solution

November 5, 2024

1 Decision Tree Implelentation

1.1 Step-1: Importing Dataset

1.2 Step-2: Parsing the Dataset

```
[26]: from ucimlrepo import fetch_ucirepo

# fetch dataset
lenses = fetch_ucirepo(id=58)

# data (as pandas dataframes)
X = lenses.data.features
y = lenses.data.targets

# variable information
# print(lenses.variables)

dataframe = lenses.data['original']
print(lenses.data['original'])
```

	id	age	spectacle_prescription	astigmatic	class
1	1	1	1	1	3
2	1	1	1	2	2
3	1	1	2	1	3
4	1	1	2	2	1
5	1	2	1	1	3
6	1	2	1	2	2
7	1	2	2	1	3
8	1	2	2	2	1
9	2	1	1	1	3
10	2	1	1	2	2
11	2	1	2	1	3
12	2	1	2	2	1
13	2	2	1	1	3
14	2	2	1	2	2
15	2	2	2	1	3
16	2	2	2	2	3
17	3	1	1	1	3

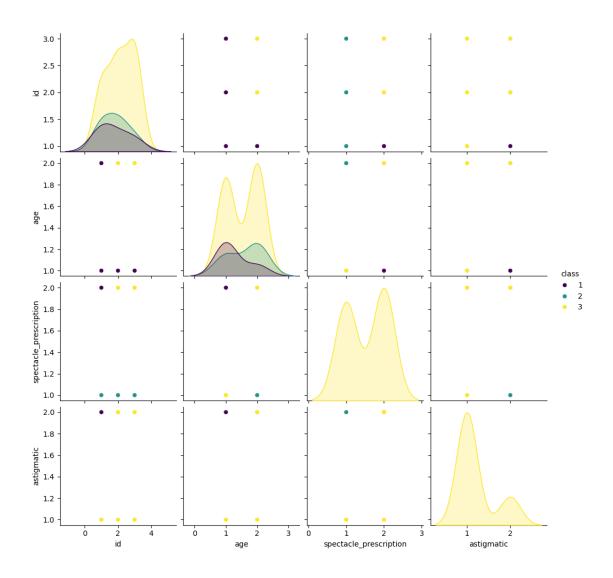
```
18
    3
                               1
                                          2
                                                3
         1
19
   3
         1
                               2
                                          1
                                                3
                               2
20
   3
         1
                                          2
                                                1
21
    3
         2
                               1
                                          1
                                                3
         2
                                          2
                                                2
22
   3
                               1
23
    3
         2
                               2
                                          1
                                                3
    3
         2
                               2
                                          2
                                                 3
24
```

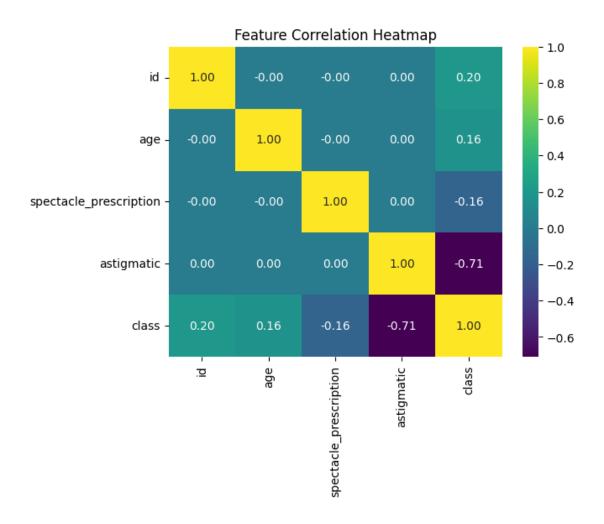
1.3 Step-3: Visual Representations

```
[27]: import pandas as pd
  import seaborn as sns
  import matplotlib.pyplot as plt

sns.pairplot(dataframe, hue='class', palette='viridis')
  plt.show()

corr = dataframe.corr()
  sns.heatmap(corr, annot=True, cmap='viridis', fmt='.2f')
  plt.title('Feature Correlation Heatmap')
  plt.show()
```





1.4 Step-4: Train the Classifier

```
clf = clf.fit(X_train, y_train)

# Predict the response
y_pred = clf.predict(X_test)
```

1.5 Step-5: Write a function to descend the tree for a given instance

```
[29]: class TreeNode:
          def __init__(self, feature_index=None, threshold=None, left=None, u
       ⇒right=None, value=None):
              self.feature_index = feature_index # Index of the feature to check (e.
       \rightarrow q., 0 for 'age')
              self.threshold = threshold
                                                   # Threshold value for the feature
              self.left = left
                                                  # Left child node
              self.right = right
                                                  # Right child node
              self.value = value
                                                   # Predicted class (for leaf nodes
       \rightarrow only)
      def descend tree(node, instance):
          Descends the decision tree for a given instance and returns the prediction.
          Parameters:
          - node (TreeNode): The root node of the decision tree.
          - instance (list): A list of feature values for a single instance (e.g., \Box
       \hookrightarrow [1, 2, 1]).
          Returns:
          - The predicted class at the leaf node.
          while node.left is not None and node.right is not None: # Traverse until au
              feature_value = instance[node.feature_index] # Get feature value from
       instance
              # Check the threshold to determine the next branch
              if feature_value <= node.threshold:</pre>
                  node = node.left
              else:
                  node = node.right
          return node.value # Return prediction at the leaf node
      # Example decision tree setup (this would normally be constructed by a training_
       →algorithm)
```

Predicted class: 2

1.6 Step-6: Persist the tree data structure so it can be recalled without building the tree; then use it in any application.

```
[30]: import pickle
      def save_tree(tree, filename):
          Saves the decision tree to a file.
          Parameters:
          - tree (TreeNode): The root node of the decision tree.
          - filename (str): The name of the file to save the tree.
          HHHH
          with open(filename, 'wb') as f:
              pickle.dump(tree, f)
          print(f"Tree saved to {filename}")
      def load_tree(filename):
          Loads the decision tree from a file.
          Parameters:
          - filename (str): The name of the file to load the tree from.
          Returns:
          - TreeNode: The root node of the loaded decision tree.
          with open(filename, 'rb') as f:
              tree = pickle.load(f)
          print(f"Tree loaded from {filename}")
          return tree
```

```
# Save the tree to a file
save_tree(root, 'decision_tree.pkl')

# Later in your application, load the tree
loaded_tree = load_tree('decision_tree.pkl')
```

Tree saved to decision_tree.pkl
Tree loaded from decision_tree.pkl