

# Interpreting Linear Equations

## Broad categories of applications

linear model or function  $y = Ax$

some broad categories of applications:

- ▶ estimation or inversion
- ▶ control or design
- ▶ mapping or transformation

(this list is not exclusive; can have combinations . . .)

## Estimation or inversion

$$y = Ax$$

- ▶  $y_i$  is  $i$ th measurement or sensor reading (which we know)
- ▶  $x_j$  is  $j$ th parameter to be estimated or determined
- ▶  $a_{ij}$  is sensitivity of  $i$ th sensor to  $j$ th parameter

sample problems:

- ▶ find  $x$ , given  $y$
- ▶ find all  $x$ 's that result in  $y$  (*i.e.*, all  $x$ 's consistent with measurements)
- ▶ if there is no  $x$  such that  $y = Ax$ , find  $x$  s.t.  $y \approx Ax$  (*i.e.*, if the sensor readings are inconsistent, find  $x$  which is almost consistent)

## Control or design

$$y = Ax$$

- ▶  $x$  is vector of design parameters or inputs (which we can choose)
- ▶  $y$  is vector of results, or outcomes
- ▶  $A$  describes how input choices affect results

sample problems:

- ▶ find  $x$  so that  $y = y_{\text{des}}$
- ▶ find all  $x$ 's that result in  $y = y_{\text{des}}$  (*i.e.*, find all designs that meet specifications)
- ▶ among  $x$ 's that satisfy  $y = y_{\text{des}}$ , find a small one (*i.e.*, find a small or efficient  $x$  that meets specifications)

## Mapping or transformation

- ▶  $x$  is mapped or transformed to  $y$  by linear function  $y = Ax$

sample problems:

- ▶ determine if there is an  $x$  that maps to a given  $y$
- ▶ (if possible) find *an*  $x$  that maps to  $y$
- ▶ find *all*  $x$ 's that map to a given  $y$
- ▶ if there is only one  $x$  that maps to  $y$ , find it (*i.e.*, decode or undo the mapping)

## Matrix multiplication as mixture of columns

write  $A \in \mathbb{R}^{m \times n}$  in terms of its columns

$$A = [ a_1 \quad a_2 \quad \cdots \quad a_n ]$$

where  $a_j \in \mathbb{R}^m$ . Then then  $y = Ax$  means

$$y = x_1 a_1 + x_2 a_2 + \cdots + x_n a_n$$

( $x_j$ 's are scalars,  $a_j$ 's are  $m$ -vectors)

- ▶  $y$  is a (linear) combination or mixture of the columns of  $A$
- ▶ coefficients of  $x$  give coefficients of mixture
- ▶ each column of  $A$  represents an *actuator*

## Geometric interpretation of control

example:  $A = \begin{bmatrix} 1 & -1 \\ 2 & 1 \end{bmatrix}$ ,  $x = \begin{bmatrix} 1 \\ -0.5 \end{bmatrix}$ ,

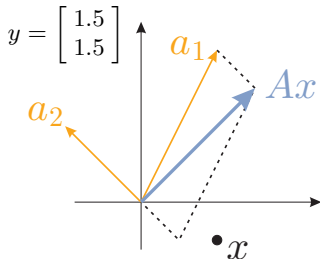
$$Ax = a_1 + (-0.5)a_2 = \begin{bmatrix} 1.5 \\ 1.5 \end{bmatrix}$$

another example:

$$a_j = Ae_j$$

where  $e_j$  is the  $j$ th unit vector:

$$e_1 = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \quad e_2 = \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}, \dots, \quad e_n = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ n \end{bmatrix}$$



- ▶  $j$ th column of  $A$  gives response to unit  $j$ th input

## Matrix multiplication as inner product with rows

write  $A$  in terms of its rows:

$$A = \begin{bmatrix} \tilde{a}_1^\top \\ \tilde{a}_2^\top \\ \vdots \\ \tilde{a}_m^\top \end{bmatrix}$$

where  $\tilde{a}_i \in \mathbb{R}^n$

then  $y = Ax$  can be written as

$$y = \begin{bmatrix} \tilde{a}_1^\top x \\ \tilde{a}_2^\top x \\ \vdots \\ \tilde{a}_m^\top x \end{bmatrix}$$

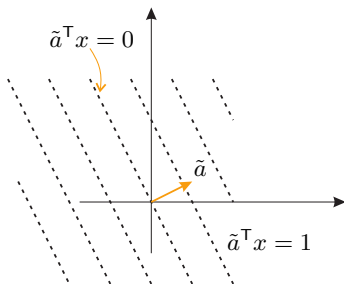
- ▶  $y_i = \tilde{a}_i^\top x$ , so that  $y_i$  is inner product of  $i$ th row of  $A$  with  $x$
- ▶ each row of  $A$  represents a *sensor*



## Geometric interpretation of estimation

$$a_i^T x = \text{constant}$$

is a (hyper-)plane in  $\mathbb{R}^n$  normal to  $a_i$ .

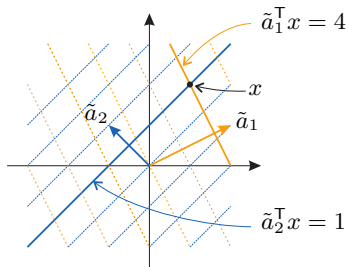


if  $Ax = y$  then  $x$  is on intersection of hyperplanes  $a_i^T x = y_i$

$$A = \begin{bmatrix} 2 & 1 \\ -1 & 1 \end{bmatrix}$$

$$x = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

$$y = \begin{bmatrix} 4 \\ 1 \end{bmatrix}$$

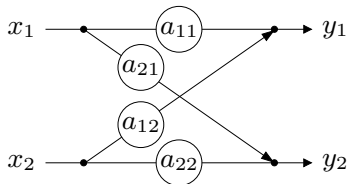


## Block diagram representation

$y = Ax$  can be represented by a *signal flow graph* or *block diagram* e.g. for  $m = n = 2$ , we represent

$$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

as



- ▶  $a_{ij}$  is the gain along the path from  $j$ th input to  $i$ th output
- ▶ (by not drawing paths with zero gain) shows sparsity structure of  $A$  (e.g., diagonal, block upper triangular, arrow ...)

## Example: block upper triangular matrices

$$A = \begin{bmatrix} A_{11} & A_{12} \\ 0 & A_{22} \end{bmatrix}$$

where  $A_{11} \in \mathbb{R}^{m_1 \times n_1}$ ,  $A_{12} \in \mathbb{R}^{m_1 \times n_2}$ ,  $A_{21} \in \mathbb{R}^{m_2 \times n_1}$ ,  $A_{22} \in \mathbb{R}^{m_2 \times n_2}$

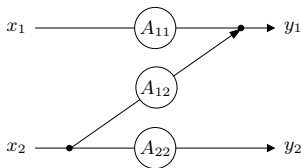
partition  $x$  and  $y$  conformably, (so that  $x_1 \in \mathbb{R}^{n_1}$ ,  $x_2 \in \mathbb{R}^{n_2}$ ,  $y_1 \in \mathbb{R}^{m_1}$ ,  $y_2 \in \mathbb{R}^{m_2}$ )

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}, \quad y = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}$$

then

$$y_1 = A_{11}x_1 + A_{12}x_2$$

$$y_2 = A_{22}x_2,$$



... no path from  $x_1$  to  $y_2$ , so  $y_2$  doesn't depend on  $x_1$

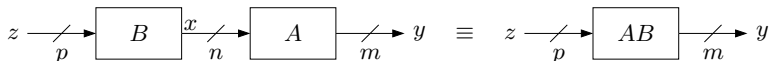
## Matrix multiplication as composition

for  $A \in \mathbb{R}^{m \times n}$  and  $B \in \mathbb{R}^{n \times p}$ ,  $C = AB \in \mathbb{R}^{m \times p}$  where

$$c_{ij} = \sum_{k=1}^n a_{ik} b_{kj}$$

*composition interpretation*

$y = Cz$  represents composition of  $y = Ax$  and  $x = Bz$



(note that  $B$  is on left in block diagram)

## Column and row interpretations

can write product  $C = AB$  as

$$C = [c_1 \quad \cdots \quad c_p] = AB = [Ab_1 \quad \cdots \quad Ab_p]$$

*i.e.*,  $i$ th column of  $C$  is  $A$  acting on  $i$ th column of  $B$

similarly we can write

$$C = \begin{bmatrix} \tilde{c}_1^\top \\ \vdots \\ \tilde{c}_m^\top \end{bmatrix} = AB = \begin{bmatrix} \tilde{a}_1^\top B \\ \vdots \\ \tilde{a}_m^\top B \end{bmatrix}$$

*i.e.*,  $i$ th row of  $C$  is  $i$ th row of  $A$  acting (on left) on  $B$

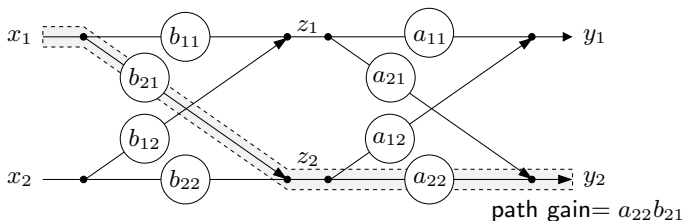
## Inner product interpretation

$$c_{ij} = \tilde{a}_i^\top b_j = \langle \tilde{a}_i, b_j \rangle$$

*i.e.*, entries of  $C$  are inner products of rows of  $A$  and columns of  $B$

- ▶  $c_{ij} = 0$  means  $i$ th row of  $A$  is orthogonal to  $j$ th column of  $B$
- ▶ **Gram matrix** of vectors  $f_1, \dots, f_n$  defined as  $G_{ij} = f_i^\top f_j$   
(gives inner product of each vector with the others)
- ▶  $G = [f_1 \ \cdots \ f_n]^\top [f_1 \ \cdots \ f_n]$

## Matrix multiplication interpretation via paths



- ▶  $a_{ik}b_{kj}$  is gain of path from input  $j$  to output  $i$  via  $k$
- ▶  $c_{ij}$  is sum of gains over **all** paths from input  $j$  to output  $i$