

Run each cell one after the other.

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
from matplotlib import *
import sys
from sklearn.feature_selection import SelectKBest
from sklearn.feature_selection import chi2
from scipy.stats import bernoulli
from sklearn.model_selection import train_test_split
from sklearn.feature_selection import mutual_info_classif
import pandas as pd
from sklearn.feature_selection import chi2
import numpy as np
from sklearn.ensemble import ExtraTreesClassifier
```

```
dataframe_renewal=pd.read_csv("lease_renewal.csv")
dataframe_renewal.head()
```

```
   lease_id  no_rent_change  rent_change_10  rent_change_20
lease_length_2 \
0  HPA0001             0             0             0
0
1  HPA0002             0             0             0
0
2  HPA0003             0             0             0
0
3  HPA0004             0             0             0
0
4  HPA0005             0             0             0
0
```

```
   lease_length_3  lease_length_1  age_range_under_24  age_range_24_29
\
0             0             0             0             0
1             0             0             0             0
2             0             0             0             0
3             0             0             0             0
4             0             0             0             0
```

```
   age_range_30_39  age_range_40_49  age_range_50_59  age_range_60 \
```

0	0	0	0	0
1	0	0	0	0
2	0	0	0	0
3	0	0	0	0
4	0	0	0	0

	NoFinesViolations	PositiveSurvey	LatePayments	HOA_mandatory
Renewed				
0	0	1	1	0
1				
1	0	0	0	0
0				
2	0	1	0	0
0				
3	0	1	1	0
0				
4	0	0	0	0
0				

dataframe_renewal.describe()

	no_rent_change	rent_change_10	rent_change_20	lease_length_2
\				
count	79850.000000	79850.000000	79850.000000	79850.000000
mean	0.221428	0.023532	0.581866	0.245172
std	0.415210	0.151586	0.493256	0.430192
min	0.000000	0.000000	0.000000	0.000000
25%	0.000000	0.000000	0.000000	0.000000
50%	0.000000	0.000000	1.000000	0.000000
75%	0.000000	0.000000	1.000000	0.000000
max	1.000000	1.000000	1.000000	1.000000

	lease_length_3	lease_length_1	age_range_under_24
age_range_24_29 \			
count	79850.000000	79850.000000	79850.000000
79850.000000			
mean	0.057495	0.524859	0.038309
0.091947			
std	0.232788	0.499385	0.191943
0.288953			
min	0.000000	0.000000	0.000000
0.000000			

25%	0.000000	0.000000	0.000000
0.000000			
50%	0.000000	1.000000	0.000000
0.000000			
75%	0.000000	1.000000	0.000000
0.000000			
max	1.000000	1.000000	1.000000
1.000000			

	age_range_30_39	age_range_40_49	age_range_50_59	age_range_60
\				
count	79850.000000	79850.000000	79850.000000	79850.000000
mean	0.149192	0.108604	0.059136	0.020977
std	0.356280	0.311143	0.235880	0.143308
min	0.000000	0.000000	0.000000	0.000000
25%	0.000000	0.000000	0.000000	0.000000
50%	0.000000	0.000000	0.000000	0.000000
75%	0.000000	0.000000	0.000000	0.000000
max	1.000000	1.000000	1.000000	1.000000

	NoFinesViolations	PositiveSurvey	LatePayments	HOA_mandatory
\				
count	79850.000000	79850.000000	79850.000000	79850.000000
mean	0.139249	0.269142	0.566399	0.164133
std	0.346208	0.443517	0.495575	0.370398
min	0.000000	0.000000	0.000000	0.000000
25%	0.000000	0.000000	0.000000	0.000000
50%	0.000000	0.000000	1.000000	0.000000
75%	0.000000	1.000000	1.000000	0.000000
max	1.000000	1.000000	1.000000	1.000000

Renewed

```
count    79850.000000
mean         0.195892
std         0.396888
min         0.000000
25%         0.000000
50%         0.000000
75%         0.000000
max         1.000000
```

```
dataframe_renewal.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 79850 entries, 0 to 79849
```

```
Data columns (total 18 columns):
```

#	Column	Non-Null Count	Dtype
0	lease_id	79850 non-null	object
1	no_rent_change	79850 non-null	int64
2	rent_change_10	79850 non-null	int64
3	rent_change_20	79850 non-null	int64
4	lease_length_2	79850 non-null	int64
5	lease_length_3	79850 non-null	int64
6	lease_length_1	79850 non-null	int64
7	age_range_under_24	79850 non-null	int64
8	age_range_24_29	79850 non-null	int64
9	age_range_30_39	79850 non-null	int64
10	age_range_40_49	79850 non-null	int64
11	age_range_50_59	79850 non-null	int64
12	age_range_60	79850 non-null	int64
13	NoFinesViolations	79850 non-null	int64
14	PositiveSurvey	79850 non-null	int64
15	LatePayments	79850 non-null	int64
16	H0A_mandatory	79850 non-null	int64
17	Renewed	79850 non-null	int64

```
dtypes: int64(17), object(1)
```

```
memory usage: 11.0+ MB
```

```
import pandas_profiling
```

```
dataframe_renewal.profile_report()
```

```
{"version_major":2,"version_minor":0,"model_id":"8346ea3721d04578b6dc45e92388cc7d"}
```

```
{"version_major":2,"version_minor":0,"model_id":"65df5a7935c64a5a9b20957d84c5bdf5"}
```

```
{"version_major":2,"version_minor":0,"model_id":"e58cbcf8c691458097a294976b4b213f"}
```

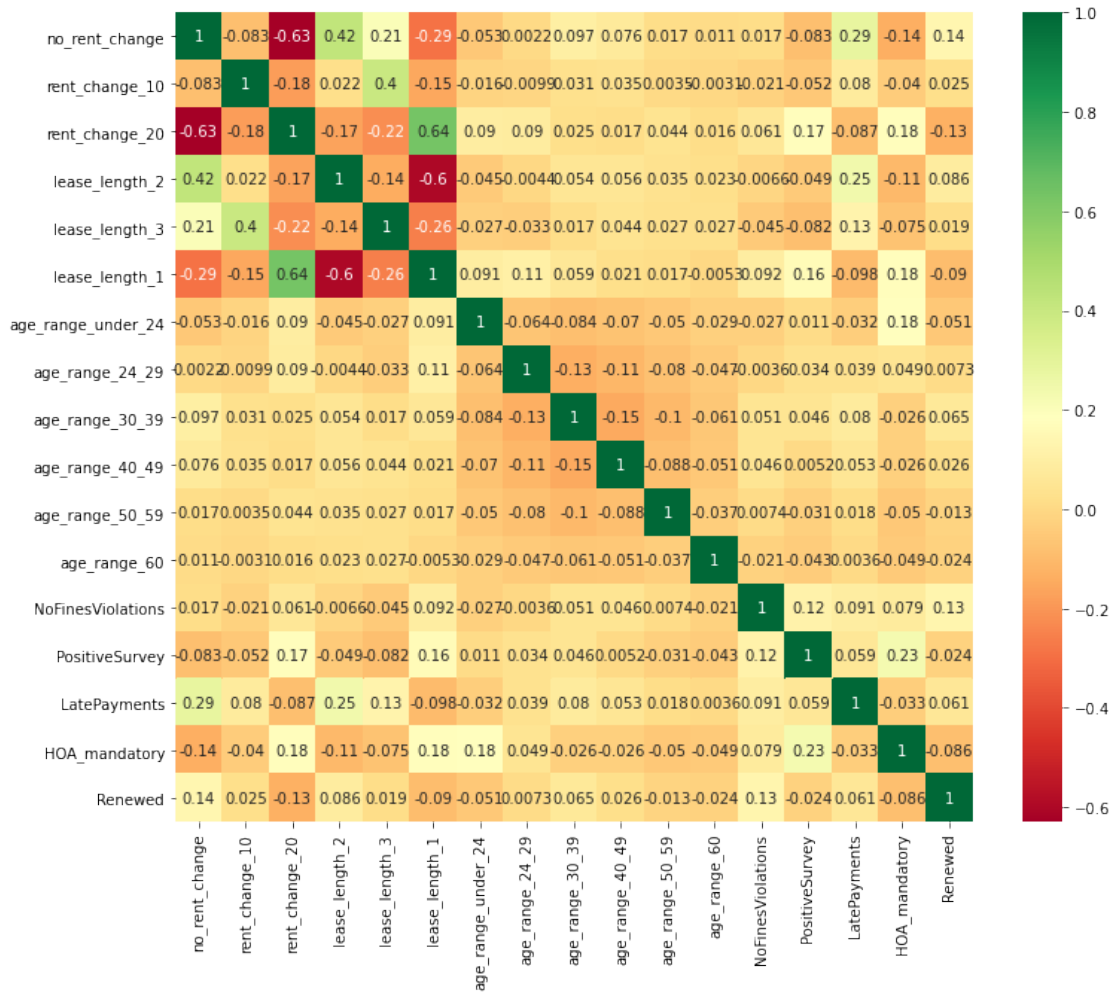
```
<IPython.core.display.HTML object>
```

```
dataframe_renewal.isnull().sum()
```

```
lease_id            0
no_rent_change      0
rent_change_10      0
rent_change_20      0
lease_length_2      0
lease_length_3      0
lease_length_1      0
age_range_under_24  0
age_range_24_29     0
age_range_30_39     0
age_range_40_49     0
age_range_50_59     0
age_range_60        0
NoFinesViolations  0
PositiveSurvey      0
LatePayments        0
HOA_mandatory       0
Renewed             0
dtype: int64
```

HeatMap visualization

```
plt.figure(figsize=(12,10)) # on this line I just set the size of  
figure to 12 by 10.  
p=sns.heatmap(dataframe_renewal.corr(), annot=True,cmap='RdYlGn') #  
seaborn has very simple solution for heatmap
```



Observations:

1) Those who were on their lease for first term were most likely had an increase of 20 ppercentage of rent

2) Those who were on their lease for second term were most liekly had no increase in their rent.

3) Those who were on their lease for third term were most likely had an increase of 10 percenatge of rent.

Bar graph to show the count of every value of each independent features to that of the dependent feature.

```
for x in range(1,len(dataframe_renewal.columns)-1):
```

```
pd.crosstab(dataframe_renewal.iloc[:,x],dataframe_renewal['Renewed']).  
plot(kind="bar",stacked=True)  
print(dataframe_renewal.iloc[:,x].value_counts())
```

```
0    62169
```

```
1    17681
```

```
Name: no_rent_change, dtype: int64
```

```
0    77971
```

```
1     1879
```

```
Name: rent_change_10, dtype: int64
```

```
1    46462
```

```
0    33388
```

```
Name: rent_change_20, dtype: int64
```

```
0    60273
```

```
1    19577
```

```
Name: lease_length_2, dtype: int64
```

```
0    75259
```

```
1     4591
```

```
Name: lease_length_3, dtype: int64
```

```
1    41910
```

```
0    37940
```

```
Name: lease_length_1, dtype: int64
```

```
0    76791
```

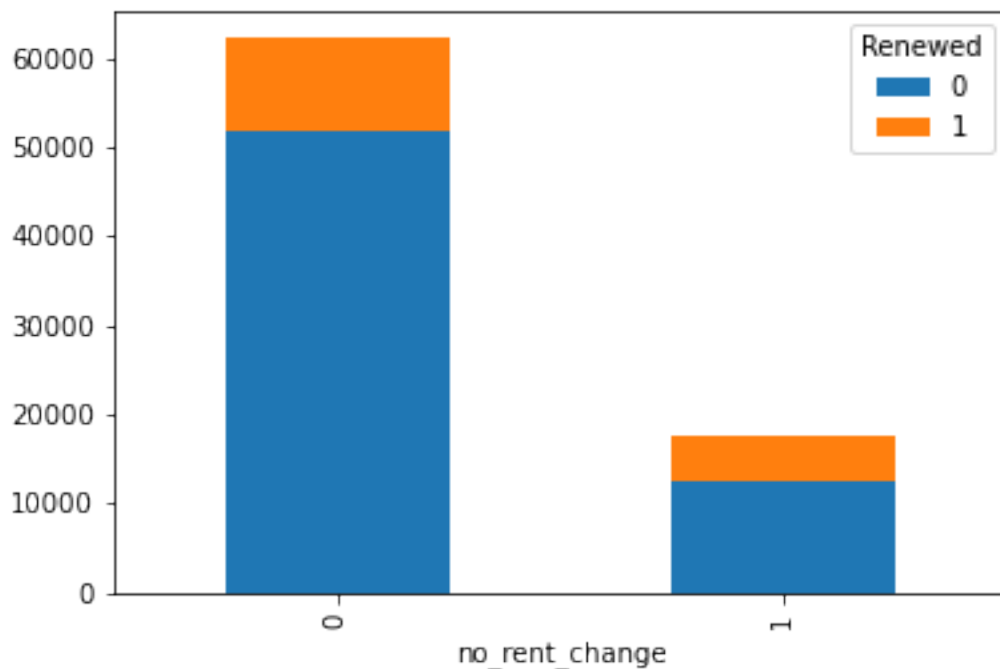
```
1     3059
```

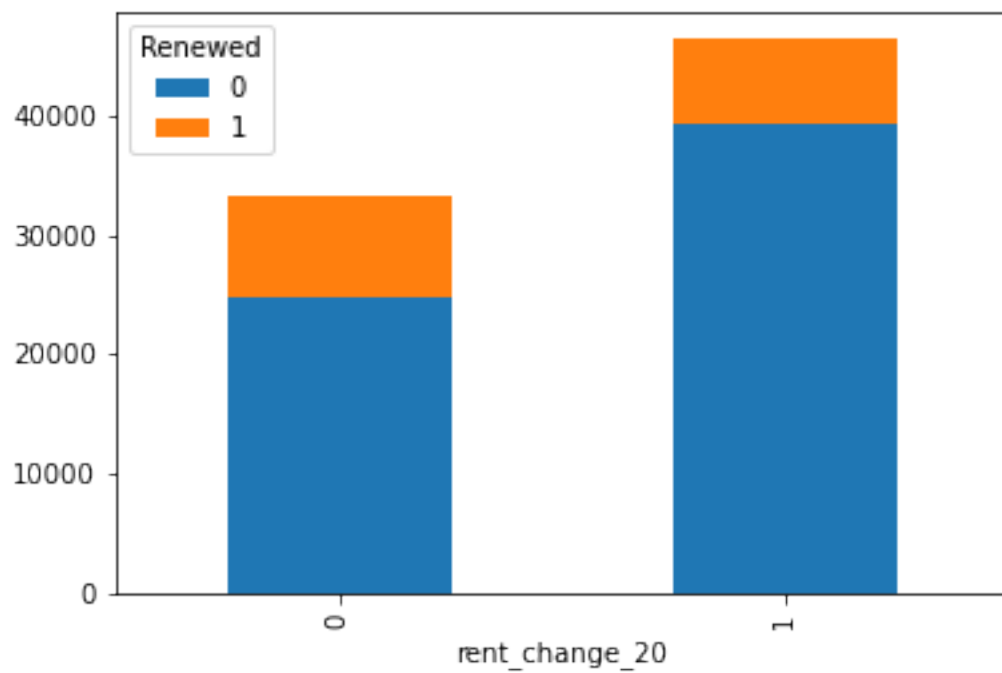
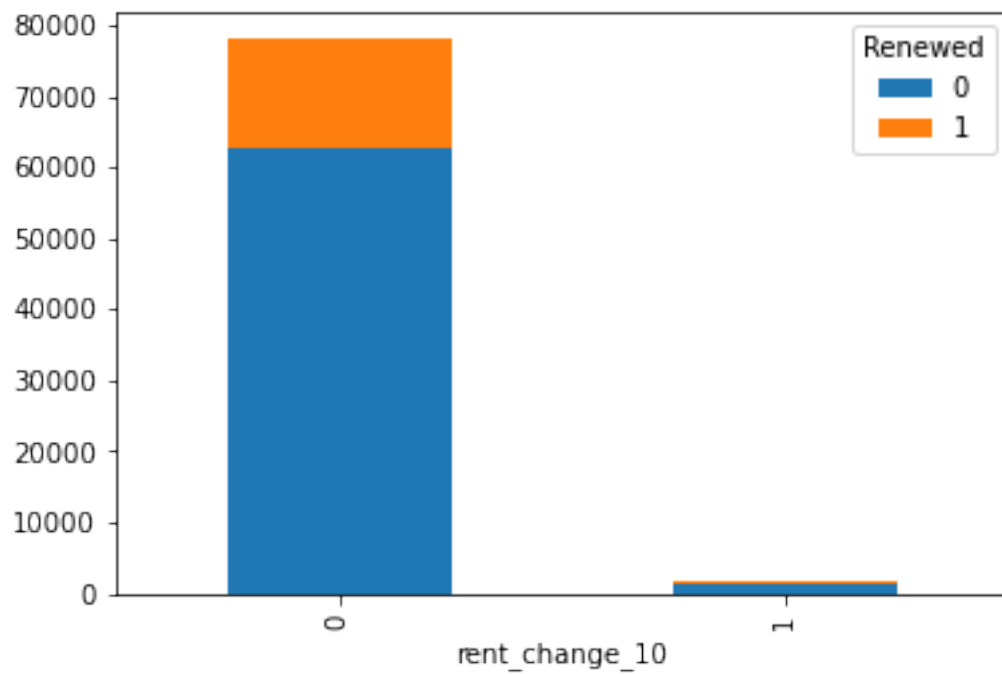
```
Name: age_range_under_24, dtype: int64
```

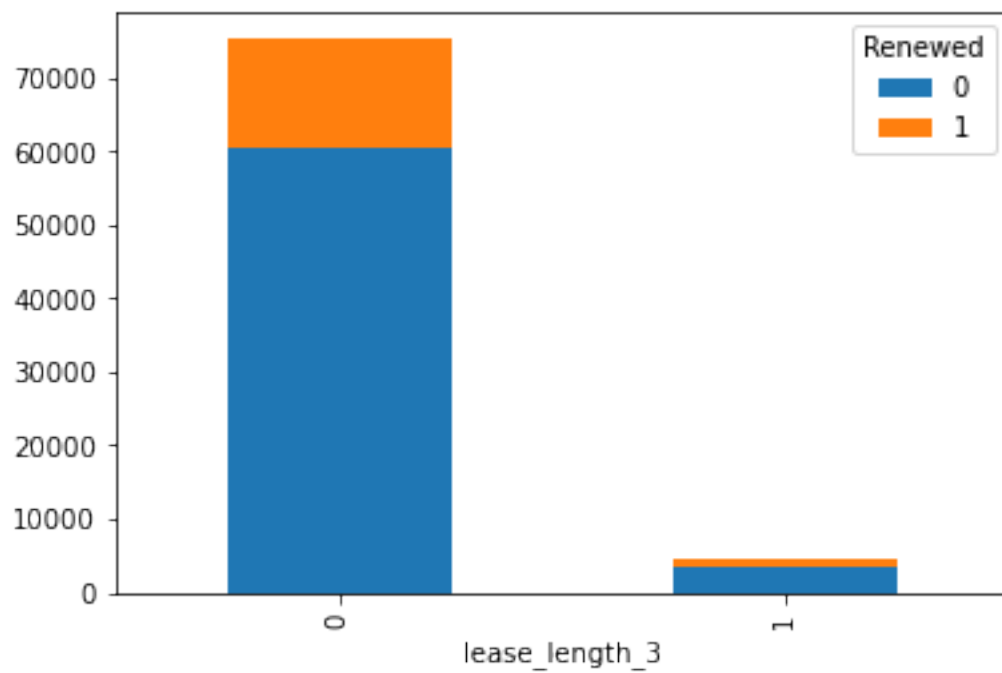
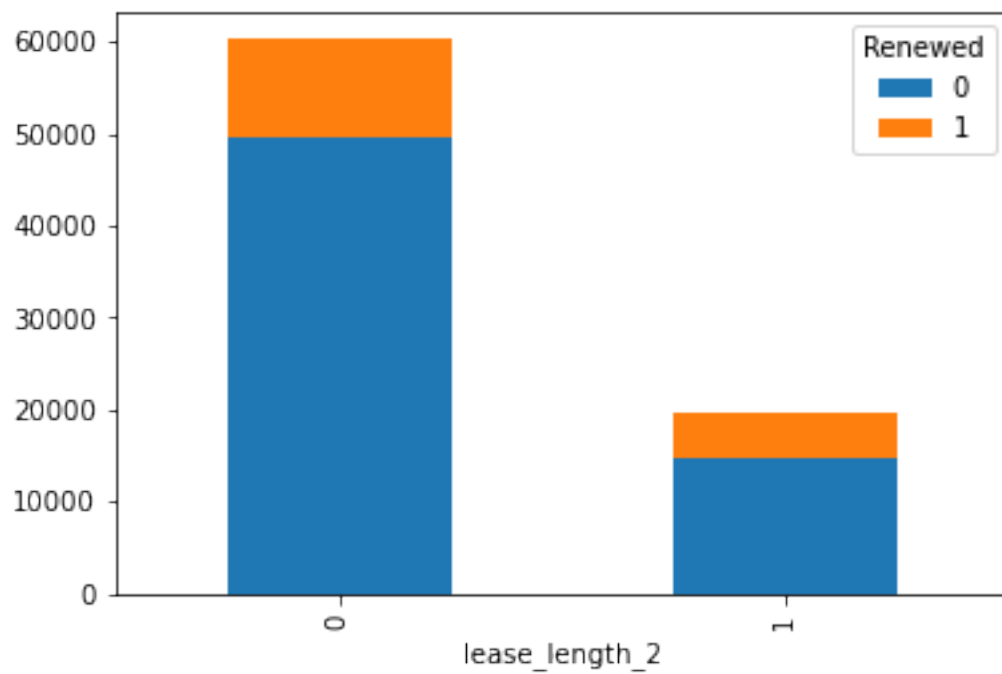
```
0    72508
```

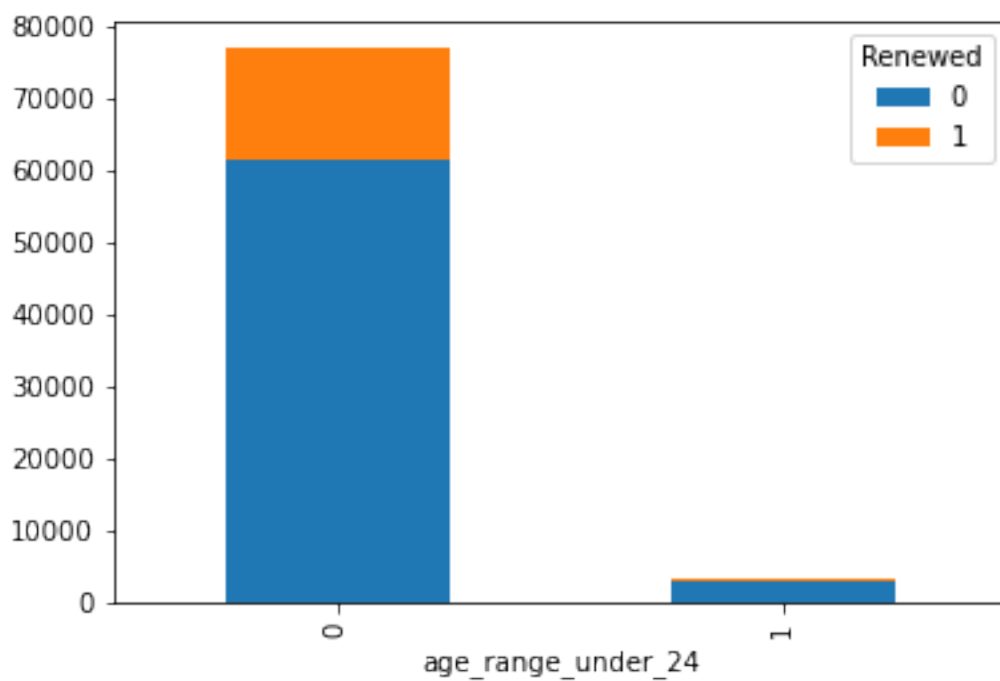
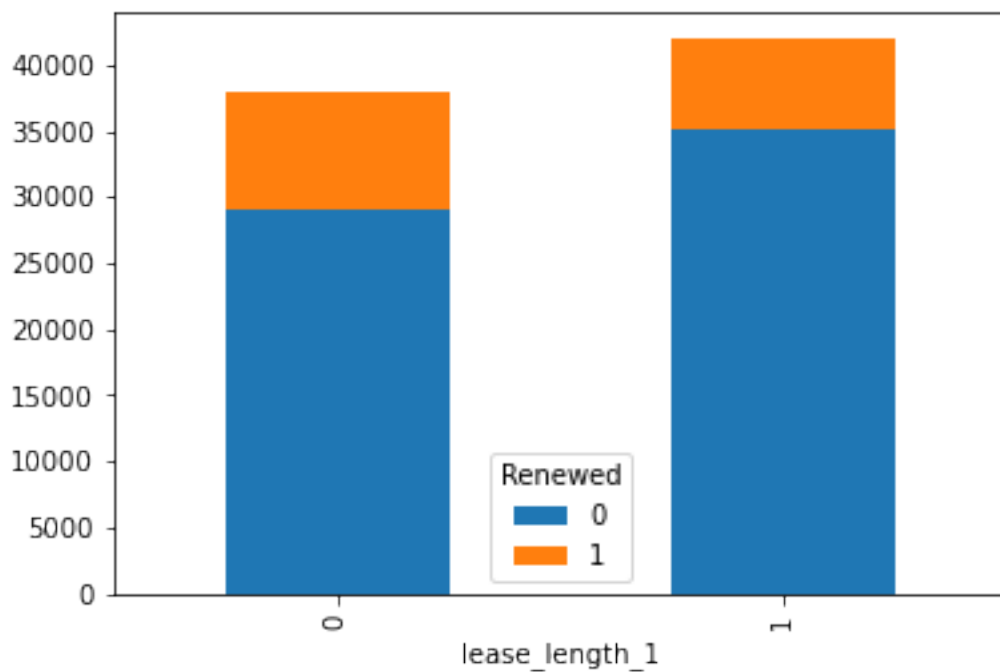
```
1     7342
```

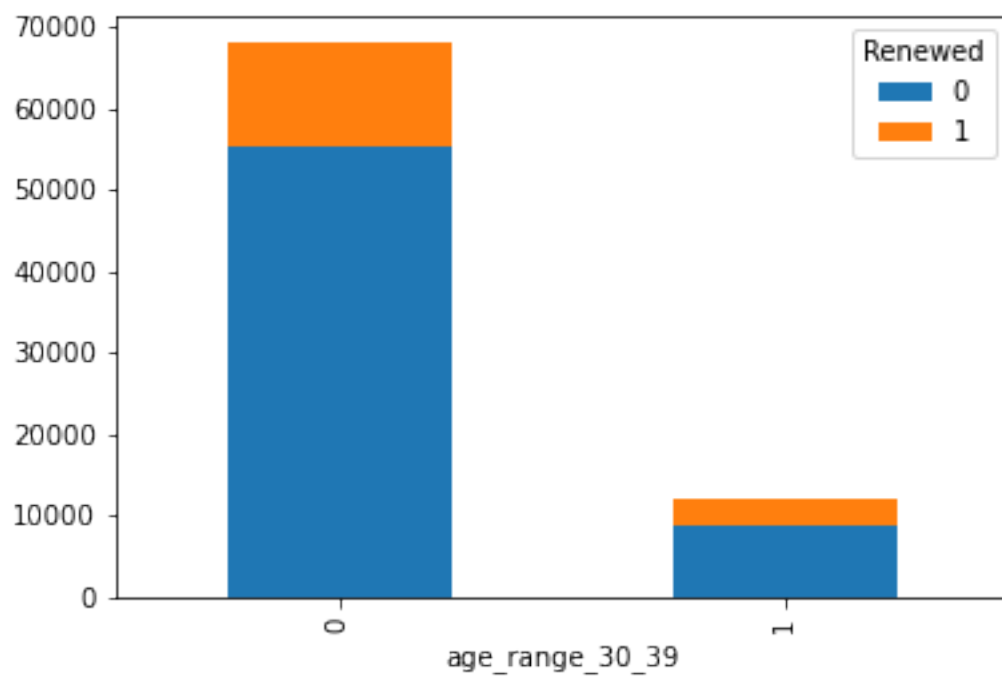
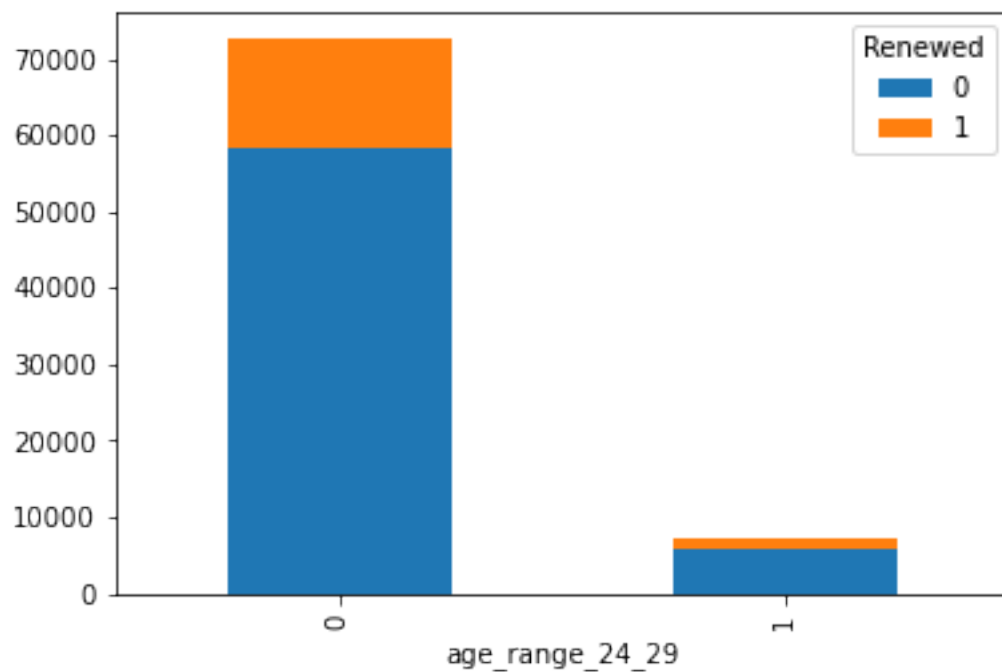
Name: age_range_24_29, dtype: int64
0 67937
1 11913
Name: age_range_30_39, dtype: int64
0 71178
1 8672
Name: age_range_40_49, dtype: int64
0 75128
1 4722
Name: age_range_50_59, dtype: int64
0 78175
1 1675
Name: age_range_60, dtype: int64
0 68731
1 11119
Name: NoFinesViolations, dtype: int64
0 58359
1 21491
Name: PositiveSurvey, dtype: int64
1 45227
0 34623
Name: LatePayments, dtype: int64
0 66744
1 13106
Name: HOA_mandatory, dtype: int64

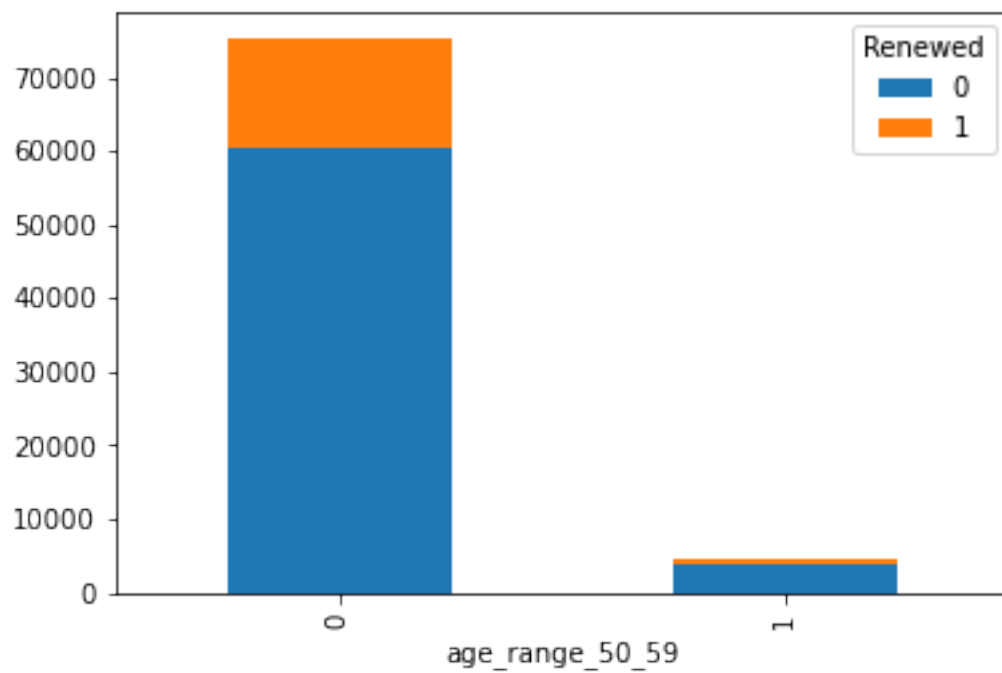
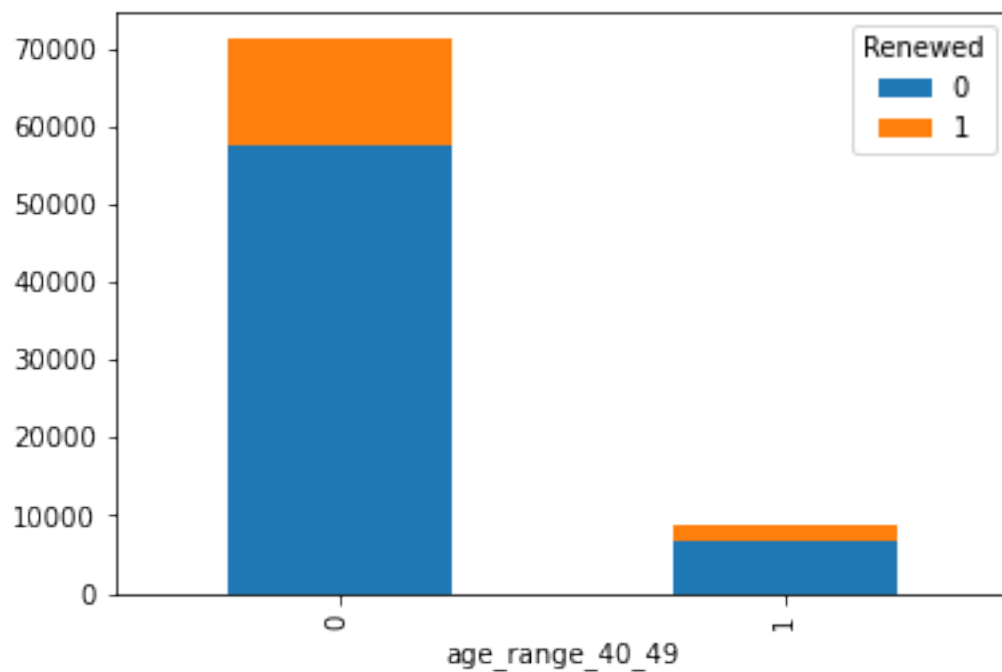


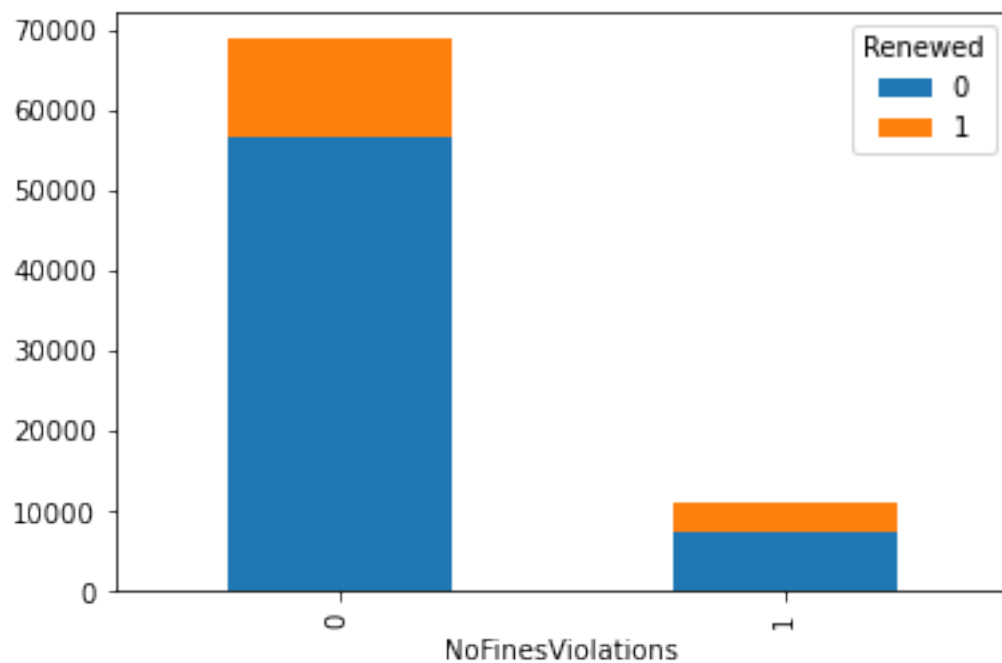
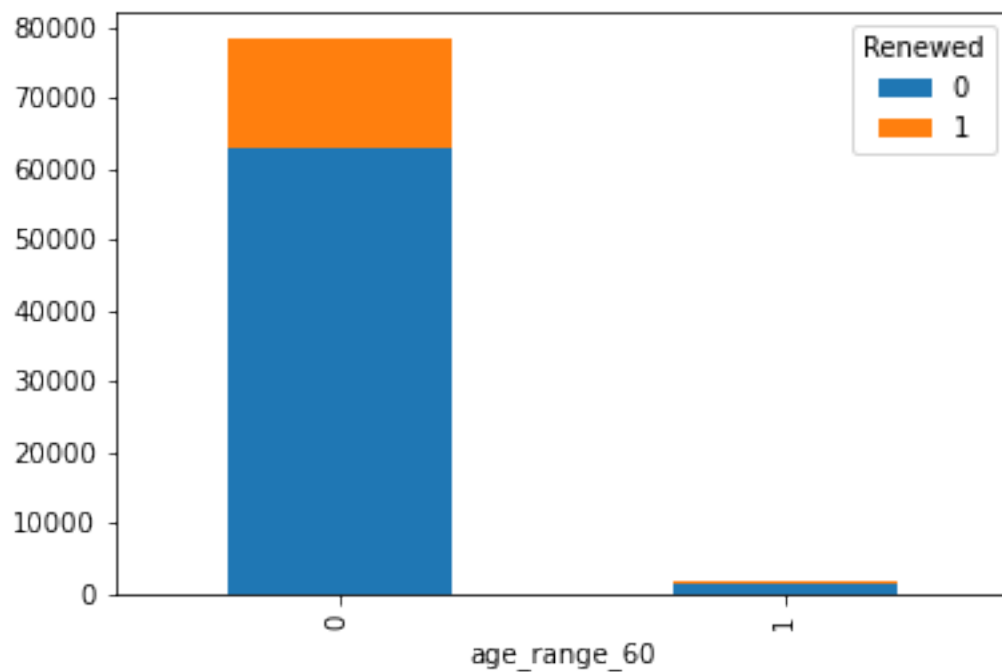


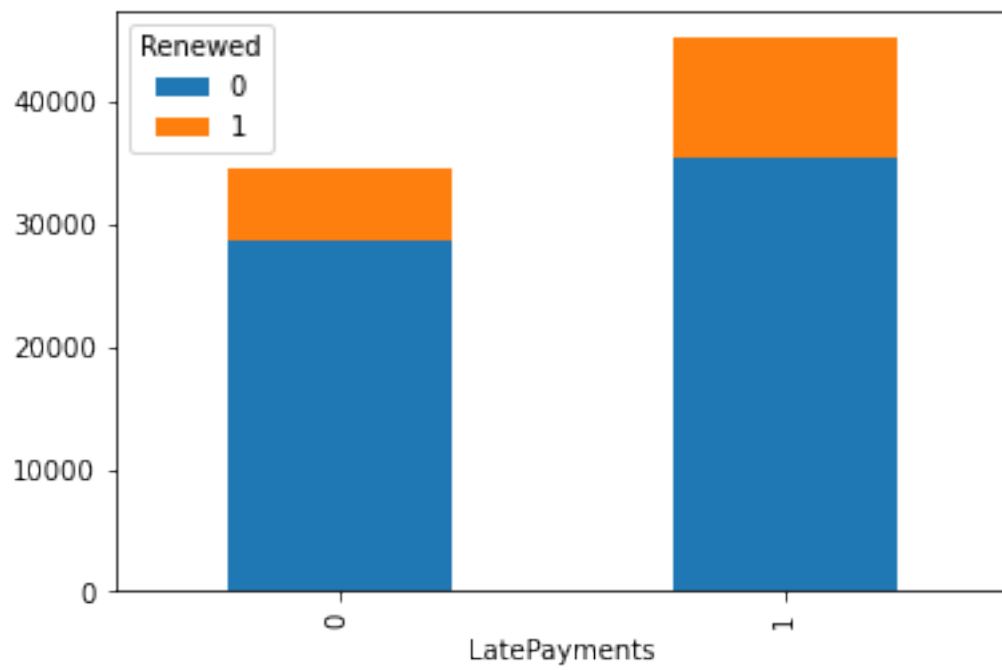
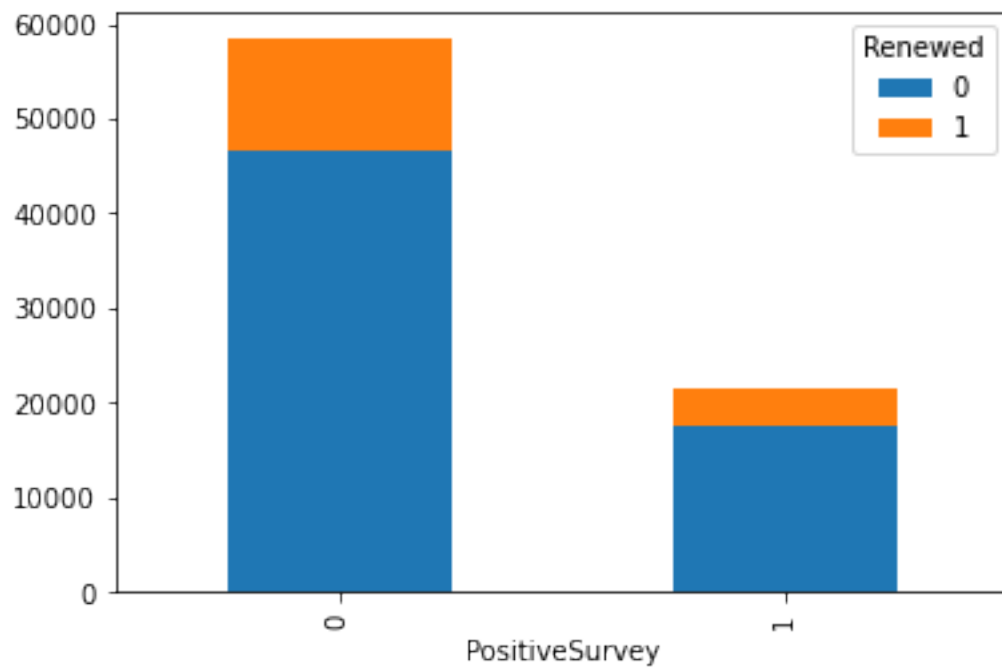


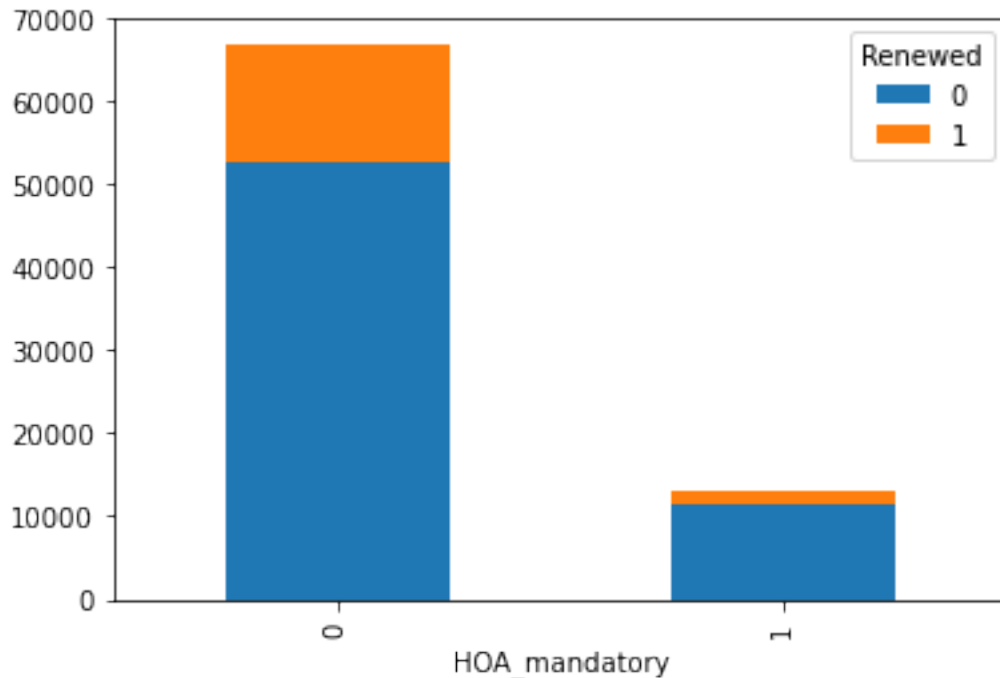






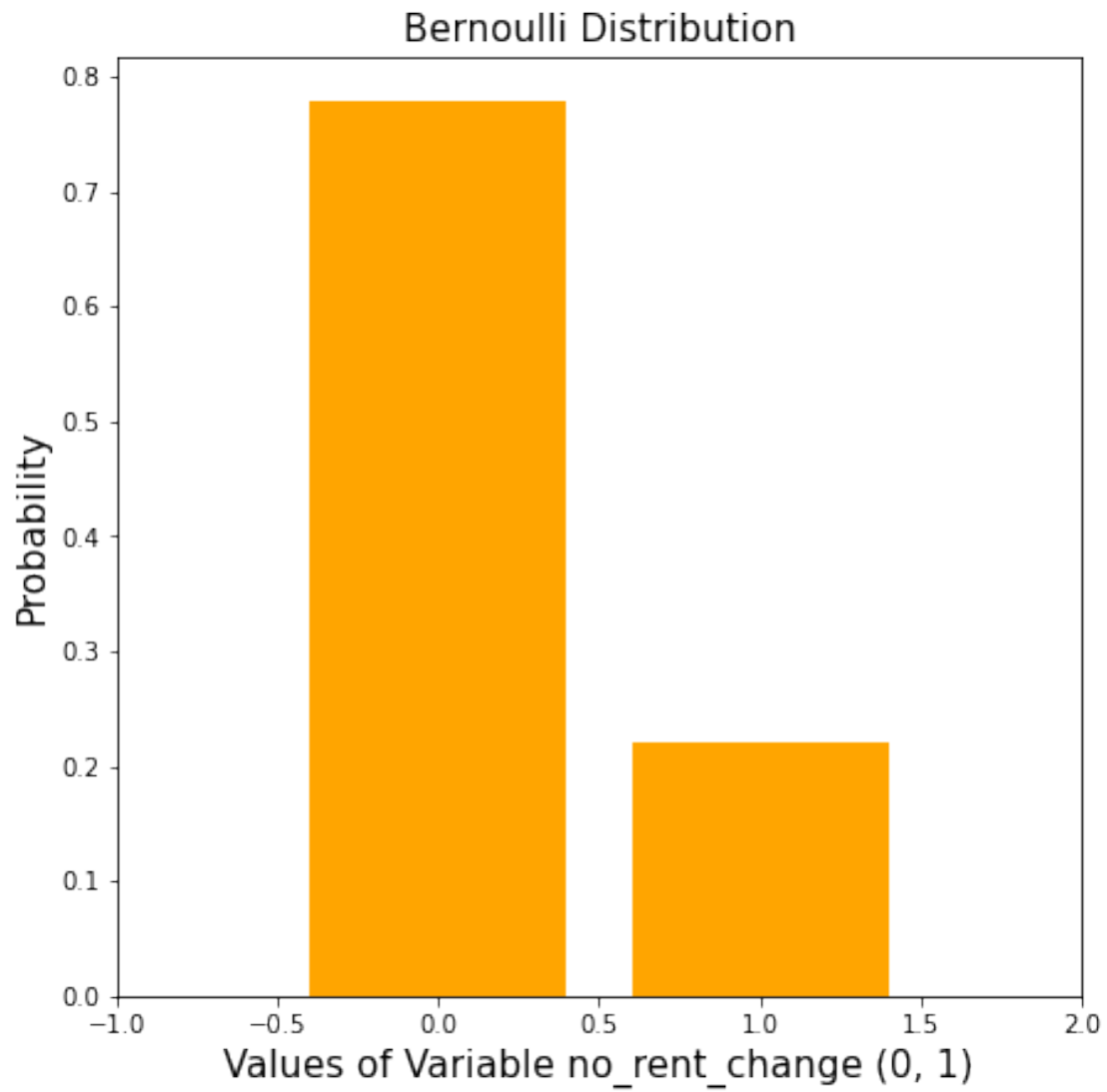


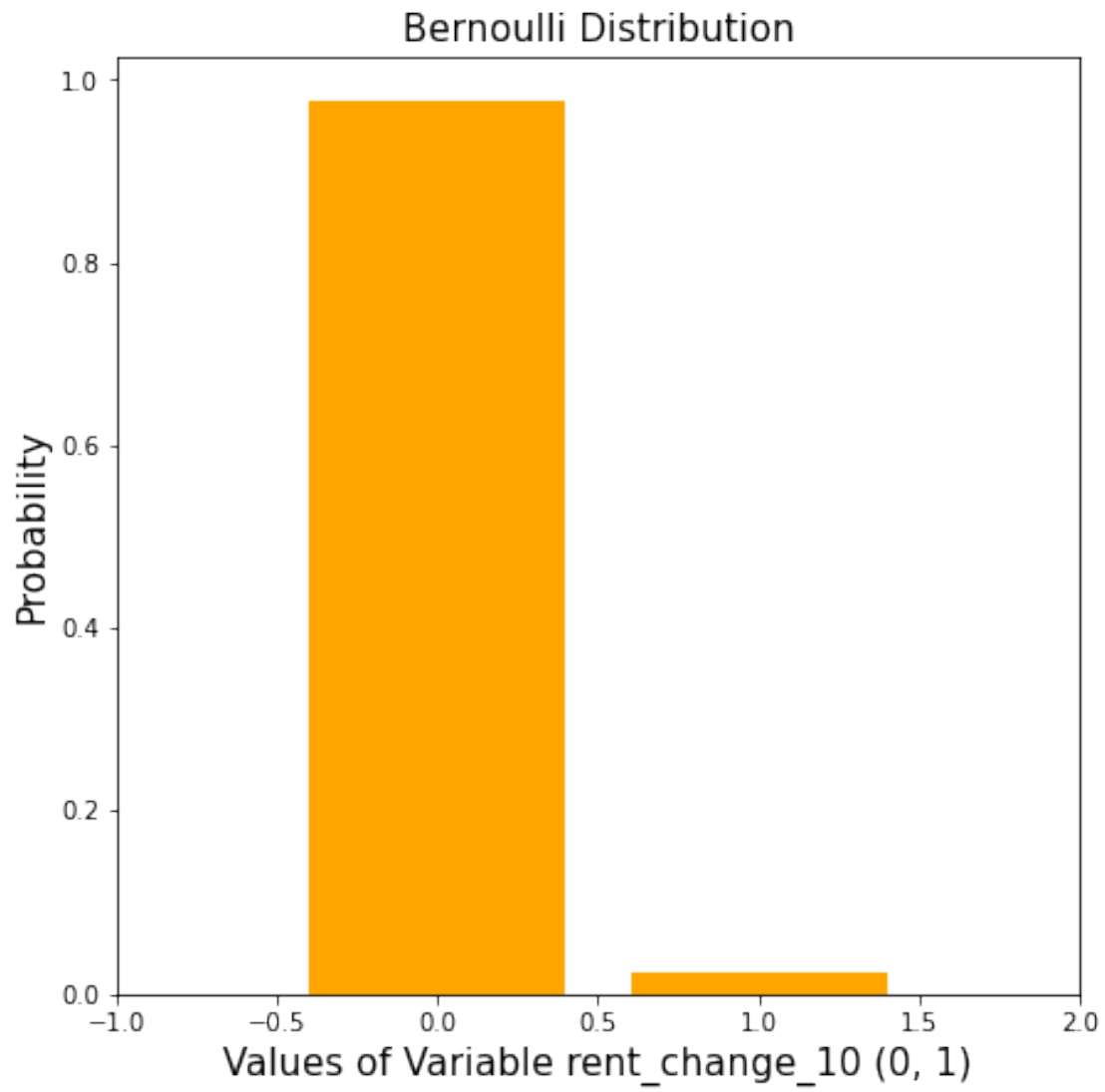


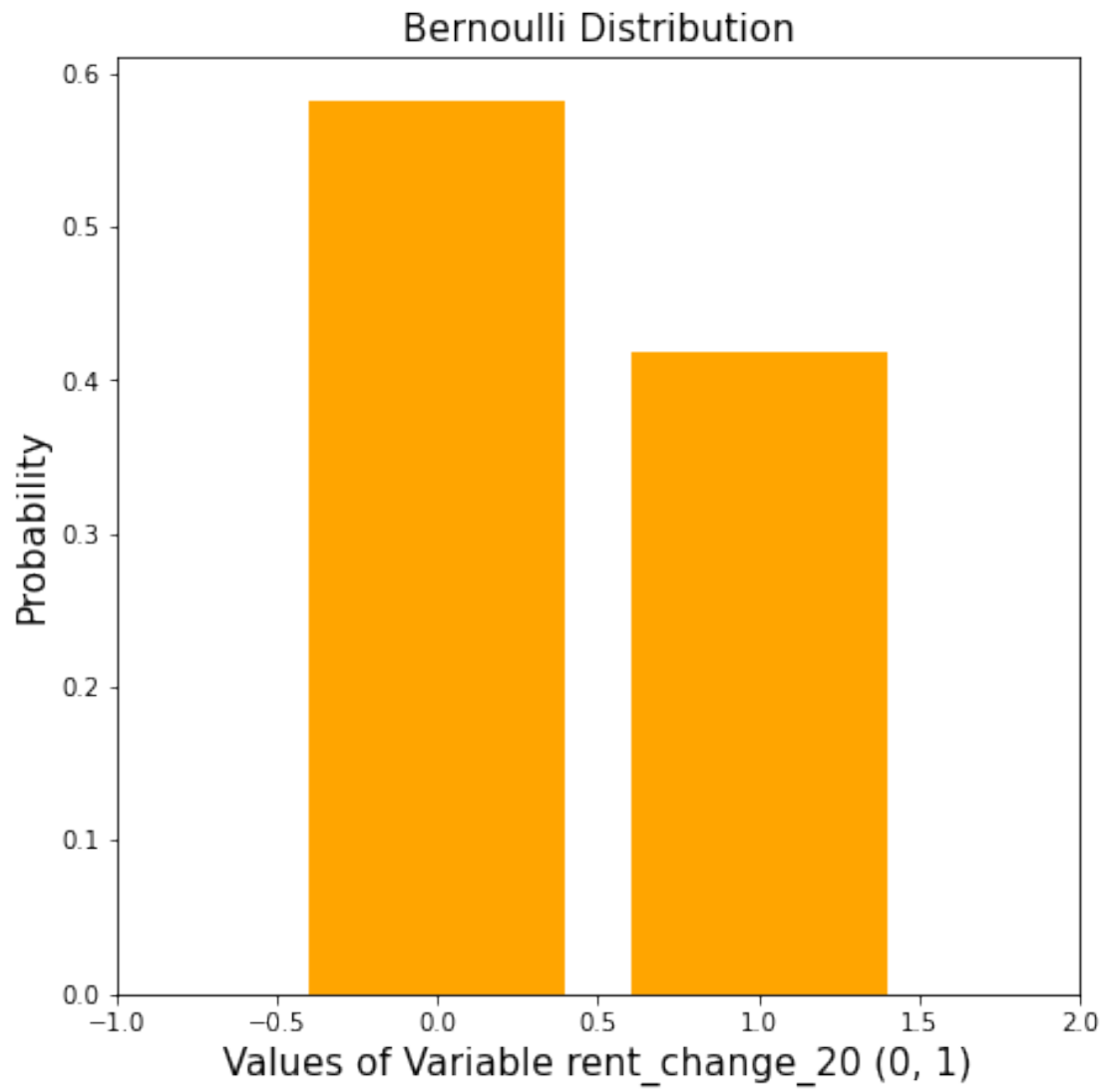


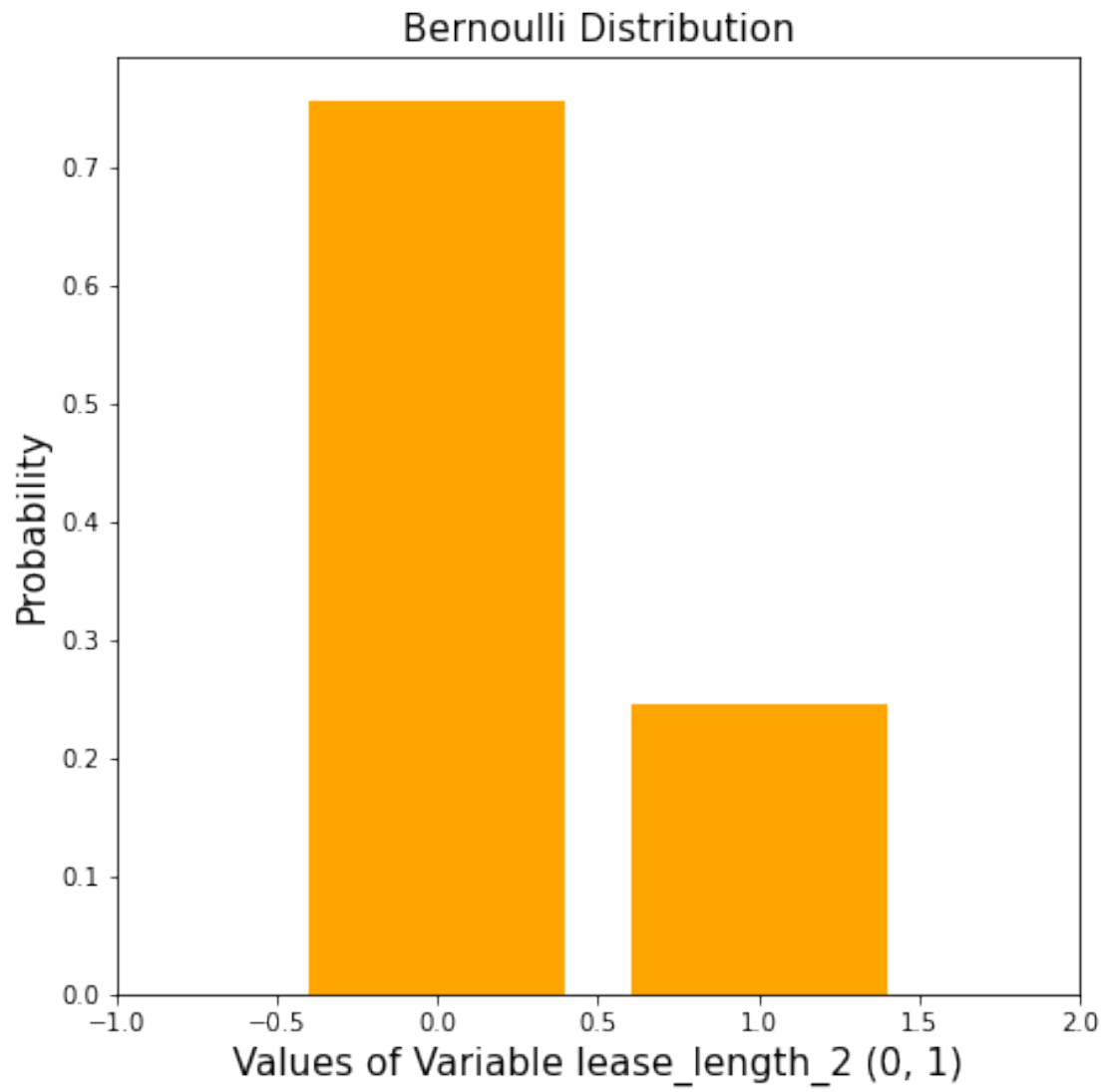
plotting the probability of each features of 0's and 1's.

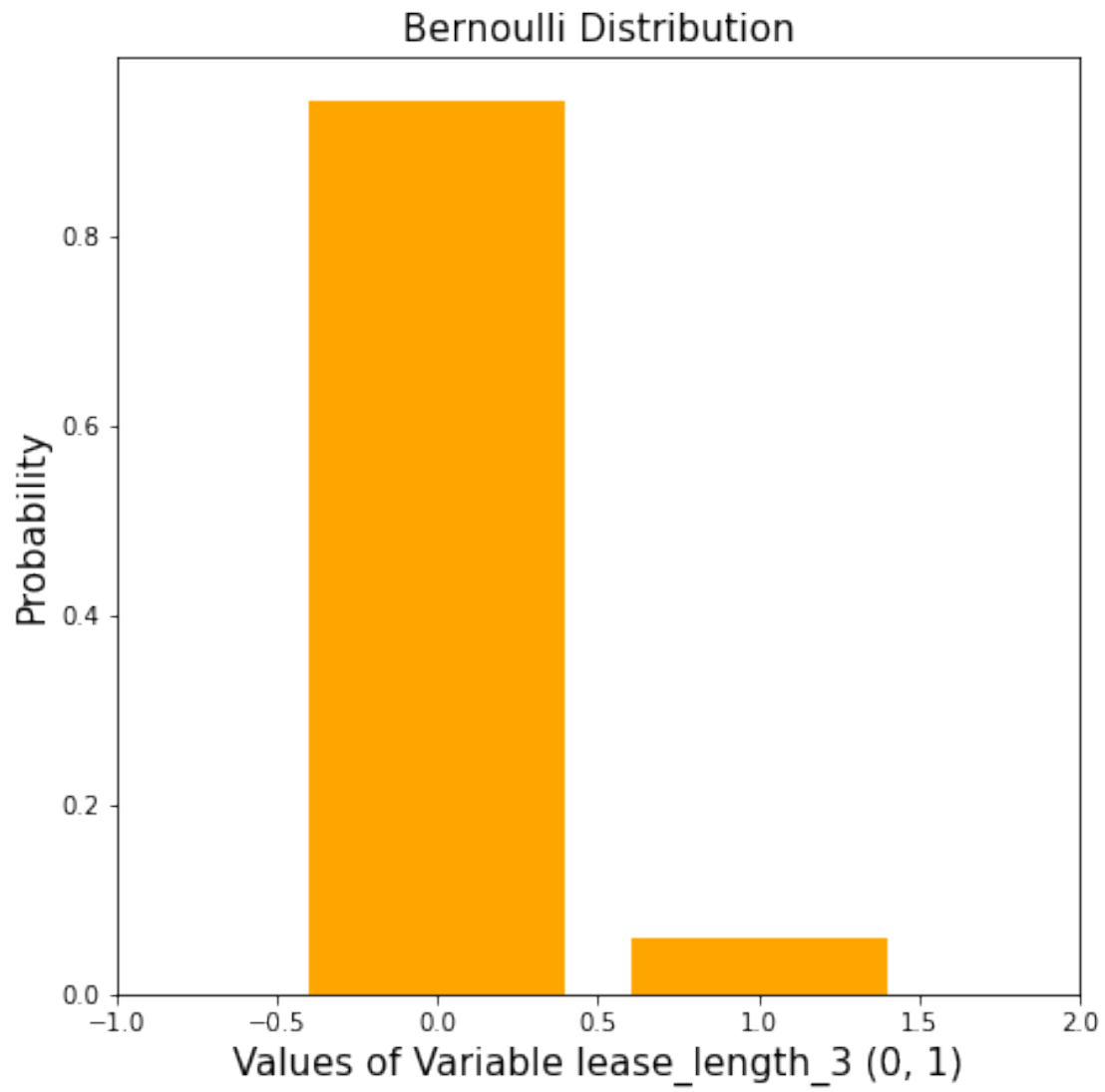
```
for x in dataframe_renewal.columns[1:]:
    val=dataframe_renewal[x].value_counts().to_list()[1]
    bd=bernoulli(val/79850)
    uniqueval= dataframe_renewal[x].unique().tolist()
    plt.figure(figsize=(7,7))
    plt.xlim(-1, 2)
    plt.bar(uniqueval, bd.pmf(uniqueval), color='orange')
    plt.title('Bernoulli Distribution', fontsize='15')
    plt.xlabel('Values of Variable '+x+' (0, 1)', fontsize='15')
    plt.ylabel('Probability', fontsize='15')
    plt.show()
```

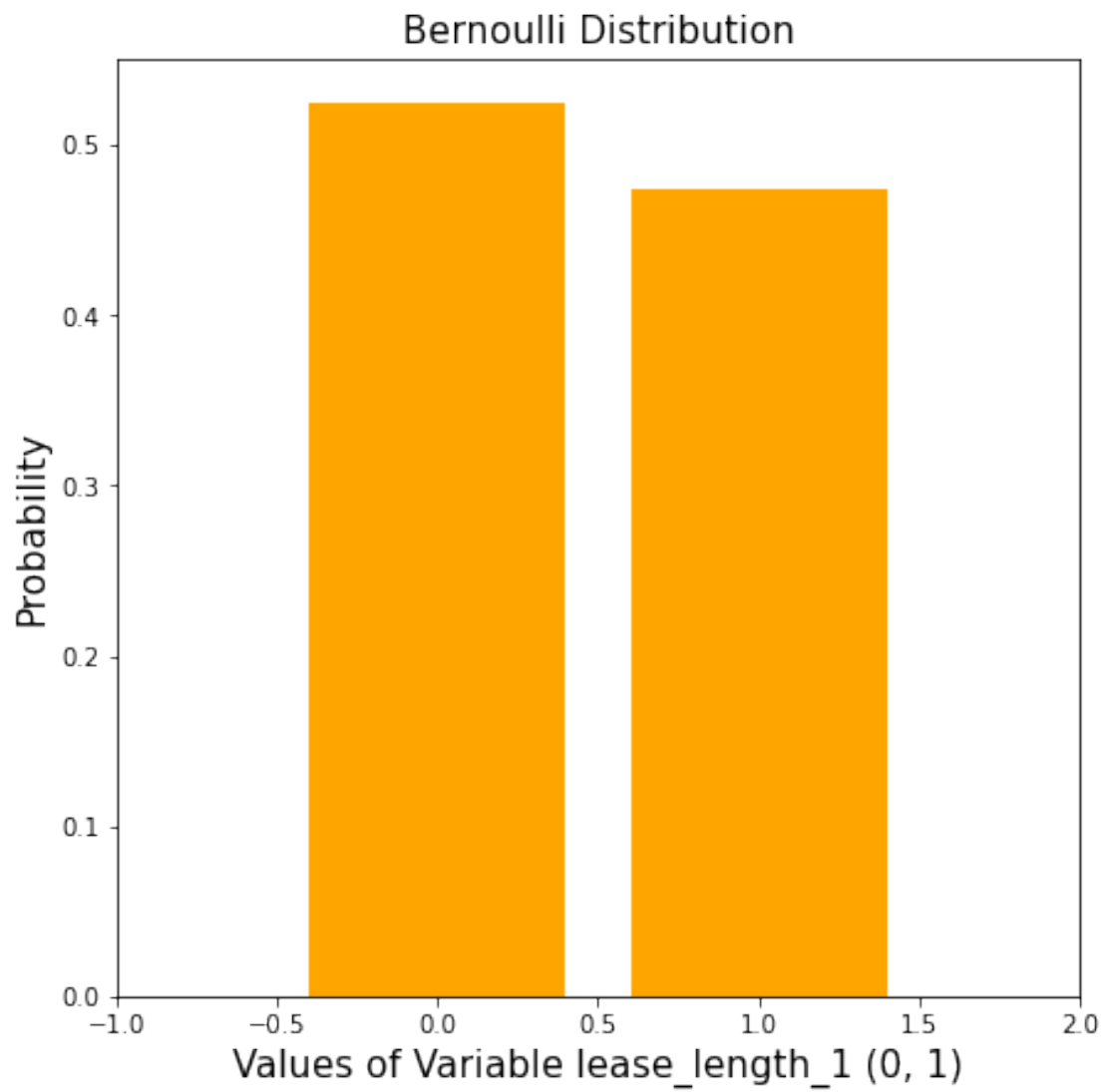



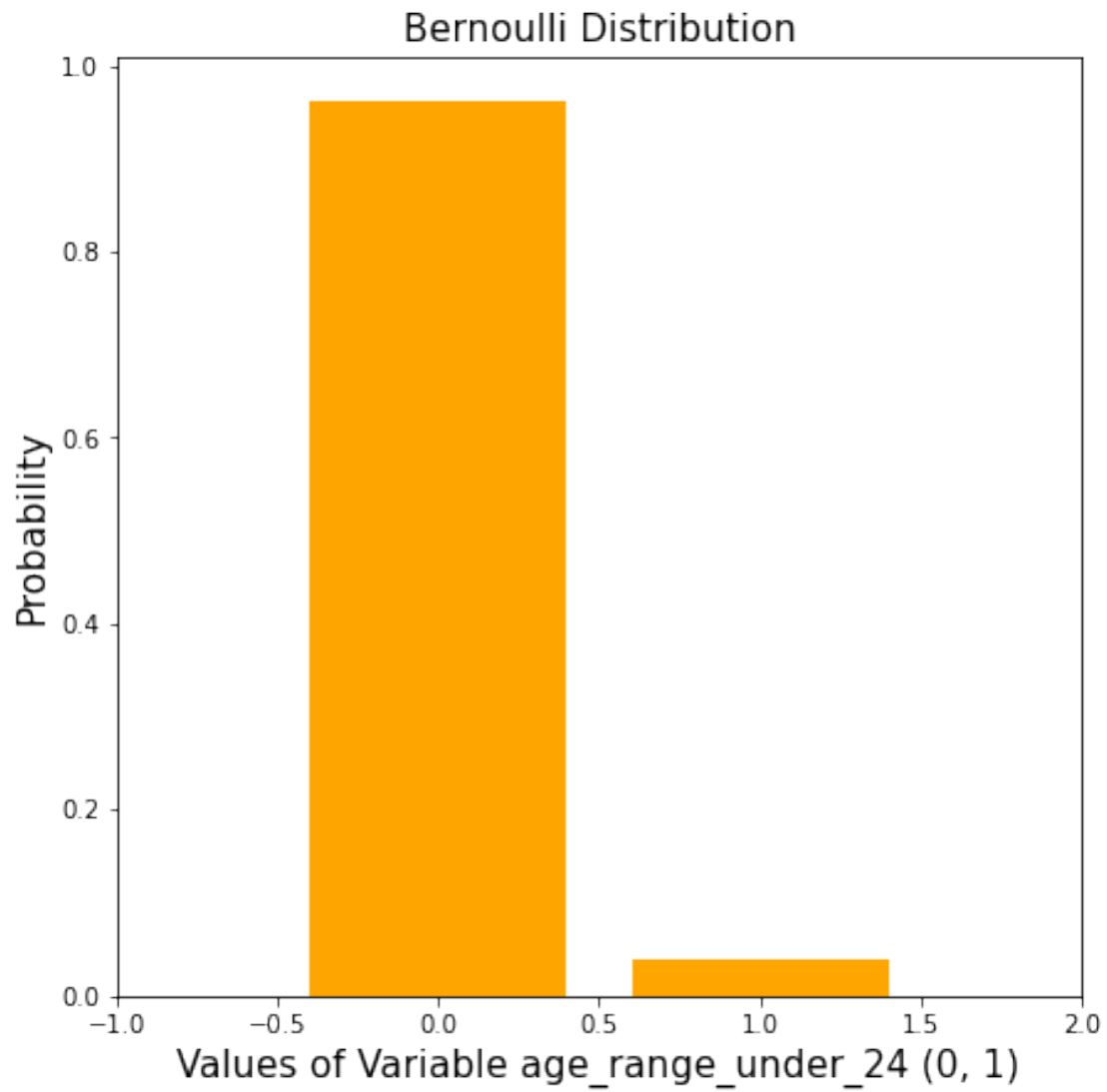


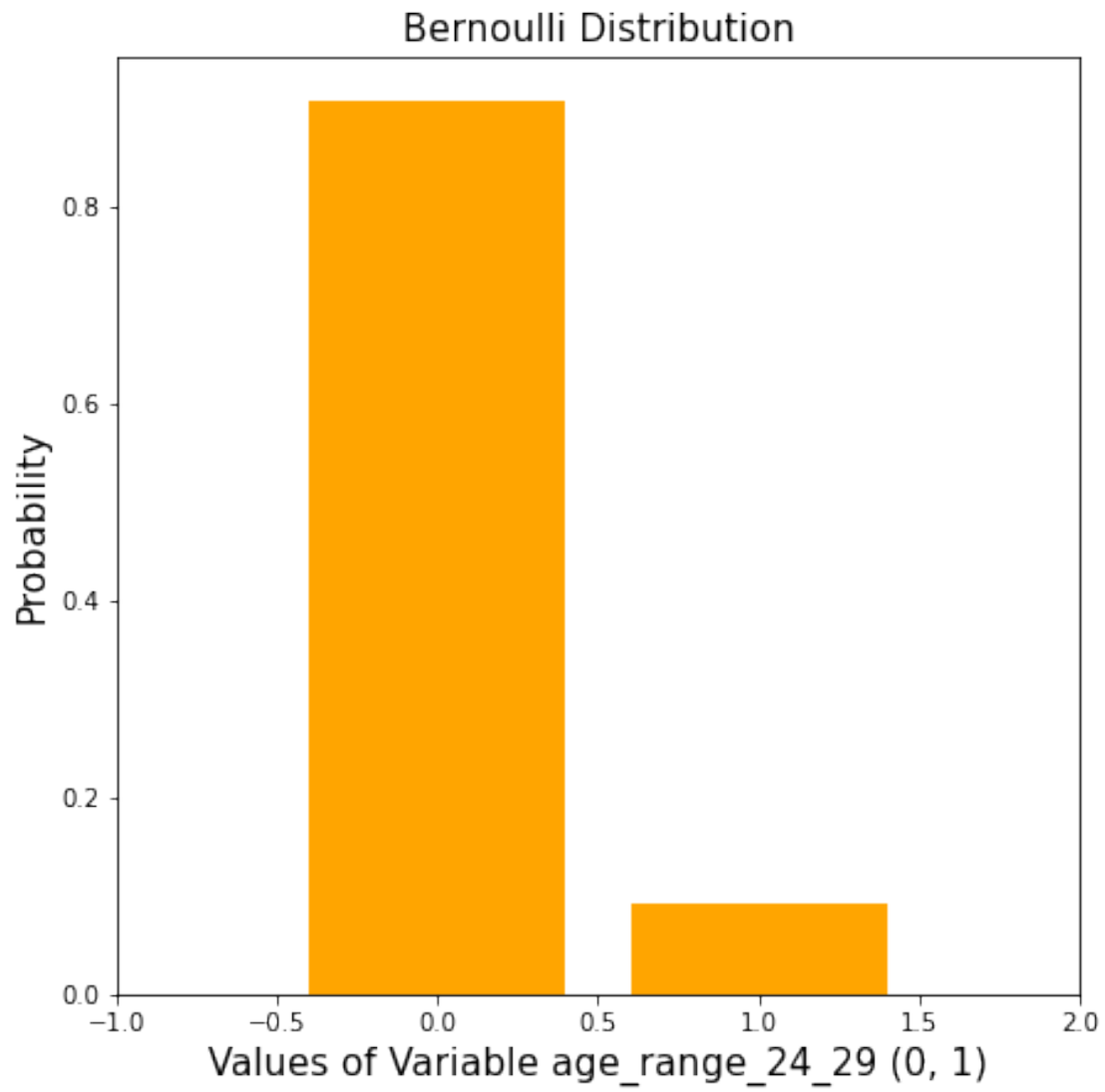


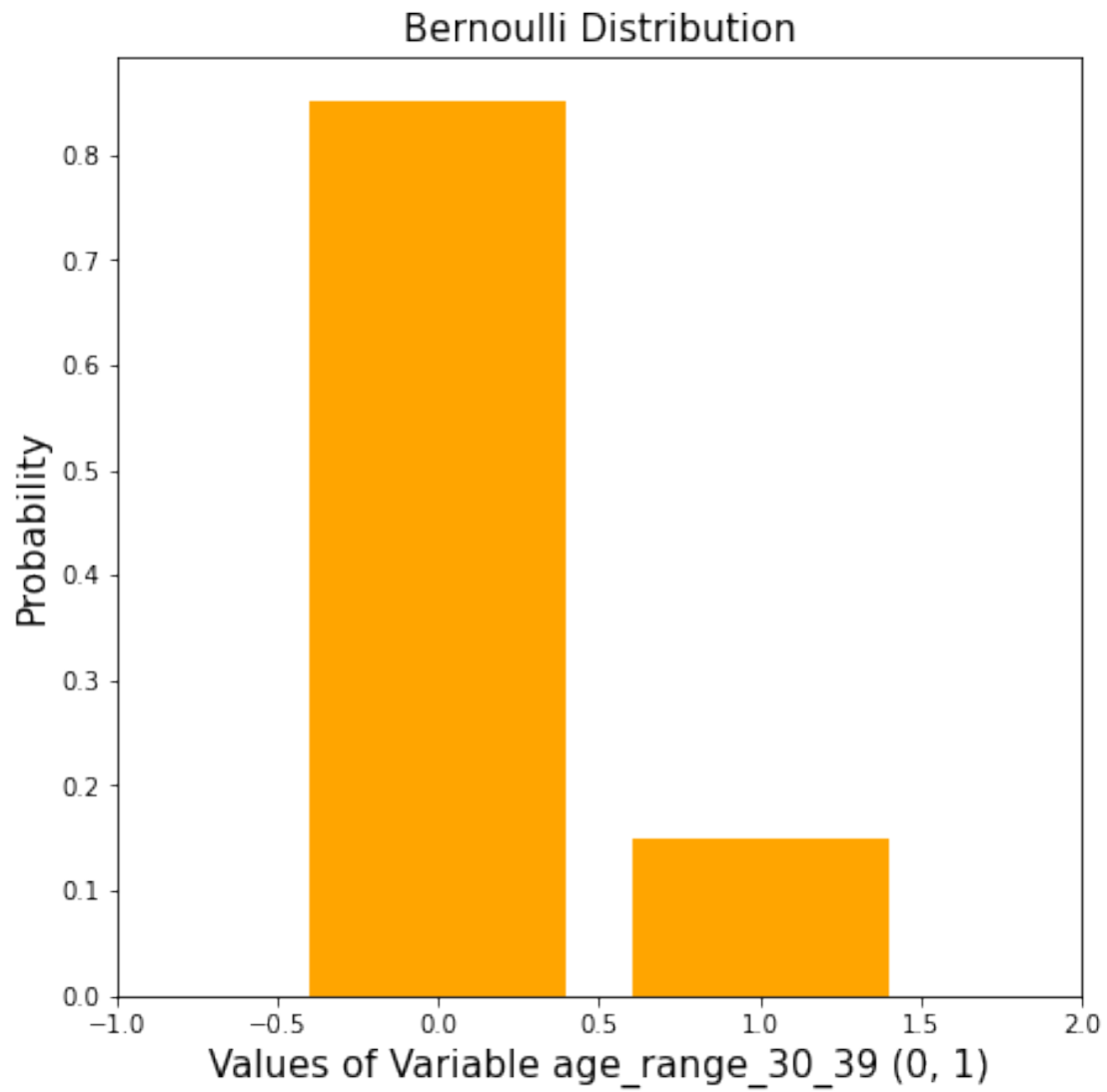


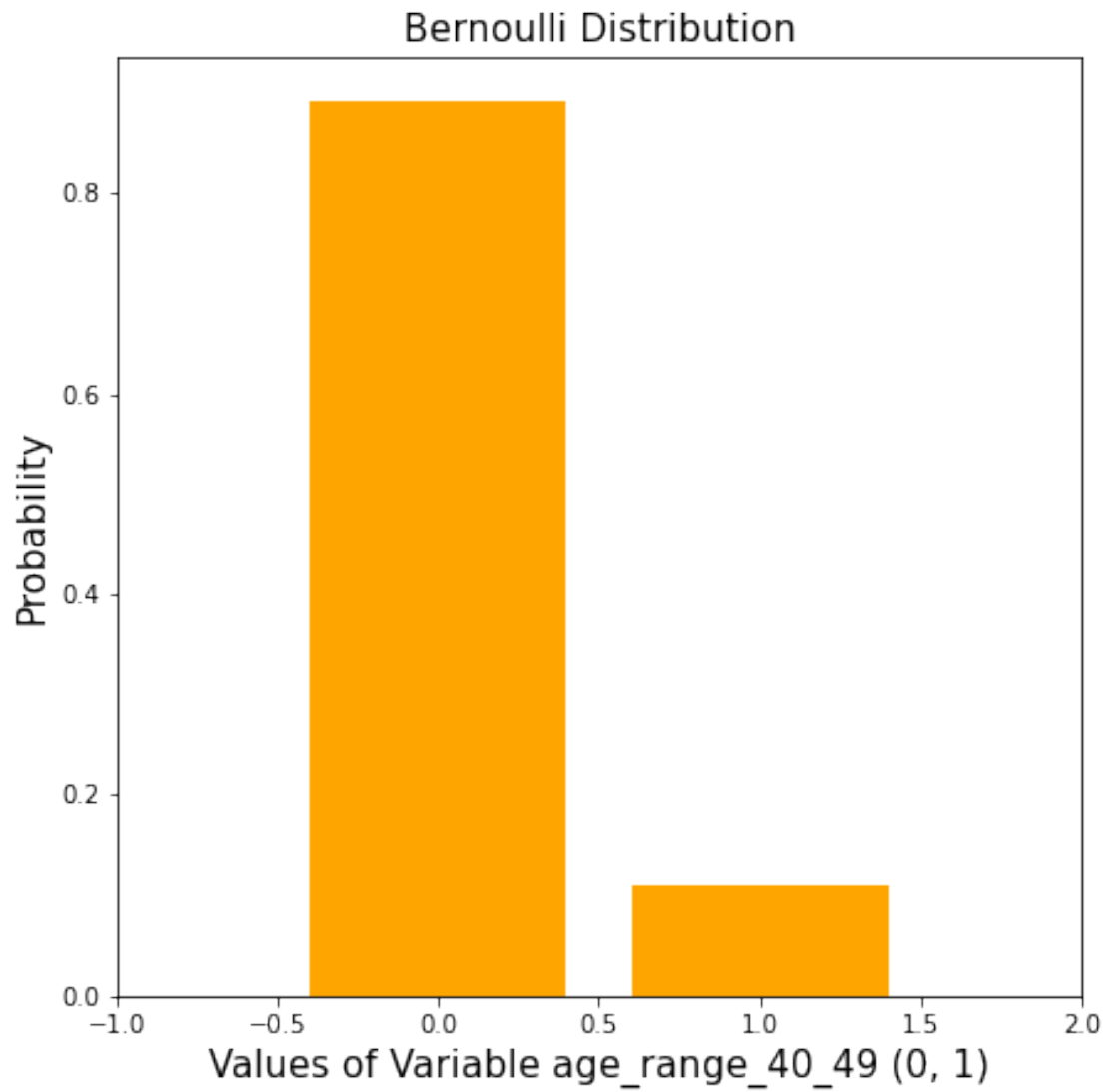


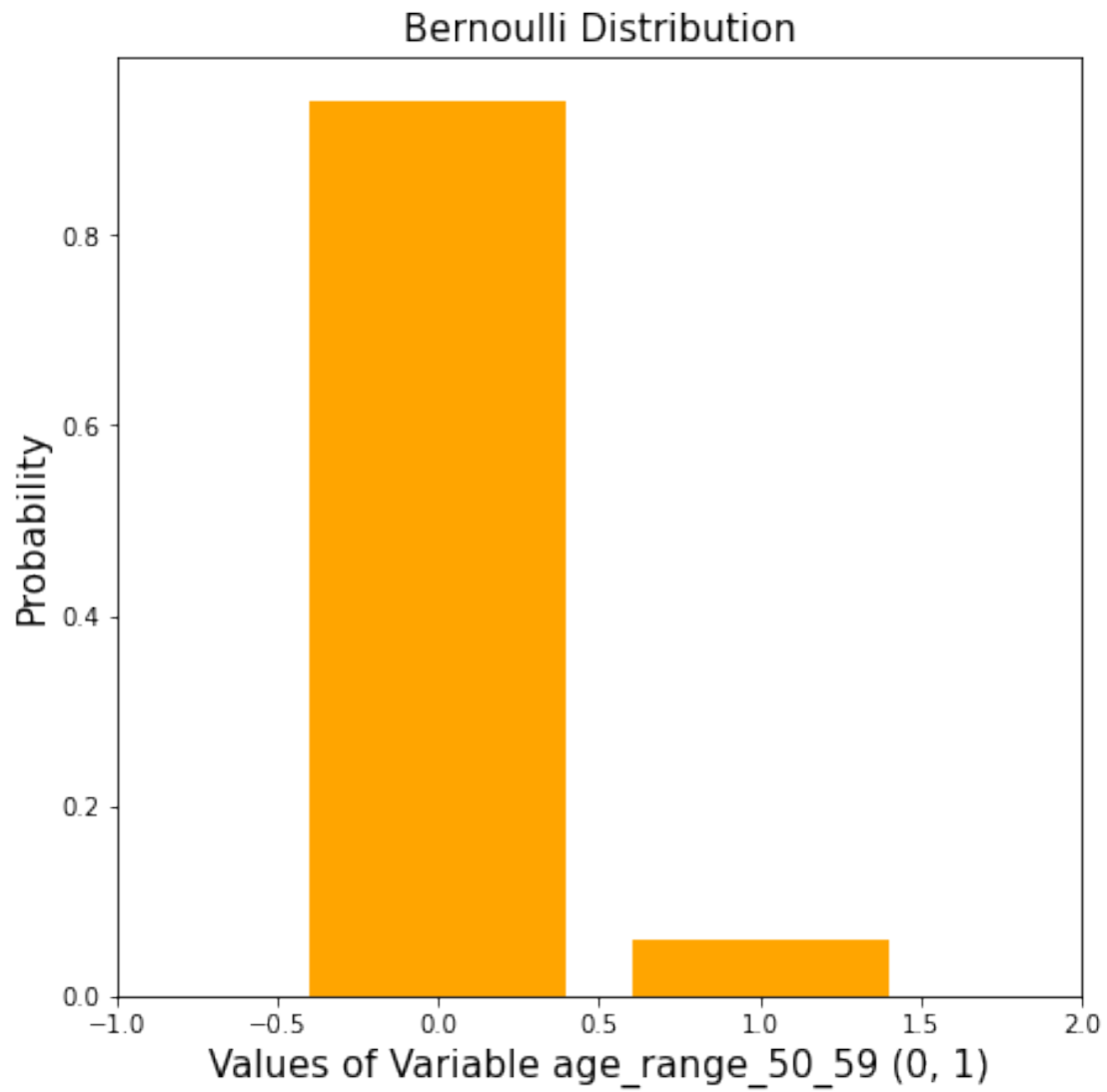


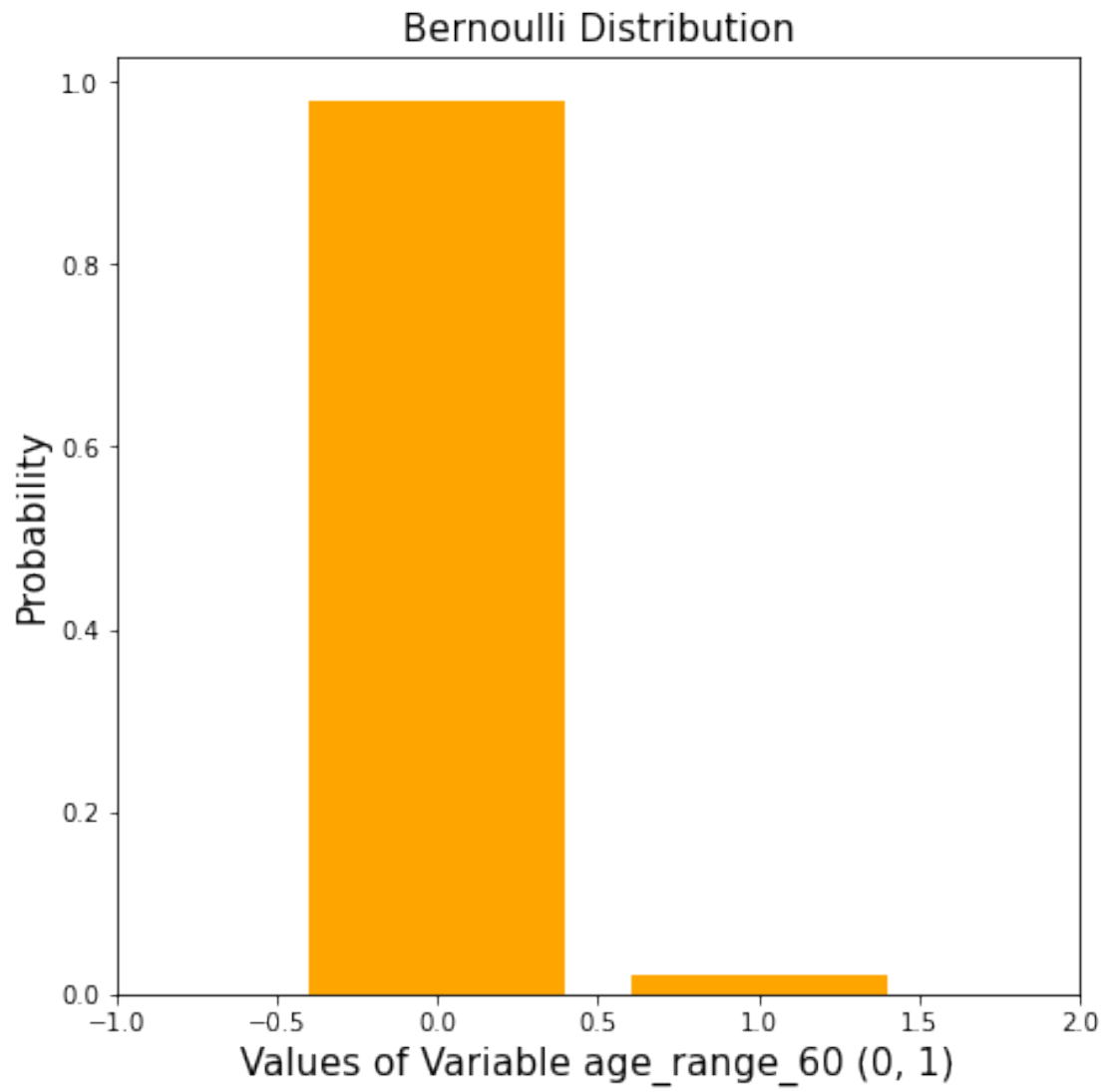


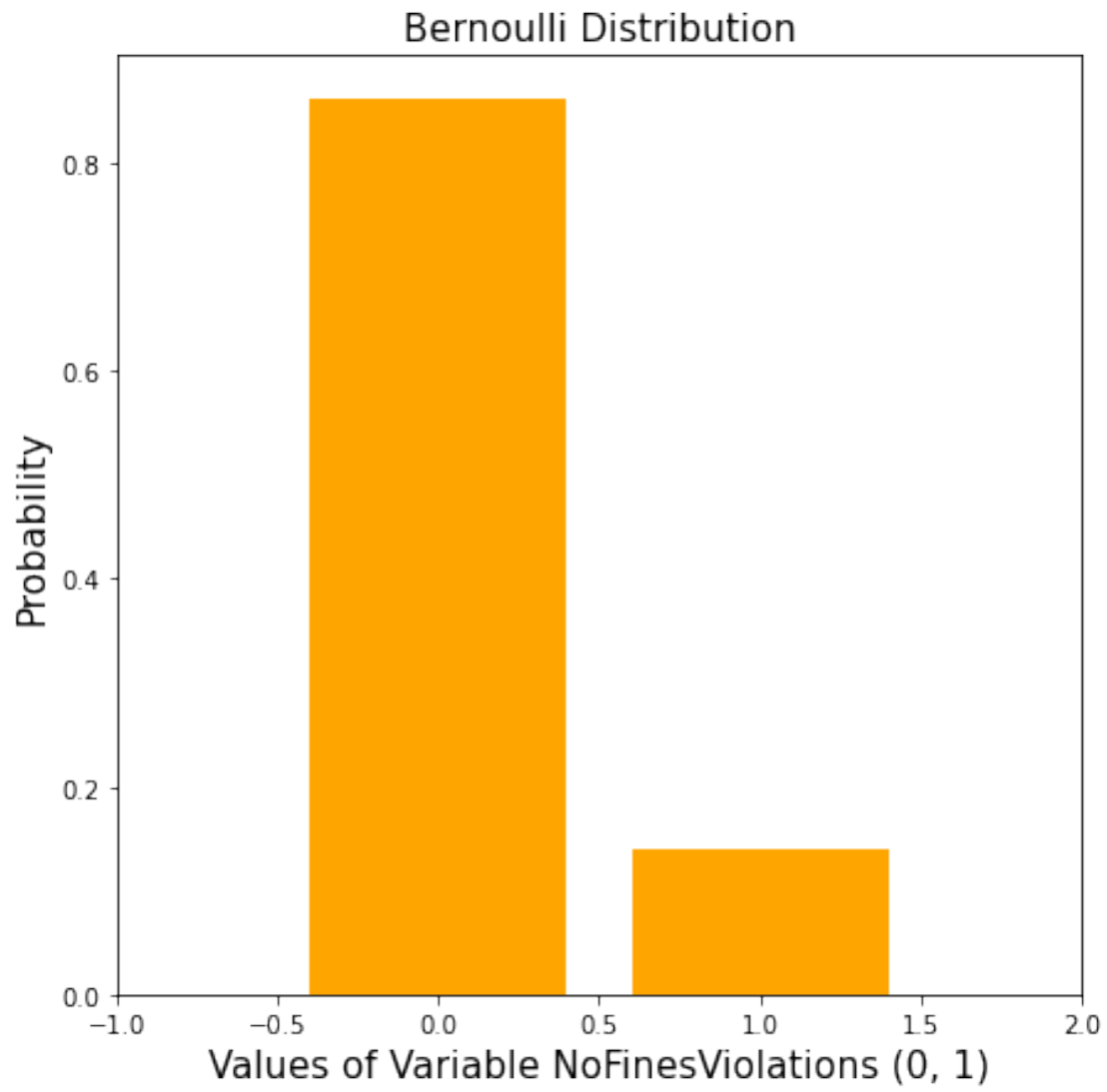


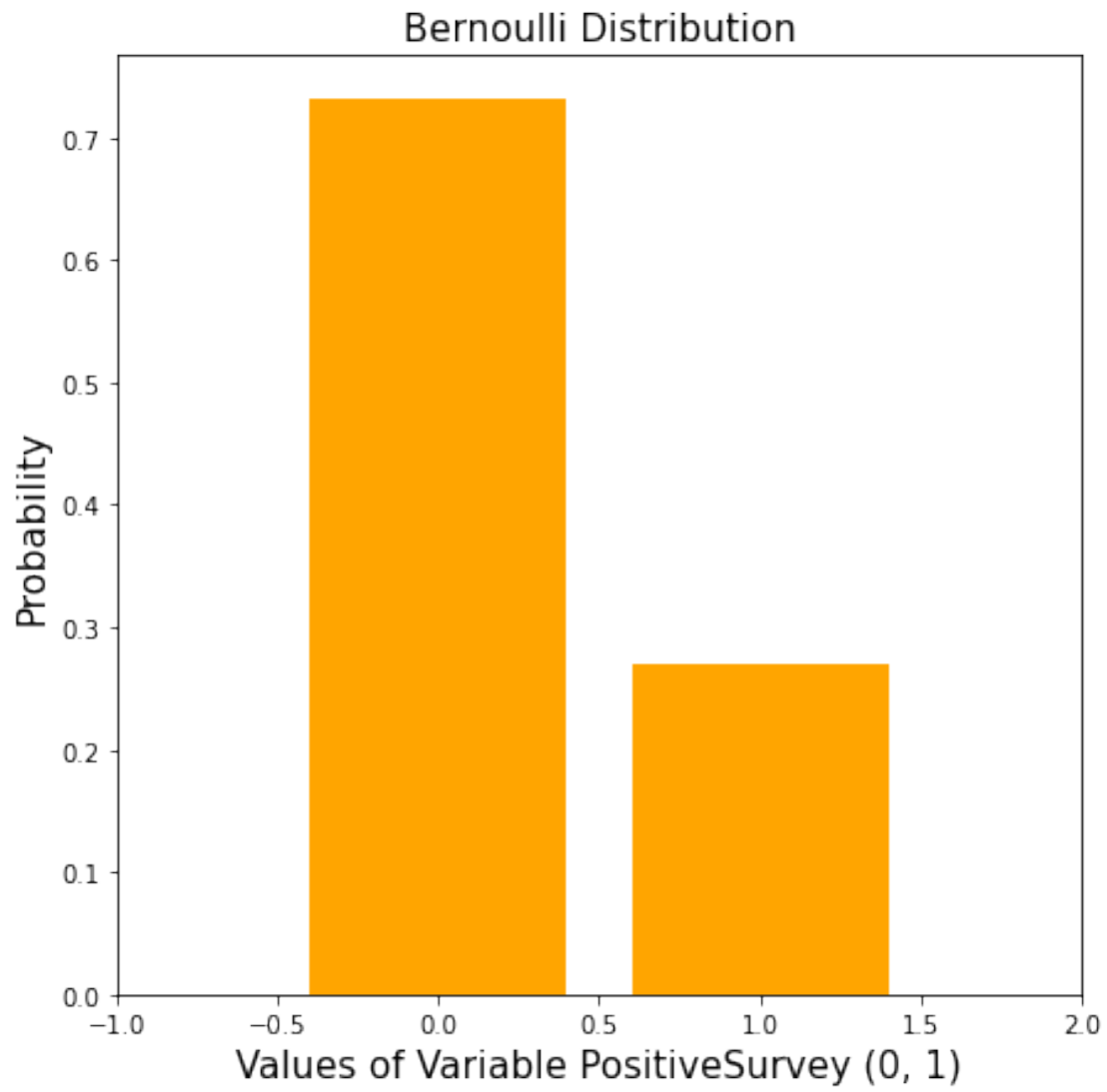


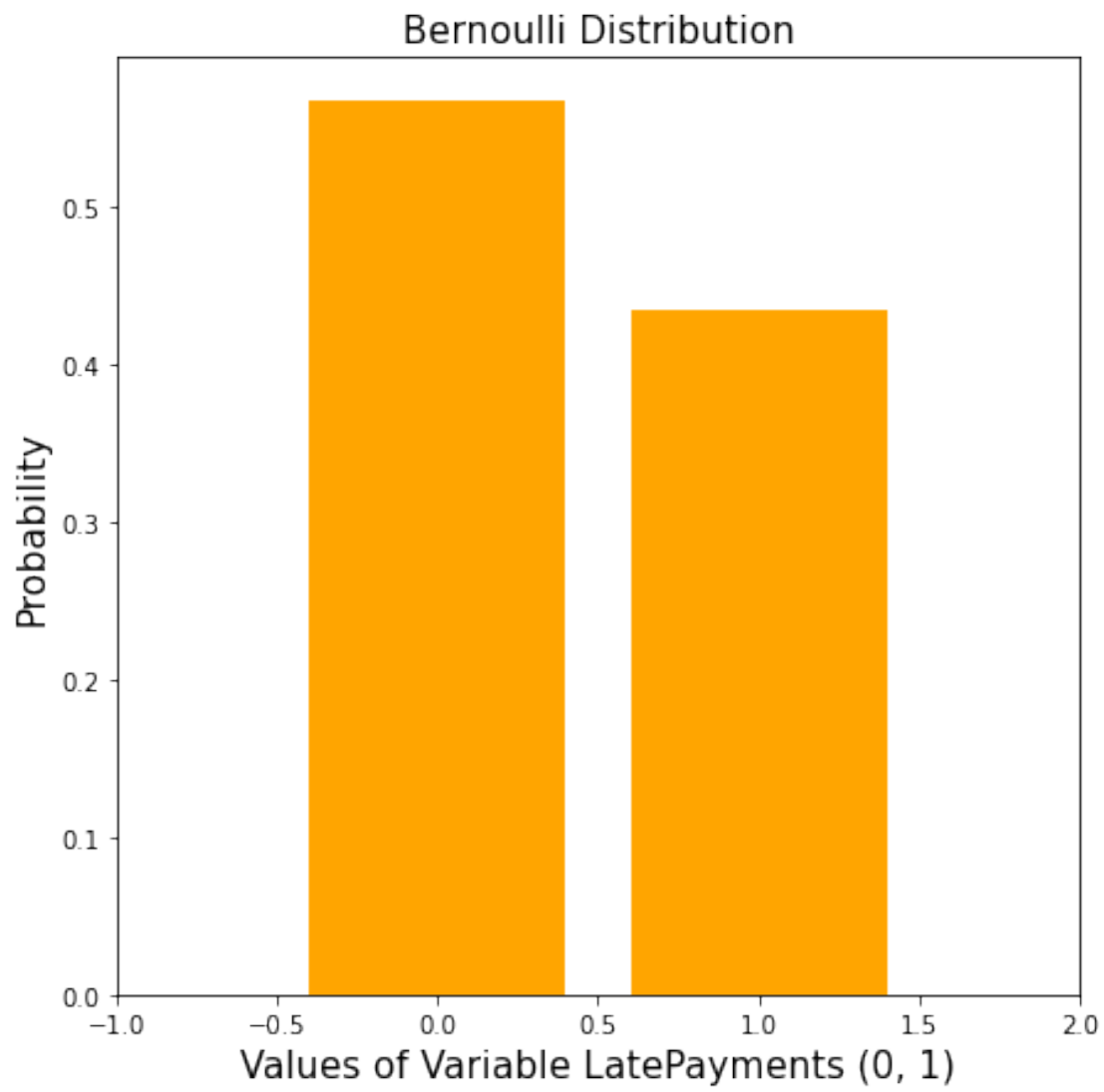


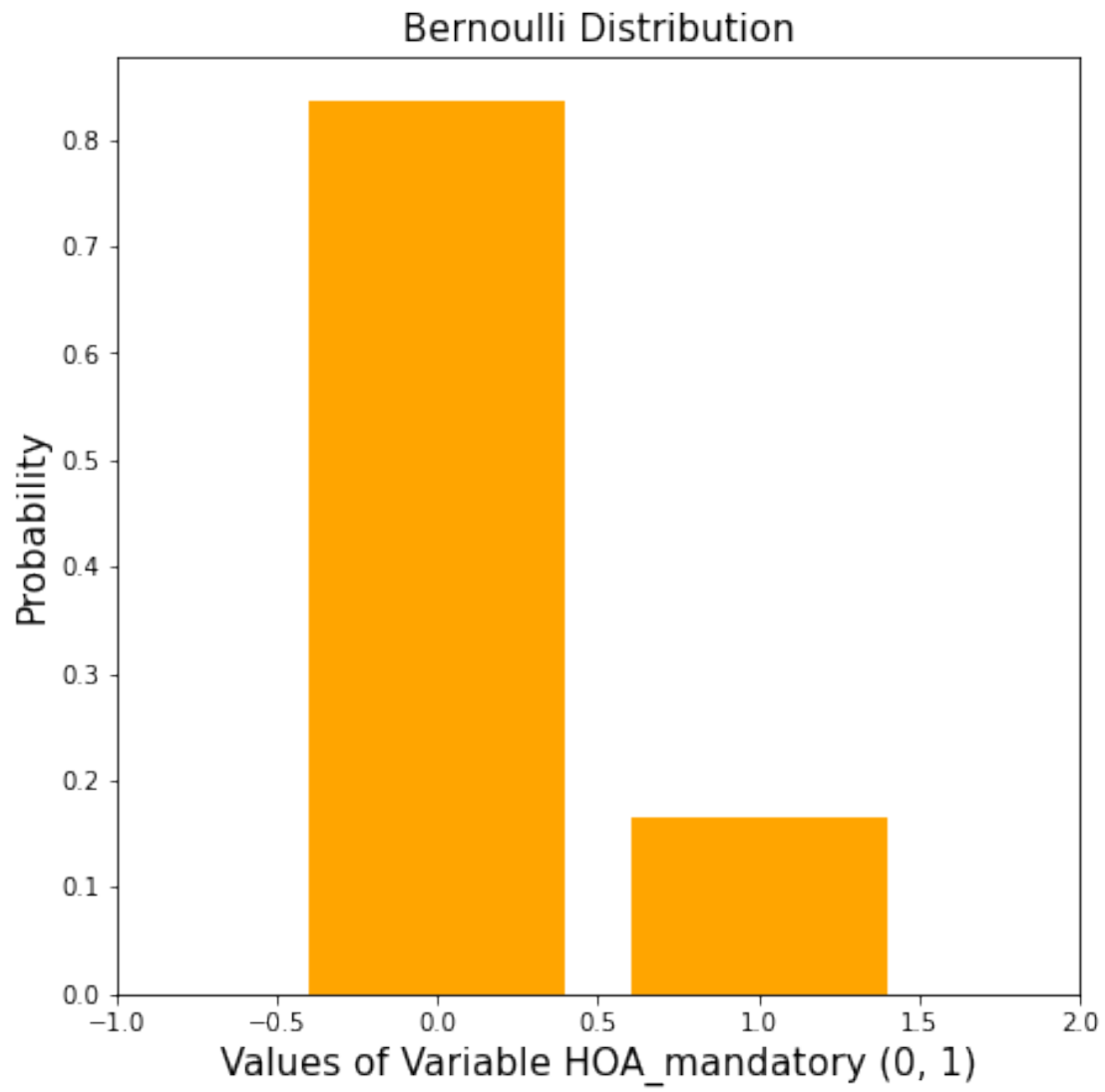


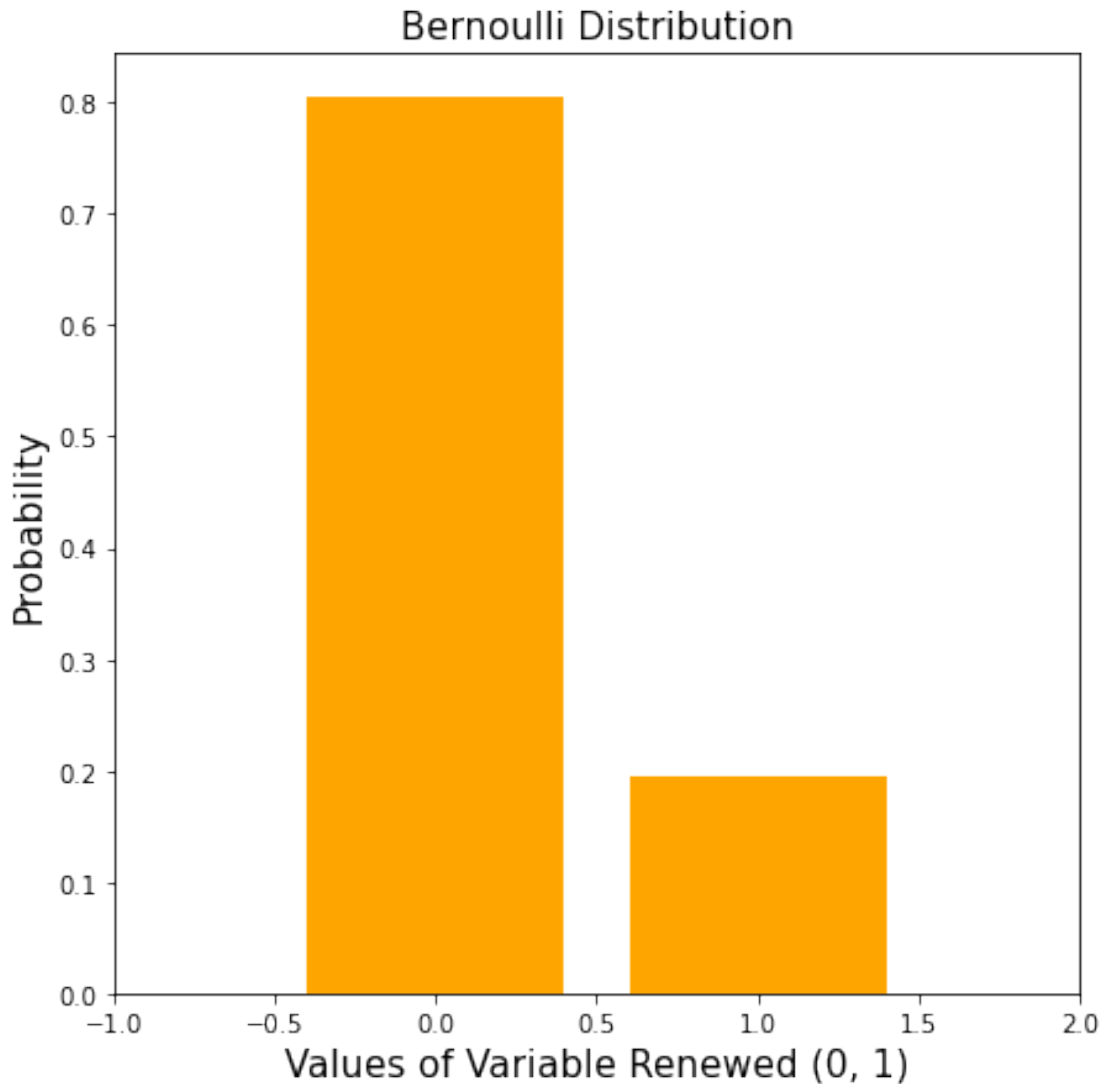












```
dataframe_renewal.iloc[:,len(dataframe_renewal.columns)-  
1].value_counts()
```

```
0    64208  
1    15642  
Name: Renewed, dtype: int64
```

19.58 percentage of total residents have renewed their lease further.

Using select k best features for feature selection

```
X = dataframe_renewal.iloc[:,1:17]  
y = dataframe_renewal.iloc[:,-1]  
bestfeatures = SelectKBest(score_func=chi2, k=10)
```

```

fit = bestfeatures.fit(X,y)
dfscores = pd.DataFrame(fit.scores_)
dfcolumns = pd.DataFrame(X.columns)
featureScores = pd.concat([dfcolumns,dfscores],axis=1)
featureScores.columns = ['Specs','Score']
print(featureScores.nlargest(10,'Score'))

```

	Specs	Score
0	no_rent_change	1171.663873
12	NoFinesViolations	1141.363078
2	rent_change_20	546.303728
15	HOA_mandatory	499.392675
3	lease_length_2	443.921557
5	lease_length_1	305.374923
8	age_range_30_39	289.719040
6	age_range_under_24	203.623655
14	LatePayments	129.735829
9	age_range_40_49	48.059799

```

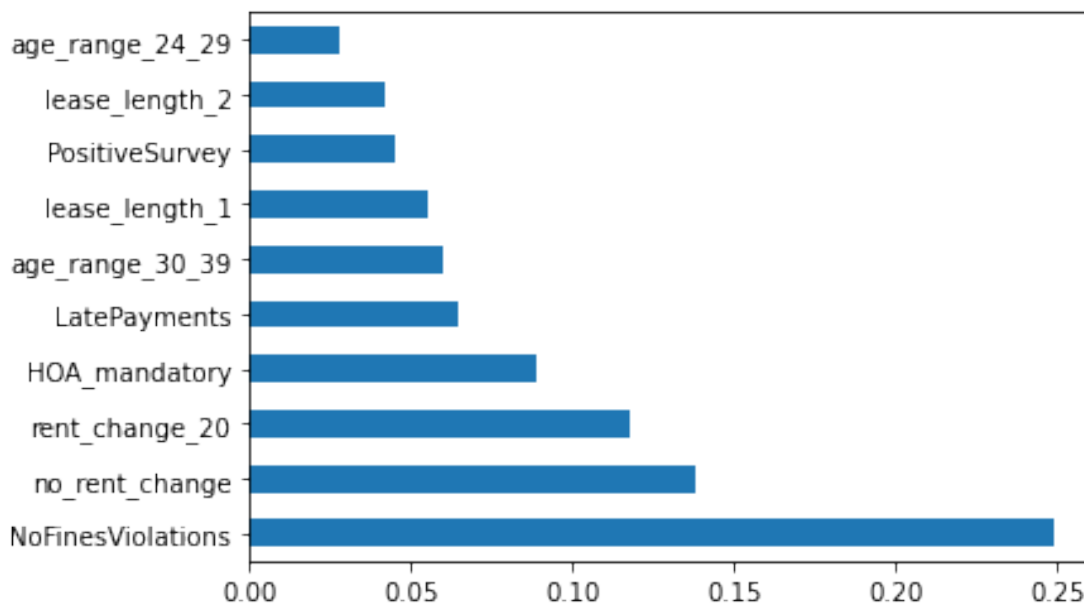
model = ExtraTreesClassifier()
model.fit(X,y)
print(model.feature_importances_)
feat_importances = pd.Series(model.feature_importances_,
index=X.columns)
feat_importances.nlargest(10).plot(kind='barh')
plt.show()

```

```

[0.13798823 0.01716024 0.11788869 0.04157493 0.01907466 0.05523355
 0.01978089 0.02763472 0.05943724 0.02532366 0.01872536 0.01239429
 0.24916766 0.04495751 0.06458268 0.08907567]

```



Observation: No violations, no_rent_change, rent_change_20, LatePayments were the highest impacting independent features

VarianceThreshold methods for feature selection

Setting the variance 10 percent. So features with less than 0.1 will have a value false.

```
from sklearn.feature_selection import VarianceThreshold
var_thres=VarianceThreshold(threshold=0.1)
var_thres.fit(dataframe_renewal.iloc[:,1:])

VarianceThreshold(threshold=0.1)

var_thres.get_support()

array([ True, False,  True,  True, False,  True, False, False,  True,
        False, False, False,  True,  True,  True,  True,  True])

dataframe_renewal.iloc[:,1:].columns

Index(['no_rent_change', 'rent_change_10', 'rent_change_20',
       'lease_length_2',
       'lease_length_3', 'lease_length_1', 'age_range_under_24',
       'age_range_24_29', 'age_range_30_39', 'age_range_40_49',
       'age_range_50_59', 'age_range_60', 'NoFinesViolations',
       'PositiveSurvey', 'LatePayments', 'HOA_mandatory', 'Renewed'],
      dtype='object')
```

Observations

Below features have low variance and these features do not exist in model feature_importances.

This indicates that these

features have less impact on the dependent variable

```
dataframe_renewal['rent_change_10'].value_counts()
```

```
0    77971
1     1879
Name: rent_change_10, dtype: int64
```

```
dataframe_renewal['lease_length_3'].value_counts()
```

```
0    75259
```

```
1     4591
```

```
Name: lease_length_3, dtype: int64
```

```
dataframe_renewal['age_range_under_24'].value_counts()
```

```
0    76791
```

```
1     3059
```

```
Name: age_range_under_24, dtype: int64
```

```
dataframe_renewal['age_range_40_49'].value_counts()
```

```
0    71178
```

```
1     8672
```

```
Name: age_range_40_49, dtype: int64
```

```
dataframe_renewal['age_range_50_59'].value_counts()
```

```
0    75128
```

```
1     4722
```

```
Name: age_range_50_59, dtype: int64
```

```
dataframe_renewal['age_range_60'].value_counts()
```

```
0    78175
```

```
1     1675
```

```
Name: age_range_60, dtype: int64
```

Feature Selection-Information gain - mutual information

```
# Train test split to avoid overfitting
```

```
X_train,X_test,y_train,y_test=train_test_split(dataframe_renewal.drop(
labels=['Renewed','lease_id'], axis=1),
```

```
dataframe_renewal['Renewed'],test_size=0.3,random_state=100)
```

```
X_train.head()
```

	no_rent_change	rent_change_10	rent_change_20	lease_length_2
56768	0	1	0	0
17813	0	0	0	0
31528	1	0	0	1
30122	0	0	1	0
11090	0	0	1	0

	lease_length_3	lease_length_1	age_range_under_24
age_range_24_29 \			
56768	1	0	0
0			
17813	0	0	0
0			
31528	0	0	1
0			
30122	0	1	0
1			
11090	0	1	0
0			

	age_range_30_39	age_range_40_49	age_range_50_59	age_range_60
\				
56768	0	0	0	0
17813	0	0	0	0
31528	0	0	0	0
30122	0	0	0	0
11090	0	0	0	0

	NoFinesViolations	PositiveSurvey	LatePayments	H0A_mandatory
56768	1	1	1	1
17813	0	0	0	0
31528	0	0	0	1
30122	0	1	0	0
11090	0	1	0	0

Determine the mutual information

```
mutual_info = mutual_info_classif(X_train, y_train)
mutual_info
```

```
array([9.84239784e-03, 3.23077301e-03, 1.13970946e-02, 6.10303841e-03,
       0.00000000e+00, 6.56789683e-03, 3.78394536e-04, 6.59842213e-04,
       1.43756562e-03, 8.74230344e-05, 0.00000000e+00, 0.00000000e+00,
       8.51132478e-03, 3.01542900e-03, 5.20284414e-03, 5.63194753e-
03])
```

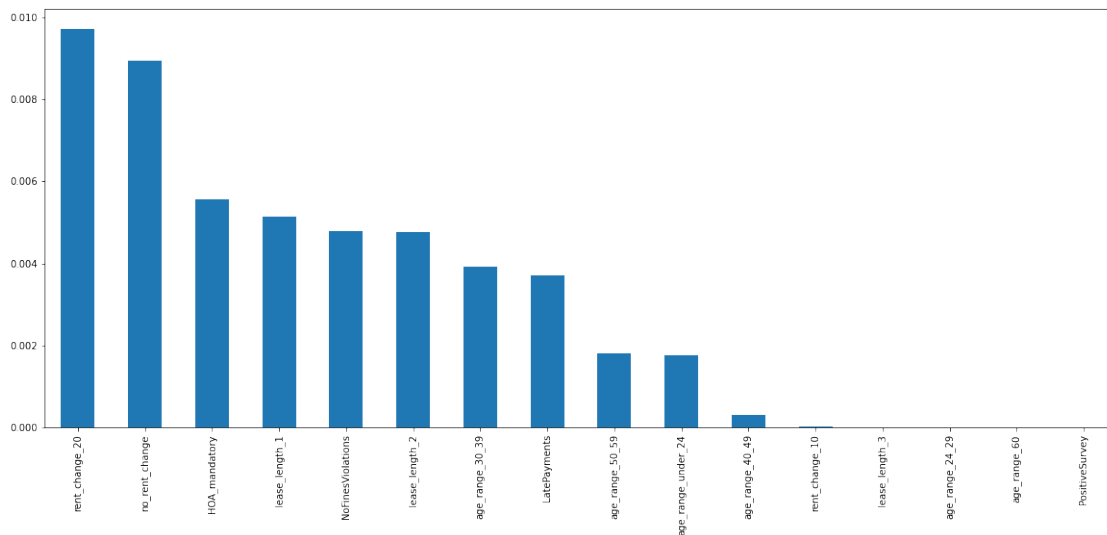
```
mutual_info = pd.Series(mutual_info)
mutual_info.index = X_train.columns
mutual_info.sort_values(ascending=False)
```

```
rent_change_20      0.009722
no_rent_change      0.008932
HOA_mandatory       0.005558
lease_length_1      0.005143
NoFinesViolations   0.004779
lease_length_2      0.004757
age_range_30_39     0.003907
LatePayments        0.003695
age_range_50_59     0.001798
age_range_under_24  0.001766
age_range_40_49     0.000301
rent_change_10      0.000017
lease_length_3      0.000000
age_range_24_29     0.000000
age_range_60        0.000000
PositiveSurvey       0.000000
dtype: float64
```

#let's plot the ordered mutual_info values per feature

```
mutual_info.sort_values(ascending=False).plot.bar(figsize=(20, 8))
```

<AxesSubplot:>



Observations:

rent_change_20,no_rent_change,NoFinesViolations,lease_length_1,H
OA_mandatory are the top features making an highest impact

Chisquare Test For Feature Selection

```
f_p_values=chi2(X_train,y_train)
```

```
f_p_values
```

```
(array([780.14183501, 26.56792927, 379.4836922 , 318.49720696,  
        11.89742588, 226.91716465, 140.62197392,  1.81715198,  
        182.77954308, 38.86398527,  6.5276809 , 40.63339053,  
        810.65116529, 16.17612522, 88.28999108, 338.3505866 ]),  
array([1.12111804e-171, 2.54432455e-007, 1.61201754e-084,  
3.07782890e-071,  
        5.62115719e-004, 2.80334734e-051, 1.94627633e-032,  
1.77652658e-001,  
        1.19828020e-041, 4.54386850e-010, 1.06208343e-002,  
1.83641596e-010,  
        2.60810485e-178, 5.77169889e-005, 5.65287623e-021,  
1.45910325e-075]))
```

```
p_values=pd.Series(f_p_values[1])
```

```
p_values.index=X_train.columns
```

```
p_values
```

no_rent_change	1.121118e-171
rent_change_10	2.544325e-07
rent_change_20	1.612018e-84
lease_length_2	3.077829e-71
lease_length_3	5.621157e-04
lease_length_1	2.803347e-51
age_range_under_24	1.946276e-32
age_range_24_29	1.776527e-01
age_range_30_39	1.198280e-41
age_range_40_49	4.543868e-10
age_range_50_59	1.062083e-02
age_range_60	1.836416e-10
NoFinesViolations	2.608105e-178
PositiveSurvey	5.771699e-05
LatePayments	5.652876e-21
H0A_mandatory	1.459103e-75
dtype:	float64

```
p_values.sort_index(ascending=False)
```

rent_change_20	1.612018e-84
rent_change_10	2.544325e-07
no_rent_change	1.121118e-171
lease_length_3	5.621157e-04
lease_length_2	3.077829e-71
lease_length_1	2.803347e-51
age_range_under_24	1.946276e-32
age_range_60	1.836416e-10
age_range_50_59	1.062083e-02
age_range_40_49	4.543868e-10
age_range_30_39	1.198280e-41
age_range_24_29	1.776527e-01
PositiveSurvey	5.771699e-05
NoFinesViolations	2.608105e-178
LatePayments	5.652876e-21
HOA_mandatory	1.459103e-75

dtype: float64

Observation

NoFinesViolations, no_rent_change, rent_change_20, HOA_mandatory, lease_length_2 are the top features having the highest impact

By looking at all the above steps we see that no rent change, no fines violated, Hoa mandatory, rentchange20 are the features having the most impact on dependent feature.

Calculating the percentage impact on the renewed feature by no_rent_change

```
listval_no_rent_change=[]
for itr in range(len(dataframe_renewal.index)):
    if dataframe_renewal.Renewed[itr]==1:

listval_no_rent_change.append(dataframe_renewal.no_rent_change[itr])
print(listval_no_rent_change.count(1))
print(listval_no_rent_change.count(0))

5270
10372
```


Observation:

From no_rent_change around 34 percent of the residents renew if no_rent_change = 1 that means most of the people who renewed has their rent increased

Below clearly indicates that there are no records with no_rent_change = 1 and rent_change =1 which indicates: if there is no change in rent then there is no change in the rent as well

```
check=[]
for itr in range(len(dataframe_renewal.index)):
    if (dataframe_renewal.no_rent_change[itr]==1 and
(dataframe_renewal.rent_change_10[itr]==1 or
dataframe_renewal.rent_change_20[itr]==1)) :
        check.append(dataframe_renewal.no_rent_change[itr])
print(len(check))
```

0

```
check=[]
for itr in range(len(dataframe_renewal.index)):
    if (dataframe_renewal.no_rent_change[itr]==0 and
(dataframe_renewal.rent_change_10[itr]==0 and
dataframe_renewal.rent_change_20[itr]==0)) :
        check.append(dataframe_renewal.no_rent_change[itr])
print(len(check))
```

13828

13828 records exists that indicates: There was a change in the rent and neither it is increased by 10 nor 20 percent.

Calculating the percentage impact on the renewed feature by rent_change_20

```
listval_rent_change_20=[]
for itr in range(len(dataframe_renewal.index)):
    if dataframe_renewal.Renewed[itr]==1:

listval_rent_change_20.append(dataframe_renewal.rent_change_20[itr])
```

```
print(listval_rent_change_20.count(1))
print(listval_rent_change_20.count(0))
```

```
7102
8540
```

From rent_change_20 around 55 percent of the residents renew if rent_change_20 = 0

that means people prefer having no change is rent although not a high percentage

Calculating the percentage impact on the renewed feature by NoFinesViolations

```
listval_NoFinesViolations=[]
for itr in range(len(dataframe_renewal.index)):
    if dataframe_renewal.Renewed[itr]==1:
```

```
listval_NoFinesViolations.append(dataframe_renewal.NoFinesViolations[itr])
print(listval_NoFinesViolations.count(1))
print(listval_NoFinesViolations.count(0))
```

```
3592
12050
```

From NoFinesViolations 77 percentage of the residents renew if NoFinesViolations = 0

that means most of the people who renewed the lease have violations

Calculating the percentage impact on the renewed feature by Hoa mandatory

```
listval_HOA_mandatory=[]
for itr in range(len(dataframe_renewal.index)):
    if dataframe_renewal.Renewed[itr]==1:
```

```
listval_HOA_mandatory.append(dataframe_renewal.HOA_mandatory[itr])
```

```
print(listval_HOA_mandatory.count(1))  
print(listval_HOA_mandatory.count(0))
```

```
1552  
14090
```

From HOA_mandatory around 90 percentage of residents renew if
HOA_mandatory = 0

that means highest people prefer having No mandatory fees on the
lease