

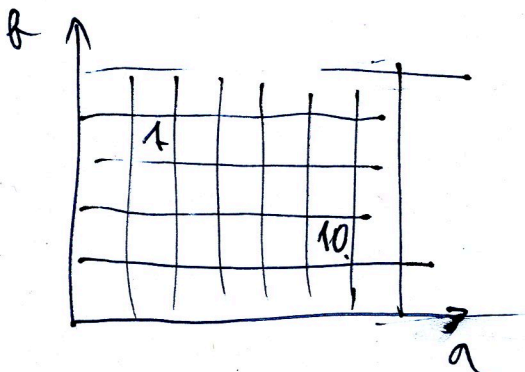
# Finding co-linear points

$$y = ax + \beta$$

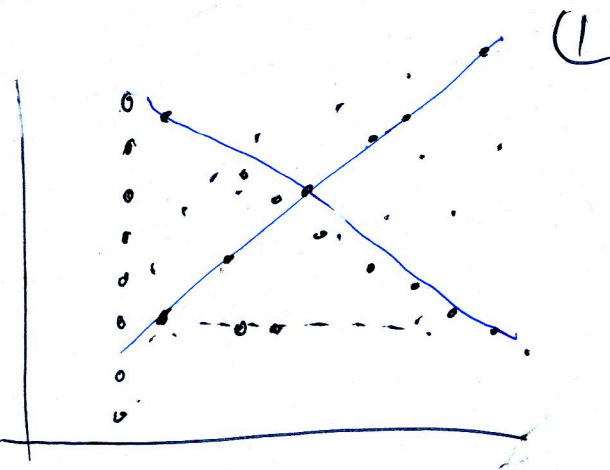
Brute force:

take all  $i, j$  pairs:

$$a_{ij} = \frac{y_i - y_j}{x_i - x_j} \Rightarrow \beta_{ij} = y_i - a_{ij}x_j$$



# of pairs  $\binom{N}{2} = \frac{N(N-1)}{2} = \mathcal{O}(N^2)$

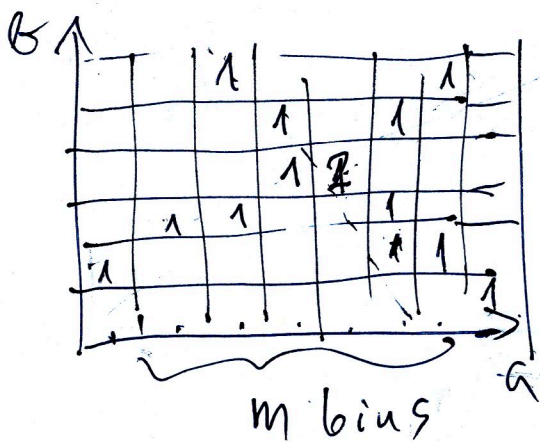


# Hough - TRANSFORM

$$y_0 = ax_0 + \beta \Rightarrow$$

$$\beta = -ax_0 + y_0$$

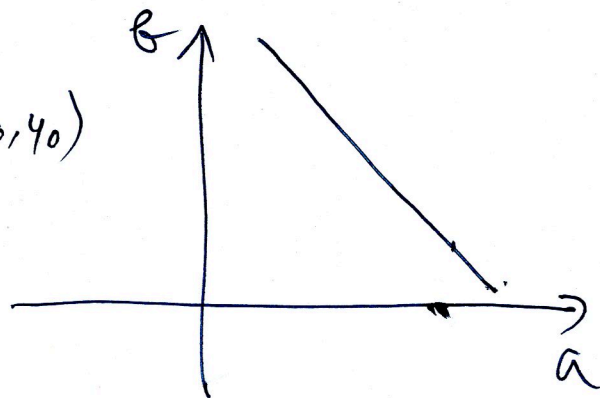
Loop through all points



$$a_k = k \quad (k = 1..m) \Rightarrow \beta_k(x_0, y_0)$$

$$x_0, y_0$$

$$\mathcal{O}(N \cdot m) \leftarrow \text{fixed}$$



$10^8$  or  $10^{12}$   $\uparrow$  cost is LINEAR in  $N$

Better parametrization of the line

$$x \cos \theta + y \sin \theta = s \quad (s, \theta)$$

$$\frac{y}{x-x_0} = \tan(\theta - 90) = \frac{\sin(\theta - 90)}{\cos(\theta - 90)}$$

$$= -\frac{\cos \theta}{\sin \theta} = -\frac{1}{\tan \theta}$$

$$s = -x_0 \sin(\theta - 90) = x_0 \cos \theta$$

$$y = ax + b$$

$$y = (x - x_0) \cdot \left(-\frac{1}{\tan \theta}\right) = -\frac{1}{\tan \theta} \cdot x + x_0 \frac{1}{\tan \theta}$$

$$y = -\frac{\cos \theta}{\sin \theta} \cdot x + \frac{s}{\cos \theta} \frac{\cos \theta}{\sin \theta} \quad \left/ \begin{matrix} x \sin \theta \\ \end{matrix} \right.$$

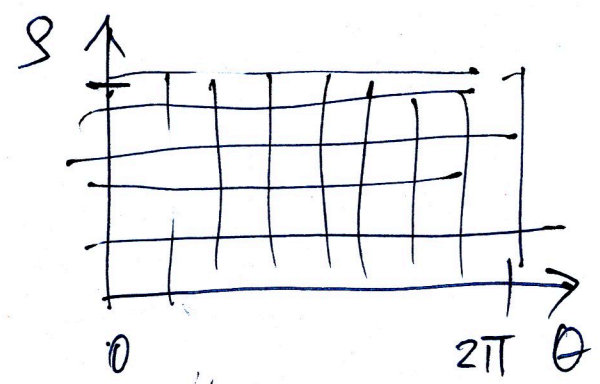
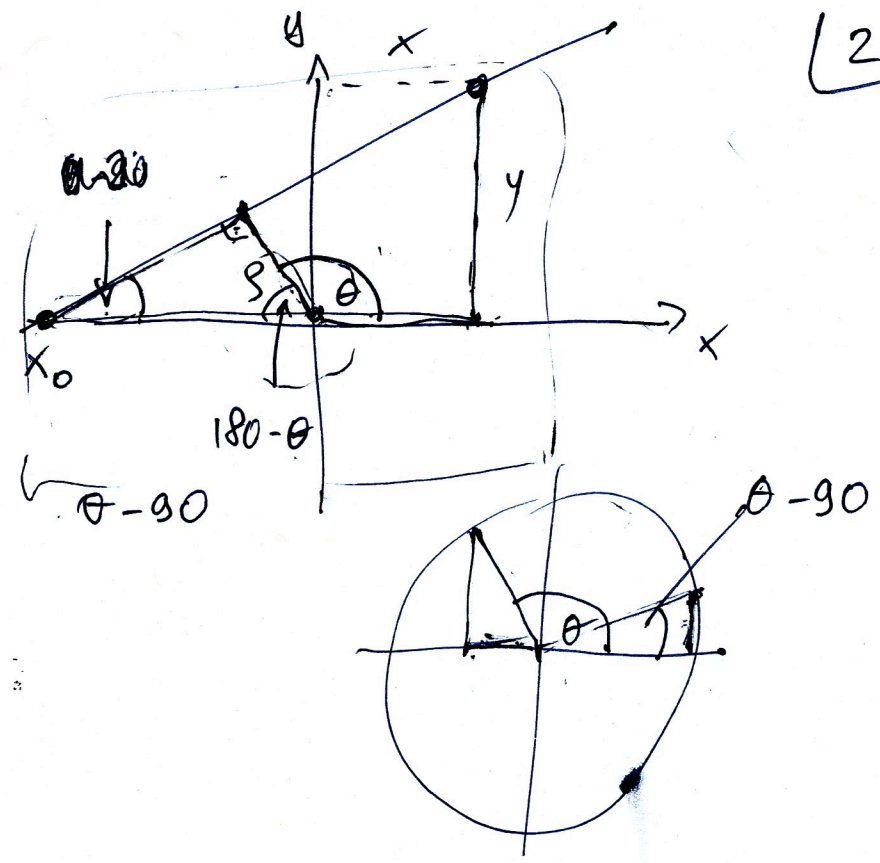
cost

$$y \sin \theta = -x \cos \theta + s$$

$$\boxed{s = x \cos \theta + y \sin \theta}$$

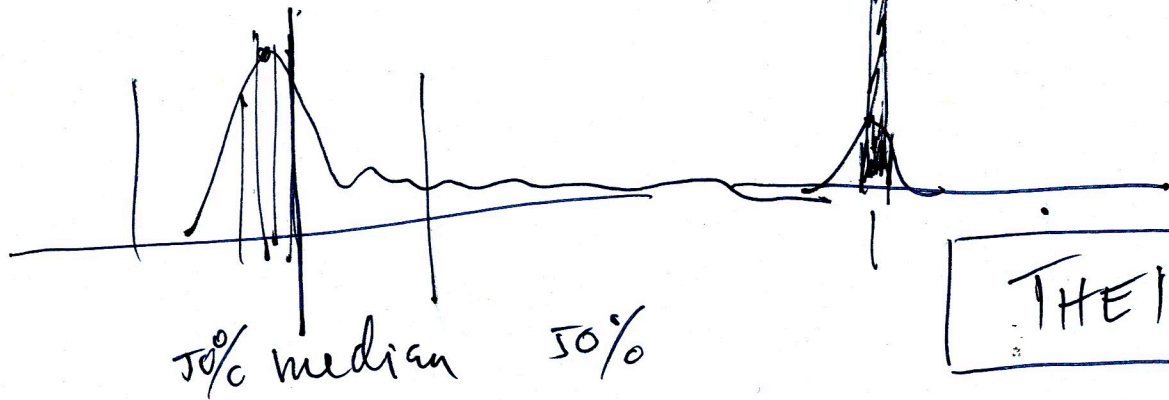
$\theta$  (N.m)

$N \log N$   $N^3$  ~~ew~~



# Least squares (LS) regression

$$\chi^2 = \sum [y_i - ax_i - b]^2 \frac{1}{\sigma_i^2}$$



THEIL-SUN

$$a_{ij} = \left( \frac{y_j - y_i}{x_j - x_i} \right)$$

$$\binom{N}{2} = \frac{N(N-1)}{2} \text{ different slopes}$$

$$a = \underset{i \neq j}{\text{median}}(a_{ij})$$

$$\text{median}(y_i - ax_i) = b$$

~ 30% points are outliers, ~~that~~  
right answers!

Robust fit

scipy.stats

Siegel

For every pair  $i, j$  we can compute

$$a_{ij} = \left( \frac{y_j - y_i}{x_j - x_i} \right)$$

$$a_i = \text{median}_{j \neq i} (a_{ij})$$

$$\beta_i = \text{median}_{j \neq i} \left( \frac{x_j y_i - x_i y_j}{x_j - x_i} \right)$$

$$a = \text{median}_i (a_i)$$

$$\beta = \text{median}_i (\beta_i)$$

Siegel slopes

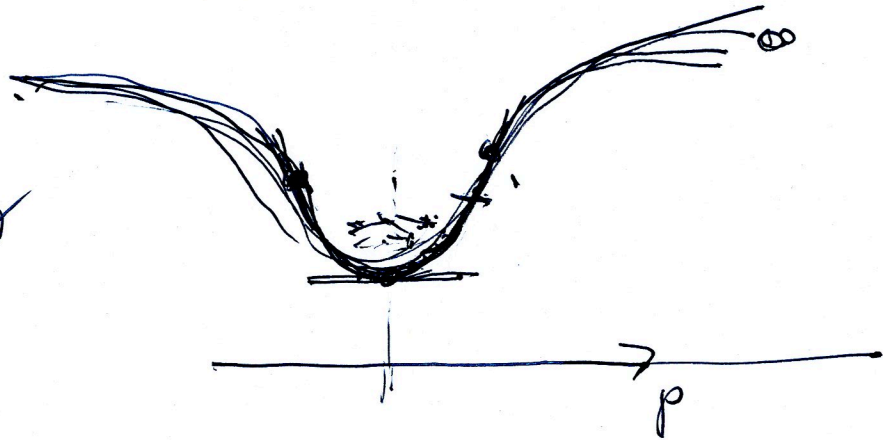
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50% points can be bad...

# Robust least squares

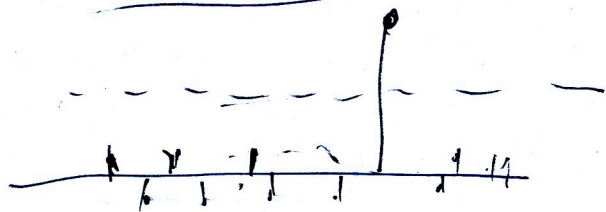
$$\chi^2 = \sum_{i=0}^N r_i^2$$

$$\frac{\partial f}{\partial \tau} = 0$$



(5)

## Gradient descent.



Robust cost function :

$$\ln\left(1 + \frac{r_i^2}{\sigma^2}\right)$$

$$\sum \ln\left(1 + \frac{r_i^2}{\sigma^2}\right)$$

$$\ln(100) \sim 5$$

$$\ln(1+x) = x + \dots$$

$\sigma$  is x over between parabolic cost function and the log.

# Vector Norms

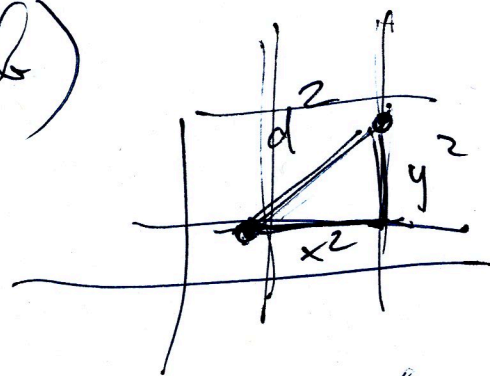
$$\vec{x} = (x_1, \dots, x_n)$$

$$\|\vec{x}\| = \sqrt{x_1^2 + x_2^2 + \dots + x_n^2}$$

Euclidean norm  
L2 norm.

L2  $\|\vec{x}\|_2$

L1  $\|\vec{x}\|_1 = \sum |x_i|$  Manhattan norm (taxicab)



Lp  $\|\vec{x}\|_p = \left[ \sum_{i=1}^n (x_i)^p \right]^{1/p}$  p-norm

$$[1, -2, 4, 3]$$

$$\vec{x} = (x, y) \quad L_2(\vec{x}) = 1$$

L1:  $1 + 2 + 4 + 3 = 10$

L2:  $\sqrt{1 + 4 + 16 + 9} = \sqrt{30}$

L $\infty$   $\lim_{p \rightarrow \infty} (4^p)^{1/p} = 4 \Rightarrow \max(x_i)$

