Matrix Inverses

Geometric Algorithms
Lecture 10

CAS CS 132

Objectives

1. Define a few more important matrix operations
2. Motivate and define matrix inverses
3. Application: Adjacency Matrices

Keywords

Matrix Transpose

Inner Product

Matrix Power

Square Matrix

Matrix Inverse

Invertible Transformation

1–1 Correspondence

numpy.linalg.inv

eterminant

Invertible Matrix Theorem

Recap Problem

Suppose that A, $B = \begin{bmatrix} \mathbf{b}_1 & \mathbf{b}_2 & \mathbf{b}_3 \end{bmatrix}$ and $C = [\mathbf{c}_1 \ \ \mathbf{c}_2 \ \ \mathbf{c}_3]$ are matrices such that

$$A(B + 5I) = C$$

Find a solution to the equation $A\mathbf{x} = \mathbf{c}_2$.

Answer: $\mathbf{b}_2 + 5\mathbf{e}_2$

More Matrix Operations

Transpose (Pictorially)

$$\begin{bmatrix} 1 & 2 & 3 & 4 & 5 \\ 1 & 2 & 3 & 4 & 5 \\ 1 & 2 & 3 & 4 & 5 \\ 1 & 2 & 3 & 4 & 5 \end{bmatrix} \longrightarrow \begin{bmatrix} 1 & 1 & 1 & 1 \\ 2 & 2 & 2 & 2 \\ 3 & 3 & 3 & 3 \\ 4 & 4 & 4 & 4 \\ 5 & 5 & 5 & 5 \end{bmatrix}$$

Transpose (Pictorially)

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Transpose

Definition. For a $m \times n$ matrix A, the **transpose of** A, written A^T, is the $n \times m$ matrix such that

$$(A^T)_{ij} = A_{ji}$$

Example.

$$\begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix}^T = \begin{bmatrix} 1 & 0 \\ -1 & 1 \end{bmatrix}$$

Algebraic Properties (Transpose)

$(A^T)^T = A$

$(A + B)^T = A^T + B^T$

$(cA)^T = cA^T$ (where c is a scalar)

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

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$(AB)^T = B^T A^T$ Important: the order reverses!

Challenge Problem (Not In-Class)

Show that $(AB)^T = B^T A^T$.

Example: $\left(\begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \right)^T$

Transposes and Inner Products

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is $\mathbf{u}^T\mathbf{v}$ defined?

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$1 \times n \qquad n \times 1 \qquad 1 \times 1$

$$\begin{bmatrix} u_1 & u_2 & u_3 & u_4 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \\ v_3 \\ v_4 \end{bmatrix} = \quad ?$$

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$1 \times n$ $\quad n \times 1 \quad$ 1×1

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$$\begin{bmatrix} u_1 & u_2 & u_3 & u_4 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \\ v_3 \\ v_4 \end{bmatrix} = u_1v_1 + u_2v_2 + u_3v_3 + u_4v_4$$

Transposes and Inner Products

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Definition. The **inner product** of two vectors $\mathbf{u}$ and $\mathbf{v}$ in $\mathbb{R}^n$ is

$$\langle \mathbf{u}, \mathbf{v} \rangle = \mathbf{u} \cdot \mathbf{v} = \mathbf{u}^T\mathbf{v}$$

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(we want $A^0 A^k = A^{0+k} = A^k$)

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1. AB is not necessarily equal to BA, even if both are defined.

2. If $AB = AC$ then it is not necessary that $B = C$.

3. If $AB = 0$ (the zero matrix) it is not necessarily the case that $A = 0$ or $B = 0$.

Question

Find two nonzero 2×2 matrices A and B such that $AB = 0$.

Challenge. *Choose A and B such that they have <u>all</u> nonzero entries.*

Answer

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

So Far: Matrix Operations

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`transpose` A^T

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`transpose`	A^T
`scaling`	cA

So Far: Matrix Operations

`transpose` A^T

`scaling` cA

`addition (subtraction)` $A + B$ $A + (-1)B = A - B$

So Far: Matrix Operations

transpose	A^T	
scaling	cA	
addition (subtraction)	$A + B$	$A + (-1)B = A - B$
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transpose	A^T	
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multiplication (powers)	AB	A^k

What's missing?

Matrix Inverses

Basic Algebra

$$2x = 10$$

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$\frac{1}{2}$ is the **reciprocal** or **multiplicative inverse** of 2.

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Basic Algebra

$$1x = 5$$

How do we solve this equation?

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How do we solve this equation?

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Wouldn't it be nice...

$$A^{-1}A\mathbf{x} = A^{-1}\mathbf{b}$$

How do we solve this equation?

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$$I\mathbf{x} = A^{-1}\mathbf{b}$$

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Do all matrices have inverses?

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No.

When does a matrix have an inverse?

Square Matrices

Definition. A $m \times n$ matrix A is **square** if $m = n$

$$\begin{bmatrix} * & * & * & * \\ * & * & * & * \\ * & * & * & * \\ * & * & * & * \end{bmatrix}$$

i.e., it has same number of rows as columns.

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» that can have inverses.

Dimension Tracking

$$A \quad \mathbf{x} = \mathbf{b}$$

Dimension Tracking

$$A^{-1}A \quad \mathbf{x} = A^{-1}\mathbf{b}$$

Dimension Tracking

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The only way for the dimensions to make sense is if A is square

Matrix Inverses

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Example. $$\begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}^{-1} = \begin{bmatrix} 1 & 0 \\ -1 & 1 \end{bmatrix}$$

Example: Geometric

Reflection across the x_1–axis in $\mathbb{R}^2$ is it's own inverse.

Verify:

Example: No inverse

$$\begin{bmatrix} 1 & 2 & -1 \\ 0 & 3 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$

`Verify:`

Inverses are Unique

Theorem. If B and C are inverses of A, then $B = C$.

Verify:

Inverses are Unique

Theorem. If B and C are inverses of A, then $B = C$.

Verify:

If A is invertible, then we write A^{-1} for *the* inverse of A.

Solutions for Invertible Matrix Equations

Theorem. For a $n \times n$ matrix A, if A is invertible then

$$A\mathbf{x} = \mathbf{b}$$

has a unique solution for any choice of $\mathbf{b}$.

Verify:

Unique Solutions

If $A\mathbf{x} = \mathbf{b}$ has a unique solution for any choice of $\mathbf{b}$, then it has

» exactly one solution for any choice of $\mathbf{b}$

Unique Solutions

If $A\mathbf{x} = \mathbf{b}$ has a unique solution for any choice of $\mathbf{b}$, then it has

» at least one solution for any choice of $\mathbf{b}$

» at most one solution for any choice of $\mathbf{b}$

Unique Solutions

If $A\mathbf{x} = \mathbf{b}$ has a unique solution for any choice of $\mathbf{b}$, then it has

» T is onto

» T is one-to-one

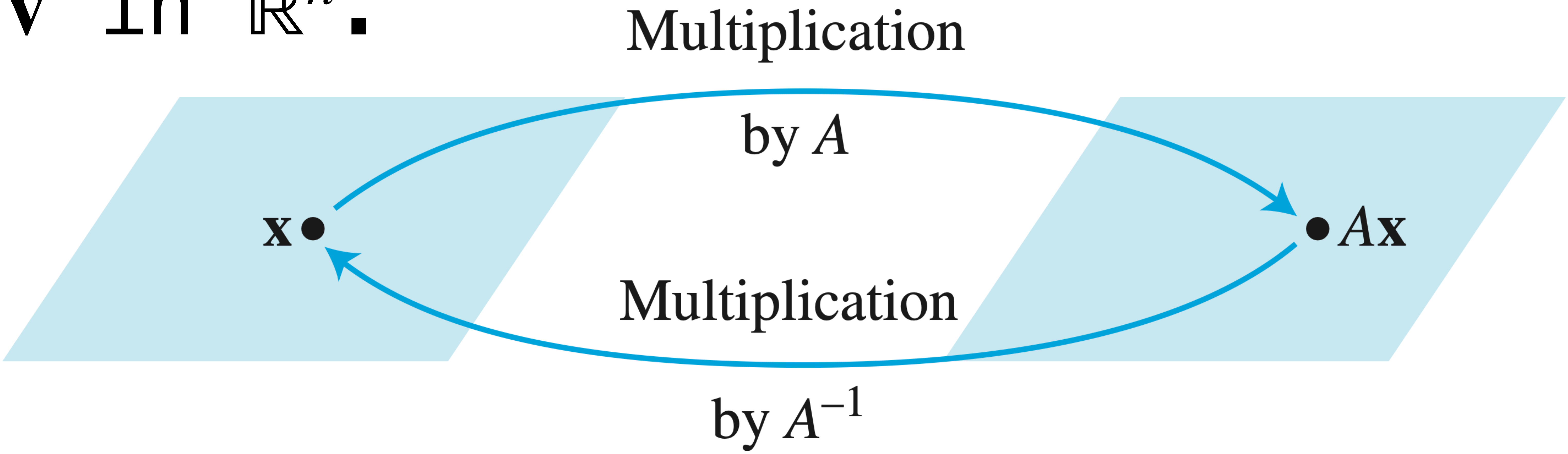
where T is implemented by A

Connection to Transformations

Definition. A linear transformation $T : \mathbb{R}^n \to \mathbb{R}^n$ is **invertible** if there is a linear transformation S such that

$$S(T(\mathbf{v})) = \mathbf{v} \quad \text{and} \quad T(S(\mathbf{v})) = \mathbf{v}$$

for any $\mathbf{v}$ in $\mathbb{R}^n$.



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Non-Example. Projection onto the x_1-axis.

Connection to Transformations

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Definition. A transformation $T: \mathbb{R}^n \to \mathbb{R}^n$ is a **one-to-one correspondence** (bijection) if any vector $\mathbf{b}$ in $\mathbb{R}^n$ is the image of **exactly** one vector $\mathbf{v}$ in $\mathbb{R}^n$ (where $T(\mathbf{v}) = \mathbf{b}$).

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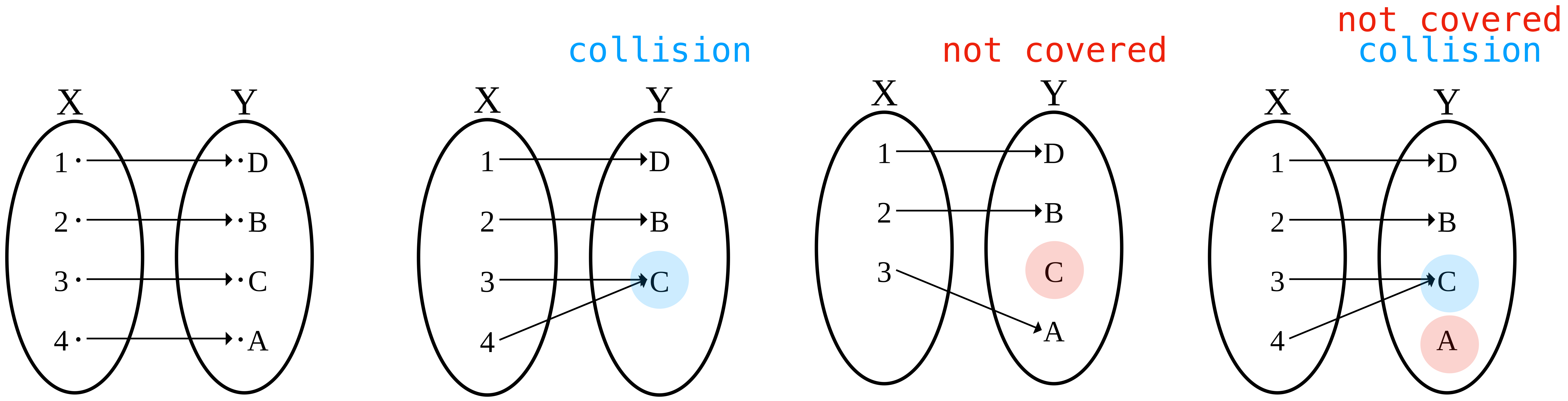
Invertible transformations are 1-1 correspondences.

Kinds of Transformations (Pictorially)



1–1 correspondence

onto, not 1–1

1–1 not onto

not 1–1, not onto

Computing Matrix Inverses

In General

$$A \begin{bmatrix} \mathbf{b}_1 & \mathbf{b}_2 & \mathbf{b}_3 \end{bmatrix} = I$$

Can we solve for each $\mathbf{b}_i$?:

How To: Matrix Inverses

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Question. Find the inverse of an invertible $n \times n$ matrix A.

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Solution. Solve the equation $A\mathbf{x} = \mathbf{e}_i$ for every standard basis vector. Put those solutions $\mathbf{s}_1, \mathbf{s}_2, \ldots, \mathbf{s}_n$ into a single matrix

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$$[\mathbf{s}_1 \quad \mathbf{s}_2 \quad \cdots \quad \mathbf{s}_n]$$

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Solution. Row reduce the matrix $[A \quad I]$ to a matrix $[I \quad B]$. Then B is the inverse of A.

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Solution. Row reduce the matrix $[A \quad I]$ to a matrix $[I \quad B]$. Then B is the inverse of A.

This is really the same thing. It's a simultaneous reduction.

How To: Matrix Inverse Computationally

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Solution. Use `numpy.linalg.inv()`

Warning: this only works if the matrix is invertible.

demo

Special Case: 2×2 Matrice Inverses

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The inverse is defined only if the determinant is nonzero.

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(see the notes on linear transformations for more information about determinants)

Example

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Is the above matrix invertible?

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Is the above matrix invertible?

No. The determinant is $(-6)(-7) - 14(3) = 42 - 42 = 0$

Algebra of Matrix Inverses

Algebraic Properties (Matrix Inverses)

Theorem. For a $n \times n$ invertible matrix A

$$(A^{-1})^{-1} = A$$

Verify:

Algebraic Properties (Matrix Inverses)

Theorem. For a $n \times n$ invertible matrix A, the matrix A^T is invertible and

$$(A^T)^{-1} = (A^{-1})^T$$

Verify:

Algebraic Properties (Matrix Inverses)

Theorem. For a $n \times n$ invertible matrices A and B, the matrix AB is invertible and

$$(AB)^{-1} = B^{-1}A^{-1}$$

Verify:

Question

Suppose that A *is a* $n \times n$ *invertible matrix such that* $A = A^T$ *and* B *is a* $m \times n$ *matrix.*

Simplify the expression $A(BA^{-1})^T$ *using the algebraic properties we've seen.*

Answer: B^T

$$A(BA^{-1})^T$$

$$A = A^T$$

Invertible Matrix Theorem

High Level

How do we know if a matrix is invertible?

By connecting everything we've said so far.

Invertible Matrix Theorem (IMT)

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6. A has n pivots (per row and per column)
7. A is row equivalent to I

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13. $\mathbf{x} \mapsto A\mathbf{x}$ is a one-to-one correspondence
14. $\mathbf{x} \mapsto A\mathbf{x}$ is invertible

We get a lot of information for free

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Theorem. If A is square, then

A **is 1–1** if and only if A **is onto**

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Warning. Remember this only applies square matrices.

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Theorem. If A is square, then

A **is invertible** $\equiv$ $A\mathbf{x} = \mathbf{0}$ **implies** $\mathbf{x} = \mathbf{0}$

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Theorem. If A is square, then

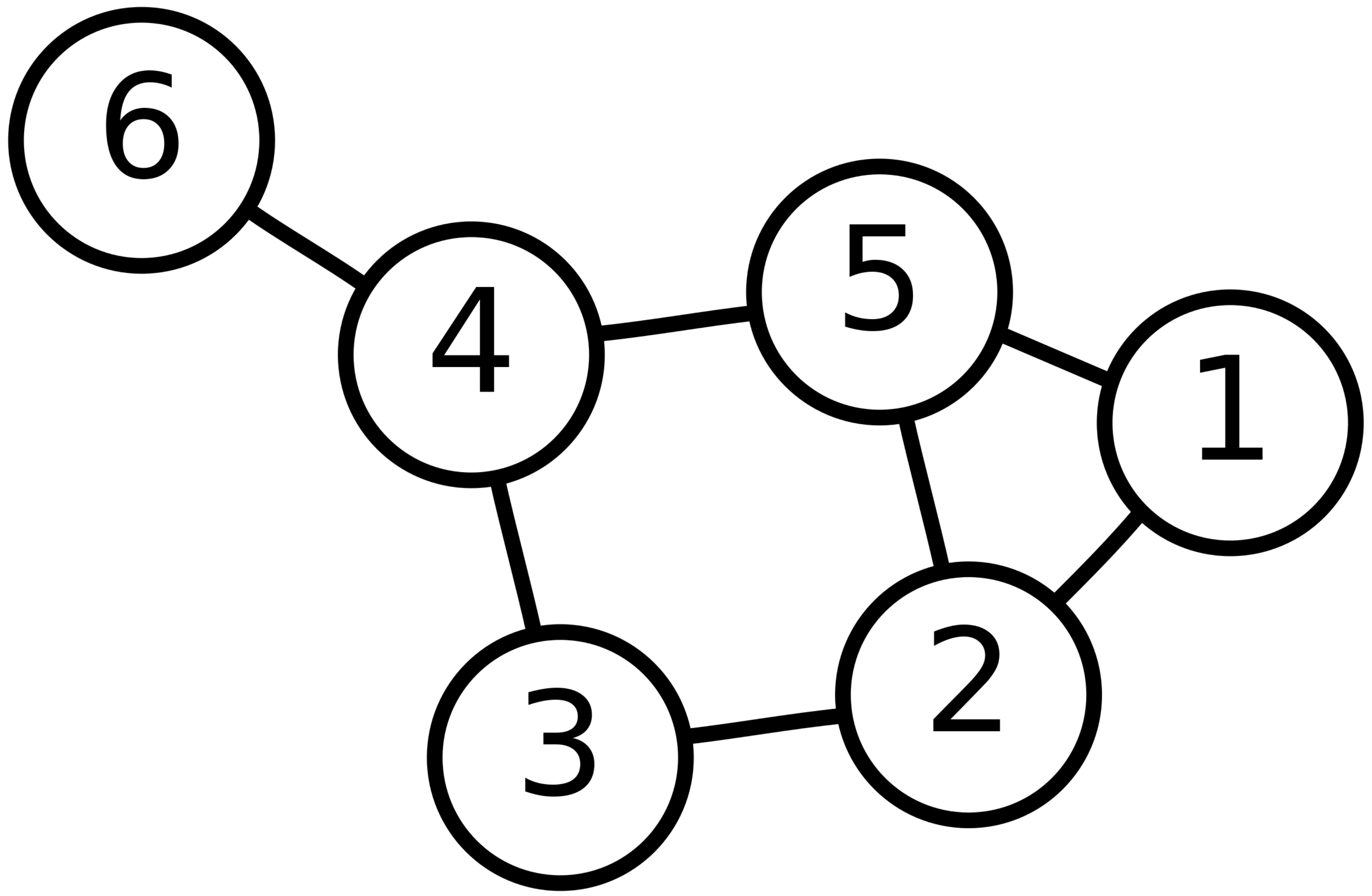
$$A \textbf{ is invertible} \quad \equiv \quad A\mathbf{x} = \mathbf{0} \textbf{ implies } \mathbf{x} = \mathbf{0}$$

Invertibility is completely determined by how A behaves on $\mathbf{0}$.

Application: Adjacency Matrices

Graphs

Definition (Informal). An **undirected graph** is a collection of nodes with edges between them.



How do we represent these in computers?

Adjacency Matrices

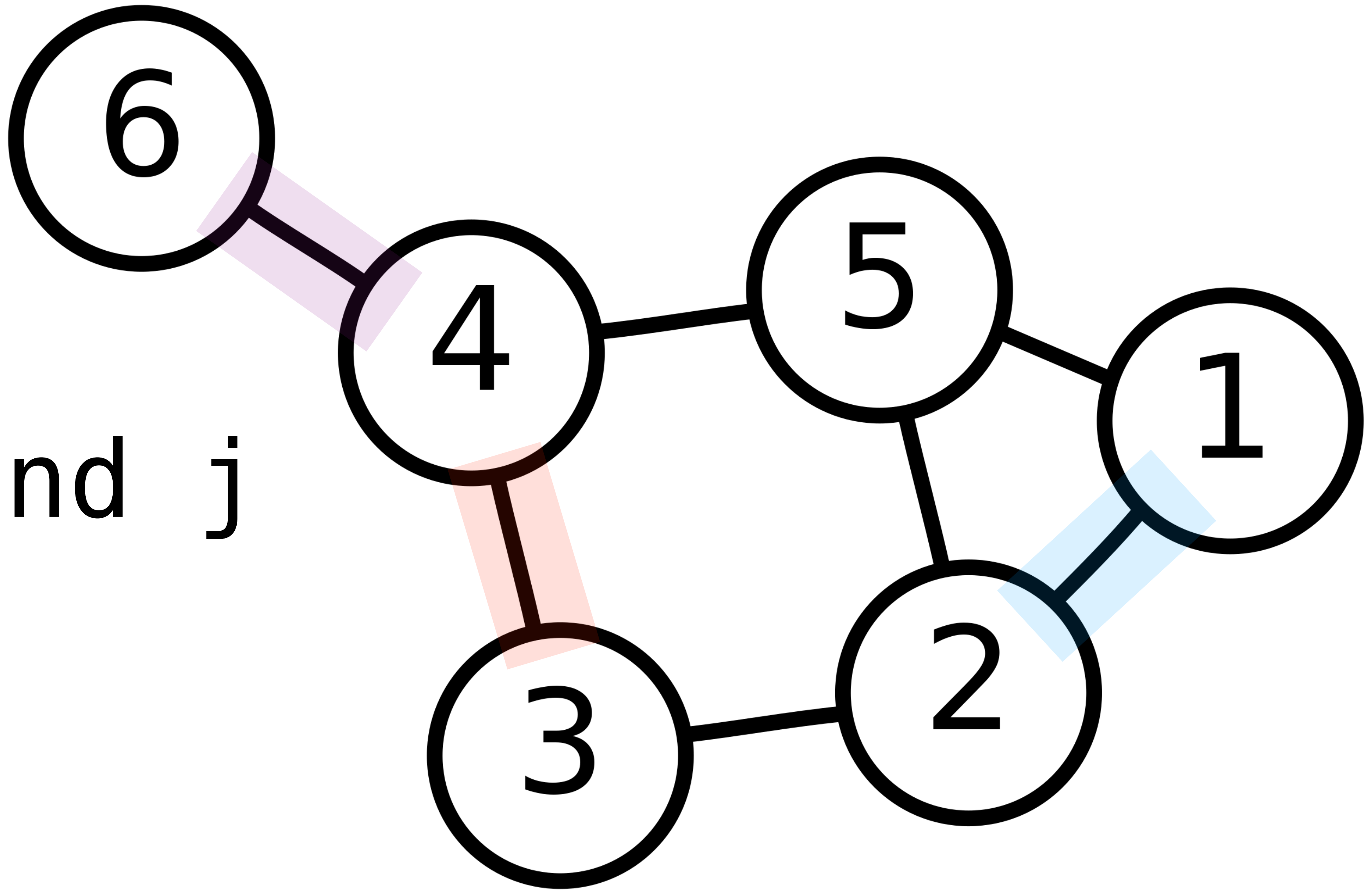
For an undirected graph G we can create the **adjacency matrix** A for G where:

$$A_{ij} = \begin{cases} 1 & \text{there is an edge between i and j} \\ 0 & \text{otherwise} \end{cases}$$

A_{12} A_{34} A_{46}

A_{21} A_{43} A_{64}

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



Spectral Graph Theory

Once we have an adjacency matrix, we can do linear algebra on graphs.

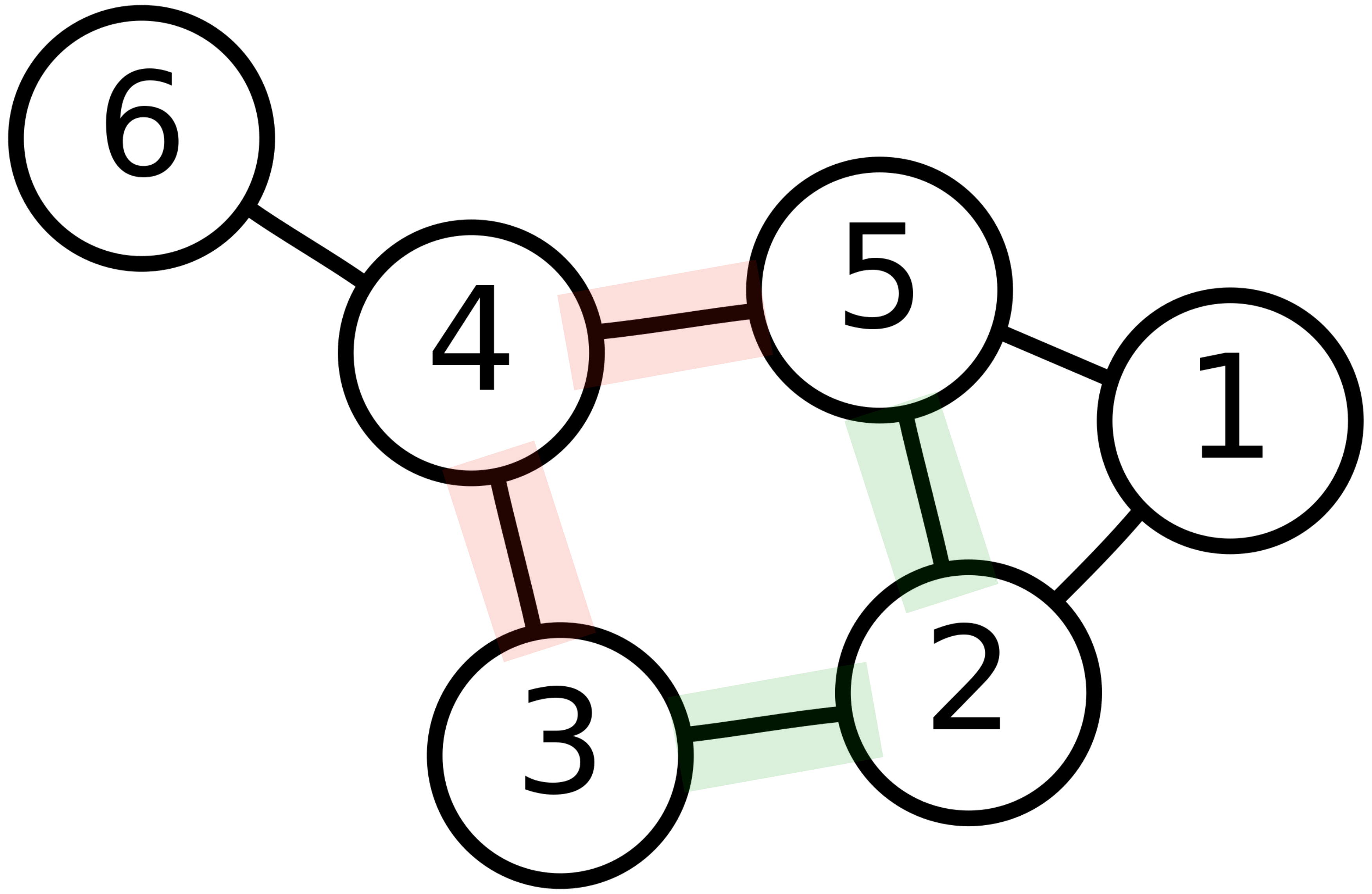
Example: Squared Adjacency Matrices

Given an adjacency matrix A

Can we interpret anything meaningful from A^2?

Example: Squared Adjacency Matrices

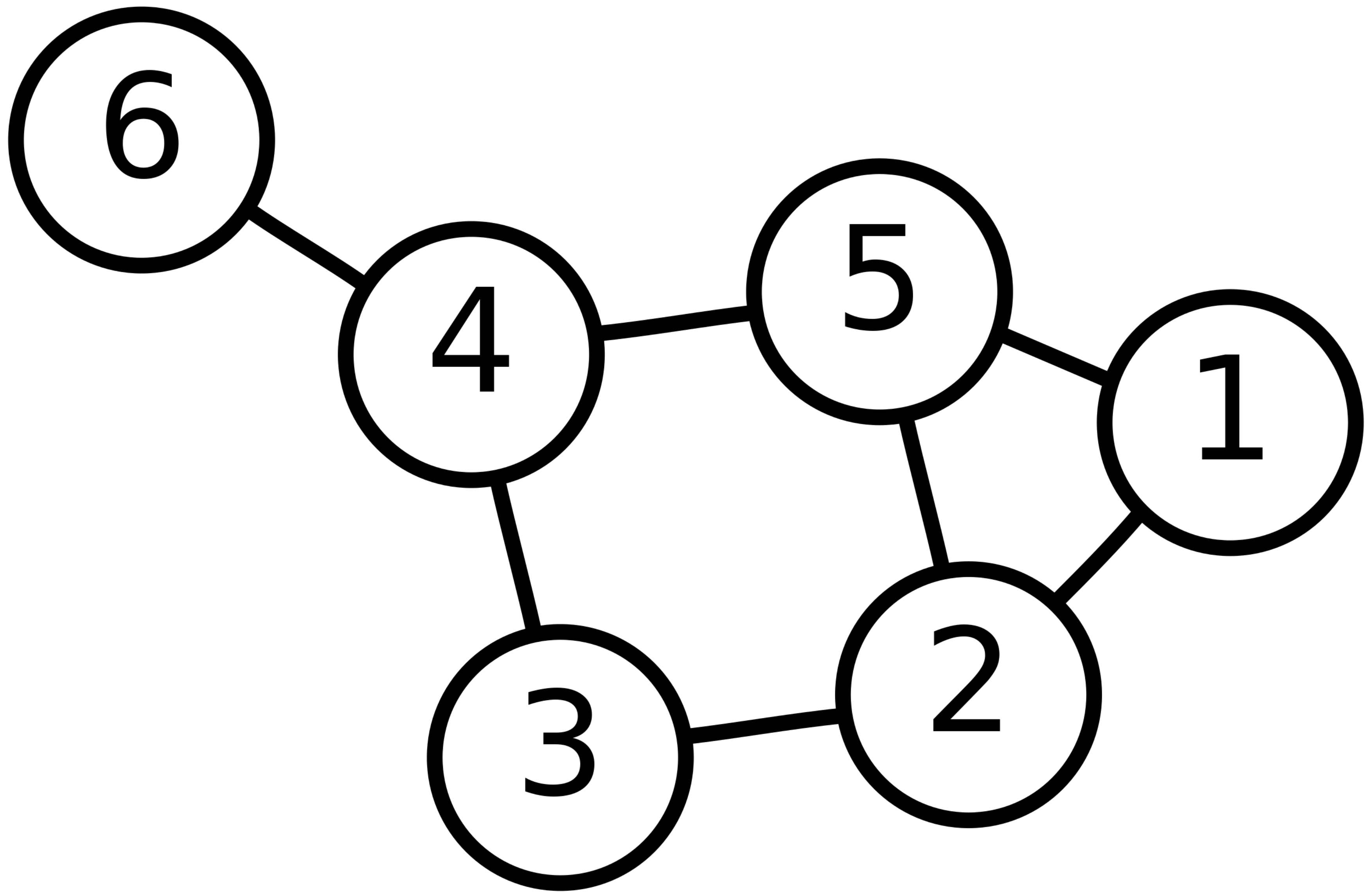
$$\begin{bmatrix} 0 & 1 & 0 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 1 \\ 1 & 1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



$$(A^2)_{53} = 1(0) + 1(1) + 0(0) + 1(1) + 0(0) + 0(0) = 2$$

Example: Squared Adjacency Matrices

$$(A^2)_{ij} = A_{i1}A_{1j} + A_{i2}A_{2j} + \ldots + A_{in}A_{nj}$$



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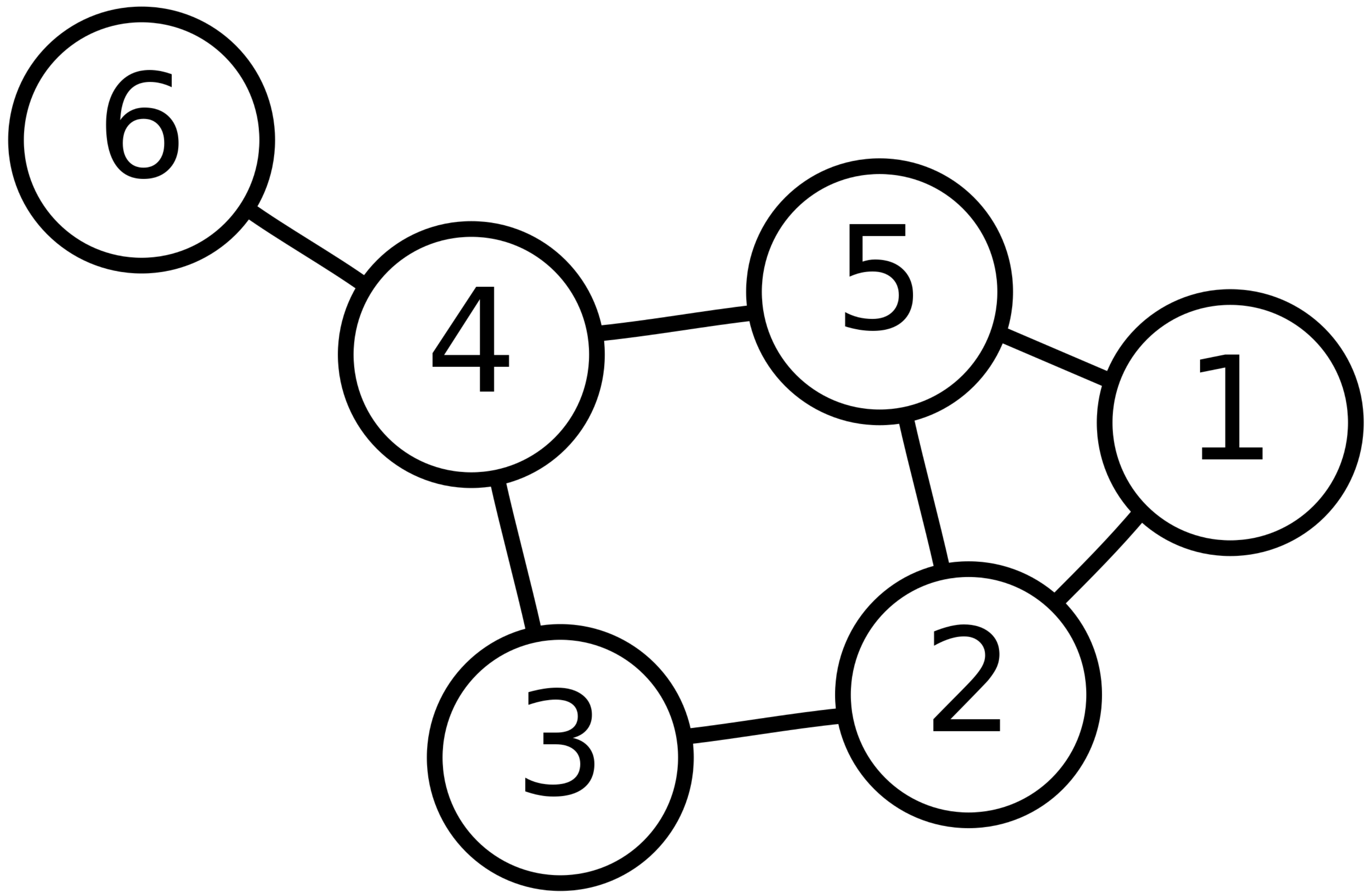
$$A_{ik}A_{kj} = \begin{cases} 1 & \texttt{there are edges from i to k and k to j} \\ 0 & \texttt{otherwise} \end{cases}$$



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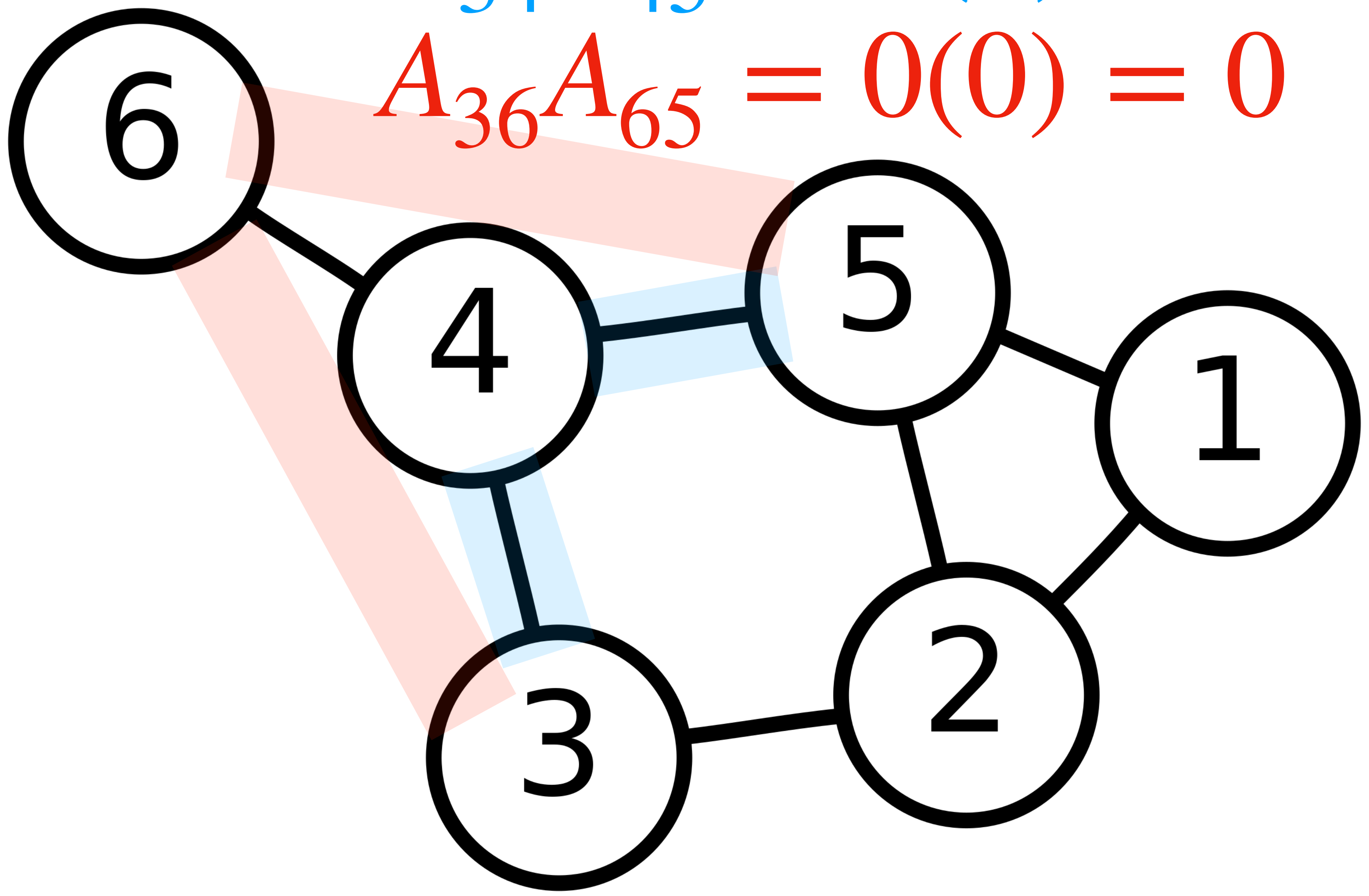

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$$(A^2)_{ij} = \boxed{\texttt{number of 2-step paths from i to j}}$$

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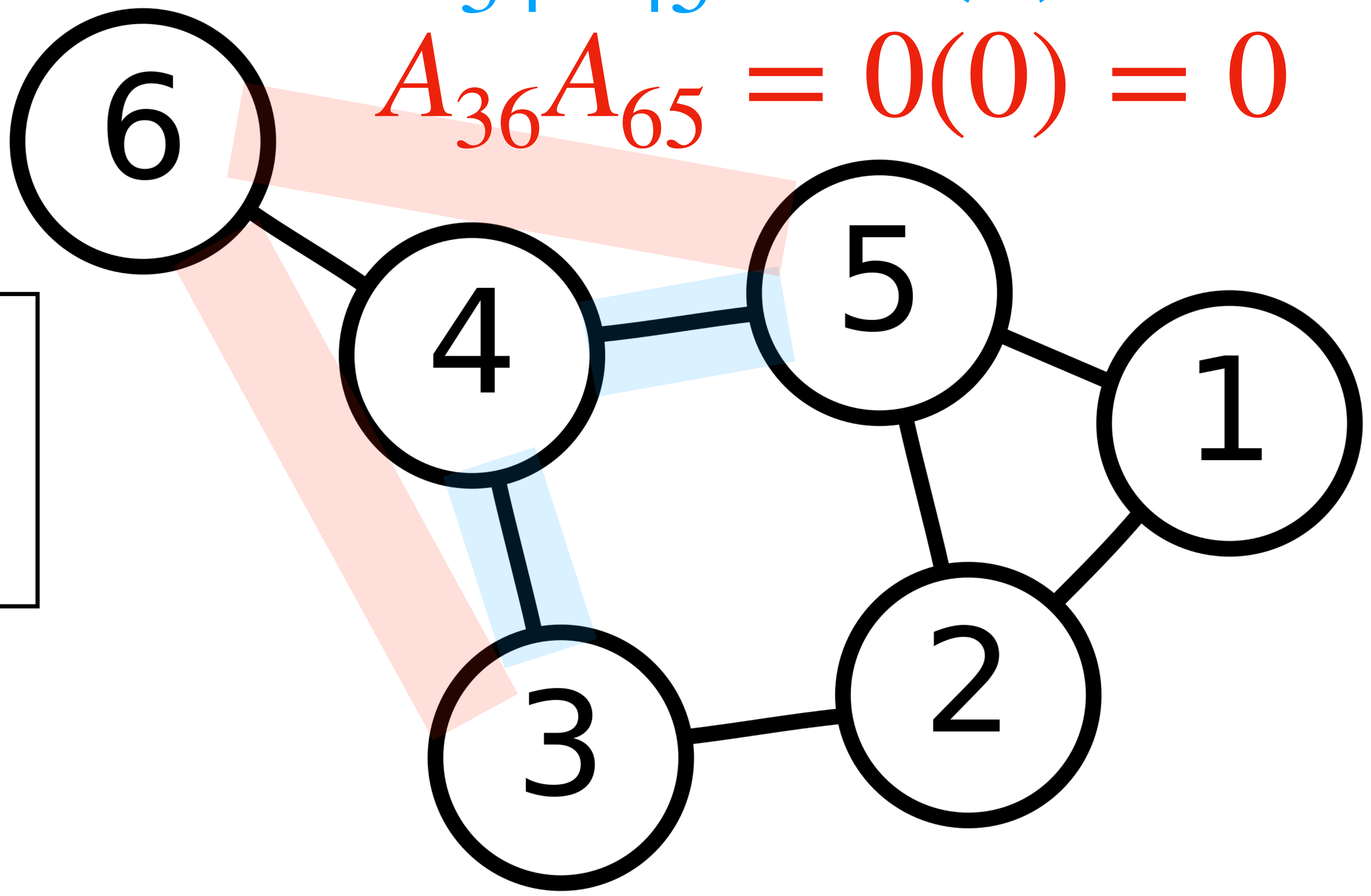
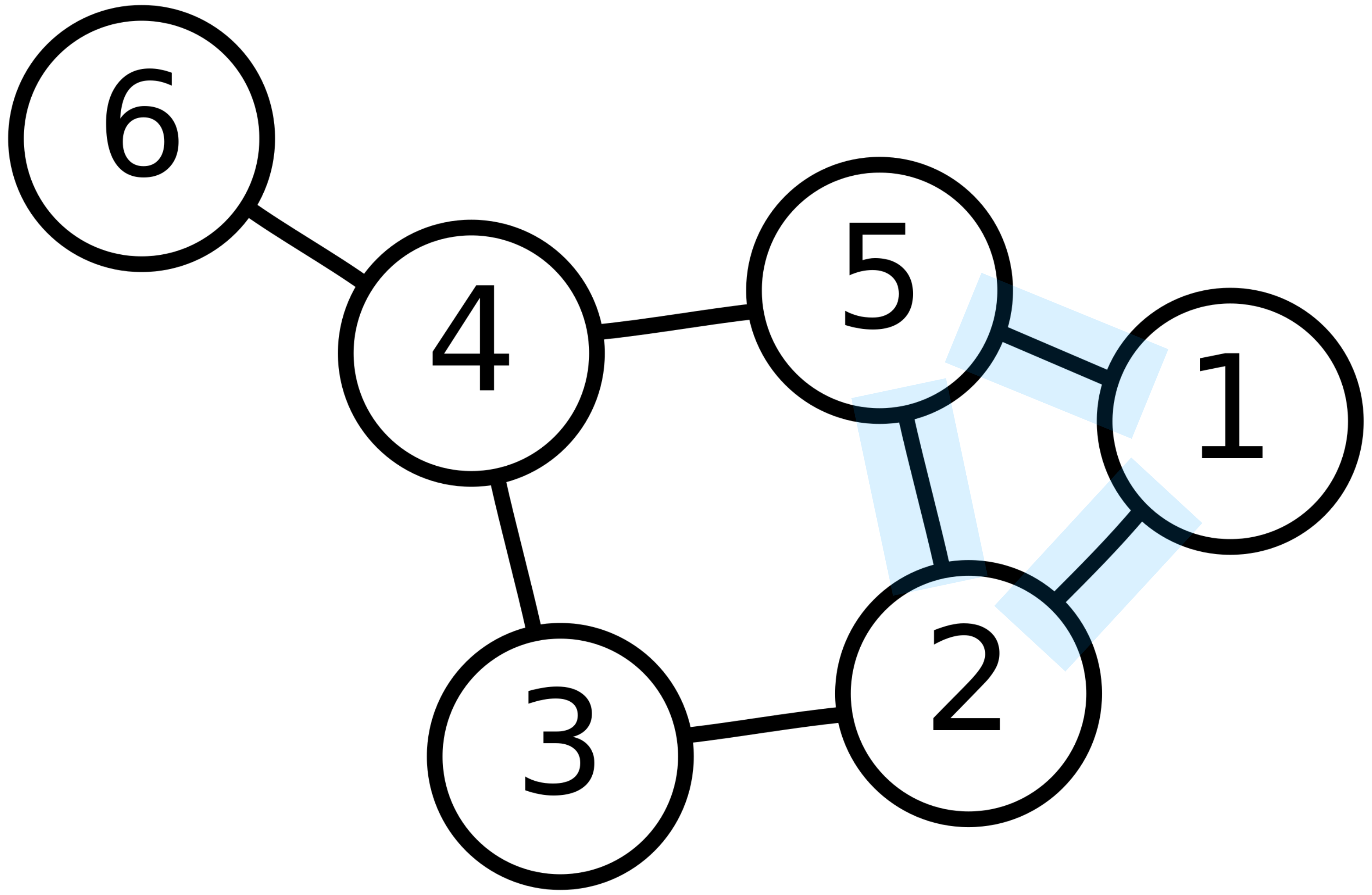


Application: Triangle Counting

A **triangle** in an undirected graph is a set of three distinct nodes with edges between every pair of nodes.

Triangles in a social network represent mutual friends and tight cohesion (among other things)



Application: Triangle Counting

Theorem. For an adjacency matrix A, the number of triangle containing the edge (i, j) is

$$(A^2)_{ij} A_{ij}$$

Application: Triangle Counting

```
FUNCTION tri_count(A):
    compute A^2
    count ← sum of (A^2)_ij A_ij for all distinct i and j
    RETURN count / 6      # why divided by 6?
```

Summary

We can solve matrix equations by inverting the matrix, though not all matrices have inverses.

We can compute matrix inverses a simultaneous row reduction.

We can connect all the concepts we've defined so far by thinking about them in terms of invertibility (for square matrices).