# LIFE EXPECTANCY DOES IMMUNIZATION MATTER?

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### **Outline**

- Context and Problem statement
- Data Wrangling
- Exploratory data analysis
- Modelling
- Conclusion

#### **Context and Problem statement**

#### Contexte

- Current pandemic and vaccination.
- Past 15 years, development of health section and reduction human mortality as compared to the past 30 years.
- does immunization matter?
- Global Health Observatory (GHO) data repository keeps track of the health status as well as many other related factors for all countries.
- The dataset (life expectancy, health factors for 193 countries) 2000-2015.

within a month, the present project assessed the contribution and the relationship of each feature on life expectancy with a special focus on immunization factors, and develop a regression model to predict life expectancy.

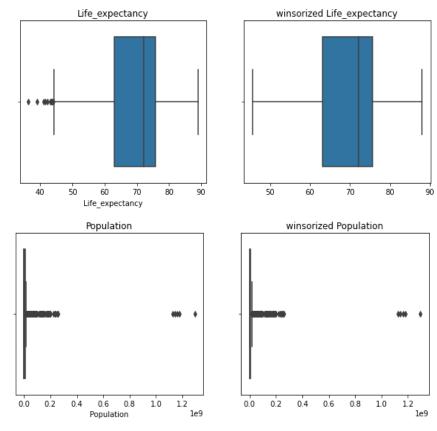
**Constraints**: The dataset has important missing values to handle (table 1). This could impact the model depending on the imputation technique chosen.

#### **Outliers visualization and treatment**

Winsorize method to treat Outliers

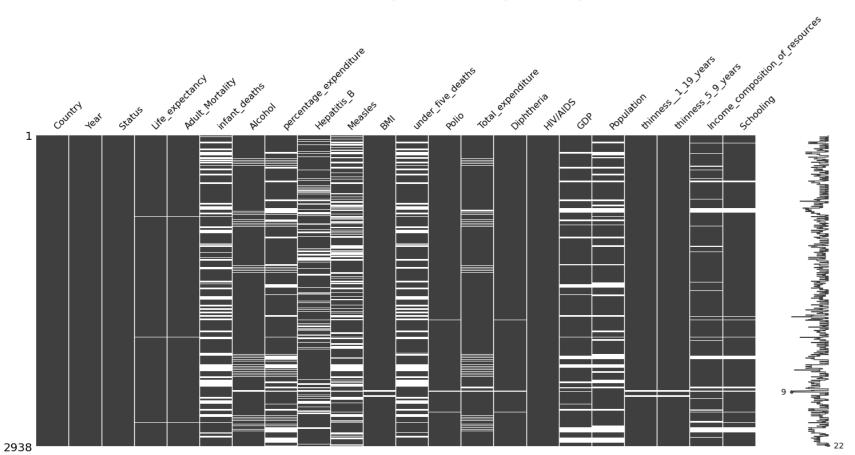
Dataset has 2938 observations and 22 columns (21 are independent variables)

Predicting variables were then divided into several broad categories:Immunization related factors, Mortality factors, Economical factors, and Social factors.

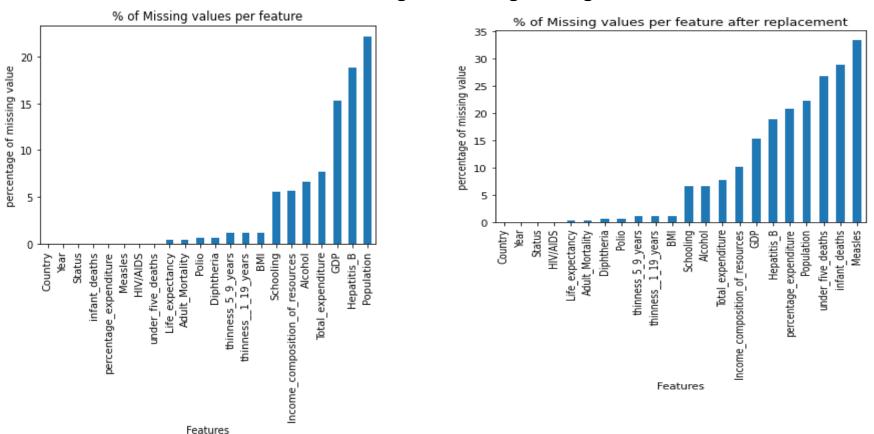


Box plot before and after winsorize application (0.01, 002)

#### **Assessing and treating Missing Value**

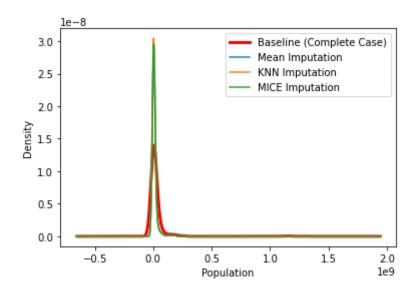


#### **Assessing and treating Missing Value**



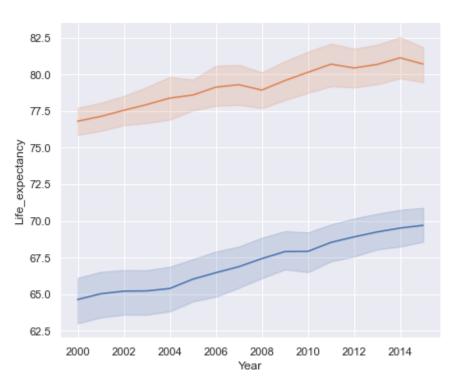
Original state of data with missing value and after the replacement of the uncommon type with NAN

#### **Assessing and treating Missing Value**

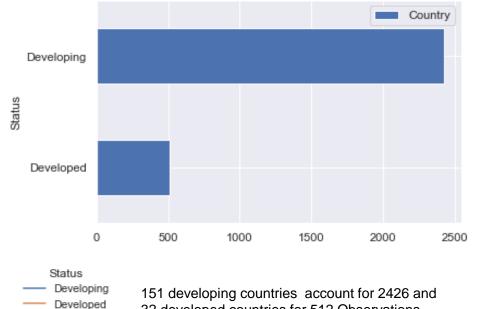


The best imputation technique is: MICE Imputation

What is the trend of life expectancy?

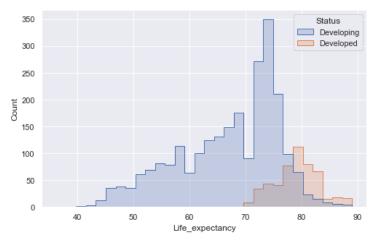


Count of observations in the dataset developed vs developing country

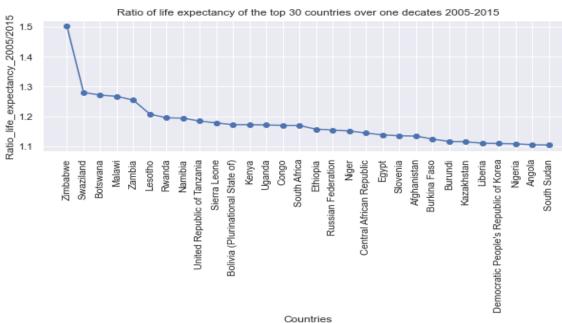


32 developed countries for 512 Observations

#### How does the distribution of life expectancy look like?

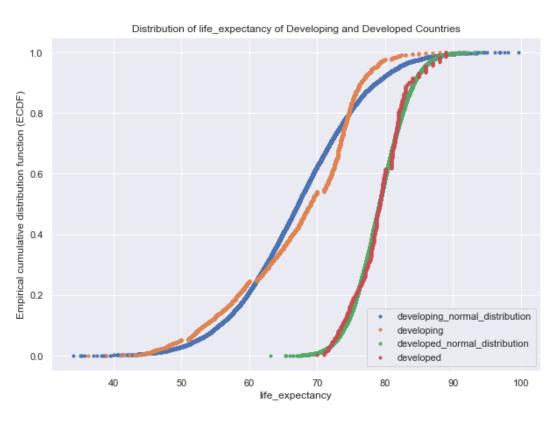


#### How was the life expectancy over one decade (2005 to 2015)?



Min in developed countries is 69 /developing countries is 39. Maxi in developed countries is 89, the same in developing countries.

### **Statistical Analysis**



Confidence Interval of Life expectancy at 95%

\*CI developing countries : [64.834, 68.364]

\*CI developed countries: [77.911, 79.451]

### **Statistical Analysis**

#### 1- state the hypothesis

ho: mean avg of LE \_developed = mean avg of LE \_developing

h1: mean avg of LE \_developed != mean avg of LE \_developing

#### 2- state the significance level (here we set the threshold for the test)

alpha = 0.05 or 5% for two tail test

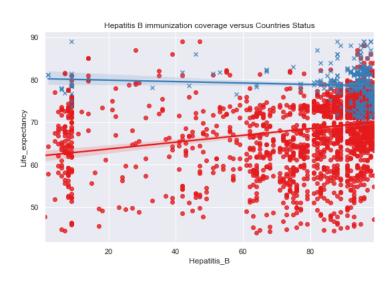
#### 3- identify the test statistic

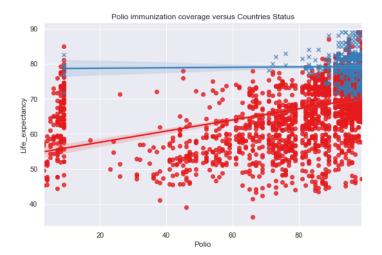
we conduct a Z test for 2 independants samples,

#### 4- Conclusion:

From This result, we reject the null hypothesis, we found that there is a significant difference between the mean average life expectancy of developed countries to that of developing countries.

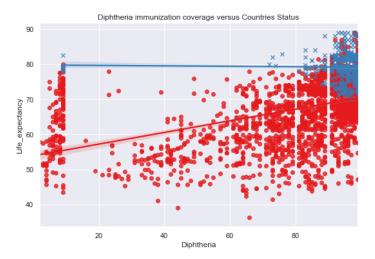
#### Immunization and life expectancy





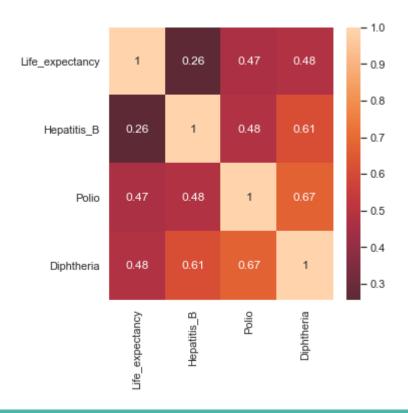
Status Developing Developed

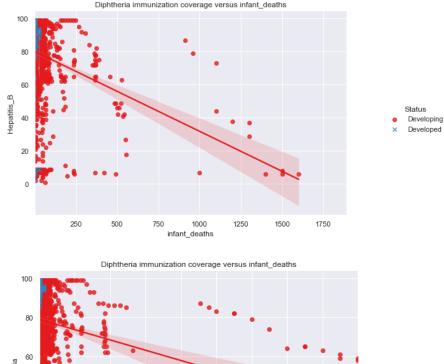
Status Developing Developed

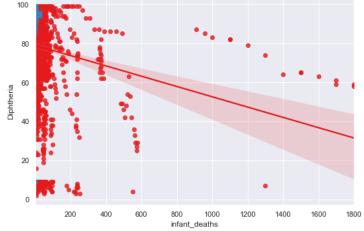


Developed

#### Immunization and life expectancy, and mortality factors

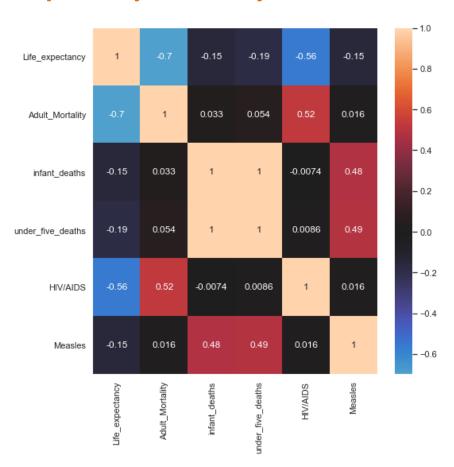




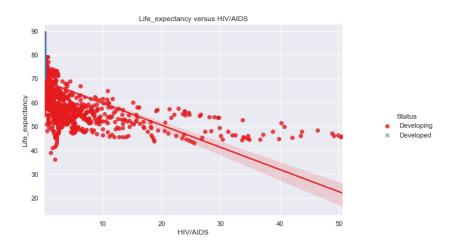


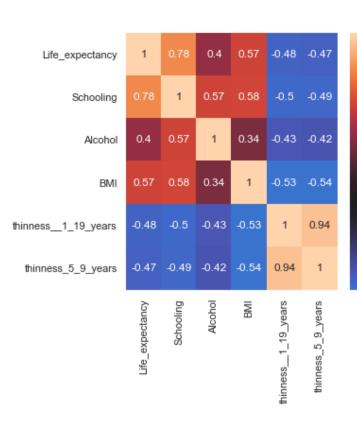
Status Developing

Developed

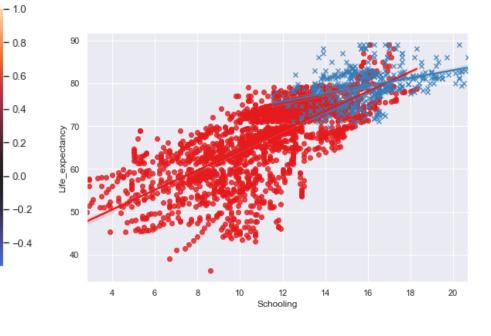


#### Mortality\_factors and Life\_expectancy



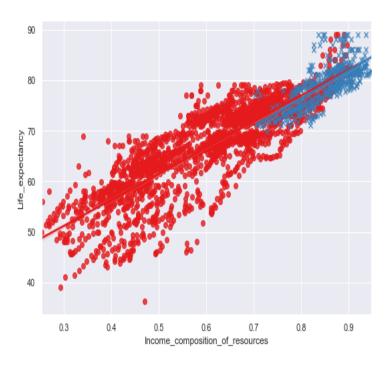


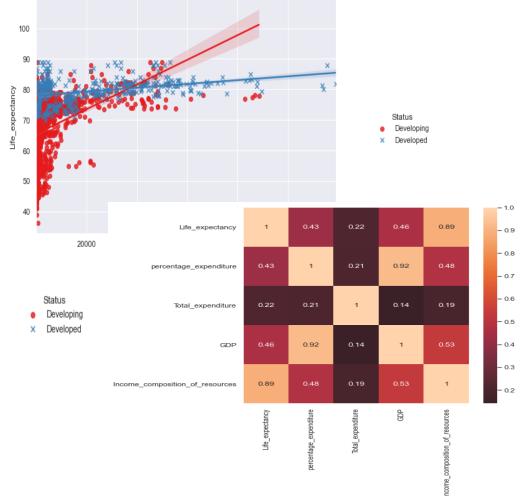
#### Social Factors and life\_expectancy



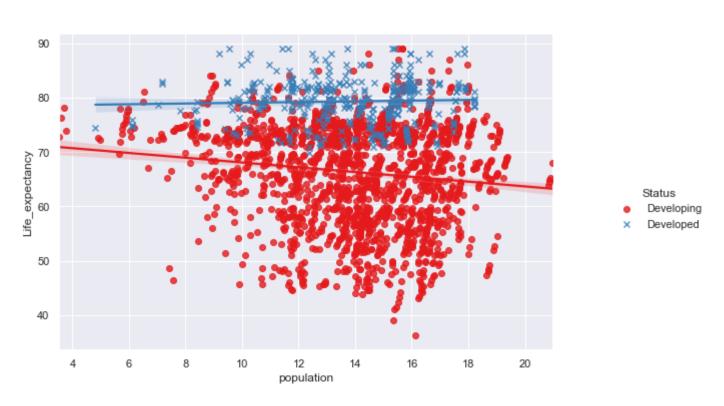
Status Developing Developed

#### **Economical\_factors and Life\_expectancy**





#### **Population and life expectancy**



## **Preprocessing the data**

Label Encoder of categorical variable with one hot\_encoder

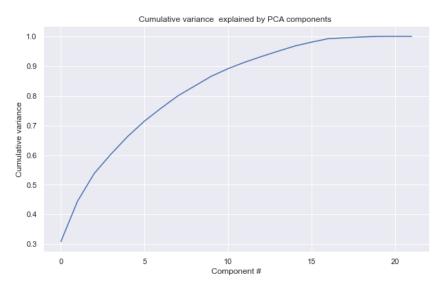
Imputing missing value (Mice imputation)

Divide in test set and train set (30%, 70%)

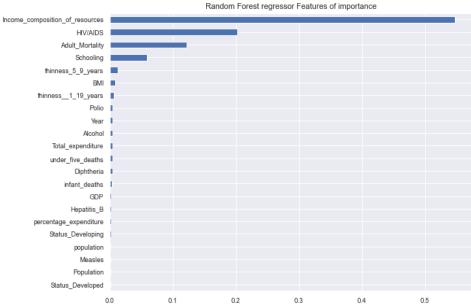
Scaling the dataset

PCA transformation

# **Preprocessing the data**



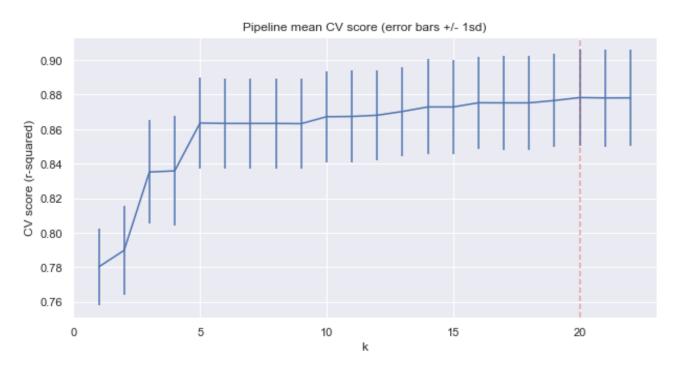
Note: The first five components seem to account for over 75% of the variance, and 10 components is 92% of the variance



# Modelling

```
pipe = make_pipeline(
    IterativeImputer(),
    StandardScaler(),
    SelectKBest(f_regression),
    LinearRegression()
```

{'selectkbest\_k': 20}



The above suggests a good value for k is 20

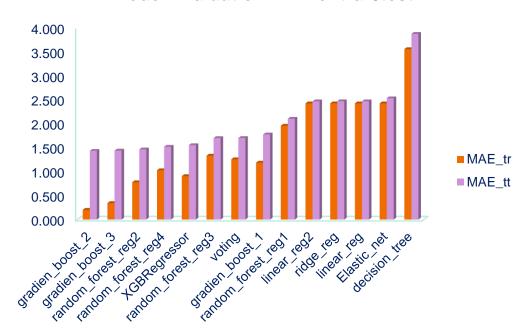
# Modeling

| model         | model_definition   |  |  |
|---------------|--|--|--|
|               | Pipeline(steps=[('iterativeimputer', IterativeImputer()),                  |  |  |
| linear_reg    | at 0x0000025E60390C10>)),  ('linearregression', LinearRegression())])      |  |  |
| linear_reg2   | Pipeline(steps=[('iterativeimputer', IterativeImputer()),                  |  |  |
| ridge_reg     | Pipeline(steps=[('iterativeimputer', IterativeImputer()),                  |  |  |
| Elastic_net   | ElasticNet(alpha=0.0001, I1_ratio=0.4)                                     |  |  |
|               | DecisionTreeRegressor(max_depth=4, max_features=0.2, min_samples_leaf=0.1, |  |  |
| decision_tree | random_state=1)  |  |  |

| model              | model_definition   |  |  |  |
|--------------------|--|--|--|--|
|                    | RandomForestRegressor(max_depth=4, max_features=0.4, n_estimators=200,             |  |  |  |
| random_forest_reg1 | n_jobs=-1, random_state=1)   |  |  |  |
|                    | RandomForestRegressor(max_depth=10, max_features=0.4, n_estimators=2               |  |  |  |
| random_forest_reg2 | n_jobs=-1, random_state=1)   |  |  |  |
|                    | RandomForestRegressor(max_depth=7, max_features=0.3, n_jobs=-1,                    |  |  |  |
| random_forest_reg3 | random_state=1)  |  |  |  |
|                    | RandomForestRegressor(max_depth=8, max_features=0.6, n_estimators=200,             |  |  |  |
| random_forest_reg4 | random_state=1)  |  |  |  |
| gradien_boost_1    | GradientBoostingRegressor(n_estimators=150, random_state=1)                        |  |  |  |
|                    | GradientBoostingRegressor(learning_rate=0.0824999999999999999999999999999999999999 |  |  |  |
|                    | max_depth=10,  |  |  |  |
|                    | max_features=0.60000000000001, min_samples_leaf=8,                                 |  |  |  |
| gradien_boost_2    | min_samples_split=10, n_estimators=118)  |  |  |  |
|                    | GradientBoostingRegressor(learning_rate=0.0824999999999999, max_depth=7,           |  |  |  |
|                    | max_features=0.8, min_samples_leaf=4,  |  |  |  |
|                    | min_samples_split=12, n_estimators=150,  |  |  |  |
| gradien_boost_3    | random_state=1)  |  |  |  |
|                    | XGBRegressor(base_score=0.5, booster='gbtree', colsample_bylevel=1,                |  |  |  |
|                    | colsample_bynode=1, colsample_bytree=1, enable_categorical=False,                  |  |  |  |
|                    | gamma=0, gpu_id=-1, importance_type=None,  |  |  |  |
|                    | interaction_constraints=", learning_rate=0.04, max_delta_step=0,                   |  |  |  |
|                    | max_depth=5, min_child_weight=1, missing=nan,                                      |  |  |  |
|                    | monotone_constraints='()', n_estimators=200, n_jobs=-1,                            |  |  |  |
|                    | num_parallel_tree=1, predictor='auto', random_state=0, reg_alpha=0,                |  |  |  |
|                    | reg_lambda=1, scale_pos_weight=1, subsample=1, tree_method='exact',                |  |  |  |
| XGBRegressor       | validate_parameters=1, verbosity=None)   |  |  |  |
|                    | VotingRegressor(estimators=[('gb',   |  |  |  |
|                    | GradientBoostingRegressor(random_state=47)),                                       |  |  |  |
|                    | ('rf', RandomForestRegressor(random_state=47)),                                    |  |  |  |
| voting             | ('Ir', LinearRegression())])   |  |  |  |

# Modeling





#### R square score in train set and test set

| model            | R2_tr | R2_tt |
|------------------|-------|-------|
| gradien_boost_2  | 0.999 | 0.950 |
| gradien_boost_3  | 0.997 | 0.947 |
| random_forest_re |       |       |
| g2               | 0.986 | 0.945 |
| random_forest_re |       |       |
| g4               | 0.976 | 0.942 |
| XGBRegressor     | 0.982 | 0.941 |
| random_forest_re |       |       |
| g3               | 0.962 | 0.931 |
| voting           | 0.965 | 0.930 |
| gradien_boost_1  | 0.970 | 0.929 |
| random_forest_re |       |       |
| g1               | 0.919 | 0.900 |
| linear_reg2      | 0.882 | 0.868 |
| linear_reg       | 0.882 | 0.868 |
| ridge_reg        | 0.882 | 0.868 |
| Elastic_net      | 0.882 | 0.863 |
| decision_tree    | 0.743 | 0.682 |

### **Conclusion**

- Life expectancy has increased over years in both developed and developing countries
- The mean average of the life expectancy of developed countries is generally higher compared to that of developing countries
- However, the ratio of LE over the decade of 2005 to 2015 showed that life expectancy in developing countries has greatly increased.
- It has been highlighted that immunization has impacted the improvement of life expectancy in a developing country, as well as the reduction in infant deaths.
- The analysis revealed that economic factors play an important role in the system, it is why countries
  with higher income resources and GDP tend to have high life expectancy even if the population is big.
  In developing countries, an increase in the population tends to impact negatively life expectancy.
- Many (14) regression models have been developed to predict expectancy, the chosen one is Gradient boost with MAE of 0.202 on train set and 1.431 on the test set, R square is 0.94 on the test set.