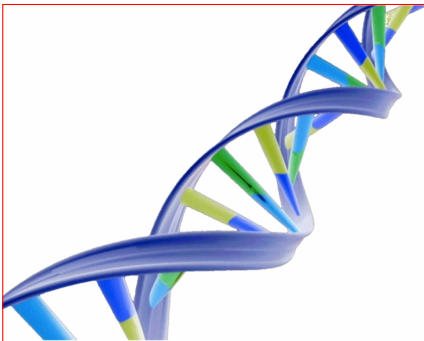


Holy Grail of Communication: *Error Correction Coding*

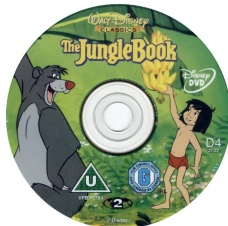
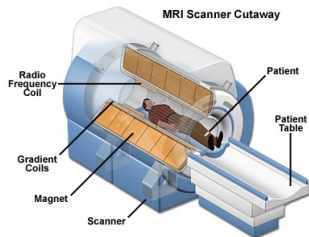


Sibi Raj B Pillai
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Subject:EE113-RollNo

Outline

- Three Systems: Communication, Storage, MRI
- Some Common Ground
- Linear Solvers
- Theory to Practice
- Conclusion

Mission Critical



Courtesy: Google Images

Pick the ODD one out



10×10 gms



10×11 gms



10×10 gms



10×10 gms

Pick the ODD one out



10×10 gms



10×11 gms

...

10×10 gms



10×10 gms



Pick the ODD one out



10×10 gms



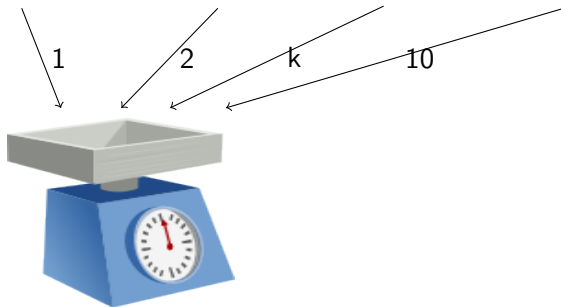
10×11 gms

...



10×10 gms

10×10 gms



Pick the ODD one out



10×10 gms



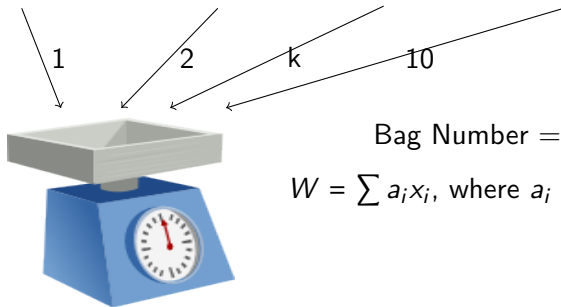
10×11 gms

...



10×10 gms

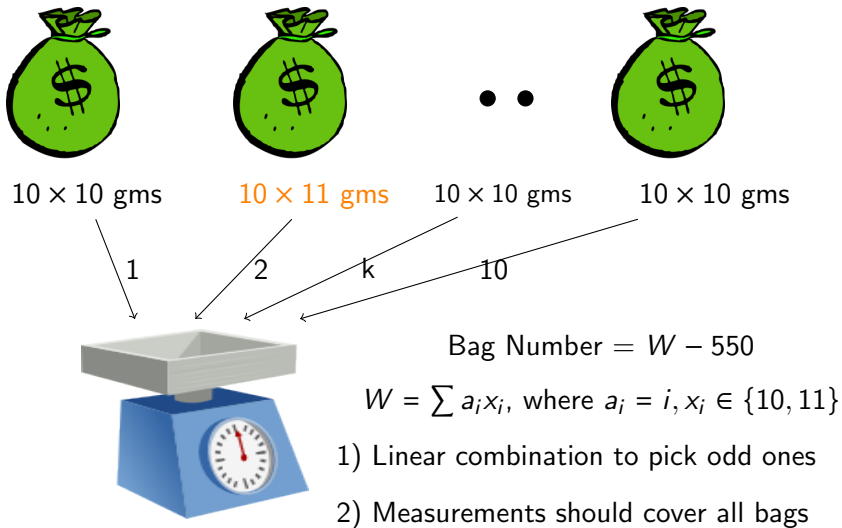
10×10 gms



Bag Number = $W - 550$

$W = \sum a_i x_i$, where $a_i = i$, $x_i \in \{10, 11\}$

Pick the ODD one out



Solving Equations

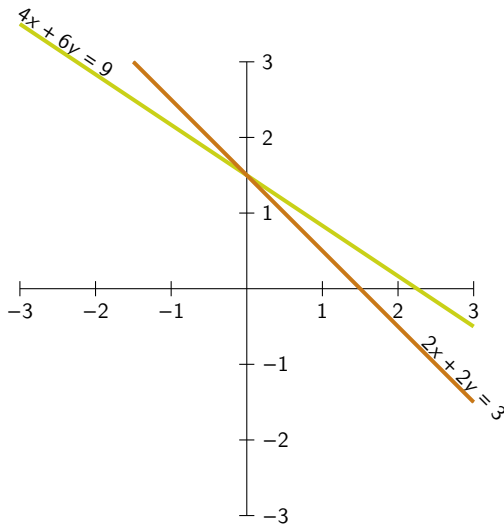
$$4x_1 + 6x_2 = 9$$

$$2x_1 + 2x_2 = 3$$

$$\begin{bmatrix} 9 \\ 3 \end{bmatrix} = \begin{bmatrix} 4 & 6 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$\begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

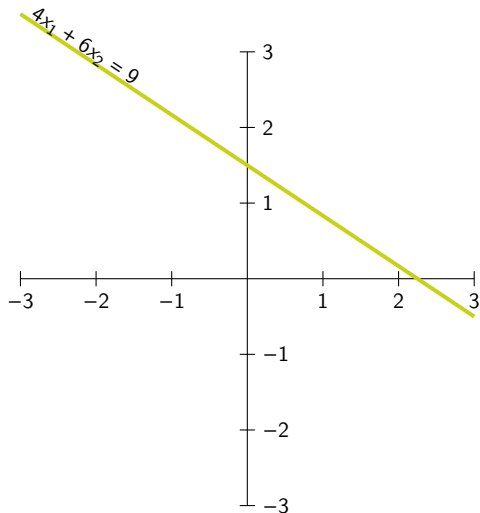
$$\bar{x} = A^{-1} \bar{c}$$



More Unknowns

$$4x_1 + 6x_2 = 9$$

- ▶ Many Solutions in general!
- ▶ Which one(s) do we need?



Linear Solvers

- ▶ We have M linear equations and N unknowns.

$$y_j = \sum_i a_{ji} x_i \quad , \quad 0 \leq i \leq M-1.$$

$$\begin{matrix} \bar{y} \\ M \times 1 \end{matrix} = \begin{matrix} A \\ M \times N \end{matrix} \begin{matrix} \bar{x} \\ N \times 1 \end{matrix}$$

- ▶ An **under-determined** set of equations, $M < N$.
- ▶ However, assume x to be **sparse** (a few *odd* values).
- ▶ **Sparsity** s represents the number of non-zero entries of \bar{x} .

System and Objectives

$$\begin{bmatrix} y_1 \\ y_2 \\ \cdot \\ \cdot \\ y_M \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & \cdot & \cdot & a_{1N} \\ a_{21} & a_{22} & \cdot & \cdot & a_{2N} \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ a_{M1} & a_{M2} & \cdot & \cdot & a_{MN} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \cdot \\ \cdot \\ x_N \end{bmatrix}$$

$$y_i = \langle \vec{a}_i, \vec{x} \rangle$$

$$\bar{y} = \sum_j x_j \bar{a}_j$$

System and Objectives

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- Our aim is to find the sparse signal(s) \hat{x} satisfying the above.

System and Objectives

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$$\bar{y} = \sum_j x_j \bar{a}_j$$

- ▶ Our aim is to find the sparse signal(s) \hat{x} satisfying the above.
- ▶ Need to design **matrix** A , as well as a **recovery algorithm**.

Design Example 1

- $N = 8, M = 1, s = 1, x_i \in \{0, 1\} :$

$$\begin{bmatrix} a_1 & a_2 & a_3 & a_4 & a_5 & a_6 & a_7 & a_8 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \\ x_8 \end{bmatrix} = y_1$$

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- Solution:

$$A = \begin{bmatrix} 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 \end{bmatrix}$$

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If $y_1 = j$, declare $x_j = 1$ and all others are zero.

Design Example 2

- $N = 8, M = 2, s = 1, x_i \in \{0, 1, 2 \dots\}$:

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} & a_{15} & a_{16} & a_{17} & a_{18} \\ a_{21} & a_{22} & a_{23} & a_{24} & a_{25} & a_{26} & a_{27} & a_{28} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \\ x_8 \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}$$

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- Solution:

$$A = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 \end{bmatrix}$$

Set $j = \frac{y_2}{y_1}$ and declare $x_j = y_1$; all others are zero.

Design Example 3

- $N = 8, M = 4, s = 2, x_i \in \{0, 1, \dots, 9\}$:

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} & a_{15} & a_{16} & a_{17} & a_{18} \\ a_{21} & a_{22} & a_{23} & a_{24} & a_{25} & a_{26} & a_{27} & a_{28} \\ a_{31} & a_{32} & a_{33} & a_{34} & a_{35} & a_{36} & a_{37} & a_{38} \\ a_{41} & a_{42} & a_{43} & a_{44} & a_{45} & a_{46} & a_{47} & a_{48} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \\ x_8 \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix}$$

Design Example 3

- $N = 8, M = 4, s = 2, x_i \in \{0, 1, \dots, 9\}$:

$$\begin{bmatrix} 1 & 2^0 & 3^0 & 4^0 & 5^0 & 6^0 & 7^0 & 8^0 \\ 1 & 2^1 & 3^1 & 4^1 & 5^1 & 6^1 & 7^1 & 8^1 \\ 1 & 2^2 & 3^2 & 4^2 & 5^2 & 6^2 & 7^2 & 8^2 \\ 1 & 2^3 & 3^3 & 4^3 & 5^3 & 6^3 & 7^3 & 8^3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \\ x_8 \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix}$$

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- Challenge:

Design an integer-valued A with $0 \leq a_{ij} \leq 10$ for this problem.

Quick Recap

- ▶ A set of equations

$$\underset{M \times 1}{\bar{y}} = \underset{M \times N}{A} \underset{N \times 1}{\bar{x}}$$

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- ▶ We wish to get back $x \in \mathbb{R}^N$ from this M measurements.
- ▶ Under-determined in general, little hope of recovering x .
- ▶ However, we wish to recover a sparse input x .
- ▶ **Goal:** Design the matrix A and a recovery strategy.

CD Writing



Discrete Fourier Transform

Recall our **Vander Monde** matrix (with $N' = N - 1$)

$$F = \begin{bmatrix} \alpha_0^0 & \alpha_1^0 & \alpha_2^0 & \cdots & \alpha_{N'}^0 \\ \alpha_0^1 & \alpha_1^1 & \alpha_2^1 & \cdots & \alpha_{N'}^1 \\ \alpha_0^2 & \alpha_1^2 & \alpha_2^2 & \cdots & \alpha_{N'}^2 \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ \alpha_0^{N'} & \alpha_1^{N'} & \alpha_2^{N'} & \cdots & \alpha_{N'}^{N'} \end{bmatrix} \quad (1)$$

Erasure Coding

x_{11}	x_{12}	x_{13}	x_{14}	x_{15}	...	x_{1n}
x_{21}	x_{22}	x_{23}	x_{24}	x_{25}	...	x_{2n}
x_{31}	x_{32}	x_{33}	x_{34}	x_{35}	...	x_{3n}
x_{41}	x_{42}	x_{43}	x_{44}	x_{45}	...	x_{4n}
x_{51}	x_{52}	x_{53}	x_{54}	x_{55}	...	x_{5n}
x_{61}	x_{62}	x_{63}	x_{64}	x_{65}	...	x_{6n}
\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots
x_{L1}	x_{L2}	x_{L3}	x_{L4}	x_{L5}	...	x_{Ln}

- ▶ Data symbols (**byte**)

$$\vec{d} = [d_1, \dots, d_k]^T$$

- ▶ Stored symbols (**int**).

$$\vec{x} = [x_1, \dots, x_n]^T$$

- ▶ Read values (**int/?**).

$$\vec{y} = [y_1, \dots, y_n]^T$$

- ▶ Upto m erasures/row.

Erasure Coding

x_{11}	x_{12}	x_{13}	[?]	x_{15}	...	x_{1n}
x_{21}	x_{22}	x_{23}	x_{24}	x_{25}	...	x_{2n}
[?]	[?]	x_{33}	x_{34}	[?]	...	x_{3n}
x_{41}	x_{42}	x_{43}	x_{44}	[?]	...	x_{4n}
x_{51}	[?]	x_{53}	x_{54}	x_{55}	...	[?]
x_{61}	x_{62}	x_{63}	[?]	x_{65}	...	[?]
\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots
x_{L1}	x_{L2}	x_{L3}	[?]	x_{L5}	...	x_{Ln}

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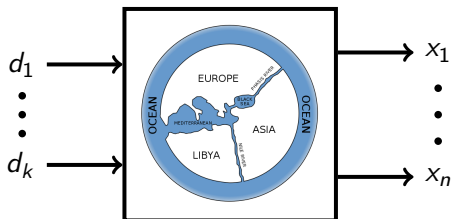
$$\vec{x} = [x_1, \dots, x_n]^T$$

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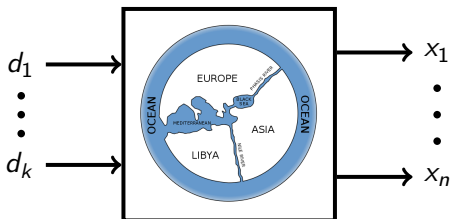
$$\vec{y} = [y_1, \dots, y_n]^T$$

- ▶ Upto m erasures/row.

Encoding Strategy



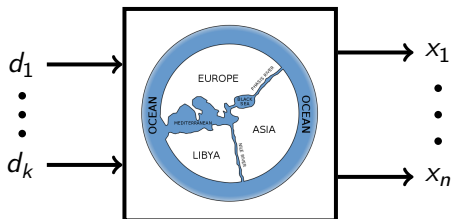
Encoding Strategy



Using our *Vandermonde* matrix F .

$$\bar{x} = F\bar{d}, \text{ where } \bar{d}^T = [d_1, \dots, d_k, 0, \dots, 0].$$

Encoding Strategy

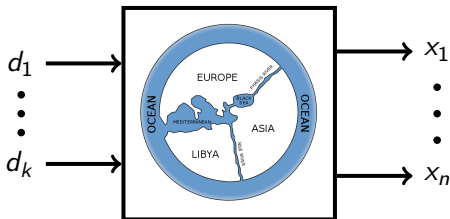


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Decoding possible if $k \leq n - m$.

Encoding Strategy



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Efficiency:

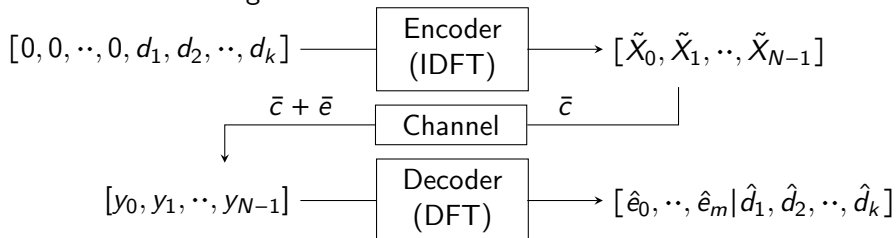
$$\frac{k}{n} = 1 - \frac{m}{n}$$

Correcting Errors

► Communication Channel



► Channel Coding



► If there are no errors, $\hat{e}_i = 0$ and $\hat{d}_i = d_i, \forall i$.

How it works

- ▶ For s -errors, consider the first $m = 2s$ rows of Fourier matrix.

$$\begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & \alpha_1^1 & \alpha_2^1 & \alpha_3^1 & \alpha_4^1 & \alpha_5^1 & \alpha_6^1 & \alpha_7^1 \\ 1 & \alpha_1^2 & \alpha_2^2 & \alpha_3^2 & \alpha_4^2 & \alpha_5^2 & \alpha_6^2 & \alpha_7^2 \\ 1 & \alpha_1^3 & \alpha_2^3 & \alpha_3^3 & \alpha_4^3 & \alpha_5^3 & \alpha_6^3 & \alpha_7^3 \end{bmatrix} \begin{bmatrix} e_1 \\ e_2 \\ \vdots \\ e_8 \end{bmatrix} = \begin{bmatrix} X_1 \\ X_2 \\ \vdots \\ X_m \end{bmatrix}$$

- ▶ Pick **any** $2s$ columns from this restricted matrix.

$$B = \begin{bmatrix} 1 & 1 & 1 & 1 \\ \alpha_k^1 & \alpha_l^1 & \alpha_m^1 & \alpha_n^1 \\ \alpha_k^2 & \alpha_l^2 & \alpha_m^2 & \alpha_n^2 \\ \alpha_k^3 & \alpha_l^3 & \alpha_m^3 & \alpha_n^3 \end{bmatrix}$$

- ▶ $\det(B) \neq 0$ if $\alpha_i \neq \alpha_j \Rightarrow$ cols. linearly independent.

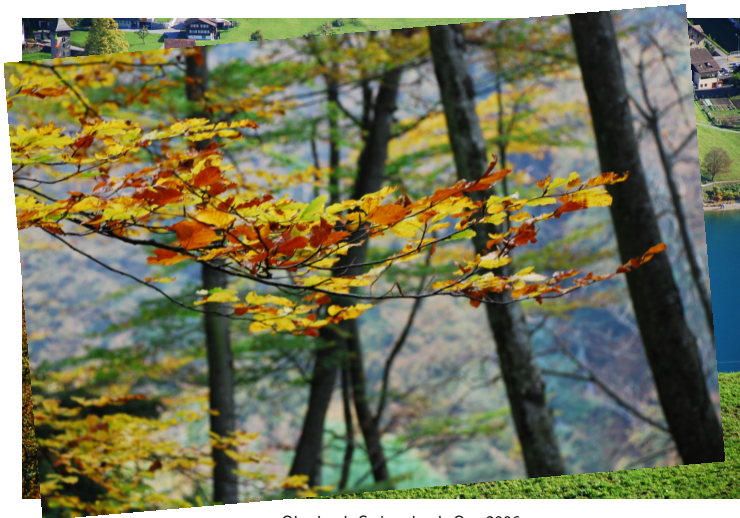
Why so many Pixels

Why so many Pixels



Lungern, Switzerland, Oct 2006

Why so many Pixels



Oberland, Switzerland, Oct 2006

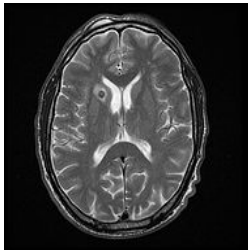
Why so many Pixels



EPFL, Switzerland, May 2007

Magnetic Resonance Imaging

MRI image, source:wikipedia



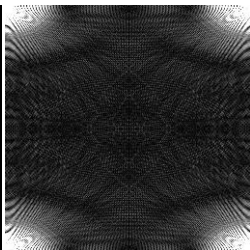
Magnetic Resonance Imaging

Shepp-Logan Phantom 256×256



Magnetic Resonance Imaging

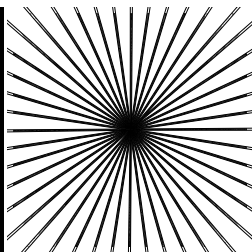
Shepp-Logan Phantom 256×256



Frequency grid

Magnetic Resonance Imaging

Shepp-Logan Phantom 256×256



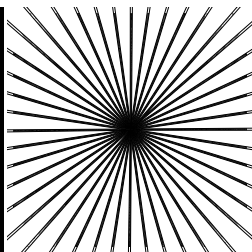
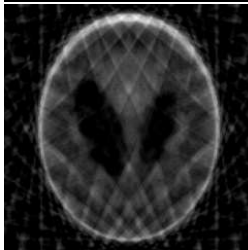
Frequency grid

Magnetic Resonance Imaging

Shepp-Logan Phantom 256×256



L_2 minimization



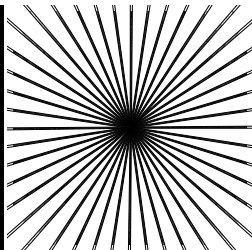
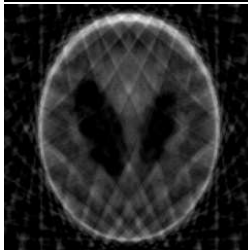
Frequency grid

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L_2 minimization



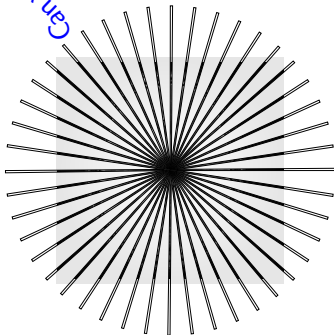
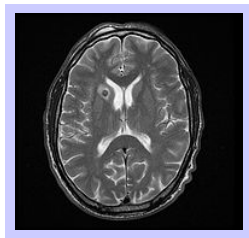
Frequency grid

CS reconstruction



CS Magic?

Can we get back the original



- ▶ Surely you are joking Mr. Xxxman.
- ▶ Linear equations seem to do wonders here.

Conclusion

- ▶ We discussed three problems
 1. Space communication.
 2. CD Information storage and retrieval.
 3. MRI Imaging
- ▶ Somewhat cute that linear solvers are the key to all three.
- ▶ Comments and queries to bsraj@ee.iitb.ac.in