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Starting point :

- Outcome measurement Y (also called dependent variable, response, target) ;
- Vector of p predictor measurements X (also called inputs, regressors, covariates, features, independent variables). X is a matrix of dimension (N,p) , where n is the number of measurements ;
- In the **regression problem**, Y is quantitative (e.g price, blood pressure) ;
- We have training data $(x_1, y_1), \dots, (x_N, y_N)$. These are observations (examples, instances) of these measurements.

Linear Regression model with one variable

$$Y_i = \beta_0 + \beta_1 X_1$$

$$Price_{house30} = K + \beta_1 Surface_{House30}$$

Figure (TBD) In fact, we could imagine the price depends from several factors, so we come with Linear Regression with several variables :

$$Price_{house30} =$$

$$K + \beta_1 Surface_{House30} + \beta_2 NbOfRooms_{House30} + \beta_3 Location_{House30}$$

In general :

$$Y_i = h(X^i) = \beta_0 + \beta_1 X_{i,1} + \beta_2 X_{i,2} + \dots + \beta_p X_{i,p}$$

Traditionally p is called the number of features. We will use matrix notation, so there will be double indices.

- The parameters in the linear regression model are very easy to interpret.
- $\beta_j, 1 \leq j \leq p$ is the average increase in Y when X_j is increased by one and all other X_i are held constant.
- Vocabulary : β_0 is the intercept (i.e. the average value for Y if all the X 's are zero), β_j is the slope for the j th variable X_j

- Historical method : least square regression ;
- Modern method : numerical iterative process : gradient descent and a huge family of similar algorithms (Maths : (Numerical)(Convex or not) Optimization.

Cost function, traditionally noted $J(\beta)$ is given by :

$$J(\beta) = \frac{1}{2n} \sum_{i=1}^n (h(X^i) - Y_i)^2$$

n is N explained before.

Explanation,
Solution in One dim
Solution in p dim



- Level on a curve or surface, direction of steepest descent
- Stochastic approach (SGD)

- The goal of Unsupervised Learning is to discover interesting things about the measurements : is there an informative way to visualize the data ? Can we discover subgroups among the variables or among the observations ?
- We discuss two methods :
 - **principal components analysis (PCA)**, a tool used for data visualization or data pre-processing before supervised techniques are applied, and
 - **clustering**, a broad class of methods for discovering unknown subgroups in data.

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image Yann Le Cun ML cake