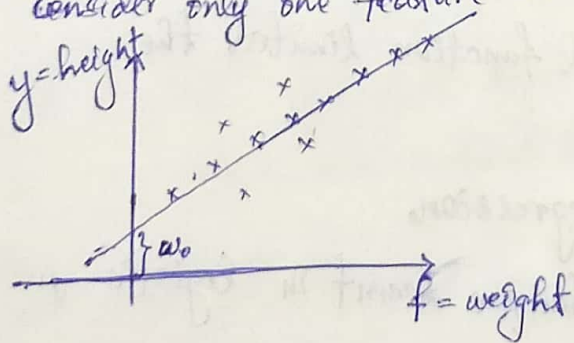


## Linear Regression

↳ does actual regressions.

Ex → predict height  $\in \mathbb{R}$  given weight, gender, ethnicity, hair color, etc.

consider only one feature → weight for simplicity.



linear regression → find a line that <sup>best</sup> fits the given data.

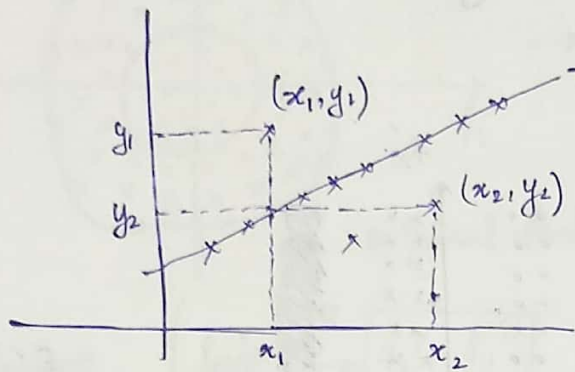
$$\text{height} = w_1 * \text{weight} + w_0$$

$$y = mx + c$$

In general, we have to predict a hyperplane that best fits the given data.

Now, what does best-fit mean?

→ let's say we have got the model  $\pi(w_1, w_0)$ .



let the pred<sup>n</sup> by model be denoted by  $\hat{y}$ .  
 so, for the model to best fit, the error must be minimum i.e. diff b/w pred<sup>n</sup> & actual values

this must be min. for best fit.

$$\sum_{i=1}^n (y_i - \hat{y}_i)^2 \text{ — squared errors}$$

### Mathematical formulation

$$(w^*, w_0^*) = \underset{w, w_0}{\operatorname{argmin}} \sum_{i=1}^n (y_i - \hat{y}_i)^2 \quad \text{where } \hat{y}_i = w^T x_i + w_0$$

with regularization, we have

$$(w^*, w_0^*) = \underset{w, w_0}{\operatorname{argmin}} \sum_{i=1}^n (y_i - \hat{y}_i)^2 + \lambda \|w\|_2^2 \quad \hat{y}_i = w^T x_i + w_0$$

$$(w^*, w_0^*) = \underset{w, w_0}{\operatorname{argmin}} \sum_{i=1}^n [y_i - (w^T x_i + w_0)]^2 + \lambda \|w\|_2^2$$

### Real world cases

Almost same as that of logistic regression.

Outliers → In logistic regression, sigmoid function limited the impact of outliers.

But nothing such here in linear regression.

So, we making use of filtering technique learnt in logistic regression, filter out outliers.