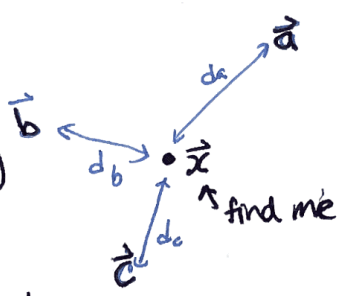


Summary: Last Lecture

Multiple Beacons A, B, C sing different (orthogonal) songs at once. Their positions known.



Each has a shift of  $\{N_a, N_b, N_c\}$  in # samples

↳ convert to distance by  $d_a = N_a \cdot T_s \cdot v$  → distance!  
# samples    Time/sample    speed (m/s)

Received signal at  $\vec{x}$  is  $\vec{r} = \vec{s}_a^{(N_a)} + \vec{s}_b^{(N_b)} + \vec{s}_c^{(N_c)}$   
shift  
Song of beacon a    beacon index

How to find shifts from received signal? Cross-correlate  $\vec{r}$  with  $\vec{s}_a, \vec{s}_b, \vec{s}_c$

Cross-correlation of  $\vec{r}$  with  $\vec{s}_a$   
 $\vec{p}_{\vec{r}, \vec{s}_a} = C_{\vec{s}_a} \cdot \vec{r}$   
circulant matrix of  $\vec{s}_a$

Find peak/max. value of  $\vec{p}_{\vec{r}, \vec{s}_a}$   
 ↳ index of max. value is shift  $N_a$

\* Then, repeat for  $\vec{s}_b, \vec{s}_c$

$= C_{\vec{s}_a} [\vec{s}_a^{(N_a)} + \vec{s}_b^{(N_b)} + \vec{s}_c^{(N_c)}]$

$= C_{\vec{s}_a} \vec{s}_a^{(N_a)} + C_{\vec{s}_a} \vec{s}_b^{(N_b)} + C_{\vec{s}_a} \vec{s}_c^{(N_c)}$   
want large @  $N_a$     want small

Cross-corr.  $\vec{s}_a$  with  $\vec{s}_b$  shifted → should not be corr.

Want:  $\vec{p}_{\vec{r}, \vec{s}_a}[m]$  to be max. @  $m = N_a$   
mth element of corr. vector is just an inner product.

$\vec{p}_{\vec{r}, \vec{s}_b}[j]$  to be max. @  $j = N_b$

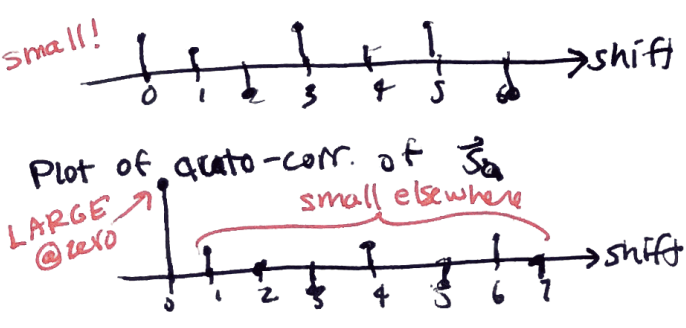
$\vec{p}_{\vec{r}, \vec{s}_c}[i]$  to be max. @  $i = N_c$

~~received signal~~

Conditions:

- 1)  $\vec{s}_a$  auto-corr @ non-zero shifts << than auto-corr @ zero
- 2)  $\vec{p}_{\vec{s}_a, \vec{s}_b}$  very small (almost orthogonal)

Plots of cross-corr.  $\vec{s}_a$  and  $\vec{s}_b$ :



How should I generate songs?

e.g. by fair coin flips (cs 70)

$$\|\vec{x} - \vec{a}\|^2 = d_a^2 \rightarrow \|\vec{x}\|^2 - 2\vec{a}^T \vec{x} + \|\vec{a}\|^2 = d_a^2 \quad (1) \quad \leftarrow \text{solve? will give circle, nonlinear!} \quad (2)$$

$$\|\vec{x} - \vec{b}\|^2 = d_b^2 \rightarrow \|\vec{x}\|^2 - 2\vec{b}^T \vec{x} + \|\vec{b}\|^2 = d_b^2 \quad (2) \quad \text{Need more info...}$$

$$\|\vec{x} - \vec{c}\|^2 = d_c^2 \rightarrow \|\vec{x}\|^2 - 2\vec{c}^T \vec{x} + \|\vec{c}\|^2 = d_c^2 \quad (3)$$

To get rid of non-linear  $\|\vec{x}\|^2$  term, try:

$$(2) - (1) \quad -2\vec{b}^T \vec{x} + 2\vec{a}^T \vec{x} + \|\vec{b}\|^2 - \|\vec{a}\|^2 = d_b^2 - d_a^2$$

$$(3) - (1) \quad -2\vec{c}^T \vec{x} + 2\vec{a}^T \vec{x} + \|\vec{c}\|^2 - \|\vec{a}\|^2 = d_c^2 - d_a^2$$

Write as matrix-vector multiply (LINEAR!)

$$\begin{bmatrix} (\vec{a} - \vec{b})^T \\ (\vec{a} - \vec{c})^T \end{bmatrix} \begin{bmatrix} \vec{x} \end{bmatrix} = \begin{bmatrix} \frac{1}{2}(d_b^2 - d_a^2) - \frac{1}{2}(\|\vec{b}\|^2 - \|\vec{a}\|^2) \\ \frac{1}{2}(d_c^2 - d_a^2) - \frac{1}{2}(\|\vec{c}\|^2 - \|\vec{a}\|^2) \end{bmatrix}$$

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

$$\text{so } \vec{x} = A^{-1} \vec{b}$$

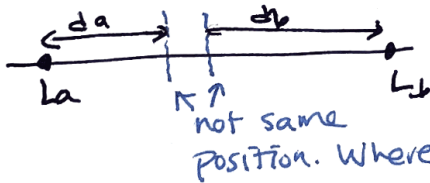
\* solves when  $\vec{a} - \vec{b}$  and  $\vec{a} - \vec{c}$  are linearly independent! (i.e. not collinear)

→ if dependent, only 1 eqn (of a line) → ∞ sol!

### Imperfect Measurements:

what if 3 circles don't have 1 perfect intersection?

e.g. 1D case



\* "inconsistent meas." case from M1

solutions:

- take more measurements of same kind. + average?
- take more 'diverse' measurements. ← different (e.g. more beacons) ← How to average?
- weight meas. by how accurate you think they are

Example: Estimate  $\hat{x}$

$$e_1(\hat{x}) = |a_1 \hat{x} - b_1|$$

$$e_2(\hat{x}) = |a_2 \hat{x} - b_2|$$

error for meas. 1

$$\text{maybe minimize } e_{\text{abs}}(\hat{x}) = e_1(\hat{x}) + e_2(\hat{x}) = |a_1 \hat{x} - b_1| + |a_2 \hat{x} - b_2|$$

$$\text{or minimize square: } e_{\text{sq}}(\hat{x}) = e_1^2(\hat{x}) + e_2^2(\hat{x}) = |a_1 \hat{x} - b_1|^2 + |a_2 \hat{x} - b_2|^2$$

use this!

$$\rightarrow = (a_1 \hat{x} - b_1)^2 + (a_2 \hat{x} - b_2)^2$$

How to minimize? Take derivative!

$$\frac{\partial}{\partial \hat{x}} (e^2(\hat{x})) = \frac{d}{d\hat{x}} \left[ (a_1 \hat{x} - b_1)^2 \right] + \frac{d}{d\hat{x}} (a_2 \hat{x} - b_2)^2 = 0$$

$$2a_1(a_1 \hat{x} - b_1) + 2a_2(a_2 \hat{x} - b_2) = 0$$

$$a_1^2 \hat{x} + a_2^2 \hat{x} - a_1 b_1 - a_2 b_2 = 0$$

$$\hat{x} = \frac{a_1 b_1 + a_2 b_2}{a_1^2 + a_2^2}$$

Does it make sense?

looks like inner prod!

looks like norm!

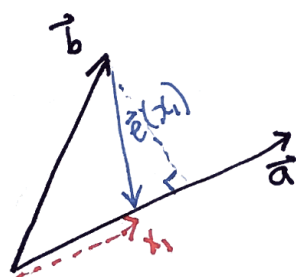
Together, looks like projection!

$$\hat{x} = \frac{\langle \vec{a}, \vec{b} \rangle}{\|\vec{a}\|^2}$$

How is min. error related to projection?

Error should be orthogonal to vector space!

$$\vec{a} = \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} \quad \vec{b} = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$$



$$\|x\| = \|a x - b\|$$

soln is restricted to be along a dir.

$$\begin{aligned} \vec{\hat{b}} &= \perp \text{proj of } \vec{b} \text{ onto } \vec{a} \\ &= \frac{\langle \vec{a}, \vec{b} \rangle}{\|\vec{a}\|^2} \end{aligned}$$

orthogonal projection onto allowed space will ALWAYS be one with smallest error!

PROOF

Assume I have some subspace

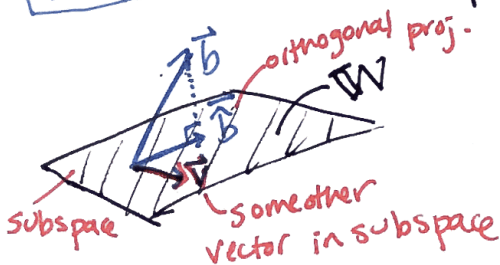
$W \in \mathbb{R}^N$   
↑  
subspace

, and some vector  $\vec{b} \notin W$  but  $\vec{b} \in \mathbb{R}^N$   
↑      ↑  
vector    not in subspace  $W$

$\vec{\hat{b}}$  = orthogonal proj. of  $\vec{b}$  onto  $W$  ← prove it's the proj. w smallest error.

How?

Let's draw a picture.



want to minimize  $\|\vec{v} - \vec{b}\|$

$$\text{or, } \vec{\hat{b}} = \underset{\vec{v}}{\text{arg min}} \|\vec{v} - \vec{b}\| \rightarrow \|\vec{b} - \vec{v}\| > \|\vec{b} - \vec{\hat{b}}\|$$

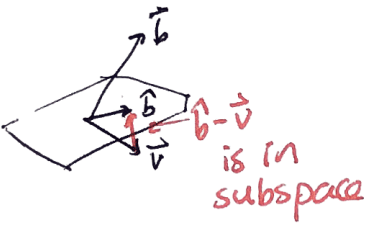
means the value of  $\vec{v}$  at which we have min.  $\|\vec{v} - \vec{b}\|$

equiv.

How to prove this?

Start with error vector:  $\vec{b} - \vec{v} = \vec{b} - \vec{v} + \vec{\hat{b}} - \vec{\hat{b}}$   
 want  $\hat{b}$  somewhere...  $= (\vec{b} - \vec{\hat{b}}) + (\vec{\hat{b}} - \vec{v})$

$$\begin{aligned} \|\vec{b} - \vec{v}\|^2 &= (\vec{b} - \vec{v})^T (\vec{b} - \vec{v}) \\ &= [(\vec{b} - \vec{\hat{b}})^T + (\vec{\hat{b}} - \vec{v})^T] [(\vec{b} - \vec{\hat{b}}) + (\vec{\hat{b}} - \vec{v})] \\ &= \|\vec{b} - \vec{\hat{b}}\|^2 + \|\vec{\hat{b}} - \vec{v}\|^2 + 2(\vec{b} - \vec{\hat{b}})^T (\vec{\hat{b}} - \vec{v}) \end{aligned}$$



so  $\|\vec{b} - \vec{v}\|^2 = \|\vec{b} - \vec{\hat{b}}\|^2 + \|\vec{\hat{b}} - \vec{v}\|^2$   
 ↑  
 $> 0$

$\|\vec{b} - \vec{v}\|^2 \rightarrow \|\vec{b} - \vec{\hat{b}}\|^2$  so  $\hat{b}$  perpendicular is min. error sol'n!  
 $\vec{v} \in W, \vec{v} \neq \vec{\hat{b}}$  error to subspace means EVERY vector

I have  $A\vec{x} = \vec{b}$   
 $\begin{bmatrix} \phantom{0} \end{bmatrix} \begin{bmatrix} \phantom{0} \end{bmatrix} = \begin{bmatrix} \phantom{0} \end{bmatrix}$

(A no longer square!)

# measurements = m  
 # unknowns = n

OVERDETERMINED  
 $m > n$   
 "extra measurement"

Extra measurements may not be redundant if they contain noise! (or error)

- Two cases: 1)  $\vec{b} \in \text{span}(A)$   
 2)  $\vec{b} \notin \text{span}(A)$  ← what do I do?

want  $\vec{x}$  to minimize  $\|A\vec{x} - \vec{b}\|$

what is subspace?  $\text{span}(A) = \text{span}\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n\}$   
 what is measurement?  $\vec{b}$

so  $\vec{\hat{b}} = A\vec{x}$  orthogonal proj. of  $\vec{b}$  onto  $W$

$\vec{b} - \vec{\hat{b}} \perp W$

recall: all cols of A are in subspace, so:

$$\left. \begin{aligned} \vec{b} - \vec{\hat{b}} \perp \vec{a}_1 &\rightarrow \vec{a}_1^T (\vec{b} - \vec{\hat{b}}) = 0 \\ \vec{b} - \vec{\hat{b}} \perp \vec{a}_2 &\rightarrow \vec{a}_2^T (\vec{b} - \vec{\hat{b}}) = 0 \\ \vdots &\vdots \\ \vec{b} - \vec{\hat{b}} \perp \vec{a}_n &\rightarrow \vec{a}_n^T (\vec{b} - \vec{\hat{b}}) = 0 \end{aligned} \right\} \begin{bmatrix} \vec{a}_1^T \\ \vec{a}_2^T \\ \vdots \\ \vec{a}_n^T \end{bmatrix} (\vec{b} - \vec{\hat{b}}) = 0$$

$$A^T (\vec{b} - \vec{\hat{b}}) = 0$$

$$\begin{aligned} A^T \vec{\hat{b}} &= A^T \vec{b} \\ (A^T A) \vec{x} &= A^T \vec{b} \end{aligned}$$

Least-squares solution to  $Ax = b$

$$\vec{x} = (A^T A)^{-1} A^T \vec{b}$$

iff  $A^T A$  invertible