

# Linear regression

## import library

```
In [ ]: 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
```

## load data points

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

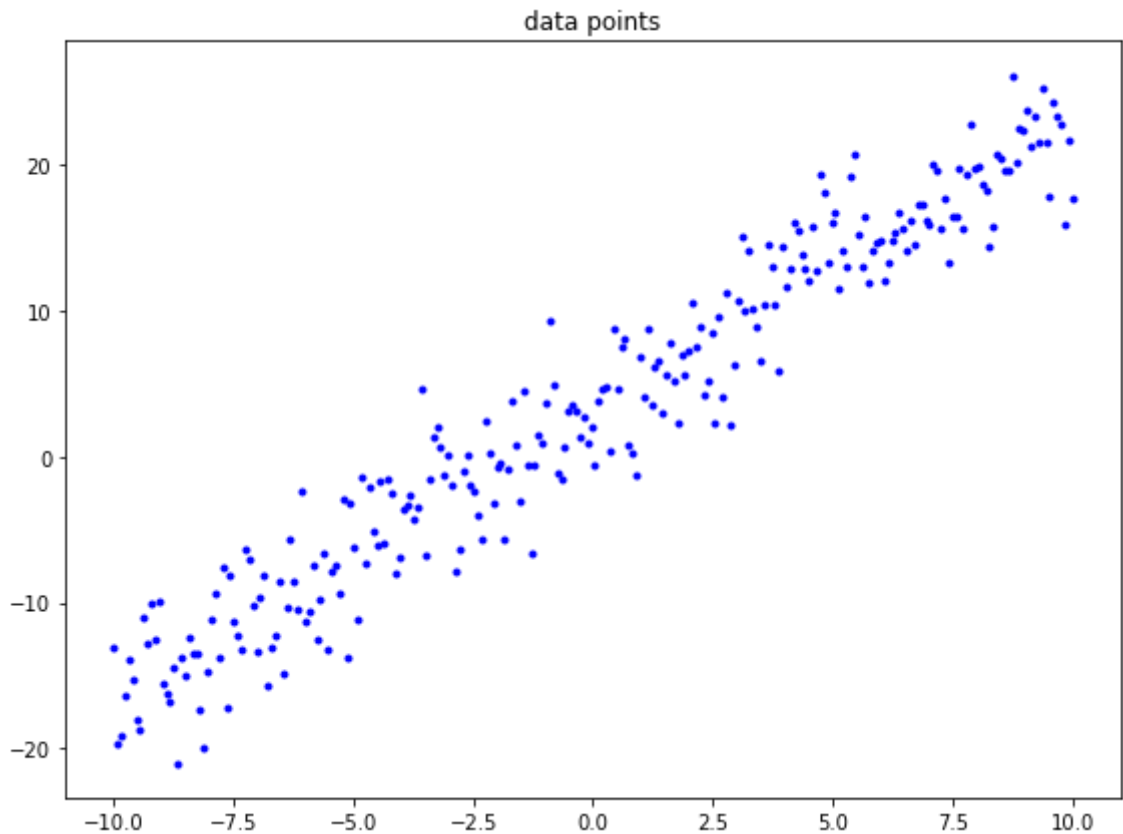
```
In [ ]: filename = 'assignment_06_data.csv'
data = np.loadtxt(filename, delimiter = ',')

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

```
In [ ]: 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`

```
In [ ]: def compute_loss(x, y, theta0, theta1):

    n          = len(x)
    loss        = 0

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

    residual = 0

    residual = compute_residual(x, y, theta0, theta1)
    loss = np.inner(residual, residual) / (n*2)

    #
    # ++++++

    return loss
```

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

- useful functions: `np.inner`

```
In [ ]: def compute_gradient_theta0(x, y, theta0, theta1):

    n          = len(x)
    derivative  = 0

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

    for i in range(n):
        derivative += np.inner(((theta0 + theta1*x[i]) - y[i]).T, 1)

    derivative = derivative/n

    #
```

```
# ++++++

return derivative
```

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

- useful functions: `np.inner`

```
In [ ]: def compute_gradient_theta1(x, y, theta0, theta1):

    n          = len(x)
    derivative  = 0

    # ++++++
    # complete the blanks
    #
    for i in range(n):
        derivative += np.inner(((theta0 + theta1*x[i]) - y[i]).T, x[i])

    derivative = derivative/n
    #
    # ++++++

    return derivative
```

## gradient descent for the optimization

```
In [ ]: 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
            list_loss[i]   = loss
```

## functions for presenting the results

```
In [ ]: def function_result_01():

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

```
In [ ]: def function_result_02():

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

```
In [ ]: def function_result_03():

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

```
In [ ]: 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()
```

```
In [ ]: 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

In [ ]:

```

number_result = 5

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

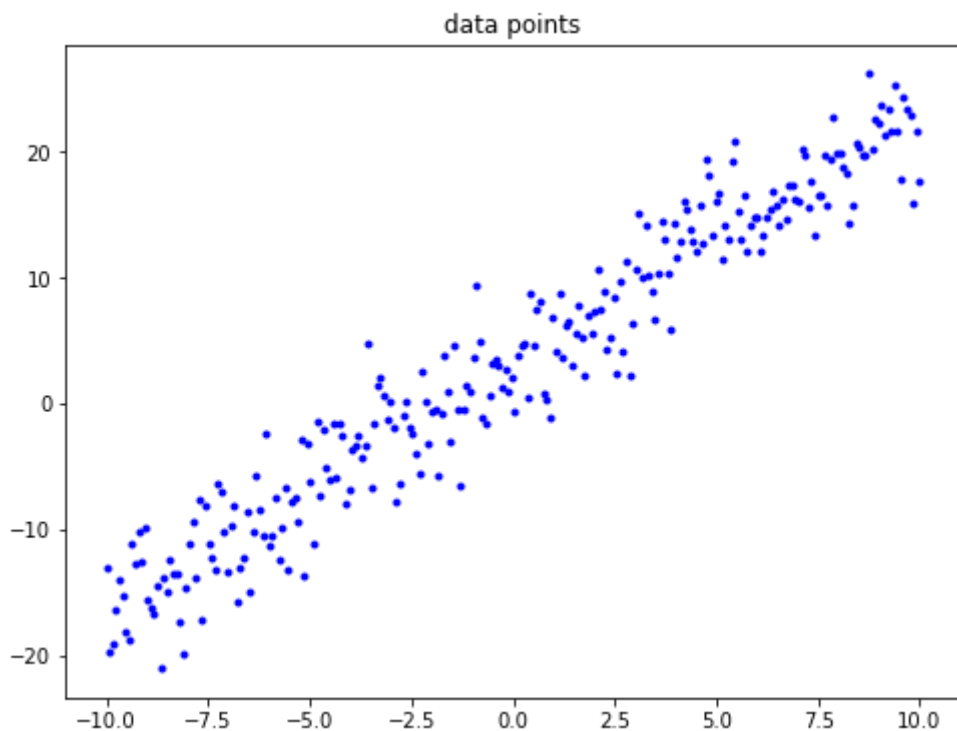
    print('*****')
    print(title)
    print('*****')
    eval(name_function)

```

```

*****
## [RESULT 01]
*****

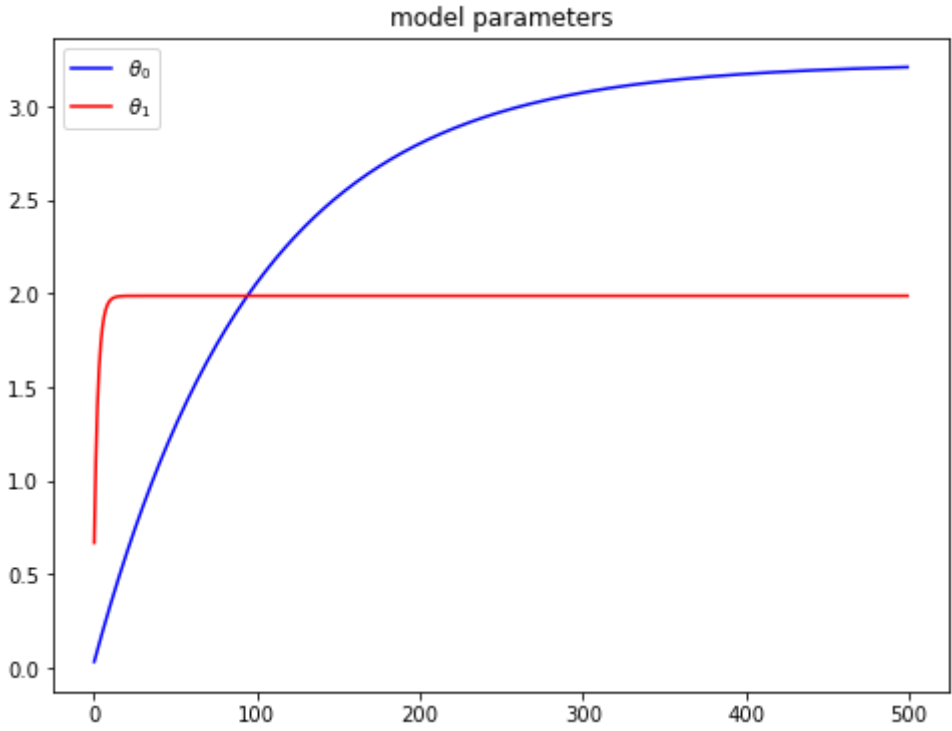
```



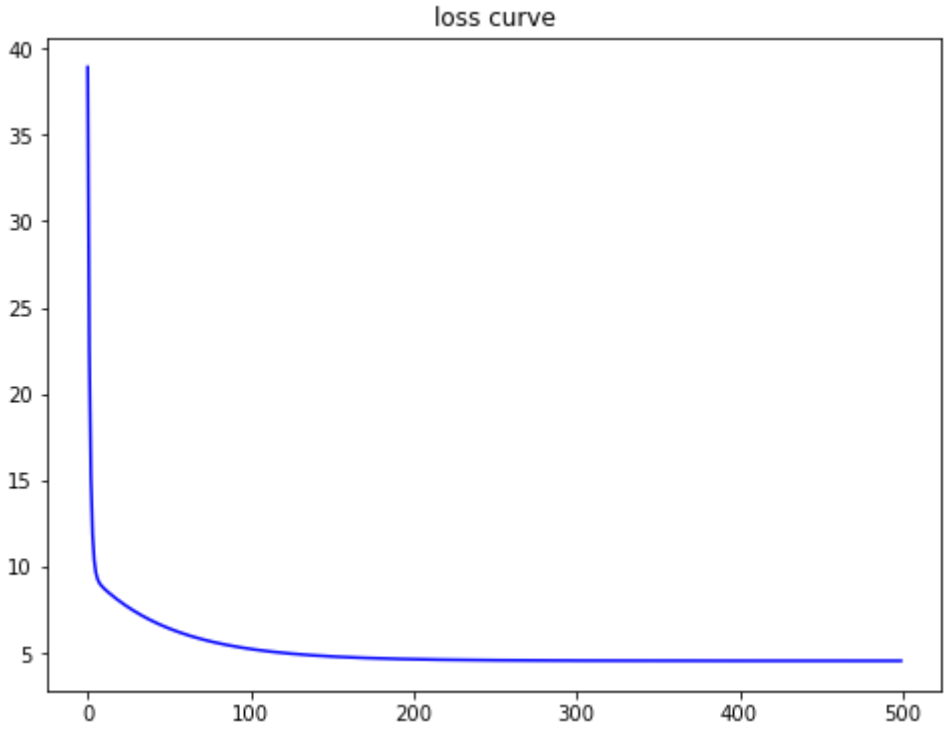
```

*****
## [RESULT 02]
*****

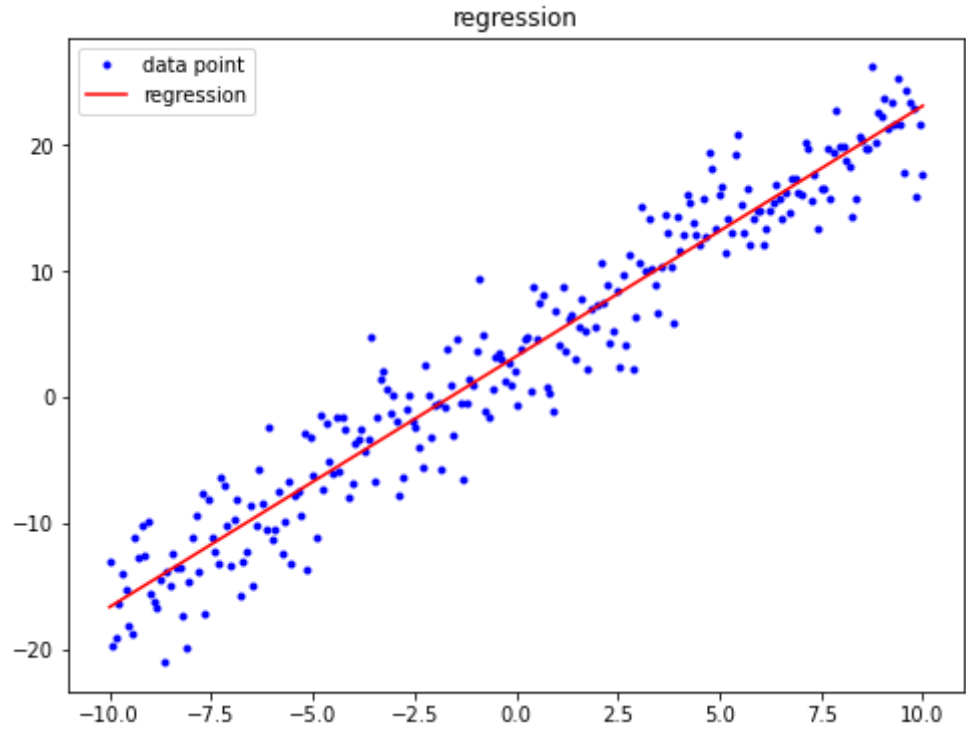
```



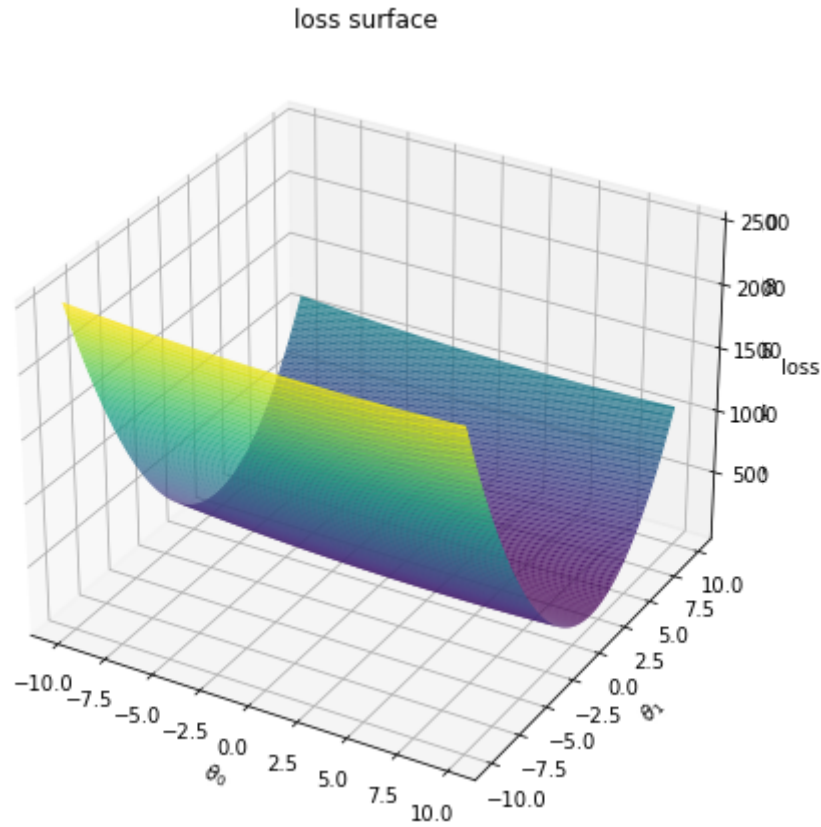
```
*****  
## [RESULT 03]  
*****
```



```
*****  
## [RESULT 04]  
*****
```



```
*****  
## [RESULT 05]  
*****
```



In [ ]: