## Assignment\_20\_ML on different dataset with commenting

## January 22, 2023

```
[]: # Import the necessary libraries
     # import libraries
     import pandas as pd
     import numpy as np
     import seaborn as sns
     import matplotlib.pyplot as plt
[]: # We have imported dataset of world
     df = pd.read_csv('world_population.csv')
[]: # check the data for understanding of data
[]:
          Rank CCA3
                      Country/Territory
                                                    Capital Continent
            36
                AFG
                                                      Kabul
                                                                 Asia
     0
                            Afghanistan
     1
           138
                ALB
                                Albania
                                                    Tirana
                                                               Europe
     2
                DZA
            34
                                Algeria
                                                   Algiers
                                                               Africa
     3
           213
                ASM
                         American Samoa
                                                 Pago Pago
                                                              Oceania
     4
           203
                AND
                                          Andorra la Vella
                                Andorra
                                                               Europe
     229
           226
                WLF
                      Wallis and Futuna
                                                  Mata-Utu
                                                              Oceania
     230
           172
                ESH
                         Western Sahara
                                                               Africa
                                                  El Aaiún
     231
            46
                YEM
                                  Yemen
                                                      Sanaa
                                                                 Asia
     232
            63
                                 Zambia
                                                               Africa
                7.MB
                                                    Lusaka
     233
            74
                ZWE
                               Zimbabwe
                                                    Harare
                                                               Africa
                                                                2010 Population \
          2022 Population
                           2020 Population
                                              2015 Population
     0
                                                                        28189672
                  41128771
                                    38972230
                                                      33753499
     1
                   2842321
                                     2866849
                                                       2882481
                                                                         2913399
     2
                  44903225
                                    43451666
                                                      39543154
                                                                        35856344
     3
                     44273
                                       46189
                                                         51368
                                                                           54849
                     79824
     4
                                       77700
                                                         71746
                                                                           71519
     229
                     11572
                                       11655
                                                         12182
                                                                           13142
                                                                          413296
     230
                                                        491824
                    575986
                                      556048
     231
                  33696614
                                    32284046
                                                      28516545
                                                                        24743946
     232
                  20017675
                                    18927715
                                                      16248230
                                                                        13792086
```

233	16320537	1566966	66 14	154937	12839771	
	2000 Population	1990 Populatio	on 1980 Popu	lation	1970 Population	\
0	19542982	1069479	96 12	486631	10752971	
1	3182021	329506	66 2	941651	2324731	
2	30774621	2551807	74 18	739378	13795915	
3	58230	478:	18	32886	27075	
4	66097	5356	69	35611	19860	
	•••	•••	•••		•••	
229	14723	1349	54	11315	9377	
230	270375	17852	29	116775	76371	
231	18628700	1337512	21 9	204938	6843607	
232	9891136	768640	01 5	720438	4281671	
233	11834676	1011389	93 7	049926	5202918	
		sity (per km²)	Growth Rate	World	Population Perce	ntage
0	652230	63.0587	1.0257			0.52
1	28748	98.8702	0.9957			0.04
2	2381741	18.8531	1.0164			0.56
3	199	222.4774	0.9831			0.00
4	468	170.5641	1.0100			0.00
	***	***	•••		***	
229	142	81.4930	0.9953			0.00
230	266000	2.1654	1.0184			0.01
231	527968	63.8232	1.0217			0.42
232	752612	26.5976	1.0280			0.25
233	390757	41.7665	1.0204			0.20

[234 rows x 17 columns]

## []: # Verified the data type and Null values df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 234 entries, 0 to 233
Data columns (total 17 columns):

#	Column	Non-Null Count	Dtype	
"	OOT UIIII	NON NULL COUNT	Боурс	
0	Rank	234 non-null	int64	
1	CCA3	234 non-null	object	
2	Country/Territory	234 non-null	object	
3	Capital	234 non-null	object	
4	Continent	234 non-null	object	
5	2022 Population	234 non-null	int64	
6	2020 Population	234 non-null	int64	
7	2015 Population	234 non-null	int64	
8	2010 Population	234 non-null	int64	
9	2000 Population	234 non-null	int64	

```
10 1990 Population
                                      234 non-null
                                                      int64
                                                      int64
     11 1980 Population
                                     234 non-null
     12 1970 Population
                                      234 non-null
                                                      int64
     13 Area (km<sup>2</sup>)
                                      234 non-null
                                                      int64
     14 Density (per km<sup>2</sup>)
                                      234 non-null
                                                      float64
     15 Growth Rate
                                      234 non-null
                                                      float64
     16 World Population Percentage 234 non-null
                                                      float64
    dtypes: float64(3), int64(10), object(4)
    memory usage: 31.2+ KB
[]: # Printed columns name for X and y
     df.columns
[]: Index(['Rank', 'CCA3', 'Country/Territory', 'Capital', 'Continent',
            '2022 Population', '2020 Population', '2015 Population',
            '2010 Population', '2000 Population', '1990 Population',
            '1980 Population', '1970 Population', 'Area (km²)', 'Density (per km²)',
            'Growth Rate', 'World Population Percentage'],
           dtype='object')
[]: # defining variables for X and y
     X = df[[
            '2022 Population', '2020 Population', '2015 Population',
            '2010 Population', '2000 Population', '1990 Population']]
     a = df['World Population Percentage'] # y is replaced with a in this step and_
      ⇒in next
[]: # Imported required Libraries for Machine learning
     from sklearn.linear_model import LogisticRegression
     from sklearn.svm import SVC
     from sklearn.tree import DecisionTreeClassifier
     from sklearn.ensemble import RandomForestClassifier
     from sklearn.neighbors import KNeighborsClassifier
     from sklearn.metrics import accuracy score, f1 score, precision score,
      →recall_score
     from sklearn.model_selection import train_test_split
[]: # We changed the values by encoding as the y is countinous
     from sklearn import preprocessing
     from sklearn import utils
     lab = preprocessing.LabelEncoder()
     y = lab.fit_transform(a)
[]: y
[]: array([39, 4, 40, 0, 0, 34, 0, 0, 41, 3, 0, 27, 10, 12, 1, 2, 62,
            0, 11, 14, 1, 16, 0, 1, 14, 4, 3, 63, 0, 1, 9, 25, 15, 18,
            28, 37, 1, 0, 7, 19, 22, 69, 44, 1, 0, 6, 5, 13, 0, 2, 12,
```

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7, 1, 0, 13, 56, 20, 57, 8, 2, 5, 2, 2, 59, 0, 0, 1, 7,
           48, 0, 0, 3, 3, 5, 52, 32, 0, 12,
                                                 0, 0, 0, 0, 19, 0, 16,
            3, 1, 14, 12, 9, 11, 0, 68, 66, 54, 40, 6, 0, 10, 46, 28,
           59, 0, 13, 21, 45, 0, 5, 8, 9, 2, 7, 3, 7, 9, 0, 3,
            1, 29, 23, 33, 1, 25, 1, 0, 0, 6, 2, 0, 60, 0, 4, 0, 4,
            1, 0, 36, 31, 45, 3, 0, 30, 19, 0, 7, 9, 27, 64, 0, 27,
            0, 7, 6, 65, 0, 7, 6, 12, 9, 33, 58, 38, 12, 4, 3, 7,
           22, 61, 16, 0, 0, 0, 0, 0, 0, 0, 0, 35, 19, 9, 0, 10,
            7, 0, 7, 3, 1, 19, 47, 44, 13, 43, 24, 42, 1, 12, 10, 25, 26,
           11, 49, 51, 2, 10, 0, 0, 2, 14, 53, 8, 0, 0, 42, 38, 11, 50,
           67, 0, 4, 33, 0, 0, 28, 55, 0, 1, 32, 22, 17], dtype=int64)
[]: #Splited the dataset in two parts for test and train, we used the random state
     ⇔to get \
    # the same results each time, if we select none everytime results will be
     X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,_
      →random_state=42)
[]: # we selected the Machine leaning models that we can use
    models = [LogisticRegression(), SVC(), DecisionTreeClassifier(),
     →RandomForestClassifier(), KNeighborsClassifier()]
    model_names = ['Logistic Regression', 'SVM', 'Decision Tree', 'Random Forest', |
    # this we loopthrough each model and save the accuracy score in model name
    models scores = []
    for model, model name in zip(models, model names):
        model.fit(X_train, y_train)
        y_pred = model.predict(X_test)
        accuracy = accuracy_score(y_test, y_pred)
        models_scores.append([model_name,accuracy])
    \# used the lambda function to loopthroug each score type and sort from a to z
    sorted_models = sorted(models_scores, key=lambda x: x[1], reverse=True)
    for model in sorted_models:
        print("Accuracy Score: ",f'{model[0]} : {model[1]:.2f}')
    c:\Users\muham\AppData\Local\Programs\Python\Python311\Lib\site-
    packages\sklearn\linear_model\_logistic.py:458: ConvergenceWarning: lbfgs failed
    to converge (status=1):
    STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
    Increase the number of iterations (max iter) or scale the data as shown in:
       https://scikit-learn.org/stable/modules/preprocessing.html
    Please also refer to the documentation for alternative solver options:
       https://scikit-learn.org/stable/modules/linear_model.html#logistic-
    regression
      n_iter_i = _check_optimize_result(
```

```
Accuracy Score: Random Forest: 0.60
    Accuracy Score: Decision Tree: 0.57
    Accuracy Score: KNN: 0.40
    Accuracy Score: SVM : 0.15
    Accuracy Score: Logistic Regression: 0.00
[]: models = [LogisticRegression(), SVC(), DecisionTreeClassifier(),
      →RandomForestClassifier(), KNeighborsClassifier()]
    model_names = ['Logistic Regression', 'SVM', 'Decision Tree', 'Random Forest',

¬'KNN']
    models_scores = []
    for model, model_name in zip(models, model_names):
        model.fit(X train, y train)
        y_pred = model.predict(X_test)
        Precision = precision_score(y_test, y_pred, average='micro') # Included_
      →average='micro' for calcluating averaging score of each value
        models_scores.append([model_name,Precision])
    sorted_models = sorted(models_scores, key=lambda x: x[1], reverse=True)
    for model in sorted_models:
        print("Precision Score: ", f'{model[0]} : {model[1]:.2f}')
    c:\Users\muham\AppData\Local\Programs\Python\Python311\Lib\site-
    packages\sklearn\linear_model\_logistic.py:458: ConvergenceWarning: lbfgs failed
    to converge (status=1):
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    Increase the number of iterations (max_iter) or scale the data as shown in:
        https://scikit-learn.org/stable/modules/preprocessing.html
    Please also refer to the documentation for alternative solver options:
        https://scikit-learn.org/stable/modules/linear_model.html#logistic-
    regression
      n_iter_i = _check_optimize_result(
    Precision Score: Random Forest: 0.62
    Precision Score: Decision Tree: 0.60
    Precision Score: KNN: 0.40
    Precision Score: SVM: 0.15
    Precision Score: Logistic Regression: 0.00
[]: models = [LogisticRegression(), SVC(), DecisionTreeClassifier(),
     →RandomForestClassifier(), KNeighborsClassifier()]
    model_names = ['Logistic Regression', 'SVM', 'Decision Tree', 'Random Forest',
      ∽'KNN']
    models scores = []
    for model, model_name in zip(models, model_names):
        model.fit(X_train, y_train)
        y_pred = model.predict(X_test)
```

```
Recall = recall_score(y_test, y_pred, average='micro')
        models_scores.append([model_name,Recall])
     sorted_models = sorted(models_scores, key=lambda x: x[1], reverse=True)
     for model in sorted_models:
        print("Precision Score: ", f'{model[0]} : {model[1]:.2f}')
    c:\Users\muham\AppData\Local\Programs\Python\Python311\Lib\site-
    packages\sklearn\linear_model\_logistic.py:458: ConvergenceWarning: lbfgs failed
    to converge (status=1):
    STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
    Increase the number of iterations (max_iter) or scale the data as shown in:
        https://scikit-learn.org/stable/modules/preprocessing.html
    Please also refer to the documentation for alternative solver options:
        https://scikit-learn.org/stable/modules/linear_model.html#logistic-
    regression
      n_iter_i = _check_optimize_result(
    Precision Score: Decision Tree: 0.60
    Precision Score: Random Forest: 0.60
    Precision Score: KNN: 0.40
    Precision Score: SVM: 0.15
    Precision Score: Logistic Regression: 0.00
[]: models = [LogisticRegression(), SVC(), DecisionTreeClassifier(),
     →RandomForestClassifier(), KNeighborsClassifier()]
     model_names = ['Logistic Regression', 'SVM', 'Decision Tree', 'Random Forest', |
      ⇔'KNN']
     models_scores = []
     for model, model_name in zip(models, model_names):
        model.fit(X_train, y_train)
        y_pred = model.predict(X_test)
        F1 = f1_score(y_test, y_pred, average='micro')
        models_scores.append([model_name,F1])
     sorted_models = sorted(models_scores, key=lambda x: x[1], reverse=True)
     for model in sorted_models:
        print("F1 Score: ",f'{model[0]} : {model[1]:.2f}')
    c:\Users\muham\AppData\Local\Programs\Python\Python311\Lib\site-
    packages\sklearn\linear_model\_logistic.py:458: ConvergenceWarning: lbfgs failed
    to converge (status=1):
    STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
    Increase the number of iterations (max_iter) or scale the data as shown in:
        https://scikit-learn.org/stable/modules/preprocessing.html
    Please also refer to the documentation for alternative solver options:
        https://scikit-learn.org/stable/modules/linear_model.html#logistic-
```

## regression

n\_iter\_i = \_check\_optimize\_result(

F1 Score: Random Forest : 0.60 F1 Score: Decision Tree : 0.55

F1 Score: KNN : 0.40 F1 Score: SVM : 0.15

F1 Score: Logistic Regression: 0.00