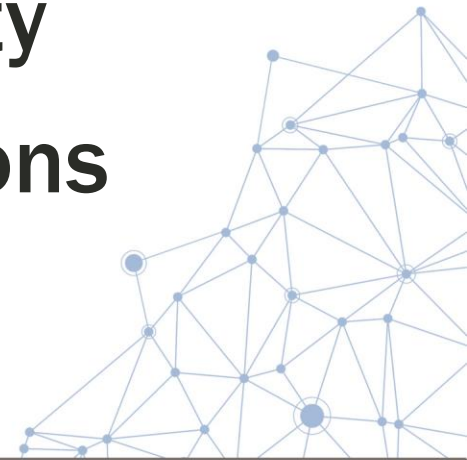
# Conclusions

- Multiple good data sets exist for primary trigger development.
- Preliminary analysis shows promising correlations between these data sets and food insecurity.
- Not all drought events lead to food crises and not all food crises are caused by drought.
- Secondary triggers help ensure crises are not missed.
- Understanding seasonality is critical to understanding which indicators are most indicative of food crises and fit for purpose in risk finance.



## Next Steps

- Review additional data including rainfall, production and yield, food price and other data.
- Refine primary and secondary trigger indexes designs
- Finalize comprehensive evaluation of primary and secondary trigger correlations to food insecurity
- Develop and fully test trigger mechanism options



**Thanks and questions?**

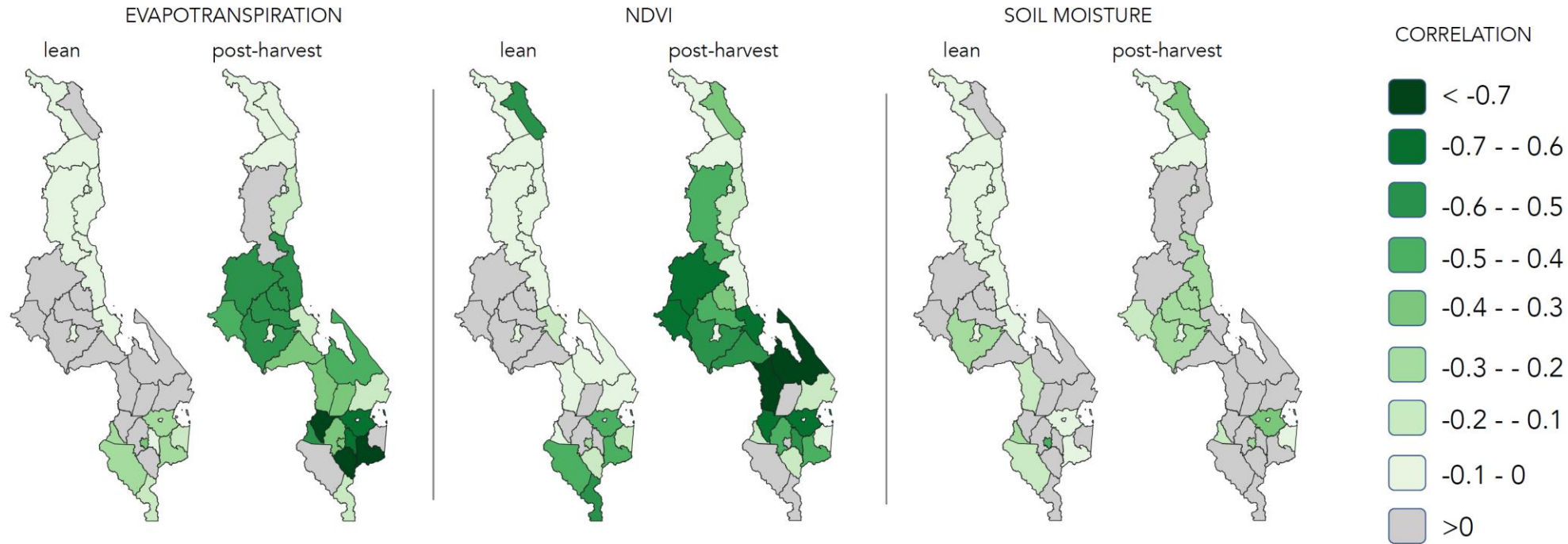


## Additional Slides

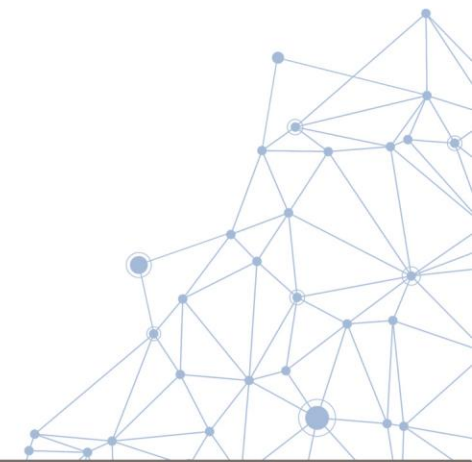


Review and analyze potential primary and secondary trigger indicators and data

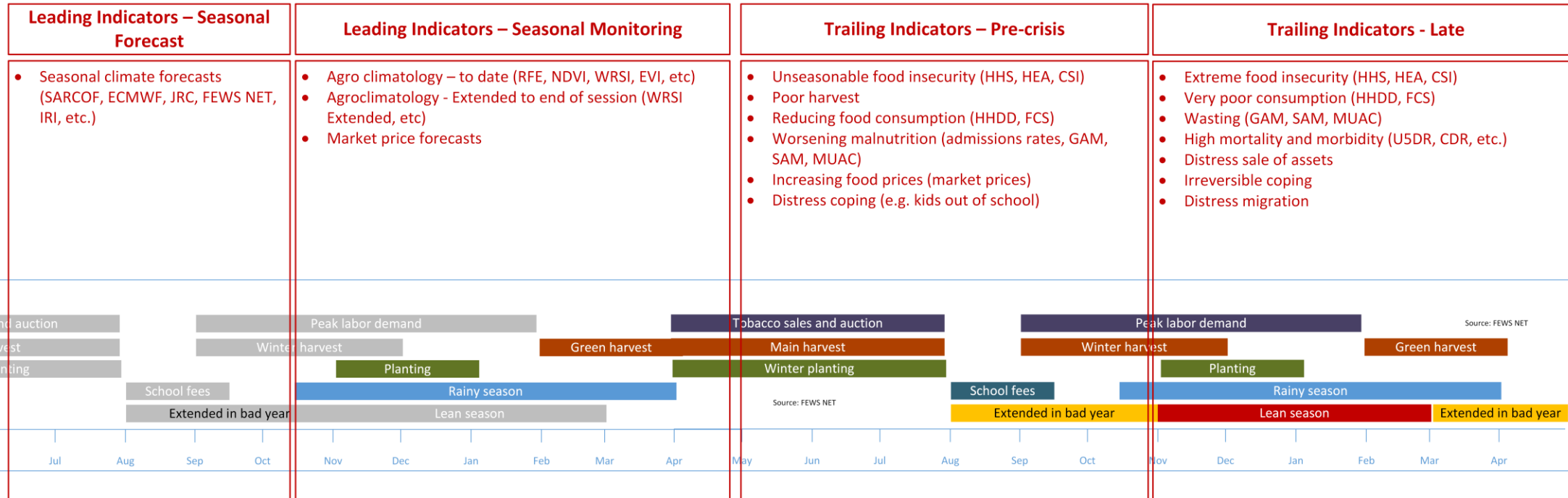
# Remote Sensing Data Correlations to Food Insecurity



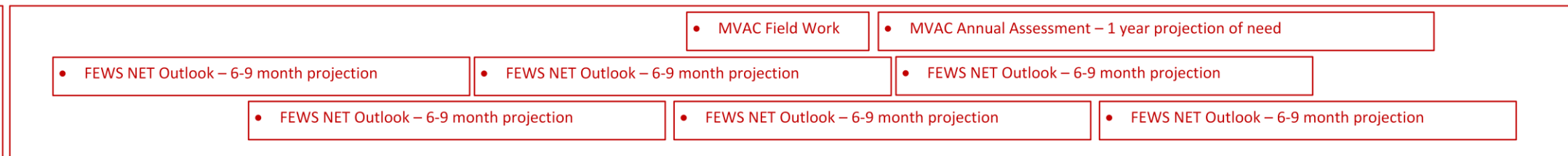
These maps show the correlations between remote sensing indicators of drought and food security severity, comparing food security during the post harvest and lean season periods with the previous 3-month average values.



# Aligning Indicators through seasonal cycles

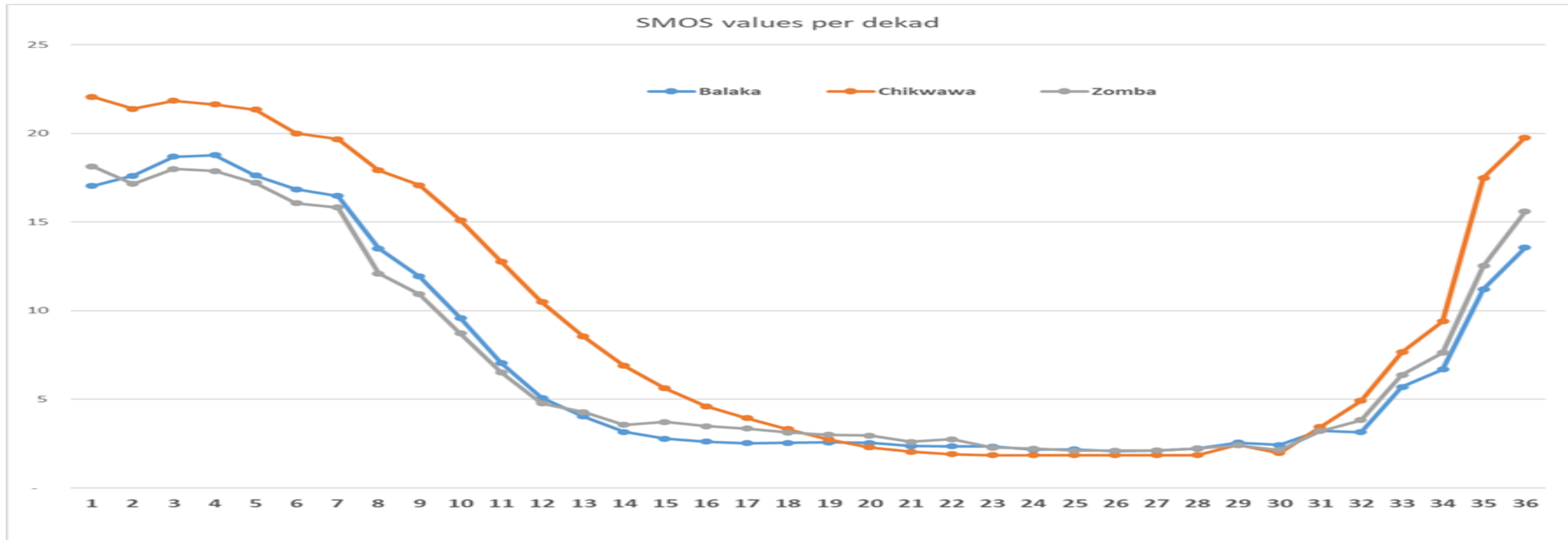


## Food Security Assessments and Forecasts



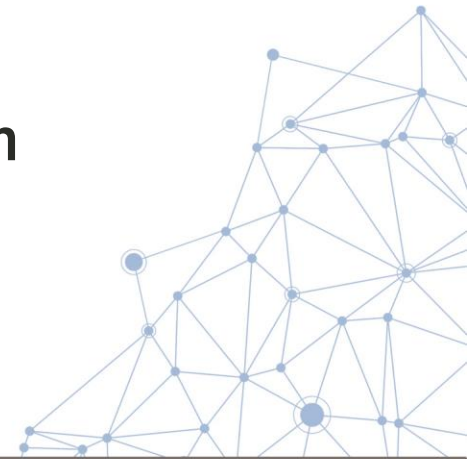
# Seasonal Trend of satellite datasets

- Modelling of satellite based historical datasets (varying between 10-38 years) of Rainfall, Vegetation Index, Soil-Moisture and Evapotranspiration was done, based on the selected satellite datasets.
- Overall, the selected satellite datasets indicate the seasonal trends and detect major historical droughts



# Selected Drought Indices

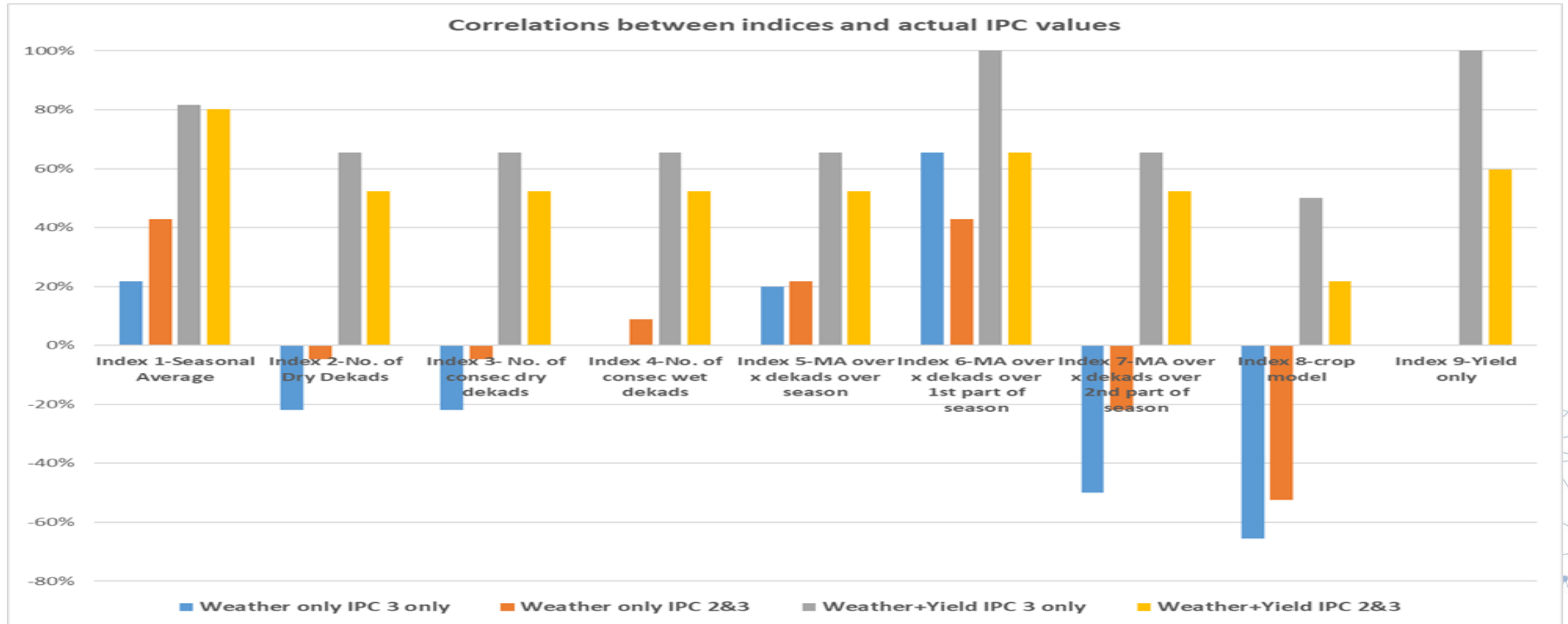
1. Index 1- seasonal drought (rainfall is low over the course of the rainy season)
2. Index 2- Many dry dekads (10-days) in the season
3. Index 3- Many consecutive dry dekads in the season
4. Index 4- Low number of consecutive wet dekads in the season
5. Index 5- Low rainfall over 30 consecutive days during the season
6. Index 6- Low rainfall over 30 consecutive days over 21<sup>st</sup> Nov- 10<sup>th</sup> Feb
7. Index 7- Low rainfall over 30 consecutive days over 11<sup>th</sup> Jan- 31<sup>st</sup> March
8. Index 8- Low rainfall for specific phases of crop cycle for Maize





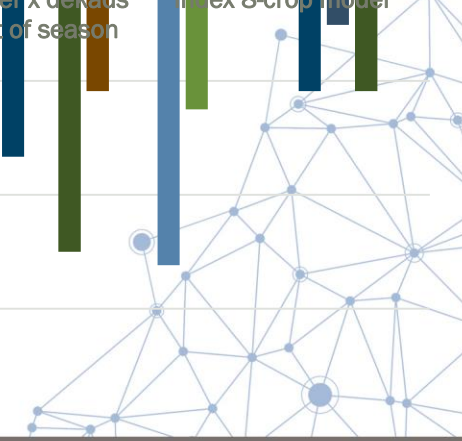
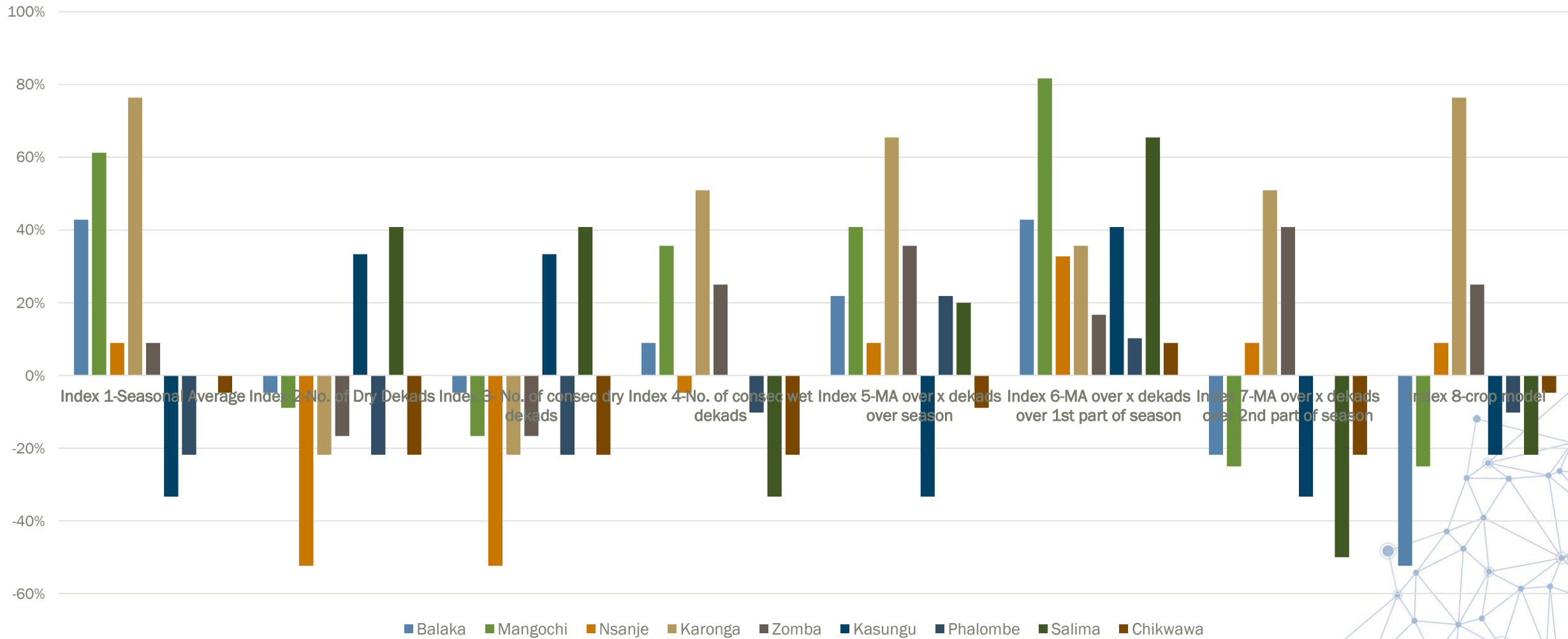
# Soil-Moisture Correlations for Balaka district

Overall Index 6 (low rainfall/soil-moisture etc.) over 20-30 days over end of Nov to early Feb, seems to be performing well so far, in terms of correlations to IPC values.



# Correlations between drought indices and IPC values for multiple districts

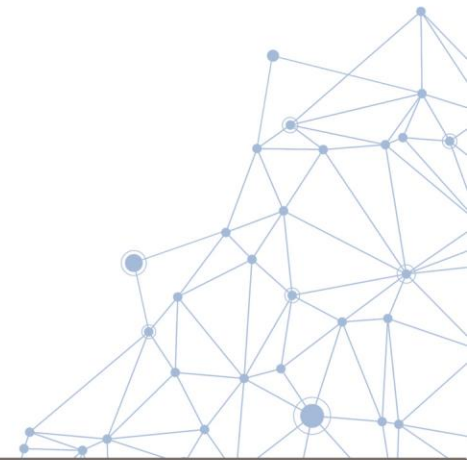
Correlations between different drought indices and IPC 2 or 3 events per district



# Drought Index Example

## Index 6: Dry spell definition (based on TAMSAT)

- a) **Early Dry spell-** Over 1<sup>st</sup> November to 31<sup>st</sup> January, if the total rainfall over 20 consecutive days is less than 70% of the normal rainfall for the same 20-day time period, then early dry spell triggers
  
- b) **Late Dry spell-** Over 1<sup>st</sup> February to 31<sup>st</sup> March, if the total rainfall over 30 consecutive days is less than 70% of the normal rainfall for the same 30-day time period, then late dry spell triggers



# Historical Drought Index Triggers for Balaka, based on TAMSAT rainfall data

