

UN Big Data Hackathon 2022

Youth Track

By HRS Tech

The Team



Our Members



Ahmad **Afif** Mauludi

aafifmauludi@gmail.com



Razis Aji Saputro

razisaji1995@gmail.com



Ulan Surjayanto

ultisa222@gmail.com



Bayu Ade Krisna Setiawan

mrbayuade@gmail.com

Outline

01

The Team

02

Introduction
& Research
Question

03

Dataset

04

Methodology

05

Exploratory
Data
Analysis

06

Model

07

Dashboard /
Visualization

08

Key Finding (s) &
Recommendation (s)

09

Conclusions

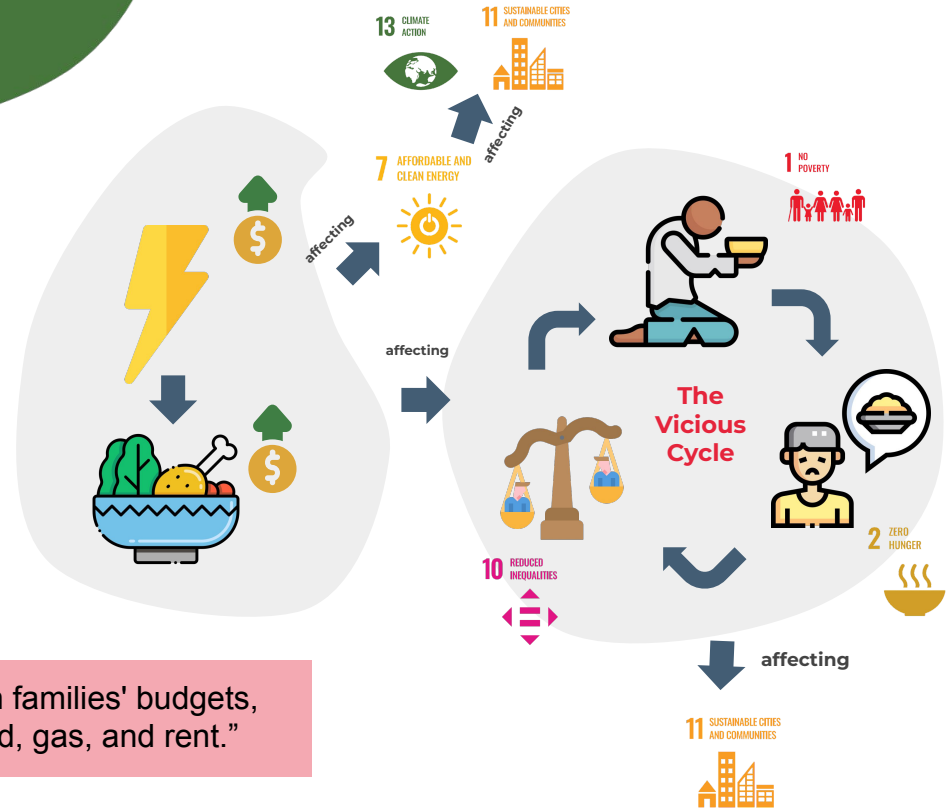
10

Next
Improvement

Introduction

Theme: The rise of food and energy prices affecting vicious cycles of poverty, hunger, and inequalities

“Rising food and energy prices are putting a strain on families' budgets, making it more difficult to afford necessities such as food, gas, and rent.”



Introduction

Globally, domestic food price inflation remains high. Data from June to September 2022 show that almost all low-income and middle-income countries experienced high inflation.

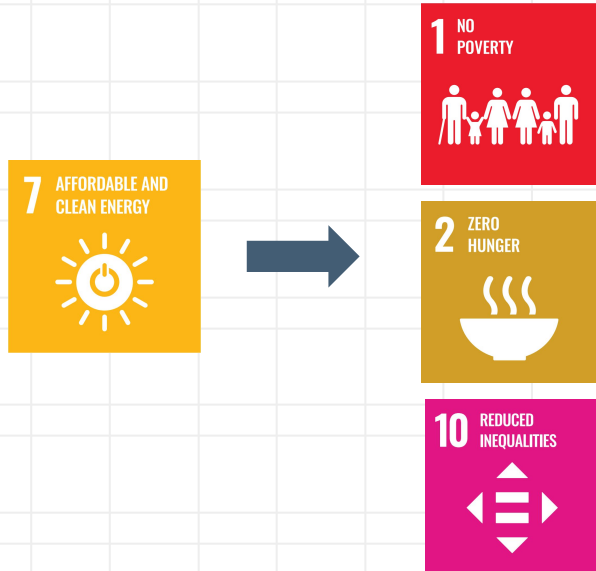
84.2% of low-income countries, 88.9% of lower-middle-income countries, and 93% of upper-middle-income countries experienced inflation levels above 5%, with many experiencing double-digit inflation.

The proportion of high-income countries experiencing high food price inflation has risen to 87.5%.

Source: <https://www.worldbank.org/en/topic/agriculture/brief/food-security-update>

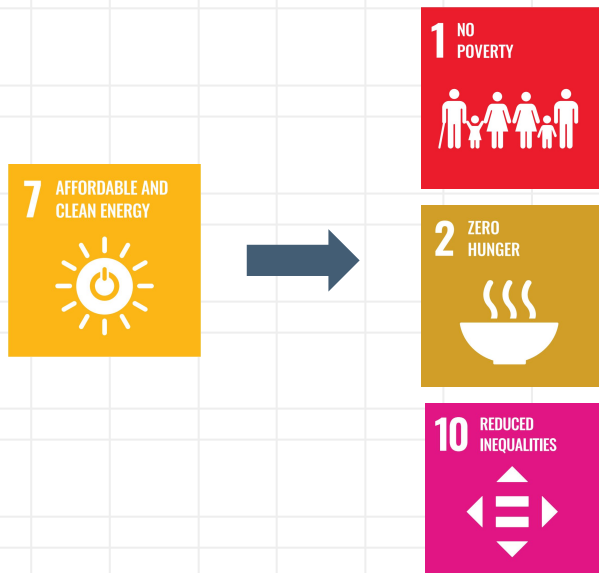


What we are trying to solve



Creating a tool based on data and visualizing the correlation between poverty, hunger, inequalities, and energy consumption & prices to support the decision-makers.

Research Question



#1 How is the correlation between **energy prices** & **energy consumption***?

#2 How is the impact between renewable & non-renewable **energy prices** on **poverty, hunger and inequality**?

#3 How is the impact between renewable & non-renewable **energy consumption*** on **poverty, hunger and inequality**?

*consumption per capita

Poverty

- Poverty Headcount Ratio National Poverty Lines
- Multidimensional poverty index
- Multidimensional poverty headcount ratio, household (% of total households) etc

Source: World Bank

Hunger (Food Insecurity)

- Prevalence of severe food insecurity in the population (%)
- Prevalence of moderate or severe food insecurity in the population (%)
- Prevalence of undernourishment (% of population)

Source: World Bank

Inequalities

- Gini Index

Source: World Bank

Energy

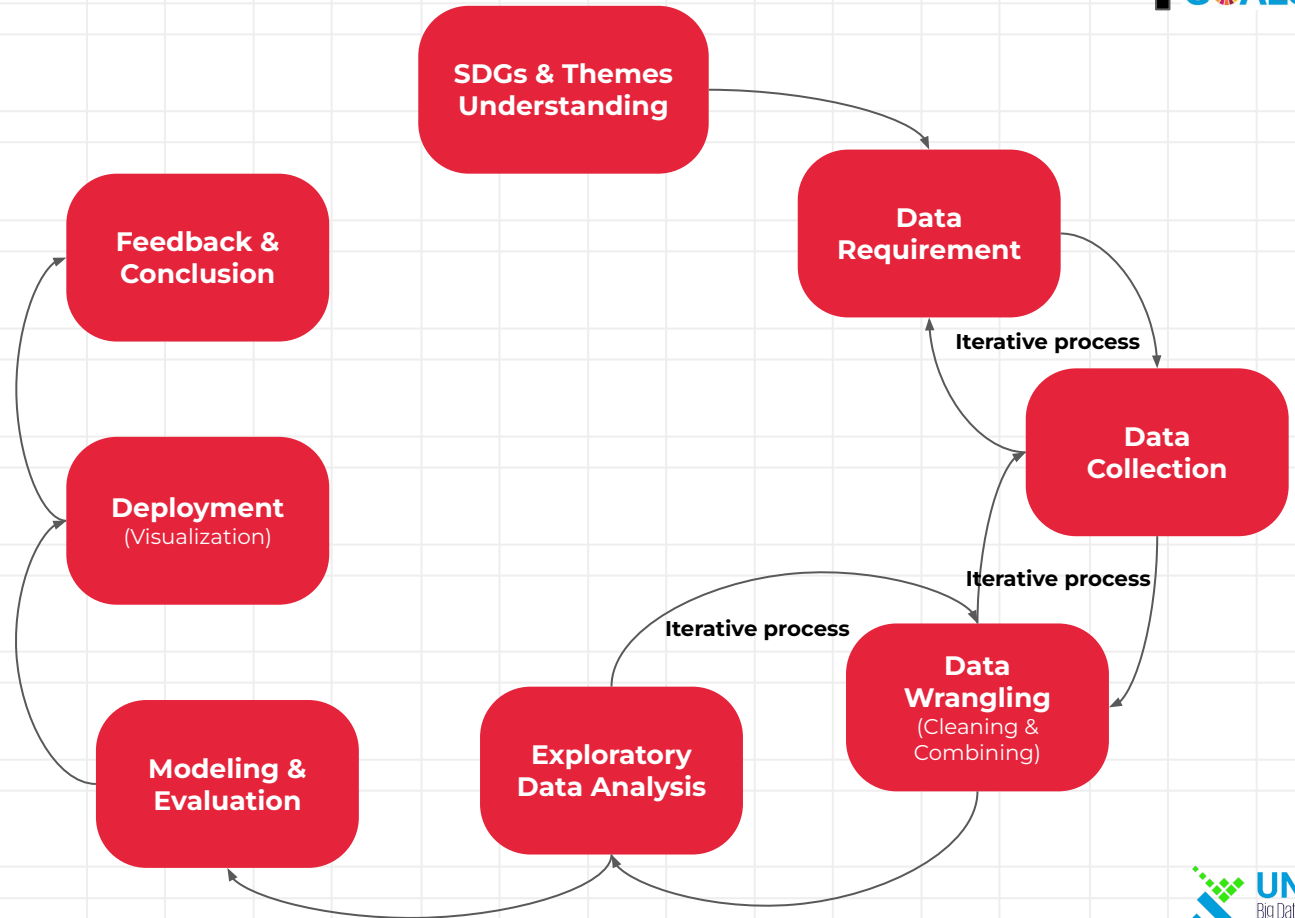
- Energy Consumption
- Energy Price
- etc

Source: Our World in Data & World Bank

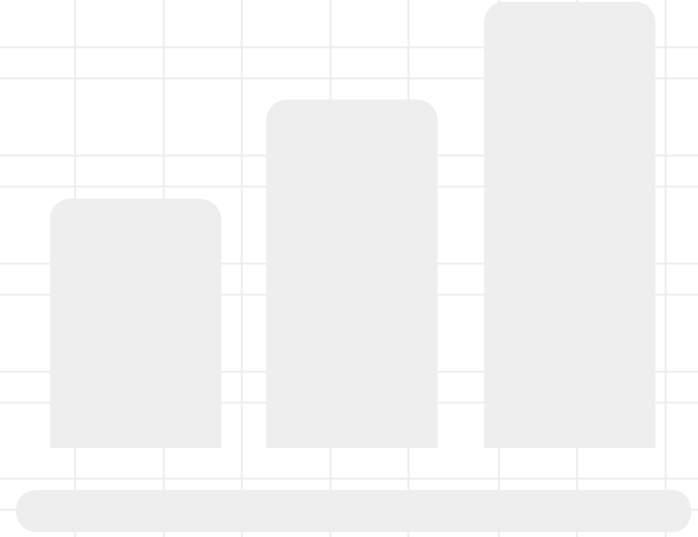
Data Documentation

Final Combined Data Set

Methodology



Exploratory Data Analysis (EDA)



Data Info

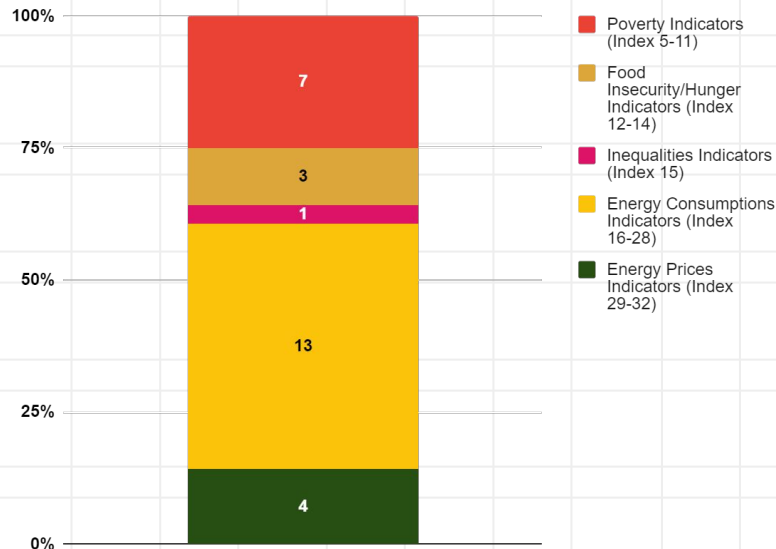
33 Columns & 16476 Rows

2 Categorical, 30 Numerical, and 1 Date Feature (s)

our_df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 16476 entries, 0 to 16475
Data columns (total 33 columns):
#   Column                                     Non-Null Count  Dtype
---  -
0   country                                   16476 non-null  object
1   region                                   16251 non-null  object
2   year                                      16476 non-null  datetime64[ns]
3   population                               16371 non-null  float64
4   gdp                                       11004 non-null  float64
5   SI_POV.NAHC                             960 non-null    float64
6   SI_POV.GAPS                             1668 non-null   float64
7   SI_POV.UHIC.GP                          1668 non-null   float64
8   SI_POV.UHIC                             1667 non-null   float64
9   SI_POV.HDIH                             431 non-null    float64
10  SI_POV.HDIH.MA                          357 non-null    float64
11  SI_POV.HDIH.FE                          357 non-null    float64
12  SN.ITK.SVFI.ZS                          665 non-null    float64
13  SN.ITK.HSFI.ZS                          671 non-null    float64
14  SN.ITK.DEFC.ZS                          2816 non-null   float64
15  SI_POV.GINI                             1663 non-null   float64
16  energy_per_capita                       9521 non-null   float64
17  biofuel_cons_per_capita                 838 non-null    float64
18  coal_cons_per_capita                    4196 non-null   float64
19  fossil_energy_per_capita                 4193 non-null   float64
20  gas_energy_per_capita                    4219 non-null   float64
21  hydro_energy_per_capita                  4215 non-null   float64
22  low_carbon_energy_per_capita             4215 non-null   float64
23  nuclear_energy_per_capita                4058 non-null   float64
24  oil_energy_per_capita                    4222 non-null   float64
25  other_renewables_energy_per_capita       4172 non-null   float64
26  renewables_energy_per_capita             4215 non-null   float64
27  solar_energy_per_capita                  4172 non-null   float64
28  wind_energy_per_capita                   4172 non-null   float64
29  EP.PIP.DESL.CD                          1582 non-null   float64
30  NY.COAL.RT.ZS                           6910 non-null   float64
31  NY.PETR.RT.ZS                           7054 non-null   float64
32  NY.TOTL.RT.ZS                           7145 non-null   float64

dtypes: datetime64[ns](1), float64(30), object(2)
memory usage: 4.1+ MB
```

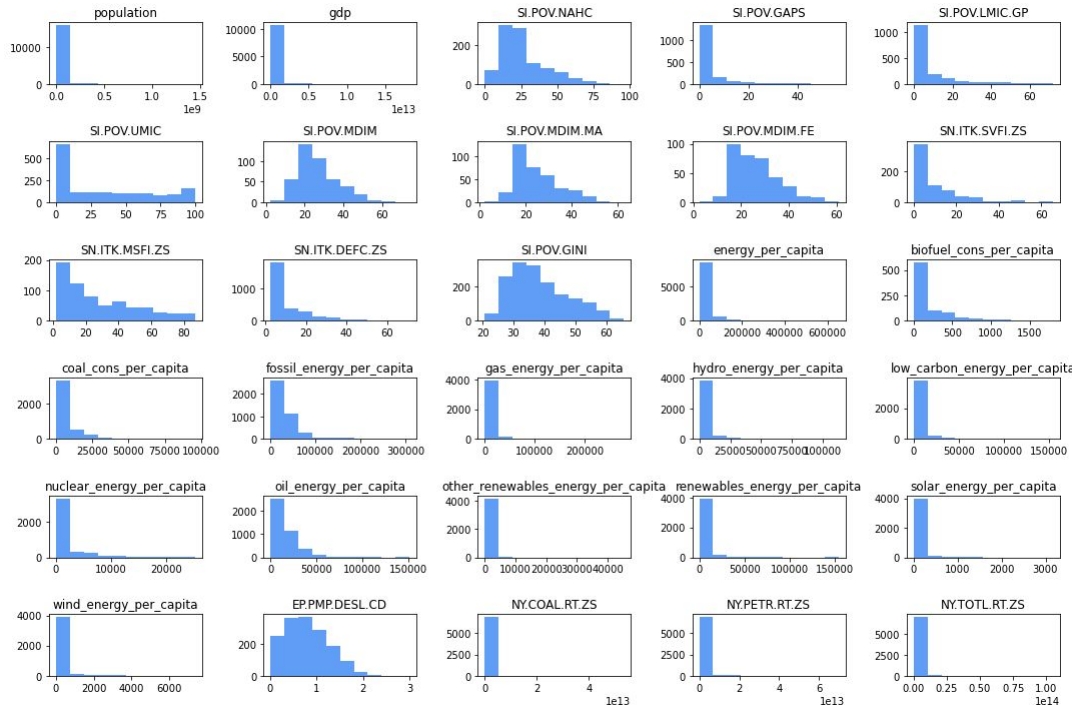


The Indicators

[Data Documentation](#)

Data Distribution

numerical



categorical

	count	unique	top	freq
country	16476	221	Zimbabwe	122
region	16251	6	Asia	4401

```
value counts dari column country
Zimbabwe    122
Morocco      122
Ecuador      122
Myanmar      122
Netherlands  122
...
Eritrea      27
Palestine    24
Timor        18
Montenegro   17
South Sudan  10
Name: country, Length: 221, dtype: int64
```

```
value counts dari column region
Asia    4401
Africa  4139
Americas 3378
Europe  3355
Oceania  938
0        40
Name: region, dtype: int64
```

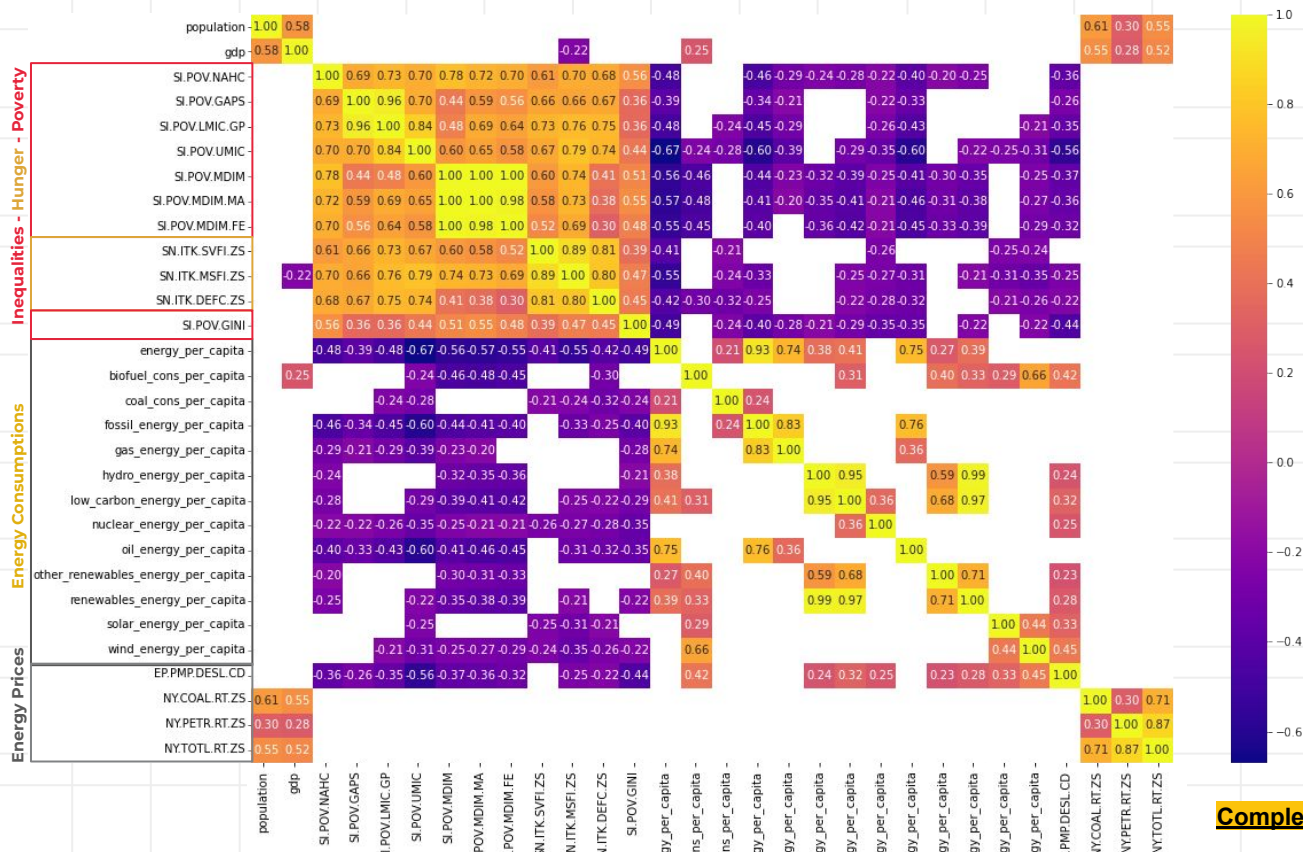
date

year range is between 1900-01-01 00:00:00 and 2021-01-01 00:00:00

[Completed Code .ipynb](#)

[Completed Code .html](#)

Heat Map Correlation



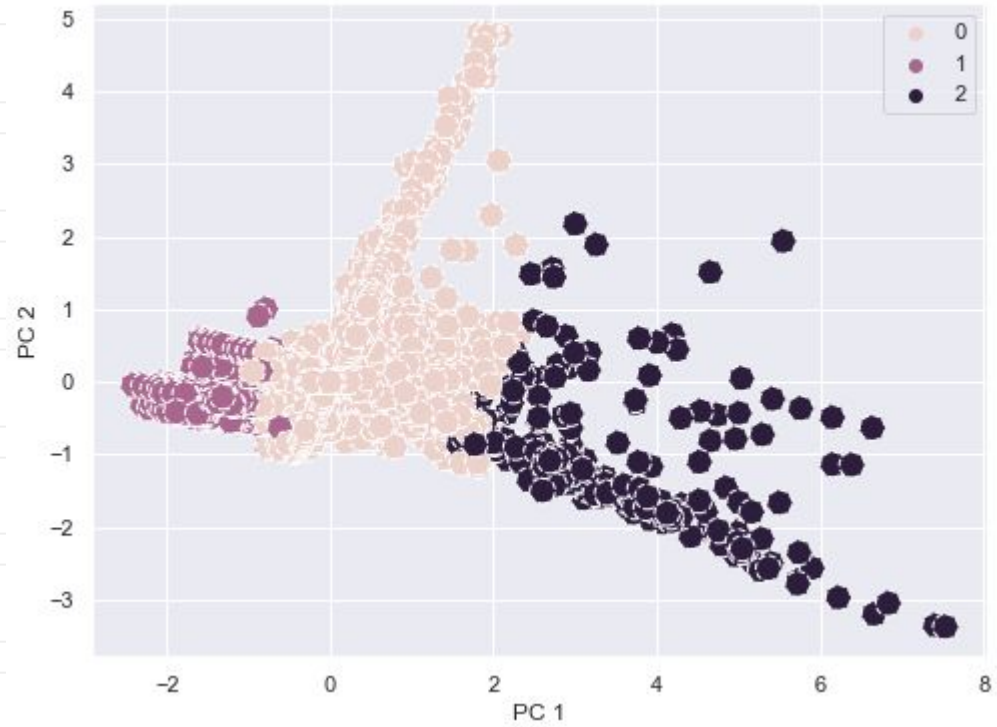
Poverty, Hunger, and Inequality have **Negatives Correlation** with **Energy Consumption** (both renewable & non-renewable)

The vicious cycle (Poverty, Hunger, and Inequality) **positively correlates**.

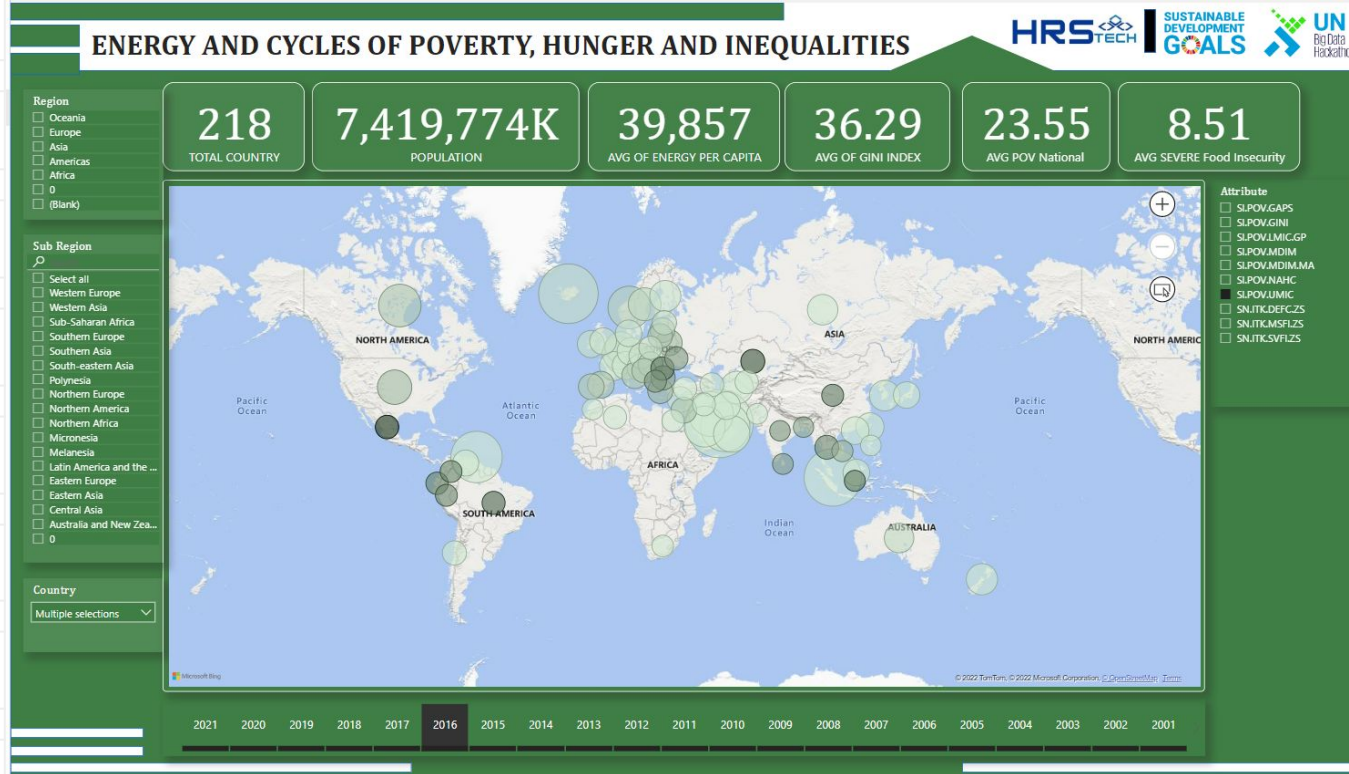
The Pump price for diesel fuel (EP.PMP.DESL.CD) is **negatively correlated** with **The Vicious Cycles** and **positively correlated** with some **renewable energy consumption**

Modeling

We use clustering to creating a cluster based on poverty, hunger and inequalities index into three clusters.



Dashboard / Visualization

[Dashboard Link](#)

Key Finding (s) & Recommendation (s)

#1 When fossil energy prices are volatile, investors tend to put money into the renewable energy industry. An additional set of actions to save energy, diversify supplies, and replace fossil fuels by hastening the deployment of renewable energy. Increasing energy efficiency and savings, as well as expanding renewables, are expected to lower energy prices while hastening the green transition.

#2 The rise of diesel fuel prices surprisingly has a negative correlation with “the vicious cycle”. It means the higher the price, the poverty, hunger, and inequality indicators would be lower!

#3 Consumption of renewable and nonrenewable energies per capita correlates negatively with poverty, hunger, and inequality. Energy bills consume a significant portion of consumers' income, limiting their ability to cover other expenses, and resulting in energy poverty.

The Conclusions

#1 The rise of the diesel fuel price is positively correlated with renewable energy consumption.

#2 The rise of diesel fuel prices has a negative correlation with “the vicious cycle” .

#3 The higher the vicious cycle, lower the energy consumption.

Next Improvement

1. Completing the *null* data on dataset for future analysis.
2. Collect and add **Renewable Energy Potential Data** for more completed analysis and decision support tools on how to utilize and prioritize the renewable energy accelerating manufacturer based on poverty, hunger, and inequalities indicators.
3. **Deployment** of “the model” on visualization and predicting the missing and future indicators.
4. Creating tools based on the model to customized and predict certain indicators.



HRS  TECH

SUSTAINABLE
DEVELOPMENT
GOALS

THANKYOU

Appendix 1 : Color Palette Inspiration



SDGs #1



SDGs #2



SDGs #7



SDGs #10



SDGs #11



SDGs #13

2022 Theme

Using Big Data and Data Science to develop ideas and solutions to address Global Challenges and help achieve Sustainable Development Goals; notably to support policies caused by:

- The disruption to Global Value Chains and Economic Globalization due to disasters, conflicts, restrictions, blockages
- The impact of Climate Change on society as part of monitoring SDG 13
- The rise of food and energy prices affecting vicious cycles of poverty, hunger, and inequalities



#E5243B



#DDA63A



#3F7E44



#ffffff



#000000



#808284



#DD1367

Appendix 2 : The Assets

Logo Usage



[SDGs Brand Guidelines](#)



Font Usage

Montserrat - Normal

Montserrat - Bold

Montserrat - Black

Icons



[source](#)



[source](#)



[source](#)



[source](#)



[source](#)

Appendix 3 : SDGs Logo

1 NO
POVERTY



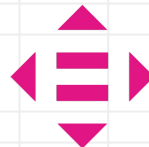
2 ZERO
HUNGER



7 AFFORDABLE AND
CLEAN ENERGY



10 REDUCED
INEQUALITIES



11 SUSTAINABLE CITIES
AND COMMUNITIES



13 CLIMATE
ACTION



1 NO
POVERTY



2 ZERO
HUNGER



7 AFFORDABLE AND
CLEAN ENERGY



10 REDUCED
INEQUALITIES



11 SUSTAINABLE CITIES
AND COMMUNITIES



13 CLIMATE
ACTION

