Decision Trees and K-nn

Problem 1 [Decision Trees using Scikit-learn]:

The objective is to build a model that can predict the appropriate medication/Drug type (target variable) for patients suffering from a specific illness based on their Age, Sex, Blood Pressure, and Cholesterol levels. [Drag.csv]

1. Data Preprocessing: Perform data preprocessing steps by handling missing data [count how many missing values occurs + handle empty cell by your own way] and encoding categorical variables.

2. First experiment: Training and Testing with Fixed Train-Test Split Ratio:
    - Divide the dataset into a training set and a testing set (30% of the samples).
    - Repeat this experiment five times with different random splits of the data into training and test sets.
    - Report the sizes and accuracies of the decision trees in each experiment, Compare the results of different models and select the one that achieves the highest overall performance.

3. Second experiment: Training and Testing with a Range of Train-Test Split Ratios:
   Consider training set sizes in the range of 30% to 70% (increments of 10%). Start with a training set size of 30% and increase it by 10% until you reach 70%.
   For each training set size:
    - Run the experiment with five different random seeds.
    - Calculate the mean, maximum, and minimum accuracy at each training set size.
    - Measure the mean, maximum, and minimum tree size.
    - Store the statistics in a report.
    - Create two plots: one showing accuracy against training set size and another showing the number of nodes in the final tree against training set size.

Problem 2 [KNN]:

Use the diabetes.csv data to implement your own simple KNN classifier using python, (Don't use any built-in functions), divide your data into 70% for training and 30% for testing. [diabetes.csv]

1. Objective:
    - Perform multiple iterations of k (e.g., 5 iterations each different k value ex. K=2,3,4…) on the dataset.
    - Use Euclidean distance for computing distances between instances.
2. Data preprocessing:
    - Normalize each feature column separately for training and test objects using Log Transformation or Min-Max Scaling.
3. Break ties using Distance-Weighted Voting:
    - When there is a tie, consider the distances between the test instance and the tied classes' neighbors.
    - Assign higher weights to closer neighbors and use these weights to break the tie, reflecting the idea that closer neighbors might have a stronger influence on the classification decision.
4. Output:
    - **For each iteration**, output the value of k and the following summary information:
        - Number of correctly classified instances. ○ Total number of instances in the test set.
        - Accuracy.
    - At the end of all iterations, output the average accuracy across all iterations.

Output Example:

k value: 3
Number of correctly classified instances: 418
Total number of instances: 658
Accuracy: 63%

Min/Max Scaling:

Formula: $x' = \frac{x - \min(x)}{\max(x) - \min(x)}$