The code and the results (final and intermediate) for question 2, part b)

```
# the code used for stochastic gredient descent
    import numpy as np
    b = np.array([[1], [1]], dtype='float')
    x = np.array([[1.5,2], [3,2.5], [4.5,3]])
    y = np.array([[10],[15.5],[21]])
    moo = 0.1
    epochs = 2
    def grad(x, y,b):
        return 2*(x.T@x@b-x.T@y)
    def mse(x,y,b):
        return (x@b - y)**2
    for epoch in range(epochs):
        for i in range(len(x)):
           x_i = x[i, :].reshape((1,2))
           y_i = y[i, :].reshape((1,1))
           gradient = -grad(x_i, y_i,b)
           err = mse(x_i, y_i,b)
           print(f'b: {b}')
           print(f'mse: {err}')
           b = b - moo * gradient
    print(f'final b is : {b}')
27
```

```
epoch 0 - step 0
b: [[1.]
[1.]]
mse: [[42.25]]
epoch 0 - step 1
_____
b: [[-0.95]
[-1.6]]
mse: [[499.5225]]
epoch 0 - step 2
_____
b: [[-14.36]
[-12.775]]
mse: [[15362.363025]]
epoch 1 - step 0
_____
b: [[-125.9105]
[ -87.142 ]]
mse: [[139240.73592506]]
epoch 1 - step 1
    -----
b: [[-237.855425]
[-236.4019 ]]
mse: [[1742587.51104455]]
epoch 1 - step 2
_____
b: [[-1029.89804 ]
[ -896.4374125]]
mse: [[53946871.72456145]]
final b is : [[-7640.26611575]
[-5303.349463
```

The code and the results (final and intermediate) for question 2, part c)

```
# the code used for stochastic gredient descent with adagrad
    import numpy as np
    b = np.array([[1], [1]], dtype='float')
    x = np.array([[1.5,2], [3,2.5], [4.5,3]])
    y = np.array([[10],[15.5],[21]])
    moo = 0.1
    epochs = 2
    epsilon = 1e-8
    def grad(x, y,b):
        return 2*(x.T@x@b-x.T@y)
    def mse(x,y,b):
        return (x@b - y)**2
     gradian_history = np.array([[0],[0]], dtype='float')
    for epoch in range(epochs):
        for i in range(len(x)):
            x_i = x[i, :].reshape((1,2))
            y_i = y[i, :].reshape((1,1))
            gradient = -grad(x_i, y_i,b)
            err = mse(x_i, y_i,b)
            print(f'b: {b}')
            print(f'mse: {err}')
            gradian_history += gradient**2
            step_size = moo / (np.sqrt(gradian_history) + epsilon)
            b = b - step_size * gradient
26
    print(f'final b is : {b}')
```

```
epoch 0 - step 0
_____
b: [[1.]
[1.]]
mse: [[42.25]]
epoch 0 - step 1
_____
b: [[0.9]
[0.9]]
mse: [[111.30249999]]
epoch 0 - step 2
_____
b: [[0.8044319 ]
[0.81030368]]
mse: [[223.47694926]]
epoch 1 - step 0
_____
b: [[0.71471464]
[0.72667635]]
mse: [[55.86927667]]
epoch 1 - step 1
_____
b: [[0.69992617]
[0.69982432]]
mse: [[135.7378943]]
epoch 1 - step 2
_____
b: [[0.65805937]
[0.65346741]]
mse: [[258.51271487]]
final b is : [[0.59256638]
[0.59257312]]
```