



Lecture 09

Math 22 Summer 2017
July 10, 2017



- ▶ X-hour Tuesday (§2.3)
- ▶ “Quiz” Wednesday to practice for Midterm
- ▶ Midterm will cover material through §2.3



- ▶ Practice problems about inverses
- ▶ More about matrices: §2.1/§2.2 powers and transpose
- ▶ §1.10 Linear difference equations

§2.2 Classwork



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- Find A^{-1} for $A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$.



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- ▶ Use A^{-1} from the previous part to solve $A\mathbf{x} = \mathbf{b}$ for $\mathbf{b} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$.



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- ▶ Find the sequence of elementary matrices that transform the above A to I_2 .

§2.1 Powers of a matrix



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What is the k -th power of a *diagonal matrix*?

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What's an example?

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(d) $(AB)^T = B^T A^T$.



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- (c) For $\lambda \in \mathbb{R}$, $(\lambda A)^T = \lambda A^T$
- (d) $(AB)^T = B^T A^T$.

You can check the first 3 as an exercise, but let's look at the last part together.

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Writing individual matrix entries more concisely we have:

$$(AB)_{ij} = \sum_{k=1}^3 a_{ik} b_{kj}, \quad \text{and} \quad (B^T A^T)_{ij} = \sum_{k=1}^3 b_{ik} a_{jk}.$$

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How can we generalize this to finish the proof?

Matrix multiplication redux



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The previous proof illustrates the many ways to view matrix multiplication...

$$A_{m \times n} B_{n \times p} = (AB)_{m \times p}.$$



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$$AB = \left[A \cdot \text{col}_1(B) \ \cdots \ A \cdot \text{col}_p(B) \right].$$

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Then we call the sequence $\{\mathbf{x}_k\}_{k=0}^{\infty}$ together with the *transition matrix* A a **linear difference equation**.

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What is the transition matrix A that describes this system as a linear difference equation?

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The matrix that describes the linear difference equation from the previous slide is given by

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<http://sagecell.sagemath.org/?q=hgoihs>