

Now the big moment: matrix, vector
multiplication

$$2a - 3b + 5c = 3$$

$$-a + b + 2c = 5$$

$$a \begin{bmatrix} 2 \\ -1 \end{bmatrix} + b \begin{bmatrix} -3 \\ 1 \end{bmatrix} + c \begin{bmatrix} 5 \\ 2 \end{bmatrix} = \begin{bmatrix} 3 \\ 5 \end{bmatrix}$$

↓

$$\begin{bmatrix} 2 & -3 & 5 \\ -1 & 1 & 2 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 3 \\ 5 \end{bmatrix}$$

$$Ax = b$$

must match

$$m \begin{bmatrix} \\ \\ \end{bmatrix} \begin{bmatrix} \\ \\ \end{bmatrix} n$$

A couple of perspectives

1) column perspective

$$\begin{matrix} [a_1 \dots a_k] \\ A \end{matrix} \begin{matrix} \begin{bmatrix} x_1 \\ \vdots \\ x_k \end{bmatrix} \\ x \end{matrix} = x_1 a_1 + \dots + x_k a_k$$

"linear combination of columns of A , where coefficients come from x "

$$\begin{bmatrix} 2 & -3 & 5 \\ -1 & 1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ -1 \end{bmatrix} + 2 \begin{bmatrix} -3 \\ 1 \end{bmatrix} + \begin{bmatrix} 5 \\ 2 \end{bmatrix} = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$$

2) Row perspective

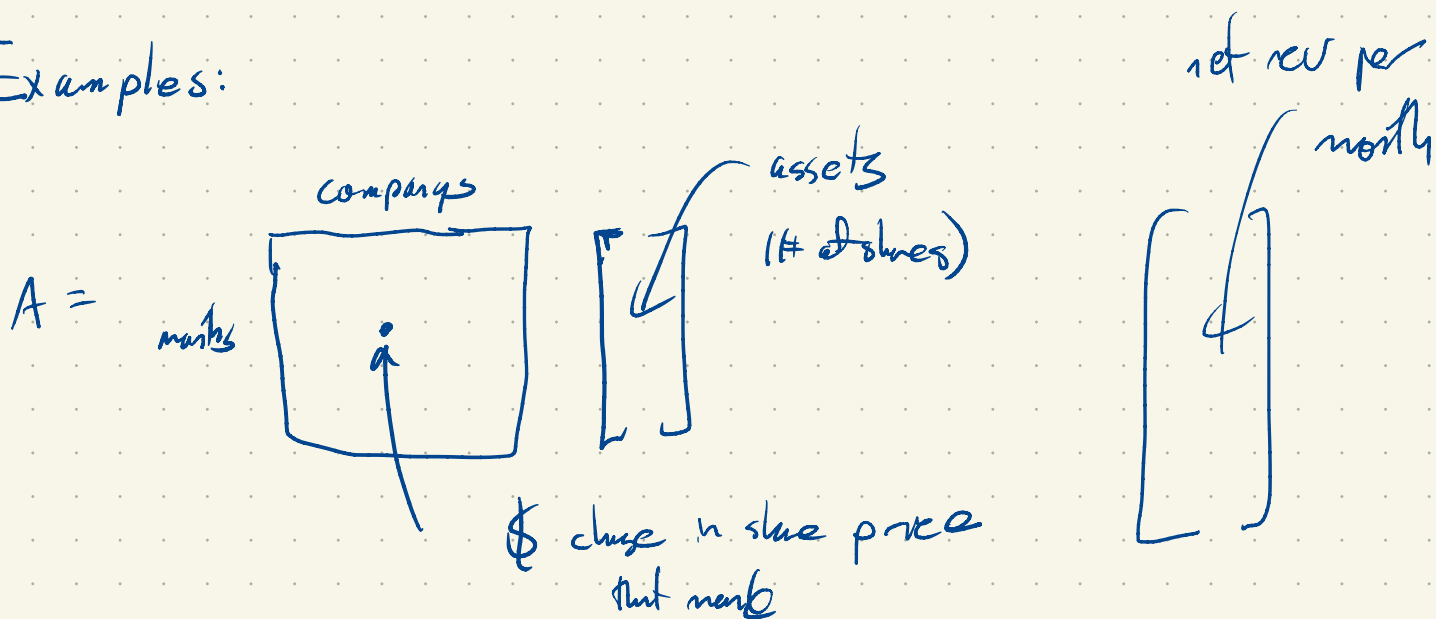
$$Ax = \begin{bmatrix} b_1^T \\ \vdots \\ b_m^T \end{bmatrix} x = \begin{bmatrix} b_1^T x \\ \vdots \\ b_m^T x \end{bmatrix}$$

The j^{th} entry of Ax is row j dotted with x .

$$\begin{bmatrix} 2 & -3 & 5 \\ -1 & 1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 - 6 + 5 \\ -1 + 2 + 2 \end{bmatrix} = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$$

$$(Ax)_i = \sum_{j=1}^n A_{ij} x_j \quad A \text{ } m \times n, \quad x \in \mathbb{R}^n$$

Examples:



$$\begin{bmatrix} a_1 & \dots & a_k \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_k \end{bmatrix} = \underbrace{x_1 a_1 + \dots + x_k a_k}_{\text{mixed signal w/ weights}}$$

↑ audio time signals ↑ weights

$$n \begin{bmatrix} -1 & 1 & & & & & 0 \\ & -1 & 1 & & & & \\ & & -1 & 1 & & & \\ & & & -1 & 1 & & \\ 0 & & & & & \ddots & \\ & & & & & & -1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_{n+1} \end{bmatrix} = \begin{bmatrix} x_2 - x_1 \\ \vdots \\ x_{n+1} - x_n \end{bmatrix} \quad \text{"discrete derivative"}$$

$$\begin{bmatrix} 1 & t_1 & t_1^2 & \dots & t_1^n \\ \vdots & \vdots & \vdots & \dots & \vdots \\ 1 & t_k & t_k^2 & \dots & t_k^n \end{bmatrix} \begin{bmatrix} c_0 \\ \vdots \\ c_n \end{bmatrix} = \begin{bmatrix} c_0 + c_1 t_1 + c_2 t_1^2 + \dots + c_n t_1^n \\ \vdots \end{bmatrix}$$

↑
Vandermonde, neat lingo!

MISC:

$$0x = 0$$

$$Ix = x$$

$$\begin{bmatrix} a_1 & \dots & a_n \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = x_1 a_1 + \dots + x_n a_n$$

$$A e_k = [a_1 \dots a_n] e_k = 0 a_1 + \dots + 1 a_k + \dots + 0 a_n = a_k$$

(selects k th column)

Average of columns: $A \vec{1}$

$$(A+B)x = Ax + Bx$$

$$(cA)x = c(Ax) = A(cx)$$

what as nice
as can be!

$$A(x+y) = Ax + Ay$$

$$A(cx) = c(Ax)$$

$$\left. \begin{array}{l} A(x+y) = Ax + Ay \\ A(cx) = c(Ax) \end{array} \right\} A(\alpha x + \beta y) = \alpha Ax + \beta Ay (!)$$

Given a matrix A , $m \times n$,

we can associate with it a map $\mathbb{R}^n \rightarrow \mathbb{R}^m$

$$x \mapsto Ax$$

Things you can represent:

all the above examples

assets \longrightarrow revenue per month for months

signal \longrightarrow discrete derivative

coeffs \longrightarrow polynomial evaluated at points

vector \longrightarrow itself (I)

Here's some more

$$\begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} x_2 \\ x_3 \\ x_1 \end{bmatrix}$$

↑
permutation
matrix

row $i \longrightarrow$ row j

if column i is e_j