The Characteristic Equation Geometric Algorithms Lecture 19

CAS CS 132

Practice Problem

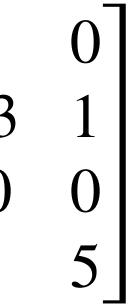
Determine the dimension of the eigenspace of A for the eigenvalue 4.

(try not to do any row reductions)

 $\begin{bmatrix} 5 & 2 & 3 & 0 \\ -1 & 2 & -3 & 1 \\ 2 & 4 & 10 & 0 \\ 1 & 2 & 3 & 5 \end{bmatrix}$



 $\begin{bmatrix} 5 & 2 & 3 & 0 \\ -1 & 2 & -3 & 1 \\ 2 & 4 & 10 & 0 \\ 1 & 2 & 3 & 5 \end{bmatrix}$



Objectives

- 2. Get a primer on <u>determinants</u>
- verify them)

1. Briefly recap eigenvalues and eigenvectors

3. Determine how to <u>find eigenvalues</u> (not just

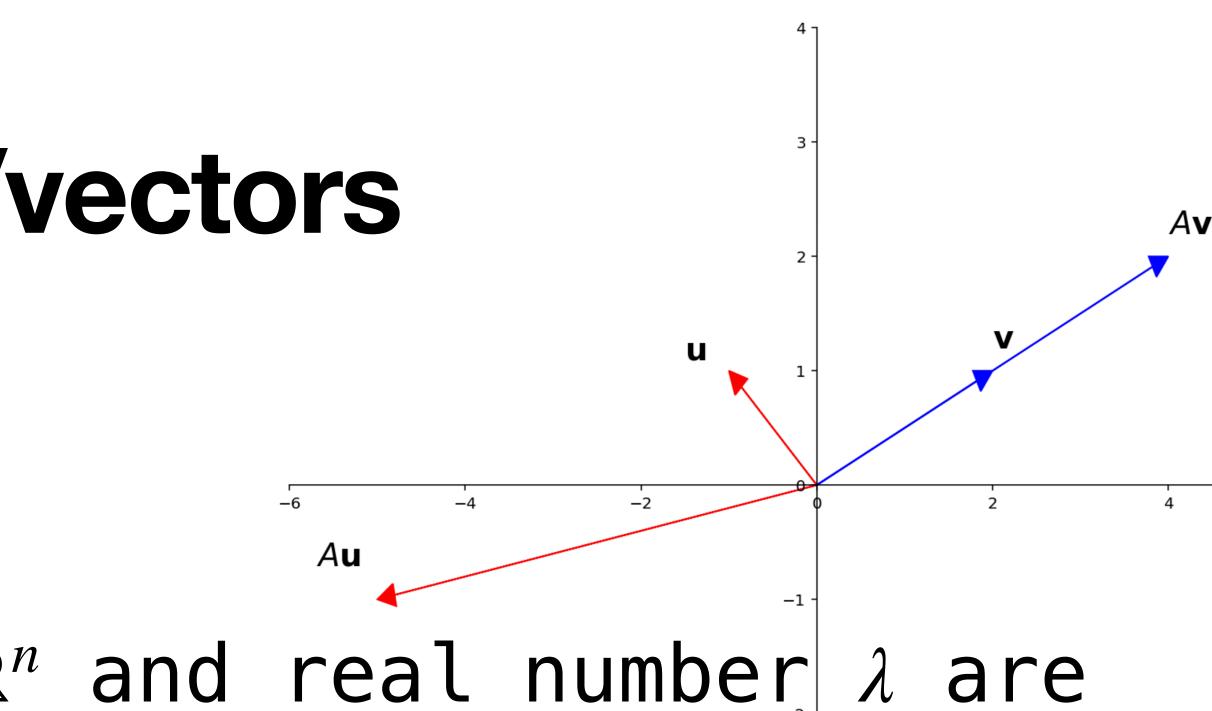
Keyword

eigenvectors eigenvalues eigenspaces eigenbases determinant characteristic equation polynomial roots triangular matrices multiplicity

Recap

Recall: Eigenvalues/vectors

A nonzero vector v in \mathbb{R}^n and real number λ are an eigenvector and eigenvalue for a $n \times n$ matrix A if

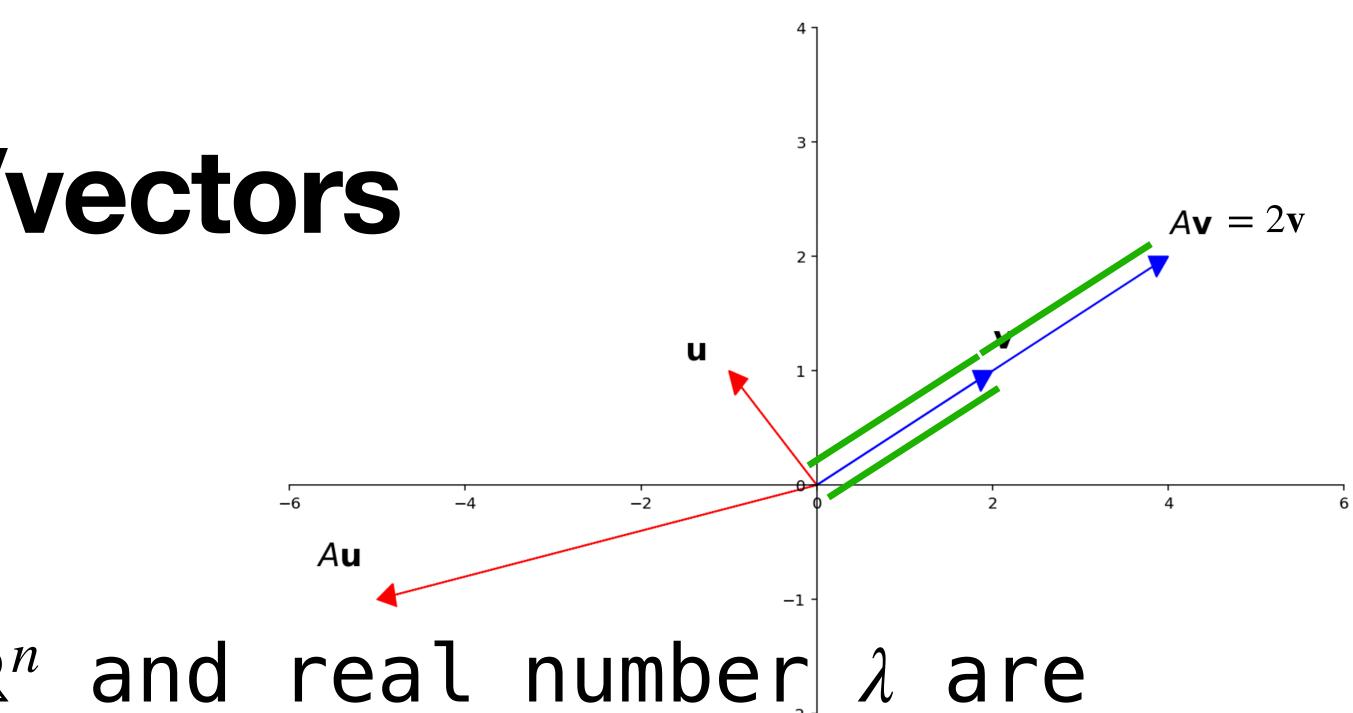


 $A\mathbf{v} = \lambda \mathbf{v}$



Recall: Eigenvalues/vectors

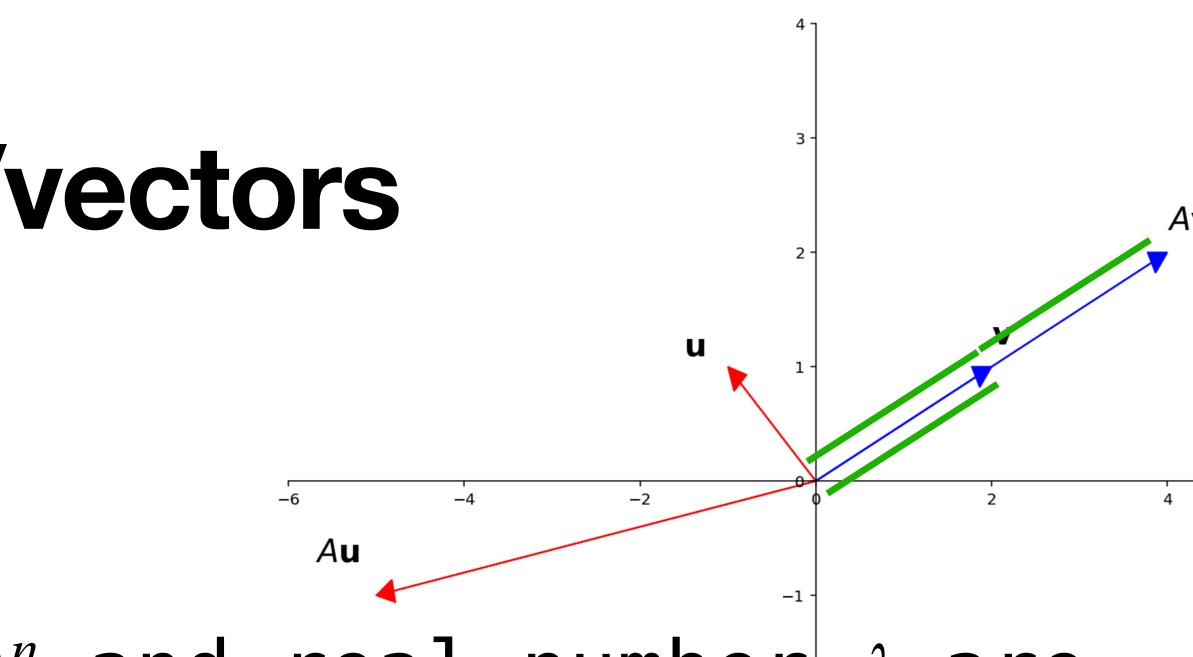
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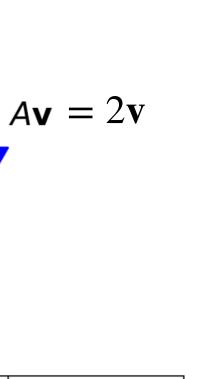
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an eigenvector and eigenvalue for a $n \times n$ matrix

 $A\mathbf{v} = \lambda \mathbf{v}$

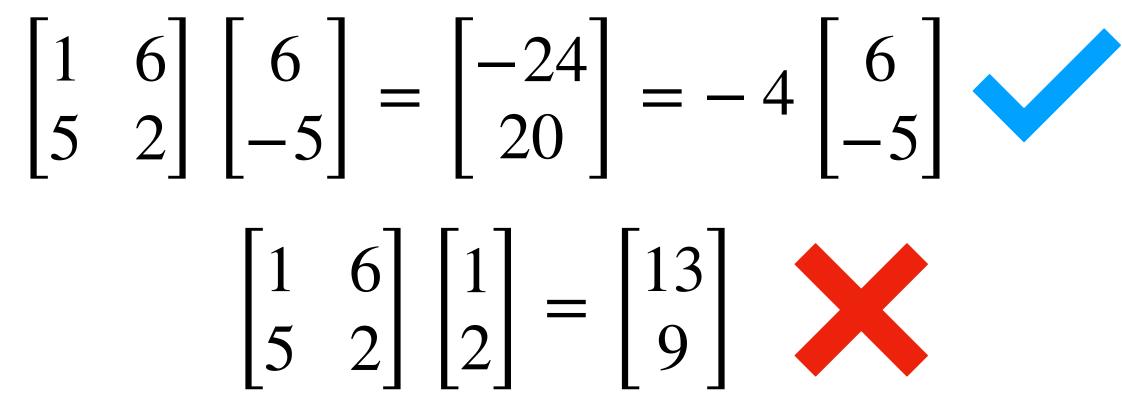
v is "just scaled" by A, not rotated



Question. Determine if v is an eigenvector of A and determine the corresponding eigenvalues.

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- Solution. Easy. Work out the matrix-vector multiplication.

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- $(A \lambda I)\mathbf{x} = \mathbf{0}$

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that $A - \lambda I$ is **not** invertible (by IMT).

- $(A \lambda I)\mathbf{x} = \mathbf{0}$
- If we don't need the vector we can just show

Question. Find a basis for the eigenspace of A corresponding to λ .

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(we did this for our recap problem)

How do eigenvectors relate to linear dynamical systems?

Recall: (Closed-Form) Solutions

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A (closed-form) solution of a linear dynamical system $\mathbf{v}_{i+1} = A\mathbf{v}_i$ is an expression for \mathbf{v}_k which is does not contain A^k or previously defined terms

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A (closed-form) solution of a linear dynamical system $\mathbf{v}_{i+1} = A\mathbf{v}_i$ is an expression for \mathbf{v}_k which is does not contain A^k or previously defined terms

In other word, it does not depend on A^k and is not **recursive**

Example $\mathbf{v}_{k} = \begin{vmatrix} 1 & 1 \\ 1 & 0 \end{vmatrix} \mathbf{v}_{k-1} \qquad \mathbf{v}_{0} = \begin{vmatrix} 1 \\ 1 \end{vmatrix}$

Determine a closed for the above linear dynamical system.



initial state is an eigenvector:

- It's easy to give a closed-form solution if the
 - $\mathbf{v}_k = A^k \mathbf{v}_0 = \lambda^k \mathbf{v}_0$



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It's easy to give a closed-form solution if the No dependence on A^k or \mathbf{v}_{k-1} $\mathbf{v}_k = A^k \mathbf{v}_0 = \lambda^k \mathbf{v}_0$



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eigenvectors.

- It's easy to give a closed-form solution if the
 - No dependence on A^k or \mathbf{v}_{k-1} $\mathbf{v}_k = A^k \mathbf{v}_0 = \lambda^k \mathbf{v}_0$
- <u>The Key Point.</u> This is still true of sums of



Solutions in terms of eigenvectors

Let's simplify $A^k \mathbf{v}$, given we have eigenvectors $\mathbf{b}_1, \mathbf{b}_2$ for A which span all of \mathbb{R}^2 :

Eigenvectors and Growth in the Limit

if \mathbf{v}_0 can be written in terms of eigenvectors $\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_k$ of A with eigenvalues

term, the system grows exponentially in λ_1). Verify:

- **Theorem.** For a linear dynamical system A with initial state \mathbf{v}_0 ,
 - $\lambda_1 > \lambda_2 \dots \geq \lambda_k$
- then $\mathbf{v}_k \sim \lambda_1^k c_1 \mathbf{b}_1$ for some constant c_1 (in other words, in the long

Definition. An eigenbasis of \mathbb{R}^n for a $n \times n$ eigenvectors of A.

matrix A is a basis of \mathbb{R}^n made up entirely of

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Not all matrices have eigenbases.

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Eigenbases and Growth in the Limit

Theorem. For a linear dynamical system A with

eigenvalue of A and b_1 is its eigenvalue.

- initial state v_0 , if A has an eigenbasis $b_1, ..., b_k$, then
 - $\mathbf{v}_k \sim \lambda_1^k c_1 \mathbf{b}_1$
- for some constant c_1 , where where λ_1 is the largest

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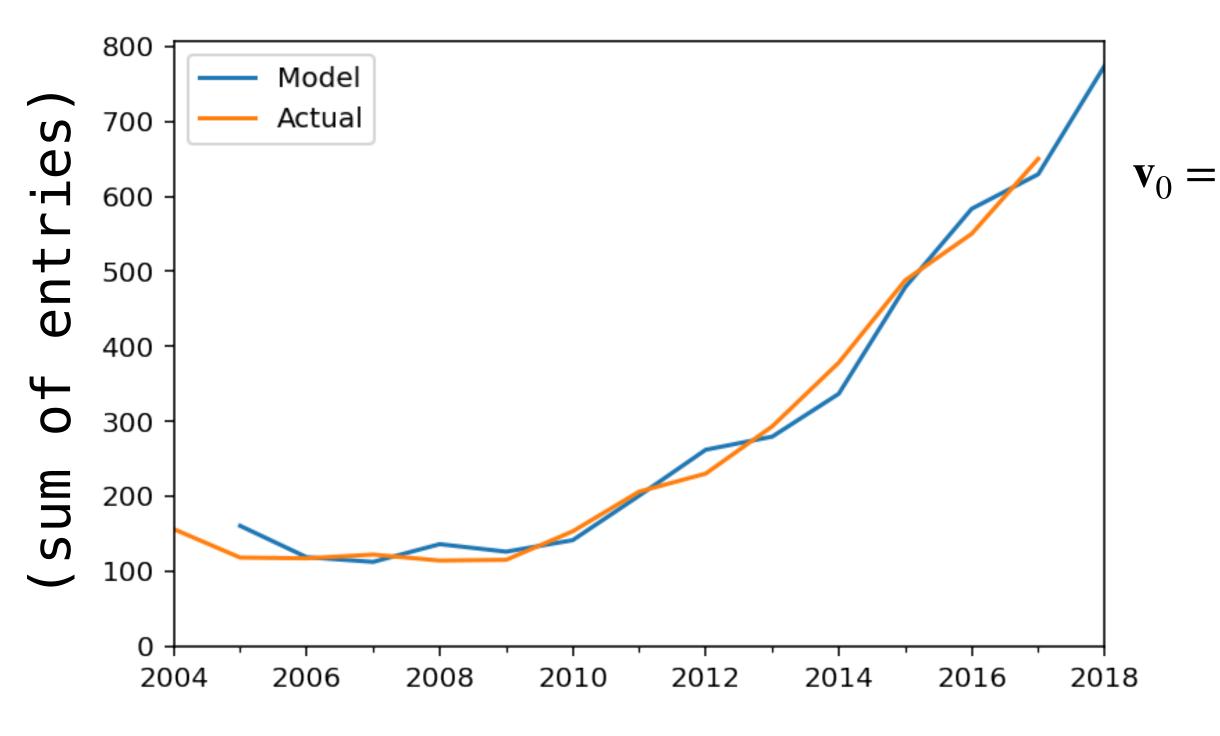
initial state v_0 , if A has an eigenbasis b_1, \ldots, b_k , then

$$\mathbf{v}_k \sim \lambda_1^k c_1 \mathbf{b}_1$$

- for some constant c_1 , where where λ_1 is the **largest**
 - The largest eigenvalue describes the long-term exponential behavior of the system.

Example: CS Major Growth

see the notes for more details



This is clearly exponential. If we want to "extract" the exponent, we need to look at the <u>largest eigenvalue</u>.

$v_{0,1}$							enrolled		
$v_{0,2}$ $v_{0,3}$ $v_{0,4}$	=	#	of	year	2	students	enrolled	in	20
<i>v</i> _{0,3}							enrolled		
$v_{0,4}$		#	of	year	4	students	enrolled	in	20

$$\mathbf{v}_k = A^k \mathbf{v}_0$$

(A is determined by least squares)

024 024 024 024

moving on...

Finding Eigenvalues

Finding Eigenvalues

Question. Determine the eigenvalues of A, along with their associated eigenspaces.

Finding Eigenvalues

with their associated eigenspaces.

in the equation

Question. Determine the eigenvalues of A, along

Solution (Idea). Can we somehow "solve for λ "

 $(A - \lambda I)\mathbf{x} = \mathbf{0}$

Determinants

An Aside: Determinants are Mysterious

Determinants are strangely polarizing

Some people love them, some people hate them

We'll only scratch the surface...

Down with Determinants!

Sheldon Axler

det

 \wedge

102 (1995), 139-154.

ry writing from the Mathematical Association of America.

without determinants. The standard proof that a square matrix of complex numbers has an eigenvalue uses or erminants, this allows us to define the multiplicity of an eigenvalue and to prove that the number of eigenva haracteristic and minimal polynomials and then prove that they behave as expected. This leads to an easy p determinants, this paper gives a simple proof of the finite-dimensional spectral theorem.

this paper. The book is intended to be a text for a second course in linear algebra.

A determinant is a number associated with a matrix.

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Notation. We will write det(A) for the determinant of A.

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entries of A.

In broad strokes, it's a big sum of products of

A Scary-Looking Definition (we won't use)

$$det(A) = \sum_{\sigma \in S_n} (-1)$$

We can think of this function as a procedure:

```
1 FUNCTION det(A):
    total = 0
3
      s = 1 IF (# of swaps necessary) is even ELSE -1
4
5
6
    RETURN total
```

 $^{\text{sgn}(\sigma)}A_{1\sigma(1)}A_{2\sigma(2)}\dots A_{n\sigma(n)}$

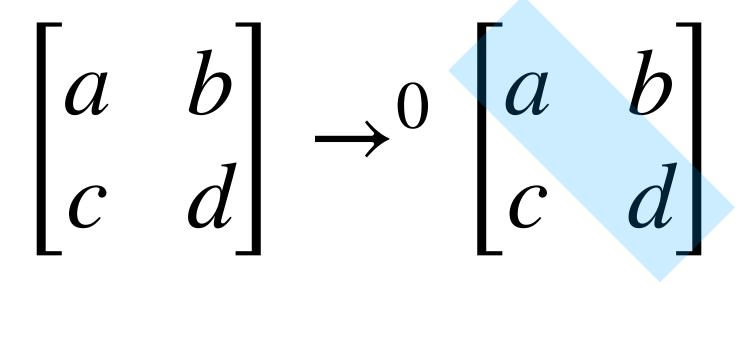
FOR all matrix B we can get by swapping a bunch of rows of A: total += s * (product of the diagonal entries of B)

The Determinant of 2×2 Matrices

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

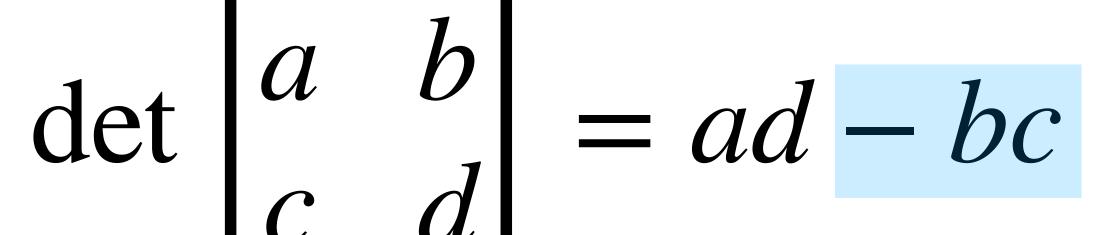
The Determinant of 2×2 Matrices

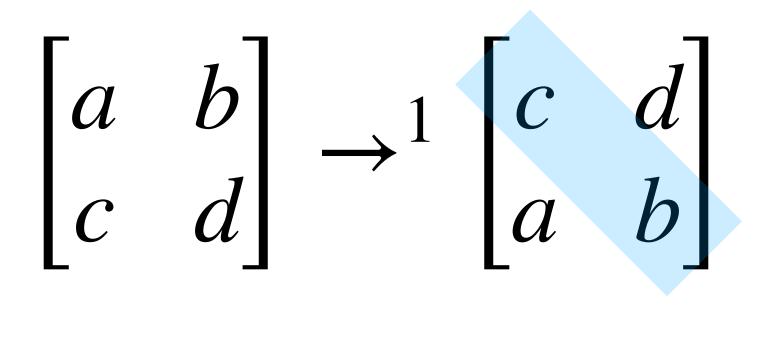
$\det \begin{vmatrix} a & b \\ c & d \end{vmatrix} = \frac{ad}{b} - bc$



 $(-1)^{0}ad$

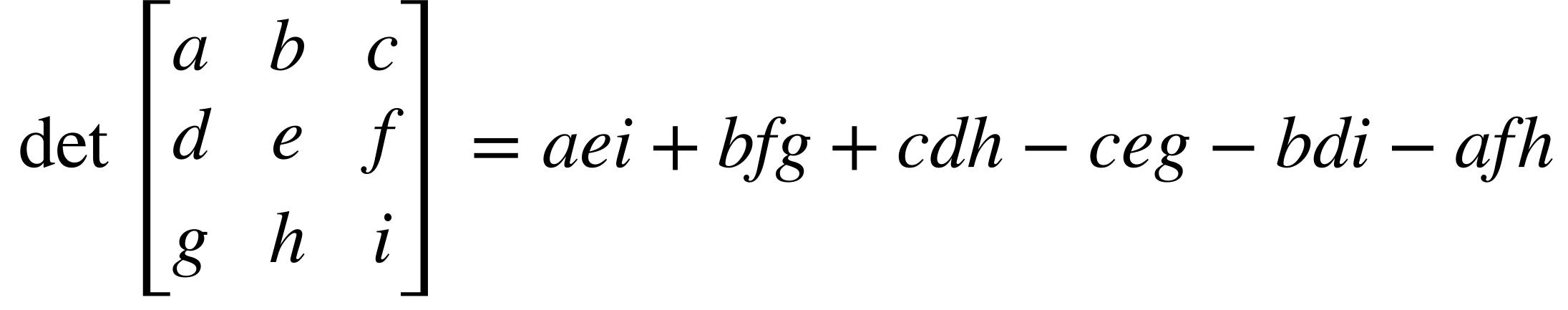
The Determinant of 2×2 Matrices



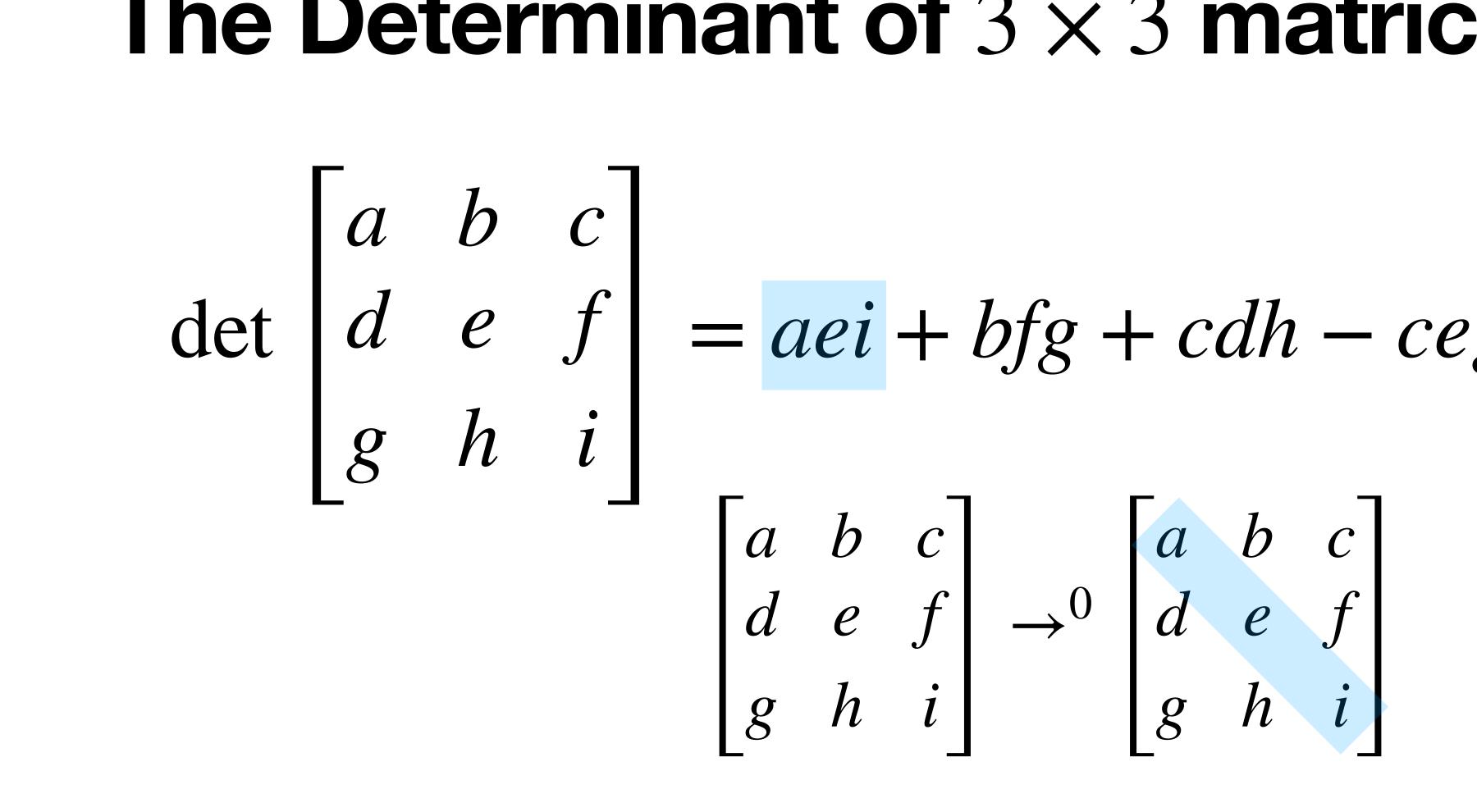


 $(-1)^{1}cb$

The Determinant of 3 × 3 matrices



The Determinant of 3 x 3 matrices

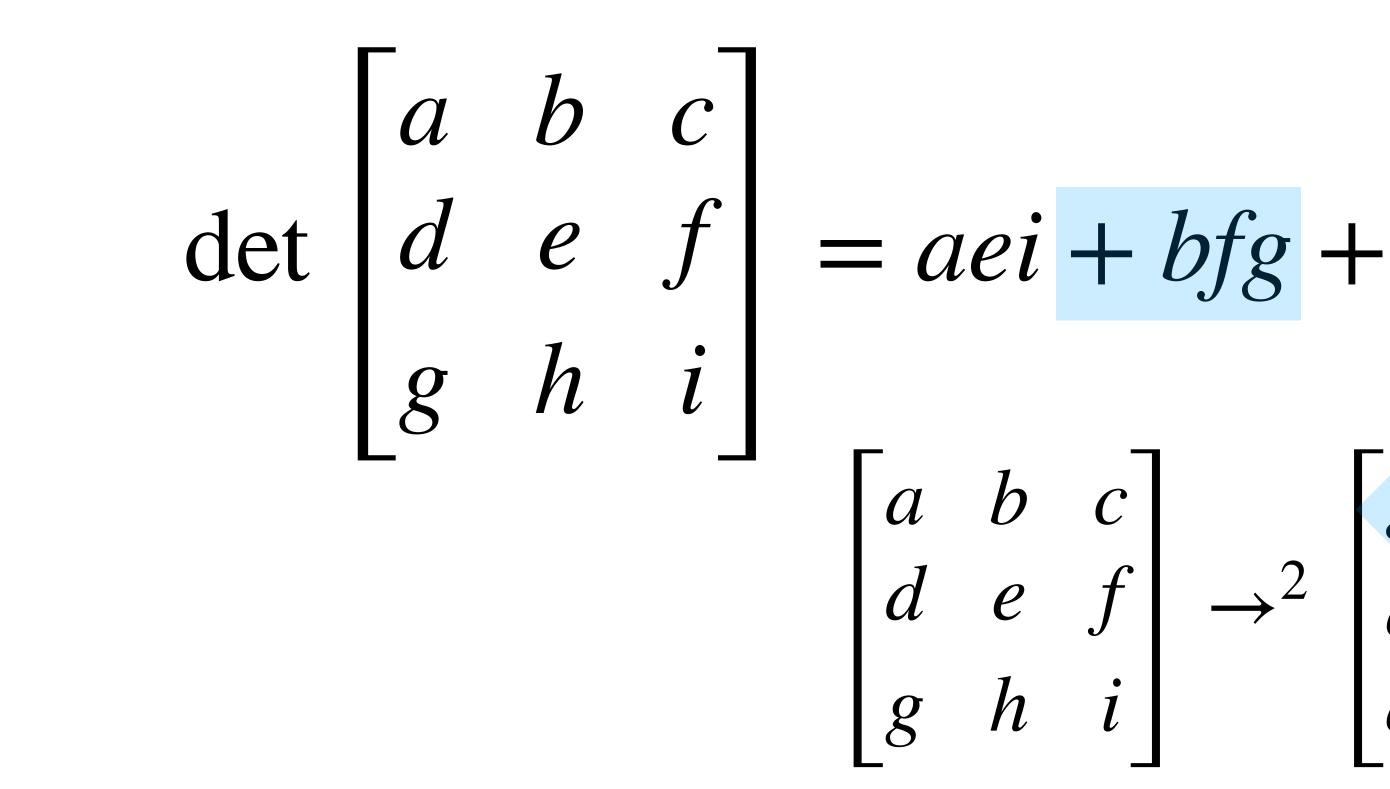


$$bfg + cdh - ceg - bdi - afh$$

$$\rightarrow^{0} \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix}$$

 $(-1)^{0}aei$

The Determinant of 3×3 matrices

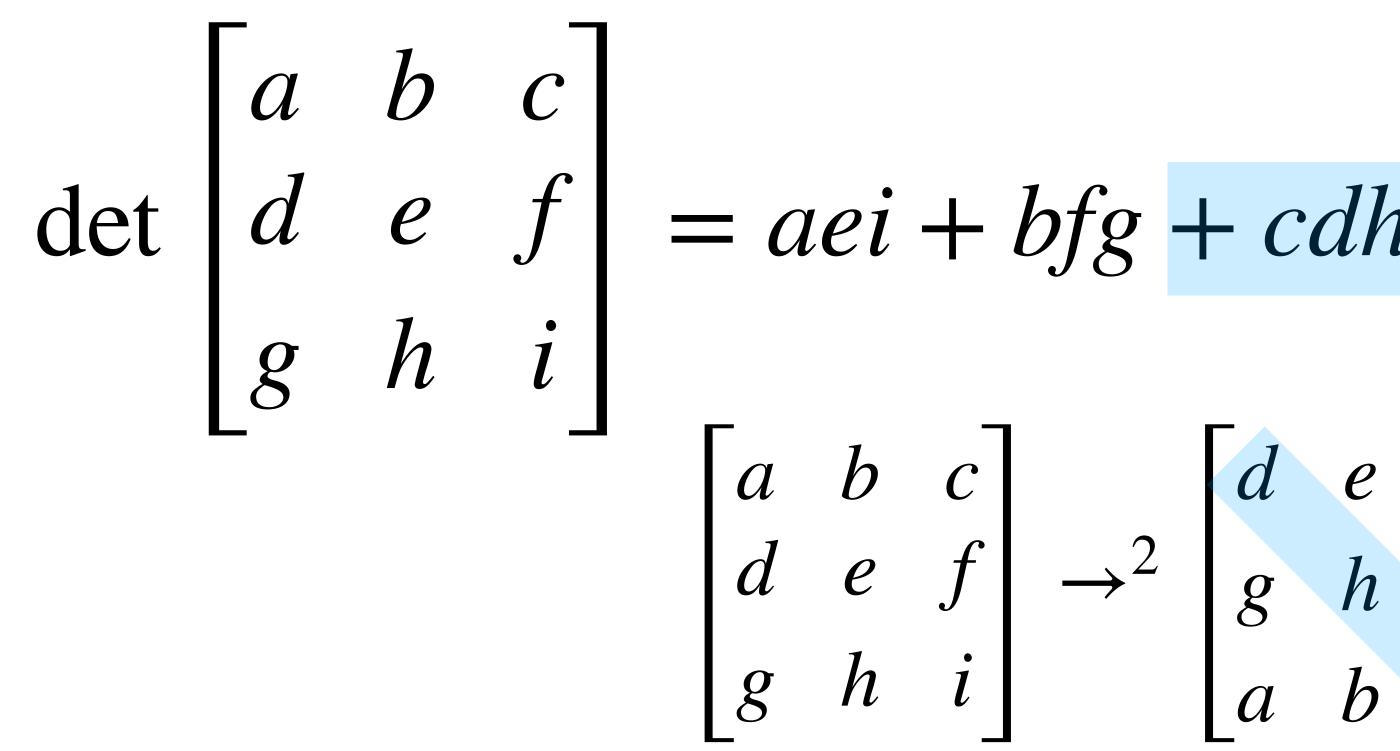


$$bfg + cdh - ceg - bdi - afh$$

$$\rightarrow^{2} \begin{bmatrix} g & h & i \\ a & b & c \\ d & e & f \end{bmatrix}$$

 $(-1)^2 gbf$

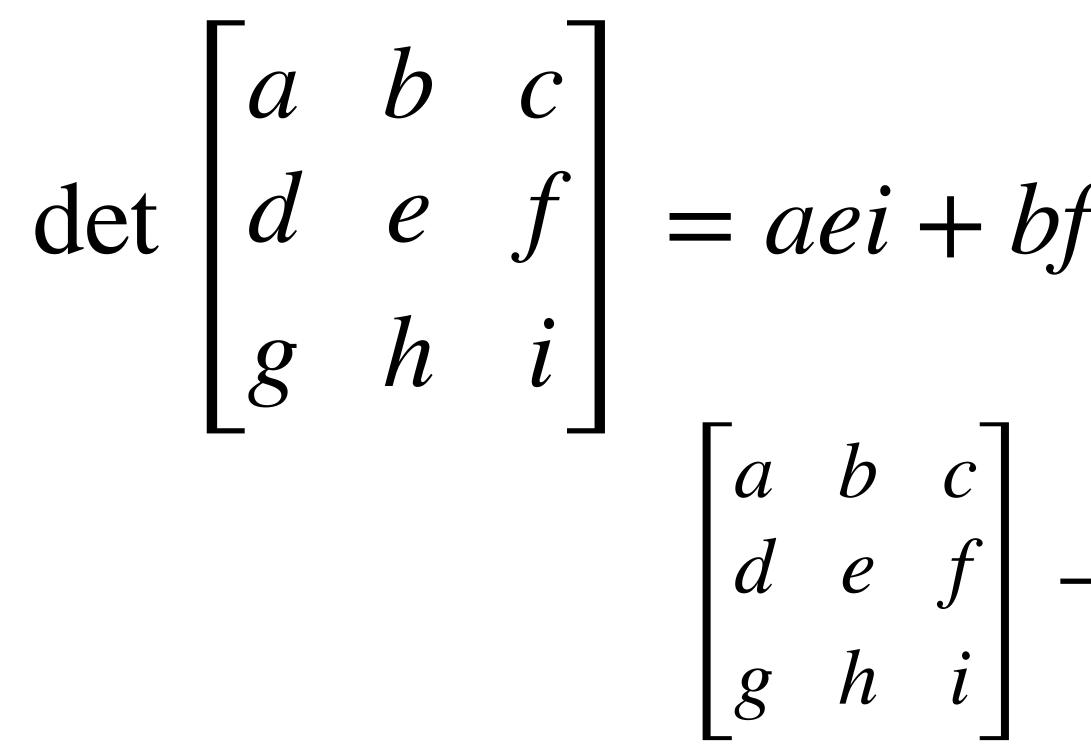
The Determinant of 3 x 3 matrices



$$\rightarrow^{2} \begin{bmatrix} d & e & f \\ g & h & i \\ a & b & c \end{bmatrix}$$

 $(-1)^2 dhc$

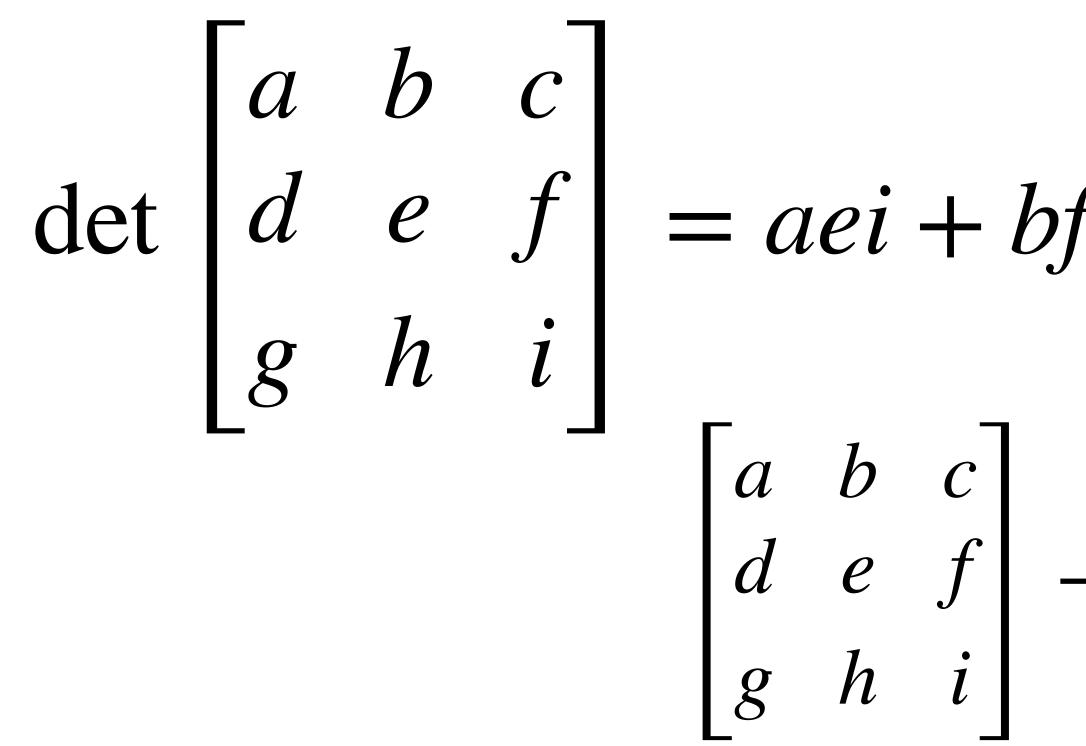
The Determinant of 3×3 matrices



$$\rightarrow^{1} \begin{bmatrix} g & h & i \\ d & e & f \\ a & b & c \end{bmatrix}$$

 $(-1)^1 gec$

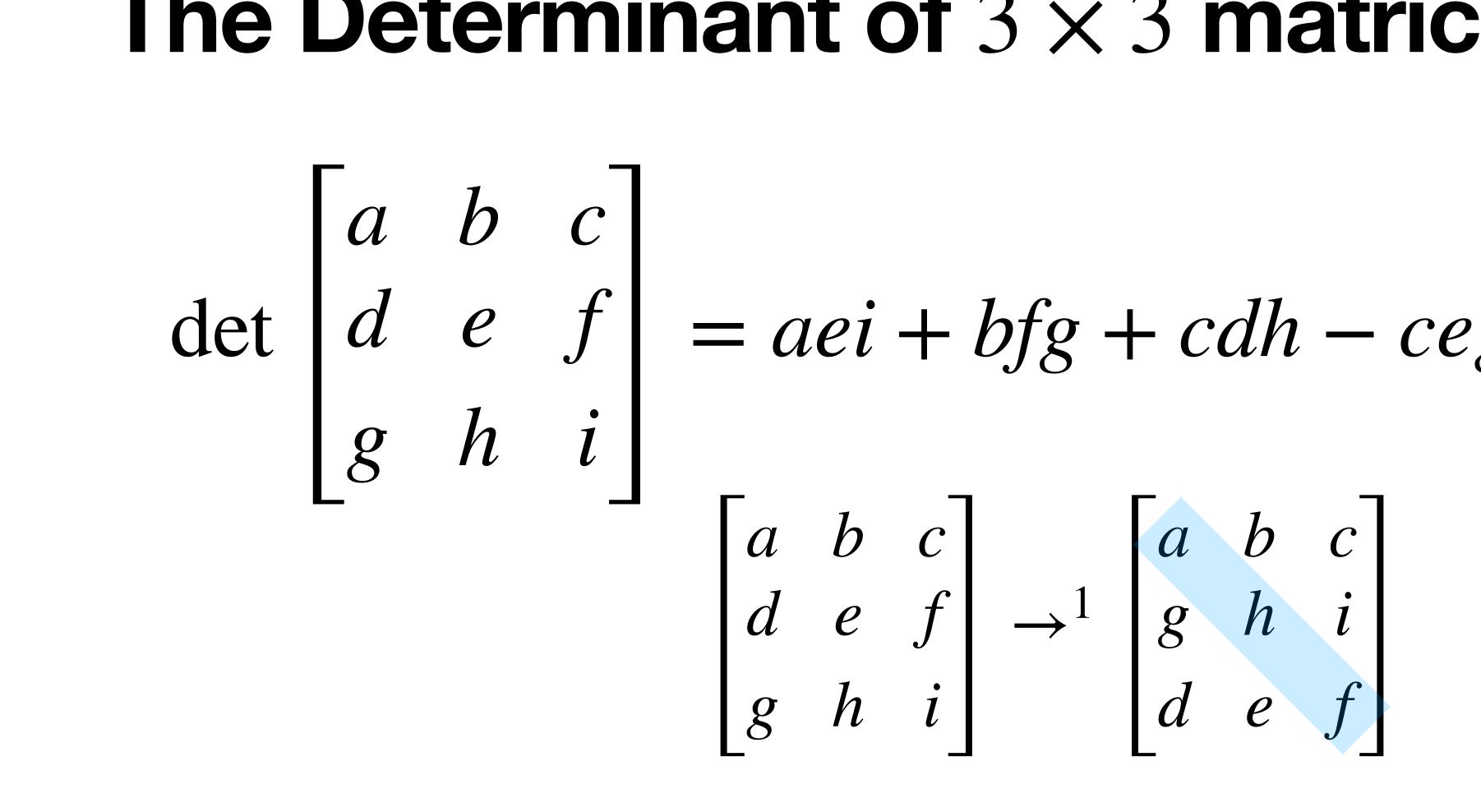
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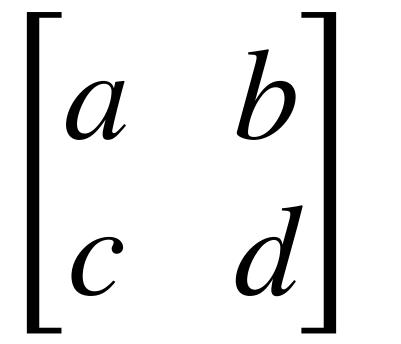


$$\rightarrow^{1} \begin{bmatrix} a & b & c \\ g & h & i \\ d & e & f \end{bmatrix}$$

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Another Perspective

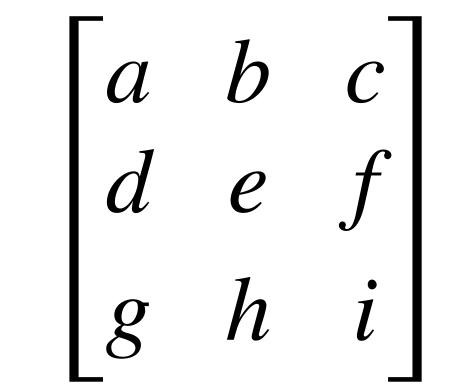
Let's row reduce an arbitrary 2×2 matrix:





Another Perspective

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Theorem. A matrix is i $det(A) \neq 0$.

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So we can yet again extend the IMT:

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- $det(A) \neq 0$.
- So we can yet again extend the IMT:
- » A is invertible
- $\Rightarrow \det(A) \neq 0$
- » 0 is not an eigenvalue

These must be all true or all false.

Theorem. A matrix is invertible if and only if

Determinants (the definition we'll use) $det(A) = \frac{(-1)^s}{c} U_{11} U_{22} \dots U_{nn}$

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by the above equation, where

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Determinants (the definition we'll use) $det(A) = \frac{(-1)^{s} \text{ product of diagonal entries}}{U_{11}U_{22}...U_{nn}}$ C 0 if A is not invertible

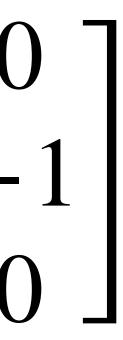
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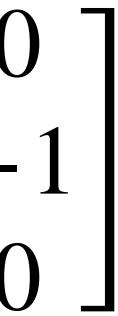


$\begin{bmatrix} 1 & 5 & 0 \\ 2 & 4 & -1 \\ 0 & -2 & 0 \end{bmatrix}$ Example Let's find the determinant of this matrix:



Example (Again)

$\begin{bmatrix} 1 & 5 & 0 \\ 2 & 4 & -1 \\ 0 & -2 & 0 \end{bmatrix}$ Let's find the determinant of this matrix again but with a different sequence of row operations:



The definition holds no matter which sequence of row operations you use.

Question. Determine the determinant of a matrix A. Solution.

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- 2. Keep track of the number of row swaps you used, call this s, and the product of all scalings, call this c
- 3. Determine the product of entries along the diagonal of U_{i} , call this P.
- 4. The determinant of A is $\frac{(-1)^{s}P}{d}$.

The Shorter Version

Beyond small matrices, we'll just use a computer With NumPy:

numpy.linalg.det(A)

Properties of Determinants

Properties of Determinants (1) det(AB) = det(A) det(B)

It follows that AB is
A and B are invertible
(we won't verify this)

It follows that AB is invertible if and only if

Example Question

Use the fact that det(AB) = det(A) det(B) to give an expression for $det(A^{-1})$ in terms of det(A). Hint. What is det(I)?

Properties of Determinants (2)

A is invertible.

(we also won't verify this)

$det(A^T) = det(A)$

It follows that A^T is invertible if and only if

Example Question

If $A^{-1} = A^T$, then what are the possible values of det(A)?



Properties of Determinants (3)

product of entries along the diagonal. Verify:

Theorem. If A is triangular, then det(A) is the



Find the determinant of the above matrix.

$\begin{bmatrix} 1 & 5 & -4 \\ -1 & -5 & 5 \\ -2 & -8 & 7 \end{bmatrix}$



Characteristic Equation



The determinant of a matrix A is an <u>arithmetic</u> <u>expression</u> written in terms of the entries of A.



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But a matrix may not have numbers as entries. We might think of the matrix $A - \lambda I$ has having polynomials as entries.

Then $det(A - \lambda I)$ is a **polynomial**.





A root of a polynomial p(x) is a value r such that p(r) = 0.

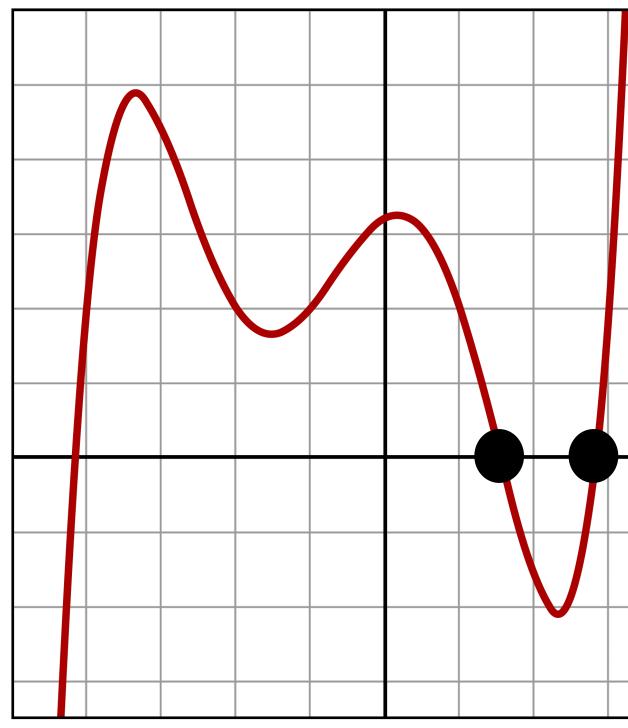




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- A root of a polynomial p(x) is a value r such that p(r) = 0.
- (A polynomial may have many roots)
- If r is a root of p(x), then it is possible to find a polynomial q(x)such that
 - p(x) = (x r)q(x)





Definition. The characteristic polynomial of a variable λ .

matrix A is $det(A - \lambda I)$ viewed as a polynomial in the

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roots.

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This is a polynomial with the eigenvalues of A as

Definition. The characteristic polynomial of a variable λ .

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So we can "solve" for the eigenvalues in the equation

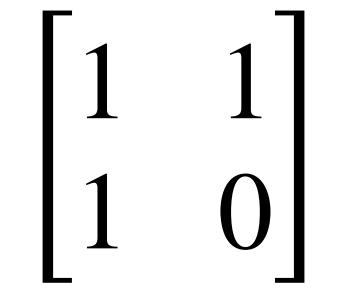
- matrix A is $det(A \lambda I)$ viewed as a polynomial in the
- This is a polynomial with the eigenvalues of A as

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

Example: 2 × 2 Matrix

Let's find the characteristic polynomial of this matrix:





A Special Linear Dynamical System $\mathbf{v}_{k+1} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \mathbf{v}_k \qquad \mathbf{v}_0 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

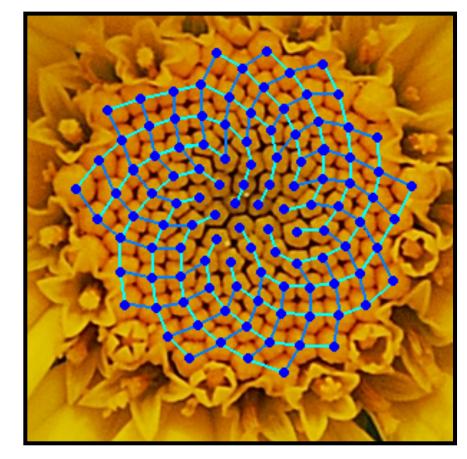
What does this system represent?:

Consider the system given by the above matrix.

Fibonacci Numbers

 $F_0 = 0$ **define** fib(n): $F_1 = 1$ $F_k = F_{k-1} + F_{k-2}$ return curr

recurrence relation.



curr, next $\leftarrow 0$, 1 repeat n times: curr, next ← next, curr + next

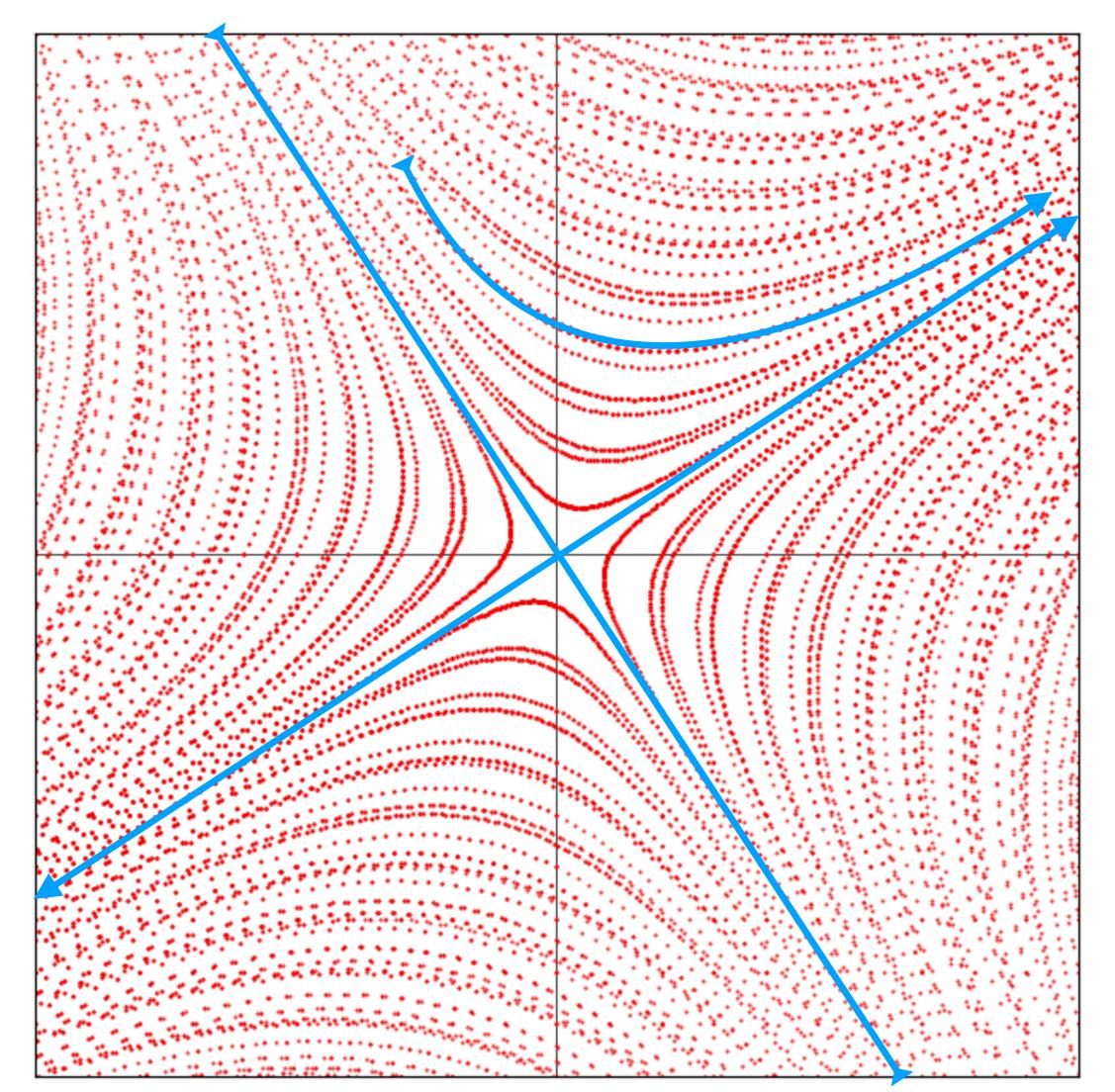
The Fibonacci numbers are defined in terms of a

They seem to crop-up in nature, engineering, etc.

https://commons.wikimedia.org/wiki/File:FibonacciChamomile.PNG



Recall: The Picture



The eigenvalue of this matrix is the golden ratio



Golden Ratio $\varphi = \frac{1 + \sqrt{5}}{2} \qquad \frac{F_{k+1}}{F_k} \to \varphi \text{ as } k \to \infty$

This is the largest eigenvalue of $\begin{vmatrix} 1 & 1 \\ 1 & 0 \end{vmatrix}$

The "long term behavior" is the Fibonacci sequence is defined by the golden ratio:

Example: Triangular matrix

The characteristic polynomial of a triangular matrix comes <u>pre-factored</u>:

 $\begin{bmatrix} 1 & -3 & 0 & 6 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & 4 \end{bmatrix}$



How To: Finding Eigenvalues

How To: Finding Eigenvalues

Question. Find all eigenvalues of the matrix A.

How To: Finding Eigenvalues

polynomial of A.

Question. Find all eigenvalues of the matrix A. Solution. Find the roots of the characteristic

An Observation: Multiplicity $\lambda^1(\lambda -$

In the examples so far, we've seen a number appear as a root multiple times.

This is called the multiplicity of the root.

$$1)^2(\lambda - 4)^1$$
 multiplicities

Is the multiplicity meaningful in this context?

Multiplicity and Dimension

for the eigenvalue λ is <u>at most</u> the multiplicity of λ in det $(A - \lambda I)$.

> The multiplicity is an upper bound on "how large" the eigenspace is.

Theorem. The dimension of the eigenspace of A

Example

Let A be a 5×5 matrix with characteristic polynomial $(x - 1)^3(x - 3)(x + 5)$.

- » What is rank(A)?

» What is the minimum possible rank of A - I?