▼ Neural Network Model - 다중분류

```
import warnings
warnings.filterwarnings('ignore')
```

▼ 실습용 데이터 설정

iris.csv

```
import seaborn as sns

DF = sns.load_dataset('iris')
```

pandas DataFrame

DF.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 5 columns):
    Column
                Non-Null Count Dtype
   sepal_length 150 non-null
                                  float64
0
    sepal_width 150 non-null
                                  float64
 1
2 petal_length 150 non-null
                                  float64
                                  float64
    petal_width 150 non-null
                 150 non-null
                                  object
    species
dtypes: float64(4), object(1)
memory usage: 6.0+ KB
```

DF.head(3)

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa

▼ I. 탐색적 데이터 분석

▼ 1) 빈도분석

```
DF.species.value_counts()
```

```
setosa 50
virginica 50
versicolor 50
```

Name: species, dtype: int64

▼ 2) 분포 시각화

```
import matplotlib.pyplot as plt
import seaborn as sns
sns.pairplot(hue = 'species', data = DF)
plt.show()
```

II. Data Preprocessing

→ 1) Data Set

```
X = DF[['sepal_length', 'sepal_width', 'petal_length', 'petal_width']]
y = DF['species']
```

→ 2) Train & Test Split

• 7:3

→ III. Modeling

▼ 1) Train_Data로 모델 생성

• hidden_layer_sizes : 은닉층 노드의 개수

activation : 활성화 함수
solver : 최적화 기법

• max_iter: 학습 반복 횟수

```
MLPClassifier (activation='logistic', alpha=0.0001, batch_size='auto', beta_1=0.9, beta_2=0.999, early_stopping=False, epsilon=1e-08, hidden_layer_sizes=5, learning_rate='constant', learning_rate_init=0.001, max_fun=15000, max_iter=5000, momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True, power_t=0.5, random_state=2045, shuffle=True, solver='adam', tol=0.0001, validation_fraction=0.1, verbose=False, warm_start=False)
```

▼ 2) Test_Data에 Model 적용

```
y_hat = Model_NN.predict(X_test)
```

→ 3) Confusion Matrix

→ 4) Accuracy

```
print('%.8f' % accuracy_score(y_test, y_hat))
```

1.00000000

→ 5) Classification Report

support	f1-score	recall	precision	
17 14 14	1.00000 1.00000 1.00000	1.00000 1.00000 1.00000	1.00000 1.00000 1.00000	setosa versicolor virginica
45 45 45	1.00000 1.00000 1.00000	1.00000	1.00000	accuracy macro avg weighted avg

#

#

#

The End

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#

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