## 线性回归算法

```
In [1]: import numpy as np
         import matplotlib.pyplot as plt
In [2]: x = np.array([10, 20, 30, 40, 50])
         y = np.array([150, 220, 225, 300, 330])
In [3]: plt.scatter(x, y, color='blue')
         plt.xlim(5, 60)
         plt.ylim(80, 400)
         plt.show()
         400
          350
          300
         250
         200
         150
         100
                10
                        20
                                30
                                       40
                                               50
                                                       60
```

## 最小二乘法

```
In [4]: import matplotlib.image as mpimg
```

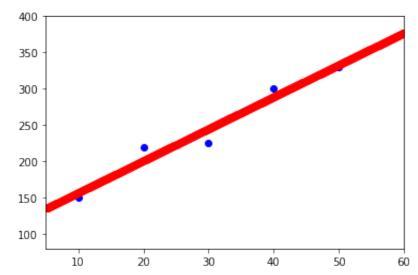
```
In [5]: | img = mpimg.imread('formula1.png')
           img.shape
           plt.imshow(img)
           plt.axis('off')
           plt.show()
                 Q = \sum_{i=1}^{n} (y^{(i)} - ax^{(i)} - b)^{2}
                当a, b取什么值时Q最小?
                a = \frac{\sum_{i=1}^{m} (x^{(i)} - \bar{x})(y^{(i)} - \bar{y})}{\sum_{i=1}^{m} (x^{(i)} - \bar{x})^2} \qquad b = \bar{y} - a\bar{x}
 In [6]: x_{mean} = np.mean(x)
           y_mean = np.mean(y)
 In [7]: u = 0 # 分子
           v = 0 # 分母
 In [8]: for x_i, y_i in zip(x, y):
                u += (x_i - x_mean) * (y_i - y_mean)
                v += (x_i - x_mean) ** 2
 In [9]: | a = u / v
           b = y_mean - a * x_mean
In [10]: a
Out[10]: 4.4
In [11]: b
Out[11]: 113.0
In [12]: | x_predict = 55
           y_predict = a * x_predict + b
In [13]: y_predict
```

Out[13]: 355.0

In [14]: x2 = np.linspace(5, 60, num = 1000)

```
In [15]: plt.scatter(x, y, color='blue')
    plt.xlim(5, 60)
    plt.ylim(80, 400)

plt.scatter(x2, a * x2 + b, color='red')
    plt.show()
```



```
In [16]: from ML.linear import LinearRegression
```

In [17]: | lr = LinearRegression()

In [18]: lr.fit(x, y)

Out[18]: <ML.linear.LinearRegression at 0x118a40198>

In [21]: lr.predict(np.array([55]))

Out[21]: array([355.])