# EECS 16A Designing Information Devices and Systems I Spring 2015 Note 5

#### Lecture notes by Dasol Yoon (02/03/2015)

### Imaging

#### **Examples of imaging:**

- Get readings from the brain, then construct an image from the information
- Milk, Juice, and Empty bottles arranged in a 3x3 carton.
  - Example carton:

(See Figure 1)

If we shine light through the carton, we can try to figure out what types of bottles are in that row or column. A Milk bottle adds 3 units, a Juice bottle adds 2 units, and an Empty bottle adds 1 unit. So in our example carton, we would have 5 units total for the first row.

(See Figure 2)

- We have a closed carton, and we want to figure out what types of bottles are inside. A way that we can do this is by shining light on each row, column, and diagonal.

(See Figure 3)

- To solve this, we just combine these equations:

$$rows: \begin{cases} x_{1,1} + x_{1,2} + x_{1,3} = 5\\ x_{2,1} + x_{2,2} + x_{2,3} = 6\\ x_{3,1} + x_{3,2} + x_{3,3} = 4 \end{cases} \quad columns: \begin{cases} x_{1,1} + x_{2,1} + x_{3,1} = 6\\ x_{1,2} + x_{2,2} + x_{3,2} = 3\\ x_{1,3} + x_{2,3} + x_{3,3} = 6 \end{cases} \quad diagonals: \begin{cases} x_{1,1} + x_{2,2} + x_{3,3} = 5\\ x_{1,3} + x_{2,2} + x_{3,1} = 5 \end{cases}$$

to get the result.

(See Figure 4)

We have a 9x9 matrix that we want to image. We can take the following steps:

- Shine light onto the image
- Collect reflected light
- Collect measurement that says whether the pixel is black or white

Μ	J	0		
Μ	J	0		
Μ	0	J		

Figure 1: An example carton.



Figure 2: Shining a light through row 1 of the example carton.



Figure 3: Shining lights through our mystery carton.

Μ	0	0		
J	0	Μ		
0	0	J		

Figure 4: Our solved carton.

• Ideally, we would shine light on every pixel:

/ 1	0	0	0	0	0	0	0	0 \	$(x_1)$		$(x_1)$
0	1	0	0	0	0	0	0	0	<i>x</i> <sub>2</sub>		<i>x</i> <sub>2</sub>
0	0	1	0	0	0	0	0	0	<i>x</i> <sub>3</sub>		<i>x</i> <sub>3</sub>
0	0	0	1	0	0	0	0	0	<i>x</i> <sub>4</sub>		<i>x</i> 4
0	0	0	0	1	0	0	0	0	<i>x</i> <sub>5</sub>	=	<i>x</i> 5
0	0	0	0	0	1	0	0	0	<i>x</i> <sub>6</sub>		<i>x</i> <sub>6</sub>
0	0	0	0	0	0	1	0	0	<i>x</i> <sub>7</sub>		<i>x</i> <sub>7</sub>
0	0	0	0	0	0	0	1	0	<i>x</i> <sub>8</sub>		<i>x</i> <sub>8</sub>
0	0	0	0	0	0	0	0	1/	$\left( x_{9} \right)$		$\left( x_{9} \right)$

This is like the action of taking a picture and representing it with the matrix multiplication

• What if the photodetector had more constraints?

1	1	1	0	0	0	0	0	0	0 \	$(x_1)$		$(x_1+x_2)$
	0	1	1	0	0	0	0	0	0	<i>x</i> <sub>2</sub>		$x_2 + x_3$
	0	0	1	1	0	0	0	0	0	<i>x</i> <sub>3</sub>		$x_3 + x_4$
	0	0	0	1	1	0	0	0	0	<i>x</i> <sub>4</sub>		$x_4 + x_5$
	0	0	0	0	1	1	0	0	0	<i>x</i> 5	=	$x_5 + x_6$
	0	0	0	0	0	1	1	0	0	<i>x</i> <sub>6</sub>		$x_6 + x_7$
	0	0	0	0	0	0	1	1	0	<i>x</i> <sub>7</sub>		$x_7 + x_8$
	0	0	0	0	0	0	0	1	1	$x_8$		$x_8 + x_9$
/	0	0	0	0	0	0	0	0	1 /	$\left( x_{9} \right)$		$\begin{pmatrix} x_9 \end{pmatrix}$

We do not want to multiply by an 8x9 matrix because we could be losing information!

### Linear Equations

$$\begin{pmatrix} a_{1,1} & a_{1,2} & a_{1,3} \\ a_{2,1} & a_{2,2} & a_{2,3} \\ a_{3,1} & a_{3,2} & a_{3,3} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix}$$
 We want to try to get an invertible matrix.

 $a_{1,1}x_1 + a_{1,2}x_2 + a_{1,3}x_3 = b_1 \qquad a_{2,1}x_1 + a_{2,2}x_2 + a_{2,3}x_3 = b_2 \qquad a_{3,1}x_1 + a_{3,2}x_2 + a_{3,3}x_3 = b_3$ (1)

$$\begin{pmatrix} 0.5 & 0.5 & 0 & 0 \\ 0 & 0.5 & 0.5 & 0 \\ 0 & 0 & 0.5 & 0.5 \\ 0 & 0 & 0 & 0.5 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 0.5 \\ 0.5 \\ 0.5 \\ 0.5 \end{pmatrix}$$
This would produce an all – gray bar.  

$$\begin{pmatrix} 0.9 & 0.1 & 0 & 0 \\ 0 & 0.9 & 0.1 & 0 \\ 0 & 0 & 0.9 & 0.1 \\ 0 & 0 & 0 & 0.9 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 0.1 \\ 0.9 \\ 0.1 \\ 0.9 \end{pmatrix}$$
This would be closer to a gradient.

## Vector Spaces

A **Basis** is the minimum number of representations of the vectors in a vector space. A **vector space**  $(\mathbb{V}, \mathbb{F})$  is a set of vectors  $\mathbb{V}$ , a set of scalars  $\mathbb{F}$ , and two operators:

- Vector Addition
  - Associative:  $\vec{u} + (\vec{v} + \vec{w}) = (\vec{u} + \vec{v}) + \vec{w}$
  - Commutative:  $\vec{u} + \vec{v} = \vec{v} + \vec{u}$
  - Additive Identity:  $\vec{0}$  such that  $\vec{v} + \vec{0} = \vec{v}$
  - Additive Inverse:  $\vec{v}$ ,  $\vec{w}$  such that  $\vec{v} + \vec{w} = 0$
- Scalar Multiplication
  - Associative:  $\vec{u} \cdot (\vec{v} \cdot \vec{w}) = (\vec{u} \cdot \vec{v}) \cdot \vec{w}$
  - Commutative:  $(\lambda \cdot \vec{u}) \cdot \vec{v} = \lambda \cdot (\vec{u} \cdot \vec{v})$
  - Multiplicative Identity:  $1 \cdot \vec{v} = \vec{v}$
  - Zero:  $\vec{0} \cdot \vec{v} = \vec{0}$

Distributive Law: Scalar multiplication will distribute across vector addition.