

```
In [7]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
import matplotlib.ticker as mtick
```

```
In [9]: telecom_cust = pd.read_csv("C:\\Users\\kankk\\OneDrive\\Desktop\\WA_Fn-UseC_-Telco-Customer-Churn.csv")
```

```
In [10]: telecom_cust.head(10)
```

```
Out[10]:
```

	customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	InternetService	OnlineSecurity	...	DevicePr
0	7590-VHVEG	Female	0	Yes	No	1	No	No phone service	DSL	No	...	
1	5575-GNVDE	Male	0	No	No	34	Yes	No	DSL	Yes	...	
2	3668-QPYBK	Male	0	No	No	2	Yes	No	DSL	Yes	...	
3	7795-CFOCW	Male	0	No	No	45	No	No phone service	DSL	Yes	...	
4	9237-HQITU	Female	0	No	No	2	Yes	No	Fiber optic	No	...	
5	9305-CDSKC	Female	0	No	No	8	Yes	Yes	Fiber optic	No	...	
6	1452-KIOVK	Male	0	No	Yes	22	Yes	Yes	Fiber optic	No	...	
7	6713-OKOMC	Female	0	No	No	10	No	No phone service	DSL	Yes	...	
8	7892-POOKP	Female	0	Yes	No	28	Yes	Yes	Fiber optic	No	...	
9	6388-TABGU	Male	0	No	Yes	62	Yes	No	DSL	Yes	...	

10 rows × 21 columns

```
In [11]: telecom_cust.columns.values
```

```
Out[11]: array(['customerID', 'gender', 'SeniorCitizen', 'Partner', 'Dependents',
        'tenure', 'PhoneService', 'MultipleLines', 'InternetService',
        'OnlineSecurity', 'OnlineBackup', 'DeviceProtection',
        'TechSupport', 'StreamingTV', 'StreamingMovies', 'Contract',
        'PaperlessBilling', 'PaymentMethod', 'MonthlyCharges',
        'TotalCharges', 'Churn'], dtype=object)
```

```
In [12]: # Checking the data types of all the columns
telecom_cust.dtypes
```

```
Out[12]: customerID      object
gender      object
SeniorCitizen  int64
Partner      object
Dependents    object
tenure      int64
PhoneService  object
MultipleLines object
InternetService object
OnlineSecurity object
OnlineBackup  object
DeviceProtection object
TechSupport   object
StreamingTV   object
StreamingMovies object
Contract      object
PaperlessBilling object
PaymentMethod object
MonthlyCharges float64
TotalCharges  object
Churn         object
dtype: object
```

```
In [13]: # Converting Total Charges to a numerical data type.
telecom_cust.TotalCharges = pd.to_numeric(telecom_cust.TotalCharges, errors='coerce')
telecom_cust.isnull().sum()
```

```
Out[13]: 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   11
Churn         0
dtype: int64
```

```
In [14]: #Removing missing values
telecom_cust.dropna(inplace = True)
#Remove customer IDs from the data set
df2 = telecom_cust.iloc[:,1:]
#Convertin the predictor variable in a binary numeric variable
df2['Churn'].replace(to_replace='Yes', value=1, inplace=True)
df2['Churn'].replace(to_replace='No', value=0, inplace=True)

#Let's convert all the categorical variables into dummy variables
df_dummies = pd.get_dummies(df2)
df_dummies.head()
```

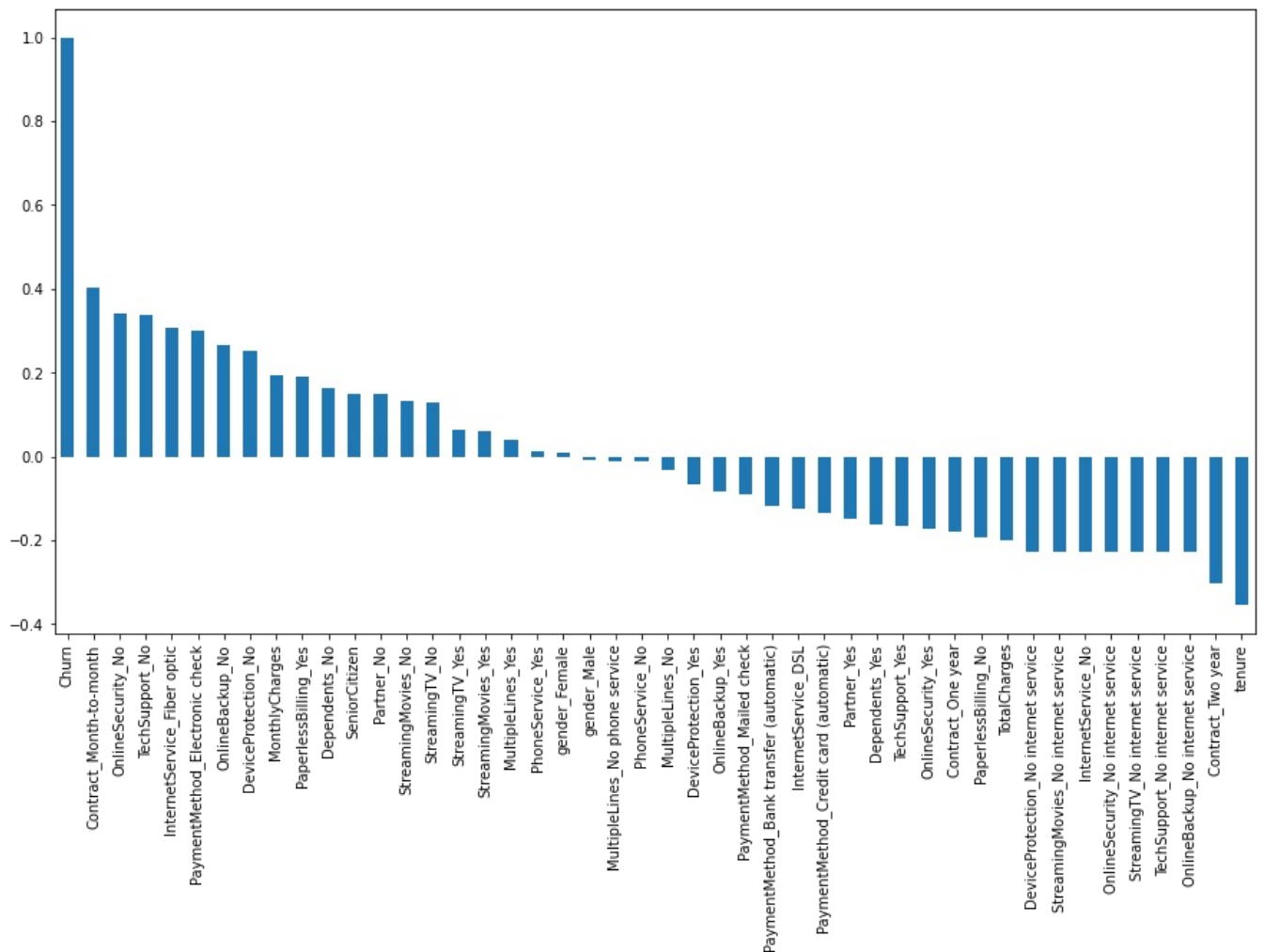
```
Out[14]:
```

	SeniorCitizen	tenure	MonthlyCharges	TotalCharges	Churn	gender_Female	gender_Male	Partner_No	Partner_Yes	Dependents_No	...	!
0	0	1	29.85	29.85	0	1	0	0	1	1	...	
1	0	34	56.95	1889.50	0	0	1	1	0	1	...	
2	0	2	53.85	108.15	1	0	1	1	0	1	...	
3	0	45	42.30	1840.75	0	0	1	1	0	1	...	
4	0	2	70.70	151.65	1	1	0	1	0	1	...	

5 rows × 46 columns

```
In [15]: #Get Correlation of "Churn" with other variables:
plt.figure(figsize=(15,8))
df_dummies.corr()['Churn'].sort_values(ascending = False).plot(kind='bar')
```

```
Out[15]: <AxesSubplot:>
```



```
In [16]: colors = ['brown', 'pink']
ax = (telecom_cust['gender'].value_counts()*100.0 / len(telecom_cust)).plot(kind='bar',
                                     stacked = True,
                                     rot = 0,
                                     color = colors)

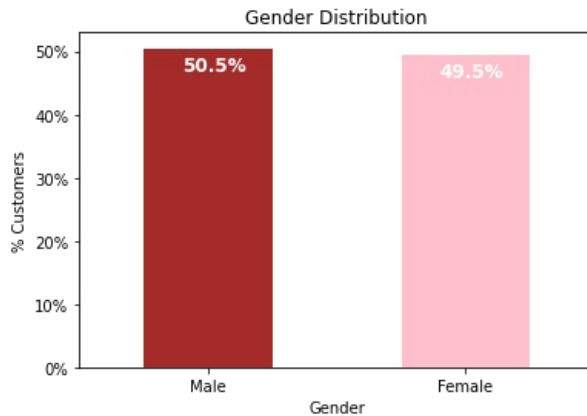
ax.yaxis.set_major_formatter(mtick.PercentFormatter())
ax.set_ylabel('% Customers')
ax.set_xlabel('Gender')
ax.set_ylabel('% Customers')
ax.set_title('Gender Distribution')

# create a list to collect the plt.patches data
totals = []
```

```
# find the values and append to list
for i in ax.patches:
    totals.append(i.get_width())

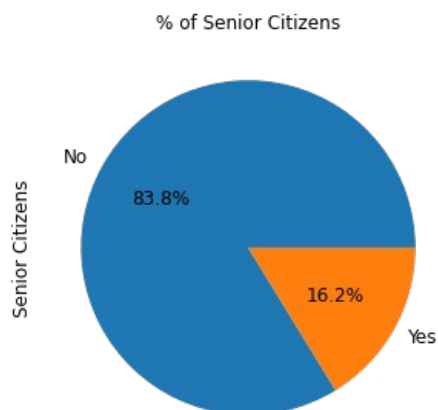
# set individual bar labels using above list
total = sum(totals)

for i in ax.patches:
    # get_width pulls left or right; get_y pushes up or down
    ax.text(i.get_x()+.15, i.get_height()-3.5, \
            str(round((i.get_height()/total), 1))+ '%',
            fontsize=12,
            color='white',
            weight = 'bold')
```



```
In [17]: ax = (telecom_cust['SeniorCitizen'].value_counts()*100.0 /len(telecom_cust))\
.plot.pie(autopct='%.1f%%', labels = ['No', 'Yes'],figsize =(5,5), fontsize = 12 )
ax.yaxis.set_major_formatter(mtick.PercentFormatter())
ax.set_ylabel('Senior Citizens',fontsize = 12)
ax.set_title('% of Senior Citizens', fontsize = 12)
```

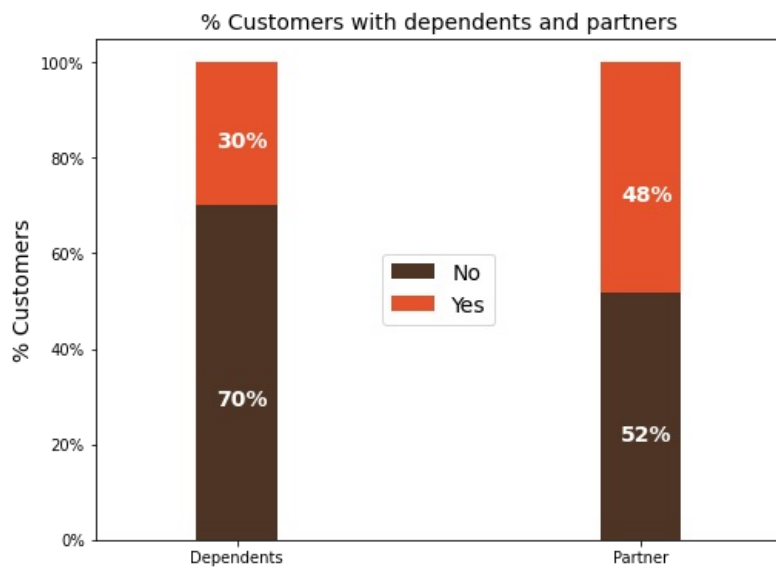
```
Out[17]: Text(0.5, 1.0, '% of Senior Citizens')
```



```
In [18]: df2 = pd.melt(telecom_cust, id_vars=['customerID'], value_vars=['Dependents','Partner'])
df3 = df2.groupby(['variable','value']).count().unstack()
df3 = df3*100/len(telecom_cust)
colors = ['#4D3425', '#E4512B']
ax = df3.loc[:, 'customerID'].plot.bar(stacked=True, color=colors,
                                     figsize=(8,6), rot = 0,
                                     width = 0.2)

ax.yaxis.set_major_formatter(mtick.PercentFormatter())
ax.set_ylabel('% Customers',size = 14)
ax.set_xlabel('')
ax.set_title('% Customers with dependents and partners',size = 14)
ax.legend(loc = 'center',prop={'size':14})

for p in ax.patches:
    width, height = p.get_width(), p.get_height()
    x, y = p.get_xy()
    ax.annotate('{:.0f}%'.format(height), (p.get_x()+.25*width, p.get_y()+.4*height),
               color = 'white',
               weight = 'bold',
               size = 14)
```

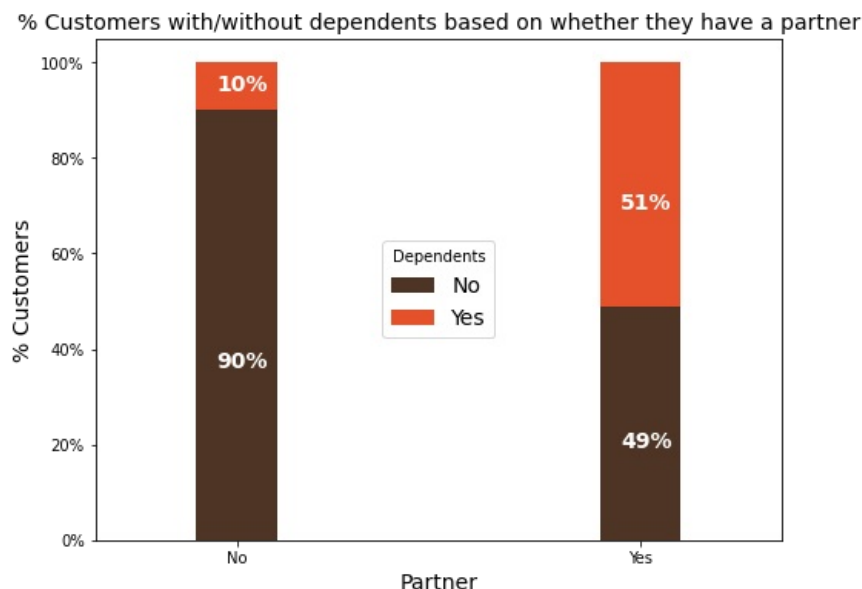


```
In [19]: colors = ['#4D3425', '#E4512B']
partner_dependents = telecom_cust.groupby(['Partner', 'Dependents']).size().unstack()

ax = (partner_dependents.T*100.0 / partner_dependents.T.sum()).T.plot(kind='bar',
                                                                    width = 0.2,
                                                                    stacked = True,
                                                                    rot = 0,
                                                                    figsize = (8,6),
                                                                    color = colors)

ax.yaxis.set_major_formatter(mtick.PercentFormatter())
ax.legend(loc='center', prop={'size':14}, title = 'Dependents', fontsize =14)
ax.set_ylabel('% Customers', size = 14)
ax.set_title('% Customers with/without dependents based on whether they have a partner', size = 14)
ax.xaxis.label.set_size(14)

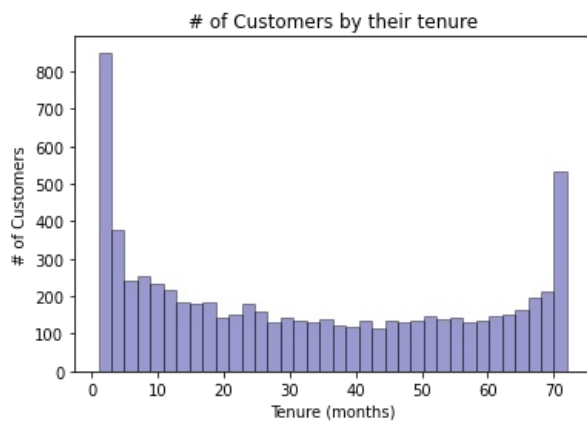
# Code to add the data labels on the stacked bar chart
for p in ax.patches:
    width, height = p.get_width(), p.get_height()
    x, y = p.get_xy()
    ax.annotate(' {:.0f}%'.format(height), (p.get_x()+.25*width, p.get_y()+.4*height),
                color = 'white',
                weight = 'bold',
                size = 14)
```



```
In [20]: ax = sns.distplot(telecom_cust['tenure'], hist=True, kde=False,
                           bins=int(180/5), color = 'darkblue',
                           hist_kws={'edgecolor':'black'},
                           kde_kws={'linewidth': 4})
ax.set_ylabel('# of Customers')
ax.set_xlabel('Tenure (months)')
ax.set_title('# of Customers by their tenure')
```

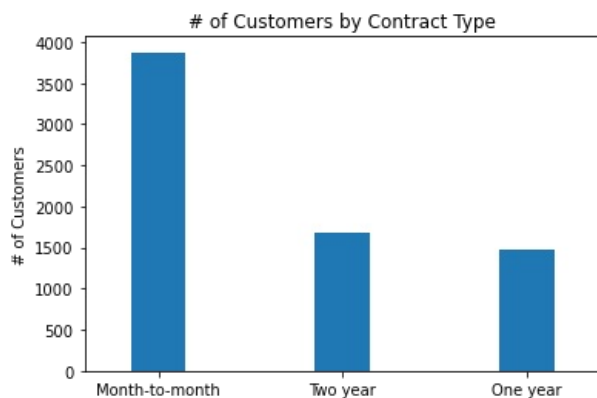
C:\Users\kankk\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

Out[20]: Text(0.5, 1.0, '# of Customers by their tenure')



```
In [21]: ax = telecom_cust['Contract'].value_counts().plot(kind = 'bar',rot = 0, width = 0.3)
ax.set_ylabel('# of Customers')
ax.set_title('# of Customers by Contract Type')
```

```
Out[21]: Text(0.5, 1.0, '# of Customers by Contract Type')
```



```
In [22]: fig, (ax1,ax2,ax3) = plt.subplots(nrows=1, ncols=3, sharey = True, figsize = (20,6))

ax = sns.distplot(telecom_cust[telecom_cust['Contract']=='Month-to-month']['tenure'],
                  hist=True, kde=False,
                  bins=int(180/5), color = 'turquoise',
                  hist_kws={'edgecolor':'black'},
                  kde_kws={'linewidth': 4},
                  ax=ax1)
ax.set_ylabel('# of Customers')
ax.set_xlabel('Tenure (months)')
ax.set_title('Month to Month Contract')

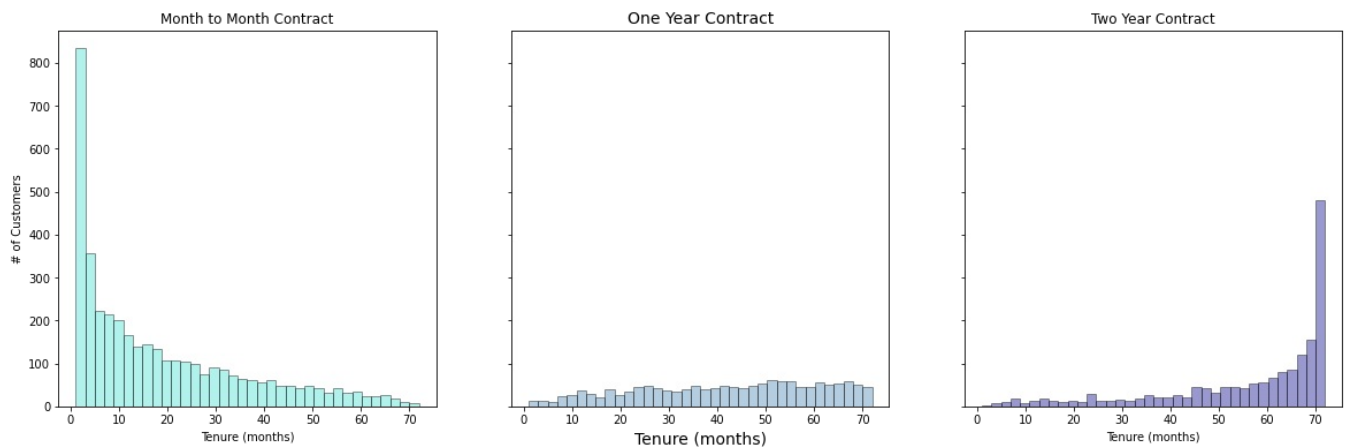
ax = sns.distplot(telecom_cust[telecom_cust['Contract']=='One year']['tenure'],
                  hist=True, kde=False,
                  bins=int(180/5), color = 'steelblue',
                  hist_kws={'edgecolor':'black'},
                  kde_kws={'linewidth': 4},
                  ax=ax2)
ax.set_xlabel('Tenure (months)',size = 14)
ax.set_title('One Year Contract',size = 14)

ax = sns.distplot(telecom_cust[telecom_cust['Contract']=='Two year']['tenure'],
                  hist=True, kde=False,
                  bins=int(180/5), color = 'darkblue',
                  hist_kws={'edgecolor':'black'},
                  kde_kws={'linewidth': 4},
                  ax=ax3)

ax.set_xlabel('Tenure (months)')
ax.set_title('Two Year Contract')
```

C:\Users\kankk\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

```
Out[22]: Text(0.5, 1.0, 'Two Year Contract')
```



Let us now look at the distribution of various services used by customers

```
In [23]: telecom_cust.columns.values
```

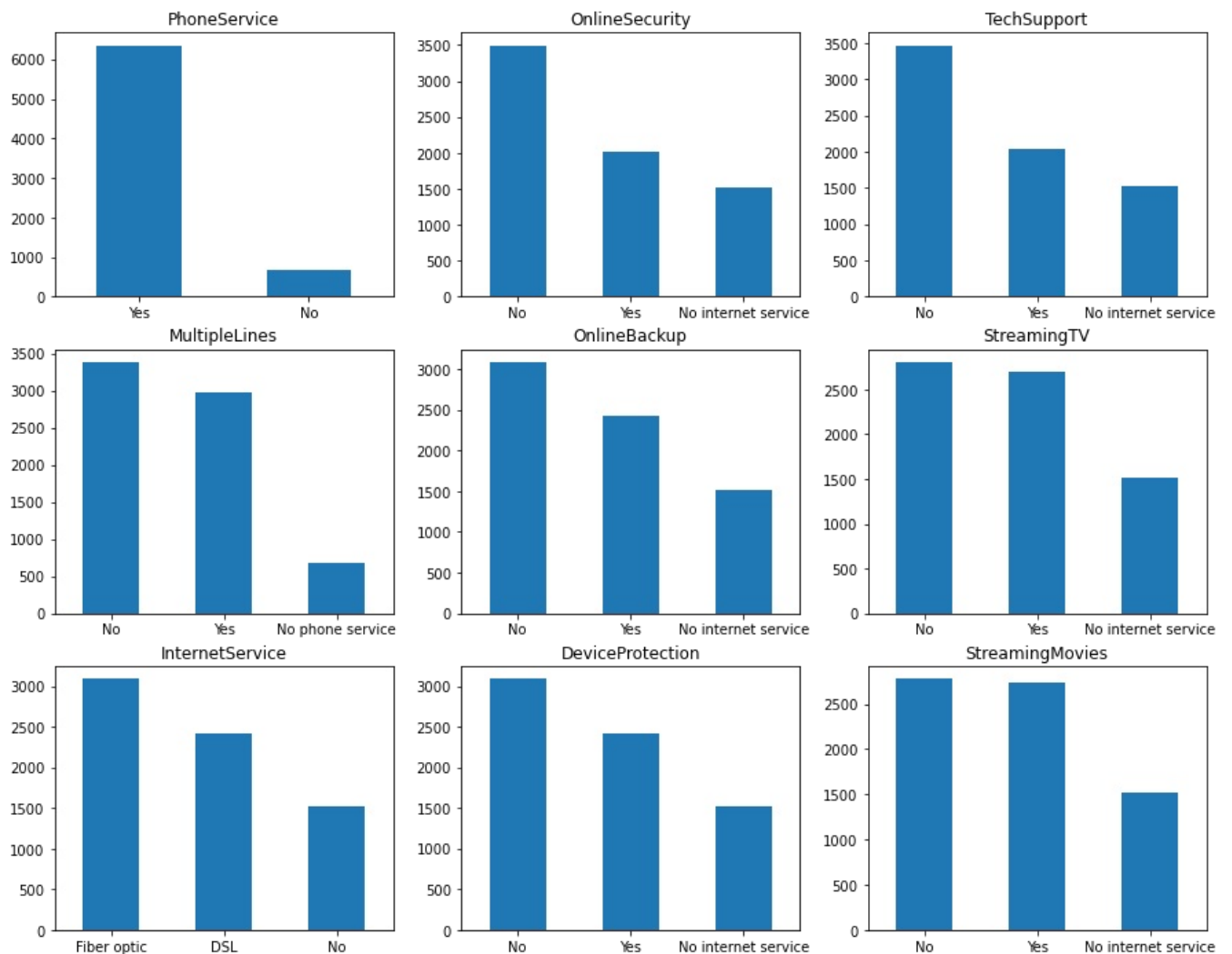
```
Out[23]: array(['customerID', 'gender', 'SeniorCitizen', 'Partner', 'Dependents',
       'tenure', 'PhoneService', 'MultipleLines', 'InternetService',
       'OnlineSecurity', 'OnlineBackup', 'DeviceProtection',
       'TechSupport', 'StreamingTV', 'StreamingMovies', 'Contract',
       'PaperlessBilling', 'PaymentMethod', 'MonthlyCharges',
       'TotalCharges', 'Churn'], dtype=object)
```

```
In [24]: services = ['PhoneService', 'MultipleLines', 'InternetService', 'OnlineSecurity',
                    'OnlineBackup', 'DeviceProtection', 'TechSupport', 'StreamingTV', 'StreamingMovies']

fig, axes = plt.subplots(nrows = 3, ncols = 3, figsize = (15,12))
for i, item in enumerate(services):
    if i < 3:
        ax = telecom_cust[item].value_counts().plot(kind = 'bar', ax=axes[i,0], rot = 0)

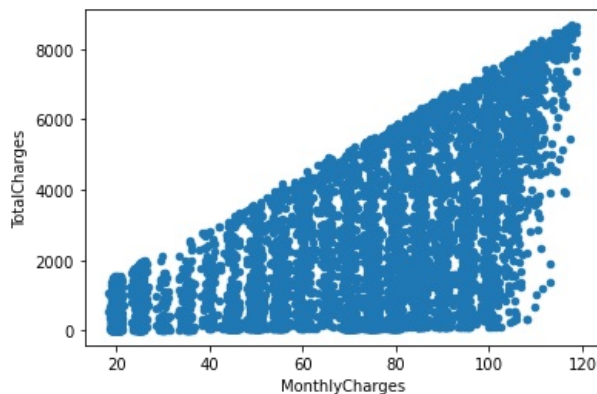
    elif i >=3 and i < 6:
        ax = telecom_cust[item].value_counts().plot(kind = 'bar', ax=axes[i-3,1], rot = 0)

    elif i < 9:
        ax = telecom_cust[item].value_counts().plot(kind = 'bar', ax=axes[i-6,2], rot = 0)
    ax.set_title(item)
```



```
In [25]: telecom_cust[['MonthlyCharges', 'TotalCharges']].plot.scatter(x = 'MonthlyCharges',
y='TotalCharges')
```

```
Out[25]: <AxesSubplot: xlabel='MonthlyCharges', ylabel='TotalCharges'>
```



Finally, let's take a look at our predictor variable (Churn) and understand its interaction with other important variables as was found out in the correlation plot.

```
In [26]: colors = ['#4D3425', '#E4512B']
ax = (telecom_cust['Churn'].value_counts()*100.0 / len(telecom_cust)).plot(kind='bar',
stacked = True,
rot = 0,
color = colors,
figsize = (8,6))

ax.yaxis.set_major_formatter(mtick.PercentFormatter())
ax.set_ylabel('% Customers', size = 14)
ax.set_xlabel('Churn', size = 14)
ax.set_title('Churn Rate', size = 14)

# create a list to collect the plt.patches data
totals = []

# find the values and append to list
for i in ax.patches:
    totals.append(i.get_width())
```



```
# set individual bar labels using above list
total = sum(totals)

for i in ax.patches:
    # get width pulls left or right; get_y pushes up or down
    ax.text(i.get_x()+.15, i.get_height()-4.0, \
            str(round((i.get_height()/total), 1))+'%',
            fontsize=12,
            color='white',
            weight = 'bold',
            size = 14)
```

-----  
**TypeError** Traceback (most recent call last)

```
Input In [26], in <cell line: 22>()
    20 total = sum(totals)
    22 for i in ax.patches:
    23     # get width pulls left or right; get_y pushes up or down
--> 24     ax.text(i.get_x()+.15, i.get_height()-4.0, \
    25             str(round((i.get_height()/total), 1))+'%',
    26             fontsize=12,
    27             color='white',
    28             weight = 'bold',
    29             size = 14)
```

File ~\anaconda3\lib\site-packages\matplotlib\axes\\_axes.py:659, in Axes.text(self, x, y, s, fontdict, \*\*kwargs)

```
    598 """
    599 Add text to the Axes.
    600
    (...)
    649 >>> text(x, y, s, bbox=dict(facecolor='red', alpha=0.5))
    650 """
    651 effective_kwargs = {
    652     'verticalalignment': 'baseline',
    653     'horizontalalignment': 'left',
    (...)
    657     **kwargs,
    658 }
--> 659 t = mtext.Text(x, y, text=s, **effective_kwargs)
    660 t.set_clip_path(self.patch)
    661 self._add_text(t)
```

File ~\anaconda3\lib\site-packages\matplotlib\text.py:160, in Text.\_\_init\_\_(self, x, y, text, color, verticalalignment, horizontalalignment, multialignment, fontproperties, rotation, linespacing, rotation\_mode, usetex, wrap, transform\_rotates\_text, parse\_math, \*\*kwargs)

```
    158 self._linespacing = linespacing
    159 self.set_rotation_mode(rotation_mode)
--> 160 self.update(kwargs)
```

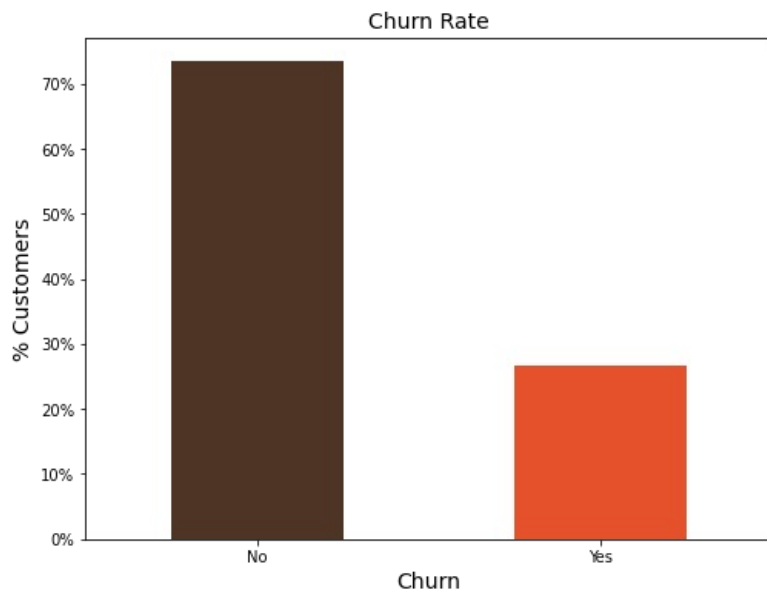
File ~\anaconda3\lib\site-packages\matplotlib\text.py:164, in Text.update(self, kwargs)

```
    162 def update(self, kwargs):
    163     # docstring inherited
--> 164     kwargs = cbook.normalize_kwargs(kwargs, Text)
    165     sentinel = object() # bbox can be None, so use another sentinel.
    166     # Update fontproperties first, as it has lowest priority.
```

File ~\anaconda3\lib\site-packages\matplotlib\cbook\\_\_init\_\_.py:1739, in normalize\_kwargs(kw, alias\_mapping)

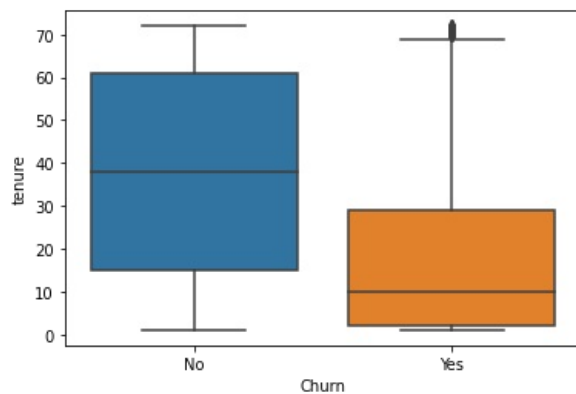
```
    1737 canonical = to_canonical.get(k, k)
    1738 if canonical in canonical_to_seen:
--> 1739     raise TypeError(f"Got both {canonical_to_seen[canonical]!r} and "
    1740                     f"{k!r}, which are aliases of one another")
    1741 canonical_to_seen[canonical] = k
    1742 ret[canonical] = v
```

**TypeError:** Got both 'fontsize' and 'size', which are aliases of one another



```
In [27]: sns.boxplot(x = telecom_cust.Churn, y = telecom_cust.tenure)
```

```
Out[27]: <AxesSubplot:xlabel='Churn', ylabel='tenure'>
```



Churn by Contract Type

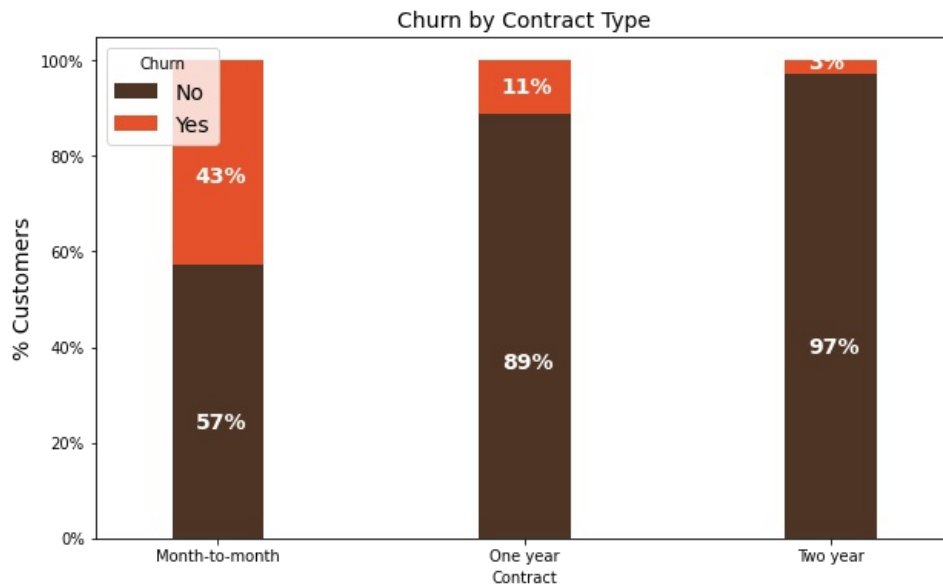
```
In [28]: colors = ['#4D3425', '#E4512B']
contract_churn = telecom_cust.groupby(['Contract', 'Churn']).size().unstack()

ax = (contract_churn.T*100.0 / contract_churn.T.sum()).T.plot(kind='bar',
                                                             width = 0.3,
                                                             stacked = True,
                                                             rot = 0,
                                                             figsize = (10,6),
                                                             color = colors)

ax.yaxis.set_major_formatter(mtick.PercentFormatter())
ax.legend(loc='best',prop={'size':14},title = 'Churn')
ax.set_ylabel('% Customers',size = 14)
ax.set_title('Churn by Contract Type',size = 14)

# Code to add the data labels on the stacked bar chart
for p in ax.patches:
```

```
width, height = p.get_width(), p.get_height()
x, y = p.get_xy()
ax.annotate('{:.0f}%'.format(height), (p.get_x()+.25*width, p.get_y()+.4*height),
            color = 'white',
            weight = 'bold',
            size = 14)
```



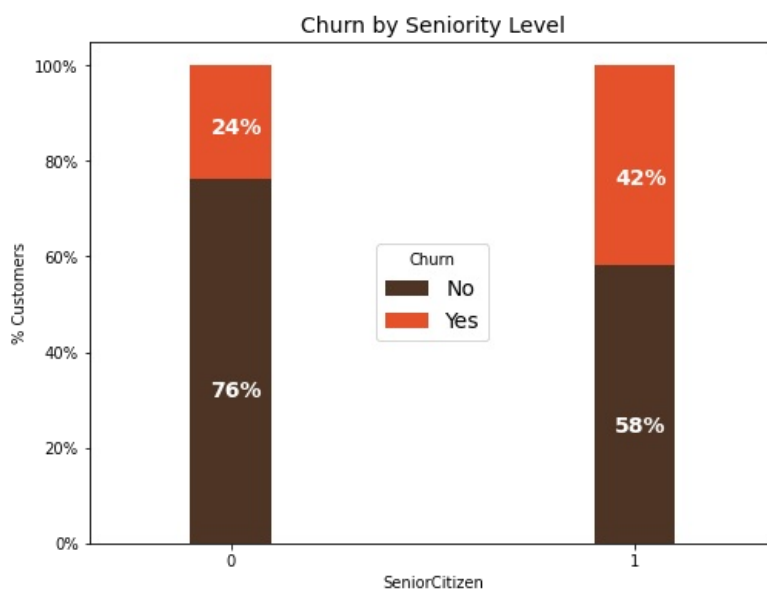
Churn by Seniority

```
In [29]: colors = ['#4D3425', '#E4512B']
seniority_churn = telecom_cust.groupby(['SeniorCitizen', 'Churn']).size().unstack()

ax = (seniority_churn.T*100.0 / seniority_churn.T.sum()).T.plot(kind='bar',
                                                                width = 0.2,
                                                                stacked = True,
                                                                rot = 0,
                                                                figsize = (8,6),
                                                                color = colors)

ax.yaxis.set_major_formatter(mtick.PercentFormatter())
ax.legend(loc='center',prop={'size':14},title = 'Churn')
ax.set_ylabel('% Customers')
ax.set_title('Churn by Seniority Level',size = 14)

# Code to add the data labels on the stacked bar chart
for p in ax.patches:
    width, height = p.get_width(), p.get_height()
    x, y = p.get_xy()
    ax.annotate('{:.0f}%'.format(height), (p.get_x()+.25*width, p.get_y()+.4*height),
                color = 'white',
                weight = 'bold',size =14)
```

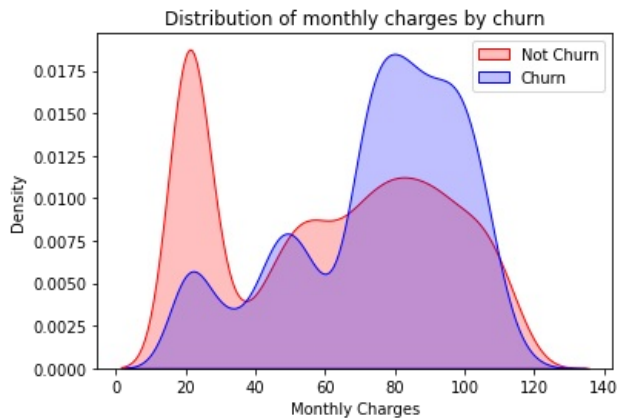


Churn by Monthly Charges

```
In [31]: ax = sns.kdeplot(telecom_cust.MonthlyCharges[(telecom_cust["Churn"] == 'No') ],
                        color="Red", shade = True)
ax = sns.kdeplot(telecom_cust.MonthlyCharges[(telecom_cust["Churn"] == 'Yes') ],
                  ax=ax, color="Blue", shade= True)
```

```
ax.legend(["Not Churn","Churn"],loc='upper right')
ax.set_ylabel('Density')
ax.set_xlabel('Monthly Charges')
ax.set_title('Distribution of monthly charges by churn')
```

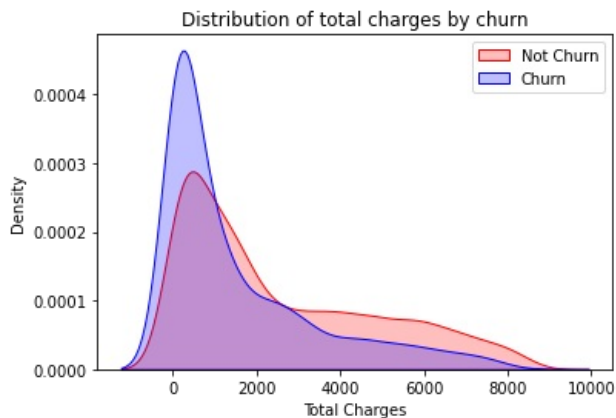
Out[31]: Text(0.5, 1.0, 'Distribution of monthly charges by churn')



Churn by Total Charges

```
In [32]: ax = sns.kdeplot(telecom_cust.TotalCharges[(telecom_cust["Churn"] == 'No') ],
                        color="Red", shade = True)
ax = sns.kdeplot(telecom_cust.TotalCharges[(telecom_cust["Churn"] == 'Yes') ],
                ax=ax, color="Blue", shade= True)
ax.legend(["Not Churn","Churn"],loc='upper right')
ax.set_ylabel('Density')
ax.set_xlabel('Total Charges')
ax.set_title('Distribution of total charges by churn')
```

Out[32]: Text(0.5, 1.0, 'Distribution of total charges by churn')



After going through the above EDA we will develop some predictive models and compare them. We will develop Logistic Regression, Random Forest, SVM, ADA Boost, XG Boost 1. Logistic Regression

```
In [33]: y = df_dummies['Churn'].values
X = df_dummies.drop(columns = ['Churn'])

# Scaling all the variables to a range of 0 to 1
from sklearn.preprocessing import MinMaxScaler
features = X.columns.values
scaler = MinMaxScaler(feature_range = (0,1))
scaler.fit(X)
X = pd.DataFrame(scaler.transform(X))
X.columns = features
```

```
In [36]: # Create Train & Test Data
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=101)
```

```
In [37]: # Running logistic regression model
from sklearn.linear_model import LogisticRegression
model = LogisticRegression()
result = model.fit(X_train, y_train)
```

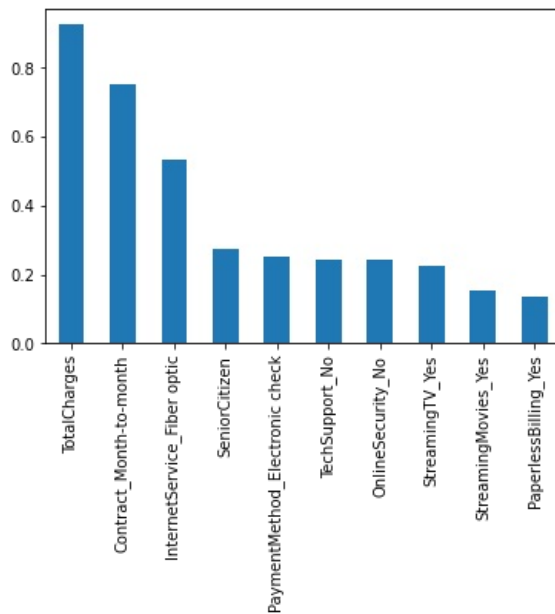
```
In [38]: from sklearn import metrics
prediction_test = model.predict(X_test)
# Print the prediction accuracy
print (metrics.accuracy_score(y_test, prediction_test))
```

0.8075829383886256

```
In [39]: # To get the weights of all the variables
```

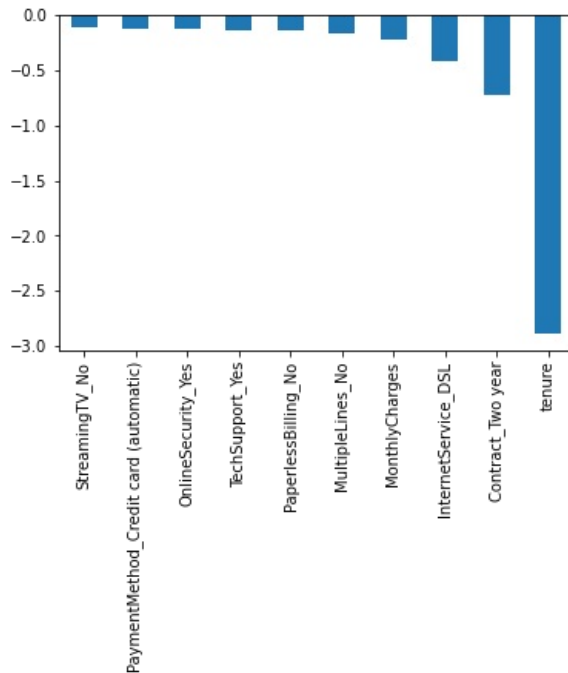
```
weights = pd.Series(model.coef_[0],
                    index=X.columns.values)
print (weights.sort_values(ascending = False)[:10].plot(kind='bar'))
```

AxesSubplot(0.125,0.125;0.775x0.755)



```
In [40]: print(weights.sort_values(ascending = False)[-10:].plot(kind='bar'))
```

AxesSubplot(0.125,0.125;0.775x0.755)



Observations We can see that some variables have a negative relation to our predicted variable (Churn), while some have positive relation. Negative relation means that likeliness of churn decreases with that variable. Let us summarize some of the interesting features below:

As we saw in our EDA, having a 2 month contract reduces chances of churn. 2 month contract along with tenure have the most negative relation with Churn as predicted by logistic regressions

Having DSL internet service also reduces the probability of Churn

Lastly, total charges, monthly contracts, fibre optic internet services and seniority can lead to higher churn rates. This is interesting because although fibre optic services are faster, customers are likely to churn because of it. I think we need to explore more to better understand why this is happening.

Any hypothesis on the above would be really helpful!