

## Project Development Phase Model Performance Test

Date	19 November 2022
Team ID	PNT2022TMID32547
Project Name	Project – Early Detection of Chronic Kidney Disease using Machine Learning
Maximum Marks	10 Marks

### Model Performance Testing:

Project team shall fill the following information in model performance testing template.

S.No.	Parameter	Values	Screenshot																														
1.	Metrics	<p><b>Regression Model:</b> MAE - , MSE - , RMSE - , R2 score -</p> <p><b>Classification Model:</b> Confusion Matrix - , Accuracy Score- &amp; Classification Report -</p>	<p>Classification Model</p> <p>Accuracy</p> <pre>[ ] from sklearn.metrics import get_scorer  ypred = pipe.predict(xtrain.values) scorer = get_scorer('accuracy') score = scorer(pipe, xtest.values, ytest.values) print("Accuracy for training data:",accuracy_score(ytrain,ypred)) print("Accuracy for test data:",score)</pre> <p>Accuracy for training data: 1.0 Accuracy for test data: 1.0</p> <p>Confusion Matrix</p> <pre>[ ] ypred = pipe.predict(xtest.values)  cm = confusion_matrix(ytest, ypred) print(cm)</pre> <pre>[[30  0]  [ 0 12]]</pre> <p>Classification Report</p> <pre>[ ] print(classification_report(ytest, ypred))</pre> <table><thead><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr></thead><tbody><tr><td>ckd</td><td>1.00</td><td>1.00</td><td>1.00</td><td>30</td></tr><tr><td>notckd</td><td>1.00</td><td>1.00</td><td>1.00</td><td>12</td></tr><tr><td>accuracy</td><td></td><td></td><td>1.00</td><td>42</td></tr><tr><td>macro avg</td><td>1.00</td><td>1.00</td><td>1.00</td><td>42</td></tr><tr><td>weighted avg</td><td>1.00</td><td>1.00</td><td>1.00</td><td>42</td></tr></tbody></table>		precision	recall	f1-score	support	ckd	1.00	1.00	1.00	30	notckd	1.00	1.00	1.00	12	accuracy			1.00	42	macro avg	1.00	1.00	1.00	42	weighted avg	1.00	1.00	1.00	42
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2.	Tune the Model	Hyperparameter Tuning - Validation Method -	<div>Hyperparameter Tuning</div> <div><pre>[ ] #n_estimators parameters = {     'n_estimators': [10, 20, 50, 100, 200, 500, 1000, 1200, 1500, 1800, 1900, 2000, 2100, 3000] } clf = GridSearchCV(extra_trees_classifier, parameters, cv=5, verbose=3, n_jobs=-1) pipe = make_pipeline(column_trans, clf)  pipe.fit(xtrain.values, ytrain.values.ravel());  pred=pipe.predict(xtest.values) print(classification_report(ytest, pred))</pre></div> <div>Fitting 5 folds for each of 14 candidates, totalling 70 fits</div> <div><table><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr><tr><td>ckd</td><td>1.00</td><td>1.00</td><td>1.00</td><td>29</td></tr><tr><td>notckd</td><td>1.00</td><td>1.00</td><td>1.00</td><td>13</td></tr><tr><td>accuracy</td><td></td><td></td><td>1.00</td><td>42</td></tr><tr><td>macro avg</td><td>1.00</td><td>1.00</td><td>1.00</td><td>42</td></tr><tr><td>weighted avg</td><td>1.00</td><td>1.00</td><td>1.00</td><td>42</td></tr></table></div> <div><pre>[ ] #criterion parameters = {     'criterion': ['gini', 'entropy'] } clf = GridSearchCV(extra_trees_classifier, parameters, cv=5, verbose=3, n_jobs=-1) pipe = make_pipeline(column_trans, clf)  pipe.fit(xtrain.values, ytrain.values.ravel());  pred=pipe.predict(xtest.values) print(classification_report(ytest, pred))</pre></div> <div>Fitting 5 folds for each of 2 candidates, totalling 10 fits</div> <div><table><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr><tr><td>ckd</td><td>1.00</td><td>1.00</td><td>1.00</td><td>29</td></tr><tr><td>notckd</td><td>1.00</td><td>1.00</td><td>1.00</td><td>13</td></tr><tr><td>accuracy</td><td></td><td></td><td>1.00</td><td>42</td></tr><tr><td>macro avg</td><td>1.00</td><td>1.00</td><td>1.00</td><td>42</td></tr><tr><td>weighted avg</td><td>1.00</td><td>1.00</td><td>1.00</td><td>42</td></tr></table></div> <div><pre>[ ] #max_depth parameters = {     'max_depth': [1, 2, 5, 8, 13, 21, 34, 53, 54, 55, 89, None] } clf = GridSearchCV(extra_trees_classifier, parameters, cv=5, verbose=3, n_jobs=-1) pipe = make_pipeline(column_trans, clf)  pipe.fit(xtrain.values, ytrain.values.ravel());  pred=pipe.predict(xtest.values) print(classification_report(ytest, pred))</pre></div> <div>Fitting 5 folds for each of 12 candidates, totalling 60 fits</div> <div><table><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr><tr><td>ckd</td><td>1.00</td><td>1.00</td><td>1.00</td><td>29</td></tr><tr><td>notckd</td><td>1.00</td><td>1.00</td><td>1.00</td><td>13</td></tr><tr><td>accuracy</td><td></td><td></td><td>1.00</td><td>42</td></tr><tr><td>macro avg</td><td>1.00</td><td>1.00</td><td>1.00</td><td>42</td></tr><tr><td>weighted avg</td><td>1.00</td><td>1.00</td><td>1.00</td><td>42</td></tr></table></div> <div><pre>[ ] #min_samples_leaf parameters = {     'min_samples_leaf': [1, 2, 5, 8, 13, 21, 34, 55, 89, 144, 233, 377] } clf = GridSearchCV(extra_trees_classifier, parameters, cv=5, verbose=3, n_jobs=-1) pipe = make_pipeline(column_trans, clf)  pipe.fit(xtrain.values, ytrain.values.ravel());  pred=pipe.predict(xtest.values) print(classification_report(ytest, pred))</pre></div> <div>Fitting 5 folds for each of 12 candidates, totalling 60 fits</div> <div><table><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr><tr><td>ckd</td><td>1.00</td><td>1.00</td><td>1.00</td><td>29</td></tr><tr><td>notckd</td><td>1.00</td><td>1.00</td><td>1.00</td><td>13</td></tr><tr><td>accuracy</td><td></td><td></td><td>1.00</td><td>42</td></tr><tr><td>macro avg</td><td>1.00</td><td>1.00</td><td>1.00</td><td>42</td></tr><tr><td>weighted avg</td><td>1.00</td><td>1.00</td><td>1.00</td><td>42</td></tr></table></div>		precision	recall	f1-score	support	ckd	1.00	1.00	1.00	29	notckd	1.00	1.00	1.00	13	accuracy			1.00	42	macro avg	1.00	1.00	1.00	42	weighted avg	1.00	1.00	1.00	42		precision	recall	f1-score	support	ckd	1.00	1.00	1.00	29	notckd	1.00	1.00	1.00	13	accuracy			1.00	42	macro avg	1.00	1.00	1.00	42	weighted avg	1.00	1.00	1.00	42		precision	recall	f1-score	support	ckd	1.00	1.00	1.00	29	notckd	1.00	1.00	1.00	13	accuracy			1.00	42	macro avg	1.00	1.00	1.00	42	weighted avg	1.00	1.00	1.00	42		precision	recall	f1-score	support	ckd	1.00	1.00	1.00	29	notckd	1.00	1.00	1.00	13	accuracy			1.00	42	macro avg	1.00	1.00	1.00	42	weighted avg	1.00	1.00	1.00	42
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### Validation Method

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[ ] cv_scores = cross_val_score(pipe, xtrain, ytrain, cv=5 )
print("Cross Validation average score: %.2f" % cv_scores.mean())
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