|                  | <pre>from sklearn.tree import DecisionTreeClassifier from sklearn.naive_bayes import MultinomialNB from sklearn.metrics import classification_report, confusion_matrix, accuracy_score</pre>  |
|------------------|---|
| n [ ]:           | <pre># data training di definisikan  dt_train = {</pre>   |
|                  | 'no' : [1, 2, 3, 4, 5, 6, 7, 8, 9, 10], 'warna' : ['Hitam', 'Putih', 'Cokelat', 'Hitam', 'Cokelat', 'Cokelat', 'Putih', 'Putih', 'Cokelat'], 'bulu' : ['Panjang', 'Panjang', 'Pendek', 'Gimbal', 'Gimbal', 'Panjang', 'Pendek', 'Gimbal'], 'makanan_kesukaan' : ['Wortel', 'Wortel', 'Kangkung', 'Kangkung', 'Rumput', 'Rumput', 'Wortel', 'Kangkung', 'Kangkung'], 'jenis' : ['A', 'A', 'B', 'B', 'C', 'A', 'B', 'C', 'C'],  |
| n [ ]:           | # data testing di definisikan   |
|                  | <pre>dt_test = {   'no'</pre>   |
|                  | 'prediksi_jenis' : ['?', '?', '?', '?'], 'jenis_sebenarnya' : ['A', 'B', 'C', 'C', 'B'] }   |
| n [ ]:           | <pre># membandingkan dt_training dengan dt_testing  dt_train = pd.DataFrame(dt_train)</pre>   |
| n [ ]·           | <pre>dt_test = pd.DataFrame(dt_test) dt_hasil = dt_test.copy()</pre>  |
| . [ ].           | <pre># run dt_training  dt_train</pre>  |
| ut[]:            | nowarnabulumakanan_kesukaanjenis01HitamPanjangWortelA12PutihPanjangWortelA  |
|                  | 2 3 Cokelat Pendek Kangkung B 3 4 Hitam Pendek Kangkung B 4 5 Hitam Gimbal Rumput C   |
|                  | <ul> <li>5 6 Cokelat Gimbal Rumput A</li> <li>6 7 Cokelat Panjang Rumput B</li> <li>7 8 Putih Pendek Wortel B</li> </ul>  |
|                  | 89PutihPendekKangkungC910CokelatGimbalKangkungC   |
| n [ ]:           | <pre># run dt_testing dt_test</pre>   |
| ut[ ]:           | nowarnabulumakanan_kesukaanprediksi_jenisjenis_sebenarnya01HitamGimbalRumput?A12PutihPendekWortel?B   |
|                  | 23CokelatPanjangKangkung?C34PutihPanjangRumput?C  |
| n [ ]:           | 4 5 Cokelat Gimbal Kangkung ? B # dt_train mengubah label kategori manjadi nilai number   |
|                  | <pre>for cols in dt_train.select_dtypes(include='object') :    dt_train[cols] = dt_train[cols].astype('category').cat.codes</pre>   |
| n [ ]:           | <pre># dt_test mengubah label kategori manjadi nilai number  for cols in dt_test.select_dtypes(include='object') :     dt_test[cols] = dt_test[cols].astype('category').cat.codes</pre>   |
| n [ ]:           | <pre># run dt_training setelah diubah menjadi number dt_train</pre>   |
| ut[ ]:           | no warna bulu makanan_kesukaan jenis  0 1 1 1 2 0   |
|                  | 1       2       2       1       2       0         2       3       0       2       0       1         3       4       1       2       0       1   |
|                  | 4       5       1       0       1       2         5       6       0       0       1       0         6       7       0       1       1       1   |
|                  | 7       8       2       2       1         8       9       2       2       0       2   |
| n [ ]:           | 9 10 0 0 0 2  # run dt_testing setelah diubah menjadi number  |
| ut[ ]:           | no warna bulu makanan_kesukaan prediksi_jenis jenis_sebenarnya  |
|                  | 0       1       1       0       0         1       2       2       2       0       1         2       3       0       1       0       0       2   |
|                  | 3       4       2       1       1       0       2         4       5       0       0       0       0       1   |
| n [ ]:           | <pre># set feature dan target variabel  feature_cols = ['warna', 'bulu', 'makanan_kesukaan'] x_train = dt_train[feature_cols] y_train = dt_train.jenis x_test = dt_test[feature_cols] y_test = dt_test.jenis_sebenarnya</pre>   |
| n [ ]:           | <pre># running decision tree algorithm  clf = DecisionTreeClassifier(max_depth=2)</pre>   |
| n [ ]:           | <pre>clf = clf.fit(x_train, y_train)  y_prediksi = clf.predict(x_test) result = confusion_matrix(y_test, y_prediksi)</pre>  |
|                  | <pre>print("Confusion Matrix:") print(result) result1 = classification_report(y_test, y_prediksi) print("Classification Report:") print(result1) result2 = accuracy_score(y_test, y_prediksi) print("Accuracy:", result2)</pre>   |
|                  | Confusion Matrix: [[0 0 1] [0 1 1] [0 0 2]] Classification Report:  |
|                  | precision recall f1-score support  0 0.00 0.00 0.00 1 1 1.00 0.50 0.67 2  |
|                  | 2 0.50 1.00 0.67 2  accuracy 0.60 5 macro avg 0.50 0.50 0.44 5 weighted avg 0.60 0.60 0.53 5  |
|                  | Accuracy: 0.6  D:\program\Python39\lib\site-packages\sklearn\metrics\_classification.py:1308: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.   |
|                  | <pre>_warn_prf(average, modifier, msg_start, len(result)) D:\program\Python39\lib\site-packages\sklearn\metrics\_classification.py:1308: UndefinedMetricWarning: Precision and F-score are ill-defined and being set t 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.     _warn_prf(average, modifier, msg_start, len(result)) D:\program\Python39\lib\site-packages\sklearn\metrics\_classification.py:1308: UndefinedMetricWarning: Precision and F-score are ill-defined and being set t</pre> |
| n [ ]:           | <pre>0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.    _warn_prf(average, modifier, msg_start, len(result))  # running naive bayes algorithm</pre>  |
|                  | <pre>clf2 = MultinomialNB() clf2 = clf2.fit(x_train, y_train)</pre>   |
| n [ ]:           | <pre>y_prediksi2 = clf2.predict(x_test) result = confusion_matrix(y_test, y_prediksi2) print("Confusion Matrix:") print(result) result1 = classification_report(y_test, y_prediksi2)</pre>  |
|                  | <pre>print("Classification Report:") print(result1) result2 = accuracy_score(y_test, y_prediksi2) print("Accuracy:", result2)</pre>   |
|                  | Confusion Matrix:<br>[[1 0 0]<br>[0 2 0]<br>[0 1 1]]  |
|                  | Classification Report:     precision recall f1-score support  0 1.00 1.00 1.00 1  |
|                  | 2 1.00 0.50 0.67 2  accuracy 0.80 5 macro avg 0.89 0.83 0.82 5  |
| n [ ]:           | weighted avg 0.87 0.80 0.79 5  Accuracy: 0.8  # hasil setelah dilakukannya decision tree dan naive bayes  |
| -                | # masil setelah dilakukannya decision tree dan naive bayes  # mengambil variabel dari algoritma decision tree  dt_hasil['prediksi_jenis_decision_tree'] = y_prediksi  # mengambil variabel dari algoritma naive bayes   |
|                  |   |
|                  | <pre>dt_hasil['prediksi_jenis_naive_bayes'] = y_prediksi2  dt_hasil['prediksi_jenis_decision_tree'].replace({0:'A', 1:'B', 2:'C'}, inplace=True)  dt_hasil['prediksi_jenis_naive_bayes'].replace({0:'A', 1:'B', 2:'C'}, inplace=True)</pre>   |
| n [ ]:           | <pre>dt_hasil['prediksi_jenis_naive_bayes'] = y_prediksi2  dt_hasil['prediksi_jenis_decision_tree'].replace({0:'A', 1:'B', 2:'C'}, inplace=True)</pre>  |
|                  | <pre>dt_hasil['prediksi_jenis_naive_bayes'] = y_prediksi2  dt_hasil['prediksi_jenis_decision_tree'].replace({0:'A', 1:'B', 2:'C'}, inplace=True)  dt_hasil['prediksi_jenis_naive_bayes'].replace({0:'A', 1:'B', 2:'C'}, inplace=True)  # run dt_hasil  dt_hasil  no warna bulu makanan_kesukaan prediksi_jenis jenis_sebenarnya prediksi_jenis_decision_tree prediksi_jenis_naive_bayes  0 1 Hitam Gimbal Rumput ? A C A</pre>  |
| n [ ]:<br>ut[ ]: | <pre>dt_hasil['prediksi_jenis_naive_bayes'] = y_prediksi2  dt_hasil['prediksi_jenis_decision_tree'].replace({0:'A', 1:'B', 2:'C'}, inplace=True)  dt_hasil['prediksi_jenis_naive_bayes'].replace({0:'A', 1:'B', 2:'C'}, inplace=True)  # run dt_hasil  dt_hasil  no warna bulu makanan_kesukaan prediksi_jenis_jenis_sebenarnya prediksi_jenis_decision_tree prediksi_jenis_naive_bayes</pre>   |

In [ ]: # import library