```
In [1]: #Name:Pawar Ved Balasaheb(T512037)
In [2]: import math
        import numpy as np
        import pandas as pd
        import tensorflow as tf
        import matplotlib.pyplot as plt
        from tensorflow.keras import Model
        from tensorflow.keras import Sequential
        from tensorflow.keras.optimizers import Adam
        from sklearn.preprocessing import StandardScaler
        from tensorflow.keras.layers import Dense, Dropout
        from sklearn.model_selection import train_test_split
        from tensorflow.keras.losses import MeanSquaredLogarithmicError
In [3]: df = np.loadtxt('https://raw.githubusercontent.com/jbrownlee/Datasets/master/pim
In [4]: df
Out[4]: array([[ 6.
                       , 148.
                                  72.
                                                 0.627, 50.
                                                                   1.
                                                                        ],
                                        , ...,
                                                 0.351, 31. ,
                 1.
                     , 85. , 66.
                                                                   0.
               [
                                       , ...,
               [ 8. , 183. , 64.
                                                 0.672, 32.
                                                                  1.
                                                                       ],
                      , 121. , 72.
                                                 0.245, 30. , 0.
                                                                       ],
                 5.
                     , 126.
                                                 0.349, 47.
                                                             , 1.
               1.
                              , 60.
                                        , ...,
                                                                      ],
                                                 0.315, 23. , 0.
                       , 93. , 70.
               1.
                                                                      ]])
In [5]: df.shape
Out[5]: (768, 9)
In [6]: x = df[:,:8]
        y = df[:,8]
In [7]: from sklearn.model_selection import train_test_split
        X_train, X_temp, y_train, y_temp = train_test_split(x, y, test_size=0.2, random_
        X_test, X_val, y_test, y_val = train_test_split(X_temp, y_temp, test_size=0.5, r
In [8]: print(f"x train shape{X_train.shape}")
        print(f"y train shape{y_train.shape}")
        print(f"x test shape{X test.shape}")
        print(f"y test shape{y test.shape}")
        print(f"x val shape{X_val.shape}")
        print(f"y val shape{y_val.shape}")
      x train shape(614, 8)
      y train shape(614,)
      x test shape(77, 8)
      y test shape(77,)
      x val shape(77, 8)
      y val shape(77,)
In [9]: from sklearn.preprocessing import StandardScaler
        scaler = StandardScaler()
        X_train = scaler.fit_transform(X_train)
```

X_test = scaler.transform(X_test)

```
X_{val} = scaler.transform(X_{val})
        from collections import Counter
In [10]:
         Counter(y)
Out[10]: Counter({1.0: 268, 0.0: 500})
        import seaborn as sns
In [11]:
In [12]: # sns.countplot(y)
        from tensorflow.keras.models import Sequential
In [14]: model = Sequential([
         tf.keras.layers.InputLayer(8,),
         Dense(50, activation='relu'),
         Dense(50, activation='relu'),
         Dense(50,activation='relu'),
         Dense(50,activation='relu'),
         Dense(1,activation='sigmoid')
         ])
        2022-11-12 19:23:28.405404: I tensorflow/core/common_runtime/process_util.cc:146]
       Creating new thread pool with default inter op setting: 2. Tune using inter_op_pa
       rallelism_threads for best performance.
In [15]: model.summary()
       Model: "sequential"
                                                           Param #
       Layer (type)
                                   Output Shape
        ______
       dense (Dense)
                                   (None, 50)
                                                           450
       dense_1 (Dense)
                                   (None, 50)
                                                           2550
       dense 2 (Dense)
                                   (None, 50)
                                                           2550
       dense_3 (Dense)
                                   (None, 50)
                                                           2550
       dense_4 (Dense)
                                   (None, 1)
                                                           51
        ______
       Total params: 8,151
       Trainable params: 8,151
       Non-trainable params: 0
In [16]: opt = tf.keras.optimizers.Adam(learning_rate=0.0001)
         model.compile(loss='binary crossentropy', optimizer=opt, metrics=['accuracy'])
In [17]: history = model.fit(x=x,y=y,epochs=300, batch_size=50,validation_data=(X_val,y_v
        2022-11-12 19:23:28.666100: I tensorflow/compiler/mlir/mlir graph optimization pa
       ss.cc:185] None of the MLIR Optimization Passes are enabled (registered 2)
```

```
Epoch 1/300
16/16 [==============] - 1s 20ms/step - loss: 4.3687 - accuracy:
0.6484 - val_loss: 0.7085 - val_accuracy: 0.3766
Epoch 2/300
0.5820 - val_loss: 0.7094 - val_accuracy: 0.4156
Epoch 3/300
0.4987 - val_loss: 0.7087 - val_accuracy: 0.4286
Epoch 4/300
0.6276 - val loss: 0.7089 - val accuracy: 0.4286
Epoch 5/300
0.6341 - val_loss: 0.7092 - val_accuracy: 0.4156
Epoch 6/300
0.6406 - val_loss: 0.7088 - val_accuracy: 0.4156
Epoch 7/300
0.6471 - val_loss: 0.7083 - val_accuracy: 0.4026
Epoch 8/300
0.6562 - val_loss: 0.7080 - val_accuracy: 0.4156
Epoch 9/300
0.6641 - val_loss: 0.7077 - val_accuracy: 0.4156
Epoch 10/300
0.6706 - val loss: 0.7072 - val accuracy: 0.4156
Epoch 11/300
0.6667 - val_loss: 0.7070 - val_accuracy: 0.3896
Epoch 12/300
0.6849 - val_loss: 0.7066 - val_accuracy: 0.3896
Epoch 13/300
0.6823 - val_loss: 0.7061 - val_accuracy: 0.4026
Epoch 14/300
0.6888 - val loss: 0.7054 - val accuracy: 0.4026
Epoch 15/300
16/16 [============== ] - 0s 5ms/step - loss: 0.5979 - accuracy:
0.6823 - val_loss: 0.7049 - val_accuracy: 0.4156
Epoch 16/300
0.6901 - val loss: 0.7043 - val accuracy: 0.4156
Epoch 17/300
16/16 [============== ] - 0s 6ms/step - loss: 0.5915 - accuracy:
0.6927 - val_loss: 0.7039 - val_accuracy: 0.4156
Epoch 18/300
0.6797 - val_loss: 0.7033 - val_accuracy: 0.4416
Epoch 19/300
0.6888 - val_loss: 0.7029 - val_accuracy: 0.4416
Epoch 20/300
0.7044 - val_loss: 0.7021 - val_accuracy: 0.4675
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Epoch 21/300
0.7070 - val_loss: 0.7018 - val_accuracy: 0.4675
Epoch 22/300
0.7044 - val_loss: 0.7014 - val_accuracy: 0.4545
Epoch 23/300
0.7109 - val_loss: 0.7009 - val_accuracy: 0.4675
Epoch 24/300
0.7096 - val loss: 0.7004 - val accuracy: 0.4805
Epoch 25/300
0.7122 - val_loss: 0.7001 - val_accuracy: 0.4935
Epoch 26/300
0.7135 - val_loss: 0.6995 - val_accuracy: 0.5325
Epoch 27/300
0.7331 - val_loss: 0.6993 - val_accuracy: 0.5195
Epoch 28/300
0.6979 - val_loss: 0.6987 - val_accuracy: 0.5195
Epoch 29/300
0.7005 - val_loss: 0.6983 - val_accuracy: 0.5195
Epoch 30/300
0.7292 - val loss: 0.6980 - val accuracy: 0.5195
Epoch 31/300
0.7109 - val_loss: 0.6975 - val_accuracy: 0.5065
Epoch 32/300
0.7174 - val_loss: 0.6971 - val_accuracy: 0.5065
Epoch 33/300
0.7227 - val_loss: 0.6967 - val_accuracy: 0.4935
Epoch 34/300
0.7357 - val loss: 0.6963 - val accuracy: 0.5065
Epoch 35/300
16/16 [============== ] - 0s 5ms/step - loss: 0.5552 - accuracy:
0.7214 - val_loss: 0.6960 - val_accuracy: 0.4935
Epoch 36/300
0.7279 - val loss: 0.6955 - val accuracy: 0.5065
Epoch 37/300
16/16 [============== ] - 0s 5ms/step - loss: 0.5540 - accuracy:
0.7174 - val_loss: 0.6952 - val_accuracy: 0.5065
Epoch 38/300
0.7357 - val_loss: 0.6950 - val_accuracy: 0.4805
Epoch 39/300
0.7214 - val_loss: 0.6945 - val_accuracy: 0.4675
Epoch 40/300
0.7266 - val_loss: 0.6941 - val_accuracy: 0.4675
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Epoch 41/300
0.7201 - val_loss: 0.6935 - val_accuracy: 0.4675
Epoch 42/300
0.7331 - val_loss: 0.6930 - val_accuracy: 0.4675
Epoch 43/300
0.7214 - val_loss: 0.6927 - val_accuracy: 0.5065
Epoch 44/300
0.7227 - val loss: 0.6923 - val accuracy: 0.4935
Epoch 45/300
0.7318 - val_loss: 0.6919 - val_accuracy: 0.5065
Epoch 46/300
0.7135 - val_loss: 0.6917 - val_accuracy: 0.5065
Epoch 47/300
0.7292 - val_loss: 0.6909 - val_accuracy: 0.5325
Epoch 48/300
0.7344 - val_loss: 0.6908 - val_accuracy: 0.5325
Epoch 49/300
0.7266 - val_loss: 0.6904 - val_accuracy: 0.5455
Epoch 50/300
0.7279 - val loss: 0.6902 - val accuracy: 0.5455
Epoch 51/300
0.7305 - val_loss: 0.6898 - val_accuracy: 0.5455
Epoch 52/300
0.7344 - val_loss: 0.6893 - val_accuracy: 0.5455
Epoch 53/300
0.7318 - val_loss: 0.6887 - val_accuracy: 0.5455
Epoch 54/300
0.7487 - val loss: 0.6885 - val accuracy: 0.5584
Epoch 55/300
16/16 [============== ] - 0s 5ms/step - loss: 0.5327 - accuracy:
0.7305 - val_loss: 0.6882 - val_accuracy: 0.5584
Epoch 56/300
0.7422 - val loss: 0.6879 - val accuracy: 0.5584
Epoch 57/300
16/16 [============== ] - 0s 5ms/step - loss: 0.5343 - accuracy:
0.7331 - val_loss: 0.6875 - val_accuracy: 0.5584
Epoch 58/300
0.7370 - val_loss: 0.6869 - val_accuracy: 0.5714
Epoch 59/300
0.7409 - val_loss: 0.6862 - val_accuracy: 0.5584
Epoch 60/300
0.7461 - val_loss: 0.6859 - val_accuracy: 0.5714
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Epoch 61/300
0.7383 - val_loss: 0.6855 - val_accuracy: 0.5714
Epoch 62/300
0.7318 - val_loss: 0.6852 - val_accuracy: 0.5714
Epoch 63/300
0.7370 - val_loss: 0.6850 - val_accuracy: 0.5714
Epoch 64/300
0.7344 - val loss: 0.6847 - val accuracy: 0.5714
Epoch 65/300
0.7448 - val_loss: 0.6844 - val_accuracy: 0.5584
Epoch 66/300
0.7461 - val_loss: 0.6840 - val_accuracy: 0.5584
Epoch 67/300
0.7422 - val_loss: 0.6836 - val_accuracy: 0.5584
Epoch 68/300
0.7396 - val_loss: 0.6832 - val_accuracy: 0.5584
Epoch 69/300
0.7396 - val_loss: 0.6830 - val_accuracy: 0.5584
Epoch 70/300
0.7383 - val loss: 0.6825 - val accuracy: 0.5584
Epoch 71/300
0.7422 - val_loss: 0.6824 - val_accuracy: 0.5584
Epoch 72/300
0.7383 - val_loss: 0.6822 - val_accuracy: 0.5195
Epoch 73/300
0.7344 - val_loss: 0.6817 - val_accuracy: 0.5065
Epoch 74/300
0.7396 - val loss: 0.6814 - val accuracy: 0.5195
Epoch 75/300
16/16 [============== ] - 0s 5ms/step - loss: 0.5220 - accuracy:
0.7240 - val_loss: 0.6810 - val_accuracy: 0.5065
Epoch 76/300
0.7383 - val loss: 0.6804 - val accuracy: 0.5325
Epoch 77/300
0.7318 - val_loss: 0.6804 - val_accuracy: 0.5325
Epoch 78/300
0.7344 - val_loss: 0.6802 - val_accuracy: 0.5325
Epoch 79/300
0.7279 - val_loss: 0.6798 - val_accuracy: 0.5325
Epoch 80/300
0.7344 - val_loss: 0.6795 - val_accuracy: 0.5325
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Epoch 81/300
0.7526 - val_loss: 0.6792 - val_accuracy: 0.5584
Epoch 82/300
0.7565 - val_loss: 0.6788 - val_accuracy: 0.5584
Epoch 83/300
0.7344 - val_loss: 0.6785 - val_accuracy: 0.5584
Epoch 84/300
0.7448 - val loss: 0.6783 - val accuracy: 0.5714
Epoch 85/300
0.7448 - val_loss: 0.6779 - val_accuracy: 0.5844
Epoch 86/300
0.7227 - val_loss: 0.6779 - val_accuracy: 0.5844
Epoch 87/300
0.7370 - val_loss: 0.6775 - val_accuracy: 0.5844
Epoch 88/300
0.7357 - val_loss: 0.6771 - val_accuracy: 0.5844
Epoch 89/300
0.7422 - val_loss: 0.6769 - val_accuracy: 0.5844
Epoch 90/300
0.7357 - val loss: 0.6768 - val accuracy: 0.5844
Epoch 91/300
0.7552 - val_loss: 0.6766 - val_accuracy: 0.5844
Epoch 92/300
0.7344 - val_loss: 0.6764 - val_accuracy: 0.5844
Epoch 93/300
0.7487 - val_loss: 0.6762 - val_accuracy: 0.5974
Epoch 94/300
0.7370 - val loss: 0.6760 - val accuracy: 0.5974
Epoch 95/300
16/16 [============== ] - 0s 6ms/step - loss: 0.5042 - accuracy:
0.7487 - val_loss: 0.6758 - val_accuracy: 0.5974
Epoch 96/300
0.7461 - val loss: 0.6754 - val accuracy: 0.5974
Epoch 97/300
0.7474 - val_loss: 0.6752 - val_accuracy: 0.5974
Epoch 98/300
0.7474 - val_loss: 0.6752 - val_accuracy: 0.5974
Epoch 99/300
0.7461 - val_loss: 0.6750 - val_accuracy: 0.5974
Epoch 100/300
0.7474 - val_loss: 0.6749 - val_accuracy: 0.5974
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Epoch 101/300
16/16 [============== ] - 0s 5ms/step - loss: 0.5015 - accuracy:
0.7669 - val_loss: 0.6749 - val_accuracy: 0.6104
Epoch 102/300
0.7435 - val_loss: 0.6748 - val_accuracy: 0.6104
Epoch 103/300
0.7266 - val_loss: 0.6744 - val_accuracy: 0.6104
Epoch 104/300
0.7578 - val loss: 0.6742 - val accuracy: 0.6104
Epoch 105/300
0.7695 - val_loss: 0.6742 - val_accuracy: 0.6104
Epoch 106/300
0.7461 - val_loss: 0.6738 - val_accuracy: 0.6104
Epoch 107/300
0.7604 - val_loss: 0.6737 - val_accuracy: 0.6104
Epoch 108/300
0.7513 - val_loss: 0.6737 - val_accuracy: 0.6104
Epoch 109/300
0.7331 - val_loss: 0.6735 - val_accuracy: 0.6104
Epoch 110/300
0.7513 - val loss: 0.6731 - val accuracy: 0.6104
Epoch 111/300
0.7565 - val_loss: 0.6735 - val_accuracy: 0.6104
Epoch 112/300
0.7448 - val_loss: 0.6737 - val_accuracy: 0.6104
Epoch 113/300
0.7292 - val_loss: 0.6727 - val_accuracy: 0.6104
Epoch 114/300
0.7357 - val loss: 0.6725 - val accuracy: 0.6104
Epoch 115/300
16/16 [============== ] - 0s 5ms/step - loss: 0.4999 - accuracy:
0.7422 - val_loss: 0.6728 - val_accuracy: 0.6104
Epoch 116/300
0.7487 - val loss: 0.6727 - val accuracy: 0.6104
Epoch 117/300
16/16 [============== ] - 0s 5ms/step - loss: 0.5224 - accuracy:
0.7292 - val_loss: 0.6725 - val_accuracy: 0.6104
Epoch 118/300
0.7474 - val_loss: 0.6723 - val_accuracy: 0.6104
Epoch 119/300
0.7643 - val_loss: 0.6724 - val_accuracy: 0.6104
Epoch 120/300
0.7552 - val_loss: 0.6719 - val_accuracy: 0.6104
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Epoch 121/300
16/16 [============== ] - Os 6ms/step - loss: 0.4995 - accuracy:
0.7461 - val_loss: 0.6718 - val_accuracy: 0.6104
Epoch 122/300
0.7539 - val_loss: 0.6721 - val_accuracy: 0.6104
Epoch 123/300
16/16 [============= ] - 0s 5ms/step - loss: 0.4979 - accuracy:
0.7539 - val_loss: 0.6720 - val_accuracy: 0.6104
Epoch 124/300
0.7513 - val loss: 0.6719 - val accuracy: 0.6104
Epoch 125/300
0.7565 - val_loss: 0.6718 - val_accuracy: 0.6104
Epoch 126/300
0.7565 - val_loss: 0.6716 - val_accuracy: 0.6104
Epoch 127/300
0.7578 - val_loss: 0.6717 - val_accuracy: 0.6104
Epoch 128/300
0.7474 - val_loss: 0.6720 - val_accuracy: 0.6104
Epoch 129/300
0.7526 - val_loss: 0.6717 - val_accuracy: 0.6104
Epoch 130/300
0.7539 - val loss: 0.6713 - val accuracy: 0.6104
Epoch 131/300
16/16 [============== ] - Os 5ms/step - loss: 0.4905 - accuracy:
0.7604 - val_loss: 0.6712 - val_accuracy: 0.6104
Epoch 132/300
0.7591 - val_loss: 0.6712 - val_accuracy: 0.6104
Epoch 133/300
0.7461 - val_loss: 0.6713 - val_accuracy: 0.6104
Epoch 134/300
0.7604 - val loss: 0.6716 - val accuracy: 0.6104
Epoch 135/300
16/16 [============== ] - 0s 5ms/step - loss: 0.4909 - accuracy:
0.7552 - val_loss: 0.6714 - val_accuracy: 0.6104
Epoch 136/300
0.7552 - val loss: 0.6711 - val accuracy: 0.6104
Epoch 137/300
0.7448 - val_loss: 0.6715 - val_accuracy: 0.6104
Epoch 138/300
0.7669 - val_loss: 0.6713 - val_accuracy: 0.6104
Epoch 139/300
0.7578 - val_loss: 0.6715 - val_accuracy: 0.6104
Epoch 140/300
0.7578 - val_loss: 0.6714 - val_accuracy: 0.6104
```

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Epoch 141/300
0.7266 - val_loss: 0.6715 - val_accuracy: 0.6104
Epoch 142/300
0.7604 - val_loss: 0.6708 - val_accuracy: 0.6104
Epoch 143/300
16/16 [============= ] - 0s 5ms/step - loss: 0.4852 - accuracy:
0.7656 - val_loss: 0.6712 - val_accuracy: 0.6104
Epoch 144/300
0.7682 - val loss: 0.6710 - val accuracy: 0.6104
Epoch 145/300
0.7721 - val_loss: 0.6708 - val_accuracy: 0.6104
Epoch 146/300
0.7500 - val_loss: 0.6714 - val_accuracy: 0.6104
Epoch 147/300
0.7604 - val_loss: 0.6712 - val_accuracy: 0.6104
Epoch 148/300
0.7773 - val_loss: 0.6708 - val_accuracy: 0.6104
Epoch 149/300
0.7474 - val_loss: 0.6716 - val_accuracy: 0.6104
Epoch 150/300
0.7799 - val loss: 0.6715 - val accuracy: 0.6104
Epoch 151/300
0.7695 - val_loss: 0.6714 - val_accuracy: 0.6104
Epoch 152/300
0.7656 - val_loss: 0.6720 - val_accuracy: 0.6104
Epoch 153/300
0.7526 - val_loss: 0.6713 - val_accuracy: 0.6104
Epoch 154/300
0.7604 - val loss: 0.6716 - val accuracy: 0.6104
Epoch 155/300
16/16 [============== ] - 0s 5ms/step - loss: 0.4977 - accuracy:
0.7539 - val_loss: 0.6713 - val_accuracy: 0.6104
Epoch 156/300
0.7565 - val loss: 0.6711 - val accuracy: 0.6104
Epoch 157/300
0.7487 - val_loss: 0.6713 - val_accuracy: 0.6104
Epoch 158/300
0.7552 - val_loss: 0.6714 - val_accuracy: 0.6104
Epoch 159/300
0.7630 - val_loss: 0.6714 - val_accuracy: 0.6104
Epoch 160/300
0.7630 - val_loss: 0.6719 - val_accuracy: 0.6104
```

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Epoch 161/300
0.7747 - val_loss: 0.6723 - val_accuracy: 0.6104
Epoch 162/300
0.7773 - val_loss: 0.6721 - val_accuracy: 0.6104
Epoch 163/300
16/16 [============= ] - 0s 5ms/step - loss: 0.4768 - accuracy:
0.7760 - val_loss: 0.6720 - val_accuracy: 0.6104
Epoch 164/300
0.7409 - val loss: 0.6719 - val accuracy: 0.6104
Epoch 165/300
0.7799 - val_loss: 0.6720 - val_accuracy: 0.6104
Epoch 166/300
0.7656 - val_loss: 0.6721 - val_accuracy: 0.6104
Epoch 167/300
0.7682 - val_loss: 0.6729 - val_accuracy: 0.6104
Epoch 168/300
0.7578 - val_loss: 0.6719 - val_accuracy: 0.6104
Epoch 169/300
0.7734 - val_loss: 0.6726 - val_accuracy: 0.6104
Epoch 170/300
0.7578 - val loss: 0.6730 - val accuracy: 0.6104
Epoch 171/300
16/16 [============== ] - 0s 5ms/step - loss: 0.4750 - accuracy:
0.7786 - val_loss: 0.6728 - val_accuracy: 0.6104
Epoch 172/300
0.7878 - val_loss: 0.6727 - val_accuracy: 0.6104
Epoch 173/300
0.7526 - val_loss: 0.6728 - val_accuracy: 0.6104
Epoch 174/300
0.7812 - val loss: 0.6736 - val accuracy: 0.6104
Epoch 175/300
16/16 [============== ] - 0s 5ms/step - loss: 0.4859 - accuracy:
0.7721 - val_loss: 0.6734 - val_accuracy: 0.6104
Epoch 176/300
0.7591 - val loss: 0.6730 - val accuracy: 0.6104
Epoch 177/300
0.7682 - val_loss: 0.6730 - val_accuracy: 0.6104
Epoch 178/300
0.7799 - val_loss: 0.6738 - val_accuracy: 0.6104
Epoch 179/300
0.7721 - val_loss: 0.6737 - val_accuracy: 0.6104
Epoch 180/300
0.7773 - val_loss: 0.6740 - val_accuracy: 0.6104
```

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Epoch 181/300
0.7799 - val_loss: 0.6742 - val_accuracy: 0.6104
Epoch 182/300
0.7721 - val_loss: 0.6742 - val_accuracy: 0.6104
Epoch 183/300
16/16 [============= ] - 0s 13ms/step - loss: 0.4688 - accuracy:
0.7839 - val_loss: 0.6747 - val_accuracy: 0.6104
Epoch 184/300
16/16 [============ ] - 0s 10ms/step - loss: 0.4740 - accuracy:
0.7721 - val loss: 0.6749 - val accuracy: 0.6104
Epoch 185/300
16/16 [============] - 0s 11ms/step - loss: 0.4693 - accuracy:
0.7643 - val_loss: 0.6750 - val_accuracy: 0.6104
Epoch 186/300
0.7786 - val_loss: 0.6749 - val_accuracy: 0.6104
Epoch 187/300
0.7773 - val_loss: 0.6752 - val_accuracy: 0.6104
Epoch 188/300
0.7839 - val_loss: 0.6751 - val_accuracy: 0.6104
Epoch 189/300
0.7682 - val_loss: 0.6748 - val_accuracy: 0.6104
Epoch 190/300
0.7812 - val loss: 0.6753 - val accuracy: 0.6104
Epoch 191/300
0.7773 - val_loss: 0.6754 - val_accuracy: 0.6104
Epoch 192/300
0.7865 - val_loss: 0.6756 - val_accuracy: 0.6104
Epoch 193/300
0.7773 - val_loss: 0.6761 - val_accuracy: 0.6104
Epoch 194/300
0.7760 - val loss: 0.6765 - val accuracy: 0.6104
Epoch 195/300
16/16 [============== ] - 0s 6ms/step - loss: 0.4657 - accuracy:
0.7773 - val_loss: 0.6766 - val_accuracy: 0.6104
Epoch 196/300
0.7839 - val loss: 0.6761 - val accuracy: 0.6104
Epoch 197/300
0.7799 - val_loss: 0.6765 - val_accuracy: 0.6104
Epoch 198/300
0.7682 - val_loss: 0.6771 - val_accuracy: 0.6104
Epoch 199/300
0.7630 - val_loss: 0.6774 - val_accuracy: 0.6104
Epoch 200/300
0.7917 - val_loss: 0.6779 - val_accuracy: 0.6104
```

```
Epoch 201/300
16/16 [============== ] - Os 6ms/step - loss: 0.4694 - accuracy:
0.7839 - val_loss: 0.6778 - val_accuracy: 0.6104
Epoch 202/300
0.7917 - val_loss: 0.6774 - val_accuracy: 0.6104
Epoch 203/300
0.7799 - val_loss: 0.6773 - val_accuracy: 0.6104
Epoch 204/300
0.7734 - val loss: 0.6789 - val accuracy: 0.6104
Epoch 205/300
0.7500 - val_loss: 0.6793 - val_accuracy: 0.6104
Epoch 206/300
0.7643 - val_loss: 0.6792 - val_accuracy: 0.6104
Epoch 207/300
0.7578 - val_loss: 0.6783 - val_accuracy: 0.6104
Epoch 208/300
0.7734 - val_loss: 0.6786 - val_accuracy: 0.6104
Epoch 209/300
0.7773 - val_loss: 0.6783 - val_accuracy: 0.6104
Epoch 210/300
0.7734 - val loss: 0.6793 - val accuracy: 0.6104
Epoch 211/300
0.7669 - val_loss: 0.6795 - val_accuracy: 0.6104
Epoch 212/300
0.7852 - val_loss: 0.6808 - val_accuracy: 0.6104
Epoch 213/300
0.7682 - val_loss: 0.6803 - val_accuracy: 0.6104
Epoch 214/300
0.7878 - val loss: 0.6814 - val accuracy: 0.6104
Epoch 215/300
0.7630 - val_loss: 0.6807 - val_accuracy: 0.6104
Epoch 216/300
0.7891 - val loss: 0.6821 - val accuracy: 0.6104
Epoch 217/300
16/16 [============== ] - 0s 5ms/step - loss: 0.4693 - accuracy:
0.7708 - val_loss: 0.6820 - val_accuracy: 0.6104
Epoch 218/300
0.7812 - val_loss: 0.6822 - val_accuracy: 0.6104
Epoch 219/300
0.7917 - val_loss: 0.6819 - val_accuracy: 0.6104
Epoch 220/300
0.7839 - val_loss: 0.6822 - val_accuracy: 0.6104
```

```
Epoch 221/300
16/16 [============== ] - 0s 5ms/step - loss: 0.4633 - accuracy:
0.7760 - val_loss: 0.6825 - val_accuracy: 0.6104
Epoch 222/300
0.7656 - val_loss: 0.6828 - val_accuracy: 0.6104
Epoch 223/300
0.7917 - val_loss: 0.6832 - val_accuracy: 0.6104
Epoch 224/300
0.7604 - val loss: 0.6846 - val accuracy: 0.6104
Epoch 225/300
0.7812 - val_loss: 0.6840 - val_accuracy: 0.6104
Epoch 226/300
0.7539 - val_loss: 0.6838 - val_accuracy: 0.6104
Epoch 227/300
0.7891 - val_loss: 0.6840 - val_accuracy: 0.6104
Epoch 228/300
0.7852 - val_loss: 0.6843 - val_accuracy: 0.6104
Epoch 229/300
0.7773 - val_loss: 0.6844 - val_accuracy: 0.6104
Epoch 230/300
0.7839 - val loss: 0.6847 - val accuracy: 0.6104
Epoch 231/300
0.7773 - val_loss: 0.6855 - val_accuracy: 0.6104
Epoch 232/300
0.7773 - val_loss: 0.6855 - val_accuracy: 0.6104
Epoch 233/300
0.7839 - val_loss: 0.6862 - val_accuracy: 0.6104
Epoch 234/300
0.7826 - val loss: 0.6859 - val accuracy: 0.6104
Epoch 235/300
16/16 [============== ] - 0s 5ms/step - loss: 0.4556 - accuracy:
0.7826 - val_loss: 0.6872 - val_accuracy: 0.6104
Epoch 236/300
0.7865 - val loss: 0.6871 - val accuracy: 0.6104
Epoch 237/300
16/16 [============== ] - 0s 5ms/step - loss: 0.4797 - accuracy:
0.7643 - val_loss: 0.6866 - val_accuracy: 0.6104
Epoch 238/300
0.7604 - val_loss: 0.6869 - val_accuracy: 0.6104
Epoch 239/300
0.7526 - val_loss: 0.6863 - val_accuracy: 0.6104
Epoch 240/300
0.7812 - val_loss: 0.6872 - val_accuracy: 0.6104
```

```
Epoch 241/300
16/16 [============== ] - 0s 5ms/step - loss: 0.4634 - accuracy:
0.7852 - val_loss: 0.6871 - val_accuracy: 0.6104
Epoch 242/300
0.7786 - val_loss: 0.6880 - val_accuracy: 0.6104
Epoch 243/300
0.7865 - val_loss: 0.6890 - val_accuracy: 0.6104
Epoch 244/300
0.7917 - val loss: 0.6888 - val accuracy: 0.6104
Epoch 245/300
0.7943 - val_loss: 0.6892 - val_accuracy: 0.6104
Epoch 246/300
0.7878 - val_loss: 0.6896 - val_accuracy: 0.6104
Epoch 247/300
0.7799 - val_loss: 0.6896 - val_accuracy: 0.6104
Epoch 248/300
0.7943 - val_loss: 0.6904 - val_accuracy: 0.6104
Epoch 249/300
0.7865 - val_loss: 0.6902 - val_accuracy: 0.6104
Epoch 250/300
0.7799 - val loss: 0.6917 - val accuracy: 0.6104
Epoch 251/300
0.7891 - val_loss: 0.6909 - val_accuracy: 0.6104
Epoch 252/300
0.7839 - val_loss: 0.6914 - val_accuracy: 0.6104
Epoch 253/300
0.7917 - val_loss: 0.6918 - val_accuracy: 0.6104
Epoch 254/300
0.7878 - val loss: 0.6923 - val accuracy: 0.6104
Epoch 255/300
16/16 [============== ] - 0s 5ms/step - loss: 0.4424 - accuracy:
0.7956 - val_loss: 0.6928 - val_accuracy: 0.6104
Epoch 256/300
0.7917 - val loss: 0.6935 - val accuracy: 0.6104
Epoch 257/300
16/16 [============== ] - 0s 6ms/step - loss: 0.4498 - accuracy:
0.7891 - val_loss: 0.6942 - val_accuracy: 0.6104
Epoch 258/300
0.7865 - val_loss: 0.6938 - val_accuracy: 0.6104
Epoch 259/300
0.7826 - val_loss: 0.6942 - val_accuracy: 0.6104
Epoch 260/300
0.7773 - val_loss: 0.6949 - val_accuracy: 0.6104
```

```
Epoch 261/300
0.7904 - val_loss: 0.6956 - val_accuracy: 0.6104
Epoch 262/300
0.7826 - val_loss: 0.6963 - val_accuracy: 0.6104
Epoch 263/300
16/16 [============= ] - 0s 5ms/step - loss: 0.4478 - accuracy:
0.7799 - val_loss: 0.6959 - val_accuracy: 0.6104
Epoch 264/300
0.7799 - val loss: 0.6970 - val accuracy: 0.6104
Epoch 265/300
0.7865 - val_loss: 0.6968 - val_accuracy: 0.6104
Epoch 266/300
0.7982 - val_loss: 0.6969 - val_accuracy: 0.6104
Epoch 267/300
0.7969 - val_loss: 0.6972 - val_accuracy: 0.6104
Epoch 268/300
0.7826 - val_loss: 0.6974 - val_accuracy: 0.6104
Epoch 269/300
0.7904 - val_loss: 0.6990 - val_accuracy: 0.6104
Epoch 270/300
0.7878 - val loss: 0.6991 - val accuracy: 0.6104
Epoch 271/300
0.7904 - val_loss: 0.6997 - val_accuracy: 0.6104
Epoch 272/300
0.7956 - val_loss: 0.6999 - val_accuracy: 0.6104
Epoch 273/300
0.7865 - val_loss: 0.7003 - val_accuracy: 0.6104
Epoch 274/300
0.8021 - val loss: 0.7012 - val accuracy: 0.6104
Epoch 275/300
16/16 [============== ] - 0s 5ms/step - loss: 0.4388 - accuracy:
0.8034 - val_loss: 0.7021 - val_accuracy: 0.6104
Epoch 276/300
0.7721 - val loss: 0.7025 - val accuracy: 0.6104
Epoch 277/300
16/16 [============== ] - 0s 5ms/step - loss: 0.4388 - accuracy:
0.7865 - val_loss: 0.7020 - val_accuracy: 0.6104
Epoch 278/300
0.7904 - val_loss: 0.7037 - val_accuracy: 0.6104
Epoch 279/300
0.7852 - val_loss: 0.7032 - val_accuracy: 0.6104
Epoch 280/300
0.7982 - val_loss: 0.7040 - val_accuracy: 0.6104
```

```
Epoch 281/300
16/16 [============== ] - 0s 5ms/step - loss: 0.4400 - accuracy:
0.7878 - val_loss: 0.7038 - val_accuracy: 0.6104
Epoch 282/300
0.7917 - val_loss: 0.7044 - val_accuracy: 0.6104
Epoch 283/300
16/16 [============= ] - 0s 5ms/step - loss: 0.4441 - accuracy:
0.7839 - val_loss: 0.7065 - val_accuracy: 0.6104
Epoch 284/300
0.7786 - val loss: 0.7043 - val accuracy: 0.6104
Epoch 285/300
0.7969 - val_loss: 0.7052 - val_accuracy: 0.6104
Epoch 286/300
0.7812 - val_loss: 0.7061 - val_accuracy: 0.6104
Epoch 287/300
0.7799 - val_loss: 0.7075 - val_accuracy: 0.6104
Epoch 288/300
0.7930 - val_loss: 0.7064 - val_accuracy: 0.6104
Epoch 289/300
0.7891 - val_loss: 0.7066 - val_accuracy: 0.6104
Epoch 290/300
0.7865 - val loss: 0.7081 - val accuracy: 0.6104
Epoch 291/300
0.7969 - val_loss: 0.7075 - val_accuracy: 0.6104
Epoch 292/300
0.7865 - val_loss: 0.7088 - val_accuracy: 0.6104
Epoch 293/300
0.7995 - val_loss: 0.7098 - val_accuracy: 0.6104
Epoch 294/300
0.7995 - val loss: 0.7106 - val accuracy: 0.6104
Epoch 295/300
16/16 [============== ] - 0s 5ms/step - loss: 0.4384 - accuracy:
0.7865 - val_loss: 0.7120 - val_accuracy: 0.6104
Epoch 296/300
0.7695 - val loss: 0.7113 - val accuracy: 0.6104
Epoch 297/300
0.7852 - val_loss: 0.7110 - val_accuracy: 0.6104
Epoch 298/300
0.8021 - val_loss: 0.7113 - val_accuracy: 0.6104
Epoch 299/300
0.7891 - val_loss: 0.7128 - val_accuracy: 0.6104
Epoch 300/300
0.8034 - val_loss: 0.7133 - val_accuracy: 0.6104
```

```
losses = pd.DataFrame(model.history.history)
In [18]:
          losses.plot()
Out[18]: <AxesSubplot:>
        4.5
                                                  loss
        4.0
                                                   accuracy
                                                  val_loss
        3.5
                                                   val accuracy
        3.0
        2.5
        2.0
        1.5
        1.0
        0.5
                            100
                                   150
                                           200
                     50
                                                  250
              0
                                                          300
In [19]: model.evaluate(x,y)
                               =========] - 0s 2ms/step - loss: 0.4277 - accuracy:
        24/24 [=======
        0.8099
Out[19]: [0.42774689197540283, 0.8098958134651184]
```

In [20]: y_pred = model.predict(X_test)

In [21]: y_pred

```
Out[21]: array([[0.29241186],
                 [0.25659525],
                 [0.2345556],
                 [0.25635856],
                 [0.2683779],
                 [0.26534295],
                 [0.24615443],
                 [0.18696874],
                 [0.23531413],
                 [0.26499653],
                 [0.25147843],
                 [0.19693151],
                 [0.2525043],
                 [0.2924193],
                 [0.24108097],
                 [0.24926525],
                 [0.24154398],
                 [0.30027997],
                 [0.26890767],
                 [0.18842933],
                 [0.27719277],
                 [0.23296693],
                 [0.25262323],
                 [0.23384169],
                 [0.26180875],
                 [0.2531795],
                 [0.21080497],
                 [0.2493422],
                 [0.27968198],
                 [0.22347766],
                 [0.2744137],
                 [0.22154438],
                 [0.24633941],
                 [0.26000386],
                 [0.30314064],
                 [0.1814619],
                 [0.17811224],
                 [0.22696063],
                 [0.23095623],
                 [0.24577609],
                 [0.21299466],
                 [0.25943238],
                 [0.26748422],
                 [0.28173897],
                 [0.21895227],
                 [0.25581098],
                 [0.2772973],
                 [0.26322144],
                 [0.27639455],
                 [0.2651384],
                 [0.29406345],
                 [0.19382304],
                 [0.24121556],
                 [0.2256059],
                 [0.2787133],
                 [0.28141072],
                 [0.22622645],
                 [0.30917984],
                  [0.23824352],
                 [0.25481486],
```

```
[0.25906867],
[0.2787506],
[0.20846003],
[0.2344867],
[0.27582753],
[0.24831259],
[0.29489657],
[0.26952785],
[0.26609173],
[0.26449886],
[0.3253206],
[0.20729351],
[0.27737445],
[0.27037466],
[0.2857654],
[0.21922398],
[0.2420043 ]], dtype=float32)
```