PROGRAM NO. – 1

Write a program in python to read the contents of .csv file using panda module.

<u>OUTPUT</u> –

	Duration	Pulse	Maxpulse	Calories
0	60	110	130	409.1
1	60	117	145	479.0
2	60	103	135	340.0
3	45	109	175	282.4
4	45	117	148	406.0
5	60	102	127	300.0

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Index(['Duration', 'Pulse', 'Maxpulse', 'Calories'], dtype='object')

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≺bound	method	DataFra	me.info of	Duration	Pulse	Maxpulse	Calories
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1	60	117
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3	45	109
4	45	117
5	60	102

	Duration	Pulse	Maxpulse	Calories
5	60	102	127	300.0

PROGRAM NO. – 1

Write a program in python to read the contents of .csv file using panda module.

```
#prints the dataframe
import pandas as pd
import numpy as np
df = pd.read_csv('data.csv')
#returns first 3 rows of dataframe
df.head(3)
#returns last 2 rows of dataframe
df.tail(2)
#returns first 5 rows of dataframe
df.head()
#returns the columns
df.columns
#prints information about the data
df.info()
#primts the information about datafrmae
df.info
#prints rows with values greater than 50
df[df.Duration>50]
#prints the row with min Pulse value
df[df.Pulse==df.Pulse.min()]
#removes columns
df1 = df.drop(columns=['Calories', 'Maxpulse'])
df1
#renames columns
df = df.rename(columns={'Duration':'Period'})
df
```

PROGRAM NO. - 2

Write a program for scaling the data using min-max scaling and standard scaling method

OUTPUT -

```
[[ 0.97596444 -1.61155897]
[-0.66776515 0.08481889]
[-1.28416374 1.10264561]
[ 0.97596444 0.42409446]]
```

```
[[1. 0. ]
[0.27272727 0.625 ]
[0. 1. ]
[1. 0.75 ]]
```

PROGRAM NO. -2

Write a program for scaling the data using min-max scaling and standard scaling method

```
# STANDARD SCALING METHOD
# import module
from sklearn.preprocessing import StandardScaler
# create data
data = [[11, 2], [3, 7], [0, 10], [11, 8]]
# compute required values
scaler = StandardScaler()
model = scaler.fit(data)
scaled data = model.transform(data)
# print scaled data
print(scaled_data)
# MIN-MAX SCALING
# import module
from sklearn.preprocessing import MinMaxScaler
# create data
data = [[11, 2], [3, 7], [0, 10], [11, 8]]
# scale features
scaler = MinMaxScaler()
model=scaler.fit(data)
scaled data=model.transform(data)
# print scaled features
print(scaled_data)
```

PROGRAM NO. – 3

Using train test split method of module model_selection split the data into training and testing.

OUTPUT -

Training set size: 120 Testing set size: 30

PROGRAM NO. – 3

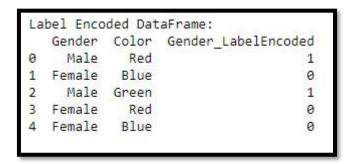
Using train test split method of module model_selection split the data into training and testing.

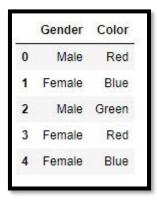
```
from sklearn.model_selection import train_test_split
from sklearn.datasets import load_iris
#load the iris datesheet
iris = load iris()
#split the dataset into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(iris.data, iris.target,
test_size = 0.2, random_state=2)
#print the size of the training and testing sets
print('Training set size:', len(X_train))
print('Testing set size:', len(X_test))
iris.target
```

PROGRAM NO. – 4

Write a program to convert the categorical data into numerical values using label encoder, 1-hot encoder.

OUTPUT -





	Gender	Color	Clr_Blue	Clr_Green	Clr_Red
0	Male	Red	0	0	1
1	Female	Blue	1	0	0
2	Male	Green	0	1	0
3	Female	Red	0	0	1
4	Female	Blue	1	0	0

PROGRAM NO. – 4

Write a program to convert the categorical data into numerical values using label encoder, 1-hot encoder.

```
#LABEL ENCODER
from sklearn.preprocessing import LabelEncoder
label encoder = LabelEncoder()
df['Gender_LabelEncoded'] =
label_encoder.fit_transform(df['Gender'])
print("Label Encoded DataFrame:")
print(df)
df2 = df.drop(columns=['Gender_LabelEncoded'])
df2
#ONE-HOT ENCODER
one_hot_encoded = pd.get_dummies(df['Color'], prefix='Clr')
df2 = pd.concat([df2, one_hot_encoded], axis=1)
print("\nOne-Hot Encoded DataFrame:")
print(df2)
df2
```

PROGRAM NO. – 5

Write a program to scale the data using robust scaling, max absolute scaling and minmax scaling within range.

OUTPUT -

0r	Original DataFrame:					
	Feature1	Feature2	Feature3			
0	23	752	4368			
1	43	764	5183			
2	95	194	4271			
3	30	261	6378			
4	57	746	1440			
5	48	743	1761			
6	88	736	3369			
7	14	315	2389			
8	89	823	1356			
9	39	547	4165			

```
Min-Max Scaled DataFrame:
Feature1 Feature2 Feature3
0 0.111111 0.887122 0.599761
1 0.358025 0.906200 0.762047
2 1.000000 0.0000000 0.580446
3 0.197531 0.106518 1.000000
4 0.530864 0.877583 0.016726
5 0.419753 0.872814 0.080645
6 0.913580 0.861685 0.400836
7 0.000000 0.192369 0.205695
8 0.925926 1.000000 0.0000000
9 0.308642 0.561208 0.559339
```

Standard Scaled DataFrame: Feature1 Feature2 Feature3 0 -1.075802 0.717120 0.558037 1 -0.348909 0.769624 1.063370 2 1.541013 -1.724326 0.497893 3 -0.821389 -1.431178 1.804318 4 0.159916 0.690868 -1.257442 5 -0.167185 0.677742 -1.058409 6 1.286601 0.647115 -0.061384 7 -1.402904 -1.194909 -0.669024 8 1.322946 1.027770 -1.309526 9 -0.494287 -0.179827 0.432168

Robust Scaled DataFrame: Feature1 Feature2 Feature3 0 -0.468750 0.033113 0.247758 1 -0.052083 0.064901 0.583737 2 1.031250 -1.445033 0.207771 3 -0.322917 -1.267550 1.076368 4 0.239583 0.017219 -0.959291 5 0.052083 0.009272 -0.826961 6 0.885417 -0.009272 -0.164073 7 -0.656250 -1.124503 -0.568072 8 0.906250 0.221192 -0.993919

9 -0.135417 -0.509934 0.164073

```
Min-Max Scaled DataFrame within Range:
   Feature1
             Feature2
                       Feature3
   1.111111
              8.871224
                         5.997611
   3.580247
              9.062003
                       7.620470
1
2 10.000000
              0.000000
                       5.804460
3
   1.975309
             1.065183 10.000000
   5.308642
             8.775835
                       0.167264
5
   4.197531
              8.728140
                        0.806452
6
   9.135802
              8.616852
                        4.008363
7
   0.000000
             1.923688
                         2.056949
8
   9.259259 10.000000
                         0.000000
9
   3.086420
              5.612083
                         5.593389
```

MaxAbs Scaled DataFrame:

Feature1 Feature2 Feature3 0 0.242105 0.913730 0.684854 1 0.452632 0.928311 0.812637 2 1.000000 0.235723 0.669646 3 0.315789 0.317132 1.000000 4 0.600000 0.906440 0.225776 5 0.505263 0.902795 0.276105 6 0.926316 0.894289 0.528222 7 0.147368 0.382746 0.374569 8 0.936842 1.000000 0.212606 9 0.410526 0.664642 0.653026

PROGRAM NO. – 5

Write a program to scale the data using robust scaling, max absolute scaling and minmax scaling within range.

```
#ORIGINAL DATAFRAME
import pandas as pd
import numpy as np
data = {
  'Feature1': np.random.randint(1,100,10),
  'Feature2': np.random.randint(100,1000,10),
  'Feature3': np.random.randint(1000,10000,10)
}
df = pd.DataFrame(data)
print("Original DataFrame:")
print(df)
#MinMaxScaler
from sklearn.preprocessing import MinMaxScaler
minmax scaler = MinMaxScaler()
minmax_scaled = minmax_scaler.fit_transform(df)
minmax_df = pd.DataFrame(minmax_scaled, columns=df.columns)
print("\nMin-Max Scaled DataFrame:")
print(minmax_df)
```

#Standardisation or z-scrore

from sklearn.preprocessing import StandardScaler standard_scaler = StandardScaler() standard_scaled = standard_scaler.fit_transform(df) standard_df = pd.DataFrame(standard_scaled, columns=df.columns) print("\nStandard Scaled DataFrame:") print(standard_df)

#Robust Scaling

from sklearn.preprocessing import RobustScaler
robust_scaler = RobustScaler()
robust_scaled = robust_scaler.fit_transform(df)
robust_df = pd.DataFrame(robust_scaled, columns=df.columns)
print("\nRobust Scaled DataFrame:")
print(robust_df)

#Max Absolute Scaling

from sklearn.preprocessing import MaxAbsScaler
maxabs_scaler = MaxAbsScaler()
maxabs_scaled = maxabs_scaler.fit_transform(df)
maxabs_df = pd.DataFrame(maxabs_scaled, columns=df.columns)
print("\nMaxAbs Scaled DataFrame:")
print(maxabs_df)

#MinMax Scaler within range $minmax_range = (0,10)$ minmax_range_scaler = MinMaxScaler(feature_range=minmax_range) minmax_range_scaled = minmax_range_scaler.fit_transform(df) minmax_range_df = pd.DataFrame(minmax_range_scaled, columns=df.columns) print("\nMin-Max Scaled DataFrame within Range:") print(minmax_range_df)