

```
In [1]: !pip install scikit_learn
```

```
Requirement already satisfied: scikit_learn in c:\users\anany\anaconda3\lib\site-packages (1.0.2)  
Requirement already satisfied: joblib>=0.11 in c:\users\anany\anaconda3\lib\site-packages (from scikit_learn) (1.1.0)  
Requirement already satisfied: threadpoolctl>=2.0.0 in c:\users\anany\anaconda3\lib\site-packages (from scikit_learn) (2.2.0)  
Requirement already satisfied: numpy>=1.14.6 in c:\users\anany\anaconda3\lib\site-packages (from scikit_learn) (1.21.5)  
Requirement already satisfied: scipy>=1.1.0 in c:\users\anany\anaconda3\lib\site-packages (from scikit_learn) (1.9.1)
```

```
In [2]: from sklearn.datasets import load_iris  
import pandas as pd  
import numpy as np
```

```
In [3]: dataset = load_iris()
```

```
In [7]: #summarise dataset  
print(dataset.data)  
print(dataset.target)
```

```
[[5.1 3.5 1.4 0.2]
[4.9 3. 1.4 0.2]
[4.7 3.2 1.3 0.2]
[4.6 3.1 1.5 0.2]
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[4.8 3.4 1.6 0.2]
[4.8 3. 1.4 0.1]
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[5.6 2.9 3.6 1.3]
[6.7 3.1 4.4 1.4]
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[5.8 2.7 4.1 1.]
[6.2 2.2 4.5 1.5]
[5.6 2.5 3.9 1.1]
[5.9 3.2 4.8 1.8]
[6.1 2.8 4. 1.3]
[6.3 2.5 4.9 1.5]
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[6.4 2.9 4.3 1.3]
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[6.8 2.8 4.8 1.4]
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[6. 2.7 5.1 1.6]
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[5.6 3. 4.1 1.3]
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[5.5 2.6 4.4 1.2]
[6.1 3. 4.6 1.4]
[5.8 2.6 4. 1.2]
[5. 2.3 3.3 1.]
[5.6 2.7 4.2 1.3]
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[6.2 2.9 4.3 1.3]
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[6.3 3.3 6. 2.5]
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[7.1 3. 5.9 2.1]
[6.3 2.9 5.6 1.8]
[6.5 3. 5.8 2.2]
[7.6 3. 6.6 2.1]
[4.9 2.5 4.5 1.7]
[7.3 2.9 6.3 1.8]
[6.7 2.5 5.8 1.8]
[7.2 3.6 6.1 2.5]
[6.5 3.2 5.1 2.]
[6.4 2.7 5.3 1.9]
[6.8 3. 5.5 2.1]
[5.7 2.5 5. 2.]
[5.8 2.8 5.1 2.4]
[6.4 3.2 5.3 2.3]
[6.5 3. 5.5 1.8]
[7.7 3.8 6.7 2.2]
[7.7 2.6 6.9 2.3]
[6. 2.2 5. 1.5]
[6.9 3.2 5.7 2.3]
[5.6 2.8 4.9 2.]
[7.7 2.8 6.7 2.]
[6.3 2.7 4.9 1.8]
[6.7 3.3 5.7 2.1]
[7.2 3.2 6. 1.8]
[6.2 2.8 4.8 1.8]
[6.1 3. 4.9 1.8]

[illegible]

```
In [5]: X=pd.DataFrame(dataset.data, columns=dataset.feature_names)
X
```

Out[5]:	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)
0	5.1	3.5	1.4	0.2
1	4.9	3.0	1.4	0.2
2	4.7	3.2	1.3	0.2
3	4.6	3.1	1.5	0.2
4	5.0	3.6	1.4	0.2
...
145	6.7	3.0	5.2	2.3
146	6.3	2.5	5.0	1.9
147	6.5	3.0	5.2	2.0
148	6.2	3.4	5.4	2.3
149	5.9	3.0	5.1	1.8

150 rows \times 4 columns

```
In [6]: Y=dataset.target
Y
```

```
Out[6]: array([0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
        0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
        0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
        1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
        1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, 2, 2, 2, 2,
        2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2,
        2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2])
```

```
In [10]: #Splitting dataset into train and test
from sklearn.model_selection import train_test_split
X_train, X_test, Y_train, Y_test = train_test_split(X,Y,test_size=0.25,random_state=0)
print(X_train.shape)
print(X_test.shape)
```

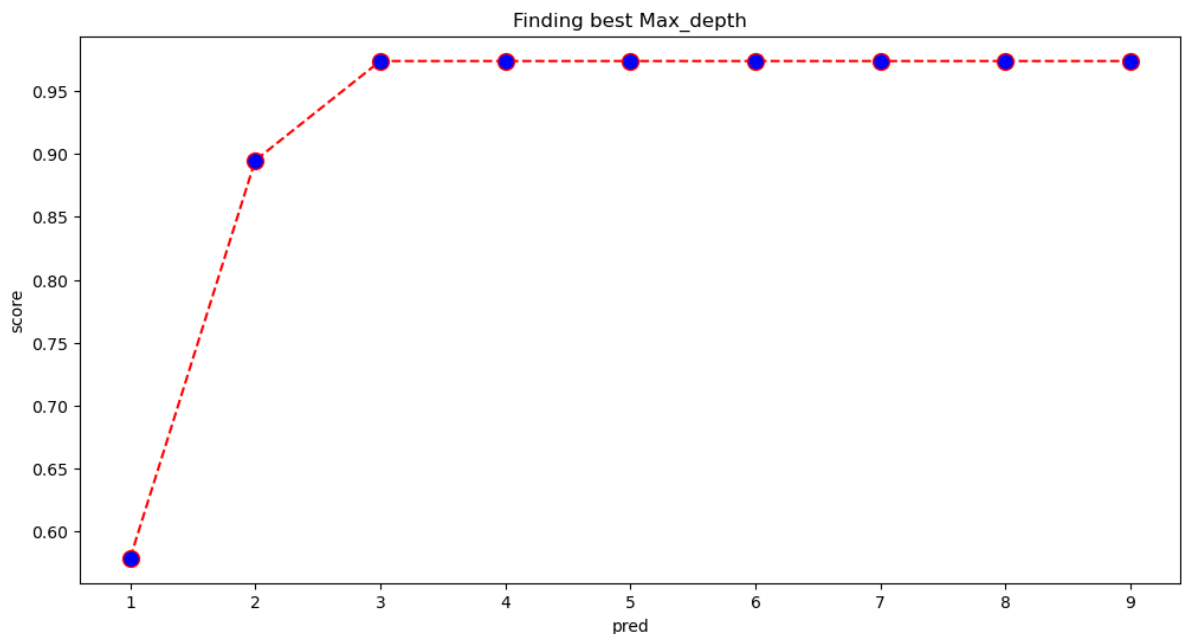
```
(112, 4)
```

```
(38, 4)
```

```
In [17]: #find best max depth value
accuracy=[]
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy_score
import matplotlib.pyplot as plt

for i in range(1,10):
    model=DecisionTreeClassifier(max_depth=i,random_state=0)
    model.fit(X_train,Y_train)
    pred=model.predict(X_test)
    score=accuracy_score(Y_test,pred)
    accuracy.append(score)

plt.figure(figsize=(12,6))
plt.plot(range(1,10),accuracy,color='red',linestyle='dashed',marker='o',markerfacecolor='blue')
plt.title('Finding best Max_depth')
plt.xlabel('pred')
plt.ylabel('score')
plt.show()
```



```
In [18]: #training
from sklearn.tree import DecisionTreeClassifier
model=DecisionTreeClassifier(criterion='entropy',max_depth=3,random_state=0)
model.fit(X_train,Y_train)
```

```
Out[18]: DecisionTreeClassifier(criterion='entropy', max_depth=3, random_state=0)
```

```
In [19]: ##Prediction
Y_pred=model.predict(X_test)
print(np.concatenate((Y_pred.reshape(len(Y_pred),1),Y_test.reshape(len(Y_test),1)),

[[2 2]
 [1 1]
 [0 0]
 [2 2]
 [0 0]
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 [2 1]]
```

```
In [20]: #Accuracy score

from sklearn.metrics import accuracy_score
print("Accuracy of the model: {0}%".format(accuracy_score(Y_test,Y_pred)*100))

Accuracy of the model: 97.36842105263158%
```