

# Customer Survival Analysis

## Importing Libraries

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from scipy.stats import norm
import statsmodels.api as st
from sklearn.preprocessing import LabelEncoder
labelencoder = LabelEncoder()

#Lifelines is a survival analysis package
from lifelines import KaplanMeierFitter
from lifelines.statistics import multivariate_logrank_test
from lifelines.statistics import logrank_test
from lifelines import CoxPHFitter
```

## Data Preparation

```
In [2]: df = pd.read_csv("C:/Data/Telco-Customer-Churn.csv")
df.head()
```

Out[2]:

	customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	Mul
<b>0</b>	7590-VHVEG	Female	0	Yes	No	1	No	
<b>1</b>	5575-	Male	0	No	No	34	Yes	
<b>2</b>	3668-QPYBK	Male	0	No	No	2	Yes	
<b>3</b>	7795-	Male	0	No	No	45	No	
<b>4</b>	9237-HQITU	Female	0	No	No	2	Yes	

5 rows × 21 columns

```
In [3]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 7043 entries, 0 to 7042
Data columns (total 21 columns):
customerID      7043 non-null object
gender          7043 non-null object
SeniorCitizen    7043 non-null int64
Partner          7043 non-null object
Dependents       7043 non-null object
tenure           7043 non-null int64
PhoneService     7043 non-null object
MultipleLines    7043 non-null object
InternetService  7043 non-null object
OnlineSecurity   7043 non-null object
OnlineBackup      7043 non-null object
DeviceProtection 7043 non-null object
TechSupport       7043 non-null object
StreamingTV       7043 non-null object
StreamingMovies   7043 non-null object
Contract          7043 non-null object
PaperlessBilling  7043 non-null object
PaymentMethod     7043 non-null object
MonthlyCharges    7043 non-null float64
TotalCharges      7043 non-null object
Churn             7043 non-null object
dtypes: float64(1), int64(2), object(18)
memory usage: 1.1+ MB
```

```
In [4]: df.Churn = labelencoder.fit_transform(df.Churn)
df.Churn.value_counts()
```

```
Out[4]: 0    5174
1    1869
Name: Churn, dtype: int64
```

```
In [5]: eventvar = df['Churn']
timevar = df['tenure']
```

```
In [6]: categorical = ['gender', 'SeniorCitizen', 'Partner', 'Dependents', 'PhoneService',
                    'InternetService', 'OnlineSecurity', 'OnlineBackup', 'DeviceProtection',
                    'TechSupport', 'StreamingTV', 'StreamingMovies', 'Contract',
                    'PaperlessBilling', 'PaymentMethod']

survivaldata = pd.get_dummies(df, columns = categorical, drop_first= True)
survivaldata.head()
```

Out[6]:

	customerID	tenure	MonthlyCharges	TotalCharges	Churn	gender_Male	SeniorCitiz
0	7590-VHVEG	1	29.85	29.85	0	0	0
1	5575-QPYBK	34	56.95	1889.5	0	1	1
2	3668-QPYBK	2	53.85	108.15	1	1	1
3	7795-HQITU	45	42.30	1840.75	0	1	1
4	9237-HQITU	2	70.70	151.65	1	0	0

5 rows × 32 columns



In [7]:

```
survivaldata.drop(['customerID', 'tenure', 'Churn'], axis = 1, inplace= True)
survivaldata = st.add_constant(survivaldata, prepend=False)
survivaldata.head()
```

Out[7]:

	MonthlyCharges	TotalCharges	gender_Male	SeniorCitizen_1	Partner_Yes	Dependen
0	29.85	29.85	0	0	0	1
1	56.95	1889.5	1	0	0	0
2	53.85	108.15	1	0	0	0
3	42.30	1840.75	1	0	0	0
4	70.70	151.65	0	0	0	0

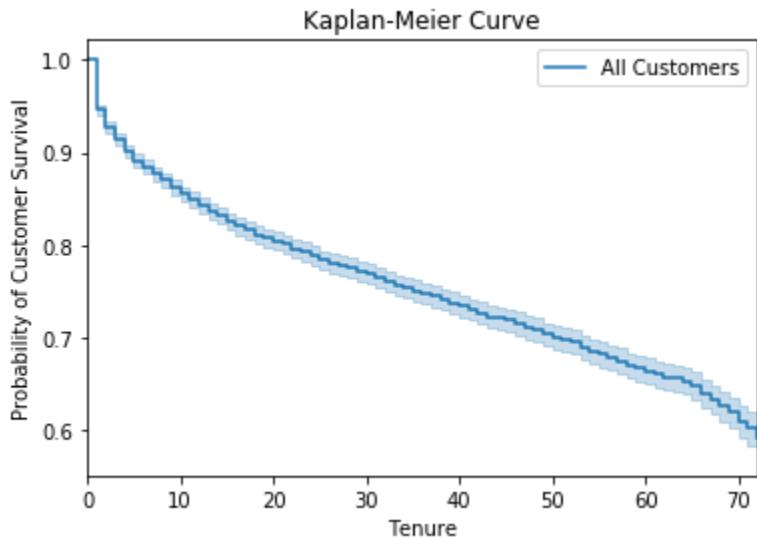
5 rows × 30 columns



## Survival Analysis

### Kaplan-Meier Curve

```
#Create a KaplanMeier object, imported from Lifelines
kmf = KaplanMeierFitter()
#Calculate the K-M curve for all groups
kmf.fit(timevar,event_observed = eventvar,label = "All Customers")
#Plot the curve and assign labels
kmf.plot()
plt.ylabel('Probability of Customer Survival')
plt.xlabel('Tenure')
plt.title('Kaplan-Meier Curve');
```



## Log-Rank Test

```
In [90]: male = (survivaldata['gender_Male'] == 1)
female = (survivaldata['gender_Male'] == 0)

plt.figure()
ax = plt.subplot(1,1,1)

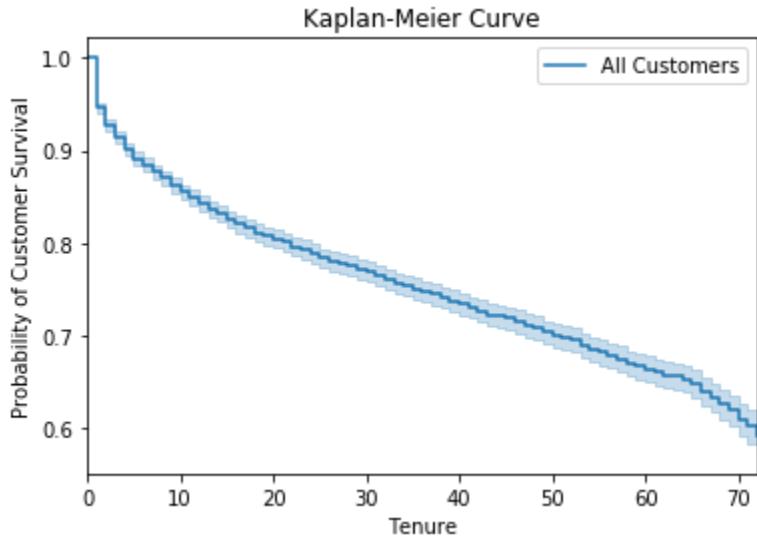
kmf.fit(timevar[male],event_observed = eventvar[male],label = "Male")
plot1 = kmf.plot(ax = ax)

kmf.fit(timevar[female],event_observed = eventvar[female],label = "Female")
plot2 = kmf.plot(ax = plot1)

plt.title('Survival of customers: Gender')
plt.xlabel('Tenure')
plt.ylabel('Survival Probability')
plt.yticks(np.linspace(0,1,11))
groups = logrank_test(timevar[male], timevar[female], event_observed_A=eventvar[
groups.print_summary()

<lifelines.StatisticalResult>
    t_0 = -1
null_distribution = chi squared
degrees_of_freedom = 1

---
test_statistic      p   -log2(p)
      0.53  0.47      1.09
```



## Senior Citizen

```
In [91]: SeniorCitizen = (survivaldata['SeniorCitizen_1'] == 1)
no_SeniorCitizen = (survivaldata['SeniorCitizen_1'] == 0)

plt.figure()
ax = plt.subplot(1,1,1)

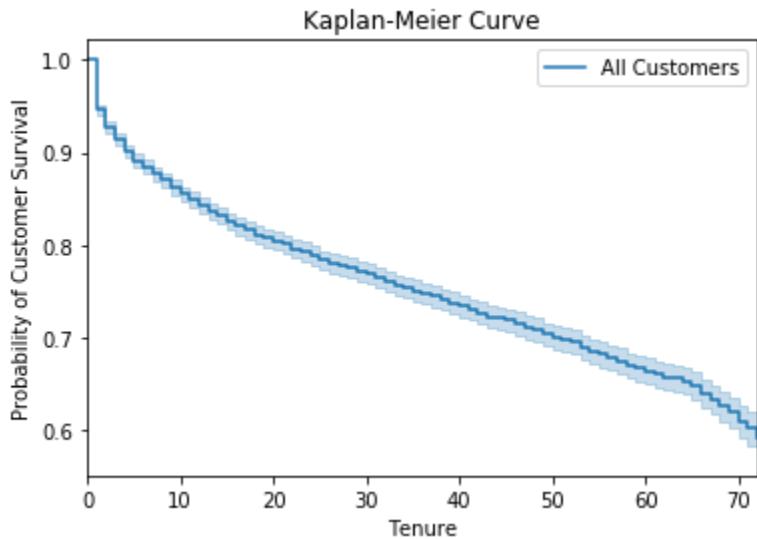
kmf.fit(timevar[SeniorCitizen], event_observed = eventvar[SeniorCitizen], label =
plot1 = kmf.plot(ax = ax)

kmf.fit(timevar[no_SeniorCitizen], event_observed = eventvar[no_SeniorCitizen], la
plot2 = kmf.plot(ax = plot1)

plt.title('Survival of customers: Senior Citizen')
plt.xlabel('Tenure')
plt.ylabel('Survival Probability')
plt.yticks(np.linspace(0,1,11))
groups = logrank_test(timevar[SeniorCitizen], timevar[no_SeniorCitizen], event_o
groups.print_summary()

<lifelines.StatisticalResult>
    t_0 = -1
    null_distribution = chi squared
    degrees_of_freedom = 1

    ...
    test_statistic      p   -log2(p)
    109.49 <0.005      82.71
```



## Partner

```
In [92]: partner = (survivaldata['Partner_Yes'] == 1)
no_partner = (survivaldata['Partner_Yes'] == 0)

plt.figure()
ax = plt.subplot(1,1,1)

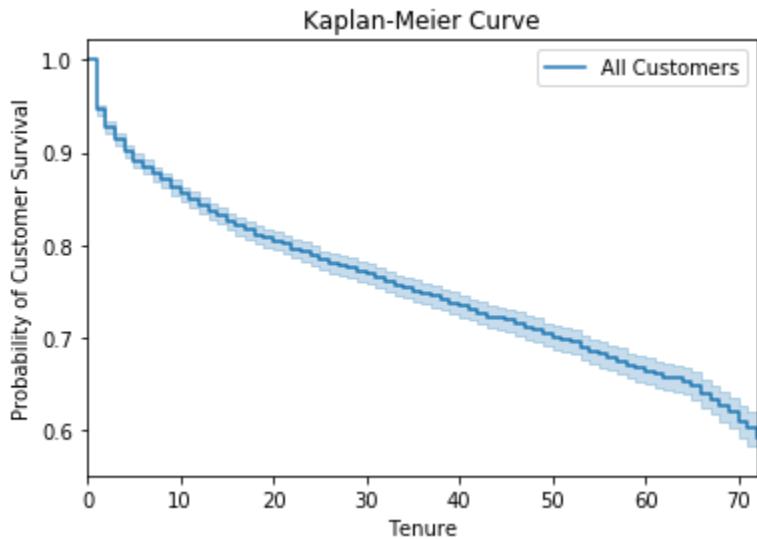
kmf.fit(timevar[partner],event_observed = eventvar[partner],label = "Has partner")
plot1 = kmf.plot(ax = ax)

kmf.fit(timevar[no_partner],event_observed = eventvar[no_partner],label = "Does not have partner")
plot2 = kmf.plot(ax = plot1)

plt.title('Survival of customers: Partner')
plt.xlabel('Tenure')
plt.ylabel('Survival Probability')
plt.yticks(np.linspace(0,1,11))
groups = logrank_test(timevar[partner], timevar[no_partner], event_observed_A=eventvar[partner], event_observed_B=eventvar[no_partner])
groups.print_summary()

<lifelines.StatisticalResult>
    t_0 = -1
    null_distribution = chi squared
    degrees_of_freedom = 1

    ...
    test_statistic      p   -log2(p)
        423.54 <0.005      310.21
```



## Dependents

```
In [93]: Dependents = (survivaldata['Dependents_Yes'] == 1)
no_Dependents = (survivaldata['Dependents_Yes'] == 0)

plt.figure()
ax = plt.subplot(1,1,1)

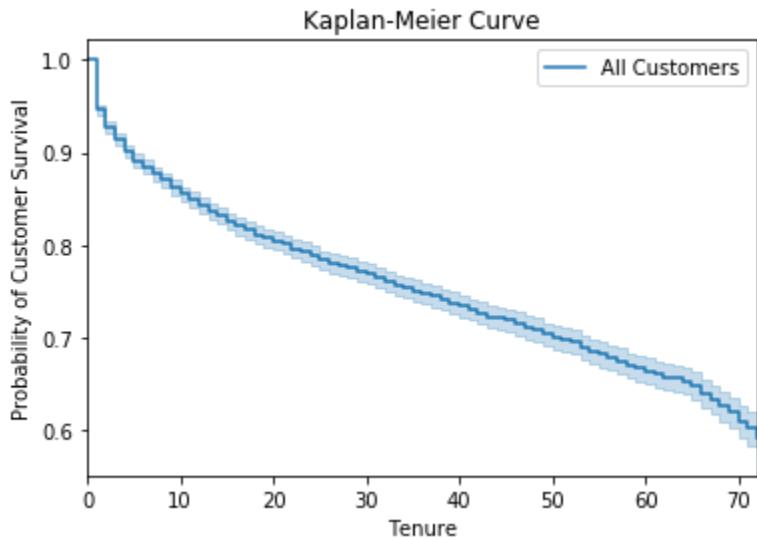
kmf.fit(timevar[Dependents], event_observed = eventvar[Dependents], label = "Has dependents")
plot1 = kmf.plot(ax = ax)

kmf.fit(timevar[no_Dependents], event_observed = eventvar[no_Dependents], label = "No dependents")
plot2 = kmf.plot(ax = plot1)

plt.title('Survival of customers: Dependents')
plt.xlabel('Tenure')
plt.ylabel('Survival Probability')
plt.yticks(np.linspace(0,1,11))
groups = logrank_test(timevar[Dependents], timevar[no_Dependents], event_observed = eventvar)
groups.print_summary()

<lifelines.StatisticalResult>
    t_0 = -1
    null_distribution = chi squared
    degrees_of_freedom = 1

    ...
    test_statistic      p   -log2(p)
    232.70 <0.005     172.12
```



## PhoneService

```
In [94]: PhoneService = (survivaldata['PhoneService_Yes'] == 1)
no_PhoneService = (survivaldata['PhoneService_Yes'] == 0)

plt.figure()
ax = plt.subplot(1,1,1)

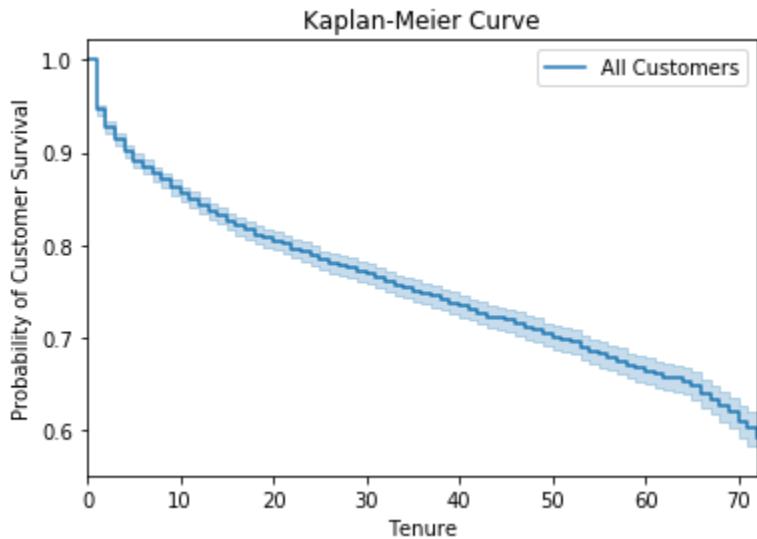
kmf.fit(timevar[PhoneService], event_observed = eventvar[PhoneService], label = "H")
plot1 = kmf.plot(ax = ax)

kmf.fit(timevar[no_PhoneService], event_observed = eventvar[no_PhoneService], label = "O")
plot2 = kmf.plot(ax = plot1)

plt.title('Survival of customers: Phone Service')
plt.xlabel('Tenure')
plt.ylabel('Survival Probability')
plt.yticks(np.linspace(0,1,11))
groups = logrank_test(timevar[PhoneService], timevar[no_PhoneService], event_observed = eventvar[PhoneService])
groups.print_summary()

<lifelines.StatisticalResult>
    t_0 = -1
    null_distribution = chi squared
    degrees_of_freedom = 1

    ...
    test_statistic      p   -log2(p)
        0.43  0.51      0.97
```



## MultipleLines

```
In [95]: no_phone = (survivaldata['MultipleLines_No phone service'] == 1)
multiLines = (survivaldata['MultipleLines_Yes'] == 1)
no_multiLines = ((survivaldata['MultipleLines_Yes'] == 0) & (survivaldata['Multi

plt.figure()
ax = plt.subplot(1,1,1)

kmf.fit(timevar[no_phone],event_observed = eventvar[no_phone],label = "No Phone
plot1 = kmf.plot(ax = ax)

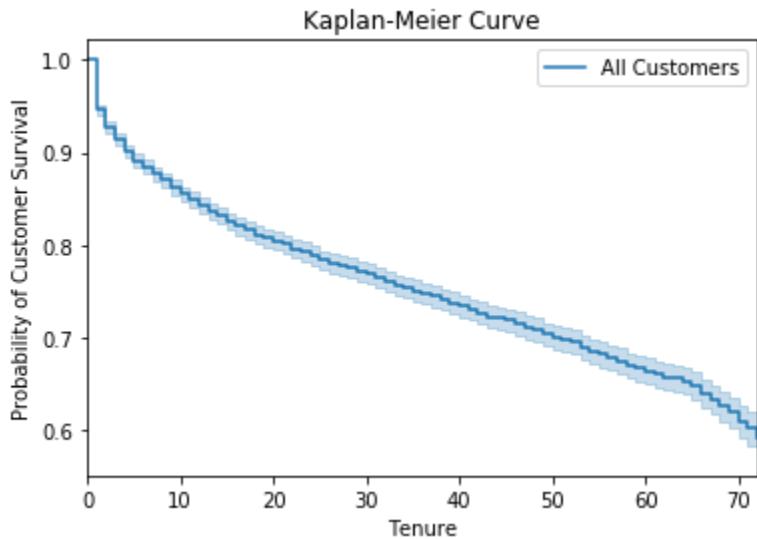
kmf.fit(timevar[multiLines],event_observed = eventvar[multiLines],label = "Multi
plot2 = kmf.plot(ax = plot1)

kmf.fit(timevar[no_multiLines],event_observed = eventvar[no_multiLines],label =
plot3 = kmf.plot(ax = plot2)

plt.title('Survival of customers: Mutliple Lines')
plt.xlabel('Tenure')
plt.ylabel('Survival Probability')
plt.yticks(np.linspace(0,1,11))
twoplusgroups_logrank = multivariate_logrank_test(df['tenure'], df['MultipleLine
twoplusgroups_logrank.print_summary()

<lifelines.StatisticalResult>
    t_0 = -1
    null_distribution = chi squared
    degrees_of_freedom = 2
    alpha = 0.95

---
    test_statistic      p   -log2(p)
    30.97 <0.005      22.34
```



## Internet Service

```
In [96]: Fiber_optic = (survivaldata['InternetService_Fiber optic'] == 1)
No_Service = (survivaldata['InternetService_No'] == 1)
DSL = ((survivaldata['InternetService_Fiber optic'] == 0) & (survivaldata['InternetService_No'] == 0))

plt.figure()
ax = plt.subplot(1,1,1)

kmf.fit(timevar[Fiber_optic],event_observed = eventvar[Fiber_optic],label = "Fiber optic")
plot1 = kmf.plot(ax = ax)

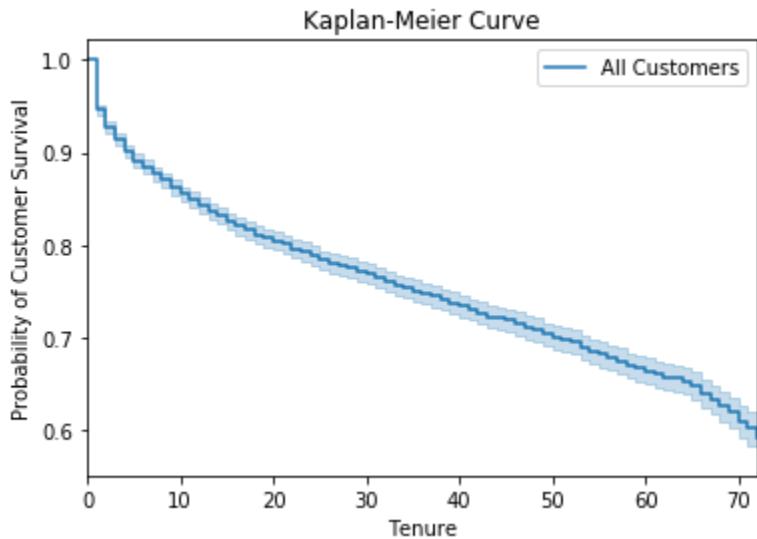
kmf.fit(timevar[No_Service],event_observed = eventvar[No_Service],label = "No Service")
plot2 = kmf.plot(ax = plot1)

kmf.fit(timevar[DSL],event_observed = eventvar[DSL],label = "DSL")
plot3 = kmf.plot(ax = plot2)

plt.title('Survival of customers: Internet Service')
plt.xlabel('Tenure')
plt.ylabel('Survival Probability')
plt.yticks(np.linspace(0,1,11))
twoplusgroups_logrank = multivariate_logrank_test(df['tenure'], df['InternetService'], groupby='InternetService')
twoplusgroups_logrank.print_summary()

<lifelines.StatisticalResult>
    t_0 = -1
    null_distribution = chi squared
    degrees_of_freedom = 2
    alpha = 0.95

---
test_statistic      p   -log2(p)
      520.12 <0.005     375.19
```



## Online Security

```
In [97]: no_internetService = (survivaldata['OnlineSecurity_No internet service'] == 1)
onlineSecurity = (survivaldata['OnlineSecurity_Yes'] == 1)
no_onlineSecurity = ((survivaldata['OnlineSecurity_No internet service'] == 0) &
                     plt.figure()
ax = plt.subplot(1,1,1)

kmf.fit(timevar[no_internetService],event_observed = eventvar[no_internetService])
plot1 = kmf.plot(ax = ax)

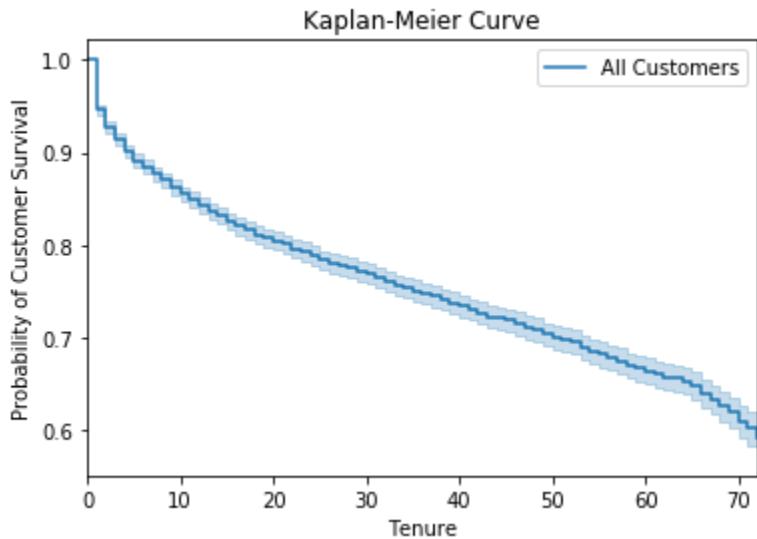
kmf.fit(timevar[onlineSecurity],event_observed = eventvar[onlineSecurity],label = 'Online Security')
plot2 = kmf.plot(ax = plot1)

kmf.fit(timevar[no_onlineSecurity],event_observed = eventvar[no_onlineSecurity]),
plot3 = kmf.plot(ax = plot2)

plt.title('Survival of customers: Online Security')
plt.xlabel('Tenure')
plt.ylabel('Survival Probability')
plt.yticks(np.linspace(0,1,11))
twoplusgroups_logrank = multivariate_logrank_test(df['tenure'], df['OnlineSecurity'])
twoplusgroups_logrank.print_summary()

<lifelines.StatisticalResult>
    t_0 = -1
    null_distribution = chi squared
    degrees_of_freedom = 2
    alpha = 0.95

    ...
    test_statistic      p -log2(p)
    1013.86 <0.005      731.35
```



## Online Backup

```
In [98]: no_internetService = (survivaldata['OnlineBackup_No internet service'] == 1)
onlineBackup = (survivaldata['OnlineBackup_Yes'] == 1)
no_onlineBackup = ((survivaldata['OnlineBackup_No internet service'] == 0) & (survivaldata['OnlineBackup_Yes'] == 0))

plt.figure()
ax = plt.subplot(1,1,1)

kmf.fit(timevar[no_internetService], event_observed = eventvar[no_internetService])
plot1 = kmf.plot(ax = ax)

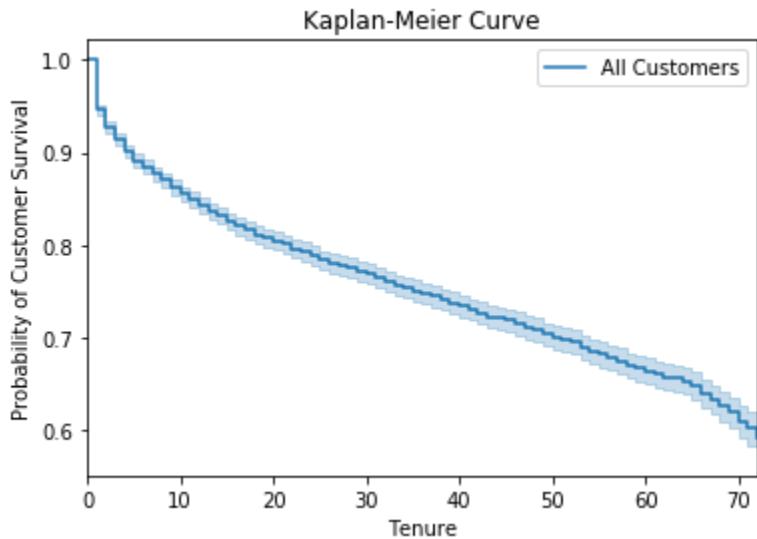
kmf.fit(timevar[onlineBackup], event_observed = eventvar[onlineBackup], label = "0")
plot2 = kmf.plot(ax = plot1)

kmf.fit(timevar[no_onlineBackup], event_observed = eventvar[no_onlineBackup], label = "1")
plot3 = kmf.plot(ax = plot2)

plt.title('Survival of customers: Online Backup')
plt.xlabel('Tenure')
plt.ylabel('Survival Probability')
plt.yticks(np.linspace(0,1,11))
twoplusgroups_logrank = multivariate_logrank_test(df['tenure'], df['OnlineBackup'])
twoplusgroups_logrank.print_summary()

<lifelines.StatisticalResult>
    t_0 = -1
    null_distribution = chi squared
    degrees_of_freedom = 2
    alpha = 0.95

    ...
    test_statistic      p   -log2(p)
    821.34 <0.005      592.47
```



## Device Protection

```
In [99]: no_internetService = (survivaldata['DeviceProtection_No internet service'] == 1)
DeviceProtection = (survivaldata['DeviceProtection_Yes'] == 1)
no_DeviceProtection = ((survivaldata['DeviceProtection_No internet service'] == 0) &
                      (survivaldata['DeviceProtection_Yes'] == 0))

plt.figure()
ax = plt.subplot(1,1,1)

kmf.fit(timevar[no_internetService], event_observed = eventvar[no_internetService])
plot1 = kmf.plot(ax = ax)

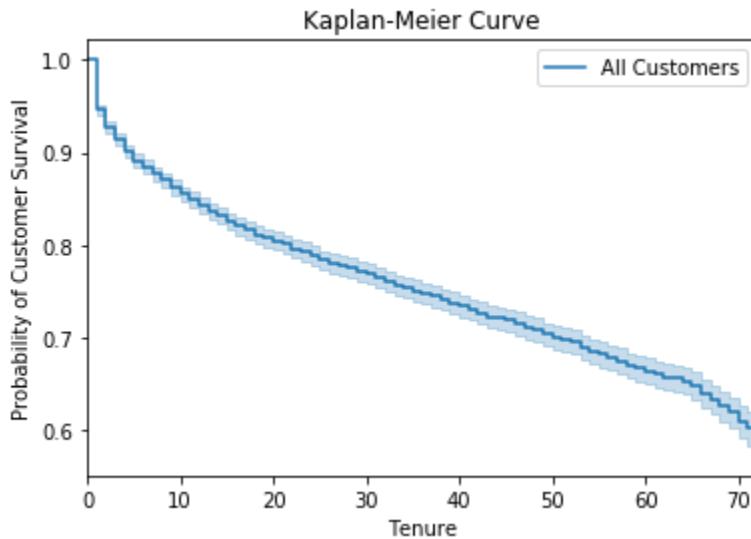
kmf.fit(timevar[DeviceProtection], event_observed = eventvar[DeviceProtection], label='Device Protection')
plot2 = kmf.plot(ax = plot1)

kmf.fit(timevar[no_DeviceProtection], event_observed = eventvar[no_DeviceProtection])
plot3 = kmf.plot(ax = plot2)

plt.title('Survival of customers: Device Protection')
plt.xlabel('Tenure')
plt.ylabel('Survival Probability')
plt.yticks(np.linspace(0,1,11))
twoplusgroups_logrank = multivariate_logrank_test(df['tenure'], df['DeviceProtection'])
twoplusgroups_logrank.print_summary()

<lifelines.StatisticalResult>
    t_0 = -1
    null_distribution = chi squared
    degrees_of_freedom = 2
    alpha = 0.95

    ...
    test_statistic      p   -log2(p)
    763.51 <0.005      550.75
```



## Tech Support

```
In [100...]: no_internetService = (survivaldata['TechSupport_No internet service'] == 1)
TechSupport = (survivaldata['TechSupport_Yes'] == 1)
no_TechSupport = ((survivaldata['TechSupport_No internet service'] == 0) & (survivaldata['TechSupport_Yes'] == 0))

plt.figure()
ax = plt.subplot(1,1,1)

kmf.fit(timevar[no_internetService], event_observed = eventvar[no_internetService])
plot1 = kmf.plot(ax = ax)

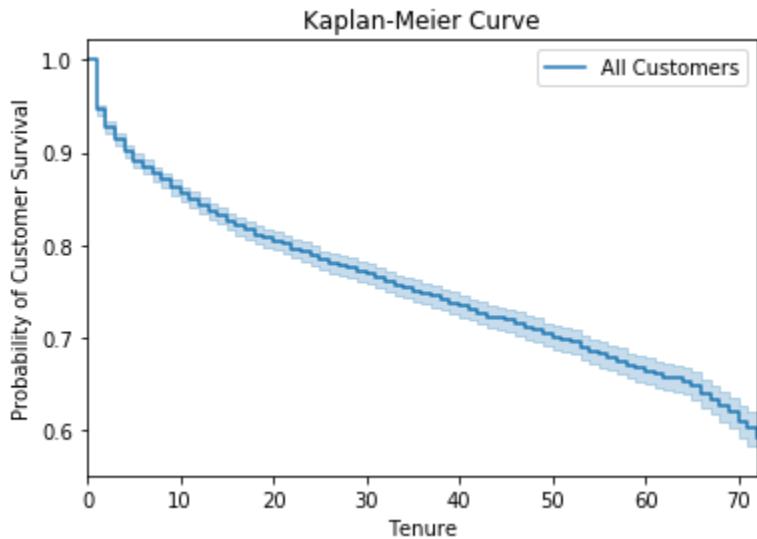
kmf.fit(timevar[TechSupport], event_observed = eventvar[TechSupport], label = "Tech Support")
plot2 = kmf.plot(ax = plot1)

kmf.fit(timevar[no_TechSupport], event_observed = eventvar[no_TechSupport], label = "No Tech Support")
plot3 = kmf.plot(ax = plot2)

plt.title('Survival of customers: Tech Support')
plt.xlabel('Tenure')
plt.ylabel('Survival Probability')
plt.yticks(np.linspace(0,1,11))
twoplusgroups_logrank = multivariate_logrank_test(df['tenure'], df['TechSupport'])
twoplusgroups_logrank.print_summary()

<lifelines.StatisticalResult>
    t_0 = -1
    null_distribution = chi squared
    degrees_of_freedom = 2
    alpha = 0.95

    ...
    test_statistic      p   -log2(p)
    989.56 <0.005      713.82
```



## Streaming TV

```
In [101...]: no_internetService = (survivaldata['StreamingTV_No internet service'] == 1)
StreamingTV = (survivaldata['StreamingTV_Yes'] == 1)
no_StreamingTV = ((survivaldata['StreamingTV_No internet service'] == 0) & (survivaldata['StreamingTV_Yes'] == 0))

plt.figure()
ax = plt.subplot(1,1,1)

kmf.fit(timevar[no_internetService], event_observed = eventvar[no_internetService])
plot1 = kmf.plot(ax = ax)

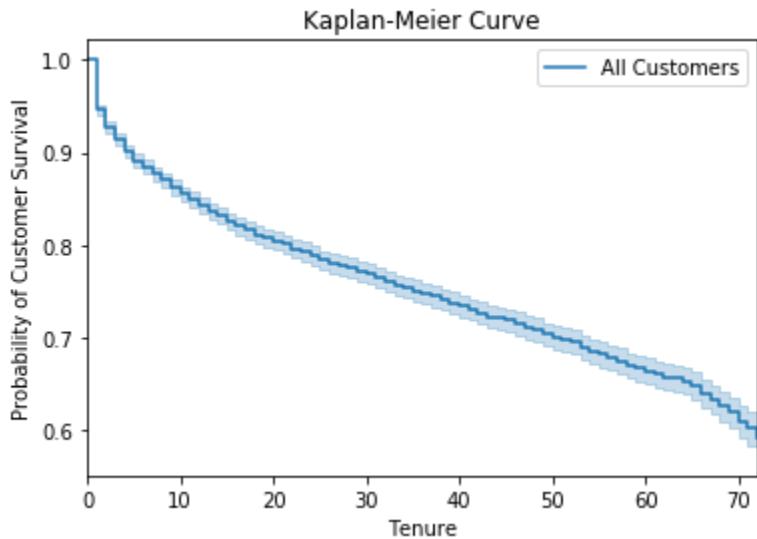
kmf.fit(timevar[StreamingTV], event_observed = eventvar[StreamingTV], label = "StreamingTV")
plot2 = kmf.plot(ax = plot1)

kmf.fit(timevar[no_StreamingTV], event_observed = eventvar[no_StreamingTV], label = "no_StreamingTV")
plot3 = kmf.plot(ax = plot2)

plt.title('Survival of customers: Streaming TV')
plt.xlabel('Tenure')
plt.ylabel('Survival Probability')
plt.yticks(np.linspace(0,1,11))
twoplusgroups_logrank = multivariate_logrank_test(df['tenure'], df['StreamingTV'])
twoplusgroups_logrank.print_summary()

<lifelines.StatisticalResult>
    t_0 = -1
    null_distribution = chi squared
    degrees_of_freedom = 2
    alpha = 0.95

    ...
    test_statistic      p   -log2(p)
    368.31 <0.005      265.68
```



## Streaming Movies

```
In [102...]: no_internetService = (survivaldata['StreamingMovies_No internet service'] == 1)
StreamingMovies = (survivaldata['StreamingMovies_Yes'] == 1)
no_StreamingMovies = ((survivaldata['StreamingMovies_No internet service'] == 0)

plt.figure()
ax = plt.subplot(1,1,1)

kmf.fit(timevar[no_internetService], event_observed = eventvar[no_internetService])
plot1 = kmf.plot(ax = ax)

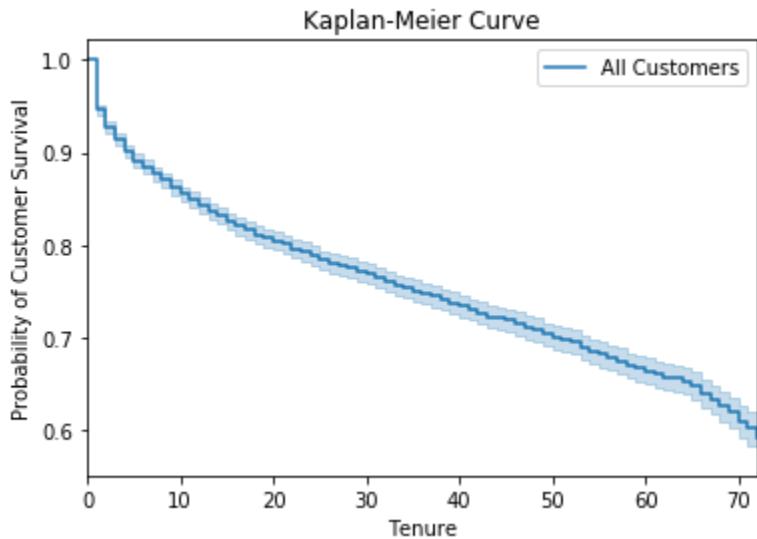
kmf.fit(timevar[StreamingMovies], event_observed = eventvar[StreamingMovies], label='StreamingMovies')
plot2 = kmf.plot(ax = plot1)

kmf.fit(timevar[no_StreamingMovies], event_observed = eventvar[no_StreamingMovies])
plot3 = kmf.plot(ax = plot2)

plt.title('Survival of customers: Streaming Movies')
plt.xlabel('Tenure')
plt.ylabel('Survival Probability')
plt.yticks(np.linspace(0,1,11))
twoplusgroups_logrank = multivariate_logrank_test(df['tenure'], df['StreamingMovies'])
twoplusgroups_logrank.print_summary()

<lifelines.StatisticalResult>
    t_0 = -1
    null_distribution = chi squared
    degrees_of_freedom = 2
    alpha = 0.95

    ...
    test_statistic      p   -log2(p)
    378.43 <0.005     272.98
```



## Contract

```
In [103...]: Contract_One_year = (survivaldata['Contract_One year'] == 1)
Contract_Two_year = (survivaldata['Contract_Two year'] == 1)
Contract_month_to_month = ((survivaldata['Contract_One year'] == 0) & (survivaldata['Contract_Two year'] == 0))

plt.figure()
ax = plt.subplot(1,1,1)

kmf.fit(timevar[Contract_One_year], event_observed = eventvar[Contract_One_year],
plot1 = kmf.plot(ax = ax))

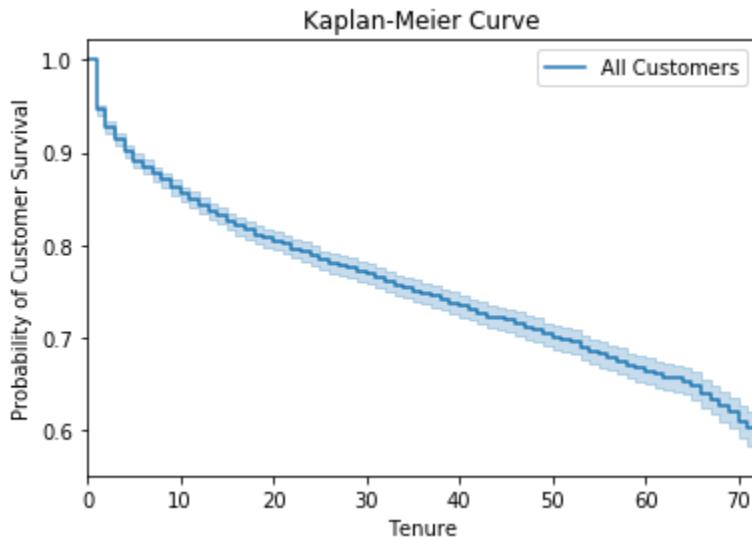
kmf.fit(timevar[Contract_Two_year], event_observed = eventvar[Contract_Two_year],
plot2 = kmf.plot(ax = plot1))

kmf.fit(timevar[Contract_month_to_month], event_observed = eventvar[Contract_month_to_month],
plot3 = kmf.plot(ax = plot2))

plt.title('Survival of customers: Contract')
plt.xlabel('Tenure')
plt.ylabel('Survival Probability')
plt.yticks(np.linspace(0,1,11))
twoplusgroups_logrank = multivariate_logrank_test(df['tenure'], df['Contract'],
twoplusgroups_logrank.print_summary())

<lifelines.StatisticalResult>
    t_0 = -1
    null_distribution = chi squared
    degrees_of_freedom = 2
    alpha = 0.95

    ...
    test_statistic      p -log2(p)
    2352.87 <0.005      inf
```



## Payment Method

```
In [104...]: automatic_Credit_Card = (survivaldata['PaymentMethod_Credit card (automatic)'] == 1)
electronic_check = (survivaldata['PaymentMethod_Electronic check'] == 1)
mailed_check = (survivaldata['PaymentMethod_Mailed check'] == 1)
automatic_Bank_Transfer = ((survivaldata['PaymentMethod_Credit card (automatic)'] == 1) & (survivaldata['PaymentMethod_Electronic check'] == 0) & (survivaldata['PaymentMethod_Mailed check'] == 0))

plt.figure()
ax = plt.subplot(1,1,1)

kmf.fit(timevar[automatic_Credit_Card], event_observed = eventvar[automatic_Credit_Card])
plot1 = kmf.plot(ax = ax)

kmf.fit(timevar[electronic_check], event_observed = eventvar[electronic_check], label = "Electronic Check")
plot2 = kmf.plot(ax = plot1)

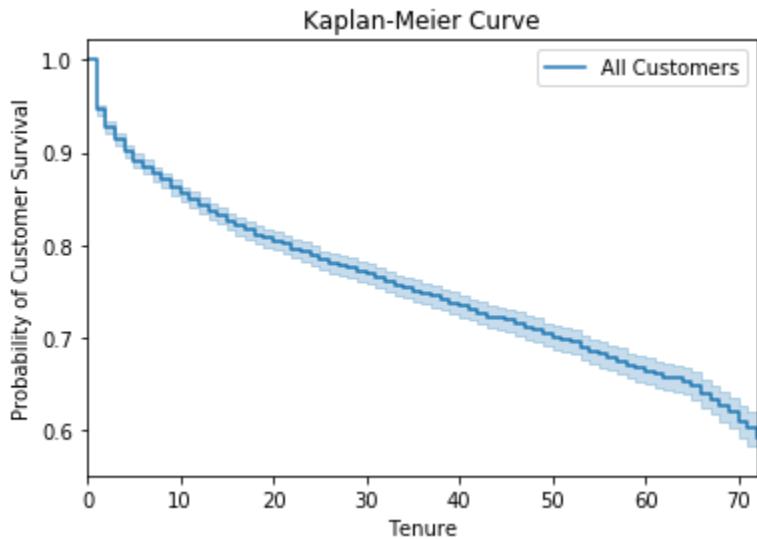
kmf.fit(timevar[mailed_check], event_observed = eventvar[mailed_check], label = "Mailed Check")
plot3 = kmf.plot(ax = plot2)

kmf.fit(timevar[automatic_Bank_Transfer], event_observed = eventvar[automatic_Bank_Transfer], label = "Automatic Transfer")
plot4 = kmf.plot(ax = plot3)

plt.title('Survival of customers: PaymentMethod')
plt.xlabel('Tenure')
plt.ylabel('Survival Probability')
plt.yticks(np.linspace(0,1,11))
twoplusgroups_logrank = multivariate_logrank_test(df['tenure'], df['PaymentMethod'])
twoplusgroups_logrank.print_summary()

<lifelines.StatisticalResult>
    t_0 = -1
    null_distribution = chi squared
    degrees_of_freedom = 3
    alpha = 0.95

    ...
    test_statistic      p   -log2(p)
    865.24 <0.005      619.58
```



## Paperless Billing

```
In [105...]: PaperlessBilling = (survivaldata['PaperlessBilling_Yes'] == 1)
no_PaperlessBilling = (survivaldata['PaperlessBilling_Yes'] == 0)

plt.figure()
ax = plt.subplot(1,1,1)

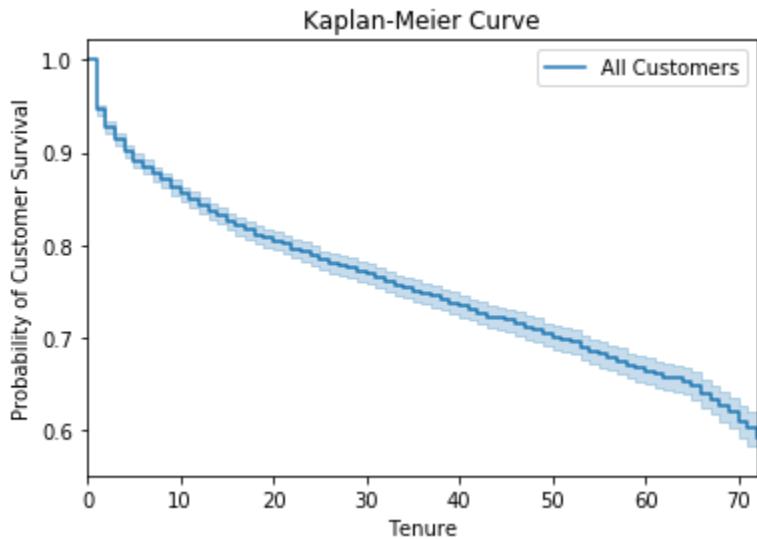
kmf.fit(timevar[PaperlessBilling], event_observed = eventvar[PaperlessBilling], label='Paperless Billing')
plot1 = kmf.plot(ax = ax)

kmf.fit(timevar[no_PhoneService], event_observed = eventvar[no_PhoneService], label='No Phone Service')
plot2 = kmf.plot(ax = plot1)

plt.title('Survival of customers: Paperless Billing')
plt.xlabel('Tenure')
plt.ylabel('Survival Probability')
plt.yticks(np.linspace(0,1,11))
groups = logrank_test(timevar[PaperlessBilling], timevar[no_PaperlessBilling], eventvar[PaperlessBilling])
groups.print_summary()

<lifelines.StatisticalResult>
    t_0 = -1
    null_distribution = chi squared
    degrees_of_freedom = 1

    ...
    test_statistic      p   -log2(p)
    189.51 <0.005     140.82
```



## Survival Regression

```
In [3]: def datapreparation(filepath):

    df = pd.read_csv(filepath)
    df.drop(["customerID"], inplace = True, axis = 1)

    df.TotalCharges = df.TotalCharges.replace(" ", np.nan)
    df.TotalCharges.fillna(0, inplace = True)
    df.TotalCharges = df.TotalCharges.astype(float)

    cols1 = ['Partner', 'Dependents', 'PaperlessBilling', 'Churn', 'PhoneService'
    for col in cols1:
        df[col] = df[col].apply(lambda x: 0 if x == "No" else 1)

    df.gender = df.gender.apply(lambda x: 0 if x == "Male" else 1)
    df.MultipleLines = df.MultipleLines.map({'No phone service': 0, 'No': 0, 'Ye
    cols2 = ['OnlineSecurity', 'OnlineBackup', 'DeviceProtection', 'TechSupport'
    for col in cols2:
        df[col] = df[col].map({'No internet service': 0, 'No': 0, 'Yes': 1})

    df = pd.get_dummies(df, columns=['InternetService', 'Contract', 'PaymentMeth
    return df

In [4]: regression_df = datapreparation("C:/Data/Telco-Customer-Churn.csv")
regression_df.head()
```

Out[4]:

	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	O
0	1	0	1	0	1	0	0	0
1	0	0	0	0	34	1	0	0
2	0	0	0	0	2	1	0	0
3	0	0	0	0	45	0	0	0
4	1	0	0	0	2	1	0	0

5 rows × 24 columns



## Survival Regression Ananlysis using Cox Proportional Hazard model

In [5]:

```
cph = CoxPHFitter()  
cph.fit(regression_df, duration_col='tenure', event_col='Churn')  
  
cph.print_summary()
```

```

<lifelines.CoxPHFitter: fitted with 7043 observations, 5174 censored>
    duration col = 'tenure'
    event col = 'Churn'
number of subjects = 7043
number of events = 1869
partial log-likelihood = -12659.69
time fit was run = 2020-09-22 14:53:48 UTC

---

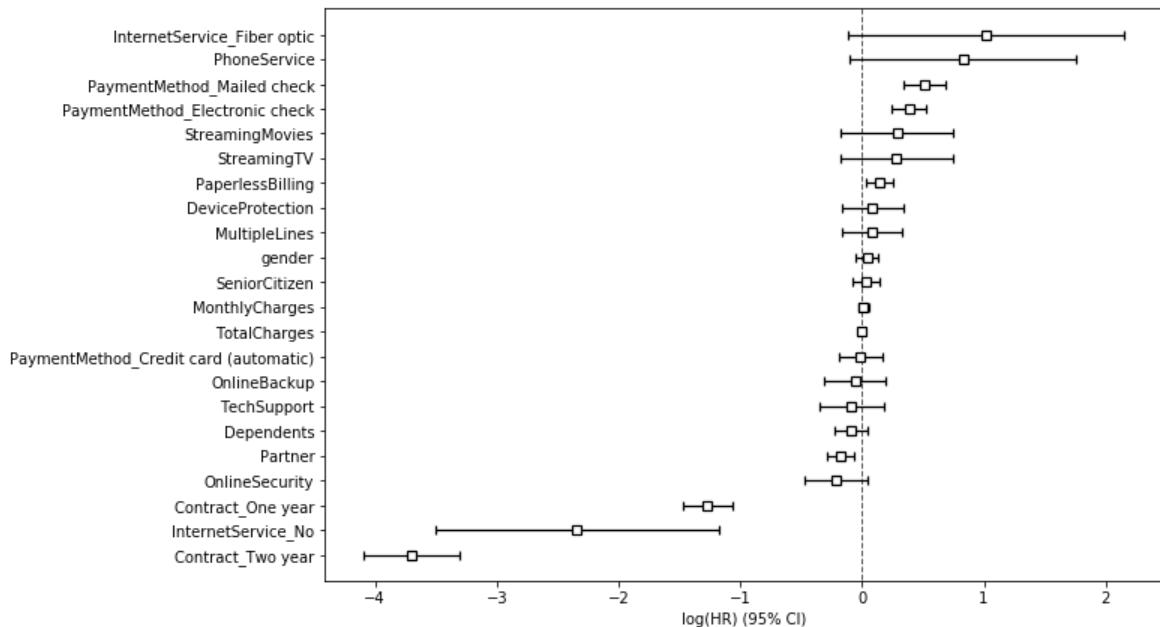

|                                       | og2(p) | lower | 0.95 | upper | 0.95 | coef  | exp(coef) | se(coef) | z      | p      | -l |
|---------------------------------------|--------|-------|------|-------|------|-------|-----------|----------|--------|--------|----|
| gender                                |        |       |      |       |      | 0.04  | 1.04      | 0.05     | 0.85   | 0.40   |    |
| SeniorCitizen                         | 1.33   | -0.05 |      | 0.13  |      | 0.03  | 1.04      | 0.06     | 0.61   | 0.54   |    |
| Partner                               | 0.88   | -0.08 |      | 0.15  |      | -0.18 | 0.84      | 0.06     | -3.23  | <0.005 |    |
| Dependents                            | 9.67   | -0.29 |      | -0.07 |      | -0.09 | 0.91      | 0.07     | -1.31  | 0.19   |    |
| PhoneService                          | 2.40   | -0.23 |      | 0.05  |      | 0.83  | 2.29      | 0.47     | 1.75   | 0.08   |    |
| MultipleLines                         | 3.63   | -0.10 |      | 1.76  |      | 0.09  | 1.09      | 0.13     | 0.69   | 0.49   |    |
| OnlineSecurity                        | 1.03   | -0.16 |      | 0.33  |      | -0.21 | 0.81      | 0.13     | -1.60  | 0.11   |    |
| OnlineBackup                          | 3.20   | -0.47 |      | 0.05  |      | -0.06 | 0.95      | 0.13     | -0.44  | 0.66   |    |
| DeviceProtection                      | 0.60   | -0.31 |      | 0.19  |      | 0.09  | 1.09      | 0.13     | 0.69   | 0.49   |    |
| TechSupport                           | 1.03   | -0.16 |      | 0.34  |      | -0.08 | 0.92      | 0.13     | -0.64  | 0.52   |    |
| StreamingTV                           | 0.93   | -0.34 |      | 0.17  |      | 0.28  | 1.32      | 0.24     | 1.19   | 0.23   |    |
| StreamingMovies                       | 2.10   | -0.18 |      | 0.74  |      | 0.29  | 1.33      | 0.24     | 1.22   | 0.22   |    |
| PaperlessBilling                      | 2.16   | -0.18 |      | 0.75  |      | 0.15  | 1.16      | 0.06     | 2.65   | 0.01   |    |
| MonthlyCharges                        | 6.95   | 0.04  |      | 0.26  |      | 0.01  | 1.01      | 0.02     | 0.57   | 0.57   |    |
| TotalCharges                          | 0.82   | -0.03 |      | 0.06  |      | -0.00 | 1.00      | 0.00     | -39.16 | <0.005 |    |
| InternetService_Fiber                 | inf    | -0.00 |      | -0.00 |      | 1.02  | 2.77      | 0.58     | 1.76   | 0.08   |    |
| InternetService_No                    | 3.67   | -0.12 |      | 2.15  |      | -2.34 | 0.10      | 0.60     | -3.93  | <0.005 |    |
| Contract_One year                     | 13.51  | -3.51 |      | -1.17 |      | -1.27 | 0.28      | 0.10     | -12.55 | <0.005 |    |
| Contract_Two year                     | 117.58 | -1.46 |      | -1.07 |      | -3.70 | 0.02      | 0.20     | -18.32 | <0.005 |    |
| PaymentMethod_Credit card (automatic) | 246.60 | -4.10 |      | -3.31 |      | -0.01 | 0.99      | 0.09     | -0.13  | 0.90   |    |
| PaymentMethod_Electronic check        | 0.16   | -0.19 |      | 0.17  |      | 0.39  | 1.47      | 0.07     | 5.31   | <0.005 |    |
| PaymentMethod_Mailed check            | 23.13  | 0.24  |      | 0.53  |      | 0.51  | 1.67      | 0.09     | 5.87   | <0.005 |    |
|                                       | 27.74  | 0.34  |      | 0.68  |      |       |           |          |        |        |    |
| Concordance                           |        |       |      |       |      |       |           |          |        |        |    |
| Log-likelihood ratio test             |        |       |      |       |      |       |           |          |        |        |    |


```

In [6]: cph.score\_

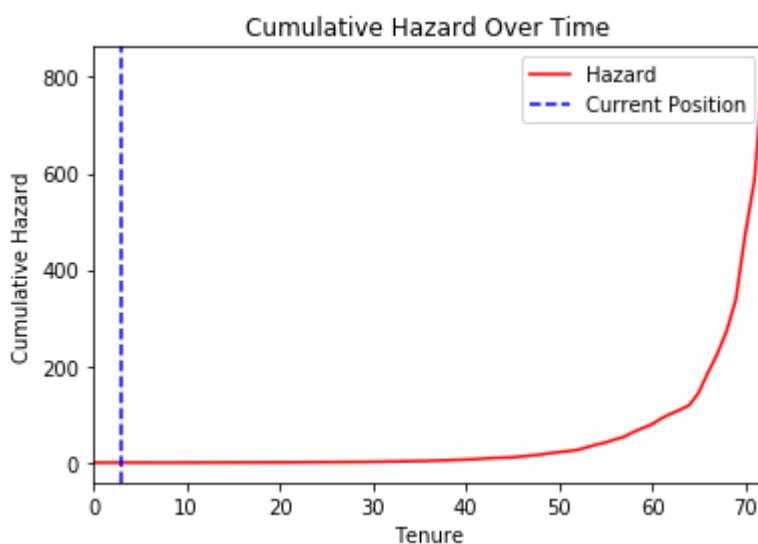
Out[6]: 0.9285636735265471

```
In [7]: fig, ax = plt.subplots(figsize = (10,7))
cph.plot(ax = ax);
```

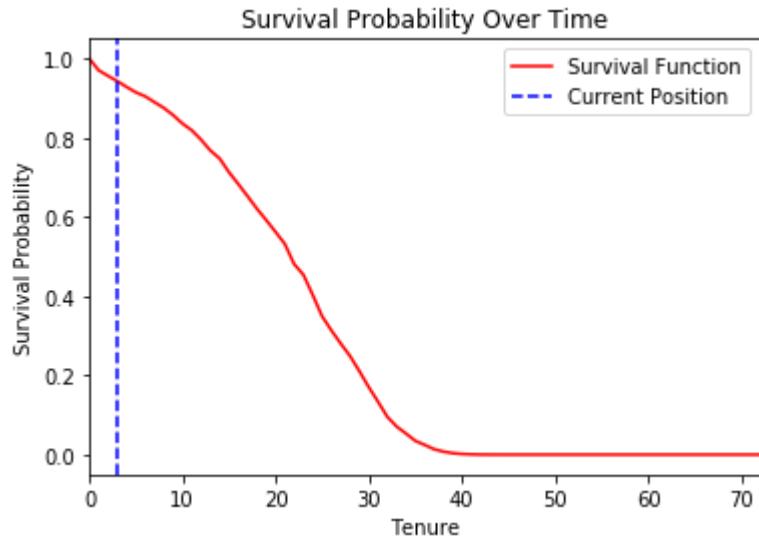


```
In [15]: test_id = regression_df.sample(1)
```

```
In [16]: fig, ax = plt.subplots()
cph.predict_cumulative_hazard(test_id).plot(ax = ax, color = 'red')
plt.axvline(x=test_id.tenure.values[0], color = 'blue', linestyle='--')
plt.legend(labels=['Hazard', 'Current Position'])
ax.set_xlabel('Tenure', size = 10)
ax.set_ylabel('Cumulative Hazard', size = 10)
ax.set_title('Cumulative Hazard Over Time');
```



```
In [17]: fig, ax = plt.subplots()
cph.predict_survival_function(test_id).plot(ax = ax, color = 'red')
plt.axvline(x=test_id.tenure.values[0], color = 'blue', linestyle='--')
plt.legend(labels=['Survival Function', 'Current Position'])
ax.set_xlabel('Tenure', size = 10)
ax.set_ylabel('Survival Probability', size = 10)
ax.set_title('Survival Probability Over Time');
```



Saving the model

```
In [8]: import pickle
pickle.dump(cph, open('survivemodel.pkl', 'wb'))
```

## Customer Lifetime Value

```
In [87]: def LTV(info):
    life = cph.predict_survival_function(info).reset_index()
    life.columns = ['Tenure', 'Probability']
    max_life = life.Tenure[life.Probability > 0.1].max()

    LTV = max_life * info['MonthlyCharges'].values[0]
    return LTV
```

```
In [89]: print('LTV of a testid is:', LTV(test_id), 'dollars.')
```

LTV of a testid is: 922.25 dollars.