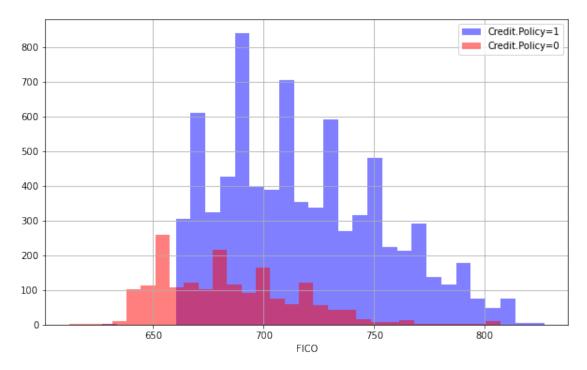
```
# import libraries
import os
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model selection import train test split
from sklearn.preprocessing import MinMaxScaler
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers 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
loan = pd.read csv('loan data.csv',encoding = "ISO-8859-1",
low memory=False)
loan.shape
(9578, 14)
loan.describe()
                          int.rate installment
                                                 log.annual.inc
       credit.policy
dti
count
         9578.000000
                      9578.000000
                                    9578.000000
                                                     9578.000000
9578.000000
mean
            0.804970
                          0.122640
                                     319.089413
                                                       10.932117
12.606679
                          0.026847
                                     207.071301
std
            0.396245
                                                        0.614813
6.883970
            0.000000
                          0.060000
                                      15.670000
                                                        7.547502
min
0.000000
                          0.103900
                                     163,770000
                                                       10.558414
25%
            1.000000
7.212500
                                     268.950000
                                                       10.928884
50%
            1.000000
                          0.122100
12.665000
75%
            1.000000
                          0.140700
                                     432.762500
                                                       11.291293
17.950000
max
            1.000000
                          0.216400
                                     940.140000
                                                       14.528354
29.960000
                    days.with.cr.line
                                           revol.bal
              fico
                                                        revol.util
       9578.000000
                           9578.000000
count
                                        9.578000e+03
                                                       9578.000000
        710.846314
                           4560.767197
                                        1.691396e+04
                                                         46.799236
mean
                                        3.375619e+04
                                                         29.014417
std
         37.970537
                           2496.930377
        612.000000
                            178.958333
                                        0.000000e+00
                                                          0.00000
min
        682.000000
                                        3.187000e+03
25%
                           2820.000000
                                                         22.600000
```

50 ⁹ 75 ⁹ max	6	707. 737. 827.	000	000	9	573	39.958 30.000 39.958	9000	1.	596000 824950 207359	9e+04	70	.300000 .900000 .000000
std 2. min 0. 25% 0. 50% 1. 75% 2.		0.0 0.0 1.0 2.3 33.0	00(57 20(00(00(00(0.5 0.6 0.6 0.6		95 3 5 9 9 9	0. 0. 0. 0. 0. 0.	ub.rec 000000 062122 262126 000000 000000 000000	9 2 5 9 9 9	9578. 0. 0. 0. 0.	y.paid 000000 160054 366676 000000 000000 000000 000000	
LO		ead(10											
lo		lit.po nual.i		y \			purp	ose	int	.rate	inst	allme	nt
0	3504			1	debt_	_consol	idat	ion	0	.1189		829.	10
1				1		cred	dit_ca	ard	0	.1071		228.	22
11 2	.0821	982143 973491 950407 999732		1	debt_	_consol	.idat:	ion	0	. 1357		366.	86
10 3	.3734			1	debt	consol	idat	ion	0	.1008		162.	34
11 4	. 3504			1	-	_	dit ca			.1426		102.	
11	. 2997						_						
5 11	.9049	068		1		cred	dit_ca	ard	0	.0788		125.	13
6 10	.7144	118			debt_	_consolidation		0.1496			194.02		
7		1			all_other		0.1114			131.22			
8	.0021			1	hor	me_impr	oveme	ent	0	. 1134		87.	19
11.4075 9 10.2035				1	debt_	_consol	idat	ion	0	.1221		84.	12
			.CO	da	ays.w:	ith.cr.	line	rev	ol.	bal	revol.	util	
ind 0	q.las 19.4	st.6mt 18 7	:hs '37	\		5639.95	8333		28	854		52.1	
0													
1 0	14.2		'07			2760.00				623		76.7	
2 1	11.6	53 6	82		4	4710.00	00000		3	511		25.6	
3 1	8.1	.0 7	12		•	2699.95	8333		33	667		73.2	
4	14.9	7 6	67		4	4066.00	0000		4	740		39.5	

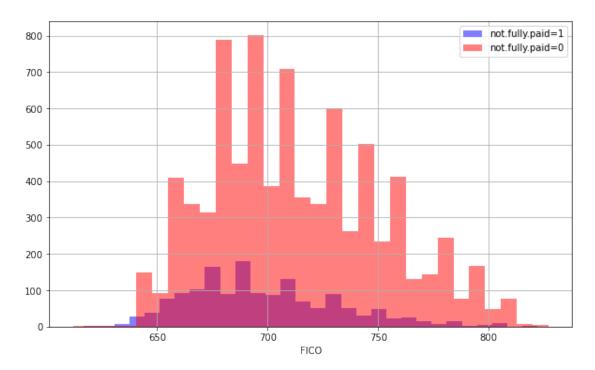
```
5
   16.98
           727
                       6120.041667
                                          50807
                                                        51.0
0
6
    4.00
           667
                       3180.041667
                                           3839
                                                        76.8
0
7
   11.08
           722
                       5116.000000
                                          24220
                                                        68.6
0
8
                                                        51.1
   17.25
           682
                       3989.000000
                                          69909
1
9
           707
                       2730.041667
                                                        23.0
   10.00
                                           5630
1
   deling.2yrs
                 pub.rec
                           not.fully.paid
0
                       0
                                         0
1
              0
2
              0
                       0
                                         0
3
              0
                       0
                                         0
4
              1
                       0
                                         0
5
              0
                       0
                                         0
6
              0
                       1
                                         1
7
              0
                       0
                                         1
8
                       0
                                         0
              0
9
              0
                       0
                                         0
loan.dtypes
credit.policy
                        int64
                       object
purpose
int.rate
                       float64
installment
                      float64
log.annual.inc
                      float64
dti
                      float64
fico
                        int64
days.with.cr.line
                      float64
revol.bal
                         int64
revol.util
                      float64
ing.last.6mths
                        int64
deling.2yrs
                         int64
pub.rec
                         int64
not.fully.paid
                         int64
dtype: object
#Transform categorical values into numerical values
obj loan = loan.select dtypes(include=['object']).copy()
obj loan.head()
               purpose
   debt_consolidation
0
1
           credit card
```

debt consolidation

```
debt_consolidation
          credit card
#obj loan[obj loan.isnull().any(axis=1)]
obj_loan["purpose"].value_counts()
debt consolidation
                      3957
all other
                      2331
credit card
                      1262
home improvement
                       629
small business
                       619
                       437
major purchase
educational
                       343
Name: purpose, dtype: int64
obj loan = obj loan.fillna({"purpose" : "credit_card"})
cleanup nums = {"purpose": {"credit card": 1, "debt consolidation":
2 }}
obj loan=obj loan.replace(cleanup nums)
obj_loan.head()
  purpose
0
        2
        1
1
2
        2
3
        2
4
        1
#clean loan = loan[:]#pd.read csv('clean loan.csv',encoding='utf-8')
#clean loan.nunique().sort values()
#EDA
plt.figure(figsize=(10,6))
loan[loan['credit.policy']==1]
['fico'].hist(alpha=0.5,color='blue',bins=30,label='Credit.Policy=1')
loan[loan['credit.policy']==0]
['fico'].hist(alpha=0.5,color='red',bins=30,label='Credit.Policy=0')
plt.legend()
plt.xlabel('FICO')
Text(0.5, 0, 'FICO')
```

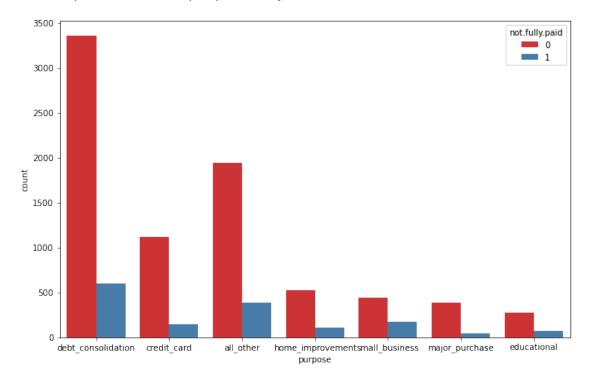


```
plt.figure(figsize=(10,6))
loan[loan['not.fully.paid']==1]
['fico'].hist(alpha=0.5,color='blue',bins=30,label='not.fully.paid=1')
loan[loan['not.fully.paid']==0]
['fico'].hist(alpha=0.5,color='red',bins=30,label='not.fully.paid=0')
plt.legend()
plt.xlabel('FICO')
Text(0.5, 0, 'FICO')
```

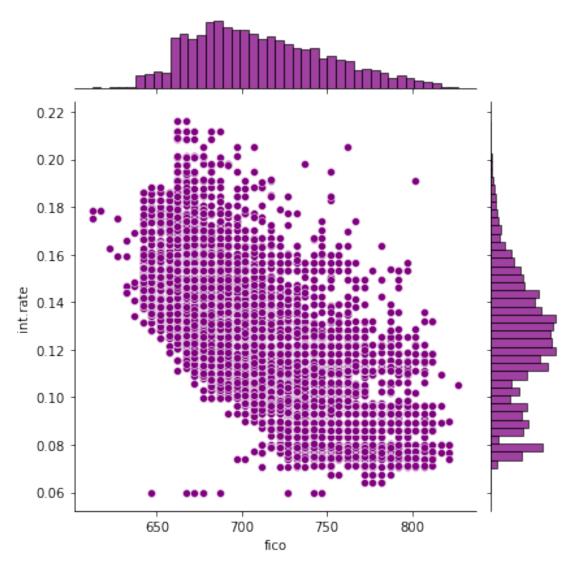


plt.figure(figsize=(11,7))
sns.countplot(x='purpose',hue='not.fully.paid',data=loan,palette='Set1
')

<AxesSubplot:xlabel='purpose', ylabel='count'>

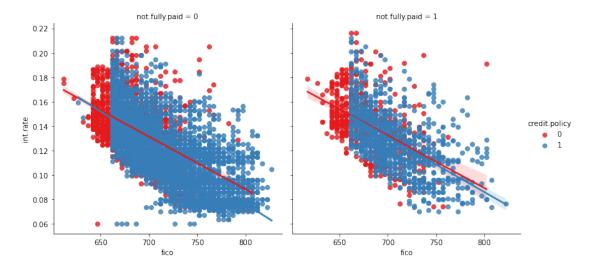


sns.jointplot(x='fico',y='int.rate',data=loan,color='purple')



<seaborn.axisgrid.FacetGrid at 0x1bb34622d60>

<Figure size 792x504 with 0 Axes>



loan_num = loan.select_dtypes(include = ['float64','int64'])
loan_num.head()

	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

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

	pub.rec	not.fully.paid
0	0	0
1	0	0
2	0	0

```
0
3
         0
         0
                         0
#loan num.hist(figsize=(16,20), bins=50, xlabelsize=8, ylabelsize=8);
#for i in range(0, len(loan num.columns),5):
    #sns.pairplot(data=loan num,
x_vars=loan_num.columns[i:i+5],y_vars=['log.annual.inc'])
#loan num corr = loan num.corr()['int.rate'][:-1] # -1 because the
latest row is SalePrice
#golden_features_list = loan_num_corr[abs(loan_num_corr) >
0.5].sort values(ascending=False)
#print("There is {} strongly correlated values with rate:\
n{}".format(len(golden features list), golden features list))
#correlation
cor matrix = loan.corr().abs()
print(cor matrix)
                   credit.policy int.rate
                                             installment
log.annual.inc \
credit.policy
                        1.000000
                                  0.294089
                                                0.058770
0.034906
int.rate
                        0.294089
                                  1.000000
                                                0.276140
0.056383
installment
                        0.058770
                                  0.276140
                                                1.000000
0.448102
log.annual.inc
                        0.034906
                                  0.056383
                                                0.448102
1.000000
                        0.090901
                                  0.220006
dti
                                                0.050202
0.054065
                                                0.086039
fico
                        0.348319
                                  0.714821
0.114576
                        0.099026
days.with.cr.line
                                  0.124022
                                                0.183297
0.336896
revol.bal
                        0.187518
                                  0.092527
                                                0.233625
0.372140
revol.util
                        0.104095
                                                0.081356
                                  0.464837
0.054881
inq.last.6mths
                        0.535511
                                  0.202780
                                                0.010419
0.029171
                        0.076318
deling.2yrs
                                  0.156079
                                                0.004368
0.029203
                        0.054243
                                  0.098162
                                                0.032760
pub.rec
0.016506
not.fully.paid
                        0.158119
                                  0.159552
                                                0.049955
0.033439
                                  fico days.with.cr.line
                                                           revol.bal \
                        dti
credit.policy
                   0.090901
                             0.348319
                                                 0.099026
                                                            0.187518
```

<pre>int.rate installment log.annual.inc dti fico days.with.cr.line revol.bal revol.util inq.last.6mths delinq.2yrs pub.rec not.fully.paid</pre>	0.050202 0.08 0.054065 0.11 1.000000 0.24 0.241191 1.06 0.060101 0.26 0.188748 0.01 0.337109 0.54 0.029189 0.18 0.021792 0.21 0.006209 0.14	14821 36039 14576 11191 30000 53880 15553 11289 35293 16340 17592	0.124022 0.183297 0.336896 0.060101 0.263880 1.000000 0.229344 0.024239 0.041736 0.081374 0.071826 0.029237	0.092527 0.233625 0.372140 0.188748 0.015553 0.229344 1.000000 0.203779 0.022394 0.033243 0.031010 0.053699
<pre>pub.rec \</pre>	revol.util ir	nq.last.6mths	delinq.2yrs	
credit.policy	0.104095	0.535511	0.076318	0.054243
int.rate	0.464837	0.202780	0.156079	0.098162
installment	0.081356	0.010419	0.004368	0.032760
log.annual.inc	0.054881	0.029171	0.029203	0.016506
dti	0.337109	0.029189	0.021792	0.006209
fico	0.541289	0.185293	0.216340	0.147592
days.with.cr.line	0.024239	0.041736	0.081374	0.071826
revol.bal	0.203779	0.022394	0.033243	0.031010
revol.util	1.000000	0.013880	0.042740	0.066717
inq.last.6mths	0.013880	1.000000	0.021245	0.072673
delinq.2yrs	0.042740	0.021245	1.000000	0.009184
pub.rec	0.066717	0.072673	0.009184	1.000000
not.fully.paid	0.082088	0.149452	0.008881	0.048634
<pre>credit.policy int.rate installment log.annual.inc dti fico</pre>	not.fully.paid 0.158119 0.159552 0.049955 0.033439 0.037362 0.149666) 2 5 9		

```
days.with.cr.line
                          0.029237
revol.bal
                          0.053699
revol.util
                          0.082088
ing.last.6mths
                          0.149452
deling.2yrs
                          0.008881
pub.rec
                          0.048634
not.fully.paid
                          1.000000
upper tri =
cor matrix.where(np.triu(np.ones(cor matrix.shape), k=1).astype(np.bool
print(upper_tri)
                    credit.policy
                                              installment
                                    int.rate
log.annual.inc
credit.policy
                              NaN
                                    0.294089
                                                   0.05877
0.034906
int.rate
                              NaN
                                         NaN
                                                   0.27614
0.056383
installment
                              NaN
                                         NaN
                                                       NaN
0.448102
log.annual.inc
                              NaN
                                         NaN
                                                       NaN
NaN
dti
                              NaN
                                         NaN
                                                       NaN
NaN
                                                       NaN
fico
                              NaN
                                         NaN
NaN
                                                       NaN
days.with.cr.line
                              NaN
                                         NaN
NaN
revol.bal
                              NaN
                                                       NaN
                                         NaN
NaN
revol.util
                              NaN
                                         NaN
                                                       NaN
NaN
inq.last.6mths
                              NaN
                                         NaN
                                                       NaN
NaN
deling.2yrs
                              NaN
                                         NaN
                                                       NaN
NaN
                                                       NaN
pub.rec
                              NaN
                                         NaN
NaN
not.fully.paid
                              NaN
                                         NaN
                                                       NaN
NaN
                                                             revol.bal
                         dti
                                   fico
                                         days.with.cr.line
credit.policy
                    0.090901
                              0.348319
                                                   0.099026
                                                              0.187518
int.rate
                    0.220006
                              0.714821
                                                   0.124022
                                                              0.092527
installment
                    0.050202
                              0.086039
                                                   0.183297
                                                              0.233625
log.annual.inc
                    0.054065
                              0.114576
                                                   0.336896
                                                              0.372140
dti
                         NaN
                              0.241191
                                                   0.060101
                                                              0.188748
fico
                         NaN
                                    NaN
                                                   0.263880
                                                              0.015553
days.with.cr.line
                         NaN
                                    NaN
                                                        NaN
                                                              0.229344
```

revol.bal revol.util inq.last.6mths delinq.2yrs pub.rec not.fully.paid	NaN NaN NaN NaN NaN NaN	NaN NaN NaN NaN NaN NaN	NaN NaN NaN NaN NaN	NaN NaN NaN NaN NaN
pub.rec \	revol.util	inq.last.6mths	delinq.2yrs	
credit.policy	0.104095	0.535511	0.076318	0.054243
int.rate	0.464837	0.202780	0.156079	0.098162
installment	0.081356	0.010419	0.004368	0.032760
log.annual.inc	0.054881	0.029171	0.029203	0.016506
dti	0.337109	0.029189	0.021792	0.006209
fico	0.541289	0.185293	0.216340	0.147592
days.with.cr.line	0.024239	0.041736	0.081374	0.071826
revol.bal	0.203779	0.022394	0.033243	0.031010
revol.util	NaN	0.013880	0.042740	0.066717
inq.last.6mths	NaN	NaN	0.021245	0.072673
delinq.2yrs	NaN	NaN	NaN	0.009184
pub.rec	NaN	NaN	NaN	NaN
not.fully.paid	NaN	NaN	NaN	NaN
<pre>credit.policy int.rate installment log.annual.inc dti fico days.with.cr.line revol.bal</pre>	not.fully.pa 0.1581 0.1595 0.0499 0.0334 0.0373 0.1496 0.0292 0.0536	19 52 55 39 62 66 37		

0.082088

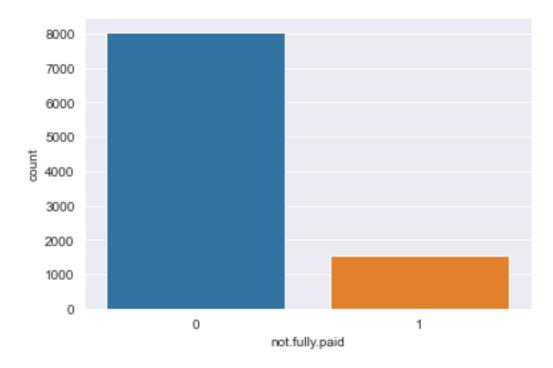
0.149452 0.008881

revol.util

inq.last.6mths delinq.2yrs

```
0.048634
pub.rec
not.fully.paid
                               NaN
C:\Users\MANISH~1\AppData\Local\Temp/ipykernel 20832/3477284062.py:1:
DeprecationWarning: `np.bool` is a deprecated alias for the builtin
`bool`. To silence this warning, use `bool` by itself. Doing this will
not modify any behavior and is safe. If you specifically wanted the
numpy scalar type, use `np.bool ` here.
Deprecated in NumPy 1.20; for more details and guidance:
https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations
  upper tri =
cor matrix.where(np.triu(np.ones(cor matrix.shape), k=1).astype(np.bool
final data.corr()
plt.figure(
        figsize=[16,12]
sns.heatmap(
        data=final data.corr(),
        cmap='viridis',
        annot=False,
        fmt='.2g'
)
                                           Traceback (most recent call
NameError
last)
C:\Users\MANISH~1\AppData\Local\Temp/ipykernel 20832/3027186847.py in
<module>
----> 1 final data.corr()
      2 plt.figure(
      3
                figsize=[16,12]
      4)
      5 sns.heatmap(
NameError: name 'final data' is not defined
loan.describe().transpose()
                    count
                                                   std
                                                                min
                                    mean
credit.policy
                   9578.0
                                0.804970
                                              0.396245
                                                          0.000000
                   9578.0
                                              0.026847
int.rate
                                0.122640
                                                          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
```

```
ing.last.6mths
                    9578.0
                                1.577469
                                               2.200245
                                                           0.000000
deling.2yrs
                    9578.0
                                0.163708
                                               0.546215
                                                           0.000000
                                               0.262126
pub.rec
                    9578.0
                                0.062122
                                                           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
                     163.770000
                                  268.950000
                                                 432.762500
installment
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
                                 8596,000000
                    3187.000000
                                               18249.500000
1.207359e+06
revol.util
                      22.600000
                                   46.300000
                                                  70.900000
1.190000e+02
ing.last.6mths
                       0.000000
                                    1.000000
                                                   2,000000
3.300000e+01
deling.2yrs
                       0.000000
                                    0.000000
                                                   0.000000
1.300000e+01
                                    0.000000
                                                   0.000000
pub.rec
                       0.000000
5.000000e+00
not.fully.paid
                       0.000000
                                    0.000000
                                                   0.000000
1.000000e+00
loan['not.fully.paid'].isnull().mean()
loan.groupby('not.fully.paid')['not.fully.paid'].count()/len(loan)
not.fully.paid
     0.839946
1
     0.160054
Name: not.fully.paid, dtype: float64
sns.set style('darkgrid')
sns.countplot(x='not.fully.paid', data=loan)
<AxesSubplot:xlabel='not.fully.paid', ylabel='count'>
```

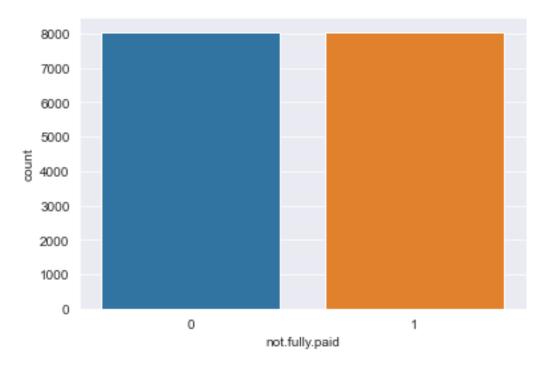


```
count_class_0, count_class_1 = loan['not.fully.paid'].value_counts()
loan_0 = loan[loan['not.fully.paid'] == 0]
loan_1 = loan[loan['not.fully.paid'] == 1]
loan_1_over = loan_1.sample(count_class_0, replace=True)
loan_test_over = pd.concat([loan_0, loan_1_over], axis=0)
print('Random over-sampling:')
print(loan_test_over['not.fully.paid'].value_counts())

sns.set_style('darkgrid')
sns.countplot(x='not.fully.paid', data=loan_test_over)

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

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



```
col_fea = ['purpose']
final_data =
pd.get_dummies(loan_test_over,columns=col_fea,drop_first=True)
final_data.info()
```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 16090 entries, 0 to 8878
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	ing.last.6mths	16090 non-null	int64
10	deling.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: 2.3 MB
to train = final data[final data['not.fully.paid'].isin([0,1])]
to pred = final data[final data['not.fully.paid'] == 2]
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)
scaler = MinMaxScaler()
X train = scaler.fit transform(X train)
X test = scaler.transform(X test)
model = Sequential()
model.add(
        Dense(19, activation='relu')
model.add(
        Dense(10, activation='relu')
model.add(
        Dense(5, activation='relu')
)
model.add(
        Dense(1, activation='sigmoid')
)
model.compile(
        optimizer='adam',
        loss='binary crossentropy',
        metrics=['accuracy']
)
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
accuracy: 0.4922 - val loss: 0.6920 - val accuracy: 0.5250
Epoch 2/200
accuracy: 0.5556 - val loss: 0.6878 - val accuracy: 0.5770
Epoch 3/200
accuracy: 0.5867 - val loss: 0.6817 - val accuracy: 0.5981
Epoch 4/200
accuracy: 0.6083 - val loss: 0.6748 - val accuracy: 0.6089
Epoch 5/200
accuracy: 0.6144 - val loss: 0.6663 - val accuracy: 0.6093
Epoch 6/200
accuracy: 0.6156 - val_loss: 0.6567 - val_accuracy: 0.6085
Epoch 7/200
accuracy: 0.6225 - val loss: 0.6521 - val accuracy: 0.6134
Epoch 8/200
accuracy: 0.6205 - val loss: 0.6471 - val accuracy: 0.6143
Epoch 9/200
accuracy: 0.6233 - val loss: 0.6464 - val accuracy: 0.6159
Epoch 10/200
accuracy: 0.6239 - val_loss: 0.6438 - val_accuracy: 0.6184
Epoch 11/200
accuracy: 0.6253 - val loss: 0.6441 - val accuracy: 0.6211
Epoch 12/200
accuracy: 0.6302 - val loss: 0.6422 - val accuracy: 0.6182
Epoch 13/200
accuracy: 0.6314 - val loss: 0.6409 - val accuracy: 0.6215
Epoch 14/200
accuracy: 0.6332 - val loss: 0.6414 - val accuracy: 0.6196
```

```
Epoch 15/200
accuracy: 0.6338 - val loss: 0.6398 - val accuracy: 0.6219
Epoch 16/200
accuracy: 0.6322 - val loss: 0.6417 - val accuracy: 0.6194
Epoch 17/200
accuracy: 0.6353 - val loss: 0.6392 - val accuracy: 0.6250
Epoch 18/200
accuracy: 0.6322 - val loss: 0.6400 - val accuracy: 0.6221
Epoch 19/200
accuracy: 0.6338 - val loss: 0.6385 - val accuracy: 0.6244
Epoch 20/200
accuracy: 0.6329 - val_loss: 0.6389 - val_accuracy: 0.6219
Epoch 21/200
accuracy: 0.6343 - val loss: 0.6377 - val accuracy: 0.6267
Epoch 22/200
accuracy: 0.6362 - val loss: 0.6376 - val accuracy: 0.6244
Epoch 23/200
accuracy: 0.6359 - val_loss: 0.6374 - val_accuracy: 0.6261
Epoch 24/200
accuracy: 0.6349 - val_loss: 0.6369 - val_accuracy: 0.6261
Epoch 25/200
accuracy: 0.6362 - val loss: 0.6395 - val accuracy: 0.6217
Epoch 26/200
accuracy: 0.6370 - val loss: 0.6355 - val accuracy: 0.6285
Epoch 27/200
accuracy: 0.6396 - val loss: 0.6368 - val accuracy: 0.6250
Epoch 28/200
accuracy: 0.6383 - val loss: 0.6349 - val accuracy: 0.6259
Epoch 29/200
accuracy: 0.6417 - val loss: 0.6345 - val accuracy: 0.6273
Epoch 30/200
accuracy: 0.6406 - val loss: 0.6347 - val accuracy: 0.6294
Epoch 31/200
```

```
accuracy: 0.6425 - val loss: 0.6350 - val accuracy: 0.6256
Epoch 32/200
accuracy: 0.6406 - val loss: 0.6340 - val accuracy: 0.6271
Epoch 33/200
accuracy: 0.6415 - val loss: 0.6331 - val accuracy: 0.6290
Epoch 34/200
accuracy: 0.6456 - val loss: 0.6329 - val accuracy: 0.6263
Epoch 35/200
accuracy: 0.6449 - val loss: 0.6332 - val accuracy: 0.6275
Epoch 36/200
accuracy: 0.6425 - val loss: 0.6322 - val accuracy: 0.6256
Epoch 37/200
accuracy: 0.6433 - val loss: 0.6318 - val accuracy: 0.6265
Epoch 38/200
accuracy: 0.6430 - val loss: 0.6328 - val accuracy: 0.6285
Epoch 39/200
accuracy: 0.6449 - val loss: 0.6308 - val accuracy: 0.6275
Epoch 40/200
accuracy: 0.6469 - val loss: 0.6340 - val accuracy: 0.6321
Epoch 41/200
accuracy: 0.6439 - val loss: 0.6310 - val accuracy: 0.6267
Epoch 42/200
accuracy: 0.6461 - val loss: 0.6323 - val accuracy: 0.6325
Epoch 43/200
accuracy: 0.6490 - val loss: 0.6323 - val accuracy: 0.6343
Epoch 44/200
accuracy: 0.6485 - val loss: 0.6344 - val accuracy: 0.6321
Epoch 45/200
accuracy: 0.6478 - val loss: 0.6307 - val accuracy: 0.6337
Epoch 46/200
accuracy: 0.6488 - val_loss: 0.6299 - val_accuracy: 0.6288
Epoch 47/200
accuracy: 0.6504 - val loss: 0.6336 - val accuracy: 0.6306
Epoch 48/200
```

```
accuracy: 0.6496 - val loss: 0.6304 - val accuracy: 0.6304
Epoch 49/200
accuracy: 0.6494 - val loss: 0.6304 - val accuracy: 0.6335
Epoch 50/200
accuracy: 0.6495 - val loss: 0.6296 - val accuracy: 0.6333
Epoch 51/200
accuracy: 0.6520 - val loss: 0.6295 - val accuracy: 0.6364
Epoch 52/200
accuracy: 0.6530 - val loss: 0.6295 - val accuracy: 0.6348
Epoch 53/200
accuracy: 0.6541 - val loss: 0.6297 - val accuracy: 0.6352
Epoch 54/200
accuracy: 0.6498 - val loss: 0.6289 - val accuracy: 0.6354
Epoch 55/200
accuracy: 0.6520 - val loss: 0.6291 - val accuracy: 0.6370
Epoch 56/200
accuracy: 0.6537 - val loss: 0.6286 - val accuracy: 0.6356
Epoch 57/200
accuracy: 0.6536 - val loss: 0.6299 - val accuracy: 0.6370
Epoch 58/200
accuracy: 0.6542 - val loss: 0.6298 - val accuracy: 0.6321
Epoch 59/200
accuracy: 0.6568 - val loss: 0.6289 - val accuracy: 0.6379
Epoch 60/200
accuracy: 0.6546 - val loss: 0.6282 - val accuracy: 0.6424
Epoch 61/200
accuracy: 0.6587 - val loss: 0.6276 - val accuracy: 0.6358
Epoch 62/200
accuracy: 0.6585 - val loss: 0.6290 - val accuracy: 0.6370
Epoch 63/200
44/44 [=============] - Os 2ms/step - loss: 0.6057 -
accuracy: 0.6585 - val loss: 0.6282 - val accuracy: 0.6381
Epoch 64/200
```

```
accuracy: 0.6576 - val loss: 0.6282 - val accuracy: 0.6383
Epoch 65/200
accuracy: 0.6579 - val loss: 0.6277 - val accuracy: 0.6358
Epoch 66/200
accuracy: 0.6591 - val loss: 0.6270 - val accuracy: 0.6406
Epoch 67/200
accuracy: 0.6577 - val loss: 0.6268 - val accuracy: 0.6428
Epoch 68/200
accuracy: 0.6593 - val loss: 0.6271 - val accuracy: 0.6428
Epoch 69/200
accuracy: 0.6596 - val loss: 0.6267 - val accuracy: 0.6397
Epoch 70/200
accuracy: 0.6607 - val loss: 0.6281 - val accuracy: 0.6422
Epoch 71/200
accuracy: 0.6587 - val loss: 0.6262 - val accuracy: 0.6424
Epoch 72/200
accuracy: 0.6607 - val loss: 0.6274 - val accuracy: 0.6418
Epoch 73/200
accuracy: 0.6620 - val loss: 0.6275 - val accuracy: 0.6410
Epoch 74/200
accuracy: 0.6591 - val loss: 0.6294 - val accuracy: 0.6383
Epoch 75/200
accuracy: 0.6611 - val loss: 0.6267 - val accuracy: 0.6422
Epoch 76/200
accuracy: 0.6589 - val loss: 0.6268 - val accuracy: 0.6408
Epoch 77/200
accuracy: 0.6616 - val loss: 0.6286 - val accuracy: 0.6364
Epoch 78/200
accuracy: 0.6590 - val loss: 0.6271 - val accuracy: 0.6397
Epoch 79/200
accuracy: 0.6618 - val_loss: 0.6276 - val_accuracy: 0.6366
Epoch 80/200
accuracy: 0.6627 - val loss: 0.6267 - val accuracy: 0.6404
Epoch 81/200
```

```
accuracy: 0.6632 - val loss: 0.6267 - val accuracy: 0.6426
Epoch 82/200
accuracy: 0.6625 - val loss: 0.6319 - val accuracy: 0.6370
Epoch 83/200
accuracy: 0.6606 - val loss: 0.6294 - val accuracy: 0.6399
Epoch 84/200
accuracy: 0.6616 - val loss: 0.6310 - val accuracy: 0.6385
Epoch 85/200
accuracy: 0.6614 - val loss: 0.6257 - val accuracy: 0.6437
Epoch 86/200
accuracy: 0.6620 - val loss: 0.6268 - val accuracy: 0.6362
Epoch 87/200
accuracy: 0.6654 - val loss: 0.6271 - val accuracy: 0.6430
Epoch 88/200
accuracy: 0.6631 - val_loss: 0.6277 - val_accuracy: 0.6395
Epoch 89/200
accuracy: 0.6654 - val loss: 0.6258 - val accuracy: 0.6457
Epoch 90/200
accuracy: 0.6623 - val loss: 0.6284 - val accuracy: 0.6358
Epoch 91/200
accuracy: 0.6623 - val loss: 0.6265 - val accuracy: 0.6393
Epoch 92/200
accuracy: 0.6651 - val loss: 0.6297 - val accuracy: 0.6375
Epoch 93/200
accuracy: 0.6631 - val loss: 0.6261 - val accuracy: 0.6451
Epoch 94/200
accuracy: 0.6632 - val loss: 0.6261 - val accuracy: 0.6420
Epoch 95/200
accuracy: 0.6639 - val loss: 0.6266 - val_accuracy: 0.6389
Epoch 96/200
accuracy: 0.6645 - val loss: 0.6246 - val accuracy: 0.6443
Epoch 97/200
accuracy: 0.6665 - val loss: 0.6255 - val accuracy: 0.6428
```

```
Epoch 98/200
accuracy: 0.6639 - val loss: 0.6294 - val accuracy: 0.6370
Epoch 99/200
accuracy: 0.6630 - val loss: 0.6254 - val accuracy: 0.6433
Epoch 100/200
accuracy: 0.6694 - val loss: 0.6260 - val accuracy: 0.6445
Epoch 101/200
accuracy: 0.6624 - val loss: 0.6270 - val accuracy: 0.6422
Epoch 102/200
accuracy: 0.6651 - val loss: 0.6262 - val accuracy: 0.6441
Epoch 103/200
accuracy: 0.6657 - val_loss: 0.6251 - val_accuracy: 0.6470
Epoch 104/200
accuracy: 0.6657 - val loss: 0.6247 - val accuracy: 0.6466
Epoch 105/200
accuracy: 0.6675 - val loss: 0.6256 - val accuracy: 0.6401
Epoch 106/200
accuracy: 0.6668 - val_loss: 0.6249 - val_accuracy: 0.6480
Epoch 107/200
accuracy: 0.6669 - val_loss: 0.6258 - val_accuracy: 0.6451
Epoch 108/200
accuracy: 0.6670 - val loss: 0.6251 - val accuracy: 0.6462
Epoch 109/200
accuracy: 0.6702 - val loss: 0.6243 - val accuracy: 0.6466
Epoch 110/200
44/44 [============== ] - Os 2ms/step - loss: 0.5971 -
accuracy: 0.6674 - val loss: 0.6266 - val accuracy: 0.6404
Epoch 111/200
accuracy: 0.6654 - val loss: 0.6276 - val accuracy: 0.6404
Epoch 112/200
accuracy: 0.6679 - val_loss: 0.6242 - val_accuracy: 0.6472
Epoch 113/200
accuracy: 0.6669 - val loss: 0.6255 - val accuracy: 0.6439
Epoch 114/200
```

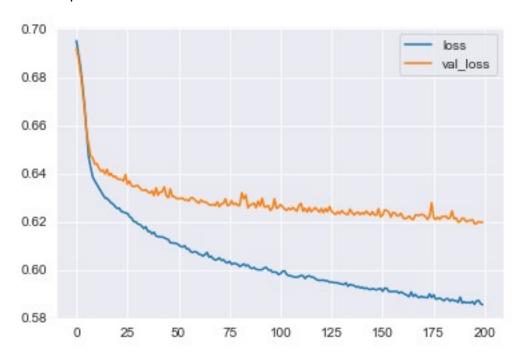
```
accuracy: 0.6675 - val loss: 0.6240 - val accuracy: 0.6495
Epoch 115/200
accuracy: 0.6655 - val loss: 0.6258 - val accuracy: 0.6433
Epoch 116/200
accuracy: 0.6692 - val loss: 0.6241 - val accuracy: 0.6466
Epoch 117/200
accuracy: 0.6663 - val loss: 0.6249 - val accuracy: 0.6468
Epoch 118/200
accuracy: 0.6710 - val loss: 0.6258 - val accuracy: 0.6410
Epoch 119/200
accuracy: 0.6696 - val loss: 0.6248 - val accuracy: 0.6459
Epoch 120/200
accuracy: 0.6714 - val loss: 0.6240 - val accuracy: 0.6464
Epoch 121/200
accuracy: 0.6682 - val loss: 0.6254 - val accuracy: 0.6447
Epoch 122/200
accuracy: 0.6734 - val loss: 0.6237 - val_accuracy: 0.6464
Epoch 123/200
accuracy: 0.6702 - val_loss: 0.6254 - val_accuracy: 0.6482
Epoch 124/200
accuracy: 0.6722 - val loss: 0.6241 - val accuracy: 0.6433
Epoch 125/200
accuracy: 0.6716 - val loss: 0.6262 - val accuracy: 0.6389
Epoch 126/200
accuracy: 0.6702 - val_loss: 0.6242 - val_accuracy: 0.6445
Epoch 127/200
accuracy: 0.6713 - val loss: 0.6235 - val accuracy: 0.6428
Epoch 128/200
accuracy: 0.6710 - val loss: 0.6230 - val accuracy: 0.6484
Epoch 129/200
accuracy: 0.6702 - val loss: 0.6240 - val accuracy: 0.6439
Epoch 130/200
accuracy: 0.6714 - val loss: 0.6232 - val accuracy: 0.6435
```

```
Epoch 131/200
accuracy: 0.6702 - val loss: 0.6245 - val accuracy: 0.6422
Epoch 132/200
accuracy: 0.6726 - val loss: 0.6235 - val accuracy: 0.6455
Epoch 133/200
accuracy: 0.6680 - val loss: 0.6231 - val accuracy: 0.6480
Epoch 134/200
accuracy: 0.6708 - val loss: 0.6228 - val accuracy: 0.6478
Epoch 135/200
accuracy: 0.6704 - val loss: 0.6251 - val accuracy: 0.6437
Epoch 136/200
accuracy: 0.6697 - val_loss: 0.6238 - val_accuracy: 0.6462
Epoch 137/200
accuracy: 0.6717 - val loss: 0.6227 - val accuracy: 0.6538
Epoch 138/200
accuracy: 0.6734 - val loss: 0.6235 - val accuracy: 0.6493
Epoch 139/200
accuracy: 0.6719 - val_loss: 0.6239 - val_accuracy: 0.6433
Epoch 140/200
accuracy: 0.6754 - val_loss: 0.6228 - val_accuracy: 0.6476
Epoch 141/200
accuracy: 0.6720 - val loss: 0.6237 - val accuracy: 0.6468
Epoch 142/200
accuracy: 0.6726 - val loss: 0.6231 - val accuracy: 0.6464
Epoch 143/200
accuracy: 0.6725 - val loss: 0.6243 - val accuracy: 0.6466
Epoch 144/200
accuracy: 0.6730 - val loss: 0.6233 - val accuracy: 0.6484
Epoch 145/200
accuracy: 0.6739 - val_loss: 0.6227 - val_accuracy: 0.6478
Epoch 146/200
accuracy: 0.6722 - val loss: 0.6241 - val accuracy: 0.6466
Epoch 147/200
```

```
accuracy: 0.6764 - val loss: 0.6227 - val accuracy: 0.6414
Epoch 148/200
accuracy: 0.6728 - val loss: 0.6227 - val accuracy: 0.6491
Epoch 149/200
accuracy: 0.6718 - val loss: 0.6253 - val accuracy: 0.6437
Epoch 150/200
accuracy: 0.6725 - val loss: 0.6219 - val accuracy: 0.6472
Epoch 151/200
accuracy: 0.6721 - val loss: 0.6227 - val accuracy: 0.6462
Epoch 152/200
accuracy: 0.6747 - val loss: 0.6248 - val accuracy: 0.6435
Epoch 153/200
accuracy: 0.6715 - val loss: 0.6246 - val accuracy: 0.6445
Epoch 154/200
accuracy: 0.6714 - val loss: 0.6239 - val accuracy: 0.6472
Epoch 155/200
accuracy: 0.6739 - val loss: 0.6218 - val accuracy: 0.6476
Epoch 156/200
accuracy: 0.6729 - val loss: 0.6237 - val accuracy: 0.6464
Epoch 157/200
accuracy: 0.6710 - val loss: 0.6235 - val accuracy: 0.6420
Epoch 158/200
accuracy: 0.6713 - val loss: 0.6225 - val accuracy: 0.6466
Epoch 159/200
accuracy: 0.6725 - val loss: 0.6224 - val accuracy: 0.6472
Epoch 160/200
accuracy: 0.6746 - val loss: 0.6235 - val accuracy: 0.6449
Epoch 161/200
accuracy: 0.6715 - val loss: 0.6216 - val accuracy: 0.6524
Epoch 162/200
accuracy: 0.6740 - val_loss: 0.6211 - val_accuracy: 0.6507
Epoch 163/200
accuracy: 0.6761 - val loss: 0.6215 - val accuracy: 0.6472
Epoch 164/200
```

```
accuracy: 0.6721 - val loss: 0.6222 - val accuracy: 0.6443
Epoch 165/200
accuracy: 0.6699 - val loss: 0.6211 - val accuracy: 0.6466
Epoch 166/200
accuracy: 0.6747 - val loss: 0.6208 - val accuracy: 0.6495
Epoch 167/200
accuracy: 0.6700 - val loss: 0.6227 - val accuracy: 0.6509
Epoch 168/200
accuracy: 0.6765 - val loss: 0.6227 - val accuracy: 0.6497
Epoch 169/200
accuracy: 0.6759 - val loss: 0.6225 - val accuracy: 0.6445
Epoch 170/200
accuracy: 0.6743 - val loss: 0.6230 - val accuracy: 0.6484
Epoch 171/200
accuracy: 0.6736 - val loss: 0.6227 - val accuracy: 0.6459
Epoch 172/200
accuracy: 0.6729 - val loss: 0.6220 - val accuracy: 0.6491
Epoch 173/200
accuracy: 0.6756 - val loss: 0.6209 - val accuracy: 0.6505
Epoch 174/200
accuracy: 0.6721 - val loss: 0.6222 - val accuracy: 0.6453
Epoch 175/200
accuracy: 0.6734 - val loss: 0.6278 - val accuracy: 0.6441
Epoch 176/200
accuracy: 0.6719 - val loss: 0.6215 - val accuracy: 0.6489
Epoch 177/200
accuracy: 0.6742 - val loss: 0.6209 - val accuracy: 0.6499
Epoch 178/200
accuracy: 0.6756 - val loss: 0.6218 - val accuracy: 0.6468
Epoch 179/200
44/44 [=============] - Os 2ms/step - loss: 0.5882 -
accuracy: 0.6724 - val loss: 0.6209 - val accuracy: 0.6476
Epoch 180/200
```

```
accuracy: 0.6718 - val loss: 0.6221 - val accuracy: 0.6507
Epoch 181/200
accuracy: 0.6768 - val loss: 0.6220 - val accuracy: 0.6503
Epoch 182/200
accuracy: 0.6732 - val loss: 0.6223 - val accuracy: 0.6414
Epoch 183/200
accuracy: 0.6747 - val loss: 0.6211 - val accuracy: 0.6449
Epoch 184/200
accuracy: 0.6769 - val loss: 0.6215 - val accuracy: 0.6499
Epoch 185/200
accuracy: 0.6735 - val loss: 0.6244 - val accuracy: 0.6368
Epoch 186/200
accuracy: 0.6767 - val loss: 0.6211 - val accuracy: 0.6441
Epoch 187/200
accuracy: 0.6752 - val loss: 0.6214 - val accuracy: 0.6468
Epoch 188/200
accuracy: 0.6744 - val loss: 0.6198 - val accuracy: 0.6491
Epoch 189/200
accuracy: 0.6714 - val loss: 0.6200 - val accuracy: 0.6453
Epoch 190/200
accuracy: 0.6759 - val loss: 0.6214 - val accuracy: 0.6474
Epoch 191/200
accuracy: 0.6754 - val loss: 0.6211 - val accuracy: 0.6474
Epoch 192/200
accuracy: 0.6763 - val loss: 0.6199 - val accuracy: 0.6474
Epoch 193/200
accuracy: 0.6743 - val loss: 0.6203 - val accuracy: 0.6501
Epoch 194/200
accuracy: 0.6766 - val loss: 0.6205 - val accuracy: 0.6459
Epoch 195/200
accuracy: 0.6749 - val_loss: 0.6208 - val_accuracy: 0.6439
Epoch 196/200
accuracy: 0.6784 - val loss: 0.6191 - val accuracy: 0.6484
Epoch 197/200
```



```
model_new.add(Dropout(0.2))
model new.add(
     Dense(5, 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
binary accuracy: 0.5133 - val loss: 0.6907 - val binary accuracy:
0.5473
Epoch 2/200
binary accuracy: 0.5391 - val loss: 0.6839 - val binary accuracy:
0.5925
Epoch 3/200
binary accuracy: 0.5639 - val loss: 0.6770 - val binary accuracy:
0.6056
Epoch 4/200
binary accuracy: 0.5889 - val loss: 0.6711 - val binary accuracy:
0.6051
Epoch 5/200
binary_accuracy: 0.5969 - val_loss: 0.6686 - val_binary_accuracy:
0.6062
Epoch 6/200
```

```
binary accuracy: 0.6030 - val loss: 0.6669 - val binary accuracy:
0.6076
Epoch 7/200
binary_accuracy: 0.5979 - val_loss: 0.6659 - val_binary_accuracy:
0.6122
Epoch 8/200
binary accuracy: 0.6013 - val loss: 0.6648 - val binary accuracy:
0.6101
Epoch 9/200
binary accuracy: 0.6076 - val loss: 0.6620 - val binary accuracy:
0.6053
Epoch 10/200
binary_accuracy: 0.6037 - val_loss: 0.6590 - val_binary_accuracy:
0.6145
Epoch 11/200
binary accuracy: 0.6115 - val loss: 0.6565 - val binary accuracy:
0.6157
Epoch 12/200
binary accuracy: 0.6076 - val loss: 0.6550 - val binary accuracy:
0.6159
Epoch 13/200
binary accuracy: 0.6129 - val loss: 0.6535 - val binary accuracy:
0.6138
Epoch 14/200
binary accuracy: 0.6093 - val loss: 0.6523 - val binary accuracy:
0.6153
Epoch 15/200
binary accuracy: 0.6175 - val loss: 0.6504 - val binary accuracy:
0.6201
Epoch 16/200
binary accuracy: 0.6165 - val loss: 0.6494 - val binary accuracy:
0.6244
Epoch 17/200
binary_accuracy: 0.6155 - val_loss: 0.6486 - val_binary_accuracy:
0.6281
Epoch 18/200
binary accuracy: 0.6164 - val loss: 0.6478 - val binary accuracy:
```

```
0.6265
Epoch 19/200
binary accuracy: 0.6164 - val loss: 0.6471 - val binary accuracy:
0.6236
Epoch 20/200
binary accuracy: 0.6135 - val loss: 0.6473 - val binary accuracy:
0.6246
Epoch 21/200
binary_accuracy: 0.6165 - val_loss: 0.6466 - val_binary_accuracy:
0.6312
Epoch 22/200
binary accuracy: 0.6190 - val loss: 0.6456 - val binary accuracy:
0.6283
Epoch 23/200
binary accuracy: 0.6252 - val loss: 0.6452 - val binary accuracy:
0.6327
Epoch 24/200
binary accuracy: 0.6224 - val loss: 0.6438 - val binary accuracy:
0.6314
Epoch 25/200
binary accuracy: 0.6170 - val loss: 0.6434 - val binary accuracy:
0.6306
Epoch 26/200
binary accuracy: 0.6215 - val loss: 0.6435 - val binary accuracy:
0.6317
Epoch 27/200
binary accuracy: 0.6201 - val loss: 0.6423 - val binary accuracy:
0.6329
Epoch 28/200
binary accuracy: 0.6267 - val loss: 0.6424 - val binary accuracy:
0.6267
Epoch 29/200
binary accuracy: 0.6219 - val loss: 0.6418 - val binary accuracy:
0.6323
Epoch 30/200
binary accuracy: 0.6208 - val loss: 0.6410 - val binary accuracy:
0.6343
Epoch 31/200
```

```
binary accuracy: 0.6196 - val loss: 0.6407 - val binary accuracy:
0.6352
Epoch 32/200
binary_accuracy: 0.6256 - val_loss: 0.6398 - val_binary_accuracy:
0.6370
Epoch 33/200
binary accuracy: 0.6257 - val loss: 0.6394 - val binary accuracy:
0.6350
Epoch 34/200
binary accuracy: 0.6294 - val loss: 0.6399 - val binary accuracy:
0.6352
Epoch 35/200
binary_accuracy: 0.6216 - val_loss: 0.6387 - val_binary_accuracy:
0.6346
Epoch 36/200
binary accuracy: 0.6291 - val loss: 0.6384 - val binary accuracy:
0.6354
Epoch 37/200
binary accuracy: 0.6275 - val loss: 0.6382 - val_binary_accuracy:
0.6348
Epoch 38/200
binary accuracy: 0.6297 - val loss: 0.6372 - val binary accuracy:
0.6331
Epoch 39/200
binary accuracy: 0.6354 - val loss: 0.6377 - val binary accuracy:
0.6317
Epoch 40/200
binary accuracy: 0.6278 - val loss: 0.6367 - val binary accuracy:
0.6312
Epoch 41/200
binary accuracy: 0.6293 - val loss: 0.6359 - val binary accuracy:
0.6348
Epoch 42/200
binary_accuracy: 0.6279 - val_loss: 0.6361 - val_binary_accuracy:
0.6327
Epoch 43/200
binary accuracy: 0.6303 - val loss: 0.6366 - val binary accuracy:
```

```
0.6294
Epoch 44/200
binary accuracy: 0.6291 - val loss: 0.6349 - val binary accuracy:
0.6304
Epoch 45/200
binary accuracy: 0.6308 - val loss: 0.6359 - val binary accuracy:
0.6329
Epoch 46/200
binary_accuracy: 0.6363 - val_loss: 0.6349 - val_binary_accuracy:
0.6341
Epoch 47/200
binary accuracy: 0.6329 - val loss: 0.6350 - val binary accuracy:
0.6325
Epoch 48/200
binary accuracy: 0.6351 - val_loss: 0.6344 - val_binary_accuracy:
0.6323
Epoch 49/200
binary accuracy: 0.6297 - val loss: 0.6340 - val binary accuracy:
0.6329
Epoch 50/200
binary accuracy: 0.6278 - val loss: 0.6338 - val binary accuracy:
0.6366
Epoch 51/200
binary accuracy: 0.6297 - val loss: 0.6350 - val binary accuracy:
0.6348
Epoch 52/200
binary accuracy: 0.6344 - val loss: 0.6338 - val binary accuracy:
0.6304
Epoch 53/200
binary accuracy: 0.6346 - val loss: 0.6332 - val binary accuracy:
0.6360
Epoch 54/200
binary accuracy: 0.6322 - val loss: 0.6333 - val binary accuracy:
0.6300
Epoch 55/200
binary accuracy: 0.6394 - val loss: 0.6334 - val binary accuracy:
0.6389
```

```
Epoch 56/200
binary_accuracy: 0.6350 - val_loss: 0.6332 - val_binary_accuracy:
0.6372
Epoch 57/200
binary accuracy: 0.6394 - val loss: 0.6324 - val binary accuracy:
0.6343
Epoch 58/200
binary accuracy: 0.6361 - val loss: 0.6320 - val binary accuracy:
0.6304
Epoch 59/200
binary accuracy: 0.6356 - val loss: 0.6324 - val binary accuracy:
0.6362
Epoch 60/200
binary accuracy: 0.6364 - val loss: 0.6323 - val binary accuracy:
0.6348
Epoch 61/200
binary accuracy: 0.6382 - val loss: 0.6325 - val binary accuracy:
0.6339
Epoch 62/200
binary_accuracy: 0.6374 - val_loss: 0.6327 - val_binary_accuracy:
0.6304
Epoch 63/200
binary accuracy: 0.6356 - val loss: 0.6315 - val binary accuracy:
0.6412
Epoch 64/200
binary accuracy: 0.6421 - val loss: 0.6313 - val binary accuracy:
0.6422
Epoch 65/200
binary accuracy: 0.6402 - val loss: 0.6312 - val binary accuracy:
0.6375
Epoch 66/200
binary accuracy: 0.6388 - val loss: 0.6313 - val binary accuracy:
0.6370
Epoch 67/200
binary accuracy: 0.6421 - val loss: 0.6320 - val binary accuracy:
0.6410
Epoch 68/200
```

```
binary accuracy: 0.6444 - val loss: 0.6305 - val binary accuracy:
0.6381
Epoch 69/200
binary accuracy: 0.6399 - val loss: 0.6309 - val binary accuracy:
0.6437
Epoch 70/200
binary accuracy: 0.6366 - val loss: 0.6316 - val binary accuracy:
0.6399
Epoch 71/200
binary accuracy: 0.6353 - val loss: 0.6306 - val binary accuracy:
0.6428
Epoch 72/200
binary accuracy: 0.6406 - val loss: 0.6303 - val binary accuracy:
0.6383
Epoch 73/200
binary accuracy: 0.6423 - val loss: 0.6315 - val binary accuracy:
0.6430
Epoch 74/200
binary accuracy: 0.6391 - val_loss: 0.6311 - val_binary_accuracy:
0.6435
Epoch 75/200
binary accuracy: 0.6437 - val loss: 0.6301 - val binary accuracy:
0.6408
Epoch 76/200
binary accuracy: 0.6390 - val loss: 0.6300 - val binary accuracy:
0.6379
Epoch 77/200
binary accuracy: 0.6399 - val loss: 0.6301 - val binary accuracy:
0.6385
Epoch 78/200
binary accuracy: 0.6400 - val loss: 0.6290 - val binary accuracy:
0.6377
Epoch 79/200
binary accuracy: 0.6418 - val loss: 0.6296 - val binary accuracy:
0.6397
Epoch 80/200
binary accuracy: 0.6449 - val loss: 0.6289 - val binary accuracy:
0.6435
```

```
Epoch 81/200
binary_accuracy: 0.6410 - val_loss: 0.6287 - val_binary_accuracy:
0.6395
Epoch 82/200
binary accuracy: 0.6449 - val_loss: 0.6286 - val_binary_accuracy:
0.6368
Epoch 83/200
binary accuracy: 0.6427 - val loss: 0.6286 - val binary accuracy:
0.6422
Epoch 84/200
binary accuracy: 0.6433 - val loss: 0.6282 - val binary accuracy:
0.6437
Epoch 85/200
binary accuracy: 0.6446 - val loss: 0.6282 - val binary accuracy:
0.6410
Epoch 86/200
binary accuracy: 0.6473 - val loss: 0.6286 - val binary accuracy:
0.6414
Epoch 87/200
binary_accuracy: 0.6400 - val_loss: 0.6281 - val_binary_accuracy:
0.6428
Epoch 88/200
binary_accuracy: 0.6437 - val loss: 0.6282 - val binary accuracy:
0.6424
Epoch 89/200
binary accuracy: 0.6425 - val loss: 0.6275 - val binary accuracy:
0.6389
Epoch 90/200
binary accuracy: 0.6451 - val loss: 0.6273 - val binary accuracy:
0.6422
Epoch 91/200
binary accuracy: 0.6420 - val loss: 0.6273 - val binary accuracy:
0.6441
Epoch 92/200
binary accuracy: 0.6482 - val loss: 0.6270 - val binary accuracy:
0.6424
Epoch 93/200
```

```
binary accuracy: 0.6504 - val loss: 0.6273 - val binary accuracy:
0.6447
Epoch 94/200
binary accuracy: 0.6469 - val loss: 0.6274 - val binary accuracy:
0.6433
Epoch 95/200
binary accuracy: 0.6458 - val loss: 0.6268 - val binary accuracy:
0.6428
Epoch 96/200
binary accuracy: 0.6440 - val loss: 0.6271 - val binary accuracy:
0.6428
Epoch 97/200
binary accuracy: 0.6448 - val loss: 0.6277 - val binary accuracy:
0.6410
Epoch 98/200
binary accuracy: 0.6433 - val loss: 0.6267 - val binary accuracy:
0.6445
Epoch 99/200
binary_accuracy: 0.6407 - val_loss: 0.6269 - val_binary_accuracy:
0.6428
Epoch 100/200
binary accuracy: 0.6420 - val loss: 0.6262 - val binary accuracy:
0.6430
Epoch 101/200
binary accuracy: 0.6456 - val loss: 0.6256 - val binary accuracy:
0.6457
Epoch 102/200
binary accuracy: 0.6408 - val loss: 0.6261 - val binary accuracy:
0.6414
Epoch 103/200
binary_accuracy: 0.6417 - val_loss: 0.6257 - val_binary_accuracy:
0.6439
Epoch 104/200
binary accuracy: 0.6497 - val loss: 0.6257 - val binary accuracy:
0.6478
Epoch 105/200
binary accuracy: 0.6399 - val loss: 0.6264 - val binary accuracy:
```

```
0.6443
Epoch 106/200
binary accuracy: 0.6444 - val loss: 0.6260 - val binary accuracy:
0.6428
Epoch 107/200
binary accuracy: 0.6447 - val loss: 0.6250 - val binary accuracy:
0.6455
Epoch 108/200
binary_accuracy: 0.6453 - val_loss: 0.6264 - val_binary_accuracy:
0.6404
Epoch 109/200
binary accuracy: 0.6489 - val loss: 0.6266 - val binary accuracy:
0.6443
Epoch 110/200
binary accuracy: 0.6513 - val loss: 0.6260 - val binary accuracy:
0.6443
Epoch 111/200
binary accuracy: 0.6465 - val loss: 0.6258 - val binary accuracy:
0.6422
Epoch 112/200
binary accuracy: 0.6494 - val loss: 0.6245 - val binary accuracy:
0.6470
Epoch 113/200
binary accuracy: 0.6459 - val loss: 0.6249 - val binary accuracy:
0.6445
Epoch 114/200
binary accuracy: 0.6485 - val loss: 0.6244 - val binary accuracy:
0.6480
Epoch 115/200
binary accuracy: 0.6467 - val loss: 0.6260 - val binary accuracy:
0.6445
Epoch 116/200
binary accuracy: 0.6463 - val loss: 0.6243 - val binary accuracy:
0.6493
Epoch 117/200
binary accuracy: 0.6488 - val loss: 0.6242 - val binary accuracy:
0.6453
Epoch 118/200
```

```
binary accuracy: 0.6473 - val loss: 0.6263 - val binary accuracy:
0.6418
Epoch 119/200
binary_accuracy: 0.6456 - val_loss: 0.6248 - val_binary_accuracy:
0.6470
Epoch 120/200
binary accuracy: 0.6518 - val loss: 0.6247 - val binary accuracy:
0.6443
Epoch 121/200
binary accuracy: 0.6472 - val loss: 0.6256 - val binary accuracy:
0.6439
Epoch 122/200
binary_accuracy: 0.6539 - val_loss: 0.6251 - val_binary_accuracy:
Epoch 123/200
binary accuracy: 0.6468 - val loss: 0.6248 - val binary accuracy:
0.6491
Epoch 124/200
binary accuracy: 0.6480 - val loss: 0.6242 - val binary accuracy:
0.6501
Epoch 125/200
binary accuracy: 0.6521 - val loss: 0.6238 - val binary accuracy:
0.6491
Epoch 126/200
binary accuracy: 0.6510 - val loss: 0.6247 - val binary accuracy:
0.6441
Epoch 127/200
binary accuracy: 0.6542 - val loss: 0.6243 - val binary accuracy:
0.6499
Epoch 128/200
binary accuracy: 0.6497 - val loss: 0.6250 - val binary accuracy:
0.6443
Epoch 129/200
binary_accuracy: 0.6499 - val_loss: 0.6250 - val_binary_accuracy:
0.6503
Epoch 130/200
binary accuracy: 0.6542 - val loss: 0.6236 - val binary accuracy:
```

```
0.6480
Epoch 131/200
binary accuracy: 0.6538 - val loss: 0.6239 - val binary accuracy:
0.6455
Epoch 132/200
binary accuracy: 0.6530 - val loss: 0.6238 - val binary accuracy:
0.6486
Epoch 133/200
binary_accuracy: 0.6450 - val_loss: 0.6244 - val_binary_accuracy:
0.6503
Epoch 134/200
binary accuracy: 0.6477 - val loss: 0.6231 - val binary accuracy:
0.6499
Epoch 135/200
binary accuracy: 0.6531 - val loss: 0.6235 - val binary accuracy:
0.6501
Epoch 136/200
binary accuracy: 0.6544 - val loss: 0.6226 - val binary accuracy:
0.6509
Epoch 137/200
binary accuracy: 0.6513 - val loss: 0.6231 - val binary accuracy:
0.6511
Epoch 138/200
binary accuracy: 0.6475 - val loss: 0.6233 - val binary accuracy:
0.6486
Epoch 139/200
binary accuracy: 0.6511 - val loss: 0.6227 - val binary accuracy:
0.6472
Epoch 140/200
binary accuracy: 0.6530 - val loss: 0.6239 - val binary accuracy:
0.6441
Epoch 141/200
binary accuracy: 0.6513 - val loss: 0.6238 - val binary accuracy:
0.6478
Epoch 142/200
binary accuracy: 0.6464 - val loss: 0.6231 - val binary accuracy:
0.6482
Epoch 143/200
```

```
binary accuracy: 0.6490 - val loss: 0.6231 - val binary accuracy:
0.6480
Epoch 144/200
binary_accuracy: 0.6493 - val_loss: 0.6231 - val_binary_accuracy:
0.6489
Epoch 145/200
binary accuracy: 0.6539 - val loss: 0.6223 - val binary accuracy:
0.6538
Epoch 146/200
binary accuracy: 0.6492 - val loss: 0.6227 - val binary accuracy:
0.6524
Epoch 147/200
binary_accuracy: 0.6494 - val_loss: 0.6228 - val_binary_accuracy:
0.6518
Epoch 148/200
binary accuracy: 0.6470 - val loss: 0.6234 - val binary accuracy:
0.6491
Epoch 149/200
binary accuracy: 0.6571 - val loss: 0.6228 - val binary accuracy:
0.6522
Epoch 150/200
binary accuracy: 0.6529 - val loss: 0.6226 - val binary accuracy:
0.6497
Epoch 151/200
binary accuracy: 0.6479 - val loss: 0.6233 - val binary accuracy:
0.6478
Epoch 152/200
binary accuracy: 0.6526 - val loss: 0.6229 - val binary accuracy:
0.6522
Epoch 153/200
binary accuracy: 0.6479 - val loss: 0.6224 - val binary accuracy:
0.6505
Epoch 154/200
binary_accuracy: 0.6522 - val_loss: 0.6222 - val_binary_accuracy:
0.6553
Epoch 155/200
binary accuracy: 0.6488 - val loss: 0.6231 - val binary accuracy:
```

```
0.6530
Epoch 156/200
binary accuracy: 0.6563 - val loss: 0.6234 - val binary accuracy:
0.6497
Epoch 157/200
binary accuracy: 0.6506 - val loss: 0.6225 - val binary accuracy:
0.6497
Epoch 158/200
binary accuracy: 0.6569 - val loss: 0.6230 - val binary accuracy:
0.6501
Epoch 159/200
binary accuracy: 0.6546 - val loss: 0.6220 - val binary accuracy:
0.6501
Epoch 160/200
binary accuracy: 0.6576 - val loss: 0.6228 - val binary accuracy:
0.6491
Epoch 161/200
binary accuracy: 0.6490 - val loss: 0.6218 - val_binary_accuracy:
0.6551
Epoch 162/200
binary accuracy: 0.6552 - val loss: 0.6230 - val binary accuracy:
0.6507
Epoch 163/200
binary_accuracy: 0.6519 - val_loss: 0.6227 - val binary accuracy:
0.6447
Epoch 164/200
binary accuracy: 0.6552 - val loss: 0.6229 - val binary accuracy:
0.6486
Epoch 165/200
binary accuracy: 0.6552 - val loss: 0.6234 - val binary accuracy:
0.6464
Epoch 166/200
binary accuracy: 0.6499 - val loss: 0.6226 - val binary accuracy:
0.6513
Epoch 167/200
binary accuracy: 0.6548 - val loss: 0.6222 - val binary accuracy:
0.6522
```

```
Epoch 168/200
binary_accuracy: 0.6560 - val_loss: 0.6228 - val_binary_accuracy:
0.6501
Epoch 169/200
binary accuracy: 0.6527 - val loss: 0.6218 - val binary accuracy:
0.6507
Epoch 170/200
binary accuracy: 0.6506 - val loss: 0.6225 - val binary accuracy:
0.6509
Epoch 171/200
binary accuracy: 0.6523 - val loss: 0.6220 - val binary accuracy:
0.6495
Epoch 172/200
binary accuracy: 0.6602 - val loss: 0.6217 - val binary accuracy:
0.6540
Epoch 173/200
binary accuracy: 0.6556 - val loss: 0.6213 - val binary accuracy:
0.6520
Epoch 174/200
binary_accuracy: 0.6552 - val_loss: 0.6228 - val_binary_accuracy:
0.6470
Epoch 175/200
binary_accuracy: 0.6569 - val loss: 0.6223 - val binary accuracy:
0.6470
Epoch 176/200
binary accuracy: 0.6574 - val loss: 0.6219 - val binary accuracy:
0.6507
Epoch 177/200
binary accuracy: 0.6559 - val loss: 0.6224 - val binary accuracy:
0.6466
Epoch 178/200
binary accuracy: 0.6536 - val loss: 0.6218 - val_binary_accuracy:
0.6505
Epoch 179/200
binary accuracy: 0.6538 - val loss: 0.6214 - val binary accuracy:
0.6486
Epoch 180/200
```

```
binary accuracy: 0.6530 - val loss: 0.6211 - val binary accuracy:
0.6528
Epoch 181/200
binary accuracy: 0.6523 - val loss: 0.6206 - val binary accuracy:
0.6534
Epoch 182/200
binary accuracy: 0.6559 - val loss: 0.6221 - val binary accuracy:
0.6476
Epoch 183/200
binary accuracy: 0.6522 - val loss: 0.6213 - val binary accuracy:
0.6505
Epoch 184/200
binary accuracy: 0.6510 - val loss: 0.6216 - val binary accuracy:
0.6459
Epoch 185/200
binary accuracy: 0.6521 - val loss: 0.6211 - val binary accuracy:
0.6549
Epoch 186/200
binary accuracy: 0.6568 - val loss: 0.6216 - val binary accuracy:
0.6476
Epoch 187/200
binary accuracy: 0.6558 - val loss: 0.6202 - val binary accuracy:
0.6571
Epoch 188/200
binary accuracy: 0.6595 - val loss: 0.6207 - val binary accuracy:
0.6532
Epoch 189/200
binary accuracy: 0.6596 - val loss: 0.6209 - val binary accuracy:
0.6555
Epoch 190/200
binary_accuracy: 0.6598 - val_loss: 0.6209 - val_binary_accuracy:
0.6503
Epoch 191/200
binary accuracy: 0.6615 - val loss: 0.6204 - val binary accuracy:
0.6518
Epoch 192/200
binary accuracy: 0.6595 - val loss: 0.6205 - val binary accuracy:
0.6534
```

```
Epoch 193/200
binary_accuracy: 0.6540 - val_loss: 0.6211 - val_binary_accuracy:
0.6511
Epoch 194/200
binary accuracy: 0.6605 - val loss: 0.6203 - val binary accuracy:
0.6511
Epoch 195/200
binary accuracy: 0.6611 - val loss: 0.6203 - val binary accuracy:
0.6544
Epoch 196/200
binary accuracy: 0.6538 - val loss: 0.6197 - val binary accuracy:
0.6515
Epoch 197/200
binary accuracy: 0.6647 - val loss: 0.6202 - val binary accuracy:
0.6542
Epoch 198/200
binary accuracy: 0.6580 - val loss: 0.6196 - val binary accuracy:
0.6571
Epoch 199/200
binary_accuracy: 0.6576 - val_loss: 0.6198 - val_binary_accuracy:
0.6520
Epoch 200/200
binary accuracy: 0.6581 - val loss: 0.6192 - val binary accuracy:
0.6530
<keras.callbacks.History at 0x1bb38b62160>
pd.DataFrame(model new.history.history)[['loss','val loss']].plot()
<AxesSubplot:>
```

