3647SHDS-attached python

December 8, 2022

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[44]: import numpy as np
      import pandas as pd
      from sklearn.model_selection import StratifiedKFold
      from sklearn.metrics import confusion_matrix, accuracy_score,_
       ⇔classification_report
      from sklearn.svm import SVC
      from sklearn.linear_model import LogisticRegression
      from sklearn.neighbors import KNeighborsClassifier
      from sklearn.ensemble import RandomForestClassifier
      from xgboost import XGBClassifier
      from xgboost import plot_importance
      import seaborn as sns
      from matplotlib import pyplot
      from collections import Counter
      from imblearn.over_sampling import RandomOverSampler
[47]: df = pd.read csv("../Documents/PHS/Statistics for HDS/module assessment/part2/
      ⇔child_cleaned.csv")
      df = df.drop("Unnamed: 0",axis=1)
      df
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[47]:
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      1846
      [1847 rows x 9 columns]
[63]: def get_dataset(return_X=True, over_sampling = True):
          X = df.to_numpy()[:,:-2]
          y = df["disease"]
          if over_sampling:
              ros = RandomOverSampler(random_state=0)
              X_resampled, y_resampled = ros.fit_resample(X, y)
              return X_resampled, np.array(y_resampled, dtype=int)
          return X, np.array(y, dtype=int)
      len(get_dataset()[1])
[63]: 3272
[51]: def data_split():
          X, y = get_dataset()
          split_indices = []
          skf = StratifiedKFold(n_splits=5, shuffle=True, random_state=42)
          for train_index, test_index in skf.split(X, y):
              split_indices.append((train_index, test_index))
          return split_indices
[56]: def train_and_test(model_name, svm_kernel='rbf'):
          X, y = get_dataset()
          split_indices = data_split()
          y_pre, y_tru = np.array([]), np.array([])
          for train_indices, test_indices in split_indices:
              X_train, X_test = X[train_indices], X[test_indices]
              y_train, y_test = y[train_indices], y[test_indices]
              if model_name=='LR':
                  model = LogisticRegression(solver='lbfgs', max_iter=10000)
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[38]: def plot_heatmap(res, model_name):
    print('Model:', model_name)
    classes = [0,1]
    y_pred, y_test = res['pre'].astype(int), res['tru'].astype(int)

    vis = confusion_matrix(y_test, y_pred)
    accu = accuracy_score(y_test, y_pred)

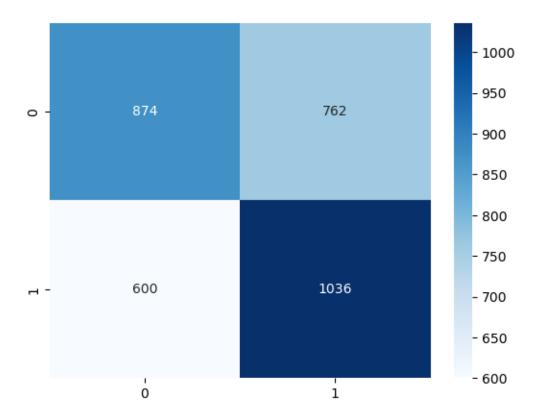
    #print(classification_report(y_test, y_pred))
    print('Accuracy: %.2f%'' % (accu*100))

    pd_vis = pd.DataFrame(vis, columns=classes, index=classes)
    ax = sns.heatmap(pd_vis, cmap="Blues", annot=True, fmt='d')
```

[64]: res_lr = train_and_test('LR')
plot_heatmap(res_lr, 'Multinominal Logistic Regression')

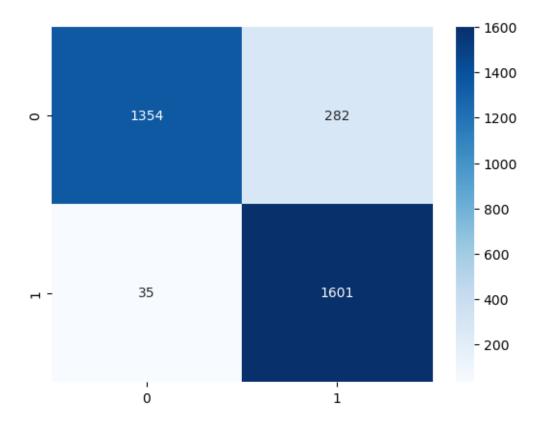
Model: Multinominal Logistic Regression

Accuracy: 58.37%



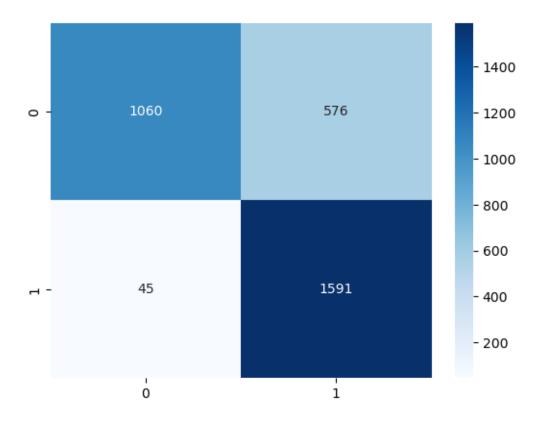
```
[58]: res_xgboost = train_and_test('XGBoost')
plot_heatmap(res_xgboost, 'XGBoost')
```

Model: XGBoost Accuracy: 90.31%



Model: KNN

Accuracy: 81.02%



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[60]: res_svm_rbf = train_and_test('SVM', 'rbf')
plot_heatmap(res_svm_rbf, 'SVM, kernel: rbf')
```

Model: SVM, kernel: rbf

Accuracy: 59.50%

