

# SINGLE PIXEL CAMERA

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# OUTLINE

- ✻ Modern Day Mega Pixel Cameras
- ✻ Riddle to Compact Storage
- ✻ Fundamental Principles
- ✻ Single-Pixel Camera
- ✻ Conclusions

# MODERN CAMERAS

☼ 5 MP is common

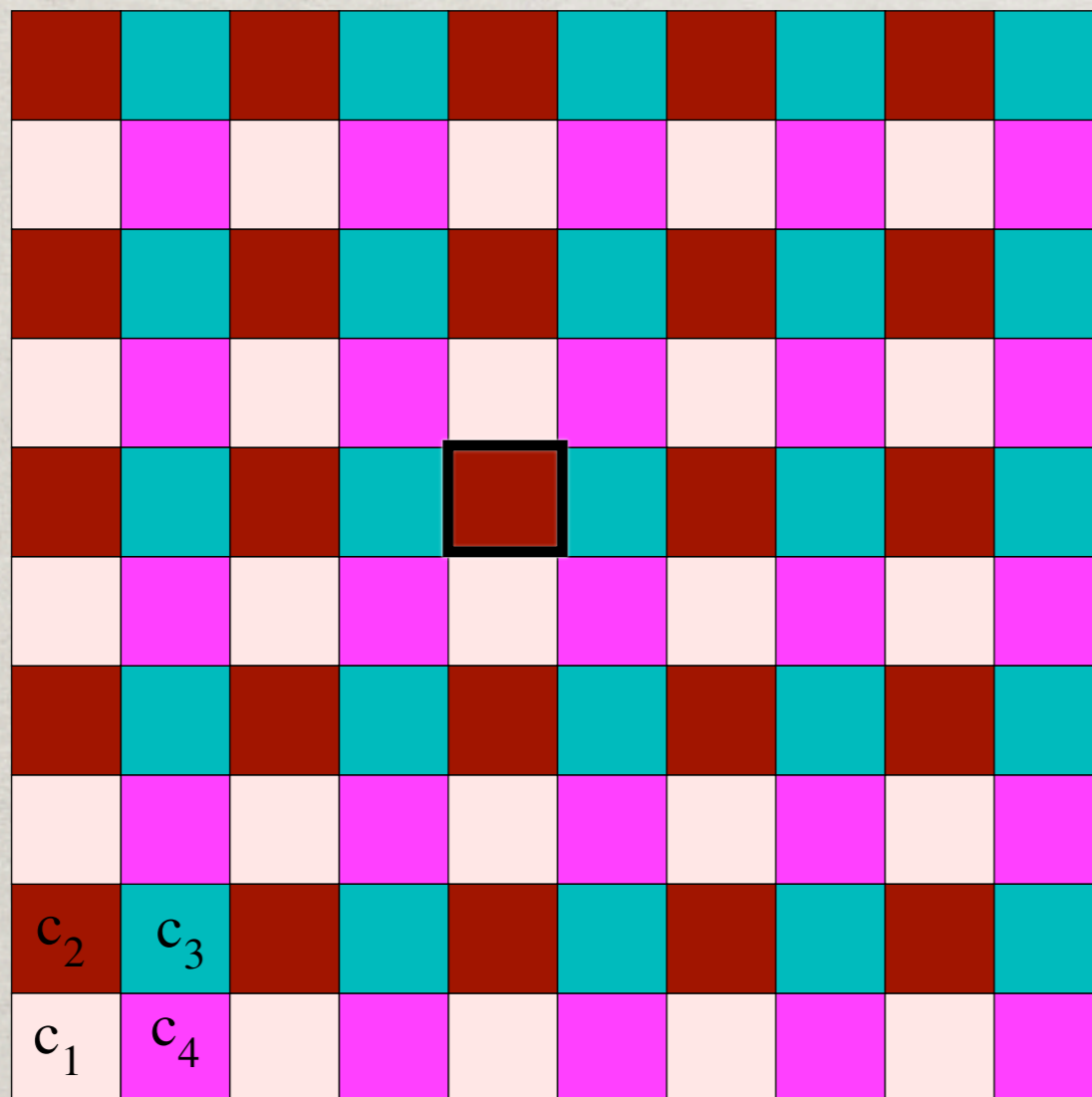


☼ 10-15 MP is also popular



# WHAT IS PIXEL ?

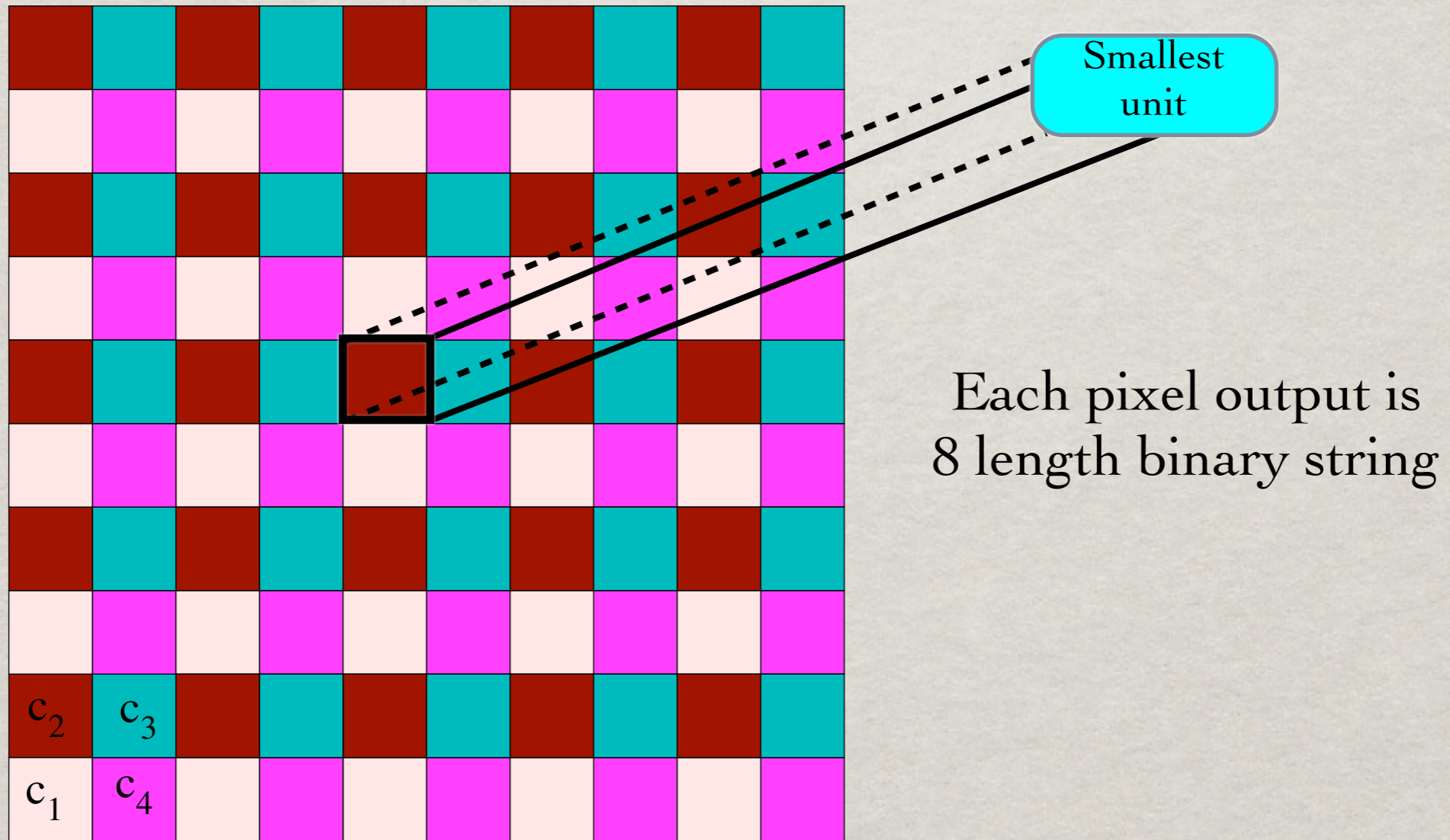
- ☼ Smallest visual-sensor resolution





# WHAT IS PIXEL ?

☼ Smallest visual-sensor resolution



# TOTAL STORAGE

- ✿ Each Pixel - roughly 8 bits
- ✿ 5MPixel - 5 MB of data per picture
- ✿ 1GB can store only 200 images
- ✿ But you can store many more
- ✿ Each picture - 200-440 KB
- ✿ Where is the **MAGIC** ?

# NATURAL IMAGES



# NATURAL IMAGES



# NATURAL IMAGES



- ✻ There is lot of local image correlation(similarity)
- ✻ Some simpler representation

# HISTORY OF THIS IMAGE



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**Alexander Sawchuk** in summer of 73, then an assistant professor of electrical engineering at the [University of Southern California](#) was hurriedly searching the lab for a good image to scan for a colleague's conference paper. They got tired of their stock of usual test images, dull stuff dating back to television standards work in the early 1960s. They wanted something glossy to ensure good output dynamic range, and they wanted a human face. Just then, somebody happened to walk in with a recent issue of *Playboy*.

**David C. Munson, editor-in-chief, January 1996 *IEEE Transactions on Image Processing***  
First, the image contains a nice mixture of detail, flat regions, shading, and texture that do a good job of testing various image processing algorithms. It is a good test image! Second, the Lena image is a picture of an attractive woman.

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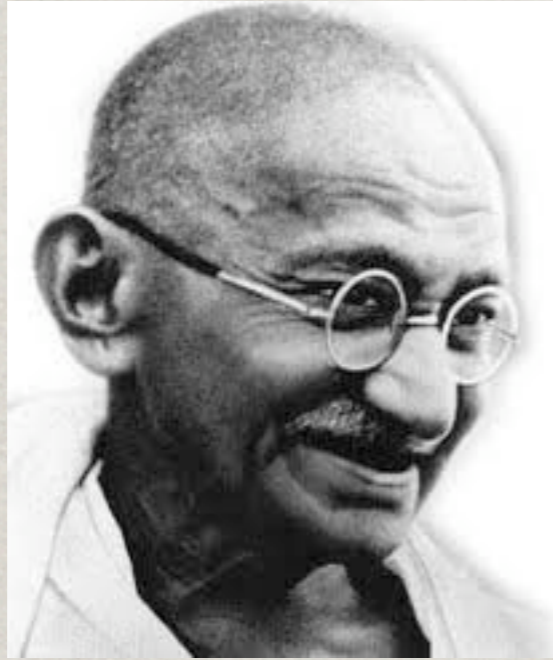
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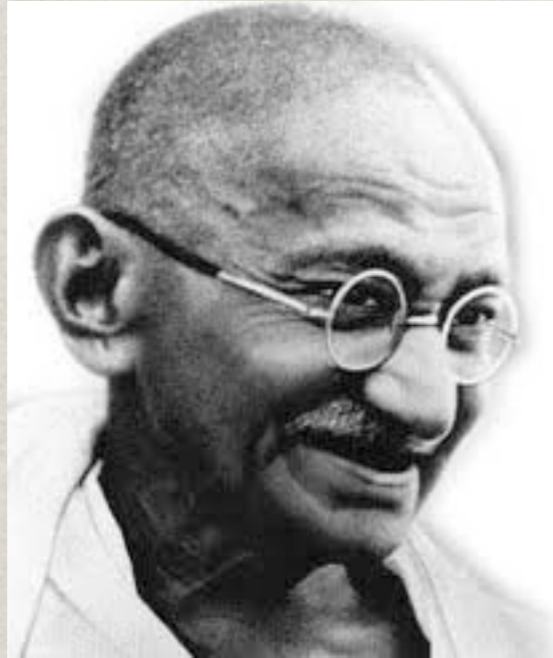
**Scientists do have an eye for beauty!**

# ANALOGY

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# ANALOGY



☀ Succinct Representation

☀ Reversible ?



# NATURAL IMAGES

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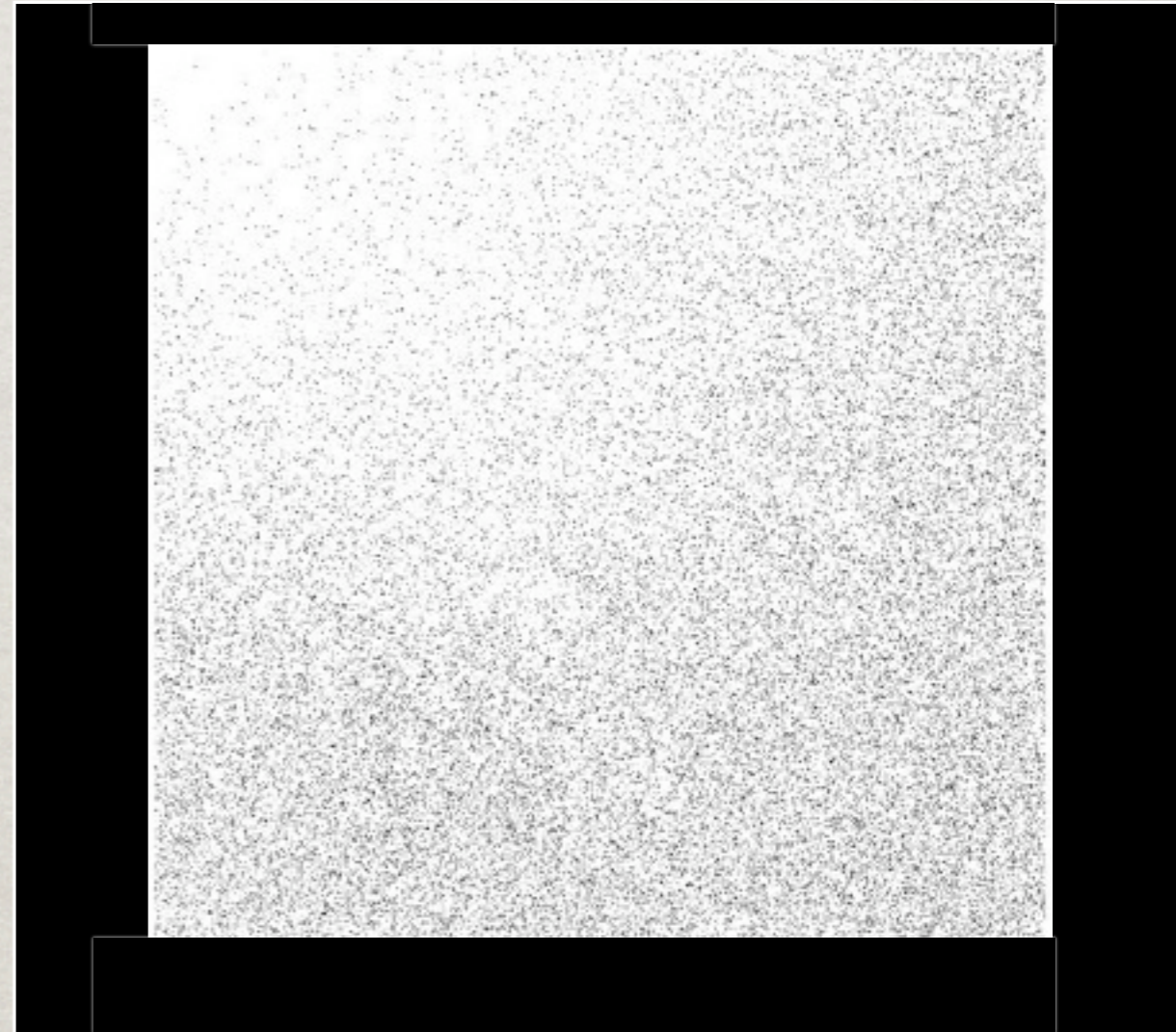


- ✿ There is lot of local image correlation (similarity)
- ✿ Some simpler representation
- ✿ **Reversible**

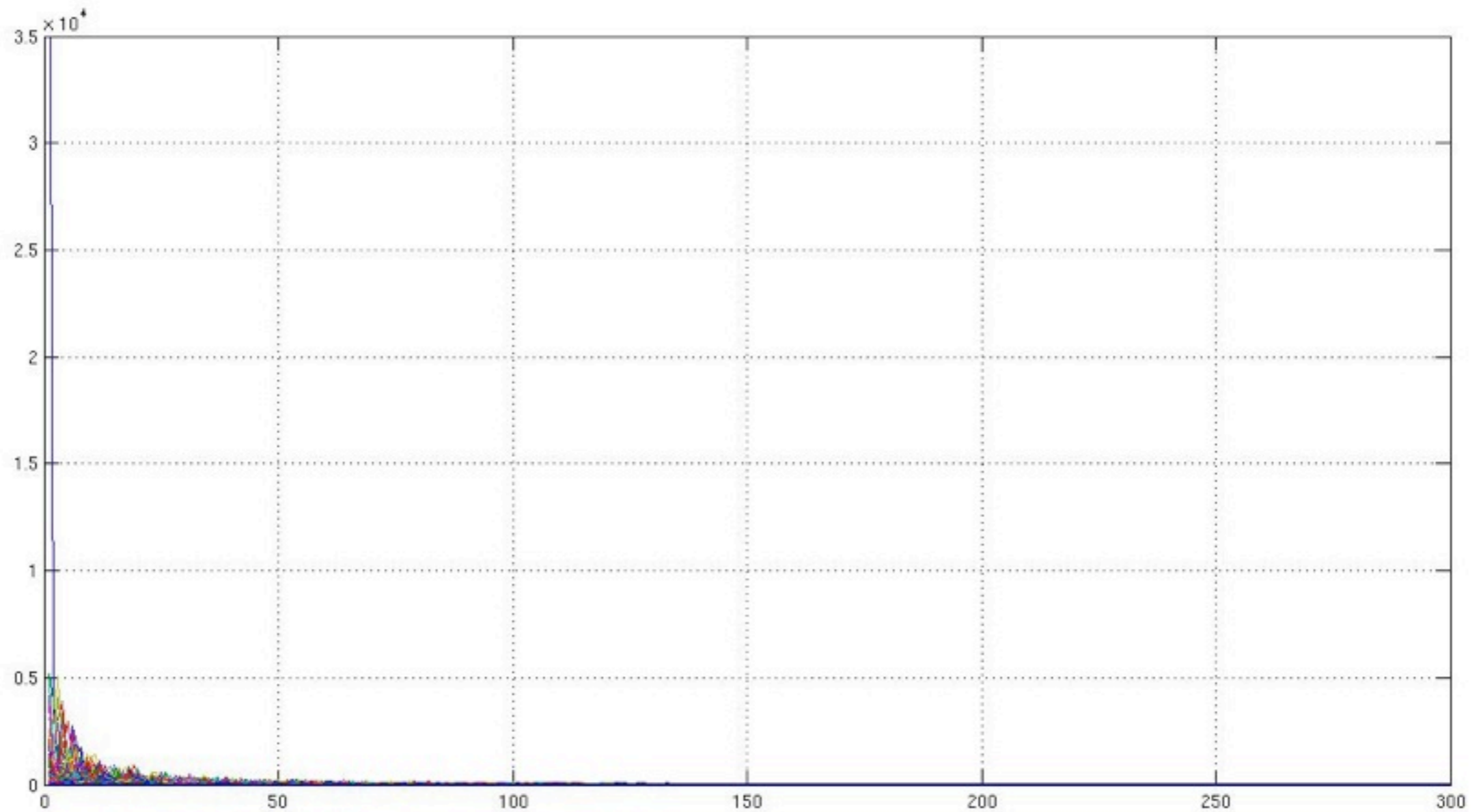
# MAGIC UNRAVELLED

✻ Discrete Cosine/Wavelet Transform

# DCT



# DCT



# WHAT IS DCT ?

$$X_{k_1, k_2} = \sum_{n_1=0}^{N_1-1} \sum_{n_2=0}^{N_2-1} x_{n_1, n_2} \cos \left[ \frac{\pi}{N_1} \left( n_1 + \frac{1}{2} \right) k_1 \right] \cos \left[ \frac{\pi}{N_2} \left( n_2 + \frac{1}{2} \right) k_2 \right].$$

- ✻ Similar to Fourier Transform
- ✻ With no sine pulses
- ✻ Uses only real values (comp. to FT)
- ✻ Easier than K-L Transform (optimal)

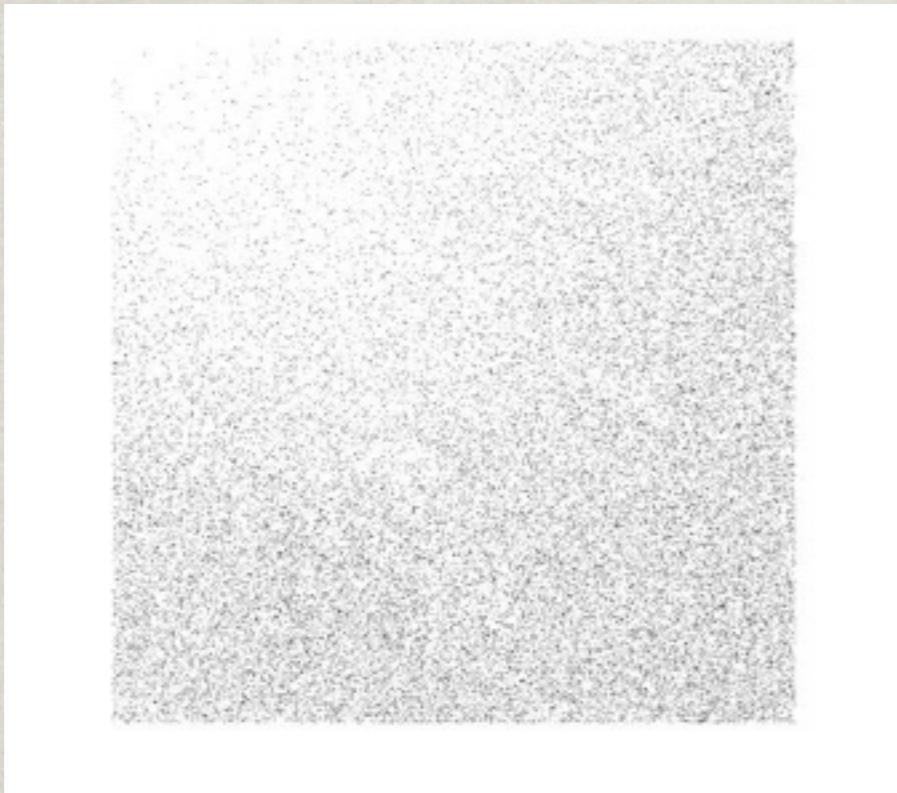
# WHY DCT ?

- ✻ Decorrelates (dis-similar) the data
- ✻ Energy compaction property
- ✻ Reduces entropy
- ✻ Useful for audio/image compression

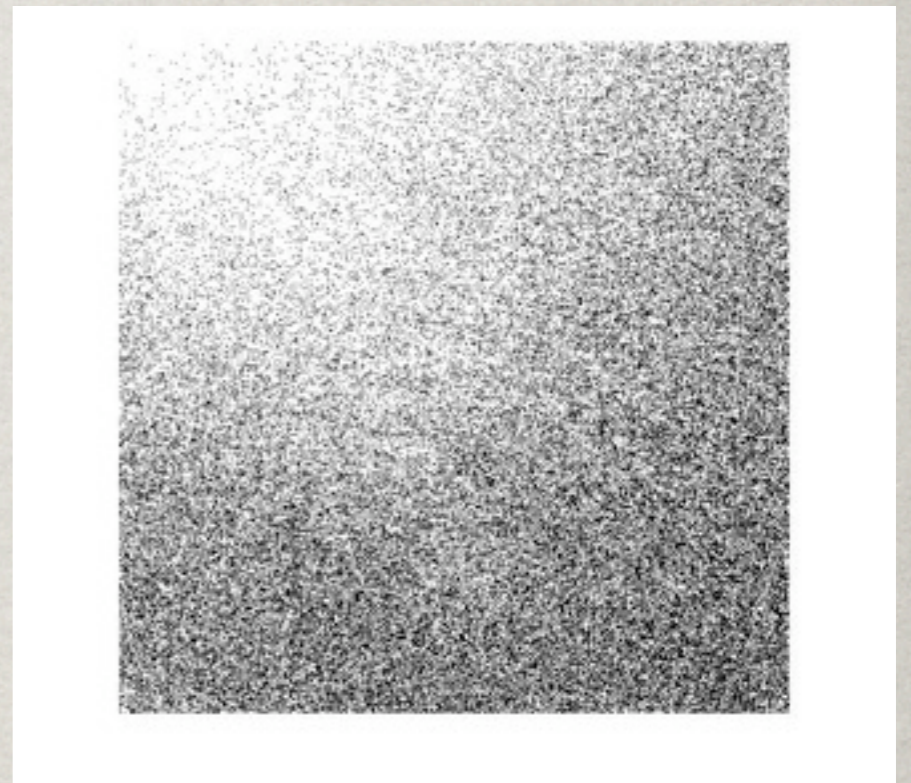


# COMPRESSED DCT

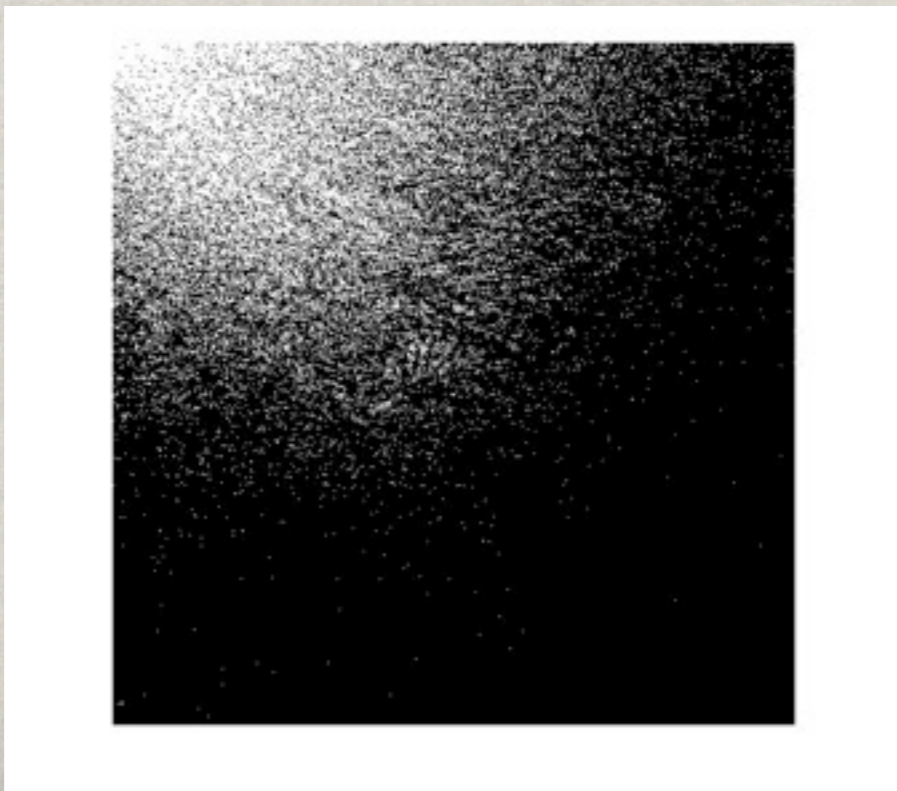
Orig.



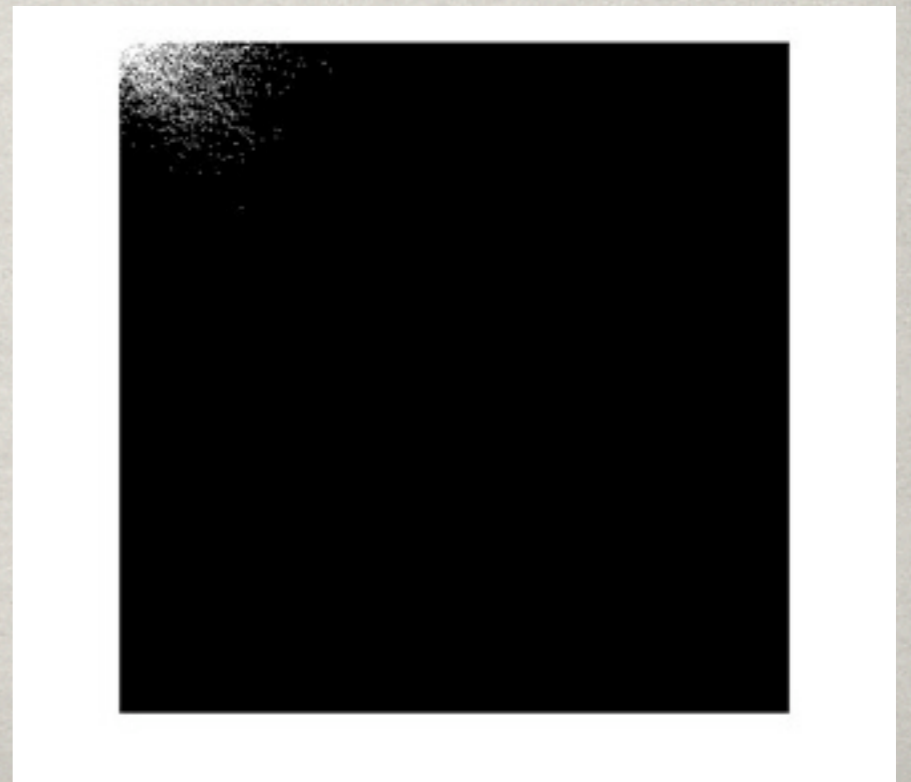
Orig.  
<1 removed



Orig.  
<10 removed



Orig.  
<100 removed



# UNCOMPRESSED-IDCT

Orig.



Rec.  
<1 removed



Rec.  
<10 removed



Rec.  
<100 removed



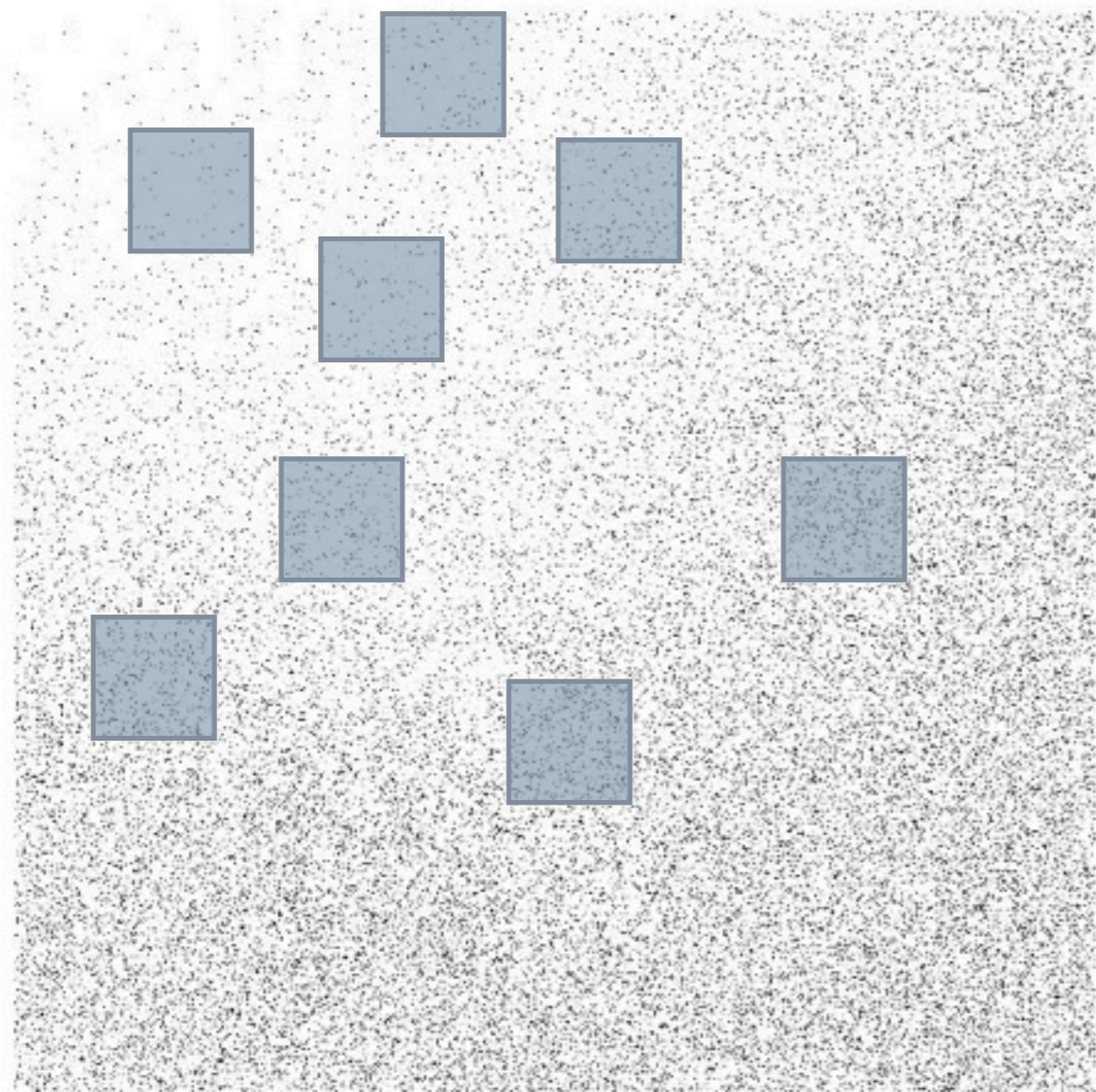
# DIGITAL IMAGE STORAGE

- ✱ Store only the significant DCT components
- ✱ Use IDCT to recover the image
- ✱ Significant savings in storage
- ✱ Almost no perceptible image quality loss

# PHILOSOPHICAL LESSON ?

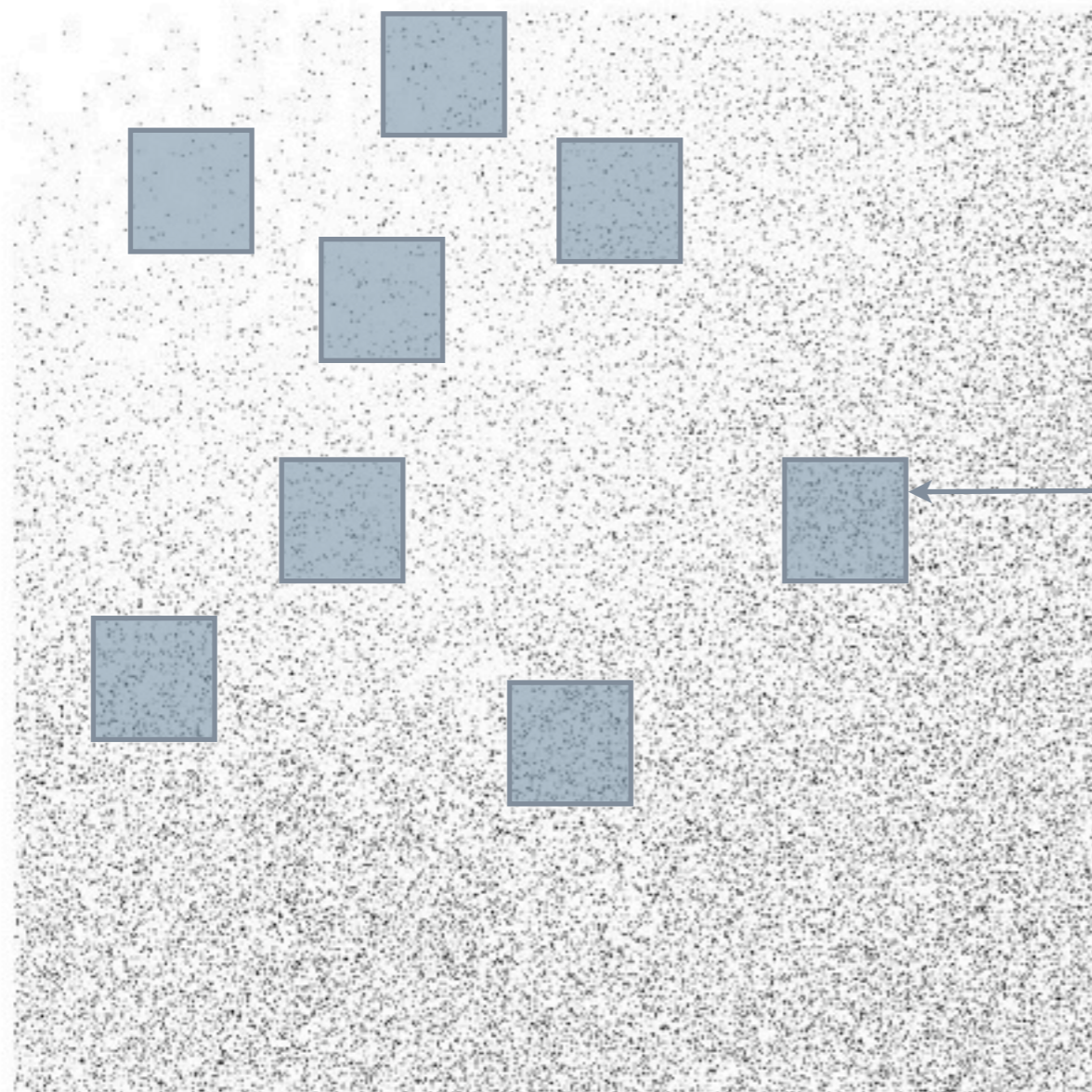
- ✻ Conventional photography
  - ✻ collects a lot
  - ✻ throws away a lot
- ✻ Can we collect as much as we need
  - ✻ Ans: **Compressed sensing**

# NAIVE APPROACH FOR COMPRESSION



Select some fixed  
number of entries of  
DCT

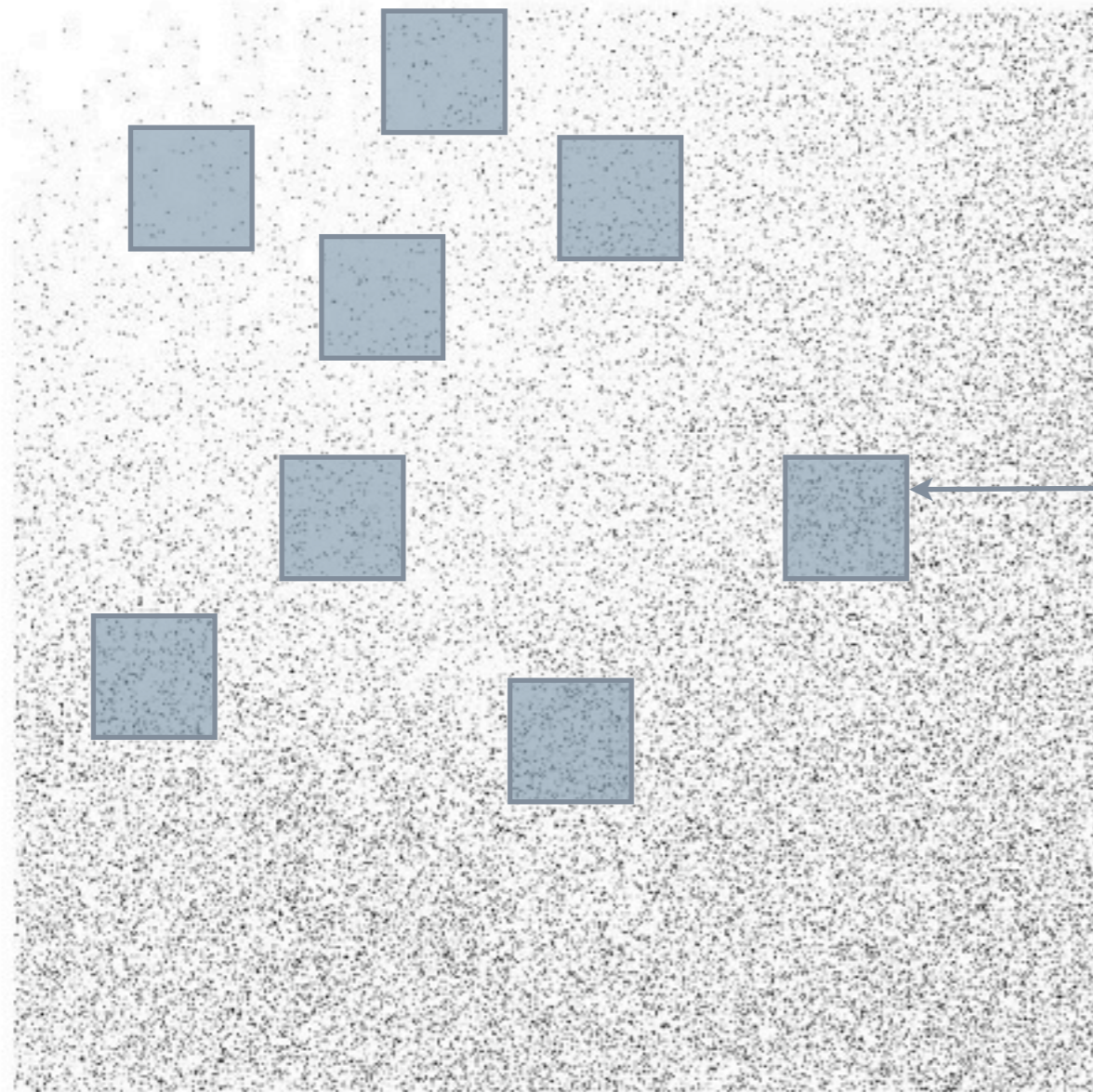
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Option: Take random linear combinations



# COMPRESSED SENSING



# COMPRESSED SENSING

vector  $x$  - DCT image

$$x = D \times I$$



# COMPRESSED SENSING

$$\underbrace{\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1N} \\ \vdots & \vdots & \dots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mN} \end{bmatrix}}_A$$

vector  $x$  - DCT image

$$x = D \times I$$

Matrix  $A$  for  
random linear combinations  
of  $x$



# COMPRESSED SENSING

$$y = \underbrace{\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1N} \\ \vdots & \vdots & \dots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mN} \end{bmatrix}}_A$$

$$m \ll N$$

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Observation vector  $y$



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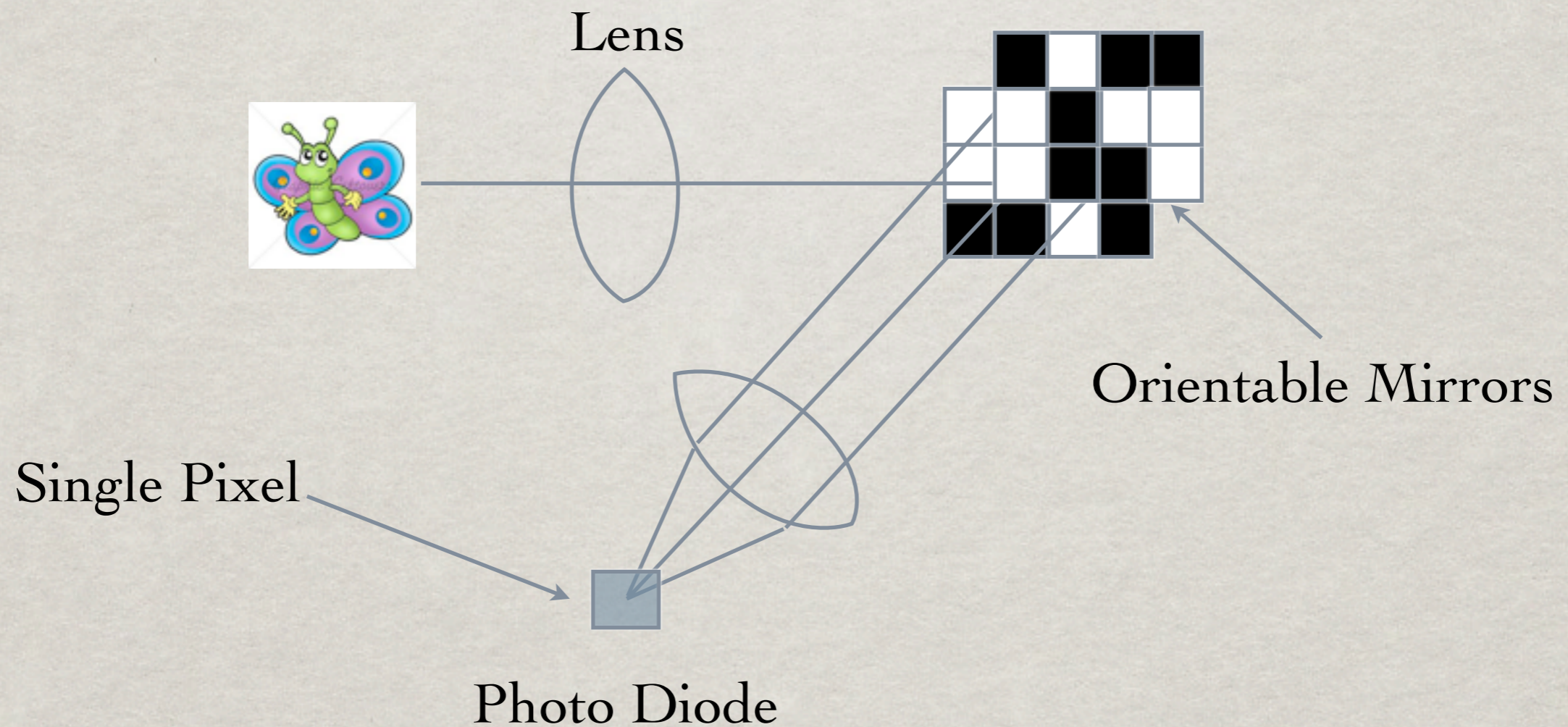
## Recovery

find  $x$

$$y = Ax$$

$$|x|_0 \leq s$$

# SINGLE PIXEL CAMERA ASSEMBLY



No. of Measurements  $m \sim s \log N/s$

# SINGLE PIXEL CAMERA

- ✻ Single pixel used  $m$  times
- ✻ Each time taking a different linear combination of  $x$
- ✻ **Hope** : We can recover  $I$  !

# IMAGE RECONSTRUCTION



Original



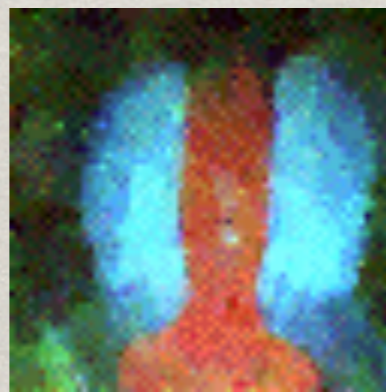
65536 Pixels  
 $m=1300$



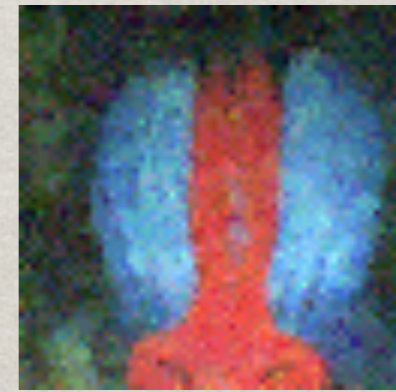
65536 Pixels  
 $m=3300$



Original



4096 Pixels  
 $m=800$

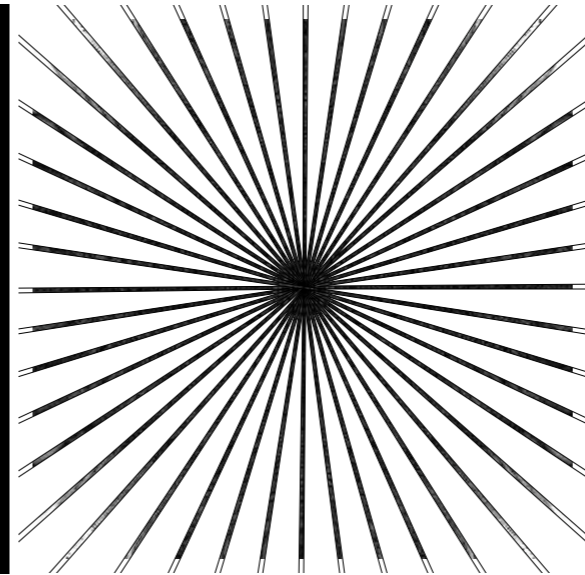


4096 Pixels  
 $m=1600$



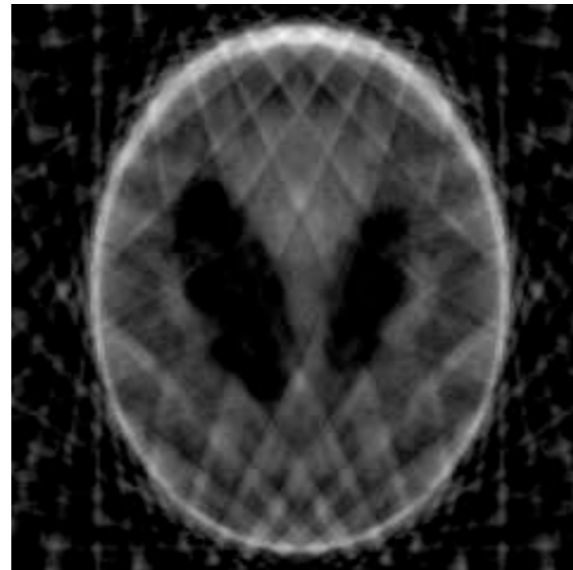
# MAGNETIC RESONANCE IMAGING

Shepp-Logan Phantom 256 × 256



Frequency grid

$L_2$  minimization



CS reconstruction

We can recover MRI with 10 % measurements

# WHY CS ?

- ✱ No extraordinary gain in photography
- ✱ CCDs are cheap (because silicon responds to visible light)
- ✱ Real gain is in non-visible spectrum image acquisition for which CCDs are costly
- ✱ Use of exotic detectors (not possible with MP CCDs)

# SIMPLE GEOMETRY

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Let image be  $x_1$   
 $x_2$

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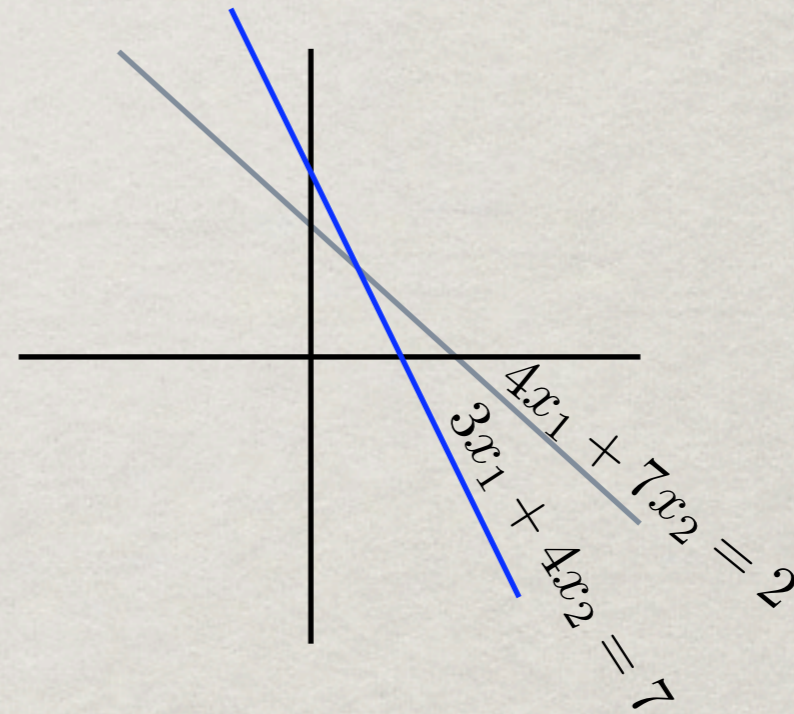
$$\begin{bmatrix} 7 \\ 2 \end{bmatrix} = \begin{bmatrix} 3 & 4 \\ 4 & 7 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

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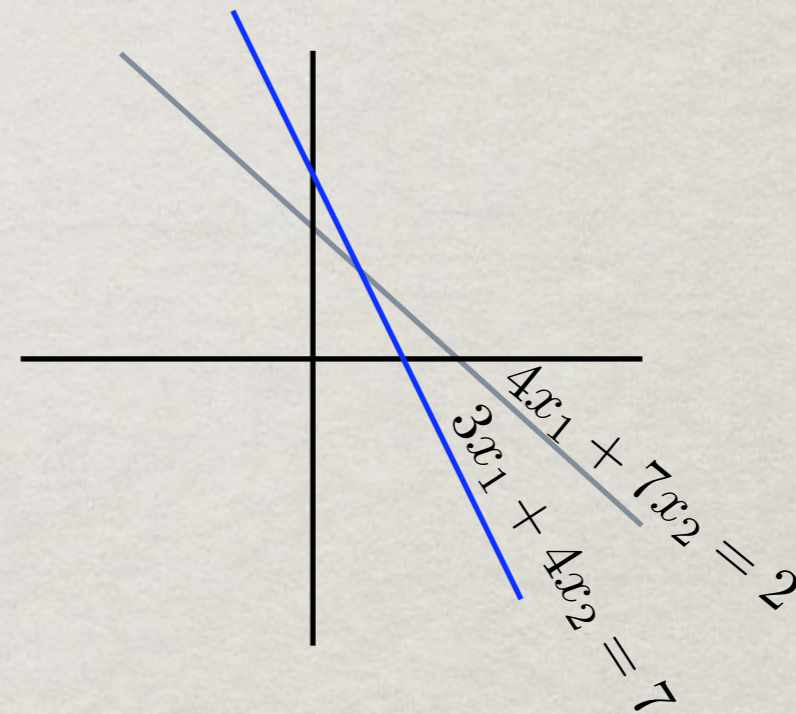


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Soln.  $x_1, x_2?$  Easy



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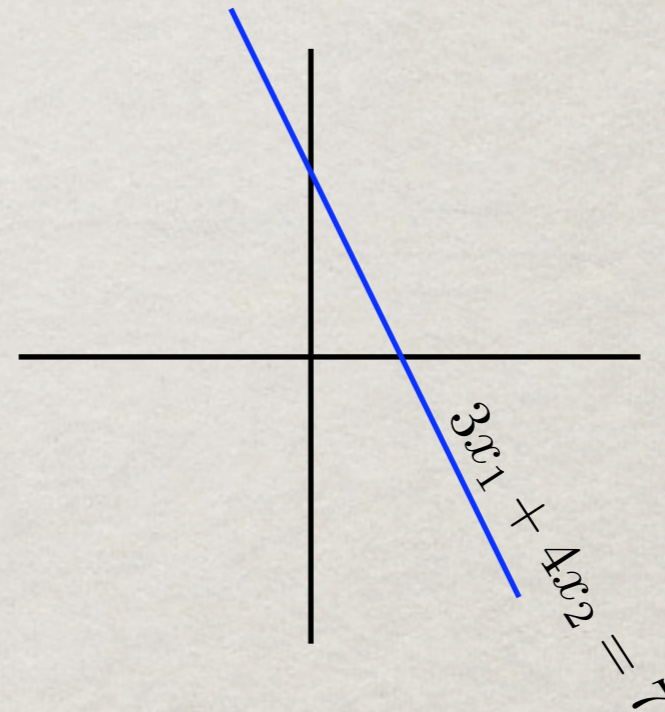
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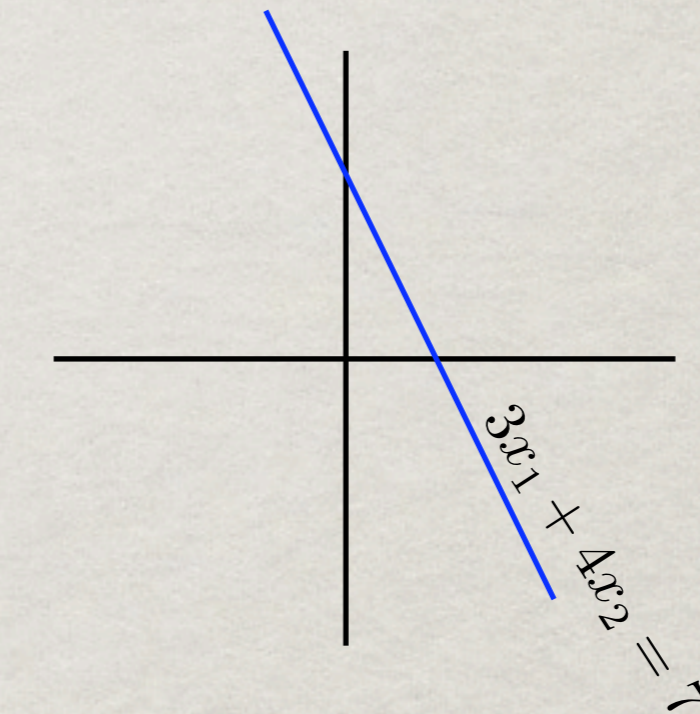
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Too many Solutions

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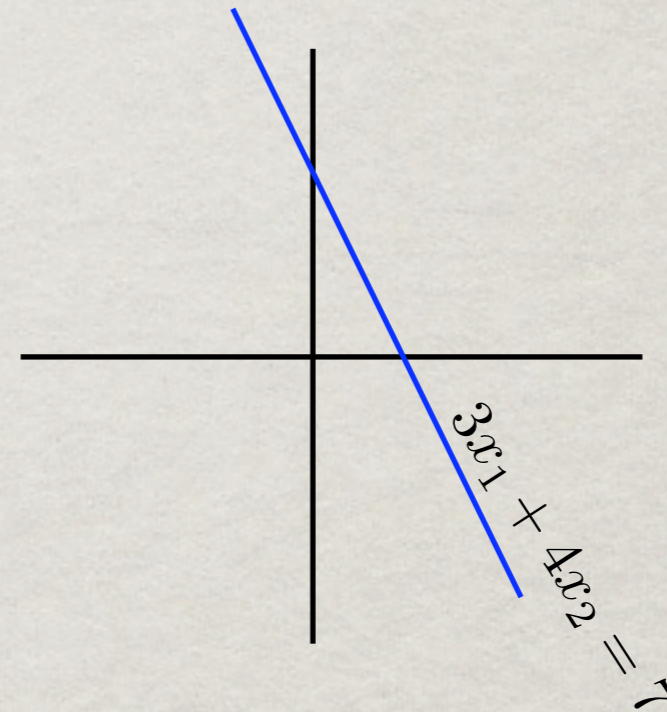
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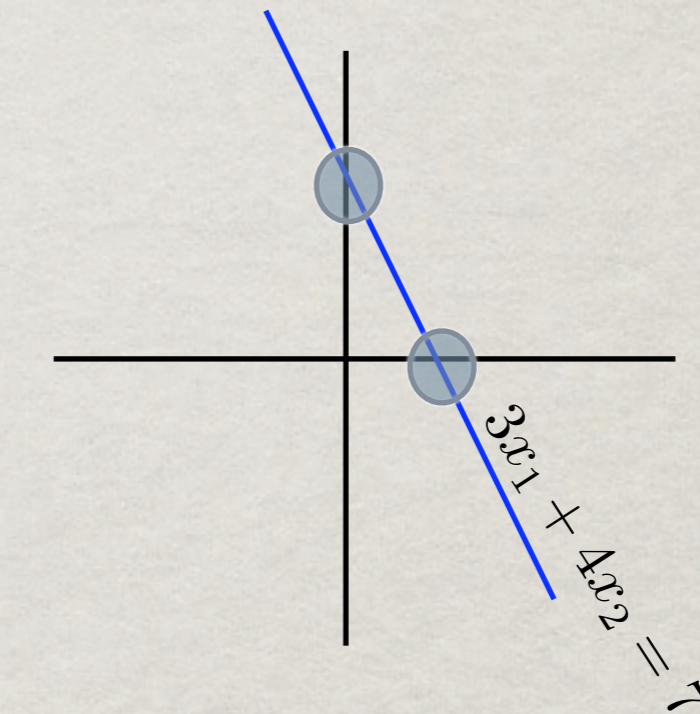
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Too many Solutions

Need more structure !

Two solutions even if soln is 1-sparse

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Real Image

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$$\underbrace{\begin{bmatrix} x'_1 \\ x'_2 \\ x'_3 \end{bmatrix}}_{x'} = \begin{bmatrix} 0 \\ b \\ 0 \end{bmatrix}$$

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$$y \neq Ax'$$

# RETURN TO CS

$$\begin{array}{l} \text{find } x \\ y = Ax \\ |x|_0 \leq s \end{array} \quad \text{equivalent to} \quad \begin{array}{l} \min \|x\|_0 \\ \text{s.t. } y = Ax \end{array}$$

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- ✱ Solving the  $l_0$  problem has exponential complexity

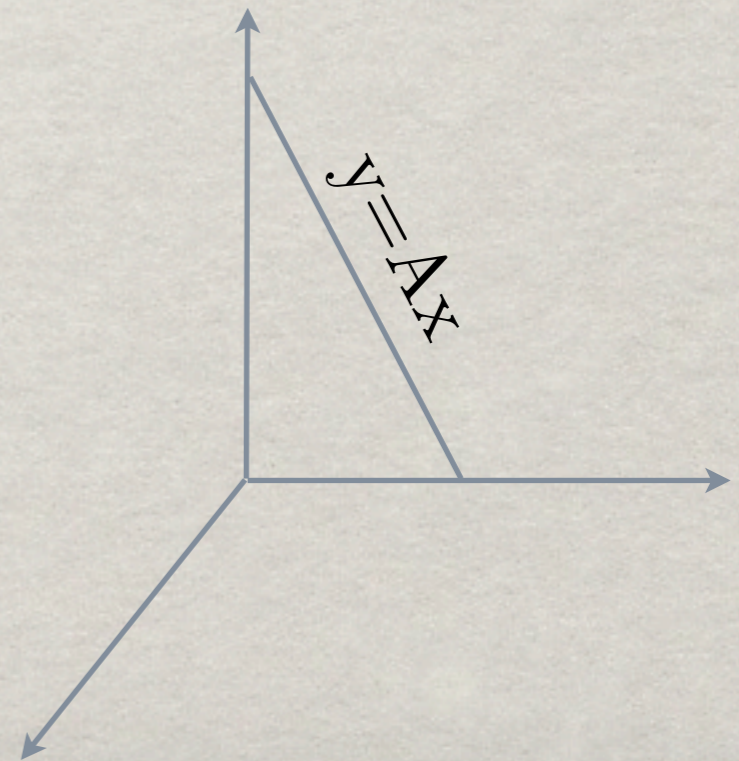
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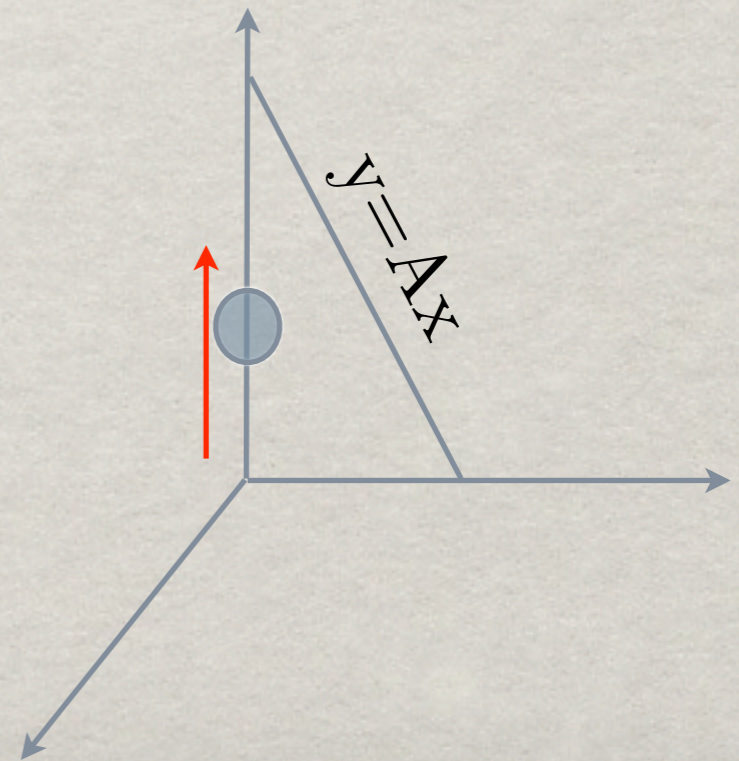
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find  $x$

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equivalent to

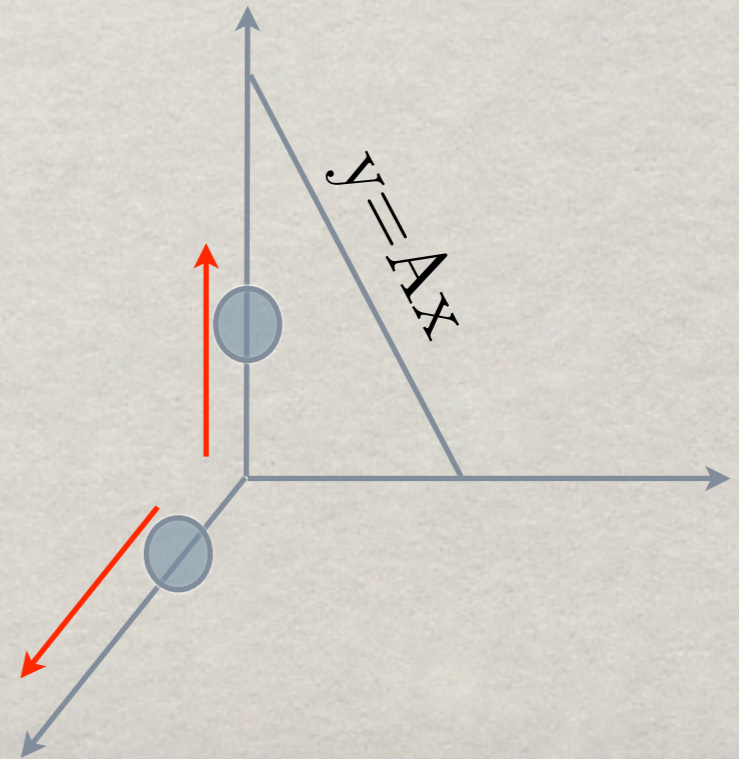
$$\min \|x\|_0$$

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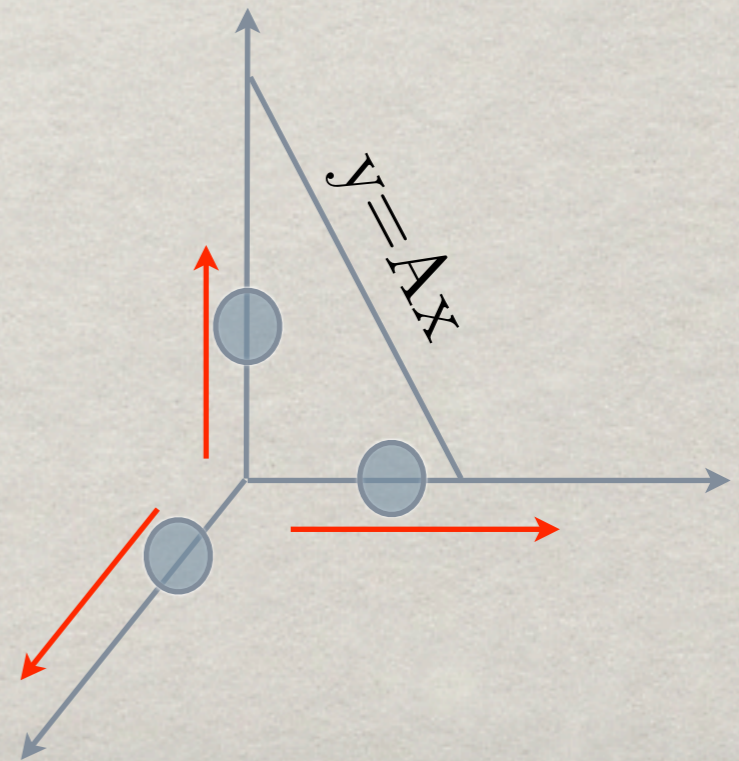
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# RELAXATION

$$\begin{aligned} \min \quad & \|x\|_2 \\ \text{s.t.} \quad & y = Ax \end{aligned}$$

$\|x\|_2 = r$  - sphere of radius  $r$

# RELAXATION

$$\begin{aligned} \min \quad & \|x\|_2 \\ \text{s.t.} \quad & y = Ax \end{aligned}$$

$\|x\|_2 = r$  - sphere of radius  $r$

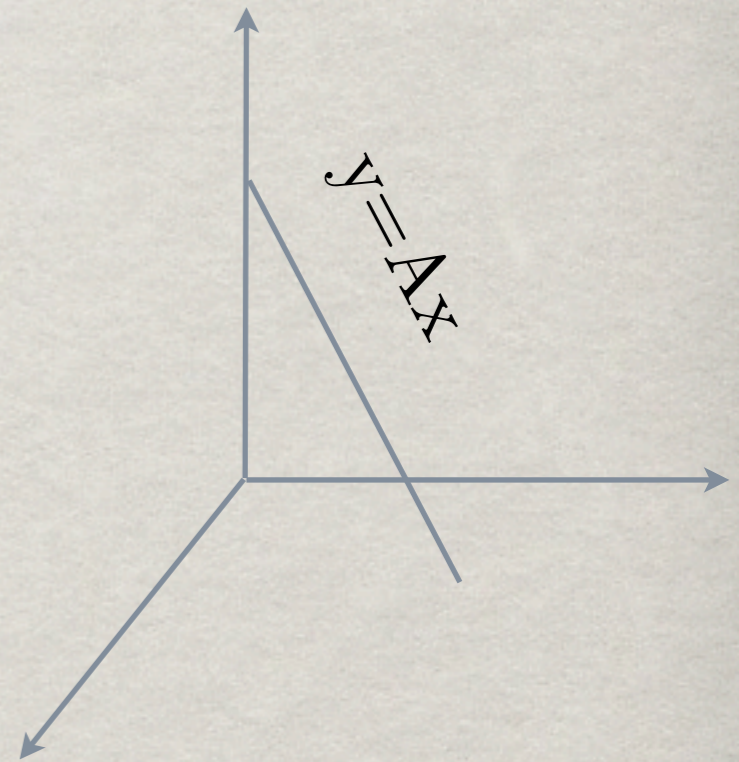
$$\|x\|_2 = \sqrt{\sum_{i=1}^N x_i^2}$$

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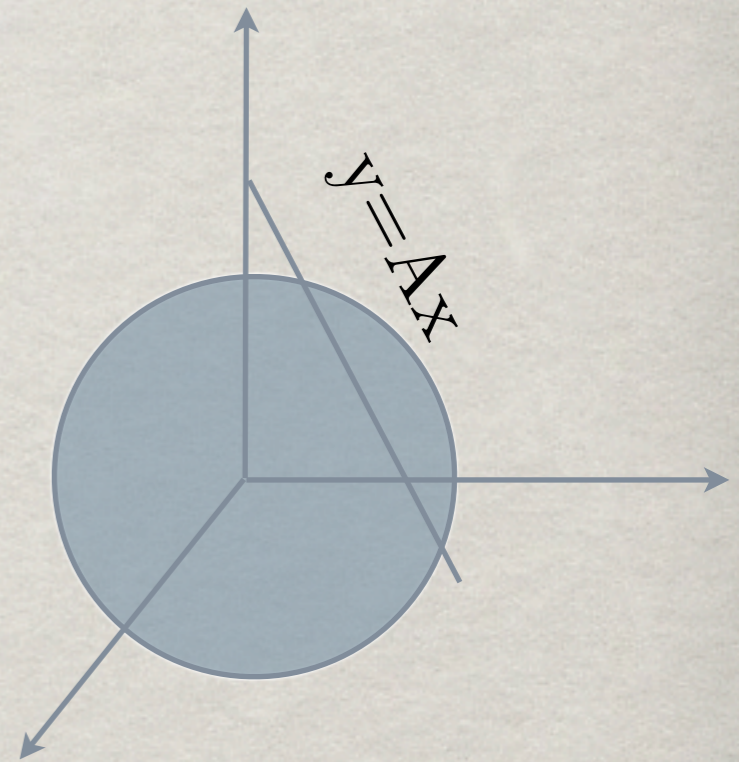


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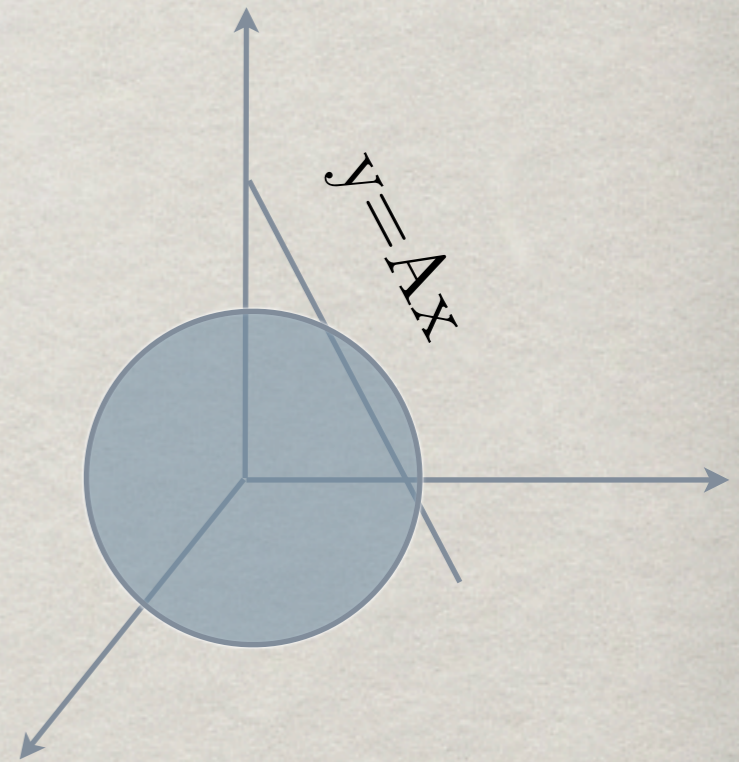


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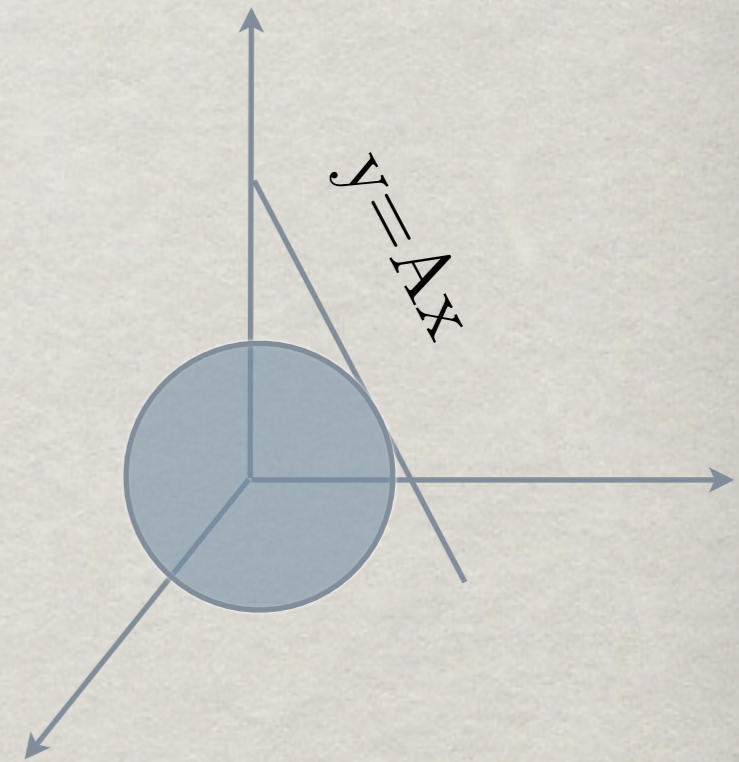


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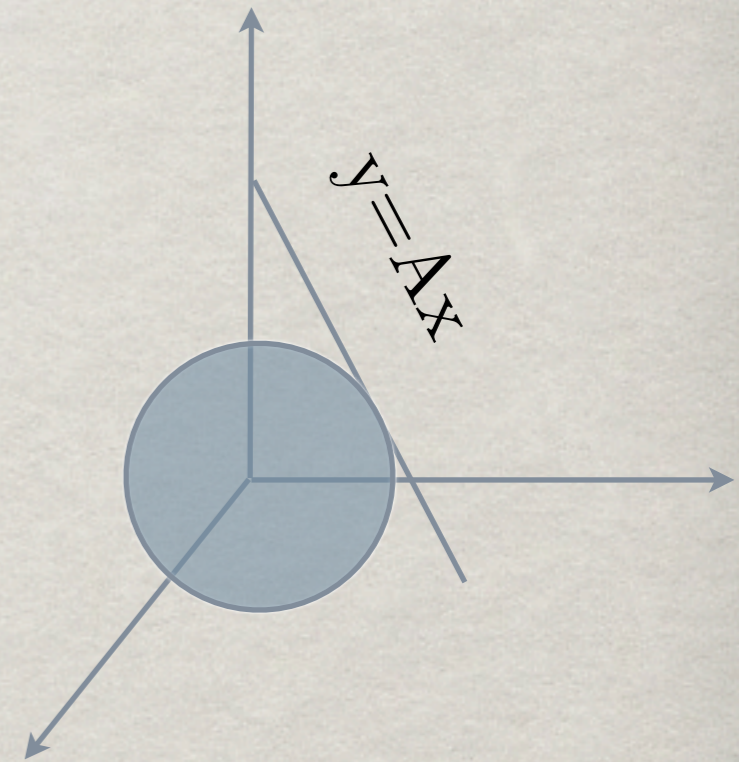
Non-Sparse Solution

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Non-Sparse Solution

- ✿ Simple to solve but does not serve the purpose

# RELAXATION

$$\begin{aligned} \min \quad & \|x\|_1 \\ \text{s.t.} \quad & y = Ax \end{aligned}$$

$\|x\|_1 = r$  - polyhedron

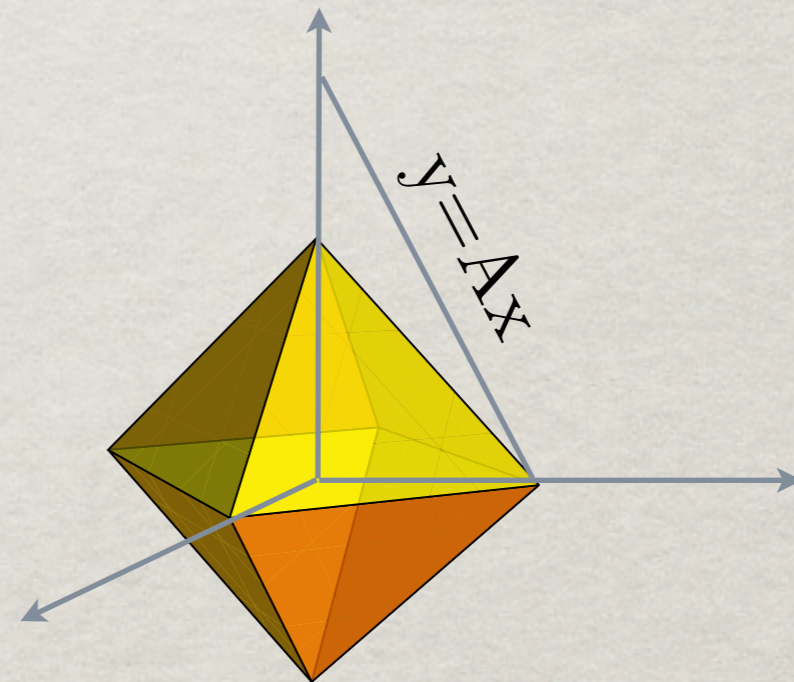
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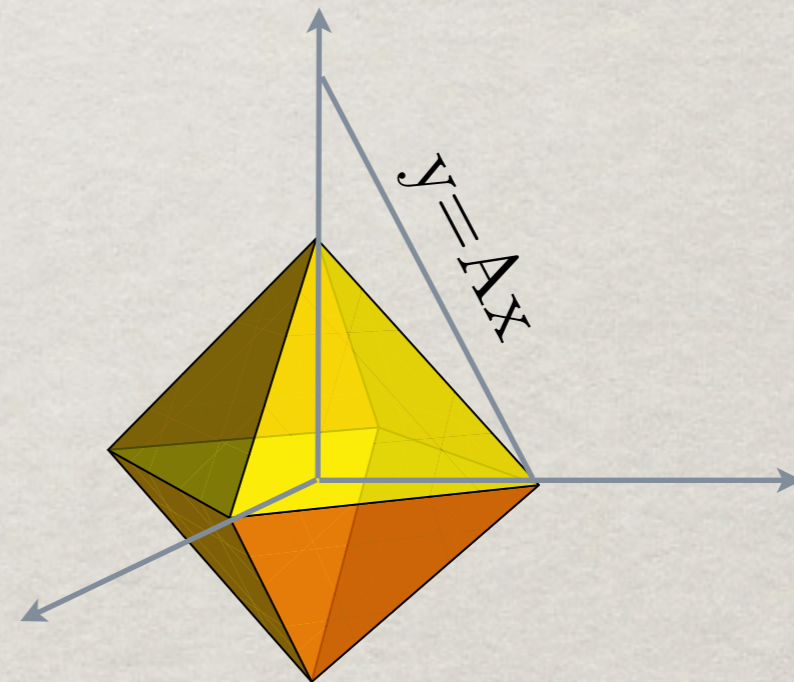


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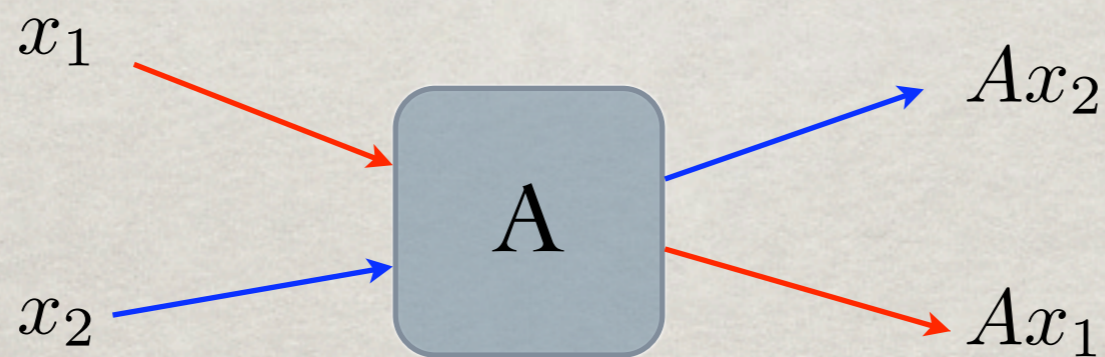
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No competing sparse solutions

# TECHNICAL CONDITIONS

- ✪ Restricted Isometry Property: Any  $2s$  columns of  $A$  are nearly orthogonal [CandesTao'05]



$2s$  RIP  $\Rightarrow$  Pairwise distances  
are preserved for  $s$ -sparse signals



# EXAMPLES OF A-MATRICES

- ✱ Each entry independent coin flips
- ✱ Columns dist. uniformly on unit sphere
- ✱ Independent Gaussian entries
- ✱ Exact recovery if  $m \sim s \log N/s$

# CONCLUSIONS

- ✻ Questioned the conventional wisdom
- ✻ Exciting area of research

**THANK YOU**

# SIMPLIFICATION

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