The Invertible Matrix Theorem

Linear Algebra MATH 2076



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An $n \times n$ matrix A is *invertible* if and only if there is another $n \times n$ matrix C with A = C = C A.



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If E = I, then $F = A^{-1}$.



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When does $A\vec{x} = \vec{0}$ have a *unique* solution?

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What's important: any one of last 5 statements true \implies A is invertible.



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Section 2.3

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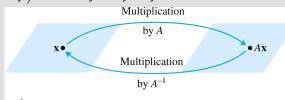
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Above says $T^{-1} = S$.



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If A is square and invertible, then so is A^T and $A^T = (A^{-1})^T$.