

# PerceptronAdeline

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## 1 ANN

### 1.1 Adeline class yang dibuat tanpa library

```
[1]: class CustomAdaline(object):

    def __init__(self, n_iterations=100, random_state=1, learning_rate=0.01):
        self.n_iterations = n_iterations
        self.random_state = random_state
        self.learning_rate = learning_rate

    '''
    Batch Gradient Descent

    1. Weights are updated considering all training examples.
    2. Learning of weights can continue for multiple iterations
    3. Learning rate needs to be defined
    '''

    def fit(self, X, y):
        rgen = np.random.RandomState(self.random_state)
        self.coef_ = rgen.normal(loc=0.0, scale=0.01, size=1 + X.shape[1])
        for _ in range(self.n_iterations):
            activation_function_output = self.activation_function(self.
→net_input(X))
            errors = y - activation_function_output
            self.coef_[1:] = self.coef_[1:] + self.learning_rate*X.T.
→dot(errors)
            self.coef_[0] = self.coef_[0] + self.learning_rate*errors.sum()

    '''
    Net Input is sum of weighted input signals
    '''

    def net_input(self, X):
        weighted_sum = np.dot(X, self.coef_[1:]) + self.coef_[0]
        return weighted_sum

    '''
```

```

Activation function is fed the net input. As the activation function is
an identity function, the output from activation function is same as the
input to the function.
'''
def activation_function(self, X):
    return X

'''
Prediction is made on the basis of output of activation function
'''
def predict(self, X):
    return np.where(self.activation_function(self.net_input(X)) >= 0.0, 1, 0)

'''
Model score is calculated based on comparison of
expected value and predicted value
'''
def score(self, X, y):
    misclassified_data_count = 0
    for xi, target in zip(X, y):
        output = self.predict(xi)
        if(target != output):
            misclassified_data_count += 1
    total_data_count = len(X)
    self.score_ = (total_data_count - misclassified_data_count)/
→total_data_count
    return self.score_

```

## 1.2 Perceptron class yang dibuat tanpa library

```

[2]: import numpy as np
#
# Perceptron implementation
#
class CustomPerceptron(object):

    def __init__(self, n_iterations=100, random_state=1, learning_rate=0.01):
        self.n_iterations = n_iterations
        self.random_state = random_state
        self.learning_rate = learning_rate

    '''
    Stochastic Gradient Descent

    1. Weights are updated based on each training examples.
    2. Learning of weights can continue for multiple iterations
    3. Learning rate needs to be defined

```

```

'''
def fit(self, X, y):
    rgen = np.random.RandomState(self.random_state)
    self.coef_ = rgen.normal(loc=0.0, scale=0.01, size=1 + X.shape[1])
    self.errors_ = []
    for _ in range(self.n_iterations):
        errors = 0
        for xi, expected_value in zip(X, y):
            predicted_value = self.predict(xi)
            self.coef_[1:] = self.coef_[1:] + self.learning_rate *
→(expected_value - predicted_value) * xi
            self.coef_[0] = self.coef_[0] + self.learning_rate *
→(expected_value - predicted_value) * 1
            update = self.learning_rate * (expected_value - predicted_value)
            errors += int(update != 0.0)
            self.errors_.append(errors)
'''

Net Input is sum of weighted input signals
'''
def net_input(self, X):
    weighted_sum = np.dot(X, self.coef_[1:]) + self.coef_[0]
    return weighted_sum

'''

Activation function is fed the net input and the unit step function
is executed to determine the output.
'''
def activation_function(self, X):
    weighted_sum = self.net_input(X)
    return np.where(weighted_sum >= 0.0, 1, 0)

'''

Prediction is made on the basis of output of activation function
'''
def predict(self, X):
    return self.activation_function(X)

'''

Model score is calculated based on comparison of
expected value and predicted value
'''
def score(self, X, y):
    misclassified_data_count = 0
    for xi, target in zip(X, y):
        output = self.predict(xi)
        if(target != output):
            misclassified_data_count += 1

```

```

        total_data_count = len(X)
        self.score_ = (total_data_count - misclassified_data_count)/
→total_data_count
        return self.score_

```

### 1.3 Implementasi code untuk Perceptron dan Adeline

Pada implementasinya, kelas custom model perceptron dan adeline akan dipanggil

```

[3]: from sklearn.datasets import load_breast_cancer
      from sklearn.model_selection import train_test_split
      import numpy as np

      data = load_breast_cancer()
      X = data.data
      y = data.target
      #
      # Create training and test split
      #
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,
→random_state=42, stratify=y)
      #
      # Instantiate CustomPerceptron
      #
      adaline = CustomAdaline(n_iterations = 10)
      #
      # Fit the model
      #
      adaline.fit(X_train, y_train)
      #
      # Score the model
      #
      adaline.score(X_test, y_test),adaline.score(X_train, y_train)

```

[3]: (0.6257309941520468, 0.628140703517588)

```

[4]: import numpy as np
      from sklearn import datasets
      from sklearn.model_selection import train_test_split
      #
      # Load the data set
      #
      bc = datasets.load_breast_cancer()
      X = bc.data
      y = bc.target
      #
      # Create training and test split

```

```
#
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,
→random_state=42, stratify=y)
#
# Instantiate CustomPerceptron
#
prcptrn = CustomPerceptron(n_iterations=10)
#
# Fit the model
#
prcptrn.fit(X_train, y_train)
#
# Score the model
#
prcptrn.score(X_test, y_test), prcptrn.score(X_train, y_train)
```

[4]: (0.8888888888888888, 0.9120603015075377)

## 1.4 SKLEARN Implementation

### 1.4.1 Implementasi code untuk Perceptron dan Adeline dengan menggunakan Library

**Perceptron** Import fungsi Perceptron yang berada di Library sklearn, kemudian deklarasikan variable model perceptron lalu melakukan process training.

```
[5]: from sklearn.linear_model import Perceptron
p = Perceptron()
p.fit(X_train, y_train)
```

[5]: Perceptron()

```
[6]: from sklearn.metrics import accuracy_score
predictions_train = p.predict(X_train)
predictions_test = p.predict(X_test)
train_score = accuracy_score(predictions_train, y_train)
print("score on train data: ", train_score)
test_score = accuracy_score(predictions_test, y_test)
print("score on test data: ", test_score)
```

score on train data: 0.9271356783919598  
score on test data: 0.8830409356725146

### MLP Implementation

```
[7]: from sklearn.neural_network import MLPClassifier
clf = MLPClassifier()
clf.fit(X_train, y_train)
```

/Users/istiqomah/opt/anaconda3/lib/python3.8/site-packages/sklearn/neural\_network/\_multilayer\_perceptron.py:686:

ConvergenceWarning: Stochastic Optimizer: Maximum iterations (200) reached and the optimization hasn't converged yet.

```
warnings.warn(
```

```
[7]: MLPClassifier()
```

```
[8]: from sklearn.metrics import accuracy_score
      predictions_train = clf.predict(X_train)
      predictions_test = clf.predict(X_test)
      train_score = accuracy_score(predictions_train, y_train)
      print("score on train data: ", train_score)
      test_score = accuracy_score(predictions_test, y_test)
      print("score on test data: ", test_score)
```

```
score on train data:  0.9396984924623115
```

```
score on test data:  0.9064327485380117
```

```
[9]: from sklearn.neural_network import MLPClassifier
      clf = MLPClassifier(hidden_layer_sizes=(150,100))
      clf.fit(X_train, y_train)
```

```
[9]: MLPClassifier(hidden_layer_sizes=(150, 100))
```

```
[10]: from sklearn.metrics import accuracy_score
       predictions_train = clf.predict(X_train)
       predictions_test = clf.predict(X_test)
       train_score = accuracy_score(predictions_train, y_train)
       print("score on train data: ", train_score)
       test_score = accuracy_score(predictions_test, y_test)
       print("score on test data: ", test_score)
```

```
score on train data:  0.9296482412060302
```

```
score on test data:  0.9122807017543859
```

```
[ ]:
```