

(a)

1) `tf.keras.layers.dense`

- Role: Fully(Dense) connected layer, 즉 완전연결계층을 만들어주는 역할을 합니다.

- Arguments and their roles:

Units: output space의 차원

Activation function: input->hidden, hidden->output layer 사이에서의 activation function을 무엇으로 할지 설정한다.

use_bias: bias 항을 사용할지 여부

kernel_initializer: weight 값들에 대한 initializer

bias_initializer: bias 값들에 대한 initializer

kernel(or bias, activity)_regularizer: kernel weights(or bias, activation)에 대한 정규화

kernel_constraint: kernel weight에 제약을 가하는 것

bias_constraint: 마찬가지로 bias에 제약을 가하는 것

2) `tf.keras.layers.Dropout`

- Role: 정규화의 방식중 하나인 드롭아웃을 쉽게 구현해주는 역할을 한다. Input에 적용.

- Arguments and their roles:

rate: 몇 퍼센트의 input을 dropout 시킬 것인지 설정 (0~1)

noise_shape: the shape of the binary dropout mask

seed: 임의의 seed. 정수값

(b)

기초 작업

Importing the libraries

from numpy import loadtxt

```
import numpy as np

import tensorflow as tf

import os

from keras.models import Sequential

from keras.layers import Dense

from keras.layers import Dropout

from tensorflow.keras.layers.experimental import preprocessing


# Change Working Directory

os.getcwd()

os.chdir("C:/Users/david/Downloads")


# load the dataset

dataset_train = loadtxt('mistletoeTrain.csv', delimiter=',')


# split into input (X) and output (y) variables

x_train = dataset_train[:,0:4]

y_train = dataset_train[:,4]

print(x_train.shape)

(20000, 4)

print(y_train.shape)

(20000, )


# normalization

normalizer = preprocessing.Normalization()

normalizer.adapt(np.array(x_train))
```

```
print(normalizer.mean.numpy())

model = tf.keras.models.Sequential([

    normalizer,

    tf.keras.layers.Dense(16, activation='softmax'),

    tf.keras.layers.Dense(32, activation='softmax'),

    tf.keras.layers.Dense(16, activation='softmax'),

    tf.keras.layers.Dropout(0.2),

    tf.keras.layers.Dense(1, activation='sigmoid')

])
```

```
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
=		
normalization (Normalization (None, 4))		9
<hr/>		
dense (Dense)	(None, 16)	80
<hr/>		
dense_1 (Dense)	(None, 32)	544
<hr/>		
dense_2 (Dense)	(None, 16)	528
<hr/>		
dropout (Dropout)	(None, 16)	0
<hr/>		
dense_3 (Dense)	(None, 1)	17

=====

Total params: 1,178

Trainable params: 1,169

Non-trainable params: 9

(c)

Compiling the model

```
model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])
```

Fitting the neural network to the training set

```
model.fit(x_train, y_train, epochs = 50)
```

Epoch 1/50

625/625 [=====] - 2s 2ms/step - loss: 0.4860 - accuracy: 0.8984

Epoch 2/50

625/625 [=====] - 1s 2ms/step - loss: 0.3392 - accuracy: 0.9051

Epoch 3/50

625/625 [=====] - 1s 2ms/step - loss: 0.3197 - accuracy: 0.9051

Epoch 4/50

625/625 [=====] - 1s 2ms/step - loss: 0.3130 - accuracy: 0.9051

Epoch 5/50

625/625 [=====] - 1s 2ms/step - loss: 0.3066 - accuracy:
0.9051

Epoch 6/50

625/625 [=====] - 1s 2ms/step - loss: 0.3008 - accuracy:
0.9051

Epoch 7/50

625/625 [=====] - 1s 2ms/step - loss: 0.2972 - accuracy:
0.9051

Epoch 8/50

625/625 [=====] - 1s 2ms/step - loss: 0.2948 - accuracy:
0.9051

Epoch 9/50

625/625 [=====] - 1s 2ms/step - loss: 0.2914 - accuracy:
0.9051

Epoch 10/50

625/625 [=====] - 1s 2ms/step - loss: 0.2893 - accuracy:
0.9051

Epoch 11/50

625/625 [=====] - 1s 2ms/step - loss: 0.2888 - accuracy:
0.9051

Epoch 12/50

625/625 [=====] - 1s 2ms/step - loss: 0.2867 - accuracy:
0.9051

Epoch 13/50

625/625 [=====] - 1s 2ms/step - loss: 0.2858 - accuracy:
0.9051

Epoch 14/50

625/625 [=====] - 1s 2ms/step - loss: 0.2854 - accuracy:
0.9051

Epoch 15/50

625/625 [=====] - 1s 2ms/step - loss: 0.2846 - accuracy: 0.9051

Epoch 16/50

625/625 [=====] - 1s 2ms/step - loss: 0.2834 - accuracy: 0.9051

Epoch 17/50

625/625 [=====] - 1s 2ms/step - loss: 0.2832 - accuracy: 0.9051

Epoch 18/50

625/625 [=====] - 1s 2ms/step - loss: 0.2829 - accuracy: 0.9051

Epoch 19/50

625/625 [=====] - 1s 2ms/step - loss: 0.2822 - accuracy: 0.9051

Epoch 20/50

625/625 [=====] - 1s 2ms/step - loss: 0.2809 - accuracy: 0.9051

Epoch 21/50

625/625 [=====] - 1s 2ms/step - loss: 0.2808 - accuracy: 0.9051

Epoch 22/50

625/625 [=====] - 1s 2ms/step - loss: 0.2808 - accuracy: 0.9051

Epoch 23/50

625/625 [=====] - 1s 2ms/step - loss: 0.2802 - accuracy: 0.9051

Epoch 24/50

625/625 [=====] - 1s 2ms/step - loss: 0.2799 - accuracy:
0.9051

Epoch 25/50

625/625 [=====] - 1s 2ms/step - loss: 0.2788 - accuracy:
0.9051

Epoch 26/50

625/625 [=====] - 1s 2ms/step - loss: 0.2787 - accuracy:
0.9051

Epoch 27/50

625/625 [=====] - 1s 2ms/step - loss: 0.2790 - accuracy:
0.9051

Epoch 28/50

625/625 [=====] - 1s 2ms/step - loss: 0.2781 - accuracy:
0.9051

Epoch 29/50

625/625 [=====] - 1s 2ms/step - loss: 0.2778 - accuracy:
0.9051

Epoch 30/50

625/625 [=====] - 1s 2ms/step - loss: 0.2779 - accuracy:
0.9051

Epoch 31/50

625/625 [=====] - 1s 2ms/step - loss: 0.2774 - accuracy:
0.9051

Epoch 32/50

625/625 [=====] - 1s 2ms/step - loss: 0.2771 - accuracy:
0.9051

Epoch 33/50

625/625 [=====] - 1s 2ms/step - loss: 0.2768 - accuracy:
0.9051

Epoch 34/50

625/625 [=====] - 1s 2ms/step - loss: 0.2764 - accuracy: 0.9051

Epoch 35/50

625/625 [=====] - 1s 2ms/step - loss: 0.2761 - accuracy: 0.9051

Epoch 36/50

625/625 [=====] - 1s 2ms/step - loss: 0.2753 - accuracy: 0.9051

Epoch 37/50

625/625 [=====] - 1s 2ms/step - loss: 0.2764 - accuracy: 0.9051

Epoch 38/50

625/625 [=====] - 1s 2ms/step - loss: 0.2755 - accuracy: 0.9051

Epoch 39/50

625/625 [=====] - 1s 2ms/step - loss: 0.2755 - accuracy: 0.9051

Epoch 40/50

625/625 [=====] - 1s 2ms/step - loss: 0.2754 - accuracy: 0.9051

Epoch 41/50

625/625 [=====] - 1s 2ms/step - loss: 0.2755 - accuracy: 0.9051

Epoch 42/50

625/625 [=====] - 1s 2ms/step - loss: 0.2750 - accuracy: 0.9051

Epoch 43/50

625/625 [=====] - 1s 2ms/step - loss: 0.2754 - accuracy:
0.9051

Epoch 44/50

625/625 [=====] - 1s 2ms/step - loss: 0.2751 - accuracy:
0.9051

Epoch 45/50

625/625 [=====] - 1s 2ms/step - loss: 0.2743 - accuracy:
0.9051

Epoch 46/50

625/625 [=====] - 1s 2ms/step - loss: 0.2745 - accuracy:
0.9051

Epoch 47/50

625/625 [=====] - 1s 2ms/step - loss: 0.2746 - accuracy:
0.9051

Epoch 48/50

625/625 [=====] - 1s 2ms/step - loss: 0.2744 - accuracy:
0.9051

Epoch 49/50

625/625 [=====] - 1s 2ms/step - loss: 0.2739 - accuracy:
0.9051

Epoch 50/50

625/625 [=====] - 1s 2ms/step - loss: 0.2748 - accuracy:
0.9051

Accuracy는 0.9051이다.

(d)

```
dataset_test = loadtxt('mistletoeTest.csv', delimiter=',')
```

```
x_test = dataset_train[:,0:4]
```

```
y_test = dataset_train[:,4]
```

```
model.evaluate(x_test,y_test, verbose=True)
```

```
170/170 [=====] - 1s 2ms/step - loss: 0.4532 - accuracy:  
0.8205
```

```
Out[20]: [0.45324593782424927, 0.8204750418663025]
```

Accuracy는 0.8205이다. 이때 test data에서는 dropout을 하지 않는다.