## module 불러오기

Out[3]:

array(['malignant', 'benign'], dtype='<U9')</pre>

```
In [ ]:
    # data load
 1
    from sklearn.datasets import load breast cancer
 3
 4
    # train test split
    from sklearn.model selection import train test split
 5
    # model
 7
    from sklearn.tree import DecisionTreeClassifier
 8
    from sklearn.ensemble import RandomForestClassifier
    from sklearn.svm import SVC
10
    from sklearn.linear model import SGDClassifier
11
    from sklearn.linear model import LogisticRegression
12
13
    # report
14
15 from sklearn.metrics import classification report
breast cancer 데이터 불러오기
In [ ]:
 1 breast cancer = load breast cancer()
    print(breast cancer)
In [ ]:
    breast cancer.feature names
Out[16]:
array(['mean radius', 'mean texture', 'mean perimeter', 'mean area',
       'mean smoothness', 'mean compactness', 'mean concavity',
       'mean concave points', 'mean symmetry', 'mean fractal dimension',
       'radius error', 'texture error', 'perimeter error', 'area error',
       'smoothness error', 'compactness error', 'concavity error',
       'concave points error', 'symmetry error',
       'fractal dimension error', 'worst radius', 'worst texture',
       'worst perimeter', 'worst area', 'worst smoothness',
       'worst compactness', 'worst concavity', 'worst concave points',
       'worst symmetry', 'worst fractal dimension'], dtype='<U23')
In [ ]:
 1 breast_cancer.target_names
```

```
In [ ]:

1  data = breast_cancer.data
2  target = breast_cancer.target
```

# train, test 데이터 분리

```
In [ ]:
    X train, X test, y train, y test = train test split(data, target, test size=0.2, random state=7)
    X train
Out[5]:
array([[1.242e+01, 1.504e+01, 7.861e+01, ..., 4.052e-02, 2.901e-01,
        6.783e-02],
       [1.825e+01, 1.998e+01, 1.196e+02, ..., 1.932e-01, 3.063e-01,
       8.368e-02],
       [1.454e+01, 2.754e+01, 9.673e+01, ..., 1.712e-01, 4.218e-01,
        1.341e-01],
       [1.727e+01, 2.542e+01, 1.124e+02, ..., 1.739e-01, 2.500e-01,
       7.944e-02],
       [1.185e+01, 1.746e+01, 7.554e+01, ..., 9.140e-02, 3.101e-01,
       7.007e-02],
       [1.403e+01, 2.125e+01, 8.979e+01, ..., 7.963e-02, 2.226e-01,
        7.617e-02]])
```

# 각 모델들 학습 시키기

```
In [ ]:
```

```
# DecisionTree
model_tree = DecisionTreeClassifier()
model_tree.fit(X_train, y_train)
```

#### Out[6]:

```
In [ ]:
```

```
# RandomForest
model_random_forest = RandomForestClassifier()
model_random_forest.fit(X_train, y_train)
```

#### Out[7]:

```
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None, criterion='gini', max_depth=None, max_features='auto', max_leaf_nodes=None, max_samples=None, min_impurity_decrease=0.0, min_impurity_split=None, min_samples_leaf=1, min_samples_split=2, min_weight_fraction_leaf=0.0, n_estimators=100, n_jobs=None, oob_score=False, random_state=None, verbose=0, warm start=False)
```

#### In [ ]:

```
1 # SVM
2 model_svc = SVC()
3 model_svc.fit(X_train, y_train)
```

#### Out[8]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

#### In [ ]:

```
1 # SGD
2 model_sgd = SGDClassifier()
3 model_sgd.fit(X_train, y_train)
```

#### Out[9]:

```
SGDClassifier(alpha=0.0001, average=False, class_weight=None, early_stopping=False, epsilon=0.1, eta0=0.0, fit_intercept=True, l1_ratio=0.15, learning_rate='optimal', loss='hinge', max_iter=1000, n_iter_no_change=5, n_jobs=None, penalty='l2', power_t=0.5, random_state=None, shuffle=True, tol=0.001, validation_fraction=0.1, verbose=0, warm_start=False)
```

```
In [ ]:
```

```
# Logistic Regression
model_LR = LogisticRegression()
model_LR.fit(X_train, y_train)
```

/usr/local/lib/python3.7/dist-packages/sklearn/linear\_model/\_logistic.py:940: ConvergenceWarning: lbfgs failed to converge (status=1): STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max\_iter) or scale the data as shown in:
 https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-learn.org/stable/modules/preprocessing.html)

Please also refer to the documentation for alternative solver options:

https://scikit-learn.org/stable/modules/linear\_model.html#logistic-regression (h
ttps://scikit-learn.org/stable/modules/linear\_model.html#logistic-regression)
extra\_warning\_msg=\_LOGISTIC\_SOLVER\_CONVERGENCE\_MSG)

#### Out[10]:

LogisticRegression(C=1.0, class\_weight=None, dual=False, fit\_intercept=True, intercept\_scaling=1, l1\_ratio=None, max\_iter=100, multi\_class='auto', n\_jobs=None, penalty='l2', random\_state=None, solver='lbfgs', tol=0.0001, verbose=0, warm start=False)

# 모델 사용해보고 평가하기

#### In [ ]:

1 # Decision Tree
2 y\_pred = model\_tree.predict(X\_test)
3 print(classification\_report(y\_pred, y\_test))

	precision	recall	f1-score	support
0	0.92	0.88	0.90	41
1	0.93	0.96	0.95	73
accuracy			0.93	114
macro avg	0.93	0.92	0.92	114
weighted avg	0.93	0.93	0.93	114

#### In [ ]:

# Random Forest
y\_pred = model\_random\_forest.predict(X\_test)
print(classification\_report(y\_pred, y\_test))

	precision	recall	f1-score	support
0	0.87	0.94	0.91	36
1	0.97	0.94	0.95	78
accuracy			0.94	114
macro avg	0.92	0.94	0.93	114
weighted avg	0.94	0.94	0.94	114

```
In [ ]:
```

```
# SVM
y_pred = model_svc.predict(X_test)
print(classification_report(y_pred, y_test))
```

	precision	recall	f1-score	support
0	0.77	0.97	0.86	31
1	0.99	0.89	0.94	83
accuracy			0.91	114
macro avg	0.88	0.93	0.90	114
weighted avg	0.93	0.91	0.92	114

### In [ ]:

```
# SGD
y_pred = model_sgd.predict(X_test)
print(classification_report(y_pred, y_test))
```

	precision	recall	f1-score	support
0	0.92	0.77	0.84	47
1	0.85	0.96	0.90	67
accuracy			0.88	114
macro avg	0.89	0.86	0.87	114
weighted avg	0.88	0.88	0.87	114

#### In [ ]:

```
# Logistic Regression
y_pred = model_LR.predict(X_test)
print(classification_report(y_pred, y_test))
```

	precision	recall	f1-score	support
0 1	0.87 0.97	0.94 0.94	0.91 0.95	36 78
accuracy macro avg	0.92	0.94	0.94 0.93	114 114
weighted avg	0.94	0.94	0.94	114

Random Forest, logistic Regression을 사용한다. 암은 반드시 발견되어야 하고, 한명의 환자도 놓치면 안되며 recall이 안정적인 모델을 택하여야한다.

### In [ ]:

1

### In [ ]:

1

In [ ]:

1