MidTerm Project

Predicting Life Expectancy using Linear Regression

Oct 30th, 2024 Yoji

Agenda

- 1. Dataset overview
- 2. Key findings from EDA
- 3. Feature scaling
- 4. Regression models
- 5. Cross validation
- 6. Challenges

Dataset overview

Dataset information

- Dataset name: Life Expectancy (WHO)
- Dataset link: https://www.kaggle.com/datasets/kumarajarshi/life-expectancywho/data

Dataset size

- 22 columns
- 2938 records

Key findings from EDA

Some data had null-values so I had to fill up these data

- Apply mean imputation for int and float columns.
- Apply mode imputation for object columns.

Perform one-hot-encoding for country column.

Feature scaling

Correlation matrix

Find out top 5 features based on the correlation matrix.

```
Life expectancy 1.000000
Schooling 0.715066
Adult Mortality 0.696359
Income composition of resources 0.692483
BMI 0.559255
HIV/AIDS 0.556457
```

Feature scaling

For this mid-term-project, I used 3 type feature scaling and 2 types of dataset. So totaling, it is 6 kind of feature scaling

- StandardScaler with entire data
- StandardScaler with top5-feature
- MinMaxScaler with entire data
- MinMaxScaler with top5-feature
- RobustScaler with entire data
- RobustScaler with top5-feature

And, compare result

Regression models

Train-test split

Before starting to make models, split data into train-data and test-data. This time I split data into 60%-40% and 80%-20% in order to compare results.

Perform PCA

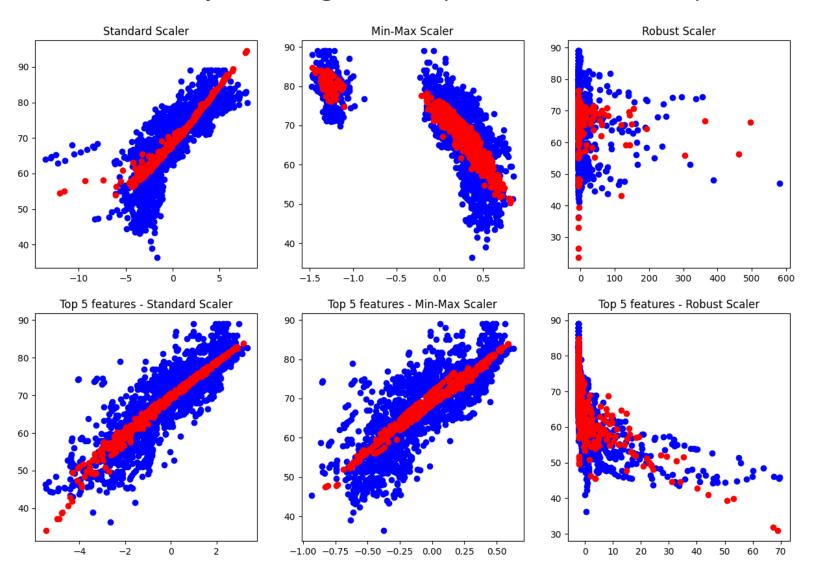
In order to warp into 2d numbers, perform PCA.

Make models

Make models using with data which was performed PCA.

Prediction

These scatter plot shows train data and prediction data. x-axis:features, y-axis:target, blue-plot:train-data, red-plot:test-data



Evaluation with MAE, MSE, RMSE, and R2

Train-data:60%, Test-data:40%

```
Model
                                    MAE
                                               MSE
                                                         RMSE
                                                                     R2
               Standard Scaler 4.558195
                                         32.513194
                                                     5.702034
                                                               0.641470
                Min-Max Scaler 9.579367
                                         139.613058 11.815797 -0.539541
                 Robust Scaler 8.596135
                                         125.391163 11.197820 -0.382714
Top5 features - Standard Scaler 9.791548
                                         154.255161 12.419950 -0.701003
Top5 features - Min-Max Scaler
                               9.992961
                                         156.235747 12.499430 -0.722843
 Top5 features - Robust Scaler
                               9.977972
                                         160.161032 12.655474 -0.766128
```

Train-data:80%, Test-data:20%

Model	MAE	MSE	RMSE	R2
Standard Scaler	4.545390	32.062129	5.662343	0.636536
Min-Max Scaler	9.991245	157.483100	12.549227	-0.785267
Robust Scaler	8.077685	114.153075	10.684244	-0.294067
Top5 features - Standard Scaler	9.796859	156.941338	12.527623	-0.779125
Top5 features - Min-Max Scaler	9.770935	150.164734	12.254172	-0.702304
Top5 features - Robust Scaler	9.849289	158.176033	12.576805	-0.793122

Cross validation

Perform KFold cross-validation(splits: 5)

```
Model Score(MSE)
Standard Scaler 32.398139
Min-Max Scaler 36.830030
Robust Scaler 55.933006
Top5 features - Standard Scaler 20.863351
Top5 features - Min-Max Scaler 28.347514
Top5 features - Robust Scaler 21.614507
```

Challenges

There was a lot of way to do feature scaling, split-data, and extract data. But I couldn't which way is most proper. So this time, I tried to perform everyway as far as I could and compare results. After all, I could figure out which one is most-fit-way.