

# Welcome to EECS 16A!

## Designing Information Devices and Systems I

Ana Arias and Miki Lustig

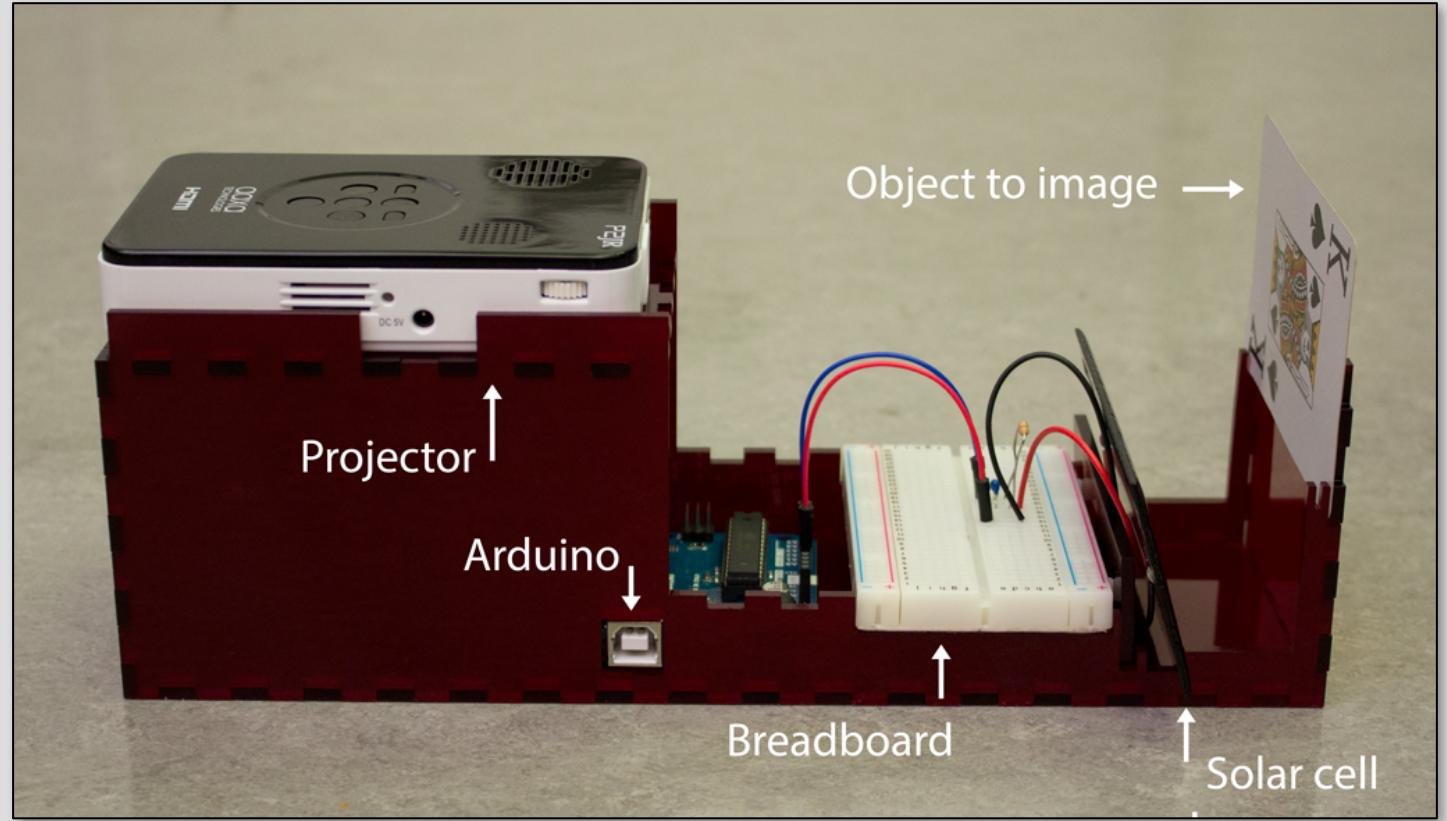


2022

Lecture 1A  
Tomography and Linear Equations



# Module 1: Imaging



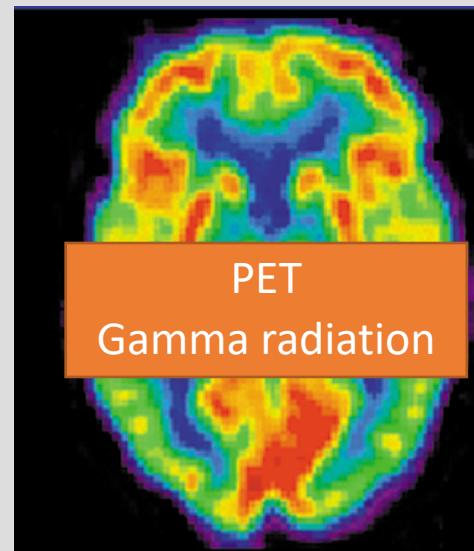
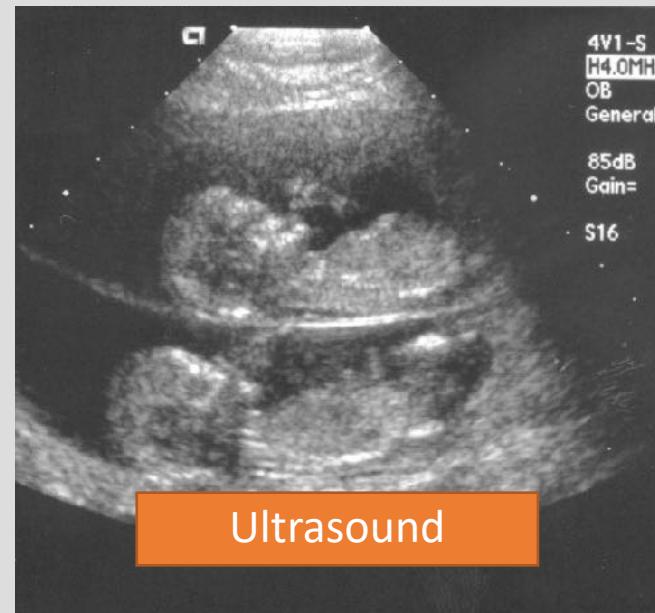
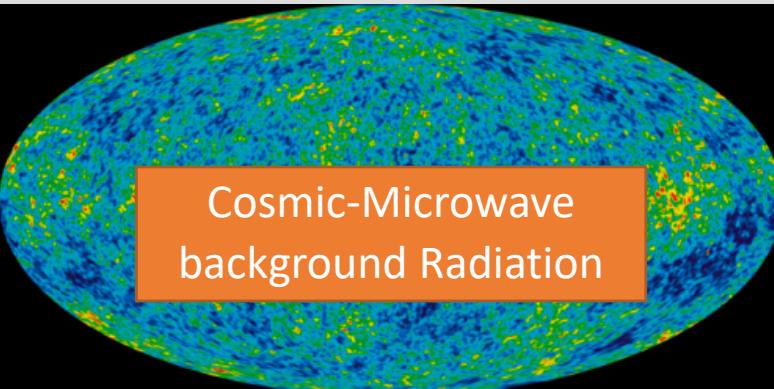
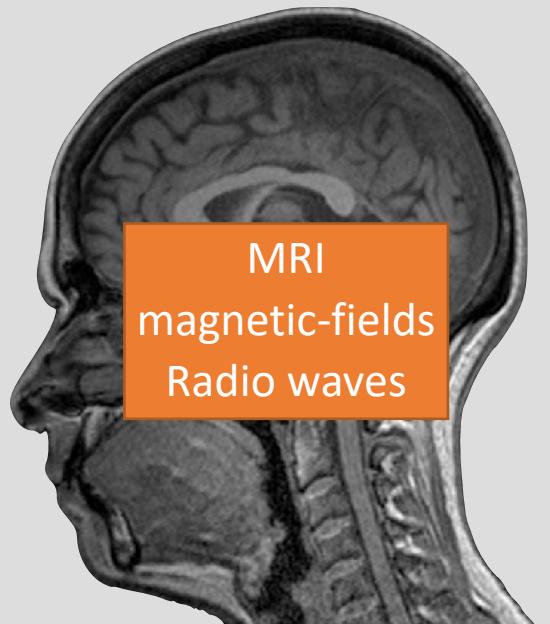
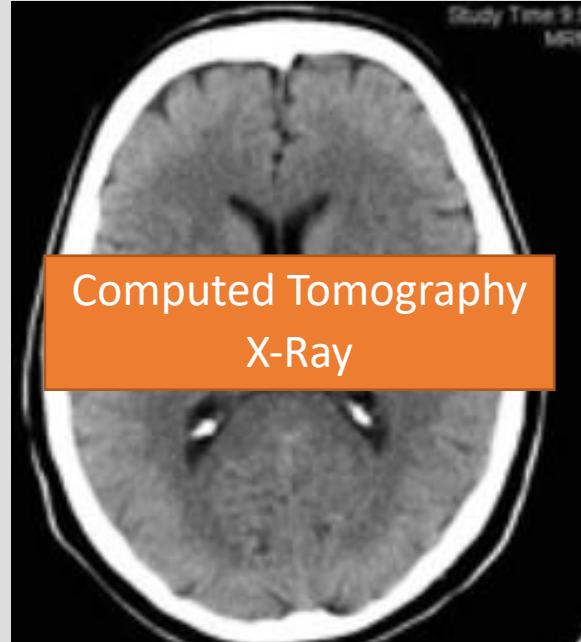
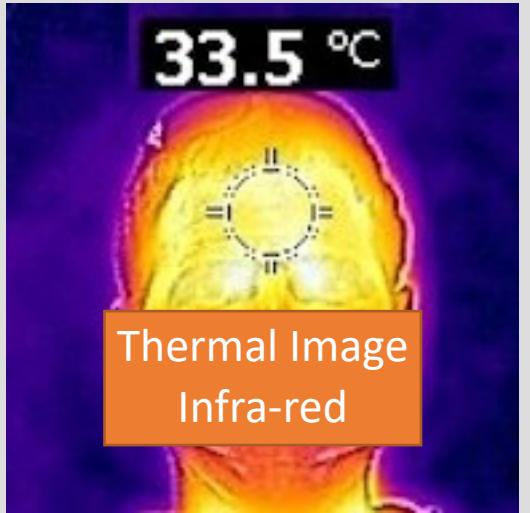
# Image

Merriam-Webster: A visual representation of something

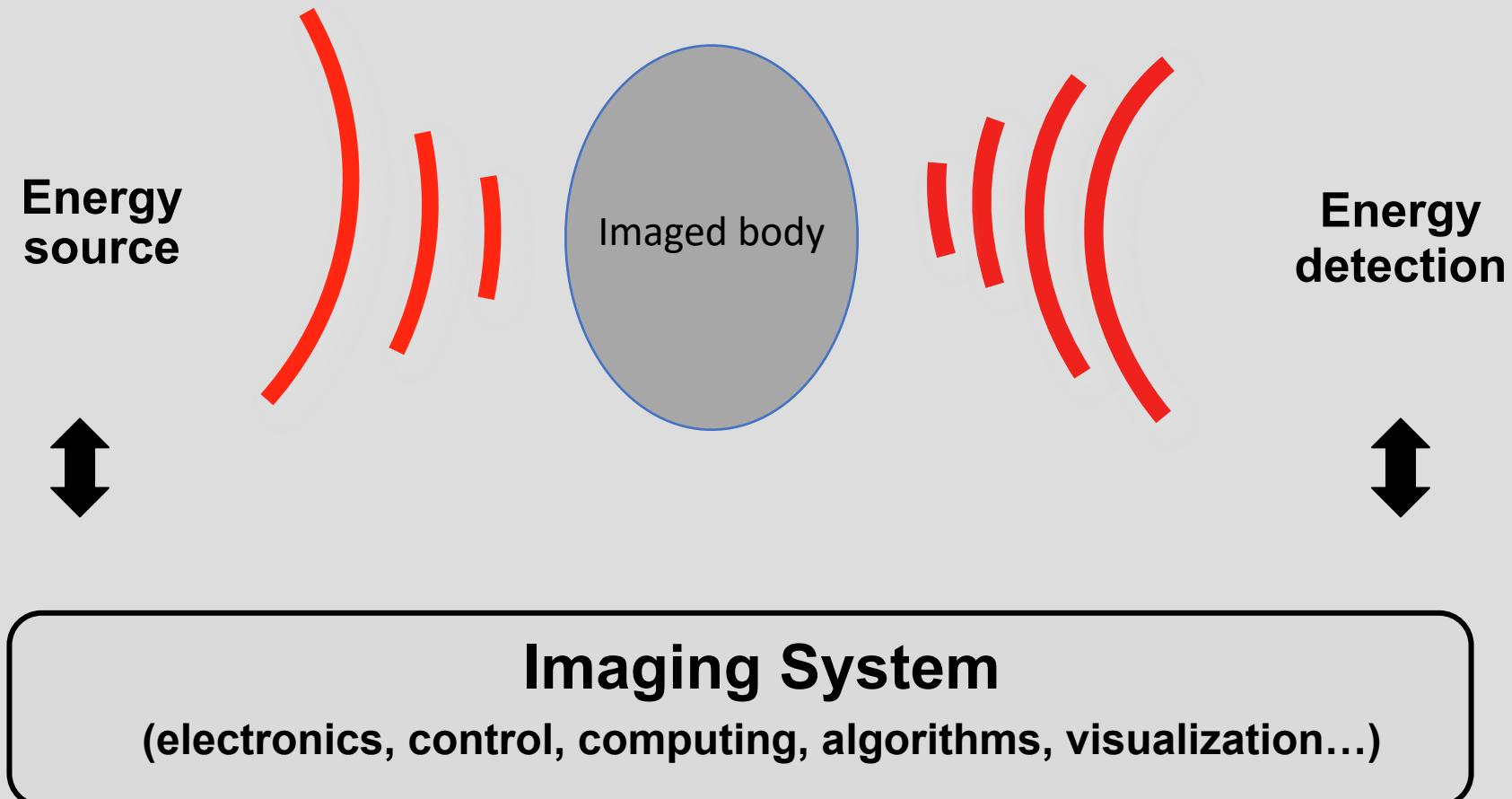
# Imaging

Merriam-Webster: the action or the process of producing an image

# Different Images



# Imaging Systems in General

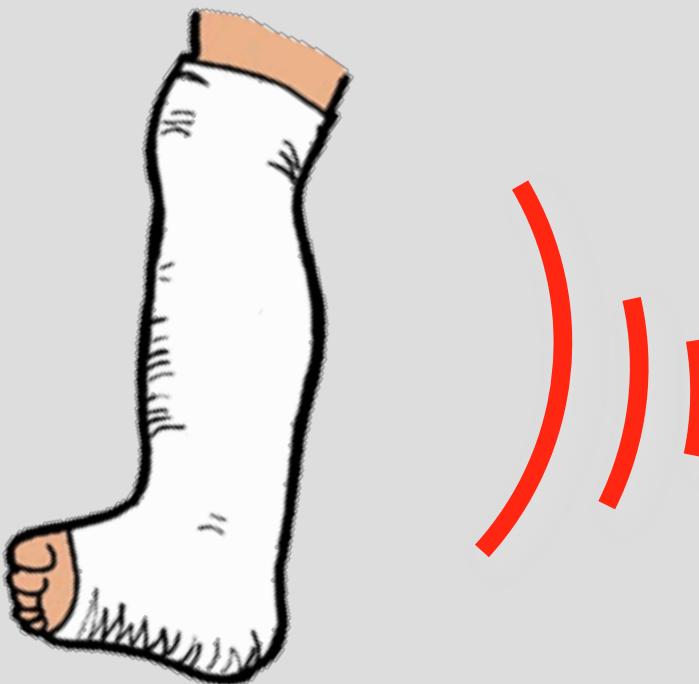


“Medical imaging” circa 1632

“The Anatomy Lesson of Dr. Nicolaes Tulp”, Rembrandt  
Mauritshuis, The Hague



# Projection Xray



# Projection Xray



# Tomography



‘tomo’ – slice  
‘graphy’ – to write

Assume it is not desirable to slice open leg. How does tomography visualizes cross-sectional slices?

# From Projections

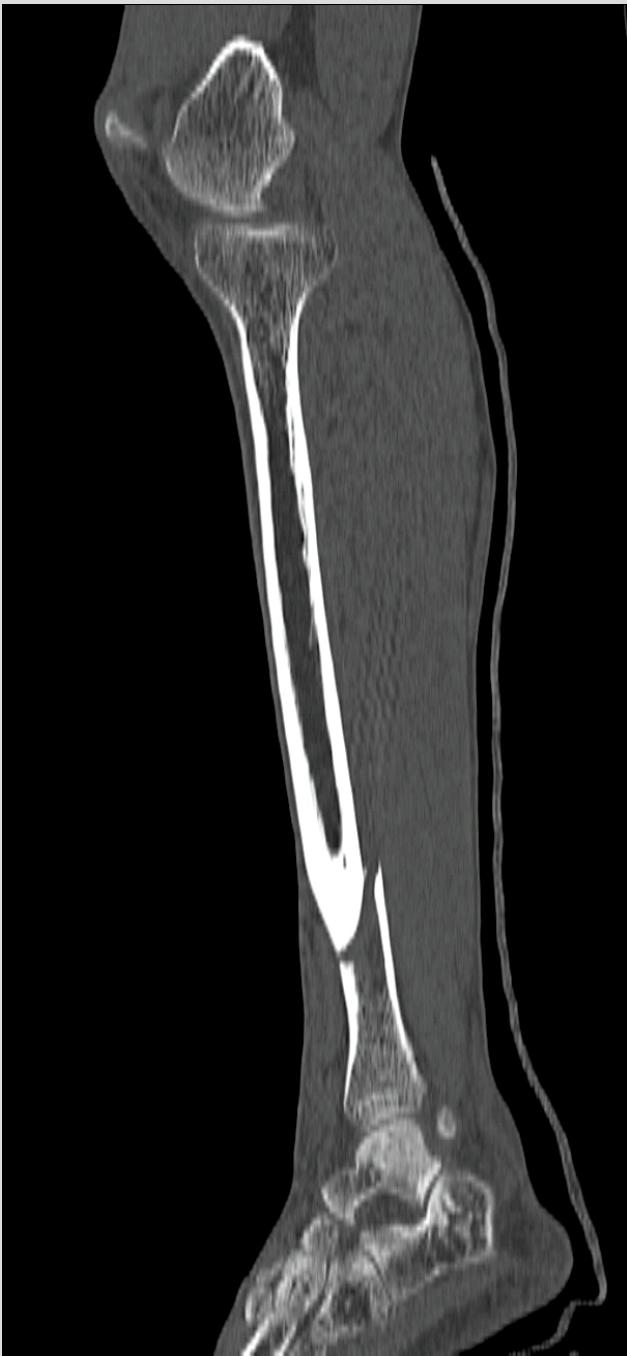
## Projections



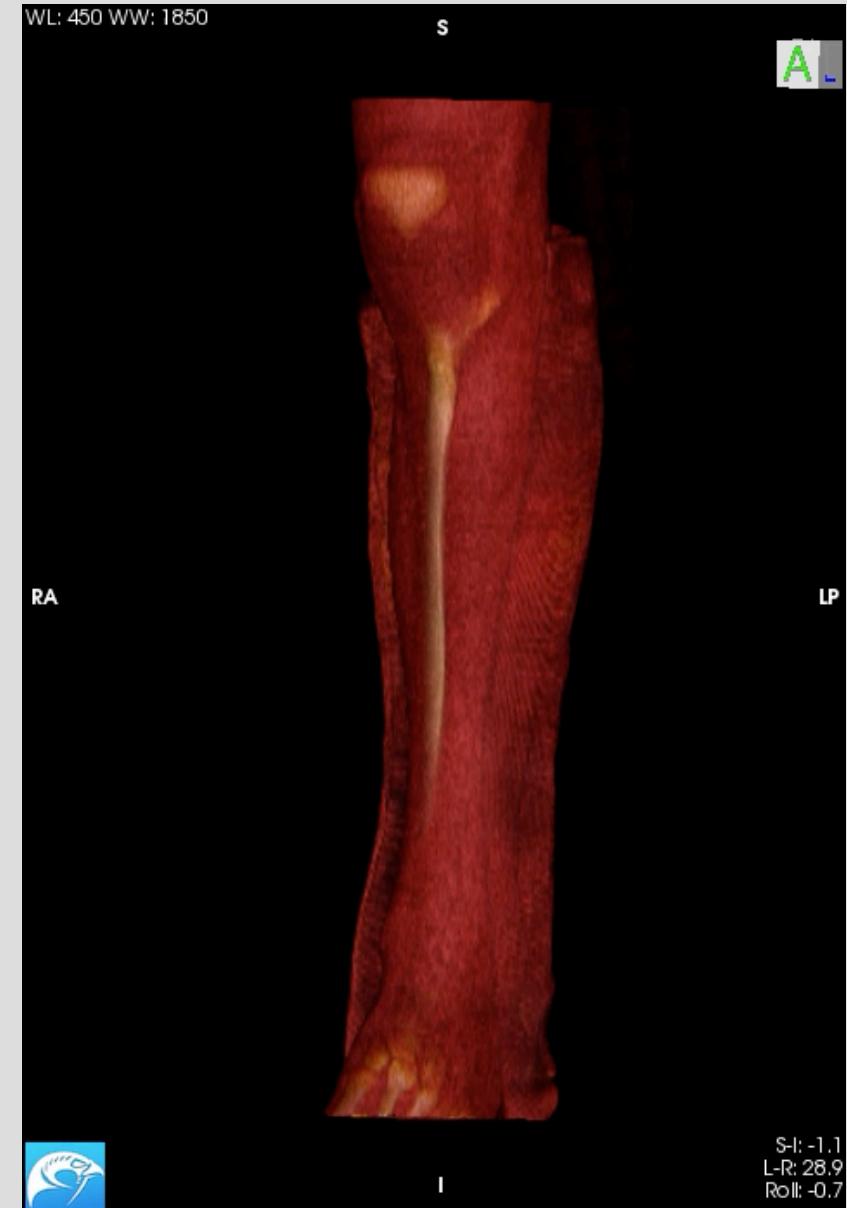
## Axial Slices



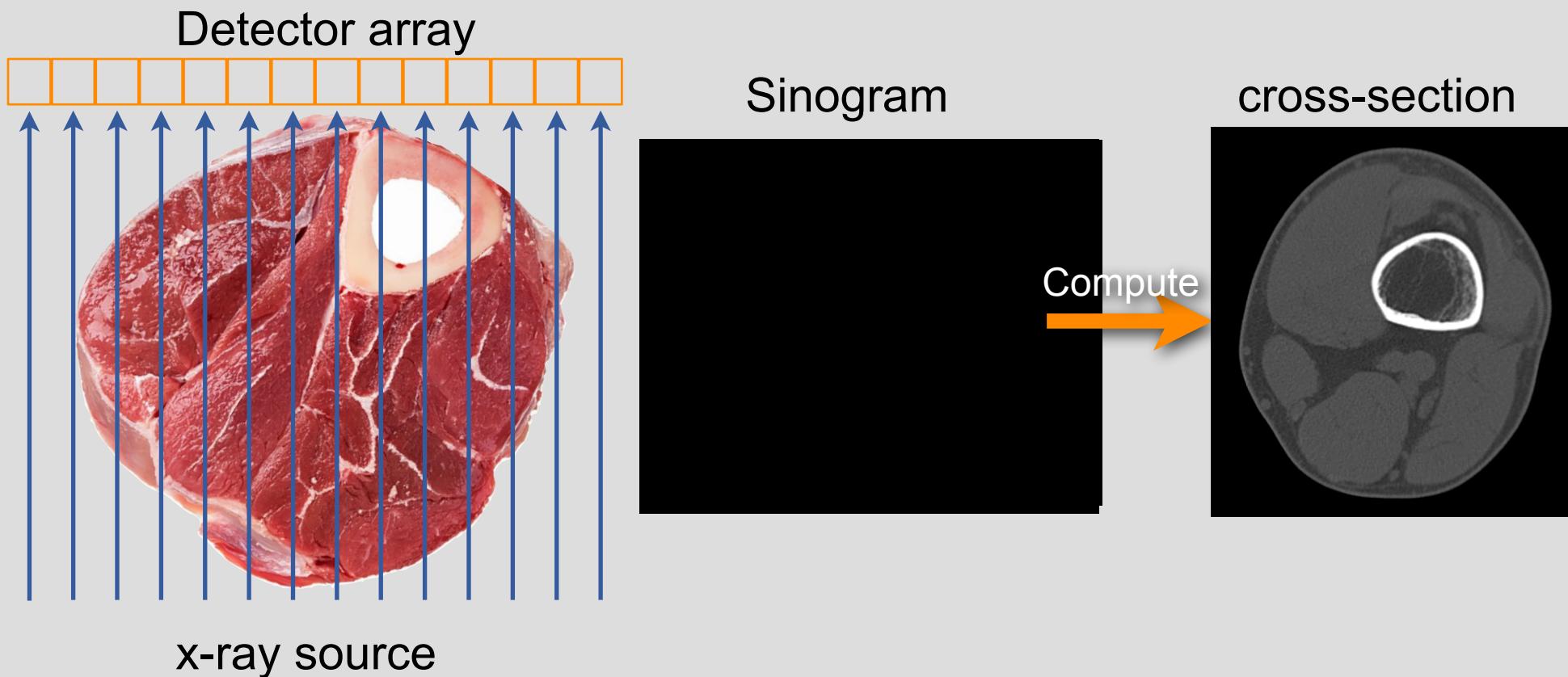
## Sagittal Slices



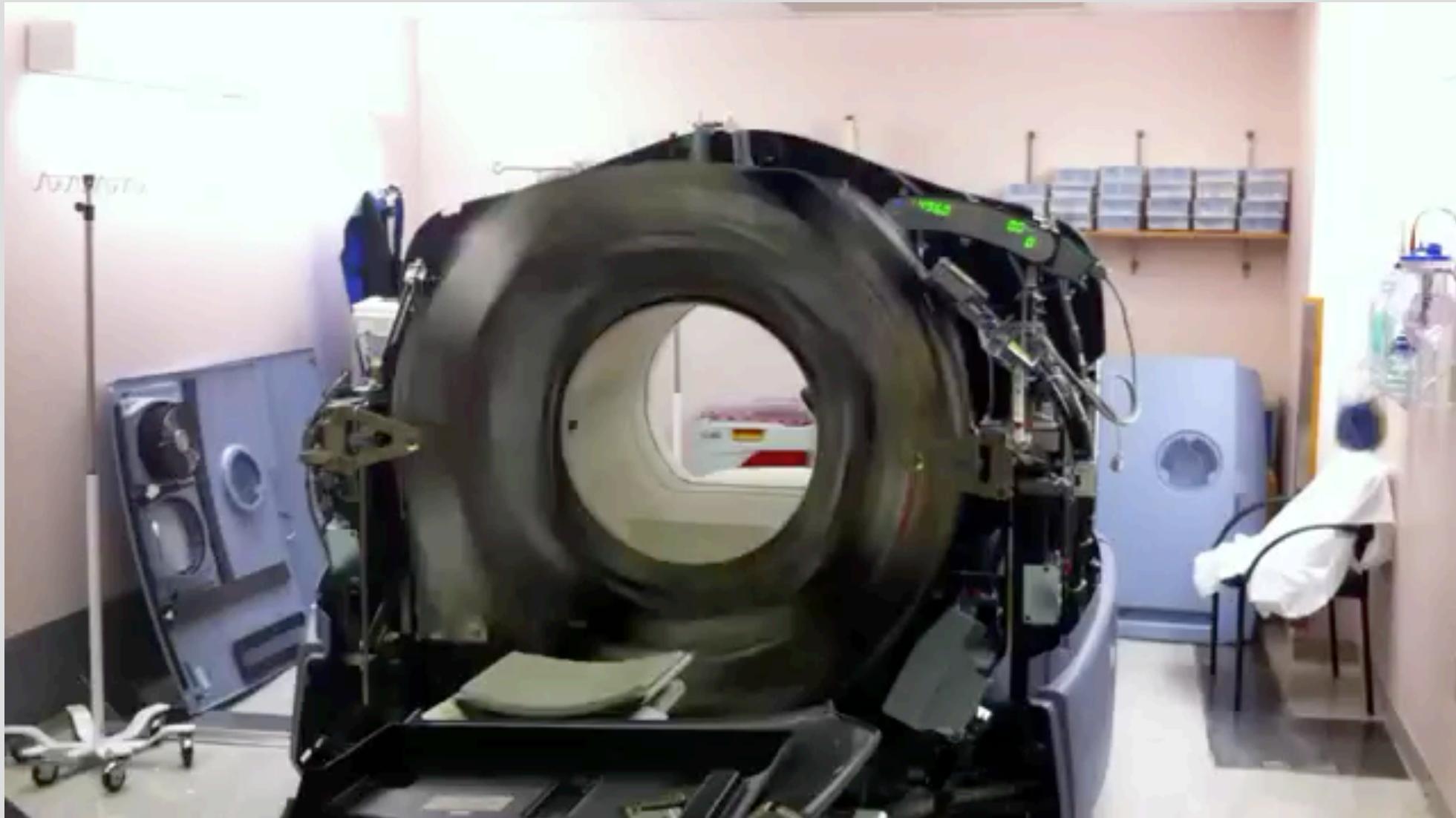
# 3D Rendering from Slices



# Computed Tomography



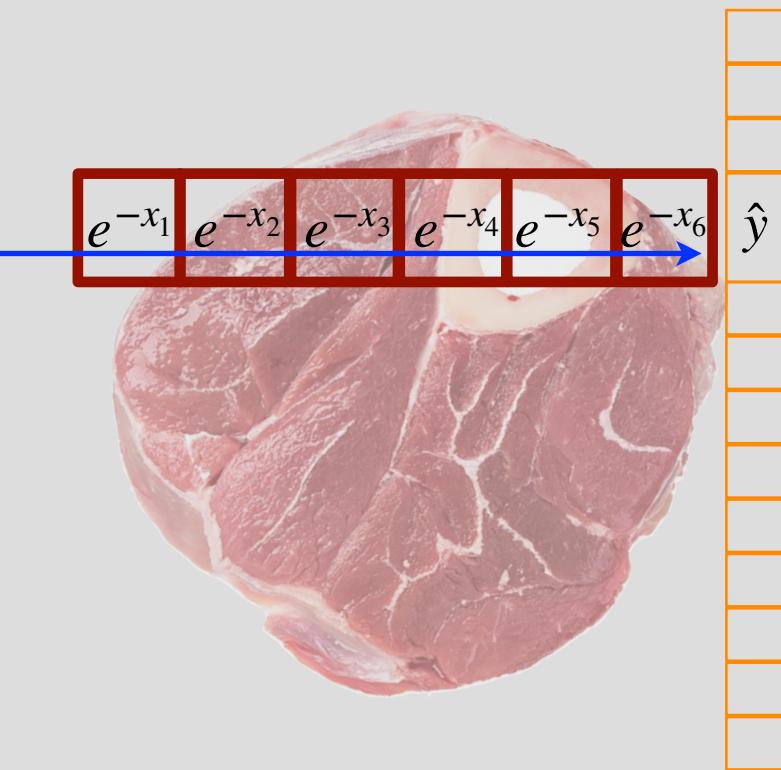
# Computed Tomography



<http://www.youtube.com/watch?v=4gkIQHM19aY&feature=related>

# Modeling Tomography

power=1



$$1 \cdot e^{-(x_1+x_2+x_3+x_4+x_5+x_6)} = \hat{y}$$

$$\log\{e^{-(x_1+x_2+x_3+x_4+x_5+x_6)}\} = \log\{\hat{y}\}$$

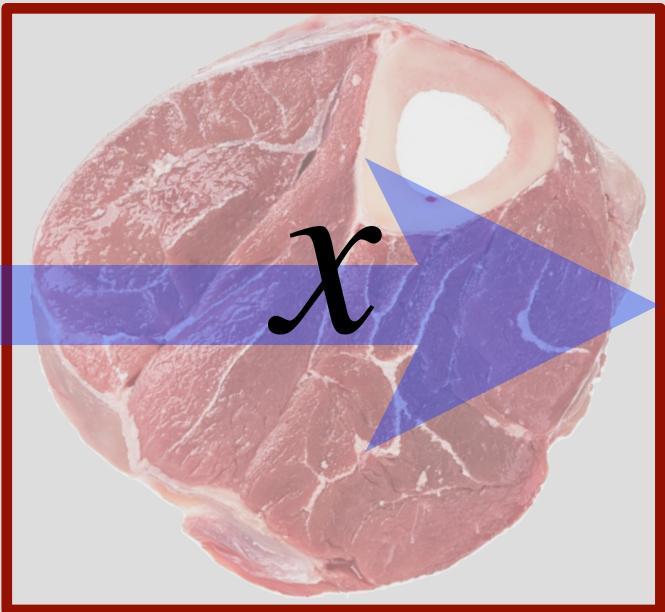
$$y = -\log\{\hat{y}\}$$

$$x_1 + x_2 + x_3 + x_4 + x_5 + x_6 = y$$

.... or  $y$  is the sum of x-ray attenuation coefficients along a line

# Modeling Tomography

power=1



Unknown



Measurement

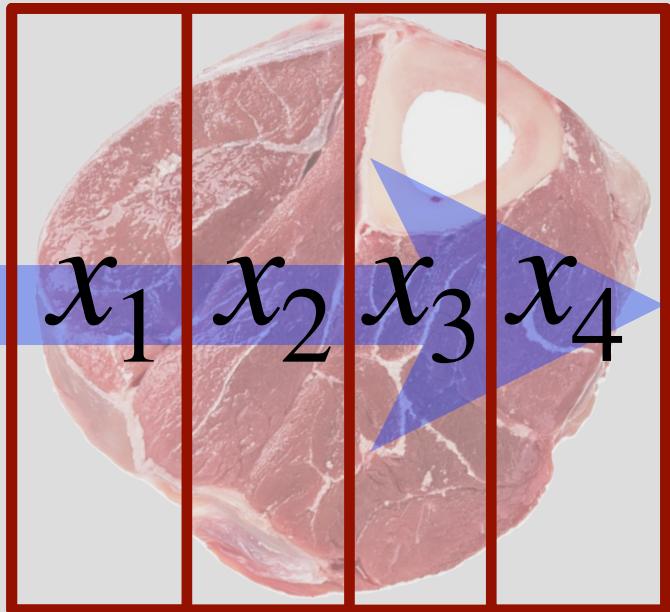
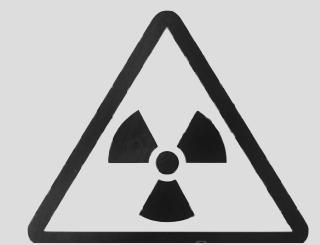
$$y = 1 \cdot x$$

↓

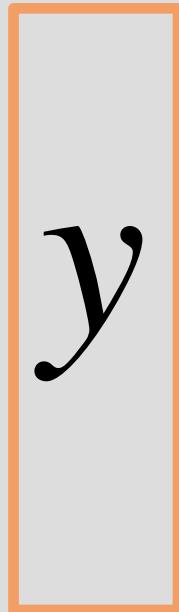
$$x = y$$

# Modeling Tomography

power=1



Unknown



Measurement

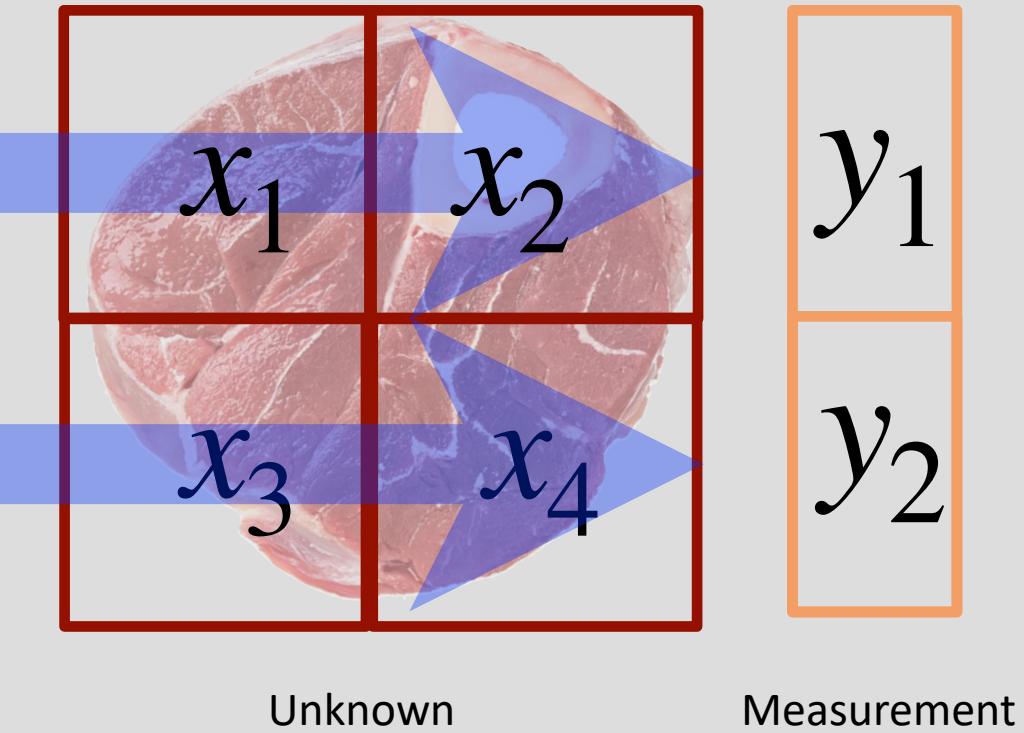
$$y = x_1 + x_2 + x_3 + x_4$$



1 equation 4 unknowns!



# Modeling Tomography



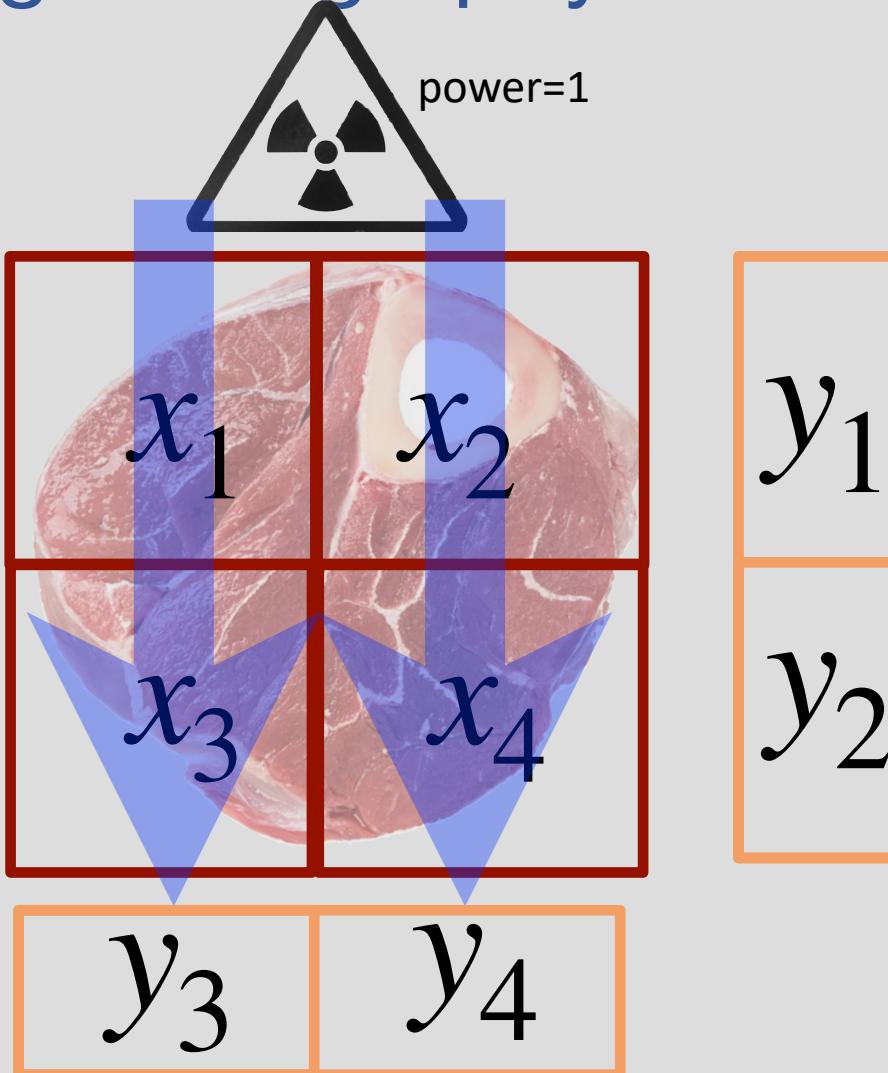
$$y_1 = x_1 + x_2$$
$$y_2 = x_3 + x_4$$



2 equation 4 unknowns!



# Modeling Tomography



$$y_1 = x_1 + x_2$$

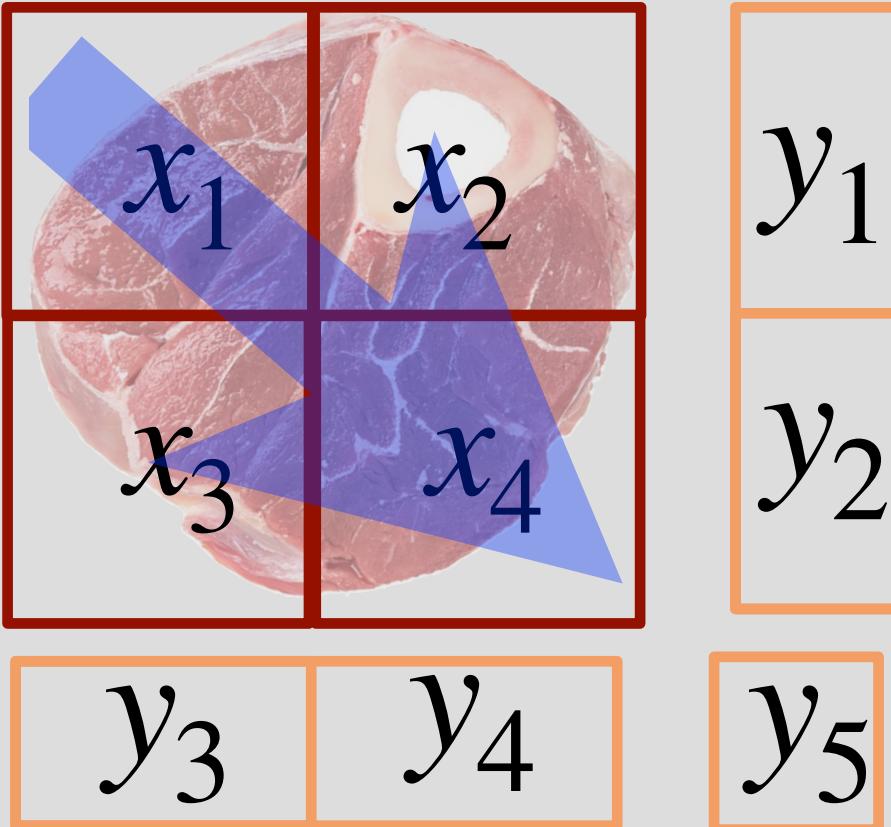
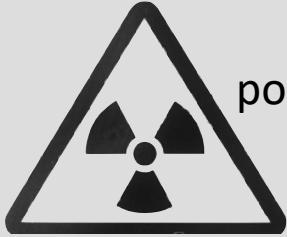
$$y_2 = \quad \quad \quad x_3 + x_4$$

$$y_3 = x_1 \quad \quad + x_3$$

$$y_4 = \quad + x_2 \quad \quad + x_4$$

Can we solve this?

# Modeling Tomography



$$y_1 = x_1 + x_2$$

$$y_2 = \quad \quad \quad x_3 + x_4$$

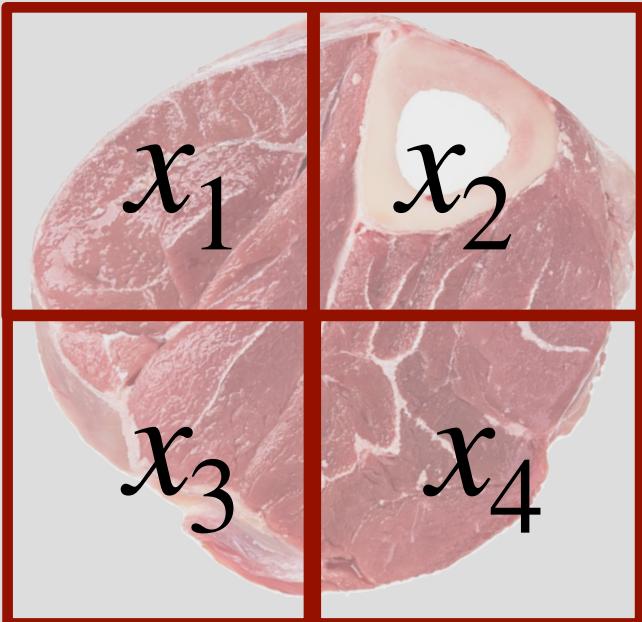
$$y_3 = x_1 \quad \quad + x_3$$

$$y_4 = \quad + x_2 \quad \quad + x_4$$

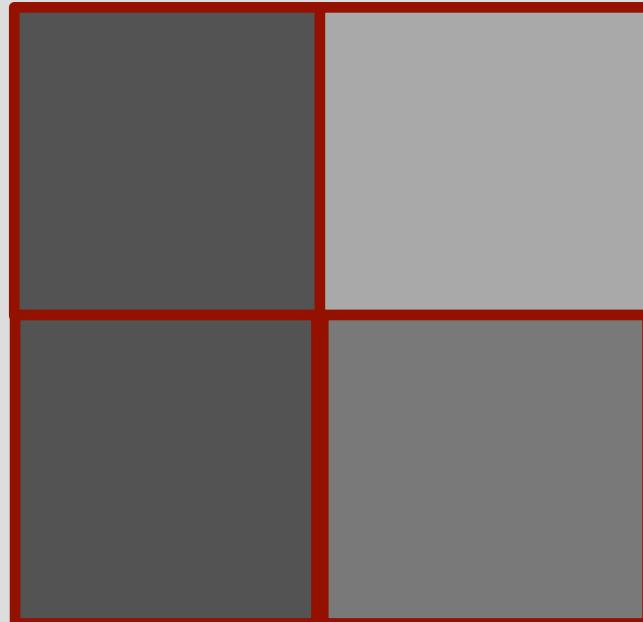
$$y_5 \approx \sqrt{2}x_1 \quad \quad + \sqrt{2}x_4$$

May be able to solve this!

# Modeling Tomography



Possible reconstruction



Blurred version of :



# All our measurements are (converted to) linear

What does that mean?

Each variable ( $x$ ) is multiplied by a scalar to contribute to the measurement

$$y_1 = x_1 + x_2$$

$$y_2 = \qquad \qquad x_3 + x_4$$

$$y_3 = x_1 \qquad + x_3$$

$$y_4 = \qquad + x_2 \qquad + x_4$$

$$y_5 = \sqrt{2}x_1 \qquad \qquad + \sqrt{2}x_4$$

This is called a  
system of linear equations

Linear Algebra is what  
we need to solve it!

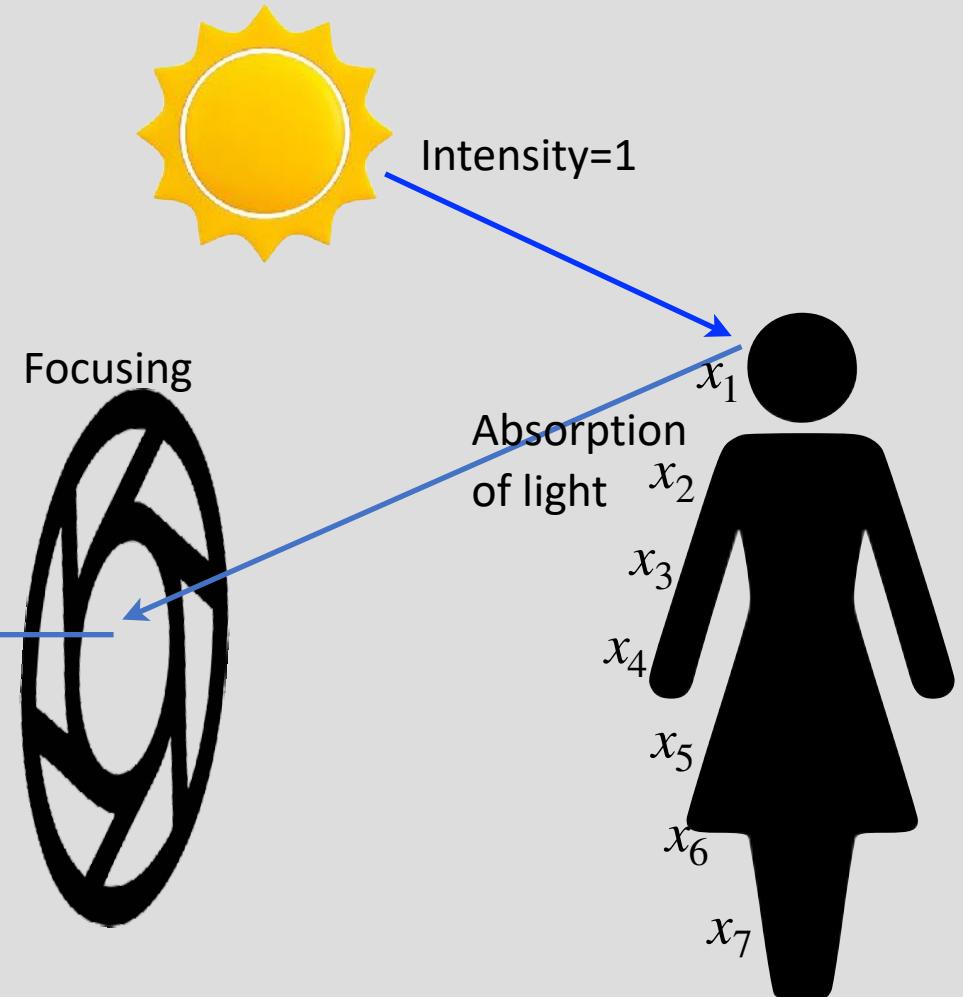
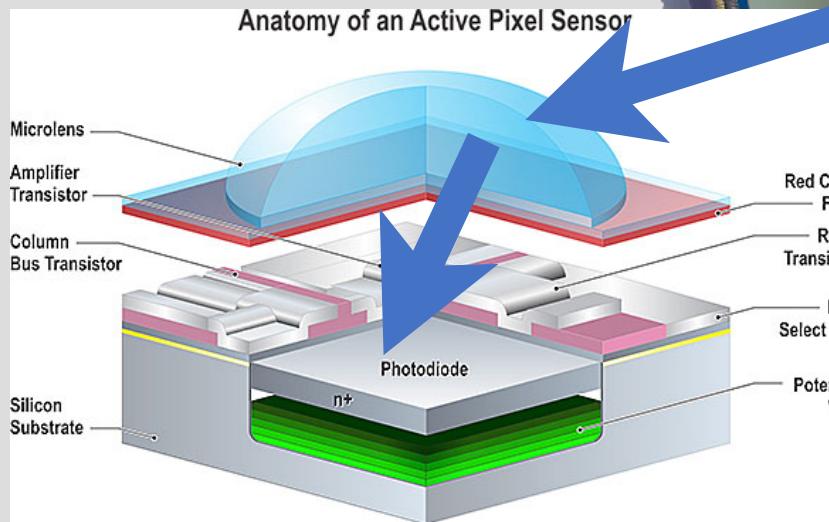
# Camera Model

Lens maps image onto sensor

Each pixel is sensed separately

$$y_i = 1 \cdot x_i$$

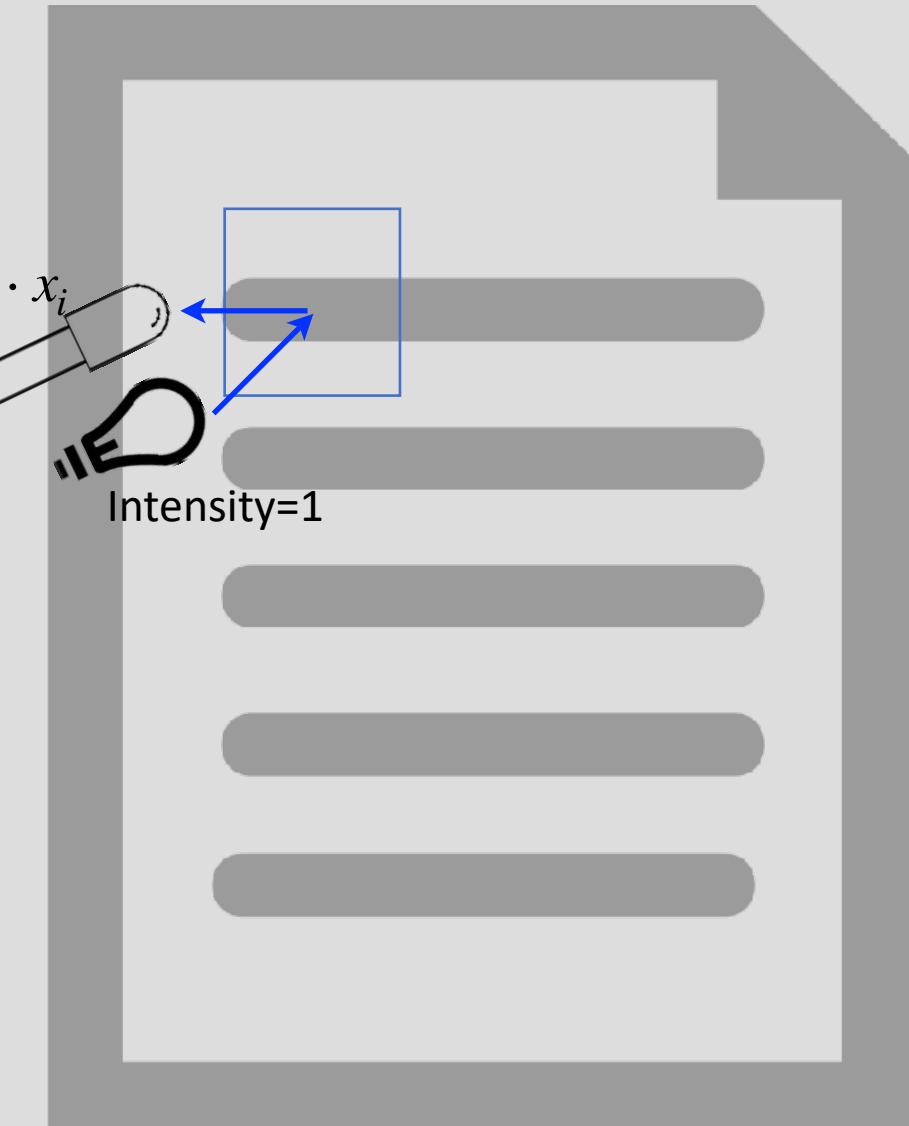
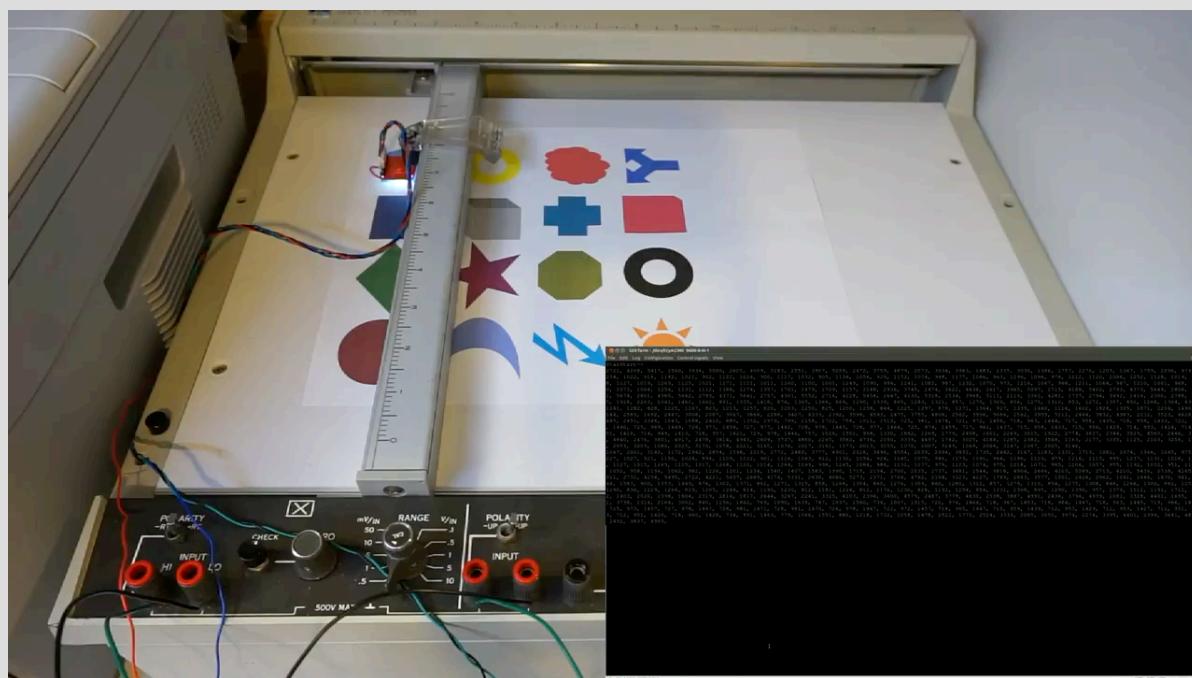
All pixels sensed in parallel



# Single Pixel Scanner

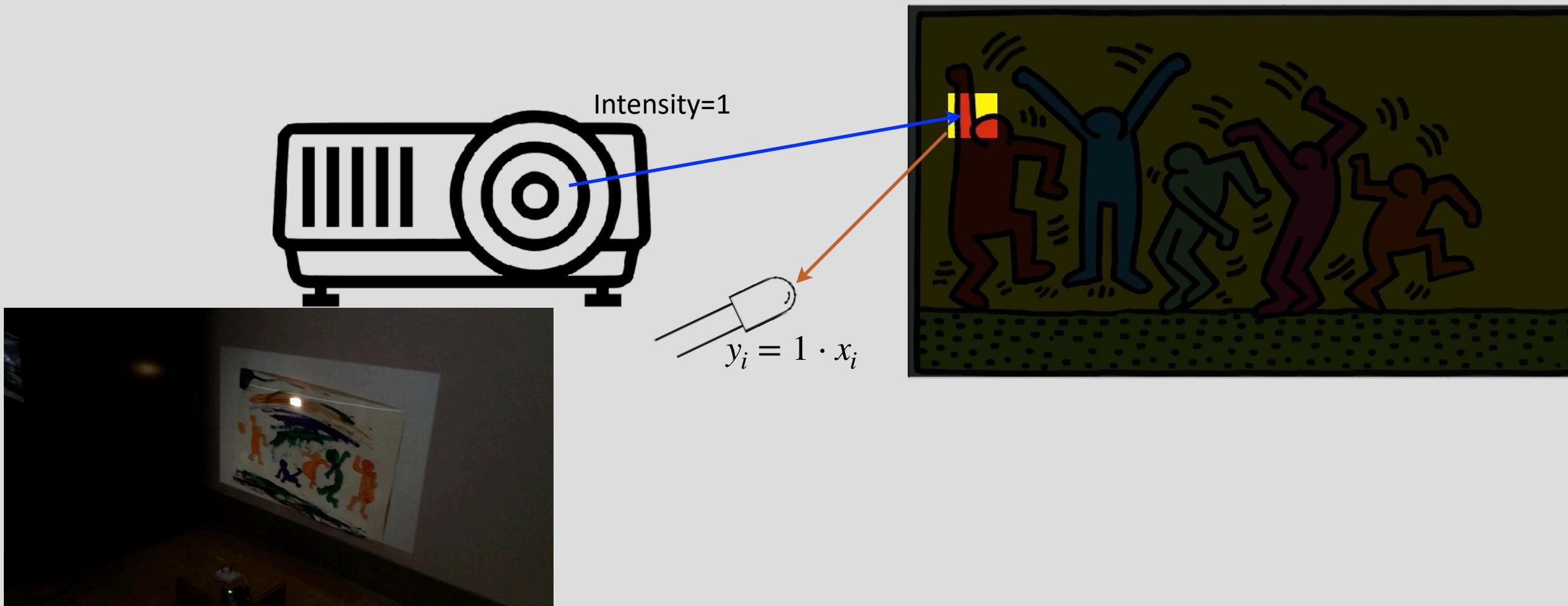
- What if we had only a single sensor?
- How can we create an image?

<https://www.youtube.com/watch?v=U5PwsVqHT8Y>



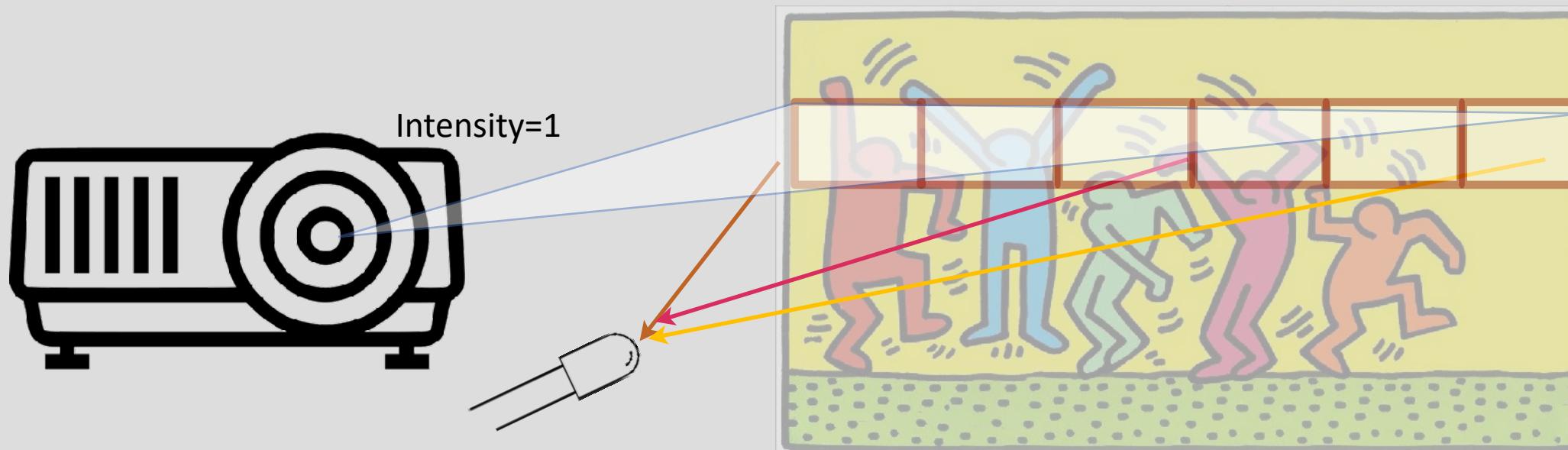
# Non-moving Single Pixel Camera

- Use a projector to illuminate pixels
- Sense reflected light with a sensor



# Non-moving Single Pixel Camera

- Use a projector to illuminate several pixels!
- Sense reflected light with a sensor
- Make many measurements and solve the equations!

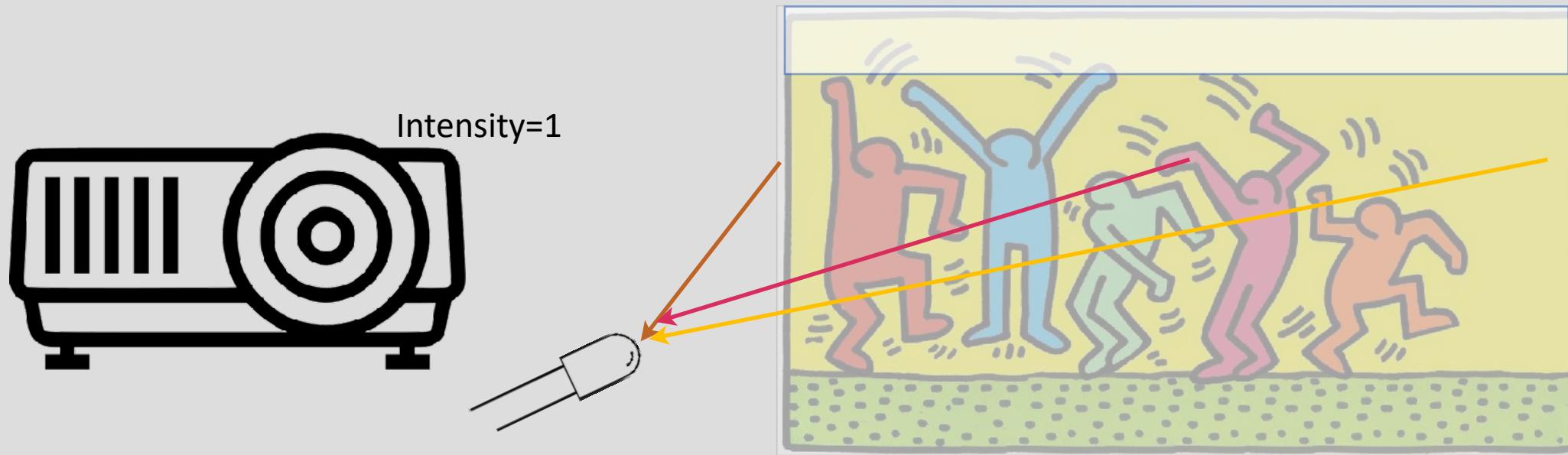


$$y_1 = x_1 + x_2 + x_3 + x_4 + x_5 + x_6$$

Similar math as Tomography!

# Non-moving Single Pixel Camera

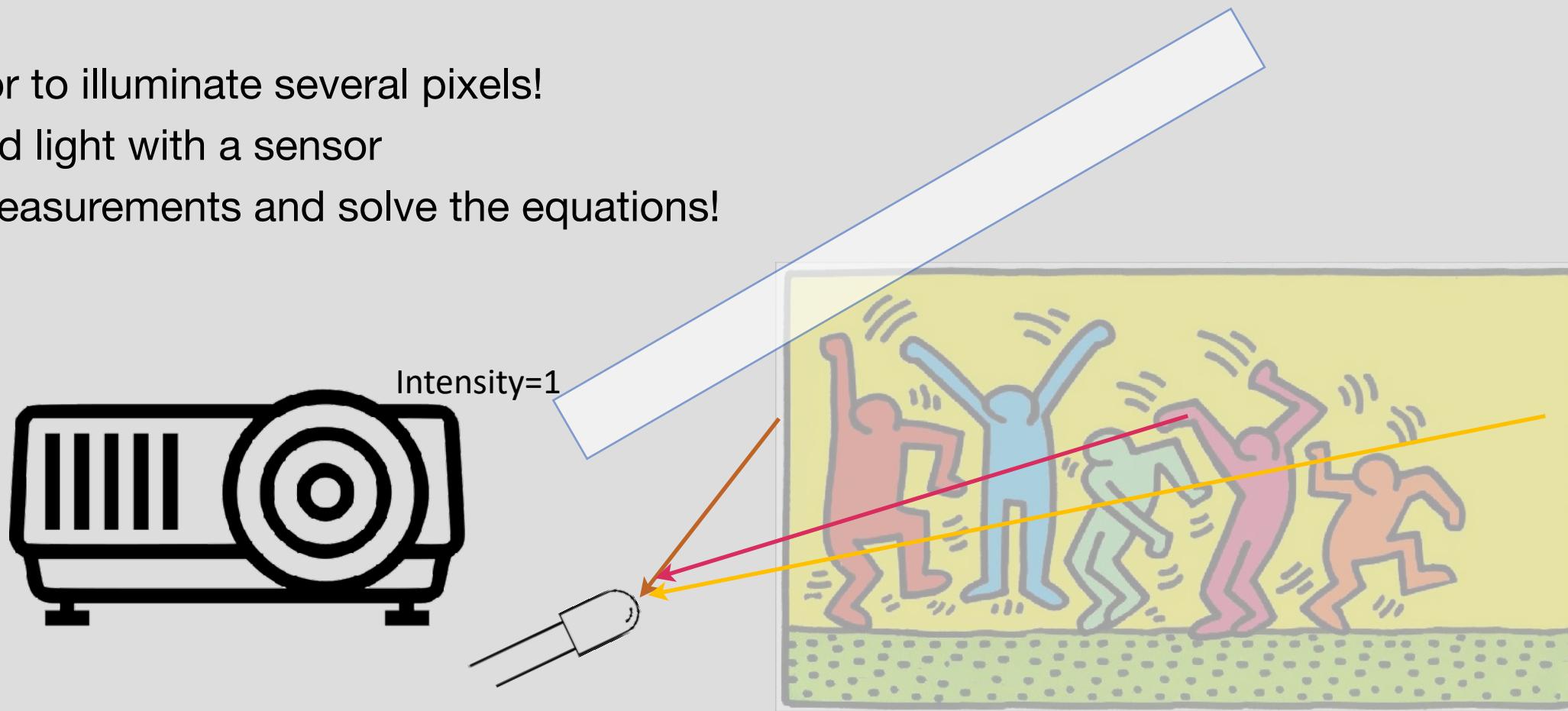
- Use a projector to illuminate several pixels!
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# Non-moving Single Pixel Camera

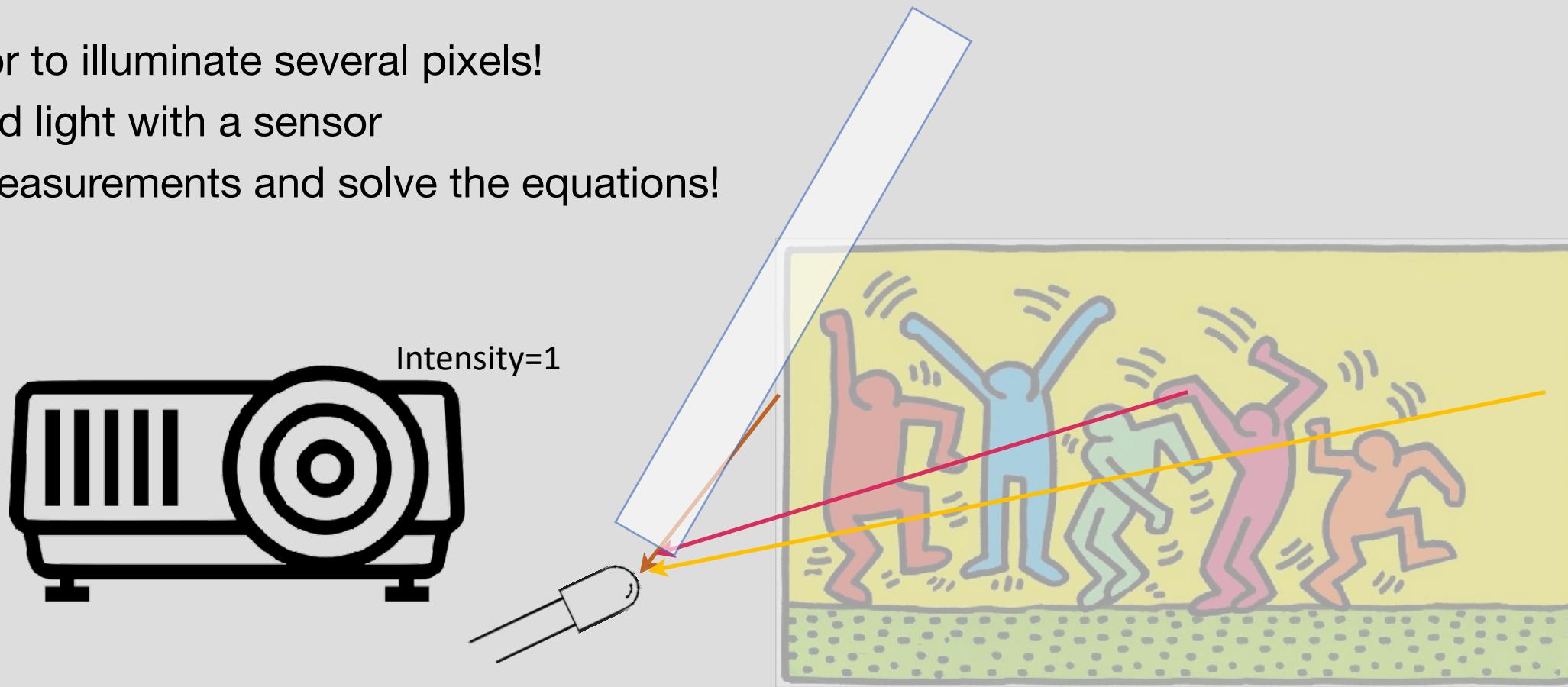
- Use a projector to illuminate several pixels!
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Similar math as Tomography!

# Non-moving Single Pixel Camera

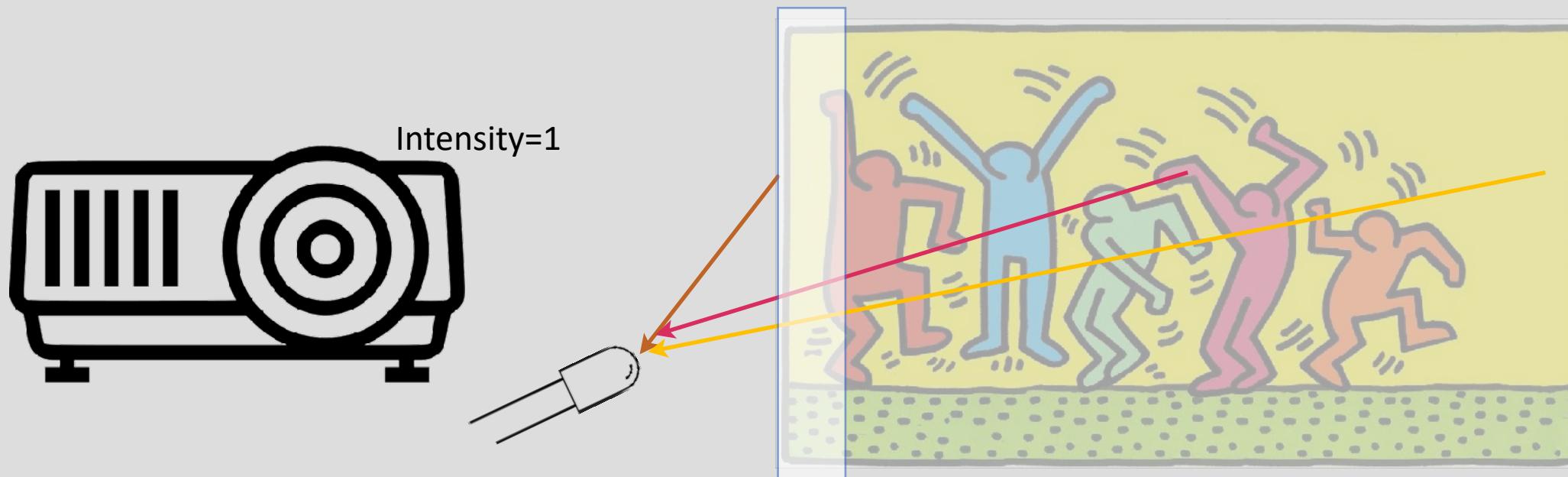
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Similar math as Tomography!

# Non-moving Single Pixel Camera

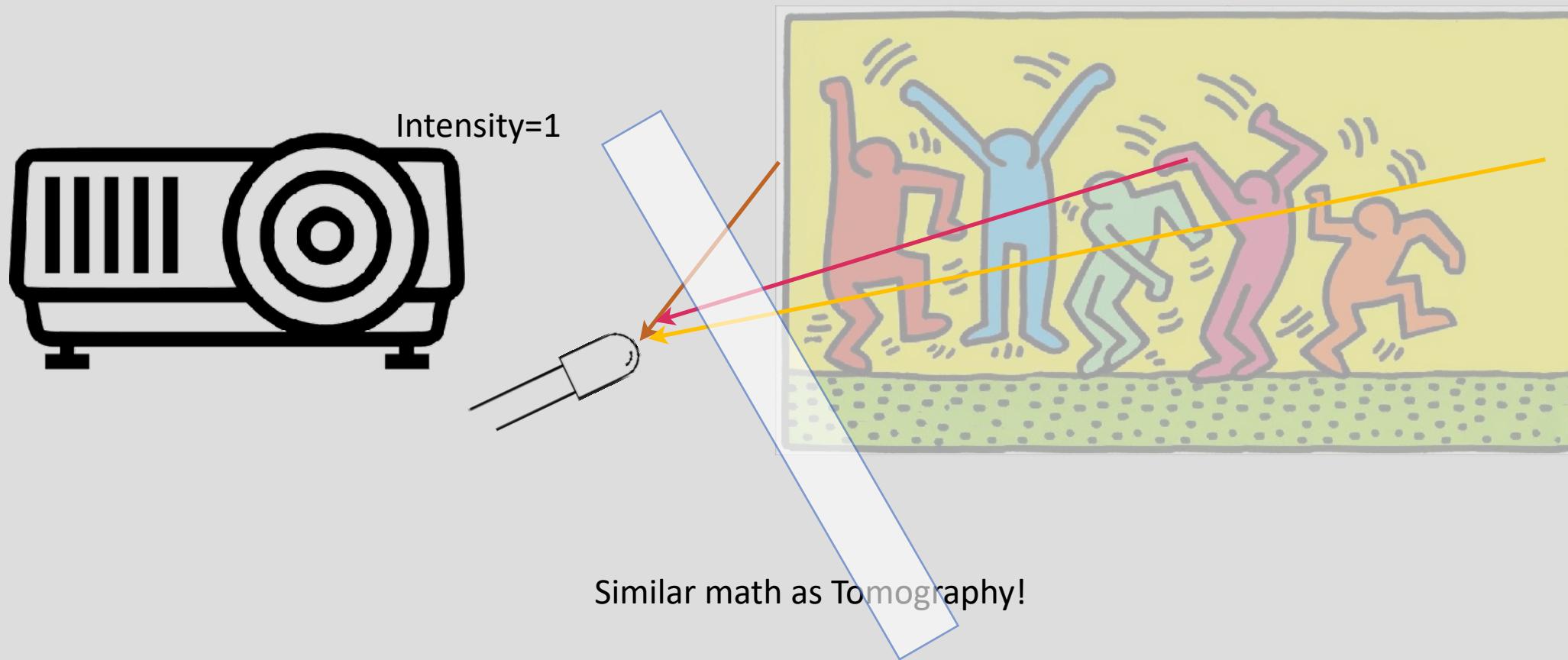
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Similar math as Tomography!

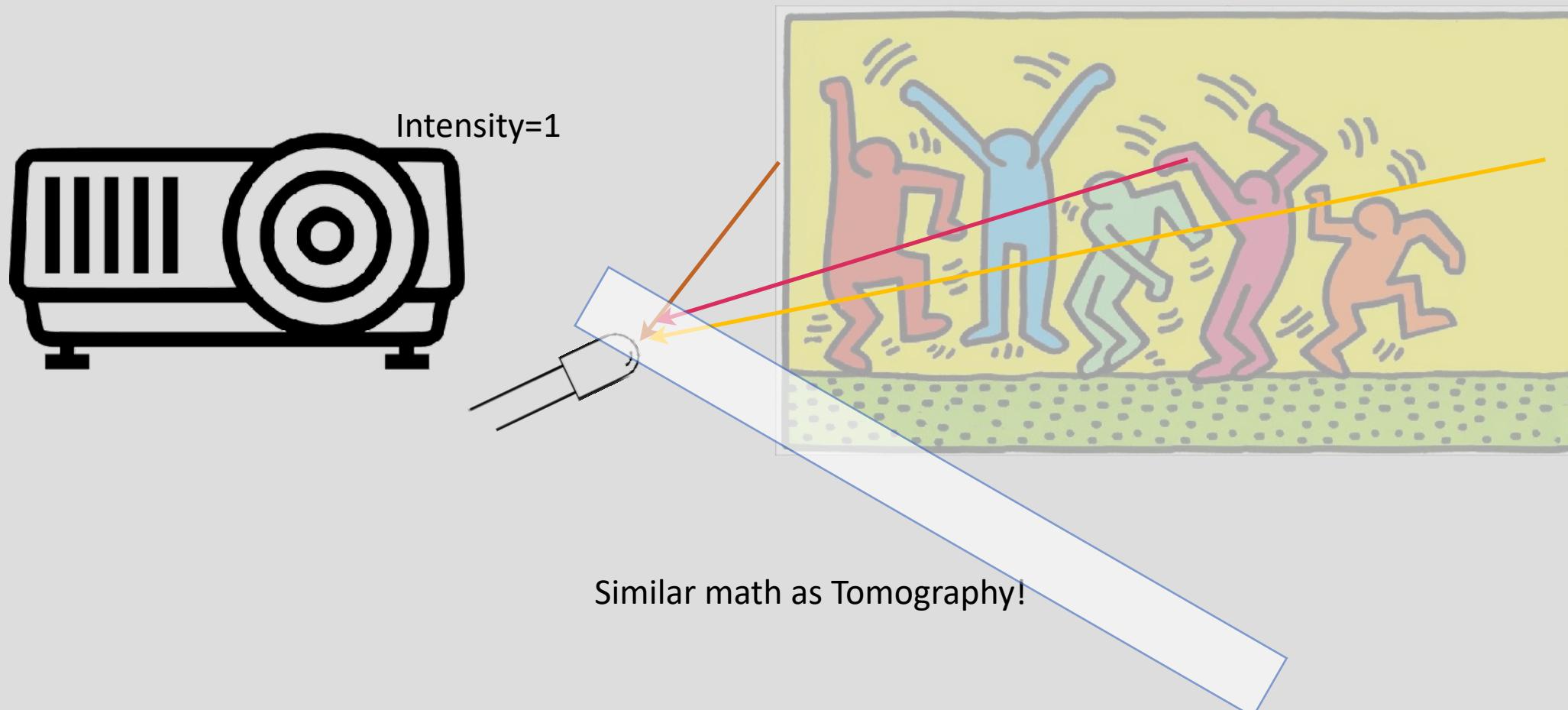
# Non-moving Single Pixel Camera

- Use a projector to illuminate several pixels!
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- Make many measurements and solve the equations!

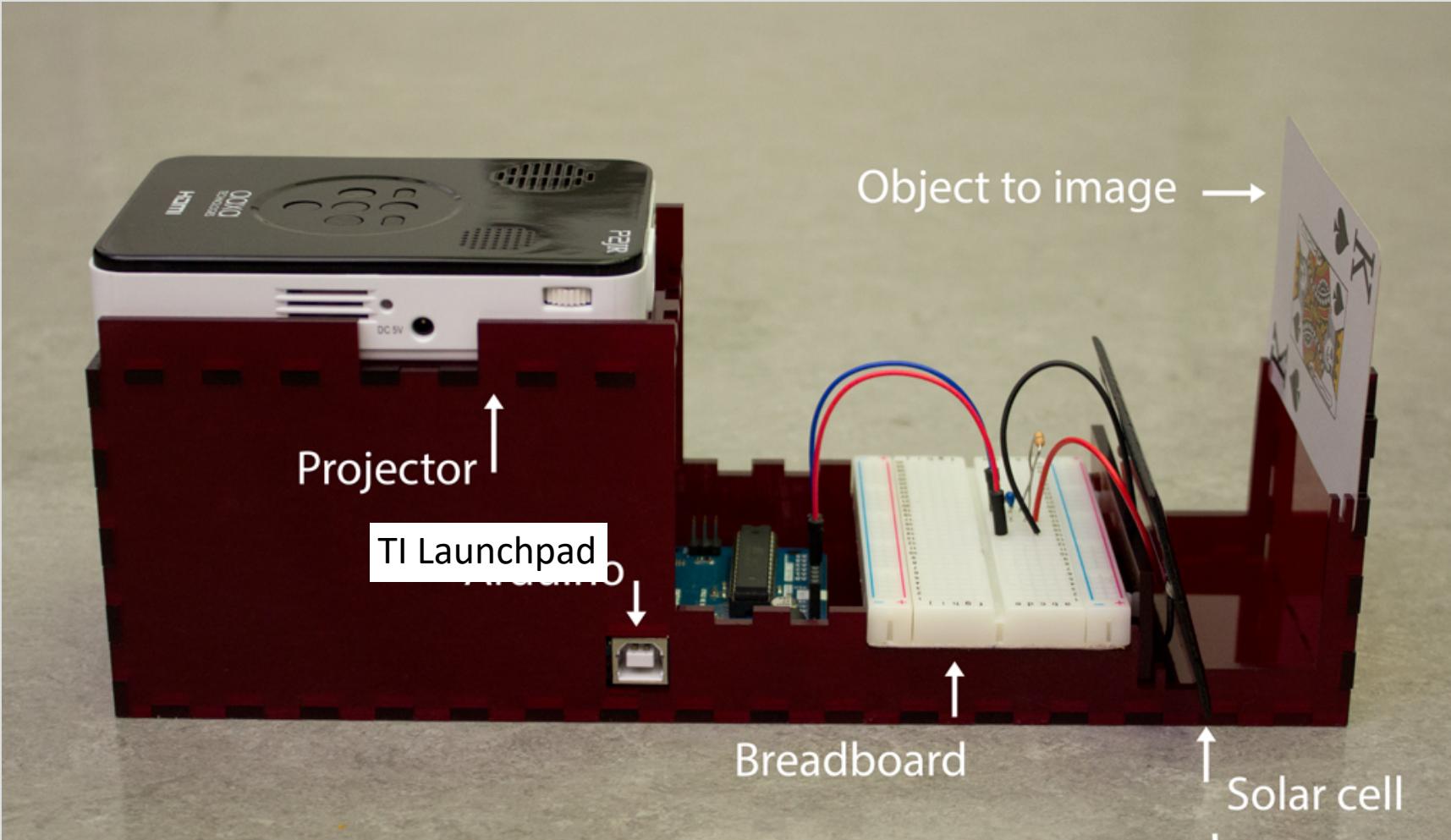


# Non-moving Single Pixel Camera

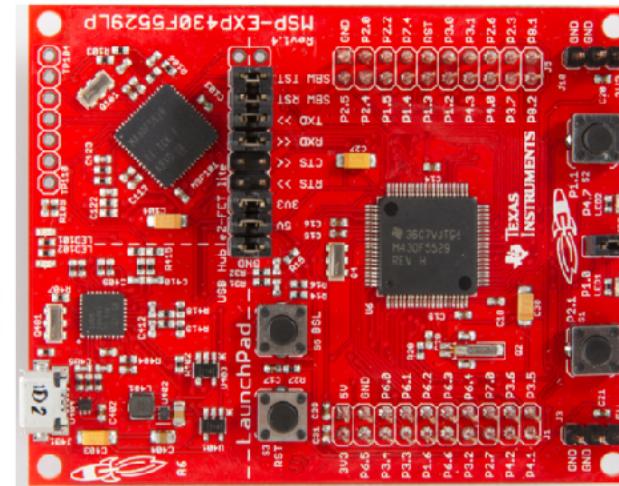
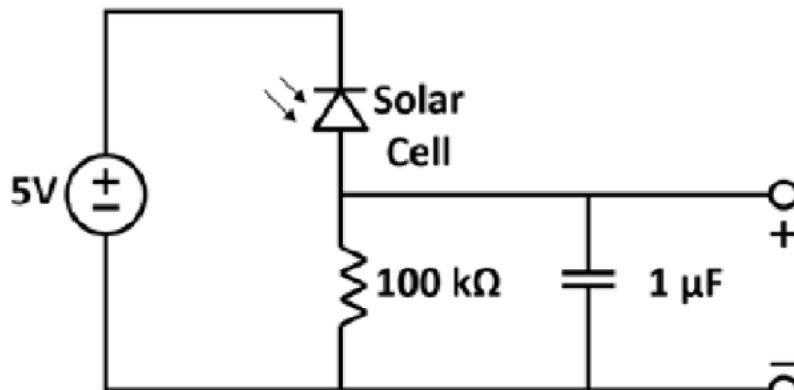
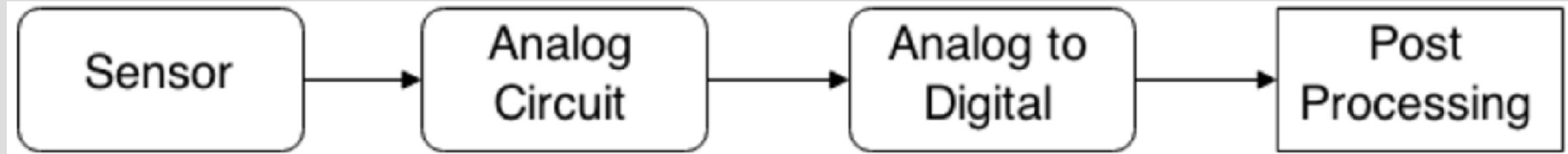
- Use a projector to illuminate several pixels!
- Sense reflected light with a sensor
- Make many measurements and solve the equations!



# Imaging Lab #1 Setup



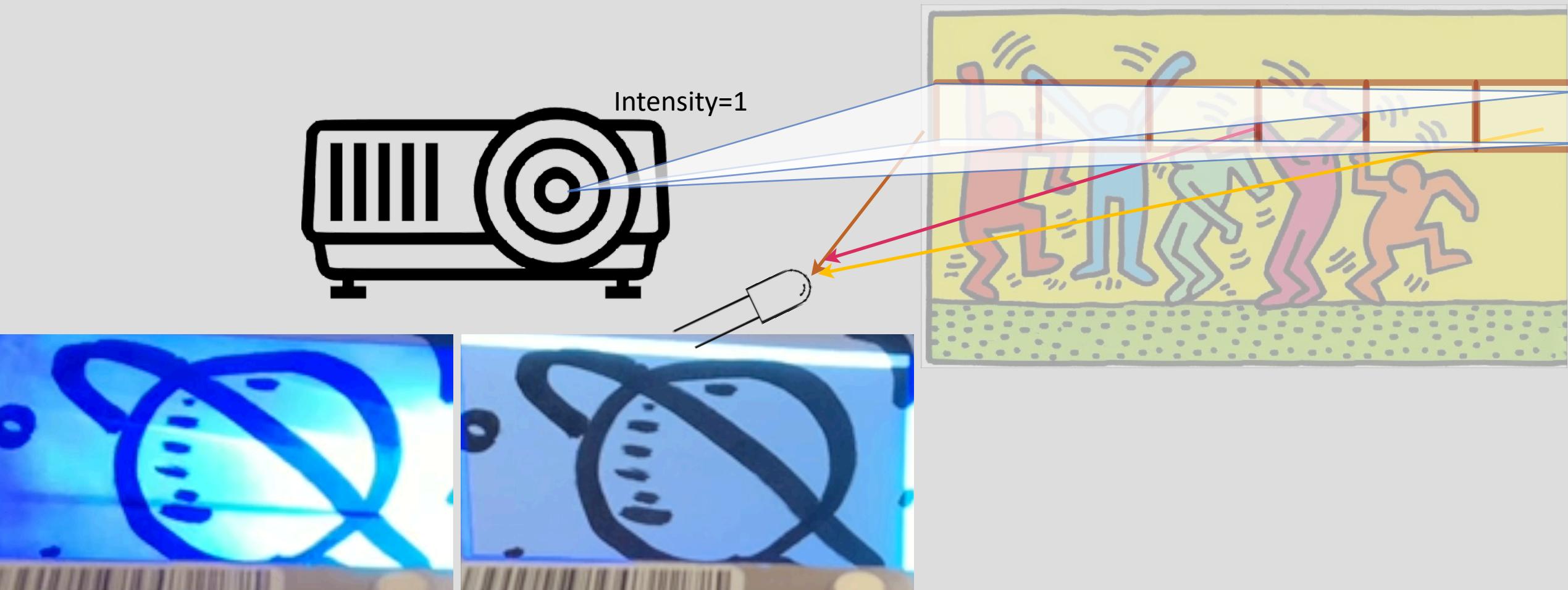
# Imaging Lab #1



# IP[y]: IPython

# Non-moving Single Pixel Camera

- How many measurements do you need?
- What are the best patterns?



# What is linear algebra?

- The study of linear functions and linear equations, typically using vectors and matrices
- Linearity is not always applicable, but can be a good first-order approximation
- There exist good fast algorithms to solve these problems

# Linear Equations

- Definition:

Consider:  $f(x_1, x_2, \dots, x_N) : \mathbb{R}^n \rightarrow \mathbb{R}$

*f is linear if the following identity holds:*

(1) *Homogeneity:*

$$f(\alpha x_1, \dots, \alpha x_N) = \alpha f(x_1, \dots, x_N)$$

(2) *Super Position (distributivity): if  $x_i = y_i + z_i$ , then*

$$f(y_1 + z_1, \dots, y_N + z_N) = f(y_1, \dots, y_N) + f(z_1, \dots, z_N)$$

*Claim: linear functions can always be expressed as:*

$$f(x_1, x_2, \dots, x_N) = c_1 x_1 + c_2 x_2 + \dots + c_N x_N$$

# Proof for $\mathbb{R}^2$

- $f(x_1, x_2) : \mathbb{R}^2 \Rightarrow \mathbb{R}$  is linear. Need to prove:  $f(x_1, x_2) = c_1x_1 + c_2x_2$

*Trick:*

$$x_1 = 1 \cdot x_1 + 0 \cdot x_2$$

$$x_2 = 0 \cdot x_1 + 1 \cdot x_2$$

So,

$$\begin{aligned} f(x_1, x_2) &= f(\cancel{x_1} \cdot 1 + x_2 \cdot 0, \cancel{x_1} \cdot 0 + x_2 \cdot 1) \\ &= f(\cancel{x_1} \cdot 1, \cancel{x_1} \cdot 0) + f(x_2 \cdot 0, x_2 \cdot 1) \\ &= f(\mathbf{x}_1 \cdot 1, \mathbf{x}_1 \cdot 0) + f(\mathbf{x}_2 \cdot 0, \mathbf{x}_2 \cdot 1) \\ &= x_1 f(1, 0) + x_2 f(0, 1) \\ &\quad \text{\scriptsize \textcolor{blue}{\cancel{x_1}} \qquad \qquad \qquad \textcolor{blue}{\cancel{x_2}}} \\ &= c_1 x_1 + c_2 x_2 \end{aligned}$$

# Linear Set of Equations

- Consider the set of M linear equations with N variables:

$$a_{11}x_1 + a_{12}x_2 + \cdots + a_{1N}x_N = b_1$$

$$a_{21}x_1 + a_{22}x_2 + \cdots + a_{2N}x_N = b_2$$

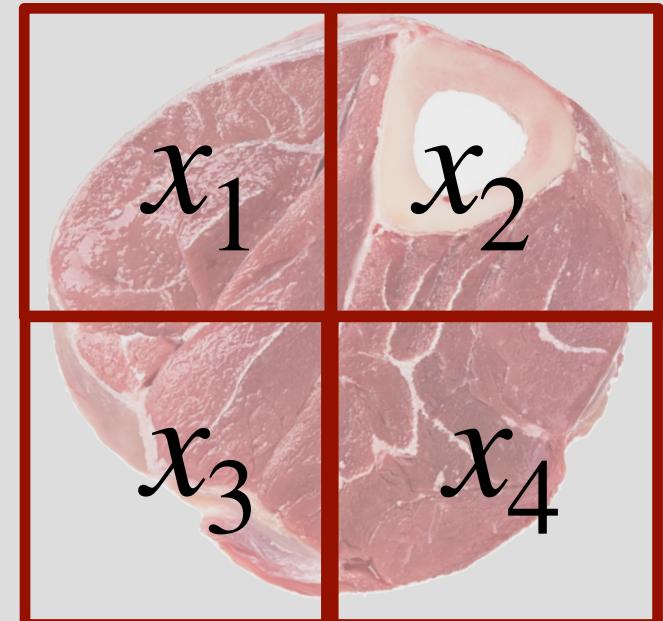
$$\vdots \qquad \vdots$$

$$a_{M1}x_1 + a_{M2}x_2 + \cdots + a_{MN}x_N = b_M$$

- Can be written compactly using augmented matrix:

$$\left[ \begin{array}{cccc|c} a_{11} & a_{12} & \cdots & a_{1N} & b_1 \\ a_{21} & a_{22} & \cdots & a_{2N} & b_2 \\ \vdots & & & \vdots & \vdots \\ a_{M1} & a_{M2} & \cdots & a_{MN} & b_M \end{array} \right]$$

# Back to Tomography



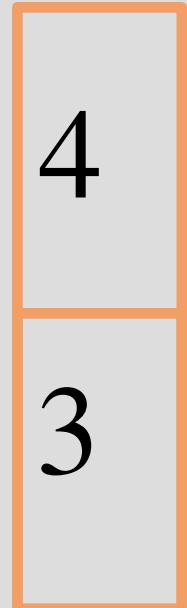
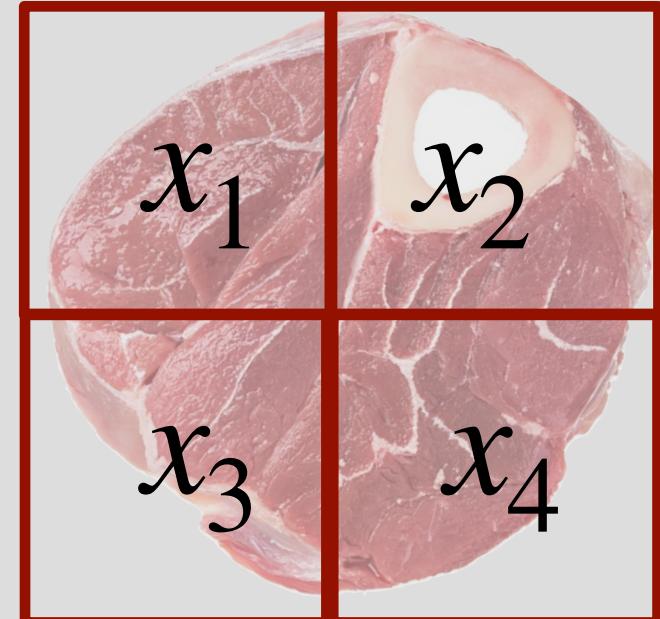
2

5

$$3\sqrt{2}$$

$$\begin{aligned}1 \cdot x_1 + 1 \cdot x_2 + 0 \cdot x_3 + 0 \cdot x_4 &= 4 \\0 \cdot x_1 + 0 \cdot x_2 + 1 \cdot x_3 + 1 \cdot x_4 &= 3 \\1 \cdot x_1 + 0 \cdot x_2 + 1 \cdot x_3 + 0 \cdot x_4 &= 2 \\0 \cdot x_1 + 1 \cdot x_2 + 0 \cdot x_3 + 1 \cdot x_4 &= 5 \\\sqrt{2}x_1 + 0 \cdot x_2 + 0 \cdot x_3 + \sqrt{2}x_4 &= 3\sqrt{2}\end{aligned}$$

# Back to Tomography



$3\sqrt{2}$

How do we solve it?

$$1 \cdot x_1 + 1 \cdot x_2 + 0 \cdot x_3 + 0 \cdot x_4 = 4$$

$$0 \cdot x_1 + 0 \cdot x_2 + 1 \cdot x_3 + 1 \cdot x_4 = 3$$

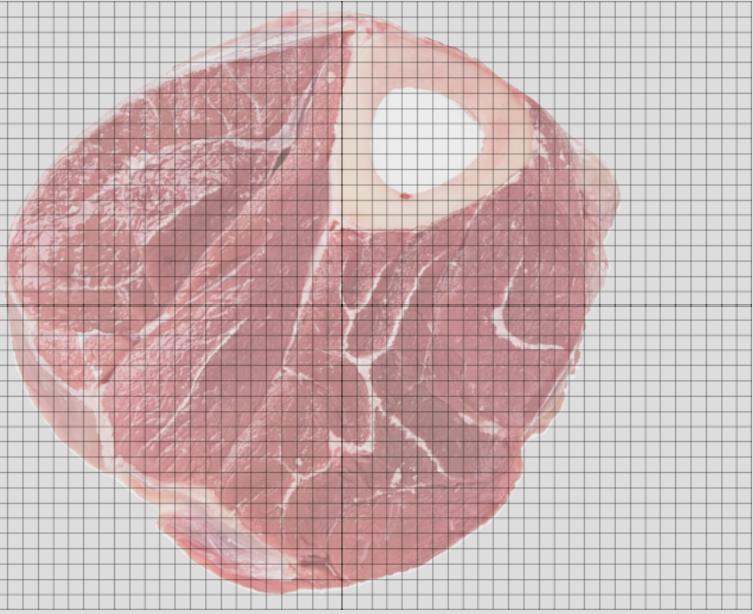
$$1 \cdot x_1 + 0 \cdot x_2 + 1 \cdot x_3 + 0 \cdot x_4 = 2$$

$$0 \cdot x_1 + 1 \cdot x_2 + 0 \cdot x_3 + 1 \cdot x_4 = 5$$

$$\sqrt{2}x_1 + 0 \cdot x_2 + 0 \cdot x_3 + \sqrt{2}x_4 = 3\sqrt{2}$$

$$\left[ \begin{array}{cccc|c} 1 & 1 & 0 & 0 & 4 \\ 0 & 0 & 1 & 1 & 3 \\ 1 & 0 & 1 & 0 & 2 \\ 0 & 1 & 0 & 1 & 5 \\ \hline \sqrt{2} & 0 & 0 & \sqrt{2} & 3\sqrt{2} \end{array} \right]$$

# Back to Tomography



# How do we systematically solve it?

# Algorithm for solving linear equations

- Three basic operations that don't change a solution:
  1. Multiply an equation with *nonzero* scalar
  2. Adding a scalar constant multiple of one equation to another
  3. Swapping equations

# Algorithm for solving linear equations

- Three basic operations that don't change a solution:
  1. Multiply an equation with *nonzero* scalar
  2. Adding a scalar constant multiple of one equation to another
  3. Swapping equations

$$(1) \quad x + y = 2$$

$$(2) \quad 3x + 2y = 5$$

and

$$(1) \quad 3x + 2y = 5$$

$$(2) \quad x + y = 2$$

Have the same solution

Proof: Pretty obvious!

# Algorithm for solving linear equations

- Three basic operations that don't change a solution:

## 1. Multiply an equation with *nonzero* scalar

$$2x + 3y = 4 \text{ has the same solution as: } 4x + 6y = 8$$

Proof for N=2:

Let  $ax + by = c$ , with solution  $x_0, y_0$   
 $\Rightarrow ax_0 + by_0 = c$

Show that  $\beta ax + \beta by = \beta c$ ,  
has the same solution.

Substitute  $x_0, y_0$  for  $x, y$ :

$$\beta ax_0 + \beta by_0 = \beta c$$

$$\beta(ax_0 + by_0) = \beta c$$

$$\beta c = \beta c \quad \text{But is it the only solution?}$$

$\beta ax + \beta by = \beta c$ , with solution:  $x_1, y_1$   
 $\Rightarrow \beta ax_1 + \beta by_1 = \beta c$

Show that  $ax + by = c$ ,  
has the same solution.....

Since  $\beta \neq 0$ ....

$$\beta ax_1 + \beta by_1 = \beta c \Rightarrow ax_1 + by_1 = c$$

SOLUTION OF ONE, IMPLIES THE OTHER  
AND VICE-VERSA!