

knn-star-classify

[4]: *#importing libraries and creating dataframe of the dataset.*

```
import numpy as np
import pandas as pd
df=pd.read_csv('/content/star_classification.csv')
df
```

```
[4]:
```

	obj_ID	alpha	delta	u	g	r \
0	1.237661e+18	135.689107	32.494632	23.87882	22.27530	20.39501
1	1.237665e+18	144.826101	31.274185	24.77759	22.83188	22.58444
2	1.237661e+18	142.188790	35.582444	25.26307	22.66389	20.60976
3	1.237663e+18	338.741038	-0.402828	22.13682	23.77656	21.61162
4	1.237680e+18	345.282593	21.183866	19.43718	17.58028	16.49747
...
99995	1.237679e+18	39.620709	-2.594074	22.16759	22.97586	21.90404
99996	1.237679e+18	29.493819	19.798874	22.69118	22.38628	20.45003
99997	1.237668e+18	224.587407	15.700707	21.16916	19.26997	18.20428
99998	1.237661e+18	212.268621	46.660365	25.35039	21.63757	19.91386
99999	1.237661e+18	196.896053	49.464643	22.62171	21.79745	20.60115

	i	z	run_ID	rerun_ID	cam_col	field_ID	spec_obj_ID \
0	19.16573	18.79371	3606	301	2	79	6.543777e+18
1	21.16812	21.61427	4518	301	5	119	1.176014e+19
2	19.34857	18.94827	3606	301	2	120	5.152200e+18
3	20.50454	19.25010	4192	301	3	214	1.030107e+19
4	15.97711	15.54461	8102	301	3	137	6.891865e+18
...
99995	21.30548	20.73569	7778	301	2	581	1.055431e+19
99996	19.75759	19.41526	7917	301	1	289	8.586351e+18
99997	17.69034	17.35221	5314	301	4	308	3.112008e+18
99998	19.07254	18.62482	3650	301	4	131	7.601080e+18
99999	20.00959	19.28075	3650	301	4	60	8.343152e+18

	class	redshift	plate	MJD	fiber_ID
0	GALAXY	0.634794	5812	56354	171
1	GALAXY	0.779136	10445	58158	427
2	GALAXY	0.644195	4576	55592	299

3	GALAXY	0.932346	9149	58039	775
4	GALAXY	0.116123	6121	56187	842
...
99995	GALAXY	0.000000	9374	57749	438
99996	GALAXY	0.404895	7626	56934	866
99997	GALAXY	0.143366	2764	54535	74
99998	GALAXY	0.455040	6751	56368	470
99999	GALAXY	0.542944	7410	57104	851

[100000 rows x 18 columns]

```
[5]: df.head()
```

```
[5]:
```

	obj_ID	alpha	delta	u	g	r	\
0	1.237661e+18	135.689107	32.494632	23.87882	22.27530	20.39501	
1	1.237665e+18	144.826101	31.274185	24.77759	22.83188	22.58444	
2	1.237661e+18	142.188790	35.582444	25.26307	22.66389	20.60976	
3	1.237663e+18	338.741038	-0.402828	22.13682	23.77656	21.61162	
4	1.237680e+18	345.282593	21.183866	19.43718	17.58028	16.49747	

	i	z	run_ID	rerun_ID	cam_col	field_ID	spec_obj_ID	\
0	19.16573	18.79371	3606	301	2	79	6.543777e+18	
1	21.16812	21.61427	4518	301	5	119	1.176014e+19	
2	19.34857	18.94827	3606	301	2	120	5.152200e+18	
3	20.50454	19.25010	4192	301	3	214	1.030107e+19	
4	15.97711	15.54461	8102	301	3	137	6.891865e+18	

	class	redshift	plate	MJD	fiber_ID
0	GALAXY	0.634794	5812	56354	171
1	GALAXY	0.779136	10445	58158	427
2	GALAXY	0.644195	4576	55592	299
3	GALAXY	0.932346	9149	58039	775
4	GALAXY	0.116123	6121	56187	842

```
[6]: df.tail()
```

```
[6]:
```

	obj_ID	alpha	delta	u	g	r	\
99995	1.237679e+18	39.620709	-2.594074	22.16759	22.97586	21.90404	
99996	1.237679e+18	29.493819	19.798874	22.69118	22.38628	20.45003	
99997	1.237668e+18	224.587407	15.700707	21.16916	19.26997	18.20428	
99998	1.237661e+18	212.268621	46.660365	25.35039	21.63757	19.91386	
99999	1.237661e+18	196.896053	49.464643	22.62171	21.79745	20.60115	

	i	z	run_ID	rerun_ID	cam_col	field_ID	spec_obj_ID	\
99995	21.30548	20.73569	7778	301	2	581	1.055431e+19	
99996	19.75759	19.41526	7917	301	1	289	8.586351e+18	
99997	17.69034	17.35221	5314	301	4	308	3.112008e+18	

99998	19.07254	18.62482	3650	301	4	131	7.601080e+18
99999	20.00959	19.28075	3650	301	4	60	8.343152e+18

	class	redshift	plate	MJD	fiber_ID
99995	GALAXY	0.000000	9374	57749	438
99996	GALAXY	0.404895	7626	56934	866
99997	GALAXY	0.143366	2764	54535	74
99998	GALAXY	0.455040	6751	56368	470
99999	GALAXY	0.542944	7410	57104	851

```
[7]: df.shape
```

```
[7]: (100000, 18)
```

```
[8]: df.columns
```

```
[8]: Index(['obj_ID', 'alpha', 'delta', 'u', 'g', 'r', 'i', 'z', 'run_ID',
          'rerun_ID', 'cam_col', 'field_ID', 'spec_obj_ID', 'class', 'redshift',
          'plate', 'MJD', 'fiber_ID'],
          dtype='object')
```

```
[9]: #checking for missing values
```

```
df.isna().sum()
```

```
[9]: obj_ID      0
     alpha      0
     delta      0
     u          0
     g          0
     r          0
     i          0
     z          0
     run_ID     0
     rerun_ID   0
     cam_col    0
     field_ID   0
     spec_obj_ID 0
     class      0
     redshift    0
     plate      0
     MJD        0
     fiber_ID    0
     dtype: int64
```

```
[10]: df.dtypes
```

```
[10]: obj_ID      float64
      alpha      float64
      delta      float64
      u          float64
      g          float64
      r          float64
      i          float64
      z          float64
      run_ID     int64
      rerun_ID   int64
      cam_col    int64
      field_ID   int64
      spec_obj_ID float64
      class      object
      redshift   float64
      plate      int64
      MJD        int64
      fiber_ID   int64
      dtype: object
```

```
[11]: df.drop(['obj_ID'],axis=1,inplace=True)
      df
```

```
[11]:
```

	alpha	delta	u	g	r	i	\
0	135.689107	32.494632	23.87882	22.27530	20.39501	19.16573	
1	144.826101	31.274185	24.77759	22.83188	22.58444	21.16812	
2	142.188790	35.582444	25.26307	22.66389	20.60976	19.34857	
3	338.741038	-0.402828	22.13682	23.77656	21.61162	20.50454	
4	345.282593	21.183866	19.43718	17.58028	16.49747	15.97711	
...	
99995	39.620709	-2.594074	22.16759	22.97586	21.90404	21.30548	
99996	29.493819	19.798874	22.69118	22.38628	20.45003	19.75759	
99997	224.587407	15.700707	21.16916	19.26997	18.20428	17.69034	
99998	212.268621	46.660365	25.35039	21.63757	19.91386	19.07254	
99999	196.896053	49.464643	22.62171	21.79745	20.60115	20.00959	

	z	run_ID	rerun_ID	cam_col	field_ID	spec_obj_ID	class	\
0	18.79371	3606	301	2	79	6.543777e+18	GALAXY	
1	21.61427	4518	301	5	119	1.176014e+19	GALAXY	
2	18.94827	3606	301	2	120	5.152200e+18	GALAXY	
3	19.25010	4192	301	3	214	1.030107e+19	GALAXY	
4	15.54461	8102	301	3	137	6.891865e+18	GALAXY	
...		
99995	20.73569	7778	301	2	581	1.055431e+19	GALAXY	
99996	19.41526	7917	301	1	289	8.586351e+18	GALAXY	
99997	17.35221	5314	301	4	308	3.112008e+18	GALAXY	
99998	18.62482	3650	301	4	131	7.601080e+18	GALAXY	

```
99999  19.28075    3650        301         4        60  8.343152e+18  GALAXY
```

```

      redshift  plate    MJD  fiber_ID
0    0.634794   5812  56354      171
1    0.779136  10445  58158      427
2    0.644195   4576  55592      299
3    0.932346   9149  58039      775
4    0.116123   6121  56187      842
...
99995  0.000000   9374  57749      438
99996  0.404895   7626  56934      866
99997  0.143366   2764  54535       74
99998  0.455040   6751  56368      470
99999  0.542944   7410  57104      851
```

```
[100000 rows x 17 columns]
```

```
[12]: #Splitting the dataframe into input and output features
```

```
x=df.drop(['class'],axis=1).values
x
```

```
[12]: array([[1.35689107e+02, 3.24946318e+01, 2.38788200e+01, ...,
           5.81200000e+03, 5.63540000e+04, 1.71000000e+02],
          [1.44826101e+02, 3.12741849e+01, 2.47775900e+01, ...,
           1.04450000e+04, 5.81580000e+04, 4.27000000e+02],
          [1.42188790e+02, 3.55824442e+01, 2.52630700e+01, ...,
           4.57600000e+03, 5.55920000e+04, 2.99000000e+02],
          ...,
          [2.24587407e+02, 1.57007074e+01, 2.11691600e+01, ...,
           2.76400000e+03, 5.45350000e+04, 7.40000000e+01],
          [2.12268621e+02, 4.66603653e+01, 2.53503900e+01, ...,
           6.75100000e+03, 5.63680000e+04, 4.70000000e+02],
          [1.96896053e+02, 4.94646428e+01, 2.26217100e+01, ...,
           7.41000000e+03, 5.71040000e+04, 8.51000000e+02]])
```

```
[13]: y=df['class'].values
y
```

```
[13]: array(['GALAXY', 'GALAXY', 'GALAXY', ..., 'GALAXY', 'GALAXY', 'GALAXY'],
          dtype=object)
```

```
[14]: #Splitting the features into training and testing datas.
```

```

from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.
↪30,random_state=42)
```

```
x_train
```

```
[14]: array([[1.30932167e+02, 4.31341083e+00, 2.01307000e+01, ...,  
          1.18800000e+03, 5.26500000e+04, 1.65000000e+02],  
          [2.25429599e+02, 3.31720833e+01, 1.98631500e+01, ...,  
          2.93500000e+03, 5.46520000e+04, 3.74000000e+02],  
          [2.19173525e+02, 5.55117400e+01, 1.98478500e+01, ...,  
          3.29600000e+03, 5.49090000e+04, 4.75000000e+02],  
          ...,  
          [1.56991726e+02, 3.86124564e+01, 2.39568400e+01, ...,  
          3.26200000e+03, 5.48840000e+04, 5.41000000e+02],  
          [5.58294316e+01, 9.76439658e+00, 1.77922400e+01, ...,  
          2.67900000e+03, 5.43680000e+04, 2.87000000e+02],  
          [1.89902619e+02, 3.37795907e+01, 2.49314200e+01, ...,  
          3.97100000e+03, 5.53220000e+04, 4.50000000e+02]])
```

```
[15]: x_test
```

```
[15]: array([[1.69568898e+01, 3.64613009e+00, 2.33354200e+01, ...,  
          4.31200000e+03, 5.55110000e+04, 4.95000000e+02],  
          [2.40063240e+02, 6.13413060e+00, 1.78603300e+01, ...,  
          2.17500000e+03, 5.46120000e+04, 3.48000000e+02],  
          [3.08872221e+01, 1.18870964e+00, 1.81891100e+01, ...,  
          7.33200000e+03, 5.66830000e+04, 9.43000000e+02],  
          ...,  
          [2.09415904e+02, 4.98478278e+01, 2.29654700e+01, ...,  
          7.43200000e+03, 5.71070000e+04, 8.67000000e+02],  
          [2.26833308e+02, 2.61099640e+01, 1.94337400e+01, ...,  
          2.15400000e+03, 5.45390000e+04, 2.54000000e+02],  
          [4.69558448e+01, 8.55015509e-01, 2.26332100e+01, ...,  
          1.06600000e+03, 5.25890000e+04, 3.78000000e+02]])
```

```
[16]: #Normalization by Standard Scaler
```

```
from sklearn.preprocessing import StandardScaler  
scaler=StandardScaler()  
scaler.fit(x_train)  
x_train=scaler.transform(x_train)  
x_test=scaler.transform(x_test)  
x_train
```

```
[16]: array([[ -0.48390605, -1.00945814, -0.86643846, ..., -1.33420844,  
          -1.62135443, -1.04171225],  
          [ 0.49756661,  0.45809196, -0.98519609, ..., -0.74322585,  
          -0.51517188, -0.27483231],  
          [ 0.43258954,  1.59413054, -0.99198732, ..., -0.62110523,  
          -0.37316943,  0.09576517],
```

```
...,
[-0.21324531,  0.73475126,  0.83187338, ..., -0.6326069 ,
 -0.3869829 ,  0.33793778],
[-1.26394079, -0.73225915, -1.90441259, ..., -0.82982662,
 -0.67209288, -0.59405984],
[ 0.128575 ,  0.48898554,  1.26446099, ..., -0.39276336,
 -0.14497093,  0.00403312]])
```

```
[17]: x_test
```

```
[17]: array([[ -1.6676802 , -1.04339136,  0.55604319, ..., -0.27740843,
 -0.04054111,  0.16915081],
 [ 0.64955506, -0.91686907, -1.87418942, ..., -1.00032188,
 -0.53727343, -0.37023364],
 [-1.52299647, -1.16835857, -1.72825358, ...,  0.74421002,
  0.60703429,  1.81298913],
 ...,
 [ 0.33124459,  1.30610361,  0.39183318, ...,  0.77803845,
  0.84131071,  1.5341237 ],
 [ 0.51214587,  0.098962 , -1.17579866, ..., -1.00742585,
 -0.57760876, -0.71514615],
 [-1.35610396, -1.18532792,  0.24435267, ..., -1.37547912,
 -1.65505929, -0.26015518]])
```

```
[18]: #Model creation
```

```
from sklearn.neighbors import KNeighborsClassifier
knn=KNeighborsClassifier(n_neighbors=7)
knn.fit(x_train,y_train)
y_pred=knn.predict(x_test)
y_pred
```

```
[18]: array(['GALAXY', 'STAR', 'STAR', ..., 'STAR', 'QSO', 'STAR'], dtype=object)
```

```
[19]: y_test
```

```
[19]: array(['GALAXY', 'STAR', 'STAR', ..., 'STAR', 'QSO', 'STAR'], dtype=object)
```

```
[20]: y_train
```

```
[20]: array(['GALAXY', 'STAR', 'STAR', ..., 'STAR', 'GALAXY', 'GALAXY'],
 dtype=object)
```

```
[21]: #Performance Evaluation
```

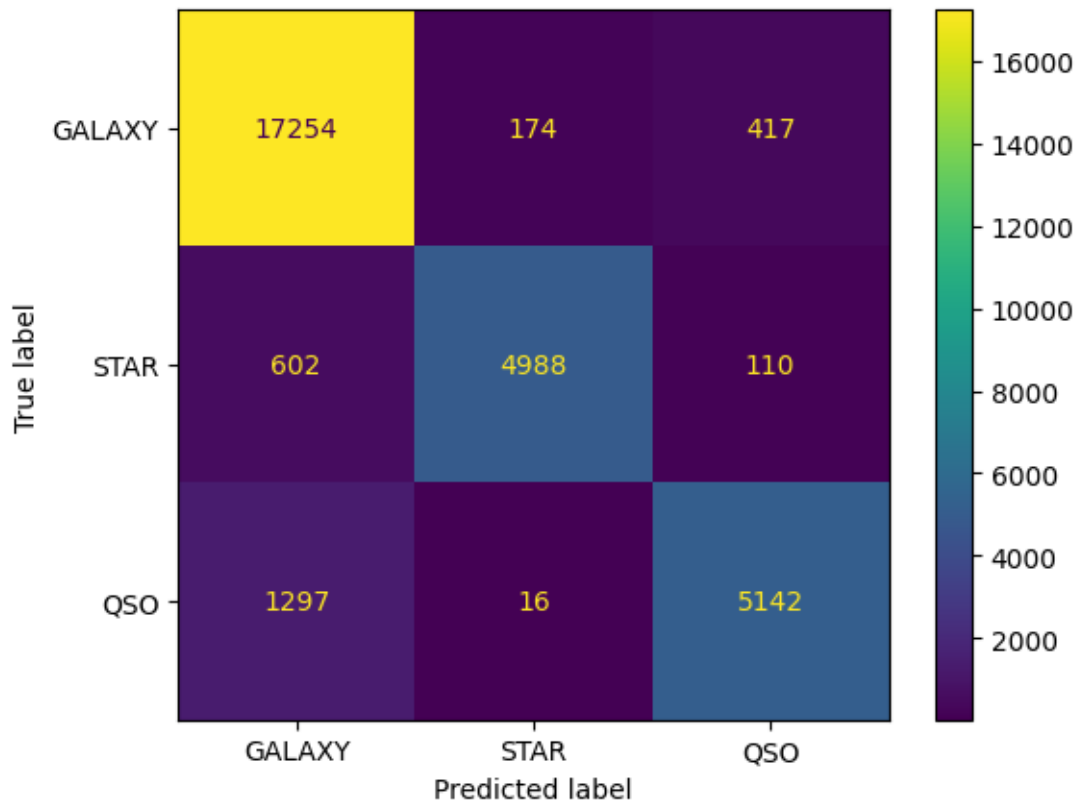
```
from sklearn.metrics import confusion_matrix
cm=confusion_matrix(y_test,y_pred)
```

```
cm
```

```
[21]: array([[17254,   174,   417],
           [  602, 4988,   110],
           [1297,    16, 5142]])
```

```
[22]: from sklearn.metrics import ConfusionMatrixDisplay
labels=['GALAXY','STAR','QSO']
cmd=ConfusionMatrixDisplay(cm,display_labels=labels)
cmd.plot()
```

```
[22]: <sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at
0x7b365f8da0b0>
```



```
[23]: from sklearn.metrics import accuracy_score
score=accuracy_score(y_test,y_pred)
score
```

```
[23]: 0.9128
```



```
[24]: from sklearn.metrics import classification_report
report=classification_report(y_test,y_pred)
print(report)
```

	precision	recall	f1-score	support
GALAXY	0.90	0.97	0.93	17845
QSO	0.96	0.88	0.92	5700
STAR	0.91	0.80	0.85	6455
accuracy			0.91	30000
macro avg	0.92	0.88	0.90	30000
weighted avg	0.91	0.91	0.91	30000