Lab Exercise – 7: Data Similarity - Dissimilarity & Handling Missing Values

Note:

- * Prepare a PDF document and name the file as "Lab7 RegisterNo.pdf".
- * PDF file should consist Question No, Code, and Result for each Question.
- * File Should be headed with your Register number, Slot number, Lab Exercise number.

1. Develop User Defined functions to calculate dissimilarity matrices for Nominal attributes, Binary attributes, Numeric attributes, Ordinal attributes, Mixed attributes. Apply these functions on the following data:

Obj Id	A1	A2	A3	A4	A5	A6	A7	A8	A9	A10
1	4	D3	0	24	T	Y	P	poor	A	14
2	5	D1	1	36	F	Y	N	average	В	16
3	6	D4	1	43	T	N	P	good	С	18
4	7	D6	0	13	T	N	N	average	Е	12
5	8	D1	1	22	F	N	P	poor	В	21

Hint:

- All Binary attributes are symmetric.
- All Numeric attributes are in Euclidean space
- Use the following ordinal relationships in the following attributes:

A9:
$$A < B < C < D < E$$

A9: A < B < C < D < E **A2:** D6 < D5 < D4 < D3 <

2. Calculate Term Frequency Vector (Document Vectors) for the following documents and calculate Cosine Similarity between every pair of documents to identify which documents are more similar.

Data 1:

doc1 = "I want to start learning to charge something in life"

doc2 = "reading something about life no one else knows"

doc3 = "Never stop learning"

doc4 = "life learning"

Data 2:

doc1 = "Mr. Imran Khan win the president seat after winning the National election 2020-2021. Though he lost the support of some republican friends, Imran Khan is friends with President Nawaz Sharif"

doc2 = "President Imran Khan says Nawaz Sharif had no political interference is the election outcome. He claimed President Nawaz Sharif is a friend who had nothing to do with the election"

doc3 = "Post elections, Vladimir Nawaz Sharif win the president seat of Russia. President Nawaz Sharif had served as the Prime Minister earlier in his political career"

3.

- a. Load *crx.data* into a data frame and do the following operations: (The data has no headers)
 - ➤ Change the column names to A1 to A16
 - > Replace all '?' marks with np.nan
 - Convert A2 and A14 attributes to float data type
 - Convert '+' to 1 and "-" to 0 of A16 attribute
 - Replace values of "A3, A8, A9, A10" attributes to np.nan in 50 random objects
 - > Save the file as Transformed crx.csv

b. Ignoring missing values:

- ➤ Load the Credit Approval Data Set Transformed crx.csv
- Calculate the percentage of missing values for each variable and sort them in ascending order
- Remove the observations with missing data in any of the variables
- ➤ Print and compare the size of the original and complete case datasets
- c. Performing mean and median imputation:
 - ➤ Load the Credit Approval Data Set Transformed crx.csv
 - ➤ Replace the missing values with the median in five numerical variables 'A2', 'A3', 'A8', 'A11', 'A15' using pandas
 - ➤ Replace the missing values with the mean in five numerical variables 'A2', 'A3', 'A8', 'A11', 'A15' using pandas
 - > Use SimpleImputer() of scikit-learn to fill the missing values with median and mean, separately.
- d. Performing mode or frequent category imputation:
 - Load the Credit Approval Data Set Transformed crx.csv
 - Replace the missing values with the mode in the attributes 'A4', 'A5', 'A6', 'A7' using pandas
 - ➤ Use SimpleImputer() of scikit-learn to fill the missing values with mode.
- e. Performing most probable value imputation: [Optional]
 - ➤ Load the Credit Approval Data Set Transformed_crx.csv
 - Replace the missing values with probable values using linear regression in the attributes 'A2', 'A3', 'A8', 'A11', 'A15' using pandas