

Practical No.3

Aim:- Feature scaling dummification.

- Apply feature-scaling techniques like standardization and normalization to numerical features.
- Perform feature dummification to convert categorical variables into numerical representations.

Importing Libraries

```
In [2]: import pandas as pd
from sklearn.preprocessing import StandardScaler, MinMaxScaler, OneHotEncoder
from sklearn.compose import ColumnTransformer
```

Define the data

```
In [3]: data = {
    'Product': ['Apple_Juice', 'Banana_Smoothie', 'Orange_Jam', 'Grape_Jelly',
               'Kiwi_Juice', 'Mango_Pickle', 'Pineapple_Sorbet',
               'Strawberry_Yoghurt', 'Blueberry_Pie', 'Cherry_Salsa'],
    'Category': ['Apple', 'Banana', 'Orange', 'Grape', 'Kiwi', 'Mango',
                'Pineapple', 'Strawberry', 'Blueberry', 'Cherry'],
    'Sales': [1200, 1700, 2200, 1400, 2000, 1000, 1500, 1800, 1300, 1600],
    'Cost': [600, 850, 1100, 700, 1000, 500, 750, 900, 650, 800],
    'Profit': [600, 850, 1100, 700, 1000, 500, 750, 900, 650, 800]
}
```

```
In [4]: #Create a DataFrame
df = pd.DataFrame(data)
#Display the original dataset
print("Original Dataset:")
print(df)
```

Original Dataset:

	Product	Category	Sales	Cost	Profit
0	Apple_Juice	Apple	1200	600	600
1	Banana_Smoothie	Banana	1700	850	850
2	Orange_Jam	Orange	2200	1100	1100
3	Grape_Jelly	Grape	1400	700	700
4	Kiwi_Juice	Kiwi	2000	1000	1000
5	Mango_Pickle	Mango	1000	500	500
6	Pineapple_Sorbet	Pineapple	1500	750	750
7	Strawberry_Yoghurt	Strawberry	1800	900	900
8	Blueberry_Pie	Blueberry	1300	650	650
9	Cherry_Salsa	Cherry	1600	800	800

Feature Scaling(Standardization and Normalization)

```
In [5]: numeric_columns = ['Sales', 'Cost', 'Profit']
        scaler_std = StandardScaler()
        scaler_normal = MinMaxScaler()
        df_scaled_std = pd.DataFrame(scaler_std.fit_transform(df[numeric_columns]), columns = numeric_columns)
        df_scaled_normal = pd.DataFrame(scaler_normal.fit_transform(df[numeric_columns]), columns = numeric_columns)
```

```
In [6]: df_scaled_std
```

```
Out[6]:
```

	Sales	Cost	Profit
0	-1.058873	-1.058873	-1.058873
1	0.372036	0.372036	0.372036
2	1.802946	1.802946	1.802946
3	-0.486509	-0.486509	-0.486509
4	1.230582	1.230582	1.230582
5	-1.631237	-1.631237	-1.631237
6	-0.200327	-0.200327	-0.200327
7	0.658218	0.658218	0.658218
8	-0.772691	-0.772691	-0.772691
9	0.085855	0.085855	0.085855

```
In [7]: #Combine the scaled numeric features with the categorical values
        df_scaled = pd.concat([df_scaled_std, df.drop(numeric_columns, axis = 1)], axis = 1)
        #Display the dataset after feature scaling
        print("\nDataset after Feature Scaling:")
        print(df_scaled)
```

```
Dataset after Feature Scaling:
```

	Sales	Cost	Profit	Product	Category
0	-1.058873	-1.058873	-1.058873	Apple_Juice	Apple
1	0.372036	0.372036	0.372036	Banana_Smoothie	Banana
2	1.802946	1.802946	1.802946	Orange_Jam	Orange
3	-0.486509	-0.486509	-0.486509	Grape_Jelly	Grape
4	1.230582	1.230582	1.230582	Kiwi_Juice	Kiwi
5	-1.631237	-1.631237	-1.631237	Mango_Pickle	Mango
6	-0.200327	-0.200327	-0.200327	Pineapple_Sorbet	Pineapple
7	0.658218	0.658218	0.658218	Strawberry_Yoghurt	Strawberry
8	-0.772691	-0.772691	-0.772691	Blueberry_Pie	Blueberry
9	0.085855	0.085855	0.085855	Cherry_Salsa	Cherry

Feature Dummification(Convert Categorical Columns to numerical representation)

```

In [11]: #Identify categorical columns
categorical_columns = ['Product', 'Category']

#Create a column transformer for dummification
preprocessor = ColumnTransformer(
    transformers=[
        ('categorical', OneHotEncoder(), categorical_columns)
    ], remainder='passthrough')

In [12]: #Apply the column transformer to the dataset
df_dummified = pd.DataFrame(preprocessor.fit_transform(df))

In [13]: #Display the dataset after feature dummification
print("\nDataset after Feature Dummification:")
print(df_dummified)

```

Dataset after Feature Dummification:

```

0
0  (0, 0)\t1.0\n  (0, 10)\t1.0\n  (0, 20)\t1200...
1  (0, 1)\t1.0\n  (0, 11)\t1.0\n  (0, 20)\t1700...
2  (0, 7)\t1.0\n  (0, 17)\t1.0\n  (0, 20)\t2200...
3  (0, 4)\t1.0\n  (0, 14)\t1.0\n  (0, 20)\t1400...
4  (0, 5)\t1.0\n  (0, 15)\t1.0\n  (0, 20)\t2000...
5  (0, 6)\t1.0\n  (0, 16)\t1.0\n  (0, 20)\t1000...
6  (0, 8)\t1.0\n  (0, 18)\t1.0\n  (0, 20)\t1500...
7  (0, 9)\t1.0\n  (0, 19)\t1.0\n  (0, 20)\t1800...
8  (0, 2)\t1.0\n  (0, 12)\t1.0\n  (0, 20)\t1300...
9  (0, 3)\t1.0\n  (0, 13)\t1.0\n  (0, 20)\t1600...

```