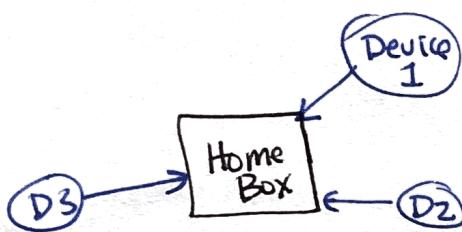


- Last time:
- IOT problem setup
  - OMP algorithm development ← This time!

IOT → Internet of Things.



New: at most  $K$  devices 'ON' at once

→ may not care what shifts are (if not locationing)

→ but want value of song strength from each device → "message"

e.g. Temperature Value  
from smart thermometer

encoded by  $\vec{s} \rightarrow \alpha \vec{s}$   
temp.

Looked @ example situation w  $n$  2,000 devices → each has unique song  $\vec{s}_0, \vec{s}_1, \dots, \vec{s}_{1999}$

device ID  
of length  $m$  400 samples →  $\vec{s}_i \in \mathbb{R}^m$  dim of song

- Want to:
- ① Identify 'on' devices
  - ② Estimate message values

Received signal at homebox is sum of shifted & scaled songs. How to find them? cross-corr., but in particular way.

### Recap of OMP:

Unknowns:  $\vec{x} \in \mathbb{R}^n$

$m < n$

Songs:  $S = \{\vec{s}_0, \vec{s}_1, \dots, \vec{s}_{n-1}\} \in \mathbb{R}^m$

device ID  
all the songs

Messages:  $\alpha_0, \alpha_1, \dots, \alpha_{n-1}$  but a lot of these are zero

Sparsity:  $K \leftarrow$  max. # devices talking @ once

↪ device IDs for 'ON' devices:  $i_1, i_2, \dots, i_K$

↪ Shifts for 'ON' devices:  $N_{i1}, N_{i2}, \dots, N_{iK}$

Measurement:  
(received signal)  
@ Home Box

$\vec{b} \in \mathbb{R}^m$

$$\vec{b} = \sum_i \alpha_i \vec{s}_i^{(N_i)} + \vec{\eta}$$

↑ msg      ↑ shift      ↑ noise

When to use OMP? to recover estimate of a 'sparse' unknown signal (msgs)  
 $\# \text{unknowns} > \# \text{measurements}$

Let's us recover answer from fewer meas. than unknowns!

Why? sparsity is a 'prior' that adds extra information.. sort

What do I need? songs are all approx. orthogonal.

- Algorithm development:
- this is a stretch for 16A (i.e. hard!)
  - ~~but~~ helps understand algorithm to develop it step-by-step.
  - also, shows you how to think to design your own algorithms in future  
 ↳ cause you are leaders, don't want to just use existing algos, but to invent better ones!

Algo VI.

→ cross-corr  $\vec{b}$  w all songs

→ pick out peaks to find 'on' devices

CONS: non-orthog. means small msg. gets drowned by big one! can't find small stuff.

Fix: subtract out one at a time.

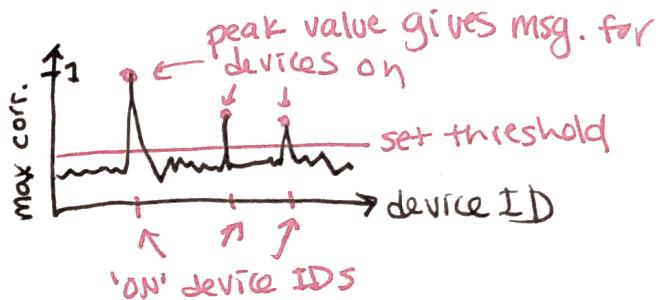
Problem: Noise

Fix: use LS to solve

$$\vec{b} = \begin{bmatrix} \vec{z}_1 & \vec{z}_2 & \dots & \vec{z}_k \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_k \end{bmatrix}$$

meas

set of shifted songs



messages

where  $\vec{z}_k = \vec{s}_{ik}^{(Nik)}$

correctly shifted song

SOLVE FOR ME) by LS:  $\hat{\alpha} = (\vec{z}^T \vec{z})^{-1} \vec{z}^T \vec{b}$

LS solution for msgs

Key Idea: This 'grass' between peaks is not necessarily noise, but rather due to non-orthogonality! (if orthogonal  $\rightarrow$  zero, no problem!)

'peeling' songs off one-by-one

To subtract out loudest song: → pick largest peak/ $\alpha$  value  $\alpha_{ii}$  of msg.

$$\vec{b}' = \vec{b} - \alpha_{ii} \vec{z}_i$$

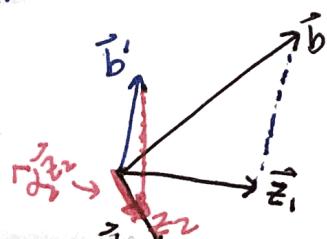
'residue'

what's left of meas. after effect of 1st device is removed

→ then do this over and over until done!

CONS: if  $\vec{z}_1, \vec{z}_2$  (songs) not orthogonal, can't just do one at a time & add components

Fix: need to project onto entire subspace of  $\vec{z}_1, \vec{z}_2$ ! (subtract both together!)



# as us finally to Orthogonal Matching Pursuit (OMP) Algorithm:

→ at each step, keep track of identified 'ON' songs (+shifts)

then remove effect of 'found' devices & repeat...

What if less devices ( $< K$ ) are on?

↳ Stop early by checking if  $\|\vec{y}\|$  less than some small threshold value.

So, Find peaks one-by-one, but remove all together at each step!

OMP Algo pseudo code:

OMP ( $S, \vec{b}, K, th$ ) {  
 songs      meas      sparsity      threshold  
 inputs                     for stopping}

Initialization:  $\vec{y} = \vec{b}$   
 start w/ meas

$F = [\text{empty}]$       index of iteration  
 Found Devices      IDs       $A_0 = [ ]$       Songs found so far  
 j=1      iter #

$\vec{x} = \vec{0}$       init  $\vec{w}$  zeros  
 Device messages (SOLUTION)

Algorithm: while ( $j \leq K$ ) and ( $\|\vec{y}\| > th$ ) {

$[i, N_i] = \text{find-song}(S, \vec{y})$   
 device ID      shift  
 outputs      function      inputs

some fn. that finds loudest song

Newly discovered:  $\vec{s}_i = \vec{s}_i(N_i)$  Song of identified device w/ proper shift. Does it have to be shifted?  
 Yes, to remove properly.

$F = F \cup \{i\}$  add identified device to list of on devices

'Augment' matrix  $A_j = [A_{j-1} \mid \vec{z}_j]$  add new song to set

proxy to  $\vec{x}$   $\vec{a}_j = [A_j^T A_j]^{-1} A_j^T \vec{b}$  projection of meas. onto subspace of all the songs we know on so far...

e.g.  $A = \begin{bmatrix} 1 & 1 \\ \vec{z}_1 & \vec{z}_2 \end{bmatrix}$ , solve  $\vec{b} = A \hat{\vec{\alpha}}$   
 using least squares:

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

at each iter.  
 add a new col  
 from 'found' loudest song

↑ project meas.  
 onto entire 'known'  
 (so far) subspace  
 to est. msg.

$$\vec{b}_j = A_j \vec{a}_j = A_j \left[ A_j^T A_j \right]^{-1} A_j^T \vec{b}$$

part of meas. that  
is NOT EXPLAINED:

$$\vec{y} = \vec{b} - \vec{b}_j$$

When does  $\vec{b}_j = \vec{b}$ ? When I found everything!

so  $\vec{y} \rightarrow 0$  and  $\|\vec{y}\| < \text{th}$

threshold test for  
stopping

$\vec{x}[F] = \vec{a}_j$  update msg. values for 'on' devices  
with their current estimates. (rest stay zero)

$$j = j + 1$$

} update iteration counter.

Am I done? Yes!  $\vec{x}$  is solution! Yay!

But what was 'find-song' function? Finds device ID & shift of loudest song.

find-song( $S, \vec{y}$ ) {

max-correl = zeros (# songs, 1) } initialize

shifts = zeros (# songs, 1) }

for ( $i = 1$ : # songs)  
from 1 to # songs

$\rho = \text{corr}(\vec{y}, \vec{s}_i)$

max-correl[i] = max(abs( $\rho$ ))

shifts[i] = argmax(abs( $\rho$ )) ← find its shift

*pick out max corr (for each song)*

What if msg = -1?

Large negative corr., so just  
use abs to ignore sign!

return  
these as  
outputs

j = argmax(max-correl)

$N_j = \text{shifts}[j]$

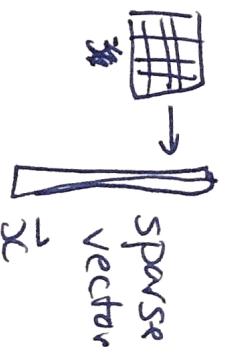
}

0	$[ ]$	$[ ]$	$\vec{0}$	$\vec{b}$
1	$\vec{z}_1$	$A_1 = \begin{bmatrix} \vec{z}_1 \\ \vec{z}_1 \end{bmatrix}$	$\vec{b}_1 = A_1(A_1^T A_1)^{-1} A_1^T \vec{b}$	$\vec{b} - \vec{b}_1$
2	$\vec{z}_2$	$A_2 = \begin{bmatrix} \vec{z}_1 & \vec{z}_2 \\ \vec{z}_1 & \vec{z}_2 \end{bmatrix}$	$\vec{b}_2 = A_2(A_2^T A_2)^{-1} A_2^T \vec{b}$	$\vec{b} - \vec{b}_2$
		$\vdots$	$\vdots$	$\vdots$
j	$\vec{z}_j$	$A_j = \begin{bmatrix} \vec{z}_1 & \vec{z}_2 & \cdots & \vec{z}_j \\ \vec{z}_1 & \vec{z}_2 & \cdots & \vec{z}_j \end{bmatrix}$	$\vec{b}_j = A_j(A_j^T A_j)^{-1} A_j^T \vec{b}$	$\vec{b} - \vec{b}_j$
			$\vec{b} - \vec{b}_j$	$\vec{b}$
			$\cdots$	$\vec{b}$

Recall: Imaging Lab

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

Now say  $\vec{x}$  is sparse image



What can OMP do for you here?

We can take fewer measurements!

But how to make orthogonal? does it work?