Danny_Xia_HW5

February 9, 2025

```
[69]: import numpy as np
      import pandas as pd
      from sklearn.model_selection import train_test_split
      from sklearn.linear_model import LogisticRegression
      from sklearn.model_selection import KFold
      from sklearn.model selection import cross val score
      from sklearn.feature_selection import SequentialFeatureSelector
      import matplotlib.pyplot as plt
[70]: df = pd.read_csv("heart.csv")
      display(df)
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[303 rows x 14 columns]
[71]: X=df.drop('target', axis=1)
      y=df['target']
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=103/303,__
       →random_state=42)
[72]: def ErrorEstimate(X_train_selected, X_test_selected, y_train, y_test):
          model = LogisticRegression(max_iter=3000,penalty = None)
          model.fit(X_train_selected, y_train)
          kf= KFold(n_splits=5, shuffle=True)
          cv= cross_val_score(model, X, y, cv=kf)
          print(f'Error From K-Fold Cross-Validation with {k} features::')
          print(1 - np.mean(cv))
          error = 1 - np.mean(cv)
          accuracy = model.score(X_test_selected, y_test)
          print(f"Accuracy: {accuracy}")
          test_error = 1 - accuracy
          return error, test_error
[95]: model = LogisticRegression(max_iter=3000, penalty='11', solver='liblinear')
      model.fit(X_train, y_train)
      coefs = model.coef_[0]
      abs_coefs = np.abs(coefs)
      cv_errors = []
      test_errors = []
      for k in range(0,13):
          indices = np.argpartition(abs_coefs, -k)[-k:]
          X_train_selected = X_train.iloc[:, indices]
          X_test_selected = X_test.iloc[:, indices]
          if k == 2:
              best_columns = X_train.columns[indices].tolist()
          cross_error, test_error = ErrorEstimate(X_train_selected, X_test_selected,_

y_train, y_test)

          cv_errors.append(cross_error)
          test_errors.append(test_error)
```

features=list(range(1,14))

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plt.figure
plt.plot(features, cv_errors, marker='o',linestyle='-', color='r',u
  →label='5-Fold Cross Validation Error')
plt.plot(features, test_errors ,marker='o', linestyle='-', color='b',_
  ⇔label='Test Error')
plt.xlabel('K')
plt.ylabel('Error')
plt.title('Error Estimation Per value of K')
plt.legend()
plt.grid(True)
plt.show()
Estimated Error From 5-Fold Cross-Validation with 0 features::
0.17158469945355181
Accuracy with 0 features: 0.8155339805825242
Estimated Error From 5-Fold Cross-Validation with 1 features::
0.171639344262295
Accuracy with 1 features: 0.6990291262135923
Estimated Error From 5-Fold Cross-Validation with 2 features::
0.17491803278688534
Accuracy with 2 features: 0.6990291262135923
Estimated Error From 5-Fold Cross-Validation with 3 features::
0.18513661202185783
Accuracy with 3 features: 0.7184466019417476
Estimated Error From 5-Fold Cross-Validation with 4 features::
0.1750273224043717
Accuracy with 4 features: 0.7281553398058253
Estimated Error From 5-Fold Cross-Validation with 5 features::
0.17158469945355193
Accuracy with 5 features: 0.7572815533980582
Estimated Error From 5-Fold Cross-Validation with 6 features::
0.17486338797814205
Accuracy with 6 features: 0.7864077669902912
Estimated Error From 5-Fold Cross-Validation with 7 features::
0.16169398907103827
Accuracy with 7 features: 0.8155339805825242
Estimated Error From 5-Fold Cross-Validation with 8 features::
0.1654098360655738
Accuracy with 8 features: 0.7961165048543689
Estimated Error From 5-Fold Cross-Validation with 9 features::
0.1618579234972678
Accuracy with 9 features: 0.7864077669902912
Estimated Error From 5-Fold Cross-Validation with 10 features::
0.1680874316939891
Accuracy with 10 features: 0.8058252427184466
Estimated Error From 5-Fold Cross-Validation with 11 features::
```

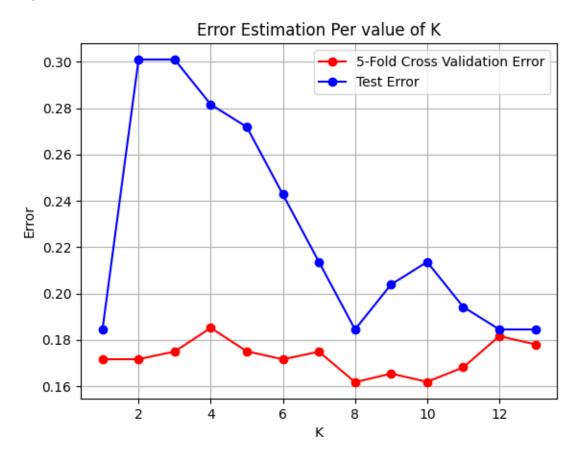
0.18158469945355193

Accuracy with 11 features: 0.8155339805825242

Estimated Error From 5-Fold Cross-Validation with 12 features::

0.17803278688524604

Accuracy with 12 features: 0.8155339805825242



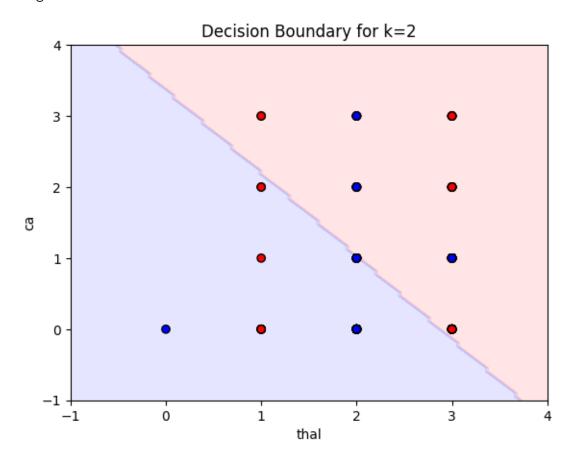
```
[98]: from matplotlib.colors import ListedColormap

def plot_decision_boundary(X, y, model, selected_features):
    X_selected = X[selected_features]
    x_min, x_max = X_selected.iloc[:, 0].min() - 1, X_selected.iloc[:, 0].max()_U
    + 1
    y_min, y_max = X_selected.iloc[:, 1].min() - 1, X_selected.iloc[:, 1].max()_U
    + 1
    xx, yy = np.meshgrid(np.linspace(x_min, x_max, 100), np.linspace(y_min,_U
    -y_max, 100))

Z = model.predict(np.c_[xx.ravel(), yy.ravel()])
    Z = Z.reshape(xx.shape)
```

['thal', 'ca']

/Library/Frameworks/Python.framework/Versions/3.12/lib/python3.12/site-packages/sklearn/utils/validation.py:2739: UserWarning: X does not have valid feature names, but LogisticRegression was fitted with feature names warnings.warn(



[]: