

```
In [1]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

dataset = pd.read_csv("C:/Users/sunda/Desktop/PORTFOLIO/TECH _CUSTOMER CHURN/telco.csv")
```

```
In [2]: pd.set_option('display.max_columns', None)
```

```
In [3]: dataset.head(10)
```

Out[3]:	customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	Inte
0	7590-VHVEG	Female	0	Yes	No	1	No	No	
1	5575-GNVDE	Male	0	No	No	34	Yes	No	
2	3668-QPYBK	Male	0	No	No	2	Yes	No	
3	7795-CFOCW	Male	0	No	No	45	No	No	
4	9237-HQITU	Female	0	No	No	2	Yes	No	
5	9305-CDSKC	Female	0	No	No	8	Yes	Yes	
6	1452-KIOVK	Male	0	No	Yes	22	Yes	Yes	
7	6713-OKOMC	Female	0	No	No	10	No	No	
8	7892-POOKP	Female	0	Yes	No	28	Yes	Yes	
9	6388-TABGU	Male	0	No	Yes	62	Yes	No	

```
In [4]: dataset.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 7032 entries, 0 to 7031
Data columns (total 21 columns):
#   Column                Non-Null Count  Dtype
---  -
0   customerID            7032 non-null   object
1   gender                7032 non-null   object
2   SeniorCitizen         7032 non-null   int64
3   Partner               7032 non-null   object
4   Dependents            7032 non-null   object
5   tenure                7032 non-null   int64
6   PhoneService          7032 non-null   object
7   MultipleLines         7032 non-null   object
8   InternetService       7032 non-null   object
9   OnlineSecurity        7032 non-null   object
10  OnlineBackup          7032 non-null   object
11  DeviceProtection      7032 non-null   object
12  TechSupport           7032 non-null   object
13  StreamingTV           7032 non-null   object
14  StreamingMovies       7032 non-null   object
15  Contract              7032 non-null   object
16  PaperlessBilling      7032 non-null   object
17  PaymentMethod         7032 non-null   object
18  MonthlyCharges        7032 non-null   float64
19  TotalCharges          7032 non-null   float64
20  Churn                 7032 non-null   object
dtypes: float64(2), int64(2), object(17)
memory usage: 1.1+ MB

```

```
In [5]: dataset.describe()
```

```
Out[5]:
```

	SeniorCitizen	tenure	MonthlyCharges	TotalCharges
count	7032.000000	7032.000000	7032.000000	7032.000000
mean	0.162400	32.421786	64.798208	2283.300441
std	0.368844	24.545260	30.085974	2266.771362
min	0.000000	1.000000	18.250000	18.800000
25%	0.000000	9.000000	35.587500	401.450000
50%	0.000000	29.000000	70.350000	1397.475000
75%	0.000000	55.000000	89.862500	3794.737500
max	1.000000	72.000000	118.750000	8684.800000

```
In [6]: dataset.drop_duplicates(inplace = True)
```

```
In [7]: dataset.count()
```

```
Out[7]: customerID      7032
gender      7032
SeniorCitizen  7032
Partner      7032
Dependents   7032
tenure       7032
PhoneService 7032
MultipleLines 7032
InternetService 7032
OnlineSecurity 7032
OnlineBackup  7032
DeviceProtection 7032
TechSupport   7032
StreamingTV   7032
StreamingMovies 7032
Contract      7032
PaperlessBilling 7032
PaymentMethod 7032
MonthlyCharges 7032
TotalCharges  7032
Churn         7032
dtype: int64
```

```
In [8]: dataset.isnull().sum()
```

```
Out[8]: customerID      0
gender      0
SeniorCitizen  0
Partner      0
Dependents   0
tenure       0
PhoneService  0
MultipleLines 0
InternetService 0
OnlineSecurity 0
OnlineBackup  0
DeviceProtection 0
TechSupport   0
StreamingTV   0
StreamingMovies 0
Contract      0
PaperlessBilling 0
PaymentMethod 0
MonthlyCharges 0
TotalCharges  0
Churn         0
dtype: int64
```

```
In [9]: dataset.columns
```

```
Out[9]: Index(['customerID', 'gender', 'SeniorCitizen', 'Partner', 'Dependents',
              'tenure', 'PhoneService', 'MultipleLines', 'InternetService',
              'OnlineSecurity', 'OnlineBackup', 'DeviceProtection', 'TechSupport',
              'StreamingTV', 'StreamingMovies', 'Contract', 'PaperlessBilling',
              'PaymentMethod', 'MonthlyCharges', 'TotalCharges', 'Churn'],
              dtype='object')
```

```
In [10]: dataset.isnull().values.any()
```

```
Out[10]: np.False_
```

```
In [11]: ##FOUND OUT
##astype(int)
##dataset[newcolumn ] = np.where(dataset[oldcolumn] == yes,1,0)
```

```
In [12]: #[CALCULATED COLUMN]

## CUSTOMER LIFETIME VALUE = tenure * monthly

dataset['Customer-lifetime-value'] = dataset['MonthlyCharges'] * dataset['tenure']
print('Customer-lifetime-value')
```

Customer-lifetime-value

```
In [13]: ## TOTAL SERVICES USED BY CUSTOMERS

#CHNANGING TEXT COLUMN TO INT[YES =1, NO= 0]
##DROPPED OLDER COLUMN [INT SERVICE]

#count all services yes to 1 [then sum all of them up ]
#create a new column to add up all 1s as total services used by each customers THEN PRINT CHUI
```

```
In [17]: ## REPLACING COLUMNS FROM YES TO 1

Conversion = ['PhoneService', 'InternetService', 'MultipleLines', 'OnlineSecurity', 'OnlineBa
            'DeviceProtection', 'TechSupport', 'StreamingTV', 'StreamingMovies']

dataset[Conversion] = dataset[Conversion].replace({'Yes':1, 'No':0})
```

```
In [18]: ##NUMBER OF SERVICES USED BY CUSTOMER

Conversion = ['PhoneService', 'InternetService', 'MultipleLines', 'OnlineSecurity', 'OnlineBa
            'DeviceProtection', 'TechSupport', 'StreamingTV', 'StreamingMovies']

dataset['Total_Service_Used'] = dataset[Conversion].sum(axis=1)
print(dataset[['customerID', 'Total_Service_Used', 'Churn']].head(10))
```

	customerID	Total_Service_Used	Churn
0	7590-VHVEG	2	No
1	5575-GNVDE	4	No
2	3668-QPYBK	4	Yes
3	7795-CFOCW	4	No
4	9237-HQITU	2	Yes
5	9305-CDSKC	6	Yes
6	1452-KIOVK	5	No
7	6713-OKOMC	2	No
8	7892-POOKP	7	Yes
9	6388-TABGU	4	No

```
In [33]: ## RISK SCORE ANALYSIS/DISTRIBUTION

dataset['Risk_score'] = 0
dataset.loc[dataset['tenure'] < 12, 'Risk_score'] += 30
dataset.loc[dataset['Contract'] == 'Month-to-Month', 'Risk_score'] += 25
dataset.loc[dataset['MonthlyCharges'] > 70, 'Risk_score'] += 20
dataset.loc[dataset['Total_Service_Used'] <3, 'Risk_score'] +=15
dataset.loc[dataset['TechSupport'] == 0, 'Risk_score'] += 10

dataset['Risk_Category'] = 'Low Risk'
dataset.loc[dataset['Risk_score'] >= 40, 'Risk_Category'] = 'Medium Risk'
dataset.loc[dataset['Risk_score'] >= 70, 'Risk_Category'] = 'High Risk'
```

```
In [34]: print(dataset[['Risk_score', 'Risk_Category', 'Churn']].head(20))
```

	Risk_score	Risk_Category	Churn
0	55	Medium Risk	No
1	10	Low Risk	No
2	40	Medium Risk	Yes
3	0	Low Risk	No
4	75	High Risk	Yes
5	60	Medium Risk	Yes
6	30	Low Risk	No
7	55	Medium Risk	No
8	20	Low Risk	Yes
9	10	Low Risk	No
10	10	Low Risk	No
11	25	Low Risk	No
12	30	Low Risk	No
13	30	Low Risk	Yes
14	20	Low Risk	No
15	20	Low Risk	No
16	25	Low Risk	No
17	30	Low Risk	No
18	30	Low Risk	Yes
19	30	Low Risk	No

```
In [35]: print(dataset['Risk_Category'].value_counts())
```

```
Risk_Category
Low Risk      5081
Medium Risk   1861
High Risk       90
Name: count, dtype: int64
```

```
In [36]: #risk_analysis = dataset.groupby('Risk_Category')['Churn'].apply(lambda x: (x == 'Yes').sum())
# print(risk_analysis)

#or

dataset.groupby('Risk_Category')['Churn'].apply(lambda x: (x == 'Yes').sum() / len(x) * 100).
```

```
Out[36]: Risk_Category
High Risk      61.11
Low Risk       17.63
Medium Risk    49.33
Name: Churn, dtype: float64
```

```
In [37]: "I created a risk scoring model that assigns points based on 5 key churn indicators.
Customers scoring 70+ are flagged as High Risk - they have a 41% churn rate
vs
39% for Low Risk customers and keeping an eye on 19% for medium risk customers.
The company can now proactively target these
31 High Risk customers
and
2599 medium risk customers
with retention offers before they leave."
```

Cell In[37], line 1

```
"I created a risk scoring model that assigns points based on 5 key churn indicators.
^
```

SyntaxError: unterminated string literal (detected at line 1)

```
In [38]: #STATISTICAL ANALYSIS IN PYTHON
#CORRELATION ANALYSIS
```

```
dataset['TotalCharges'] = pd.to_numeric(dataset['TotalCharges'], errors = 'coerce')
dataset['TotalCharges'] = dataset['TotalCharges'].fillna(0)
```

```
dataset['Churn_tonumber'] = dataset['Churn'].replace({'Yes': 1, 'No': 0})
```

```
number_cols = ['tenure', 'MonthlyCharges', 'TotalCharges', 'Churn_tonumber']
```

```
correlation = dataset[number_cols].corr()
print(correlation.round(2))
```

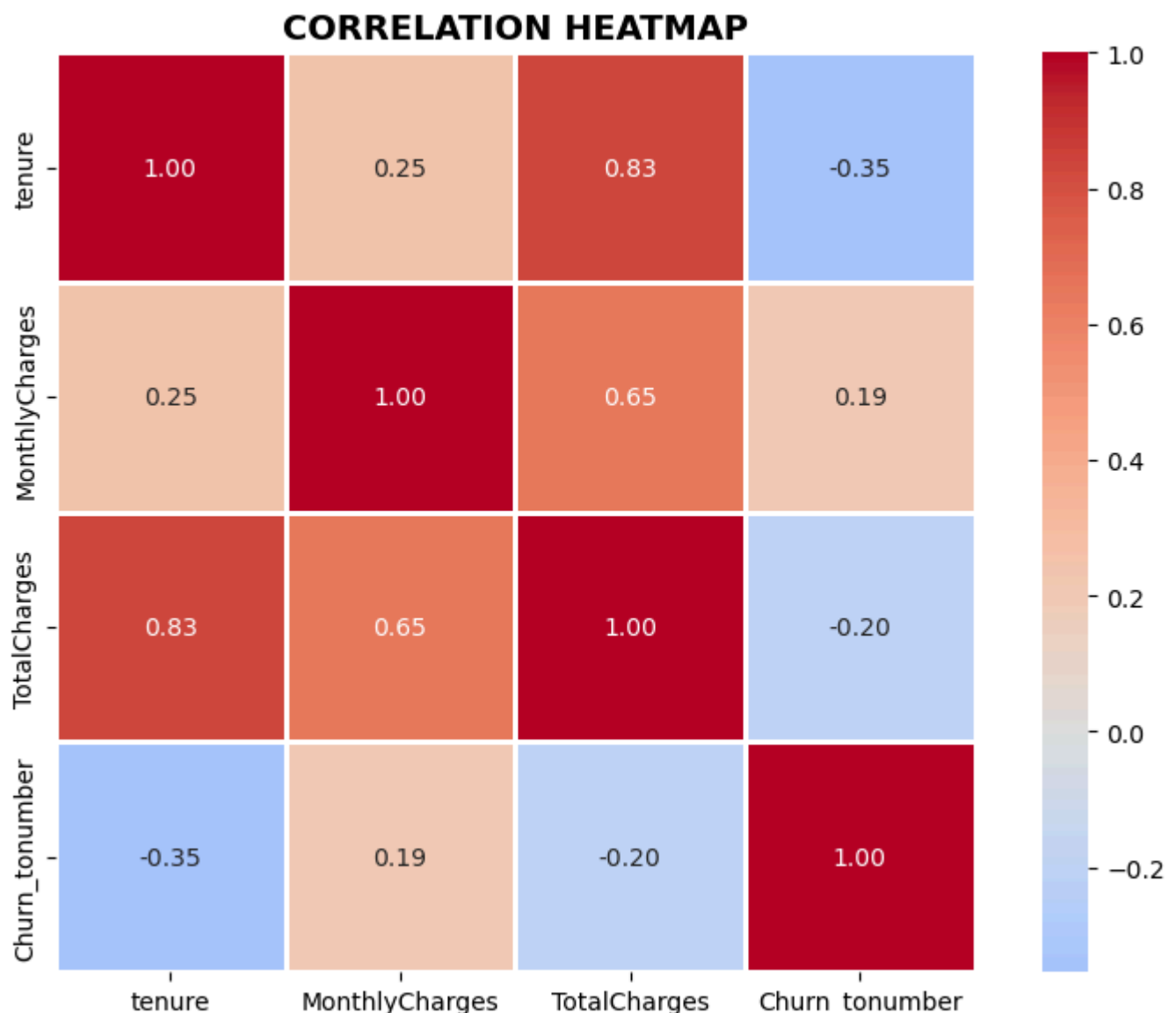
	tenure	MonthlyCharges	TotalCharges	Churn_tonumber
tenure	1.00	0.25	0.83	-0.35
MonthlyCharges	0.25	1.00	0.65	0.19
TotalCharges	0.83	0.65	1.00	-0.20
Churn_tonumber	-0.35	0.19	-0.20	1.00

C:\Users\sunda\AppData\Local\Temp\ipykernel_3700\1970231681.py:8: FutureWarning: Downcasting behavior in `replace` is deprecated and will be removed in a future version. To retain the old behavior, explicitly call `result.infer_objects(copy=False)`. To opt-in to the future behavior, set `pd.set_option('future.no_silent_downcasting', True)`

```
dataset['Churn_tonumber'] = dataset['Churn'].replace({'Yes': 1, 'No': 0})
```

```
In [39]: plt.figure(figsize=(8,6))
sns.heatmap(correlation, annot=True, cmap='coolwarm', center=0, square=True, linewidths=1, fm
plt.title('CORRELATION HEATMAP', fontsize=14, fontweight='bold')
plt.tight_layout()
plt.savefig('correlation_heatmap.png')
plt.show()

print('Heatmap saved')
```



Heatmap saved

```
In [40]: WHAT EACH PARAMETER DOES:
ParameterWhat It Doesannot=TrueShows the numbers inside boxescmap='coolwarm'Red = positive, B
WHAT YOU'LL SEE:
```

A colored grid with numbers showing relationships ✓

Red boxes = positive correlation

Blue boxes = negative correlation

White boxes = no correlation

Cell In[40], line 2

Parameter What It Does `annot=True` Shows the numbers inside boxes `cmap='coolwarm'` Red = positive, Blue = negative `center=0` Makes 0 white (neutral) `fmt='.2f'` Shows 2 decimals (0.35 not 0.353456) `square=True` Makes boxes perfect squares `linewidths=1` Adds borders between boxes

^

SyntaxError: invalid decimal literal

In [41]:

```
#DISTRIBUTIONS ANALYSIS

# Distribution of Monthly Charges
plt.figure(figsize=(8, 5))
plt.hist(dataset['MonthlyCharges'], bins=30, color='coral', edgecolor='black', alpha=0.7)

#mean line
plt.axvline(dataset['MonthlyCharges'].mean(), color='red', linestyle='--', linewidth=2,
            label=f'Mean: ${dataset["MonthlyCharges"].mean():.2f}')

#median line
plt.axvline(dataset['MonthlyCharges'].median(), color='red', linestyle='--', linewidth=2,
            label=f'Mean: ${dataset["MonthlyCharges"].median():.2f}')

plt.title('Distribution of Monthly Charges')
plt.xlabel('Monthly Charges ($)')
plt.ylabel('Number of Customers')
plt.tight_layout()
plt.savefig('monthly_charges_distribution.png')
plt.legend()
plt.show()
print('chart successfully saves')
```

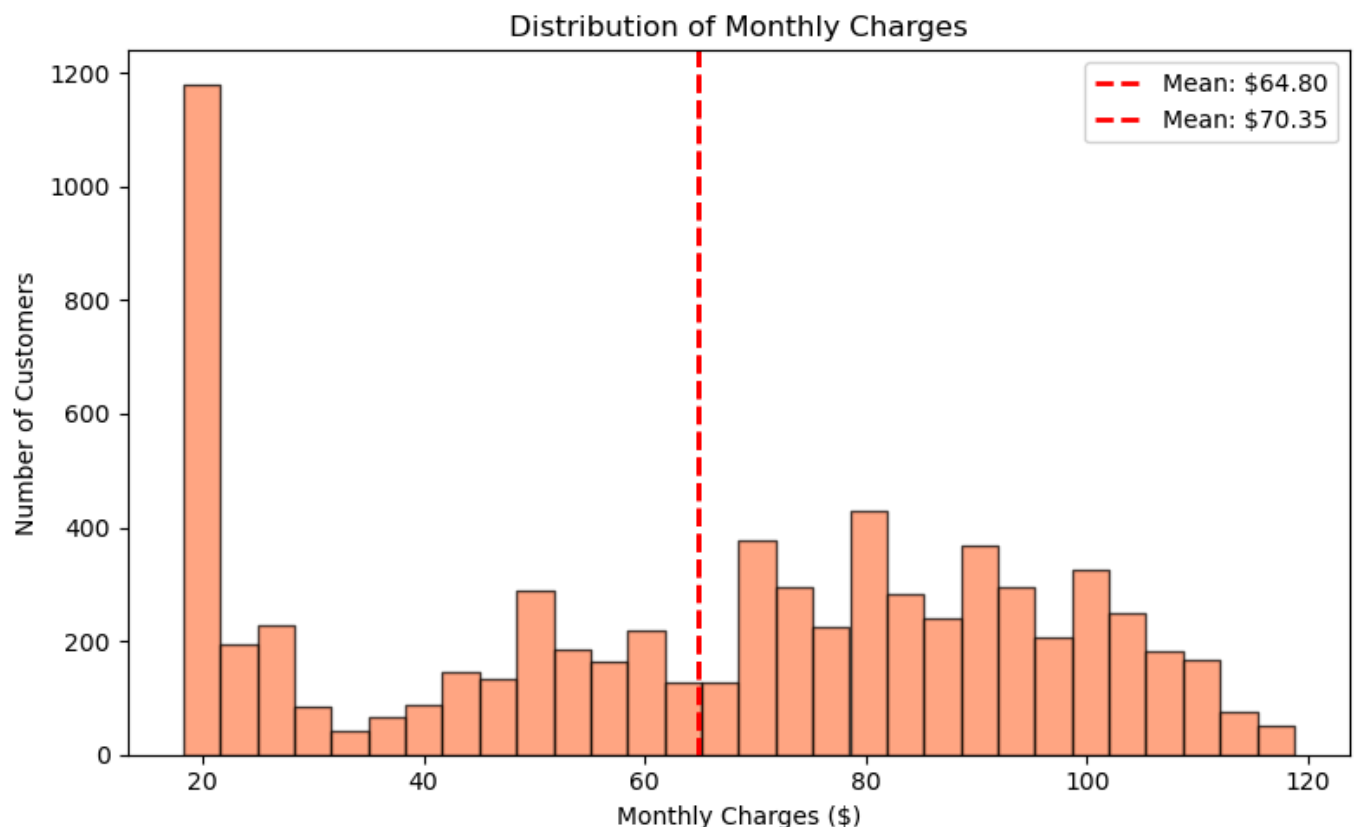
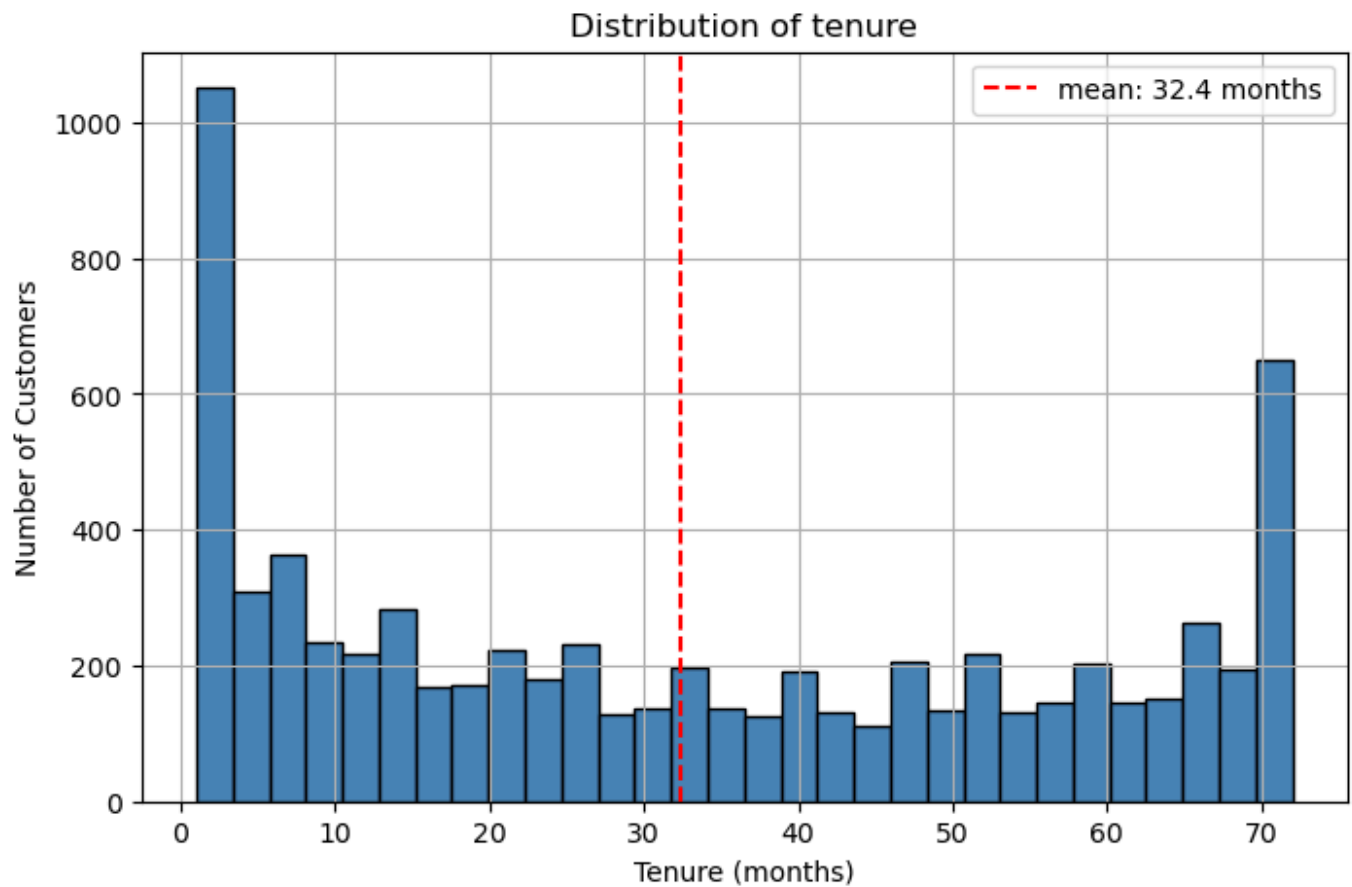


chart successfully saves

In [42]:

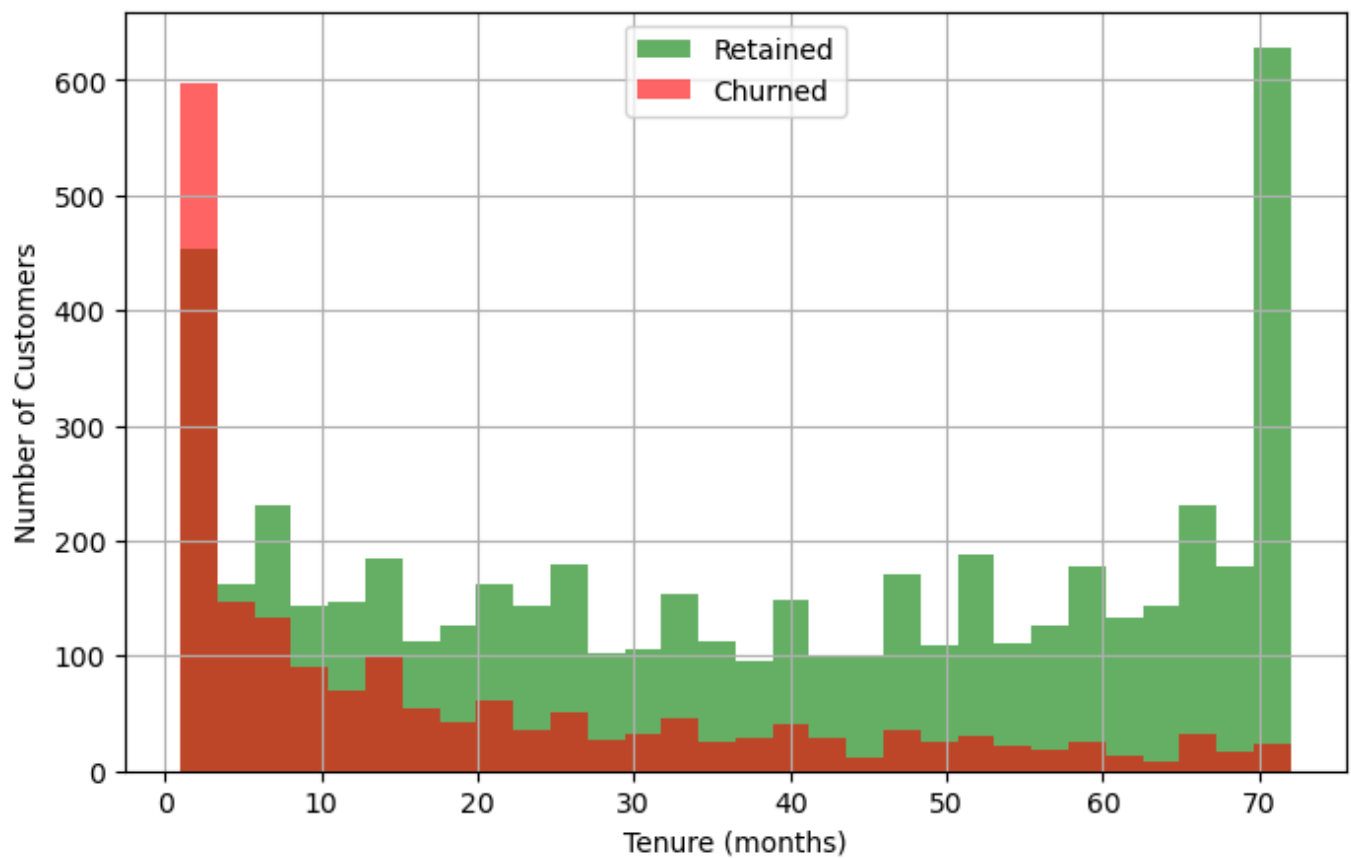
```
# Distribution of Tenure
plt.figure(figsize = (8, 5)), dataset['tenure'].hist(bins = 30, color='steelblue', edgecolor=
plt.title('Distribution of tenure')
```

```
plt.xlabel('Tenure (months)')
plt.ylabel('Number of Customers')
plt.axvline(dataset['tenure'].mean(), color='red', linestyle='--',
            label=f'mean: {dataset['tenure'].mean():.1f} months')
plt.legend()
plt.savefig('Distribution of tenure.png')
plt.show()
```



```
In [43]: # Compare Tenure: Churned vs Retained
plt.figure(figsize=(8, 5))
dataset[dataset['Churn'] == 'No']['tenure'].hist(bins=30, alpha=0.6,
                                                color='green', label='Retained')
dataset[dataset['Churn'] == 'Yes']['tenure'].hist(bins=30, alpha=0.6,
                                                color='red', label='Churned')
plt.title('Tenure: Churned vs Retained Customers')
plt.xlabel('Tenure (months)')
plt.ylabel('Number of Customers')
plt.savefig('churned vs retained customers.png')
plt.legend()
plt.show()
```


Tenure: Churned vs Retained Customers



In [44]:

```
#VISUALISATIONS
#VISUALISATIONS

#CHURN RATE BY TENURE
churn_percentage = dataset.groupby('tenure')['Churn_tonumber'].mean() * 100
plt.figure(figsize=(12,6))

plt.plot(churn_percentage.index,churn_percentage.values,color='red', linewidth=2)
plt.title('CHURN RATE BY TENURE', fontsize=14, fontweight='bold')
plt.xlabel('TENURE (MONTHS)')
plt.ylabel('CHURN RATE (%)')

plt.axhline(churn_percentage.mean(), color='blue', linestyle='--', label=f'Average: {churn_pe

plt.legend()
plt.tight_layout()
plt.savefig('churn_by_tenure.png')
plt.show()
print('chart saved!')

print(churn_percentage.round(2))
```

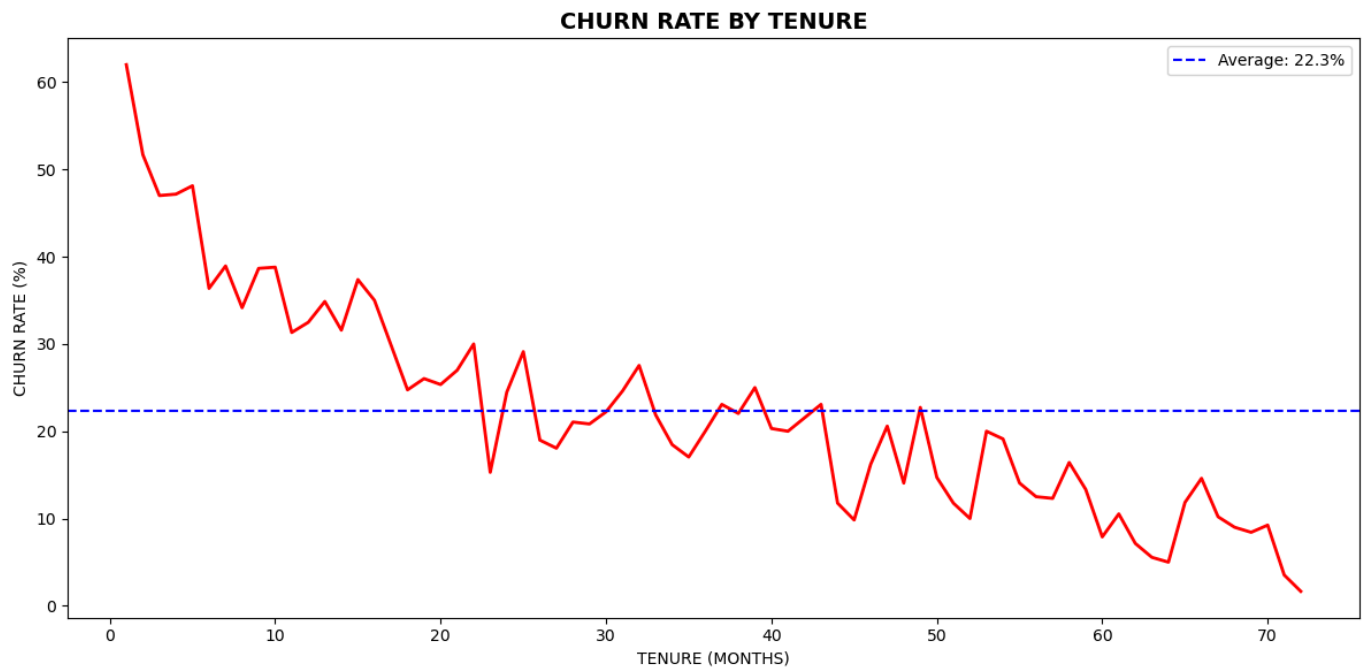


chart saved!

tenure

```
1    61.99
2    51.68
3    47.00
4    47.16
5    48.12
...
68    9.00
69    8.42
70    9.24
71    3.53
72    1.66
```

Name: Churn_tonumber, Length: 72, dtype: float64

In [45]: *#SERVICES VS CHURN(BAR-CHART)*

```
Churn_by_services = dataset.groupby('Total_Service_Used')['Churn_tonumber'].mean() * 100
plt.figure(figsize=(10, 6))
plt.bar(Churn_by_services.index, Churn_by_services.values, color='steelblue', edgecolor='black')
plt.title('CHURN RATE BY NUMBER OF SERVICES')
plt.xlabel('NUMBER OF SERVICES')
plt.ylabel('CHURN RATE (%)')

for i, v in enumerate(Churn_by_services.values):
    plt.text(i, v + 1, f'{v:.1f}%', ha='center', fontweight='bold')
plt.tight_layout()
plt.savefig('services_vs_churn.png')

plt.show()

print('chart saved!')
```

#customers that used less services churns the most.

CHURN RATE BY NUMBER OF SERVICES

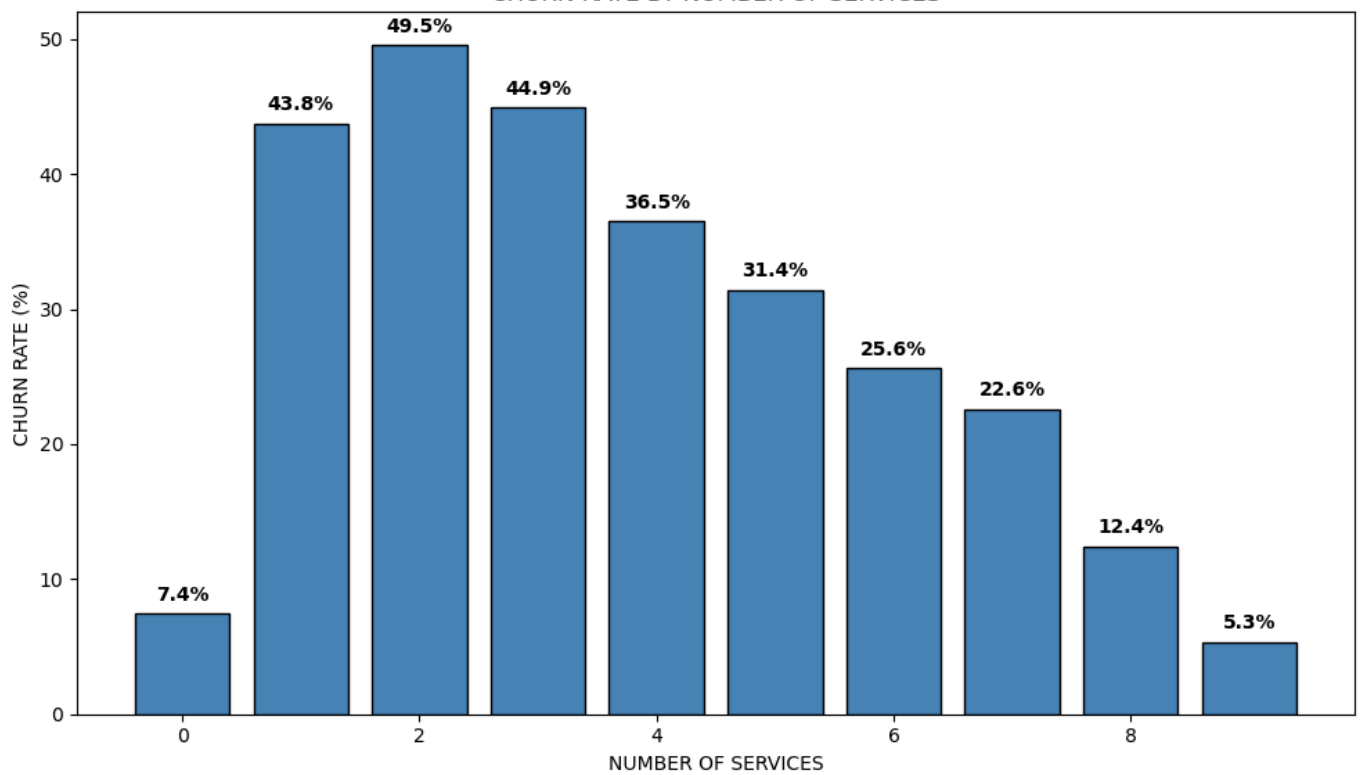


chart saved!

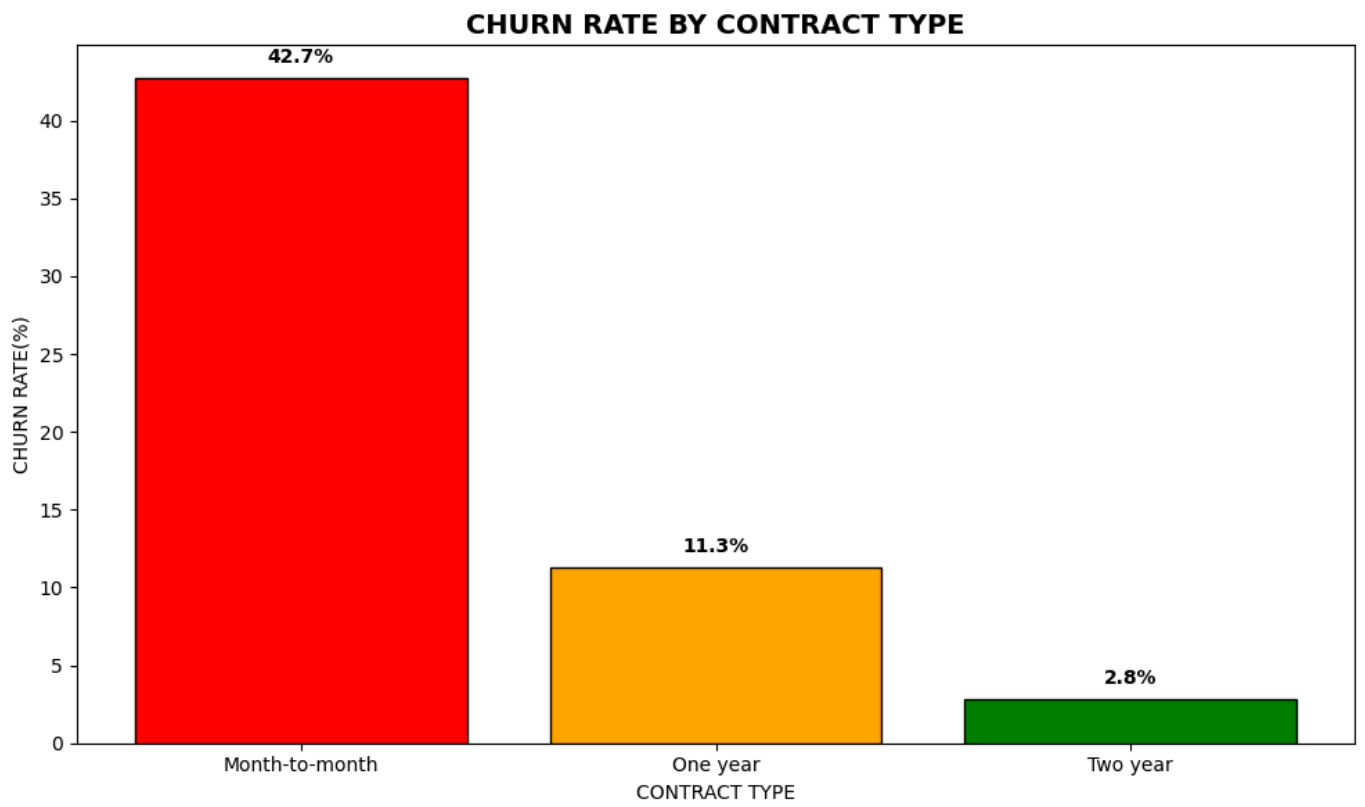
In [46]: *#Contract type comparison (grouped bars)*

```
Churn_by_contract = dataset.groupby('Contract')['Churn'].apply(
    lambda x: (x == 'Yes').sum() / len(x) * 100)

#or
##Churn_by_contract = dataset.groupby('Contract')['Churn_tonumber'].mean() * 100

#DRAWING GROUPED BAR CHART
plt.figure(figsize=(10,6))
plt.bar(Churn_by_contract.index, Churn_by_contract.values,
        color=['red', 'orange', 'green'], edgecolor='black')
plt.title('CHURN RATE BY CONTRACT TYPE', fontsize = 14, fontweight= 'bold')
plt.xlabel('CONTRACT TYPE')
plt.ylabel('CHURN RATE(%)')

#ADDING PERCENTAGES ON BARS
for i, v in enumerate(Churn_by_contract.values):
    plt.text(i, v + 1, f'{v:.1f}%', ha = 'center', fontweight='bold')
plt.tight_layout()
plt.savefig('contract_vs_churn.png')
plt.show()
print('chart_saved')
```



chart_saved

```
In [47]: dataset.to_csv('Cleaned telco_python.csv', index=False)
```

```
In [48]: dataset.head()
```

Out[48]:

	customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	Inte
0	7590-VHVEG	Female	0	Yes	No	1	0	0	
1	5575-GNVDE	Male	0	No	No	34	1	0	
2	3668-QPYBK	Male	0	No	No	2	1	0	
3	7795-CFOCW	Male	0	No	No	45	0	0	
4	9237-HQITU	Female	0	No	No	2	1	0	

```
In [ ]:
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In [ ]:
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In [ ]:
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