

# Lending Club Loan Data Analysis Project

January 13, 2024

```
[1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Dropout
from tensorflow.keras.callbacks import EarlyStopping
from tensorflow.keras.models import load_model
from sklearn.metrics import confusion_matrix, classification_report
from pickle import dump, load
```

```
%matplotlib inline
```

```
2024-01-13 19:19:44.157756: I tensorflow/core/util/port.cc:110] oneDNN custom
operations are on. You may see slightly different numerical results due to
floating-point round-off errors from different computation orders. To turn them
off, set the environment variable `TF_ENABLE_ONEDNN_OPTS=0`.
2024-01-13 19:19:44.197890: I tensorflow/core/platform/cpu_feature_guard.cc:182]
This TensorFlow binary is optimized to use available CPU instructions in
performance-critical operations.
To enable the following instructions: AVX2 AVX512F AVX512_VNNI FMA, in other
operations, rebuild TensorFlow with the appropriate compiler flags.
```

```
VOC-NOTICE: GPU memory for this assignment is capped at 1024MiB
```

```
2024-01-13 19:19:46.032287: E
tensorflow/compiler/xla/stream_executor/cuda/cuda_driver.cc:268] failed call to
cuInit: CUDA_ERROR_NO_DEVICE: no CUDA-capable device is detected
```

```
[2]: df = pd.read_csv('loan_data.csv')
```

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

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9578 entries, 0 to 9577
```

Data columns (total 14 columns):

#	Column	Non-Null Count	Dtype
0	credit.policy	9578 non-null	int64
1	purpose	9578 non-null	object
2	int.rate	9578 non-null	float64
3	installment	9578 non-null	float64
4	log.annual.inc	9578 non-null	float64
5	dti	9578 non-null	float64
6	fico	9578 non-null	int64
7	days.with.cr.line	9578 non-null	float64
8	revol.bal	9578 non-null	int64
9	revol.util	9578 non-null	float64
10	inq.last.6mths	9578 non-null	int64
11	delinq.2yrs	9578 non-null	int64
12	pub.rec	9578 non-null	int64
13	not.fully.paid	9578 non-null	int64

dtypes: float64(6), int64(7), object(1)

memory usage: 1.0+ MB

```
[3]: credit.policy      purpose  int.rate  installment  log.annual.inc  \
0      1  debt_consolidation  0.1189      829.10      11.350407
1      1      credit_card    0.1071      228.22      11.082143
2      1  debt_consolidation  0.1357      366.86      10.373491
3      1  debt_consolidation  0.1008      162.34      11.350407
4      1      credit_card    0.1426      102.92      11.299732

      dti  fico  days.with.cr.line  revol.bal  revol.util  inq.last.6mths  \
0  19.48  737      5639.958333      28854      52.1      0
1  14.29  707      2760.000000      33623      76.7      0
2  11.63  682      4710.000000      3511      25.6      1
3   8.10  712      2699.958333      33667      73.2      1
4  14.97  667      4066.000000      4740      39.5      0

      delinq.2yrs  pub.rec  not.fully.paid
0      0      0      0
1      0      0      0
2      0      0      0
3      0      0      0
4      1      0      0
```

```
[4]: df.describe().transpose()
```

	count	mean	std	min	\
credit.policy	9578.0	0.804970	0.396245	0.000000	
int.rate	9578.0	0.122640	0.026847	0.060000	
installment	9578.0	319.089413	207.071301	15.670000	

log.annual.inc	9578.0	10.932117	0.614813	7.547502
dti	9578.0	12.606679	6.883970	0.000000
fico	9578.0	710.846314	37.970537	612.000000
days.with.cr.line	9578.0	4560.767197	2496.930377	178.958333
revol.bal	9578.0	16913.963876	33756.189557	0.000000
revol.util	9578.0	46.799236	29.014417	0.000000
inq.last.6mths	9578.0	1.577469	2.200245	0.000000
delinq.2yrs	9578.0	0.163708	0.546215	0.000000
pub.rec	9578.0	0.062122	0.262126	0.000000
not.fully.paid	9578.0	0.160054	0.366676	0.000000

	25%	50%	75%	max
credit.policy	1.000000	1.000000	1.000000	1.000000e+00
int.rate	0.103900	0.122100	0.140700	2.164000e-01
installment	163.770000	268.950000	432.762500	9.401400e+02
log.annual.inc	10.558414	10.928884	11.291293	1.452835e+01
dti	7.212500	12.665000	17.950000	2.996000e+01
fico	682.000000	707.000000	737.000000	8.270000e+02
days.with.cr.line	2820.000000	4139.958333	5730.000000	1.763996e+04
revol.bal	3187.000000	8596.000000	18249.500000	1.207359e+06
revol.util	22.600000	46.300000	70.900000	1.190000e+02
inq.last.6mths	0.000000	1.000000	2.000000	3.300000e+01
delinq.2yrs	0.000000	0.000000	0.000000	1.300000e+01
pub.rec	0.000000	0.000000	0.000000	5.000000e+00
not.fully.paid	0.000000	0.000000	0.000000	1.000000e+00

```
[5]: df['not.fully.paid'].isnull().mean()
```

```
[5]: 0.0
```

```
[6]: df1=pd.get_dummies(df, columns=['purpose'])
```

```
[7]: df1['log.annual.inc'] = np.exp(df1['log.annual.inc'])
```

```
[8]: df1.head()
```

```
[8]:
```

	credit.policy	int.rate	installment	log.annual.inc	dti	fico	\
0	1	0.1189	829.10	85000.000385	19.48	737	
1	1	0.1071	228.22	65000.000073	14.29	707	
2	1	0.1357	366.86	31999.999943	11.63	682	
3	1	0.1008	162.34	85000.000385	8.10	712	
4	1	0.1426	102.92	80799.999636	14.97	667	

	days.with.cr.line	revol.bal	revol.util	inq.last.6mths	delinq.2yrs	\
0	5639.958333	28854	52.1	0	0	
1	2760.000000	33623	76.7	0	0	
2	4710.000000	3511	25.6	1	0	

3	2699.958333	33667	73.2	1	0
4	4066.000000	4740	39.5	0	1

	pub.rec	not.fully.paid	purpose_all_other	purpose_credit_card	\
0	0	0	0	0	
1	0	0	0	1	
2	0	0	0	0	
3	0	0	0	0	
4	0	0	0	1	

	purpose_debt_consolidation	purpose_educational	purpose_home_improvement	\
0	1	0	0	
1	0	0	0	
2	1	0	0	
3	1	0	0	
4	0	0	0	

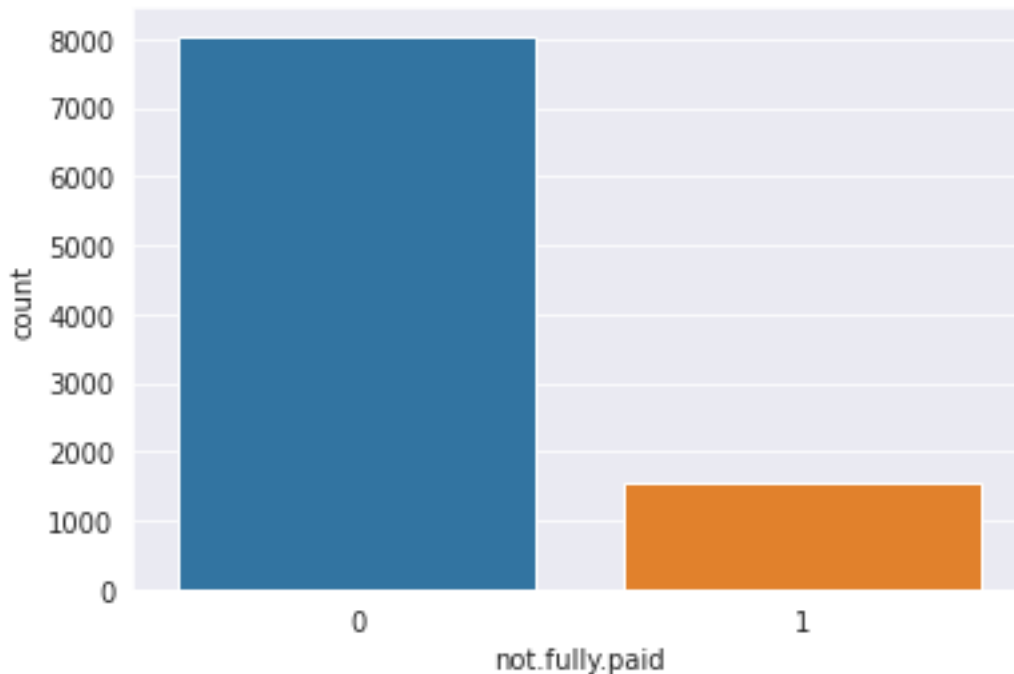
	purpose_major_purchase	purpose_small_business
0	0	0
1	0	0
2	0	0
3	0	0
4	0	0

```
[9]: df.groupby('not.fully.paid')['not.fully.paid'].count()/len(df)
```

```
[9]: not.fully.paid
0    0.839946
1    0.160054
Name: not.fully.paid, dtype: float64
```

```
[10]: sns.set_style('darkgrid')
sns.countplot(x='not.fully.paid', data=df)
```

```
[10]: <AxesSubplot: xlabel='not.fully.paid', ylabel='count'>
```



```
[11]: count_class_0, count_class_1 = df['not.fully.paid'].value_counts()
```

```
[12]: df_0 = df[df['not.fully.paid'] == 0]
      df_1 = df[df['not.fully.paid'] == 1]
```

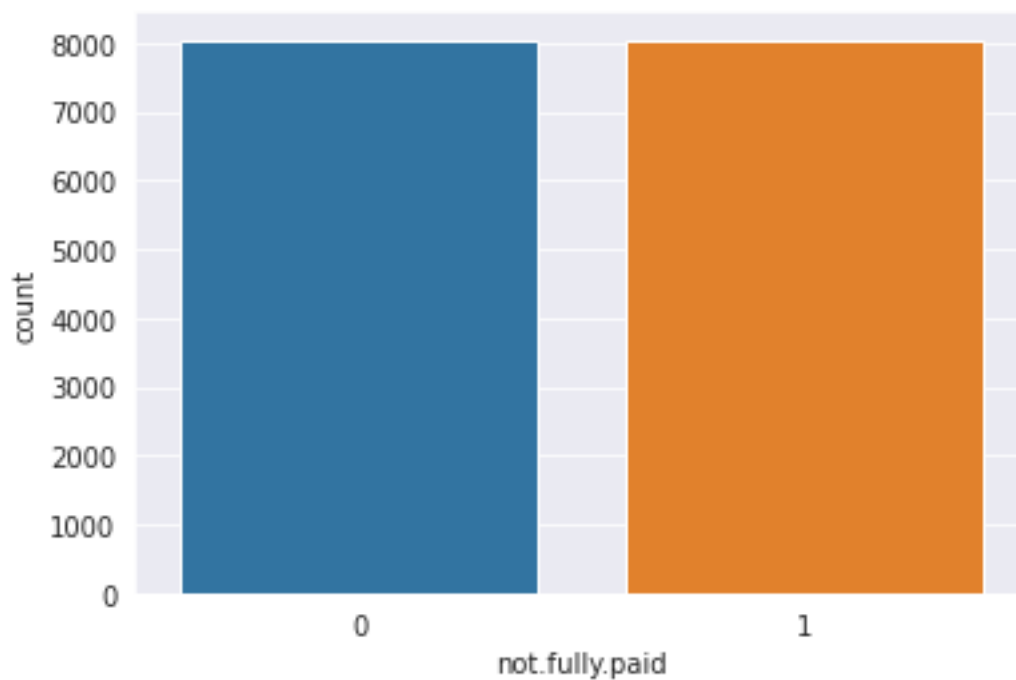
```
[13]: df_1_over = df_1.sample(count_class_0, replace=True)
      df_test_over = pd.concat([df_0, df_1_over], axis=0)
```

```
[14]: print('Random over-sampling:')
      print(df_test_over['not.fully.paid'].value_counts())
```

```
Random over-sampling:
0      8045
1      8045
Name: not.fully.paid, dtype: int64
```

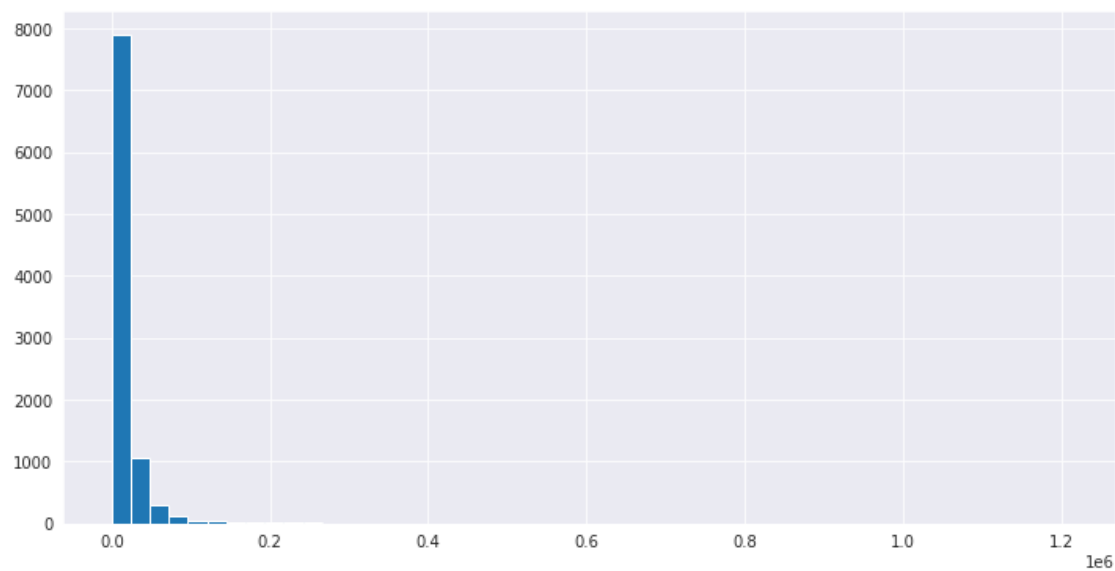
```
[15]: sns.set_style('darkgrid')
      sns.countplot(x='not.fully.paid', data=df_test_over)
```

```
[15]: <AxesSubplot: xlabel='not.fully.paid', ylabel='count'>
```



```
[16]: df['revol.bal'].hist(figsize=[12,6], bins=50)
```

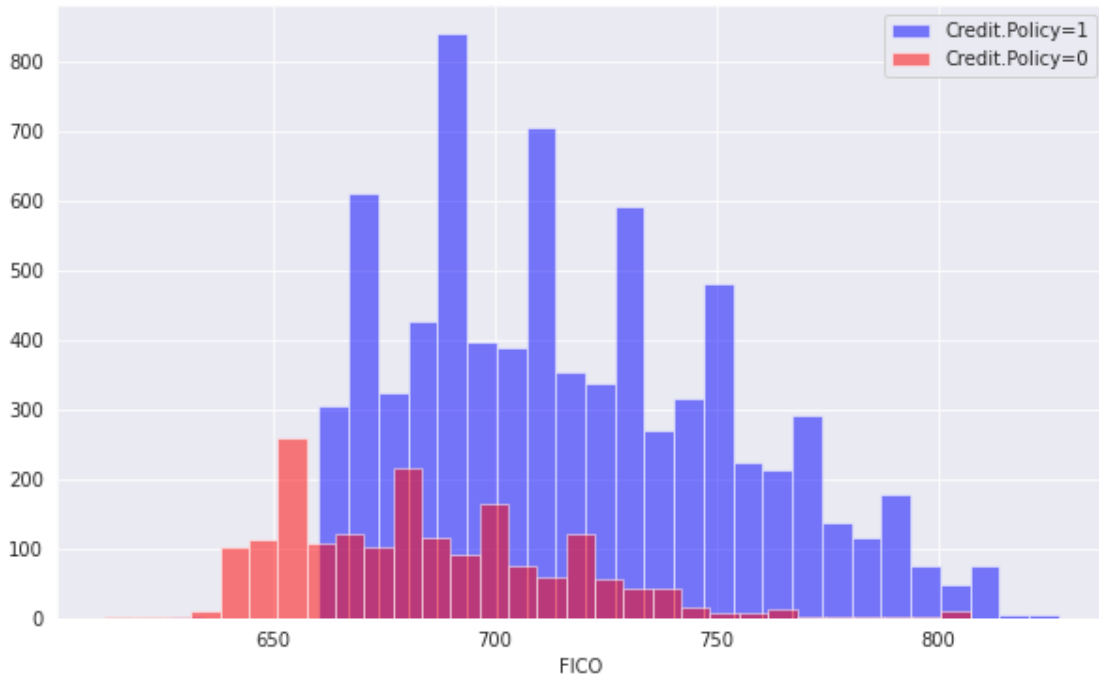
```
[16]: <AxesSubplot: >
```



```
[17]: df1=pd.get_dummies(df, columns=['purpose'])
```

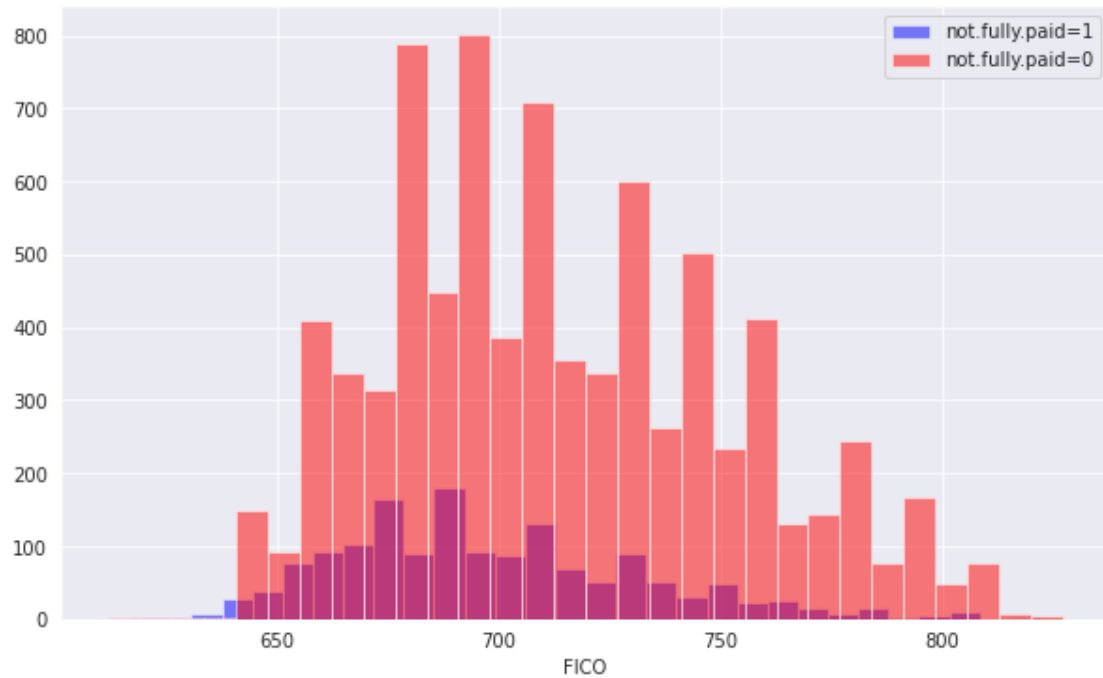
```
[18]: plt.figure(figsize=(10,6))
df[df['credit.policy']==1]['fico'].hist(alpha=0.5,color='blue',bins=30,label='Credit.Policy=1')
df[df['credit.policy']==0]['fico'].hist(alpha=0.5,color='red',bins=30,label='Credit.Policy=0')
plt.legend()
plt.xlabel('FICO')
```

[18]: Text(0.5, 0, 'FICO')



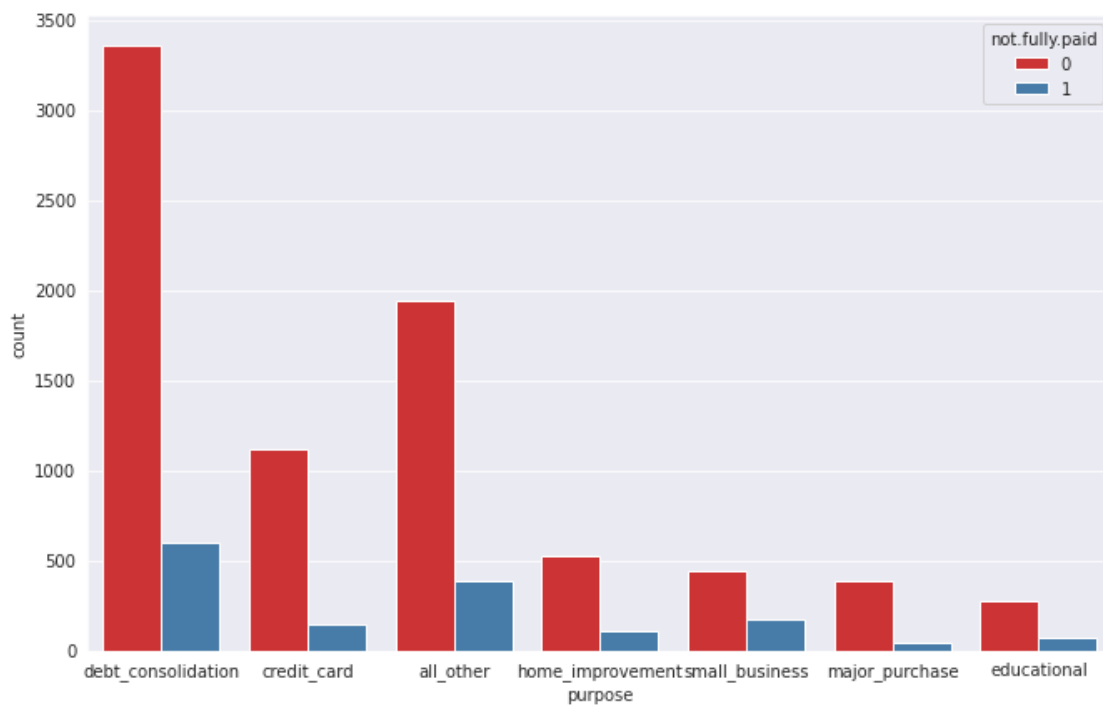
```
[19]: plt.figure(figsize=(10,6))
df[df['not.fully.paid']==1]['fico'].hist(alpha=0.5,color='blue',
                                           bins=30,label='not.fully.paid=1')
df[df['not.fully.paid']==0]['fico'].hist(alpha=0.5,color='red',
                                           bins=30,label='not.fully.paid=0')
plt.legend()
plt.xlabel('FICO')
```

[19]: Text(0.5, 0, 'FICO')



```
[20]: plt.figure(figsize=(11,7))
      sns.countplot(x='purpose',hue='not.fully.paid',data=df,palette='Set1')
```

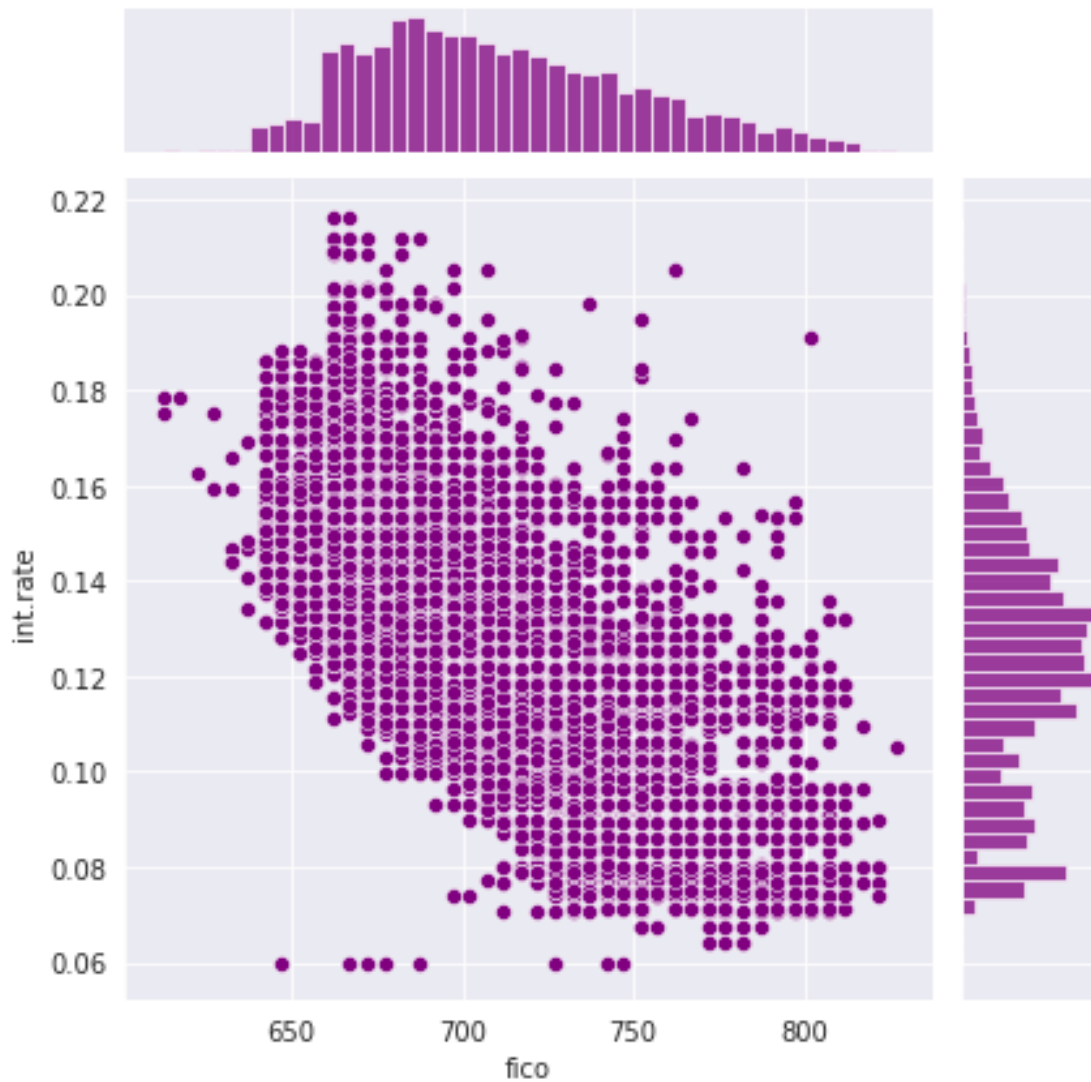
```
[20]: <AxesSubplot: xlabel='purpose', ylabel='count'>
```





```
[21]: sns.jointplot(x='fico',y='int.rate',data=df,color='purple')
```

```
[21]: <seaborn.axisgrid.JointGrid at 0x7fef47e06320>
```



```
[22]: plt.figure(figsize=(11,7))
sns.lmplot(y='int.rate',x='fico',data=df,hue='credit.policy',
          col='not.fully.paid',palette='Set1')
```

```
[22]: <seaborn.axisgrid.FacetGrid at 0x7fef479e6740>
```

<Figure size 792x504 with 0 Axes>



```
[24]: #cat =df_test_over
cat = ['purpose']
```

```
[25]: #final_data = pd.get_dummies(df,columns=cat,drop_first=True)
final_data = pd.get_dummies(df_test_over,columns=cat,drop_first=True)
```

```
[26]: final_data.info()
final_data.head()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 16090 entries, 0 to 7990
Data columns (total 19 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   credit.policy                        16090 non-null  int64
1   int.rate                            16090 non-null  float64
2   installment                         16090 non-null  float64
3   log.annual.inc                     16090 non-null  float64
4   dti                                 16090 non-null  float64
5   fico                               16090 non-null  int64
6   days.with.cr.line                  16090 non-null  float64
7   revol.bal                          16090 non-null  int64
8   revol.util                         16090 non-null  float64
9   inq.last.6mths                    16090 non-null  int64
10  delinq.2yrs                       16090 non-null  int64
11  pub.rec                           16090 non-null  int64
12  not.fully.paid                    16090 non-null  int64
13  purpose_credit_card               16090 non-null  uint8
14  purpose_debt_consolidation        16090 non-null  uint8
15  purpose_educational               16090 non-null  uint8
```

```

16 purpose_home_improvement    16090 non-null uint8
17 purpose_major_purchase      16090 non-null uint8
18 purpose_small_business      16090 non-null uint8
dtypes: float64(6), int64(7), uint8(6)
memory usage: 1.8 MB

```

```

[26]: credit.policy  int.rate  installment  log.annual.inc    dti  fico  \
0          1    0.1189      829.10      11.350407  19.48  737
1          1    0.1071      228.22      11.082143  14.29  707
2          1    0.1357      366.86      10.373491  11.63  682
3          1    0.1008      162.34      11.350407   8.10  712
4          1    0.1426      102.92      11.299732  14.97  667

      days.with.cr.line  revol.bal  revol.util  inq.last.6mths  delinq.2yrs  \
0      5639.958333      28854      52.1              0              0
1      2760.000000      33623      76.7              0              0
2      4710.000000       3511      25.6              1              0
3      2699.958333      33667      73.2              1              0
4      4066.000000       4740      39.5              0              1

      pub.rec  not.fully.paid  purpose_credit_card  purpose_debt_consolidation  \
0          0          0          0          1
1          0          0          1          0
2          0          0          0          1
3          0          0          0          1
4          0          0          1          0

      purpose_educational  purpose_home_improvement  purpose_major_purchase  \
0              0              0              0
1              0              0              0
2              0              0              0
3              0              0              0
4              0              0              0

      purpose_small_business
0              0
1              0
2              0
3              0
4              0

```

```
[27]: final_data.corr()
```

```

[27]: credit.policy  int.rate  installment  \
credit.policy      1.000000 -0.289283    0.061706
int.rate          -0.289283  1.000000    0.268816
installment       0.061706  0.268816    1.000000

```

log.annual.inc	0.021497	0.076230	0.474401
dti	-0.094707	0.193884	0.018823
fico	0.378253	-0.678045	0.117787
days.with.cr.line	0.092575	-0.096885	0.182100
revol.bal	-0.188060	0.090163	0.254101
revol.util	-0.093077	0.418919	0.045253
inq.last.6mths	-0.547679	0.185112	-0.021168
delinq.2yrs	-0.063473	0.152646	-0.002454
pub.rec	-0.072808	0.105766	-0.023658
not.fully.paid	-0.200215	0.225666	0.068163
purpose_credit_card	0.002024	-0.042292	0.007798
purpose_debt_consolidation	0.039930	0.075892	0.106183
purpose_educational	-0.029810	-0.015291	-0.103938
purpose_home_improvement	-0.017452	-0.041250	0.035959
purpose_major_purchase	0.028011	-0.060634	-0.051060
purpose_small_business	-0.002790	0.183748	0.187774

	log.annual.inc	dti	fico \
credit.policy	0.021497	-0.094707	0.378253
int.rate	0.076230	0.193884	-0.678045
installment	0.474401	0.018823	0.117787
log.annual.inc	1.000000	-0.038214	0.110237
dti	-0.038214	1.000000	-0.216070
fico	0.110237	-0.216070	1.000000
days.with.cr.line	0.344314	0.091449	0.257549
revol.bal	0.398760	0.179443	0.002416
revol.util	0.070953	0.322415	-0.501762
inq.last.6mths	0.032401	0.028583	-0.193626
delinq.2yrs	0.012771	-0.039978	-0.203008
pub.rec	0.022025	0.028230	-0.157496
not.fully.paid	-0.041956	0.051480	-0.211935
purpose_credit_card	0.081272	0.075680	-0.008011
purpose_debt_consolidation	-0.036586	0.174768	-0.140090
purpose_educational	-0.133568	-0.030793	-0.020571
purpose_home_improvement	0.107927	-0.088411	0.092760
purpose_major_purchase	-0.025726	-0.085678	0.058652
purpose_small_business	0.111604	-0.065699	0.071830

	days.with.cr.line	revol.bal	revol.util \
credit.policy	0.092575	-0.188060	-0.093077
int.rate	-0.096885	0.090163	0.418919
installment	0.182100	0.254101	0.045253
log.annual.inc	0.344314	0.398760	0.070953
dti	0.091449	0.179443	0.322415
fico	0.257549	0.002416	-0.501762
days.with.cr.line	1.000000	0.279911	0.013468
revol.bal	0.279911	1.000000	0.181225

revol.util	0.013468	0.181225	1.000000
inq.last.6mths	-0.029292	0.014048	-0.026215
delinq.2yrs	0.085697	-0.027238	-0.048792
pub.rec	0.064314	-0.037237	0.077071
not.fully.paid	-0.046594	0.057401	0.117790
purpose_credit_card	0.056971	0.060689	0.082524
purpose_debt_consolidation	-0.002262	-0.013797	0.190038
purpose_educational	-0.056818	-0.038605	-0.049580
purpose_home_improvement	0.059564	-0.016523	-0.117841
purpose_major_purchase	-0.028091	-0.060458	-0.111932
purpose_small_business	0.046484	0.091854	-0.061126

	inq.last.6mths	delinq.2yrs	pub.rec \
credit.policy	-0.547679	-0.063473	-0.072808
int.rate	0.185112	0.152646	0.105766
installment	-0.021168	-0.002454	-0.023658
log.annual.inc	0.032401	0.012771	0.022025
dti	0.028583	-0.039978	0.028230
fico	-0.193626	-0.203008	-0.157496
days.with.cr.line	-0.029292	0.085697	0.064314
revol.bal	0.014048	-0.027238	-0.037237
revol.util	-0.026215	-0.048792	0.077071
inq.last.6mths	1.000000	0.017783	0.103086
delinq.2yrs	0.017783	1.000000	-0.002125
pub.rec	0.103086	-0.002125	1.000000
not.fully.paid	0.184368	0.009774	0.061461
purpose_credit_card	-0.032431	-0.004863	0.030776
purpose_debt_consolidation	-0.054588	-0.014405	0.044514
purpose_educational	0.027542	0.008499	-0.019363
purpose_home_improvement	0.070591	-0.015685	-0.000424
purpose_major_purchase	0.003339	0.007183	-0.016291
purpose_small_business	0.036802	0.023252	-0.001695

	not.fully.paid	purpose_credit_card \
credit.policy	-0.200215	0.002024
int.rate	0.225666	-0.042292
installment	0.068163	0.007798
log.annual.inc	-0.041956	0.081272
dti	0.051480	0.075680
fico	-0.211935	-0.008011
days.with.cr.line	-0.046594	0.056971
revol.bal	0.057401	0.060689
revol.util	0.117790	0.082524
inq.last.6mths	0.184368	-0.032431
delinq.2yrs	0.009774	-0.004863
pub.rec	0.061461	0.030776
not.fully.paid	1.000000	-0.067058

purpose_credit_card	-0.067058	1.000000
purpose_debt_consolidation	-0.023673	-0.300718
purpose_educational	0.037387	-0.075815
purpose_home_improvement	0.016186	-0.099161
purpose_major_purchase	-0.042900	-0.074202
purpose_small_business	0.099555	-0.109618

	purpose_debt_consolidation	purpose_educational \
credit.policy	0.039930	-0.029810
int.rate	0.075892	-0.015291
installment	0.106183	-0.103938
log.annual.inc	-0.036586	-0.133568
dti	0.174768	-0.030793
fico	-0.140090	-0.020571
days.with.cr.line	-0.002262	-0.056818
revol.bal	-0.013797	-0.038605
revol.util	0.190038	-0.049580
inq.last.6mths	-0.054588	0.027542
delinq.2yrs	-0.014405	0.008499
pub.rec	0.044514	-0.019363
not.fully.paid	-0.023673	0.037387
purpose_credit_card	-0.300718	-0.075815
purpose_debt_consolidation	1.000000	-0.171807
purpose_educational	-0.171807	1.000000
purpose_home_improvement	-0.224714	-0.056653
purpose_major_purchase	-0.168152	-0.042393
purpose_small_business	-0.248412	-0.062628

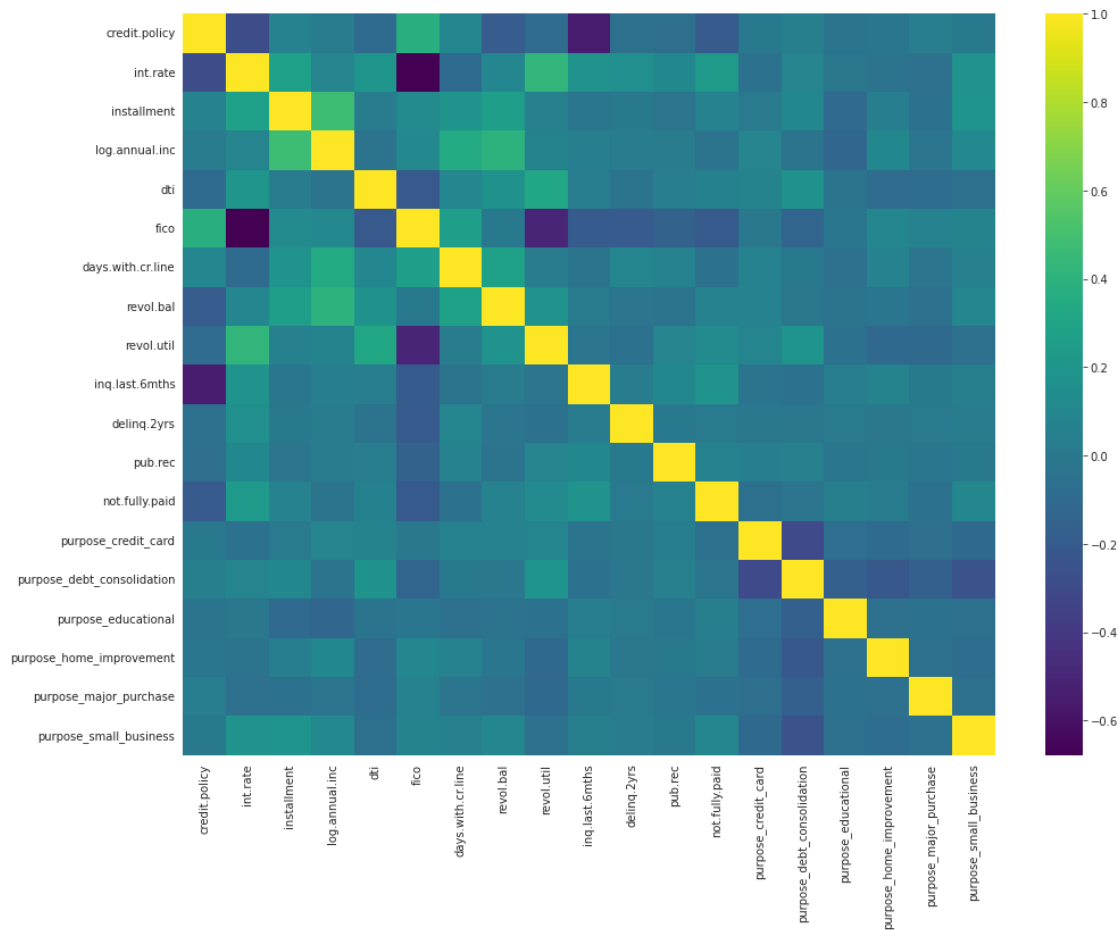
	purpose_home_improvement	purpose_major_purchase \
credit.policy	-0.017452	0.028011
int.rate	-0.041250	-0.060634
installment	0.035959	-0.051060
log.annual.inc	0.107927	-0.025726
dti	-0.088411	-0.085678
fico	0.092760	0.058652
days.with.cr.line	0.059564	-0.028091
revol.bal	-0.016523	-0.060458
revol.util	-0.117841	-0.111932
inq.last.6mths	0.070591	0.003339
delinq.2yrs	-0.015685	0.007183
pub.rec	-0.000424	-0.016291
not.fully.paid	0.016186	-0.042900
purpose_credit_card	-0.099161	-0.074202
purpose_debt_consolidation	-0.224714	-0.168152
purpose_educational	-0.056653	-0.042393
purpose_home_improvement	1.000000	-0.055448
purpose_major_purchase	-0.055448	1.000000

purpose_small_business	-0.081913	-0.061295
	purpose_small_business	
credit.policy	-0.002790	
int.rate	0.183748	
installment	0.187774	
log.annual.inc	0.111604	
dti	-0.065699	
fico	0.071830	
days.with.cr.line	0.046484	
revol.bal	0.091854	
revol.util	-0.061126	
inq.last.6mths	0.036802	
delinq.2yrs	0.023252	
pub.rec	-0.001695	
not.fully.paid	0.099555	
purpose_credit_card	-0.109618	
purpose_debt_consolidation	-0.248412	
purpose_educational	-0.062628	
purpose_home_improvement	-0.081913	
purpose_major_purchase	-0.061295	
purpose_small_business	1.000000	

```
[28]: plt.figure(
        figsize=[16,12]
    )

    sns.heatmap(
        data=final_data.corr(),
        cmap='viridis',
        annot=False,
        fmt='.2g'
    )
```

```
[28]: <AxesSubplot: >
```



```
[29]: #focus on the grids of yellow or very light green. After comparing with the
      ↪ feature description again,  revol.bal,day.with.cr.line,installment can
      ↪ represent by annual income. revol.util can represent by int.rate,
to_drop2 = ['revol.bal', 'days.with.cr.line', 'installment', 'revol.util']

final_data.drop(to_drop2, axis=1, inplace=True)
```

```
[30]: final_data.isnull().mean()
```

```
[30]: credit.policy      0.0
      int.rate          0.0
      log.annual.inc    0.0
      dti              0.0
      fico             0.0
      revol.util       0.0
      inq.last.6mths   0.0
      delinq.2yrs      0.0
      pub.rec          0.0
```



```
not.fully.paid          0.0
purpose_credit_card     0.0
purpose_debt_consolidation 0.0
purpose_educational     0.0
purpose_home_improvement 0.0
purpose_major_purchase   0.0
purpose_small_business   0.0
dtype: float64
```

```
[31]: to_train = final_data[final_data['not.fully.paid'].isin([0,1])]
      to_pred = final_data[final_data['not.fully.paid'] == 2]
```

```
[32]: X = to_train.drop('not.fully.paid', axis=1).values
      y = to_train['not.fully.paid'].values

      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,
      ↪random_state = 101)
```

```
[34]: scaler = MinMaxScaler()
      X_train = scaler.fit_transform(X_train)
      X_test = scaler.transform(X_test)
```

```
[35]: X_train.shape
```

```
[35]: (11263, 15)
```

```
[36]: model = Sequential()

      model.add(
          Dense(94, activation='relu')
      )

      model.add(
          Dense(30, activation='relu')
      )

      model.add(
          Dense(15, activation='relu')
      )

      model.add(
          Dense(1, activation='sigmoid')
      )

      model.compile(
          optimizer='adam',
```

```

        loss='binary_crossentropy',
        metrics=['accuracy']
    )

```

```

[37]: early_stop = EarlyStopping(
        monitor='val_loss',
        mode='min',
        verbose=1,
        patience=25
    )

    model.fit(
        X_train,
        y_train,
        epochs=200,
        batch_size=256,
        validation_data=(X_test, y_test),
        callbacks=[early_stop]
    )

```

```

Epoch 1/200
44/44 [=====] - 1s 5ms/step - loss: 0.6743 - accuracy:
0.5922 - val_loss: 0.6592 - val_accuracy: 0.6091
Epoch 2/200
44/44 [=====] - 0s 2ms/step - loss: 0.6504 - accuracy:
0.6204 - val_loss: 0.6509 - val_accuracy: 0.6095
Epoch 3/200
44/44 [=====] - 0s 2ms/step - loss: 0.6443 - accuracy:
0.6222 - val_loss: 0.6479 - val_accuracy: 0.6143
Epoch 4/200
44/44 [=====] - 0s 2ms/step - loss: 0.6409 - accuracy:
0.6228 - val_loss: 0.6458 - val_accuracy: 0.6124
Epoch 5/200
44/44 [=====] - 0s 2ms/step - loss: 0.6401 - accuracy:
0.6275 - val_loss: 0.6441 - val_accuracy: 0.6159
Epoch 6/200
44/44 [=====] - 0s 2ms/step - loss: 0.6370 - accuracy:
0.6283 - val_loss: 0.6435 - val_accuracy: 0.6180
Epoch 7/200
44/44 [=====] - 0s 2ms/step - loss: 0.6357 - accuracy:
0.6349 - val_loss: 0.6419 - val_accuracy: 0.6236
Epoch 8/200
44/44 [=====] - 0s 2ms/step - loss: 0.6348 - accuracy:
0.6277 - val_loss: 0.6443 - val_accuracy: 0.6161
Epoch 9/200
44/44 [=====] - 0s 2ms/step - loss: 0.6329 - accuracy:
0.6319 - val_loss: 0.6399 - val_accuracy: 0.6261

```

Epoch 10/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6325 - accuracy:  
0.6361 - val\_loss: 0.6395 - val\_accuracy: 0.6242  
Epoch 11/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6317 - accuracy:  
0.6330 - val\_loss: 0.6407 - val\_accuracy: 0.6279  
Epoch 12/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6314 - accuracy:  
0.6367 - val\_loss: 0.6390 - val\_accuracy: 0.6225  
Epoch 13/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6296 - accuracy:  
0.6386 - val\_loss: 0.6400 - val\_accuracy: 0.6176  
Epoch 14/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6292 - accuracy:  
0.6370 - val\_loss: 0.6377 - val\_accuracy: 0.6335  
Epoch 15/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6281 - accuracy:  
0.6413 - val\_loss: 0.6380 - val\_accuracy: 0.6308  
Epoch 16/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6282 - accuracy:  
0.6363 - val\_loss: 0.6372 - val\_accuracy: 0.6275  
Epoch 17/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6265 - accuracy:  
0.6430 - val\_loss: 0.6367 - val\_accuracy: 0.6341  
Epoch 18/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6247 - accuracy:  
0.6410 - val\_loss: 0.6369 - val\_accuracy: 0.6333  
Epoch 19/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6240 - accuracy:  
0.6430 - val\_loss: 0.6355 - val\_accuracy: 0.6339  
Epoch 20/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6233 - accuracy:  
0.6481 - val\_loss: 0.6358 - val\_accuracy: 0.6377  
Epoch 21/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6230 - accuracy:  
0.6421 - val\_loss: 0.6351 - val\_accuracy: 0.6381  
Epoch 22/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6213 - accuracy:  
0.6475 - val\_loss: 0.6358 - val\_accuracy: 0.6339  
Epoch 23/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6221 - accuracy:  
0.6464 - val\_loss: 0.6342 - val\_accuracy: 0.6333  
Epoch 24/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6194 - accuracy:  
0.6504 - val\_loss: 0.6399 - val\_accuracy: 0.6269  
Epoch 25/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6229 - accuracy:  
0.6465 - val\_loss: 0.6349 - val\_accuracy: 0.6343

Epoch 26/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6186 - accuracy: 0.6495 - val\_loss: 0.6345 - val\_accuracy: 0.6246  
Epoch 27/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6175 - accuracy: 0.6482 - val\_loss: 0.6328 - val\_accuracy: 0.6346  
Epoch 28/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6172 - accuracy: 0.6488 - val\_loss: 0.6336 - val\_accuracy: 0.6290  
Epoch 29/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6162 - accuracy: 0.6489 - val\_loss: 0.6329 - val\_accuracy: 0.6383  
Epoch 30/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6159 - accuracy: 0.6505 - val\_loss: 0.6331 - val\_accuracy: 0.6341  
Epoch 31/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6142 - accuracy: 0.6541 - val\_loss: 0.6315 - val\_accuracy: 0.6372  
Epoch 32/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6146 - accuracy: 0.6497 - val\_loss: 0.6312 - val\_accuracy: 0.6323  
Epoch 33/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6134 - accuracy: 0.6533 - val\_loss: 0.6305 - val\_accuracy: 0.6381  
Epoch 34/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6119 - accuracy: 0.6536 - val\_loss: 0.6318 - val\_accuracy: 0.6323  
Epoch 35/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6101 - accuracy: 0.6573 - val\_loss: 0.6320 - val\_accuracy: 0.6343  
Epoch 36/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6112 - accuracy: 0.6534 - val\_loss: 0.6305 - val\_accuracy: 0.6360  
Epoch 37/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6098 - accuracy: 0.6576 - val\_loss: 0.6298 - val\_accuracy: 0.6391  
Epoch 38/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6095 - accuracy: 0.6545 - val\_loss: 0.6321 - val\_accuracy: 0.6387  
Epoch 39/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6073 - accuracy: 0.6567 - val\_loss: 0.6293 - val\_accuracy: 0.6404  
Epoch 40/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6062 - accuracy: 0.6575 - val\_loss: 0.6287 - val\_accuracy: 0.6399  
Epoch 41/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6065 - accuracy: 0.6590 - val\_loss: 0.6324 - val\_accuracy: 0.6288

Epoch 42/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6041 - accuracy:  
0.6606 - val\_loss: 0.6279 - val\_accuracy: 0.6341  
Epoch 43/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6041 - accuracy:  
0.6579 - val\_loss: 0.6280 - val\_accuracy: 0.6364  
Epoch 44/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6074 - accuracy:  
0.6523 - val\_loss: 0.6310 - val\_accuracy: 0.6314  
Epoch 45/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6037 - accuracy:  
0.6596 - val\_loss: 0.6269 - val\_accuracy: 0.6433  
Epoch 46/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6011 - accuracy:  
0.6658 - val\_loss: 0.6297 - val\_accuracy: 0.6395  
Epoch 47/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6037 - accuracy:  
0.6571 - val\_loss: 0.6371 - val\_accuracy: 0.6331  
Epoch 48/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5996 - accuracy:  
0.6644 - val\_loss: 0.6300 - val\_accuracy: 0.6408  
Epoch 49/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5991 - accuracy:  
0.6625 - val\_loss: 0.6285 - val\_accuracy: 0.6410  
Epoch 50/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5959 - accuracy:  
0.6702 - val\_loss: 0.6272 - val\_accuracy: 0.6418  
Epoch 51/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5955 - accuracy:  
0.6697 - val\_loss: 0.6264 - val\_accuracy: 0.6404  
Epoch 52/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5940 - accuracy:  
0.6684 - val\_loss: 0.6265 - val\_accuracy: 0.6406  
Epoch 53/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5939 - accuracy:  
0.6658 - val\_loss: 0.6267 - val\_accuracy: 0.6406  
Epoch 54/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5916 - accuracy:  
0.6718 - val\_loss: 0.6270 - val\_accuracy: 0.6422  
Epoch 55/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5911 - accuracy:  
0.6753 - val\_loss: 0.6315 - val\_accuracy: 0.6350  
Epoch 56/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5909 - accuracy:  
0.6725 - val\_loss: 0.6253 - val\_accuracy: 0.6404  
Epoch 57/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5892 - accuracy:  
0.6775 - val\_loss: 0.6275 - val\_accuracy: 0.6433

Epoch 58/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5887 - accuracy: 0.6751 - val\_loss: 0.6246 - val\_accuracy: 0.6397  
Epoch 59/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5890 - accuracy: 0.6742 - val\_loss: 0.6282 - val\_accuracy: 0.6377  
Epoch 60/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5899 - accuracy: 0.6726 - val\_loss: 0.6259 - val\_accuracy: 0.6476  
Epoch 61/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5860 - accuracy: 0.6786 - val\_loss: 0.6235 - val\_accuracy: 0.6449  
Epoch 62/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5859 - accuracy: 0.6766 - val\_loss: 0.6263 - val\_accuracy: 0.6408  
Epoch 63/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5844 - accuracy: 0.6767 - val\_loss: 0.6285 - val\_accuracy: 0.6433  
Epoch 64/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5821 - accuracy: 0.6781 - val\_loss: 0.6272 - val\_accuracy: 0.6457  
Epoch 65/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5830 - accuracy: 0.6819 - val\_loss: 0.6235 - val\_accuracy: 0.6426  
Epoch 66/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5815 - accuracy: 0.6771 - val\_loss: 0.6253 - val\_accuracy: 0.6497  
Epoch 67/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5798 - accuracy: 0.6844 - val\_loss: 0.6270 - val\_accuracy: 0.6399  
Epoch 68/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5796 - accuracy: 0.6827 - val\_loss: 0.6257 - val\_accuracy: 0.6491  
Epoch 69/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5782 - accuracy: 0.6842 - val\_loss: 0.6232 - val\_accuracy: 0.6532  
Epoch 70/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5769 - accuracy: 0.6837 - val\_loss: 0.6254 - val\_accuracy: 0.6522  
Epoch 71/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5788 - accuracy: 0.6835 - val\_loss: 0.6264 - val\_accuracy: 0.6522  
Epoch 72/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5751 - accuracy: 0.6834 - val\_loss: 0.6226 - val\_accuracy: 0.6482  
Epoch 73/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5759 - accuracy: 0.6869 - val\_loss: 0.6205 - val\_accuracy: 0.6470

Epoch 74/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5723 - accuracy:  
0.6916 - val\_loss: 0.6230 - val\_accuracy: 0.6470  
Epoch 75/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5740 - accuracy:  
0.6907 - val\_loss: 0.6281 - val\_accuracy: 0.6433  
Epoch 76/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5757 - accuracy:  
0.6821 - val\_loss: 0.6207 - val\_accuracy: 0.6451  
Epoch 77/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5762 - accuracy:  
0.6837 - val\_loss: 0.6259 - val\_accuracy: 0.6482  
Epoch 78/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5704 - accuracy:  
0.6912 - val\_loss: 0.6205 - val\_accuracy: 0.6480  
Epoch 79/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5695 - accuracy:  
0.6899 - val\_loss: 0.6203 - val\_accuracy: 0.6493  
Epoch 80/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5683 - accuracy:  
0.6896 - val\_loss: 0.6214 - val\_accuracy: 0.6520  
Epoch 81/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5689 - accuracy:  
0.6956 - val\_loss: 0.6173 - val\_accuracy: 0.6542  
Epoch 82/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5660 - accuracy:  
0.6963 - val\_loss: 0.6193 - val\_accuracy: 0.6549  
Epoch 83/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5660 - accuracy:  
0.6933 - val\_loss: 0.6204 - val\_accuracy: 0.6507  
Epoch 84/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5654 - accuracy:  
0.6983 - val\_loss: 0.6185 - val\_accuracy: 0.6592  
Epoch 85/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5651 - accuracy:  
0.6897 - val\_loss: 0.6207 - val\_accuracy: 0.6551  
Epoch 86/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5624 - accuracy:  
0.6986 - val\_loss: 0.6193 - val\_accuracy: 0.6538  
Epoch 87/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5609 - accuracy:  
0.7000 - val\_loss: 0.6198 - val\_accuracy: 0.6573  
Epoch 88/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5627 - accuracy:  
0.6953 - val\_loss: 0.6187 - val\_accuracy: 0.6486  
Epoch 89/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5587 - accuracy:  
0.6979 - val\_loss: 0.6161 - val\_accuracy: 0.6522

Epoch 90/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5659 - accuracy: 0.6901 - val\_loss: 0.6167 - val\_accuracy: 0.6495  
Epoch 91/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5567 - accuracy: 0.7013 - val\_loss: 0.6167 - val\_accuracy: 0.6555  
Epoch 92/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5573 - accuracy: 0.7016 - val\_loss: 0.6179 - val\_accuracy: 0.6524  
Epoch 93/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5543 - accuracy: 0.7032 - val\_loss: 0.6158 - val\_accuracy: 0.6592  
Epoch 94/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5538 - accuracy: 0.7040 - val\_loss: 0.6205 - val\_accuracy: 0.6524  
Epoch 95/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5545 - accuracy: 0.7067 - val\_loss: 0.6237 - val\_accuracy: 0.6497  
Epoch 96/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5525 - accuracy: 0.6991 - val\_loss: 0.6134 - val\_accuracy: 0.6491  
Epoch 97/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5517 - accuracy: 0.7023 - val\_loss: 0.6268 - val\_accuracy: 0.6468  
Epoch 98/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5543 - accuracy: 0.7080 - val\_loss: 0.6131 - val\_accuracy: 0.6559  
Epoch 99/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5520 - accuracy: 0.7047 - val\_loss: 0.6217 - val\_accuracy: 0.6501  
Epoch 100/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5483 - accuracy: 0.7106 - val\_loss: 0.6117 - val\_accuracy: 0.6629  
Epoch 101/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5504 - accuracy: 0.7063 - val\_loss: 0.6118 - val\_accuracy: 0.6563  
Epoch 102/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5489 - accuracy: 0.7074 - val\_loss: 0.6145 - val\_accuracy: 0.6584  
Epoch 103/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5493 - accuracy: 0.7074 - val\_loss: 0.6266 - val\_accuracy: 0.6511  
Epoch 104/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5500 - accuracy: 0.7054 - val\_loss: 0.6191 - val\_accuracy: 0.6555  
Epoch 105/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5453 - accuracy: 0.7137 - val\_loss: 0.6135 - val\_accuracy: 0.6611



Epoch 106/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5453 - accuracy: 0.7098 - val\_loss: 0.6094 - val\_accuracy: 0.6580  
Epoch 107/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5437 - accuracy: 0.7100 - val\_loss: 0.6169 - val\_accuracy: 0.6507  
Epoch 108/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5443 - accuracy: 0.7082 - val\_loss: 0.6135 - val\_accuracy: 0.6563  
Epoch 109/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5416 - accuracy: 0.7106 - val\_loss: 0.6081 - val\_accuracy: 0.6590  
Epoch 110/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5395 - accuracy: 0.7162 - val\_loss: 0.6112 - val\_accuracy: 0.6592  
Epoch 111/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5420 - accuracy: 0.7110 - val\_loss: 0.6098 - val\_accuracy: 0.6598  
Epoch 112/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5431 - accuracy: 0.7120 - val\_loss: 0.6121 - val\_accuracy: 0.6621  
Epoch 113/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5415 - accuracy: 0.7114 - val\_loss: 0.6085 - val\_accuracy: 0.6658  
Epoch 114/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5416 - accuracy: 0.7112 - val\_loss: 0.6053 - val\_accuracy: 0.6644  
Epoch 115/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5403 - accuracy: 0.7131 - val\_loss: 0.6103 - val\_accuracy: 0.6652  
Epoch 116/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5365 - accuracy: 0.7169 - val\_loss: 0.6131 - val\_accuracy: 0.6698  
Epoch 117/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5374 - accuracy: 0.7138 - val\_loss: 0.6070 - val\_accuracy: 0.6671  
Epoch 118/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5369 - accuracy: 0.7154 - val\_loss: 0.6068 - val\_accuracy: 0.6725  
Epoch 119/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5357 - accuracy: 0.7193 - val\_loss: 0.6070 - val\_accuracy: 0.6718  
Epoch 120/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5384 - accuracy: 0.7099 - val\_loss: 0.6068 - val\_accuracy: 0.6625  
Epoch 121/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5340 - accuracy: 0.7174 - val\_loss: 0.6097 - val\_accuracy: 0.6689

Epoch 122/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5331 - accuracy:  
0.7185 - val\_loss: 0.6031 - val\_accuracy: 0.6739  
Epoch 123/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5326 - accuracy:  
0.7187 - val\_loss: 0.6076 - val\_accuracy: 0.6609  
Epoch 124/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5348 - accuracy:  
0.7168 - val\_loss: 0.6027 - val\_accuracy: 0.6673  
Epoch 125/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5314 - accuracy:  
0.7187 - val\_loss: 0.6059 - val\_accuracy: 0.6671  
Epoch 126/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5281 - accuracy:  
0.7244 - val\_loss: 0.6057 - val\_accuracy: 0.6679  
Epoch 127/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5305 - accuracy:  
0.7224 - val\_loss: 0.6027 - val\_accuracy: 0.6663  
Epoch 128/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5288 - accuracy:  
0.7239 - val\_loss: 0.6036 - val\_accuracy: 0.6718  
Epoch 129/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5262 - accuracy:  
0.7233 - val\_loss: 0.6066 - val\_accuracy: 0.6733  
Epoch 130/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5265 - accuracy:  
0.7260 - val\_loss: 0.6084 - val\_accuracy: 0.6656  
Epoch 131/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5299 - accuracy:  
0.7237 - val\_loss: 0.6016 - val\_accuracy: 0.6723  
Epoch 132/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5257 - accuracy:  
0.7219 - val\_loss: 0.6048 - val\_accuracy: 0.6689  
Epoch 133/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5261 - accuracy:  
0.7227 - val\_loss: 0.6080 - val\_accuracy: 0.6694  
Epoch 134/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5252 - accuracy:  
0.7242 - val\_loss: 0.6029 - val\_accuracy: 0.6766  
Epoch 135/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5237 - accuracy:  
0.7279 - val\_loss: 0.6063 - val\_accuracy: 0.6654  
Epoch 136/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5227 - accuracy:  
0.7283 - val\_loss: 0.5995 - val\_accuracy: 0.6714  
Epoch 137/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5200 - accuracy:  
0.7301 - val\_loss: 0.6055 - val\_accuracy: 0.6673

Epoch 138/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5213 - accuracy:  
0.7283 - val\_loss: 0.6009 - val\_accuracy: 0.6779  
Epoch 139/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5212 - accuracy:  
0.7245 - val\_loss: 0.6001 - val\_accuracy: 0.6710  
Epoch 140/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5168 - accuracy:  
0.7322 - val\_loss: 0.6003 - val\_accuracy: 0.6727  
Epoch 141/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5162 - accuracy:  
0.7284 - val\_loss: 0.6030 - val\_accuracy: 0.6723  
Epoch 142/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5178 - accuracy:  
0.7304 - val\_loss: 0.6001 - val\_accuracy: 0.6710  
Epoch 143/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5183 - accuracy:  
0.7290 - val\_loss: 0.6045 - val\_accuracy: 0.6814  
Epoch 144/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5184 - accuracy:  
0.7282 - val\_loss: 0.6156 - val\_accuracy: 0.6700  
Epoch 145/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5184 - accuracy:  
0.7297 - val\_loss: 0.6017 - val\_accuracy: 0.6700  
Epoch 146/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5191 - accuracy:  
0.7237 - val\_loss: 0.5985 - val\_accuracy: 0.6729  
Epoch 147/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5140 - accuracy:  
0.7316 - val\_loss: 0.6109 - val\_accuracy: 0.6671  
Epoch 148/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5160 - accuracy:  
0.7311 - val\_loss: 0.5973 - val\_accuracy: 0.6708  
Epoch 149/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5207 - accuracy:  
0.7241 - val\_loss: 0.6061 - val\_accuracy: 0.6737  
Epoch 150/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5138 - accuracy:  
0.7329 - val\_loss: 0.6112 - val\_accuracy: 0.6725  
Epoch 151/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5158 - accuracy:  
0.7286 - val\_loss: 0.6153 - val\_accuracy: 0.6644  
Epoch 152/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5101 - accuracy:  
0.7327 - val\_loss: 0.5957 - val\_accuracy: 0.6793  
Epoch 153/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5132 - accuracy:  
0.7318 - val\_loss: 0.6027 - val\_accuracy: 0.6721

Epoch 154/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5107 - accuracy: 0.7339 - val\_loss: 0.5996 - val\_accuracy: 0.6851  
Epoch 155/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5133 - accuracy: 0.7301 - val\_loss: 0.6043 - val\_accuracy: 0.6772  
Epoch 156/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5089 - accuracy: 0.7373 - val\_loss: 0.5964 - val\_accuracy: 0.6789  
Epoch 157/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5089 - accuracy: 0.7369 - val\_loss: 0.6069 - val\_accuracy: 0.6675  
Epoch 158/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5101 - accuracy: 0.7299 - val\_loss: 0.6030 - val\_accuracy: 0.6750  
Epoch 159/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5056 - accuracy: 0.7388 - val\_loss: 0.5992 - val\_accuracy: 0.6741  
Epoch 160/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5057 - accuracy: 0.7384 - val\_loss: 0.5969 - val\_accuracy: 0.6783  
Epoch 161/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5077 - accuracy: 0.7344 - val\_loss: 0.6058 - val\_accuracy: 0.6805  
Epoch 162/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5068 - accuracy: 0.7346 - val\_loss: 0.5998 - val\_accuracy: 0.6766  
Epoch 163/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5058 - accuracy: 0.7366 - val\_loss: 0.5976 - val\_accuracy: 0.6845  
Epoch 164/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5103 - accuracy: 0.7289 - val\_loss: 0.6044 - val\_accuracy: 0.6702  
Epoch 165/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5005 - accuracy: 0.7387 - val\_loss: 0.5953 - val\_accuracy: 0.6892  
Epoch 166/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5022 - accuracy: 0.7390 - val\_loss: 0.5950 - val\_accuracy: 0.6814  
Epoch 167/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5011 - accuracy: 0.7389 - val\_loss: 0.6002 - val\_accuracy: 0.6776  
Epoch 168/200  
44/44 [=====] - 0s 2ms/step - loss: 0.4987 - accuracy: 0.7407 - val\_loss: 0.6029 - val\_accuracy: 0.6747  
Epoch 169/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5020 - accuracy: 0.7397 - val\_loss: 0.5999 - val\_accuracy: 0.6745

Epoch 170/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5003 - accuracy: 0.7404 - val\_loss: 0.6013 - val\_accuracy: 0.6793  
Epoch 171/200  
44/44 [=====] - 0s 2ms/step - loss: 0.4997 - accuracy: 0.7422 - val\_loss: 0.5958 - val\_accuracy: 0.6832  
Epoch 172/200  
44/44 [=====] - 0s 2ms/step - loss: 0.4976 - accuracy: 0.7406 - val\_loss: 0.6041 - val\_accuracy: 0.6847  
Epoch 173/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5015 - accuracy: 0.7396 - val\_loss: 0.5923 - val\_accuracy: 0.6832  
Epoch 174/200  
44/44 [=====] - 0s 2ms/step - loss: 0.4957 - accuracy: 0.7416 - val\_loss: 0.6024 - val\_accuracy: 0.6762  
Epoch 175/200  
44/44 [=====] - 0s 2ms/step - loss: 0.4990 - accuracy: 0.7384 - val\_loss: 0.6042 - val\_accuracy: 0.6756  
Epoch 176/200  
44/44 [=====] - 0s 2ms/step - loss: 0.4974 - accuracy: 0.7425 - val\_loss: 0.5979 - val\_accuracy: 0.6820  
Epoch 177/200  
44/44 [=====] - 0s 2ms/step - loss: 0.4967 - accuracy: 0.7415 - val\_loss: 0.6079 - val\_accuracy: 0.6638  
Epoch 178/200  
44/44 [=====] - 0s 2ms/step - loss: 0.4984 - accuracy: 0.7447 - val\_loss: 0.5978 - val\_accuracy: 0.6845  
Epoch 179/200  
44/44 [=====] - 0s 2ms/step - loss: 0.4948 - accuracy: 0.7453 - val\_loss: 0.5960 - val\_accuracy: 0.6853  
Epoch 180/200  
44/44 [=====] - 0s 2ms/step - loss: 0.4941 - accuracy: 0.7447 - val\_loss: 0.5952 - val\_accuracy: 0.6863  
Epoch 181/200  
44/44 [=====] - 0s 2ms/step - loss: 0.4902 - accuracy: 0.7447 - val\_loss: 0.6054 - val\_accuracy: 0.6681  
Epoch 182/200  
44/44 [=====] - 0s 2ms/step - loss: 0.4929 - accuracy: 0.7423 - val\_loss: 0.5904 - val\_accuracy: 0.6874  
Epoch 183/200  
44/44 [=====] - 0s 2ms/step - loss: 0.4944 - accuracy: 0.7422 - val\_loss: 0.6038 - val\_accuracy: 0.6679  
Epoch 184/200  
44/44 [=====] - 0s 2ms/step - loss: 0.4893 - accuracy: 0.7439 - val\_loss: 0.5933 - val\_accuracy: 0.6824  
Epoch 185/200  
44/44 [=====] - 0s 2ms/step - loss: 0.4891 - accuracy: 0.7436 - val\_loss: 0.6114 - val\_accuracy: 0.6700

```

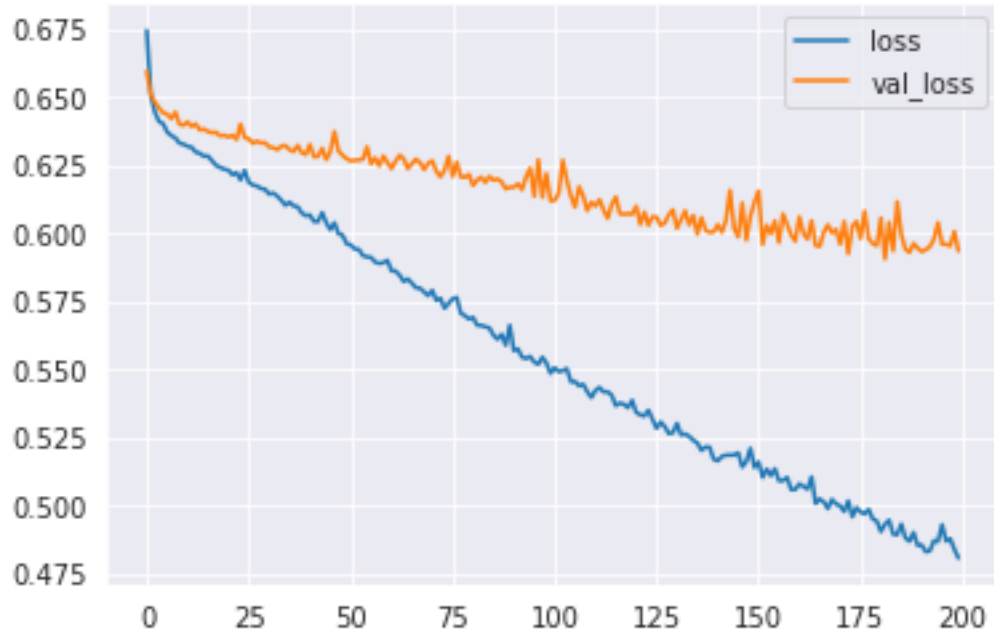
Epoch 186/200
44/44 [=====] - 0s 2ms/step - loss: 0.4930 - accuracy:
0.7432 - val_loss: 0.5994 - val_accuracy: 0.6845
Epoch 187/200
44/44 [=====] - 0s 2ms/step - loss: 0.4879 - accuracy:
0.7481 - val_loss: 0.5939 - val_accuracy: 0.6874
Epoch 188/200
44/44 [=====] - 0s 2ms/step - loss: 0.4866 - accuracy:
0.7498 - val_loss: 0.5928 - val_accuracy: 0.6911
Epoch 189/200
44/44 [=====] - 0s 2ms/step - loss: 0.4899 - accuracy:
0.7446 - val_loss: 0.5960 - val_accuracy: 0.6839
Epoch 190/200
44/44 [=====] - 0s 2ms/step - loss: 0.4852 - accuracy:
0.7518 - val_loss: 0.5943 - val_accuracy: 0.6824
Epoch 191/200
44/44 [=====] - 0s 2ms/step - loss: 0.4853 - accuracy:
0.7515 - val_loss: 0.5931 - val_accuracy: 0.6863
Epoch 192/200
44/44 [=====] - 0s 2ms/step - loss: 0.4828 - accuracy:
0.7500 - val_loss: 0.5937 - val_accuracy: 0.6837
Epoch 193/200
44/44 [=====] - 0s 2ms/step - loss: 0.4831 - accuracy:
0.7530 - val_loss: 0.5948 - val_accuracy: 0.6899
Epoch 194/200
44/44 [=====] - 0s 2ms/step - loss: 0.4869 - accuracy:
0.7478 - val_loss: 0.5976 - val_accuracy: 0.6785
Epoch 195/200
44/44 [=====] - 0s 2ms/step - loss: 0.4866 - accuracy:
0.7509 - val_loss: 0.6037 - val_accuracy: 0.6843
Epoch 196/200
44/44 [=====] - 0s 2ms/step - loss: 0.4927 - accuracy:
0.7422 - val_loss: 0.5958 - val_accuracy: 0.6874
Epoch 197/200
44/44 [=====] - 0s 2ms/step - loss: 0.4867 - accuracy:
0.7497 - val_loss: 0.5958 - val_accuracy: 0.6849
Epoch 198/200
44/44 [=====] - 0s 2ms/step - loss: 0.4876 - accuracy:
0.7472 - val_loss: 0.5951 - val_accuracy: 0.6826
Epoch 199/200
44/44 [=====] - 0s 2ms/step - loss: 0.4837 - accuracy:
0.7527 - val_loss: 0.6004 - val_accuracy: 0.6805
Epoch 200/200
44/44 [=====] - 0s 2ms/step - loss: 0.4805 - accuracy:
0.7506 - val_loss: 0.5933 - val_accuracy: 0.6874

```

[37]: <keras.src.callbacks.History at 0x7fef3d825ed0>

```
[39]: pd.DataFrame(model.history.history)[['loss', 'val_loss']].plot()
      #over fitting
```

```
[39]: <AxesSubplot: >
```



```
[44]: predictions = (model.predict(X_test) > 0.5).astype("int32")

print(
    confusion_matrix(y_test, predictions),
    '\n',
    classification_report(y_test, predictions)
)
```

```
151/151 [=====] - 0s 595us/step
```

```
[[1597 840]
```

```
 [ 669 1721]]
```

	precision	recall	f1-score	support
0	0.70	0.66	0.68	2437
1	0.67	0.72	0.70	2390
accuracy			0.69	4827
macro avg	0.69	0.69	0.69	4827
weighted avg	0.69	0.69	0.69	4827

```
[45]: model_new = Sequential()

model_new.add(
    Dense(94, activation='relu')
)

model_new.add(Dropout(0.2))

model_new.add(
    Dense(30, activation='relu')
)

model_new.add(Dropout(0.2))

model_new.add(
    Dense(15, activation='relu')
)

model_new.add(Dropout(0.2))

model_new.add(
    Dense(1, activation='sigmoid')
)

model_new.compile(
    optimizer='adam',
    loss='binary_crossentropy',
    metrics=['binary_accuracy']
)

model_new.fit(
    X_train,
    y_train,
    epochs=200,
    batch_size=256,
    validation_data=(X_test, y_test),
    callbacks=[early_stop]
)
```

Epoch 1/200

44/44 [=====] - 1s 4ms/step - loss: 0.6805 -  
binary\_accuracy: 0.5616 - val\_loss: 0.6648 - val\_binary\_accuracy: 0.6101

Epoch 2/200

44/44 [=====] - 0s 2ms/step - loss: 0.6659 -  
binary\_accuracy: 0.6033 - val\_loss: 0.6563 - val\_binary\_accuracy: 0.6143

Epoch 3/200



44/44 [=====] - 0s 2ms/step - loss: 0.6584 -  
binary\_accuracy: 0.6130 - val\_loss: 0.6523 - val\_binary\_accuracy: 0.6138  
Epoch 4/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6558 -  
binary\_accuracy: 0.6174 - val\_loss: 0.6503 - val\_binary\_accuracy: 0.6101  
Epoch 5/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6509 -  
binary\_accuracy: 0.6165 - val\_loss: 0.6485 - val\_binary\_accuracy: 0.6130  
Epoch 6/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6498 -  
binary\_accuracy: 0.6196 - val\_loss: 0.6472 - val\_binary\_accuracy: 0.6109  
Epoch 7/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6481 -  
binary\_accuracy: 0.6213 - val\_loss: 0.6457 - val\_binary\_accuracy: 0.6165  
Epoch 8/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6449 -  
binary\_accuracy: 0.6246 - val\_loss: 0.6441 - val\_binary\_accuracy: 0.6174  
Epoch 9/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6473 -  
binary\_accuracy: 0.6243 - val\_loss: 0.6441 - val\_binary\_accuracy: 0.6172  
Epoch 10/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6449 -  
binary\_accuracy: 0.6249 - val\_loss: 0.6430 - val\_binary\_accuracy: 0.6194  
Epoch 11/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6438 -  
binary\_accuracy: 0.6288 - val\_loss: 0.6426 - val\_binary\_accuracy: 0.6196  
Epoch 12/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6431 -  
binary\_accuracy: 0.6271 - val\_loss: 0.6414 - val\_binary\_accuracy: 0.6176  
Epoch 13/200  
44/44 [=====] - 0s 3ms/step - loss: 0.6403 -  
binary\_accuracy: 0.6292 - val\_loss: 0.6407 - val\_binary\_accuracy: 0.6205  
Epoch 14/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6400 -  
binary\_accuracy: 0.6299 - val\_loss: 0.6406 - val\_binary\_accuracy: 0.6221  
Epoch 15/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6391 -  
binary\_accuracy: 0.6330 - val\_loss: 0.6399 - val\_binary\_accuracy: 0.6207  
Epoch 16/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6389 -  
binary\_accuracy: 0.6329 - val\_loss: 0.6389 - val\_binary\_accuracy: 0.6211  
Epoch 17/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6381 -  
binary\_accuracy: 0.6340 - val\_loss: 0.6387 - val\_binary\_accuracy: 0.6242  
Epoch 18/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6374 -  
binary\_accuracy: 0.6395 - val\_loss: 0.6387 - val\_binary\_accuracy: 0.6219  
Epoch 19/200

44/44 [=====] - 0s 2ms/step - loss: 0.6386 -  
binary\_accuracy: 0.6313 - val\_loss: 0.6385 - val\_binary\_accuracy: 0.6217  
Epoch 20/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6336 -  
binary\_accuracy: 0.6362 - val\_loss: 0.6375 - val\_binary\_accuracy: 0.6269  
Epoch 21/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6352 -  
binary\_accuracy: 0.6359 - val\_loss: 0.6380 - val\_binary\_accuracy: 0.6279  
Epoch 22/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6355 -  
binary\_accuracy: 0.6353 - val\_loss: 0.6370 - val\_binary\_accuracy: 0.6341  
Epoch 23/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6341 -  
binary\_accuracy: 0.6332 - val\_loss: 0.6368 - val\_binary\_accuracy: 0.6256  
Epoch 24/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6326 -  
binary\_accuracy: 0.6385 - val\_loss: 0.6356 - val\_binary\_accuracy: 0.6331  
Epoch 25/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6330 -  
binary\_accuracy: 0.6352 - val\_loss: 0.6348 - val\_binary\_accuracy: 0.6321  
Epoch 26/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6315 -  
binary\_accuracy: 0.6400 - val\_loss: 0.6349 - val\_binary\_accuracy: 0.6298  
Epoch 27/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6294 -  
binary\_accuracy: 0.6375 - val\_loss: 0.6347 - val\_binary\_accuracy: 0.6335  
Epoch 28/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6317 -  
binary\_accuracy: 0.6385 - val\_loss: 0.6341 - val\_binary\_accuracy: 0.6269  
Epoch 29/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6307 -  
binary\_accuracy: 0.6385 - val\_loss: 0.6339 - val\_binary\_accuracy: 0.6271  
Epoch 30/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6286 -  
binary\_accuracy: 0.6441 - val\_loss: 0.6349 - val\_binary\_accuracy: 0.6321  
Epoch 31/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6299 -  
binary\_accuracy: 0.6398 - val\_loss: 0.6331 - val\_binary\_accuracy: 0.6306  
Epoch 32/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6289 -  
binary\_accuracy: 0.6388 - val\_loss: 0.6337 - val\_binary\_accuracy: 0.6329  
Epoch 33/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6265 -  
binary\_accuracy: 0.6484 - val\_loss: 0.6337 - val\_binary\_accuracy: 0.6343  
Epoch 34/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6280 -  
binary\_accuracy: 0.6432 - val\_loss: 0.6323 - val\_binary\_accuracy: 0.6333  
Epoch 35/200

44/44 [=====] - 0s 2ms/step - loss: 0.6241 -  
binary\_accuracy: 0.6465 - val\_loss: 0.6320 - val\_binary\_accuracy: 0.6354  
Epoch 36/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6245 -  
binary\_accuracy: 0.6481 - val\_loss: 0.6312 - val\_binary\_accuracy: 0.6387  
Epoch 37/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6233 -  
binary\_accuracy: 0.6491 - val\_loss: 0.6316 - val\_binary\_accuracy: 0.6414  
Epoch 38/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6239 -  
binary\_accuracy: 0.6433 - val\_loss: 0.6307 - val\_binary\_accuracy: 0.6393  
Epoch 39/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6238 -  
binary\_accuracy: 0.6478 - val\_loss: 0.6307 - val\_binary\_accuracy: 0.6393  
Epoch 40/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6204 -  
binary\_accuracy: 0.6514 - val\_loss: 0.6311 - val\_binary\_accuracy: 0.6379  
Epoch 41/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6224 -  
binary\_accuracy: 0.6510 - val\_loss: 0.6290 - val\_binary\_accuracy: 0.6387  
Epoch 42/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6227 -  
binary\_accuracy: 0.6509 - val\_loss: 0.6294 - val\_binary\_accuracy: 0.6395  
Epoch 43/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6192 -  
binary\_accuracy: 0.6514 - val\_loss: 0.6281 - val\_binary\_accuracy: 0.6401  
Epoch 44/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6208 -  
binary\_accuracy: 0.6449 - val\_loss: 0.6289 - val\_binary\_accuracy: 0.6379  
Epoch 45/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6183 -  
binary\_accuracy: 0.6520 - val\_loss: 0.6279 - val\_binary\_accuracy: 0.6459  
Epoch 46/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6190 -  
binary\_accuracy: 0.6507 - val\_loss: 0.6284 - val\_binary\_accuracy: 0.6437  
Epoch 47/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6201 -  
binary\_accuracy: 0.6525 - val\_loss: 0.6277 - val\_binary\_accuracy: 0.6428  
Epoch 48/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6185 -  
binary\_accuracy: 0.6470 - val\_loss: 0.6276 - val\_binary\_accuracy: 0.6383  
Epoch 49/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6164 -  
binary\_accuracy: 0.6531 - val\_loss: 0.6270 - val\_binary\_accuracy: 0.6443  
Epoch 50/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6141 -  
binary\_accuracy: 0.6607 - val\_loss: 0.6269 - val\_binary\_accuracy: 0.6430  
Epoch 51/200

44/44 [=====] - 0s 2ms/step - loss: 0.6182 -  
binary\_accuracy: 0.6561 - val\_loss: 0.6261 - val\_binary\_accuracy: 0.6455  
Epoch 52/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6148 -  
binary\_accuracy: 0.6591 - val\_loss: 0.6252 - val\_binary\_accuracy: 0.6414  
Epoch 53/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6153 -  
binary\_accuracy: 0.6591 - val\_loss: 0.6256 - val\_binary\_accuracy: 0.6368  
Epoch 54/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6144 -  
binary\_accuracy: 0.6551 - val\_loss: 0.6249 - val\_binary\_accuracy: 0.6449  
Epoch 55/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6111 -  
binary\_accuracy: 0.6569 - val\_loss: 0.6265 - val\_binary\_accuracy: 0.6464  
Epoch 56/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6133 -  
binary\_accuracy: 0.6578 - val\_loss: 0.6240 - val\_binary\_accuracy: 0.6449  
Epoch 57/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6116 -  
binary\_accuracy: 0.6567 - val\_loss: 0.6250 - val\_binary\_accuracy: 0.6443  
Epoch 58/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6103 -  
binary\_accuracy: 0.6596 - val\_loss: 0.6230 - val\_binary\_accuracy: 0.6499  
Epoch 59/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6117 -  
binary\_accuracy: 0.6639 - val\_loss: 0.6226 - val\_binary\_accuracy: 0.6480  
Epoch 60/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6091 -  
binary\_accuracy: 0.6623 - val\_loss: 0.6233 - val\_binary\_accuracy: 0.6497  
Epoch 61/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6092 -  
binary\_accuracy: 0.6600 - val\_loss: 0.6220 - val\_binary\_accuracy: 0.6412  
Epoch 62/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6096 -  
binary\_accuracy: 0.6598 - val\_loss: 0.6220 - val\_binary\_accuracy: 0.6515  
Epoch 63/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6099 -  
binary\_accuracy: 0.6615 - val\_loss: 0.6217 - val\_binary\_accuracy: 0.6491  
Epoch 64/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6069 -  
binary\_accuracy: 0.6593 - val\_loss: 0.6229 - val\_binary\_accuracy: 0.6449  
Epoch 65/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6068 -  
binary\_accuracy: 0.6683 - val\_loss: 0.6203 - val\_binary\_accuracy: 0.6501  
Epoch 66/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6052 -  
binary\_accuracy: 0.6666 - val\_loss: 0.6223 - val\_binary\_accuracy: 0.6468  
Epoch 67/200

44/44 [=====] - 0s 2ms/step - loss: 0.6085 -  
binary\_accuracy: 0.6623 - val\_loss: 0.6194 - val\_binary\_accuracy: 0.6522  
Epoch 68/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6058 -  
binary\_accuracy: 0.6700 - val\_loss: 0.6201 - val\_binary\_accuracy: 0.6509  
Epoch 69/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6042 -  
binary\_accuracy: 0.6679 - val\_loss: 0.6203 - val\_binary\_accuracy: 0.6507  
Epoch 70/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6017 -  
binary\_accuracy: 0.6686 - val\_loss: 0.6222 - val\_binary\_accuracy: 0.6426  
Epoch 71/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6027 -  
binary\_accuracy: 0.6668 - val\_loss: 0.6195 - val\_binary\_accuracy: 0.6511  
Epoch 72/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6013 -  
binary\_accuracy: 0.6693 - val\_loss: 0.6197 - val\_binary\_accuracy: 0.6596  
Epoch 73/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6009 -  
binary\_accuracy: 0.6693 - val\_loss: 0.6183 - val\_binary\_accuracy: 0.6489  
Epoch 74/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6023 -  
binary\_accuracy: 0.6729 - val\_loss: 0.6201 - val\_binary\_accuracy: 0.6495  
Epoch 75/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6016 -  
binary\_accuracy: 0.6708 - val\_loss: 0.6174 - val\_binary\_accuracy: 0.6547  
Epoch 76/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5989 -  
binary\_accuracy: 0.6711 - val\_loss: 0.6177 - val\_binary\_accuracy: 0.6549  
Epoch 77/200  
44/44 [=====] - 0s 2ms/step - loss: 0.6008 -  
binary\_accuracy: 0.6651 - val\_loss: 0.6174 - val\_binary\_accuracy: 0.6486  
Epoch 78/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5970 -  
binary\_accuracy: 0.6709 - val\_loss: 0.6170 - val\_binary\_accuracy: 0.6549  
Epoch 79/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5986 -  
binary\_accuracy: 0.6713 - val\_loss: 0.6165 - val\_binary\_accuracy: 0.6542  
Epoch 80/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5983 -  
binary\_accuracy: 0.6753 - val\_loss: 0.6172 - val\_binary\_accuracy: 0.6538  
Epoch 81/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5983 -  
binary\_accuracy: 0.6729 - val\_loss: 0.6160 - val\_binary\_accuracy: 0.6549  
Epoch 82/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5970 -  
binary\_accuracy: 0.6709 - val\_loss: 0.6143 - val\_binary\_accuracy: 0.6596  
Epoch 83/200

44/44 [=====] - 0s 2ms/step - loss: 0.5956 -  
binary\_accuracy: 0.6760 - val\_loss: 0.6142 - val\_binary\_accuracy: 0.6538  
Epoch 84/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5964 -  
binary\_accuracy: 0.6701 - val\_loss: 0.6147 - val\_binary\_accuracy: 0.6582  
Epoch 85/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5953 -  
binary\_accuracy: 0.6766 - val\_loss: 0.6157 - val\_binary\_accuracy: 0.6499  
Epoch 86/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5947 -  
binary\_accuracy: 0.6712 - val\_loss: 0.6140 - val\_binary\_accuracy: 0.6530  
Epoch 87/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5935 -  
binary\_accuracy: 0.6726 - val\_loss: 0.6132 - val\_binary\_accuracy: 0.6629  
Epoch 88/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5919 -  
binary\_accuracy: 0.6765 - val\_loss: 0.6125 - val\_binary\_accuracy: 0.6592  
Epoch 89/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5941 -  
binary\_accuracy: 0.6735 - val\_loss: 0.6106 - val\_binary\_accuracy: 0.6596  
Epoch 90/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5929 -  
binary\_accuracy: 0.6767 - val\_loss: 0.6117 - val\_binary\_accuracy: 0.6602  
Epoch 91/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5911 -  
binary\_accuracy: 0.6763 - val\_loss: 0.6104 - val\_binary\_accuracy: 0.6588  
Epoch 92/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5945 -  
binary\_accuracy: 0.6756 - val\_loss: 0.6113 - val\_binary\_accuracy: 0.6582  
Epoch 93/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5931 -  
binary\_accuracy: 0.6781 - val\_loss: 0.6112 - val\_binary\_accuracy: 0.6600  
Epoch 94/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5907 -  
binary\_accuracy: 0.6813 - val\_loss: 0.6118 - val\_binary\_accuracy: 0.6567  
Epoch 95/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5909 -  
binary\_accuracy: 0.6796 - val\_loss: 0.6108 - val\_binary\_accuracy: 0.6532  
Epoch 96/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5888 -  
binary\_accuracy: 0.6817 - val\_loss: 0.6138 - val\_binary\_accuracy: 0.6553  
Epoch 97/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5902 -  
binary\_accuracy: 0.6787 - val\_loss: 0.6099 - val\_binary\_accuracy: 0.6621  
Epoch 98/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5877 -  
binary\_accuracy: 0.6771 - val\_loss: 0.6098 - val\_binary\_accuracy: 0.6584  
Epoch 99/200

44/44 [=====] - 0s 2ms/step - loss: 0.5863 -  
binary\_accuracy: 0.6785 - val\_loss: 0.6102 - val\_binary\_accuracy: 0.6607  
Epoch 100/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5897 -  
binary\_accuracy: 0.6777 - val\_loss: 0.6100 - val\_binary\_accuracy: 0.6644  
Epoch 101/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5891 -  
binary\_accuracy: 0.6812 - val\_loss: 0.6078 - val\_binary\_accuracy: 0.6658  
Epoch 102/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5878 -  
binary\_accuracy: 0.6813 - val\_loss: 0.6085 - val\_binary\_accuracy: 0.6611  
Epoch 103/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5889 -  
binary\_accuracy: 0.6781 - val\_loss: 0.6073 - val\_binary\_accuracy: 0.6648  
Epoch 104/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5835 -  
binary\_accuracy: 0.6852 - val\_loss: 0.6089 - val\_binary\_accuracy: 0.6615  
Epoch 105/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5829 -  
binary\_accuracy: 0.6879 - val\_loss: 0.6068 - val\_binary\_accuracy: 0.6638  
Epoch 106/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5870 -  
binary\_accuracy: 0.6845 - val\_loss: 0.6089 - val\_binary\_accuracy: 0.6623  
Epoch 107/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5840 -  
binary\_accuracy: 0.6845 - val\_loss: 0.6086 - val\_binary\_accuracy: 0.6615  
Epoch 108/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5837 -  
binary\_accuracy: 0.6814 - val\_loss: 0.6085 - val\_binary\_accuracy: 0.6644  
Epoch 109/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5834 -  
binary\_accuracy: 0.6851 - val\_loss: 0.6061 - val\_binary\_accuracy: 0.6694  
Epoch 110/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5833 -  
binary\_accuracy: 0.6847 - val\_loss: 0.6056 - val\_binary\_accuracy: 0.6646  
Epoch 111/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5815 -  
binary\_accuracy: 0.6828 - val\_loss: 0.6080 - val\_binary\_accuracy: 0.6607  
Epoch 112/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5854 -  
binary\_accuracy: 0.6838 - val\_loss: 0.6059 - val\_binary\_accuracy: 0.6673  
Epoch 113/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5829 -  
binary\_accuracy: 0.6867 - val\_loss: 0.6067 - val\_binary\_accuracy: 0.6596  
Epoch 114/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5809 -  
binary\_accuracy: 0.6871 - val\_loss: 0.6064 - val\_binary\_accuracy: 0.6658  
Epoch 115/200

44/44 [=====] - 0s 2ms/step - loss: 0.5803 -  
binary\_accuracy: 0.6841 - val\_loss: 0.6053 - val\_binary\_accuracy: 0.6658  
Epoch 116/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5804 -  
binary\_accuracy: 0.6838 - val\_loss: 0.6057 - val\_binary\_accuracy: 0.6640  
Epoch 117/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5836 -  
binary\_accuracy: 0.6877 - val\_loss: 0.6090 - val\_binary\_accuracy: 0.6571  
Epoch 118/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5755 -  
binary\_accuracy: 0.6948 - val\_loss: 0.6105 - val\_binary\_accuracy: 0.6640  
Epoch 119/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5804 -  
binary\_accuracy: 0.6842 - val\_loss: 0.6097 - val\_binary\_accuracy: 0.6607  
Epoch 120/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5793 -  
binary\_accuracy: 0.6873 - val\_loss: 0.6047 - val\_binary\_accuracy: 0.6687  
Epoch 121/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5748 -  
binary\_accuracy: 0.6896 - val\_loss: 0.6043 - val\_binary\_accuracy: 0.6638  
Epoch 122/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5768 -  
binary\_accuracy: 0.6882 - val\_loss: 0.6058 - val\_binary\_accuracy: 0.6625  
Epoch 123/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5803 -  
binary\_accuracy: 0.6821 - val\_loss: 0.6066 - val\_binary\_accuracy: 0.6640  
Epoch 124/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5776 -  
binary\_accuracy: 0.6902 - val\_loss: 0.6072 - val\_binary\_accuracy: 0.6660  
Epoch 125/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5784 -  
binary\_accuracy: 0.6868 - val\_loss: 0.6067 - val\_binary\_accuracy: 0.6598  
Epoch 126/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5754 -  
binary\_accuracy: 0.6897 - val\_loss: 0.6051 - val\_binary\_accuracy: 0.6677  
Epoch 127/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5737 -  
binary\_accuracy: 0.6973 - val\_loss: 0.6040 - val\_binary\_accuracy: 0.6708  
Epoch 128/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5724 -  
binary\_accuracy: 0.6914 - val\_loss: 0.6025 - val\_binary\_accuracy: 0.6735  
Epoch 129/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5796 -  
binary\_accuracy: 0.6863 - val\_loss: 0.6050 - val\_binary\_accuracy: 0.6689  
Epoch 130/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5743 -  
binary\_accuracy: 0.6943 - val\_loss: 0.6012 - val\_binary\_accuracy: 0.6733  
Epoch 131/200



44/44 [=====] - 0s 2ms/step - loss: 0.5762 -  
binary\_accuracy: 0.6882 - val\_loss: 0.6036 - val\_binary\_accuracy: 0.6675  
Epoch 132/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5716 -  
binary\_accuracy: 0.6964 - val\_loss: 0.6029 - val\_binary\_accuracy: 0.6654  
Epoch 133/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5784 -  
binary\_accuracy: 0.6883 - val\_loss: 0.6015 - val\_binary\_accuracy: 0.6607  
Epoch 134/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5738 -  
binary\_accuracy: 0.6964 - val\_loss: 0.6014 - val\_binary\_accuracy: 0.6631  
Epoch 135/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5743 -  
binary\_accuracy: 0.6858 - val\_loss: 0.6006 - val\_binary\_accuracy: 0.6629  
Epoch 136/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5717 -  
binary\_accuracy: 0.6936 - val\_loss: 0.6000 - val\_binary\_accuracy: 0.6696  
Epoch 137/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5697 -  
binary\_accuracy: 0.6937 - val\_loss: 0.6035 - val\_binary\_accuracy: 0.6642  
Epoch 138/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5697 -  
binary\_accuracy: 0.6947 - val\_loss: 0.6001 - val\_binary\_accuracy: 0.6596  
Epoch 139/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5701 -  
binary\_accuracy: 0.6954 - val\_loss: 0.6023 - val\_binary\_accuracy: 0.6673  
Epoch 140/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5707 -  
binary\_accuracy: 0.6960 - val\_loss: 0.6002 - val\_binary\_accuracy: 0.6677  
Epoch 141/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5687 -  
binary\_accuracy: 0.6976 - val\_loss: 0.5988 - val\_binary\_accuracy: 0.6685  
Epoch 142/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5715 -  
binary\_accuracy: 0.6977 - val\_loss: 0.5981 - val\_binary\_accuracy: 0.6710  
Epoch 143/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5719 -  
binary\_accuracy: 0.6947 - val\_loss: 0.6044 - val\_binary\_accuracy: 0.6658  
Epoch 144/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5695 -  
binary\_accuracy: 0.6940 - val\_loss: 0.6002 - val\_binary\_accuracy: 0.6679  
Epoch 145/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5692 -  
binary\_accuracy: 0.6937 - val\_loss: 0.5990 - val\_binary\_accuracy: 0.6681  
Epoch 146/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5703 -  
binary\_accuracy: 0.6979 - val\_loss: 0.6001 - val\_binary\_accuracy: 0.6677  
Epoch 147/200

44/44 [=====] - 0s 2ms/step - loss: 0.5704 -  
binary\_accuracy: 0.6927 - val\_loss: 0.6027 - val\_binary\_accuracy: 0.6687  
Epoch 148/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5649 -  
binary\_accuracy: 0.6952 - val\_loss: 0.6001 - val\_binary\_accuracy: 0.6718  
Epoch 149/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5690 -  
binary\_accuracy: 0.7021 - val\_loss: 0.5993 - val\_binary\_accuracy: 0.6710  
Epoch 150/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5658 -  
binary\_accuracy: 0.6990 - val\_loss: 0.5989 - val\_binary\_accuracy: 0.6689  
Epoch 151/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5681 -  
binary\_accuracy: 0.6965 - val\_loss: 0.5964 - val\_binary\_accuracy: 0.6646  
Epoch 152/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5630 -  
binary\_accuracy: 0.7016 - val\_loss: 0.5968 - val\_binary\_accuracy: 0.6745  
Epoch 153/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5649 -  
binary\_accuracy: 0.7024 - val\_loss: 0.5978 - val\_binary\_accuracy: 0.6692  
Epoch 154/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5622 -  
binary\_accuracy: 0.7037 - val\_loss: 0.5960 - val\_binary\_accuracy: 0.6727  
Epoch 155/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5641 -  
binary\_accuracy: 0.6967 - val\_loss: 0.5968 - val\_binary\_accuracy: 0.6747  
Epoch 156/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5622 -  
binary\_accuracy: 0.7060 - val\_loss: 0.5995 - val\_binary\_accuracy: 0.6723  
Epoch 157/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5620 -  
binary\_accuracy: 0.6983 - val\_loss: 0.5965 - val\_binary\_accuracy: 0.6706  
Epoch 158/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5612 -  
binary\_accuracy: 0.6987 - val\_loss: 0.6007 - val\_binary\_accuracy: 0.6677  
Epoch 159/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5630 -  
binary\_accuracy: 0.7014 - val\_loss: 0.5991 - val\_binary\_accuracy: 0.6687  
Epoch 160/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5650 -  
binary\_accuracy: 0.6948 - val\_loss: 0.5986 - val\_binary\_accuracy: 0.6679  
Epoch 161/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5653 -  
binary\_accuracy: 0.6965 - val\_loss: 0.5947 - val\_binary\_accuracy: 0.6700  
Epoch 162/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5635 -  
binary\_accuracy: 0.6940 - val\_loss: 0.5951 - val\_binary\_accuracy: 0.6739  
Epoch 163/200

44/44 [=====] - 0s 2ms/step - loss: 0.5631 -  
binary\_accuracy: 0.6985 - val\_loss: 0.5936 - val\_binary\_accuracy: 0.6727  
Epoch 164/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5627 -  
binary\_accuracy: 0.7020 - val\_loss: 0.5936 - val\_binary\_accuracy: 0.6710  
Epoch 165/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5642 -  
binary\_accuracy: 0.6966 - val\_loss: 0.5945 - val\_binary\_accuracy: 0.6687  
Epoch 166/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5613 -  
binary\_accuracy: 0.6978 - val\_loss: 0.5933 - val\_binary\_accuracy: 0.6692  
Epoch 167/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5627 -  
binary\_accuracy: 0.6971 - val\_loss: 0.5960 - val\_binary\_accuracy: 0.6685  
Epoch 168/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5605 -  
binary\_accuracy: 0.7021 - val\_loss: 0.5949 - val\_binary\_accuracy: 0.6669  
Epoch 169/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5582 -  
binary\_accuracy: 0.7056 - val\_loss: 0.5967 - val\_binary\_accuracy: 0.6733  
Epoch 170/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5600 -  
binary\_accuracy: 0.7059 - val\_loss: 0.5941 - val\_binary\_accuracy: 0.6692  
Epoch 171/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5595 -  
binary\_accuracy: 0.6998 - val\_loss: 0.5959 - val\_binary\_accuracy: 0.6737  
Epoch 172/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5619 -  
binary\_accuracy: 0.7005 - val\_loss: 0.5944 - val\_binary\_accuracy: 0.6752  
Epoch 173/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5607 -  
binary\_accuracy: 0.7009 - val\_loss: 0.5912 - val\_binary\_accuracy: 0.6741  
Epoch 174/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5588 -  
binary\_accuracy: 0.7023 - val\_loss: 0.5950 - val\_binary\_accuracy: 0.6754  
Epoch 175/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5591 -  
binary\_accuracy: 0.7004 - val\_loss: 0.5907 - val\_binary\_accuracy: 0.6752  
Epoch 176/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5596 -  
binary\_accuracy: 0.7040 - val\_loss: 0.5910 - val\_binary\_accuracy: 0.6756  
Epoch 177/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5572 -  
binary\_accuracy: 0.7029 - val\_loss: 0.5918 - val\_binary\_accuracy: 0.6752  
Epoch 178/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5588 -  
binary\_accuracy: 0.7068 - val\_loss: 0.5905 - val\_binary\_accuracy: 0.6725  
Epoch 179/200

44/44 [=====] - 0s 2ms/step - loss: 0.5606 -  
binary\_accuracy: 0.7013 - val\_loss: 0.5895 - val\_binary\_accuracy: 0.6743  
Epoch 180/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5550 -  
binary\_accuracy: 0.7068 - val\_loss: 0.5915 - val\_binary\_accuracy: 0.6776  
Epoch 181/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5570 -  
binary\_accuracy: 0.7037 - val\_loss: 0.5918 - val\_binary\_accuracy: 0.6716  
Epoch 182/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5584 -  
binary\_accuracy: 0.7025 - val\_loss: 0.5898 - val\_binary\_accuracy: 0.6776  
Epoch 183/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5587 -  
binary\_accuracy: 0.7013 - val\_loss: 0.5904 - val\_binary\_accuracy: 0.6764  
Epoch 184/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5571 -  
binary\_accuracy: 0.7046 - val\_loss: 0.5910 - val\_binary\_accuracy: 0.6727  
Epoch 185/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5568 -  
binary\_accuracy: 0.7044 - val\_loss: 0.5916 - val\_binary\_accuracy: 0.6745  
Epoch 186/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5572 -  
binary\_accuracy: 0.7047 - val\_loss: 0.5906 - val\_binary\_accuracy: 0.6739  
Epoch 187/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5536 -  
binary\_accuracy: 0.7057 - val\_loss: 0.5932 - val\_binary\_accuracy: 0.6729  
Epoch 188/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5543 -  
binary\_accuracy: 0.7041 - val\_loss: 0.5934 - val\_binary\_accuracy: 0.6681  
Epoch 189/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5561 -  
binary\_accuracy: 0.7022 - val\_loss: 0.5876 - val\_binary\_accuracy: 0.6764  
Epoch 190/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5552 -  
binary\_accuracy: 0.7053 - val\_loss: 0.5902 - val\_binary\_accuracy: 0.6762  
Epoch 191/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5567 -  
binary\_accuracy: 0.7105 - val\_loss: 0.5890 - val\_binary\_accuracy: 0.6808  
Epoch 192/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5567 -  
binary\_accuracy: 0.6992 - val\_loss: 0.5897 - val\_binary\_accuracy: 0.6712  
Epoch 193/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5529 -  
binary\_accuracy: 0.7100 - val\_loss: 0.5896 - val\_binary\_accuracy: 0.6756  
Epoch 194/200  
44/44 [=====] - 0s 2ms/step - loss: 0.5542 -  
binary\_accuracy: 0.7042 - val\_loss: 0.5872 - val\_binary\_accuracy: 0.6731  
Epoch 195/200

```

44/44 [=====] - 0s 2ms/step - loss: 0.5514 -
binary_accuracy: 0.7116 - val_loss: 0.5897 - val_binary_accuracy: 0.6747
Epoch 196/200
44/44 [=====] - 0s 2ms/step - loss: 0.5529 -
binary_accuracy: 0.7039 - val_loss: 0.5854 - val_binary_accuracy: 0.6741
Epoch 197/200
44/44 [=====] - 0s 2ms/step - loss: 0.5491 -
binary_accuracy: 0.7060 - val_loss: 0.5895 - val_binary_accuracy: 0.6785
Epoch 198/200
44/44 [=====] - 0s 2ms/step - loss: 0.5521 -
binary_accuracy: 0.7057 - val_loss: 0.5894 - val_binary_accuracy: 0.6785
Epoch 199/200
44/44 [=====] - 0s 2ms/step - loss: 0.5541 -
binary_accuracy: 0.7068 - val_loss: 0.5876 - val_binary_accuracy: 0.6772
Epoch 200/200
44/44 [=====] - 0s 2ms/step - loss: 0.5506 -
binary_accuracy: 0.7089 - val_loss: 0.5861 - val_binary_accuracy: 0.6799

```

[45]: <keras.src.callbacks.History at 0x7fef47d28ac0>

```

[46]: #Added dropout layouts to reduce overfitting issue
pd.DataFrame(model_new.history.history)[['loss', 'val_loss']].plot()

```

[46]: <AxesSubplot: >



```
[55]: predictions_new = (model.predict(X_test) >= 0.2).astype('int')

print(
    confusion_matrix(y_test, predictions_new),
    '\n',
    classification_report(y_test, predictions_new)
)
```

151/151 [=====] - 0s 564us/step

[[ 671 1766]

[ 69 2321]]

	precision	recall	f1-score	support
0	0.91	0.28	0.42	2437
1	0.57	0.97	0.72	2390
accuracy			0.62	4827
macro avg	0.74	0.62	0.57	4827
weighted avg	0.74	0.62	0.57	4827

```
[57]: dump(scaler, open('scaler.pkl', 'wb'))
model.save('model_lending_club.h5')
```

```
[58]: later_scaler = load(open('scaler.pkl', 'rb'))
later_model = load_model('model_lending_club.h5')
```

```
[59]: X_00T = to_pred.drop('not.fully.paid', axis=1).values
to_pred.drop('not.fully.paid', axis=1).values

print(X_00T.shape)
```

(0, 15)

```
[ ]:
```