```
In [2]: import os
       import matplotlib.pyplot as plt
       import tensorflow as tf
       import pandas as pd
       import numpy as np
       print(tf. version )
       2.8.0
In [3]:
       TRAIN DATA URL = "https://storage.googleapis.com/tf-datasets/titanic/train.csv"
       TEST DATA URL = "https://storage.googleapis.com/tf-datasets/titanic/eval.csv"
       train file path = tf.keras.utils.get file("train.csv", TRAIN DATA URL)
       test file path = tf.keras.utils.get file("eval.csv", TEST DATA URL)
       column names = ['survived', 'sex', 'age', 'n siblings spouses', 'parch', 'fare',
             'class', 'deck', 'embark town', 'alone']
       df = pd.read csv(TRAIN DATA URL)
       # pd.set option('display.max rows', None)
       print(df)
       # 데이터 정보 확인
       print(df.info())
       # 수치형 데이터 확인
       print(df.describe())
       # 범주형 데이터 확인
       df.describe(include = np.object )
       # 결측치 확인
       df.isnull().sum()
          survived sex age n siblings spouses parch fare class \
                                              1 0 7.2500 Third
             0 male 22.0
       0
                1 female 38.0
                                              1
                                                    0 71.2833 First
                1 female 26.0
                                                    0 7.9250 Third
                                               0
                                                    0 53.1000 First
                1 female 35.0
                                               1
                                              0
                                                    0 8.4583 Third
                0 male 28.0
                     . . .
                                                   . . .
                                                         ... ...
                                             . . .
                                              0 0 10.5000 Second
0 0 7.0500 Third
0 0 30.0000 First
       622
                0 male 28.0
       623
                0 male 25.0
       624
                1 female 19.0
                0 female 28.0
                                              1
                                                    2 23.4500 Third
       625
                                               0 0 7.7500 Third
                0 male 32.0
       626
             deck embark town alone
          unknown Southampton n
       0
           C Cherbourg
       1
       2 unknown Southampton
                                У
       3
          C Southampton
                                n
          unknown Queenstown
                                У
       622 unknown Southampton
                                У
       623 unknown Southampton
                                У
       624
            B Southampton
       625 unknown Southampton
       626 unknown Queenstown
       [627 rows x 10 columns]
       <class 'pandas.core.frame.DataFrame'>
       RangeIndex: 627 entries, 0 to 626
       Data columns (total 10 columns):
        # Column Non-Null Count Dtype
       --- -----
                            _____
        0 survived
                           627 non-null int64
          sex
                            627 non-null object
```

```
2
                                 627 non-null
                                               float64
             age
          3
            n siblings spouses 627 non-null
                                               int64
                        627 non-null int64
            parch
                               627 non-null float64
627 non-null object
          5 fare
            class
          6
                               627 non-null object
          7
            deck
                     town 627 non-null object 627 non-null object
            embark town
          9
             alone
         dtypes: float64(2), int64(3), object(5)
         memory usage: 49.1+ KB
         None
                survived
                                age n siblings spouses parch
                                                                           fare
         count 627.000000 627.000000 627.000000 627.000000 627.000000
        mean 0.387560 29.631308
                                               0.545455 0.379585 34.385399
                0.487582 12.511818
                                               1.151090 0.792999 54.597730
         std
                                               0.000000 0.000000 0.000000
               0.000000 0.750000
0.000000 23.000000
        min
         25%
                                               0.000000 0.000000 7.895800
                                               0.000000 0.000000 15.045800
         50%
                0.000000 28.000000
                                               1.000000 0.000000 31.387500
8.000000 5.000000 512.329200
                1.000000 35.000000
         75%
        75% 1.000000 35.000000 max 1.000000 80.000000
        survived
                            0
Out[3]:
                              \cap
         sex
         age
         n siblings spouses
         parch
                              0
         fare
         class
                              0
         deck
                              0
                            0
         embark town
         alone
        dtype: int64
In [4]: # 데이터 요약
         print("전체 데이터 수:", df.shape[0] * df.shape[1])
         print(f"결측치 수: {df.isnull().sum().sum()}")
         print("총 인원 수:", df["age"].count())
         print("중복된 데이터:",df.duplicated().sum())
         전체 데이터 수: 6270
         결측치 수: 0
         총 인원 수: 627
         중복된 데이터: 69
In [16]: class names = ['UnSurvived', 'Survived']
         feature names = column names[1:]
         label name = column names[0]
         print(feature names)
         print(label name)
         batch size = 32
         train dataset = tf.data.experimental.make csv dataset(
             train file path,
            batch size,
            column names=column names,
             label name=label name,
            num epochs=1)
         test dataset = tf.data.experimental.make csv dataset(
             train file path,
            batch size,
            column names=column names,
             label name=label name,
             num epochs=1)
```

```
features, labels = next(iter(train dataset))
print(f"##features \n {features}")
['sex', 'age', 'n siblings spouses', 'parch', 'fare', 'class', 'deck', 'embark town', 'a
survived
##features
OrderedDict([('sex', <tf.Tensor: shape=(32,), dtype=string, numpy=
array([b'male', b'female', b'male', b'male', b'male', b'male',
      b'female', b'male', b'male', b'male', b'male', b'male',
      b'male', b'female', b'male', b'male', b'female', b'male',
      b'male', b'male', b'male', b'male', b'male', b'female', b'male',
      b'female', b'male', b'female'], dtype=object)>), ('age', <tf.Tensor: sha
pe=(32,), dtype=float32, numpy=
array([18., 45., 30.5, 28., 19., 22., 28., 2., 28., 24., 28.,
      23. , 15. , 48. , 31. , 26. , 41. , 32. , 28. , 23. , 20. , 35. ,
      46., 28., 31., 26., 49., 29., 32., 71., 28., 52.],
     dtype=float32)>), ('n siblings spouses', <tf.Tensor: shape=(32,), dtype=int32, num
ру=
array([0, 1, 0, 0, 0, 1, 1, 1, 0, 0, 0, 1, 1, 0, 0, 2, 0, 0, 0, 0, 0,
      0, 0, 1, 0, 0, 1, 0, 0, 0, 1], dtype=int32)>), ('parch', <tf.Tensor: shape=(32,),
dtype=int32, numpy=
0, 0, 0, 0, 0, 0, 0, 0, 0, 1], dtype=int32)>), ('fare', <tf.Tensor: shape=(32,),
dtype=float32, numpy=
array([ 8.3 , 164.8667, 8.05 , 56.4958, 8.05 ,
                                                     7.25 ,
                                  8.05 , 35.5 , 13.
       15.5
             , 26. , 13.
                     , 7.775 , 78.85 , 14.1083, 7.75 ,
        7.2292, 52.
                         8.05 ,
       7.75 , 7.55 ,
                                  7.05 , 79.2 , 10.5
             , 18.7875, 25.9292, 27.7208, 13. , 34.6542,
        6.95 , 93.5 ], dtype=float32)>), ('class', <tf.Tensor: shape=(32,), dtype=s
tring, numpy=
array([b'Third', b'First', b'Third', b'Third', b'Third', b'Third',
      b'Third', b'Second', b'Second', b'Third', b'First', b'Second',
      b'Third', b'First', b'Third', b'First', b'Third',
      b'Third', b'Third', b'Third', b'First', b'Second',
      b'First', b'Third', b'First', b'Second', b'Second', b'First',
      b'Third', b'First'], dtype=object)>), ('deck', <tf.Tensor: shape=(32,), dtype=str
ing, numpy=
array([b'unknown', b'unknown', b'unknown', b'unknown',
      b'unknown', b'unknown', b'unknown', b'unknown', b'C',
      b'unknown', b'unknown', b'C', b'unknown', b'unknown',
      b'unknown', b'unknown', b'unknown', b'unknown', b'B',
      b'unknown', b'B', b'unknown', b'D', b'unknown', b'unknown', b'A',
      b'unknown', b'B'], dtype=object)>), ('embark town', <tf.Tensor: shape=(32,), dtyp
e=string, numpy=
array([b'Southampton', b'Southampton', b'Southampton', b'Southampton',
      b'Southampton', b'Southampton', b'Queenstown', b'Southampton',
      b'Southampton', b'Southampton', b'Southampton', b'Southampton',
      b'Cherbourg', b'Southampton', b'Southampton', b'Southampton',
      b'Southampton', b'Queenstown', b'Queenstown', b'Southampton',
      b'Southampton', b'Southampton', b'Cherbourg', b'Southampton',
      b'Southampton', b'Cherbourg', b'Southampton', b'Cherbourg',
      b'Southampton', b'Cherbourg', b'Queenstown', b'Southampton'],
     dtype=object)>), ('alone', <tf.Tensor: shape=(32,), dtype=string, numpy=
array([b'y', b'n', b'y', b'y', b'y', b'n', b'n', b'n', b'y', b'y', b'y',
      b'y', b'n', b'n', b'y', b'y', b'n', b'y', b'y', b'y', b'y', b'y',
      b'y', b'y', b'n', b'y', b'y', b'n', b'y', b'y', b'y', b'n'],
     dtype=object)>)])
def remove columns(features, labels):
```

In [17]: # 불필요한 컬럼 삭제 -> embark town del (features['embark town']) return features, labels

```
test dataset = test dataset.map(remove columns)
         features, labels = next(iter(train dataset))
         print(f"features \n {features}")
         features
         OrderedDict([('sex', <tf.Tensor: shape=(32,), dtype=string, numpy=
         array([b'male', b'male', b'male', b'male', b'male', b'female', b'female',
                b'male', b'male', b'male', b'female', b'male', b'female', b'male',
                b'male', b'male', b'female', b'male', b'female',
                b'male', b'male', b'male', b'male', b'male', b'male',
                b'female', b'female', b'male', b'male'], dtype=object)>), ('age', <tf.
         Tensor: shape=(32,), dtype=float32, numpy=
         array([16., 61., 50., 22., 56., 11., 15., 17., 39., 28., 19.,
                32. , 42. , 22. , 34. , 21. , 28. , 28. , 17. , 18. , 20. , 19. ,
                28.5, 28., 33., 25., 44., 28., 14., 14., 36., 28.],
               dtype=float32)>), ('n siblings spouses', <tf.Tensor: shape=(32,), dtype=int32, num
         array([4, 0, 1, 1, 0, 4, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 2, 0, 0,
                0, 0, 0, 1, 0, 1, 1, 1, 0, 0], dtype=int32)>), ('parch', <tf.Tensor: shape=(32,),
         dtype=int32, numpy=
         array([1, 0, 0, 0, 0, 2, 0, 2, 0, 0, 2, 0, 0, 0, 0, 0, 0, 0, 0, 2, 0, 0,
                0, 0, 0, 0, 1, 2, 0, 0, 1, 0], dtype=int32)>), ('fare', <tf.Tensor: shape=(32,),
         dtype=float32, numpy=
         array([ 39.6875, 6.2375, 55.9 , 7.25 , 26.55 , 31.275 ,

    14.4542, 110.8833, 26.
    , 7.25 , 26.2833, 7.75 ,

    13.
    , 9.35 , 13.
    , 73.5 , 15.5 ,
    8.4583,

                108.9 , 262.375 , 9.8458, 8.05 , 7.2292, 7.25 ,
                 5. , 26. , 16.1 , 23.45 , 11.2417, 30.0708,
                512.3292, 7.55 ], dtype=float32)>), ('class', <tf.Tensor: shape=(32,), dtype=s
         tring, numpy=
         array([b'Third', b'Third', b'First', b'Third', b'First', b'Third',
                b'Third', b'First', b'Second', b'Third', b'First', b'Third',
                b'Second', b'Third', b'Second', b'Second', b'Third', b'Third',
                b'First', b'First', b'Third', b'Third', b'Third',
                b'First', b'Second', b'Third', b'Third', b'Third', b'Second',
                b'First', b'Third'], dtype=object)>), ('deck', <tf.Tensor: shape=(32,), dtype=str
         ing, numpy=
         array([b'unknown', b'unknown', b'E', b'unknown', b'unknown', b'unknown',
                b'unknown', b'C', b'unknown', b'unknown', b'D', b'unknown',
                b'unknown', b'unknown', b'unknown', b'unknown',
                b'unknown', b'C', b'B', b'unknown', b'unknown', b'unknown',
                b'unknown', b'B', b'unknown', b'unknown', b'unknown',
                b'unknown', b'B', b'unknown'], dtype=object)>), ('alone', <tf.Tensor: shape=(3
         2,), dtype=string, numpy=
         array([b'n', b'y', b'n', b'n', b'y', b'n', b'n', b'n', b'y', b'y', b'n',
                b'y', b'y', b'y', b'y', b'y', b'n', b'y', b'n', b'n', b'n', b'y', b'y',
               b'y', b'y', b'y', b'n', b'n', b'n', b'n', b'n', b'n', b'y'],
               dtype=object)>)])
In [18]: # 문자열 처리 -> sex, deck, alone, class를 문자열로
         def convert to int(feature, label):
             if feature['sex'] == 'male':
                 feature ['sex'] = 0.
             else:
                 feature['sex'] = 1.
             feature['sex'] = tf.cast(feature['sex'], tf.float32)
             return feature, label
         CAT COLUMNS = ['sex', 'deck', 'alone', 'class']
         NUM COLUMNS = ['age', 'fare', 'n siblings spouses', 'parch']
```

train dataset = train dataset.map(remove columns)

```
feature_cols = []

# Create IndicatorColumn for categorical features
for feature in CAT_COLUMNS:
    vocab = df[feature].unique()
    feature_cols.append(tf.feature_column.indicator_column(
        tf.feature_column.categorical_column_with_vocabulary_list(feature, vocab)))

# Create NumericColumn for numerical features
for feature in NUM_COLUMNS:
    feature_cols.append(tf.feature_column.numeric_column(feature, dtype=tf.float32))

print(feature_cols)
```

[IndicatorColumn(categorical_column=VocabularyListCategoricalColumn(key='sex', vocabular y_list=('male', 'female'), dtype=tf.string, default_value=-1, num_oov_buckets=0)), Indic atorColumn(categorical_column=VocabularyListCategoricalColumn(key='deck', vocabulary_list=('unknown', 'C', 'G', 'A', 'B', 'D', 'F', 'E'), dtype=tf.string, default_value=-1, num_oov_buckets=0)), IndicatorColumn(categorical_column=VocabularyListCategoricalColumn(key='alone', vocabulary_list=('n', 'y'), dtype=tf.string, default_value=-1, num_oov_buckets=0)), IndicatorColumn(categorical_column=VocabularyListCategoricalColumn(key='class', vocabulary_list=('Third', 'First', 'Second'), dtype=tf.string, default_value=-1, num_oov_buckets=0)), NumericColumn(key='age', shape=(1,), default_value=None, dtype=tf.float32, normalizer_fn=None), NumericColumn(key='fare', shape=(1,), default_value=None, dtype=tf.float32, normalizer_fn=None), NumericColumn(key='n_siblings_spouses', shape=(1,), default_value=None, dtype=tf.float32, normalizer_fn=None), NumericColumn(key='parch', shape=(1,), default_value=None, dtype=tf.float32, normalizer_fn=None)]

```
In [15]: # 모델 생성
model = tf.keras.Sequential()
model.add(tf.keras.layers.DenseFeatures(feature_cols))
model.add(tf.keras.layers.Dense(10, activation='relu'))
model.add(tf.keras.layers.Dense(10, activation='relu'))
model.add(tf.keras.layers.Dense(10, activation='relu'))
model.add(tf.keras.layers.Dense(1, activation='relu'))
model.add(tf.keras.layers.Dense(1, activation='sigmoid'))

model.compile(optimizer='sgd', loss='binary_crossentropy', metrics=['binary_accuracy'])
model.fit(train_dataset, epochs=100)
```

Epoch 1/1000

WARNING:tensorflow:Layers in a Sequential model should only have a single input tensor. Received: inputs=OrderedDict([('sex', <tf.Tensor 'IteratorGetNext:7' shape=(None,) dtype=string>), ('age', <tf.Tensor 'IteratorGetNext:0' shape=(None,) dtype=float32>), ('n_sib lings_spouses', <tf.Tensor 'IteratorGetNext:5' shape=(None,) dtype=int32>), ('parch', <tf.Tensor 'IteratorGetNext:6' shape=(None,) dtype=int32>), ('fare', <tf.Tensor 'IteratorGetNext:2' shape=(None,) dtype=string>), ('deck', <tf.Tensor 'IteratorGetNext:3' shape=(None,) dtype=string>), ('deck', <tf.Tensor 'IteratorGetNext:3' shape=(None,) dtype=string>), ('alone', <tf.Tensor 'IteratorGetNext:1' shape=(None,) dtype=string>)]). Consider rewriting this model with the Functional API.

WARNING:tensorflow:Layers in a Sequential model should only have a single input tensor. Received: inputs=OrderedDict([('sex', <tf.Tensor 'IteratorGetNext:7' shape=(None,) dtype=string>), ('age', <tf.Tensor 'IteratorGetNext:0' shape=(None,) dtype=float32>), ('n_sib lings_spouses', <tf.Tensor 'IteratorGetNext:5' shape=(None,) dtype=int32>), ('parch', <tf.Tensor 'IteratorGetNext:6' shape=(None,) dtype=int32>), ('fare', <tf.Tensor 'IteratorGetNext:2' shape=(None,) dtype=string>), ('deck', <tf.Tensor 'IteratorGetNext:3' shape=(None,) dtype=string>), ('deck', <tf.Tensor 'IteratorGetNext:3' shape=(None,) dtype=string>), ('alone', <tf.Tensor 'IteratorGetNext:1' shape=(None,) dtype=string>)]). Consider rewriting this model with the Functional API.

```
0.6730
Epoch 3/1000
0.6715
Epoch 4/1000
Epoch 5/1000
0.6842
Epoch 6/1000
0.6810
Epoch 7/1000
0.7002
Epoch 8/1000
0.6906
Epoch 9/1000
0.6746
Epoch 10/1000
0.7049
Epoch 11/1000
Epoch 12/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.5904 - binary accuracy:
0.6986
Epoch 13/1000
0.6938
Epoch 14/1000
0.7097
Epoch 15/1000
0.7145
Epoch 16/1000
0.6986
Epoch 17/1000
0.7129
Epoch 18/1000
0.7193
Epoch 19/1000
0.7113
Epoch 20/1000
0.6986
Epoch 21/1000
Epoch 22/1000
0.7209
Epoch 23/1000
20/20 [================= ] - Os 10ms/step - loss: 0.5798 - binary accuracy:
0.7209
Epoch 24/1000
```

```
0.7049
Epoch 25/1000
20/20 [============= ] - 0s 9ms/step - loss: 0.5809 - binary accuracy:
0.7241
Epoch 26/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5748 - binary accuracy:
Epoch 27/1000
0.7113
Epoch 28/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5829 - binary accuracy:
0.7129
Epoch 29/1000
20/20 [================== ] - Os 10ms/step - loss: 0.5744 - binary accuracy:
0.7145
Epoch 30/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5691 - binary accuracy:
0.7289
Epoch 31/1000
0.7081
Epoch 32/1000
0.7049
Epoch 33/1000
Epoch 34/1000
0.7273
Epoch 35/1000
0.7305
Epoch 36/1000
0.7161
Epoch 37/1000
0.7384
Epoch 38/1000
0.7129
Epoch 39/1000
0.7321
Epoch 40/1000
0.7384
Epoch 41/1000
0.7018
Epoch 42/1000
0.7033
Epoch 43/1000
Epoch 44/1000
0.7400
Epoch 45/1000
0.7225
Epoch 46/1000
```

```
0.7352
Epoch 47/1000
20/20 [============= ] - 0s 8ms/step - loss: 0.5693 - binary accuracy:
0.7400
Epoch 48/1000
20/20 [=============== ] - 0s 11ms/step - loss: 0.5583 - binary accuracy:
Epoch 49/1000
0.7480
Epoch 50/1000
0.6683
Epoch 51/1000
0.7193
Epoch 52/1000
0.7432
Epoch 53/1000
0.7416
Epoch 54/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.5597 - binary accuracy:
0.7464
Epoch 55/1000
Epoch 56/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5388 - binary accuracy:
0.7544
Epoch 57/1000
0.7352
Epoch 58/1000
0.7560
Epoch 59/1000
0.7273
Epoch 60/1000
0.7161
Epoch 61/1000
0.7400
Epoch 62/1000
0.7448
Epoch 63/1000
0.7384
Epoch 64/1000
0.7273
Epoch 65/1000
Epoch 66/1000
0.7337
Epoch 67/1000
0.7432
Epoch 68/1000
```

```
0.7432
Epoch 69/1000
0.7496
Epoch 70/1000
Epoch 71/1000
0.6730
Epoch 72/1000
0.7321
Epoch 73/1000
0.7560
Epoch 74/1000
0.6810
Epoch 75/1000
20/20 [================== ] - 0s 11ms/step - loss: 0.5643 - binary accuracy:
0.7305
Epoch 76/1000
0.7671
Epoch 77/1000
Epoch 78/1000
0.7592
Epoch 79/1000
0.7432
Epoch 80/1000
0.7049
Epoch 81/1000
0.7799
Epoch 82/1000
0.7496
Epoch 83/1000
0.7480
Epoch 84/1000
0.7671
Epoch 85/1000
0.7352
Epoch 86/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.5597 - binary accuracy:
0.7049
Epoch 87/1000
Epoch 88/1000
20/20 [================== ] - 0s 10ms/step - loss: 0.5408 - binary accuracy:
0.7368
Epoch 89/1000
0.7384
Epoch 90/1000
```

```
0.7847
Epoch 91/1000
20/20 [============== ] - 0s 10ms/step - loss: 0.5193 - binary accuracy:
0.7624
Epoch 92/1000
Epoch 93/1000
0.7544
Epoch 94/1000
0.7608
Epoch 95/1000
0.7735
Epoch 96/1000
0.7400
Epoch 97/1000
0.7624
Epoch 98/1000
0.7608
Epoch 99/1000
Epoch 100/1000
0.7480
Epoch 101/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5158 - binary accuracy:
0.7528
Epoch 102/1000
20/20 [================== ] - 0s 11ms/step - loss: 0.5278 - binary accuracy:
0.7560
Epoch 103/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5192 - binary accuracy:
0.7640
Epoch 104/1000
0.7560
Epoch 105/1000
0.7544
Epoch 106/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.6072 - binary accuracy:
0.6890
Epoch 107/1000
0.7081
Epoch 108/1000
0.6475
Epoch 109/1000
Epoch 110/1000
0.6683
Epoch 111/1000
0.6715
Epoch 112/1000
```

```
0.6794
Epoch 113/1000
0.6794
Epoch 114/1000
Epoch 115/1000
0.6810
Epoch 116/1000
20/20 [================ ] - 0s 10ms/step - loss: 0.5995 - binary accuracy:
0.6970
Epoch 117/1000
0.6874
Epoch 118/1000
0.6938
Epoch 119/1000
0.6810
Epoch 120/1000
20/20 [================== ] - 0s 10ms/step - loss: 0.5856 - binary accuracy:
0.6954
Epoch 121/1000
Epoch 122/1000
0.6922
Epoch 123/1000
0.6954
Epoch 124/1000
0.7033
Epoch 125/1000
0.7002
Epoch 126/1000
0.7129
Epoch 127/1000
20/20 [=============== ] - 0s 11ms/step - loss: 0.5717 - binary accuracy:
0.6970
Epoch 128/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5827 - binary accuracy:
0.7018
Epoch 129/1000
0.6970
Epoch 130/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.5663 - binary accuracy:
0.7033
Epoch 131/1000
Epoch 132/1000
0.6986
Epoch 133/1000
0.7002
Epoch 134/1000
```

```
0.7002
Epoch 135/1000
0.7065
Epoch 136/1000
Epoch 137/1000
0.6842
Epoch 138/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5731 - binary accuracy:
0.7033
Epoch 139/1000
0.7097
Epoch 140/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5529 - binary accuracy:
0.7145
Epoch 141/1000
20/20 [================== ] - Os 10ms/step - loss: 0.5922 - binary accuracy:
0.6778
Epoch 142/1000
0.7002
Epoch 143/1000
Epoch 144/1000
0.7097
Epoch 145/1000
0.7033
Epoch 146/1000
0.6922
Epoch 147/1000
0.7065
Epoch 148/1000
0.6858
Epoch 149/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5569 - binary accuracy:
0.7193
Epoch 150/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5650 - binary accuracy:
0.7097
Epoch 151/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5641 - binary accuracy:
0.7177
Epoch 152/1000
0.7129
Epoch 153/1000
Epoch 154/1000
0.7241
Epoch 155/1000
20/20 [================== ] - Os 10ms/step - loss: 0.5696 - binary accuracy:
0.7065
Epoch 156/1000
```

```
0.7049
Epoch 157/1000
0.7257
Epoch 158/1000
Epoch 159/1000
0.7065
Epoch 160/1000
0.7161
Epoch 161/1000
0.7177
Epoch 162/1000
0.6858
Epoch 163/1000
0.7177
Epoch 164/1000
0.7177
Epoch 165/1000
Epoch 166/1000
0.7018
Epoch 167/1000
0.7065
Epoch 168/1000
0.7177
Epoch 169/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5565 - binary accuracy:
0.7289
Epoch 170/1000
0.7209
Epoch 171/1000
0.6938
Epoch 172/1000
0.6794
Epoch 173/1000
0.6874
Epoch 174/1000
20/20 [================== ] - 0s 11ms/step - loss: 0.5571 - binary accuracy:
0.7081
Epoch 175/1000
Epoch 176/1000
0.7209
Epoch 177/1000
0.7177
Epoch 178/1000
```

```
0.7305
Epoch 179/1000
0.7241
Epoch 180/1000
Epoch 181/1000
0.7113
Epoch 182/1000
0.7337
Epoch 183/1000
0.7049
Epoch 184/1000
0.7113
Epoch 185/1000
0.7193
Epoch 186/1000
0.7400
Epoch 187/1000
Epoch 188/1000
0.7097
Epoch 189/1000
0.7337
Epoch 190/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5549 - binary accuracy:
0.7225
Epoch 191/1000
0.7193
Epoch 192/1000
0.6890
Epoch 193/1000
0.7400
Epoch 194/1000
0.7177
Epoch 195/1000
0.7097
Epoch 196/1000
0.7177
Epoch 197/1000
Epoch 198/1000
20/20 [================= ] - Os 10ms/step - loss: 0.5458 - binary accuracy:
0.7241
Epoch 199/1000
20/20 [================== ] - Os 11ms/step - loss: 0.5368 - binary accuracy:
0.7225
Epoch 200/1000
```

```
0.7273
Epoch 201/1000
0.7368
Epoch 202/1000
Epoch 203/1000
0.7337
Epoch 204/1000
0.7018
Epoch 205/1000
0.7129
Epoch 206/1000
0.7400
Epoch 207/1000
0.7352
Epoch 208/1000
0.7241
Epoch 209/1000
Epoch 210/1000
0.7002
Epoch 211/1000
0.7065
Epoch 212/1000
0.7368
Epoch 213/1000
0.7289
Epoch 214/1000
0.7337
Epoch 215/1000
0.6651
Epoch 216/1000
0.7321
Epoch 217/1000
0.7225
Epoch 218/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.5427 - binary accuracy:
0.7368
Epoch 219/1000
Epoch 220/1000
0.7400
Epoch 221/1000
20/20 [================= ] - 0s 16ms/step - loss: 0.5563 - binary accuracy:
0.7145
Epoch 222/1000
```

```
0.7273
Epoch 223/1000
20/20 [============= ] - 0s 9ms/step - loss: 0.5385 - binary accuracy:
0.7528
Epoch 224/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5429 - binary accuracy:
Epoch 225/1000
0.7528
Epoch 226/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5493 - binary accuracy:
0.7193
Epoch 227/1000
0.7416
Epoch 228/1000
0.7400
Epoch 229/1000
0.7193
Epoch 230/1000
0.7289
Epoch 231/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.5541 - binary accuracy:
Epoch 232/1000
0.7321
Epoch 233/1000
0.7496
Epoch 234/1000
0.7193
Epoch 235/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5380 - binary accuracy:
0.7400
Epoch 236/1000
0.7608
Epoch 237/1000
0.7416
Epoch 238/1000
0.7576
Epoch 239/1000
0.7640
Epoch 240/1000
20/20 [================== ] - Os 10ms/step - loss: 0.5181 - binary accuracy:
0.7687
Epoch 241/1000
Epoch 242/1000
0.7576
Epoch 243/1000
0.7640
Epoch 244/1000
```

```
0.7560
Epoch 245/1000
0.7656
Epoch 246/1000
Epoch 247/1000
0.7719
Epoch 248/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5606 - binary accuracy:
0.7113
Epoch 249/1000
0.7847
Epoch 250/1000
0.7656
Epoch 251/1000
0.7767
Epoch 252/1000
0.7416
Epoch 253/1000
Epoch 254/1000
0.7608
Epoch 255/1000
0.7719
Epoch 256/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5436 - binary accuracy:
0.7193
Epoch 257/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4949 - binary accuracy:
0.7671
Epoch 258/1000
20/20 [=============== ] - 0s 11ms/step - loss: 0.4867 - binary accuracy:
0.7735
Epoch 259/1000
0.7576
Epoch 260/1000
0.7656
Epoch 261/1000
0.7959
Epoch 262/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.5088 - binary accuracy:
0.7767
Epoch 263/1000
Epoch 264/1000
0.7767
Epoch 265/1000
0.7576
Epoch 266/1000
```

```
0.7608
Epoch 267/1000
0.7911
Epoch 268/1000
Epoch 269/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5099 - binary accuracy:
0.7528
Epoch 270/1000
0.8006
Epoch 271/1000
0.7751
Epoch 272/1000
20/20 [=============== ] - 0s 11ms/step - loss: 0.4851 - binary accuracy:
0.7895
Epoch 273/1000
20/20 [================= ] - Os 11ms/step - loss: 0.4912 - binary accuracy:
0.7863
Epoch 274/1000
0.7624
Epoch 275/1000
20/20 [================== ] - 0s 11ms/step - loss: 0.4868 - binary accuracy:
Epoch 276/1000
20/20 [================== ] - 0s 11ms/step - loss: 0.4831 - binary accuracy:
0.7799
Epoch 277/1000
0.7815
Epoch 278/1000
20/20 [=============== ] - 0s 11ms/step - loss: 0.5086 - binary accuracy:
0.7783
Epoch 279/1000
20/20 [=============== ] - 0s 14ms/step - loss: 0.4771 - binary accuracy:
0.7799
Epoch 280/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4826 - binary accuracy:
0.7767
Epoch 281/1000
20/20 [================== ] - 0s 11ms/step - loss: 0.5021 - binary accuracy:
0.7783
Epoch 282/1000
20/20 [=============== ] - 0s 11ms/step - loss: 0.5670 - binary accuracy:
0.7432
Epoch 283/1000
20/20 [================= ] - Os 10ms/step - loss: 0.4947 - binary accuracy:
0.7783
Epoch 284/1000
20/20 [================= ] - Os 10ms/step - loss: 0.4942 - binary accuracy:
0.7671
Epoch 285/1000
Epoch 286/1000
0.7911
Epoch 287/1000
0.7943
Epoch 288/1000
```

```
0.7687
Epoch 289/1000
20/20 [============= ] - 0s 9ms/step - loss: 0.5233 - binary accuracy:
0.7512
Epoch 290/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4873 - binary accuracy:
Epoch 291/1000
0.7735
Epoch 292/1000
20/20 [=============== ] - 0s 11ms/step - loss: 0.4800 - binary accuracy:
0.7863
Epoch 293/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.5075 - binary accuracy:
0.7640
Epoch 294/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4705 - binary accuracy:
0.7799
Epoch 295/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4687 - binary accuracy:
0.8038
Epoch 296/1000
20/20 [================= ] - Os 10ms/step - loss: 0.4926 - binary accuracy:
0.7703
Epoch 297/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.6041 - binary accuracy:
0.7209
Epoch 298/1000
20/20 [================== ] - 0s 11ms/step - loss: 0.5381 - binary accuracy:
0.7257
Epoch 299/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5090 - binary accuracy:
0.7703
Epoch 300/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4994 - binary accuracy:
0.7751
Epoch 301/1000
0.7895
Epoch 302/1000
0.7544
Epoch 303/1000
0.7895
Epoch 304/1000
0.7751
Epoch 305/1000
0.7911
Epoch 306/1000
20/20 [================== ] - 0s 10ms/step - loss: 0.4620 - binary accuracy:
0.8022
Epoch 307/1000
Epoch 308/1000
0.7959
Epoch 309/1000
20/20 [================== ] - Os 12ms/step - loss: 0.4695 - binary accuracy:
0.7815
Epoch 310/1000
```

```
0.7847
Epoch 311/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5076 - binary accuracy:
0.7751
Epoch 312/1000
20/20 [=============== ] - 0s 12ms/step - loss: 0.5191 - binary accuracy:
Epoch 313/1000
0.7959
Epoch 314/1000
0.7799
Epoch 315/1000
0.7735
Epoch 316/1000
0.7895
Epoch 317/1000
0.7943
Epoch 318/1000
0.7974
Epoch 319/1000
0.7815
Epoch 320/1000
0.8086
Epoch 321/1000
0.7767
Epoch 322/1000
0.7895
Epoch 323/1000
0.7799
Epoch 324/1000
0.7799
Epoch 325/1000
0.7815
Epoch 326/1000
0.7863
Epoch 327/1000
0.7974
Epoch 328/1000
0.7783
Epoch 329/1000
Epoch 330/1000
0.7767
Epoch 331/1000
0.7847
Epoch 332/1000
```

```
0.7735
Epoch 333/1000
0.7895
Epoch 334/1000
Epoch 335/1000
0.7767
Epoch 336/1000
0.7927
Epoch 337/1000
0.7783
Epoch 338/1000
0.7879
Epoch 339/1000
0.7943
Epoch 340/1000
0.7735
Epoch 341/1000
Epoch 342/1000
20/20 [================== ] - Os 10ms/step - loss: 0.4442 - binary accuracy:
0.7895
Epoch 343/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4852 - binary accuracy:
0.7719
Epoch 344/1000
0.7879
Epoch 345/1000
0.7879
Epoch 346/1000
0.7990
Epoch 347/1000
0.7990
Epoch 348/1000
0.8150
Epoch 349/1000
0.7863
Epoch 350/1000
0.7624
Epoch 351/1000
Epoch 352/1000
0.8006
Epoch 353/1000
0.7815
Epoch 354/1000
```

```
0.8006
Epoch 355/1000
0.7943
Epoch 356/1000
Epoch 357/1000
20/20 [=============== ] - 0s 11ms/step - loss: 0.4712 - binary accuracy:
0.8022
Epoch 358/1000
0.7863
Epoch 359/1000
0.7943
Epoch 360/1000
0.7879
Epoch 361/1000
0.7863
Epoch 362/1000
0.7974
Epoch 363/1000
Epoch 364/1000
20/20 [================= ] - 0s 11ms/step - loss: 0.4714 - binary accuracy:
0.7703
Epoch 365/1000
0.7799
Epoch 366/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4554 - binary accuracy:
0.8054
Epoch 367/1000
0.7767
Epoch 368/1000
0.7943
Epoch 369/1000
0.7990
Epoch 370/1000
0.7847
Epoch 371/1000
0.7927
Epoch 372/1000
0.8006
Epoch 373/1000
Epoch 374/1000
0.7783
Epoch 375/1000
0.7927
Epoch 376/1000
```

```
0.7767
Epoch 377/1000
0.7895
Epoch 378/1000
Epoch 379/1000
0.7895
Epoch 380/1000
0.8086
Epoch 381/1000
0.8038
Epoch 382/1000
0.7911
Epoch 383/1000
0.7799
Epoch 384/1000
0.7911
Epoch 385/1000
Epoch 386/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4671 - binary accuracy:
0.7974
Epoch 387/1000
0.7959
Epoch 388/1000
0.7783
Epoch 389/1000
0.7990
Epoch 390/1000
0.7847
Epoch 391/1000
0.7799
Epoch 392/1000
0.8022
Epoch 393/1000
0.7990
Epoch 394/1000
0.7974
Epoch 395/1000
Epoch 396/1000
0.7927
Epoch 397/1000
0.8070
Epoch 398/1000
```

```
0.7879
Epoch 399/1000
0.7927
Epoch 400/1000
Epoch 401/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4669 - binary accuracy:
0.8070
Epoch 402/1000
0.8022
Epoch 403/1000
0.7703
Epoch 404/1000
0.7990
Epoch 405/1000
0.7879
Epoch 406/1000
0.7831
Epoch 407/1000
Epoch 408/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4811 - binary accuracy:
0.7927
Epoch 409/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4640 - binary accuracy:
0.7799
Epoch 410/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4527 - binary accuracy:
0.8038
Epoch 411/1000
0.7974
Epoch 412/1000
0.7974
Epoch 413/1000
0.8038
Epoch 414/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4863 - binary accuracy:
0.7911
Epoch 415/1000
20/20 [=============== ] - 0s 11ms/step - loss: 0.4625 - binary accuracy:
0.7990
Epoch 416/1000
0.8054
Epoch 417/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4525 - binary accuracy:
Epoch 418/1000
0.7927
Epoch 419/1000
0.8022
Epoch 420/1000
```

```
0.8038
Epoch 421/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4550 - binary accuracy:
0.7974
Epoch 422/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4639 - binary accuracy:
Epoch 423/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4548 - binary accuracy:
0.7863
Epoch 424/1000
20/20 [=============== ] - 0s 11ms/step - loss: 0.4702 - binary accuracy:
0.7783
Epoch 425/1000
0.7974
Epoch 426/1000
0.8006
Epoch 427/1000
0.7959
Epoch 428/1000
0.7974
Epoch 429/1000
Epoch 430/1000
0.8038
Epoch 431/1000
0.7895
Epoch 432/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4493 - binary accuracy:
0.7974
Epoch 433/1000
0.7927
Epoch 434/1000
0.7927
Epoch 435/1000
0.7974
Epoch 436/1000
0.8070
Epoch 437/1000
0.7831
Epoch 438/1000
0.7895
Epoch 439/1000
Epoch 440/1000
0.8022
Epoch 441/1000
0.7767
Epoch 442/1000
```

```
0.7974
Epoch 443/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4471 - binary accuracy:
0.8006
Epoch 444/1000
Epoch 445/1000
0.8070
Epoch 446/1000
0.7751
Epoch 447/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4569 - binary accuracy:
0.7863
Epoch 448/1000
0.8054
Epoch 449/1000
0.7974
Epoch 450/1000
0.7927
Epoch 451/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4569 - binary accuracy:
Epoch 452/1000
0.8006
Epoch 453/1000
0.7959
Epoch 454/1000
0.8070
Epoch 455/1000
0.7911
Epoch 456/1000
0.7943
Epoch 457/1000
0.7974
Epoch 458/1000
0.7911
Epoch 459/1000
0.7911
Epoch 460/1000
20/20 [================== ] - Os 11ms/step - loss: 0.4487 - binary accuracy:
0.8070
Epoch 461/1000
Epoch 462/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4744 - binary accuracy:
0.7927
Epoch 463/1000
0.7448
Epoch 464/1000
```

```
0.7432
Epoch 465/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.5025 - binary accuracy:
0.7480
Epoch 466/1000
Epoch 467/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4669 - binary accuracy:
0.7943
Epoch 468/1000
0.8070
Epoch 469/1000
0.7895
Epoch 470/1000
0.7911
Epoch 471/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4589 - binary accuracy:
0.7990
Epoch 472/1000
0.8038
Epoch 473/1000
20/20 [================= ] - 0s 11ms/step - loss: 0.4497 - binary accuracy:
Epoch 474/1000
0.8054
Epoch 475/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4682 - binary accuracy:
0.7959
Epoch 476/1000
0.8134
Epoch 477/1000
0.8166
Epoch 478/1000
0.7911
Epoch 479/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4598 - binary accuracy:
0.7831
Epoch 480/1000
0.8118
Epoch 481/1000
0.7879
Epoch 482/1000
0.7943
Epoch 483/1000
Epoch 484/1000
0.7959
Epoch 485/1000
20/20 [================= ] - 0s 11ms/step - loss: 0.4476 - binary accuracy:
0.7943
Epoch 486/1000
```

```
0.8022
Epoch 487/1000
0.8166
Epoch 488/1000
Epoch 489/1000
20/20 [============== ] - 0s 10ms/step - loss: 0.4386 - binary accuracy:
0.8150
Epoch 490/1000
0.7863
Epoch 491/1000
0.7911
Epoch 492/1000
0.7959
Epoch 493/1000
0.8022
Epoch 494/1000
0.8086
Epoch 495/1000
20/20 [================= ] - Os 10ms/step - loss: 0.4477 - binary accuracy:
Epoch 496/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4427 - binary accuracy:
0.8006
Epoch 497/1000
0.8054
Epoch 498/1000
20/20 [=============== ] - 0s 11ms/step - loss: 0.4447 - binary accuracy:
0.8022
Epoch 499/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4736 - binary accuracy:
0.7974
Epoch 500/1000
0.7974
Epoch 501/1000
0.8038
Epoch 502/1000
0.7990
Epoch 503/1000
0.8070
Epoch 504/1000
0.7927
Epoch 505/1000
Epoch 506/1000
0.8006
Epoch 507/1000
0.8086
Epoch 508/1000
```

```
0.8022
Epoch 509/1000
0.8038
Epoch 510/1000
20/20 [=============== ] - 0s 13ms/step - loss: 0.4676 - binary accuracy:
Epoch 511/1000
0.7879
Epoch 512/1000
0.7847
Epoch 513/1000
0.7990
Epoch 514/1000
0.8118
Epoch 515/1000
0.8054
Epoch 516/1000
0.8118
Epoch 517/1000
Epoch 518/1000
0.7927
Epoch 519/1000
0.8006
Epoch 520/1000
0.8086
Epoch 521/1000
0.8006
Epoch 522/1000
0.8118
Epoch 523/1000
0.8102
Epoch 524/1000
0.7863
Epoch 525/1000
0.8054
Epoch 526/1000
0.7927
Epoch 527/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4334 - binary accuracy:
Epoch 528/1000
0.7959
Epoch 529/1000
0.8086
Epoch 530/1000
```

```
0.7959
Epoch 531/1000
0.8118
Epoch 532/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4470 - binary accuracy:
Epoch 533/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4440 - binary accuracy:
0.8070
Epoch 534/1000
0.8198
Epoch 535/1000
0.7879
Epoch 536/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4405 - binary accuracy:
0.8054
Epoch 537/1000
0.7927
Epoch 538/1000
0.7943
Epoch 539/1000
Epoch 540/1000
0.8038
Epoch 541/1000
0.8022
Epoch 542/1000
0.8166
Epoch 543/1000
0.7895
Epoch 544/1000
0.7959
Epoch 545/1000
0.7974
Epoch 546/1000
0.8150
Epoch 547/1000
0.8102
Epoch 548/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4235 - binary accuracy:
0.8166
Epoch 549/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4362 - binary accuracy:
Epoch 550/1000
0.7895
Epoch 551/1000
0.8006
Epoch 552/1000
```

```
0.8006
Epoch 553/1000
0.7959
Epoch 554/1000
Epoch 555/1000
0.8102
Epoch 556/1000
0.7959
Epoch 557/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4378 - binary accuracy:
0.8006
Epoch 558/1000
0.8118
Epoch 559/1000
0.8070
Epoch 560/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4808 - binary accuracy:
0.7847
Epoch 561/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4363 - binary accuracy:
0.8118
Epoch 562/1000
0.7911
Epoch 563/1000
0.8038
Epoch 564/1000
0.8038
Epoch 565/1000
0.8134
Epoch 566/1000
0.7895
Epoch 567/1000
0.8070
Epoch 568/1000
0.8022
Epoch 569/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4516 - binary accuracy:
0.7879
Epoch 570/1000
0.8070
Epoch 571/1000
Epoch 572/1000
0.8022
Epoch 573/1000
20/20 [================= ] - Os 10ms/step - loss: 0.4471 - binary accuracy:
0.7927
Epoch 574/1000
```

20/20 [=======] - 0s 9ms/step - loss: 0.4676 - binary_accuracy:

```
0.7959
Epoch 575/1000
0.8086
Epoch 576/1000
Epoch 577/1000
0.7943
Epoch 578/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4626 - binary accuracy:
0.7911
Epoch 579/1000
20/20 [================= ] - Os 10ms/step - loss: 0.4392 - binary accuracy:
0.8022
Epoch 580/1000
0.8086
Epoch 581/1000
0.8070
Epoch 582/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4467 - binary accuracy:
0.7911
Epoch 583/1000
0.8086
Epoch 584/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4545 - binary accuracy:
0.7990
Epoch 585/1000
0.8118
Epoch 586/1000
0.8006
Epoch 587/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4591 - binary accuracy:
0.8006
Epoch 588/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4455 - binary accuracy:
0.7927
Epoch 589/1000
0.8086
Epoch 590/1000
0.7863
Epoch 591/1000
0.7974
Epoch 592/1000
0.8102
Epoch 593/1000
Epoch 594/1000
0.8102
Epoch 595/1000
0.8134
Epoch 596/1000
```

```
0.7863
Epoch 597/1000
0.8198
Epoch 598/1000
Epoch 599/1000
0.7959
Epoch 600/1000
0.7943
Epoch 601/1000
0.8102
Epoch 602/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4377 - binary accuracy:
0.8166
Epoch 603/1000
0.8102
Epoch 604/1000
0.7767
Epoch 605/1000
0.8070
Epoch 606/1000
20/20 [================== ] - 0s 10ms/step - loss: 0.4311 - binary accuracy:
0.8134
Epoch 607/1000
0.8134
Epoch 608/1000
0.7847
Epoch 609/1000
0.7863
Epoch 610/1000
0.7927
Epoch 611/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4325 - binary accuracy:
0.8086
Epoch 612/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4534 - binary accuracy:
0.8070
Epoch 613/1000
0.7959
Epoch 614/1000
0.7943
Epoch 615/1000
Epoch 616/1000
0.8102
Epoch 617/1000
0.7863
Epoch 618/1000
```

```
0.8006
Epoch 619/1000
0.8118
Epoch 620/1000
Epoch 621/1000
0.8150
Epoch 622/1000
0.8118
Epoch 623/1000
0.8086
Epoch 624/1000
0.8086
Epoch 625/1000
0.8086
Epoch 626/1000
0.8118
Epoch 627/1000
Epoch 628/1000
0.8022
Epoch 629/1000
0.8038
Epoch 630/1000
0.7879
Epoch 631/1000
0.8086
Epoch 632/1000
0.7879
Epoch 633/1000
0.7927
Epoch 634/1000
0.7927
Epoch 635/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4384 - binary accuracy:
0.8022
Epoch 636/1000
20/20 [================== ] - Os 10ms/step - loss: 0.4320 - binary accuracy:
0.8070
Epoch 637/1000
Epoch 638/1000
0.8006
Epoch 639/1000
0.8150
Epoch 640/1000
```

```
0.8070
Epoch 641/1000
0.8054
Epoch 642/1000
Epoch 643/1000
0.7974
Epoch 644/1000
0.7831
Epoch 645/1000
20/20 [================= ] - 0s 11ms/step - loss: 0.4341 - binary accuracy:
0.8038
Epoch 646/1000
0.8118
Epoch 647/1000
0.8086
Epoch 648/1000
0.7927
Epoch 649/1000
Epoch 650/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4342 - binary accuracy:
0.8086
Epoch 651/1000
0.8054
Epoch 652/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4452 - binary accuracy:
0.8054
Epoch 653/1000
0.8150
Epoch 654/1000
0.7974
Epoch 655/1000
0.8102
Epoch 656/1000
0.8102
Epoch 657/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4347 - binary accuracy:
0.8006
Epoch 658/1000
20/20 [================= ] - Os 10ms/step - loss: 0.4380 - binary accuracy:
0.7927
Epoch 659/1000
Epoch 660/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4484 - binary accuracy:
0.8086
Epoch 661/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.5535 - binary accuracy:
0.7576
Epoch 662/1000
```

```
0.7927
Epoch 663/1000
0.7990
Epoch 664/1000
Epoch 665/1000
0.8006
Epoch 666/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4352 - binary accuracy:
0.8022
Epoch 667/1000
0.8022
Epoch 668/1000
0.8166
Epoch 669/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4668 - binary accuracy:
0.8022
Epoch 670/1000
0.8038
Epoch 671/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4499 - binary accuracy:
0.8006
Epoch 672/1000
0.7974
Epoch 673/1000
0.8022
Epoch 674/1000
0.8134
Epoch 675/1000
0.8070
Epoch 676/1000
0.8070
Epoch 677/1000
0.7959
Epoch 678/1000
0.8054
Epoch 679/1000
0.8038
Epoch 680/1000
0.8022
Epoch 681/1000
Epoch 682/1000
0.8166
Epoch 683/1000
0.8054
Epoch 684/1000
```

```
0.8166
Epoch 685/1000
0.8150
Epoch 686/1000
20/20 [============== ] - 0s 11ms/step - loss: 0.4333 - binary accuracy:
Epoch 687/1000
0.8182
Epoch 688/1000
0.7990
Epoch 689/1000
0.8214
Epoch 690/1000
0.8038
Epoch 691/1000
0.8038
Epoch 692/1000
0.8070
Epoch 693/1000
20/20 [================== ] - 0s 10ms/step - loss: 0.4232 - binary accuracy:
Epoch 694/1000
0.8038
Epoch 695/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4295 - binary accuracy:
0.8166
Epoch 696/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4251 - binary accuracy:
0.8022
Epoch 697/1000
0.7974
Epoch 698/1000
0.8022
Epoch 699/1000
0.8102
Epoch 700/1000
0.8150
Epoch 701/1000
0.7959
Epoch 702/1000
0.8038
Epoch 703/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4259 - binary accuracy:
Epoch 704/1000
0.8086
Epoch 705/1000
0.8054
Epoch 706/1000
```

```
0.8054
Epoch 707/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4553 - binary accuracy:
0.8022
Epoch 708/1000
20/20 [=============== ] - 0s 11ms/step - loss: 0.4438 - binary accuracy:
Epoch 709/1000
20/20 [============== ] - 0s 10ms/step - loss: 0.4437 - binary accuracy:
0.7911
Epoch 710/1000
0.8054
Epoch 711/1000
20/20 [================= ] - 0s 11ms/step - loss: 0.4615 - binary accuracy:
0.8006
Epoch 712/1000
0.8086
Epoch 713/1000
0.8086
Epoch 714/1000
0.8134
Epoch 715/1000
Epoch 716/1000
0.8166
Epoch 717/1000
0.8150
Epoch 718/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4296 - binary accuracy:
0.7990
Epoch 719/1000
0.8230
Epoch 720/1000
20/20 [============== ] - 0s 10ms/step - loss: 0.4331 - binary accuracy:
0.8166
Epoch 721/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4557 - binary accuracy:
0.7911
Epoch 722/1000
0.8182
Epoch 723/1000
0.8070
Epoch 724/1000
0.8054
Epoch 725/1000
Epoch 726/1000
0.7990
Epoch 727/1000
0.8054
Epoch 728/1000
```

20/20 [========] - 0s 9ms/step - loss: 0.4578 - binary_accuracy:

```
0.7990
Epoch 729/1000
0.8086
Epoch 730/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4160 - binary accuracy:
Epoch 731/1000
0.8006
Epoch 732/1000
0.8214
Epoch 733/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4468 - binary accuracy:
0.8198
Epoch 734/1000
0.7911
Epoch 735/1000
0.7990
Epoch 736/1000
0.8198
Epoch 737/1000
Epoch 738/1000
0.8150
Epoch 739/1000
0.8070
Epoch 740/1000
0.8230
Epoch 741/1000
0.8054
Epoch 742/1000
0.8070
Epoch 743/1000
0.8118
Epoch 744/1000
0.8006
Epoch 745/1000
0.8054
Epoch 746/1000
0.7990
Epoch 747/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4360 - binary accuracy:
Epoch 748/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4294 - binary accuracy:
0.8006
Epoch 749/1000
0.8038
Epoch 750/1000
```

```
0.8070
Epoch 751/1000
0.8054
Epoch 752/1000
Epoch 753/1000
20/20 [=============== ] - 0s 11ms/step - loss: 0.4247 - binary accuracy:
0.8102
Epoch 754/1000
0.7879
Epoch 755/1000
0.8070
Epoch 756/1000
20/20 [=============== ] - Os 10ms/step - loss: 0.4470 - binary accuracy:
0.7959
Epoch 757/1000
20/20 [================ ] - 0s 10ms/step - loss: 0.4206 - binary accuracy:
0.8150
Epoch 758/1000
20/20 [================= ] - Os 10ms/step - loss: 0.4540 - binary accuracy:
0.8006
Epoch 759/1000
20/20 [================= ] - Os 10ms/step - loss: 0.4494 - binary accuracy:
Epoch 760/1000
0.8214
Epoch 761/1000
0.8022
Epoch 762/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4378 - binary accuracy:
0.8102
Epoch 763/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4268 - binary accuracy:
0.8070
Epoch 764/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4320 - binary accuracy:
0.7974
Epoch 765/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4402 - binary accuracy:
0.8070
Epoch 766/1000
0.8246
Epoch 767/1000
0.8198
Epoch 768/1000
20/20 [================== ] - 0s 10ms/step - loss: 0.4277 - binary accuracy:
0.8150
Epoch 769/1000
Epoch 770/1000
20/20 [================= ] - Os 10ms/step - loss: 0.4169 - binary accuracy:
0.8166
Epoch 771/1000
20/20 [================= ] - Os 10ms/step - loss: 0.4599 - binary accuracy:
0.7863
Epoch 772/1000
```

20/20 [=======] - 0s 9ms/step - loss: 0.4072 - binary_accuracy:

```
0.8262
Epoch 773/1000
0.8118
Epoch 774/1000
Epoch 775/1000
0.8150
Epoch 776/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4330 - binary accuracy:
0.8150
Epoch 777/1000
0.8038
Epoch 778/1000
0.7911
Epoch 779/1000
0.8006
Epoch 780/1000
0.8038
Epoch 781/1000
Epoch 782/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4362 - binary accuracy:
0.8102
Epoch 783/1000
0.8006
Epoch 784/1000
0.8054
Epoch 785/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4288 - binary accuracy:
0.8166
Epoch 786/1000
0.8070
Epoch 787/1000
0.8102
Epoch 788/1000
0.8102
Epoch 789/1000
0.8086
Epoch 790/1000
0.8278
Epoch 791/1000
Epoch 792/1000
0.8022
Epoch 793/1000
20/20 [================== ] - Os 10ms/step - loss: 0.4405 - binary accuracy:
0.8054
Epoch 794/1000
```

20/20 [=======] - 0s 9ms/step - loss: 0.4305 - binary_accuracy:

```
0.8118
Epoch 795/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4314 - binary accuracy:
0.8006
Epoch 796/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4417 - binary accuracy:
Epoch 797/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4300 - binary accuracy:
0.8198
Epoch 798/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4510 - binary accuracy:
0.8022
Epoch 799/1000
20/20 [================= ] - 0s 14ms/step - loss: 0.4282 - binary accuracy:
0.8309
Epoch 800/1000
20/20 [=============== ] - 0s 17ms/step - loss: 0.4311 - binary accuracy:
0.8102
Epoch 801/1000
20/20 [================= ] - Os 10ms/step - loss: 0.4417 - binary accuracy:
0.7990
Epoch 802/1000
0.8102
Epoch 803/1000
20/20 [================== ] - Os 10ms/step - loss: 0.4183 - binary accuracy:
Epoch 804/1000
20/20 [=============== ] - 0s 11ms/step - loss: 0.4200 - binary accuracy:
0.8198
Epoch 805/1000
0.8022
Epoch 806/1000
0.8070
Epoch 807/1000
0.7927
Epoch 808/1000
0.8198
Epoch 809/1000
0.7974
Epoch 810/1000
0.8086
Epoch 811/1000
0.8054
Epoch 812/1000
0.8022
Epoch 813/1000
Epoch 814/1000
0.8150
Epoch 815/1000
0.8118
Epoch 816/1000
```

```
0.8214
Epoch 817/1000
0.8150
Epoch 818/1000
Epoch 819/1000
0.8118
Epoch 820/1000
0.8070
Epoch 821/1000
0.7895
Epoch 822/1000
0.7974
Epoch 823/1000
0.8134
Epoch 824/1000
0.7927
Epoch 825/1000
Epoch 826/1000
0.8086
Epoch 827/1000
0.8070
Epoch 828/1000
0.8150
Epoch 829/1000
0.8150
Epoch 830/1000
0.8006
Epoch 831/1000
0.8038
Epoch 832/1000
0.8214
Epoch 833/1000
0.8118
Epoch 834/1000
0.8102
Epoch 835/1000
Epoch 836/1000
0.8262
Epoch 837/1000
0.8214
Epoch 838/1000
```

```
0.8166
Epoch 839/1000
0.8150
Epoch 840/1000
Epoch 841/1000
0.8102
Epoch 842/1000
0.8086
Epoch 843/1000
0.8198
Epoch 844/1000
0.8118
Epoch 845/1000
0.8086
Epoch 846/1000
0.7911
Epoch 847/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4124 - binary accuracy:
Epoch 848/1000
0.8086
Epoch 849/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4142 - binary accuracy:
0.8214
Epoch 850/1000
0.8054
Epoch 851/1000
0.8134
Epoch 852/1000
0.8102
Epoch 853/1000
0.8214
Epoch 854/1000
0.8150
Epoch 855/1000
0.8230
Epoch 856/1000
0.8166
Epoch 857/1000
Epoch 858/1000
0.8134
Epoch 859/1000
0.7974
Epoch 860/1000
```

```
0.7974
Epoch 861/1000
0.8166
Epoch 862/1000
Epoch 863/1000
0.8118
Epoch 864/1000
20/20 [================== ] - 0s 10ms/step - loss: 0.4031 - binary accuracy:
0.8182
Epoch 865/1000
0.8198
Epoch 866/1000
0.7879
Epoch 867/1000
0.8118
Epoch 868/1000
0.8134
Epoch 869/1000
Epoch 870/1000
0.8198
Epoch 871/1000
0.8198
Epoch 872/1000
0.7911
Epoch 873/1000
0.8134
Epoch 874/1000
0.8070
Epoch 875/1000
0.8134
Epoch 876/1000
0.8054
Epoch 877/1000
0.8230
Epoch 878/1000
0.8038
Epoch 879/1000
Epoch 880/1000
0.8214
Epoch 881/1000
0.8150
Epoch 882/1000
```

```
0.8118
Epoch 883/1000
0.8182
Epoch 884/1000
Epoch 885/1000
0.8182
Epoch 886/1000
0.8182
Epoch 887/1000
0.7959
Epoch 888/1000
0.8022
Epoch 889/1000
0.8134
Epoch 890/1000
0.8262
Epoch 891/1000
Epoch 892/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4132 - binary accuracy:
0.8198
Epoch 893/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4198 - binary accuracy:
0.8246
Epoch 894/1000
0.8102
Epoch 895/1000
0.8198
Epoch 896/1000
0.8070
Epoch 897/1000
20/20 [================== ] - 0s 10ms/step - loss: 0.4372 - binary accuracy:
0.8054
Epoch 898/1000
0.8182
Epoch 899/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4310 - binary accuracy:
0.8182
Epoch 900/1000
0.8086
Epoch 901/1000
Epoch 902/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4455 - binary accuracy:
0.8102
Epoch 903/1000
20/20 [================== ] - 0s 12ms/step - loss: 0.4658 - binary accuracy:
0.7974
Epoch 904/1000
```

```
0.8054
Epoch 905/1000
0.8070
Epoch 906/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4130 - binary accuracy:
Epoch 907/1000
0.7990
Epoch 908/1000
0.8150
Epoch 909/1000
0.8086
Epoch 910/1000
0.7895
Epoch 911/1000
0.8214
Epoch 912/1000
0.8198
Epoch 913/1000
Epoch 914/1000
0.7863
Epoch 915/1000
0.8118
Epoch 916/1000
0.8150
Epoch 917/1000
0.8166
Epoch 918/1000
0.8134
Epoch 919/1000
0.8246
Epoch 920/1000
0.8134
Epoch 921/1000
0.8102
Epoch 922/1000
0.8134
Epoch 923/1000
Epoch 924/1000
0.8102
Epoch 925/1000
0.8118
Epoch 926/1000
```

```
0.8198
Epoch 927/1000
0.8038
Epoch 928/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4157 - binary accuracy:
Epoch 929/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4214 - binary accuracy:
0.8278
Epoch 930/1000
0.8038
Epoch 931/1000
0.8182
Epoch 932/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4117 - binary accuracy:
0.8118
Epoch 933/1000
20/20 [================= ] - Os 10ms/step - loss: 0.4411 - binary accuracy:
0.8086
Epoch 934/1000
0.8006
Epoch 935/1000
20/20 [================== ] - 0s 12ms/step - loss: 0.4092 - binary accuracy:
Epoch 936/1000
0.8373
Epoch 937/1000
0.8086
Epoch 938/1000
0.8198
Epoch 939/1000
0.8214
Epoch 940/1000
0.8102
Epoch 941/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4373 - binary accuracy:
0.8070
Epoch 942/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4053 - binary accuracy:
0.8214
Epoch 943/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4620 - binary accuracy:
0.7974
Epoch 944/1000
20/20 [================== ] - 0s 10ms/step - loss: 0.4050 - binary accuracy:
0.8198
Epoch 945/1000
Epoch 946/1000
0.8006
Epoch 947/1000
0.8214
Epoch 948/1000
```

```
0.8293
Epoch 949/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4138 - binary accuracy:
0.8214
Epoch 950/1000
Epoch 951/1000
0.8198
Epoch 952/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4254 - binary accuracy:
0.8246
Epoch 953/1000
0.8166
Epoch 954/1000
0.8070
Epoch 955/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4125 - binary accuracy:
0.8102
Epoch 956/1000
0.8086
Epoch 957/1000
Epoch 958/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4291 - binary accuracy:
0.8086
Epoch 959/1000
0.8214
Epoch 960/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4161 - binary accuracy:
0.8070
Epoch 961/1000
0.8102
Epoch 962/1000
0.8150
Epoch 963/1000
0.8086
Epoch 964/1000
0.8102
Epoch 965/1000
0.8182
Epoch 966/1000
0.8118
Epoch 967/1000
Epoch 968/1000
0.8118
Epoch 969/1000
0.7974
Epoch 970/1000
```

```
0.8150
Epoch 971/1000
0.8150
Epoch 972/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4091 - binary accuracy:
Epoch 973/1000
Epoch 974/1000
0.8150
Epoch 975/1000
20/20 [================= ] - Os 10ms/step - loss: 0.4270 - binary accuracy:
0.8038
Epoch 976/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4322 - binary accuracy:
0.8070
Epoch 977/1000
20/20 [================= ] - 0s 11ms/step - loss: 0.4113 - binary accuracy:
0.8166
Epoch 978/1000
0.8134
Epoch 979/1000
20/20 [================= ] - 0s 10ms/step - loss: 0.4225 - binary accuracy:
Epoch 980/1000
0.8278
Epoch 981/1000
0.8198
Epoch 982/1000
0.8150
Epoch 983/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4257 - binary accuracy:
0.8150
Epoch 984/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4549 - binary accuracy:
0.8054
Epoch 985/1000
20/20 [=============== ] - 0s 10ms/step - loss: 0.4380 - binary accuracy:
0.8166
Epoch 986/1000
0.8166
Epoch 987/1000
0.8150
Epoch 988/1000
0.8134
Epoch 989/1000
Epoch 990/1000
0.8150
Epoch 991/1000
0.8102
Epoch 992/1000
```

```
0.8214
      Epoch 993/1000
      0.8134
      Epoch 994/1000
      Epoch 995/1000
      20/20 [=============== ] - 0s 10ms/step - loss: 0.4254 - binary accuracy:
      Epoch 996/1000
      0.8150
      Epoch 997/1000
      0.8134
      Epoch 998/1000
      0.8150
      Epoch 999/1000
      20/20 [================== ] - 0s 10ms/step - loss: 0.3999 - binary accuracy:
      0.8198
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