Model Development Phase

Date	11 July 2024
Team ID	SWTID1720013031
Project Title	Prediction and Analysis of Liver Patient Data Using Machine Learning
Maximum Marks	4 Marks

Initial Model Training Code, Model Validation and Evaluation Report

The initial model training code will be showcased in the future through a screenshot. The model validation and evaluation report will include classification reports, accuracy, and confusion matrices for multiple models, presented through respective screenshots.

Initial Model Training Code:

```
x = data.iloc[:, 0:-1]
y = data.iloc[:, -1]
```

```
x_train, x_test, y_train, y_test = train_test_split(x, y, random_state=0, test_size=0.3)
```

```
x_train.shape
```

```
(408, 10)
```

```
x_test.shape
```

```
(175, 10)
```

Logistic Regression

```
lr_param_grid = {
    'penalty': ['l1', 'l2'],
    'C': [0.001, 0.01, 0.1, 1.0, 10.0],
    'solver': ['liblinear', 'saga']
}
```

```
lr_s = LogisticRegression(max_iter=1000)
```

```
lr_grid_search = GridSearchCV(estimator=lr_s, param_grid=lr_param_grid, cv=5, scoring='accuracy', verbose=1, n_jobs=-1)
```

```
lr_grid_search.fit(x_train_final, y_train)

# Get best parameters
lr_best_params = lr_grid_search.best_params_
print("Best parameters for Logistic Regression:", lr_best_params)
```

```
Fitting 5 folds for each of 20 candidates, totalling 100 fits
Best parameters for Logistic Regression: {'C': 1.0, 'penalty': 'l1', 'solver': 'liblinear'}
```

```
lr = LogisticRegression(**lr_best_params, max_iter=1000)
```

```
lr.fit(x_train_final, y_train)
```

```
▾ LogisticRegression
LogisticRegression(max_iter=1000, penalty='l1', solver='liblinear')
```

```
y_pred_lr = lr.predict(x_test_final)
```

```
lr_acc = accuracy_score(y_pred_lr, y_test)
lr_acc
```

```
0.6971428571428572
```

```
print(classification_report(y_test, y_pred_lr))
```

```
              precision    recall  f1-score   support

           1       0.72      0.93      0.81       122
           2       0.50      0.15      0.23        53

    accuracy                           0.70       175
   macro avg       0.61      0.54      0.52       175
weighted avg       0.65      0.70      0.64       175
```

```
lr_cross = cross_val_score(lr, x_train_final, y_train, scoring='accuracy', cv = 6)
lr_cross.mean()
```

```
0.7352941176470589
```

```
lr_cm = confusion_matrix(y_pred_lr,y_test)
lr_cm
```

```
array([[114,  45],
       [  8,   8]], dtype=int64)
```

Support Vector Classifier (SVC)

```
svm_param_dist = {
    'C': uniform(0.1, 10),
    'kernel': ['linear', 'rbf'],
    'gamma': ['scale', 'auto']
}
```

```
svm_s = SVC()
```

```
# Initialize GridSearchCV
svm_random_search = RandomizedSearchCV(estimator=svm_s, param_distributions=svm_param_dist, n_iter=10, cv=3, scoring='accuracy', verbose=2, n_jobs=2,
# Fit GridSearchCV
svm_random_search.fit(x_train_final, y_train)
```

```
GridSearchCV
arch = RandomizedSearchCV(estimator=svm_s, param_distributions=svm_param_dist, n_iter=10, cv=3, scoring='accuracy', verbose=2, n_jobs=2, random_state=42)
rchCV
arch.fit(x_train_final, y_train)
```

```
Fitting 3 folds for each of 10 candidates, totalling 30 fits
▸ RandomizedSearchCV
  ▸ estimator: SVC
      ▸ SVC
```

```
svm_best_params = svm_random_search.best_params_
print("Best parameters for SVC:", svm_best_params)
```

```
Best parameters for SVC: {'C': 7.41993941811405, 'gamma': 'scale', 'kernel': 'linear'}
```

```
svm = SVC(**svm_best_params)
```

```
svm.fit(x_train_final, y_train)
```

```
▾ SVC
SVC(C=7.41993941811405, kernel='linear')
```

```
y_pred_svm = svm.predict(x_test_final)
```

```
svm_acc = accuracy_score(y_pred_svm, y_test)
svm_acc
```

```
0.6971428571428572
```

```
print(classification_report(y_test, y_pred_svm))
```

```
              precision    recall  f1-score   support

           1       0.70      1.00      0.82       122
           2       0.00      0.00      0.00        53

    accuracy                           0.70       175
   macro avg       0.35      0.50      0.41       175
weighted avg       0.49      0.70      0.57       175
```

```
svm_cross = cross_val_score(svm, x_train_final, y_train, scoring='accuracy', cv = 6)
svm_cross.mean()
```

```
0.7181372549019608
```

```
svm_cm = confusion_matrix(y_pred_svm,y_test)
svm_cm
```

```
array([[122,  53],
       [  0,   0]], dtype=int64)
```

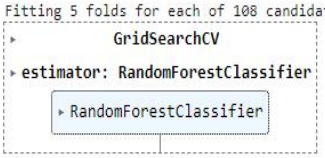
Random Forest Classifier

```
rfc_param_grid = {
    'n_estimators': [100, 200, 300],
    'max_depth': [None, 10, 20, 30],
    'min_samples_split': [2, 5, 10],
    'min_samples_leaf': [1, 2, 4]
}
```

```
rfc_s = RandomForestClassifier()
```

```
rfc_grid_search = GridSearchCV(estimator=rfc_s, param_grid=rfc_param_grid, cv=5, scoring='accuracy', verbose=1, n_jobs=-1)
rfc_grid_search.fit(x_train_final, y_train)
```

```
Fitting 5 folds for each of 108 candidates, totalling 540 fits
```




```
rfc_best_params = rfc_grid_search.best_params_
print("Best parameters for Random Forest Classifier:", rfc_best_params)
```

```
Best parameters for Random Forest Classifier: {'max_depth': 10, 'min_samples_leaf': 2, 'min_samples_split': 10, 'n_estimators': 100}
```

```
rfc = RandomForestClassifier(**rfc_best_params)
rfc.fit(x_train_final, y_train)
```

```
rfc = RandomForestClassifier(**rfc_best_params)
rfc.fit(x_train_final, y_train)
```

```
RandomForestClassifier(max_depth=10, min_samples_leaf=2, min_samples_split=10)
```

```
ypred_rfc = rfc.predict(x_test_final)
```

```
rfc_acc = accuracy_score(ypred_rfc, y_test)
rfc_acc
```

```
0.6971428571428572
```

```
print(classification_report(y_test, ypred_rfc))
```

```
              precision    recall  f1-score   support

           1       0.73      0.90      0.81       122
           2       0.50      0.23      0.31        53

    accuracy                           0.70       175
   macro avg       0.61      0.56      0.56       175
weighted avg       0.66      0.70      0.66       175
```

```
rfc_cross = cross_val_score(rfc, x_train_final, y_train, scoring='accuracy', cv = 6)
rfc_cross.mean()
```

```
0.7205882352941178
```

```
rfc_cm = confusion_matrix(ypred_rfc,y_test)
rfc_cm
```

```
array([[110,  41],
       [ 12,  12]], dtype=int64)
```

K Neighbors Classifier

```
knn_param_grid = {
    'n_neighbors': [3, 5, 7, 9],
    'weights': ['uniform', 'distance'],
    'metric': ['euclidean', 'manhattan']
}
```

```
knn_s = KNeighborsClassifier()
```

```
knn_grid_search = GridSearchCV(estimator=knn_s, param_grid=knn_param_grid, cv=5, scoring='accuracy', verbose=1, n_jobs=-1)

# Fit GridSearchCV
knn_grid_search.fit(x_train_final, y_train)
```

```
Fitting 5 folds for each of 16 candidates, totalling 80 fits
```

GridSearchCV
estimator: KNeighborsClassifier
KNeighborsClassifier

```
knn_best_params = knn_grid_search.best_params_
print("Best parameters for K Neighbors Classifier:", knn_best_params)
```

```
Best parameters for K Neighbors Classifier: {'metric': 'manhattan', 'n_neighbors': 3, 'weights': 'distance'}
```

```
knn = KNeighborsClassifier(**knn_best_params)
knn.fit(x_train_final, y_train)
```

KNeighborsClassifier
KNeighborsClassifier(metric='manhattan', n_neighbors=3, weights='distance')

```
ypred_knn = knn.predict(x_test_final)
```

```
knn_acc = accuracy_score(ypred_knn, y_test)
knn_acc
```

```
0.6628571428571428
```

```
print(classification_report(y_test, ypred_knn))
```

```
              precision    recall  f1-score   support

           1       0.74      0.80      0.77       122
           2       0.43      0.36      0.39        53

    accuracy                           0.66       175
   macro avg       0.59      0.58      0.58       175
weighted avg       0.65      0.66      0.65       175
```

```
knn_cross = cross_val_score(knn, x_train_final, y_train, scoring='accuracy', cv = 6)
knn_cross.mean()
```

```
0.7181372549019609
```

```
knn_cm = confusion_matrix(ypred_knn,y_test)
knn_cm
```

```
array([[97, 34],
       [25, 19]], dtype=int64)
```

```
pickle.dump(svm, open('svm_liver_analysis.pkl', 'wb'))
pickle.dump(rfc, open('rfc_liver_analysis.pkl', 'wb'))
pickle.dump(knn, open('knn_liver_analysis.pkl', 'wb'))
pickle.dump(lr, open('lr_liver_analysis.pkl', 'wb'))
```

Model Validation and Evaluation Report:

Model	Classification Report	Accuracy	Confusion Matrix

Logistic Regression	`print(classification_report(y_test, y_pred_lr))`
	precision recall f1-score support
	1 0.72 0.93 0.81 122
	2 0.50 0.15 0.23 53
	accuracy 0.70 175
	macro avg 0.61 0.54 0.52 175
	weighted avg 0.65 0.70 0.64 175
		69.71%
		`lr_acc = accuracy_score(y_pred_lr, y_test)`
		`lr_acc`
		0.6971428571428572
			`lr_cm = confusion_matrix(y_pred_lr,y_test)`
			`lr_cm`
			array([[114, 45],
			[8, 8]], dtype=int64)

Support Vector Classifier (SVC)	`print(classification_report(y_test, y_pred_svm))`
	precision recall f1-score support
	1 0.70 1.00 0.82 122
	2 0.00 0.00 0.00 53
	accuracy 0.70 175
	macro avg 0.35 0.50 0.41 175
	weighted avg 0.49 0.70 0.57 175
		69.71%
		`svm_acc = accuracy_score(y_pred_svm, y_test)`
		`svm_acc`
		0.6971428571428572
			`svm_cm = confusion_matrix(y_pred_svm,y_test)`
			`svm_cm`
			array([[122, 53],
			[0, 0]], dtype=int64)

Random Forest Classifier	`print(classification_report(y_test, ypred_rfc))`
	precision recall f1-score support
	1 0.73 0.90 0.81 122
	2 0.50 0.23 0.31 53
	accuracy 0.70 175
	macro avg 0.61 0.56 0.56 175
	weighted avg 0.66 0.70 0.66 175
		69.7%
		`rfc_acc = accuracy_score(ypred_rfc, y_test)`
		`rfc_acc`
		0.6971428571428572
			`rfc_cm = confusion_matrix(ypred_rfc,y_test)`
			`rfc_cm`
			array([[110, 41],
			[12, 12]], dtype=int64)

K Neighbors Classifier	`print(classification_report(y_test, ypred_knn))`
	precision recall f1-score support
	1 0.74 0.80 0.77 122
	2 0.43 0.36 0.39 53
	accuracy 0.66 175
	macro avg 0.59 0.58 0.58 175
	weighted avg 0.65 0.66 0.65 175
		66.2%
		`knn_acc = accuracy_score(ypred_knn, y_test)`
		`knn_acc`
		0.6628571428571428
			`knn_cm = confusion_matrix(ypred_knn,y_test)`
			`knn_cm`
			array([[97, 34],
			[25, 19]], dtype=int64)