

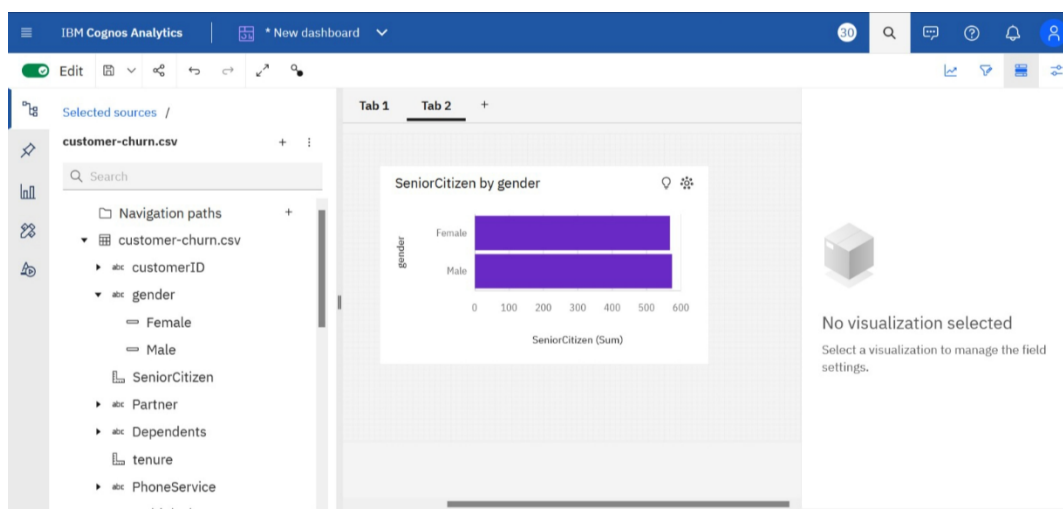
Phase 4

Customer Churn Prediction

Using IBM Cognos for data visualization with a dataset (Customer Churn Prediction) is a straightforward process that can provide valuable insights for businesses. First, you can import your dataset into Cognos, which supports various data sources, such as spreadsheets, databases, and even cloud-based data. Once your data is loaded, you can begin creating compelling visualizations. Cognos offers a wide array of visualization options, from basic bar charts and line graphs to more advanced options like heat maps, geographical maps, and interactive dashboards.

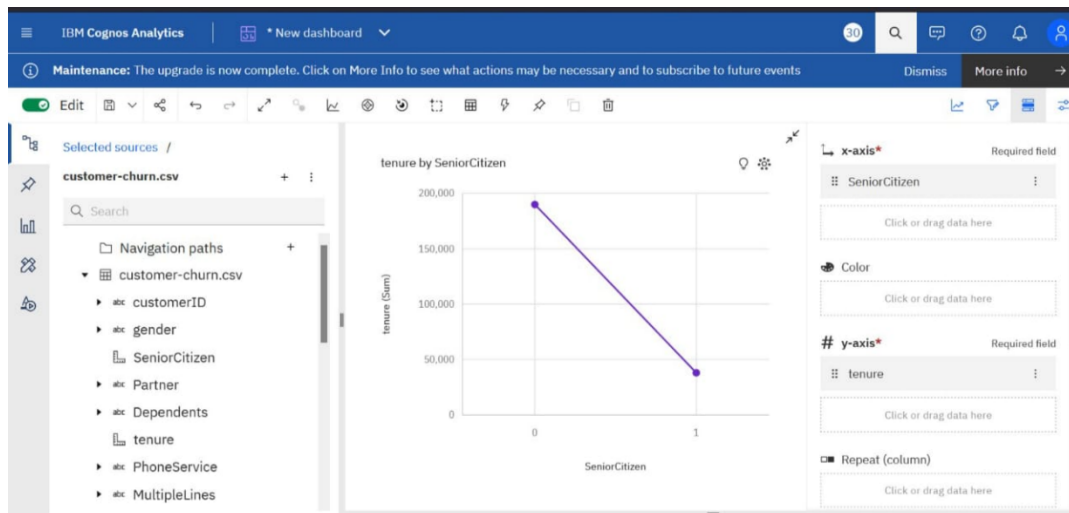
Create interactive dashboards and reports in IBM Cognos to visualize churn patterns, retention rates, and key factors influencing churn.

1.Bar Chart: A bar chart or bar graph is a chart or graph that presents categorical data with rectangular bars with heights or lengths proportional to the values that they represent. The bars can be plotted vertically or horizontally. A vertical bar chart is sometimes called a column chart.

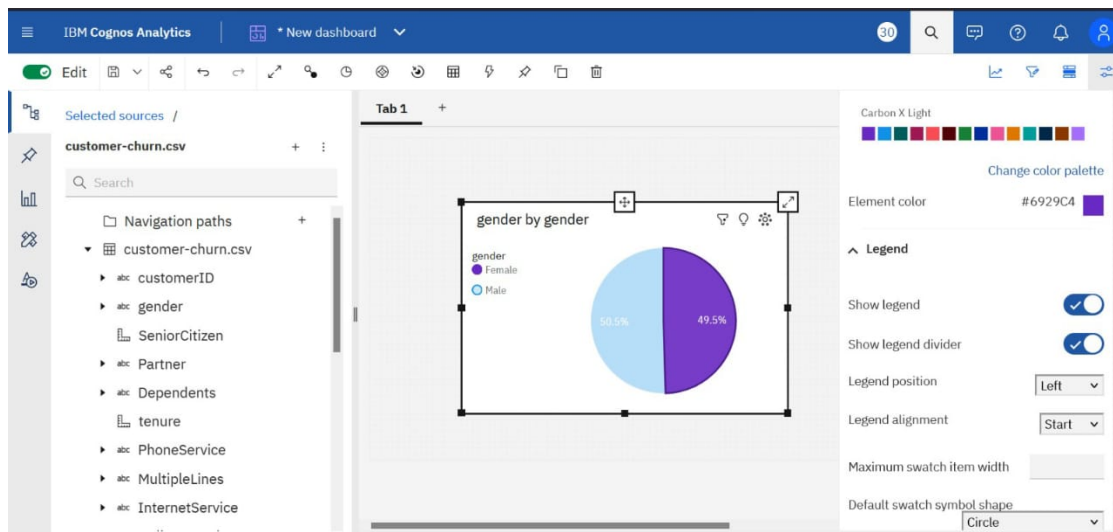


2. Line Chart: A line chart is a type of chart used to show information that changes over time. Line charts are created by plotting a series of several points

and connecting them with a straight line. Line charts are used to track changes over short and long periods.

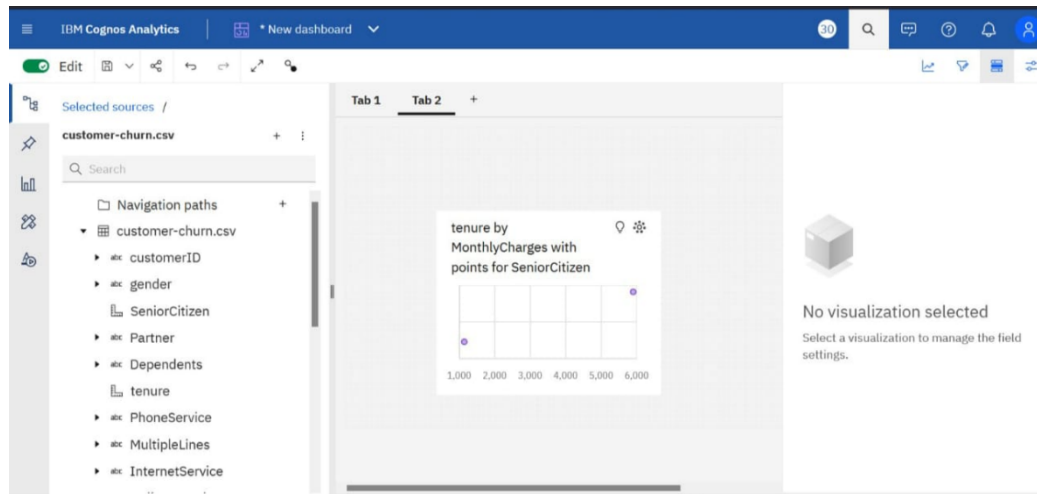


3.Pie Chart: A pie chart is a type of graph that represents the data in the circular graph. The slices of pie show the relative size of the data, and it is a type of pictorial representation of data. A pie chart requires a list of categorical variables and numerical variables. Here, the term “pie” represents the whole, and the “slices” represent the parts of the whole.

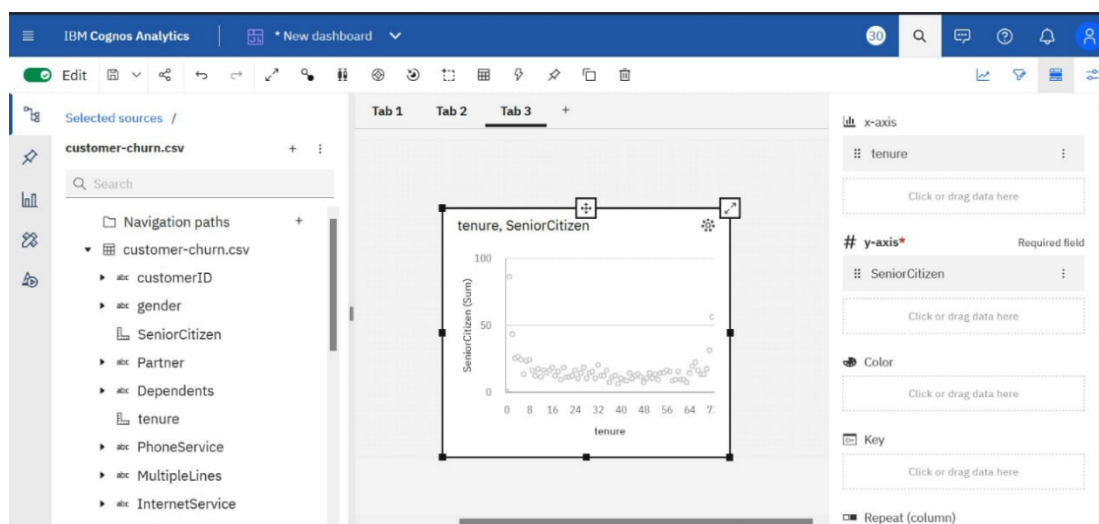


4.Box Plot: A box and whisker plot—also called a box plot—displays the five-number summary of a set of data. The five-number summary is the minimum, first quartile, median, third quartile, and maximum.

In a box plot, we draw a box from the first quartile to the third quartile. A vertical line goes through the box at the median. The whiskers go from each quartile to the minimum or maximum.



5. Scatter Plot: A scatter plot (aka scatter chart, scatter graph) uses dots to represent values for two different numeric variables. The position of each dot on the horizontal and vertical axis indicates values for an individual data point. Scatter plots are used to observe relationships between variables.



Use machine learning algorithms to build a predictive model that identifies potential churners based on historical data and relevant features.

1.importing the modules and packages.

```
[3] import pandas as pd
import numpy as np
import seaborn as sns
import plotly.express as px
import plotly.graph_objects as go
from plotly.subplots import make_subplots
import warnings
warnings.filterwarnings('ignore')

from sklearn.preprocessing import LabelEncoder
from sklearn.preprocessing import StandardScaler

from sklearn.model_selection import train_test_split
from sklearn.neighbors import KNeighborsClassifier
from sklearn.preprocessing import StandardScaler
from sklearn.preprocessing import LabelEncoder

from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.naive_bayes import GaussianNB
from sklearn.neighbors import KNeighborsClassifier
from sklearn.svm import SVC
from sklearn.neural_network import MLPClassifier
from sklearn.ensemble import AdaBoostClassifier
from sklearn.ensemble import GradientBoostingClassifier
from sklearn.ensemble import ExtraTreesClassifier
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score
from xgboost import XGBClassifier

from sklearn import metrics
from sklearn.metrics import roc_curve
from sklearn.metrics import recall_score, confusion_matrix, precision_score, f1_score, accuracy_score, classification_report

from sklearn.ensemble import VotingClassifier
```

```
[4] from sklearn.metrics import confusion_matrix, accuracy_score
from sklearn.metrics import f1_score, precision_score, recall_score, fbeta_score
from statsmodels.stats.outliers_influence import variance_inflation_factor
from sklearn.model_selection import cross_val_score
from sklearn.model_selection import GridSearchCV
from sklearn.model_selection import ShuffleSplit
from sklearn.model_selection import KFold
from sklearn import feature_selection
from sklearn import model_selection
from sklearn import metrics
from sklearn.metrics import classification_report, precision_recall_curve
from sklearn.metrics import auc, roc_auc_score, roc_curve
from sklearn.metrics import make_scorer, recall_score, log_loss
from sklearn.metrics import average_precision_score
```

```
data = pd.read_csv("/content/customer-churn.csv")
data.head()
```

	customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	InternetService	OnlineSecurity	...	DeviceProtection	TechSupport	StreamingTV	Streaming
0	7590-VHVEG	Female	0	Yes	No	1	No	No phone service	DSL	No	...	No	No	No	No
1	5575-GNVDE	Male	0	No	No	34	Yes	No	DSL	Yes	...	Yes	No	No	No
2	3668-QPYBK	Male	0	No	No	2	Yes	No	DSL	Yes	...	No	No	No	No
3	7795-CFOCW	Male	0	No	No	45	No	No phone service	DSL	Yes	...	Yes	Yes	No	No
4	9237-HQITU	Female	0	No	No	2	Yes	No	Fiber optic	No	...	No	No	No	No

2.getting the information of the dataset

```
data.info()

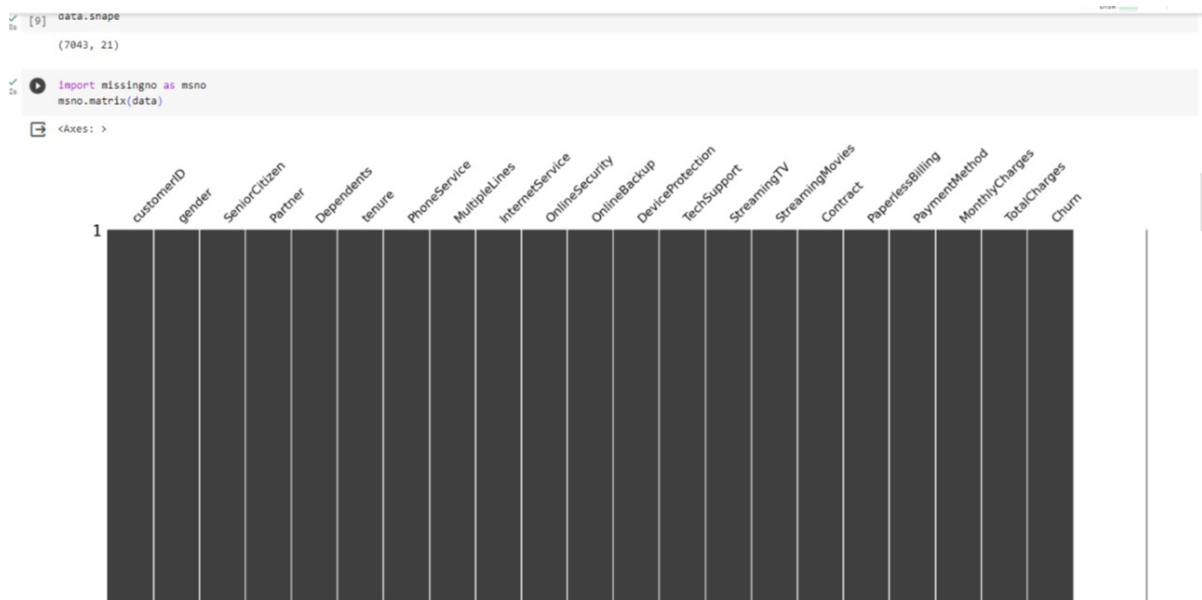
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 7043 entries, 0 to 7042
Data columns (total 21 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0   customerID            7043 non-null   object
 1   gender                7043 non-null   object
 2   SeniorCitizen         7043 non-null   int64
 3   Partner               7043 non-null   object
 4   Dependents            7043 non-null   object
 5   tenure                7043 non-null   int64
 6   PhoneService          7043 non-null   object
 7   MultipleLines         7043 non-null   object
 8   InternetService       7043 non-null   object
 9   OnlineSecurity        7043 non-null   object
10   OnlineBackup          7043 non-null   object
11   DeviceProtection      7043 non-null   object
12   TechSupport          7043 non-null   object
13   StreamingTV           7043 non-null   object
14   StreamingMovies       7043 non-null   object
15   Contract              7043 non-null   object
16   PaperlessBilling      7043 non-null   object
17   PaymentMethod         7043 non-null   object
18   MonthlyCharges        7043 non-null   float64
19   TotalCharges          7043 non-null   object
20   Churn                 7043 non-null   object
dtypes: float64(1), int64(2), object(18)
memory usage: 1.1+ MB

[9] data.shape

(7043, 21)

[10] import missingno as msno
     msno.matrix(data)
```

3.creating the matrix of the dataset



4.Analysing our dataset with the visualization of pie chart model.

```

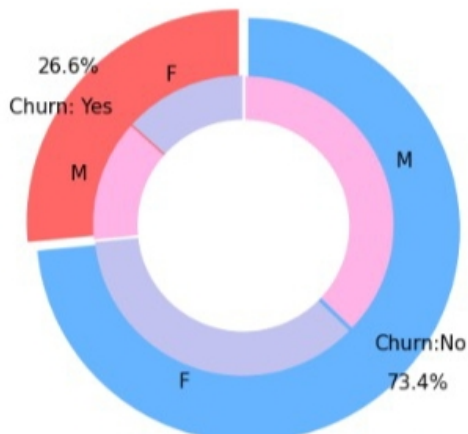
plt.figure(figsize=(6, 6))
labels = ["Churn: Yes", "Churn:No"]
values = [1869, 5163]
labels_gender = ["F", "M", "F", "M"]
sizes_gender = [939, 930, 2544, 2619]
colors = ['#ff6666', '#66b3ff']
colors_gender = ['#c2c2f0', '#ffb3e6', '#c2c2f0', '#ffb3e6']
explode = (0.3, 0.3)
explode_gender = (0.1, 0.1, 0.1, 0.1)
textprops = {"fontsize": 15}
#Plot
plt.pie(values, labels=labels, autopct='%1.1f%%', pctdistance=1.08, labeldistance=0.8, colors=colors, startangle=90, frame=True)
plt.pie(sizes_gender, labels=labels_gender, colors=colors_gender, startangle=90, explode=explode_gender, radius=7, textprops=textprops)
#Draw circle
centre_circle = plt.Circle((0,0),5,color='black', fc='white',linewidth=0)
fig = plt.gcf()
fig.gca().add_artist(centre_circle)

plt.title('Churn Distribution w.r.t Gender: Male(M), Female(F)', fontsize=15, y=1.1)

# show plot
plt.axis('equal')
plt.tight_layout()
plt.show()

```

Churn Distribution w.r.t Gender: Male(M), Female(F)



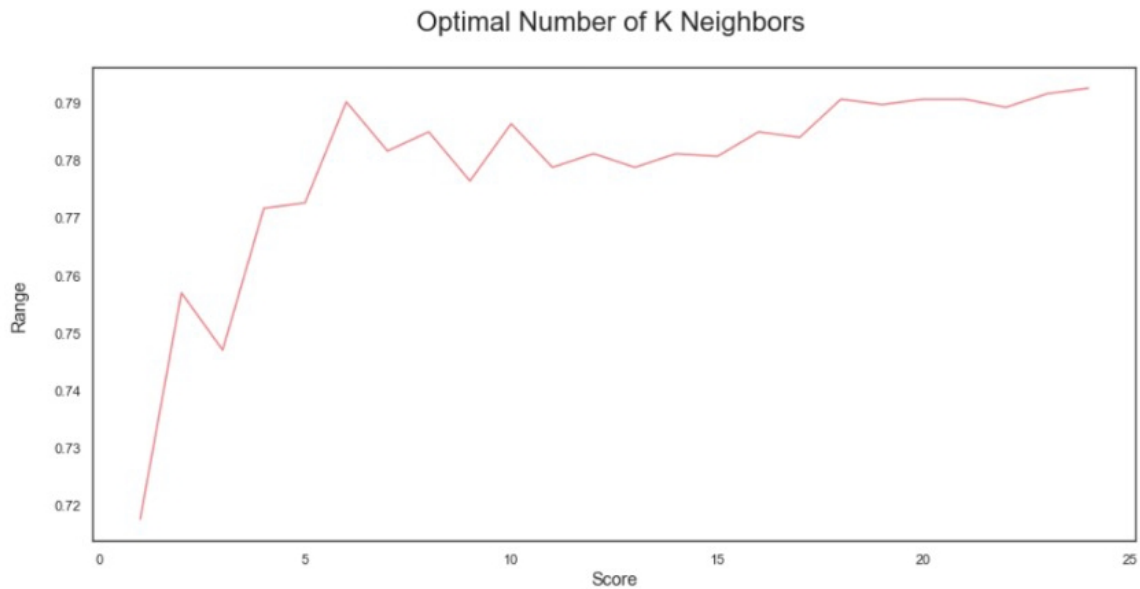
5. The k-Nearest Neighbors (KNN) algorithm is a versatile and simple machine learning model that can be applied to various classification and regression tasks. It operates on the principle of similarity, where it classifies or predicts based on the majority class or average value of its k-nearest data points in the feature space. Using the k-nearest the dataset of the customer churn is predicted.

9...

```
fig = plt.figure(figsize=(15, 7))
plt.plot(range(1,25),score_array, color = '#ec838a')
plt.ylabel('Range\n',horizontalalignment="center",fontstyle = "normal", fontsize = "large", fontfamily = "sans-serif")
plt.xlabel('Score\n',horizontalalignment="center",fontstyle = "normal", fontsize = "large", fontfamily = "sans-serif")

plt.title('Optimal Number of K Neighbors \n',horizontalalignment="center", fontstyle = "normal",fontsize = "22", fontfami
#plt.legend(loc='top right', fontsize = "medium")

plt.xticks(rotation=0, horizontalalignment="center")
plt.yticks(rotation=0, horizontalalignment="right")
plt.show()
```



6. Random Forest is a popular ensemble machine learning algorithm that is widely used for both classification and regression tasks. It's considered one of the most powerful and versatile algorithms in the field of machine learning.

```

fig = plt.figure(figsize=(15, 7))
plt.plot(range(1,100),score_array, color = '#ec838a')
plt.ylabel('Range\n',horizontalalignment="center",
fontstyle = "normal", fontsize = "large",
fontfamily = "sans-serif")
plt.xlabel('Score\n',horizontalalignment="center",
fontstyle = "normal", fontsize = "large",
fontfamily = "sans-serif")
plt.title('Optimal Number of Trees for Random Forest Model \n',horizontalalignment="center", fontstyle = "normal", fontsi
#plt.legend(loc='top right', fontsize = "medium")
plt.xticks(rotation=0, horizontalalignment="center")
plt.yticks(rotation=0, horizontalalignment="right")
plt.show()

```

Optimal Number of Trees for Random Forest Model

