



PREDISAN

Predictive Model for Food Security in Central America based on Surveys and Secondary Data

GIS4tech
Spin-Off of the University of Granada (Spain)

March 2024

OUTLINE



1. WHO ARE WE?

GIS, IA AND KNOWLEDGE TRANSFER



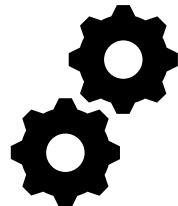
2. INTRODUCTION of the ISSUE

LACK OF DETAILED INFORMATION FOR DECISIONS OF DISTRIBUTION OF AID



3. METHODOLOGY of PREDISAN

PROCESS FOR PREDICTIVE MODEL FOR FOOD SECURITY



4. PREDICTIVE MODELS

REACHING BEYOND FOOD SECURITY FIELD DATA



5. FUTURE IMPROVEMENTS

REFLECTIONS AND AREAS FOR IMPROVEMENTS



6. DEPLOYMENT IN OTHER REGIONS

ADAPTING THE APPROACH TO OTHER REGIONS



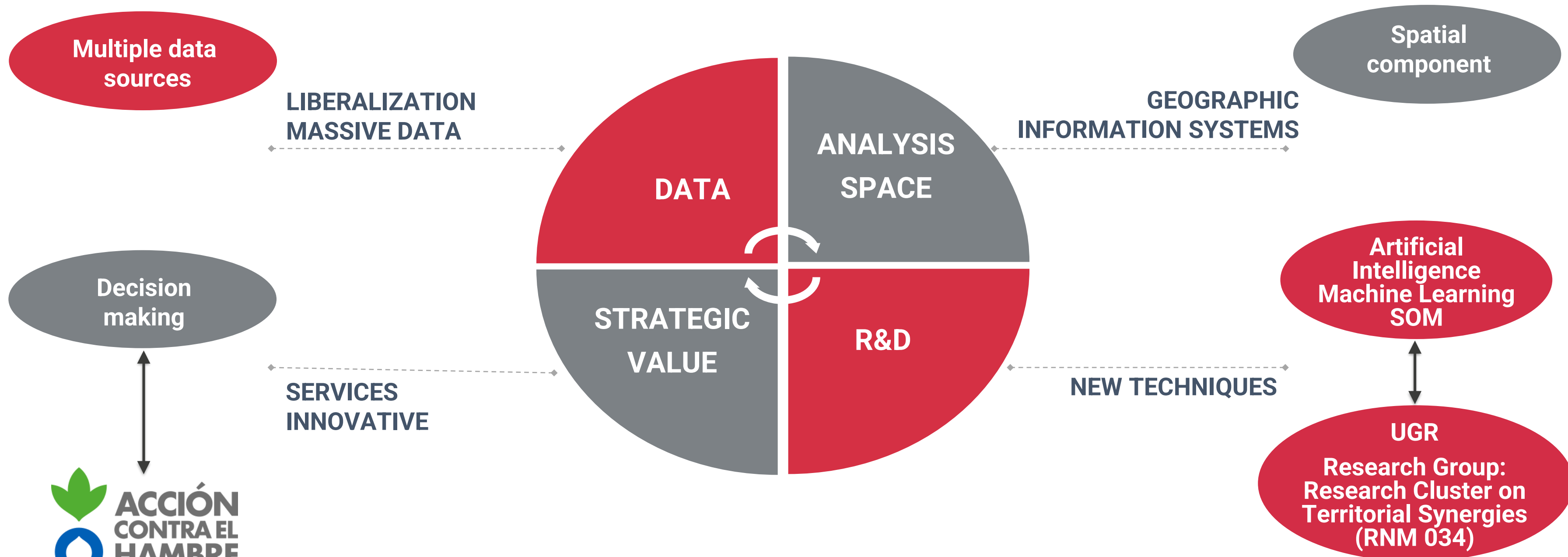
7. CONCLUDING REFLECTIONS



1. WHO ARE WE?

GIS, AI, AND KNOWLEDGE TRANSFER.

1. WHO ARE WE?



act:onaid



European Union
Civil Protection and
Humanitarian Aid

USAID
FROM THE AMERICAN PEOPLE



Spin-Off
UNIVERSIDAD DE GRANADA



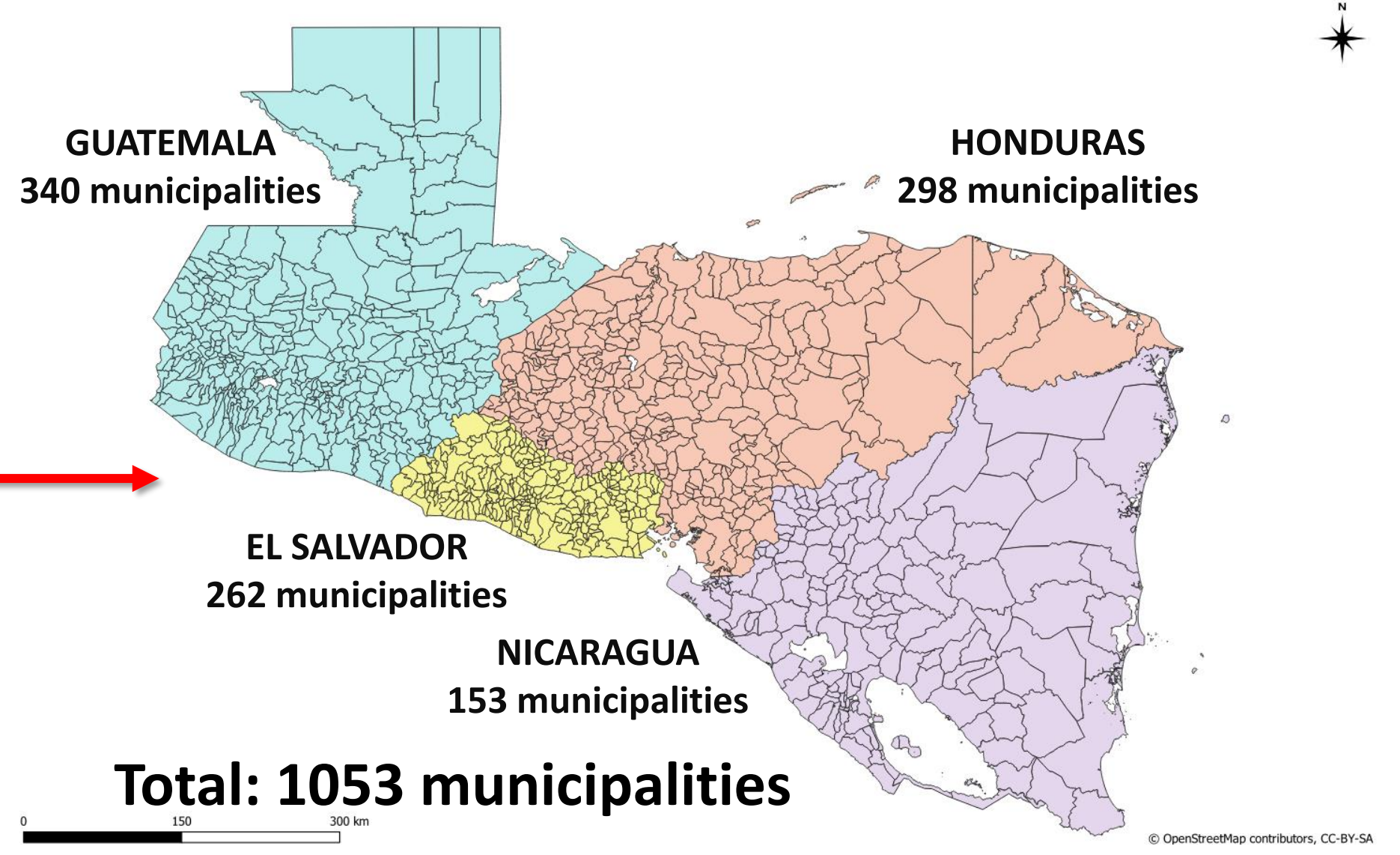
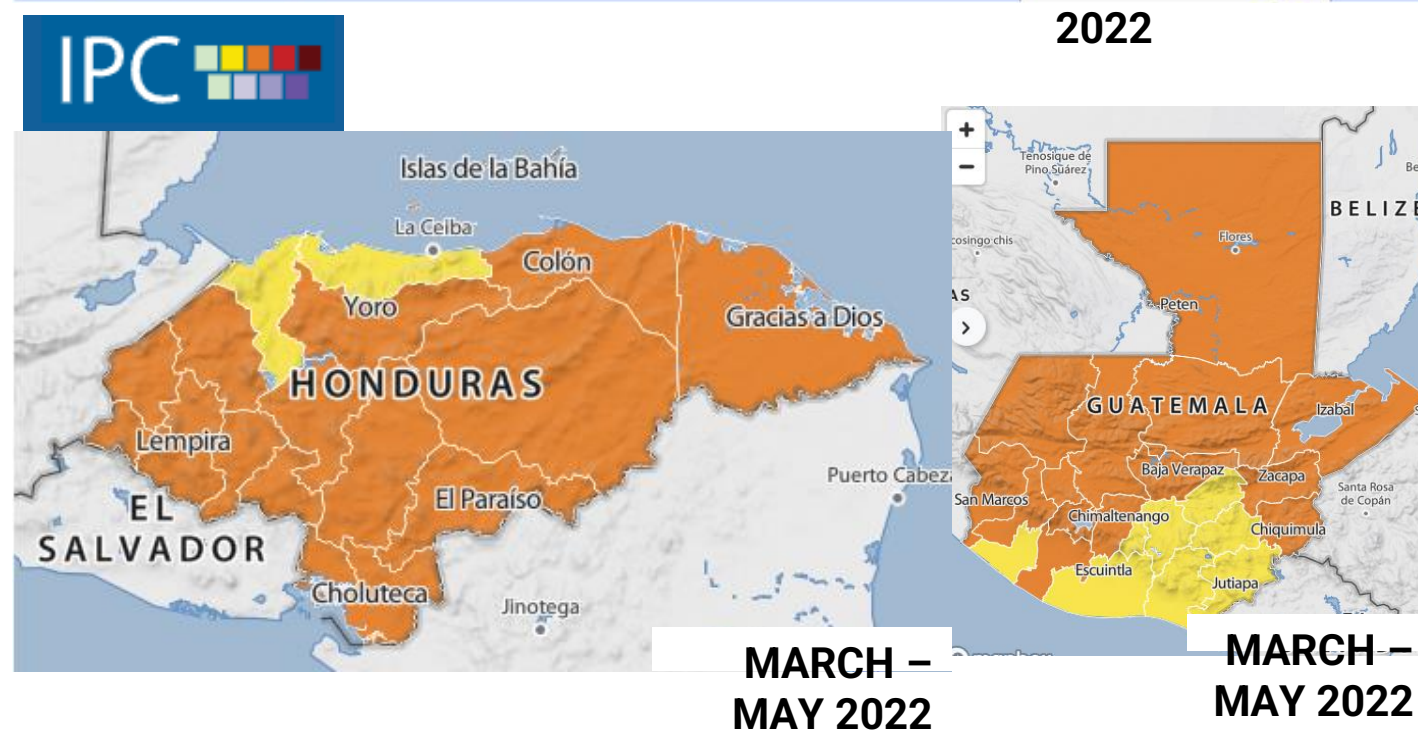
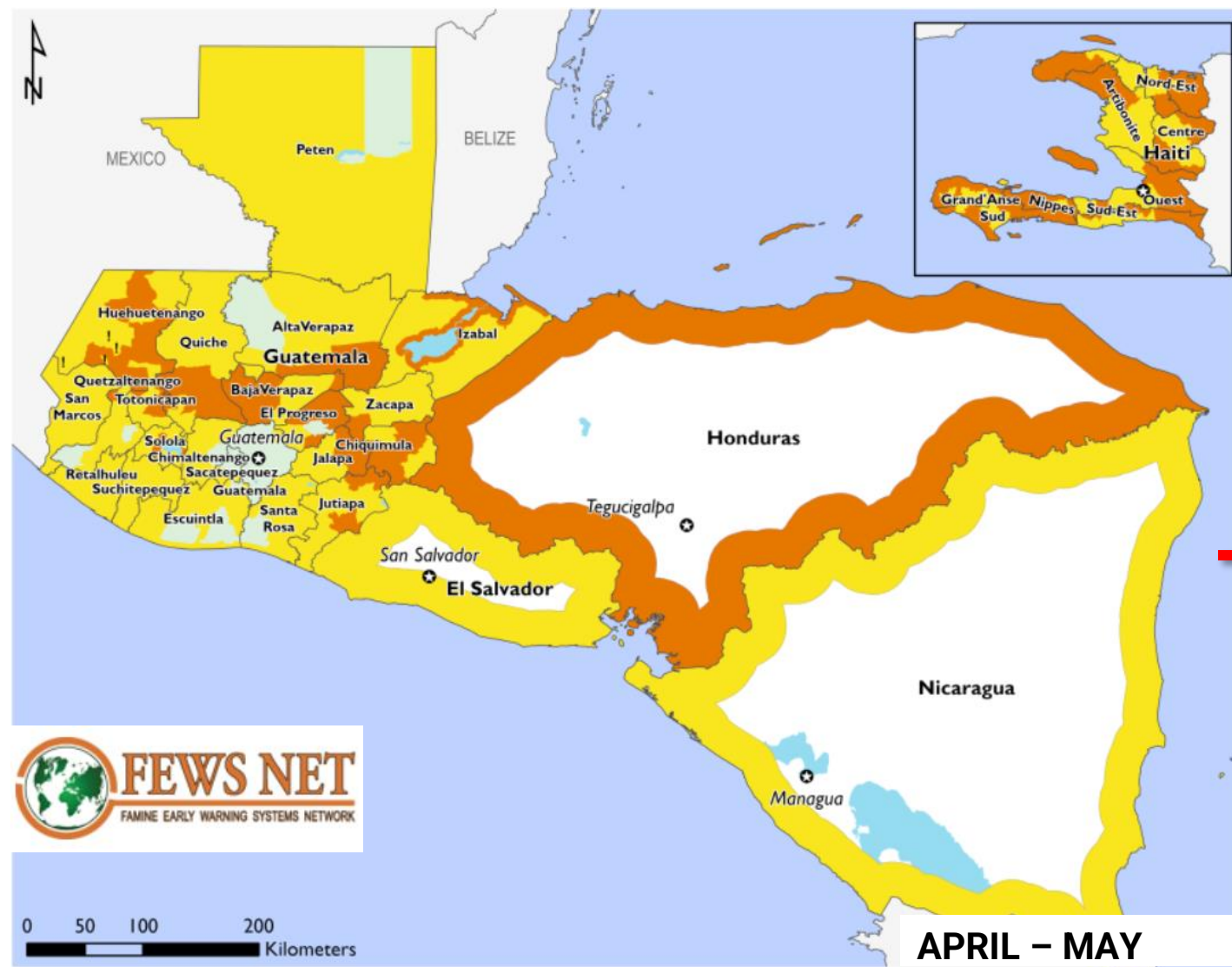


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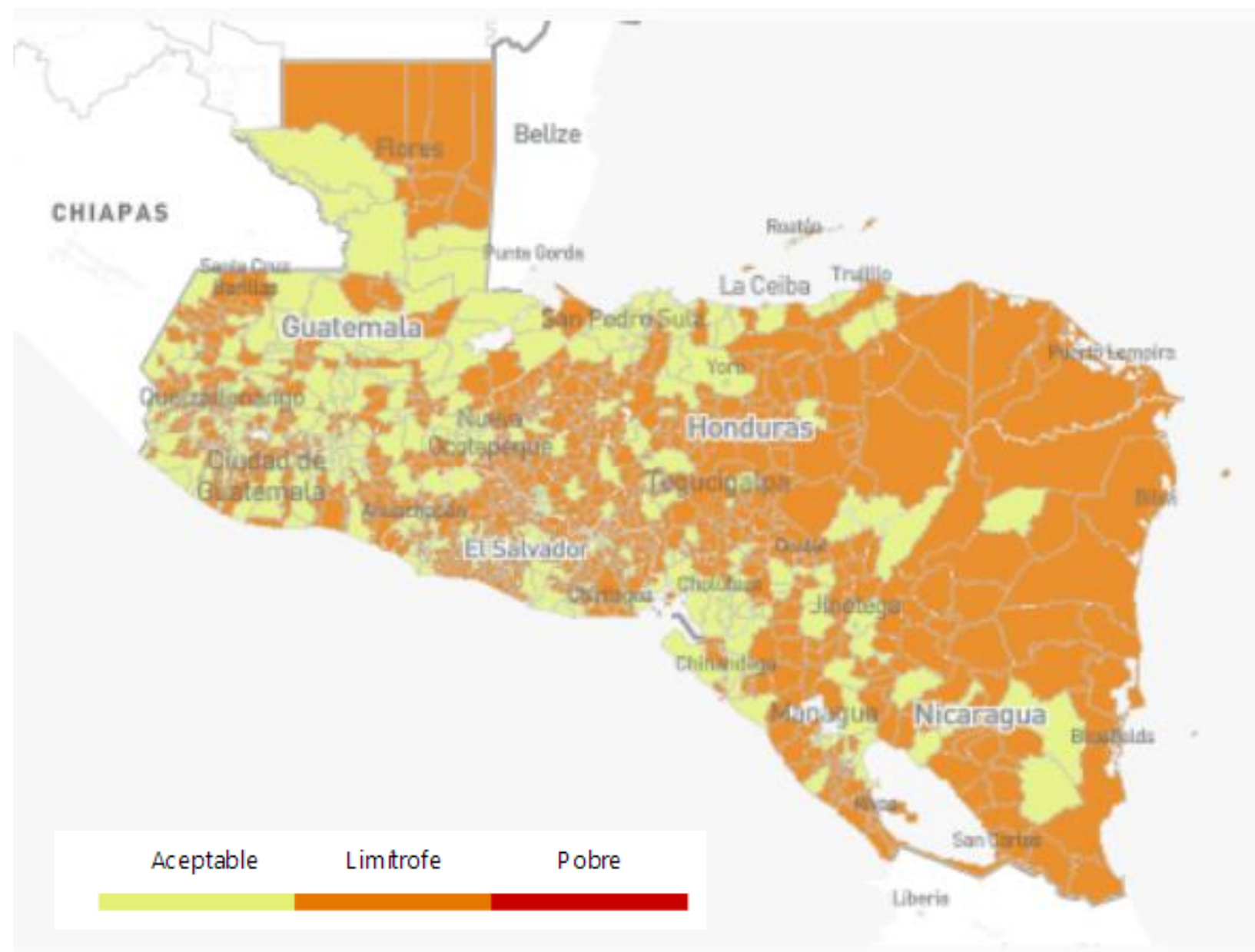


Sources: Vulnerability Indicators from FEWS NET and IPC.

2. INTRODUCTION. THE POINT

REACHING BEYOND FOOD SECURITY FIELD DATA

Predicciones Puntaje de consumo de alimentos (PCA)

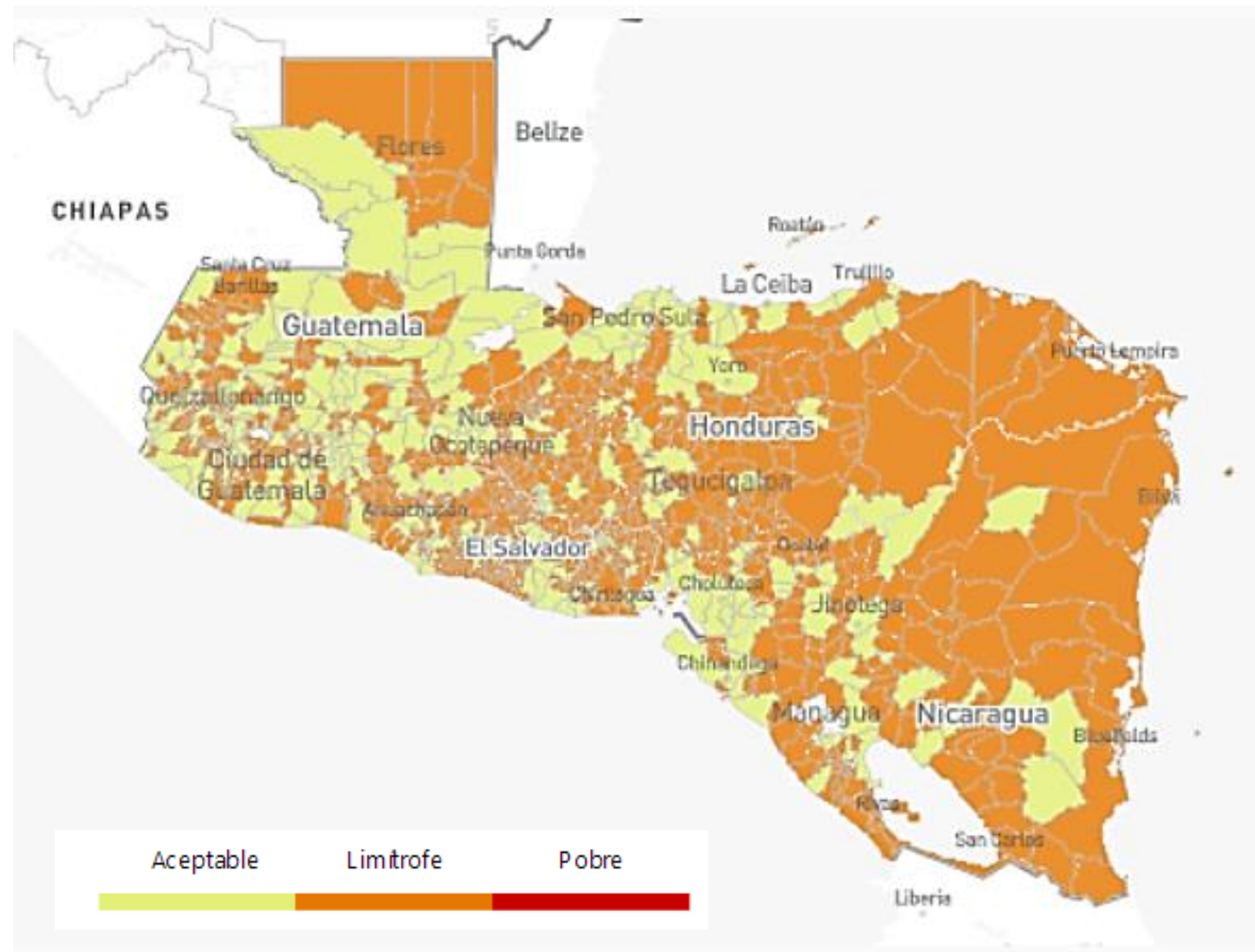


OBJECTIVE

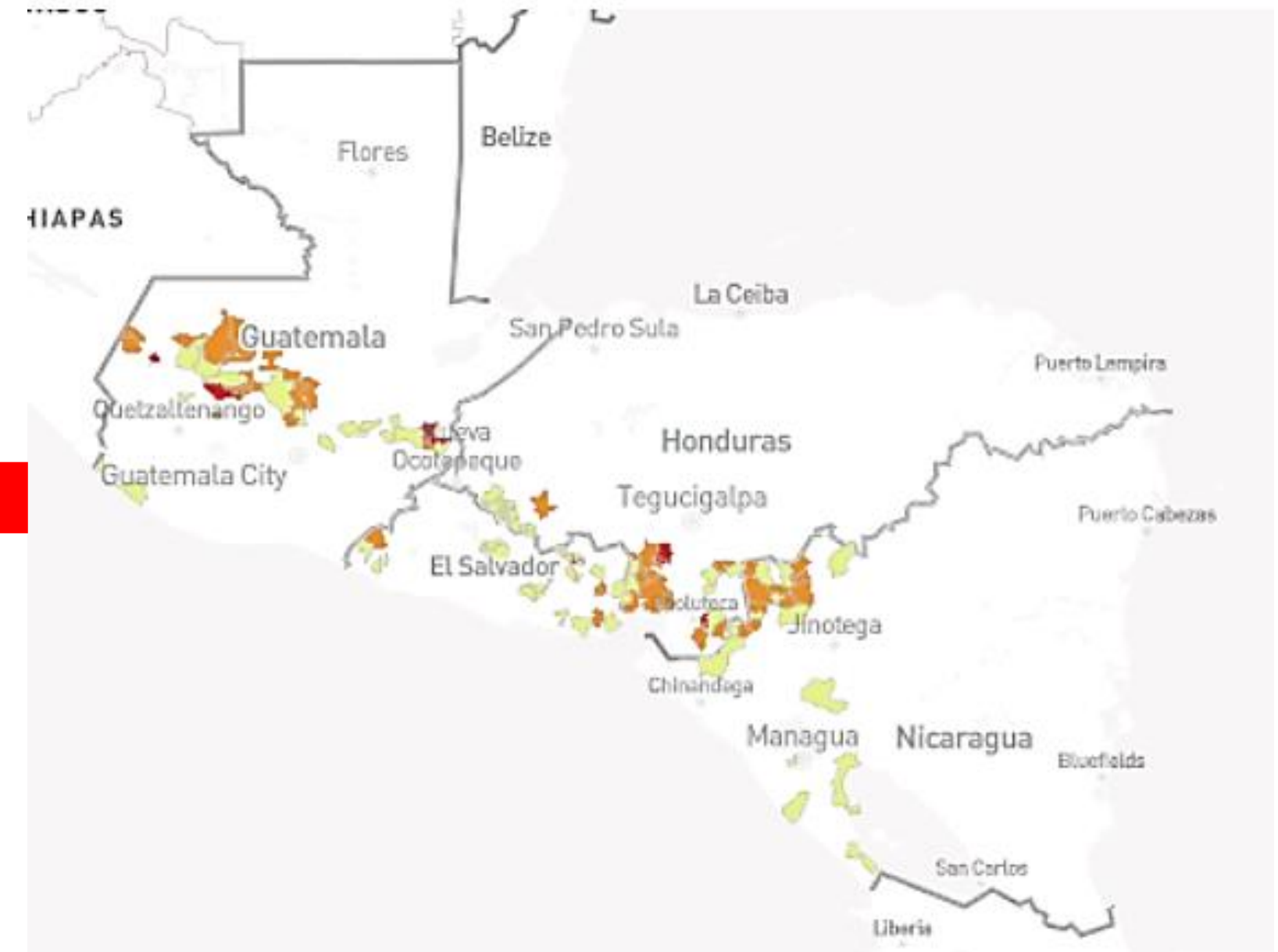
2. INTRODUCTION. THE POINT

REACHING BEYOND FOOD SECURITY FIELD DATA

Predicciones Puntaje de consumo de alimentos (PCA)



Puntaje de consumo de alimentos (PCA)



Source: PREDISAN. Food Consumption Score (FCS) October 2021

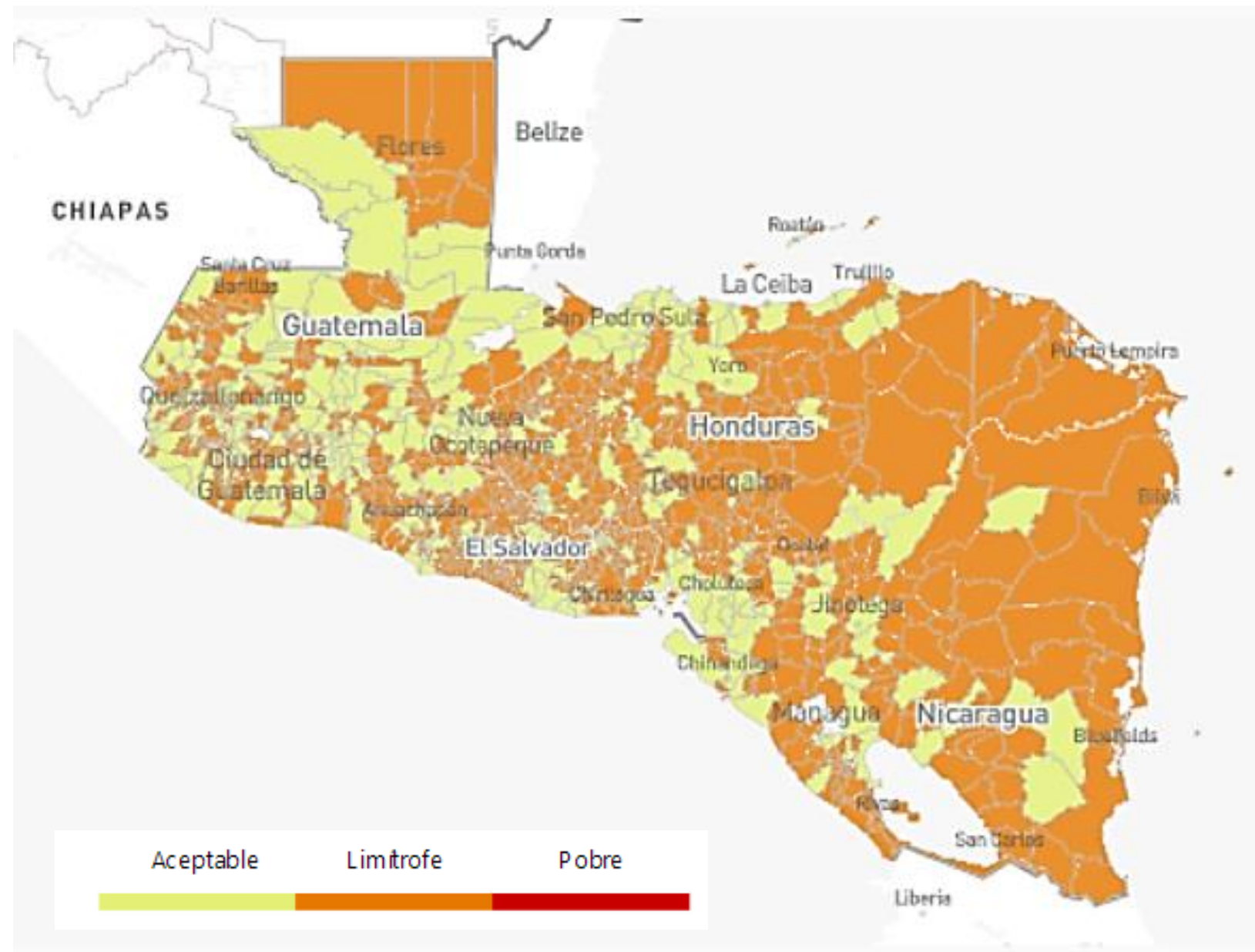
OBJECTIVE

BASIS

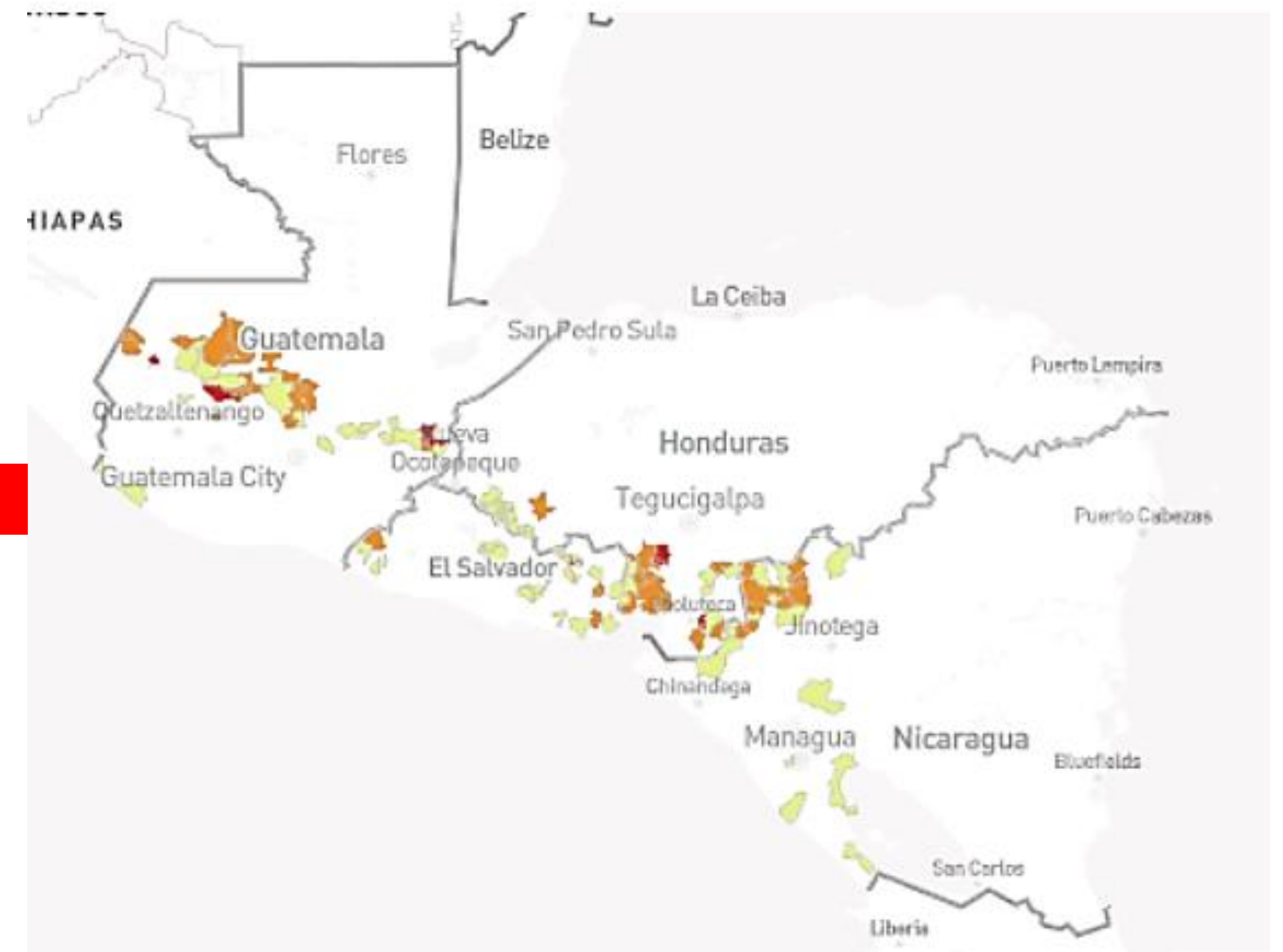
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Puntaje de consumo de alimentos (PCA)



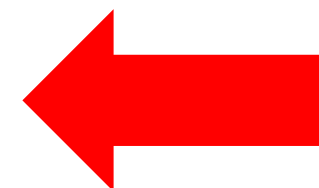
Source: PREDISAN. Food Consumption Score (FCS) October 2021

OBJECTIVE

1st. PREDICTIONS (nowcasting)

BASIS

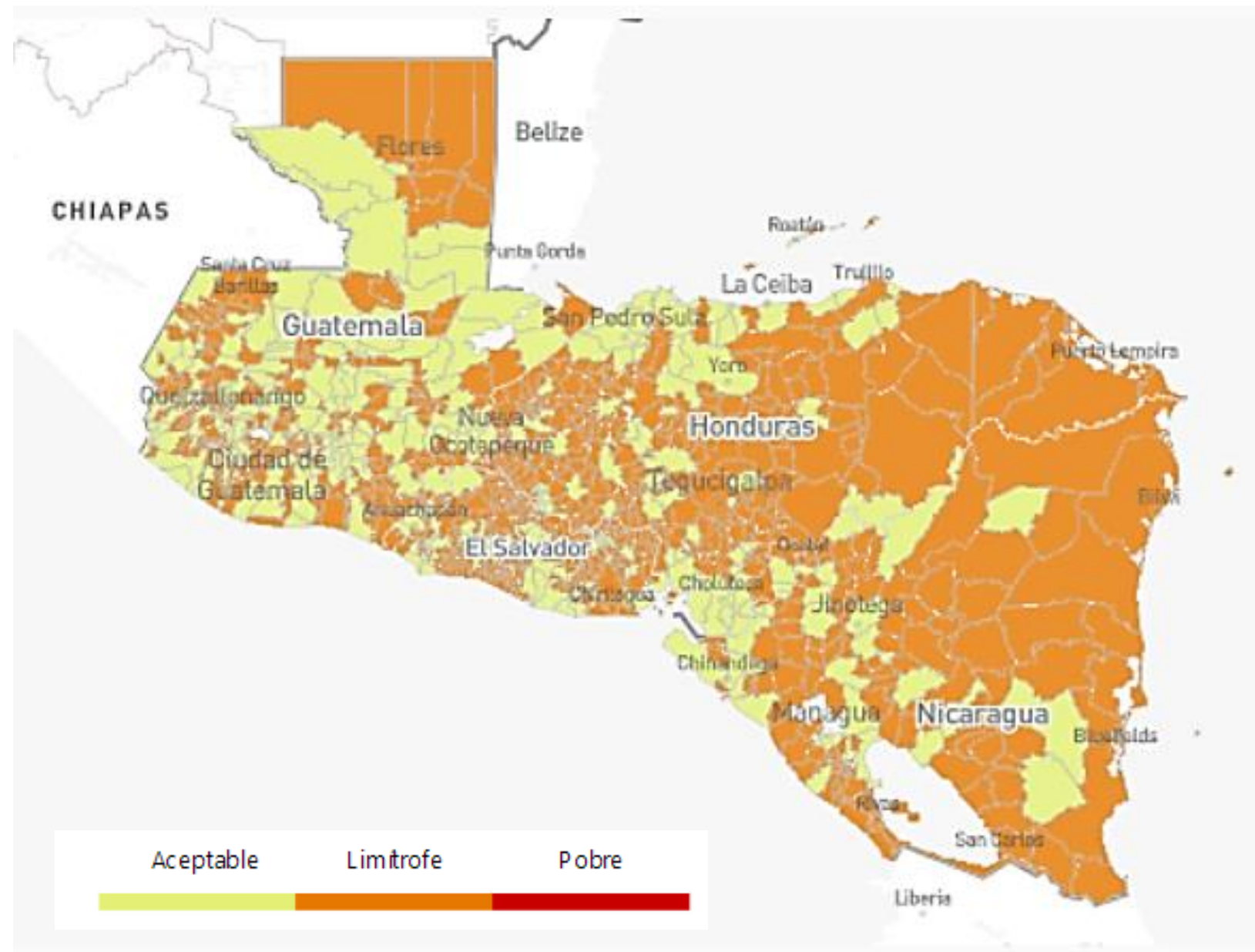
REAL DATA (NGO's field surveys)



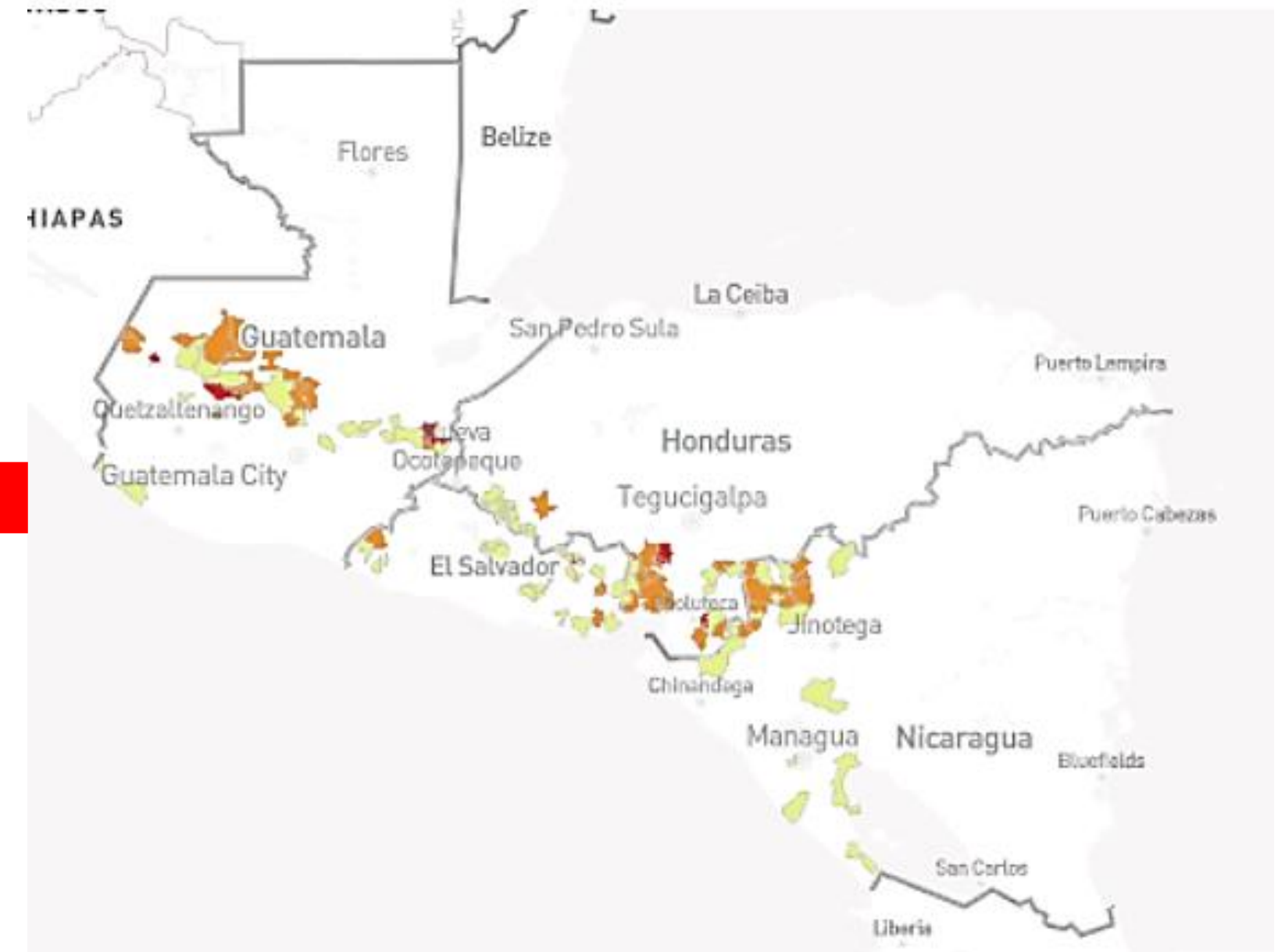
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Puntaje de consumo de alimentos (PCA)



Source: PREDISAN. Food Consumption Score (FCS) October 2021

OBJECTIVE

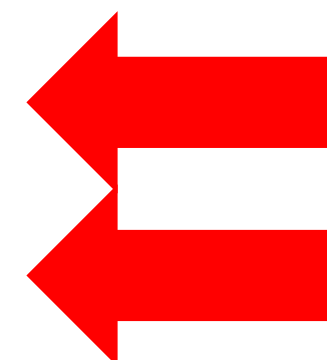
1st. PREDICTIONS (**now**casting)

2nd. PREDICTIONS (**fore**casting)

BASIS

REAL DATA (NGO's field surveys)

SCARCITY or NO REAL DATA



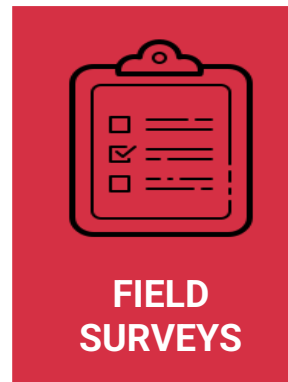


3. METHODOLOGY of PREDISAN

PROCESS FOR PREDICTIVE MODEL FOR FOOD SECURITY

3. METHODOLOGY of PREDISAN

REACHING BEYOND FOOD SECURITY FIELD DATA



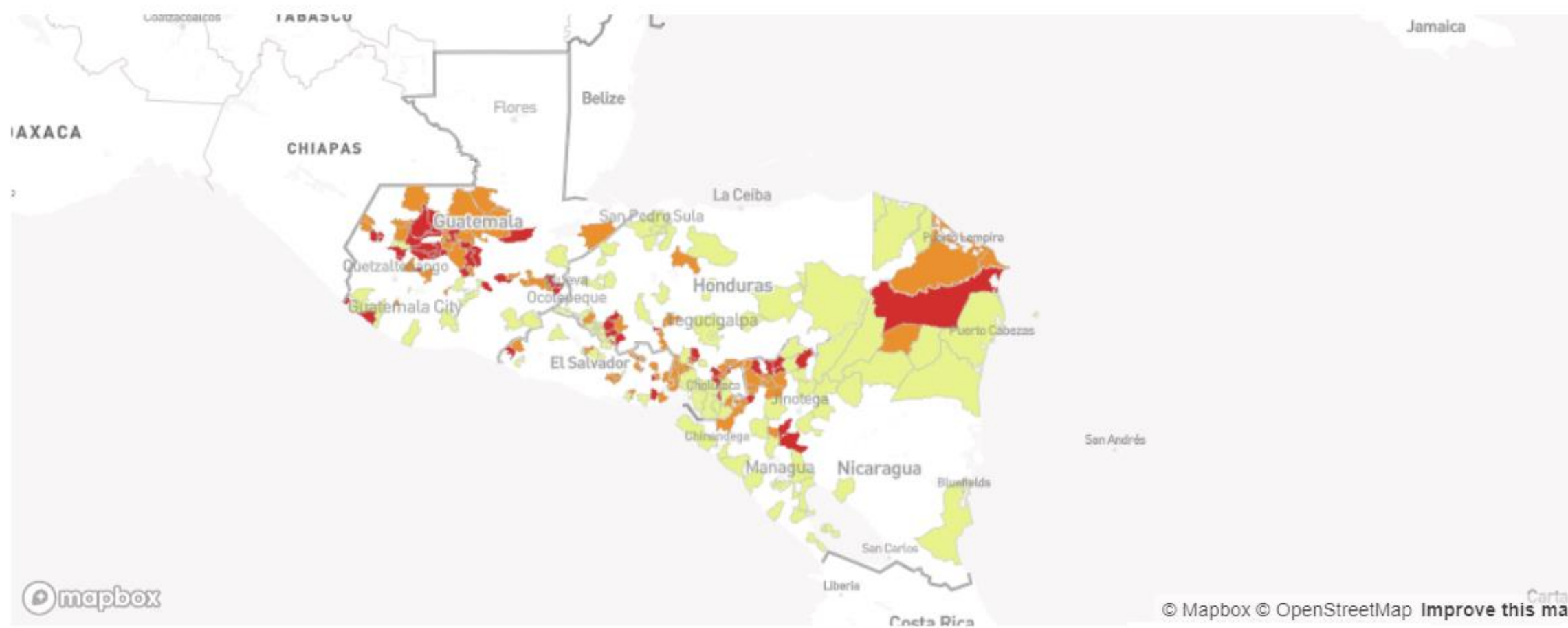
3. METHODOLOGY of PREDISAN

REACHING BEYOND FOOD SECURITY FIELD DATA



Food Consumption Score (FCS)

HDX
Descargar datos desde HDX



Aplicando la regla del 20% a la población la categoría es:
Limítrofe

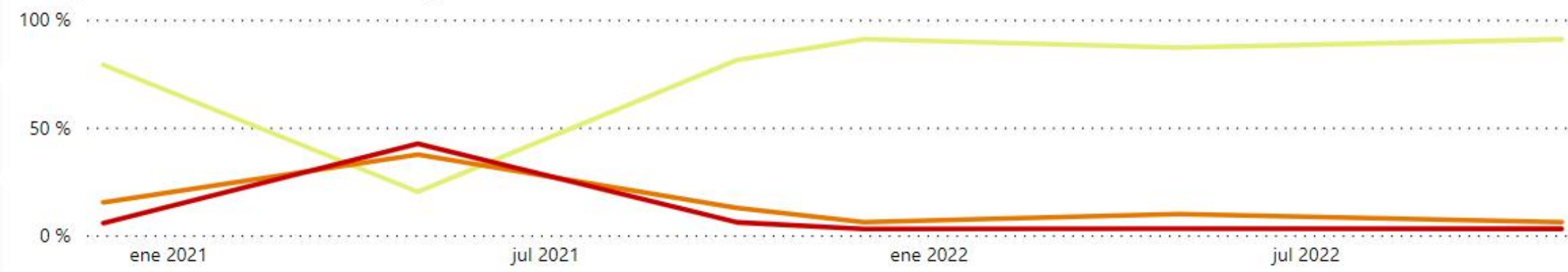
Fuente: ACH

País: Todas

Departamento, Municipio: Todas

Fecha: Selección múltiple

Comportamiento histórico de categorías



Comportamiento según el período:
diciembre de 2020

Fases	Población	%
Aceptable	15.893.246	68,81%
Limítrofe	2.508.288	17,66%
Pobre	1.462.331	13,52%

Cantidad de personas en fase Limítrofe y Pobre:
3.970.619



SOURCE: ACH. Deployed at PREDISAN by the platform authors

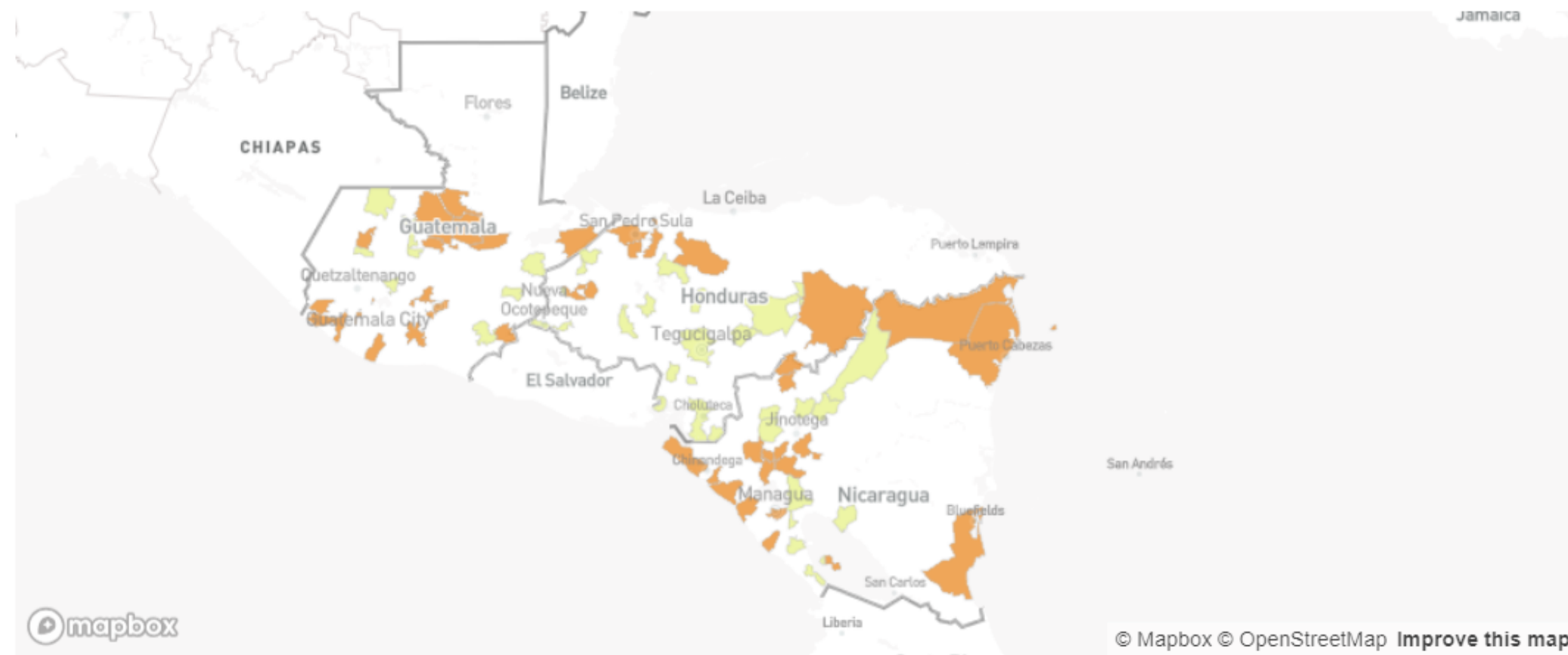
3. METHODOLOGY of PREDISAN

REACHING BEYOND FOOD SECURITY FIELD DATA



Household Hunger Scale (HHS)

HDX
 Descargar datos desde HDX



Aplicando la regla del 20% a la población la categoría es:
Moderada incidencia

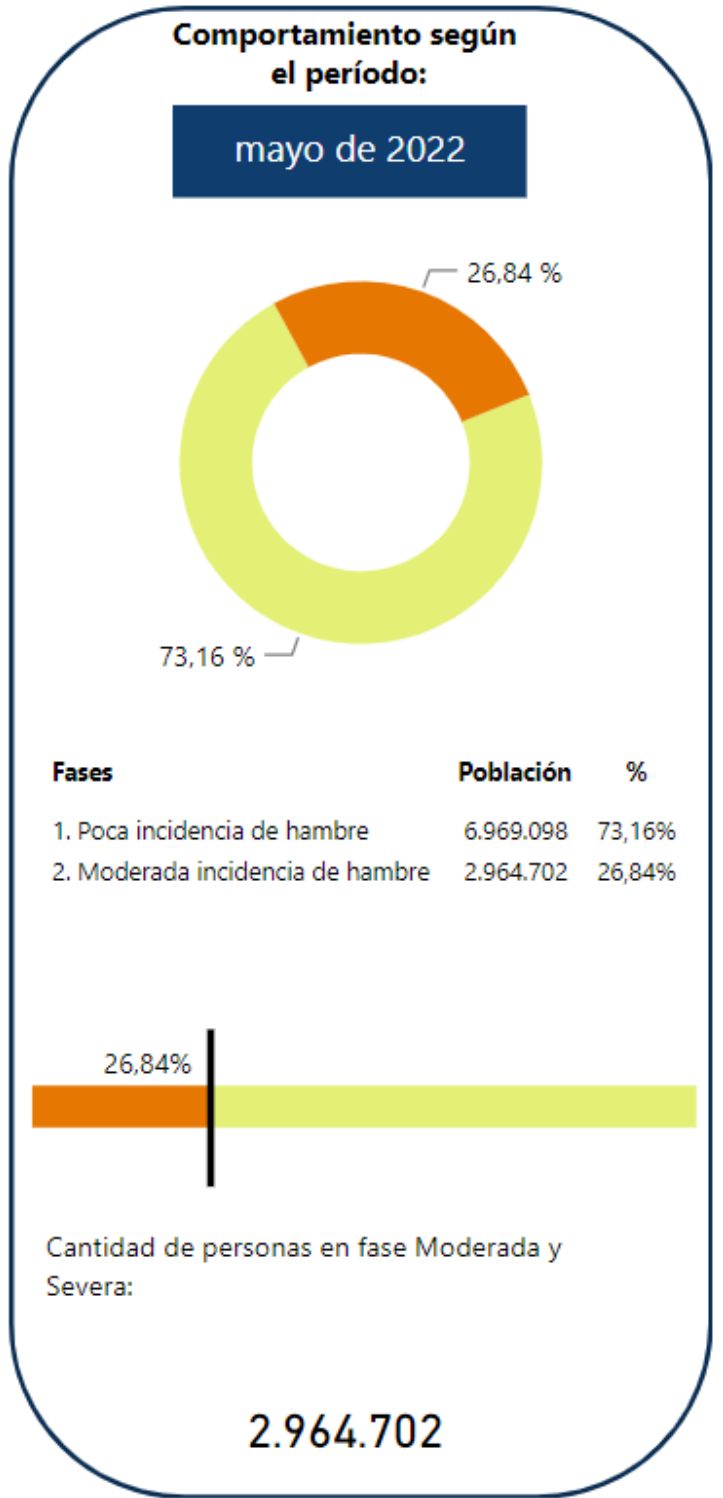
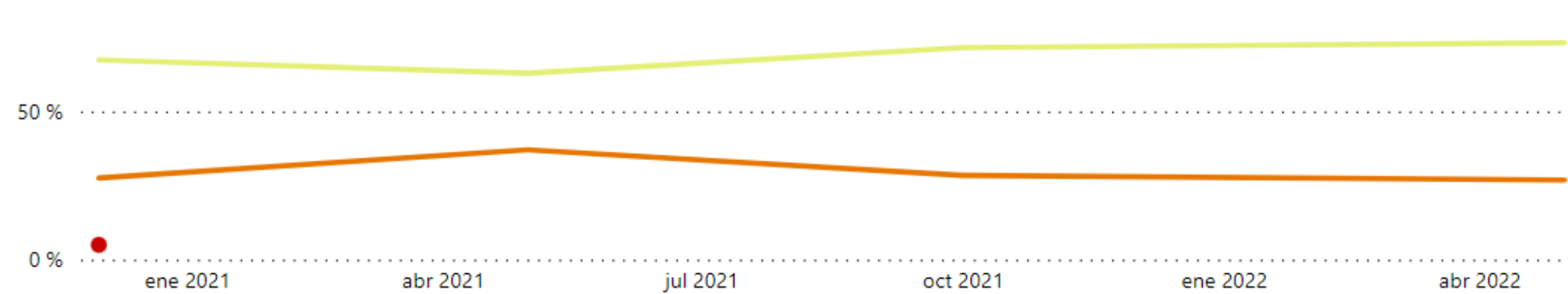
Fuente:
 Consorcio de Organizaciones Humanitarias

País:
 Todas

Departamento, Municipio:
 Todas

Fecha:
 mayo de 2022

Comportamiento histórico de categorías



SOURCE: Consorcio de Organizaciones Humanitarias (ACH and others)
 Deployed at PREDISAN by the platform authors

3. METHODOLOGY of PREDISAN

REACHING BEYOND FOOD SECURITY FIELD DATA



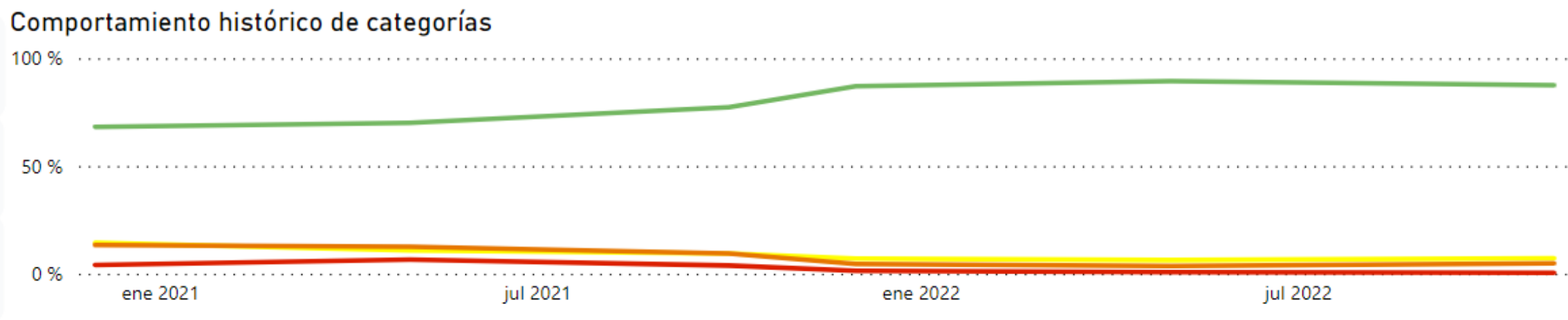
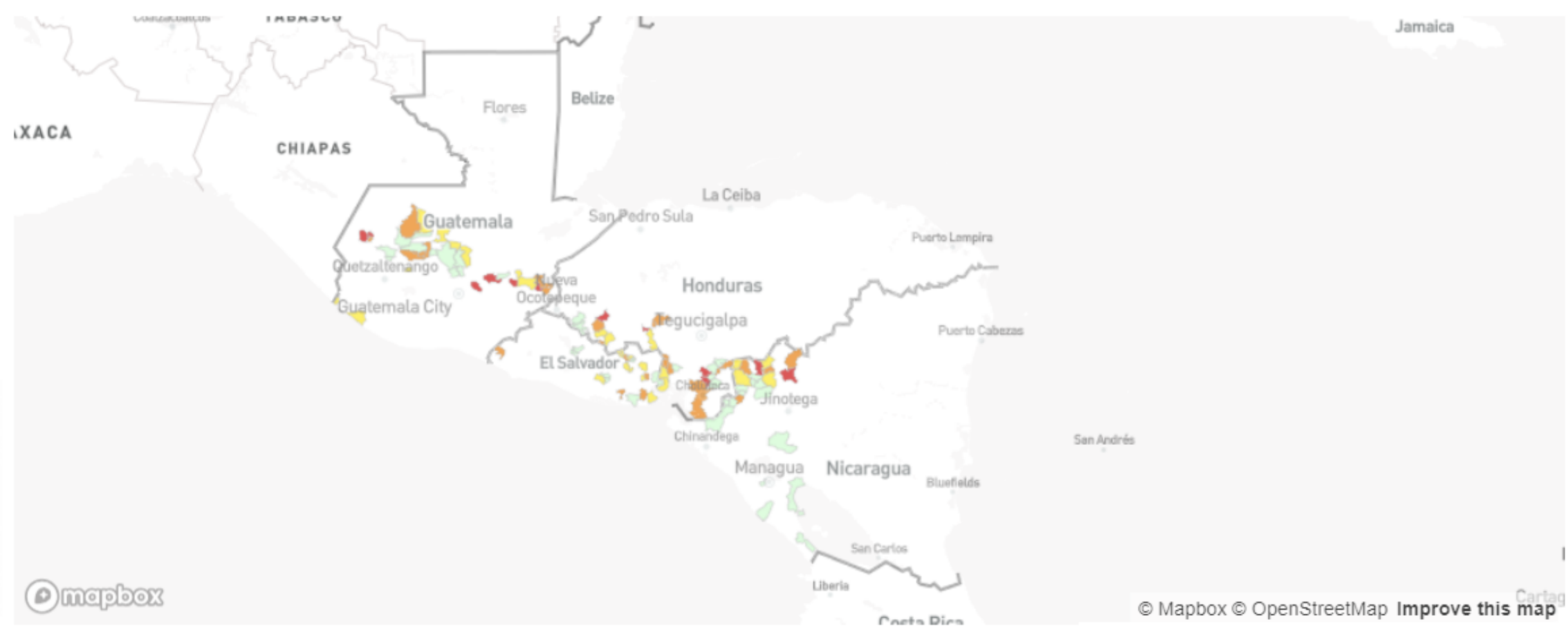
Household Dietary Diversity Score (HDDS)

Descargar datos desde HDX
 HDDS Nivel Departamental (FAO-DIEM)

Aplicando la regla del 20% a la población la categoría es:
Acentuada

Fuente:
 Consorcio de Organizaciones Humanitarias

País:
 Departamento, Municipio:
 Fecha:



Comportamiento según el período:

mayo de 2021

Fases	Población	%
Fase 1. Ninguna	1.552.351	70,02%
Fase 2. Acentuada	315.423	10,83%
Fase 3. Crisis	301.615	12,59%
Fase 4. Emergencia	135.437	6,57%

Cantidad de personas en fase Acentuada, Crisis y Emergencia:
752.475



SOURCE: Consorcio de Organizaciones Humanitarias (ACH and others)
 Deployed at PREDISAN by the platform authors

3. METHODOLOGY of PREDISAN

REACHING BEYOND FOOD SECURITY FIELD DATA



Other features

RESULTADOS DE PRIMER NIVEL



Puntaje de Consumo de Alimentos (PCA)



Escala de Hambre en el Hogar (HHS)



Índice de Estrategias de afrontamiento Reducido (rCSI)



Puntaje de Diversidad Dietética en el Hogar (HDDS)



Indicador de Estrategias de Afrontamiento de Medios de Vida (LCS)



Escala de Experiencia de Inseguridad Alimentaria (FIES)



Evolución Indicadores entre los grupos de población más excluidos

RESULTADOS DE SEGUNDO NIVEL



Desnutrición aguda por MUAC



Desnutrición aguda peso talla



Desnutrición crónica



Desnutrición Global




Ayudas

+200 QUESTION SURVEYS

3. METHODOLOGY of PREDISAN

REACHING BEYOND FOOD SECURITY FIELD DATA




SATELLITE IMAGERY
REMOTE SENSING
AGRO-CLIMATIC MODEL



FIELD
SURVEYS

3. METHODOLOGY of PREDISAN

REACHING BEYOND FOOD SECURITY FIELD DATA



**SATELLITE IMAGERY
REMOTE SENSING
AGRO-CLIMATIC MODEL**



**FIELD
SURVEYS**

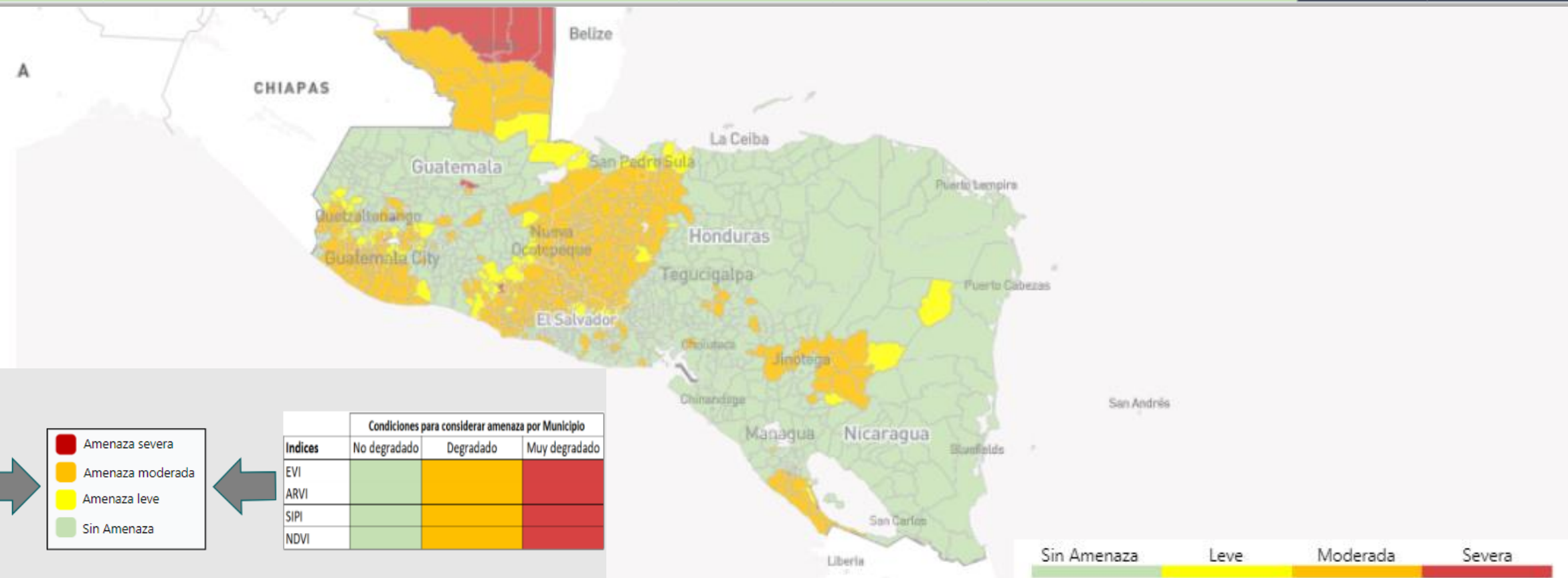
Agro-climatic Threats

PAÍS	DEPARTAMENTO	MUNICIPIO	SPI 1	SPI 3	SPI 6	SPI 9	SPI 12	EVI	ARVI	SIPI	NDVI	HÉCTAREAS DE CULTIVOS	HÉCTAREAS DE PASTO	NÚMERO DE HABITANTES	Amenazas Agroclimáticas globales
El Salvador	Sonsonate	Acajutla	Sin Amenaza	Sin Amenaza	Amenaza leve	Amenaza moderada	Amenaza moderada	Amenaza moderada	Sin Amenaza	Amenaza moderada	Sin Amenaza	964,17	1.336,83	52.359	Amenaza moderada
Guatemala	Chimaltenango	Acatenango	Sin Amenaza	Sin Amenaza	Amenaza leve	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	183,37	76,90	23.228	Sin Amenaza
Nicaragua	Leon	Achuapa	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Amenaza moderada	295,76	1.159,37	13.797	Sin Amenaza
Nicaragua	Chontales	Acoyapa	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	431,81	14.474,40	16.946	Sin Amenaza
Guatemala	Jutiapa	Agua Blanca	Sin Amenaza	Sin Amenaza	Amenaza leve	Amenaza moderada	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	905,02	354,91	16.353	Amenaza leve
El Salvador	Chalatenango	Agua Caliente	Sin Amenaza	Sin Amenaza	Amenaza leve	Amenaza moderada	Amenaza moderada	Sin Amenaza	Sin Amenaza	Amenaza moderada	Sin Amenaza	59,15	82,81	8.261	Amenaza moderada
Guatemala	Huehuetenango	Aguacatan	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	378,57	218,86	49.607	Sin Amenaza
Honduras	La Paz	Aguanqueterique	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	136,05	283,93	5.025	Sin Amenaza
El Salvador	San Salvador	Aguilares	Sin Amenaza	Sin Amenaza	Amenaza leve	Amenaza moderada	Amenaza moderada	Amenaza moderada	Sin Amenaza	Amenaza severa	Sin Amenaza	189,29	147,88	21.267	Amenaza moderada
El Salvador	Ahuachapan	Ahuachapan	Sin Amenaza	Sin Amenaza	Amenaza leve	Amenaza moderada	Amenaza moderada	Amenaza moderada	Sin Amenaza	Amenaza moderada	Sin Amenaza	1.129,80	674,33	110.511	Amenaza moderada
Honduras	Gracias A Dios	Ahuas	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	2.330,57	20.070,14	9.171	Sin Amenaza
Honduras	Comayagua	Ajuterique	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	Sin Amenaza	834,04	130,13	11.888	Sin Amenaza

País
Todas

Departamento, Municipio
Todas

Fecha
octubre de 2023



	Condiciones para considerar amenaza en SPI por Municipio			
	No sequía	Sequía moderada	Sequía severa	Sequía extrema
SPI 1 mes	Green	Green	Yellow	Orange
SPI 3 meses	Green	Green	Yellow	Orange
SPI 6 meses	Green	Yellow	Orange	Red
SPI 9 meses	Green	Yellow	Orange	Red
SPI 12 meses	Green	Yellow	Orange	Red

Indicadores	Condiciones para considerar amenaza por Municipio		
	No degradado	Degradado	Muy degradado
EVI	Green	Yellow	Red
ARVI	Green	Yellow	Red
SIPI	Green	Yellow	Red
NDVI	Green	Yellow	Red

Nota: En "No sequía" se consideran las categorías "Humedad extrema", "Humedad severa", "Humedad moderada" y "Precipitación normal".

 ¿Qué estoy viendo?

 Cultivos afectados

 Descargar datos desde HDX

3. METHODOLOGY of PREDISAN

REACHING BEYOND FOOD SECURITY FIELD DATA



SATELLITE IMAGERY
REMOTE SENSING
AGRO-CLIMATIC MODEL



FIELD
SURVEYS



TERRITORIAL
PATTERNS

3. METHODOLOGY of PREDISAN

REACHING BEYOND FOOD SECURITY FIELD DATA



SATELLITE IMAGERY
REMOTE SENSING
AGRO-CLIMATIC MODEL

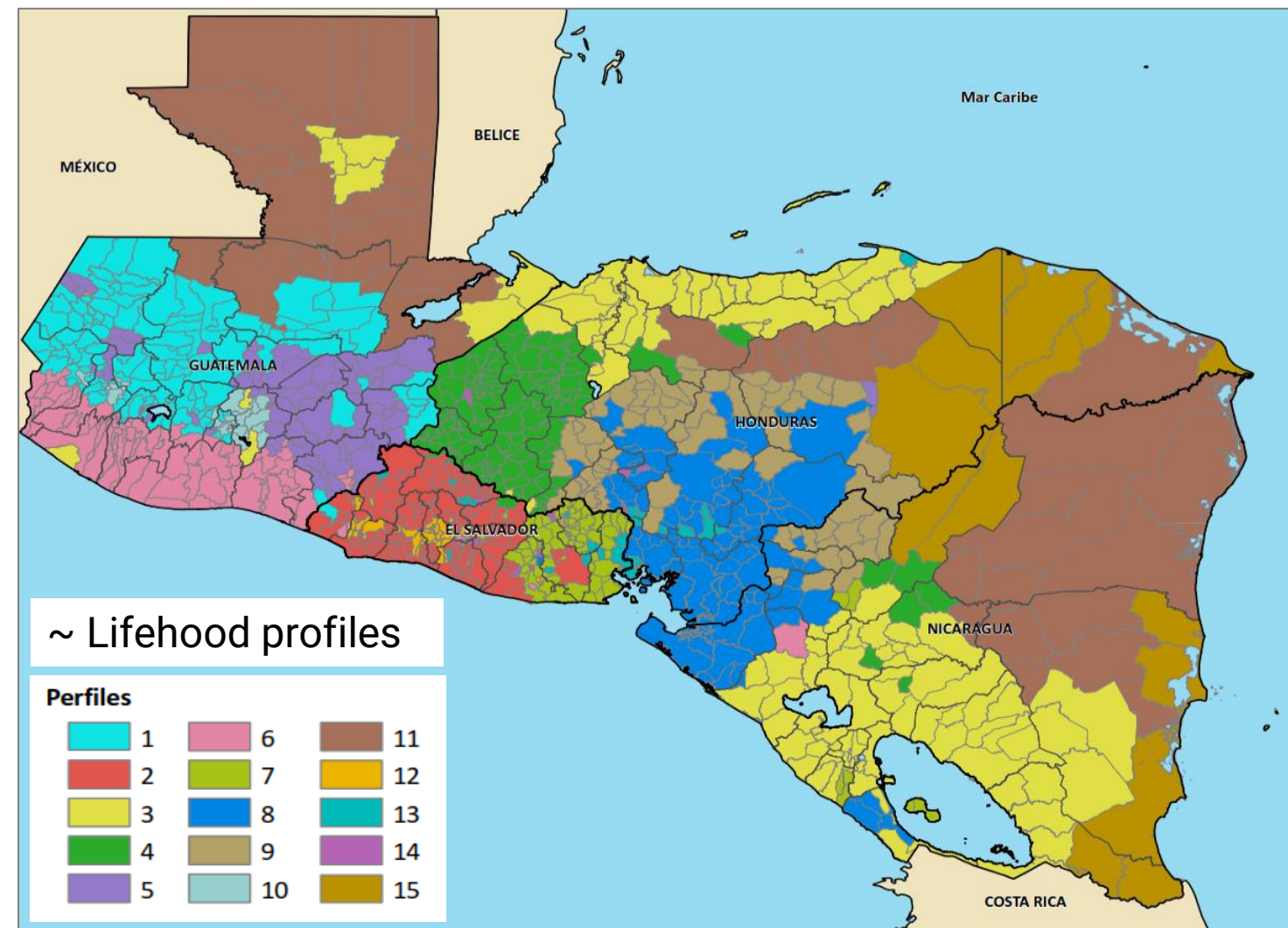
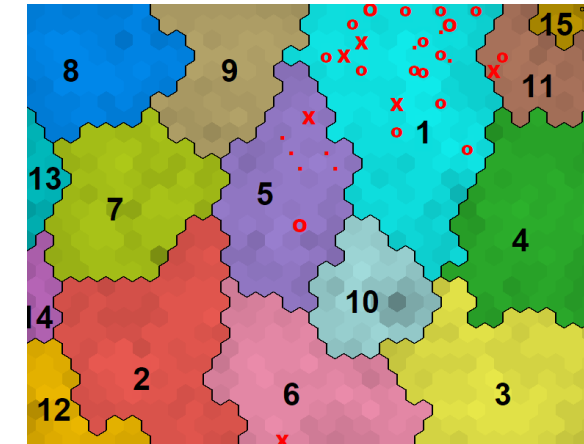


FIELD SURVEYS



TERRITORIAL PATTERNS

APPROX.
200 VARIABLES
1053 MUNICIPALITIES



3. METHODOLOGY of PREDISAN

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SATELLITE IMAGERY
REMOTE SENSING
AGRO-CLIMATIC MODEL



FIELD
SURVEYS



PUBLIC DATA
SOURCES



TERRITORIAL
PATTERNS

3. METHODOLOGY of PREDISAN

REACHING BEYOND FOOD SECURITY FIELD DATA



SATELLITE IMAGERY
REMOTE SENSING
AGRO-CLIMATIC MODEL



FIELD SURVEYS



PUBLIC DATA SOURCES



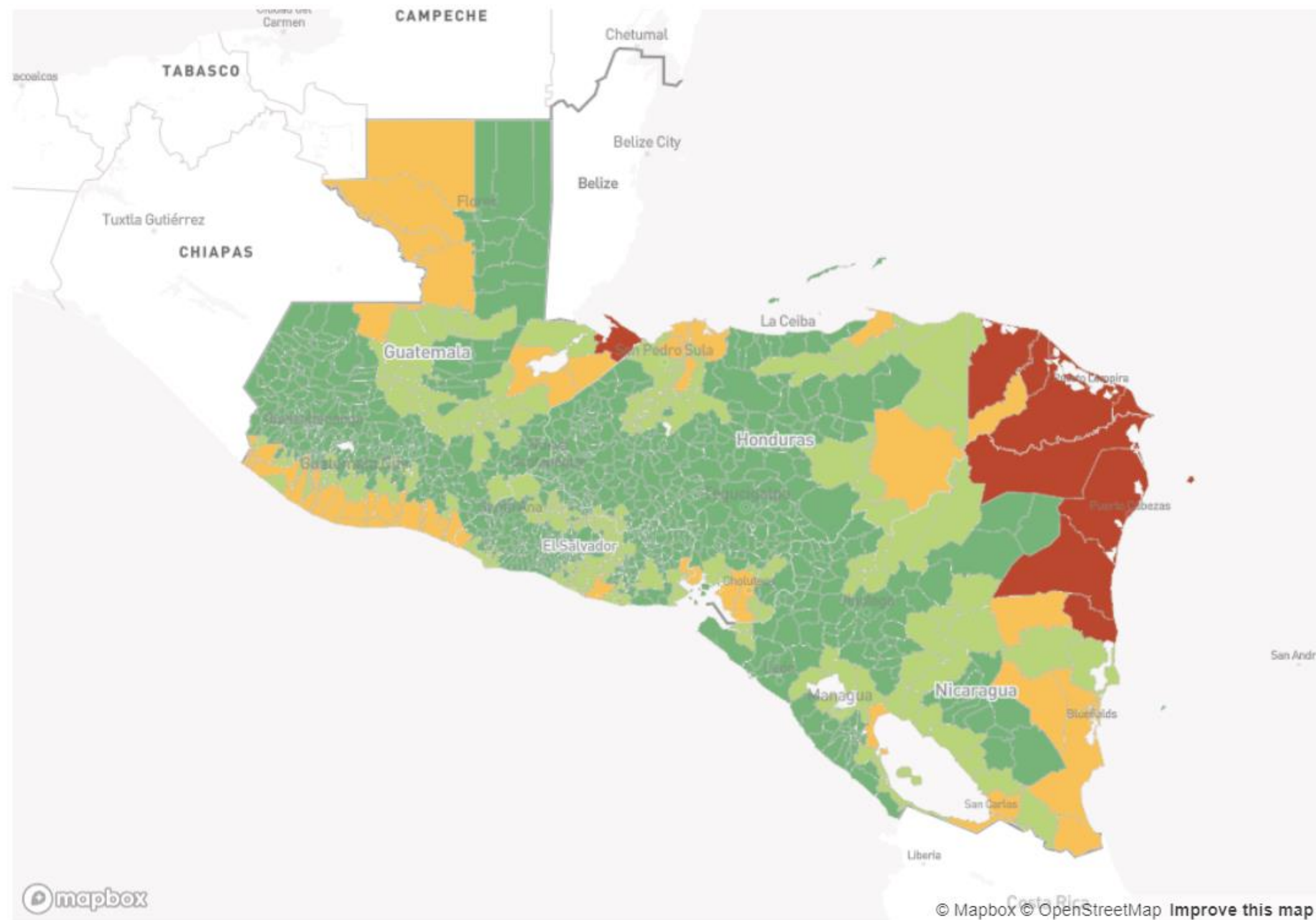
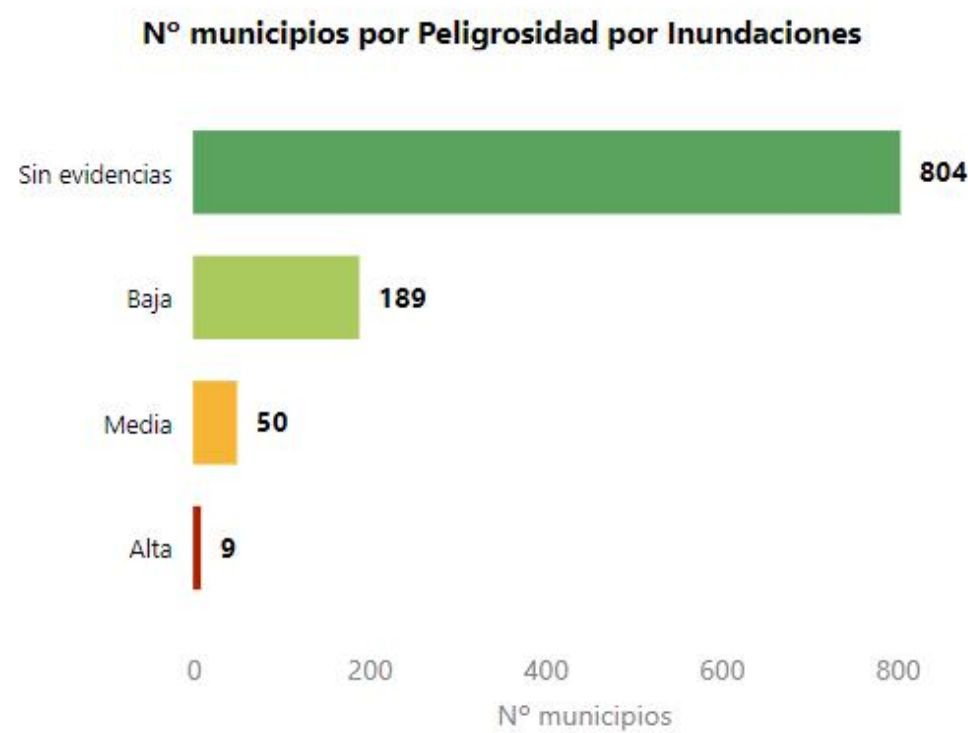
TERRITORIAL PATTERNS

NATURAL HAZARD: FLOODINGS



País
Todas

Departamento, Municipio
Todas



¿Cómo se calcula este indicador?

HDX
Descargar desde HDX



SOURCE: European Commission's Joint Research Centre Data Catalogue.
Deployed at PREDISAN by the platform authors

3. METHODOLOGY of PREDISAN

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SATELLITE IMAGERY
REMOTE SENSING
AGRO-CLIMATIC MODEL



FIELD
SURVEYS

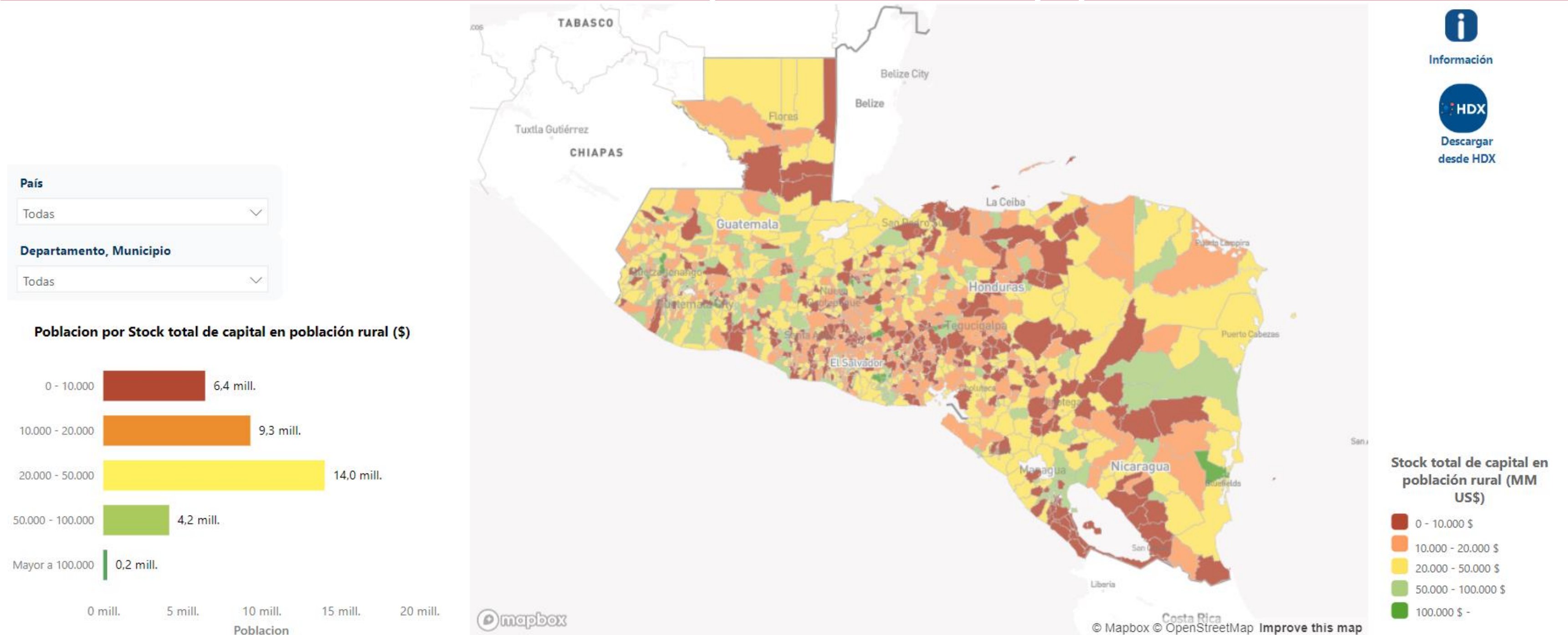


PUBLIC DATA
SOURCES



TERRITORIAL
PATTERNS

ECONOMY: Total capital stock in rural population



3. METHODOLOGY of PREDISAN

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SATELLITE IMAGERY
REMOTE SENSING
AGRO-CLIMATIC MODEL



FIELD SURVEYS



PUBLIC DATA SOURCES



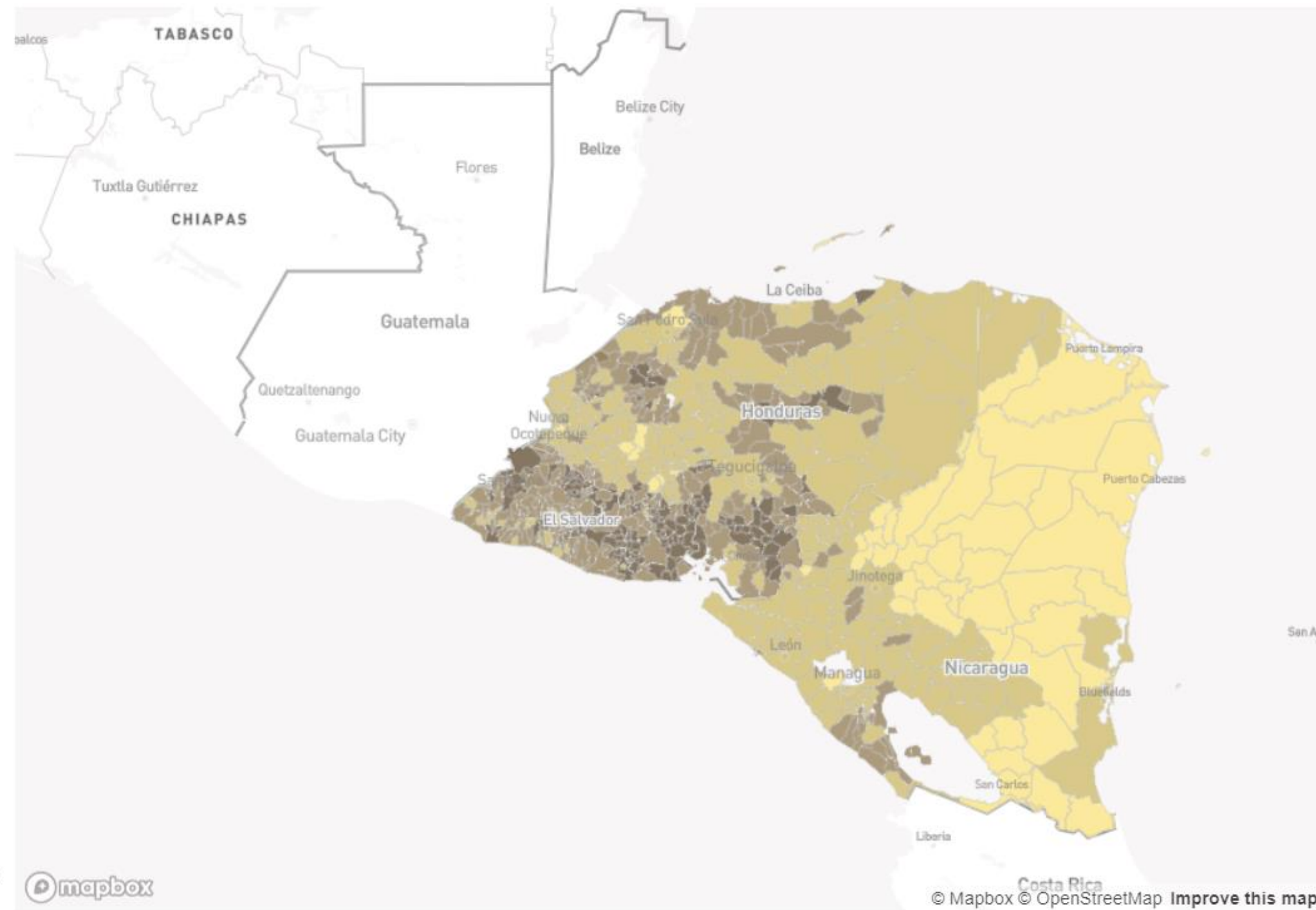
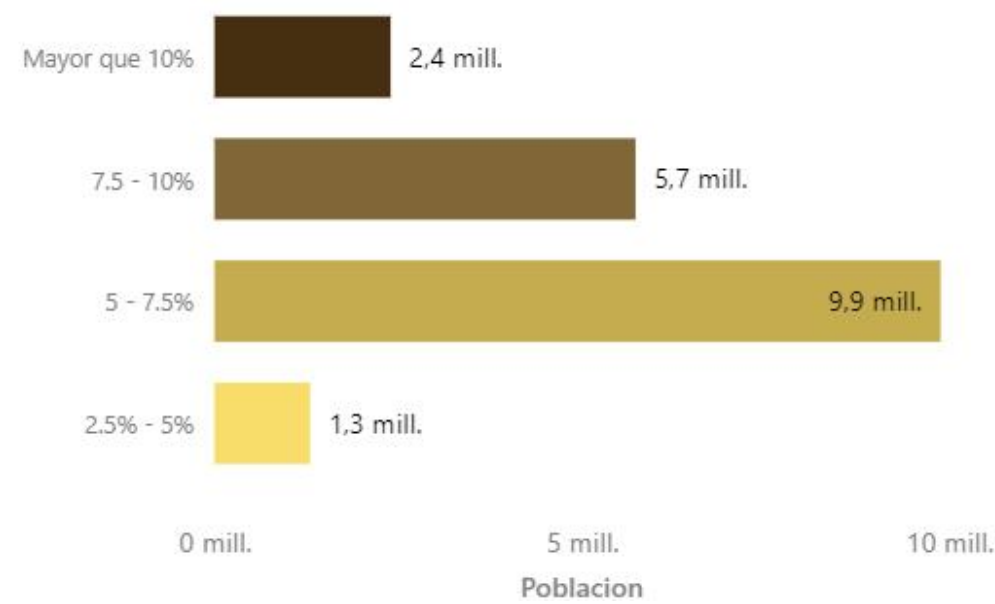
TERRITORIAL PATTERNS

SOCIAL: Percentage of elderly people

País
Todas

Departamento, Municipio
Todas

Poblacion por 3ª edad



[Información](#)

[HDX](#)
Descargar desde HDX



SOURCE: Facebook
Deployed at PREDISAN by the platform authors

3. METHODOLOGY of PREDISAN

REACHING BEYOND FOOD SECURITY FIELD DATA



SATELLITE IMAGERY
REMOTE SENSING
AGRO-CLIMATIC MODEL



FIELD
SURVEYS



PUBLIC DATA
SOURCES




NEW LAYERS
BASED ON
PUBLIC DATA
SOURCES



TERRITORIAL
PATTERNS

3. METHODOLOGY of PREDISAN

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SATELLITE IMAGERY
REMOTE SENSING
AGRO-CLIMATIC MODEL



FIELD SURVEYS



PUBLIC DATA SOURCES

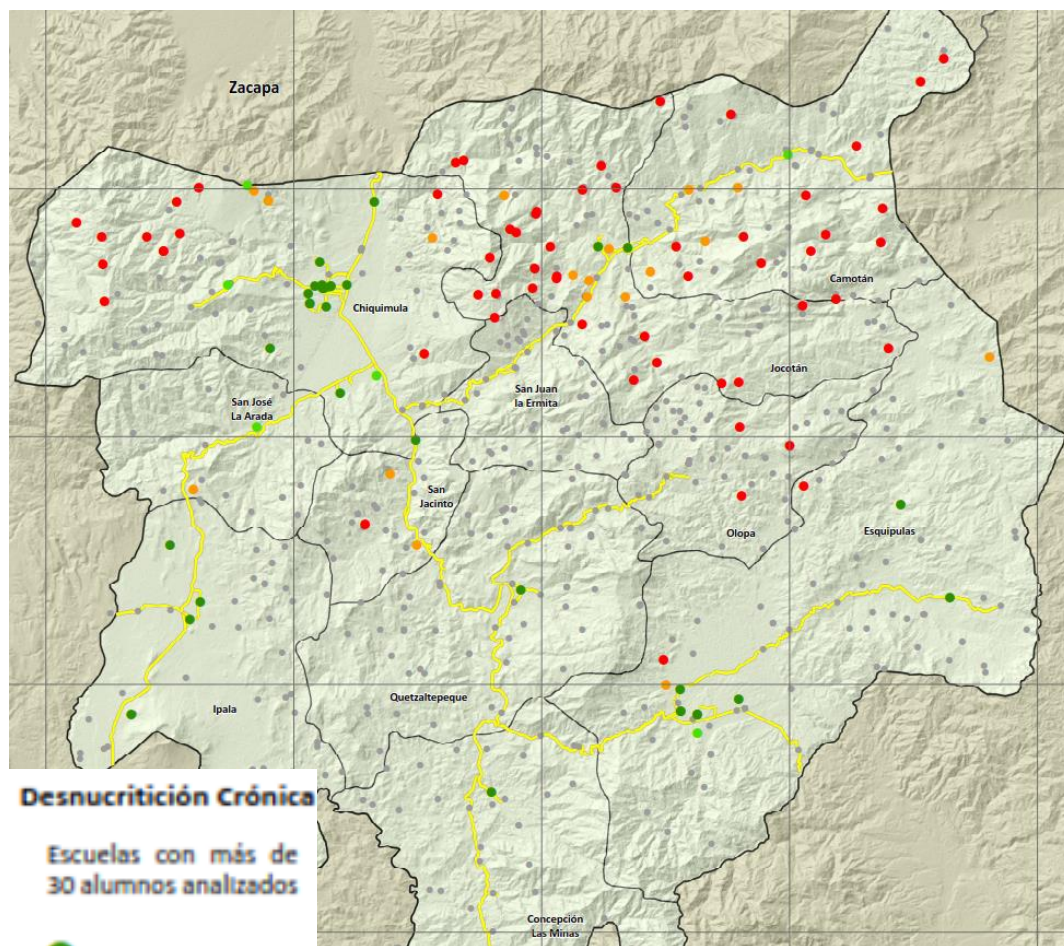


NEW LAYERS
BASED ON
PUBLIC DATA
SOURCES

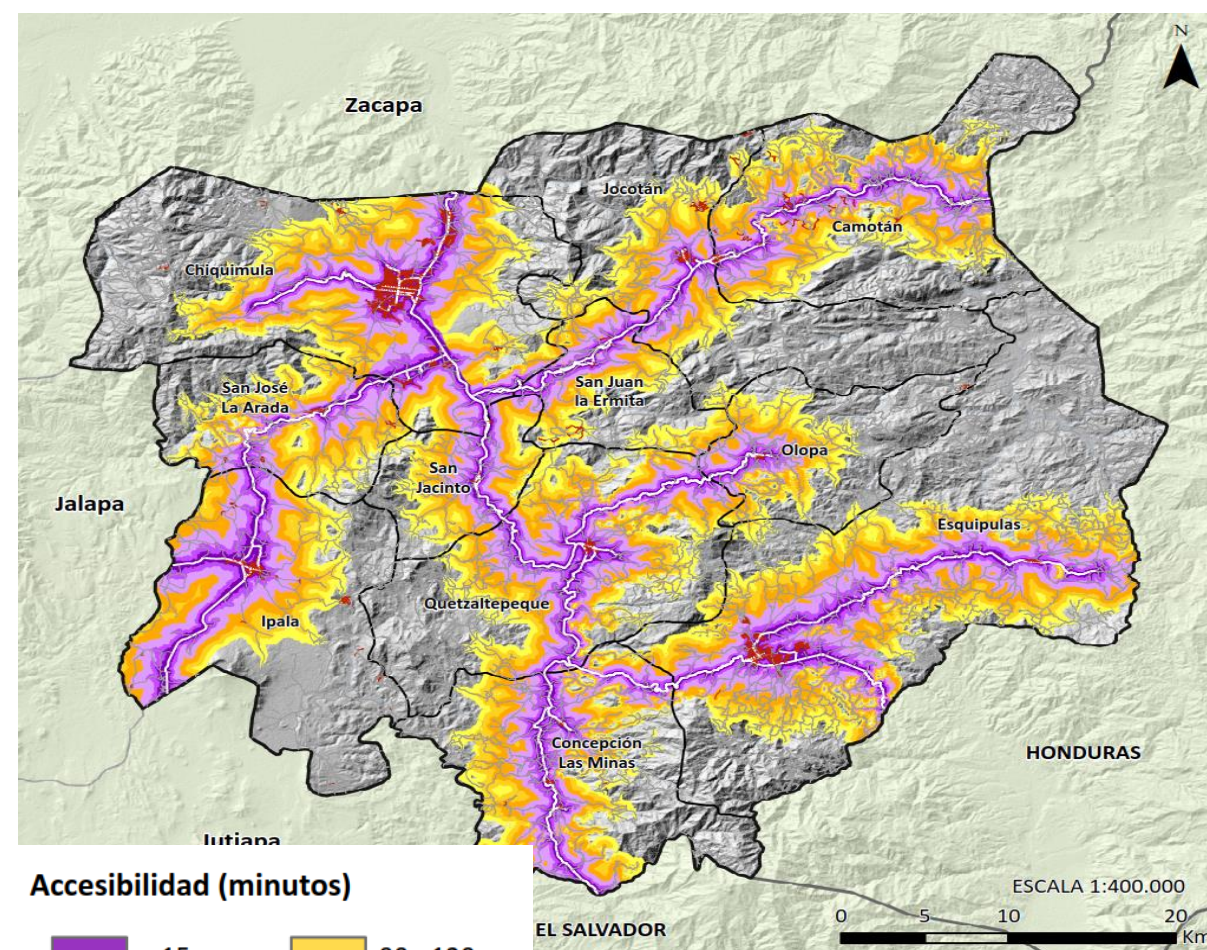


TERRITORIAL PATTERNS

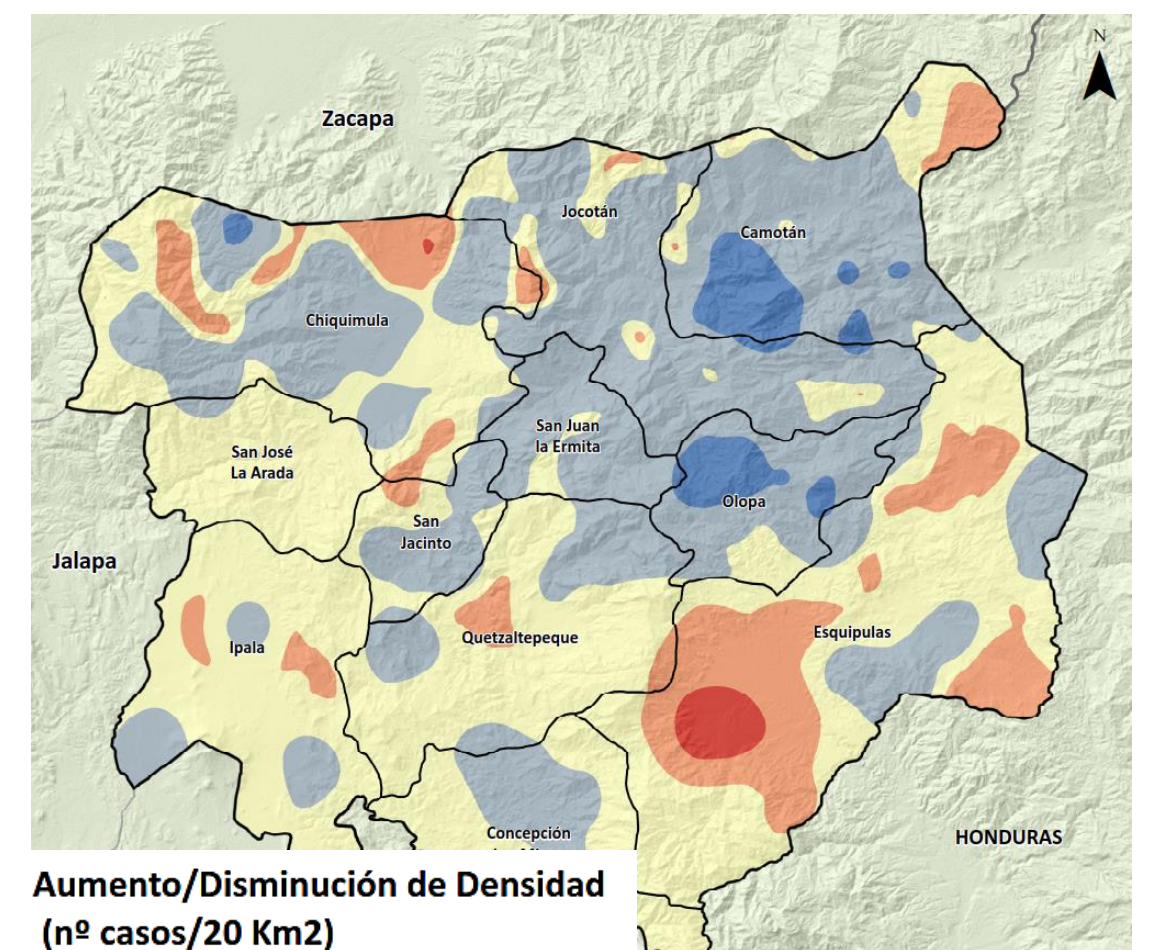
ACCESSIBILITY



Correlation of Chronic Malnutrition and distance to main roads in Chiquimula (Guatemala)




Road accessibility in Chiquimula (Guatemala)



Spatial Evolution of Acute Malnutrition 2015-2018 in Chiquimula (Guatemala)

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REMOTE SENSING
AGRO-CLIMATIC MODEL



FIELD
SURVEYS



PUBLIC DATA
SOURCES

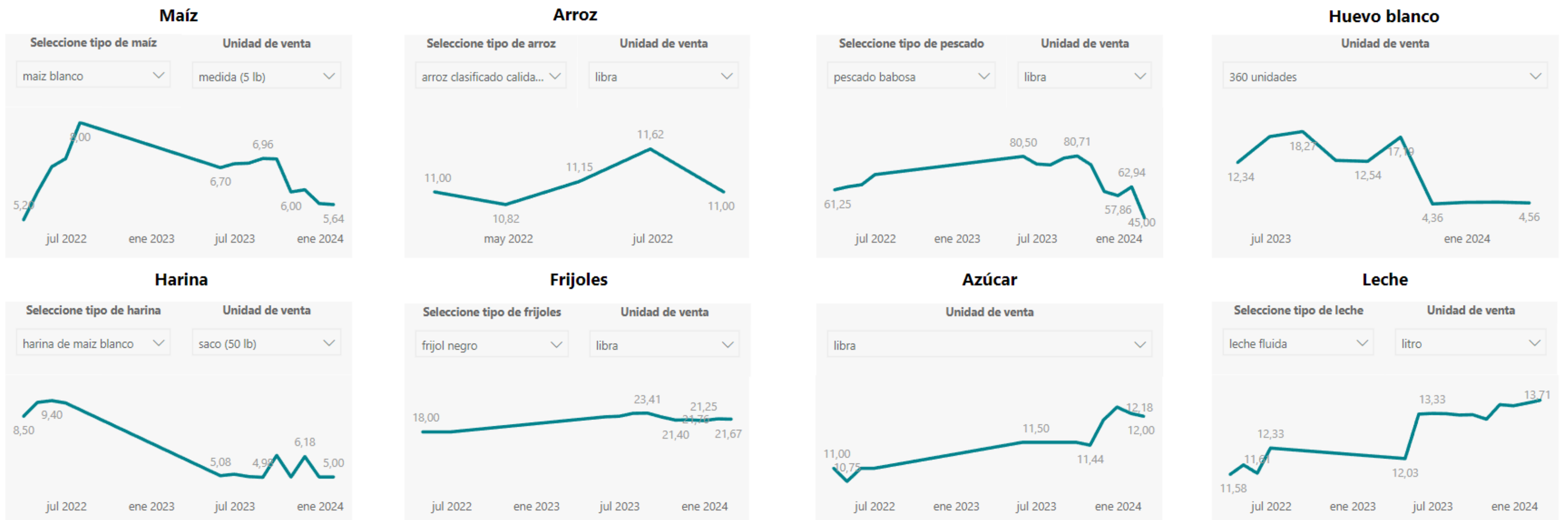


NEW LAYERS
BASED ON
PUBLIC DATA
SOURCES



TERRITORIAL
PATTERNS


FOOD PRICES



Food Price (in dollars) daily monitoring in markets and supermarkets in Honduras

3. METHODOLOGY of PREDISAN

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SATELLITE IMAGERY
REMOTE SENSING
AGRO-CLIMATIC MODEL



FIELD SURVEYS



PUBLIC DATA SOURCES

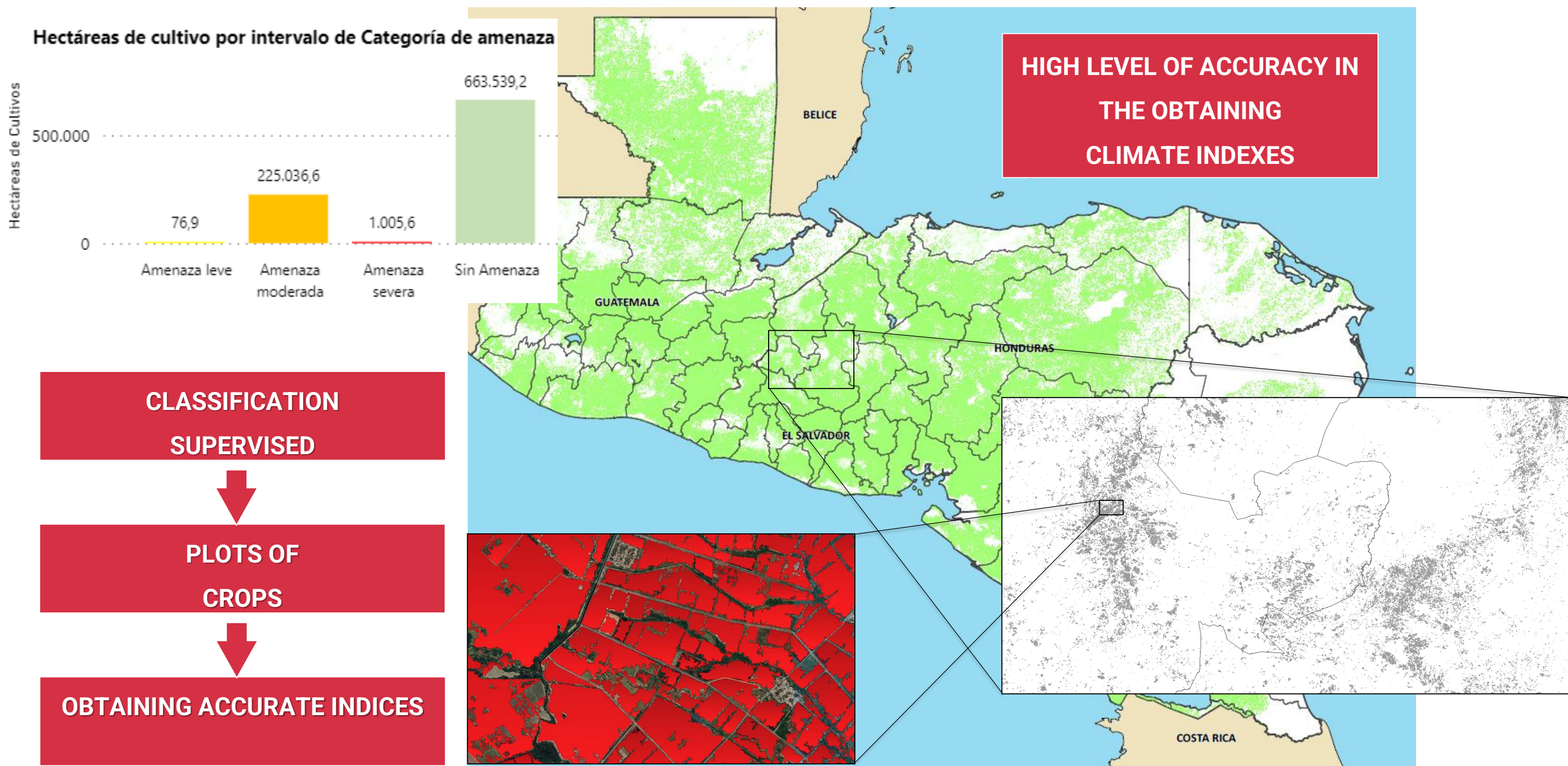


NEW LAYERS
BASED ON
PUBLIC DATA
SOURCES



TERRITORIAL PATTERNS

ACTUAL CROPS ON EACH PLOT




Obtaining indicators at the level of the crop plot in Central America

SOURCE: Prepared by PREDISAN platform authors

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SATELLITE IMAGERY
REMOTE SENSING
AGRO-CLIMATIC MODEL



FIELD SURVEYS



PUBLIC DATA SOURCES



NEW LAYERS
BASED ON
PUBLIC DATA
SOURCES



TERRITORIAL PATTERNS

VIOLENCE

Tipo de evento

- Seleccionar todo
- Alerta temprana
- Disturbios
- Enfrentamientos
- Explosiones/Violencia remota
- Protestas

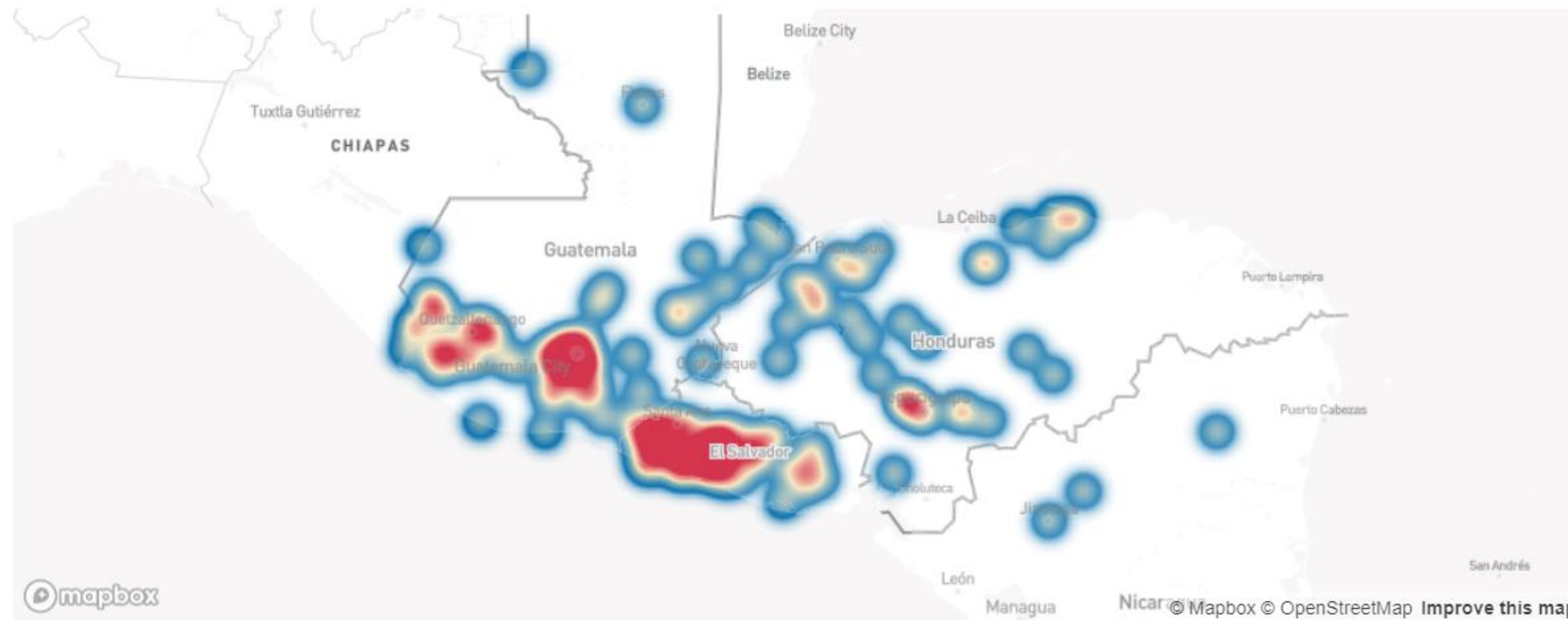
País

Todas

Departamento, Municipio

Todas

19/12/2021 08/12/2023

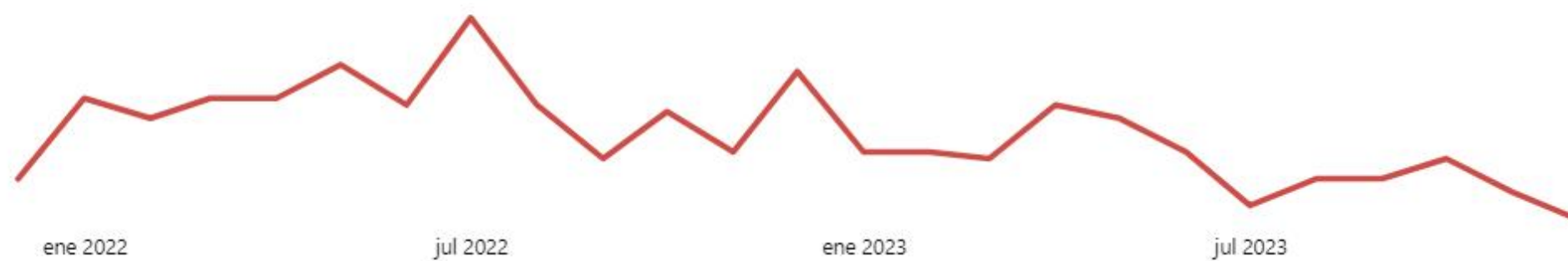


Información

Mostrar Explicación tipos de Eventos

Cantidad de eventos por Mes

Eventos ● Enfrentamientos




Cantidad de eventos por País

Eventos	El Salvador	Guatemala	Honduras	Nicaragua	Total
Enfrentamientos	77	186	75	4	342
Total	77	186	75	4	342

3. METHODOLOGY of PREDISAN

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SATELLITE IMAGERY
REMOTE SENSING
AGRO-CLIMATIC MODEL



FIELD SURVEYS



PUBLIC DATA SOURCES



NEW LAYERS
BASED ON
PUBLIC DATA
SOURCES



TERRITORIAL PATTERNS

MASS MEDIA

Nº de Medios rastreadas	Nº de Noticias clasificadas	Nº total de Noticias rastreadas
31	125	362,34 mil
Click para ver medios de comunicación	Click para ver noticias	Total de links rastreados

Medios de Comunicación

País, Departamento, Municipio: Todas

Categoría: Riesgos Naturales

- Seleccionar todo
- Ayuda
- Migración
- Política
- Riesgos Naturales
- Servicios Públicos
- Violencia
- Violencia de Género

Fecha: 2023




Categoría	Enlace Web
Riesgos Naturales	https://www.republica.gt/guatemala/recomendaciones-ante-los-sismos-en-guatemala--20231212170
Riesgos Naturales	https://www.republica.gt/guatemala/conred-mantiene-habilitados-ocho-albergues-por-bajas-temperaturas-continuan-es-diciembre-20231217480
Riesgos Naturales	https://www.quepasa.com.ve/internacionales/sismo-de-magnitud-58-sacudio-guatemala-sin-muchas-consecuencias/#respond



Media monitoring in Central America

3. METHODOLOGY of PREDISAN

REACHING BEYOND FOOD SECURITY FIELD DATA



SATELLITE IMAGERY
REMOTE SENSING
AGRO-CLIMATIC MODEL



FIELD
SURVEYS



PUBLIC DATA
SOURCES

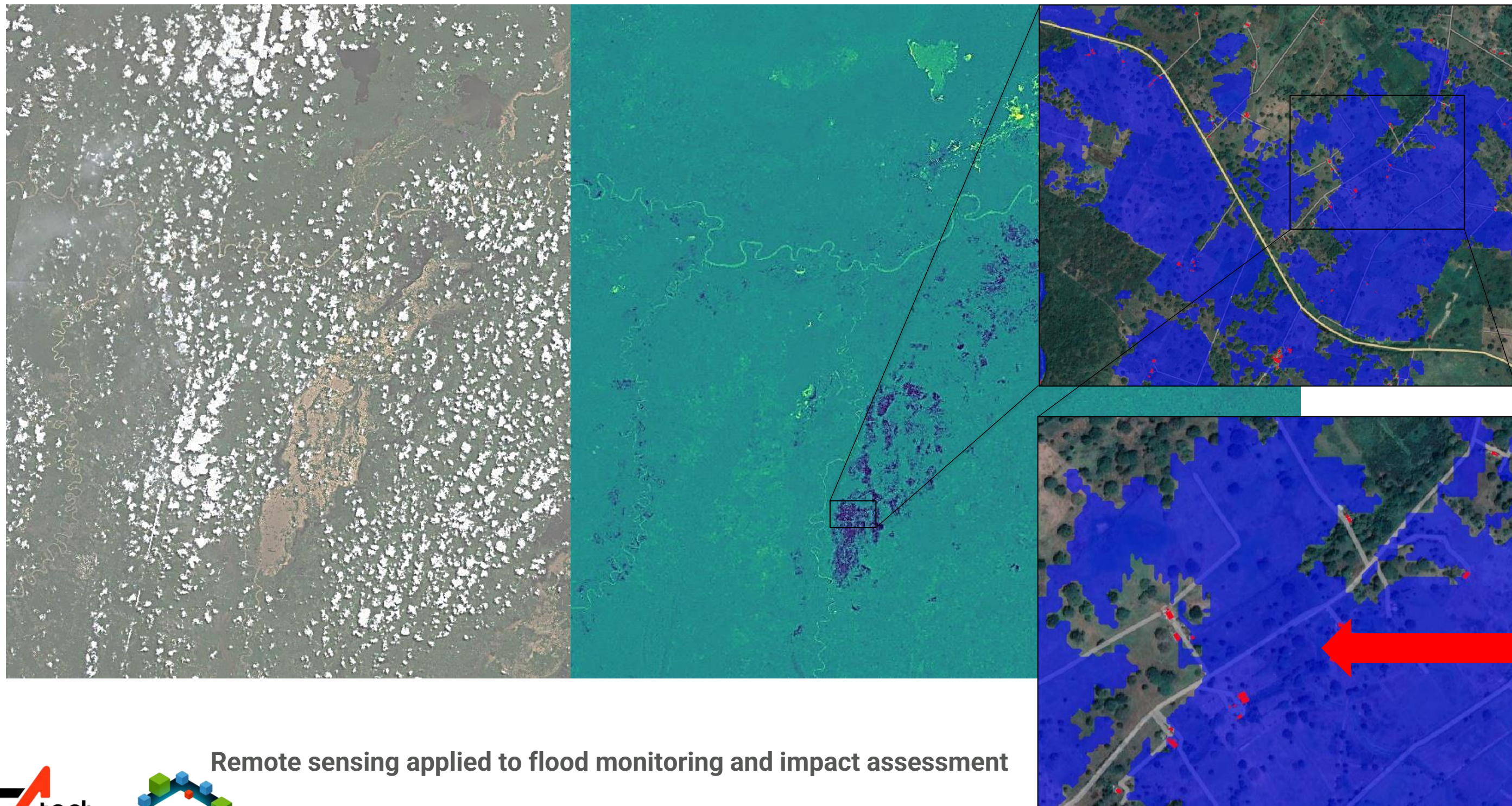


NEW LAYERS
BASED ON
PUBLIC DATA
SOURCES




TERRITORIAL
PATTERNS

WEATHER HISTORICAL EVENTS



3. METHODOLOGY of PREDISAN

REACHING BEYOND FOOD SECURITY FIELD DATA




SATELLITE IMAGERY
REMOTE SENSING
AGRO-CLIMATIC MODEL



FIELD SURVEYS



PUBLIC DATA SOURCES



NEW LAYERS
BASED ON
PUBLIC DATA
SOURCES



TERRITORIAL PATTERNS

MIGRATIONS

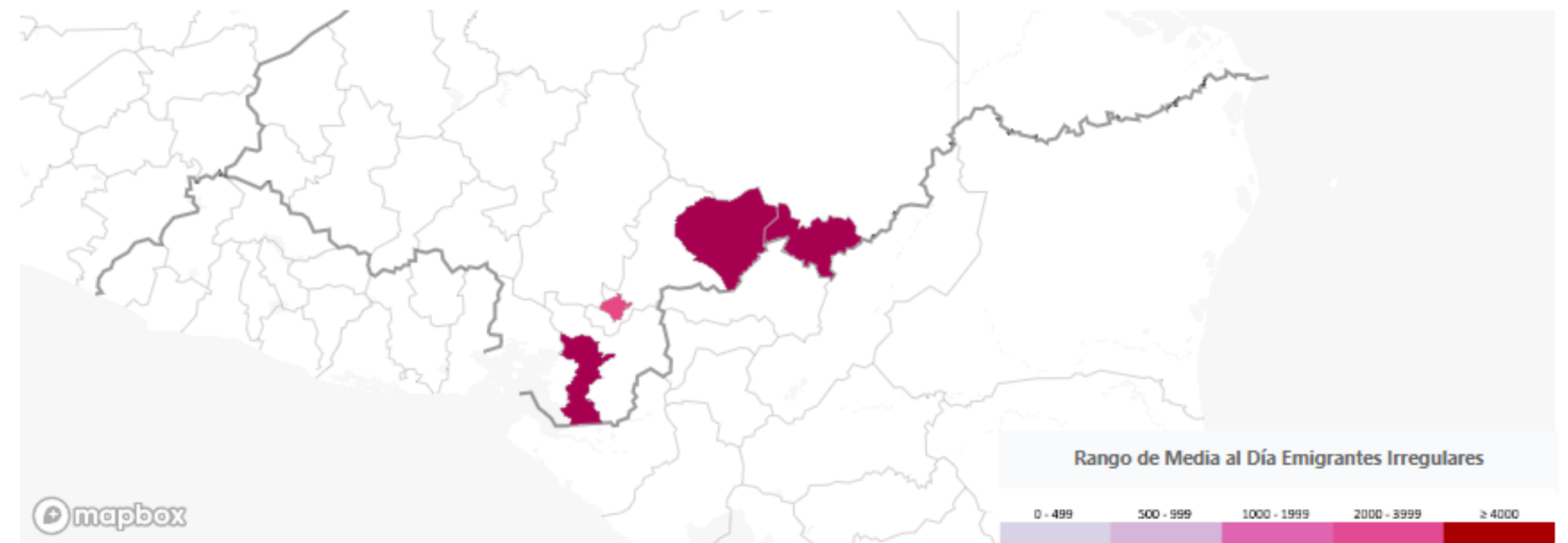
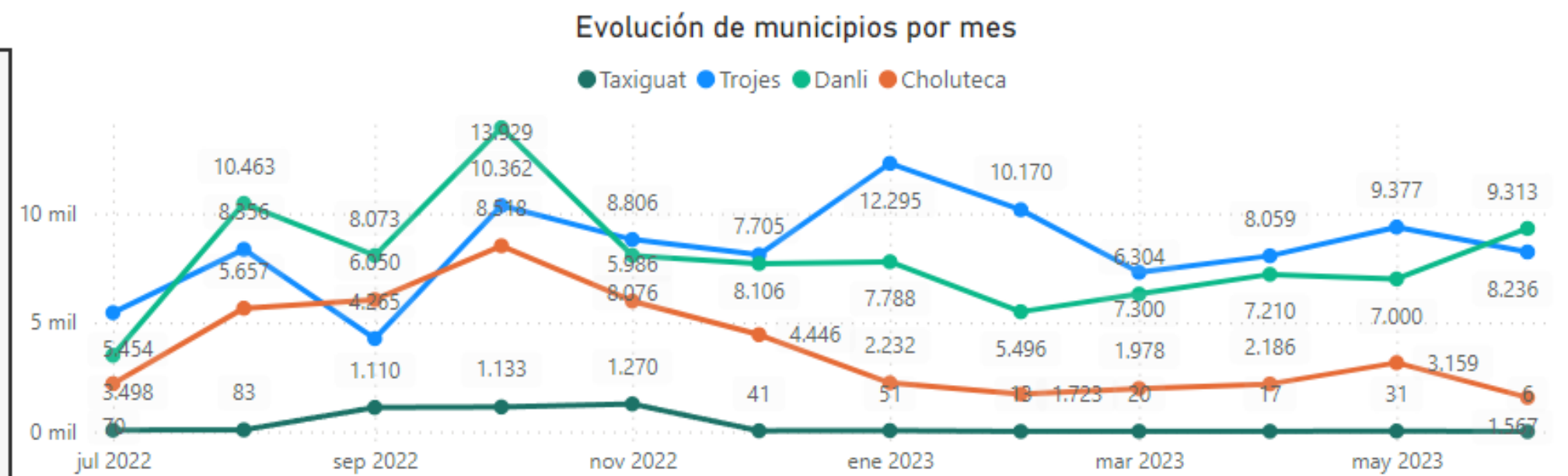
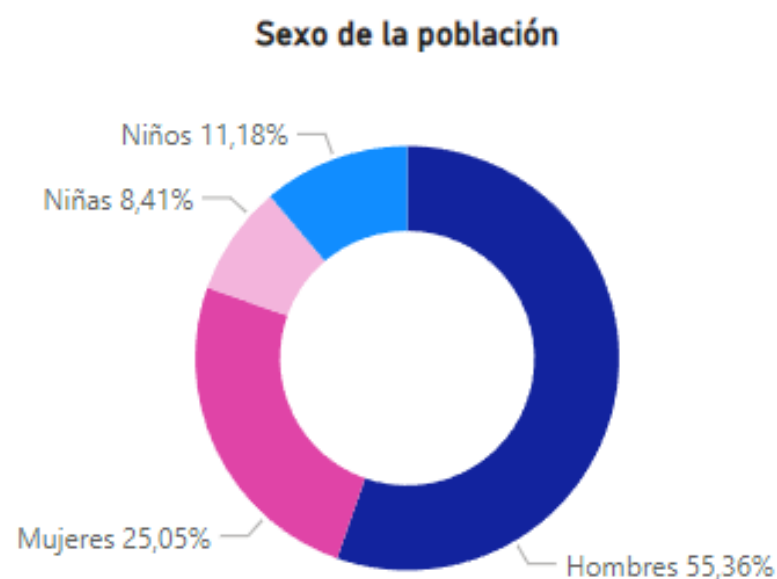
Ir a menú migraciones Honduras

04/07/2022 30/06/2023

Semana
Todas


Total de migrantes:
245.189

	Irregulares	Media/Diaria
Total	245.189	49.037,8
Choluteca	45.703	9.140,6
Danli	94.855	18.971,0
Texiguat	3.845	769,0
Trojes	100.786	20.157,2



3. METHODOLOGY of PREDISAN

REACHING BEYOND FOOD SECURITY FIELD DATA



SATELLITE IMAGERY
REMOTE SENSING
AGRO-CLIMATIC MODEL



FIELD
SURVEYS



PUBLIC DATA
SOURCES

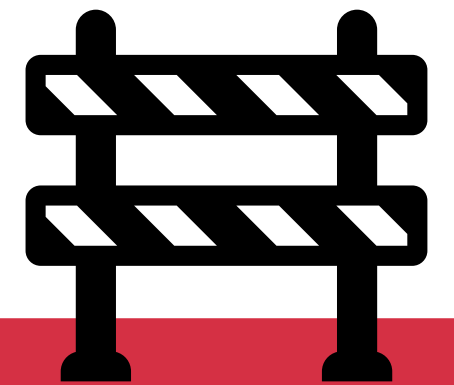
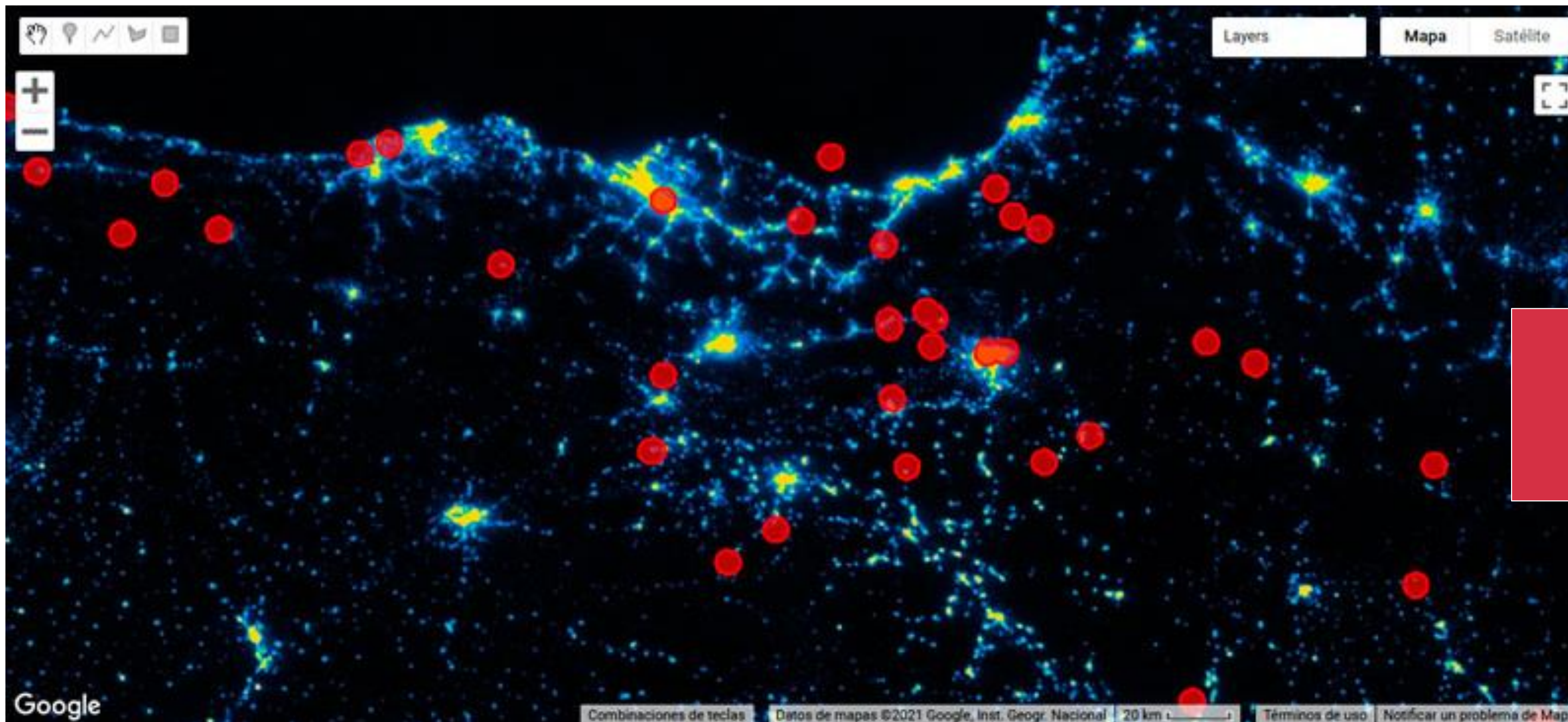


NEW LAYERS
BASED ON
PUBLIC DATA
SOURCES



TERRITORIAL
PATTERNS

NIGHT LIGHTS (MEASURE HUMAN ACTIVITIES)

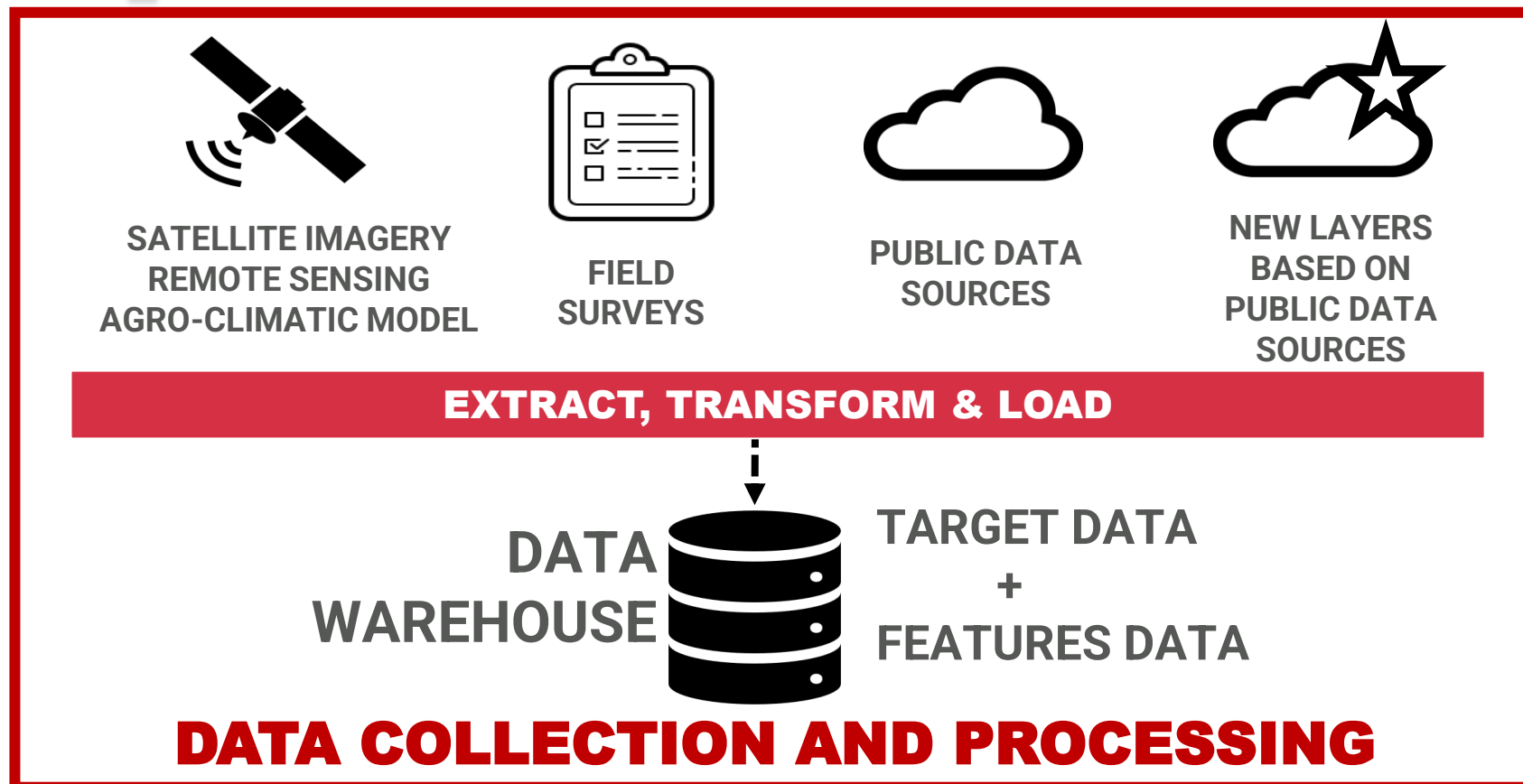


WORKING IN
PROGRESS

SOURCE: Prepared by PREDISAN platform authors

3. METHODOLOGY of PREISAN

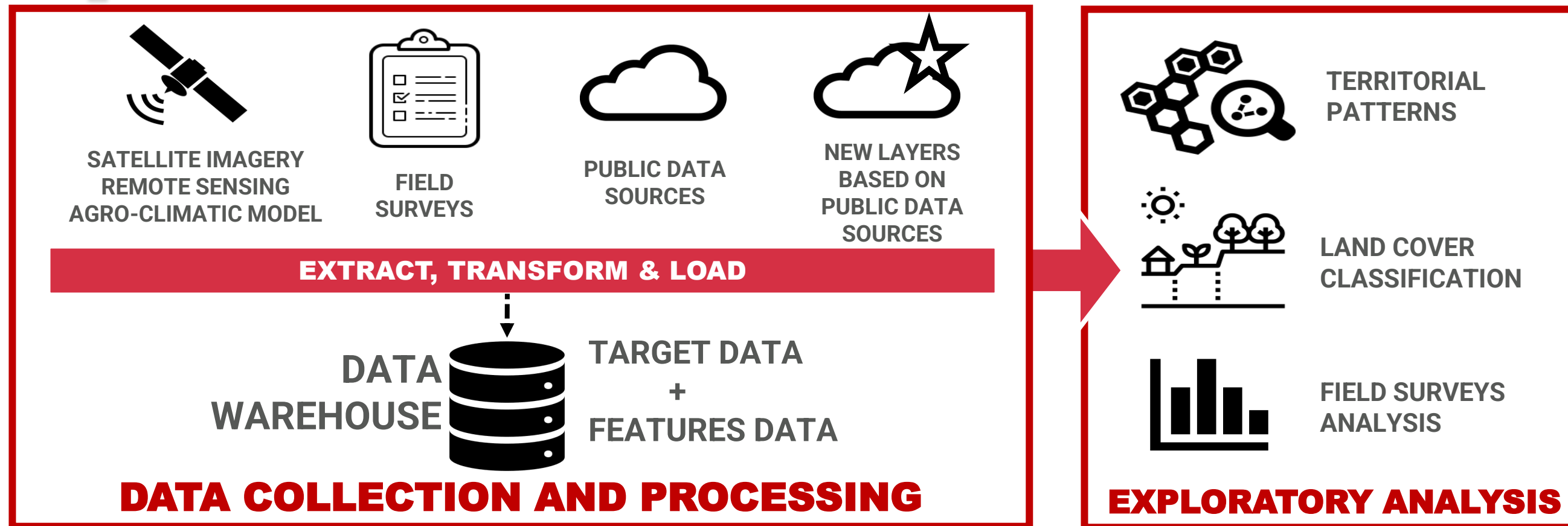
REACHING BEYOND FOOD SECURITY FIELD DATA



TERRITORIAL
PATTERNS

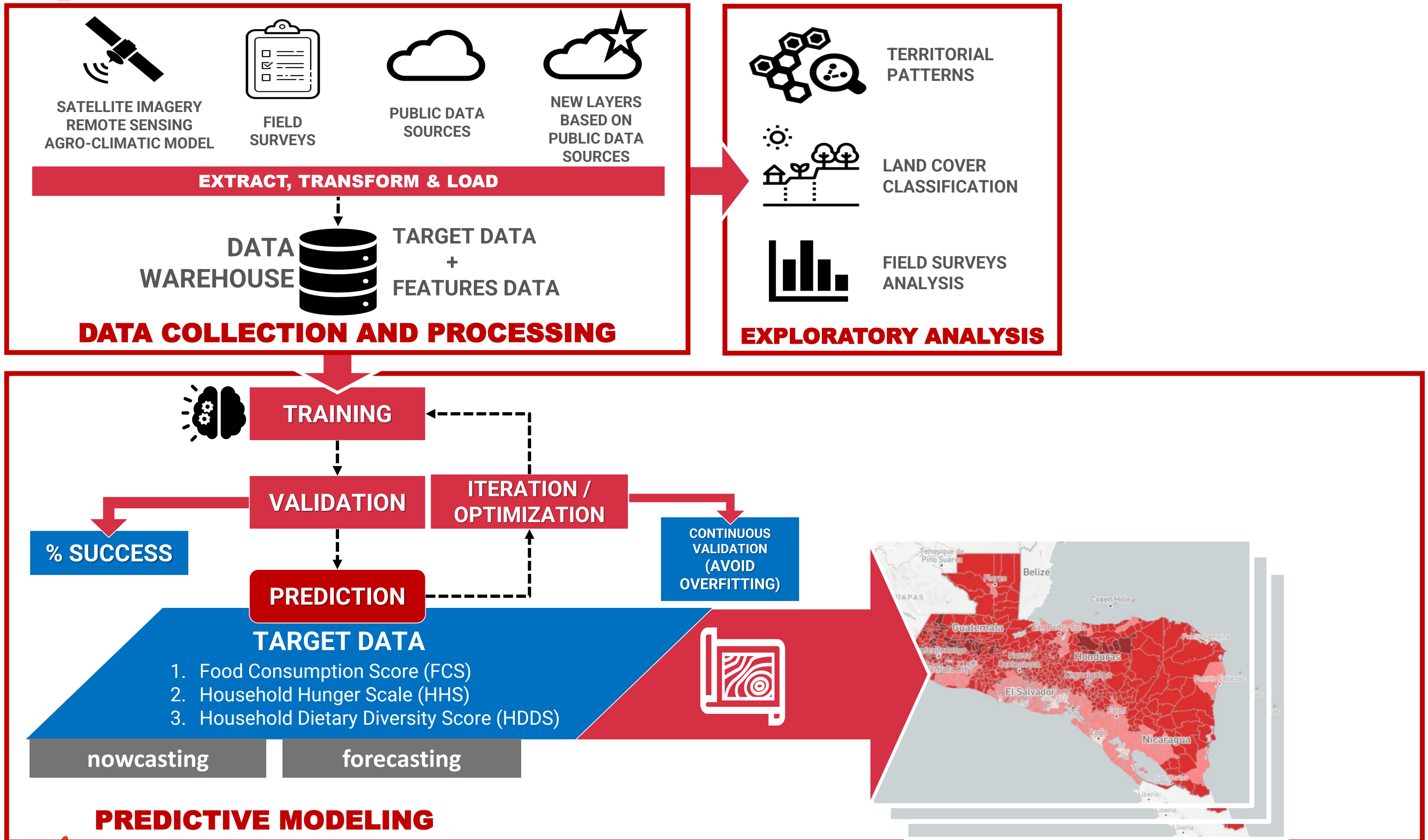
3. METHODOLOGY of PRESISAN

REACHING BEYOND FOOD SECURITY FIELD DATA



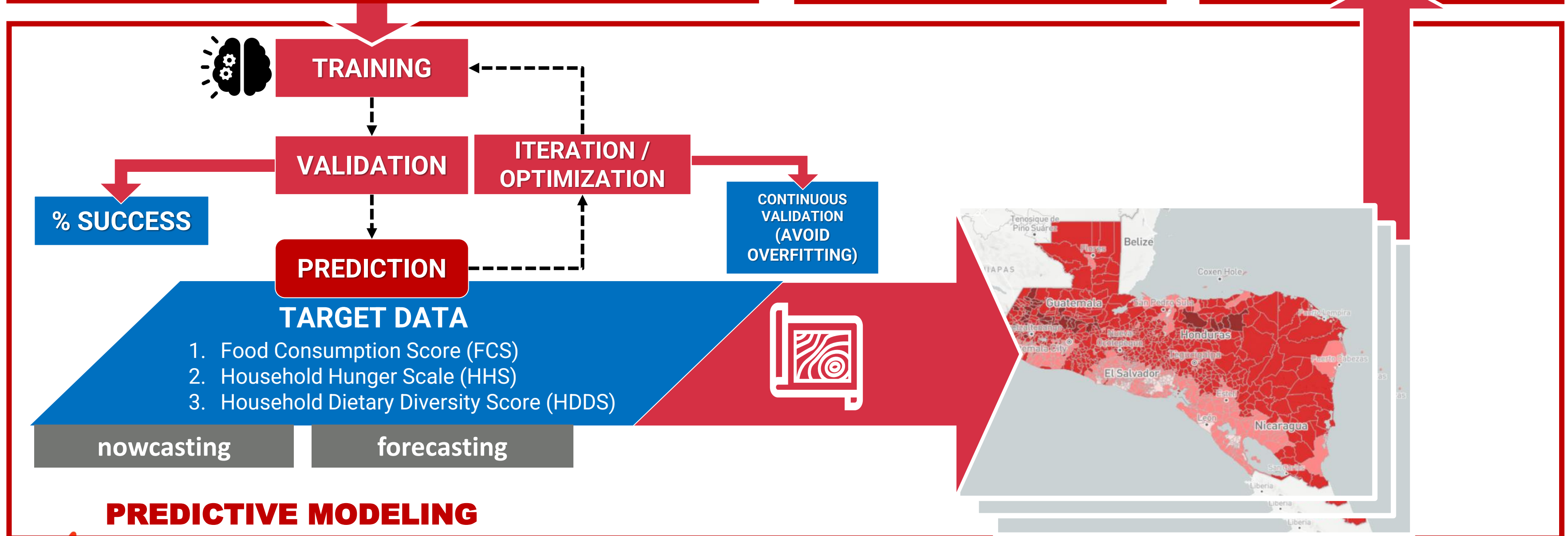
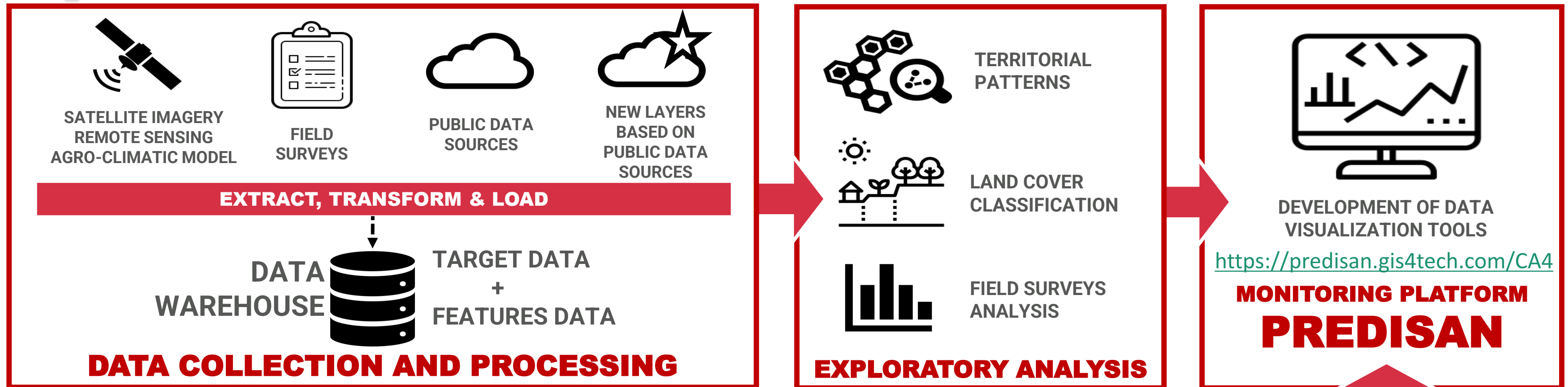
3. METHODOLOGY of PRESISAN

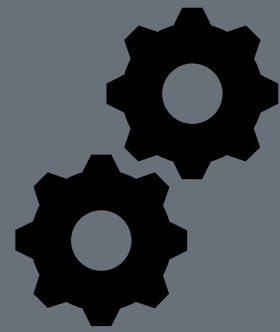
REACHING BEYOND FOOD SECURITY FIELD DATA



3. METHODOLOGY of PRESISAN

REACHING BEYOND FOOD SECURITY FIELD DATA





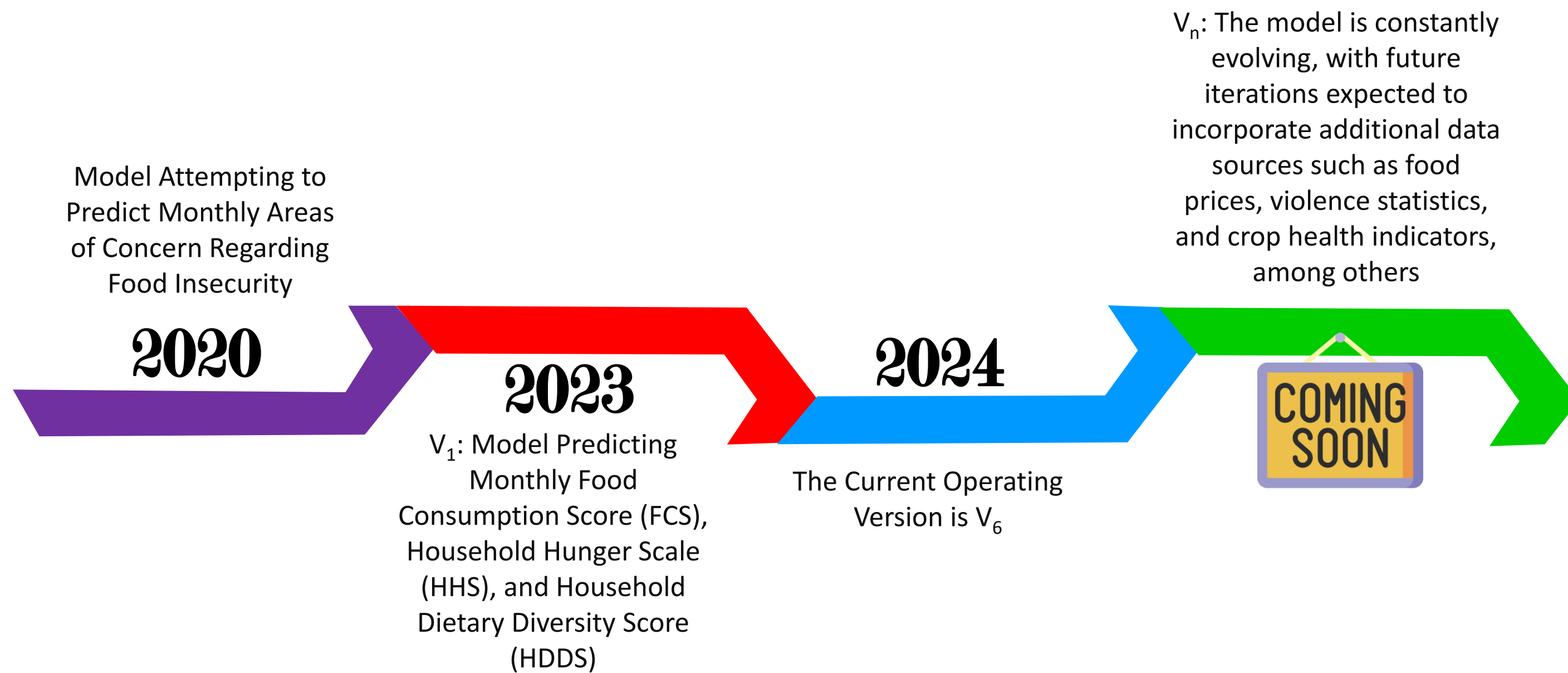
4. PREDICTIVE MODELS

THE MODEL IN DETAIL

4. PREDICTIVE MODELS

VULNERABILITY MODEL EVOLUTION

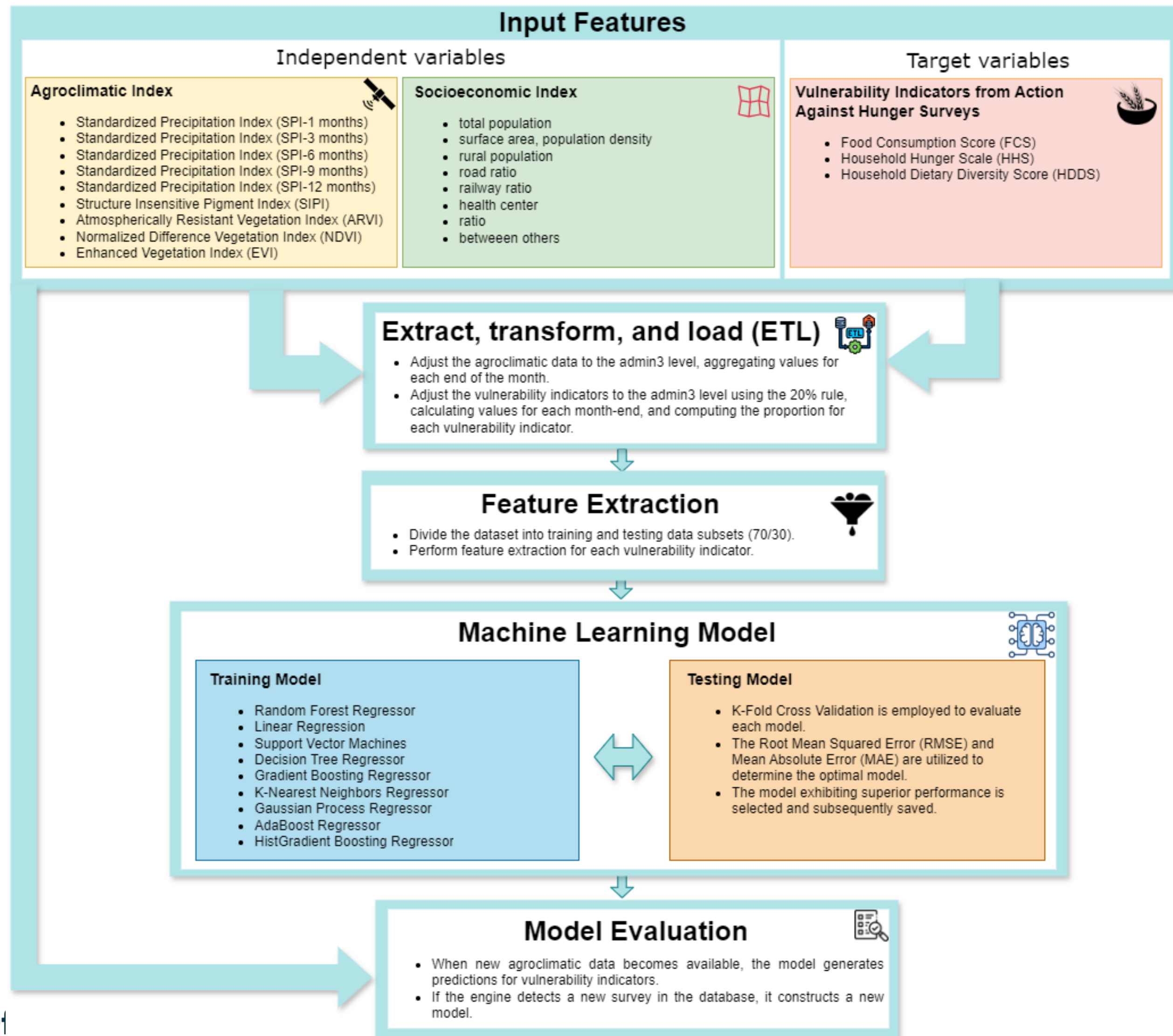
The machine learning model has evolved through iterative learning and feedback received from humanitarian organizations. Since 2020, the model has undergone several modifications, as illustrated in Fig. X, to enhance its effectiveness and performance.



Timeline Depicting the Evolution of the Machine Learning Vulnerability Model Over Time

4. PREDICTIVE MODELS

CURRENT MODEL WORKFLOW

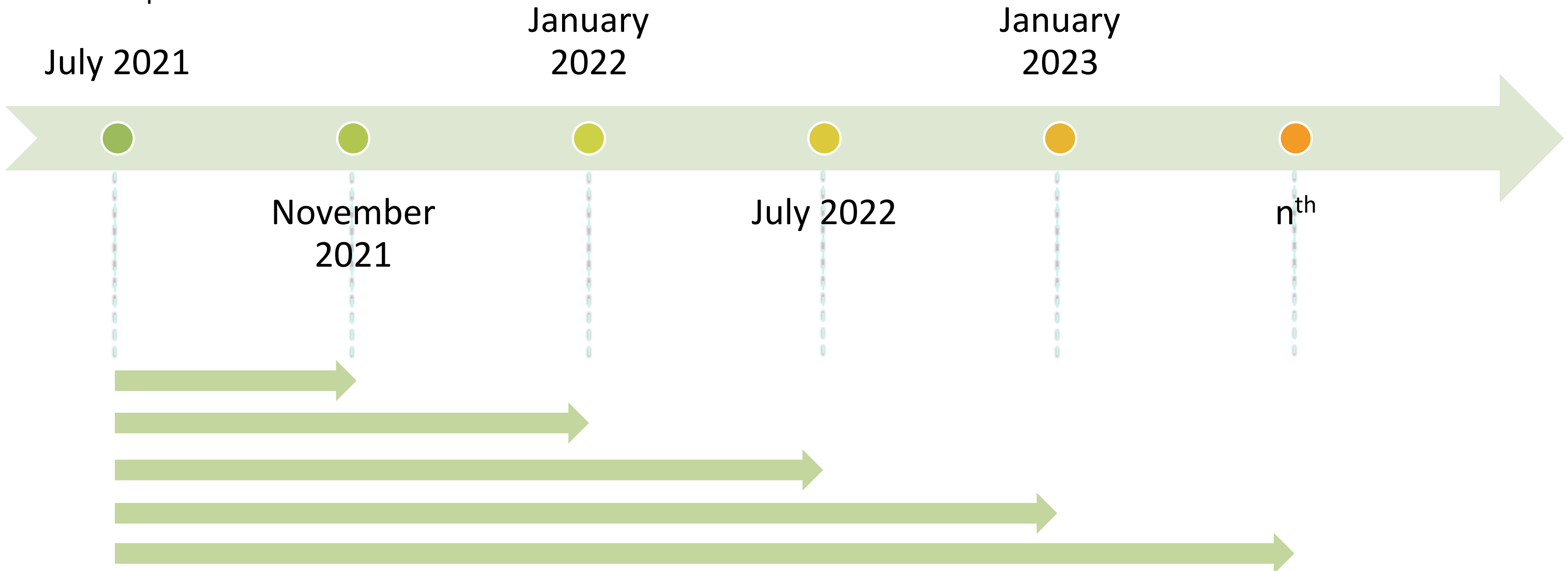


Streamlined Workflow for Machine Learning Model

4. PREDICTIVE MODELS

MACHINE LEARNING MODELS STRUCTURATION

The machine learning pipeline engine was developed with an additive approach, where it **constructs a new model upon detecting a new survey in the databases**, enabling predictions. Conversely, when the model identifies solely **new agroclimatic data**, the pipeline engine utilizes the **most recent model** to generate predictions. This process **operates automatically**, constantly monitoring for new surveys or agroclimatic data updates.

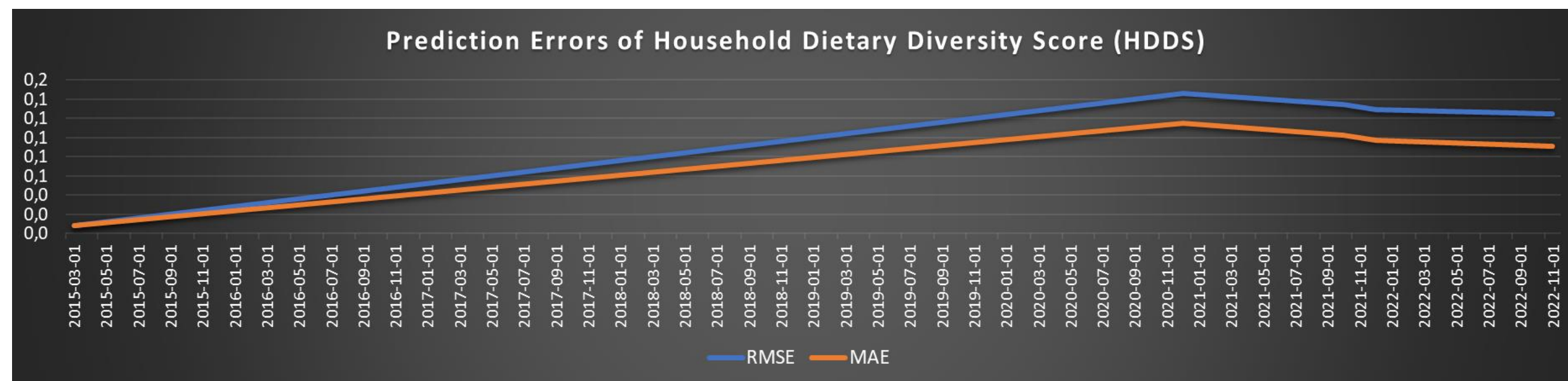
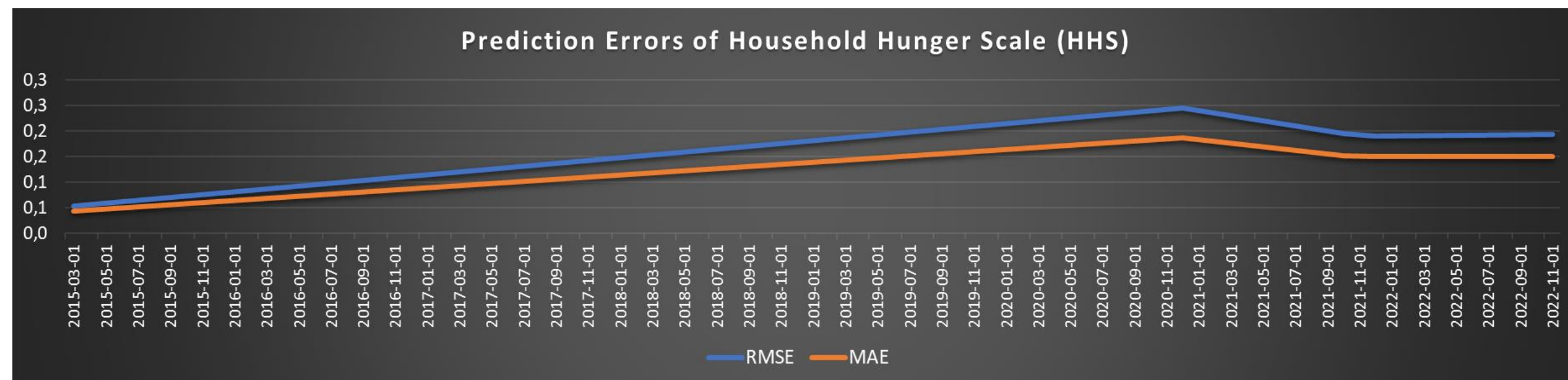
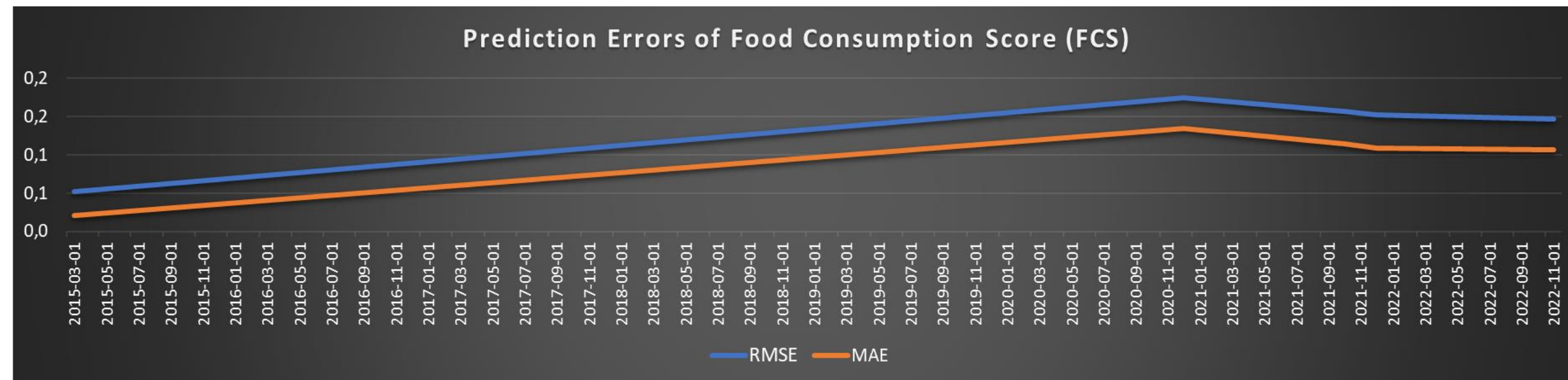


Streamlined Workflow for Machine Learning Model

4. PREDICTIVE MODELS

PERFORMANCE EVALUATION OF MACHINE LEARNING MODELS

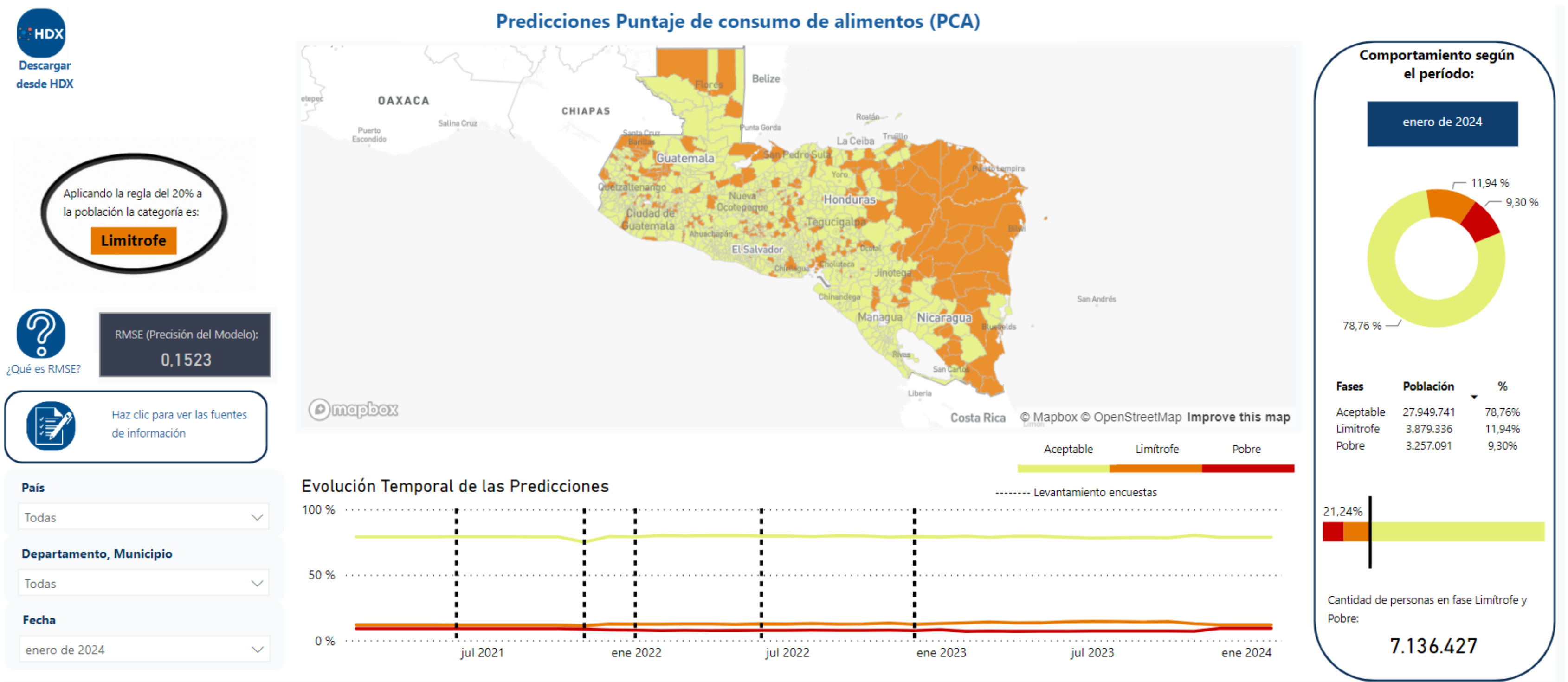
Based on the results obtained, the models exhibit RMSE and MAE ranging between 0.008 and 0.246. The HHS_2 Model yields the highest error, while the HDDS_1 model demonstrates the lowest error.



4. PREDICTIVE MODELS

PREDISAN PUBLIC DASHBOARD

The model predictions are contingent upon the availability of agroclimatic data. As a result, the model can be categorized as **Now-casting**, with a monthly periodicity. Predictions for the Food Consumption Score (FCS), Household Hunger Scale (HHS), and Household Dietary Diversity Score (HDDS) are displayed on a dashboard (<https://www.predisan.gis4tech.com/ca4>) and **accessible online 24/7**.

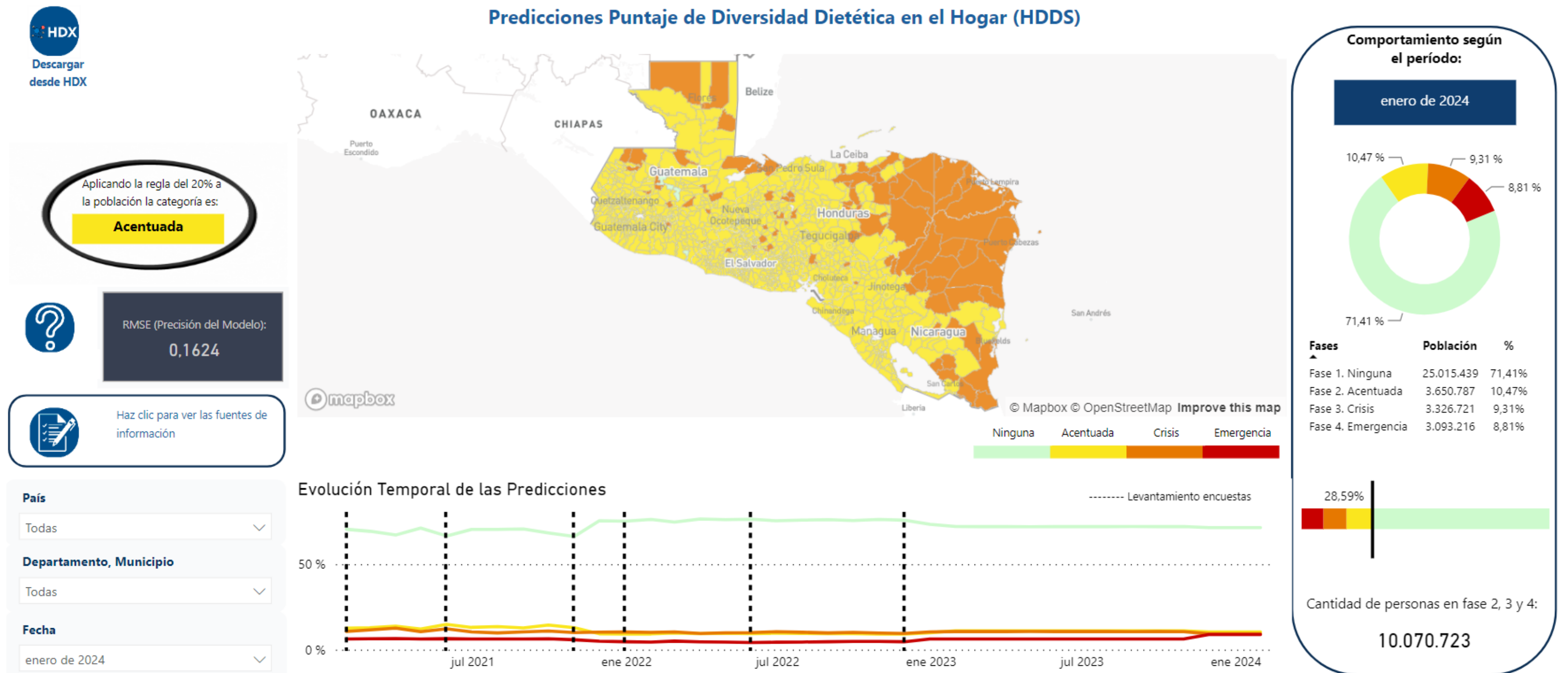


The image displays the predictions of Food Consumption Score (FCS) for the most recent months with available data

4. PREDICTIVE MODELS

PREDISAN PUBLIC DASHBOARD

The dashboard displays the vulnerability category for each municipality based on the selected indicator. Categories are assigned using the 20% rule. Additionally, temporal trends of predictions can be viewed at the bottom, while the left side provides insights into the population distribution across different vulnerability categories and the number of individuals classified under each.



The image displays the predictions of Household Dietary Diversity Score (HDDS) for the most recent months with available data



5. FUTURE IMPROVEMENTS

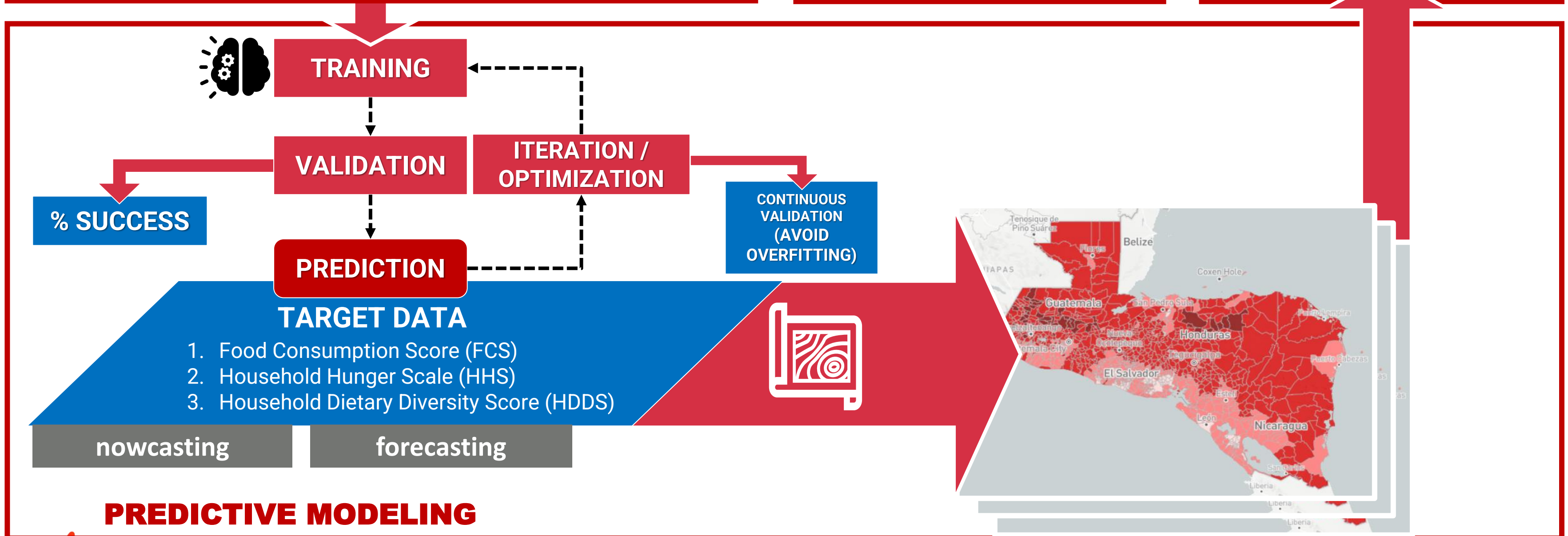
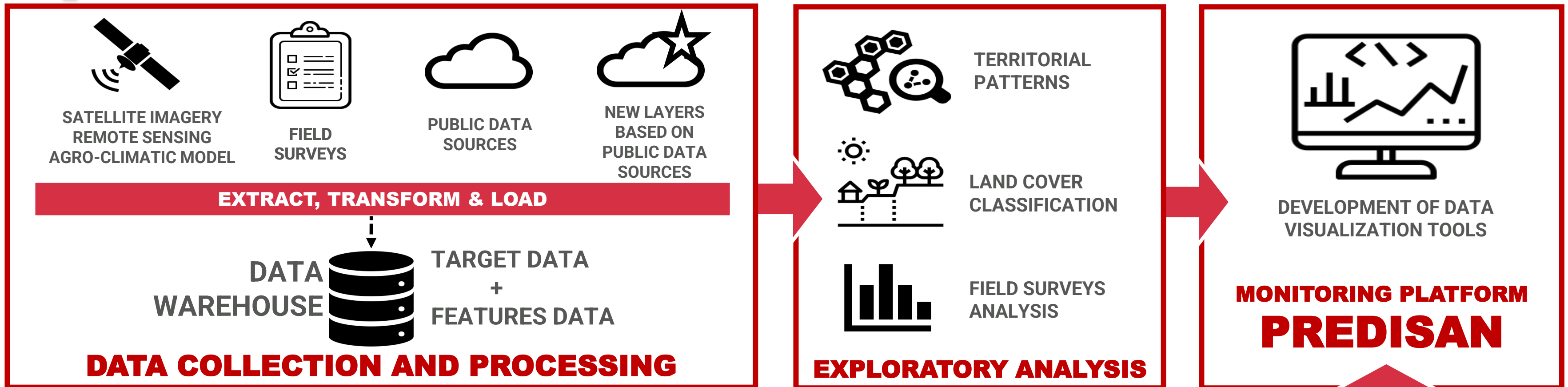
REFLECTIONS AND AREAS FOR IMPROVEMENTS

5. FUTURE IMPROVEMENTS

REFLECTIONS AND AREAS FOR IMPROVEMENTS

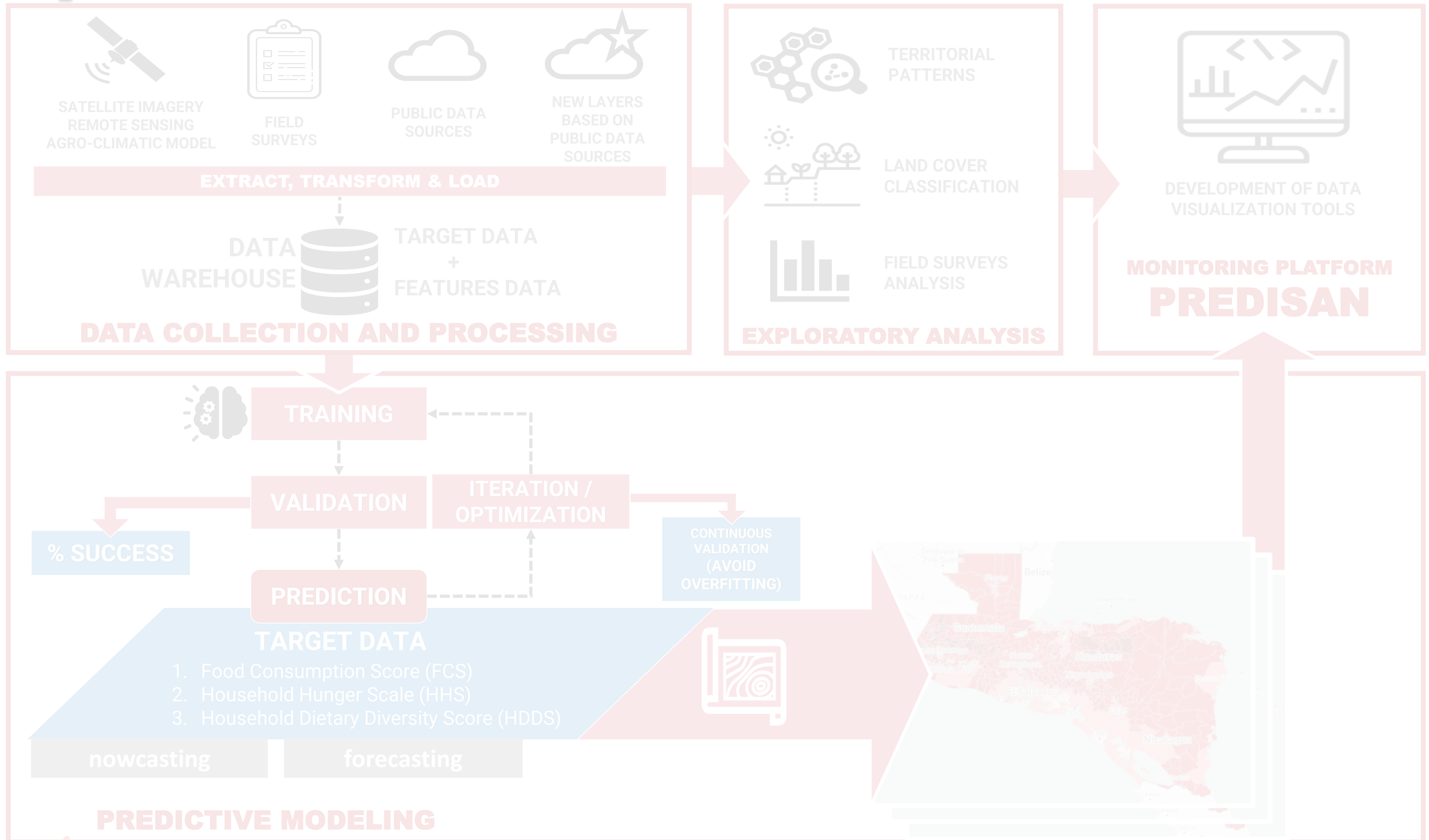
5. FUTURE IMPROVEMENTS

REFLECTIONS AND AREAS FOR IMPROVEMENTS



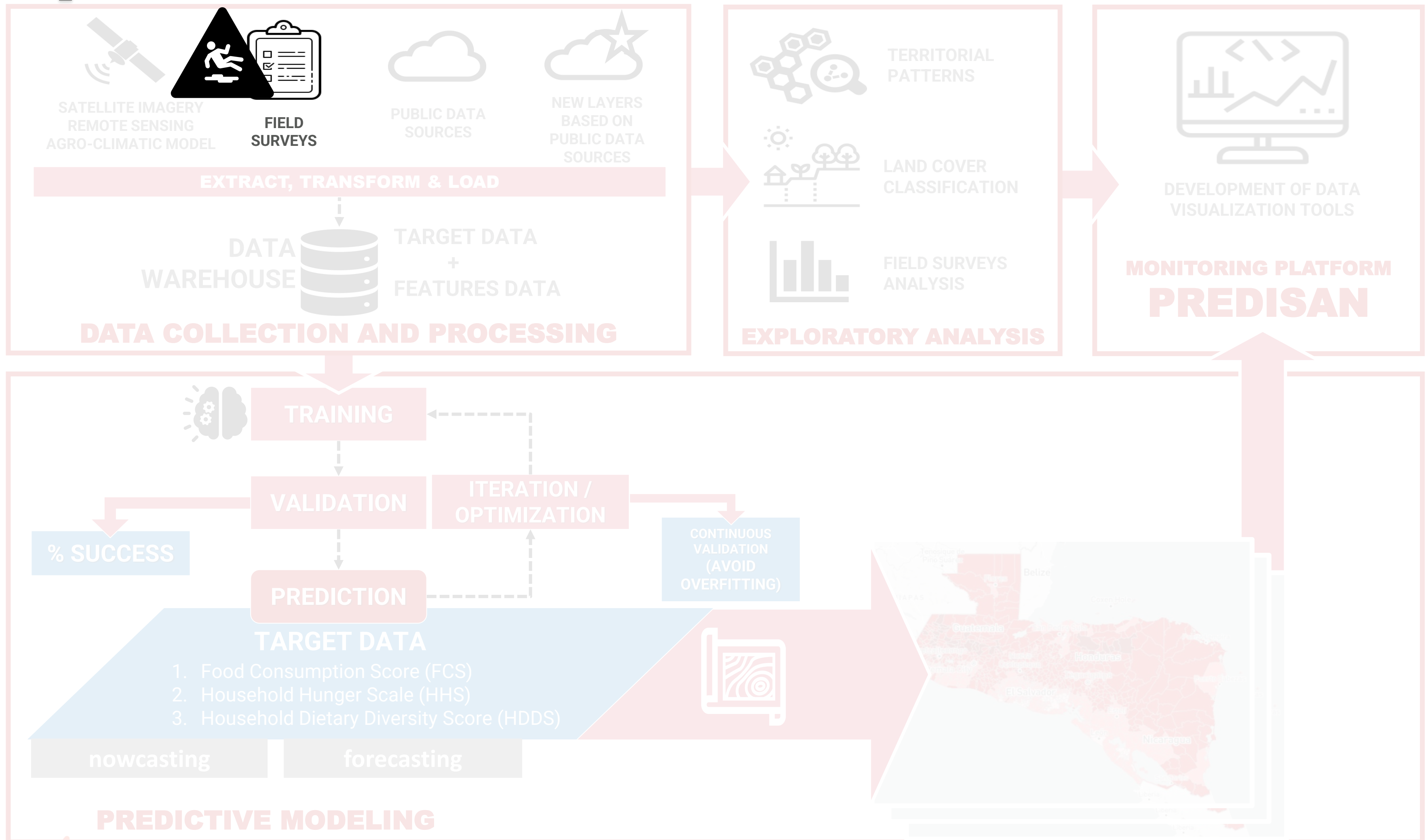
5. FUTURE IMPROVEMENTS

REFLECTIONS AND AREAS FOR IMPROVEMENTS



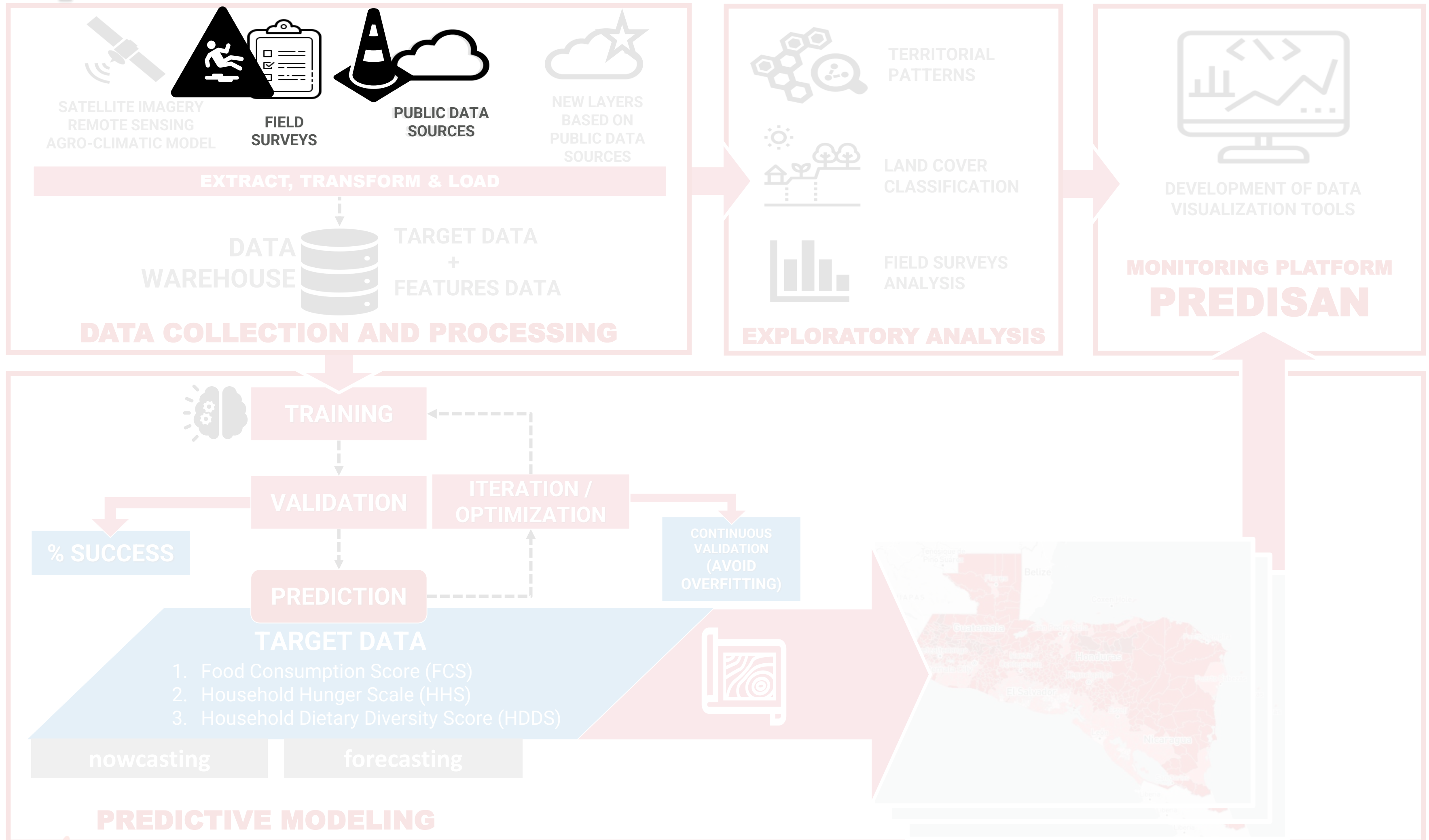
5. FUTURE IMPROVEMENTS

REFLECTIONS AND AREAS FOR IMPROVEMENTS



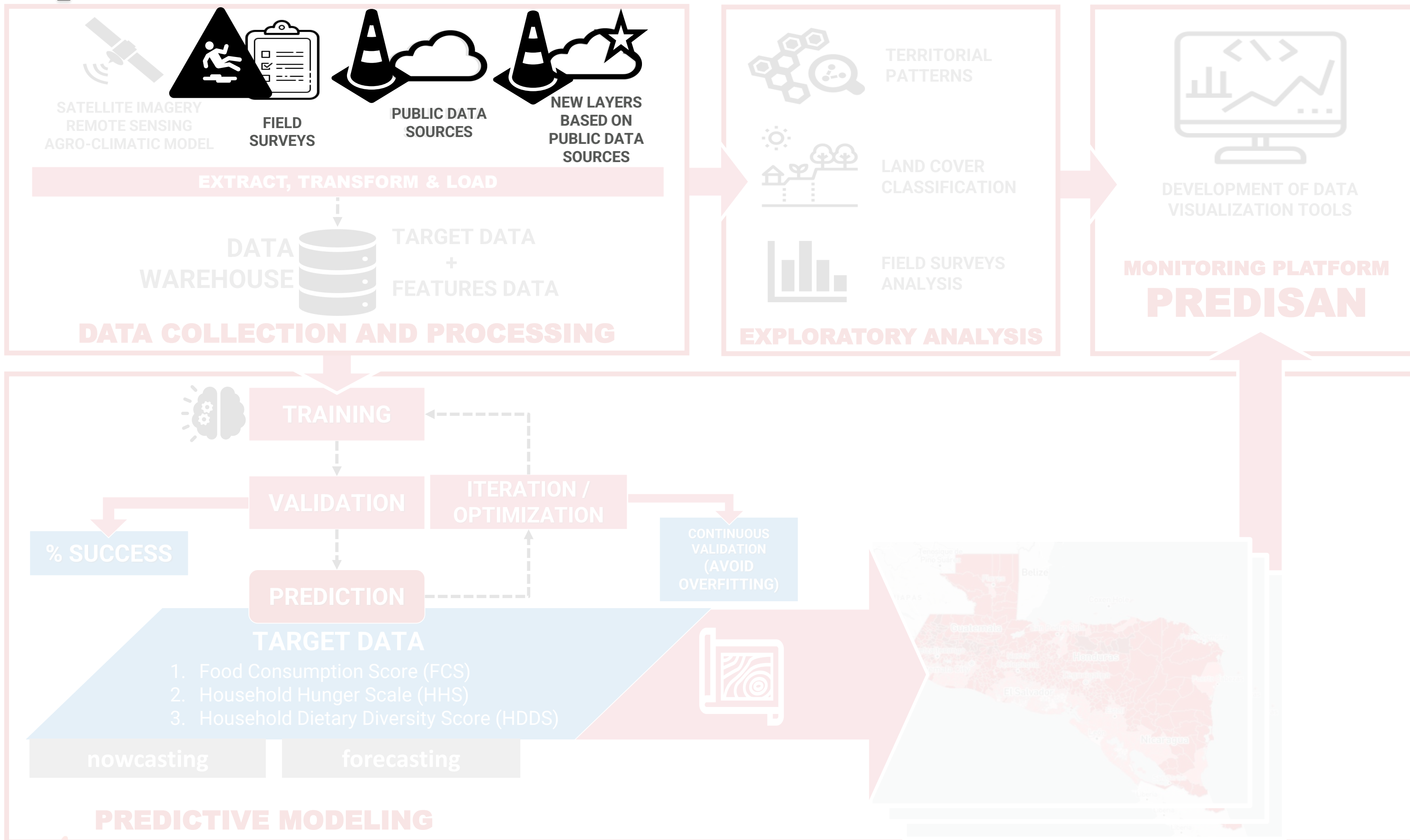
5. FUTURE IMPROVEMENTS

REFLECTIONS AND AREAS FOR IMPROVEMENTS



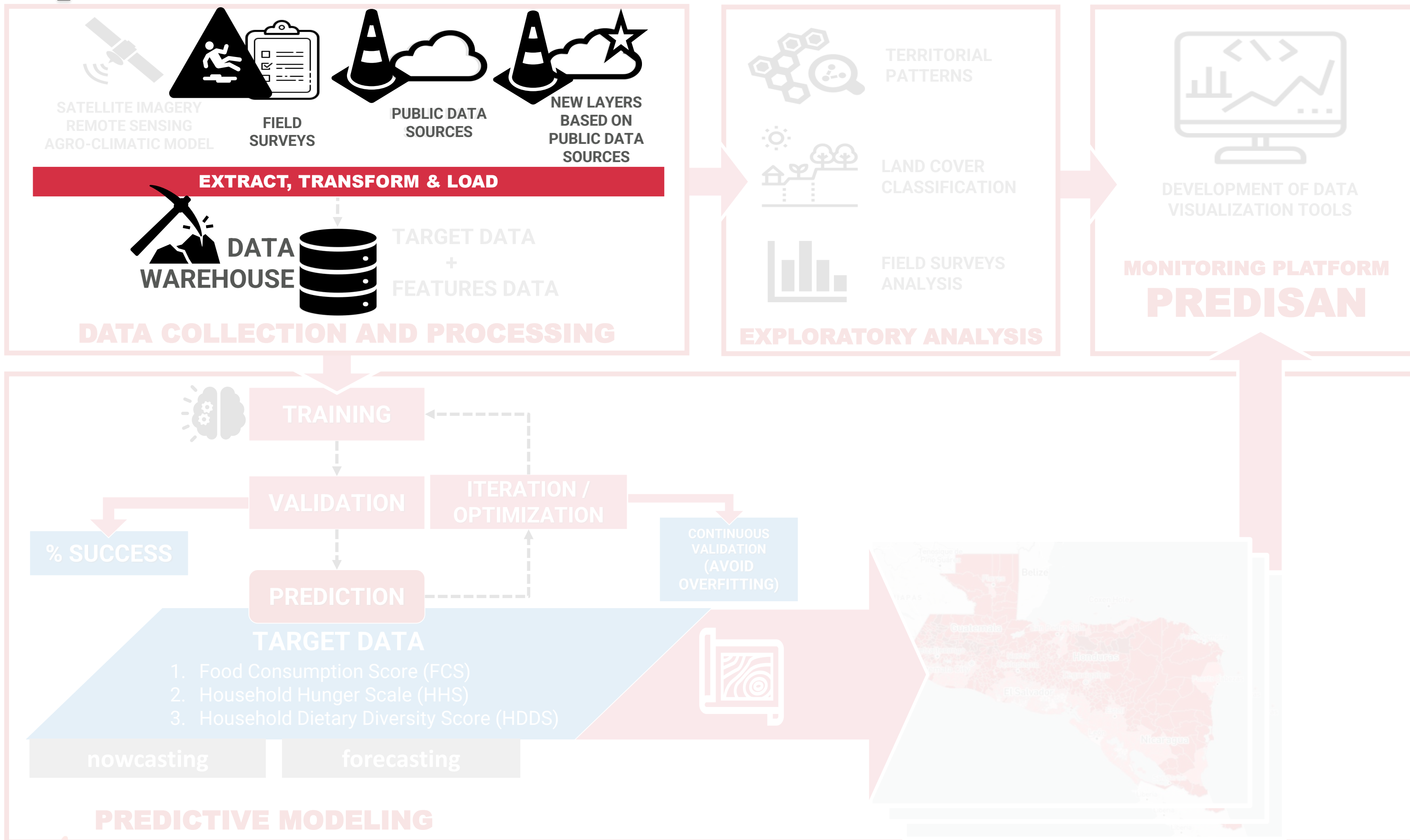
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REFLECTIONS AND AREAS FOR IMPROVEMENTS



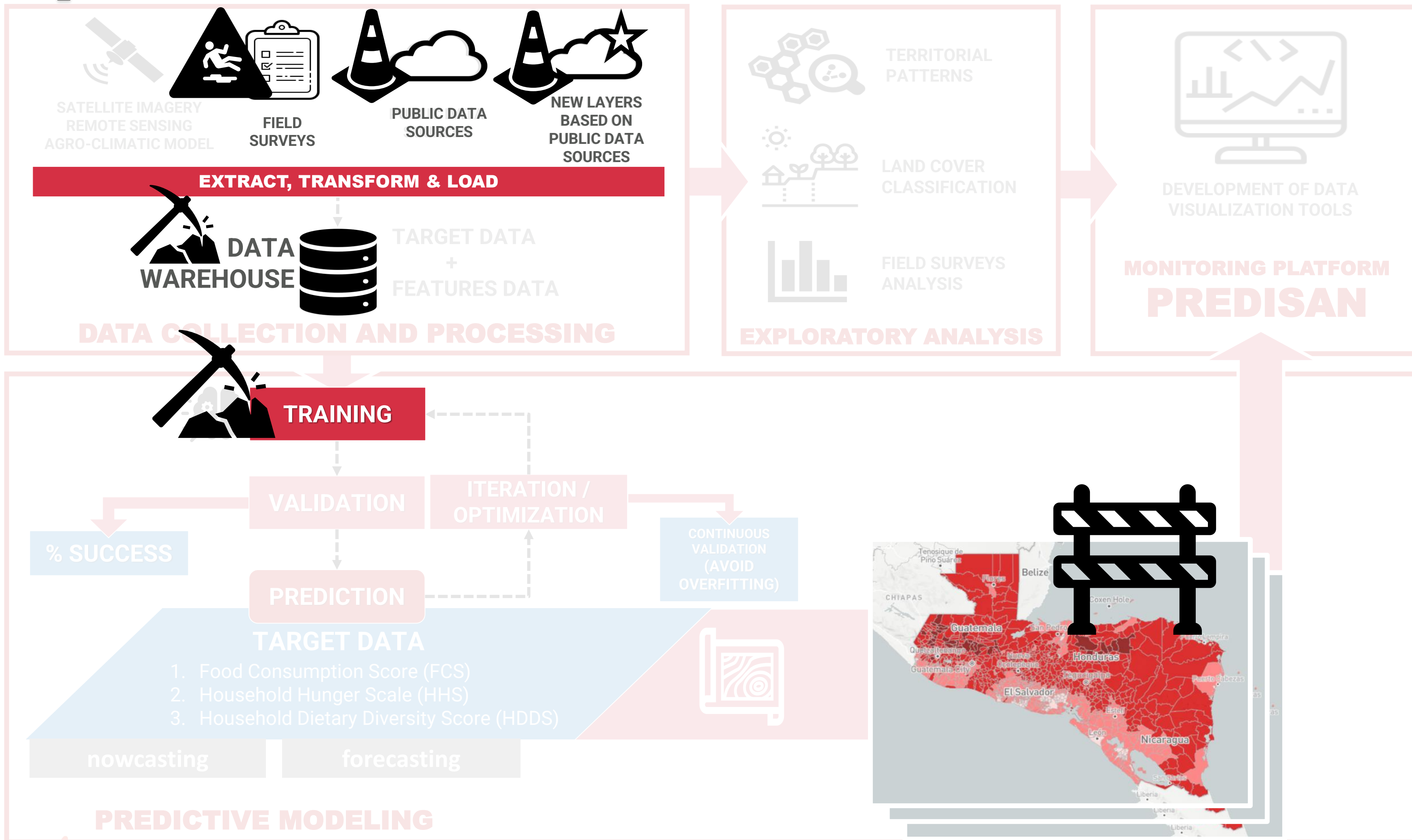
5. FUTURE IMPROVEMENTS

REFLECTIONS AND AREAS FOR IMPROVEMENTS



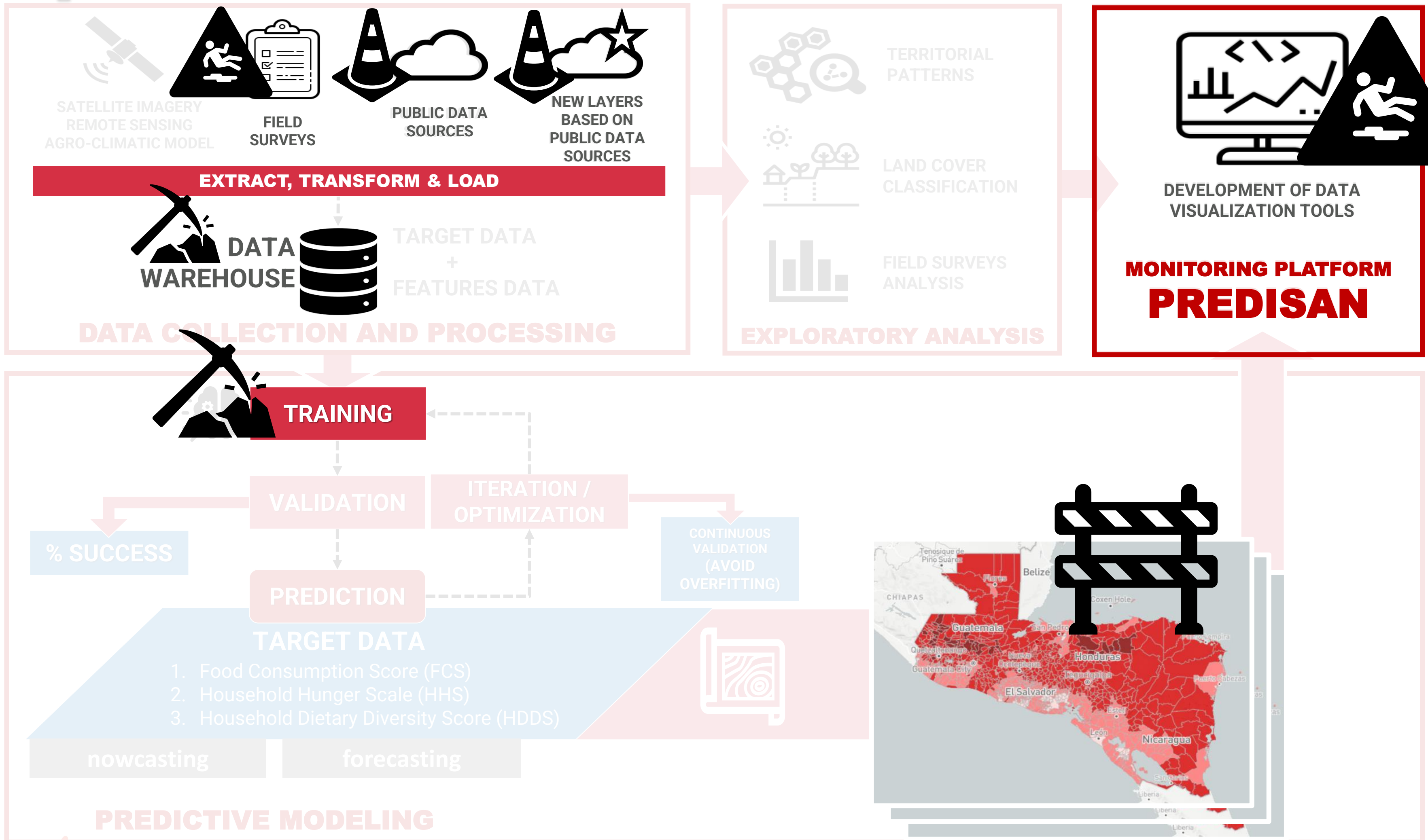
5. FUTURE IMPROVEMENTS

REFLECTIONS AND AREAS FOR IMPROVEMENTS



5. FUTURE IMPROVEMENTS

REFLECTIONS AND AREAS FOR IMPROVEMENTS





6. DEPLOYMENT IN OTHER REGIONS

ADAPTING THE APPROACH TO OTHER REGIONS

6. DEPLOYMENT IN OTHER REGIONS

ADAPTING THE APPROACH TO OTHER REGIONS


PREDISAN VENEZUELA



6. DEPLOYMENT IN OTHER REGIONS

ADAPTING THE APPROACH TO OTHER REGIONS

PREDISAN SAHEL




JUNTA DE ANDALUCÍA
PROGRAMA DE COOPERACIÓN INTERNACIONAL

UNIVERSIDAD DE GRANADA

MONITORING SYSTEM IN SAHEL

Animal Health - Animal Concentration (1)



Select your language:
EN ES

VULNERABILITY

THREATS

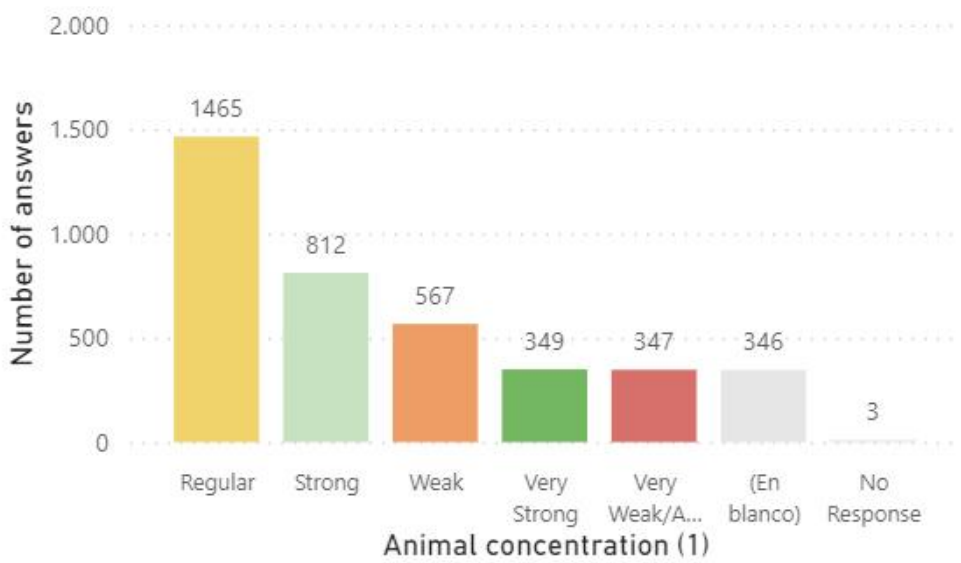
RESULTS

IMPACT

PREDICTIONS

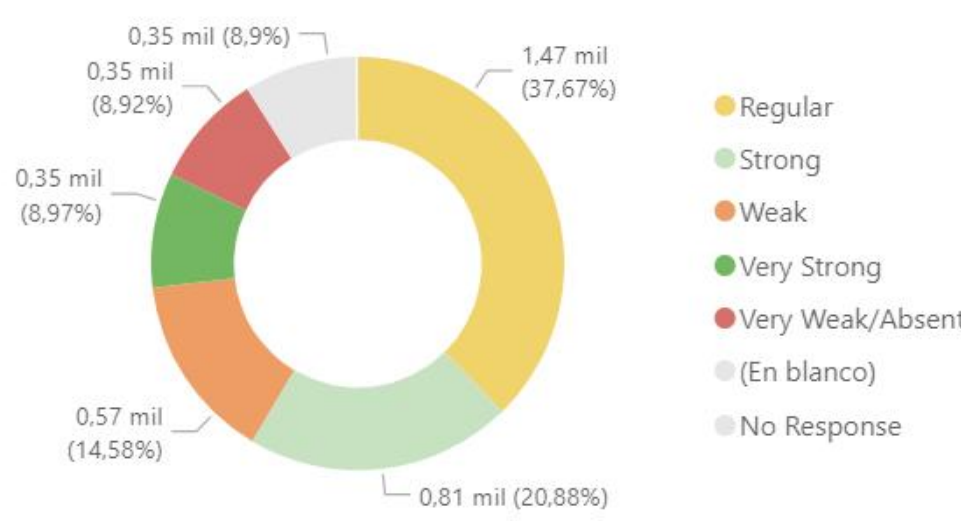
Go to reports

Animal Concentration (1)

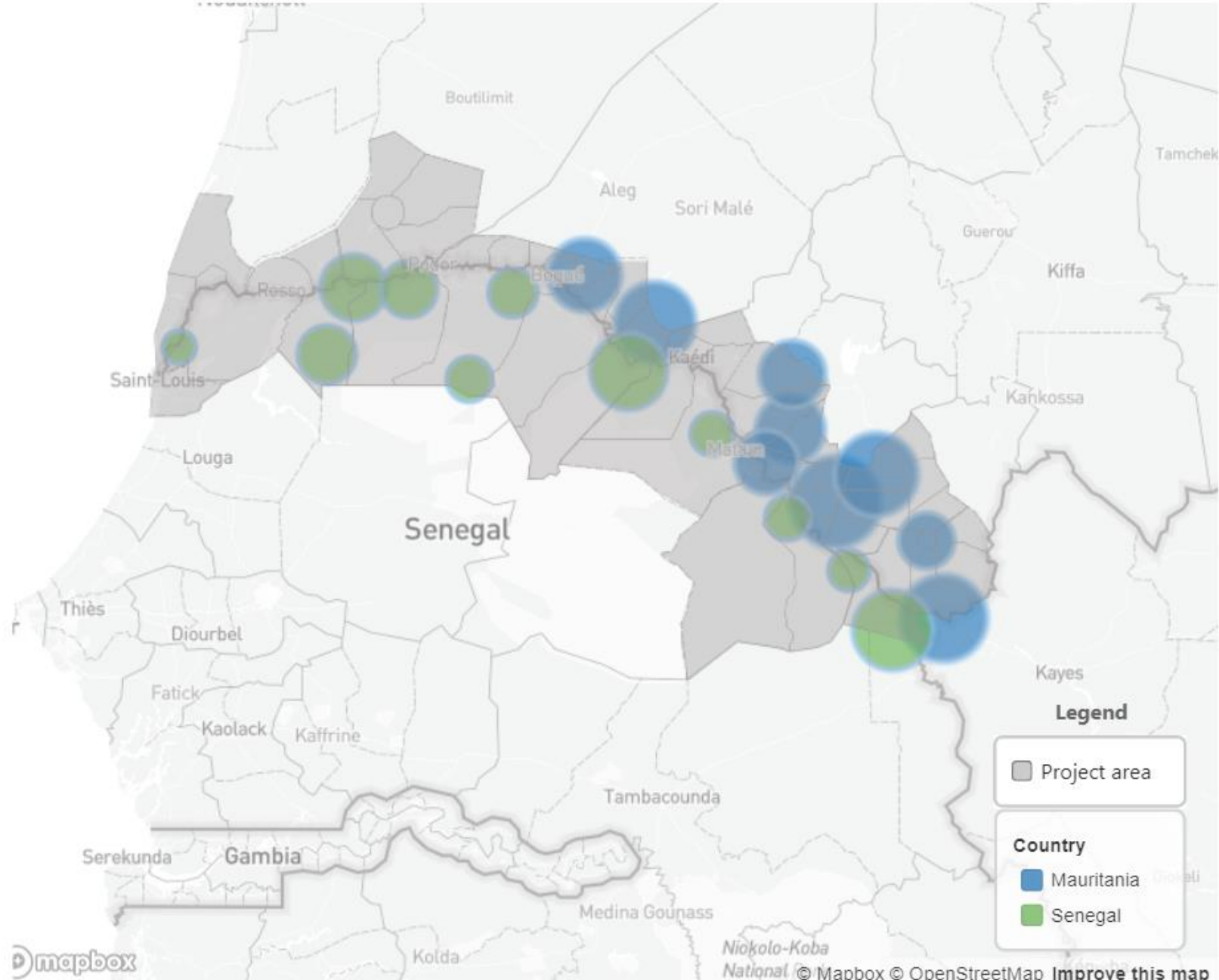


Animal concentration (1)	Number of answers
Regular	1465
Strong	812
Weak	567
Very Strong	349
Very Weak/Absent	347
(En blanco)	346
No Response	3

Animal Concentration (1)



Animal concentration (1)	Percentage
Regular	37,67%
Strong	20,88%
Weak	14,58%
Very Strong	8,97%
Very Weak/Absent	8,92%
(En blanco)	8,9%
No Response	0,35 mil



Map showing animal concentration data points across Senegal. Legend indicates Project area (grey), Mauritania (blue), and Senegal (green).

Go to overweight Large ruminants

Show explanation of categories

Clean filters

Period of time: 12/06/2018 - 03/12/2023

Select Country: Todas

Select Mougatha: Todas

Select Municipality: Todas

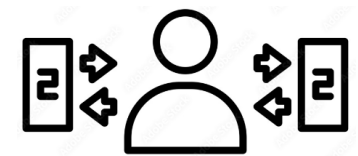
<https://predisan.gis4tech.com/sahel>



7. CONCLUDING REFLECTIONS

7. CONCLUDING REFLECTIONS

- Predictions for food security indicators have room for **improvement**.
- Despite challenges, **the predictions show logical trends** consistent with expert knowledge.
- Incorporating high-quality data and **contextual secondary information** (e.g., violence indicators, food prices) **is necessary**.
- The PREDISAN platform **provides historical data** at the municipality level, including key indicators like Food Consumption Score (**FCS**), Household Hunger Scale (**HHS**), and Household Dietary Diversity Score (**HDDS**).
- Our proposed model uses **nowcast** modeling techniques to predict food insecurity situations **24/7**.
- While our study case focuses on **Central America**, the model can be **adapted for use globally**.





Spin-Off
UNIVERSIDAD DE GRANADA

THANK YOU SO MUCH FOR YOUR ATTENTION

YOU CAN SEND YOUR INQUIRIES TO US, USING THE FOLLOWING CHANNELS:



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