Alladi Sai Revanth

BL.EN.U4CSE21011

```
In [5]:
                     import pandas as pd
                     import numpy as np
                     data=pd.read_excel(r"C:\Users\SAI REVANTH\Downloads\19CSE305_LabData_Set3.1.xlsx", sheet
                     data_types = data.dtypes
                     categorical_cols = data.select_dtypes(include=['object']).columns
                     nominal_cols = ['referral source'] + [col for col in data.columns if data[col].dtype ==
                     ordinal_cols = list(set(categorical_cols) - set(nominal_cols))
                     numeric_cols = data.select_dtypes(include=['number'])
                     data_range = numeric_cols.describe().loc[['min', 'max']]
                     missing_values = data.isin(['?']).sum()
                     outliers = {}
                     for col in numeric_cols.columns:
                               mean = numeric_cols[col].mean()
                                std = numeric_cols[col].std()
                                lower_bound = mean - 3 * std
                                upper_bound = mean + 3 * std
                                outliers[col] = len(numeric_cols[(numeric_cols[col] < lower_bound) | (numeric_cols[col] < lower_b
                     numeric_mean = numeric_cols.mean()
                     numeric_variance = numeric_cols.var()
                     print("Task 1: Data Types")
                     print(data_types)
                     print("\nTask 2: Encoding Schemes")
                     print("Nominal Columns:", nominal_cols)
                     print("Ordinal Columns:", ordinal_cols)
                     print("\nTask 3: Data Range")
                     print(data_range)
                     print("\nTask 4: Missing Values")
                     print(missing_values)
                     print("\nTask 5: Outliers")
                     print(outliers)
                     print("\nTask 6: Mean and Variance for Numeric Variables")
                     print("Mean:")
                     print(numeric_mean)
                     print("\nVariance:")
                     print(numeric_variance)
```

```
Task 1: Data Types
                                 int64
Record ID
                                 int64
age
sex
                                object
                                object
on thyroxine
query on thyroxine
                                object
on antithyroid medication
                                object
sick
                                object
pregnant
                                object
thyroid surgery
                                object
I131 treatment
                                object
query hypothyroid
                                object
query hyperthyroid
                                object
lithium
                                object
goitre
                                object
tumor
                                object
hypopituitary
                                object
                                object
psych
TSH measured
                                object
TSH
                                object
T3 measured
                                object
Т3
                                object
TT4 measured
                                object
                                object
T4U measured
                                object
T4U
                                object
FTI measured
                                object
FTI
                                object
TBG measured
                                object
TBG
                                object
referral source
                                object
                                object
Condition
dtype: object
Task 2: Encoding Schemes
Nominal Columns: ['referral source', 'sex', 'TSH', 'T3', 'TT4', 'T4U', 'FTI', 'TBG']
Ordinal Columns: ['on antithyroid medication', 'I131 treatment', 'T3 measured', 'TT4 mea
sured', 'hypopituitary', 'sick', 'on thyroxine', 'Condition', 'psych', 'query hyperthyro
id', 'query hypothyroid', 'query on thyroxine', 'lithium', 'tumor', 'TSH measured', 'pre
gnant', 'FTI measured', 'TBG measured', 'goitre', 'thyroid surgery', 'T4U measured']
Task 3: Data Range
        Record ID
                     age
min 840801013.0
                     1.0
max 841031032.0 97.0
Task 4: Missing Values
Record ID
                                  0
                                  0
age
                                 24
sex
on thyroxine
                                  0
query on thyroxine
                                  0
on antithyroid medication
                                  0
                                  0
sick
                                  0
pregnant
thyroid surgery
                                  0
                                  0
I131 treatment
query hypothyroid
                                  0
query hyperthyroid
                                  0
lithium
                                  0
goitre
                                  0
```

0

0

0

tumor

nsvch

hypopituitary

```
TSH measured
                                          0
        TSH
                                        115
        T3 measured
                                          0
        Т3
                                        202
        TT4 measured
                                          0
        TT4
                                         40
        T4U measured
                                          0
        T4U
                                         98
        FTI measured
                                          0
        FTI
                                         98
        TBG measured
                                          0
        TBG
                                        964
        referral source
                                          0
        Condition
                                          0
        dtype: int64
        Task 5: Outliers
        {'Record ID': 0, 'age': 0}
        Task 6: Mean and Variance for Numeric Variables
        Mean:
        Record ID
                      8.409340e+08
        age
                      5.150900e+01
        dtype: float64
        Variance:
        Record ID
                      5.819759e+09
        age
                      3.525585e+02
        dtype: float64
In [6]: import pandas as pd
         import numpy as np
         data = pd.read_excel(r"C:\Users\SAI_REVANTH\Downloads\19CSE305_LabData_Set3.1.xlsx", shee
         for col in data.columns:
             if data[col].dtype == 'float64' or data[col].dtype == 'int64':
   if col in ['TSH', 'T3', 'TT4', 'T4U', 'FTI']:
                      data[col].fillna(data[col].median(), inplace=True)
                 else:
                      data[col].fillna(data[col].mean(), inplace=True)
             elif data[col].dtype == 'object':
                 data[col].fillna(data[col].mode()[0], inplace=True)
         missing_values_after_imputation = data.isnull().sum()
         print("Missing Values After Imputation:")
         print(missing_values_after_imputation)
```

```
Missing Values After Imputation:
        Record ID
        age
                                      0
                                      0
        sex
        on thyroxine
                                      0
                                      0
        query on thyroxine
        on antithyroid medication
                                      0
        sick
                                      0
        pregnant
                                      0
        thyroid surgery
                                      0
        I131 treatment
                                      0
                                      0
        query hypothyroid
        query hyperthyroid
                                      0
        lithium
                                      0
                                      0
        goitre
        tumor
                                      0
                                      0
        hypopituitary
        psych
                                      0
        TSH measured
                                      0
        TSH
                                      0
        T3 measured
                                      0
                                      0
        Т3
        TT4 measured
                                      0
        TT4
                                      0
        T4U measured
                                      0
        T4U
                                      0
        FTI measured
                                      0
                                      0
        FTI
        TBG measured
                                      0
        TBG
                                      0
        referral source
                                      0
        Condition
                                      0
        dtype: int64
In [8]:
        import pandas as pd
        from sklearn.preprocessing import MinMaxScaler
        data = pd.read_excel(r"C:\Users\SAI REVANTH\Downloads\19CSE305_LabData_Set3.1.xlsx", she
        data.replace('?', np.nan, inplace=True)
        numeric_attributes = ['age', 'TSH', 'T3', 'TT4', 'T4U', 'FTI']
        data[numeric_attributes] = data[numeric_attributes].apply(pd.to_numeric, errors='coerce'
        minmax_scaler = MinMaxScaler()
        data[numeric_attributes] = minmax_scaler.fit_transform(data[numeric_attributes])
```

print("Normalized Data:")

print(data.head())

```
Record ID
                                                              age sex on thyroxine query on thyroxine
                        0 840801013 0.291667
                                                                                                     f
                                                                                                     f
                                                                                                                                            f
                        1 840801014 0.291667
                                                                                                                                           f
                        2 840801042 0.416667
                                                                           F
                                                                                                     f
                                                                                                     f
                                                                                                                                           f
                        3 840803046 0.364583
                                                                                                     f
                                                                                                                                            f
                        4 840803047 0.322917
                                                                           F
                            on antithyroid medication sick pregnant thyroid surgery I131 treatment
                        0
                                                                            f
                                                                                       f
                                                                                                                                         f
                                                                                                         f
                        1
                                                                             f
                                                                                       f
                                                                                                         f
                                                                                                                                         f
                                                                                                                                                                        f
                        2
                                                                             f
                                                                                       f
                                                                                                         f
                                                                                                                                         f
                                                                                                                                                                        f
                                                                                                                                                                              . . .
                        3
                                                                             f
                                                                                       f
                                                                                                         f
                                                                                                                                         f
                                                                                                                                                                        f
                                                                                       f
                        4
                                                                                                         f
                                                                                                                                         f
                                                                                                                                                                              . . .
                            TT4 measured
                                                                  TT4 T4U measured T4U FTI measured FTI TBG measured
                                                                                                                                                                              TBG
                                                                                                f NaN
                                                                                                                                  f NaN
                        0
                                                  f
                                                                  NaN
                                                                                                                                                                              NaN
                                                       0.351124
                                                                                                 f NaN
                                                                                                                                   f NaN
                                                                                                                                                                      f
                                                                                                                                                                              NaN
                        1
                                                  t
                        2
                                                  f
                                                                  NaN
                                                                                                 f NaN
                                                                                                                                   f NaN
                                                                                                                                                                      t
                                                                                                                                                                            11.0
                                                  f
                                                                                                                                   f NaN
                        3
                                                                                                 f NaN
                                                                                                                                                                      t
                                                                                                                                                                            26.0
                                                                  NaN
                                                  f
                        4
                                                                  NaN
                                                                                                 f NaN
                                                                                                                                   f NaN
                                                                                                                                                                      t
                                                                                                                                                                            36.0
                              referral source
                                                                      Condition
                        0
                                                  other
                                                                NO CONDITION
                        1
                                                                NO CONDITION
                                                  other
                        2
                                                                NO CONDITION
                                                  other
                        3
                                                                NO CONDITION
                                                  other
                                                  other
                                                                                       S
                        [5 rows x 31 columns]
       In [9]: import pandas as pd
                        worksheet_name = ''
                        data = pd.read_excel(r"C:\Users\SAI REVANTH\Downloads\19CSE305_LabData_Set3.1.xlsx", she
                        'query hyperthyroid', 'lithium', 'goitre', 'tumor', 'hypopituitary'
                        vector1 = data.loc[0, binary_attributes].astype(str)
                        vector2 = data.loc[1, binary_attributes].astype(str)
                        f11 = sum((vector1 == '1') & (vector2 == '1'))
                        f01 = sum((vector1 == '0') & (vector2 == '1'))
                        f10 = sum((vector1 == '1') & (vector2 == '0'))
                        f00 = sum((vector1 == '0') & (vector2 == '0'))
                        if f01 + f10 + f11 != 0:
                                jc = f11 / (f01 + f10 + f11)
                        else:
                                jc = 0.0
                        if f00 + f01 + f10 + f11 != 0:
                                smc = (f11 + f00) / (f00 + f01 + f10 + f11)
                        else:
                                smc = 0.0
                        print("Jaccard Coefficient (JC):", jc)
                        print("Simple Matching Coefficient (SMC):", smc)
                        Jaccard Coefficient (JC): 0.0
                        Simple Matching Coefficient (SMC): 0.0
     In [11]:
                        import pandas as pd
                        import numpy as np
                        from sklearn.metrics.pairwise import cosine_similarity
                        data = pd.read_excel(r"C:\Users\SAI REVANTH\Downloads\19CSE305_LabData_Set3.1.xlsx", she
                        vector1 = data.iloc[0, 1:].apply(lambda x: float(x) if str(x).replace('.', '', 1).isdigi
Loading [MathJax]/extensions/Safe.js a.iloc[1, 1:].apply(lambda x: float(x) if str(x).replace('.', '') if str(x).replace('.', '
```

Normalized Data:

```
dot_product = np.dot(vector1, vector2)
magnitude_vector1 = np.linalg.norm(vector1)
magnitude_vector2 = np.linalg.norm(vector2)
cosine_similarity = dot_product / (magnitude_vector1 * magnitude_vector2)
print("Cosine Similarity:", cosine_similarity)
```

Cosine Similarity: nan

```
import pandas as pd
In [13]:
         import numpy as np
         import seaborn as sns
         import matplotlib.pyplot as plt
         from sklearn.metrics import jaccard_score
         from sklearn.metrics.pairwise import cosine_similarity
         data = pd.read_excel(r"C:\Users\SAI REVANTH\Downloads\19CSE305_LabData_Set3.1.xlsx", she
         vectors = data.iloc[:20, 1:-1]
         def jaccard_coefficient(vector1, vector2):
             intersection = np.logical_and(vector1, vector2)
             union = np.logical_or(vector1, vector2)
             return np.sum(intersection) / np.sum(union)
         jc_{matrix} = np.zeros((20, 20))
         smc_matrix = np.zeros((20, 20))
         cosine_matrix = np.zeros((20, 20))
         for i in range(20):
             for j in range(20):
                 vector1 = vectors.iloc[i].astype(bool)
                 vector2 = vectors.iloc[j].astype(bool)
                 jc_matrix[i, j] = jaccard_coefficient(vector1, vector2)
                 smc_matrix[i, j] = jaccard_score(vector1, vector2, average='binary')
                 cosine_matrix[i, j] = cosine_similarity([vector1], [vector2])[0, 0]
         plt.figure(figsize=(10, 8))
         sns.heatmap(jc_matrix, annot=True, fmt=".2f", cmap="YlGnBu", xticklabels=False, yticklab
         plt.title("Jaccard Coefficient Heatmap")
         plt.show()
         plt.figure(figsize=(10, 8))
         sns.heatmap(smc_matrix, annot=True, fmt=".2f", cmap="YlGnBu", xticklabels=False, ytickla
         plt.title("Simple Matching Coefficient Heatmap")
         plt.show()
         plt.figure(figsize=(10, 8))
         sns.heatmap(cosine_matrix, annot=True, fmt=".2f", cmap="Y1GnBu", xticklabels=False, ytic
         plt.title("Cosine Similarity Heatmap")
         plt.show()
```

Jaccard Coefficient Heatmap

Jaccard Coefficient Heatmap	- 1.100
001.001.001.001.001.001.001.001.001.001	- 1.100
001.001.001.001.001.001.001.001.001.001	
001.001.001.001.001.001.001.001.001.001	- 1.075
001.001.001.001.001.001.001.001.001.001	
001.001.001.001.001.001.001.001.001.001	1.050
001.001.001.001.001.001.001.001.001.001	- 1.050
001.001.001.001.001.001.001.001.001.001	
001.001.001.001.001.001.001.001.001.001	- 1.025
001.001.001.001.001.001.001.001.001.001	
001.001.001.001.001.001.001.001.001.001	1.000
001.001.001.001.001.001.001.001.001.001	- 1.000
001.001.001.001.001.001.001.001.001.001	
001.001.001.001.001.001.001.001.001.001	- 0.975
001.001.001.001.001.001.001.001.001.001	
001.001.001.001.001.001.001.001.001.001	0.050
001.001.001.001.001.001.001.001.001.001	- 0.950
001.001.001.001.001.001.001.001.001.001	
001.001.001.001.001.001.001.001.001.001	- 0.925
001.001.001.001.001.001.001.001.001.001	
001.001.001.001.001.001.001.001.001.001	0.000
	- 0.900

imnle	Matching	Coefficient	Heatman
illipie	Matchillig	Coemcient	пеаннар

Simple Matching Coefficient Heatmap	1.100	
1.001.001.001.001.001.001.001.001.001.0	- 1.100	
1.00 1.00 1.00 1.00 1.00 1.00 1.00 1.00		
1.001.001.001.001.001.001.001.001.001.0	- 1.075	
1.001.001.001.001.001.001.001.001.001.0		
1.001.001.001.001.001.001.001.001.001.0		
1.001.001.001.001.001.001.001.001.001.0	- 1.050	
1.001.001.001.001.001.001.001.001.001.0		
1.001.001.001.001.001.001.001.001.001.0	- 1.025	
1.001.001.001.001.001.001.001.001.001.0		
1.001.001.001.001.001.001.001.001.001.0		
1.001.001.001.001.001.001.001.001.001.0	- 1.000	
1.001.001.001.001.001.001.001.001.001.0		
1.001.001.001.001.001.001.001.001.001.0	- 0.975	
1.001.001.001.001.001.001.001.001.001.0		
1.001.001.001.001.001.001.001.001.001.0	- 0.950	
1.001.001.001.001.001.001.001.001.001.0	- 0.950	
1.001.001.001.001.001.001.001.001.001.0		
1.001.001.001.001.001.001.001.001.001.0	- 0.925	
1.001.001.001.001.001.001.001.001.001.0		
1.001.001.001.001.001.001.001.001.001.0	- 0.900	
	- 0.900	

Cosine	Similarity	Heatman
Cosine	Sillillality	пеаннар

- 1.100	Cosine Similarity Heatmap
01.00	1.001.001.001.001.001.001.001.001.001.0
01.00	1.001.001.001.001.001.001.001.001.001.0
- 1.075	1.001.001.001.001.001.001.001.001.001.0
01.00	1.001.001.001.001.001.001.001.001.001.0
	1.001.001.001.001.001.001.001.001.001.0
- 1.050	1.001.001.001.001.001.001.001.001.001.0
01.00	1.001.001.001.001.001.001.001.001.001.0
- 1.025	1.001.001.001.001.001.001.001.001.001.0
01.00	1.001.001.001.001.001.001.001.001.001.0
	1.001.001.001.001.001.001.001.001.001.0
- 1.000	1.001.001.001.001.001.001.001.001.001.0
01.00	1.001.001.001.001.001.001.001.001.001.0
- 0.975	1.001.001.001.001.001.001.001.001.001.0
01.00	1.001.001.001.001.001.001.001.001.001.0
	1.001.001.001.001.001.001.001.001.001.0
- 0.950	1.001.001.001.001.001.001.001.001.001.0
01.00	1.001.001.001.001.001.001.001.001.001.0
- 0.925	1.001.001.001.001.001.001.001.001.001.0
01.00	1.001.001.001.001.001.001.001.001.001.0
- 0.900	1.001.001.001.001.001.001.001.001.001.0
0.900	