

```
[ ] 1 # survived : 1-생존, 0-사망
     2 # pclass : 객실등급
     3 # sex : 성별
     4 # age : 나이
     5 # sibsp : 함께 탑승한 형제자매, 아내, 남편 수
     6 # parch : 함께 탐승한 부모, 자식 수
     7 # fare : 티켓요금
     8 # embarked : 배에 탑승한 항구 이름(C = Cherbourn, Q = Queenstown, S = Southampton)
     9 # class : 객실등급 숫자
    10 # who : 성별
    11 # adult male : 성인남성
    12 # deck : 배에 탑승한 항구 이름(C = Cherbourn, Q = Queenstown, S = Southampton)
    13 # embark town : 배에 탑승한 항구 이름(C = Cherbourn, Q = Queenstown, S = Southampton)
    14 # alive : yes-생존, no-사망
    15 # alone : 동반 탑승 유무
    16
    17 titanic.info()
     1 # missing data가 있는지 다시한번 그래프를 그려서 확인
```

```
2 import missingno as mino
3 mino.matrix(titanic)
4
5 # nan값이 있는 데이터 : age, embarked, deck 는 확인 필요
```

Q

```
    boxplot

[ ] 1 data = titanic.iloc[:,1:]
    2 data.boxplot()

[ ] 1 # fare에 특이한 아웃라이어가 한개 확인됨.
    2 # 처리해주는 것이 좋을것으로 예상됨
```

## ▼ 데이터 전처리

```
[] 1 titanic = sns.load_dataset('titanic')
2
3 # 중복돈 컬럼은 우선 제외하고 진행 -> 추후에는 중복된 컬럼을 사용하여 nan 값을 보완해보는 것도 방법
4 titanic = titanic[['survived','pclass','sex','age','sibsp','parch','fare','embarked','adult_male','alone']]
5 print('초기데이터:',len(data))
초기데이터: 709
```

1 # boxplot에서 500이상인 요금은 이상치로 생각하여 제외 2 titanic[titanic['fare']>500]

$\Rightarrow$		survived	pclass	sex	age	sibsp	parch	fare	embarked	adult_male	alone
	258	1	1	female	35.0	0	0	512.3292	С	False	True
	679	1	1	male	36.0	0	1	512.3292	С	True	False
	737	1	1	male	35.0	0	0	512,3292	С	True	True

<>> **□** 

{*x*}

```
1 data = titanic.drop([258,679,737])
     1 # 잘 삭제 되었는지 확인
     2 data.iloc[:,1:].boxplot()
0
     1 # nan 값 확인
     2 # data[data.isna().any(axis=1)]
     3 data[data.age.isna()]
\supseteq
         survived pclass
                            sex age sibsp parch
                                                     fare embarked adult male alone
                                                    8.4583
                                                                                 True
      5
                0
                            male NaN
                                                                  Q
                                                                           True
     17
                                                0 13.0000
                                                                  S
                                                                           True
                                                                                  True
                            male NaN
                                          0
                                                0 7.2250
                                                                           False
                                                                                  True
     19
                        3 female NaN
                                                                  C
     26
                0
                           male NaN
                                                    7.2250
                                                                                  True
                                                                  C
                                                                           True
                                          0
                                                                                  True
     28
                        3 female NaN
                                                0 7.8792
                                                                  Q
                                                                           False
     859
                0
                        3 male NaN
                                                0 7.2292
                                                                  C
                                                                           True
                                                                                True
                                                                                  False
     863
                0
                        3 female NaN
                                                 2 69.5500
                                                                  S
                                                                           False
     868
                0
                            male NaN
                                                0 9.5000
                                                                  S
                                                                           True
                                                                                  True
                0
     878
                                                                  S
                                                                                  True
                            male NaN
                                                    7.8958
                                                                           True
     888
                0
                        3 female NaN
                                                 2 23.4500
                                                                                  False
                                                                  S
                                                                           False
    177 rows × 10 columns
```

Q

{*x*}

<>

```
Q
           1 data[data.embarked.isna()]
{x}
              survived pclass sex age sibsp parch fare embarked adult_male alone
                            1 female 38.0
                                                  0 80.0
                                                              NaN
                                                                         False True
          829
                            1 female 62.0
                                                  0 80.0
                                                              NaN
                                                                         False
                                                                              True
           1 # 우선 nan 데이터 삭제후 진행할게요. 매우는 방법도 개별로 한번 생각해서 진행해보세요.
           2 data.dropna(inplace = True)
           3 print('nan 삭제 후 데이터:',len(data))
         nan 삭제 후 데이터: 709
           1 # missing data가 있는지 다시한번 그래프를 그려서 확인
           2 import missingno as mino
           3 mino.matrix(data)
```

```
Q
            2 from sklearn.preprocessing import LabelEncoder
            4 # sex, adult male, alone
            5 # LabelEncoder 사용
            6 encoding = LabelEncoder()
            7 data['sex'] = encoding.fit transform(data['sex']) # male = 1, female = 0
            8 data['adult male'] = encoding.fit transform(data['adult male']) # True = 1, False = 0
            9 data['alone'] = encoding.fit transform(data['alone']) # True = 1, False = 0
           10
           11 # embarked
           12 # apply 사용, replace 사용
           13 data['embarked'] = data['embarked'].apply(lambda x : x.replace('C','0').replace('Q','1').replace('S','2')) # C = 0, Q = 1, S = 2
           14
           15 data = data.astype('float') # Tensor의 입력은 int형태 X, Float로 변환 필요
           16 y encoding = encoding.fit transform(data['survived'])
```

1 # 범주형변수(str -> float)

```
▼ 모델링
```

{*x*}

```
1 from sklearn.model selection import train test split
 2 from tensorflow.keras.models import Sequential
 3 from tensorflow.keras.layers import Dense
 5 X = data.iloc[:,1:]
 6 y = data.iloc[:,0] # survived
 8 X train, X test, y train, y test = train test split(X, y encoding, random state=42) # test size : default = 0.25
10 model = Sequential()
11
12 model.add(Dense(8, input dim = 9, activation = 'relu'))
13 model.add(Dense(1, activation = 'sigmoid'))
14
15 model.compile(loss = 'binary crossentropy', optimizer = 'adam', metrics = ['accuracy'])
16
17 model.fit(X train, y train, epochs = 10, batch size = 10)
 1 test loss, test acc = model.evaluate(X test,y test,verbose=2)
 3 print('test loss',test loss)
 4 print('test acc', test acc)
6/6 - 0s - loss: 0.7985 - accuracy: 0.6404 - 143ms/epoch - 24ms/step
test loss 0.7984855771064758
```

<>

test acc 0.6404494643211365

```
▼ 모델 성능개선
```

{*x*}

<>

```
1 import seaborn as sns
 3 from sklearn.preprocessing import LabelEncoder
 4 from sklearn.model selection import train test split
 5 from tensorflow.keras.models import Sequential
 6 from tensorflow.keras.layers import Dense, Dropout
 8 titanic = sns.load dataset('titanic')
9 titanic = titanic[['survived','pclass','sex','age','sibsp','parch','fare','embarked','adult male','alone']]
10 data = titanic.drop([258,679,737])
11
12 data.dropna(inplace = True)
13
14 encoding = LabelEncoder()
15 data['sex'] = encoding.fit transform(data['sex']) # male = 1, female = 0
16 data['adult male'] = encoding.fit transform(data['adult male']) # True = 1, False = 0
17 data['alone'] = encoding.fit transform(data['alone']) # True = 1, False = 0
18 data['embarked'] = data['embarked'].apply(lambda x : x.replace('C','0').replace('Q','1').replace('S','2')) # C = 0, Q = 1, S = 2
19
20 data = data.astype('float')
21 y encoding = encoding.fit transform(data['survived'])
22
23 X = data.iloc[:,1:]
24 y = data.iloc[:,0]
25
26 X train, X test, y train, y test = train test split(X, y encoding, random state=42)
27
28 model = Sequential()
29
```

```
30 model.add(Dense(512, input dim = 9, activation = 'relu'))
           31 model.add(Dense(128, activation = 'relu'))
Q
           32 model.add(Dense(64, activation = 'relu'))
           33 model.add(Dense(64, activation = 'relu'))
{x}
           34 model.add(Dense(32, activation = 'relu'))
           35 model.add(Dense(1, activation = 'sigmoid'))
           36
           37 model.compile(loss = 'binary crossentropy', optimizer = 'adam', metrics = ['accuracy'])
           38
           39 model.fit(X train, y train, epochs = 100, batch size = 12)
            1 train loss, train acc = model.evaluate(X train,y train,verbose=2)
            3 print('train loss',train loss)
            4 print('train acc',train acc)
          17/17 - 0s - loss: 0.2814 - accuracy: 0.8814 - 209ms/epoch - 12ms/step
          train loss 0.28138166666030884
           train acc 0.8813559412956238
                                                                                                                                                    1 test loss, test acc = model.evaluate(X test,y test,verbose=2)
            3 print('test loss',test loss)
            4 print('test acc', test acc)
          6/6 - 0s - loss: 0.6688 - accuracy: 0.7528 - 48ms/epoch - 8ms/step
           test loss 0.6688404083251953
          test acc 0.7528089880943298
<>
      더블클릭 또는 Enter 키를 눌러 수정
```