DTCD2

September 12, 2024

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[10]: import pandas as pd
     from sklearn.tree import DecisionTreeClassifier, plot_tree
     from sklearn.model_selection import train_test_split
     from sklearn.metrics import accuracy score
     from itertools import combinations
     import matplotlib.pyplot as plt
[11]: data = {
         'Color': ['Red', 'Blue', 'Green', 'Black', 'Blue', 'Red', 'Green', 'Black',
      'Size': ['Full', 'Medium', 'Standard', 'Full', 'Standard', 'Medium',
      'Model': ['G15 5510', 'G15 5515', 'G15 5520', 'G15 5510', 'G15 5515', 'G15_{\sqcup}
      ⇔5510', 'G15 5515', 'G15 5520', 'G15 5510', 'G15 5520'],
         'Material': ['Plastic', 'Metal', 'Plastic', 'Plastic', 'Metal', 'Plastic', '
      'Target': [1, 1, 1, 0, 1, 0, 0, 1, 0, 1]
     }
[12]: df = pd.DataFrame(data)
     df
[12]:
        Color
                  Size
                          Model Material Target
         Red
                  Full G15 5510 Plastic
                                             1
                Medium G15 5515
     1
        Blue
                                  Metal
                                             1
     2 Green Standard G15 5520 Plastic
                                             1
     3 Black
                  Full G15 5510 Plastic
                                             0
         Blue Standard G15 5515
                                  Metal
                                             1
         Red
              Medium G15 5510 Plastic
     6 Green
                  Full G15 5515
                                  Metal
                                             0
     7 Black
                Medium G15 5520 Plastic
                                             1
     8
         Red Standard G15 5510 Plastic
                                             0
                  Full G15 5520 Plastic
     9 Green
                                             1
[13]: def evaluate_feature_combination(features):
         X = pd.get dummies(df[features])
```

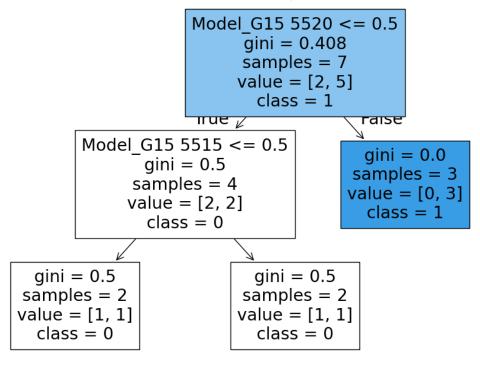
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→test_size=0.3, random_state=42)
          clf = DecisionTreeClassifier(random_state=42)
          clf.fit(X_train, y_train)
          y_pred = clf.predict(X_test)
          return accuracy score(y test, y pred), clf
[14]:
      columns = df.drop(columns=['Target']).columns
[15]: results = []
      for r in range(1, len(columns) + 1):
          for combo in combinations(columns, r):
              accuracy, clf = evaluate_feature_combination(list(combo))
              results.append({'Features': combo, 'Accuracy': accuracy, 'Model': clf})
[16]: results_df = pd.DataFrame(results)
      results_df
[16]:
                                Features Accuracy \
      0
                                (Color,) 0.333333
      1
                                 (Size,)
                                         0.333333
      2
                                (Model,) 0.666667
      3
                             (Material,) 0.000000
      4
                           (Color, Size) 0.333333
      5
                          (Color, Model) 0.333333
      6
                       (Color, Material) 0.333333
      7
                           (Size, Model) 0.333333
      8
                        (Size, Material) 0.333333
      9
                       (Model, Material) 0.666667
                    (Color, Size, Model) 0.333333
      10
                 (Color, Size, Material) 0.333333
      11
      12
                (Color, Model, Material)
                                          0.333333
                 (Size, Model, Material)
      13
                                          0.333333
          (Color, Size, Model, Material)
                                          0.333333
                                            Model
      0
          DecisionTreeClassifier(random_state=42)
          DecisionTreeClassifier(random_state=42)
      1
          DecisionTreeClassifier(random state=42)
      2
      3
          DecisionTreeClassifier(random state=42)
      4
          DecisionTreeClassifier(random_state=42)
          DecisionTreeClassifier(random state=42)
      5
      6
          DecisionTreeClassifier(random_state=42)
      7
         DecisionTreeClassifier(random state=42)
      8
          DecisionTreeClassifier(random_state=42)
          DecisionTreeClassifier(random state=42)
      9
      10 DecisionTreeClassifier(random_state=42)
```

X_train, X_test, y_train, y_test = train_test_split(X, df['Target'],_

```
11 DecisionTreeClassifier(random_state=42)
     12 DecisionTreeClassifier(random_state=42)
     13 DecisionTreeClassifier(random_state=42)
     14 DecisionTreeClassifier(random_state=42)
[17]: best result = results df.loc[results df['Accuracy'].idxmax()]
     worst result = results df.loc[results df['Accuracy'].idxmin()]
     average_accuracy = results_df['Accuracy'].mean()
[18]: print(f"Best Combination: {best_result['Features']}, Accuracy:
       ⇔{best_result['Accuracy']:.2f}")
     print(f"Worst Combination: {worst_result['Features']}, Accuracy:__
       ⇔{worst_result['Accuracy']:.2f}")
     print(f"Average Accuracy: {average_accuracy:.2f}")
     Best Combination: ('Model',), Accuracy: 0.67
     Worst Combination: ('Material',), Accuracy: 0.00
     Average Accuracy: 0.36
[19]: X_best = pd.get_dummies(df[list(best_result['Features'])])
     plt.figure(figsize=(12, 8))
     plot_tree(best_result['Model'], feature_names=X_best.columns, class_names=['0',_
       plt.title(f'Best Decision Tree (Accuracy: {best_result["Accuracy"]:.2f})')
```

Best Decision Tree (Accuracy: 0.67)

plt.show()

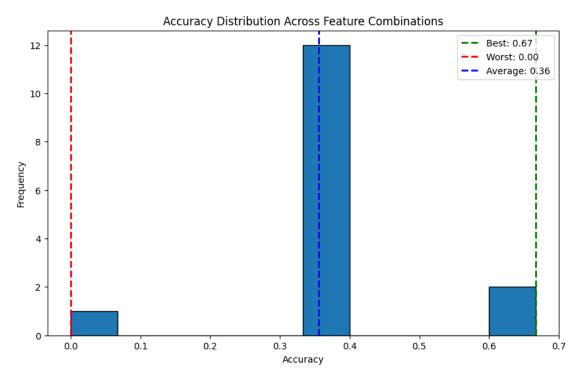


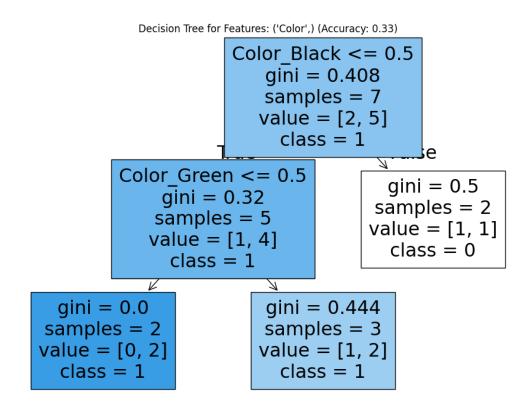
```
[20]: plt.figure(figsize=(10, 6))
      plt.hist(results_df['Accuracy'], bins=10, edgecolor='black')
      plt.axvline(best_result['Accuracy'], color='g', linestyle='dashed',_

¬linewidth=2, label=f"Best: {best_result['Accuracy']:.2f}")

      plt.axvline(worst_result['Accuracy'], color='r', linestyle='dashed',__
       →linewidth=2, label=f"Worst: {worst_result['Accuracy']:.2f}")
      plt.axvline(average_accuracy, color='b', linestyle='dashed', linewidth=2,__
       →label=f"Average: {average_accuracy:.2f}")
      plt.title('Accuracy Distribution Across Feature Combinations')
      plt.xlabel('Accuracy')
      plt.ylabel('Frequency')
      plt.legend()
      plt.show()
      for index, row in results_df.iterrows():
         plt.figure(figsize=(12, 8))
         X_combo = pd.get_dummies(df[list(row['Features'])])
         plot_tree(row['Model'], feature_names=X_combo.columns, class_names=['0',_
       plt.title(f"Decision Tree for Features: {row['Features']} (Accuracy:

¬{row['Accuracy']:.2f})")
         plt.show()
```

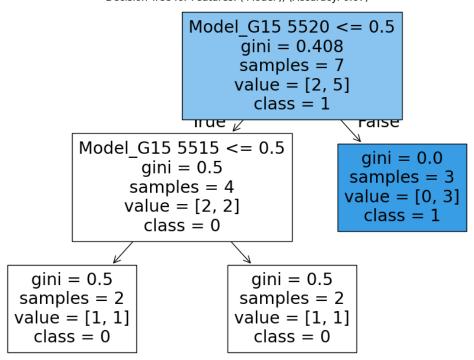




Size_Full <= 0.5
gini = 0.408
samples = 7
value = [2, 5]
class = 1

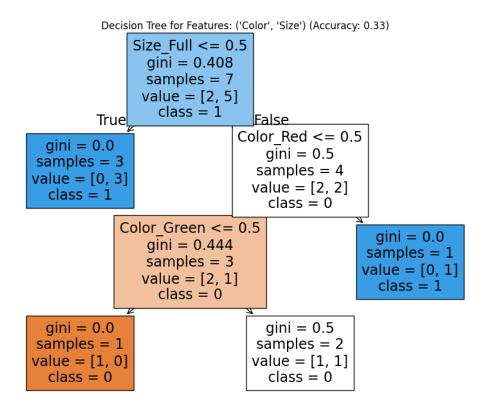
gini = 0.0 samples = 3 value = [0, 3] class = 1 gini = 0.5 samples = 4 value = [2, 2] class = 0

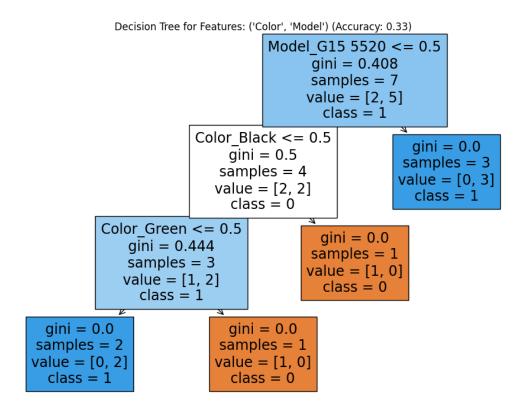
Decision Tree for Features: ('Model',) (Accuracy: 0.67)

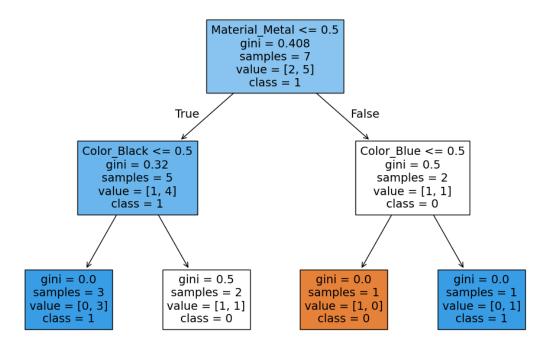


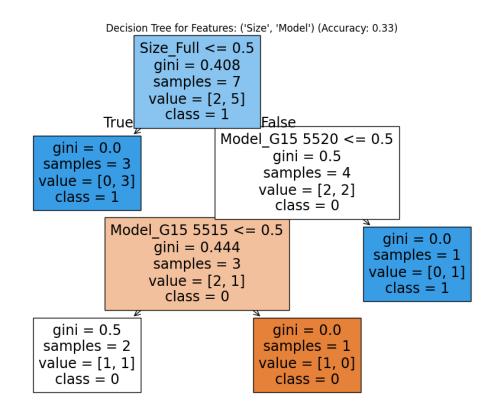
Decision Tree for Features: ('Material',) (Accuracy: 0.00)

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\label{eq:material_Metal} \begin{array}{l} \text{Material\_Metal} <= 0.5\\ \text{gini} = 0.408\\ \text{samples} = 7\\ \text{value} = [2, 5]\\ \text{class} = 1 \end{array} \begin{array}{l} \text{True} & \text{False} \\ \\ \text{gini} = 0.32\\ \text{samples} = 5\\ \text{value} = [1, 4]\\ \text{class} = 1 & \text{value} = [1, 1]\\ \text{class} = 0 \\ \end{array}
```

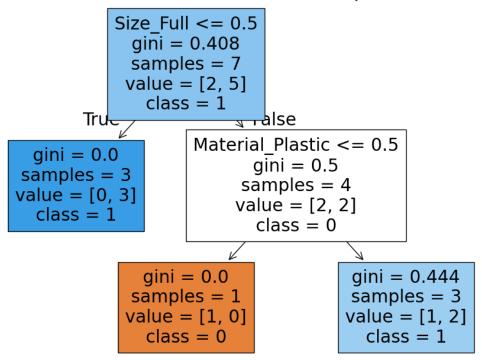




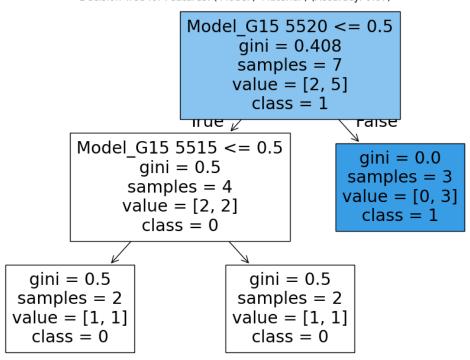


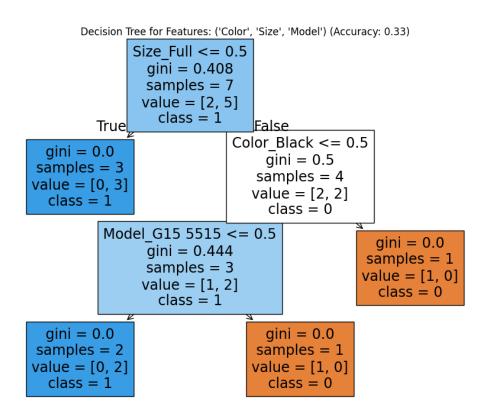


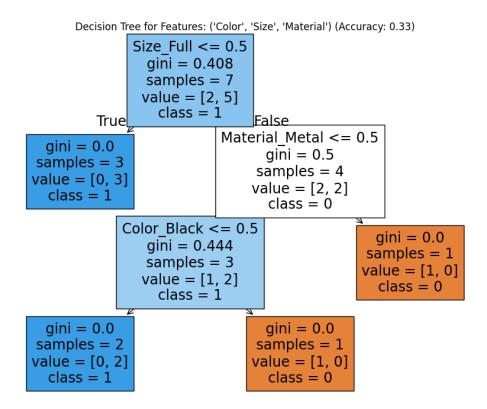
Decision Tree for Features: ('Size', 'Material') (Accuracy: 0.33)

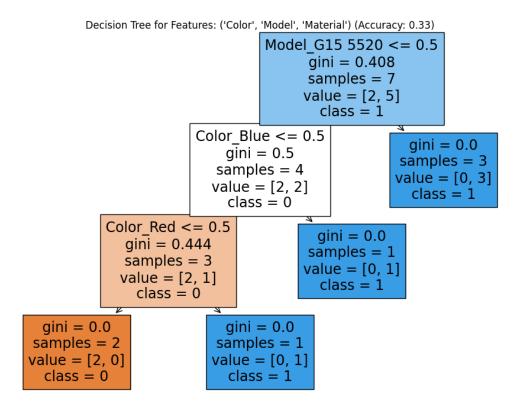


Decision Tree for Features: ('Model', 'Material') (Accuracy: 0.67)









```
Decision Tree for Features: ('Size', 'Model', 'Material') (Accuracy: 0.33)
               Size Full \leq 0.5
                 gini = 0.408
                 samples = 7
                value = [2, 5]
                   class = 1
         True
                                 False_
                           Model G15 5515 <= 0.5
  gini = 0.0
                                   gini = 0.5
samples = 3
                                 samples = 4
value = [0, 3]
                                 value = [2, 2]
  class = 1
                                   class = 0
          Model_G15 5510 <= 0.5
                                                   gini = 0.0
                 gini = 0.444
                                                  samples = 1
                 samples = 3
                                                 value = [1, 0]
                value = [1, 2]
                                                    class = 0
                   class = 1
  gini = 0.0
                                   gini = 0.5
                                 samples = 2
samples = 1
value = [0, 1]
                                 value = [1, 1]
  class = 1
                                   class = 0
```

