

For  $A^T A$  to be invertible,  $\text{Null}(A^T A)$  must be trivial. We proved last time that  $\text{Null}(A^T A) = \text{Null}(A)$ , which is nontrivial if columns of  $A$  are linearly dependent.

**REMINDER:**  $(AB)^T = B^T A^T$       $A: n \times m$       $A^T: m \times n$      cannot multiply  $A^T B^T$   
 $B: m \times k$       $B^T: k \times m$

ORTHOGONAL MATCHING PURSUIT

In a smart home, how do tell which devices are talking?

Gold Codes:  $\vec{s}_A, \vec{s}_B \quad \pm 1 \quad N = 1023$

$\langle \vec{s}_A, \vec{s}_B \rangle \approx 0$       $\langle \vec{s}_A, \vec{s}_A \rangle = N$

$$\vec{r} = \alpha_1 \vec{s}_1 + \alpha_2 \vec{s}_2 + \dots + \alpha_{10000} \vec{s}_{10000} \quad \begin{bmatrix} \vec{s}_1 & \vec{s}_2 & \dots & \vec{s}_{10000} \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_{10000} \end{bmatrix} = \begin{bmatrix} \vec{r} \\ 1023 \end{bmatrix}$$

$\text{len}(\vec{s}_n) = 1023 \uparrow$

Gaussian Elimination AND Least Squares fail here!

To solve using OMP, be greedy.

$\vec{r} = \vec{s}_A + 2\vec{s}_B \dots$  consider  $\langle \vec{r}, \vec{s}_n \rangle$ . most should be small.  
 $\hookrightarrow$  only expect  $\langle \vec{r}, \vec{s}_A \rangle$  &  $\langle \vec{r}, \vec{s}_B \rangle$  as large

1. Consider all inner products.
2. Find the maximum inner product  $\vec{s}_{\max}$   
 Guess:  $\vec{s}_{\max}$  is transmitting.
3. Project  $\vec{r}$  onto  $\vec{s}_{\max}$ :  $\text{proj}_{\vec{s}_{\max}}(\vec{r}) \rightarrow \vec{r}_{\text{new}} = \vec{r} - \text{proj}_{\vec{s}_{\max}}(\vec{r})$
4. Repeat with  $\vec{r}_{\text{new}}$  as  $\vec{r}$ .

Peel one component at a time!