

EECS 16A
August 11, 2020
Lecture 7B
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Today:

- more OMP:
- Graphical interpretation
- OMP for sparse imaging

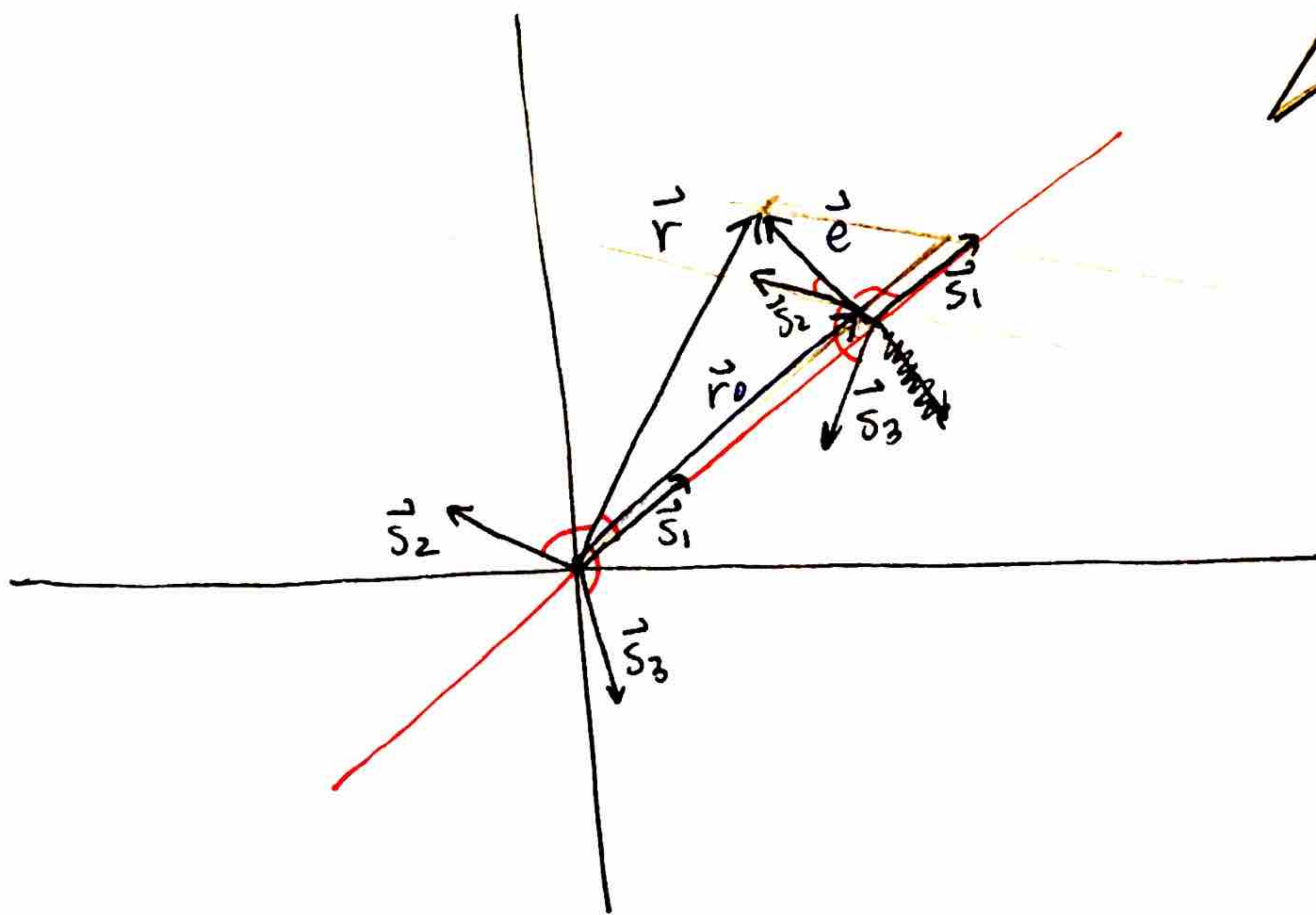
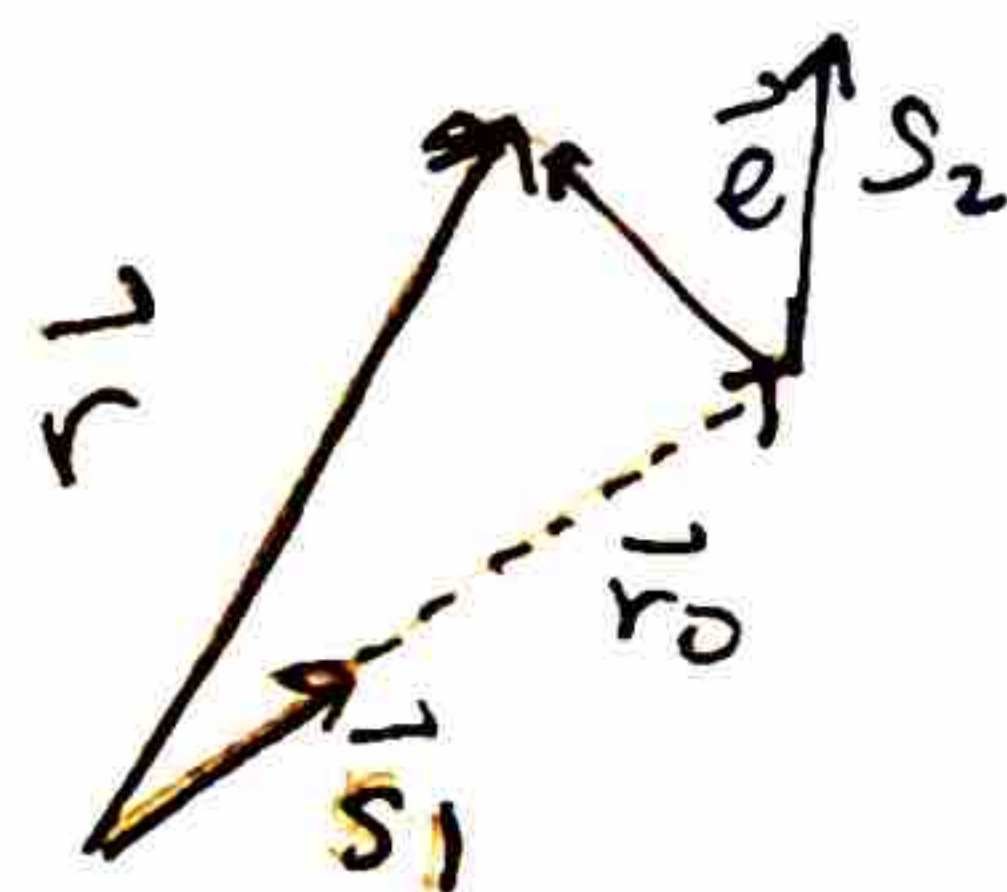
①

course evals: <https://course-evaluations.berkeley.edu>

Solve $S\vec{\alpha} = \vec{r}$ $\vec{r} \in \mathbb{R}^2$

\uparrow

$[\vec{s}_1, \vec{s}_2, \vec{s}_3]$



- ① Take inner product of \vec{r} with every \vec{s}_i and find the max abs. of inner product

$$\langle \vec{r}, \vec{s}_i \rangle = \underbrace{\|\vec{r}\| \|\vec{s}_i\|}_{a \cos \theta_i} \cos \theta_i$$

$\theta_i =$ angle between \vec{r} and \vec{s}_i

\rightarrow find θ_i close to zero (or 180°)

This ex. \vec{s}_1 has max $\langle \vec{r}, \vec{s}_i \rangle$

- ② Update matrix A and vector \vec{x}
↳ keep track of which vectors we found

$$A = \vec{s}_1 \quad \vec{x} = \alpha_1$$

No graphical interpretation

- ③ Project \vec{r} onto $\text{col}(A)$
↳ $\text{span}\{A\} = \text{span}\{\vec{s}_1\}$

- ④ Update error
 $\vec{e} = \vec{r} - \vec{r}_0$

End iter 1.

Start iter 2.

- ① inner product of \vec{e} with \vec{s}_i
↳ expect $\langle \vec{e}, \vec{s}_1 \rangle = 0$
here $\langle \vec{e}, \vec{s}_2 \rangle$ is maximum

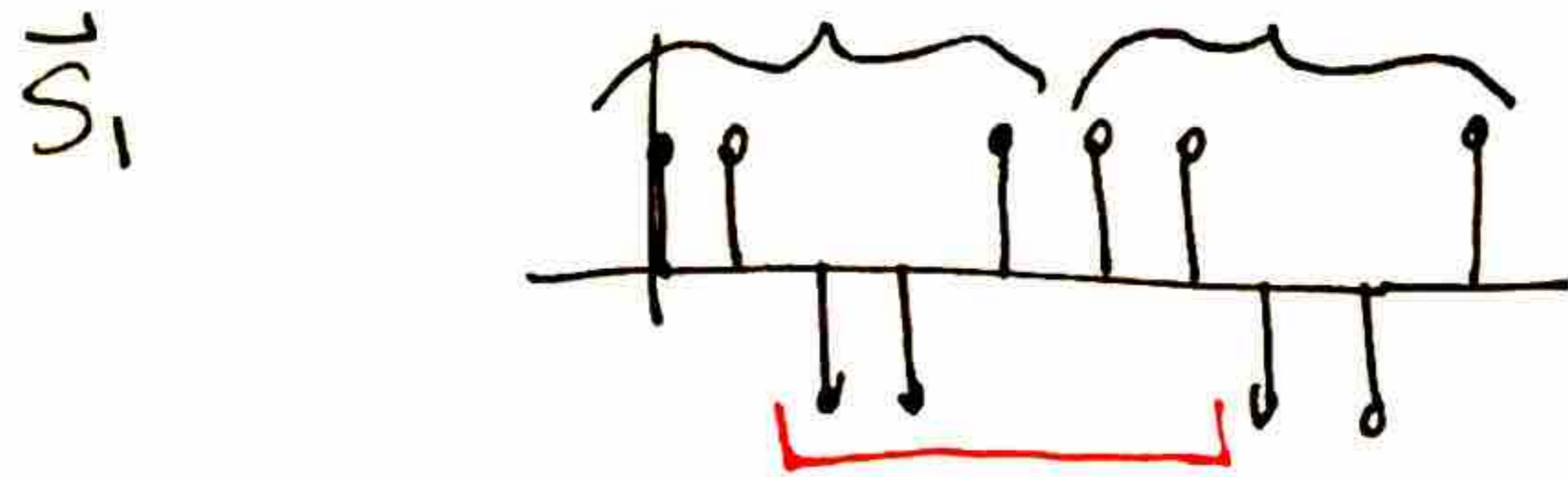
- ② Update A and \vec{x}
 $A = [\vec{s}_1, \vec{s}_2] \quad \vec{x} = \begin{bmatrix} \alpha_1 \\ \alpha_2 \end{bmatrix}$

- ③ project \vec{r} on to $\text{col}(A) = \text{span}\{\vec{s}_1, \vec{s}_2\}$
 $\hat{\vec{x}} = (A^T A)^{-1} A^T \vec{r}$
 $\vec{r}_0 = A \hat{\vec{x}}$

- ④ No more error. We're done!

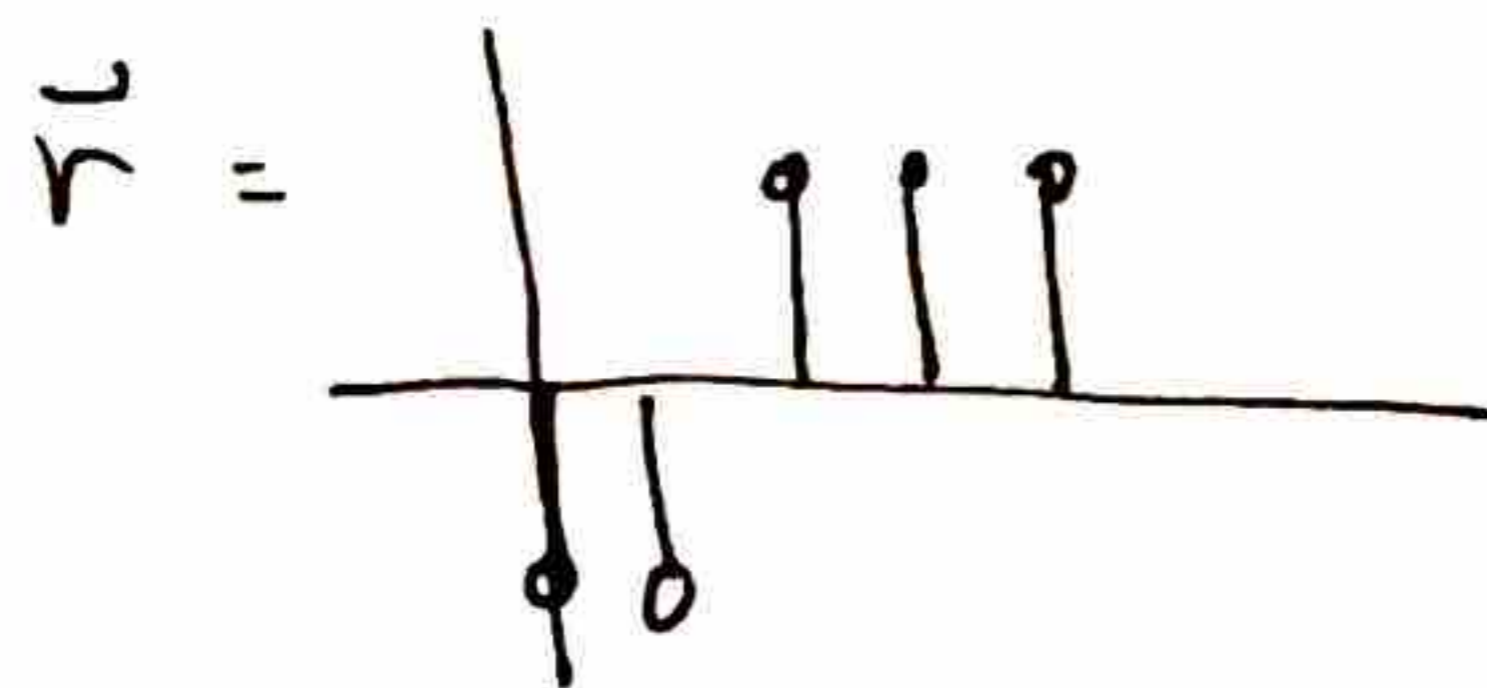
OMP with Delays

→ more realistic model



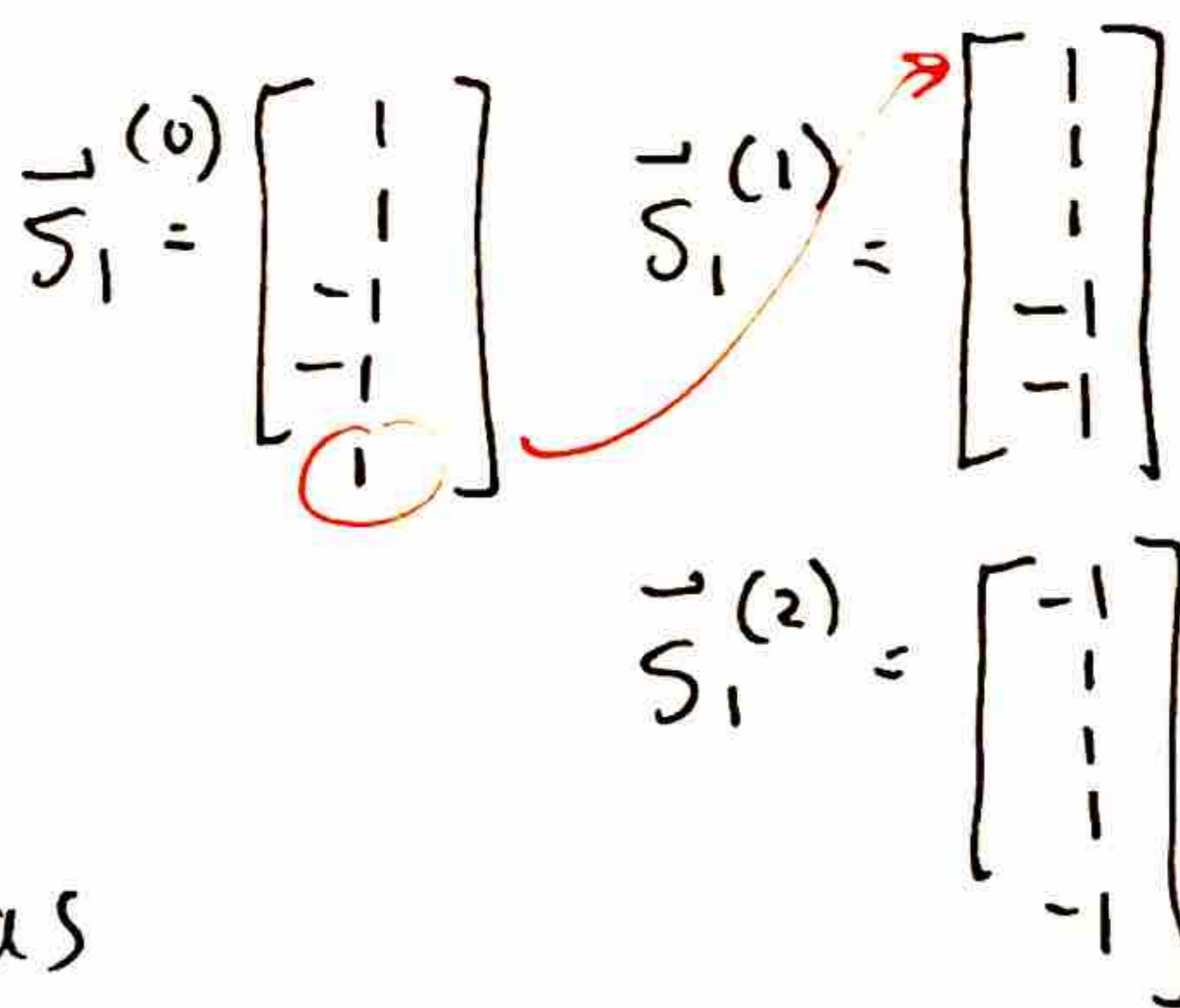
→ \vec{s}_1 now transmits a periodic signal

→ \vec{r} received signal has some unknown delay



$$\vec{r} = d_1 \vec{s}_1(\tau_1)$$

↑
unknown delay



We can still write this as matrix equation:

$$\vec{r} = \begin{bmatrix} \vec{s}_1^{(0)} & \vec{s}_1^{(1)} & \dots & \vec{s}_1^{(q)} \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_q \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ \alpha_1 \\ 0 \\ 0 \end{bmatrix}$$

→ OMP: take inner product w/ \vec{r} and every column
 ↳ how columns are shifted versions
 ↳ same as correlation.

What if we have 2 devices

$$\vec{r} = \alpha_1 \vec{s}_1(\tau_1) + \alpha_2 \vec{s}_2(\tau_2)$$

$$\vec{r} = \begin{bmatrix} \vec{s}_1^{(0)} & \vec{s}_1^{(1)} & \dots & \vec{s}_1^{(4)} & \vec{s}_2^{(0)} & \vec{s}_2^{(1)} & \dots & \vec{s}_2^{(4)} \end{bmatrix} \begin{bmatrix} \alpha_1 \\ 0 \\ 0 \\ 0 \\ 0 \\ \vdots \\ 0 \\ 0 \\ 0 \\ 0 \\ \alpha_2 \end{bmatrix}$$

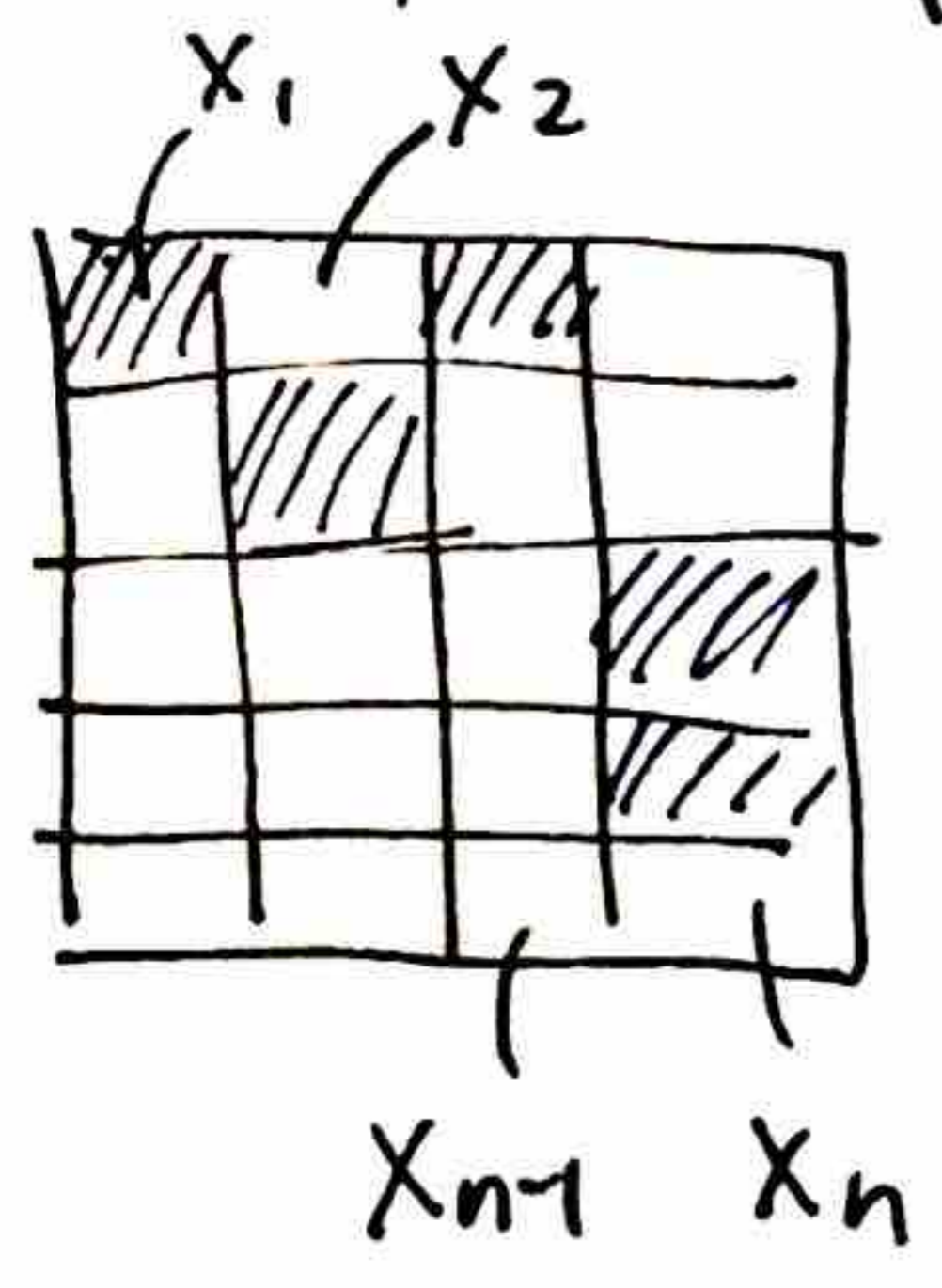
} 5 elements
} 5 elements

Same as correlation between \vec{r} and \vec{s}_1
 \vec{r} and \vec{s}_2

$$\vec{r} = \alpha_{11} \vec{s}_1^{(0)} + \alpha_{12} \vec{s}_1^{(1)} + \alpha_{13} \vec{s}_1^{(2)} + \dots + \alpha_{15} \vec{s}_1^{(4)} + \alpha_{21} \vec{s}_2^{(0)} + \alpha_{22} \vec{s}_2^{(1)} + \dots + \alpha_{25} \vec{s}_2^{(4)}$$

$\swarrow \vec{s}_1^{(4)}$
 $\nwarrow \vec{s}_2^{(4)}$

Example of OMP: Sparse Imaging



$$\begin{bmatrix} m_1 \\ m_2 \\ \vdots \\ m_n \end{bmatrix} = \begin{bmatrix} 0 & 1 & \dots & 1 \\ 1 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ \vdots & \vdots & \vdots & \vdots \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

A
} image

measurements

How many measurements for n pixels?
 → Module 1: n measurements

But, what if we know that the image is mostly zero? (image is sparse?)

- \vec{x} is sparse ex) night sky
- we can design A so that columns are nearly orthogonal fluorescent image

Example:

x_1	x_2	x_3
x_4	x_5	x_6

6 pixel image

We know that 2 pixels are non-zero
We're going to take 4 measurements.

$$\begin{bmatrix} 3 \\ 2 \\ 0 \\ 5 \end{bmatrix} = \vec{m} = \underbrace{\begin{bmatrix} 1 & 0 & 1 & 0 & 0 & 1 \\ 1 & 0 & 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 & 1 \\ 0 & 1 & 1 & 0 & 1 & 0 \end{bmatrix}}_A \vec{x}$$

\vec{a}_1 \vec{a}_2 \vec{a}_6

ITER 1

① $\langle \vec{m}, \vec{a}_i \rangle$

i	$\langle \vec{m}, \vec{a}_i \rangle$
1	5
2	8 5
3	8
4	2
5	7
6	3

$i=3$ maximizes the inner product

② Update a matrix w/ our column $B = \vec{a}_3$

③ Project \vec{m} onto the column space (B)

$$\begin{aligned} \vec{m}_0 &= B(B^T B)^{-1} B^T \vec{m} \\ &= \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix} (2)^{-1} (8) \\ &= \begin{bmatrix} 4 \\ 0 \\ 0 \\ 4 \end{bmatrix} \end{aligned}$$

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

$$\vec{m} = \begin{bmatrix} 3 \\ 2 \\ 0 \\ 5 \end{bmatrix}$$

④ Update error

$$\vec{e} = \vec{m} - \vec{m}_0 = \begin{bmatrix} -1 \\ 2 \\ 0 \\ 1 \end{bmatrix}$$

ITER 2

① calculate $\langle \vec{a}_i, \vec{e} \rangle$

i	$\langle \vec{e}, \vec{a}_i \rangle$
1	1
2	1
3	0
4	2
5	3
6	-1

← i=5 maximizes $\langle \vec{e}, \vec{a}_i \rangle$

② Update B

$$B = [\vec{a}_3, \vec{a}_5] = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 1 \end{bmatrix}$$

(3) Project \vec{m} on to $\text{col}(B)$

$$\vec{m}_0 = B \begin{bmatrix} \hat{x}_3 \\ \hat{x}_5 \end{bmatrix} \quad \begin{bmatrix} \hat{x}_3 \\ \hat{x}_5 \end{bmatrix} = (B^T B)^{-1} B^T \vec{m}$$

$$\begin{aligned} \begin{bmatrix} \hat{x}_3 \\ \hat{x}_5 \end{bmatrix} &= \left(\begin{bmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & -1 \\ 0 & 0 \\ 1 & -1 \end{bmatrix} \right)^{-1} \begin{bmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 3 \\ 2 \\ 0 \\ 5 \end{bmatrix} \\ &= \left(\begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \right)^{-1} \begin{bmatrix} 8 \\ 7 \end{bmatrix} \\ &= \frac{1}{3} \begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} 8 \\ 7 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} 9 \\ 6 \end{bmatrix} = \begin{bmatrix} 3 \\ 2 \end{bmatrix} \end{aligned}$$

$$\vec{m}_0 = \begin{bmatrix} 1 & 0 \\ 0 & -1 \\ 0 & 0 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} 3 \\ 2 \end{bmatrix} = \begin{bmatrix} 3 \\ 2 \\ 0 \\ 5 \end{bmatrix} = \vec{m}$$

$$\vec{e} = \vec{m} - \vec{m}_0 = \vec{0}$$

