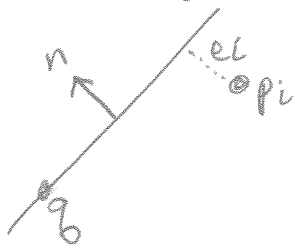


Line Fitting



scalar length

$$e_i = (p_i - q_0)^T n$$

①

Sum of squared errors

$$= \chi^2 = \sum_i n^T (p_i - q_0) (p_i - q_0)^T n$$

$$= \sum_i n^T p_i p_i^T n - n^T p_i q_0^T n - n^T q_0 p_i^T n + n^T q_0 q_0^T n$$

We will take $\frac{\partial \chi}{\partial q}$. To make this easy, rearrange each term

to start with q^T . (Because $\frac{\partial}{\partial q} q^T A = A$, $\frac{\partial}{\partial q} q^T A q = 2Aq$.)

Noting that $a^T b = b^T a$:

$$\chi^2 = \sum_i n^T p_i p_i^T n - \underbrace{q_0^T n n^T p_i - q_0^T n n^T p_i}_{\text{this term will drop out}} + q_0^T n n^T q_0$$

$$\frac{\partial \chi^2}{\partial q} = \sum_i -\cancel{q_0^T n n^T p_i} + \cancel{q_0^T n n^T p_i} = 0$$

$$= n n^T \left[\underbrace{\left(\sum_i p_i \right)}_{\text{Sure-fire way to solve is to make this term } \rightarrow 0!} + N q_0 \right] = 0$$

optimal translation
 (independent of n !)

$$\Rightarrow q_0 = \frac{1}{N} \sum_i p_i$$

Line Fitting: orientation

Recall: $\chi^2 = \sum_i n^T (p_i - g)(p_i - g)^T n$

covariance of points p , except for no $\frac{1}{n}$ term.

minimize error by picking n :

(recall: $\frac{\partial}{\partial x} x^T A x = 2Ax$)

$$\frac{\partial \chi^2}{\partial n} = \left(\sum_i (p_i - g)(p_i - g)^T \right) n = 0.$$

$$\Rightarrow An = 0.$$

This is a null-space problem, likely with no exact answer. Want to find n that gets us as close as possible,

Consider SVD of A :

$$USV^T n = 0$$

(note: $U=V$ since A is symmetric)

$$\Rightarrow \begin{bmatrix} | & | \\ u_1 & u_2 \\ | & | \end{bmatrix} \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix} \begin{bmatrix} - \\ - \\ - \\ - \end{bmatrix} n = 0$$

(note $u_1^T u_2 = 0$ because U is orthonormal)

if $n = u_2$,

(note: $\lambda_2 \leq \lambda_1$)

$$\begin{bmatrix} | & | \\ u_1 & u_2 \\ | & | \end{bmatrix} \begin{bmatrix} \lambda_1 \\ \lambda_2 \end{bmatrix} \begin{bmatrix} - \\ - \\ - \\ - \end{bmatrix} \begin{bmatrix} | \\ u_2 \\ | \\ 1 \end{bmatrix}$$

$\begin{bmatrix} 0 \\ 1 \end{bmatrix}$

$\begin{bmatrix} 0 \\ \lambda_2 \end{bmatrix}$

Key No value of n will make An shorter than u_2 !

U is orthonormal, so it won't change length.

2x2 SVD of Symmetric matrices

Let x be a unit vector, consider

$x^T A x$ max/min values are λ_i when $x = u_i$ (Prove using SVD!)
 eigenvector.

$$x^T \begin{bmatrix} a & b \\ b & c \end{bmatrix} x = au^2 + 2buv + cv^2 \quad \text{Let } x = \begin{bmatrix} u \\ v \end{bmatrix}$$

In 2D, u & v can be written in terms of one var, θ . $\begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} \cos \theta \\ \sin \theta \end{bmatrix}$

$$\lambda = a \cos^2 \theta + 2ab \cos \theta \sin \theta + c \sin^2 \theta$$

Maximize: by diff wrt θ :

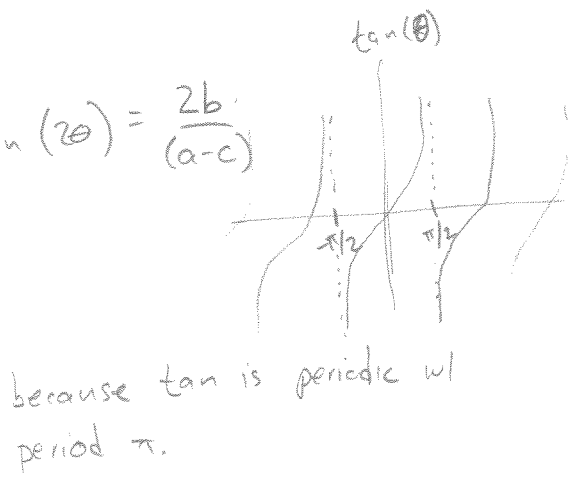
$$\frac{d\lambda}{d\theta} = -2a \cos \theta \sin \theta + 2b \cos^2 \theta - 2b \sin^2 \theta + 2c \sin \theta \cos \theta = 0$$

$$= -a \sin 2\theta + 2b \cos 2\theta + c \sin 2\theta = 0$$

$$\Rightarrow 2b \cos 2\theta = (a-c) \sin 2\theta$$

$$\Rightarrow \frac{\sin 2\theta}{\cos 2\theta} = \frac{2b}{(a-c)} \Rightarrow \tan(2\theta) = \frac{2b}{(a-c)}$$

$$\Rightarrow \theta = \frac{1}{2} \left[\arctan \left(\frac{2b}{(a-c)} \right) + \pi k \right]$$



$$= \frac{1}{2} \arctan \left(\frac{2b}{(a-c)} \right) + \frac{\pi}{2} k$$

The line with $\theta = \theta_0$ and $\theta = \theta_0 + \pi$ are the same line,

So the eigen vectors of A appear at increments of 90° !

- equiv. slants
- > The major and minor axes of con are always perpendicular
 - > Eigenvectors of sym. A are perpendicular
 - > The worst line is the line perpendicular

SVD

which eigenvector?

We want dominant vector, i.e., direction of line.

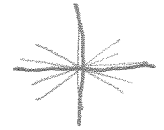
③

$$\Theta = \frac{1}{2} \arctan\left(\frac{2b}{a-c}\right) + \frac{\pi}{2} k$$

Visualize relationship between $b, (a-c)$ and the line direction.

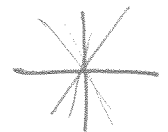
$$(a-c) > 0$$

more in the x dir than y



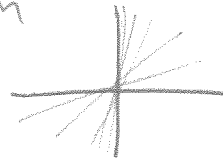
$$(a-c) < 0$$

" " " y " " x



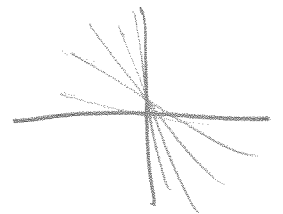
$$b > 0$$

positive correlation



$$b < 0$$

negative correlation



case#	"x" $a-c$	"y" $2b$	about Arctan to be ←
①	≥ 0	≥ 0	$0, \pi/2$
②		< 0	$-\pi/2, 0$
③	< 0	≥ 0	$\pi/2, \pi$
④		< 0	$-\pi, -\pi/2$

recall that line direction is $\frac{1}{2} \arctan$

③	①	> for each case, verify that it is in the desired region. (yes!)
④	②	> This also corresponds exactly to $\arctan 2(2b, a-c)$!

$$\text{So } \Theta_{\max} = \arctan 2(2b, a-c)$$

$$\Theta_{\text{normal}} = \Theta_{\max} + \frac{\pi}{2}$$

Computing Error for a Line

(Given g, n).

(4)

Back to e_i !

$$e_i = (p_i - g)^T n$$

$$\begin{aligned} \chi^2 &= \sum_i e_i^T e_i = \sum_i n^T (p_i - g) (p_i - g)^T n \\ &= n^T \left(\sum_i (p_i - g) (p_i - g)^T \right) n \end{aligned}$$

$$= n^T \left[\sum_i (p_i p_i^T - g p_i^T - p_i g^T + g g^T) \right] n$$

Recall that
 $g = \frac{1}{N} \sum_i p_i$!

$$= n^T \left[\sum_i (p_i p_i^T) - N g g^T - N g g^T + N g g^T \right] n$$

$$= n^T \left[\left(\sum_i p_i p_i^T \right) - N g g^T \right] n$$

We pick n to minimize this $g^T n$.

Each of these terms can be efficiently computed incrementally!

Remarks

The error is λ_{\min} of the covariance matrix!

► $x^T A x$ is min/max when x is an eigenvector of A (assume A symmetric)

► n was chosen to minimize $n^T A n$, QED.