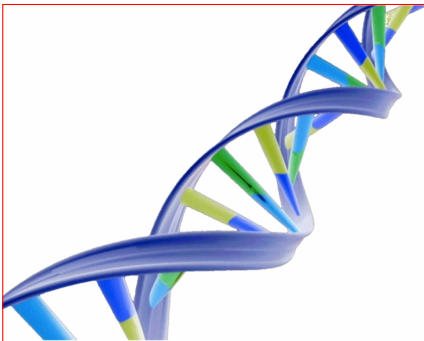


Holy Grail of Communication: *Error Correction Coding*

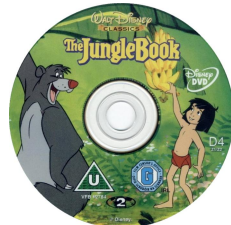
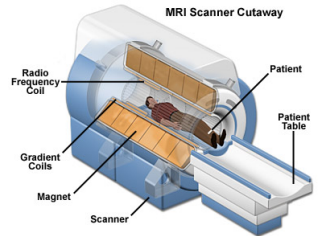


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Outline

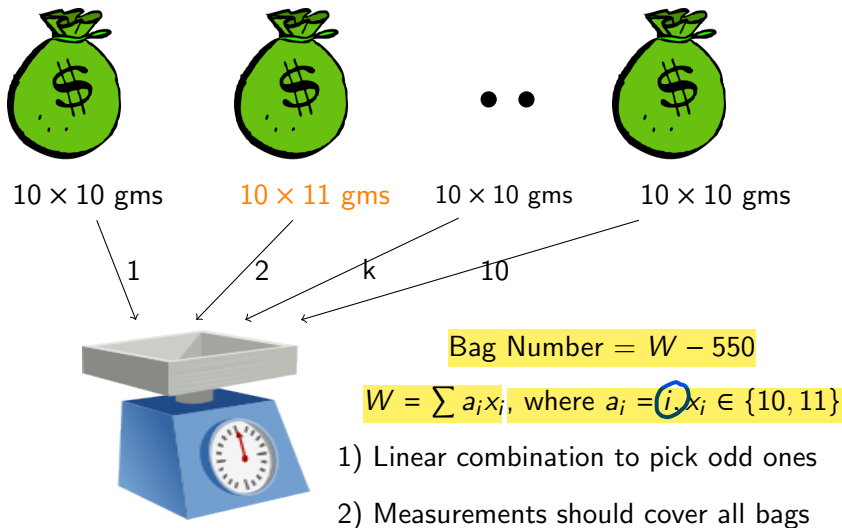
- Three Systems: Communication, Storage, MRI
- Some Common Ground
- Linear Solvers
- Theory to Practice
- Conclusion

Mission Critical



Courtesy: Google Images

Pick the ODD one out



Solving Equations

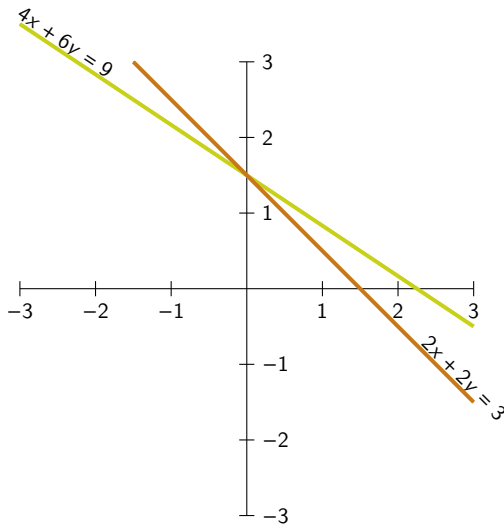
$$4x_1 + 6x_2 = 9$$

$$2x_1 + 2x_2 = 3$$

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

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

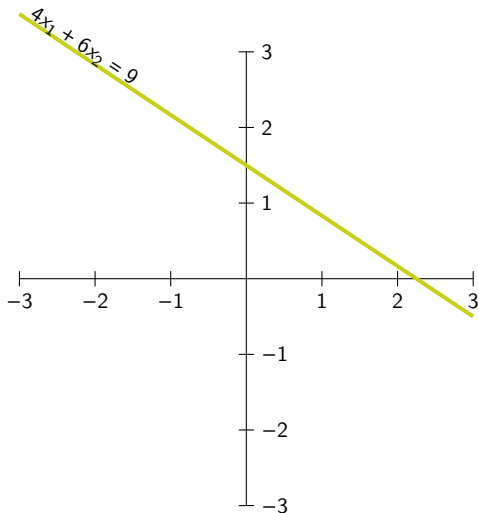
$$\bar{x} = A^{-1} \bar{c}$$



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.

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

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

- ▶ 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

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

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

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

- ▶ 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 We have to find which variable is non-zero and so accordingly, we assign the values to the coefficients.

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

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


► Solution:

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

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

Design Example 2

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



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

- 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.

$1 \cdot x_j = y_1$
 $j \cdot x_j = y_2$
 $j = \frac{y_2}{x_j} = \frac{y_2}{y_1}$
get value of x_j
get j

Design Example 3

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

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

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

Quick Recap

- ▶ A set of equations

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

- ▶ 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 (DFT)

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_{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}

- Data symbols (**byte**)

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

- Stored symbols (**int**).

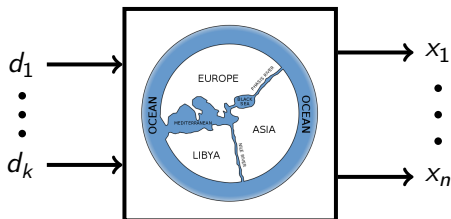
$$\bar{x} = [x_1, \dots, x_n]^T$$

- Read values (**int/?**).

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

- Upto m erasures/row.

Encoding Strategy



Using our *Vandermonde* matrix F .

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

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

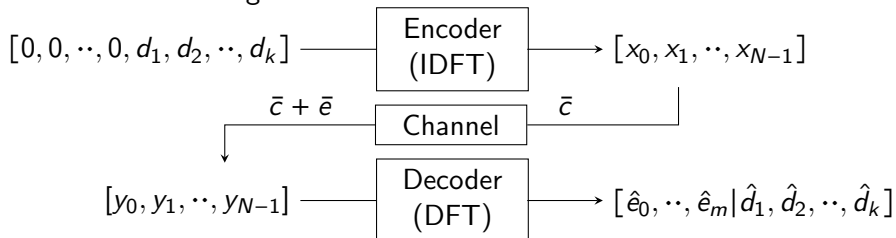
Efficiency: $\frac{k}{n} = 1 - \frac{m}{n}$

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



$m=1$

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

$$\begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & \alpha_1^1 & \alpha_2^1 & \alpha_3^1 & \alpha_4^1 & \alpha_5^1 & \alpha_6^1 & \alpha_7^1 \\ 1 & \alpha_1^2 & \alpha_2^2 & \alpha_3^2 & \alpha_4^2 & \alpha_5^2 & \alpha_6^2 & \alpha_7^2 \\ 1 & \alpha_1^3 & \alpha_2^3 & \alpha_3^3 & \alpha_4^3 & \alpha_5^3 & \alpha_6^3 & \alpha_7^3 \end{bmatrix} \begin{bmatrix} e_0 \\ e_2 \\ \vdots \\ e_7 \end{bmatrix} = \begin{bmatrix} \hat{e}_0 \\ \hat{e}_1 \\ \vdots \\ \hat{e}_3 \end{bmatrix}$$

- 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

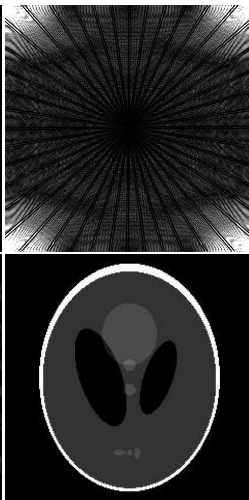
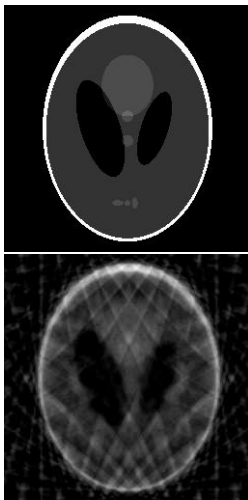


Basel, Switzerland, Oct 2006

Magnetic Resonance Imaging

Shi-Min Li, George Papanicolaou, 2006

L_2 minimization



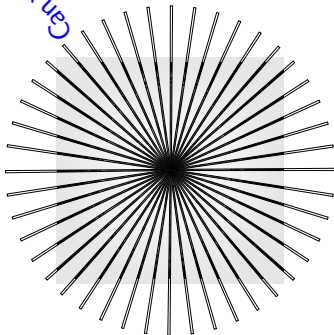
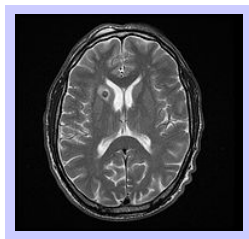
Frequency grid

CS reconstruction

Candes Romberg Tao' 2006

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