```
In [64]:
```

```
import tensorflow as tf
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import seaborn as sns
from sklearn.preprocessing import OneHotEncoder
from sklearn.compose import ColumnTransformer
from sklearn.preprocessing import StandardScaler
from keras.models import Sequential
from tensorflow.keras.layers import InputLayer
from tensorflow.keras.layers import Dense
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV
from sklearn.metrics import classification_report,confusion_matrix
```

### In [65]:

```
dataset = pd.read_csv('loan_data.csv').drop_duplicates()
dataset.head()
```

### Out[65]:

credit.policy		purpose	int.rate	installment	log.annual.inc	dti	fico	days.with.cr.line	revol.bal	revol.util	inq.la
0	1	debt_consolidation	0.1189	829.10	11.350407	19.48	737	5639.958333	28854	52.1	
1	1	credit_card	0.1071	228.22	11.082143	14.29	707	2760.000000	33623	76.7	
2	1	debt_consolidation	0.1357	366.86	10.373491	11.63	682	4710.000000	3511	25.6	
3	1	debt_consolidation	0.1008	162.34	11.350407	8.10	712	2699.958333	33667	73.2	
4	1	credit_card	0.1426	102.92	11.299732	14.97	667	4066.000000	4740	39.5	
4											···• <b>•</b>

## In [66]:

```
dataset.isnull().any()
```

### Out[66]:

credit.policy False False purpose int.rate False installment False log.annual.inc False dti False fico False days.with.cr.line False revol.bal False revol.util False inq.last.6mths False False delinq.2yrs False pub.rec False not.fully.paid dtype: bool

# In [67]:

```
dataset.shape
```

### Out[67]:

(9578, 14)

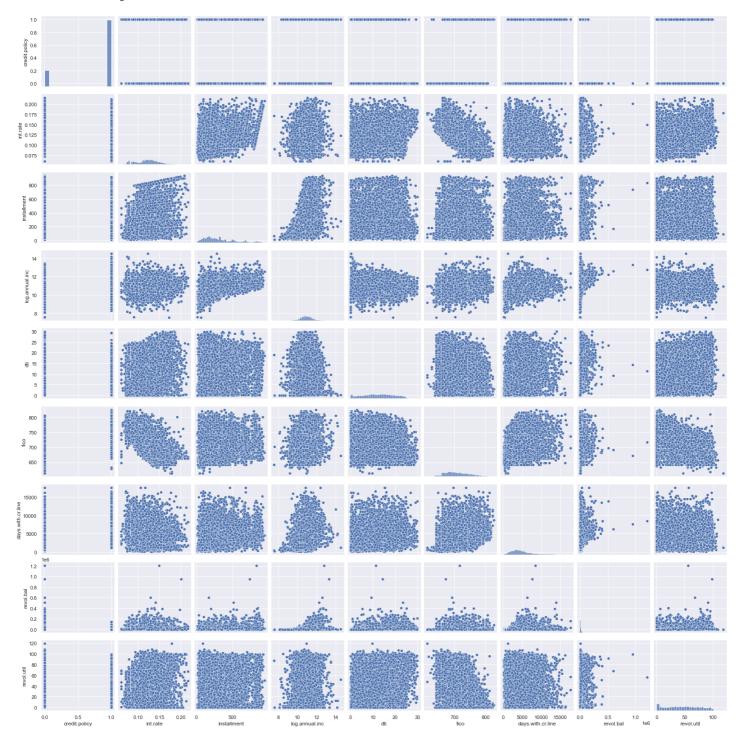
## In [68]:

```
sns.pairplot(dataset.drop(columns=['purpose','inq.last.6mths','delinq.2yrs','pub.rec','no
```

```
t.fully.paid']))
```

### Out[68]:

<seaborn.axisgrid.PairGrid at 0x196ff2b7fd0>



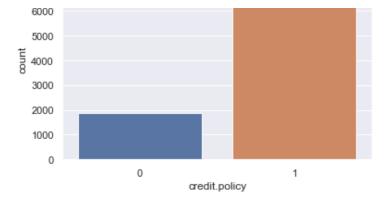
### In [69]:

```
sns.set()
cat_var = dataset.drop(columns = ['int.rate', 'installment', 'log.annual.inc', 'dti', 'fico'
,'days.with.cr.line', 'revol.bal', 'revol.util']).copy()
cat_cols = cat_var.columns.values.tolist()
for col in cat_cols:
    sns.countplot(cat_var[col])
    plt.show()
```

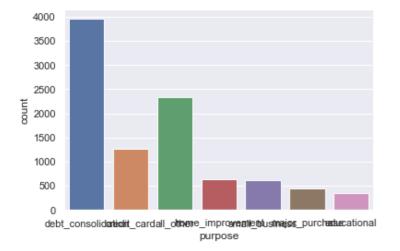
c:\Users\97156\AppData\Local\Programs\Python\Python310\lib\site-packages\seaborn\\_decorat ors.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0 .12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

warnings.warn(

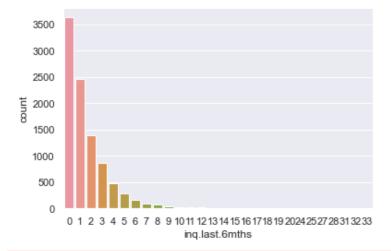
```
7000
```



c:\Users\97156\AppData\Local\Programs\Python\Python310\lib\site-packages\seaborn\\_decorat
ors.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0
.12, the only valid positional argument will be `data`, and passing other arguments witho
ut an explicit keyword will result in an error or misinterpretation.
warnings.warn(



c:\Users\97156\AppData\Local\Programs\Python\Python310\lib\site-packages\seaborn\\_decorat
ors.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0
.12, the only valid positional argument will be `data`, and passing other arguments witho
ut an explicit keyword will result in an error or misinterpretation.
warnings.warn(

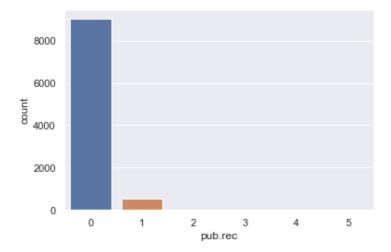


c:\Users\97156\AppData\Local\Programs\Python\Python310\lib\site-packages\seaborn\\_decorat
ors.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0
.12, the only valid positional argument will be `data`, and passing other arguments witho
ut an explicit keyword will result in an error or misinterpretation.
warnings.warn(

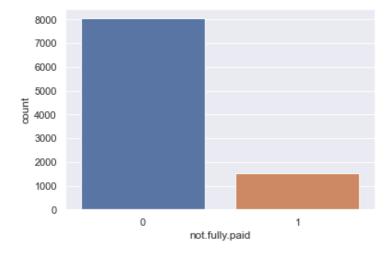




c:\Users\97156\AppData\Local\Programs\Python\Python310\lib\site-packages\seaborn\\_decorat
ors.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0
.12, the only valid positional argument will be `data`, and passing other arguments witho
ut an explicit keyword will result in an error or misinterpretation.
warnings.warn(



c:\Users\97156\AppData\Local\Programs\Python\Python310\lib\site-packages\seaborn\\_decorat
ors.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0
.12, the only valid positional argument will be `data`, and passing other arguments witho
ut an explicit keyword will result in an error or misinterpretation.
warnings.warn(



## In [70]:

```
ohe = OneHotEncoder()
ct = ColumnTransformer(transformers=[('encoder', ohe, ['purpose'])], remainder='passthro
ugh')
ct.fit_transform(dataset)
dataset = pd.get_dummies(dataset)
dataset.head()
```

### Out[70]:

	credit.policy	int.rate	installment	log.annual.inc	dti	fico	days.with.cr.line	revol.bal	revol.util	inq.last.6mths	delinq.2yrs
0	1	0.1189	829.10	11.350407	19.48	737	5639.958333	28854	52.1	0	С
1	1	0.1071	228.22	11.082143	14.29	707	2760.000000	33623	76.7	0	C
2	1	0.1357	366.86	10.373491	11.63	682	4710.000000	3511	25.6	1	С
3	1	0.1008	162.34	11.350407	8.10	712	2699.958333	33667	73.2	1	С

## In [71]:

```
f = plt.figure(figsize=(20,20))
corr = dataset.corr()
corr.style.background_gradient(cmap='coolwarm', vmin=-1, vmax=1)
```

## Out[71]:

	credit.policy	int.rate	installment	log.annual.inc	dti	fico	days.with.cr.line	revol.bal
credit.policy	1.000000	0.294089	0.058770	0.034906	0.090901	0.348319	0.099026	- 0.187518
int.rate	-0.294089	1.000000	0.276140	0.056383	0.220006	- 0.714821	-0.124022	0.092527
installment	0.058770	0.276140	1.000000	0.448102	0.050202	0.086039	0.183297	0.233625
log.annual.inc	0.034906	0.056383	0.448102	1.000000	- 0.054065	0.114576	0.336896	0.372140
dti	-0.090901	0.220006	0.050202	-0.054065	1.000000	- 0.241191	0.060101	0.188748
fico	0.348319	- 0.714821	0.086039	0.114576	- 0.241191	1.000000	0.263880	- 0.015553
days.with.cr.line	0.099026	- 0.124022	0.183297	0.336896	0.060101	0.263880	1.000000	0.229344
revol.bal	-0.187518	0.092527	0.233625	0.372140	0.188748	- 0.015553	0.229344	1.000000
revol.util	-0.104095	0.464837	0.081356	0.054881	0.337109	- 0.541289	-0.024239	0.203779
inq.last.6mths	-0.535511	0.202780	-0.010419	0.029171	0.029189	- 0.185293	-0.041736	0.022394
delinq.2yrs	-0.076318	0.156079	-0.004368	0.029203	- 0.021792	0.216340	0.081374	0.033243
pub.rec	-0.054243	0.098162	-0.032760	0.016506	0.006209	- 0.147592	0.071826	0.031010
not.fully.paid	-0.158119	0.159552	0.049955	-0.033439	0.037362	- 0.149666	-0.029237	0.053699
purpose_all_other	-0.025412	- 0.124000	-0.203103	-0.080077	- 0.125825	0.067184	-0.056574	- 0.067728
purpose_credit_card	0.003216	- 0.042109	0.000774	0.072942	0.084476	- 0.012512	0.046220	0.072316
purpose_debt_consolidation	0.020193	0.123607	0.161658	-0.026214	0.179149	- 0.154132	-0.009318	0.005785
purpose_educational	-0.031346	- 0.019618	-0.094510	-0.119799	0.035325	0.013012	-0.042621	0.034743
purpose_home_improvement	0.006036	- 0.050697	0.023024	0.116375	0.092788	0.097474	0.068087	0.003258
purpose_major_purchase	0.024281	- 0.068978	-0.079836	-0.031020	0.077719	0.067129	-0.020561	0.062395
purpose_small_business	-0.003511		0.145654	0.091540		0.063292	0.034883	0.083069
4								Þ

<Figure size 1440x1440 with 0 Axes>

We can see that the intrest rate and revolutil are highly coorelated with fico score. and that the fico score has more predictive power than the both. therefore, we will be dropping these two features

```
In [72]:
dataset = dataset.drop(columns=['int.rate','revol.util'])
In [73]:
X = dataset.iloc[:, 1:].values
y = dataset.iloc[:, 0].values
stdscl = StandardScaler()
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.25, random_state
X train[:,0:9] = stdscl.fit transform(X train[:,0:9])
X test[:,0:9] = stdscl.fit transform(X test[:,0:9])
In [74]:
def build classifier(optimizer):
   model = tf.keras.models.Sequential()
   model.add(tf.keras.layers.Dense(units = 5, activation = 'relu', input shape = (17,))
   model.add(tf.keras.layers.Dense(units = 5, activation = 'relu'))
   model.add(tf.keras.layers.Dense(units = 10, activation = 'relu'))
   model.add(tf.keras.layers.Dense(units = 1, activation = 'sigmoid'))
   model.compile(optimizer=optimizer,loss = 'binary crossentropy',metrics = ['accuracy'
])
   return model
In [83]:
classifier = tf.keras.wrappers.scikit learn.KerasClassifier(build fn=build classifier)
param grid = { 'epochs': [100,5000], 'optimizer': [tf.keras.optimizers.Adam(), tf.keras.optim
izers.RMSprop(),tf.keras.optimizers.Adagrad()]}
model = GridSearchCV(estimator=classifier,param grid=param grid,scoring='accuracy')
C:\Users\97156\AppData\Local\Temp\ipykernel 11728\2212760722.py:1: DeprecationWarning: Ke
rasClassifier is deprecated, use Sci-Keras (https://github.com/adriangb/scikeras) instead
. See https://www.adriangb.com/scikeras/stable/migration.html for help migrating.
 classifier = tf.keras.wrappers.scikit learn.KerasClassifier(build fn=build classifier)
In [76]:
X train
Out[76]:
array([[ 0.78416021, 0.95396034, -1.13324201, ..., 0.
                  1.
                            ],
      [0.31339588, -0.24590795, -0.0069299, ..., 0.
                   0.
                            ],
      [ 0.92031809, 0.11887967, 0.67264084, ...,
                            ],
      [0.39985687, -0.17043457, -0.36053951, ...,
                   0.
                            ],
      [ 1.50263404, 0.24973082, -0.50023714, ...,
                                               0.
                   1.
                            ],
      [-1.08300958, -1.31230739, 0.22735466, ..., 0.
                  0.
        0.
                           ]])
In [84]:
history = model.fit(X train, y train,epochs = 10,validation data=(X test, y test))
Epoch 1/10
- val loss: 0.3755 - val accuracy: 0.8125
- val loss: 0.3119 - val accuracy: 0.8685
Epoch 3/10
- val loss: 0.2710 - val accuracy: 0.8931
```

```
Epoch 4/10
- val loss: 0.2505 - val accuracy: 0.9035
Epoch 5/10
- val loss: 0.2408 - val accuracy: 0.9106
Epoch 6/10
- val loss: 0.2339 - val accuracy: 0.9086
Epoch 7/10
- val loss: 0.2297 - val accuracy: 0.9123
- val loss: 0.2259 - val accuracy: 0.9106
Epoch 9/10
- val loss: 0.2234 - val accuracy: 0.9144
Epoch 10/10
- val loss: 0.2216 - val accuracy: 0.9136
45/45 [========= ] - 0s 1ms/step
Epoch 1/10
- val loss: 0.5319 - val accuracy: 0.8234
Epoch 2/10
- val loss: 0.3997 - val accuracy: 0.8518
Epoch 3/10
- val loss: 0.2970 - val accuracy: 0.8956
Epoch 4/10
- val loss: 0.2550 - val accuracy: 0.9023
Epoch 5/10
- val loss: 0.2366 - val accuracy: 0.9077
Epoch 6/10
- val loss: 0.2269 - val accuracy: 0.9115
Epoch 7/10
- val loss: 0.2222 - val accuracy: 0.9132
Epoch 8/10
- val loss: 0.2180 - val accuracy: 0.9144
Epoch 9/10
- val loss: 0.2164 - val accuracy: 0.9177
Epoch 10/10
- val loss: 0.2143 - val accuracy: 0.9177
45/45 [========= ] - 0s 977us/step
Epoch 1/10
- val_loss: 0.4680 - val_accuracy: 0.8125
Epoch 2/10
- val loss: 0.3707 - val accuracy: 0.8125
Epoch 3/10
- val loss: 0.3129 - val accuracy: 0.8660
Epoch 4/10
- val loss: 0.2793 - val accuracy: 0.8756
Epoch 5/10
- val loss: 0.2503 - val accuracy: 0.9010
- val loss: 0.2361 - val accuracy: 0.9052
```

Epoch 7/10

```
- val loss: 0.2299 - val accuracy: 0.9081
Epoch 8/10
- val loss: 0.2250 - val accuracy: 0.9111
Epoch 9/10
- val loss: 0.2209 - val accuracy: 0.9127
Epoch 10/10
- val loss: 0.2181 - val accuracy: 0.9132
45/45 [=========== ] - Os 2ms/step
Epoch 1/10
- val loss: 0.4753 - val accuracy: 0.8125
Epoch 2/10
- val loss: 0.3798 - val accuracy: 0.8125
Epoch 3/10
- val loss: 0.3058 - val accuracy: 0.8810
Epoch 4/10
- val loss: 0.2778 - val accuracy: 0.9006
Epoch 5/10
- val loss: 0.2621 - val accuracy: 0.9065
Epoch 6/10
- val loss: 0.2523 - val accuracy: 0.9065
Epoch 7/10
- val loss: 0.2436 - val accuracy: 0.9090
Epoch 8/10
- val_loss: 0.2387 - val_accuracy: 0.9115
Epoch 9/10
- val loss: 0.2335 - val accuracy: 0.9115
Epoch 10/10
- val loss: 0.2297 - val accuracy: 0.9111
45/45 [=========== ] - Os 1ms/step
Epoch 1/10
- val loss: 0.4305 - val accuracy: 0.8125
Epoch 2/10
- val loss: 0.3467 - val accuracy: 0.8125
Epoch 3/10
- val loss: 0.3097 - val accuracy: 0.8718
Epoch 4/10
- val loss: 0.2835 - val accuracy: 0.9015
Epoch 5/10
- val loss: 0.2650 - val accuracy: 0.9069
Epoch 6/10
- val loss: 0.2510 - val accuracy: 0.9077
Epoch 7/10
- val loss: 0.2367 - val accuracy: 0.9127
Epoch 8/10
- val loss: 0.2285 - val accuracy: 0.9127
Epoch 9/10
- val loss: 0.2217 - val accuracy: 0.9127
Epoch 10/10
```

```
.. ...., ...r
                      - val loss: 0.2177 - val accuracy: 0.9161
45/45 [========= ] - Os 1ms/step
Epoch 1/10
- val loss: 0.4249 - val accuracy: 0.8175
Epoch 2/10
- val loss: 0.3121 - val accuracy: 0.8768
Epoch 3/10
- val loss: 0.2717 - val accuracy: 0.8944
Epoch 4/10
- val loss: 0.2537 - val accuracy: 0.8990
- val_loss: 0.2410 - val_accuracy: 0.9052
Epoch 6/10
- val_loss: 0.2294 - val_accuracy: 0.9102
Epoch 7/10
- val loss: 0.2256 - val accuracy: 0.9073
Epoch 8/10
- val loss: 0.2186 - val_accuracy: 0.9127
Epoch 9/10
- val loss: 0.2180 - val accuracy: 0.9115
Epoch 10/10
- val loss: 0.2132 - val accuracy: 0.9136
45/45 [========= ] - 0s 1ms/step
Epoch 1/10
- val loss: 0.4451 - val_accuracy: 0.8154
Epoch 2/10
- val loss: 0.3676 - val accuracy: 0.8438
Epoch 3/10
- val loss: 0.3152 - val accuracy: 0.8676
Epoch 4/10
- val loss: 0.2861 - val accuracy: 0.8868
Epoch 5/10
- val loss: 0.2705 - val accuracy: 0.8948
Epoch 6/10
- val_loss: 0.2614 - val_accuracy: 0.8981
Epoch 7/10
- val loss: 0.2546 - val_accuracy: 0.8994
Epoch 8/10
- val loss: 0.2476 - val_accuracy: 0.9027
Epoch 9/10
- val loss: 0.2426 - val accuracy: 0.9040
Epoch 10/10
- val loss: 0.2386 - val accuracy: 0.9077
45/45 [=======] - Os 1ms/step
Epoch 1/10
- val loss: 0.4057 - val accuracy: 0.8125
Epoch 2/10
- val loss: 0.3551 - val_accuracy: 0.8125
Epoch 3/10
```

```
_____
                - val loss: 0.3318 - val accuracy: 0.8493
Epoch 4/10
- val loss: 0.3192 - val accuracy: 0.8539
Epoch 5/10
- val loss: 0.3078 - val accuracy: 0.8564
Epoch 6/10
- val loss: 0.2896 - val accuracy: 0.8747
Epoch 7/10
- val loss: 0.2696 - val accuracy: 0.8990
Epoch 8/10
- val loss: 0.2575 - val accuracy: 0.9044
Epoch 9/10
- val loss: 0.2499 - val accuracy: 0.9069
Epoch 10/10
- val loss: 0.2440 - val accuracy: 0.9069
45/45 [========== ] - 1s 1ms/step
Epoch 1/10
- val loss: 0.4090 - val_accuracy: 0.8125
Epoch 2/10
- val loss: 0.3586 - val accuracy: 0.8125
Epoch 3/10
- val loss: 0.3259 - val accuracy: 0.8125
Epoch 4/10
- val loss: 0.3024 - val_accuracy: 0.8818
Epoch 5/10
- val loss: 0.2857 - val accuracy: 0.8914
Epoch 6/10
- val loss: 0.2755 - val accuracy: 0.8985
Epoch 7/10
- val loss: 0.2673 - val accuracy: 0.9023
Epoch 8/10
- val loss: 0.2597 - val accuracy: 0.9065
Epoch 9/10
- val loss: 0.2544 - val accuracy: 0.9069
Epoch 10/10
- val loss: 0.2496 - val accuracy: 0.9086
Epoch 1/10
- val_loss: 0.4286 - val_accuracy: 0.8125
Epoch 2/10
- val loss: 0.3655 - val accuracy: 0.8125
Epoch 3/10
- val loss: 0.3267 - val accuracy: 0.8693
Epoch 4/10
- val loss: 0.3050 - val accuracy: 0.8856
Epoch 5/10
- val loss: 0.2916 - val accuracy: 0.8990
Epoch 6/10
```

- val loss: 0.2823 - val accuracy: 0.9035

```
_____
       Epoch 7/10
- val loss: 0.2754 - val accuracy: 0.9044
Epoch 8/10
- val loss: 0.2724 - val accuracy: 0.9044
Epoch 9/10
- val loss: 0.2678 - val accuracy: 0.9048
Epoch 10/10
- val loss: 0.2638 - val accuracy: 0.9065
45/45 [========= ] - 0s 1ms/step
Epoch 1/10
- val loss: 0.6613 - val accuracy: 0.8163
Epoch 2/10
- val loss: 0.6493 - val accuracy: 0.8180
Epoch 3/10
- val loss: 0.6393 - val accuracy: 0.8154
Epoch 4/10
- val loss: 0.6305 - val accuracy: 0.8134
Epoch 5/10
- val loss: 0.6223 - val accuracy: 0.8134
Epoch 6/10
- val loss: 0.6145 - val accuracy: 0.8129
Epoch 7/10
- val loss: 0.6070 - val accuracy: 0.8125
Epoch 8/10
- val loss: 0.5998 - val accuracy: 0.8125
Epoch 9/10
- val loss: 0.5929 - val accuracy: 0.8125
Epoch 10/10
- val loss: 0.5861 - val accuracy: 0.8125
Epoch 1/10
- val loss: 0.7000 - val accuracy: 0.6150
- val loss: 0.6851 - val accuracy: 0.6814
Epoch 3/10
- val loss: 0.6736 - val accuracy: 0.7177
Epoch 4/10
- val loss: 0.6638 - val accuracy: 0.7474
Epoch 5/10
- val loss: 0.6551 - val accuracy: 0.7704
Epoch 6/10
- val loss: 0.6471 - val accuracy: 0.7829
Epoch 7/10
- val loss: 0.6396 - val accuracy: 0.7871
Epoch 8/10
- val loss: 0.6325 - val accuracy: 0.7937
Epoch 9/10
- val loss: 0.6255 - val accuracy: 0.8021
```

Enoch 10/10

```
- val loss: 0.6188 - val accuracy: 0.8071
45/45 [========= ] - Os 1ms/step
Epoch 1/10
- val loss: 0.5880 - val accuracy: 0.8125
Epoch 2/10
- val loss: 0.5721 - val accuracy: 0.8125
Epoch 3/10
- val loss: 0.5598 - val accuracy: 0.8125
Epoch 4/10
- val loss: 0.5496 - val accuracy: 0.8125
Epoch 5/10
- val loss: 0.5406 - val accuracy: 0.8125
Epoch 6/10
- val loss: 0.5327 - val accuracy: 0.8125
Epoch 7/10
- val loss: 0.5256 - val accuracy: 0.8125
Epoch 8/10
- val loss: 0.5191 - val accuracy: 0.8125
Epoch 9/10
- val loss: 0.5133 - val accuracy: 0.8125
Epoch 10/10
- val loss: 0.5079 - val accuracy: 0.8125
45/45 [========= ] - 0s 1ms/step
Epoch 1/10
- val loss: 0.6354 - val accuracy: 0.8125
Epoch 2/10
- val loss: 0.6209 - val accuracy: 0.8125
Epoch 3/10
- val loss: 0.6097 - val accuracy: 0.8125
Epoch 4/10
- val loss: 0.6002 - val accuracy: 0.8125
Epoch 5/10
- val loss: 0.5920 - val accuracy: 0.8125
Epoch 6/10
- val loss: 0.5846 - val accuracy: 0.8125
Epoch 7/10
- val loss: 0.5779 - val accuracy: 0.8125
Epoch 8/10
- val loss: 0.5717 - val accuracy: 0.8125
Epoch 9/10
- val loss: 0.5661 - val accuracy: 0.8125
Epoch 10/10
- val loss: 0.5608 - val accuracy: 0.8125
45/45 [========] - 0s 1ms/step
Epoch 1/10
- val loss: 0.6551 - val accuracy: 0.7428
- val loss: 0.6369 - val accuracy: 0.7875
```

Epoch 3/10

```
- val loss: 0.6231 - val accuracy: 0.8021
Epoch 4/10
- val loss: 0.6118 - val accuracy: 0.8071
Epoch 5/10
- val loss: 0.6022 - val accuracy: 0.8092
Epoch 6/10
- val loss: 0.5937 - val accuracy: 0.8113
Epoch 7/10
- val loss: 0.5862 - val accuracy: 0.8113
Epoch 8/10
- val_loss: 0.5795 - val_accuracy: 0.8121
Epoch 9/10
- val_loss: 0.5734 - val_accuracy: 0.8121
Epoch 10/10
- val loss: 0.5678 - val accuracy: 0.8121
45/45 [========= ] - 0s 1ms/step
Epoch 1/10
- val loss: 0.4557 - val accuracy: 0.8125
Epoch 2/10
- val loss: 0.3481 - val accuracy: 0.8522
Epoch 3/10
- val loss: 0.2871 - val accuracy: 0.8877
Epoch 4/10
- val_loss: 0.2606 - val_accuracy: 0.8981
Epoch 5/10
- val loss: 0.2537 - val accuracy: 0.8994
Epoch 6/10
- val loss: 0.2465 - val accuracy: 0.9027
Epoch 7/10
- val loss: 0.2434 - val accuracy: 0.9044
Epoch 8/10
- val loss: 0.2424 - val accuracy: 0.9002
Epoch 9/10
- val_loss: 0.2372 - val_accuracy: 0.9048
Epoch 10/10
- val loss: 0.2359 - val_accuracy: 0.9061
45/45 [========= ] - Os 1ms/step
Epoch 1/10
- val loss: 0.4401 - val accuracy: 0.8125
Epoch 2/10
- val loss: 0.3643 - val accuracy: 0.8138
Epoch 3/10
- val loss: 0.3197 - val accuracy: 0.8288
Epoch 4/10
- val loss: 0.2886 - val accuracy: 0.8848
Epoch 5/10
- val loss: 0.2715 - val accuracy: 0.9006
Epoch 6/10
```

```
_____
                · - -...., - - - - -
- val loss: 0.2569 - val accuracy: 0.9031
Epoch 7/10
- val loss: 0.2477 - val accuracy: 0.9010
Epoch 8/10
- val loss: 0.2384 - val accuracy: 0.9094
Epoch 9/10
- val loss: 0.2331 - val accuracy: 0.9106
Epoch 10/10
- val loss: 0.2306 - val accuracy: 0.9077
45/45 [========= ] - 0s 1ms/step
Epoch 1/10
- val_loss: 0.4578 - val_accuracy: 0.8125
Epoch 2/10
- val loss: 0.3666 - val accuracy: 0.8125
Epoch 3/10
- val loss: 0.3147 - val accuracy: 0.8806
Epoch 4/10
- val loss: 0.2685 - val_accuracy: 0.9010
Epoch 5/10
- val loss: 0.2523 - val accuracy: 0.9040
Epoch 6/10
- val loss: 0.2468 - val accuracy: 0.9052
Epoch 7/10
- val loss: 0.2436 - val_accuracy: 0.9061
Epoch 8/10
- val loss: 0.2417 - val accuracy: 0.9081
Epoch 9/10
- val loss: 0.2410 - val accuracy: 0.9052
Epoch 10/10
- val loss: 0.2399 - val accuracy: 0.9094
45/45 [========= ] - 0s 2ms/step
Epoch 1/10
- val loss: 0.4357 - val accuracy: 0.8205
Epoch 2/10
- val_loss: 0.2901 - val_accuracy: 0.8885
Epoch 3/10
- val_loss: 0.2449 - val_accuracy: 0.9031
Epoch 4/10
- val_loss: 0.2315 - val_accuracy: 0.9069
Epoch 5/10
- val loss: 0.2279 - val accuracy: 0.9069
Epoch 6/10
- val loss: 0.2253 - val accuracy: 0.9081
Epoch 7/10
- val loss: 0.2208 - val accuracy: 0.9111
Epoch 8/10
- val loss: 0.2197 - val accuracy: 0.9115
Epoch 9/10
```

- val loss: 0.2214 - val accuracy: 0.9111

```
Epoch 10/10
- val loss: 0.2158 - val accuracy: 0.9148
45/45 [=========== ] - Os 2ms/step
Epoch 1/10
- val loss: 0.4331 - val accuracy: 0.8330
Epoch 2/10
- val loss: 0.2908 - val accuracy: 0.8902
Epoch 3/10
- val loss: 0.2594 - val accuracy: 0.8994
Epoch 4/10
- val loss: 0.2524 - val accuracy: 0.9027
Epoch 5/10
- val loss: 0.2484 - val accuracy: 0.9006
Epoch 6/10
- val loss: 0.2459 - val accuracy: 0.9010
Epoch 7/10
- val loss: 0.2427 - val accuracy: 0.9027
Epoch 8/10
- val loss: 0.2419 - val accuracy: 0.9044
Epoch 9/10
- val loss: 0.2401 - val accuracy: 0.9056
Epoch 10/10
- val loss: 0.2387 - val accuracy: 0.9069
45/45 [========= ] - 0s 1ms/step
Epoch 1/10
- val loss: 0.3998 - val accuracy: 0.8605
Epoch 2/10
- val_loss: 0.3058 - val accuracy: 0.8931
Epoch 3/10
- val loss: 0.2709 - val accuracy: 0.9010
Epoch 4/10
- val loss: 0.2521 - val accuracy: 0.9073
Epoch 5/10
- val loss: 0.2421 - val accuracy: 0.9069
Epoch 6/10
- val loss: 0.2338 - val accuracy: 0.9090
Epoch 7/10
- val loss: 0.2293 - val accuracy: 0.9111
Epoch 8/10
- val loss: 0.2259 - val accuracy: 0.9102
Epoch 9/10
- val loss: 0.2221 - val accuracy: 0.9119
Epoch 10/10
- val_loss: 0.2186 - val_accuracy: 0.9119
45/45 [========= ] - Os 1ms/step
Epoch 1/10
- val loss: 0.4843 - val accuracy: 0.8301
Epoch 2/10
```

- val loss: 0.3175 - val accuracy: 0.8914

```
Epoch 3/10
- val loss: 0.2626 - val accuracy: 0.9035
Epoch 4/10
- val loss: 0.2482 - val accuracy: 0.9081
Epoch 5/10
- val loss: 0.2416 - val accuracy: 0.9090
Epoch 6/10
- val loss: 0.2372 - val accuracy: 0.9094
Epoch 7/10
- val loss: 0.2345 - val accuracy: 0.9115
Epoch 8/10
- val loss: 0.2328 - val accuracy: 0.9102
Epoch 9/10
- val loss: 0.2315 - val accuracy: 0.9069
Epoch 10/10
- val loss: 0.2287 - val accuracy: 0.9102
45/45 [======== ] - 0s 1ms/step
Epoch 1/10
- val loss: 0.3923 - val accuracy: 0.8125
Epoch 2/10
- val loss: 0.3365 - val accuracy: 0.8547
Epoch 3/10
- val loss: 0.3068 - val accuracy: 0.8898
Epoch 4/10
- val loss: 0.2884 - val accuracy: 0.8985
Epoch 5/10
- val loss: 0.2743 - val accuracy: 0.9006
Epoch 6/10
- val loss: 0.2633 - val accuracy: 0.9035
Epoch 7/10
- val loss: 0.2551 - val accuracy: 0.9065
Epoch 8/10
- val loss: 0.2486 - val accuracy: 0.9081
Epoch 9/10
- val loss: 0.2437 - val accuracy: 0.9106
Epoch 10/10
- val loss: 0.2378 - val accuracy: 0.9098
45/45 [========] - Os 1ms/step
Epoch 1/10
- val loss: 0.4530 - val accuracy: 0.8125
Epoch 2/10
- val loss: 0.3842 - val accuracy: 0.8125
Epoch 3/10
- val loss: 0.3305 - val accuracy: 0.8184
Epoch 4/10
- val loss: 0.2999 - val accuracy: 0.8864
- val loss: 0.2775 - val accuracy: 0.8981
```

Enoch 6/10

```
- val loss: 0.2634 - val accuracy: 0.8985
Epoch 7/10
- val loss: 0.2514 - val accuracy: 0.9035
Epoch 8/10
- val loss: 0.2452 - val accuracy: 0.9069
Epoch 9/10
- val loss: 0.2399 - val accuracy: 0.9094
Epoch 10/10
- val loss: 0.2319 - val accuracy: 0.9090
45/45 [=========] - 0s 1ms/step
Epoch 1/10
- val loss: 0.4148 - val accuracy: 0.8125
Epoch 2/10
- val loss: 0.3461 - val accuracy: 0.8125
Epoch 3/10
- val loss: 0.2944 - val accuracy: 0.8960
Epoch 4/10
- val loss: 0.2645 - val accuracy: 0.9081
Epoch 5/10
- val loss: 0.2472 - val accuracy: 0.9098
Epoch 6/10
- val loss: 0.2363 - val accuracy: 0.9119
Epoch 7/10
- val_loss: 0.2264 - val_accuracy: 0.9132
Epoch 8/10
- val loss: 0.2207 - val accuracy: 0.9136
Epoch 9/10
- val loss: 0.2161 - val accuracy: 0.9157
Epoch 10/10
- val loss: 0.2134 - val accuracy: 0.9177
45/45 [========] - Os 1ms/step
Epoch 1/10
- val loss: 0.6469 - val accuracy: 0.8025
Epoch 2/10
- val loss: 0.6314 - val accuracy: 0.8125
Epoch 3/10
- val loss: 0.6194 - val accuracy: 0.8125
Epoch 4/10
- val loss: 0.6092 - val accuracy: 0.8125
Epoch 5/10
- val loss: 0.6002 - val accuracy: 0.8125
Epoch 6/10
- val loss: 0.5922 - val accuracy: 0.8125
Epoch 7/10
- val loss: 0.5849 - val accuracy: 0.8125
Epoch 8/10
- val_loss: 0.5782 - val_accuracy: 0.8125
Epoch 9/10
```

```
___, ___ L
                       _____
                 .. ...., ...<sub>F</sub>
- val loss: 0.5721 - val accuracy: 0.8125
Epoch 10/10
- val loss: 0.5664 - val accuracy: 0.8125
45/45 [========= ] - 0s 1ms/step
Epoch 1/10
- val loss: 0.6661 - val accuracy: 0.7182
Epoch 2/10
- val loss: 0.6516 - val accuracy: 0.7658
Epoch 3/10
- val_loss: 0.6406 - val_accuracy: 0.7896
Epoch 4/10
- val_loss: 0.6313 - val_accuracy: 0.8046
Epoch 5/10
- val_loss: 0.6232 - val_accuracy: 0.8109
Epoch 6/10
- val loss: 0.6158 - val accuracy: 0.8125
Epoch 7/10
- val loss: 0.6088 - val_accuracy: 0.8125
Epoch 8/10
- val loss: 0.6022 - val accuracy: 0.8121
Epoch 9/10
- val loss: 0.5959 - val accuracy: 0.8125
Epoch 10/10
- val_loss: 0.5898 - val_accuracy: 0.8125
45/45 [=======] - Os 1ms/step
Epoch 1/10
- val loss: 0.6463 - val accuracy: 0.8125
Epoch 2/10
- val loss: 0.6332 - val accuracy: 0.8125
Epoch 3/10
- val loss: 0.6224 - val accuracy: 0.8125
Epoch 4/10
- val loss: 0.6129 - val accuracy: 0.8125
Epoch 5/10
- val_loss: 0.6044 - val_accuracy: 0.8125
Epoch 6/10
- val_loss: 0.5965 - val_accuracy: 0.8125
Epoch 7/10
- val_loss: 0.5892 - val_accuracy: 0.8125
Epoch 8/10
- val loss: 0.5824 - val accuracy: 0.8125
Epoch 9/10
- val loss: 0.5759 - val accuracy: 0.8125
Epoch 10/10
- val loss: 0.5699 - val_accuracy: 0.8125
45/45 [========== ] - Os 1ms/step
Epoch 1/10
- val loss: 0.6312 - val accuracy: 0.8125
Epoch 2/10
```

```
_____
                - val loss: 0.6110 - val accuracy: 0.8125
Epoch 3/10
- val loss: 0.5962 - val accuracy: 0.8125
Epoch 4/10
- val loss: 0.5842 - val accuracy: 0.8125
Epoch 5/10
- val loss: 0.5741 - val accuracy: 0.8125
Epoch 6/10
- val loss: 0.5653 - val accuracy: 0.8125
Epoch 7/10
- val loss: 0.5574 - val accuracy: 0.8125
Epoch 8/10
- val loss: 0.5503 - val accuracy: 0.8125
Epoch 9/10
- val loss: 0.5437 - val accuracy: 0.8125
Epoch 10/10
- val loss: 0.5376 - val accuracy: 0.8125
45/45 [========= ] - Os 1ms/step
Epoch 1/10
- val loss: 0.6385 - val accuracy: 0.8125
Epoch 2/10
- val loss: 0.6256 - val accuracy: 0.8125
Epoch 3/10
- val loss: 0.6158 - val_accuracy: 0.8125
Epoch 4/10
- val loss: 0.6077 - val accuracy: 0.8125
Epoch 5/10
- val loss: 0.6007 - val accuracy: 0.8125
Epoch 6/10
- val_loss: 0.5945 - val accuracy: 0.8125
Epoch 7/10
- val loss: 0.5889 - val accuracy: 0.8125
Epoch 8/10
- val loss: 0.5838 - val accuracy: 0.8125
Epoch 9/10
- val loss: 0.5791 - val accuracy: 0.8125
Epoch 10/10
- val loss: 0.5747 - val accuracy: 0.8125
45/45 [========] - 0s 1ms/step
Epoch 1/10
- val loss: 0.4275 - val accuracy: 0.8125
Epoch 2/10
- val loss: 0.3419 - val accuracy: 0.8125
Epoch 3/10
- val loss: 0.2783 - val accuracy: 0.9006
Epoch 4/10
- val loss: 0.2528 - val accuracy: 0.9077
Epoch 5/10
```

- val loss: 0.2440 - val accuracy: 0.9090

```
Epoch 6/10
225/225 [===========] - 0s 2ms/step - loss: 0.2526 - accuracy: 0.9023
- val loss: 0.2373 - val accuracy: 0.9111
Epoch 7/10
- val loss: 0.2338 - val accuracy: 0.9077
Epoch 8/10
- val loss: 0.2304 - val accuracy: 0.9102
Epoch 9/10
- val loss: 0.2266 - val accuracy: 0.9140
Epoch 10/10
- val loss: 0.2244 - val accuracy: 0.9115
In [78]:
print(model.best params )
{'epochs': 1000, 'optimizer': <keras.optimizers.optimizer v2.adam.Adam object at 0x000001
9685E6A350>}
In [79]:
print(model.best score )
0.9039388776137225
In [80]:
y pred = (model.predict(X test)>0.5).astype('int64')
75/75 [======== ] - 0s 958us/step
In [81]:
print(confusion matrix(y test, y pred))
print(classification report(y test, y pred))
[[ 331 118]
[ 65 1881]]
         precision recall f1-score
                               support
                   0.74
                          0.78
       0
             0.84
                                 449
            0.94
                   0.97
                          0.95
                                 1946
       1
  accuracy
                          0.92
                                 2395
  macro avg
            0.89
                  0.85
                         0.87
                                 2395
weighted avg
            0.92
                   0.92
                         0.92
                                 2395
```