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1. What is the overall churn rate, and how does it trend over time (if applicable)?

Churn Rate  
20%

Churn\_rate by Tenure

Tenure	Churn Rate (%)
0	23%
1	22%
2	19%
3	21%
4	21%
5	21%
6	20%
7	17%
8	19%
9	22%
10	21%

Data

Filters

Customer-Churn-Reco...  
Σ Age  
Σ Balance  
Card Type  
Churn\_rate  
Σ Complain  
Σ CreditScore  
Σ CustomerId  
Σ EstimatedSalary  
Σ Exited  
Gender  
Geography  
HasCrCard  
IsActiveMember  
Σ NumOfProducts  
Σ Point Earned  
Σ RowNumber  
Σ Satisfaction Sc...  
Surname  
Tenure

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2. How does churn vary by geography (France, Germany, Spain)?

Churn\_rate by Geography

Geography	Churn Rate (%)
Germany	32%
Spain	17%
France	16%

Customer-Churn-Reco...  
Age  
Balance  
Card Type  
Churn\_rate  
Complain  
CreditScore  
CustomerId  
EstimatedSalary  
Exited  
Gender  
Geography  
HasCrCard  
IsActiveMember  
NumOfProducts  
Point Earned  
RowNumber  
Satisfaction Sc...  
Surname  
Tenure

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3. What are the differences in churn rates between male and female customers?

Churn\_rate by Gender

Gender	Churn Rate (%)
Female	25%
Male	16%

Customer-Churn-Reco...

- $\sum$  Age
- $\sum$  Balance
- Card Type
- Churn\_rate
- $\sum$  Complain
- $\sum$  CreditScore
- $\sum$  CustomerId
- $\sum$  EstimatedSalary
- $\sum$  Exited
- Gender
- Geography
- HasCrCard
- IsActiveMember
- $\sum$  NumOfProducts
- $\sum$  Point Earned
- $\sum$  RowNumber
- $\sum$  Satisfaction Sc...
- Surname
- Tenure

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Churn\_rate by Age Group

55-64 45-54 35-44 65+ Under 25 25-34

50% 48% 18% 15% 9% 8%

4. What is the churn rate distribution across different age groups?

Customer-Churn-Re...

- $\sum$  Age
- $\sum$  Age Group
- $\sum$  Balance
- Card Type
- Churn\_rate
- $\sum$  Complain
- $\sum$  CreditScore
- $\sum$  CustomerId
- $\sum$  EstimatedSalary
- $\sum$ Exited
- Gender
- Geography
- HasCrCard
- IsActiveMemb...
- $\sum$  NumOfProdu...
- $\sum$  Point Earned
- $\sum$  RowNumber
- $\sum$  Satisfaction Sc...
- Surname

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Age Group	Churn Rate (%)
55-64	50%
45-54	48%
35-44	18%
65+	15%
Under 25	9%
25-34	8%

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**5. How does churn behavior change with tenure?**

**Churn\_rate by Tenure**

Tenure	Churn Rate (%)
0	23%
1	22%
2	19%
3	21%
4	21%
5	21%
6	20%
7	17%
8	19%
9	22%
10	21%

Format ...

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Filter pane

Filter cards

Customer-Churn-Re...

Σ Age

Age Group

Σ Balance

Card Type

Churn\_rate

Σ Complain

CreditScore

CustomerId

EstimatedSalary

Exited

Gender

Geography

HasCrCard

IsActiveMemb...

NumOfProdu...

Point Earned

RowNumber

Satisfaction Sc...

Surname

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6. What percentage of high-balance customers (e.g., >\$100K) have churned?

Churn\_rate by High Balance cust

Balance Category	Churn Rate (%)
High Balance	23%
Low Balance	20%

High Balance churn(%)

23.12

Customer-Churn-Re...

- $\sum$  Age
- $\text{Age Group}$
- $\text{Avg_Balance}$
- $\sum$  Balance
- Card Type
- $\text{Churn_rate}$
- $\sum$  Complain
- $\sum$  CreditScore
- $\sum$  CustomerId
- $\sum$  EstimatedSalary
- $\sum$  Exited
- Gender
- Geography
- $\sum$  HasCrCard
- $\text{High balance ...}$
- $\text{High Balance ...}$
- $\sum$  IsActiveMemb...
- $\sum$  NumOfProdu...
- $\sum$  Point Earned

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7. How do churn rates compare across different NumOfProducts?

Churn\_rate by NumOfProducts

NumOfProducts	Churn_rate (%)
4	100%
3	83%
1	28%
2	8%

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Customer-Churn-Re...

- $\sum$  Age
- $\text{Age Group}$
- $\text{Avg_Balance}$
- $\sum$  Balance
- Card Type
- $\text{Churn_rate}$
- $\sum$  Complain
- $\sum$  CreditScore
- $\sum$  CustomerId
- $\sum$  EstimatedSalary
- $\sum$  Exited
- Gender
- Geography
- $\sum$  HasCrCard
- $\text{High balance ...}$
- $\text{High Balance ...}$
- $\sum$  IsActiveMemb...
- NumOfProdu...
- $\sum$  Point Earned

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**8. Are inactive members (`IsActiveMember = 0`) more likely to churn?**

**Churn\_rate by Member status**

Member status	Churn_rate
Active	14%
Inactive	27%
Total	20%

Customer-Churn-Report

- $\sum$  Age
- $\text{fx}$  Age Group
- $\text{Avg}$  Balance
- $\sum$  Balance
- Card Type
- $\text{Churn\_rate}$
- $\sum$  Complain
- $\sum$  CreditScore
- $\sum$  CustomerId
- $\sum$  EstimatedSalary
- $\sum$  Exited
- Gender
- Geography
- $\sum$  HasCrCard
- $\text{High balance ...}$
- $\text{fx}$  High Balance ...
- $\sum$  IsActiveMemb...
- $\text{fx}$  Member status
- NumOfProdu...

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9. Can we build a KPI dashboard to track churn metrics like average salary, balance, and active users?

Avg\_salary all  
100.09K

Churn Rate  
20%

Geography

- France
- Germany
- Spain

Avg Balance  
76.49K

Active Users  
5151

Gender

- Female
- Male

Avg\_sal churned  
101.51K

Total Customer  
10K

Data

Search

Churn\_rate  
 $\sum$  Complain  
 $\sum$  CreditScore  
 $\sum$  CustomerId  
 $\sum$  EstimatedSalary  
 $\sum$  Exited  
Gender  
Geography  
 $\sum$  HasCrCard  
 $\sum$  High balance ...  
 $\sum$  High Balance ...  
 $\sum$  IsActiveMemb...  
 $\sum$  Member status  
NumOfProdu...  
 $\sum$  Point Earned  
 $\sum$  RowNumber  
 $\sum$  Satisfaction Sc...  
Surname  
Tenure  
 $\sum$  Total customers

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Churned vs. Retained DASHBOARD

**Churned** vs. **Retained**

Total customer: 10K

Churn Rate: 20%

Gender: Female, IsActive: Member

Satisfaction Score: 1, 2, 3, 4, 5

Churned customers by Age Group:

Age Group	Total customers
35-44	~500
45-54	~500
55-64	~200
25-34	~150
65+	~50
Under 25	~50

Churned customers by Geography:

Geography	Total customers
Germany	~800
France	~700
Spain	~400

Churned avg balance: 91.1...

Churned avg salary: 101....

Churned customers by Gender:

Retained customers by Age Group:

Age Group	Total customers
35-44	~3000
25-34	~2500
45-54	~500
Under 25	~200
55-64	~100
65+	~50

Retained customers by Geography:

Geography	Total customers
France	~4500
Spain	~2000
Germany	~1000

Retained avg balance: 72.74...

Retained avg salary: 99.73K

Retained customers by Gender:

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

[3]: df=pd.read_csv("Customer-churn-Records.csv")

[5]: df.head()

[5]:
   RowNumber CustomerId Surname CreditScore Geography Gender Age Tenure Balance NumOfProducts HasCrCard IsActiveMember EstimatedSalary Exited
0         1  15634602 Hargrave        619    France Female  42      2     0.00           1          1            1       101348.88      1
1         2  15647311     Hill        608    Spain Female  41      1   83807.86           1          0            1       112542.58      0
2         3  15619304    Onio        502    France Female  42      8  159660.80           3          1            0       113931.57      1
3         4  15701354    Boni        699    France Female  39      1     0.00           2          0            0       93826.63      0
4         5  15737888  Mitchell        850    Spain Female  43      2  125510.82           1          1            1       79084.10      0
```

◀ ▶

```
[6]: df.shape

[6]: (10000, 18)

[7]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 18 columns):
```

```
memory usage: 4.77 MB
```

```
[ ]:
```

```
[ ]: # 11. Check the percentage of missing data and handle accordingly.
```

```
[8]: df.isnull().sum()
```

```
[8]: RowNumber      0  
CustomerId      0  
Surname        0  
CreditScore     0  
Geography       0  
Gender          0  
Age             0  
Tenure          0  
Balance          0  
NumOfProducts    0  
HasCrCard        0  
IsActiveMember   0  
EstimatedSalary   0  
Exited           0  
Complain         0  
Satisfaction Score 0  
Card Type        0  
Point Earned     0  
dtype: int64
```

```
[ ]:
```



```
Balance          0
NumOfProducts    0
HasCrCard        0
IsActiveMember   0
EstimatedSalary   0
Exited           0
Complain         0
Satisfaction Score 0
Card Type        0
Point Earned     0
dtype: int64
```

```
[ ]:
```

```
[ ]: # 12. How many rows and columns are there in the dataset?
```

```
[10]: df.shape
```

```
[10]: (10000, 18)
```

```
[13]: rows, cols =df.shape
print(f"The dataset has {rows} rows and {cols} columns.")
```

```
The dataset has 10000 rows and 18 columns.
```

```
[ ]:
```



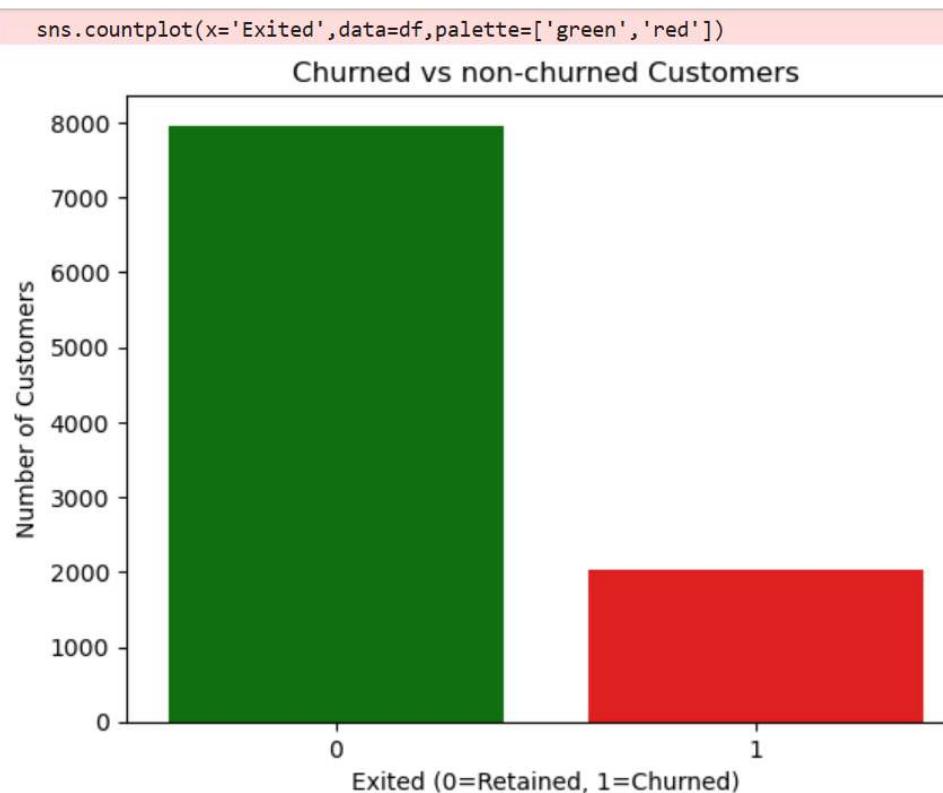
```
[ ]:
```

```
[ ]: 
[ ]: #13. What is the distribution of churned vs. non-churned customers?
[14]: df['Exited'].value_counts()
[14]: Exited
0    7962
1    2038
Name: count, dtype: int64
[15]: df['Exited'].value_counts(normalize=True)*100
[15]: Exited
0    79.62
1    20.38
Name: proportion, dtype: float64
[23]: sns.countplot(x='Exited',data=df,palette=['green','red'])
plt.title('Churned vs non-churned Customers')
plt.xlabel('Exited (0=Retained, 1=Churned)')
plt.ylabel('Number of Customers')
plt.show()

C:\Users\nayak\AppData\Local\Temp\ipykernel_19932\3929842554.py:1: FutureWarning:
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

sns.countplot(x='Exited',data=df,palette=['green','red'])
```

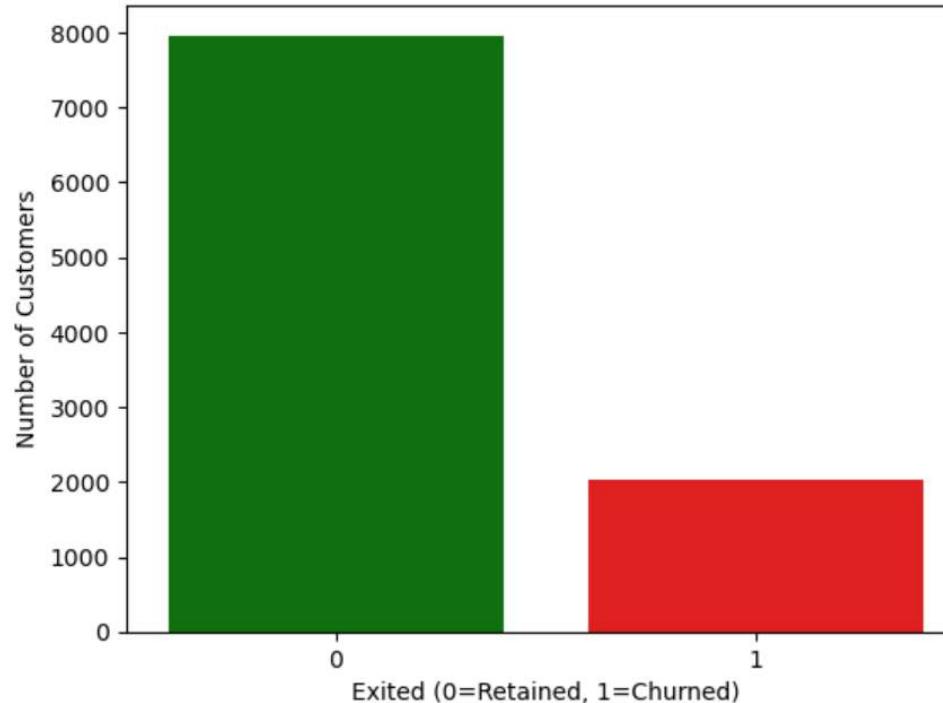
Churned vs non-churned Customers



JupyterLab Python [conda env:base] \* Trusted

```
sns.countplot(x='Exited',data=df,palette=['green','red'])
```

Churned vs non-churned Customers



```
[ ]: # About 20% customers have Churned , While 80% remain with Bank.
```



```
[ ]:
```

```
[ ]: #14. What is the distribution of EstimatedSalary of churned and retained Customers?
```



```
[ ]:
```

```
[25]: df.groupby('Exited')['EstimatedSalary'].describe()
```

```
[25]:
```

	count	mean	std	min	25%	50%	75%	max
<b>Exited</b>								
<b>0</b>	7962.0	99726.853141	57399.956717	90.07	50783.295	99620.355	148602.4450	199992.48
<b>1</b>	2038.0	101509.908783	57932.623392	11.58	51924.020	102489.335	152443.8575	199808.10

```
[26]: plt.figure(figsize=(8,5))
sns.boxplot(x='Exited',y='EstimatedSalary',data=df,palette=['green','red'])
plt.title('Distribution of EstimatedSalary- Churned vs Retained Customers')
plt.xlabel('Exited(0=Retained,1=Churned)')
plt.ylabel('Estimated Salary')
plt.show()
```

```
C:\Users\nayak\AppData\Local\Temp\ipykernel_19932\3890800965.py:2: FutureWarning:
```

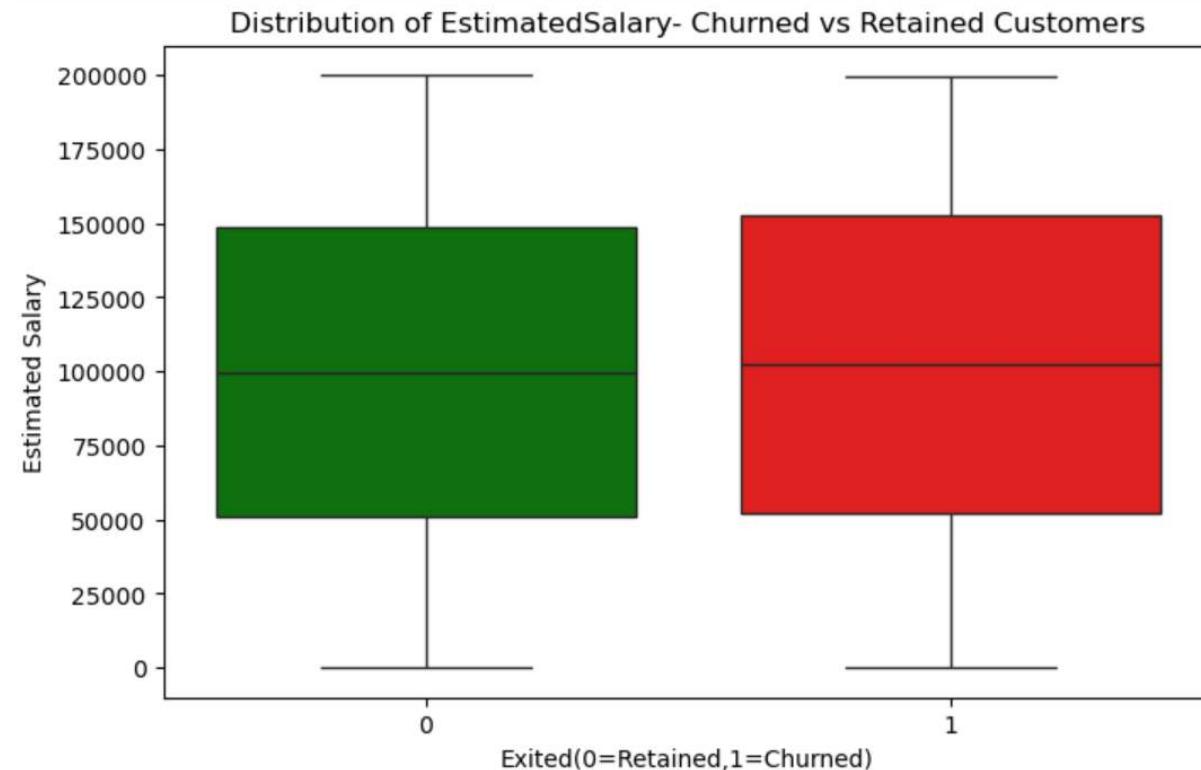
```
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.
```

```
sns.boxplot(x='Exited',y='EstimatedSalary',data=df,palette=['green','red'])
```

Distribution of EstimatedSalary- Churned vs Retained Customers

200000





```
[ ]: # Both Churned and Retained customers have similar salary distributions,So estimated salary is not a major factor in customer churn for data.
```

```
[ ]:
```

```
[ 1]:
```

```
[ ]:  
[20]: # 15. How do churn rates vary by Gender, Geography, and IsActiveMember?  
[21]: #churn rate vary by Genger  
[22]: churn_gender=df.groupby('Gender')['Exited'].mean()*100  
      print(churn_gender)
```

```
Gender  
Female    25.071539  
Male      16.474253  
Name: Exited, dtype: float64
```

```
[23]: plt.figure(figsize=(6,4))  
sns.barplot(x='Gender',y='Exited',data=df,estimator='mean',palette='pastel')  
plt.title('Churn Rate by Gender')  
plt.ylabel('Churn Rate (%)')  
plt.show()
```

```
C:\Users\nayak\AppData\Local\Temp\ipykernel_6884\2934609559.py:2: FutureWarning:
```

```
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.
```

```
sns.barplot(x='Gender',y='Exited',data=df,estimator='mean',palette='pastel')
```



[24]: #Female Customers have a higher churn Rate.

[25]: #churn rate vary by Geography

[30]: `churn_geography=df.groupby('Geography')['Exited'].mean()*100  
print(churn_geography)`

Geography  
France 16.174711  
Germany 22.442281

```
[25]: #churn rate vary by Geography
```

```
[30]: churn_geography=df.groupby('Geography')['Exited'].mean()*100
print(churn_geography)
```

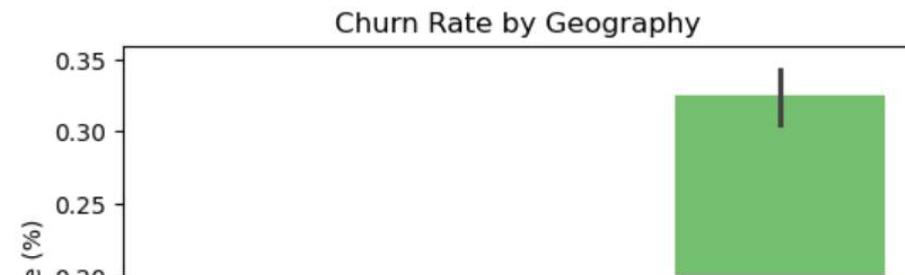
```
Geography
France      16.174711
Germany     32.443204
Spain        16.673395
Name: Exited, dtype: float64
```

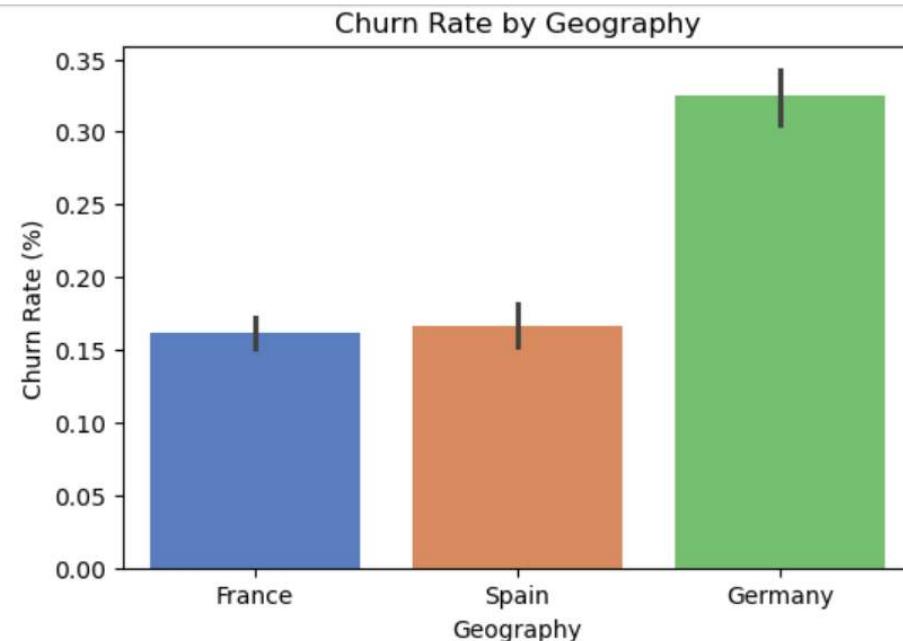
```
[26]: plt.figure(figsize=(6,4))
sns.barplot(x='Geography',y='Exited',data=df,estimator='mean',palette='muted')
plt.title('Churn Rate by Geography')
plt.ylabel('Churn Rate (%)')
plt.show()
```

```
C:\Users\nayak\AppData\Local\Temp\ipykernel_6884\707093323.py:2: FutureWarning:
```

```
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.
```

```
sns.barplot(x='Geography',y='Exited',data=df,estimator='mean',palette='muted')
```





```
[27]: #Churn rate is highest among German Customers.
```

```
[28]: #churn Rate by Active Membership
```

```
[31]: churn_Activemembership=df.groupby('IsActiveMember')['Exited'].mean()*100  
print(churn_Activemembership)
```

```
IsActiveMember  
0    26.871520  
1    14.269074
```

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Code

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[28]: *#churn Rate by Active Membership*[31]: 

```
churn_Activemembership=df.groupby('IsActiveMember')['Exited'].mean()*100
print(churn_Activemembership)
```

```
IsActiveMember
0    26.871520
1    14.269074
Name: Exited, dtype: float64
```

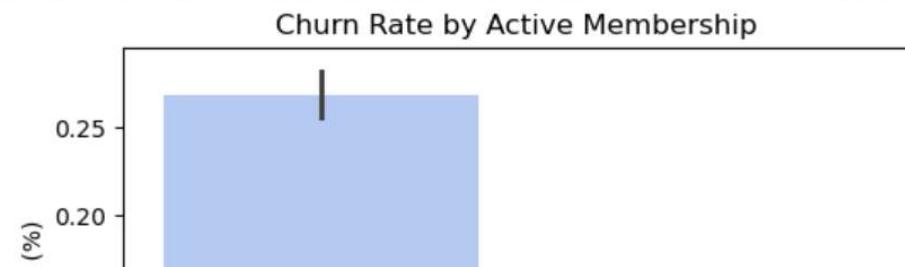
[29]: 

```
plt.figure(figsize=(6,4))
sns.barplot(x='IsActiveMember',y='Exited',data=df,estimator='mean',palette='coolwarm')
plt.title('Churn Rate by Active Membership')
plt.ylabel('Churn Rate (%)')
plt.xticks([0,1],['Inactive','Active'])
plt.show()
```

C:\Users\nayak\AppData\Local\Temp\ipykernel\_6884\2505570173.py:2: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

```
sns.barplot(x='IsActiveMember',y='Exited',data=df,estimator='mean',palette='coolwarm')
```



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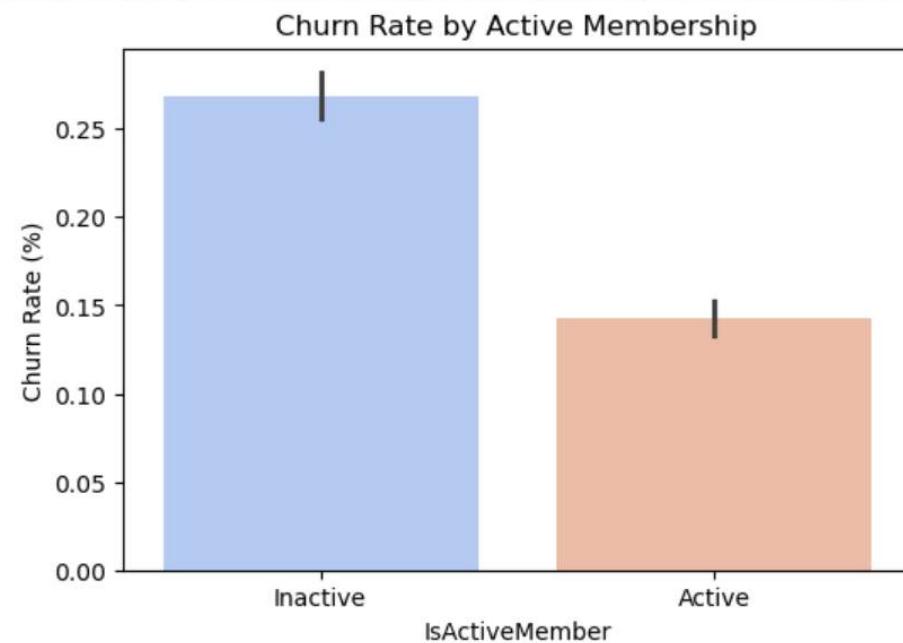
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```
plt.show()
```

```
C:\Users\nayak\AppData\Local\Temp\ipykernel_6884\2505570173.py:2: FutureWarning:
```

```
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.
```

```
sns.barplot(x='IsActiveMember',y='Exited',data=df,estimator='mean',palette='coolwarm')
```



[ ]:



[ ]:

[ ]: # 16. What is the average CreditScore, Balance, and EstimatedSalary of churned vs. retained customers?

[33]: df.head()

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101348.88	1
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	113931.57	1
3	4	15701354	Boni	699	France	Female	39	1	0.00	2	0	0	93826.63	0
4	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	79084.10	0

[34]: avg\_values=df.groupby('Exited')[['CreditScore','Balance','EstimatedSalary']].mean()  
print(avg\_values)

	CreditScore	Balance	EstimatedSalary
Exited			
0	651.837855	72742.750663	99726.853141
1	645.414622	91109.476006	101509.908783

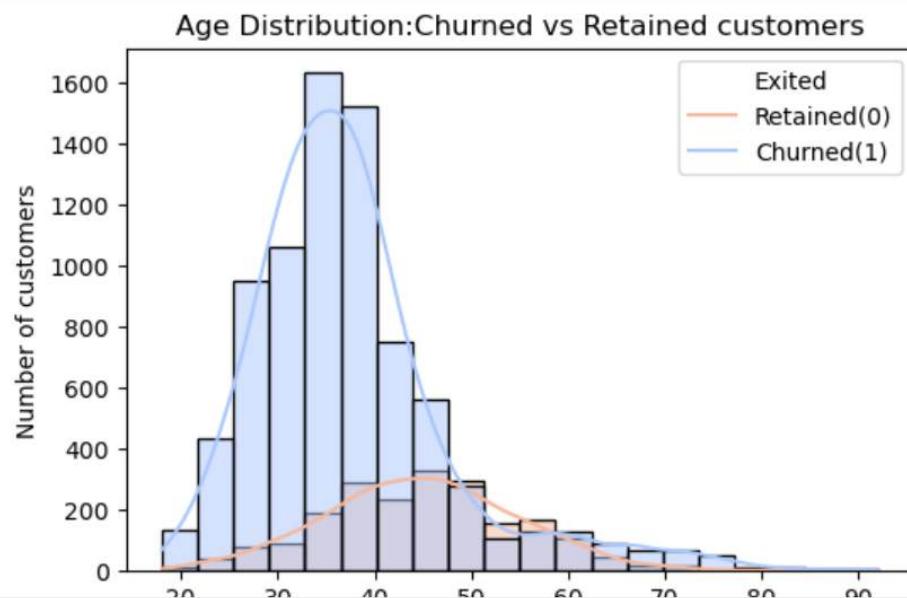
[ ]: # Note: 0 is for retained and 1 is for Churned

[ ]:

[ ]: 17. #How does Age impact churn? Plot histograms and boxplots for churned and non-churned groups.

[ ]: 17. #How does Age impact churn? Plot histograms and boxplots for churned and non-churned groups.

```
*[35]: #Histogram
plt.figure(figsize=(6,4))
sns.histplot(data=df,x='Age',hue='Exited',kde=True,bins=20,palette='coolwarm')
plt.title('Age Distribution:Churned vs Retained customers')
plt.xlabel('Age')
plt.ylabel('Number of customers')
plt.legend(title='Exited',labels=['Retained(0)', 'Churned(1)'])
plt.show()
```

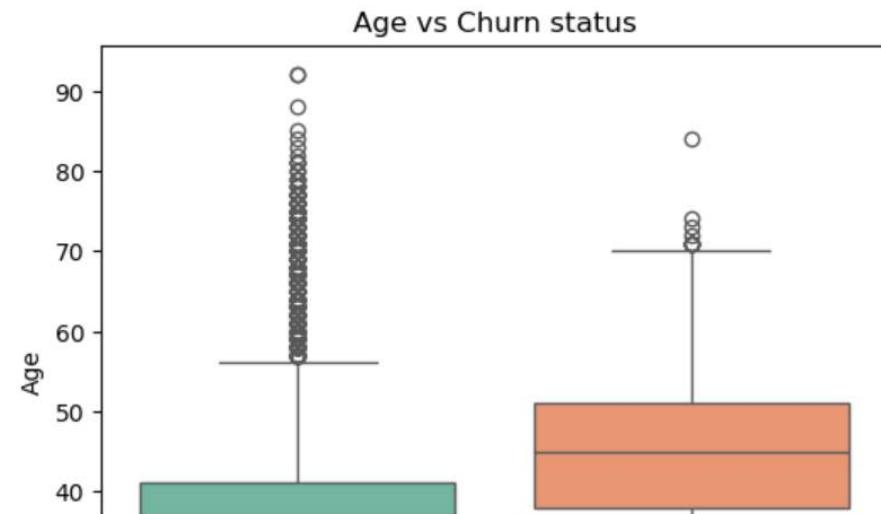


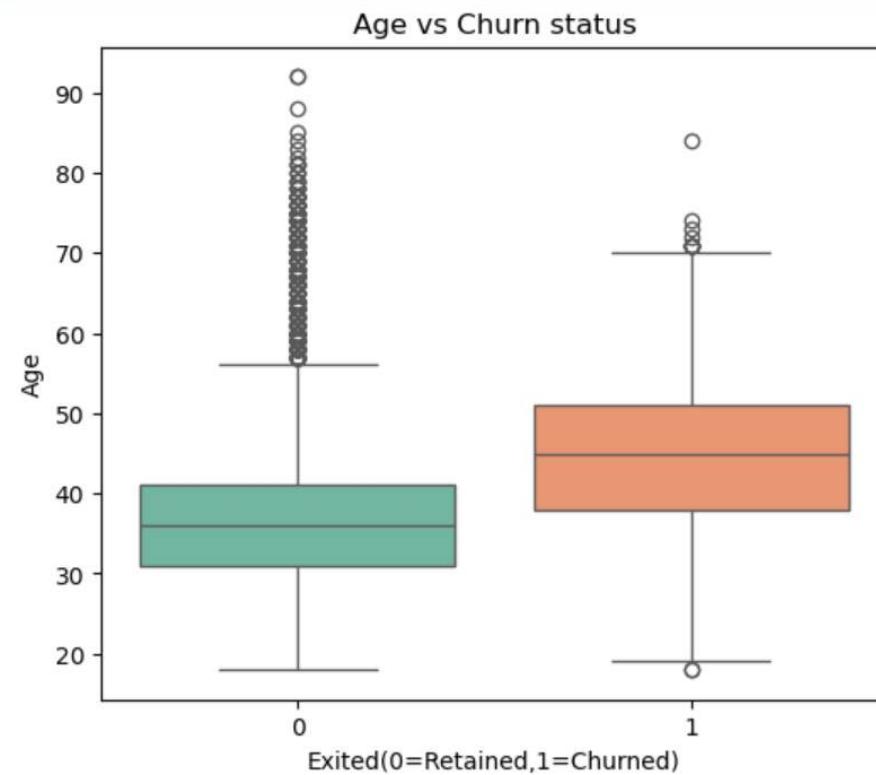
```
[37]: #Boxplot
plt.figure(figsize=(6,5))
sns.boxplot(x='Exited',y='Age',data=df,palette='Set2')
plt.title('Age vs Churn status')
plt.xlabel('Exited(0=Retained,1=Churned)')
plt.ylabel('Age')
plt.show()
```

C:\Users\nayak\AppData\Local\Temp\ipykernel\_6884\2785628990.py:3: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

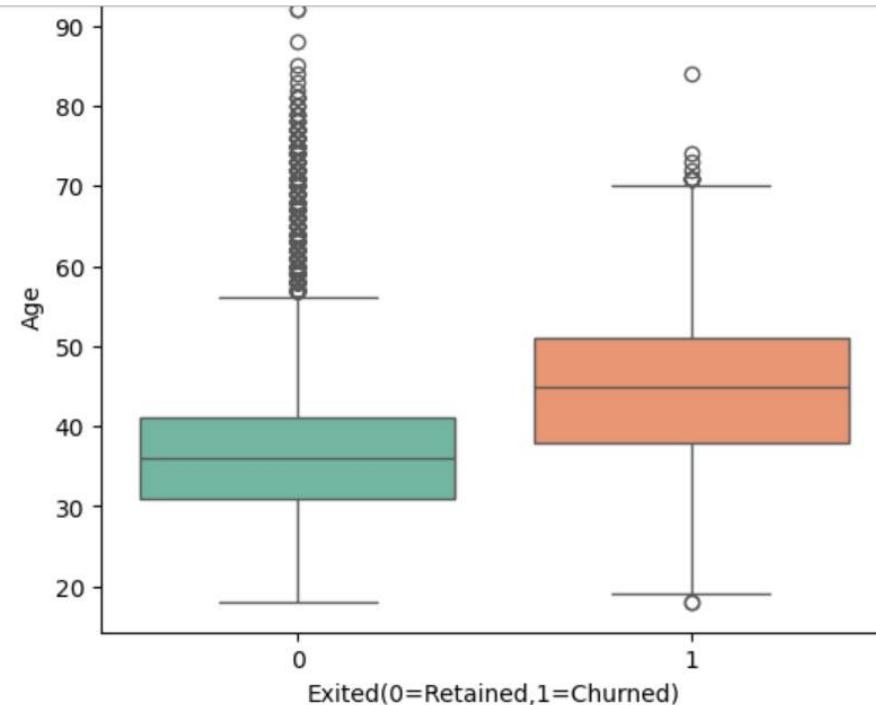
```
sns.boxplot(x='Exited',y='Age',data=df,palette='Set2')
```





```
[39]: df.groupby('Exited')['Age'].mean()
```

```
[39]: Exited
0    37.408063
1    44.835623
```



```
[39]: df.groupby('Exited')['Age'].mean()
```

```
[39]: Exited
0    37.408063
1    44.835623
Name: Age, dtype: float64
```

```
[ ]: #Churned customers have a average age og 44.8 compared to 37.4 for Retained.Older Customers are more Likely to churn.
```

[ ]: 1 44.835623  
Name: Age, dtype: float64

JupyterLab ▾ Python [conda env:base] \* ○ ≡

[ ]: #Churned customers have a average age og 44.8 compared to 37.4 for Retained.Older Customers are more likely to churn.

[ ]:

[ ]: #18. Is there any correlation among numeric features Like CreditScore, Balance, and EstimatedSalary?

[43]: #select numeric columns  
num\_cols=['CreditScore','Balance','EstimatedSalary']  
#calculate correlation matrix  
corr\_matrix=df[num\_cols].corr()  
print(corr\_matrix)

	CreditScore	Balance	EstimatedSalary
CreditScore	1.000000	0.006268	-0.001384
Balance	0.006268	1.000000	0.012797
EstimatedSalary	-0.001384	0.012797	1.000000

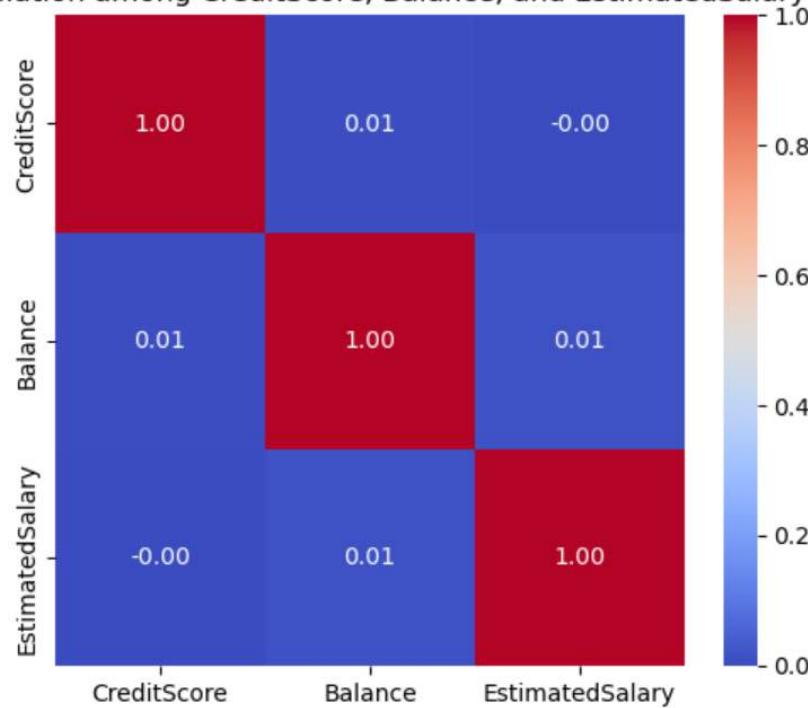
[ ]:

[44]: plt.figure(figsize=(6,5))  
sns.heatmap(corr\_matrix,annot=True,cmap='coolwarm',fmt='.2f')  
plt.title('correlation among CreditScore, Balance, and EstimatedSalary')  
plt.show()



```
[44]: plt.figure(figsize=(6,5))
sns.heatmap(corr_matrix, annot=True, cmap='coolwarm', fmt='.2f')
plt.title('correlation among CreditScore, Balance, and EstimatedSalary')
plt.show()
```

correlation among CreditScore, Balance, and EstimatedSalary



[ ]: #19. What does a heatmap reveal about feature interactions with churn?

[49]:

```
Num_cols=['CreditScore','Age','Tenure','Balance','NumOfProducts','HasCrCard','IsActiveMember','EstimatedSalary','Exited']

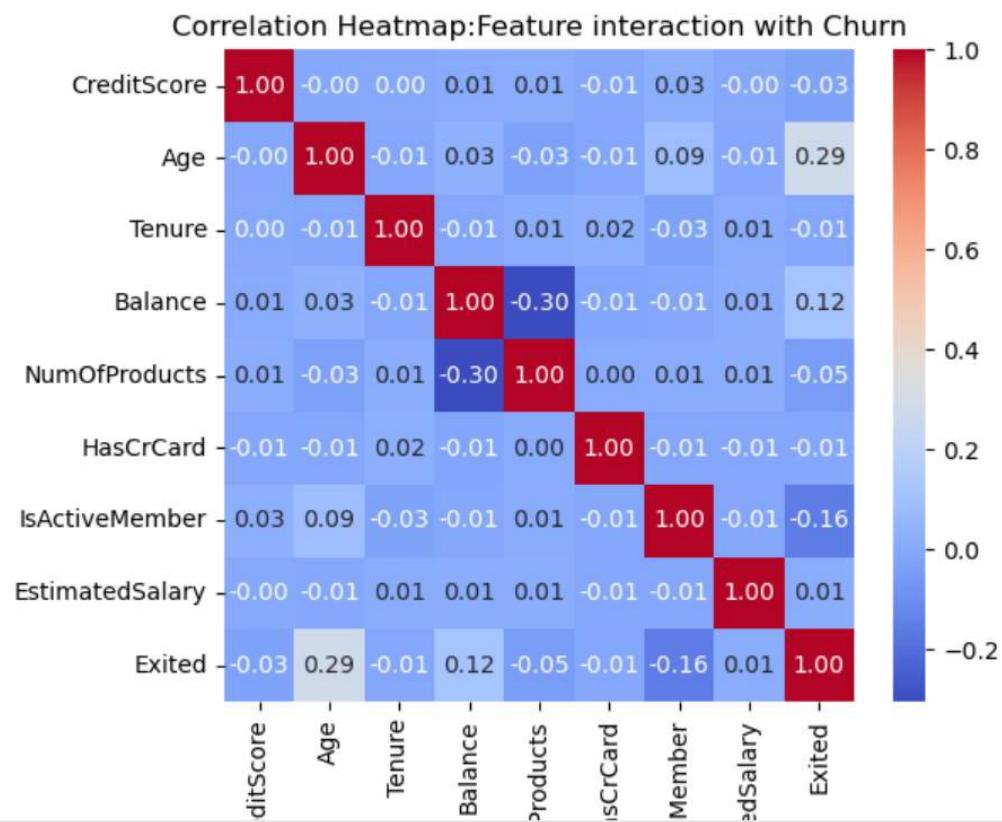
corr=df[Num_cols].corr()
print(corr)
```

	CreditScore	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited	
CreditScore	1.000000	-0.003965	0.000842	0.006268	0.012238	-0.005458	0.025651	-0.001384	-0.026771	
Age	-0.003965	1.000000	-0.009997	0.028308	-0.030680	-0.011721	0.085472	-0.007201	0.285296	
Tenure	0.000842	-0.009997	1.000000	-0.012254	0.013444	0.022583	-0.028362	0.007784	-0.013656	
Balance	0.006268	0.028308	-0.012254	1.000000	-0.304180	-0.014858	0.014204	0.012797	0.118577	
NumOfProducts	0.012238	-0.030680	0.013444	-0.304180	1.000000	0.003183	-0.011866	0.009933	-0.006976	
HasCrCard	-0.005458	-0.011721	0.022583	-0.014858	0.003183	1.000000	-0.011421	-0.011421	-0.156356	
IsActiveMember	0.025651	0.085472	-0.028362	0.010084	0.009612	1.000000	0.012490	0.012490	0.012490	
EstimatedSalary	-0.001384	-0.007201	0.007784	0.012797	0.014204	-0.009933	-0.011421	-0.011421	1.000000	
Exited	-0.026771	0.285296	-0.013656	0.118577	-0.047611	-0.011866	-0.156356	0.012490	0.012490	1.000000

[52]:

```
plt.figure(figsize=(6,5))
sns.heatmap(corr,annot=True,cmap='coolwarm',fmt='.2f')
plt.title('Correlation Heatmap:Feature interaction with Churn')
```

```
[52]: plt.figure(figsize=(6,5))
sns.heatmap(corr, annot=True, cmap='coolwarm', fmt='.2f')
plt.title('Correlation Heatmap:Feature interaction with Churn')
plt.show()
```



Credit  
Ba  
NumOfPro  
HasC  
IsActivem  
EstimatedS

```
[ ]: '''Age and Balance show a positive correlation with Churn, indicating that older customers and with higher account balance are more likely to leave the Bank. IsActivemember and Numofproducts have a negative correlation with churn, means customers who are active are less likely to Churn. EstimatedSalary and Tenure show almost no correlation with churn,meaning they have minimal influence on wheather a customer stay or leave.'''
```

```
[ ]:
```

```
[ ]: #20. Are there outliers in Balance, CreditScore, or Age that are mostly associated with churn?
```

```
[54]: numeric_cols=['CreditScore','Age','Balance']
plt.figure(figsize=(15,5))
for i, col in enumerate(numeric_cols,1):
    plt.subplot(1,3,i)
    sns.boxplot(x='Exited',y=col,data=df,palette='Set2')
    plt.title(f'cols vs Churn')
plt.tight_layout()
plt.show()
```

```
C:\Users\nayak\AppData\Local\Temp\ipykernel_6884\2120548627.py:5: FutureWarning:
```

```
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.
```

```
    sns.boxplot(x='Exited',y=col,data=df,palette='Set2')
C:\Users\nayak\AppData\Local\Temp\ipykernel_6884\2120548627.py:5: FutureWarning:
```

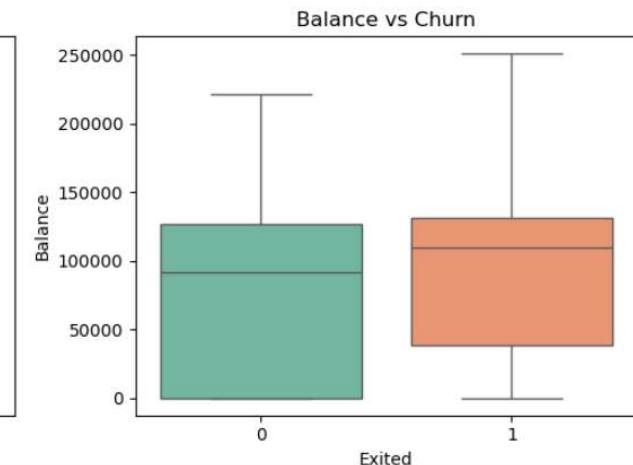
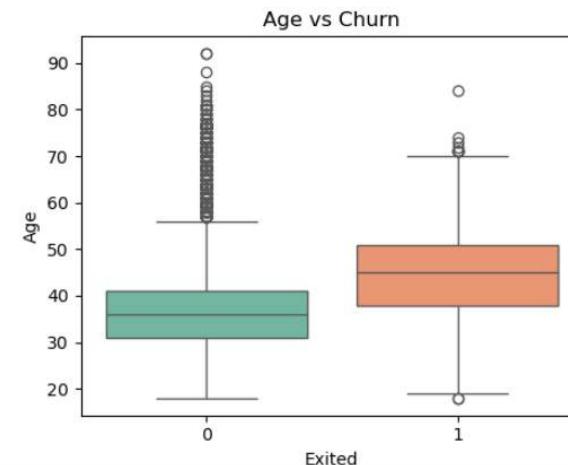
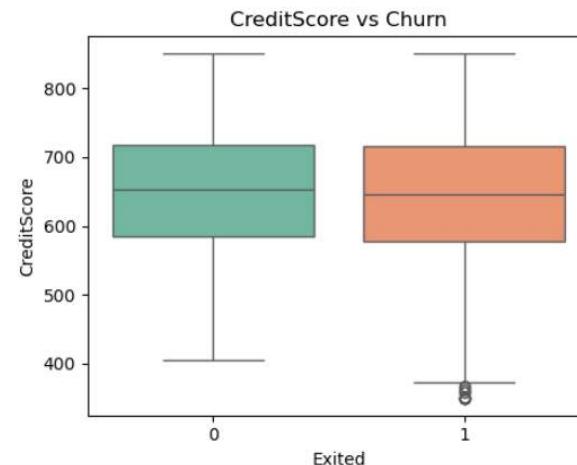
```
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.
```

JupyterLab ▾ Python [conda env:base] \* 🔍

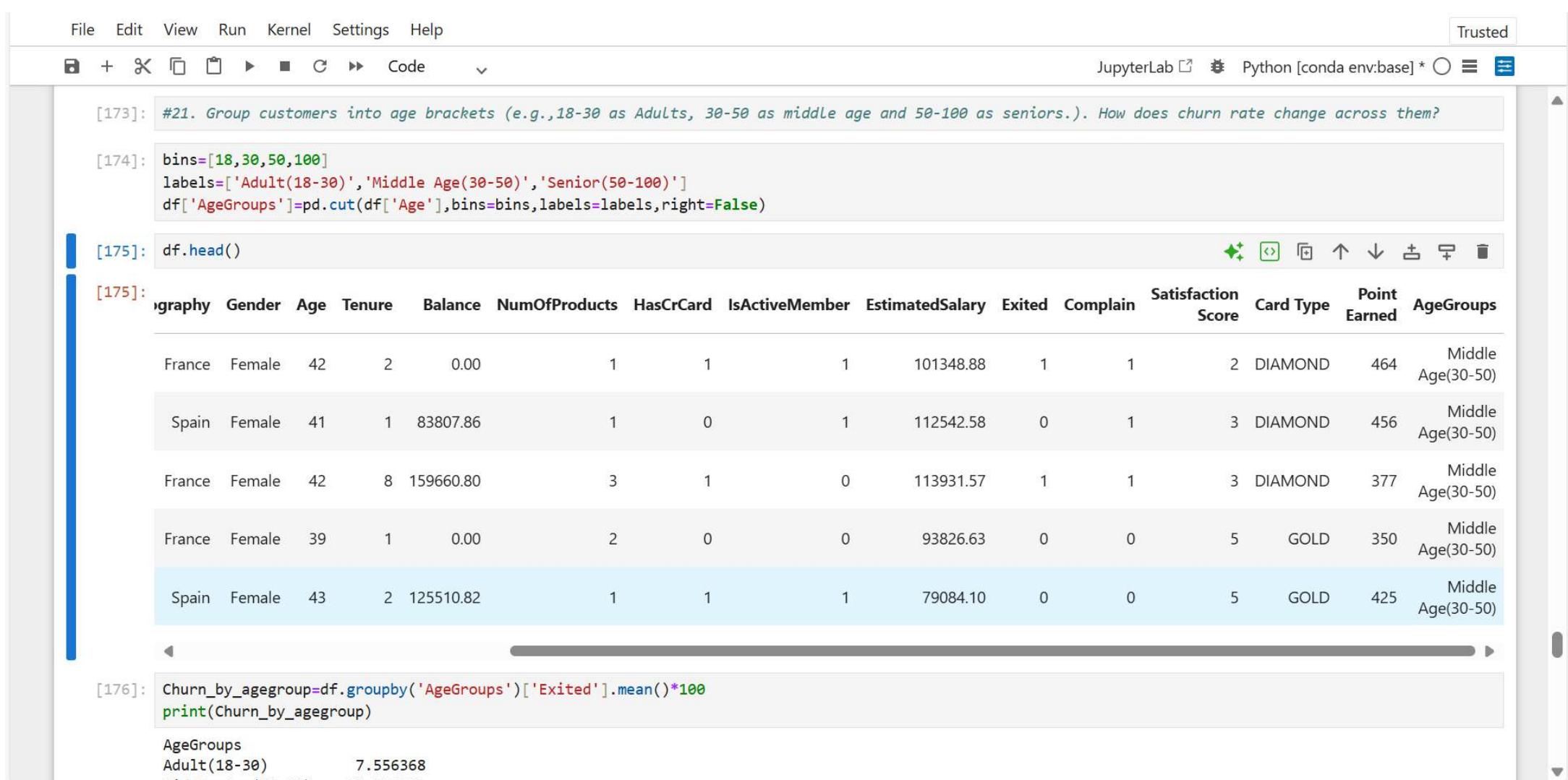
[114]: #20. Are there outliers in Balance, CreditScore, or Age that are mostly associated with churn?

```
[122]: numeric_cols=['CreditScore','Age','Balance']
plt.figure(figsize=(15,4))
for i, col in enumerate(numeric_cols,1):
    plt.subplot(1,3,i)
    sns.boxplot(x='Exited',y=col,data=df,palette='Set2')
    plt.title(f'{col} vs Churn')
plt.tight_layout()
plt.show()
```

```
sns.boxplot(x='Exited',y=col,data=df,palette='Set2')
```



[ ]:

A screenshot of a Jupyter Notebook interface. The top navigation bar includes File, Edit, View, Run, Kernel, Settings, and Help, with a Trusted badge. Below the menu is a toolbar with icons for file operations like New, Open, Save, and Run. The main area contains several code cells numbered [173] through [176]. Cell [173] contains a comment about grouping customers by age brackets. Cell [174] shows the code to create age groups using pd.cut(). Cell [175] shows the result of df.head(), displaying a sample of 5 rows from a dataset with columns like Geography, Gender, Age, Tenure, Balance, NumOfProducts, HasCrCard, IsActiveMember, EstimatedSalary, Exited, Complain, SatisfactionScore, Card Type, Point Earned, and AgeGroups. Cell [176] shows the output of a groupby operation on 'AgeGroups' to calculate the mean 'Exited' value and print it.

```
[173]: #21. Group customers into age brackets (e.g.,18-30 as Adults, 30-50 as middle age and 50-100 as seniors.). How does churn rate change across them?
```

```
[174]: bins=[18,30,50,100]
labels=['Adult(18-30)', 'Middle Age(30-50)', 'Senior(50-100)']
df['AgeGroups']=pd.cut(df['Age'],bins=bins,labels=labels,right=False)
```

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

Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited	Complain	Satisfaction Score	Card Type	Point Earned	AgeGroups
France	Female	42	2	0.00		1	1	101348.88	1	1	2	DIAMOND	464	Middle Age(30-50)
Spain	Female	41	1	83807.86		1	0	112542.58	0	1	3	DIAMOND	456	Middle Age(30-50)
France	Female	42	8	159660.80		3	1	113931.57	1	1	3	DIAMOND	377	Middle Age(30-50)
France	Female	39	1	0.00		2	0	93826.63	0	0	5	GOLD	350	Middle Age(30-50)
Spain	Female	43	2	125510.82		1	1	79084.10	0	0	5	GOLD	425	Middle Age(30-50)

```
[176]: Churn_by_agegroup=df.groupby('AgeGroups')['Exited'].mean()*100
print(Churn_by_agegroup)
```

```
AgeGroups
Adult(18-30)          7.556368
Middle Age(30-50)     10.292711
```

```
[176]: Churn_by_agegroup=df.groupby('AgeGroups')['Exited'].mean()*100
print(Churn_by_agegroup)
```

```
AgeGroups
Adult(18-30)      7.556368
Middle Age(30-50) 18.380241
Senior(50-100)    45.448029
Name: Exited, dtype: float64
```

C:\Users\nayak\AppData\Local\Temp\ipykernel\_6884\2116857766.py:1: FutureWarning: The default of observed=False is deprecated and will be changed to True in a future version of pandas. Pass observed=False to retain current behavior or observed=True to adopt the future default and silence this warning.

```
Churn_by_agegroup=df.groupby('AgeGroups')['Exited'].mean()*100
```

```
[177]: #Visulize churn rate by Age groups
```

```
[178]: plt.figure(figsize=(6,4))
sns.barplot(x=Churn_by_agegroup.index,y=Churn_by_agegroup.values,palette='Set2')
plt.title('Churn Rate by AgeGroup')
plt.ylabel('Churn Rate(%)')
plt.xlabel('Age Group')
plt.show()
```

C:\Users\nayak\AppData\Local\Temp\ipykernel\_6884\1509556778.py:2: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

```
sns.barplot(x=Churn_by_agegroup.index,y=Churn_by_agegroup.values,palette='Set2')
```

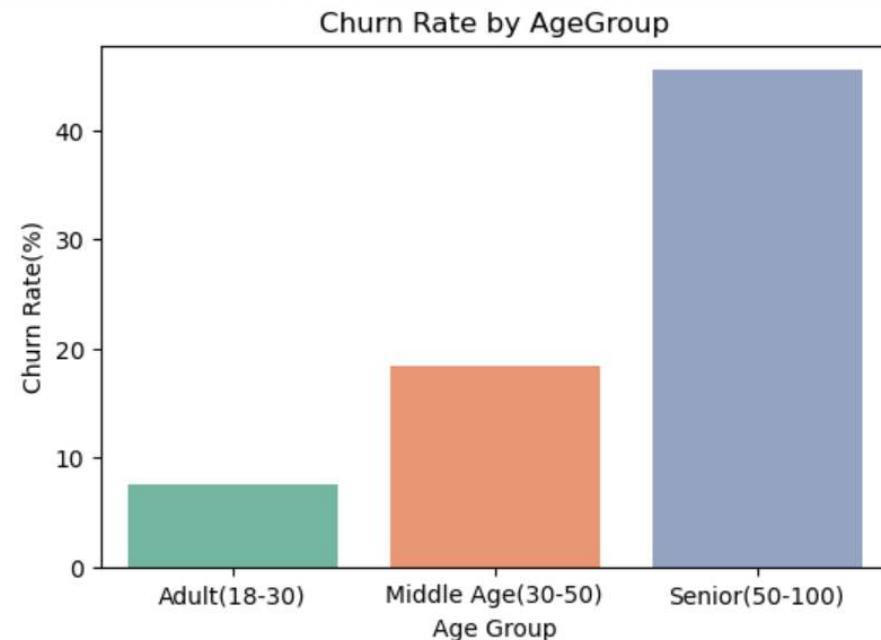
Churn Rate by AgeGroup

plt.show()

C:\Users\nayak\AppData\Local\Temp\ipykernel\_6884\1509556778.py:2: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

sns.barplot(x=Churn\_by\_agegroup.index,y=Churn\_by\_agegroup.values,palette='Set2')



[ ]:

[ ]:

```
[179]: #22. Are customers with only one product (NumOfProducts = 1) more likely to churn than those with multiple?
```

```
[180]: Churn_by_products=df.groupby('NumOfProducts')['Exited'].mean()*100  
print(Churn_by_products)
```

```
NumOfProducts  
1    27.714398  
2     7.603486  
3    82.706767  
4   100.000000  
Name: Exited, dtype: float64
```

```
[181]: plt.figure(figsize=(6,4))  
sns.barplot(x=Churn_by_products.index,y=Churn_by_products.values,palette='coolwarm')  
plt.title('Churn Rate by Number of Products')  
plt.xlabel('Number of Products')  
plt.xlabel('Churn Rate(%)')  
plt.show()
```

C:\Users\nayak\AppData\Local\Temp\ipykernel\_6884\2976775891.py:2: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

```
sns.barplot(x=Churn_by_products.index,y=Churn_by_products.values,palette='coolwarm')
```

Churn Rate by Number of Products



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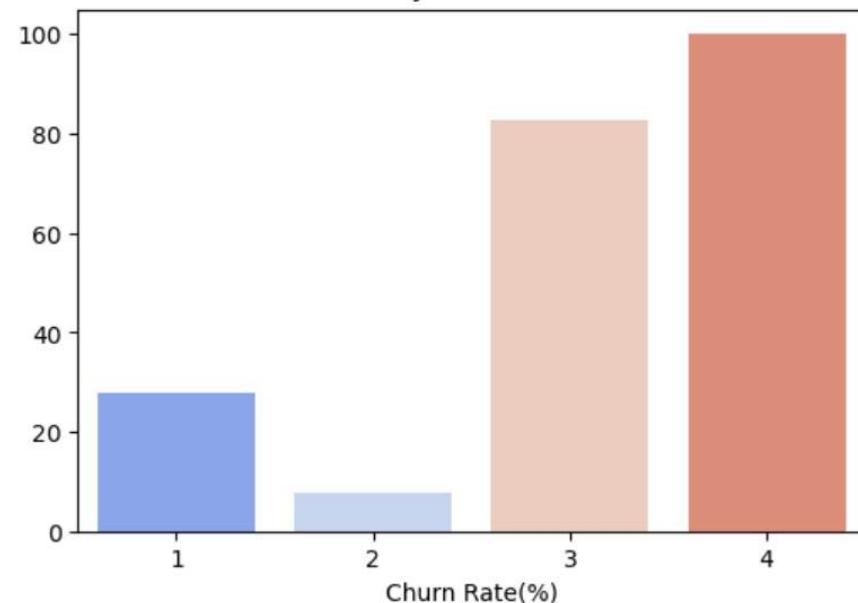
Trusted

Code

JupyterLab Python [conda env:base] \*

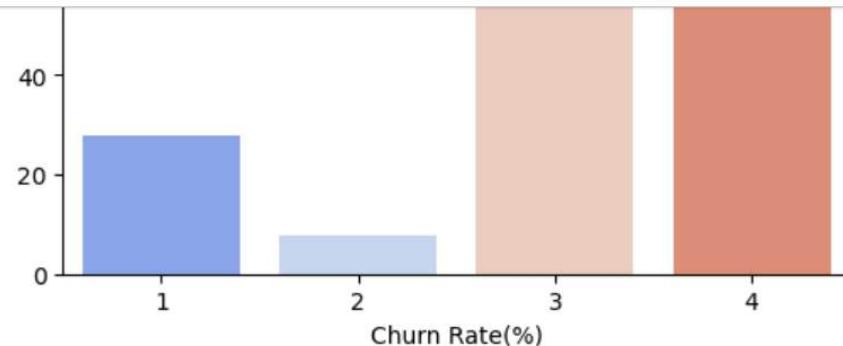
```
sns.barplot(x=Churn_by_products.index,y=Churn_by_products.values,palette='coolwarm')
```

Churn Rate by Number of Products



```
[182]: df['NumOfProducts'].value_counts()
```

```
[182]: NumOfProducts
1    5084
2    4590
3    266
4     60
Name: count, dtype: int64
```



```
[182]: df['NumOfProducts'].value_counts()
```

```
[182]: NumOfProducts
1    5084
2    4590
3     266
4      60
Name: count, dtype: int64
```

```
[184]: '''ChurnRate decreases from 1 to 2 products but spikes sharply for 3 and 4 products.That means moderate engagement(2 products) promotes loyalty.  
and under-engagement(1 product)and over-engagement(3 and 4 products)may lead to customer dissatisfaction.'''
```

```
[184]: 'ChurnRate decreases from 1 to 2 products but spikes sharply for 3 and 4 products.That means moderate engagement(2 products) promotes loyalty.\nand under-engagement(1 product)and over-engagement(3 and 4 products)may lead to customer dissatisfaction.'
```

```
[ ]:
```

```
[ ]:
```

```
[ ]:
```