

## ▼ Linear regression

### ▼ import library

```
import numpy as np
import matplotlib.image as img
import matplotlib.pyplot as plt
import matplotlib.colors as colors
from mpl_toolkits.mplot3d import Axes3D
```

```
from google.colab import drive
drive.mount('/content/drive')
```

Mounted at /content/drive

```
cd /content/drive/MyDrive/ML
```

/content/drive/MyDrive/ML

### ▼ load data points

- $\{(x_i, y_i)\}_{i=1}^n$

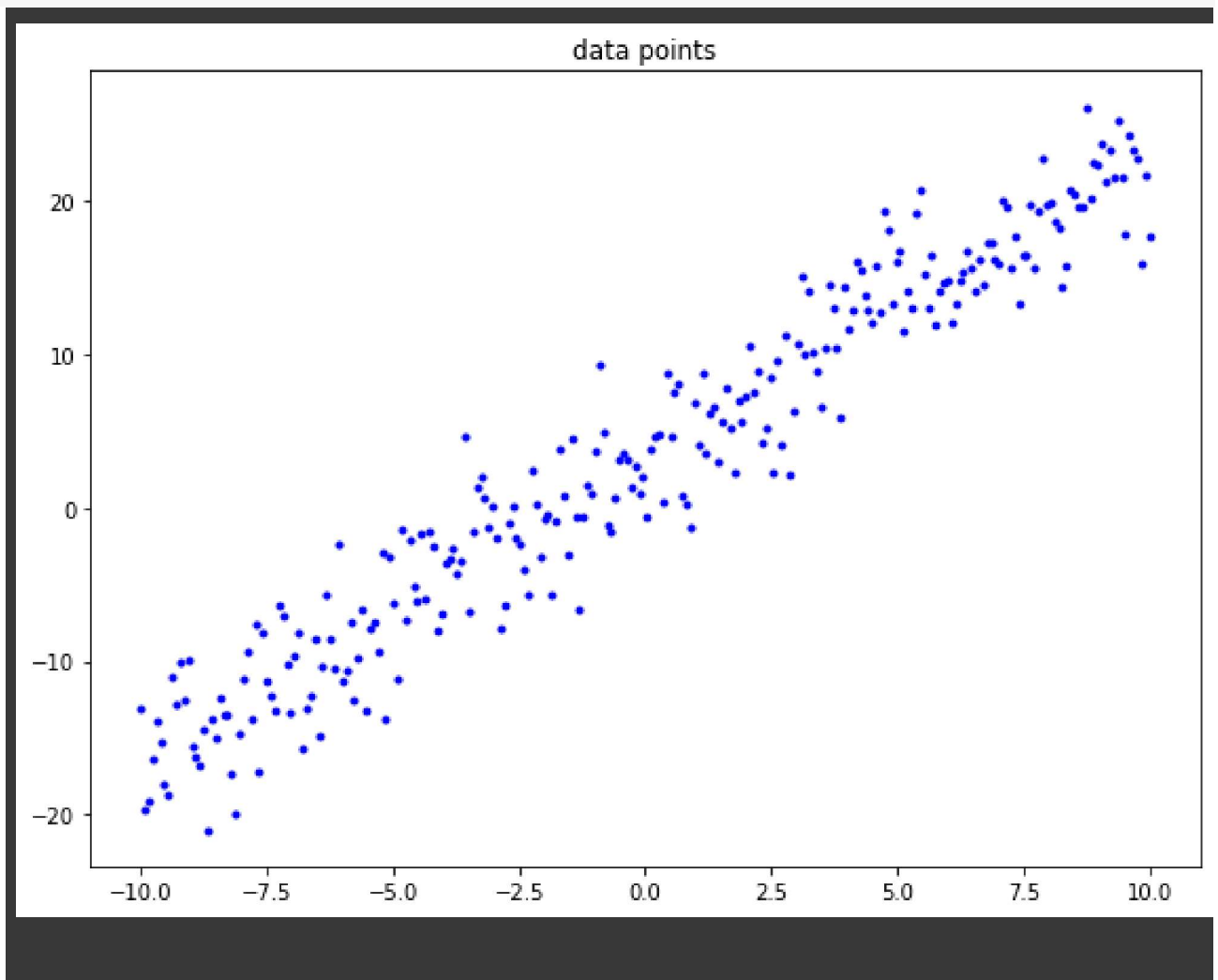
```
filename    = 'assignment_06_data.csv'
data        = np.loadtxt(filename, delimiter = ',')

x           = data[0, :]    # independent variable
y           = data[1, :]    # dependent variable

plt.figure(figsize=(8,6))

plt.plot(x, y, '.', color = 'blue')
plt.title('data points')

plt.tight_layout()
plt.show()
```



## ▼ compute the residual

```
def compute_residual(x, y, theta0, theta1):  
  
    n          = len(x)  
    residual    = np.zeros(n)  
  
    # ++++++  
    # complete the blanks  
    #  
  
    residual = theta0 + theta1*x - y  
  
    #  
    # ++++++  
  
    return residual
```

## ▼ compute the loss

- useful functions: `np.inner`

```
def compute_loss(x, y, theta0, theta1):  
  
    n          = len(x)  
    loss       = 0  
  
    # ++++++  
    # complete the blanks  
    #  
    r = compute_residual(x, y, theta0, theta1)  
    loss = np.sum(1/2/n*(np.inner(r , r)))  
  
    #  
    # ++++++  
  
    return loss
```

## ▼ compute the gradient with respect to $\theta_0$

- useful functions: `np.inner`

```
def compute_gradient_theta0(x, y, theta0, theta1):  
  
    n          = len(x)  
    derivative  = 0  
  
    # ++++++  
    # complete the blanks  
    #  
    temp = np.ones(n)  
    derivative = 1/n*(np.inner(compute_residual(x,y,theta0,theta1), temp))  
  
    #  
    # ++++++  
  
    return derivative
```

## ▼ compute the gradient with respect to $\theta_1$

- **useful functions:** `np.inner`

```
def compute_gradient_theta1(x, y, theta0, theta1):

    n          = len(x)
    derivative  = 0

    # ++++++
    # complete the blanks
    #

    derivative = 1/n*(np.inner(compute_residual(x,y,theta0,theta1),x))

    #
    # ++++++

    return derivative
```

## ▼ gradient descent for the optimization

```
number_iteration    = 500
learning_rate       = 0.01

theta0              = 0
theta1              = 0

list_theta0         = np.zeros(number_iteration)
list_theta1         = np.zeros(number_iteration)
list_loss           = np.zeros(number_iteration)

for i in range(number_iteration):

    # ++++++
    # complete the blanks
    #
    theta0 = theta0-learning_rate*compute_gradient_theta0(x,y,theta0, theta1)
    theta1 = theta1-learning_rate*compute_gradient_theta1(x,y,theta0, theta1)
    loss   = compute_loss(x,y,theta0, theta1)
    #
    # ++++++

    list_theta0[i] = theta0
```

```
list_theta1[i] = theta1
```

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## ▼ functions for presenting the results

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```
def function_result_01():  
  
    plt.figure(figsize=(8,6))  
    plt.plot(x, y, '.', color='blue')  
    plt.title('data points')  
    plt.show()
```

```
def function_result_02():  
  
    plt.figure(figsize=(8,6))  
    ax = plt.gca()  
    plt.plot(list_theta0, '-', color='blue', label=r'$w_{theta_0}$')  
    plt.plot(list_theta1, '-', color='red', label=r'$w_{theta_1}$')  
    plt.title('model parameters')  
    ax.legend()  
    plt.show()
```

```
def function_result_03():  
  
    plt.figure(figsize=(8,6))  
    plt.plot(list_loss, '-', color='blue')  
    plt.title('loss curve')  
    plt.show()
```

```
def function_result_04():  
  
    f = theta0 + theta1 * x
```

```
plt.figure(figsize=(8,6))
ax = plt.gca()
plt.plot(x, y, '.', color='blue', label='data point')
plt.plot(x, f, '-', color='red', label='regression')
plt.title('regression')
ax.legend()
plt.show()
```

```
def function_result_05():
```

```
    X0 = np.arange(-10, 10, 0.1)
    X1 = np.arange(-10, 10, 0.1)
```

```
    grid_theta0, grid_theta1 = np.meshgrid(X0, X1)
```

```
    grid_loss = np.zeros(grid_theta0.shape)
```

```
    for i, t0 in enumerate(X0):
        for j, t1 in enumerate(X1):
```

```
            grid_loss[j, i] = compute_loss(x, y, t0, t1)
```

```
    fig = plt.figure(figsize=(8,6))
```

```
    ax = fig.add_subplot(111, projection='3d')
    plt.title('loss surface')
```

```
    ax = plt.axes(projection='3d')
```

```
    ax.set_xlabel(r'$\theta_0$')
    ax.set_ylabel(r'$\theta_1$')
    ax.set_zlabel('loss')
```

```
    ax.plot_surface(grid_theta0, grid_theta1, grid_loss, rstride=1, cstride=1, cmap=
```

```
plt.tight_layout()
plt.show()
```

## ▼ results

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```
number_result = 5

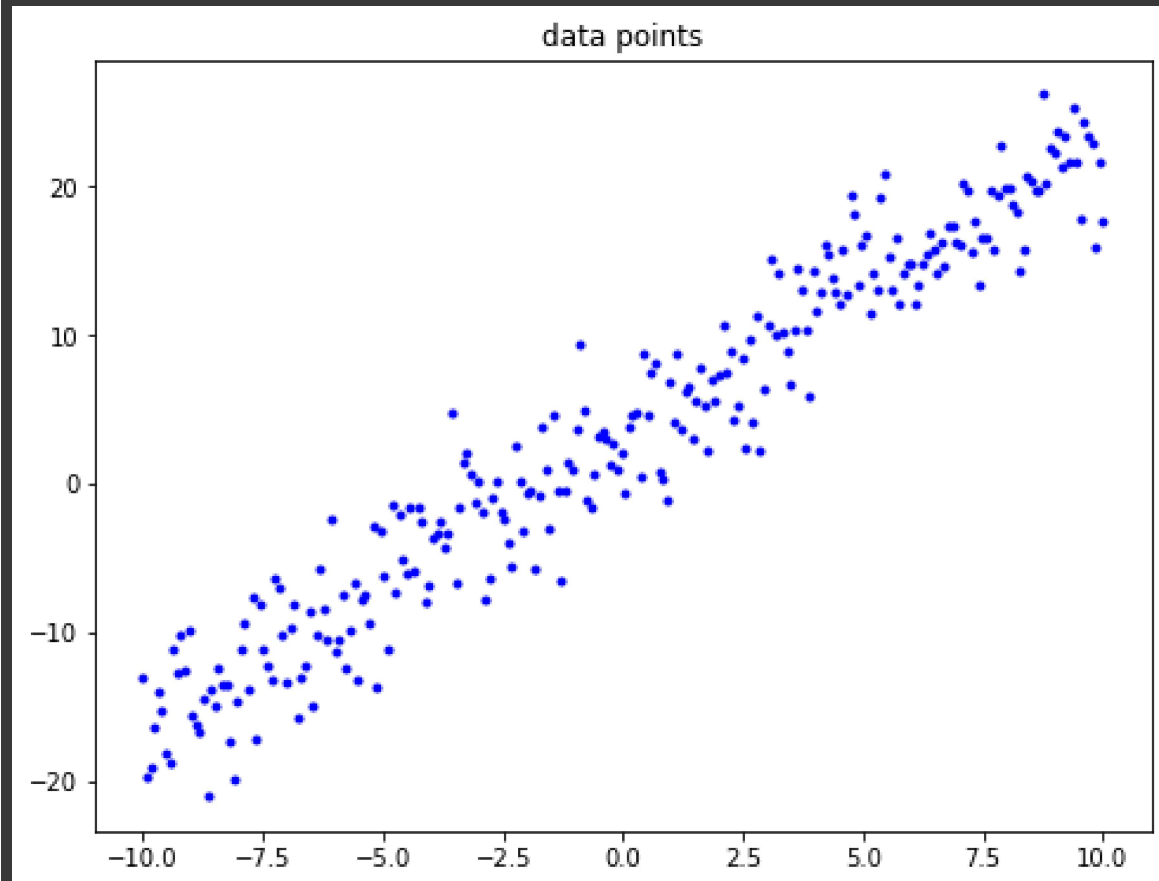
for i in range(number_result):
    title = '## [RESULT {:02d}]'.format(i+1)
    name_function = 'function_result_{:02d}()'.format(i+1)

    print('*****')
    print(title)
    print('*****')
    eval(name_function)
```

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## [RESULT 01]

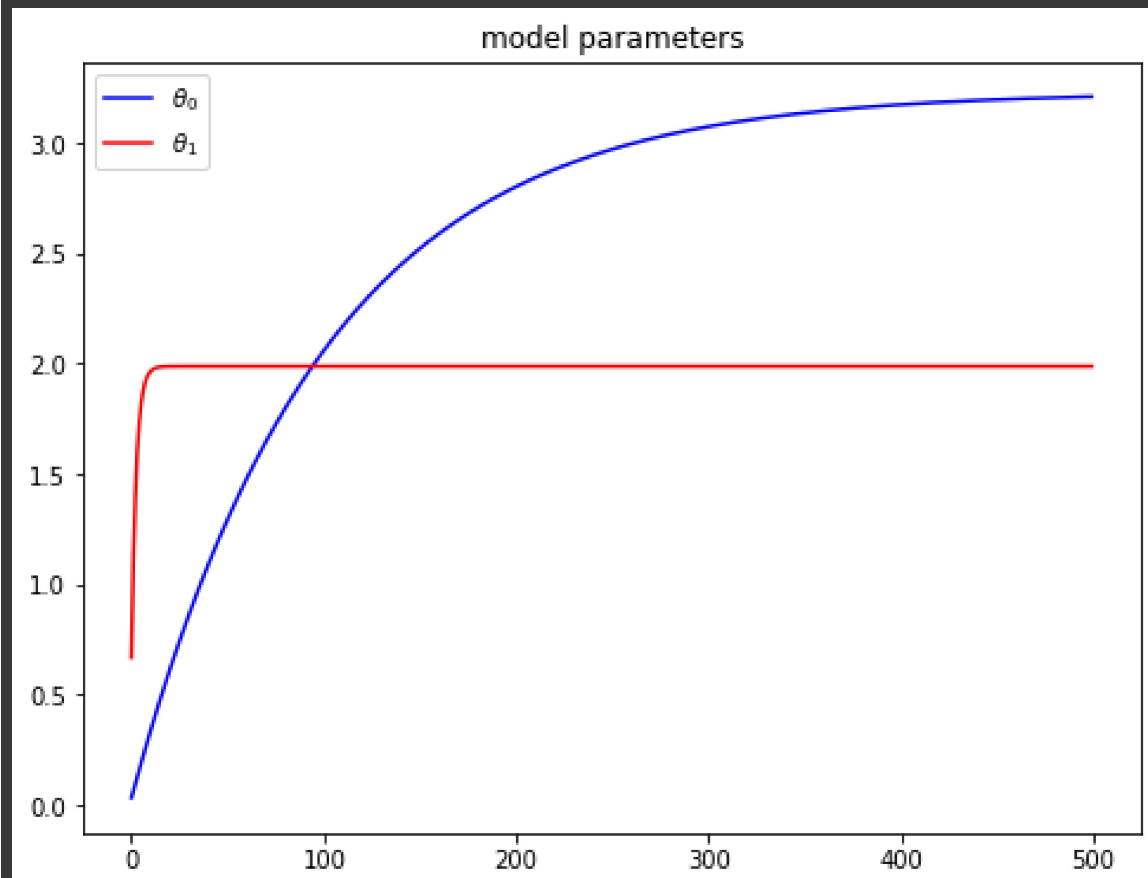
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## [RESULT 02]

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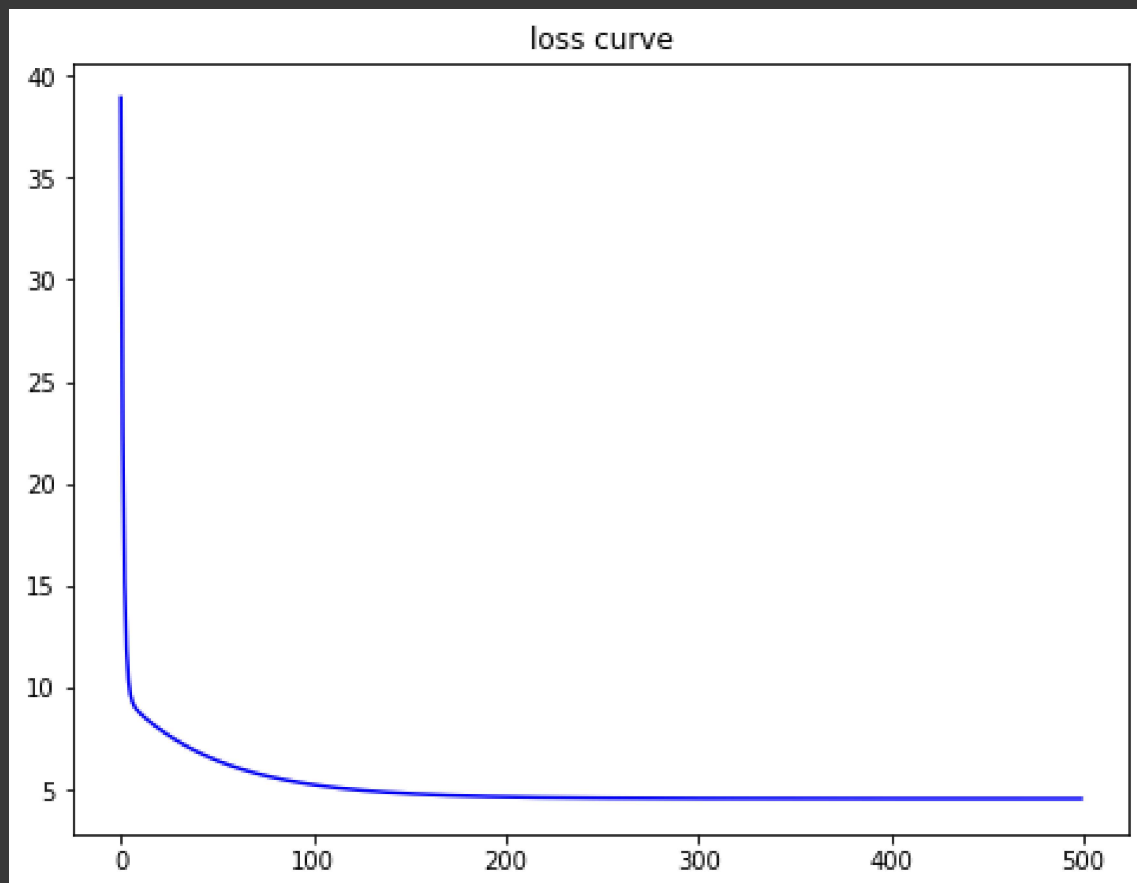


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## [RESULT 03]

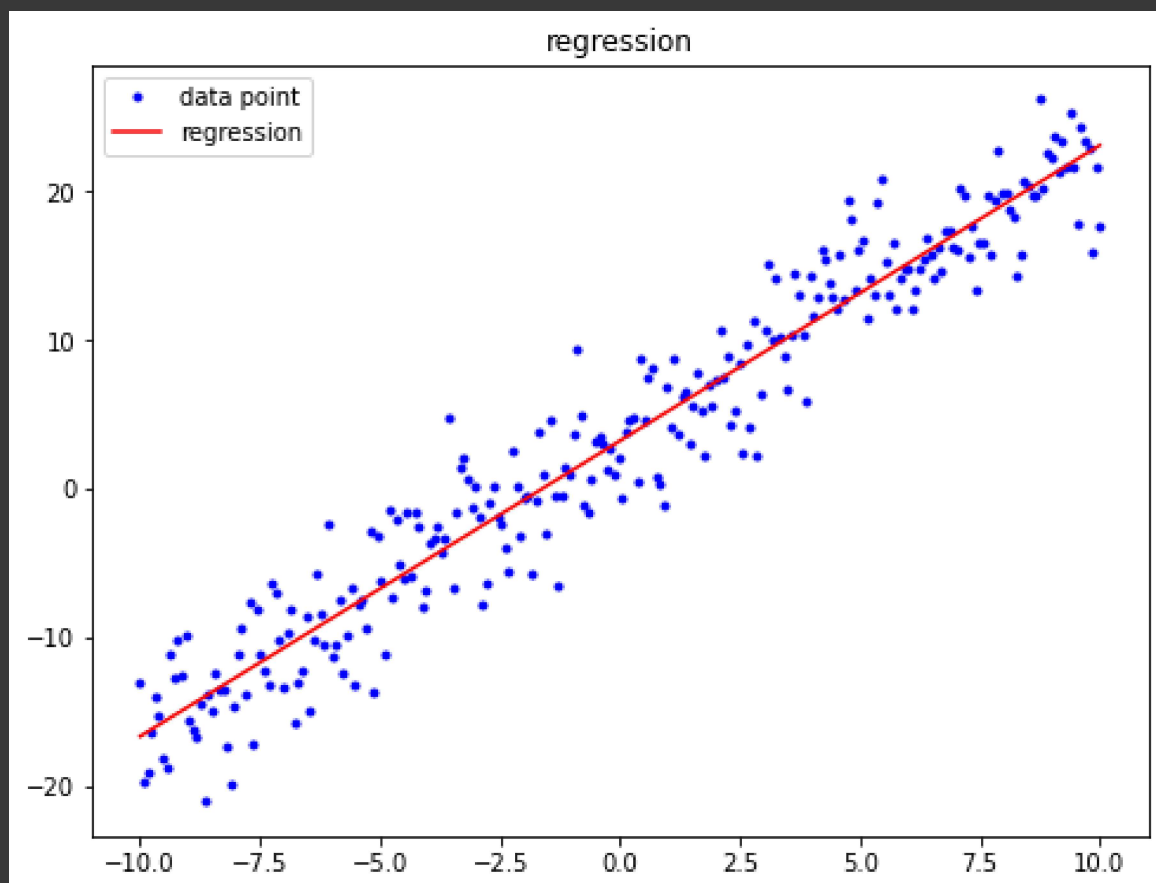
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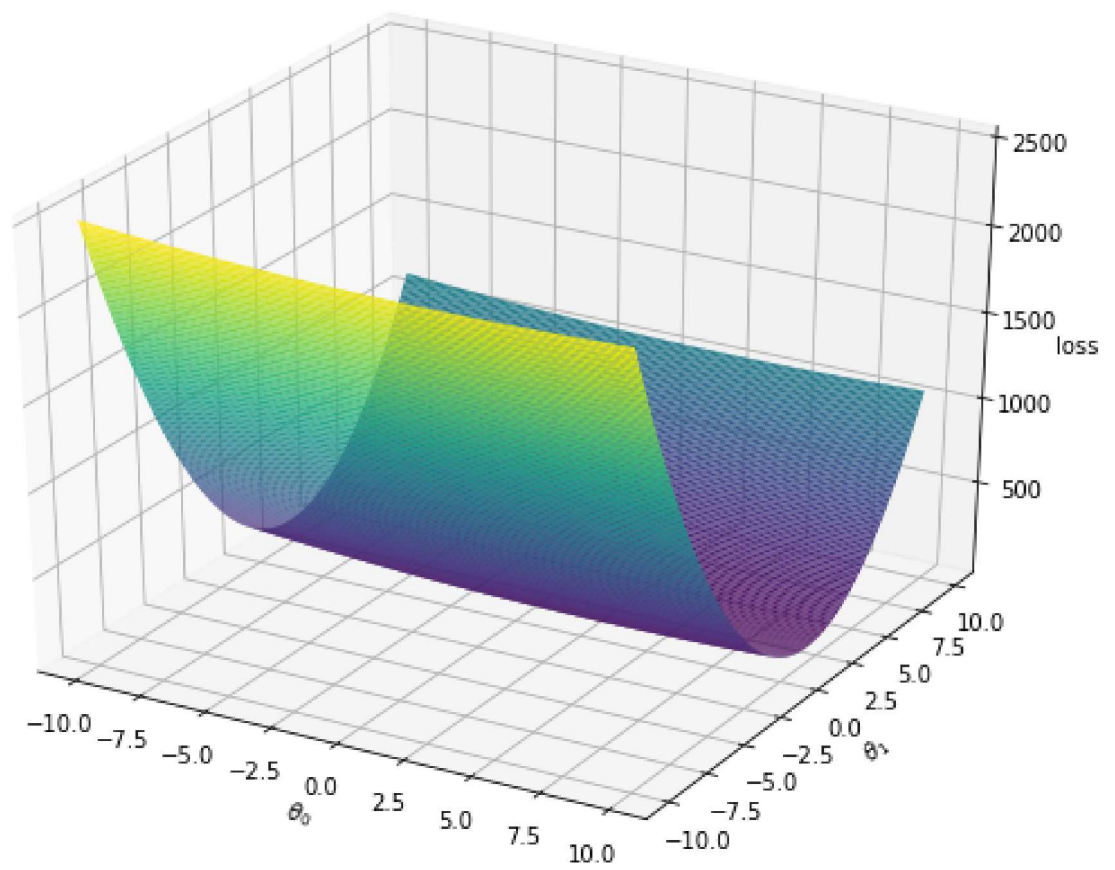
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## [RESULT 04]

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오후 9:52에 완료됨

