



Welcome to EECS 16A! Designing Information Devices and Systems I



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Announcements

- Last time:
 - Linear (in)dependance
 - Matrix Transformations
- Today:
 - Continue with Matrix transformations
 - Matrix Inverse
 - Vector spaces

Matrix Transformations

$$\begin{bmatrix} \cos 90^{\circ} & \sin 90^{\circ} \\ -\sin 90^{\circ} & \cos 90^{\circ} \end{bmatrix} \begin{bmatrix} \alpha_{1} \\ \alpha_{2} \end{bmatrix} = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$$

Linear Transformation of vectors

f: is a linear transformation if:

$$f(\alpha \overrightarrow{x}) = \alpha f(\overrightarrow{x}) \qquad \alpha \in \mathbb{R}$$
$$f(\overrightarrow{x} + \overrightarrow{y}) = f(\overrightarrow{x}) + f(\overrightarrow{y})$$

Claim: Matrix-vector multiplications satisfy linear transformation

$$A \cdot (\alpha \overrightarrow{x}) = \alpha A \overrightarrow{x}$$

Proof via explicitly writing the elements

$$A \cdot (\overrightarrow{x} + \overrightarrow{y}) = A \overrightarrow{x} + A \overrightarrow{y}$$

Vectors as states, Matrices as state transition

Vectors can represent states of a system

Example: The state of a car at time = t

 $\vec{S}(t) = \begin{cases} x(t) \\ y(t) \\ y(t) \\ y(t) \\ z \text{ velocity} \\ \theta(t) \end{cases}$

Q: Is that enough?

A: need orientation or $v_x(t), v_y(t)$

Graph Transition Matrices

Example: Reservoirs and Pumps



Pumps move water... What would the state be tomorrow?





$$x_A(t+1) = x_A(t)$$

$$x_B(t+1) = x_C(t)$$

$$x_C(t+1) = x_B(t)$$

Write as a matrix-vector multiplication:

$$\begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$



1

$$x_A(t+1) = x_A(t)$$

$$x_B(t+1) = x_C(t)$$

$$x_C(t+1) = x_B(t)$$

Write as a matrix-vector multiplication:

$$\begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$

or
$$\overrightarrow{x}(t+1) = Q\overrightarrow{x}(t)$$

What is the state after 2 times?

$$\overrightarrow{x}(t+2) = Q\overrightarrow{x}(t+1) = QQ\overrightarrow{x}(t) = Q^{2}\overrightarrow{x}(t)$$



1

$$\begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$

 B^{1}

1



$$\begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$$





 $\vec{x}(0) = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$ What is the state after at t=1, 2?





Q) What will happen if we keep going?

A) Numbers will diminish to zero









 $x(t+n) = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 2 & 0 \end{bmatrix} x(t)$

Q) What will happen if we keep going?

A) Numbers will explode to infinity



Graph Representation

Ex: Reservoirs and Pumps



Nodes

I have 3 reservoirs: A,B,C and I want to keep track of how much water is in each

When I turn on some pumps, water moves between the reservoirs.

Where the water moves and what fraction is represented by arrows. Edge weights Edges

"directed" graph because arrows have a direction

Where does the rest of the water in A go? Need to label that too...

Can you tell me how much water in each after pumps start? Need to know initial amounts





 \mathcal{O}_{C}

 $J_{R} \rightarrow C \overline{C} \rightarrow$

2 p (+41)

کرر



Example 2:

$$\begin{bmatrix} \mathcal{J}_{C_{A}}(\mathcal{H}) \\ \mathcal{J}_{C_{B}}(\mathcal{H}) \\ \mathcal{J}_{C_{B}}(\mathcal{H}) \\ \mathcal{J}_{C_{C}}(\mathcal{H}) \end{bmatrix} = \begin{bmatrix} A \to A & B \to A & C \to A \\ A \to B & B \to B & C \to B \\ A \to B & B \to B & C \to B \\ A \to C & B \to C & C \to C \end{bmatrix} \begin{bmatrix} \mathcal{J}_{C_{C}}(\mathcal{H}) \\ \mathcal{J}_{C_{C}}(\mathcal{H}) \\ \mathcal{J}_{C_{C}}(\mathcal{H}) \end{bmatrix}$$



Example 2:

 $\begin{bmatrix} J_{L_{A}}(1+1) \\ J_{L_{B}}(1+1) \\ J$



$$\begin{bmatrix} J_{L_{A}}(1+1) \\ J_{L_{A}}(1+1) \\ J_{L_{B}}(1+1) \\ J_$$







$$\begin{bmatrix} J_{L_{A}}(1+1) \\ J_{L_{A}}(1+1) \\ J_{L_{B}}(1+1) \\ J$$











$$\begin{array}{cccc}
6 \left[\mathcal{J}_{L_{A}}(1+1) \\
10 \left[\mathcal{J}_{L_{B}}(1+1) \\
2 \left[\mathcal{J}_{L_{C}}(1+1) \\
2 \left[\mathcal{J}$$



Q) Will flipping the arrows make us go back in time?





A) In general, no!

Matrix Transpose

If the elements of the matrix $A \in \mathbb{R}^{N \times M}$ are a_{ij} The elements of $A^T \in \mathbb{R}^{M \times N}$ are a_{ji} Matrix transpose is not (generally) an inverse!



Matrix Transpose

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Matrix Transpose

If the elements of the matrix $A \in \mathbb{R}^{N \times M}$ are a_{ij} The elements of $A^T \in \mathbb{R}^{M \times N}$ are a_{ji} Matrix transpose is not (generally) an inverse!



Matrix Inversion



Matrix Inverse

$$\overrightarrow{x}(t+1) = Q\overrightarrow{x}(t)$$

Is there a square matrix P such that we can go back in time?

$$\overrightarrow{x}(t) = P\overrightarrow{x}(t+1)$$

Yes, if : PQ = I

As consequence : QP = I

$$\overrightarrow{Px}(t+1) = \overrightarrow{PQx}(t)$$
$$\overrightarrow{Px}(t+1) = \overrightarrow{Ix}(t)$$

$$\overrightarrow{x}(t+1) = Q\overrightarrow{x}(t)$$

$$\overrightarrow{x}(t+1) = QP\overrightarrow{x}(t+1)$$

$$\overrightarrow{x}(t+1) = I\overrightarrow{x}(t+1)$$

Matrix Inverse - Formal definition

- Definition: Let $P, Q \in \mathbb{R}^{N \times N}$ be square matrices.
 - P is the inverse of Q if PQ = QP = I

We say that $P = Q^{-1}$ and $Q = P^{-1}$

Q: What about non-square matrices? A: EECS16B!

Computing the Matrix Inverse

$$Q$$

$$\begin{bmatrix} J_{L_{A}}(1+1) \\ J_{L_{B}}(1+1) \\ J_{L_{B}}(1+1) \\ J_{L_{C_{C}}}(1+1) \\ J_{L_{C_{C}}}$$



• Want $P = Q^{-1}$ such that $\overrightarrow{x}(t) = P\overrightarrow{x}(t+1)$

- Need:
$$QP = I$$

Computing the Matrix Inverse

Need: QP = IPose as a linear set of equations. Solve with Gaussian Elimination







Matrix Inverse via Gaussian Elimination $\begin{bmatrix} 0 & 1 & 0 & | & 1 & 0 & 0 \\ \frac{1}{2} & 0 & 1 & | & 0 & 1 & 0 \\ \frac{1}{2} & 0 & 0 & | & 0 & 0 & 1 \end{bmatrix}$ $\begin{bmatrix} 1 & 0 & 2 & 0 & 2 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & -2 & 0 & -2 & 2 \end{bmatrix}$ $\begin{bmatrix} 1 & 0 & 2 & 0 & 2 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & -1 \end{bmatrix}$ $\begin{bmatrix} \frac{1}{2} & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \\ \frac{1}{2} & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$ $\begin{bmatrix} 1 & 0 & 2 & 0 & 2 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \\ \frac{1}{2} & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$ $\begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 2 \\ 0 & 1 & 0 & 1 & 0 & 0 \end{bmatrix}$ 1 0 1 0

Let's check

$$\begin{array}{cccc}
6 \left[\mathcal{J}_{L_{A}}(1+1) \\
10 \left[\mathcal{J}_{L_{B}}(1+1) \\
2 \left[\mathcal{J}_{L_{C}}(1+1) \\
2 \left[\mathcal{J}$$





Let's check

$$\begin{array}{cccc}
6 \left[\mathcal{J}_{L_{A}}(1+1) \\
10 \left[\mathcal{J}_{L_{B}}(1+1) \\
2 \left[\mathcal{J}_{L_{C}}(1+1) \\
2 \left[\mathcal{J}$$





And now we can take any number of steps backwards!

Can we always invert a function?

- Can we always invert a function $\dots f^{-1}(f(\vec{x})) = \vec{x}$?
 - $-f(x) = x^2 ?$
 - -f(x) = ax?
 - -f(x) = Ax?

Invertibility of Linear Transformations

- Theorem: A is invertible, if and only if (iff) the columns of A are linearly independent.
 - 1. If columns of A are lin. dep. then A^{-1} does not exist
 - 2. If A^{-1} exists, then the cols. of A are linearly independent
- Proof concept: Assume linear dependence and invertibility and show that it is a contradiction
- From linear independence: $\exists \vec{\alpha} \neq 0$ such that $A\vec{\alpha} = 0$

Assume
$$A^{-1}$$
 exists
 $A^{-1}A\overrightarrow{\alpha} = A^{-1}0$
 $I\overrightarrow{\alpha} = 0$
But $\overrightarrow{\alpha} \neq 0$! Hence A^{-1} does not exist

Inverse of a 2x2 matrix



Derive via Gauss Elimination!

Equivalent Statements

- $\bullet \operatorname{Matrix} A \text{ is invertible} \\$
- • $A\overrightarrow{x} = \overrightarrow{b}$ has a unique solution
- •A has linearly independent columns (A is full rank)
- $\bullet A$ has a trivial nullspace
- The determinant of A is not zero

Today (and next time's) Jargon

- Rank a matrix A is the number of linearly independent columns
- Nullspace of a matrix A is the set of solutions to $A\vec{x} = 0$
- A vector space is a set of vectors connected by two operators (+,x)
- A vector **subspace** is a subset of vectors that have "nice properties"
- A **basis** for a vector space is a minimum set of vectors needed to represent all vectors in the space
- Dimension of a vector space is the number of basis vectors
- Column space is the span (range) of the columns of a matrix
- Row space is the span of the rows of a matrix

 ESPIRiT—an eigenvalue approach to autocalibrating parallel MRI: where SENSE meets
 834
 2014

 GRAPPA

 M Uecker, P Lai, MJ Murphy, P Virtue, M Elad, JM Pauly, SS Vasanawala, ...

 Magnetic resonance in medicine 71 (3), 990-1001

https://www.ncbi.nlm.nih.gov/pmc/articles/PMC4142121/

- Basis 3 times
- Rank 4 times
- Row space 4 times
- Columns (of a matrix) 6 times
- Subspace 17 times
- Null Space 29 times
- Eigen 87 times

Vector Space

• From Merriam Webster:

Definition of *vector space*

a set of vectors along with operations of addition and multiplication such that the set is a commutative group under addition, it includes a multiplicative inverse, and multiplication by scalars is both associative and distributive

Vector Space

A vector space, is a set of vectors and scalars (V∈ ℝ^N, F∈ ℝ) and two operators · , + that satisfy the following:

Axioms of closure

Axioms of addition

(+)

Axioms of scaling (.)

1. $\alpha \overrightarrow{x} \in \mathbb{V}$ 2. $\overrightarrow{x} + \overrightarrow{y} \in \mathbb{V}$ 3. $\vec{x} + (\vec{y} + \vec{z}) = (\vec{x} + \vec{y}) + \vec{z}$ (associativity) 4. $\vec{x} + \vec{y} = \vec{y} + \vec{x}$ (commutativity) 5. $\exists \vec{0} \in \mathbb{V}$ s.t. $\vec{x} + \vec{0} = \vec{x}$ (additive identity) 6. $\exists (-\overrightarrow{x}) \in \mathbb{V}$ s.t. $\overrightarrow{x} + (-\overrightarrow{x}) = \overrightarrow{0}$ (additive inverse) 7. $\alpha(\vec{x} + \vec{y}) = \alpha \vec{x} + \alpha \vec{y}$ (distributivity) 8. $\alpha \cdot (\beta \overrightarrow{x}) = (\alpha \beta) \cdot \overrightarrow{x}$ 9. $(\alpha + \beta)\vec{x} = \alpha\vec{x} + \beta\vec{x}$ 10. $1 \cdot \overrightarrow{x} = \overrightarrow{x}$

Vector Space

- A vector space \mathbb{V} is a set of vectors and two operators \cdot , + that satisfy the following:
 - 1. $\alpha \overrightarrow{x} \in \mathbb{V}$ 2. $\overrightarrow{x} + \overrightarrow{y} \in \mathbb{V}$ 3. $\vec{x} + (\vec{y} + \vec{z}) = (\vec{x} + \vec{y}) + \vec{z}$ (associativity) 4. $\vec{x} + \vec{y} = \vec{y} + \vec{x}$ (commutativity) 5. $\exists \vec{0} \in \mathbb{V}$ s.t. $\vec{x} + \vec{0} = \vec{x}$ (additive identity) 6. $\exists (-\overrightarrow{x}) \in \mathbb{V}$ s.t. $\overrightarrow{x} + (-\overrightarrow{x}) = \overrightarrow{0}$ 7. $\alpha(\vec{x} + \vec{y}) = \alpha \vec{x} + \alpha \vec{y}$ (distributivity) 8. $\alpha \cdot (\beta \overrightarrow{x}) = (\alpha \beta) \cdot \overrightarrow{x}$ 9. $(\alpha + \beta)\vec{x} = \alpha\vec{x} + \beta\vec{x}$ 10. $1 \cdot \overrightarrow{x} = \overrightarrow{x}$



$$\bigotimes Is \begin{bmatrix} a & b \\ c & d \end{bmatrix} \in \mathbb{R}^{2 \times 2} ?$$

 $\blacksquare \ \text{Is } \alpha \in \mathbb{R}, \alpha \geq 0 ?$

 $\sum_{n \in \mathbb{N}} \operatorname{Is} \operatorname{span} \left\{ \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\} ?$ $\sum_{n \in \mathbb{N}} \operatorname{Is} \begin{bmatrix} 0 \\ 1 \end{bmatrix} ?$ $\sum_{n \in \mathbb{N}} \operatorname{Is} \operatorname{0?}$

Subspaces

- A subspace $\mathbb U$ consists of a subset of $\mathbb V$ in vector space ($\mathbb V,\mathbb F,+\,,\,\cdot\,)$
 - $\mathbb{U}\subset\mathbb{V}$ and have 3 properties
 - 1. Contains $\overrightarrow{0}$, i.e., $\overrightarrow{0} \in \mathbb{U}$
 - 2. Closed under vector addition: $\overrightarrow{v}_1, \overrightarrow{v}_2 \in \mathbb{U}, \Rightarrow \overrightarrow{v}_1 + \overrightarrow{v}_2 \in \mathbb{U}$
 - 3. Closed under scalar multiplication: $\overrightarrow{v}_1 \in \mathbb{U}, \ \alpha \in \mathbb{F}, \ \Rightarrow \alpha \overrightarrow{v} \in \mathbb{U}$