```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

data = pd.read_csv('/content/telecom_customer_churn.csv')

data

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	Customer ID	Gender	Age	Married	Number of Dependents	City	Zip Code	Latitude	Longitude
0	0002- ORFBO	Female	37	Yes	0	Frazier Park	93225	34.827662	-118.999073
1	0003- MKNFE	Male	46	No	0	Glendale	91206	34.162515	-118.203869
2	0004- TLHLJ	Male	50	No	0	Costa Mesa	92627	33.645672	-117.922613
3	0011- IGKFF	Male	78	Yes	0	Martinez	94553	38.014457	-122.115432
4	0013- EXCHZ	Female	75	Yes	0	Camarillo	93010	34.227846	-119.079903
7038	9987- LUTYD	Female	20	No	0	La Mesa	91941	32.759327	-116.997260
7039	9992- RRAMN	Male	40	Yes	0	Riverbank	95367	37.734971	-120.954271
7040	9992- UJOEL	Male	22	No	0	Elk	95432	39.108252	-123.645121
7041	9993- LHIEB	Male	21	Yes	0	Solana Beach	92075	33.001813	-117.263628
7042	9995- HOTOH	Male	36	Yes	0	Sierra City	96125	39.600599	-120.636358

7043 rows × 38 columns

data.describe()



	Age	Number of Dependents	Zip Code	Latitude	Longitude	Number of Referrals	Ter
count	7043.000000	7043.000000	7043.000000	7043.000000	7043.000000	7043.000000	7043
mean	46.509726	0.468692	93486.070567	36.197455	-119.756684	1.951867	32
std	16.750352	0.962802	1856.767505	2.468929	2.154425	3.001199	24
min	19.000000	0.000000	90001.000000	32.555828	-124.301372	0.000000	1
25%	32.000000	0.000000	92101.000000	33.990646	-121.788090	0.000000	9
50%	46.000000	0.000000	93518.000000	36.205465	-119.595293	0.000000	29
75%	60.000000	0.000000	95329.000000	38.161321	-117.969795	3.000000	55
max	80.000000	9.000000	96150.000000	41.962127	-114.192901	11.000000	72
4							•

data.info()

<<class 'pandas.core.frame.DataFrame'>
RangeIndex: 7043 entries, 0 to 7042
Data columns (total 38 columns):

#	Column	Non-Null Count	Dtype
0	Customer ID	7043 non-null	object
1	Gender	7043 non-null	object
2	Age	7043 non-null	int64
3	Married	7043 non-null	object
4	Number of Dependents	7043 non-null	int64
5	City	7043 non-null	object
6	Zip Code	7043 non-null	int64
7	Latitude	7043 non-null	float64
8	Longitude	7043 non-null	float64
9	Number of Referrals	7043 non-null	int64
10	Tenure in Months	7043 non-null	int64
11	Offer	3166 non-null	object
12	Phone Service	7043 non-null	object
13	Avg Monthly Long Distance Charges	6361 non-null	float64
14	Multiple Lines	6361 non-null	object
15	Internet Service	7043 non-null	object
16	Internet Type	5517 non-null	object
17	Avg Monthly GB Download	5517 non-null	float64
18	Online Security	5517 non-null	object
19	Online Backup	5517 non-null	object
20	Device Protection Plan	5517 non-null	object
21	Premium Tech Support	5517 non-null	object
22	Streaming TV	5517 non-null	object
23	Streaming Movies	5517 non-null	object
24	Streaming Music	5517 non-null	object

25	Unlimited Data	5517	non-null	object
26	Contract	7043	non-null	object
27	Paperless Billing	7043	non-null	object
28	Payment Method	7043	non-null	object
29	Monthly Charge	7043	non-null	float64
30	Total Charges	7043	non-null	float64
31	Total Refunds	7043	non-null	float64
32	Total Extra Data Charges	7043	non-null	int64
33	Total Long Distance Charges	7043	non-null	float64
34	Total Revenue	7043	non-null	float64
35	Customer Status	7043	non-null	object
36	Churn Category	1869	non-null	object
37	Churn Reason	1869	non-null	object

dtypes: float64(9), int64(6), object(23)
memory usage: 2.0+ MB

data.isnull().sum()



	0
Customer ID	0
Gender	0
Age	0
Married	0
Number of Dependents	0
City	0
Zip Code	0
Latitude	0
Longitude	0
Number of Referrals	0
Tenure in Months	0
Offer	3877
Phone Service	0
Avg Monthly Long Distance Charges	682
Multiple Lines	682
Internet Service	0
Internet Type	1526
Avg Monthly GB Download	1526
Online Security	1526
Online Backup	1526
Device Protection Plan	1526
Premium Tech Support	1526
Streaming TV	1526
Streaming Movies	1526
Streaming Music	1526
Unlimited Data	1526
Contract	0
Paperless Billing	0
Payment Method	0
Monthly Charge	0

Total Charges 0
Total Refunds 0
Total Extra Data Charges 0
Total Long Distance Charges 0
Total Revenue 0
Customer Status 0
Churn Category 5174

5174

dtype: int64

data.dropna(inplace=True)

Churn Reason

data.isnull().sum()



	0
Customer ID	0
Gender	0
Age	0
Married	0
Number of Dependents	0
City	0
Zip Code	0
Latitude	0
Longitude	0
Number of Referrals	0
Tenure in Months	0
Offer	0
Phone Service	0
Avg Monthly Long Distance Charges	0
Multiple Lines	0
Internet Service	0
Internet Type	0
Avg Monthly GB Download	0
Online Security	0
Online Backup	0
Device Protection Plan	0
Premium Tech Support	0
Streaming TV	0
Streaming Movies	0
Streaming Music	0
Unlimited Data	0
Contract	0
Paperless Billing	0
Payment Method	0
Monthly Charge	0

Total Charges	0	
Total Refunds	0	
Total Extra Data Charges	0	
Total Long Distance Charges	0	
Total Revenue	0	
Customer Status	0	
Churn Category	0	
Churn Reason	0	

dtype: int64

data



	Customer ID	Gender	Age	Married	Number of Dependents	City	Zip Code	Latitude	Longitude
2	0004- TLHLJ	Male	50	No	0	Costa Mesa	92627	33.645672	-117.922613
3	0011- IGKFF	Male	78	Yes	0	Martinez	94553	38.014457	-122.115432
54	0094- OIFMO	Female	78	No	1	North Hollywood	91605	34.207295	-118.400022
67	0112- QWPNC	Male	61	Yes	0	Valyermo	93563	34.395830	-117.734568
77	0125- LZQXK	Male	42	No	0	Big Oak Flat	95305	37.818589	-120.256995
7007	9940- RHLFB	Female	53	No	0	Sunset Beach	90742	33.719221	-118.073596
7010	9944- HKVVB	Female	55	No	0	King City	93930	36.220761	-120.980777
7012	9947- OTFQU	Male	65	No	0	Covina	91722	34.097345	-117.906736
7036	9985- MWVIX	Female	53	No	0	Hume	93628	36.807595	-118.901544
7039	9992- RRAMN	Male	40	Yes	0	Riverbank	95367	37.734971	-120.954271
681 rov	vs × 38 colu	mns							
4									•

data.columns

data



	Customer ID	Gender	Age	Married	Number of Dependents	City	Zip Code	Latitude	Longitude
2	0004- TLHLJ	Male	50	No	0	Costa Mesa	92627	33.645672	-117.922613
3	0011- IGKFF	Male	78	Yes	0	Martinez	94553	38.014457	-122.115432
54	0094- OIFMO	Female	78	No	1	North Hollywood	91605	34.207295	-118.400022
67	0112- QWPNC	Male	61	Yes	0	Valyermo	93563	34.395830	-117.734568
77	0125- LZQXK	Male	42	No	0	Big Oak Flat	95305	37.818589	-120.256995
									•••
7007	. 9940- RHLFB	Female	53	No	0	Sunset Beach	90742	33.719221	-118.073596
7010	9944- HKVVB	Female	55	No	0	King City	93930	36.220761	-120.980777
7012	9947- OTFQU	Male	65	No	0	Covina	91722	34.097345	-117.906736
7036	9985- MWVIX	Female	53	No	0	Hume	93628	36.807595	-118.901544
7039	9992- RRAMN	Male	40	Yes	0	Riverbank	95367	37.734971	-120.954271
681 rc	ws × 41 colu	mns							
4									•

from urllib.request import DataHandler
from sklearn.preprocessing import MinMaxScaler, StandardScaler

'Total Revenue']

scaler = MinMaxScaler()
data[numerical_columns] = scaler.fit_transform(data[numerical_columns])
data.head()

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	Customer ID	Gender	Age	Married	Number of Dependents	City	Zip Code	Latitude	Longitud
2	0004- TLHLJ	Male	0.508197	No	0.00	Costa Mesa	92627	0.115863	0.71717
3	0011- IGKFF	Male	0.967213	Yes	0.00	Martinez	94553	0.580316	0.24119
54	0094- OIFMO	Female	0.967213	No	0.25	North Hollywood	91605	0.175570	0.66297
67	0112- QWPNC	Male	0.688525	Yes	0.00	Valyermo	93563	0.195614	0.73851
77	0125- LZQXK	Male	0.377049	No	0.00	Big Oak Flat	95305	0.559493	0.45216
5 ro	ws × 41 colu	mns							

data.isnull().sum()

0



Customer ID	0
Gender	0
Age	0
Married	0
Number of Dependents	0
City	0
Zip Code	0
Latitude	0
Longitude	0
Number of Referrals	0
Tenure in Months	0
Offer	0
Phone Service	0
Avg Monthly Long Distance Charges	0
Multiple Lines	0
Internet Service	0
Internet Type	0
Avg Monthly GB Download	0
Online Security	0
Online Backup	0
Device Protection Plan	0
Premium Tech Support	0
Streaming TV	0
Streaming Movies	0
Streaming Music	0
Unlimited Data	0
Contract	0