Project Development Phase Model Performance Test

Date	10 November 2022
Team ID	PNT2022TMID29707
Project Name	Project – WEB PHISHING DETECTION
Maximum Marks	10 Marks

Model Performance Testing:

1.Metrices Classification Model

Confusion Matrix, Accuracy Score & Classification Report

log_reg = LogisticRegression()

[24] from sklearn.linear_model import LogisticRegression

Logistic Regression

```
log_reg.fit(x_train,y_train)
    LogisticRegression()
pred3= log_reg.predict(x_test)
    s3=accuracy_score(y_test,pred3)*100
    results('LR',s3)
    print("Accuracy score :",accuracy_score(y_test,pred3)*100)
    print("Precision score :",precision_score(y_test,pred3,average="macro")*100)
    print("Confustion matrix :",confusion_matrix(y_test,pred3))
    from sklearn.metrics import classification_report
    p=log_reg.predict(X)
    print(classification_report(Y,p))
    Accuracy score : 92.67299864314789
    Precision score : 92.56074521739423
    Confustion matrix : [[ 856 86]
     [ 76 1193]]
                  precision recall f1-score support
                                 0.91
              -1
                       0.93
                                           0.92
                                                      4897
                       0.93
                                 0.95
                                           0.94
                                                      6157
                                           0.93
                                                    11054
        accuracy
   accuracy 0.93 11054
macro avg 0.93 0.93 0.93 11054
weighted avg 0.93 0.93 0.93 11054
```

2. Tune The Model Hyper Parameter Tuning

Hyper parameter tuning

```
from sklearn.linear_model import LogisticRegression
          log_reg = LogisticRegression()
          log_reg.fit(x_train,y_train)
         LogisticRegression()
     \Gamma
[43] log_reg.get_params()
          {'C': 1.0,
           'class_weight': None,
'dual': False,
            'fit_intercept': True,
           'intercept_sca__
'l1_ratio': None,
'max_iter': 100,
'ti class': 'auto',
            'intercept_scaling': 1,
           "multi_class': 'auto'
'n_jobs': None,
'penalty': 'l2',
'random_state': None,
'solver': 'lbfgs',
           'tol': 0.0001,
'verbose': 0,
'warm_start': False}
 / [46] parameters = {
           "l1_ratio":[4,5,6,7,8,9],
           "random_state":[0,1,10,42,100],
            "n_jobs":[10,20,30,40,50]
        }
 [51] from sklearn.model_selection import GridSearchCV
        grid = GridSearchCV(log_reg,parameters,cv = 5,verbose=2)
 grid.fit(X,Y)
        [CV] END ......l1_ratio=9, n_jobs=30, random_state=100; total time=
        [CV] END .......l1_ratio=9, n_jobs=30, random_state=100; total time=
                                                                            0.45
        [CV] END .........11_ratio=9, n_jobs=30, random_state=100; total time=
            [CV]
                                                                            0.45
        [CV] END ......l1_ratio=9, n_jobs=40, random_state=0; total time=
        [CV] END ......l1_ratio=9, n_jobs=40, random_state=0; total time=
        [CV] END ......l1_ratio=9, n_jobs=40, random_state=0; total time=
        [CV] END ..........11_ratio=9, n_jobs=40, random_state=0; total time=
        [CV] END ......l1_ratio=9, n_jobs=40, random_state=1; total time=
                                                                            0.45
        [CV] END ...........11_ratio=9, n_jobs=40, random_state=1; total time=
                                                                            1.15
        [CV] END ......l1_ratio=9, n_jobs=40, random_state=1; total time=
        [CV] END ...........11_ratio=9, n_jobs=40, random_state=1; total time=
                                                                            1.15
        [CV] END ...........11_ratio=9, n_jobs=40, random_state=1; total time=
                                                                            1.15
        [CV] END ......l1_ratio=9, n_jobs=40, random_state=10; total time=
                                                                            1.15
```

```
END ......11_ratio=9, n_jops=40, random_state=1; total time=
- O
     [CV]
        END ...........11_ratio=9, n_jobs=40, random_state=1; total time=
                                                         1.15
        END ........l1_ratio=9, n_jobs=40, random_state=10; total time= END .......l1_ratio=9, n_jobs=40, random_state=10; total time=
     [CV]
                                                         1.15
  ₽
     [CV]
                                                         1.15
     [CV]
        total time=
                                                         1.15
     LCV1
        END ......l1_ratio=9, n_jobs=40, random_state=10; total time=
                                                         1.15
        [CV]
                                                         1.15
        END ......l1_ratio=9, n_jobs=40, random_state=42;
     [CV]
                                                total time=
     CV1
        END ...........11_ratio=9, n_jobs=40, random_state=42;
                                                total time=
                                                         1 15
     total time=
                                                         1.15
     total time=
     1.15
                                                         1.15
     total time=
                                                         1.15
                                                total time=
      CV1
        FND
           ......l1_ratio=9, n_jobs=40, random_state=100;
                                                         1.15
     1.15
     1.15
           ......11_ratio=9, n_jobs=50, random_state=0;
      [CV]
        END
                                                total time=
                                                         1.85
     [CV] END ......l1_ratio=9, n_jobs=50, random_state=0; total time=
                                                         1.15
     1.15
     [CV]
           total time=
     CV1
        1.15
     0.45
     [CV]
           ......l1_ratio=9, n_jobs=50, random_state=1; total time=
     [CV] END ......l1_ratio=9, n_jobs=50, random_state=1;
                                               total time=
                                                         0.45
     LCV1
        END ......l1_ratio=9, n_jobs=50, random_state=1; total time=
                                                         9.45
     [cv]
        END ...........11_ratio=9, n_jobs=50, random_state=1; total time=
                                                         0.45
        [CV]
                                                         0.45
      CVI
        END ......l1_ratio=9, n_jobs=50, random_state=10;
                                                total time=
                                                         9 45
        END
           [CV]
                                                total time=
                                                         0.45
     [CV] END
           ......l1_ratio=9, n_jobs=50, random_state=10; total time=
                                                         0.45
     [CV]
        FND
           total time=
                                                         0.45
     1.15
     [CV]
        END ...........11_ratio=9, n_jobs=50, random_state=1; total time=
 0
     [CV] END ...........11_ratio=9, n_jobs=50, random_state=1; total time=
                                                         0.45
     [CV]
        END ......l1_ratio=9, n_jobs=50, random_state=10; total time=
                                                         0.45
     [CV] END ......l1_ratio=9, n_jobs=50, random_state=10; total time=
     9.45
     [CV]
        END ..........11_ratio=9, n_jobs=50, random_state=10; total time=
                                                         0.45
     [CV] END ......l1_ratio=9, n_jobs=50, random_state=42; total time=
                                                         1.15
        END ...........11_ratio=9, n_jobs=50, random_state=42; total time=
     [CV]
                                                         1.15
     [CV] END .........11_ratio=9, n_jobs=50, random_state=100; total time=
                                                         1.15
     CV1
        END ......l1_ratio=9, n_jobs=50, random_state=100; total time=
                                                         1.15
     [CV] END ............11_ratio=9, n_jobs=50, random_state=100; total time=
                                                        1.15
     [CV]
        1.15
     [CV] END ......l1_ratio=9, n_jobs=50, random_state=100; total time= 1.1s
     GridSearchCV(cv=5, estimator=LogisticRegression(),
              param_grid={'l1_ratio': [4, 5, 6, 7, 8, 9],
                      'n_jobs': [10, 20, 30, 40, 50],
'random_state': [0, 1, 10, 42, 100]},
             verbose=2)
/ [54] print(grid.best_params_)
     print(grid.best_estimator_)
     print(grid.best_score_)
     {'l1_ratio': 4, 'n_jobs': 10, 'random_state': 0}
     LogisticRegression(l1_ratio=4, n_jobs=10, random_state=0)
```

0.9225618104459195

Validation Method

Validation Method

```
#Logistic Regression
    pred3= log_reg.predict(x_test)
    s3=accuracy_score(y_test,pred3)*100
    results('LR',s3)
    print("Accuracy score of LR :",accuracy_score(y_test,pred3)*100)
    #Random Forest
    pred2=R_model.predict(x_test)
    s2=accuracy_score(y_test,pred2)*100
    print("Accuracy score of RF :",s2)
    results('RF',s2)
    #KNN
    pred5=modellin.predict(x_test)
    s5=accuracy_score(y_test,pred5)*100
    print("Accuracy score of KNN :",s5)
    results('SVM-LIN',s5)
    #SVM Linear
    pred5=modellin.predict(x_test)
    s5=accuracy_score(y_test,pred5)*100
    print("Accuracy score SVM :",s5)
    results('SVM-LIN',s5)
    #SVM Poly
    pred6 = modelpoly.predict(x_test)
    s6=accuracy_score(y_test,pred6)*100
    print("Accuracy score of SVM Poly :",s6)
    results('SVM poly', s6)
```

Accuracy score of LR: 92.67299864314789 Accuracy score of RF: 97.33152419719585 Accuracy score of KNN: 92.67299864314789 Accuracy score SVM: 92.67299864314789 Accuracy score of SVM Poly: 95.74853007688829