Least square problem for polynomial regression

import library

In [ ]:

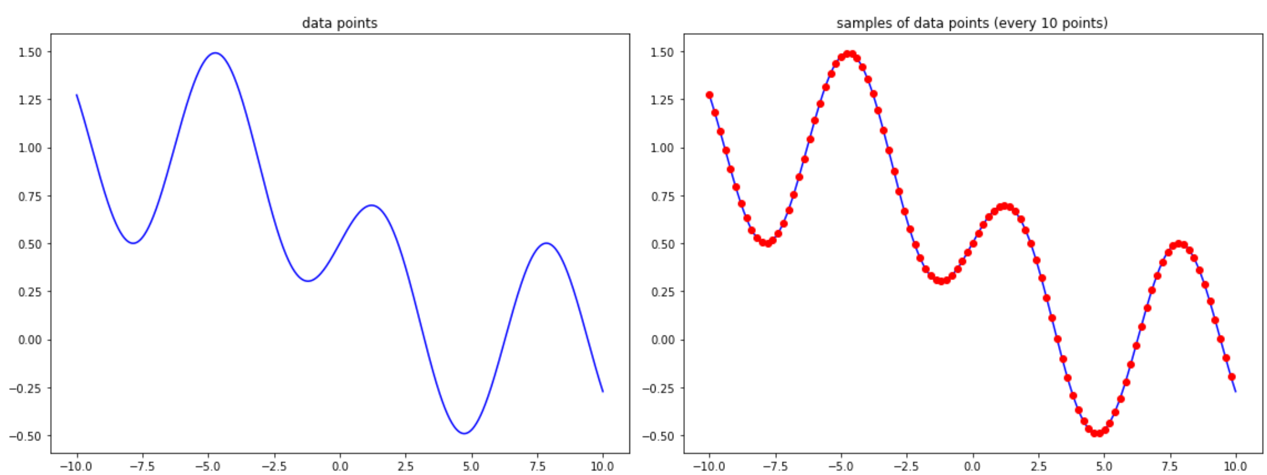
```
import numpy as np
import matplotlib.image as img
import matplotlib.pyplot as plt
import matplotlib.colors as colors
```

load data points

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

In [ ]:

```
filename    = 'assignment_05_data.csv'
data        = np.loadtxt(filename, delimiter = ',')

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

x_sample    = x[::10]
y_sample    = y[::10]

plt.figure(figsize=(16,6))

plt.subplot(121)
plt.plot(x, y, '-', color = 'blue')
plt.title('data points')

plt.subplot(122)
plt.plot(x, y, '-', color = 'blue')
plt.plot(x_sample, y_sample, 'o', color = 'red')
plt.title('samples of data points (every 10 points)')

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



construct matrix A for the polynomial regression with power $p-1$

- useful functions : `np.power`

solve a linear system of equation $Az=b$

$$A = \begin{bmatrix} x_1^0 & x_1^1 & \cdots & x_1^{p-1} \\ x_2^0 & x_2^1 & \cdots & x_2^{p-1} \\ \vdots & \vdots & \vdots & \vdots \\ x_n^0 & x_n^1 & \cdots & x_n^{p-1} \end{bmatrix}, \quad z = \begin{bmatrix} \theta_0 \\ \theta_1 \\ \vdots \\ \theta_{p-1} \end{bmatrix}, \quad b = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

In [ ]:
```
def construct_matrix_A(x, p):

    n = len(x)
    A = np.zeros([n, p])
    initial=x[...]
    # ++++++++++++++++++++++++++++++++++++++++++++++++++
    # complete the blanks
    #

    res = x
    t = []
    for i in range(0, p):
        t.append(np.power(x,i))

    A=np.array(t).T
    # +++++++++++++++++++++++++++++++++++++++++++++++++
    return A
```

construct vector b

In [ ]:
```
def construct_vector_b(y):

    n = len(y)
    b = np.zeros([n, 1])

    # ++++++++++++++++++++++++++++++++++++++++++++++++++
    # complete the blanks
    y=y.T
    b=y
    #
    # ++++++++++++++++++++++++++++++++++++++++++++++++++

    return b
```

solve the linear system of equation $Az=b$

- without regularization : $\min \frac{1}{2n} \|Az - b\|^2, \quad z = \left(A^T A\right)^{-1} A^T b$
- useful functions : `np.matmul, np.linalg.inv, np.sum`

In [ ]:
```
def solve_regression(x, y, p):

    z       = np.zeros([p, 1])
    loss    = 0

    # ++++++++++++++++++++++++++++++++++++++++++++++++++
    # complete the blanks
    #
    A=construct_matrix_A(x, p)
    A_pinv=np.linalg.pinv(A)

    b=construct_vector_b(y)
    z=np.matmul(A_pinv,b)

    n = len(x)
    loss=(1/(2*n))*(np.sum(np.square(np.subtract(np.matmul(A,z),b))))





    #
    # ++++++++++++++++++++++++++++++++++++++++++++++++++

    return z, loss
```

- with regularization : $\min \frac{1}{2n} \|Az - b\|^2 + \frac{\alpha}{2} \|z\|^2, \quad z = \left(A^T A + n\alpha I\right)^{-1} A^T b$ where I denotes identity matrix
- useful functions : `np.matmul, np.linalg.inv, np.sum`

In [ ]:
```
def solve_regression_with_regularization(x, y, p, alpha):

    z       = np.zeros([p, 1])
    loss    = 0
    n = len(x)
    # ++++++++++++++++++++++++++++++++++++++++++++++++++
    # complete the blanks
    #
    A=construct_matrix_A(x, p)
    b=construct_vector_b(y)

    A_inv=np.linalg.pinv(A)
    I=np.matmul(A,A_inv)

    z1=np.linalg.pinv(np.add(np.matmul(A,A.T),(n*alpha*I)))
    z2=np.matmul(A.T,z1)

    z=np.matmul(z2,b)
    loss=np.add((1/(2*n))*np.sum((np.square(np.subtract(np.matmul(A,z),b)))),(al


    #
```

```
    # ++++++++++++++++++++++++++++++++++++++++++++++++++

    return z, loss
```

approximate by polynomial regression

- $\hat{y} = Az^*$
- useful functions : `np.matmul`

In [ ]:
```
def approximate(x, y, p):

    n       = len(y)
    y_hat   = np.zeros([n, 1])
    loss    = 0

    # ++++++++++++++++++++++++++++++++++++++++++++++++++
    # complete the blanks
    #
    A=construct_matrix_A(x, p)
    z,loss=solve_regression(x,y,p)
    y_hat=np.matmul(A,z)
    #
    # ++++++++++++++++++++++++++++++++++++++++++++++++++

    return y_hat, loss
```

In [ ]:
```
def approximate_with_regularization(x, y, p, alpha):

    n       = len(y)
    y_hat   = np.zeros([n, 1])
    loss    = 0

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

    z,loss=solve_regression_with_regularization(x, y, p, alpha)
    A=construct_matrix_A(x, p)
    y_hat=np.matmul(A,z)


    #
    # ++++++++++++++++++++++++++++++++++++++++++++++++++

    return y_hat, 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():

    p             = 2
    (y_hat, _)  = approximate(x, y, p)

    plt.figure(figsize=(8,6))
    plt.plot(x, y, '-', color='blue')
    plt.plot(x, y_hat, '-', color='red')
    plt.show()
```

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

    p             = 4
    (y_hat, _)  = approximate(x, y, p)

    plt.figure(figsize=(8,6))
    plt.plot(x, y, '-', color='blue')
    plt.plot(x, y_hat, '-', color='red')
    plt.show()
```

In [ ]:
```
def function_result_04():

    p             = 8
    (y_hat, _)  = approximate(x, y, p)

    plt.figure(figsize=(8,6))
    plt.plot(x, y, '-', color='blue')
    plt.plot(x, y_hat, '-', color='red')
    plt.show()
```

In [ ]:
```
def function_result_05():

    p             = 16
    (y_hat, _)  = approximate(x, y, p)

    plt.figure(figsize=(8,6))
    plt.plot(x, y, '-', color='blue')
    plt.plot(x, y_hat, '-', color='red')
    plt.show()
```

In [ ]:
```
def function_result_06():

    p             = 32
    (y_hat, _)  = approximate(x, y, p)

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

```
    plt.plot(x, y, '-', color='blue')
    plt.plot(x, y_hat, '-', color='red')
    plt.show()
```

In [ ]:

```
def function_result_07():

    p             = 2
    alpha         = 0.1
    (y_hat, _)  = approximate_with_regularization(x, y, p, alpha)

    plt.figure(figsize=(8,6))
    plt.plot(x, y, '-', color='blue')
    plt.plot(x, y_hat, '-', color='red')
    plt.show()
```

In [ ]:

```
def function_result_08():

    p             = 4
    alpha         = 0.1
    (y_hat, _)  = approximate_with_regularization(x, y, p, alpha)

    plt.figure(figsize=(8,6))
    plt.plot(x, y, '-', color='blue')
    plt.plot(x, y_hat, '-', color='red')
    plt.show()
```

In [ ]:

```
def function_result_09():

    p             = 8
    alpha         = 0.1
    (y_hat, _)  = approximate_with_regularization(x, y, p, alpha)

    plt.figure(figsize=(8,6))
    plt.plot(x, y, '-', color='blue')
    plt.plot(x, y_hat, '-', color='red')
    plt.show()
```

In [ ]:

```
def function_result_10():

    p             = 16
    alpha         = 0.1
    (y_hat, _)  = approximate_with_regularization(x, y, p, alpha)

    plt.figure(figsize=(8,6))
    plt.plot(x, y, '-', color='blue')
    plt.plot(x, y_hat, '-', color='red')
    plt.show()
```

In [ ]:

```
def function_result_11():

    p             = 32
    alpha         = 0.1
    (y_hat, _)  = approximate_with_regularization(x, y, p, alpha)
```

```
    plt.figure(figsize=(8,6))
    plt.plot(x, y, '-', color='blue')
    plt.plot(x, y_hat, '-', color='red')
    plt.show()
```

In [ ]:
```
def function_result_12():

    p               = 4
    (_, loss)       = approximate(x, y, p)

    print('loss = ', loss)
```

In [ ]:
```
def function_result_13():

    p               = 16
    (_, loss)       = approximate(x, y, p)

    print('loss = ', loss)
```

In [ ]:
```
def function_result_14():

    p               = 4
    alpha           = 0.1
    (_, loss)       = approximate_with_regularization(x, y, p, alpha)

    print('loss = ', loss)
```

In [ ]:
```
def function_result_15():

    p               = 16
    alpha           = 0.1
    (_, loss)       = approximate_with_regularization(x, y, p, alpha)

    print('loss = ', loss)
```

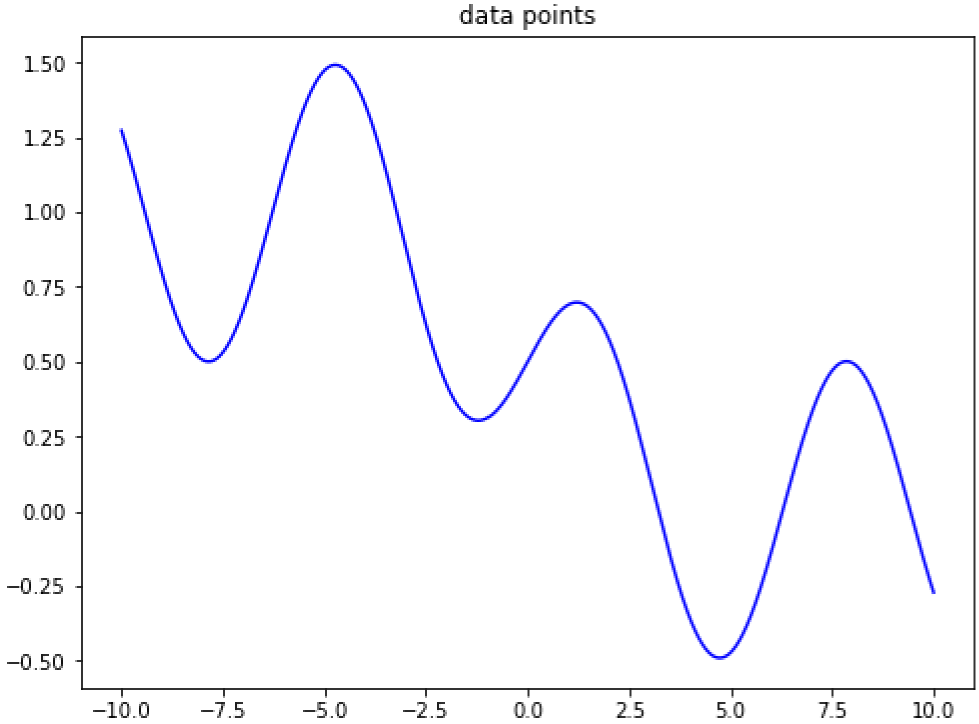
---

results

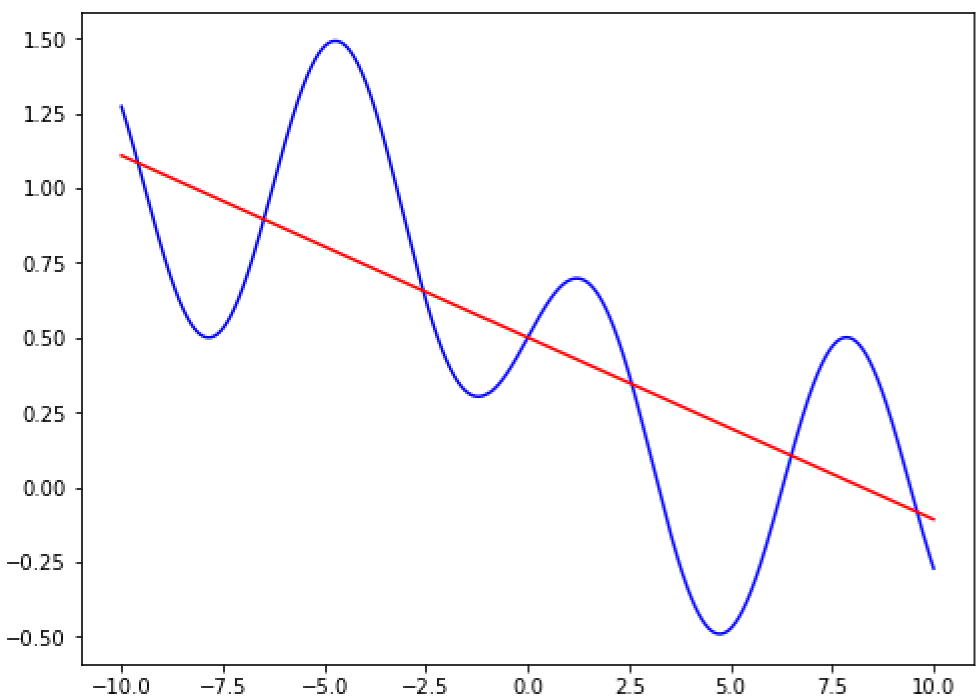
In [ ]:
```
number_result = 15

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)
```

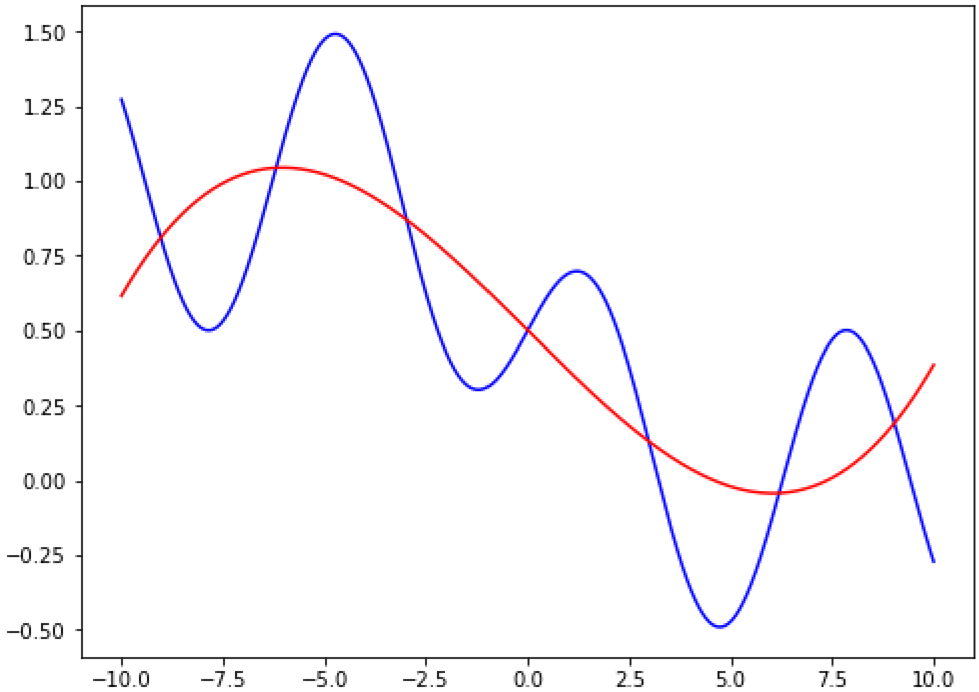
```
**************************************************
## [RESULT 01]
**************************************************
```



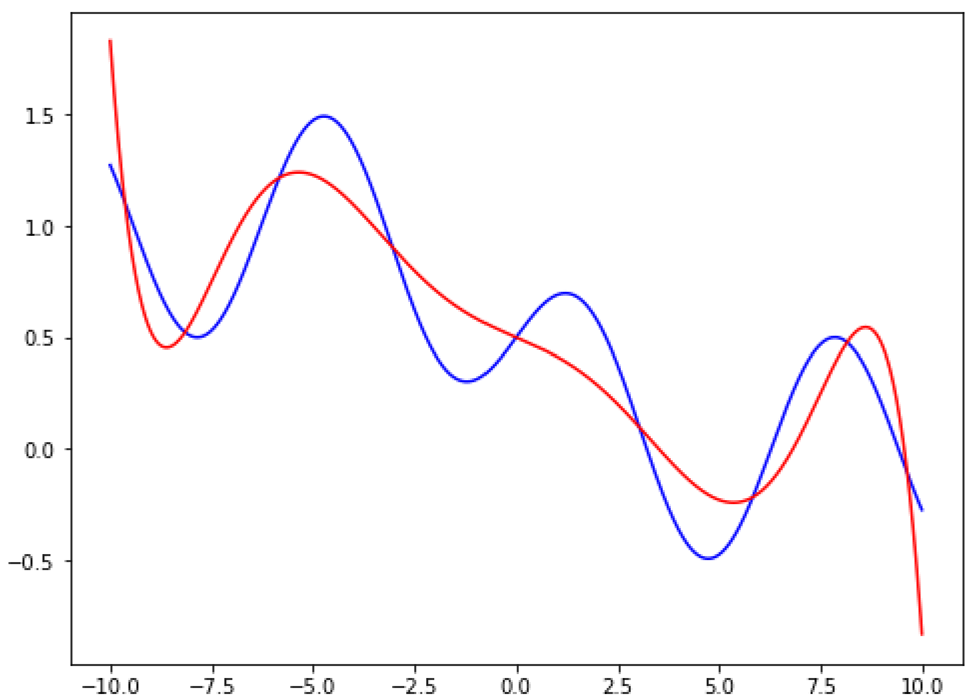

```
**************************************************
## [RESULT 02]
**************************************************
```



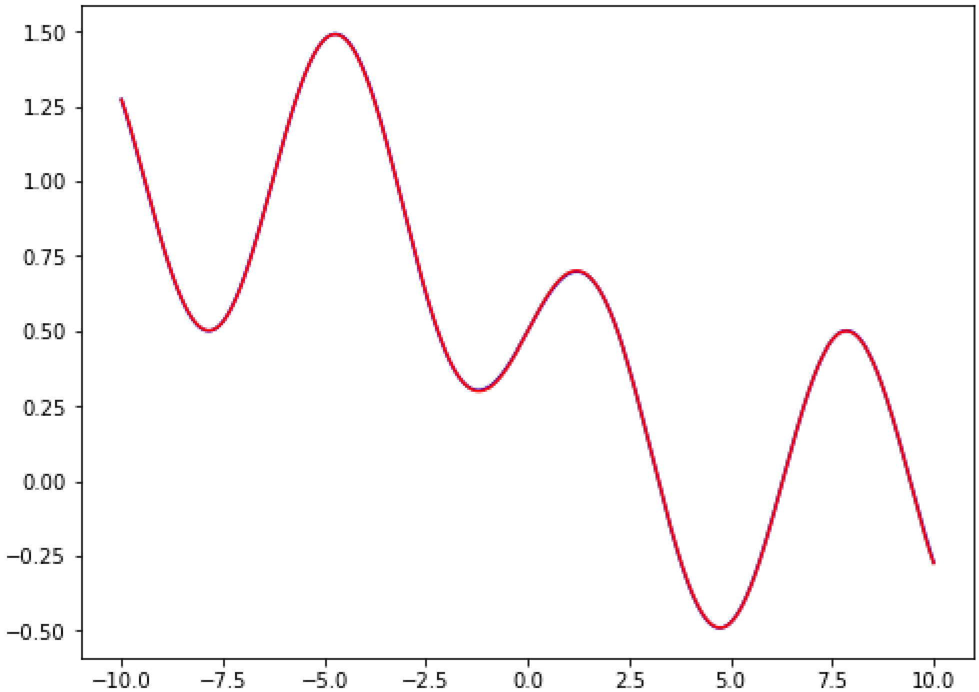

```
**************************************************
## [RESULT 03]
**************************************************
```



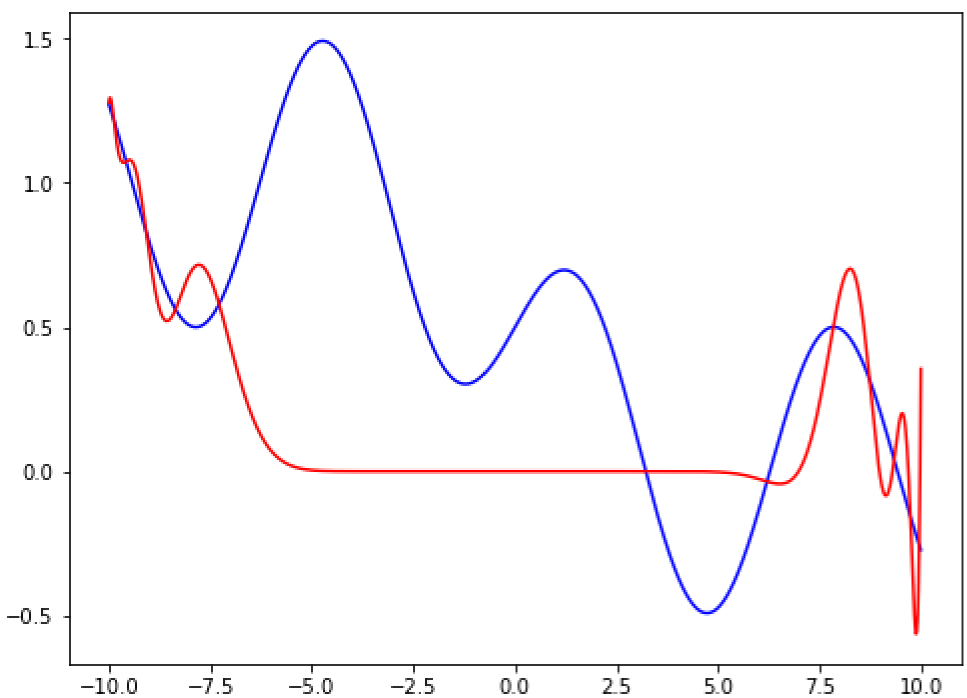

```
************************************************
## [RESULT 04]
************************************************
```



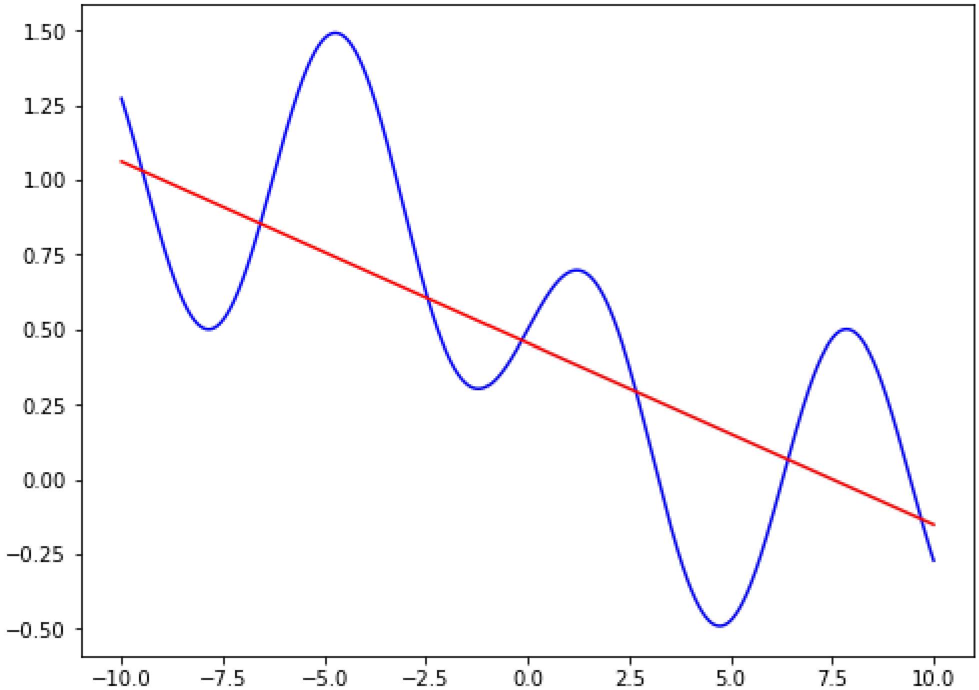

```
************************************************
## [RESULT 05]
************************************************
```



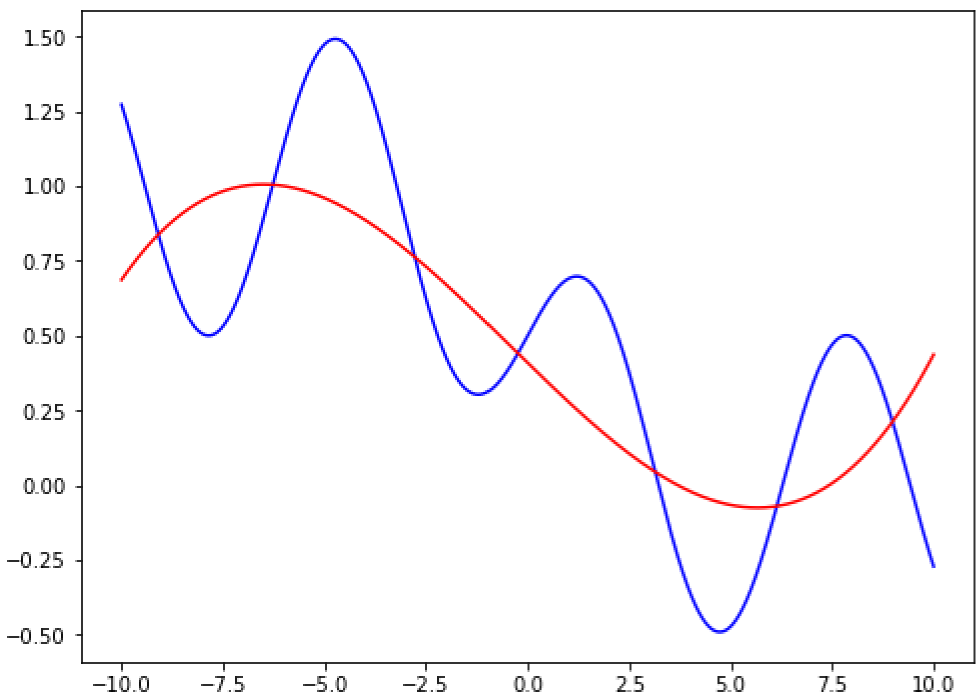

```
*************************************************
## [RESULT 06]
*************************************************
```



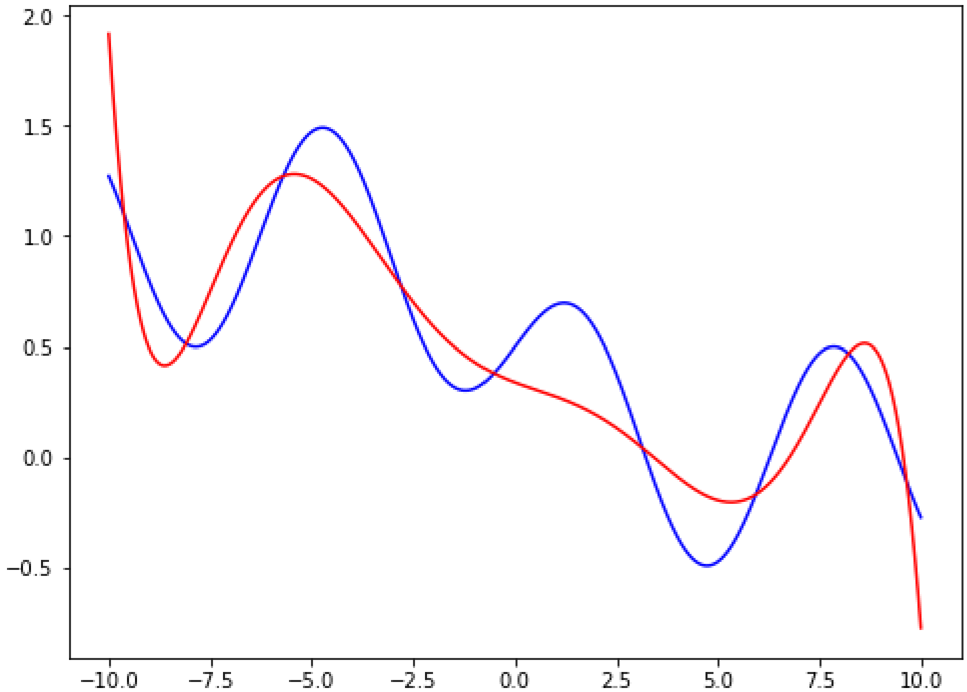

```
*************************************************
## [RESULT 07]
*************************************************
```



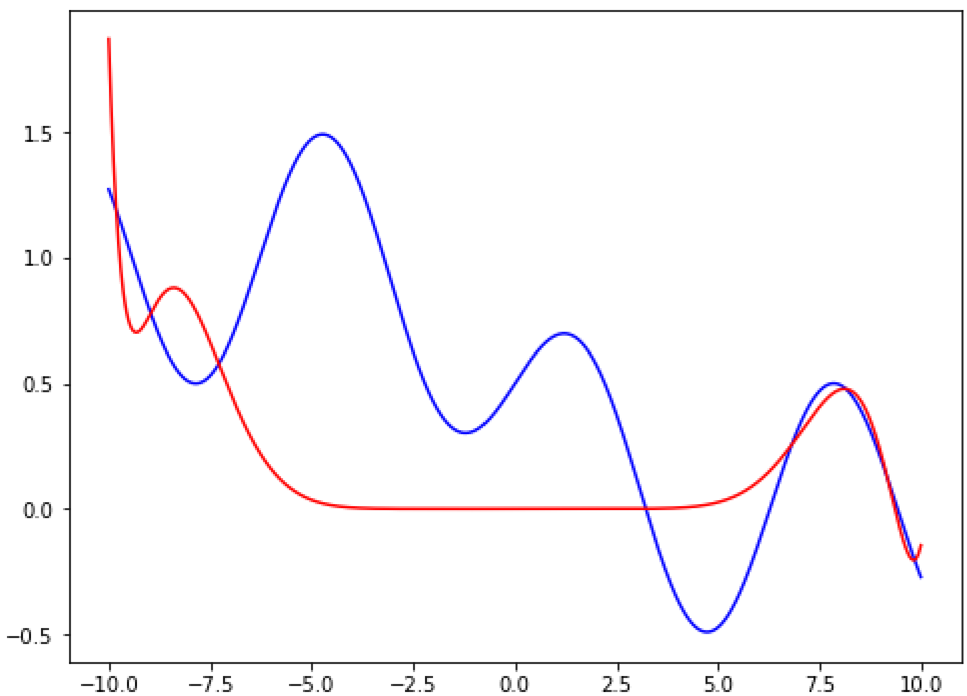

```
************************************************
## [RESULT 08]
************************************************
```



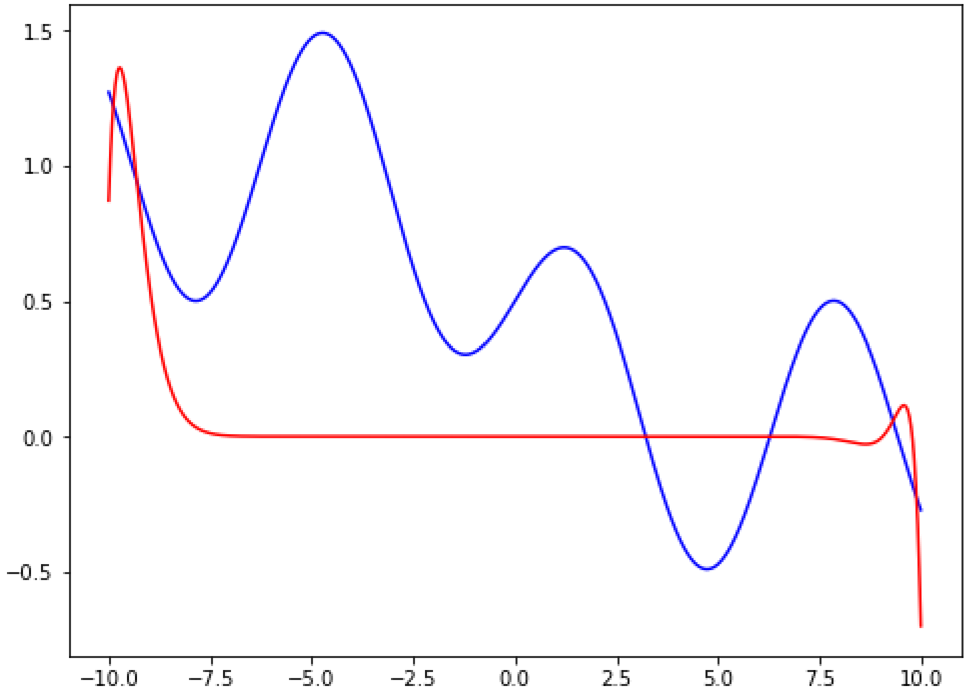

```
************************************************
## [RESULT 09]
************************************************
```




```
************************************************
## [RESULT 10]
************************************************
```




```
************************************************
## [RESULT 11]
************************************************
```




```
************************************************
## [RESULT 12]
************************************************
loss =  0.052697806863259415
************************************************
## [RESULT 13]
************************************************
loss =  1.7955542961308408e-06
************************************************
## [RESULT 14]
************************************************
loss =  0.06379867384617091
************************************************
## [RESULT 15]
************************************************
loss =  0.19084523958733982
```

In [ ]: