

딥러닝 모델 설계하기

```
In [2]: # !pip list
```

1. 환경 준비

```
In [4]: # 텐서플로 라이브러리 안에 있는 케라스 API에서 필요한 함수들을 불러옵니다.
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense

# 데이터를 다루는 데 필요한 라이브러리를 불러옵니다.
import numpy as np
```

2. 데이터 준비

```
In [6]: # 준비된 수술 환자 데이터를 불러옵니다.
Data_set = np.loadtxt("./data/ThoraricSurgery3.csv", delimiter=",")
X = Data_set[:,0:16] # 환자의 진찰 기록을 X로 지정합니다.
y = Data_set[:,16]   # 수술 1년 후 사망/생존 여부를 y로 지정합니다.
```

3. 구조 결정





















```
In [8]: # 딥러닝 모델의 구조를 결정합니다.
model = Sequential()
model.add(Dense(30, input_dim=16, activation='relu'))
model.add(Dense(1, activation='sigmoid'))
```

C:\Users\user\AppData\Roaming\Python\Python312\site-packages\keras\src\layers\core\dense.py:87: UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When using Sequential models, prefer using an `Input(shape)` object as the first layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

4. 모델 실행

```
In [10]: # 딥러닝 모델을 실행합니다.  
model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])  
history=model.fit(X, y, epochs=30, batch_size=16)
```

Epoch 1/30
30/30  1s 2ms/step - accuracy: 0.2967 - loss: 2.2655
Epoch 2/30
30/30  0s 1ms/step - accuracy: 0.8055 - loss: 0.7083
Epoch 3/30
30/30  0s 1ms/step - accuracy: 0.8124 - loss: 0.5495
Epoch 4/30
30/30  0s 1ms/step - accuracy: 0.8472 - loss: 0.4456
Epoch 5/30
30/30  0s 1ms/step - accuracy: 0.8448 - loss: 0.4388
Epoch 6/30
30/30  0s 1ms/step - accuracy: 0.8303 - loss: 0.4632
Epoch 7/30
30/30  0s 1ms/step - accuracy: 0.8448 - loss: 0.4126
Epoch 8/30
30/30  0s 1ms/step - accuracy: 0.8631 - loss: 0.4047
Epoch 9/30
30/30  0s 2ms/step - accuracy: 0.8324 - loss: 0.4672
Epoch 10/30
30/30  0s 1ms/step - accuracy: 0.8801 - loss: 0.3886
Epoch 11/30
30/30  0s 1ms/step - accuracy: 0.8581 - loss: 0.4038
Epoch 12/30
30/30  0s 1ms/step - accuracy: 0.8716 - loss: 0.3865
Epoch 13/30
30/30  0s 1ms/step - accuracy: 0.8464 - loss: 0.4365
Epoch 14/30
30/30  0s 1ms/step - accuracy: 0.8448 - loss: 0.4275
Epoch 15/30
30/30  0s 1ms/step - accuracy: 0.8717 - loss: 0.3884
Epoch 16/30
30/30  0s 1ms/step - accuracy: 0.8266 - loss: 0.4393
Epoch 17/30
30/30  0s 1ms/step - accuracy: 0.8604 - loss: 0.4109
Epoch 18/30
30/30  0s 1ms/step - accuracy: 0.8534 - loss: 0.3963
Epoch 19/30
30/30  0s 2ms/step - accuracy: 0.8335 - loss: 0.4338
Epoch 20/30
30/30  0s 2ms/step - accuracy: 0.8147 - loss: 0.4943
Epoch 21/30

```

30/30 ————— 0s 1ms/step - accuracy: 0.8522 - loss: 0.4181
Epoch 22/30
30/30 ————— 0s 1ms/step - accuracy: 0.8120 - loss: 0.4529
Epoch 23/30
30/30 ————— 0s 1ms/step - accuracy: 0.8758 - loss: 0.3673
Epoch 24/30
30/30 ————— 0s 1ms/step - accuracy: 0.8552 - loss: 0.4039
Epoch 25/30
30/30 ————— 0s 1ms/step - accuracy: 0.8428 - loss: 0.4119
Epoch 26/30
30/30 ————— 0s 1ms/step - accuracy: 0.8740 - loss: 0.3601
Epoch 27/30
30/30 ————— 0s 1ms/step - accuracy: 0.8673 - loss: 0.3975
Epoch 28/30
30/30 ————— 0s 1ms/step - accuracy: 0.8536 - loss: 0.4084
Epoch 29/30
30/30 ————— 0s 1ms/step - accuracy: 0.8923 - loss: 0.3394
Epoch 30/30
30/30 ————— 0s 1ms/step - accuracy: 0.8464 - loss: 0.4356

```

```
In [11]: print("\n Accuracy: %.4f" % (model.evaluate(X, y)[1]))
```

```
15/15 ————— 0s 2ms/step - accuracy: 0.8313 - loss: 0.4460
```

```
Accuracy: 0.8511
```

```
In [12]: print("\n loss: %.4f" % (model.evaluate(X, y)[0]))
```

```
15/15 ————— 0s 2ms/step - accuracy: 0.8313 - loss: 0.4460
```

```
loss: 0.4053
```