

Practical – 6

Shivam Tawari (A-58)

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▼ Skill Development - Practical-6

Aim:- Solving classification problems using Scikitlearn

Name: Shivam Tawari Roll no: A-58

✓ [5] `print("Shivam Tawari A-58")`

0s Shivam Tawari A-58

✓ [6] `import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import os
from sklearn.preprocessing import LabelEncoder
import seaborn as sns
from sklearn.metrics import accuracy_score
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.linear_model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier
from sklearn.svm import SVC
from sklearn.neighbors import KNeighborsClassifier`

0s

▶ `input_path='/content/iris.csv'
df=pd.read_csv(input_path)
df.head()`

0s

🔗

	Id	SepallengthCm	SepalwidthCm	PetalLengthCm	PetalWidthCm	Species
0	1	5.1	3.5	1.4	0.2	Iris-setosa
1	2	4.9	3.0	1.4	0.2	Iris-setosa
2	3	4.7	3.2	1.3	0.2	Iris-setosa
3	4	4.6	3.1	1.5	0.2	Iris-setosa
4	5	5.0	3.6	1.4	0.2	Iris-setosa

✓ [8] `df.drop('Id',axis=1,inplace=True)
df.info()`

0s

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 150 entries, 0 to 149  
Data columns (total 5 columns):  
#   Column          Non-Null Count  Dtype  
---  ---          -  
0   SepallengthCm    150 non-null   float64  
1   SepalwidthCm     150 non-null   float64  
2   PetalLengthCm    150 non-null   float64  
3   PetalWidthCm     150 non-null   float64
```

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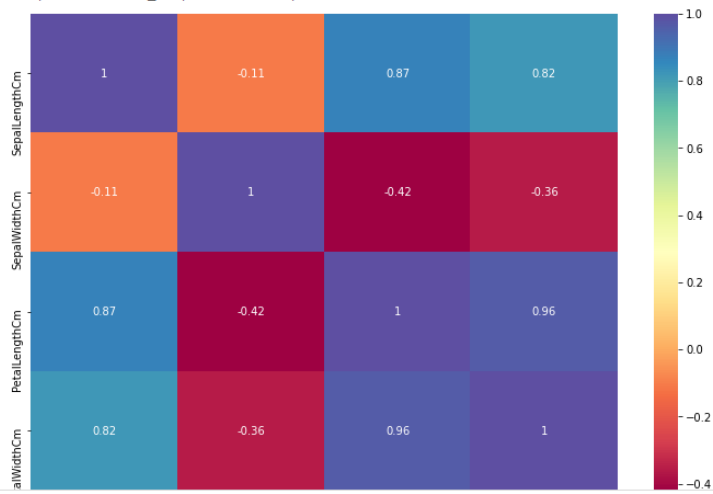
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```
plt.figure(figsize=(12,8))  
sns.heatmap(df.corr(),annot=True,cmap=plt.cm.Spectral)
```

<matplotlib.axes._subplots.AxesSubplot at 0x7f3685708190>

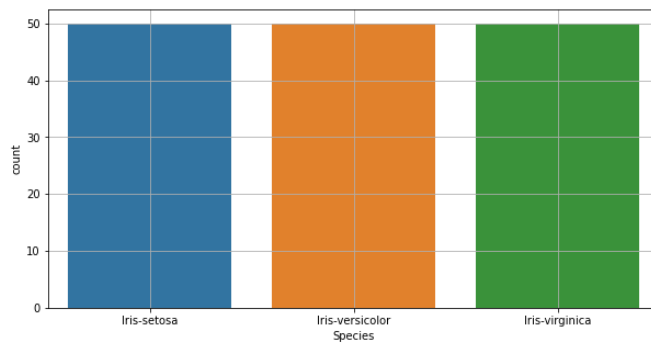


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```
[10] plt.figure(figsize=(10,5))  
sns.countplot(x='Species',data=df)  
plt.grid()
```



```
le=LabelEncoder()  
df['Species_type']=le.fit_transform(df['Species'])
```

```
[12] def scatterplot(X,Y):
```

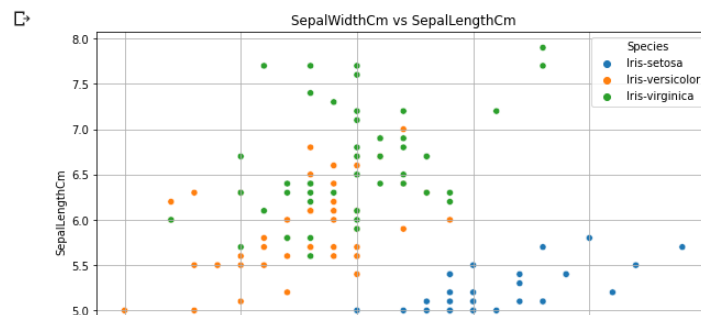
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```
✓ [12] def scatterplot(X,Y):  
0s      sns.scatterplot(data=df,x=X,y=Y,hue='Species')  
      plt.title(X+" vs "+Y)  
      plt.grid()  
      plt.show()
```

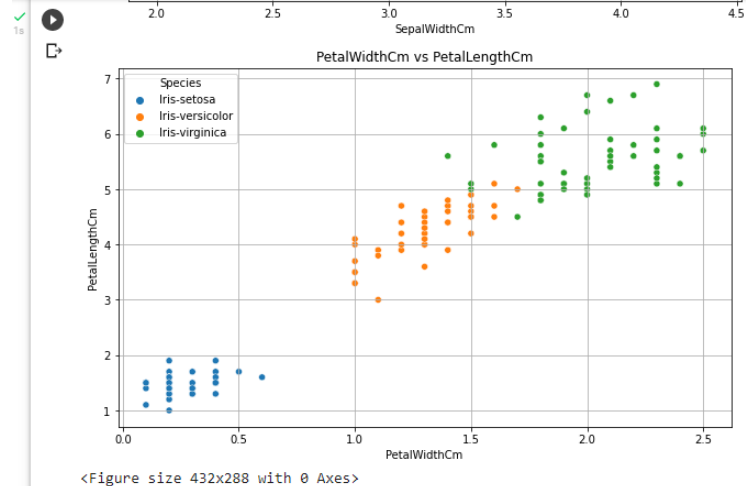
```
✓ 1s ▶ plt.figure(figsize=(10,6))  
      scatterplot('SepalWidthCm','SepalLengthCm')  
      plt.figure(figsize=(10,6))  
      scatterplot('PetalWidthCm','PetalLengthCm')  
      plt.tight_layout()
```



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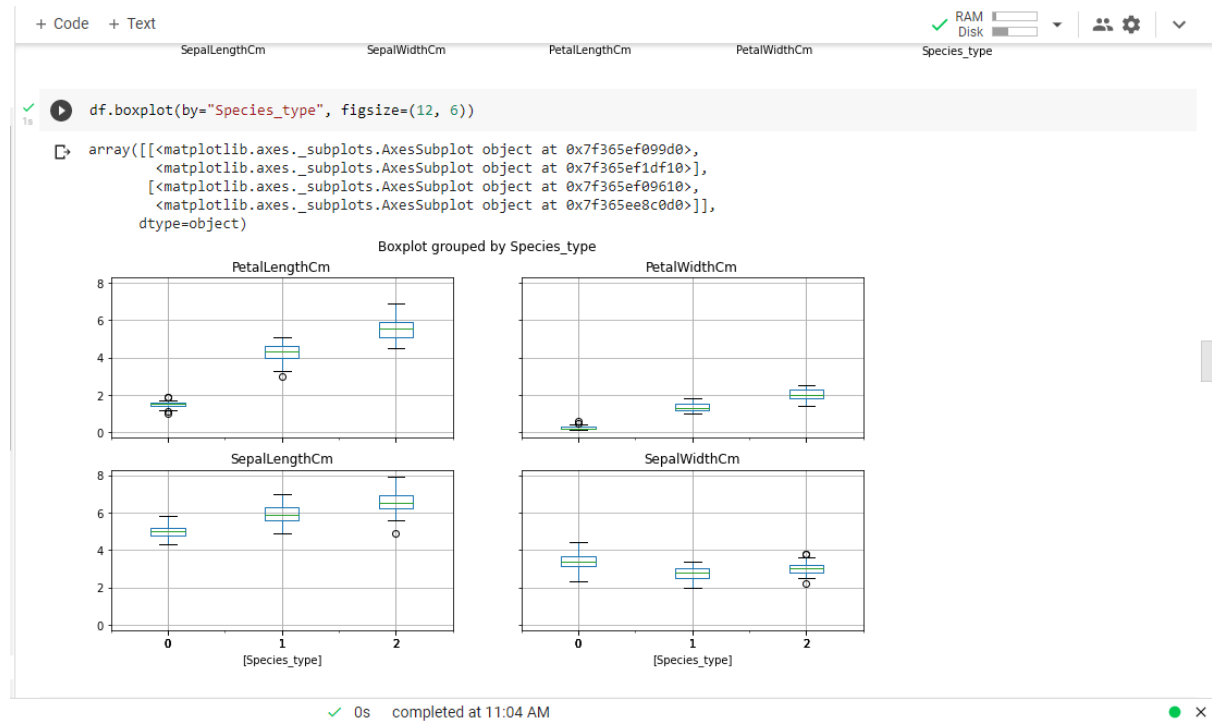
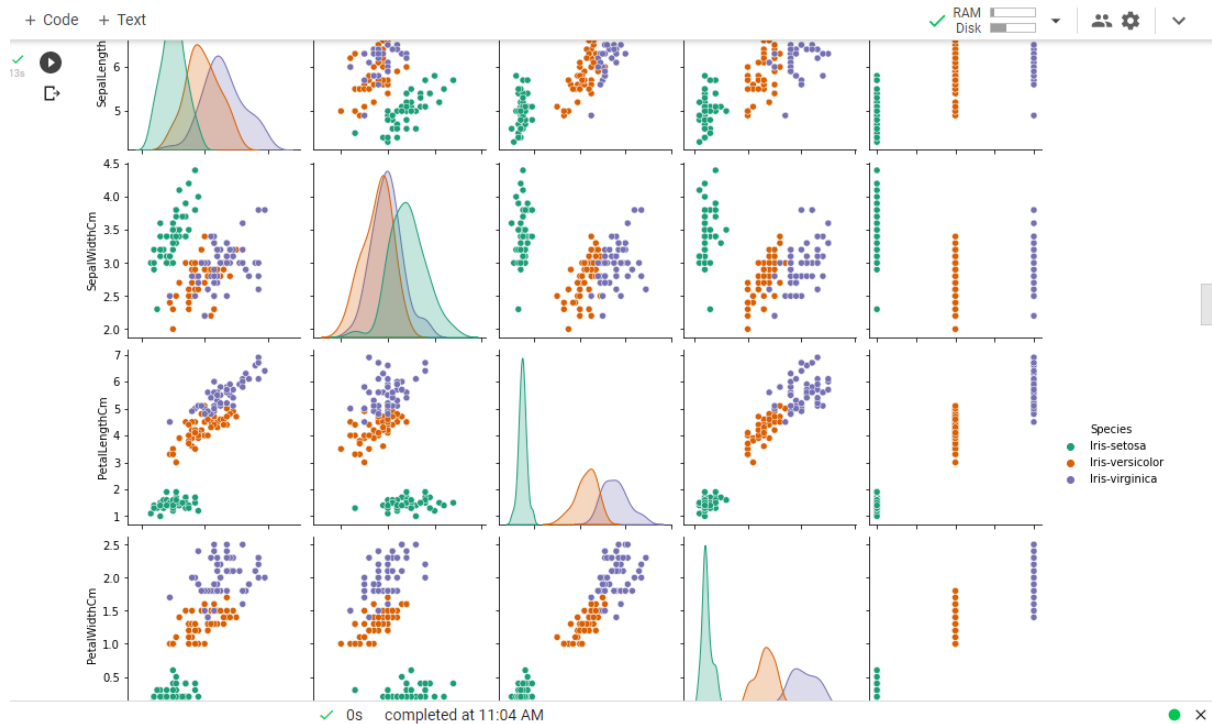


<Figure size 432x288 with 0 Axes>

```
✓ [14] sns.pairplot(df,hue='Species',palette='Dark2')  
13s
```

/usr/local/lib/python3.7/dist-packages/seaborn/distributions.py:306: UserWarning: Dataset has 0 variance; skipping density estimate.

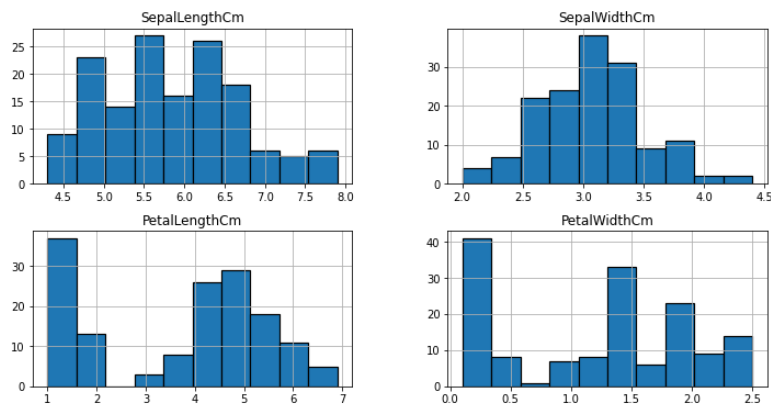
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```
✓ [16] df.drop('Species_type',axis=1).hist(edgecolor='black', linewidth=1.2)
0s fig=plt.gcf()
fig.set_size_inches(12,6)
plt.show()
```



```
✓ [17] X_train,X_test,Y_train,Y_test = train_test_split(df.drop(['Species_type','Species'],axis=1),df['Species_type'],test_size=0.25,rand=0s
```

```
✓ [18] plt.figure(figsize=(8,5))
0s sns.countplot(x=Y_test)
```

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```
✓ [19] models = ['random_forest','svm','decision_tree','logistic_regression','knn']
0s model_train_acc=[]
model_test_acc=[]
```

```
✓ [20] svm= SVC()
0s svm.fit(X_train,Y_train)
y_hat_train = svm.predict(X_train)
y_hat_test = svm.predict(X_test)
train_acc = np.round(accuracy_score(y_hat_train,Y_train),3)
test_acc = np.round(accuracy_score(y_hat_test,Y_test),3)

model_train_acc.append(train_acc)
model_test_acc.append(test_acc)

print("Accuracy Score on train data using SVM: ",train_acc)
print("Accuracy Score on test data using SVM: ",test_acc)
```

Accuracy Score on train data using SVM: 0.955
Accuracy Score on test data using SVM: 0.947

```
✓ [21] rfc = RandomForestClassifier()
0s rfc.fit(X_train,Y_train)
y_hat_train = rfc.predict(X_train)
y_hat_test = rfc.predict(X_test)
train_acc = np.round(accuracy_score(y_hat_train,Y_train),3)
test_acc = np.round(accuracy_score(y_hat_test,Y_test),3)

model_train_acc.append(train_acc)
model_test_acc.append(test_acc)
```

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```
✓ [21] print("Accuracy Score on train data using RandomForest: ",train_acc)
0s print("Accuracy Score on test data using RandomForest: ",test_acc)
```

Accuracy Score on train data using RandomForest: 1.0
Accuracy Score on test data using RandomForest: 0.895

```
✓ [22] tree = DecisionTreeClassifier()
0s tree.fit(X_train,Y_train)
y_hat_train = tree.predict(X_train)
y_hat_test = tree.predict(X_test)
train_acc = np.round(accuracy_score(y_hat_train,Y_train),3)
test_acc = np.round(accuracy_score(y_hat_test,Y_test),3)

model_train_acc.append(train_acc)
model_test_acc.append(test_acc)

print("Accuracy Score on train data using Decision tree: ",train_acc)
print("Accuracy Score on test data using Decision tree: ",test_acc)
```

Accuracy Score on train data using Decision tree: 1.0
Accuracy Score on test data using Decision tree: 0.895

```
✓ [23] lr = LogisticRegression(max_iter=200)
0s lr.fit(X_train,Y_train)
y_hat_train = lr.predict(X_train)
y_hat_test = lr.predict(X_test)
train_acc = np.round(accuracy_score(y_hat_train,Y_train),3)
test_acc = np.round(accuracy_score(y_hat_test,Y_test),3)

model_train_acc.append(train_acc)
```

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```
✓ [23] print("Accuracy Score on train data using Logistic Regression: ",train_acc)
0s print("Accuracy Score on test data using Logistic Regression: ",test_acc)
```

Accuracy Score on train data using Logistic Regression: 0.973
Accuracy Score on test data using Logistic Regression: 0.921

```
✓ [24] def select_neighbors():
0s     knn_train_acc=[]
     knn_test_acc=[]
     for i in range(1,11):
         knn = KNeighborsClassifier(n_neighbors=i)
         knn.fit(X_train,Y_train)
         y_hat_train = knn.predict(X_train)
         y_hat_test = knn.predict(X_test)
         knn_train_acc.append(accuracy_score(y_hat_train,Y_train))
         knn_test_acc.append(accuracy_score(y_hat_test,Y_test))

     return knn_train_acc,knn_test_acc
```

```
✓ [25] knn_train,knn_test = select_neighbors()
0s x = np.linspace(1,10,10)

plt.figure(figsize=(10,4))
plt.plot(x,knn_test,color='red')
plt.title("Neighbors vs Accuracy on Test Data using KNN")
plt.xticks(x)
plt.grid()

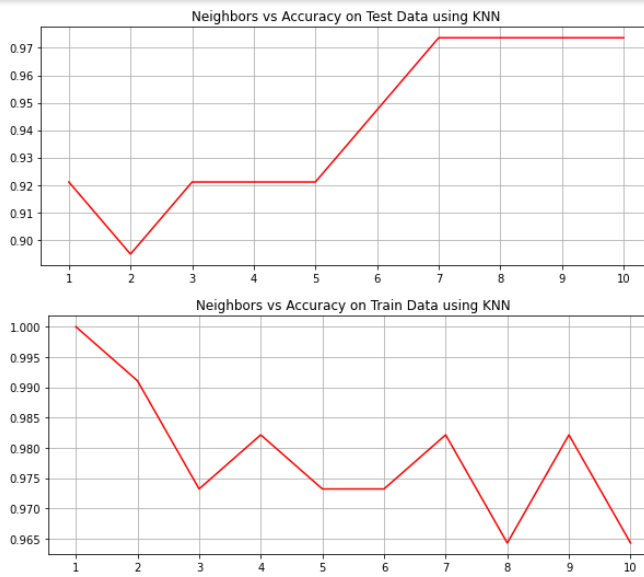
plt.figure(figsize=(10,4))
plt.plot(x,knn_train,color='red')
plt.title("Neighbors vs Accuracy on Train Data using KNN")
```

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✓ [25]
0s



✓ [26] train_acc = np.round(knn_train[6],3)
0s test_acc = np.round(knn_test[6],3)

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✓ [26]
0s

```
train_acc = np.round(knn_train[6],3)
test_acc = np.round(knn_test[6],3)

model_train_acc.append(train_acc)
model_test_acc.append(test_acc)

print("Accuracy Score on train data using KNN: ",train_acc)
print("Accuracy Score on test data using KNN: ",test_acc)
```

Accuracy Score on train data using KNN: 0.982
Accuracy Score on test data using KNN: 0.974

[]

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