```
In [1]: #1)decision tree
        #2)random forest
        #3)logistic regression
        #4)KNN
        #5)SVM
        #assignmemnt-06
        #D.prudhvi sai
In [2]: |import numpy as np
        import pandas as pd
        import matplotlib.pyplot as plt
In [3]: dataset=pd.read_csv('diabetes.csv')
In [4]: dataset.isnull().any()
Out[4]: Pregnancies
                                     False
        Glucose
                                     False
```

False

False

False

False

False

False

False

dtype: bool

In [5]: dataset

BMI

Age

BloodPressure

SkinThickness

DiabetesPedigreeFunction

Insulin

Diabetes

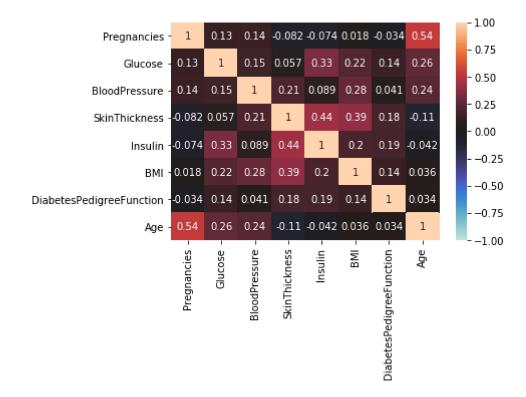
Out[5]:

<u></u>	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction
	6	148	72	35	0	33.6	0.627
,	1 1	85	66	29	0	26.6	0.351
:	2 8	183	64	0	0	23.3	0.672
;	1	89	66	23	94	28.1	0.167
•	0	137	40	35	168	43.1	2.288
•							
76	<b>3</b> 10	101	76	48	180	32.9	0.171
76	4 2	122	70	27	0	36.8	0.340
76	5 5	121	72	23	112	26.2	0.245
76	5 1	126	60	0	0	30.1	0.349
76	7 1	93	70	31	0	30.4	0.315

768 rows × 9 columns

```
In [6]: import seaborn as sns
sns.heatmap(dataset.corr(),annot=True,vmin=-1,vmax=1,center=0)
```

Out[6]: <matplotlib.axes.\_subplots.AxesSubplot at 0x29713d42888>



```
In [7]: from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
```

In [8]: dataset['Diabetes']=le.fit\_transform(dataset['Diabetes'])

In [9]: dataset

Out[9]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction
0	6	148	72	35	0	33.6	0.627
1	1	85	66	29	0	26.6	0.351
2	8	183	64	0	0	23.3	0.672
3	1	89	66	23	94	28.1	0.167
4	0	137	40	35	168	43.1	2.288
						•••	
763	10	101	76	48	180	32.9	0.171
764	2	122	70	27	0	36.8	0.340
765	5	121	72	23	112	26.2	0.245
766	1	126	60	0	0	30.1	0.349
767	1	93	70	31	0	30.4	0.315

768 rows × 9 columns

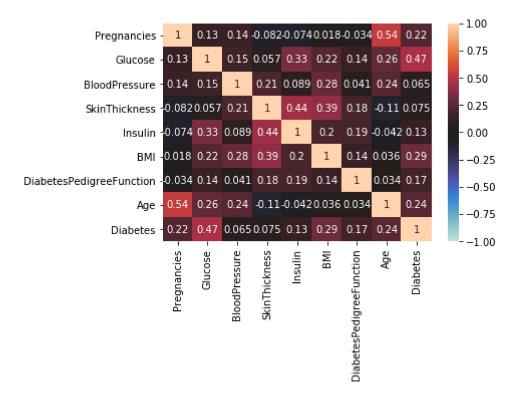
In [10]: dataset.corr()

Out[10]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВІ
Pregnancies	1.000000	0.129459	0.141282	-0.081672	-0.073535	0.01768
Glucose	0.129459	1.000000	0.152590	0.057328	0.331357	0.22107
BloodPressure	0.141282	0.152590	1.000000	0.207371	0.088933	0.28180
SkinThickness	<b>-</b> 0.081672	0.057328	0.207371	1.000000	0.436783	0.39257
Insulin	-0.073535	0.331357	0.088933	0.436783	1.000000	0.1978
ВМІ	0.017683	0.221071	0.281805	0.392573	0.197859	1.00000
DiabetesPedigreeFunction	-0.033523	0.137337	0.041265	0.183928	0.185071	0.14064
Age	0.544341	0.263514	0.239528	-0.113970	-0.042163	0.03624
Diabetes	0.221898	0.466581	0.065068	0.074752	0.130548	0.29269
4						•

```
In [11]: import seaborn as sns
    sns.heatmap(dataset.corr(),annot=True,vmin=-1,vmax=1,center=0)
```

Out[11]: <matplotlib.axes.\_subplots.AxesSubplot at 0x297146cb1c8>



In [12]: x=dataset.iloc[:,0:8]

In [13]: x.head()

Out[13]:

Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction A

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction /
0	6	148	72	35	0	33.6	0.627
1	1	85	66	29	0	26.6	0.351
2	8	183	64	0	0	23.3	0.672
3	1	89	66	23	94	28.1	0.167
4	0	137	40	35	168	43.1	2.288
4							<b>•</b>

In [14]: y=dataset.iloc[:,8]

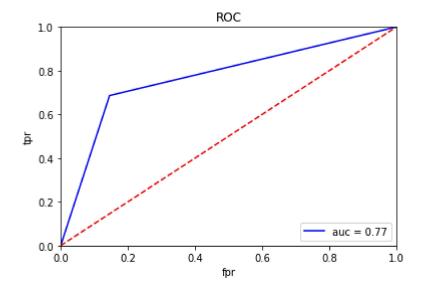
```
In [15]: |y.head()
 Out[15]: 0
                1
           1
                0
                1
                0
           3
                1
           Name: Diabetes, dtype: int32
 In [16]: x=dataset.iloc[:,0:8].values
           y=dataset.iloc[:,-1:].values
 In [17]: x
 Out[17]: array([[
                     6.
                           , 148.
                                        72.
                                                       33.6
                                                                  0.627,
                                                                          50.
                                                                                 ],
                              85.
                                                                  0.351,
                                                                          31.
                     1.
                                        66.
                                                       26.6
                                                                                 ],
                     8.
                           , 183.
                                        64.
                                                       23.3
                                                                  0.672,
                                                                          32.
                                                                                 ],
                           , 121.
                                                       26.2,
                      5.
                                        72.
                                                                  0.245,
                                                                          30.
                                                                                 ],
                                                       30.1
                                                                  0.349,
                                                                          47.
                     1.
                           , 126.
                                        60.
                                                                                 ],
                                                                  0.315,
                                                                          23.
                     1.
                              93.
                                        70.
                                                       30.4
                                                                                 ]])
 In [18]: y
                   [1],
                  [1],
                   [0],
                   [0],
                   [0],
                   [0],
                   [0],
                   [0],
                   [1],
                   [1],
                   [0],
                   [0],
                   [1],
                   [0],
                   [0],
                   [1],
                   [0],
                   [1],
                   [1],
                   [1],
 In [19]: from sklearn.model_selection import train_test_split
In [106]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2, random_state=6)
In [107]: x_train.shape
Out[107]: (614, 8)
```

```
In [108]: y_train.shape
Out[108]: (614, 1)
In [109]: from sklearn.preprocessing import StandardScaler
         sc=StandardScaler()
         x_train = sc.fit_transform(x_train)
         x_test = sc.fit_transform(x_test)
In [128]: #decision tree
         from sklearn.tree import DecisionTreeClassifier
         dtc=DecisionTreeClassifier(criterion='entropy')
         dtc.fit(x train,y train)
Out[128]: DecisionTreeClassifier(ccp_alpha=0.0, class_weight=None, criterion='entropy',
                              max depth=None, max features=None, max leaf nodes=None,
                              min impurity decrease=0.0, min impurity split=None,
                              min_samples_leaf=1, min_samples_split=2,
                              min weight fraction leaf=0.0, presort='deprecated',
                               random state=None, splitter='best')
In [129]: |y_pred = dtc.predict(x_test)
         y pred
0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0,
                0, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 1, 0,
                0, 1, 0, 0, 0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0,
                0, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 1, 1,
                0, 0, 1, 0, 1, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0,
                0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 0, 1])
```

```
In [139]: import sklearn.metrics as metrics
fpr,tpr,threshold = metrics.roc_curve(y_test,y_pred)
roc_auc = metrics.auc(fpr,tpr)

import matplotlib.pyplot as plt
plt.title("ROC")
plt.plot(fpr,tpr,'b',label = 'auc = %0.2f'%roc_auc)
plt.legend(loc='lower right')
plt.plot([0,1],[0,1],'r--')
plt.xlim([0,1])
plt.ylim([0,1])
plt.ylabel('fpr')
plt.ylabel('tpr')
```

## Out[139]: Text(0, 0.5, 'tpr')



```
In [130]: from sklearn.metrics import accuracy_score
accuracy_score(y_pred,y_test)
```

Out[130]: 0.7987012987012987

```
In [131]: from sklearn.metrics import confusion_matrix
```

```
In [132]: cm=confusion_matrix(y_test,y_pred)
```

```
In [133]: cm
```

```
Out[133]: array([[88, 15], [16, 35]], dtype=int64)
```

```
In [134]: #random forest tree
    from sklearn.ensemble import RandomForestClassifier
    rfc = RandomForestClassifier(n_estimators=100,criterion='entropy',max_depth=10,ma
```

```
In [135]: rfc.fit(x_train,y_train)
```

C:\Users\PRUDHVI\anaconda3\lib\site-packages\ipykernel\_launcher.py:1: DataConve rsionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n\_samples,), for example using ravel().

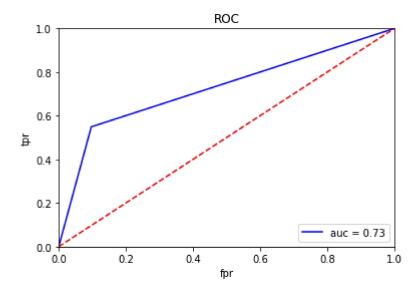
"""Entry point for launching an IPython kernel.

```
In [136]: y_pred_rfc = rfc.predict(x_test)
```

```
In [140]: import sklearn.metrics as metrics
fpr,tpr,threshold = metrics.roc_curve(y_test,y_pred_rfc)
roc_auc = metrics.auc(fpr,tpr)

import matplotlib.pyplot as plt
plt.title("ROC")
plt.plot(fpr,tpr,'b',label = 'auc = %0.2f'%roc_auc)
plt.legend(loc='lower right')
plt.plot([0,1],[0,1],'r--')
plt.xlim([0,1])
plt.ylim([0,1])
plt.xlabel('fpr')
plt.ylabel('tpr')
```

## Out[140]: Text(0, 0.5, 'tpr')

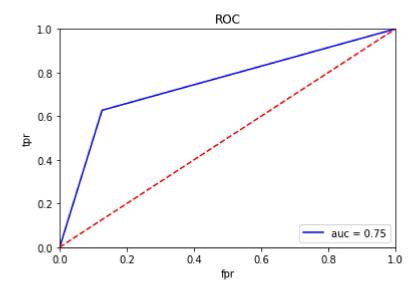


```
In [137]: | accuracy score(y test,y pred rfc)
Out[137]: 0.7857142857142857
In [141]: # Logisitic regression
          from sklearn.linear_model import LogisticRegression
          log = LogisticRegression()
          log.fit(x_train,y_train)
          y_pred_log = log.predict(x_test)
          y pred log
          C:\Users\PRUDHVI\anaconda3\lib\site-packages\sklearn\utils\validation.py:760: D
          ataConversionWarning: A column-vector y was passed when a 1d array was expecte
          d. Please change the shape of y to (n_samples, ), for example using ravel().
            y = column_or_1d(y, warn=True)
Out[141]: array([0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0,
                 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0,
                 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 1, 1, 0, 0, 0,
                 1, 1, 0, 0, 0, 1, 1, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 1,
                 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1,
                 0, 0, 1, 0, 1, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0,
                 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1])
In [142]: | accuracy score(y test,y pred log)
Out[142]: 0.7922077922077922
In [143]: | confusion matrix(y test,y pred log)
Out[143]: array([[90, 13],
                 [19, 32]], dtype=int64)
```

```
In [144]: import sklearn.metrics as metrics
fpr,tpr,threshold = metrics.roc_curve(y_test,y_pred_log)
roc_auc = metrics.auc(fpr,tpr)

import matplotlib.pyplot as plt
plt.title("ROC")
plt.plot(fpr,tpr,'b',label = 'auc = %0.2f'%roc_auc)
plt.legend(loc='lower right')
plt.plot([0,1],[0,1],'r--')
plt.xlim([0,1])
plt.ylim([0,1])
plt.ylabel('fpr')
```

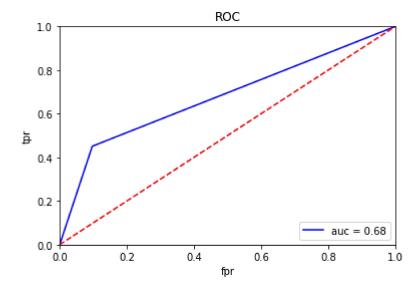
## Out[144]: Text(0, 0.5, 'tpr')



C:\Users\PRUDHVI\anaconda3\lib\site-packages\ipykernel\_launcher.py:4: DataConve rsionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n\_samples, ), for example using ravel(). after removing the cwd from sys.path.

```
In [154]: | accuracy_score(y_test,y_pred_knn)
Out[154]: 0.7532467532467533
In [155]:
          confusion_matrix(y_test,y_pred_knn)
Out[155]: array([[93, 10],
                  [28, 23]], dtype=int64)
In [156]:
          import sklearn.metrics as metrics
          fpr,tpr,threshold = metrics.roc_curve(y_test,y_pred_knn)
          roc_auc = metrics.auc(fpr,tpr)
          import matplotlib.pyplot as plt
          plt.title("ROC")
          plt.plot(fpr,tpr,'b',label = 'auc = %0.2f'%roc_auc)
          plt.legend(loc='lower right')
          plt.plot([0,1],[0,1],'r--')
          plt.xlim([0,1])
          plt.ylim([0,1])
          plt.xlabel('fpr')
          plt.ylabel('tpr')
```

## Out[156]: Text(0, 0.5, 'tpr')



```
In [157]: # SVM
         from sklearn.svm import SVC
         svm = SVC(kernel = 'poly', degree = 5)
         svm.fit(x train,y train)
         y pred_svm = svm.predict(x_test)
         y_pred_svm
         C:\Users\PRUDHVI\anaconda3\lib\site-packages\sklearn\utils\validation.py:760: D
         ataConversionWarning: A column-vector y was passed when a 1d array was expecte
         d. Please change the shape of y to (n_samples, ), for example using ravel().
           y = column_or_1d(y, warn=True)
Out[157]: array([0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 0,
                0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0,
                0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0,
                0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0,
                1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1,
                In [158]: | accuracy_score(y_test,y_pred_svm)
Out[158]: 0.7402597402597403
In [159]: |confusion_matrix(y_test,y_pred_svm)
Out[159]: array([[96, 7],
                [33, 18]], dtype=int64)
In [162]: result = [['Model', 'Accuracy'],
                  ['Decision tree', 0.7987012987012987],
                  ['Random Forest Classifier', 0.7857142857142857],
                  ['Logistic Regression', 0.7922077922077922],
                  ['K Nearest Neighbors', 0.7532467532467533],
                  ['Support Vector Machine', 0.7402597402597403]]
         result
Out[162]: [['Model', 'Accuracy'],
          ['Decision tree', 0.7987012987012987],
          ['Random Forest Classifier', 0.7857142857142857],
          ['Logistic Regression', 0.7922077922077922],
          ['K Nearest Neighbors', 0.7532467532467533],
          ['Support Vector Machine', 0.7402597402597403]]
 In [ ]:
```