

9/29/06

Note Title

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$$\min_{\vec{x}} \|A\vec{x} - \vec{y}\|^2$$

$$= \min_{\vec{x}} \underbrace{(A\vec{x} - \vec{y}, A\vec{x} - \vec{y})}$$

$$(A\vec{x}, A\vec{x}) - (A\vec{x}, \vec{y}) - (A\vec{x}, \vec{y}) + (\vec{y}, \vec{y})$$

$$\text{claim } \nabla_{\vec{x}} \underbrace{(A\vec{x}, \vec{y})} = A^T \vec{y}$$

$$\sum_i y_i [A\vec{x}]_i$$

$$= \sum_i y_i \sum_j A_{ij} x_j$$

$$= \sum_{i,j} A_{ij} y_i x_j$$

$$= \sum_{i,j} x_j A_{ij} y_i$$

$$= \sum_{i,j} x_j A_{ji}^T y_i$$

$$= \sum_j x_j \underbrace{\sum_i A_{ji}^T y_i}_{(A^T \vec{y})_j}$$

$$= (\vec{x}, A^T \vec{y})$$

So $\left[\nabla_{\vec{x}} (\vec{y}, A \vec{x}) \right]_k$

$$\frac{\partial}{\partial x_k} \left[\sum_j x_j (A^T \vec{y})_j \right]$$

zero except for $j=k$

$$\rightarrow = (A^T \vec{y})_k$$

$$\nabla_{\vec{x}} (\vec{y}, A \vec{x}) = A^T \vec{y}$$

$$\nabla_{\vec{x}}$$

$$(A\vec{x}, A\vec{x}) - (\vec{y}, A\vec{x}) - (A\vec{x}, \vec{y}) + (\vec{y}, \vec{y})$$

took care
of these
just now

\Downarrow
0

$$\nabla_{\vec{x}} (A\vec{x}, A\vec{x}) = ?$$

$\frac{d}{dx} [x \cdot x]$ ordinary
scalar variables

$$= \frac{d}{dx}(x) \cdot x + x \frac{d}{dx}(x)$$

$$\text{So } \nabla_{\vec{x}} (A\vec{x}, A\vec{x}) =$$

$$\begin{aligned} & \nabla_{\vec{x}} (\vec{x}, A^T A \vec{x}) + \nabla_{\vec{x}} (A^T A \vec{x}, \vec{x}) \\ & = 2 A^T A \vec{x} \end{aligned}$$

$$\nabla_{\vec{x}} \|A\vec{x} - \vec{y}\|^2 =$$

$$\nabla_{\vec{x}} \left\{ (A\vec{x}, A\vec{x}) - (A\vec{x}, \vec{y}) - (\vec{y}, A\vec{x}) + (\vec{y}, \vec{y}) \right\}$$

$$= 2A^T A \vec{x} - A^T \vec{y} - A^T \vec{y} + 0$$

So if we set $\nabla_{\vec{x}} \|A\vec{x} - \vec{y}\|^2 = 0$

$$\Rightarrow \boxed{A^T A \vec{x} = A^T \vec{y}} \quad \text{Normal Equation}$$

Notes!

$A^T A$ is always square and symmetric

$$A \in \mathbb{R}^{n \times m} \Rightarrow A^T \in \mathbb{R}^{m \times n}$$

So $A^T A$ involves

sum over n $m \times n \cdot n \times m$

$$\Rightarrow A^T A \in \mathbb{R}^{m \times m}$$

$$\text{and } (A^T A)_{ij} = \sum_{k=1}^n A_{ik}^T A_{kj}$$

$$= \sum_k A_{ki} A_{jk}^T$$

$$= \sum_k A_{jk}^T A_{ki}$$

$$= (A^T A)_{ji}$$

So $A^T A$ is symmetric.

Summary on
Least Squares

if $A\vec{x}=\vec{y}$ has no solution

we can seek a generalized

solution:

$$\min_{\vec{x}} \|A\vec{x}-\vec{y}\|^2$$

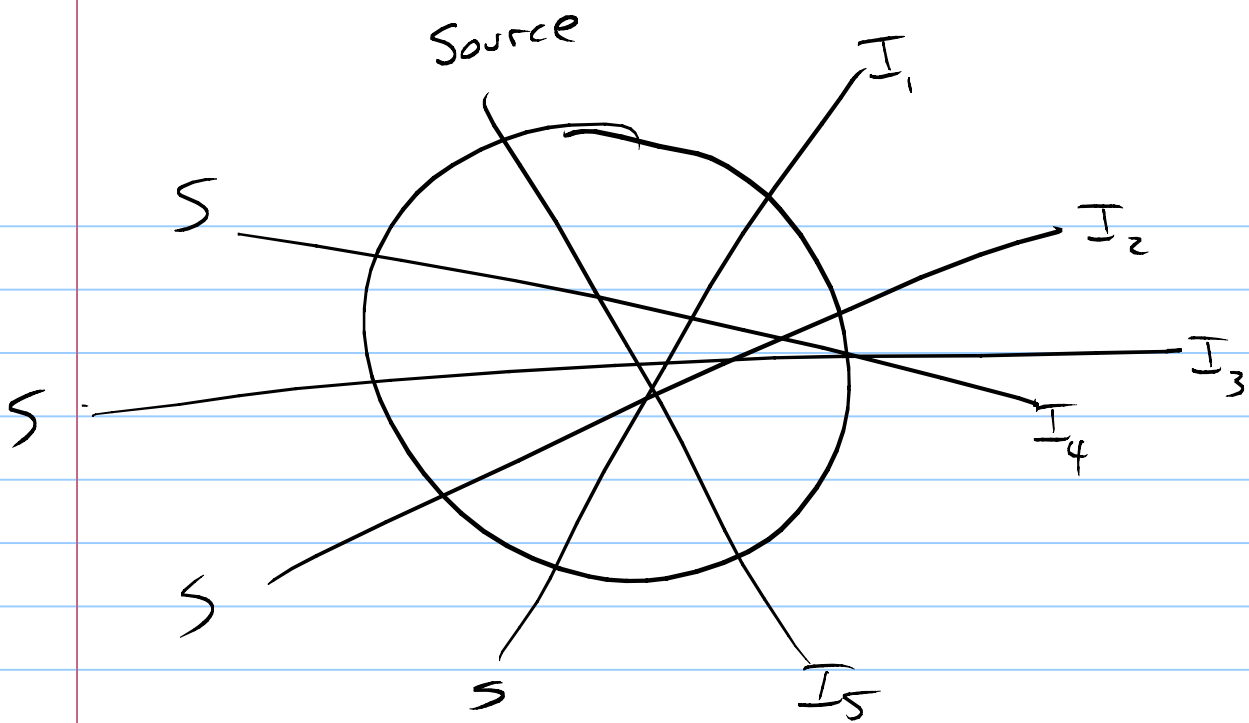
$$\Rightarrow A^T A \vec{x} = A^T \vec{y}$$

True for any A .

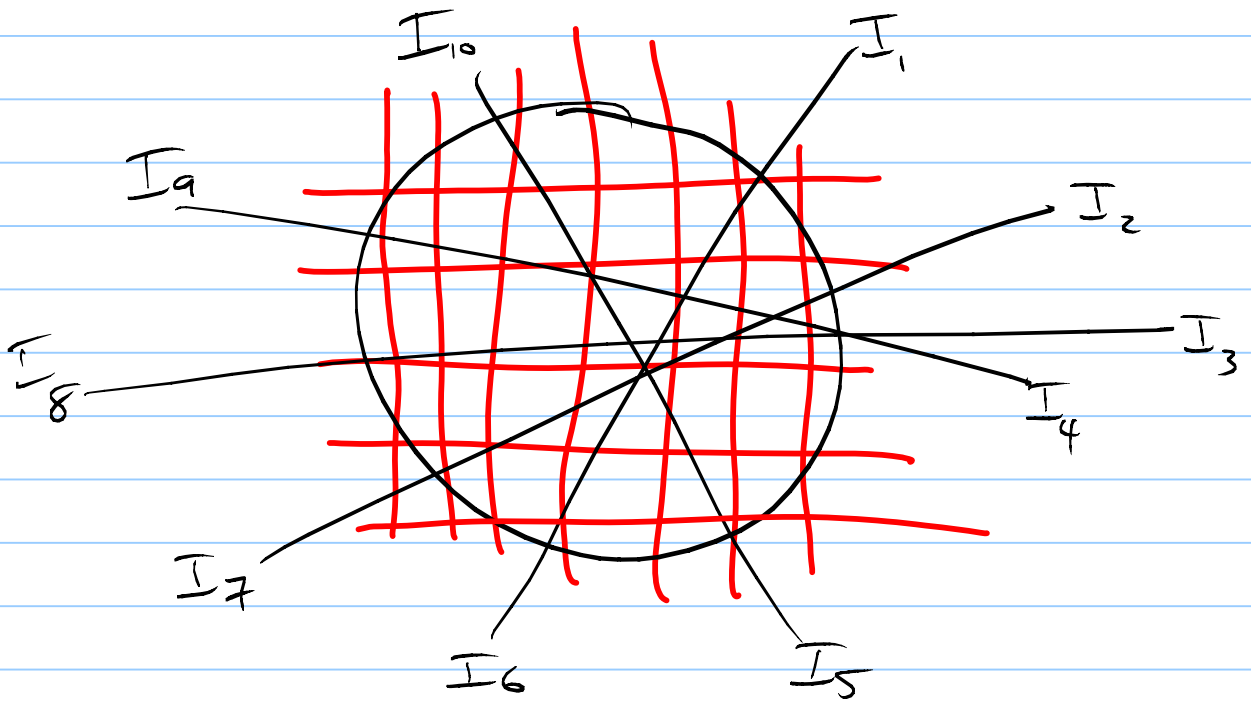
more real examples

·) GPS triangulation

·) X-ray tomography
CAT Scan



Grid the target



m cells

n data

Each msmt corresponds to
1 row of a matrix

Each parameter value (cell)
correspond to 1 column

$$J \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_m \end{bmatrix} = \begin{bmatrix} I_1 \\ I_2 \\ \vdots \\ I_N \end{bmatrix}$$

↓
absorption

↘ measured
x-ray
intensity

$$J \in \mathbb{R}^{m \times n}$$

Solve $\min_{\vec{a}} \|\vec{J}\vec{a} - \vec{I}\|^2$

Tomography

Eigenvalues / Eigenvectors

Consider the action of a generic matrix on a vector:

$$A \vec{x} \quad A \in \mathbb{R}^{n \times n} \quad \vec{x} \in \mathbb{R}^n$$

Normally this will change direction and length of \vec{x}

SUPPOSE

$$A \vec{x} = \alpha \vec{x}$$

↓

Scalar

So A maps \vec{x} into a scalar multiple of itself.

$$A\vec{x} = \alpha\vec{x} \quad \star$$

$$\Rightarrow A\vec{x} = \alpha I\vec{x}$$

$\hookrightarrow n \times n$ identity matrix

$$I \cdot \vec{x} = \vec{x}$$

$$\hookrightarrow (A - \alpha I)\vec{x} = 0$$

So, for \star to be true

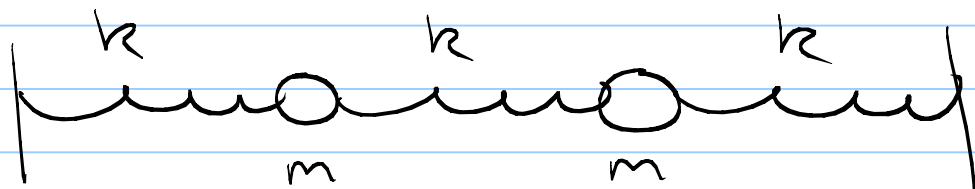
\vec{x} must be in the null

space of $A - \alpha I$

if $A - \alpha I$ has a null space, it cannot be invertible. So,

$$\boxed{\text{Det}(A - \alpha I) = 0}$$

Example



can show

$$\begin{aligned} \ddot{X}_1 &= -2\omega_0^2 X_1 + \omega_0^2 X_2 \\ \ddot{X}_2 &= -2\omega_0^2 X_2 + \omega_0^2 X_1 \end{aligned}$$

Seek solutions: $X_1 = A e^{i\omega t}$
 $X_2 = B e^{i\omega t}$

Plug into \star

$$\underbrace{\begin{pmatrix} 2\omega_0^2 & -\omega_0^2 \\ -\omega_0^2 & 2\omega_0^2 \end{pmatrix}}_{\text{matrix}} \underbrace{\begin{pmatrix} A \\ B \end{pmatrix}}_{\text{vect}} = \underbrace{\omega^2}_{\text{scalar}} \underbrace{\begin{pmatrix} A \\ B \end{pmatrix}}_{\text{vector}}$$

$$K \vec{u} = \omega^2 \vec{u}$$
$$\vec{u} = \begin{pmatrix} A \\ B \end{pmatrix} \quad K = \begin{pmatrix} 2\omega_0^2 & -\omega_0^2 \\ -\omega_0^2 & 2\omega_0^2 \end{pmatrix}$$

Examples

$$A = \begin{pmatrix} 5 & 1 \\ 1 & 5 \end{pmatrix}$$

$$Ax = \lambda Ix$$

$$\Rightarrow (A - \lambda I)x = 0$$

$$\Rightarrow \text{Det}(A - \lambda I) = 0$$

$$A - \lambda I = \begin{bmatrix} 5 & 1 \\ 1 & 5 \end{bmatrix} - \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix}$$

$$= \begin{pmatrix} 5-\lambda & 1 \\ 1 & 5-\lambda \end{pmatrix}$$

$$\text{Det}(A - \lambda I) =$$

$$(5-\lambda)^2 - 1 = 0$$

$$(5-\lambda) = \pm 1$$

$$5 \mp 1 = \lambda$$

$$\text{So } \lambda = 6 \text{ or } 4$$

now for the ξ -vectors

$$\begin{pmatrix} 5 & 1 \\ 1 & 5 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = 6 \begin{pmatrix} x \\ y \end{pmatrix}$$

$$\Rightarrow 5x + y = 6x$$

$$x + 5y = 6y$$

$$x=1, y=1$$

$$\begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

eig. vector

$$\boxed{\begin{pmatrix} 5 & 1 \\ 1 & 5 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = 6 \begin{pmatrix} 1 \\ 1 \end{pmatrix}}$$

6, $\begin{pmatrix} 1 \\ 1 \end{pmatrix}$ is an eigenvalue

eigenvector pair

Next

$$\begin{pmatrix} 5 & 1 \\ 1 & 5 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = 4 \begin{pmatrix} x \\ y \end{pmatrix}$$

$$5x + y = 4x$$

$$x + 5y = 4y$$

$$\begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

So $4, \begin{pmatrix} 1 \\ -1 \end{pmatrix}$ is the

second Σ -value/vector pair.