

# 13강. 과적합 피하기

※ 인자역전파 정략도

<실습>

```
1 from keras.models import Sequential
2 from keras.layers.core import Dense
3 from sklearn.preprocessing import LabelEncoder
4
5 import pandas as pd
6 import numpy
7 import tensorflow as tf
8
9 # seed 값 설정
10 numpy.random.seed(3)
11 tf.random.set_seed(3)
12
13 # 데이터 입력
14 df = pd.read_csv('sonar.csv', header=None)
15
16 # 데이터 개괄 보기
17 print(df.info())
18
19 # 데이터의 일부분 미리 보기
20 print(df.head())
```

(데이터 확인)

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 208 entries, 0 to 207
Data columns (total 61 columns):
#   Column  Non-Null Count  Dtype
---
```

#	Column	Non-Null Count	Dtype
0	0	208 non-null	float64
1	1	208 non-null	float64
2	2	208 non-null	float64
3	3	208 non-null	float64
4	4	208 non-null	float64
5	5	208 non-null	float64
6	6	208 non-null	float64
7	7	208 non-null	float64
8	8	208 non-null	float64
9	9	208 non-null	float64
10	10	208 non-null	float64
11	11	208 non-null	float64
12	12	208 non-null	float64
13	13	208 non-null	float64
14	14	208 non-null	float64
15	15	208 non-null	float64
16	16	208 non-null	float64
17	17	208 non-null	float64
18	18	208 non-null	float64
19	19	208 non-null	float64

```
dtypes: float64(60), object(1)
memory usage: 99.2+ KB
None
```

	0	1	2	3	4	...	56	57	58	59	60
0	0.00200	0.0371	0.0428	0.0207	0.0954	...	0.0180	0.0084	0.0090	0.0032	R
1	0.0453	0.0523	0.0843	0.0689	0.1183	...	0.0140	0.0049	0.0052	0.0044	R
2	0.0262	0.0502	0.1099	0.1083	0.0974	...	0.0316	0.0164	0.0095	0.0078	R
3	0.0100	0.0171	0.0623	0.0205	0.0205	...	0.0050	0.0044	0.0040	0.0117	R
4	0.0762	0.0666	0.0481	0.0394	0.0590	...	0.0072	0.0048	0.0107	0.0094	R

[5 rows x 61 columns]

```

1 dataset = df.values
2 X = dataset[:,0:60].astype(float) #astype 추가해야함. 버전때문,,
3 Y_obj = dataset[:,60]
4
5 # 문자열 변환
6 e = LabelEncoder()
7 e.fit(Y_obj)
8 Y = e.transform(Y_obj)
9
10 # 모델 설정
11 model = Sequential()
12 model.add(Dense(24, input_dim=60, activation='relu'))
13 model.add(Dense(10, activation='relu'))
14 model.add(Dense(1, activation='sigmoid'))
15
16 # 모델 컴파일
17 model.compile(loss='mean_squared_error',
18               optimizer='adam',
19               metrics=['accuracy'])
20
21 # 모델 실행
22 model.fit(X, Y, epochs=200, batch_size=5)
23
24 # 결과 출력
25 print("\n Accuracy: %.4f" % (model.evaluate(X, Y)[1]))

```

(결과)

```

Epoch 189/200
42/42 [=====] - 0s 1ms/step - loss: 5.9846e-04 - accuracy:
Epoch 190/200
42/42 [=====] - 0s 1ms/step - loss: 5.9450e-04 - accuracy:
Epoch 191/200
42/42 [=====] - 0s 1ms/step - loss: 5.3336e-04 - accuracy:
Epoch 192/200
42/42 [=====] - 0s 1ms/step - loss: 5.6500e-04 - accuracy:
Epoch 193/200
42/42 [=====] - 0s 1ms/step - loss: 5.4481e-04 - accuracy:
Epoch 194/200
42/42 [=====] - 0s 1ms/step - loss: 5.9656e-04 - accuracy:
Epoch 195/200
42/42 [=====] - 0s 1ms/step - loss: 7.1917e-04 - accuracy:
Epoch 196/200
42/42 [=====] - 0s 1ms/step - loss: 5.4751e-04 - accuracy:
Epoch 197/200
42/42 [=====] - 0s 1ms/step - loss: 4.3103e-04 - accuracy:
Epoch 198/200
42/42 [=====] - 0s 1ms/step - loss: 3.8134e-04 - accuracy:
Epoch 199/200
42/42 [=====] - 0s 1ms/step - loss: 4.9843e-04 - accuracy:
Epoch 200/200
42/42 [=====] - 0s 1ms/step - loss: 4.2226e-04 - accuracy:
7/7 [=====] - 0s 2ms/step - loss: 4.7309e-04 - accuracy: 1

```

Accuracy: 1.0000

• 과적합! 모델이 학습 데이터셋 안에서는 일정수준 이상의 예측정확도를 보이지만,  
새로운 데이터에 적용하면 잘 맞지 않는 것.

※ 과적합을 피하기 위해서는?

- 학습 데이터셋과 테스트 데이터셋을 랜덤히 구분하여 학습과 테스트를 병행해야 함

ex) 70개는 학습셋, 30개는 테스트셋

※ 학습이 깊어져서 학습셋 내부에서의 성공률은 높아져도 테스트셋에서 효과가  
없다면 과적합이 일어나는 것!

<실습: 학습셋과 테스트셋 구분>

```
1 dataset = df.values
2 X = dataset[:,0:60].astype(float)
3 Y_obj = dataset[:,60]
4
5 e = LabelEncoder()
6 e.fit(Y_obj)
7 Y = e.transform(Y_obj)
8
9 # 학습 셋과 테스트 셋의 구분
10 X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.3,
11                                                    random_state=seed)
12
13 model = Sequential()
14 model.add(Dense(24, input_dim=60, activation='relu'))
15 model.add(Dense(10, activation='relu'))
16 model.add(Dense(1, activation='sigmoid'))
17
18 model.compile(loss='mean_squared_error',
19               optimizer='adam',
20               metrics=['accuracy'])
21
22 model.fit(X_train, Y_train, epochs=130, batch_size=5)
23
24 # 테스트셋에 모델 적용
25 print("\n Test Accuracy: %.4f" % (model.evaluate(X_test, Y_test)[1]))
```



(결과)

```
Epoch 127/130
29/29 [=====] - 0s 1ms/step - loss: 0.0294 - accuracy: 0.96
Epoch 128/130
29/29 [=====] - 0s 2ms/step - loss: 0.0225 - accuracy: 0.96
Epoch 129/130
29/29 [=====] - 0s 1ms/step - loss: 0.0239 - accuracy: 0.97
Epoch 130/130
29/29 [=====] - 0s 1ms/step - loss: 0.0190 - accuracy: 0.96
2/2 [=====] - 0s 8ms/step - loss: 0.1386 - accuracy: 0.8254
```

Test Accuracy: 0.8254

## <모델 저장 및 재사용>

```
1 from keras.models import Sequential, load_model
```

```
1 model.save('my_model.h5') # 모델을 컴퓨터에 저장
```

```
1 del model # 테스트를 위해 메모리 내의 모델을 삭제  
2 model = load_model('my_model.h5') # 모델을 새로 불러옴  
3  
4 # 불러온 모델로 테스트 실행  
5 print("\n Test Accuracy: %.4f" % (model.evaluate(X_test, Y_test)[1]))
```

(결과)

```
2/2 [=====] - 0s 4ms/step - loss: 0.1557 - accuracy: 0.825
```

```
Test Accuracy: 0.8254
```

• 테스트가 횡단권 없음 → k접 교차 검증

↳ 데이터셋을 여러개씩 나누어 하나씩 테스트셋으로 사용하고, 나머지를 모두 합해서 학습셋으로 사용하는 방법

## <실습>

```
1 from sklearn.model_selection import StratifiedKFold
```

```
1 # 10개의 파일로 쪼갬
```

```
2 n_fold = 10
```

```
3 skf = StratifiedKFold(n_splits=n_fold, shuffle=True, random_state=seed)
```

```
4
```

```
5 # 빈 accuracy 배열
```

```
6 accuracy = []
```

```
7
```

```
8 # 모델의 설정, 컴파일, 실행
```

```
9 for train, test in skf.split(X, Y):
```

```
10     model = Sequential()
```

```
11     model.add(Dense(24, input_dim=60, activation='relu'))
```

```
12     model.add(Dense(10, activation='relu'))
```

```
13     model.add(Dense(1, activation='sigmoid'))
```

```
14     model.compile(loss='mean_squared_error',
```

```
15                     optimizer='adam',
```

```
16                     metrics=['accuracy'])
```

```
17     model.fit(X[train], Y[train], epochs=100, batch_size=5)
```

```
18     k_accuracy = "%.4f" % (model.evaluate(X[test], Y[test])[1])
```

```
19     accuracy.append(k_accuracy)
```

```
20
```

```
21 # 결과 출력
```

```
22 print("\n %.f fold accuracy:" % n_fold, accuracy)
```

(결과)

```
Epoch 96/100
```

```
38/38 [=====] - 0s 2ms/step - loss: 0.0184 - accuracy: 0.9
```

```
Epoch 97/100
```

```
38/38 [=====] - 0s 2ms/step - loss: 0.0323 - accuracy: 0.9
```

```
Epoch 98/100
```

```
38/38 [=====] - 0s 2ms/step - loss: 0.0365 - accuracy: 0.9
```

```
Epoch 99/100
```

```
38/38 [=====] - 0s 2ms/step - loss: 0.0271 - accuracy: 0.9
```

```
Epoch 100/100
```

```
38/38 [=====] - 0s 2ms/step - loss: 0.0270 - accuracy: 0.9
```

```
WARNING:tensorflow:11 out of the last 12 calls to <function Model.make_test_funcio
```

```
1/1 [=====] - 0s 118ms/step - loss: 0.1398 - accuracy: 0.8
```

```
10 fold accuracy: ['0.7143', '0.7619', '0.7619', '0.9048', '0.8095', '0.8095', '0.
```