

## Plan

Logistics

Review

Course Response Forms

Sparse Recovery

Games @ 6pm Wednesday

↳ Pizza

↳ Extra Credit

(name and # points)

## PSet 3

Nice Job!

Chat !!

## PSet 4

I'm sorry

Today: evaluation

## Project

↳ work hard tomorrow

↳ play hard tomorrow night

↳ presentation (recorded links)

↳ Gradescope

## Review

$A \in \mathbb{R}^{n \times d}$ ,  $b \in \mathbb{R}^d$

$$\begin{aligned}x^* &= \underset{x}{\operatorname{argmin}} \|Ax - b\|_2^2 \\&= (A^T A)^{-1} A^T b\end{aligned}$$

$O(nd^2)$  time

$\Pi \in \mathbb{R}^{m \times n}$   $m \approx d$

$$\begin{aligned}\tilde{x} &= \underset{x}{\operatorname{argmin}} \|\Pi A x - \Pi b\|_2^2 \\&= (A^T \Pi^T \Pi A)^{-1} A^T \Pi^T \Pi b\end{aligned}$$

$O(md^2)$

$$\|A\tilde{x} - b\|_2^2 \leq (1 + \epsilon) \|Ax^* - b\|_2^2$$

Compute  $\Pi A$ ,  $\Pi b$  fast

$$A = \begin{bmatrix} a_1 & a_2 & \dots & a_d \end{bmatrix}$$

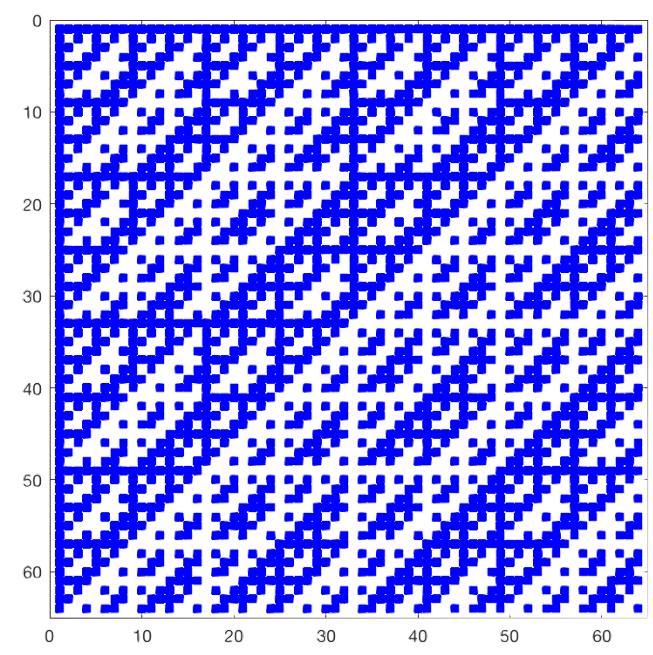
$$\Pi a_i = S H D a_i$$

$m \times n$

$$\begin{bmatrix} +1 & & & \\ & -1 & & \\ & & +1 & \\ & & & \ddots \end{bmatrix} \begin{bmatrix} 1 \\ \vdots \\ a \end{bmatrix} = \begin{bmatrix} +1 \\ -1 \\ +1 \\ +1 \end{bmatrix} * \begin{bmatrix} 1 \\ \vdots \\ a \end{bmatrix}$$

elementwise

$H$



$HD$



## Course Response Forms

What did you like? What can be improved?

↳ Content and difficulty

↳ Daily check in forms

↳ Group activities

↳ Review the next day

↳ Accessibility/receptiveness

↳ Afternoon problem solving

↳ LaTeX, self-grade

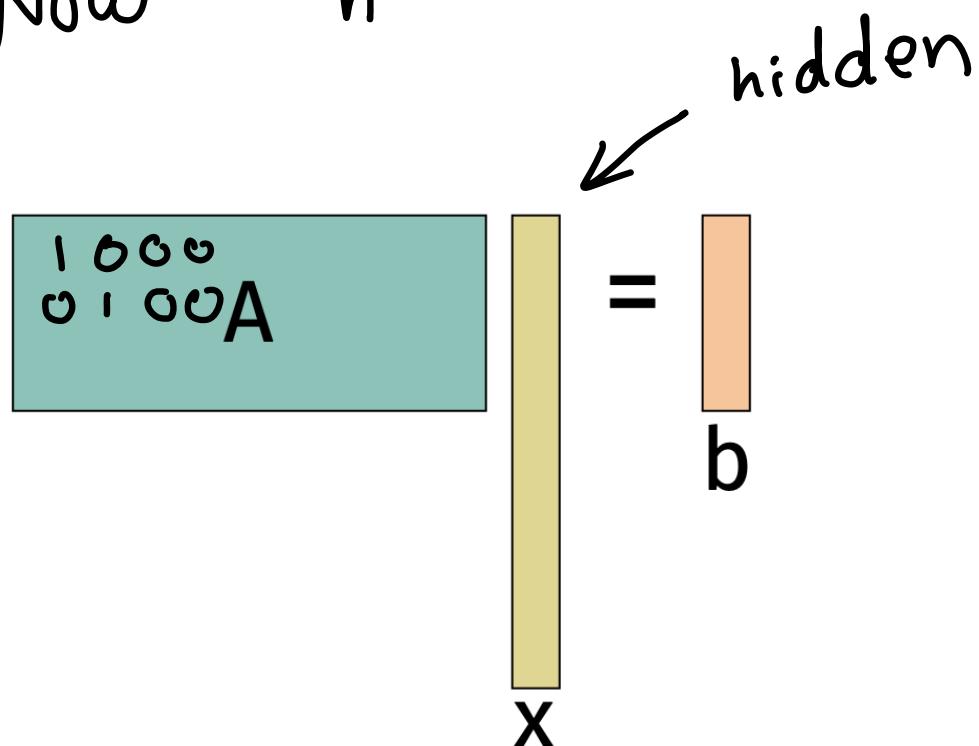
↳ Typed notes, slides

## Sparse Recovery

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

$$Ax = b$$

Now  $n \ll d$



Goal: Recover  $x$  by choosing  $A$  and observing  $Ax = b$

### Trivial solution

$n = d$  then I can recover  $x$  exactly

$A \in \mathbb{R}^{d \times d}$  is too large

### Assume $x$ is $k$ -sparse

$$\|x\|_0 = \# \text{ non-zeros in } x \leq k$$

Goal: Recover  $k$ -sparse  $x$

with only a few "measurements"

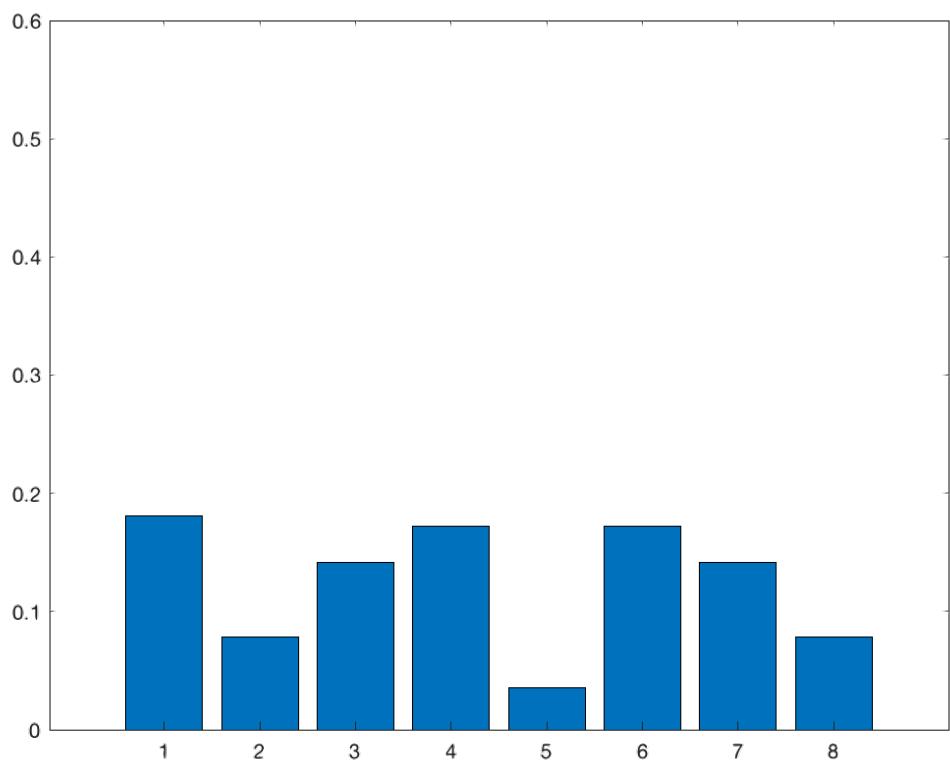
$n = \mathcal{O}(k \log k)$  measurements

## Applications

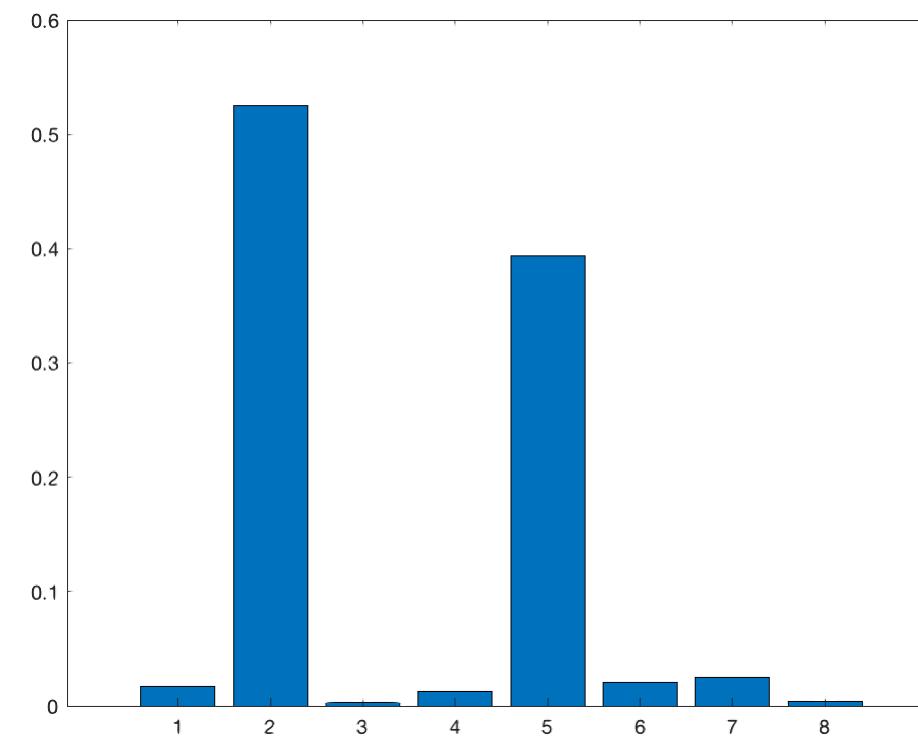
- Compress images because Fourier transform is sparse
  - Parameters that fit "Occam's razor"
  - X-rays, MRI
  - Earth exploration

# Fourier Transform

Data



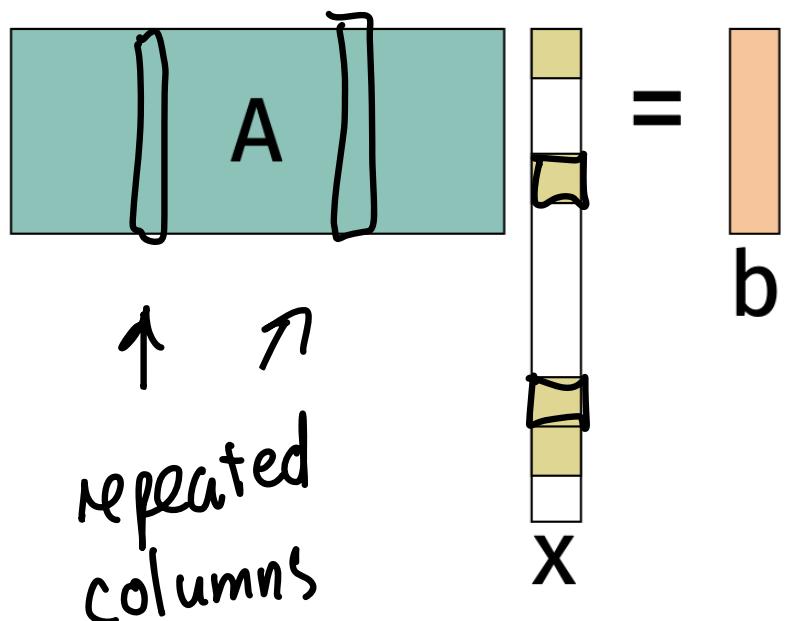
Frequency



Uncertainty Principle

Which A work?

Which definitely do not?



( $k, \epsilon$ ) - Restricted Isometry Property  
A satisfies  $(k, \epsilon)$  - RIP if  
for all  $x$  with  $\|x\|_0 \leq k$ ,

$$(1-\epsilon) \|x\|_2^2 \leq \|Ax\|_2^2 \leq (1+\epsilon) \|x\|_2^2$$

↳ Looks like JL Lemma

↳ Preserves norm of  $k$ -sparse  
(rather than discrete set  
or subspace)

$$A \in \mathbb{R}^{n \times d}$$

$$n = O\left(\frac{k \log d + \log(1/\delta)}{\epsilon^2}\right)$$

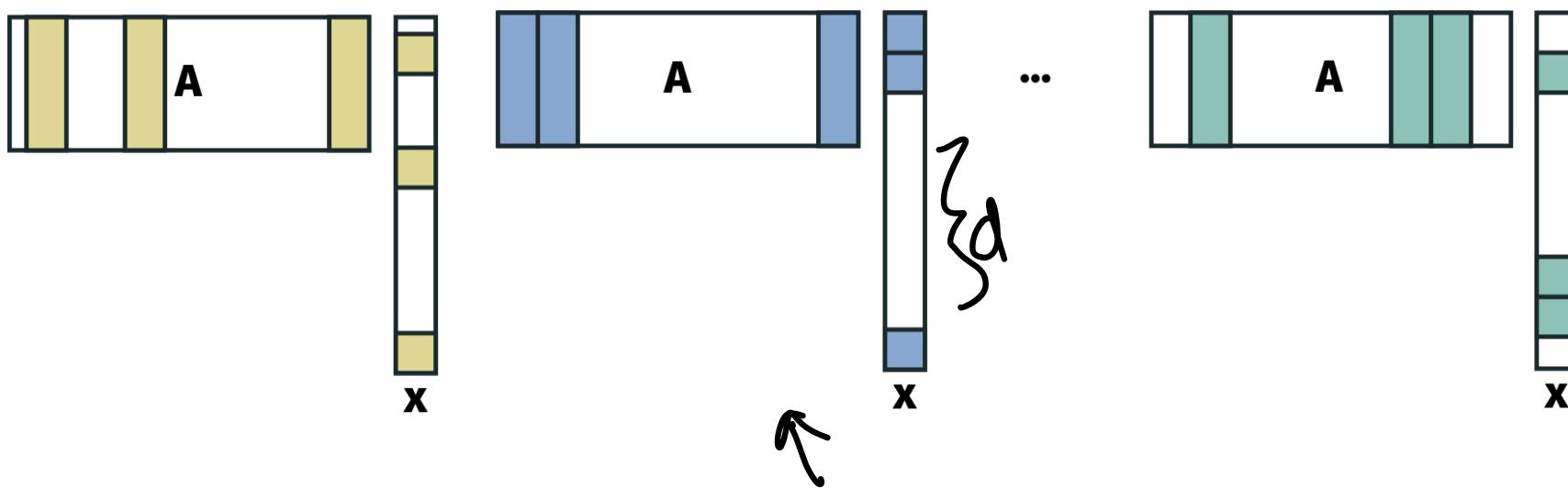
$$T = \binom{d}{k}$$

$$\delta' = \frac{\delta}{T}$$

Consider  $S_k = \{x : \|x\|_0 \leq k\}$

$$\log(1/\delta')$$

$$S_k = U_1 \cup U_2 \cup U_3 \cup \dots \cup U_T$$



Theorem : If  $A$  is  $(2\kappa, \epsilon)$ -RIP for  $\epsilon < 1$

then  $x$  is the unique minimizer

$$\left. \begin{aligned} \|y - x\|_2^2 &\leq 0 \\ &= \sum_{i=1}^d (y_i - x_i)^2 \\ \Rightarrow x &= y \end{aligned} \right\}$$

$$\min_z \|z\|_0 \text{ s.t. } Az = b$$

$\nwarrow O(d^k)$  time to find

Proof: Suppose for

contradiction that

there is a different  $y$  with

$$\|y\|_0 \leq \|x\|_0 \leq \kappa \quad \text{and} \quad Ay = b$$

$$y \neq x \quad Ay = b \quad Ax = b$$

$$\begin{aligned} Ay - Ax &= A(y - x) \\ &= b - b = 0 \end{aligned}$$

$$\|w\|_2^2(1-\epsilon) \leq \|Aw\|_2^2$$

$$\begin{aligned} \|y - x\|_2^2(1-\epsilon) &\leq \|A(y - x)\|_2^2 \\ &= 0 \end{aligned}$$

Theorem : If  $A$  is  $(3\kappa, \epsilon)$ -RIP for  $\epsilon < 1$

then  $x$  is the unique minimizer

$$\min_z \|z\|_1 \text{ s.t. } Az = b$$

↑ convex optimization solve with linear programming  
 $O(d^{3.5})$

\* Exponentially faster

\* Like relaxation

