

## Assignment 3: Individual Assignment

# Remmy Bisimbeko - B26099 - J24M19/011

My GitHub - <https://github.com/RemmyBisimbeko/Data-Science>

Instructions:

The National Coffee Research Institute generated disease resistant coffee varieties which they analysed for the quality using a cupping assessment.

You will use the provided Excel workbook labelled "Uganda Coffee\_Cupping.xlsx" Uganda Coffee\_Cupping.xlsx containing three sheets:

a) Sheet 1; Training Data b) Sheet 2: Test Data 1 c) Sheet 3: Test Data 2

Transform the datasets and make predictions on the "OVERALL SCORE" of coffee in two different areas of ORIGIN;

1. Kayunga (10 MARKS)
2. Rwenzori (10 MARKS)

```
In [ ]: # Import Libs
import pandas as pd
import numpy as np
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import train_test_split
from sklearn.metrics import mean_squared_error

In [ ]: # loading the data into a DataFrame
df = pd.read_excel('Data Sets/Uganda Coffee_Cupping.xlsx', sheet_name='Tr

In [ ]: # Preprocess the data
# Drop any unnecessary columns - or delete manually from excel
# df = df.drop(['Unnamed: 0'], axis=1)

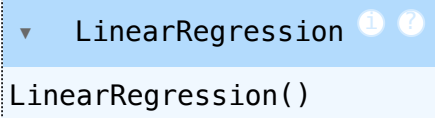
In [ ]: # I now convert categorical variables to dummy variables
df = pd.get_dummies(df, columns=['ORIGIN', 'VARIETY'])

In [ ]: # Splitting the data into features (X) and target (y)
X = df.drop(['OVERALL SCORE'], axis=1)
y = df['OVERALL SCORE']

In [ ]: # Exclude the columns that are not present in the test data
X = X.drop(['ORIGIN_Ibanda', 'ORIGIN_Mityana', 'ORIGIN_Mukono'], axis=1)

In [ ]: # Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,

In [ ]: # Train machine learning model
model = LinearRegression()
model.fit(X_train, y_train)
```

Out [ ]:  LinearRegression()

```
In [ ]: # generate predictions for the Kayunga region

kayunga_df = pd.read_excel('Data Sets/Uganda Coffee Cupping.xlsx', sheet_
# kayunga_df = kayunga_df.drop(['Unnamed: 0'], axis=1)
```

```
In [ ]: # Convert categorical variables to dummy variables with the same prefix a
kayunga_df = pd.get_dummies(kayunga_df, columns=['ORIGIN', 'VARIETY'], pr
```

```
In [ ]: # Then re-order the columns to match the training data
kayunga_df = kayunga_df[X_train.columns]
```

```
In [ ]: print(df.columns)
df.dtypes

Index(['FRAGRANCE/AROMA', 'FLAVOR', 'SALT/ ACID', 'BITTER/ SWEET',
      'AFTERTASTE', 'MOUTH FEEL', 'BALANCE', 'UNIFORMITY', 'CLEAN CUPS',
      'OVERALL SCORE', 'ORIGIN_Ibanda', 'ORIGIN_Mityana', 'ORIGIN_Mukon
o',
      'VARIETY_KR3', 'VARIETY_KR4', 'VARIETY_KR5', 'VARIETY_KR6',
      'VARIETY_KR7'],
      dtype='object')
```

```
Out [ ]: FRAGRANCE/AROMA    float64
FLAVOR                    float64
SALT/ ACID                float64
BITTER/ SWEET             float64
AFTERTASTE                float64
MOUTH FEEL                float64
BALANCE                   float64
UNIFORMITY                 int64
CLEAN CUPS                 int64
OVERALL SCORE              float64
ORIGIN_Ibanda              bool
ORIGIN_Mityana             bool
ORIGIN_Mukono              bool
VARIETY_KR3                bool
VARIETY_KR4                bool
VARIETY_KR5                bool
VARIETY_KR6                bool
VARIETY_KR7                bool
dtype: object
```

```
In [ ]: kayunga_df['OVERALL SCORE'] = np.nan

for index, row in kayunga_df.iterrows():
    # Calculate the overall score for this row based on the other columns
    kayunga_df.loc[index, 'OVERALL SCORE'] = row[['FRAGRANCE/AROMA', 'FLA

print(kayunga_df)
```

	FRAGRANCE/AROMA	FLAVOR	SALT/	ACID	BITTER/	SWEET	AFTERTASTE	\
0	7.00	6.00		6.50		6.75	6.50	
1	8.00	7.75		7.00		7.00	7.25	
2	8.25	7.25		7.00		7.00	6.75	
3	7.50	7.75		7.50		7.50	7.25	
4	7.00	7.25		8.00		8.00	7.75	
5	8.00	7.00		7.00		7.00	7.50	
6	8.00	8.00		7.00		7.00	7.25	
7	8.00	7.75		7.50		7.50	7.75	
8	7.50	7.75		7.00		7.25	7.25	
9	7.00	8.00		7.50		8.00	7.00	
10	7.25	7.00		7.00		7.00	6.50	
11	6.50	7.00		6.75		7.00	6.00	
12	7.75	7.00		7.00		7.00	7.00	
13	7.00	7.00		7.00		7.00	7.00	
14	7.00	7.25		7.00		7.25	7.25	
15	7.75	7.00		6.75		6.50	7.25	
16	8.25	7.75		7.00		7.00	7.25	
17	7.00	7.50		7.50		7.25	7.50	
18	7.25	7.50		7.00		7.25	7.00	
19	7.25	8.00		7.75		8.00	8.00	
20	7.50	7.00		7.00		7.00	7.00	
21	8.00	7.75		7.00		7.25	7.75	
22	8.00	7.50		7.25		7.50	7.00	
23	7.25	7.50		7.00		7.25	7.50	
24	7.00	7.00		6.75		6.75	7.00	

	MOUTH FEEL	BALANCE	UNIFORMITY	CLEAN	CUPS	VARIETY_KR3	VARIETY_KR4
\							
0	6.75	8.00	10		10	True	False
1	7.00	7.00	10		10	True	False
2	6.75	7.00	10		10	True	False
3	7.75	7.50	10		10	True	False
4	7.75	7.00	10		10	True	False
5	7.00	7.75	10		10	False	True
6	7.00	7.00	10		10	False	True
7	7.75	7.25	10		10	False	True
8	7.25	7.25	10		10	False	True
9	7.75	7.50	10		10	False	True
10	6.75	7.00	8		8	False	False
11	6.50	7.00	8		8	False	False
12	7.00	7.00	8		8	False	False
13	7.00	8.00	8		8	False	False
14	7.25	7.00	8		8	False	False
15	6.50	7.00	10		10	False	False
16	7.00	7.00	10		10	False	False
17	7.50	7.25	10		10	False	False
18	7.50	7.50	10		10	False	False
19	7.75	7.75	10		10	False	False
20	6.75	8.00	10		10	False	False
21	7.00	7.00	10		10	False	False
22	7.00	7.00	10		10	False	False
23	7.50	7.25	10		10	False	False
24	7.00	7.00	10		10	False	False

	VARIETY_KR5	VARIETY_KR6	VARIETY_KR7	OVERALL SCORE
0	False	False	False	7.500000
1	False	False	False	7.888889
2	False	False	False	7.777778
3	False	False	False	8.083333

4	False	False	False	8.083333
5	False	False	False	7.916667
6	False	False	False	7.916667
7	False	False	False	8.166667
8	False	False	False	7.916667
9	False	False	False	8.083333
10	True	False	False	7.166667
11	True	False	False	6.972222
12	True	False	False	7.305556
13	True	False	False	7.333333
14	True	False	False	7.333333
15	False	True	False	7.638889
16	False	True	False	7.916667
17	False	True	False	7.944444
18	False	True	False	7.888889
19	False	True	False	8.277778
20	False	False	True	7.805556
21	False	False	True	7.972222
22	False	False	True	7.916667
23	False	False	True	7.916667
24	False	False	True	7.611111

```
In [ ]: kayunga_predictions = model.predict(kayunga_df.drop(['OVERALL SCORE'], ax
# Final predictions Displayed at the bottom
```

```
In [ ]: # NOW I generate predictions for the Rwenzori region
rwezori_df = pd.read_excel('Data Sets/Uganda Coffee_Cupping.xlsx', sheet
# rwezori_df = rwezori_df.drop(['Unnamed: 0'], axis=1)
```

```
In [ ]: # Convert categorical variables to dummy variables with the same prefix a
rwezori_df = pd.get_dummies(rwezori_df, columns=['ORIGIN', 'VARIETY'],
```

```
In [ ]: # I did the reorder the columns to match the training data
rwezori_df = rwezori_df[X_train.columns]
```

```
In [ ]: # How's data lookin like?
print(df.columns)
df.dtypes
```

```
Index(['FRAGRANCE/AROMA', 'FLAVOR', 'SALT/ ACID', 'BITTER/ SWEET',
      'AFTERTASTE', 'MOUTH FEEL', 'BALANCE', 'UNIFORMITY', 'CLEAN CUPS',
      'OVERALL SCORE', 'ORIGIN_Ibanda', 'ORIGIN_Mityana', 'ORIGIN_Mukono',
      'VARIETY_KR3', 'VARIETY_KR4', 'VARIETY_KR5', 'VARIETY_KR6',
      'VARIETY_KR7'],
      dtype='object')
```

```
Out [ ]: FRAGRANCE/AROMA    float64
          FLAVOR            float64
          SALT/ ACID        float64
          BITTER/ SWEET     float64
          AFTERTASTE        float64
          MOUTH FEEL        float64
          BALANCE           float64
          UNIFORMITY        int64
          CLEAN CUPS        int64
          OVERALL SCORE     float64
          ORIGIN_Ibanda     bool
          ORIGIN_Mityana    bool
          ORIGIN_Mukono     bool
          VARIETY_KR3        bool
          VARIETY_KR4        bool
          VARIETY_KR5        bool
          VARIETY_KR6        bool
          VARIETY_KR7        bool
          dtype: object
```

```
In [ ]: # Fix for Overall Score error
        rwenzori_df['OVERALL SCORE'] = np.nan

        for index, row in rwenzori_df.iterrows():
            # Calculate the overall score for this row based on the other columns
            rwenzori_df.loc[index, 'OVERALL SCORE'] = row[['FRAGRANCE/AROMA', 'FLAVOR', 'SALT/ ACID', 'BITTER/ SWEET', 'AFTERTASTE', 'MOUTH FEEL', 'BALANCE']]

        print(rwenzori_df)
```

	FRAGRANCE/AROMA	FLAVOR	SALT/	ACID	BITTER/	SWEET	AFTERTASTE	\
0	7.25	6.75		6.75		7.50	6.75	
1	7.50	7.75		7.50		7.75	7.50	
2	7.55	8.00		7.00		7.75	8.00	
3	8.00	7.00		6.00		7.00	7.00	
4	6.50	7.00		6.75		7.00	7.00	
5	7.00	7.00		7.00		7.25	7.00	
6	7.00	7.75		7.75		7.75	7.50	
7	7.25	8.00		7.00		7.25	8.25	
8	8.00	7.00		7.00		6.00	7.00	
9	7.00	7.00		7.00		7.00	7.00	
10	6.75	7.00		6.75		7.25	7.00	
11	7.00	7.50		7.00		7.50	7.25	
12	6.75	7.75		8.00		7.75	8.00	
13	8.00	6.50		7.00		7.00	6.50	
14	7.25	7.00		7.50		7.50	7.25	
15	7.75	7.75		7.25		7.50	7.50	
16	7.25	7.50		7.75		7.25	7.25	
17	7.25	7.00		7.00		7.50	7.00	
18	7.00	8.00		6.75		7.00	7.50	
19	8.00	7.75		7.00		7.00	7.00	
20	7.25	7.00		7.25		7.00	7.00	
21	7.00	7.50		7.75		7.50	7.50	
22	7.60	7.60		7.00		6.75	7.25	
23	7.00	6.50		6.75		7.00	7.00	
24	7.50	7.50		7.00		7.00	7.50	

	MOUTH FEEL	BALANCE	UNIFORMITY	CLEAN	CUPS	VARIETY_KR3	VARIETY_KR4
\							
0	6.75	6.75	10		10	True	False
1	7.50	7.50	10		10	True	False
2	8.00	7.75	10		10	True	False
3	7.00	10.00	10		10	True	False
4	7.00	7.00	10		10	True	False
5	7.25	7.50	10		10	False	True
6	7.25	7.75	10		10	False	True
7	8.00	7.50	10		10	False	True
8	7.00	10.00	10		10	False	True
9	7.00	6.75	10		10	False	True
10	7.25	6.75	10		10	False	False
11	7.25	7.50	10		10	False	False
12	7.75	7.75	10		10	False	False
13	6.75	7.50	10		10	False	False
14	7.25	7.50	10		10	False	False
15	7.00	7.75	10		10	False	False
16	8.00	7.50	10		10	False	False
17	7.00	7.00	10		10	False	False
18	7.00	7.75	10		10	False	False
19	7.00	7.50	10		10	False	False
20	7.00	7.50	10		10	False	False
21	7.50	7.25	10		10	False	False
22	7.00	7.00	10		10	False	False
23	7.00	10.00	10		10	False	False
24	7.50	7.50	10		10	False	False

	VARIETY_KR5	VARIETY_KR6	VARIETY_KR7	OVERALL SCORE
0	False	False	False	7.611111
1	False	False	False	8.111111
2	False	False	False	8.227778
3	False	False	False	8.000000

4	False	False	False	7.583333
5	False	False	False	7.777778
6	False	False	False	8.083333
7	False	False	False	8.138889
8	False	False	False	8.000000
9	False	False	False	7.638889
10	True	False	False	7.638889
11	True	False	False	7.888889
12	True	False	False	8.194444
13	True	False	False	7.694444
14	True	False	False	7.916667
15	False	True	False	8.055556
16	False	True	False	8.055556
17	False	True	False	7.750000
18	False	True	False	7.888889
19	False	True	False	7.916667
20	False	False	True	7.777778
21	False	False	True	8.000000
22	False	False	True	7.800000
23	False	False	True	7.916667
24	False	False	True	7.944444

```
In [ ]: rwenzori_predictions = model.predict(rwenzori_df.drop(['OVERALL SCORE'],
# Final predictions all displayed at the bottom
```

```
In [ ]: # Finally, display the final result
print("Predictions for Kayunga:")
print(kayunga_predictions)
print("\nPredictions for Rwenzori:")
print(rwenzori_predictions)
```

Predictions for Kayunga:

```
[7.0234375  7.7734375  7.5234375  7.6484375  7.7734375  7.7734375
 7.7734375  8.0234375  7.6484375  7.2734375  6.8984375  6.6484375
 7.3984375  7.5234375  7.3984375  7.56054688  7.68554688  7.43554688
 7.18554688  8.06054688  7.43945312  7.81445312  7.31445312  7.43945312
 7.06445312]
```

Predictions for Rwenzori:

```
[7.0234375  7.8984375  8.1484375  7.8984375  7.1484375  7.2734375
 7.7734375  8.1484375  8.0234375  7.1484375  6.8984375  7.2734375
 7.8984375  6.8984375  7.2734375  7.81054688  7.31054688  7.06054688
 7.68554688  7.43554688  7.18945312  7.43945312  7.56445312  7.56445312
 7.56445312]
```