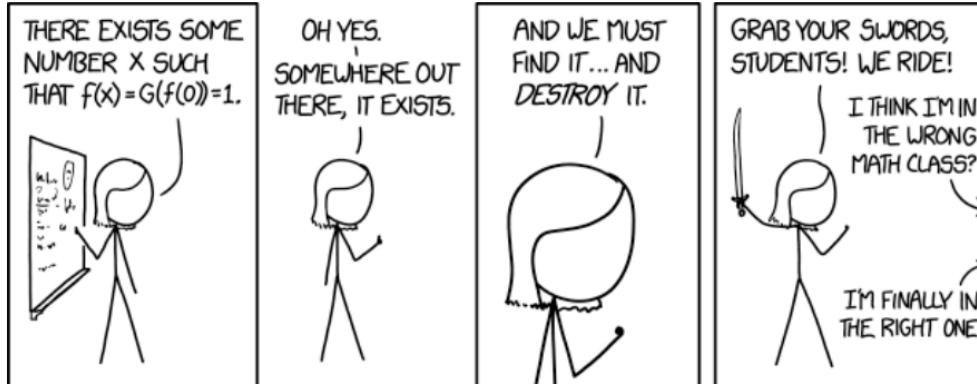


EE16A

Page Rank, Eigenvalues and Eigenspaces

Recall: Equivalent Statements:

- Matrix A is **invertible**
- $Ax=b$ has a **unique solution**
- A has linearly independent columns (A is **full rank**)
- A has a **trivial nullspace**
- The **determinant** of A is not zero



Jargon Roundup

- **range/span** of matrix A is the set of all possible linear combinations of the column vectors (all the outputs it can get to)

Jargon Roundup

- **range/span** of matrix A is the set of all possible linear combinations of the column vectors (all the outputs it can get to)
- **rank** is the dimension of the span of the columns of matrix A

$$\begin{aligned}\dim(\text{colspan}(A)) &= \# \text{ ind. cols.} \\ &= \# \text{ of cols. w/ pivots} \\ &= \dim(\text{rowspan}(A))\end{aligned}$$

$$\# \text{ lin. ind. cols } A = \# \text{ lin. indep. rows } A$$

$$A = \begin{bmatrix} & & \\ & x & x \\ x & x & x \\ x & x & x \end{bmatrix}$$

3x2

Jargon Roundup

- **range/span** of matrix A is the set of all possible linear combinations of the column vectors (all the outputs it can get to)
- **rank** is the dimension of the span of the columns of matrix A
- **nullspace** of matrix A is the set of solutions to $A\vec{x} = \vec{0}$ (all the places it can't get to)

Def'n:

$$N(A) = \left\{ \vec{x} \mid A\vec{x} = \vec{0}, \vec{x} \in \mathbb{R}^N \right\}$$

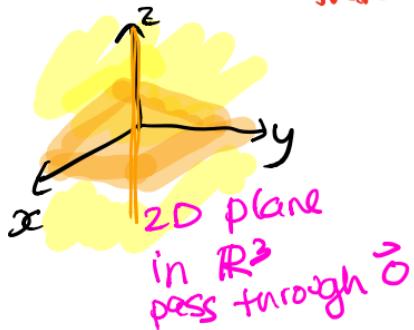
↑
Null
Space
vectors
that solve

size of
input

$$A = \begin{bmatrix} x & y & z \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

↑
lin.
ind.

\vec{x} can't
reach



$$\begin{bmatrix} A \\ M \times N \end{bmatrix} \begin{bmatrix} \vec{x} \\ N \times 1 \end{bmatrix} = \begin{bmatrix} \vec{b} \\ N \times 1 \end{bmatrix}$$

$$\text{rank}(A) = 2$$

$$\dim(N(A)) = 1$$

$$\mathbb{R}^3 \text{ 3 dim total}$$

$$2+1=3$$

"Rank-Nullity
Theorem"

$$\text{rank}(A) + \dim(N(A)) = N$$

of cols
of A

Jargon Roundup

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- **rank** is the dimension of the span of the columns of matrix A
- **nullspace** of matrix A is the set of solutions to $Ax=0$ (all the places it can't get to)
- **vector space** is a **set of vectors connected by two operators (+,x)** that obeys the **10 axioms**

A vector space \mathbb{V} is a set of vectors and two operators that satisfy the following properties:

Note 7

• Vector Addition

- Associative: $\vec{u} + (\vec{v} + \vec{w}) = (\vec{u} + \vec{v}) + \vec{w}$ for any $\vec{v}, \vec{u}, \vec{w} \in \mathbb{V}$.
- Commutative: $\vec{u} + \vec{v} = \vec{v} + \vec{u}$ for any $\vec{v}, \vec{u} \in \mathbb{V}$.
- Additive Identity: There exists an additive identity $\vec{0} \in \mathbb{V}$ such that $\vec{v} + \vec{0} = \vec{v}$ for any $\vec{v} \in \mathbb{V}$.
- Additive Inverse: For any $\vec{v} \in \mathbb{V}$, there exists $-\vec{v} \in \mathbb{V}$ such that $\vec{v} + (-\vec{v}) = \vec{0}$. We call $-\vec{v}$ the additive inverse of \vec{v} .
- Closure under vector addition: For any two vectors $\vec{v}, \vec{u} \in \mathbb{V}$, their sum $\vec{v} + \vec{u}$ must also be in \mathbb{V} .

• Scalar Multiplication

- Associative: $\alpha(\beta\vec{v}) = (\alpha\beta)\vec{v}$ for any $\vec{v} \in \mathbb{V}$, $\alpha, \beta \in \mathbb{R}$.
- Multiplicative Identity: There exists $1 \in \mathbb{R}$ where $1 \cdot \vec{v} = \vec{v}$ for any $\vec{v} \in \mathbb{V}$. We call 1 the multiplicative identity.
- Distributive in vector addition: $\alpha(\vec{u} + \vec{v}) = \alpha\vec{u} + \alpha\vec{v}$ for any $\alpha \in \mathbb{R}$ and $\vec{u}, \vec{v} \in \mathbb{V}$.
- Distributive in scalar addition: $(\alpha + \beta)\vec{v} = \alpha\vec{v} + \beta\vec{v}$ for any $\alpha, \beta \in \mathbb{R}$ and $\vec{v} \in \mathbb{V}$.
- Closure under scalar multiplication: For any vector $\vec{v} \in \mathbb{V}$ and scalar $\alpha \in \mathbb{R}$, the product $\alpha\vec{v}$ must also be in \mathbb{V} .

Jargon Roundup

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- vector **subspace** is a subset of vectors from a vector space that obey 3 properties

Definition 8.1 (Subspace): A subspace \mathbb{U} consists of a subset of the vector space \mathbb{V} that satisfies the following three properties:

Note 8

- Contains the zero vector: $\vec{0} \in \mathbb{U}$.
- Closed under vector addition: For any two vectors $\vec{v}_1, \vec{v}_2 \in \mathbb{U}$, their sum $\vec{v}_1 + \vec{v}_2$ must also be in \mathbb{U} .
- Closed under scalar multiplication: For any vector $\vec{v} \in \mathbb{U}$ and scalar $\alpha \in \mathbb{R}$, the product $\alpha\vec{v}$ must also be in \mathbb{U} .

Jargon Roundup

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- **row space** is the **span of the rows** of a matrix

Jargon Roundup

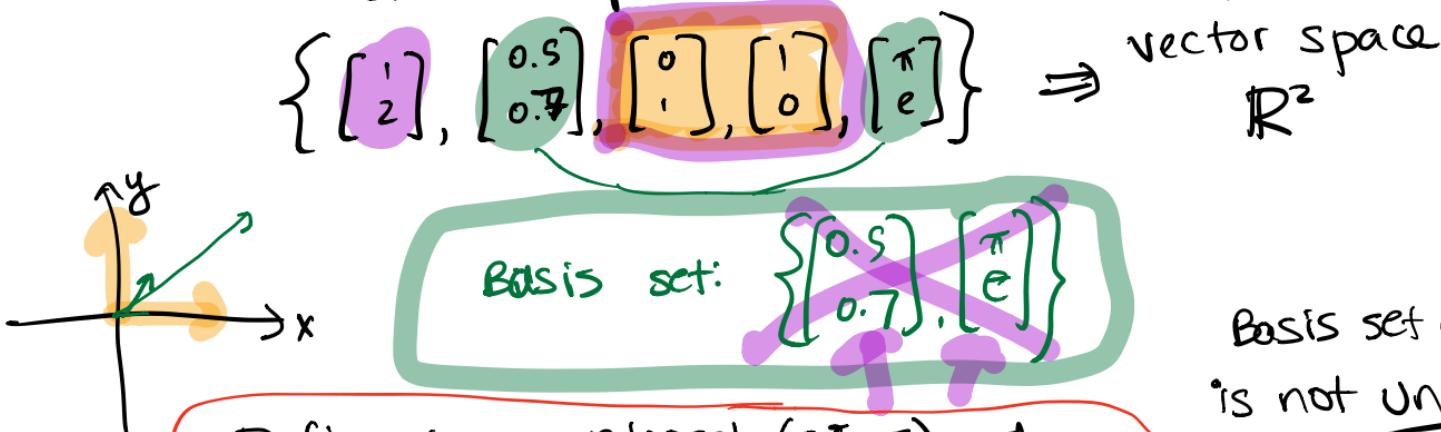
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- A **basis** for a vector space is a minimum set of vectors needed to represent all vectors in the space

Jargon Roundup

Most "efficient representation" of a vector space.



Basis set of vectors
is not unique
but dim is.

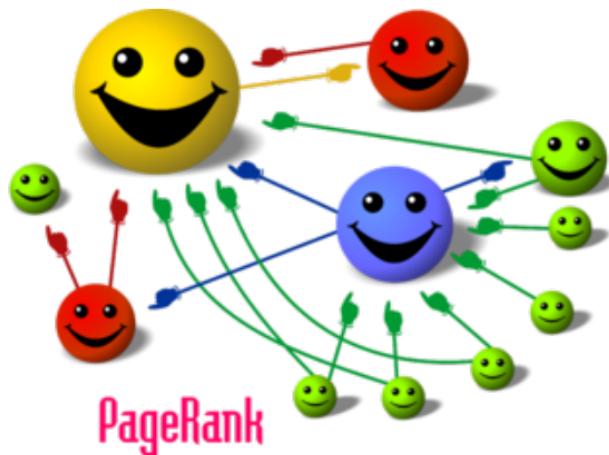
Def'n: Given a vector set (V, F) ~~is a~~
→ a set of vectors $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n\}$, is a BASIS SET
if ① $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n$ are lin. Indep.
② $\vec{v} \in V$ (any vect) $\rightarrow \vec{v} = \alpha_1 \vec{v}_1 + \alpha_2 \vec{v}_2 + \dots + \alpha_n \vec{v}_n$

$$\dim = 2$$

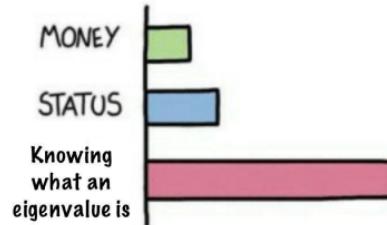
- A **basis** for a vector space is a minimum set of vectors needed to represent all vectors in the space

Today's Jargon: determinant, eigenvalue, eigenvector

- We will use Google's PageRank algorithm as a new application example to learn about these things!



WHAT GIVES PEOPLE
FEELINGS OF POWER

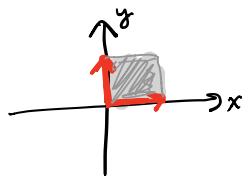


Determinant

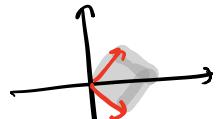
$$\det \begin{pmatrix} a & b \\ c & d \end{pmatrix} = ad - bc$$

$$\det(A) = |A|$$

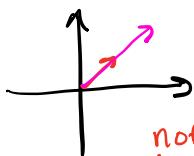
$\det(A) \neq 0$ when A invertible



$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad \text{Area} \neq 0$$

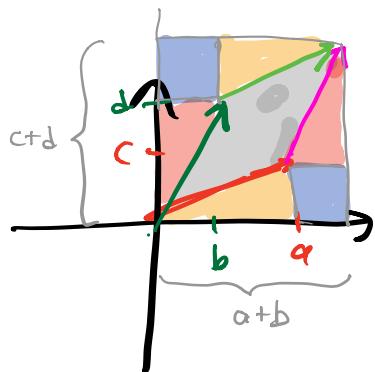


$$\begin{bmatrix} 1 & -1 \\ 1 & 0 \end{bmatrix} \neq 0$$



$$\begin{bmatrix} 1 & 2 \\ 1 & 0 \end{bmatrix} = 0$$

not invertible
 $\det(A) = 0$



$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

Compute area of parallelogram as:

$$\begin{aligned} & (a+b)(c+d) \\ & - \text{blue triangle } \times 2 \\ & - \text{orange triangle } \times 2 \\ & - \text{red triangle } \times 2 \end{aligned}$$

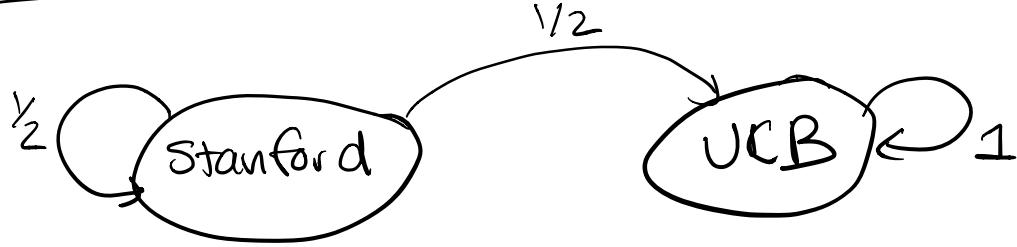
$(a+b)(c+d)$
 $b c \times 2$
 $\frac{1}{2} ac \times 2$
 $\frac{1}{2} bd \times 2$

$$\begin{aligned} \square \text{area} &= (c+d)(a+b) - 2bc - ac - bd \\ &= \cancel{ca} + \cancel{cb} + \cancel{da} + \cancel{db} - \cancel{2bc} - \cancel{ac} - \cancel{bd} \\ &= \underline{\underline{ad - bc}} \quad \text{yay!} \end{aligned}$$

Higher dimensional?

$$\begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix}$$

Page rank



$$\vec{x} = \begin{bmatrix} x_{\text{Stanford}} \\ x_{\text{UCB}} \end{bmatrix}$$

$$Q = \begin{bmatrix} 1/2 & 0 \\ 1/2 & 1 \end{bmatrix}$$

State
transition
mtx

Start with everyone @ Stanford $\vec{x}(0) = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

After one timestep: $\vec{x}(1) = Q \cdot \vec{x}(0)$

$$= \begin{bmatrix} 1/2 & 0 \\ 1/2 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 1/2 \\ 1/2 \end{bmatrix}$$

After two timesteps:

$$\vec{x}(2) = Q \cdot \vec{x}(1)$$

$$= \begin{bmatrix} 1/2 & 0 \\ 1/2 & 1 \end{bmatrix} \begin{bmatrix} 1/2 \\ 1/2 \end{bmatrix} = \begin{bmatrix} 1/4 \\ 3/4 \end{bmatrix}$$

After 3 timesteps: $\vec{x}(3) = \begin{bmatrix} 1/2 & 0 \\ 1/2 & 1 \end{bmatrix} \begin{bmatrix} 1/4 \\ 3/4 \end{bmatrix} = \begin{bmatrix} 1/8 \\ 7/8 \end{bmatrix}$

$$\vec{x}(t) = \begin{bmatrix} (1/2)^t \\ 1 - (1/2)^t \end{bmatrix}$$

$$\vec{x}(\infty) = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

Everyone's
at UCB!

What if instead $\vec{x}(0) = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

$$\vec{x}(1) = Q \cdot \vec{x}(0) = \begin{bmatrix} 1/2 & 0 \\ 1/2 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

Steady State Sol'n

In general,

$$\vec{x}_{\text{steady state}} = Q \cdot \vec{x}_{\text{steady state}}$$

$$I \vec{x}_{ss} = Q \cdot \vec{x}_{ss}$$

$$Q \cdot \vec{x}_{ss} - I \vec{x}_{ss} = \vec{0}$$

$$(Q - I) \vec{x}_{ss} = \vec{0}$$

trans. mt^x identity steady state sol'n

Want to find Null($Q - I$) \rightarrow steady state sol'n.

$$\begin{bmatrix} 1/2 & 0 \\ 1/2 & 1 \end{bmatrix} - \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \underbrace{\begin{bmatrix} -1/2 & 0 \\ 1/2 & 0 \end{bmatrix}}_{Q-I}$$

$$\text{Null}(Q - I): \begin{bmatrix} -1/2 & 0 & | & 0 \\ 1/2 & 0 & | & 0 \end{bmatrix} \xrightarrow{Q-I} \begin{bmatrix} 1 & 0 & | & 0 \\ 1/2 & 0 & | & 0 \end{bmatrix} \xrightarrow{\text{row reduction}} \begin{bmatrix} 1 & 0 & | & 0 \\ 0 & 1 & | & 0 \end{bmatrix}$$

basis x_{ss}
 I_{UCB}

$$\text{Null}(Q - I) = \begin{bmatrix} 0 \\ x_2 \end{bmatrix} = \text{span} \left\{ \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$$

Eigen space of Q
corresponding to eigenvalue
 $1 \rightarrow \text{steady state}$

Def'n: Let Q be square matrix, $\lambda \in \mathbb{R}$
 if $\vec{x} \neq 0$ such that $Q \cdot \vec{x} = \lambda \cdot \vec{x}$ "Lambda" eigenval.
 then λ is an eigenvalue of Q and $\text{Null}(Q - \lambda I)$ is
 its eigenspace corresponding to λ

Find eigenvalues & Eigspaces: $Q = \begin{bmatrix} 1/2 & 0 \\ 1/2 & 1 \end{bmatrix}$

Want to find λ, \vec{x} such that $Q \cdot \vec{x} = \lambda \cdot \vec{x}$
 matrix scalar

$$Q \cdot \vec{x} - \lambda \cdot I \cdot \vec{x} = \vec{0}$$

$$(Q - \lambda I) \vec{x} = \vec{0}$$

Find $\vec{x} \in \text{Null}(Q - \lambda I)$

$$Q - \lambda I = \begin{bmatrix} 1/2 & 0 \\ 1/2 & 1 \end{bmatrix} - \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} = \begin{bmatrix} 1/2 - \lambda & 0 \\ 1/2 & 1 - \lambda \end{bmatrix}$$

① find λ
 ② find \vec{x}

If we have non-trivial nullspace

$$\det(Q - \lambda I) = 0$$

$$(\frac{1}{2} - \lambda)(1 - \lambda) = 0$$

characteristic polynomial

$$\lambda_1 = \frac{1}{2}, \quad \lambda_2 = 1$$