Diagonalization

Geometric Algorithms
Lecture 20

CAS CS 132

Objectives

1. Finish our discussion on the characteristic polynomial
2. Motivate diagonalization via linear dynamical systems and changes of coordinate systems
3. Describe how to diagonalize a matrix

Keywords

multiplicity

similar matrices

diagonalizable matrices

change of basis

eigenbasis

Recap: Characteristic Polynomial

Recall: Determinants and Invertibility

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So by the Invertible Matrix Theorem:

$\det(A - \lambda I) = 0 \quad \equiv \quad (A - \lambda I)\mathbf{x} = \mathbf{0}$ **has nontrivial solutions**

$\equiv \quad \lambda$ is an eigenvalue of A

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So by the Invertible Matrix Theorem:

$\det(A - \lambda I) = 0$ polynomial in λ	$\equiv$	$(A - \lambda I)\mathbf{x} = \mathbf{0}$ **has nontrivial solutions**
	$\equiv$	λ is an eigenvalue of A

How To: Finding Eigenvalues

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Question. Determine the eigenvalues of A.

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Solution. Find the *roots* of the characteristic polynomial of A, which is

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viewed as a *polynomial* in λ.

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We'll also use

numpy.linalg.eig(A)

Example

$$A = \begin{bmatrix} 1 & -1 \\ 7 & -3 \end{bmatrix}$$

Example: Triangular matrix

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

The characteristic polynomial of a triangular matrix comes pre-factored:

An Observation: Multiplicity

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multiplicities

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This is called the **(algebraic) multiplicity** of the root

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This is called the **(algebraic) multiplicity** of the root

Is the multiplicity meaningful in this context?

Multiplicity and Dimension

Theorem. The dimension of the eigenspace of A for the eigenvalue λ is <u>at most</u> the multiplicity of λ in $\det(A - \lambda I)$ (and <u>at least</u> 1)

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

Example

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

» *What is* $\mathrm{rank}(A)$*?*

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

Practice Problem

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

Determine the eigenvalues and an eigenbasis for the above matrix

Answer

$$A = \begin{bmatrix} 5 & 1 \\ 4 & 2 \end{bmatrix}$$

$$(\lambda - a)(\lambda - b) = \lambda^2 - (a+b)\lambda + ab$$

$$A - \lambda I = \begin{bmatrix} 5-\lambda & 1 \\ 4 & 2-\lambda \end{bmatrix}$$

$$\det(A - \lambda I) = (5-\lambda)(2-\lambda) - 4$$
$$= \lambda^2 - 7\lambda + 6 = (\lambda - 1)(\lambda - 6)$$

$$A - I = \begin{bmatrix} 4 & 1 \\ 4 & 1 \end{bmatrix} \sim \left[\begin{array}{cc|c} 1 & 1/4 & 0 \\ 0 & 0 & 0 \end{array}\right] \Rightarrow \vec{x} = x_2 \begin{bmatrix} -1/4 \\ 1 \end{bmatrix} \Leftarrow \begin{array}{l} \lambda_1 = 1 \\ \vec{v}_1 = \begin{bmatrix} -1/4 \\ 1 \end{bmatrix} \end{array}$$

$$A - 6I = \begin{bmatrix} -1 & 1 \\ 4 & -4 \end{bmatrix} \sim \left[\begin{array}{cc|c} 1 & -1 & 0 \\ 0 & 0 & 0 \end{array}\right] \Rightarrow \vec{x} = x_2 \begin{bmatrix} 1 \\ 1 \end{bmatrix} \qquad \begin{array}{l} \lambda_2 = 6 \\ \vec{v}_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \end{array}$$

$$\left\{ \begin{bmatrix} -1/4 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\} \text{ eigenbasis}$$

Motivating Diagonalization via Linear Dynamical Systems

(briefly)

Recall: Eigenbasis

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We will be almost exclusively interested of eigenbases of $\mathbb{R}^n$ when $A \in \mathbb{R}^{n \times n}$

The Question. **When can we describe any vector in $\mathbb{R}^n$ as a unique linear combination of eigenvectors of A?**

Recall: Linear Dynamical Systems

$$
\begin{aligned}
\mathbf{v}_1 &= A\mathbf{v}_0 \\
\mathbf{v}_2 &= A\mathbf{v}_1 = A^2\mathbf{v}_0 \\
\mathbf{v}_3 &= A\mathbf{v}_2 = A^3\mathbf{v}_0 \\
\mathbf{v}_4 &= A\mathbf{v}_3 = A^4\mathbf{v}_0 \\
&\vdots
\end{aligned}
$$

A **linear dynamical system** describes a sequence of **state vectors** starting at $\mathbf{v}_0$

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$$\mathbf{v}_4 = A\mathbf{v}_3 = A^4\mathbf{v}_0$$

$$\vdots$$

multiplying by A changes the state.

A **linear dynamical system** describes a sequence of **state vectors** starting at $\mathbf{v}_0$

demo

Eigenbases and Closed-Form solutions

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Given $\mathbf{v}_k = A\mathbf{v}_{k-1} = A^k\mathbf{v}_0$, if

$$\mathbf{v}_0 = \alpha_1\mathbf{b}_1 + \alpha_2\mathbf{b}_2 + \alpha_3\mathbf{b}_3$$

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eigenvectors of A

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then

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closed-form solution

~~Verify:~~

Application: Eigenbases and Limiting Behavior

Theorem. If A has an eigenbasis with eigenvalues

$$\lambda_1 \geq \lambda_2 \ldots \geq \lambda_k$$

then $\mathbf{v}_k \sim \lambda_1^k \mathbf{u}$ for some vector $\mathbf{u}$.

In the long term, the system grows <u>exponentially in λ_1</u>.

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Sometimes, A behaves simply on $\mathscr{B}$, as in the case of eigenbases.

What we're really doing is changing our coordinate system to expose a behavior of A.

~~Recap~~: Change of Basis

Recall: Bases define Coordinate Systems

$\vec{x} = \begin{bmatrix} 1 \\ 6 \end{bmatrix}$

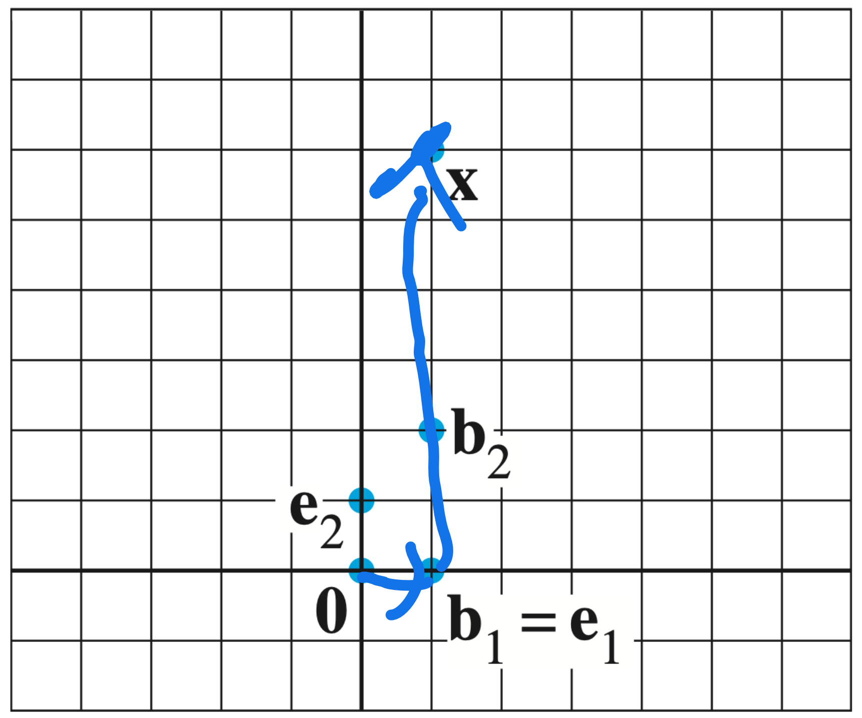


FIGURE 1 Standard graph paper.

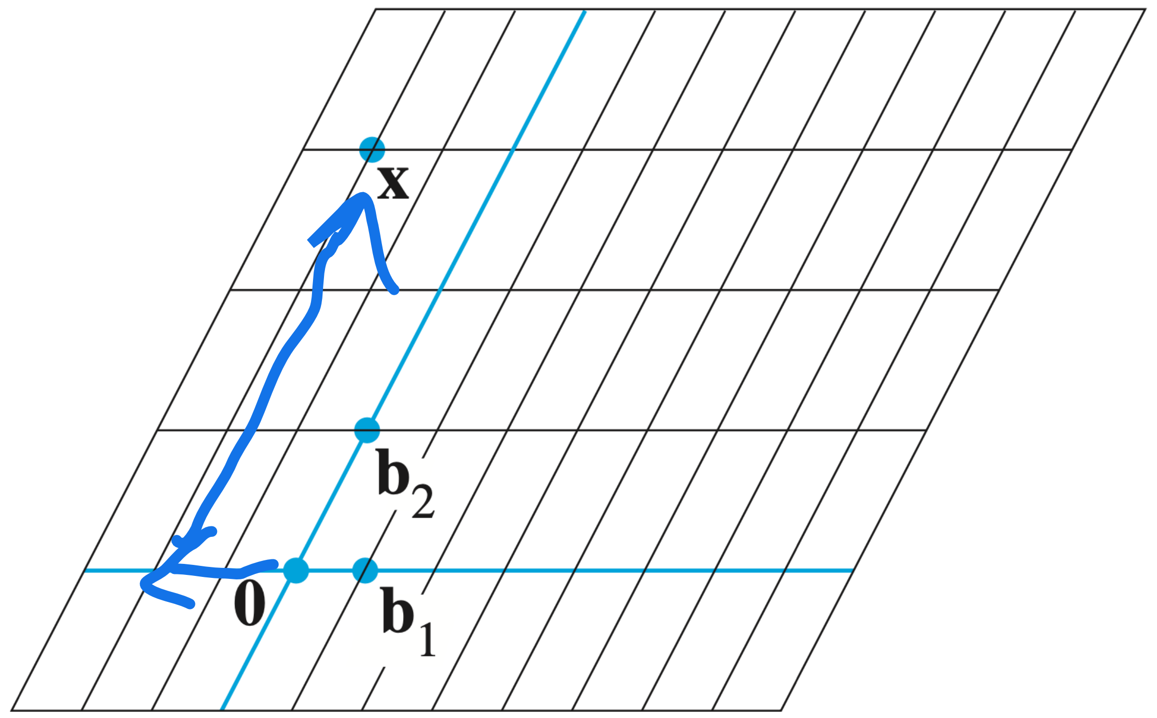


FIGURE 2 $\mathcal{B}$-graph paper.

$[\vec{x}]_{\mathcal{B}} = \begin{bmatrix} -2 \\ 3 \end{bmatrix}$

$\left[\begin{array}{cc|c} \vec{b}_1, & \vec{b}_2 & \vec{x} \end{array}\right]$

Want: $\vec{x} = a_1\vec{b}_1 + a_2\vec{b}_2$

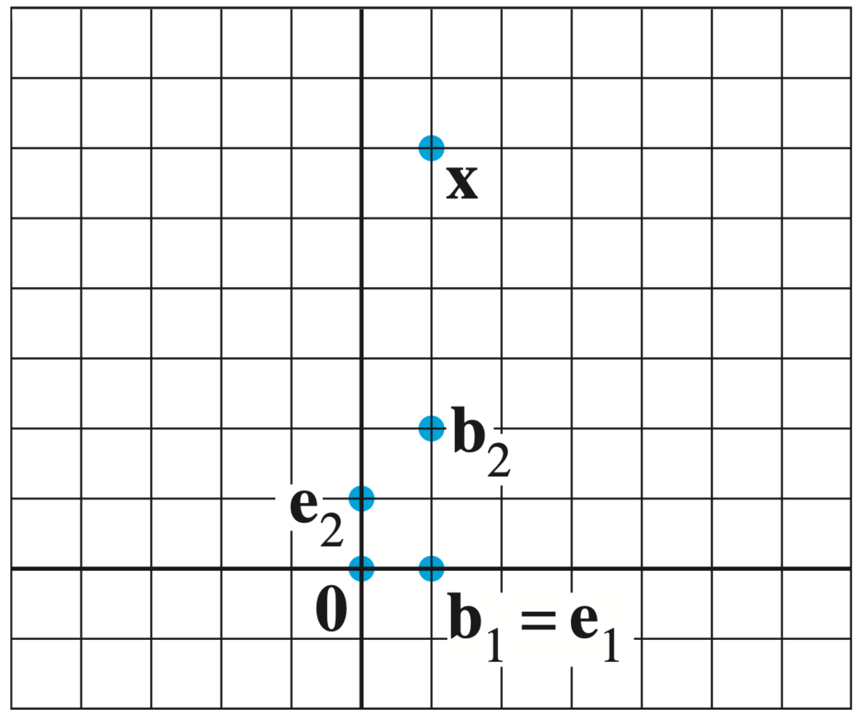
$[x]_{\mathcal{B}} = \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} \vec{b}_1 & \vec{b}_2 \end{bmatrix}^{-1} \begin{bmatrix} \vec{x} \end{bmatrix}$

COB matrix

$\longrightarrow \begin{bmatrix} \vec{b}_1 & \vec{b}_2 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} \vec{x} \end{bmatrix}$

Linear Algebra and its Applications (Lay, Lay, McDonald)

Recall: Bases define Coordinate Systems



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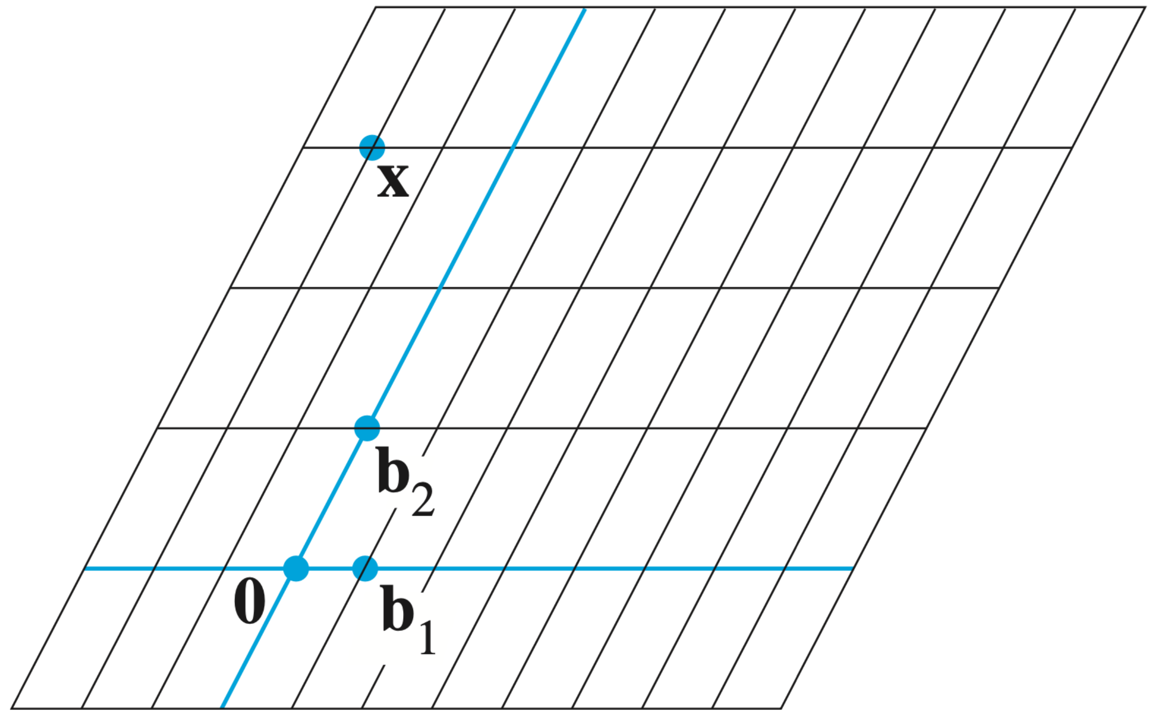

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Given a basis $\mathscr{B}$ of $\mathbb{R}^n$, there is **exactly one way** to write every vector as a linear combination of vectors in $\mathscr{B}$

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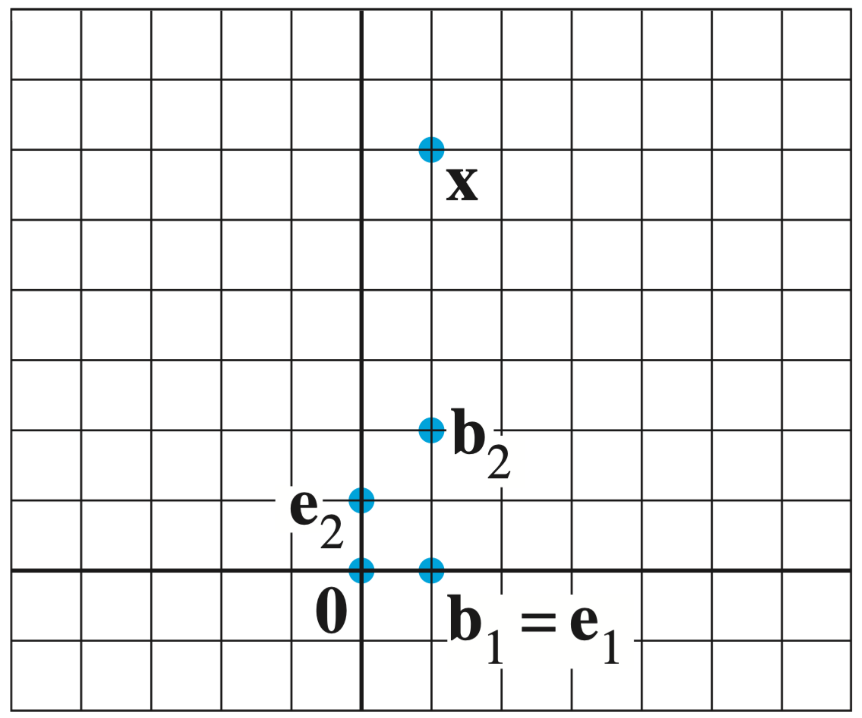

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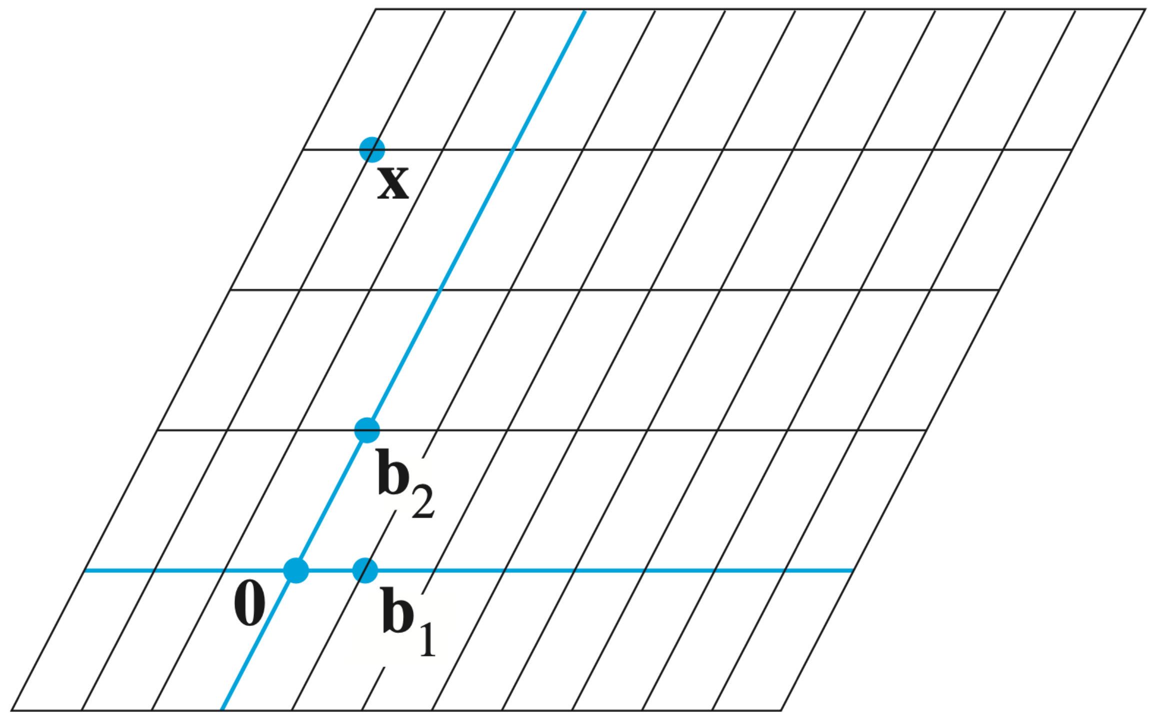

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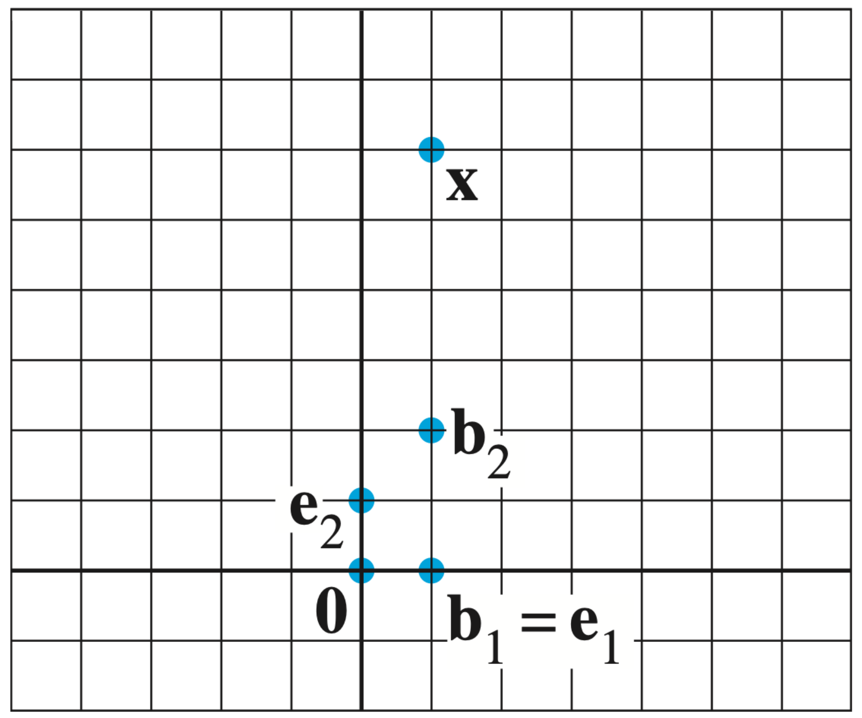

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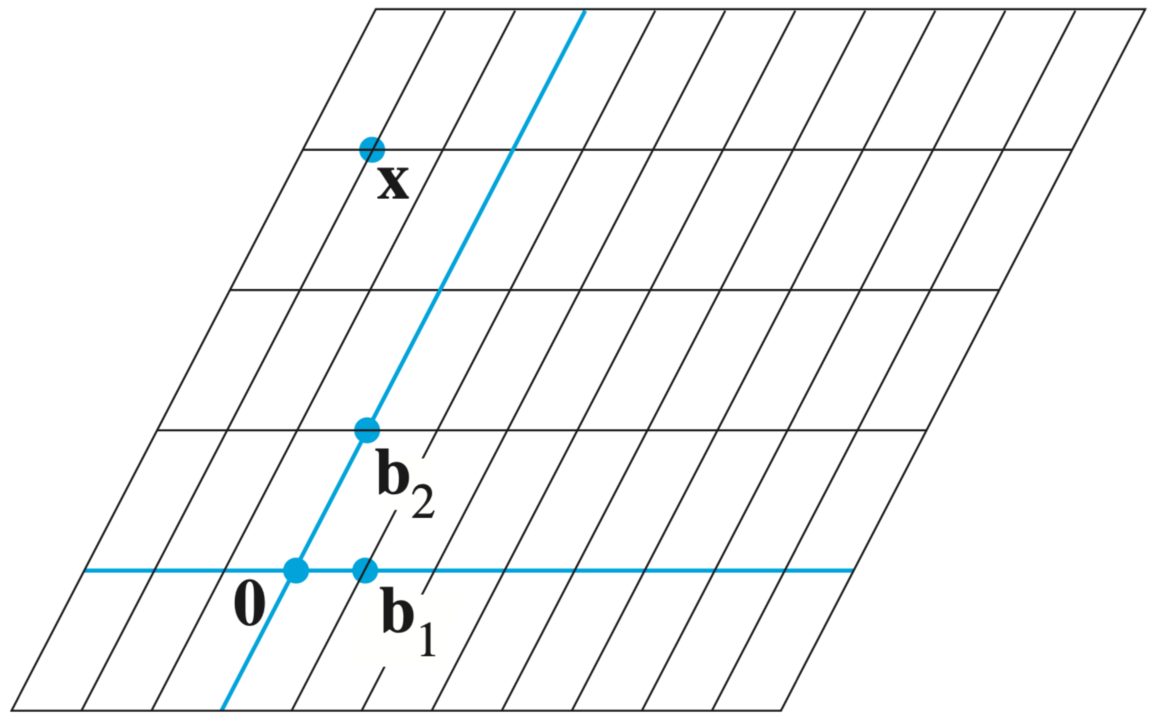

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$\mathscr{B}$ defines a "different grid for our graph paper"

Recall: Coordinate Vectors

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Let $\mathbf{v}$ be a vector in a $\mathbb{R}^n$ and let $\mathscr{B} = (\mathbf{b}_1, \mathbf{b}_2, \ldots, \mathbf{b}_n)$ be a basis of $\mathbb{R}^n$ where

$$\mathbf{v} = a_1\mathbf{b}_1 + a_2\mathbf{b}_2 + \ldots + a_n\mathbf{b}_n$$

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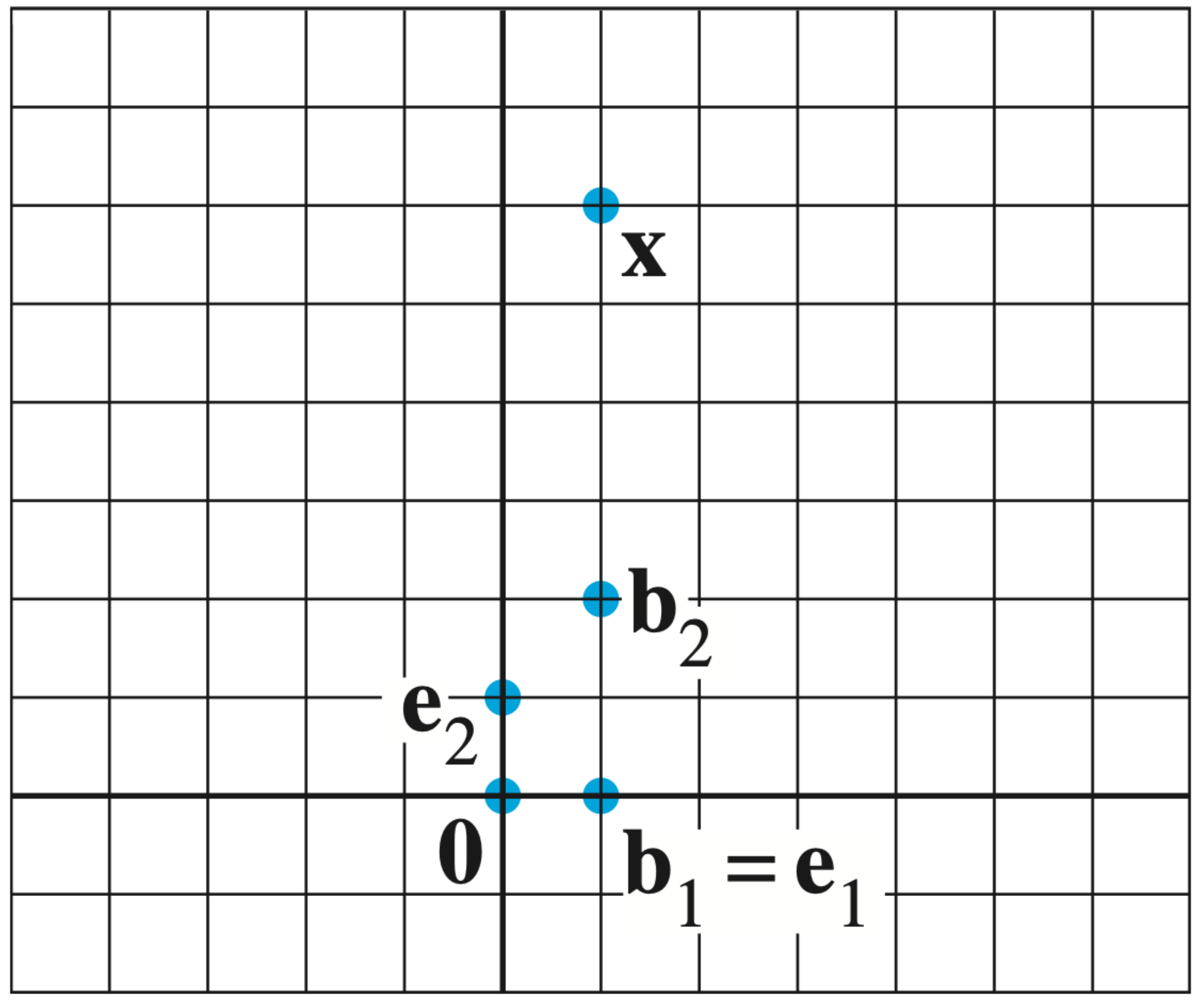
Recall: Coordinate Vectors



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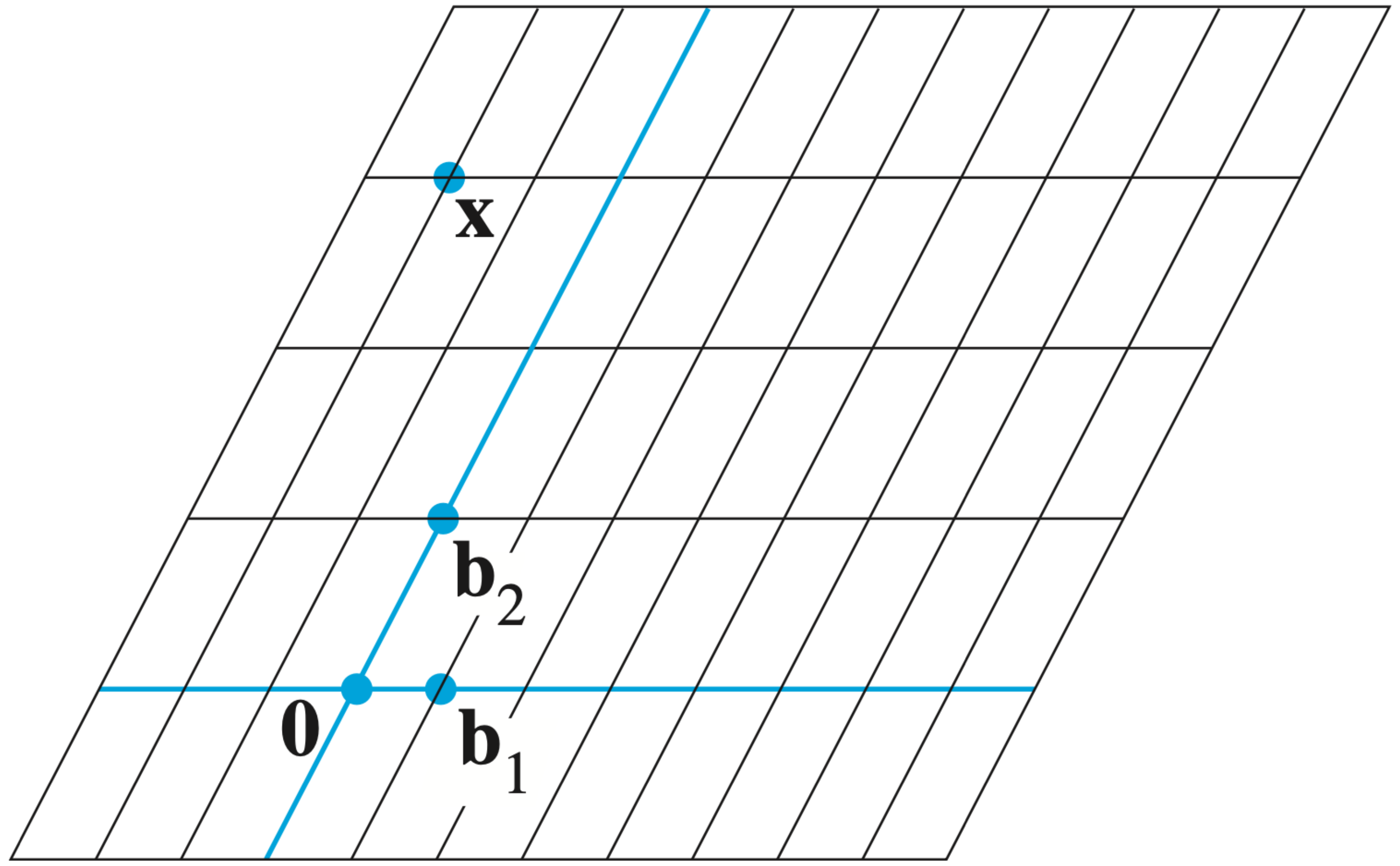


FIGURE 2 $\mathcal{B}$-graph paper.

Question (Conceptual)

We know that if a $n \times n$ matrix $B = [\mathbf{b}_1\ \mathbf{b}_2\ \ldots\ \mathbf{b}_n]$ is invertible, then the columns of B form a basis $\mathscr{B}$ of $\mathbb{R}^n$

What is the matrix that implements the transformation

$$\mathbf{x} \mapsto [\mathbf{x}]_{\mathscr{B}} = \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}$$

where $\mathbf{x} = c_1\mathbf{b}_1 + c_2\mathbf{b}_2 + \ldots + c_n\mathbf{b}_n$*?*

Change of Basis Matrix

Theorem. If $\mathscr{B} = \{\mathbf{b}_1, \mathbf{b}_2, \ldots, \mathbf{b}_n\}$ form a basis of $\mathbb{R}^n$, then

$$[\mathbf{x}]_{\mathscr{B}} = [\mathbf{b}_1\ \mathbf{b}_2\ \ \ldots\ \ \mathbf{b}_n]^{-1}\mathbf{x}$$

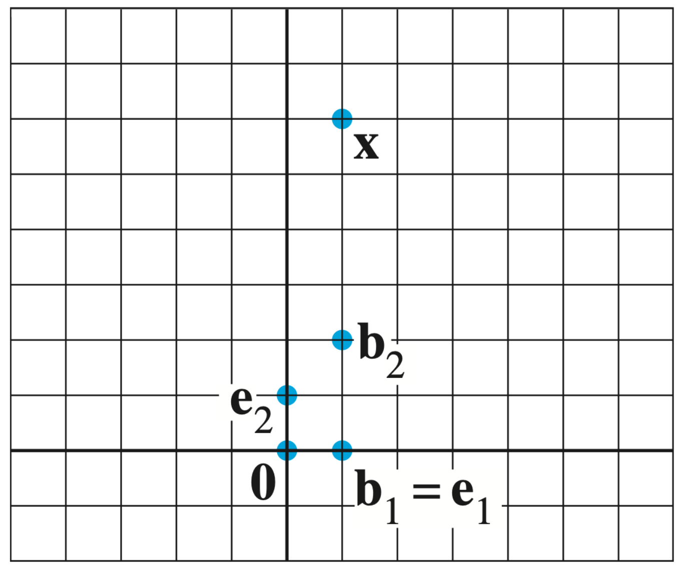
Matrix inverses perform changes of bases.

How To: Change of Basis

Question. Given a basis $\mathscr{B} = (\mathbf{b}_1, \mathbf{b}_2, \ldots, \mathbf{b}_n)$ of $\mathbb{R}^n$, find the matrix which implements $\mathbf{x} \mapsto [\mathbf{x}]_{\mathscr{B}}$.

Solution. Construct the matrix $[\mathbf{b}_1\ \mathbf{b}_2\ \ldots\ \mathbf{b}_n]^{-1}$.

Example



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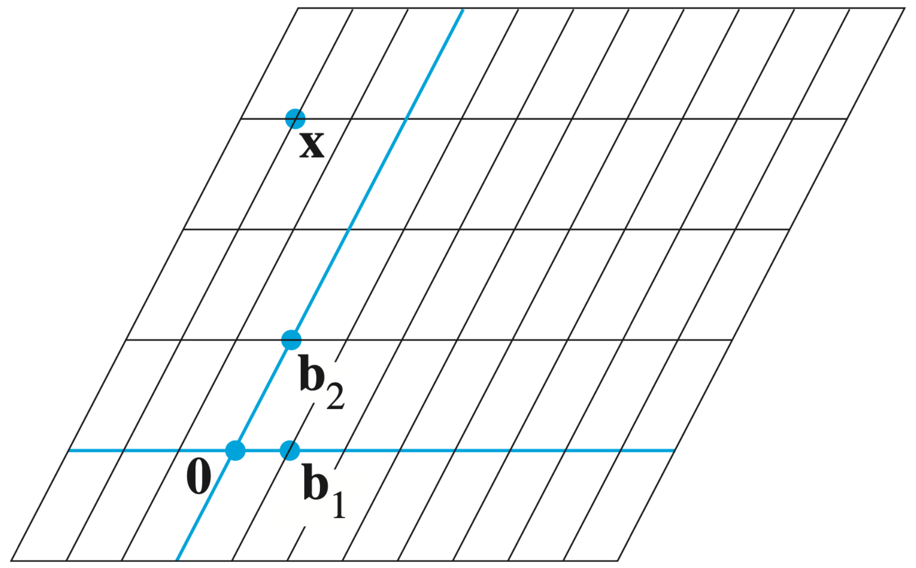

FIGURE 2 $\mathcal{B}$-graph paper.

Write the change-of-bases matrix for the basis $\left(\begin{bmatrix}1\\0\end{bmatrix}, \begin{bmatrix}1\\2\end{bmatrix}\right)$

$$\begin{bmatrix}1&1\\0&2\end{bmatrix}^{-1} = \begin{bmatrix}1&-1/2\\0&1/2\end{bmatrix}$$

$$A^{-1} = \frac{1}{\det A}\begin{bmatrix}d&-b\\-c&a\end{bmatrix} = \frac{1}{2}\begin{bmatrix}2&-1\\0&1\end{bmatrix}$$

$$\left[\begin{array}{cc|cc}1&1&1&0\\0&2&0&1\end{array}\right] \sim \left[\begin{array}{cc|cc}1&1&1&0\\0&1&0&1/2\end{array}\right] \sim \left[\begin{array}{cc|cc}1&0&1&-1/2\\0&1&0&1/2\end{array}\right]$$

Diagonalization

Diagonal Matrices

ex. $$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & -0.4 & 0 & 0 \\ 0 & 0 & 22 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Diagonal Matrices

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Definition. A $n \times n$ matrix A is **diagonal** if

$$i \neq j \quad \cancel{\text{if and only if}} \quad A_{ij} = 0$$

$$\Longrightarrow$$

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Only the diagonal entries can be nonzero ex. $$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & -0.4 & 0 & 0 \\ 0 & 0 & 22 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

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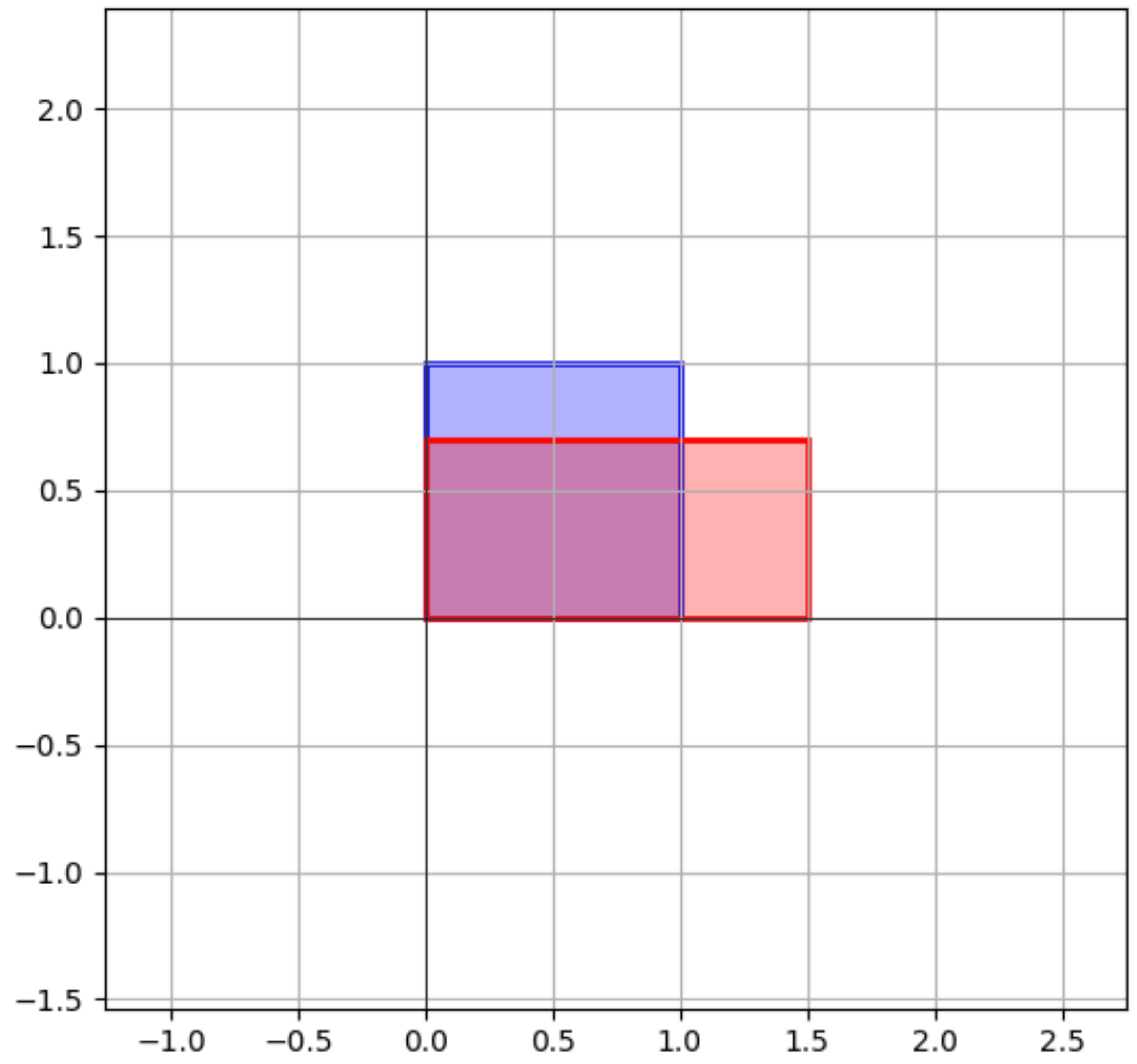
Diagonal matrices are scaling matrices

Recall: Unequal Scaling

The scaling matrix *affects each component of a vector in a simple way*

The diagonal entries <u>scale</u> each corresponding entry

$$\begin{bmatrix} 1.5 & 0 \\ 0 & 0.7 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 1.5x \\ 0.7y \end{bmatrix}$$



$$\begin{bmatrix} 1.5 & 0 \\ 0 & 0.7 \end{bmatrix}$$

High level question: unequal

When do matrices "behave" like scaling matrices "up to" change of basis?

Scaling and Eigenvectors

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The idea. Matrices behave like scaling matrices on eigenvectors.

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$$\begin{bmatrix} 2 & 0 \\ 0 & -3 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} (x\mathbf{e}_1 + y\mathbf{e}_2) = x2\mathbf{e}_1 + y(-3)\mathbf{e}_2$$

$$A \begin{bmatrix} x \\ y \end{bmatrix}_{\mathscr{B}} = A(x\mathbf{b}_1 + y\mathbf{b}_2) = x\lambda_1\mathbf{b}_1 + y\lambda_2\mathbf{b}_2$$

Scaling and Eigenvectors

The idea. Matrices behave like scaling matrices on eigenvectors.

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The fundamental question:
Can we expose this behavior in terms of a *matrix factorization*?

Recall: Matrix Factorization

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<u>Factorizations can</u>:

- make working with A easier
- expose important information about A

Similar Matrices

$$A = PBP^{-1}$$

Definition. A matrix A is **similar** to a matrix B if there is some invertible matrix P such that $A = PBP^{-1}$

A and B are the same up to a change of basis

Similar Matrices and Eigenvalues

Theorem. Similar matrices have the <u>same eigenvalues</u>.

Verify: $A = PBP^{-1}$ $\qquad B\vec{w} = \lambda\vec{w}$

$A\vec{v} = PBP^{-1}\vec{v} = \lambda\vec{v}$

$= B(P^{-1}\vec{v}) = P^{-1}(\lambda\vec{v}) = \lambda(P^{-1}\vec{v})$

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There is an invertible matrix P *and* *diagonal* *matrix* D *such that* $A = PDP^{-1}$

Diagonalizable Matrices

Definition. A matrix A is **diagonalizable** if it is similar to a diagonal matrix

There is an invertible matrix P and <u>diagonal</u> matrix D such that $A = PDP^{-1}$

unequal

Diagonalizable matrices are the same as scaling matrices up to a change of basis

$D = (I)D(I)^{-1}$

Important: Not all Matrices are Diagonalizable

This is very different from the LU factorization

We will need to figure out which matrices are diagonalizable

<u>Question</u>. Is the zero matrix diagonalizable?

$$0 = (Id)\,0\,(Id)^{-1}$$

Application: Matrix Powers

only take the power of B

Theorem. If $A = PBP^{-1}$, then $A^k = PB^kP^{-1}$

It may be easier to take the power of B (as in the case of diagonal matrices)

Verify: $A^2 = (PB\underbrace{P^{-1})(P}_{Id}BP^{-1}) = PB^2P^{-1}$

How To: Matrix Powers

Question. Given A is diagonalizable, determine A^k

Solution. Find it's diagonalization PDP^{-1} and then compute PD^kP^{-1}

Remember that

$$\begin{bmatrix} a & 0 & 0 \\ 0 & b & 0 \\ 0 & 0 & c \end{bmatrix}^k = \begin{bmatrix} a^k & 0 & 0 \\ 0 & b^k & 0 \\ 0 & 0 & c^k \end{bmatrix}$$

But how do we find the diagonalization...

Diagonalization and Eigenvectors

Suppose we have a diagonalization

$$A = PDP^{-1}$$

What do we know about it?

Columns of P are eigenvectors

(space)

$$A = [\mathbf{p}_1\ \mathbf{p}_2\ \mathbf{p}_3] \underset{\vec{e}_1 \mapsto \vec{p}_1}{} \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix} \underset{\vec{p}_1 \mapsto \vec{e}_1}{} [\mathbf{p}_1\ \mathbf{p}_2\ \mathbf{p}_3]^{-1}$$

$\vec{e}_1 \mapsto \vec{p}_1$ $\qquad$ $\vec{p}_1 \mapsto \vec{e}_1$

Verify: $A\vec{p}_1 = \begin{bmatrix} \vec{p}_1 & \vec{p}_2 & \vec{p}_3 \end{bmatrix} \begin{bmatrix} \lambda_1 & & \\ & \lambda_2 & \\ & & \lambda_3 \end{bmatrix} \begin{bmatrix} \vec{p}_1 & \vec{p}_2 & \vec{p}_3 \end{bmatrix}^{-1} \vec{p}_1$

$$= \begin{bmatrix} \vec{p}_1 & \vec{p}_2 & \vec{p}_3 \end{bmatrix} \begin{bmatrix} \lambda_1 & & \\ & \lambda_2 & \\ & & \lambda_3 \end{bmatrix} \vec{e}_1 = \begin{bmatrix} \vec{p}_1 & \vec{p}_2 & \vec{p}_3 \end{bmatrix} \lambda_1 \vec{e}_1$$

$$= \lambda_1 \vec{p}_1$$

Columns of P are eigenvectors

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In fact, the columns of P form an **eigenbasis** of $\mathbb{R}^n$ for A

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And the entries of D are the **eigenvalues** associated to each eigenvector

Columns of P are eigenvectors

$$A = \overbrace{[\mathbf{p}_1\ \mathbf{p}_2\ \mathbf{p}_3]}^{\text{eigenbasis}} \overbrace{\begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix}}^{\text{eigenvalues}} [\mathbf{p}_1\ \mathbf{p}_2\ \mathbf{p}_3]^{-1}$$

In fact, the columns of P form an **eigenbasis** of $\mathbb{R}^n$ for A

And the entries of D are the **eigenvalues** associated to each eigenvector

A diagonalization exposes a lot of information about A

The Diagonalization Theorem

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Theorem. A matrix is diagonalizable if and only if it has an eigenbasis

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(we just did the hard part, if a matrix is diagonalizable then it has an **eigenbasis**)

We can use the same recipe to go in the other direction, given an eigenbasis, we can **build a diagonalization**

Diagonalizing a Matrix

High Level

$$A = PDP^{-1}$$

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$$A = PDP^{-1}$$

The columns of P form an <u>eigenbasis</u> for A

The diagonal of D are the eigenvalues for each column of P

The matrix P^{-1} is a change of basis to this eigenbasis of A

Step 1: Eigenvalues

$$A = \begin{bmatrix} 1 & 3 & 3 \\ -3 & -5 & 3 \\ 3 & 3 & 1 \end{bmatrix}$$

Find all the eigenvalues of A

Find the roots of $\det(A - \lambda I)$

e.g.

$1 \leq \dim(\mathrm{Nul}(A+2I)) \leq 2$

$$\det(A - \lambda I) = -(\lambda - 1)(\lambda + 2)^2$$

$\lambda_1 = 1$ mult. 1

$\lambda_2 = -2$ mult. 2

Step 2: Eigenvectors

$$A = \begin{bmatrix} 1 & 3 & 3 \\ -3 & -5 & 3 \\ 3 & 3 & 1 \end{bmatrix}$$

$$\lambda_1 = 1$$

$$\lambda_2 = -2$$

Find **bases** of the corresponding eigenspaces

e.g.

$$\mathrm{Nul}(A - I) = \mathrm{span}\left\{ \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} \right\}$$

$$\mathrm{Nul}(A + 2I) = \mathrm{span}\left\{ \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} \right\}$$

Step 3: Construct P

$$A = \begin{bmatrix} 1 & 3 & 3 \\ -3 & -5 & 3 \\ 3 & 3 & 1 \end{bmatrix}$$

$$\lambda_1 = 1$$

$$\lambda_2 = -2$$

$$\text{Nul}(A - I) = \text{span}\left\{ \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} \right\}$$

$$\text{Nul}(A + 2I) = \text{span}\left\{ \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} \right\}$$

If there are n eigenvectors from the previous step they form an **eigenbasis**

Build the matrix with these vectors as the columns

e.g.

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

Step 5: Construct D

$$A = \begin{bmatrix} 1 & 3 & 3 \\ -3 & -5 & 3 \\ 3 & 3 & 1 \end{bmatrix}$$

$$\lambda_1 = 1$$

$$\lambda_2 = -2$$

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

Build the matrix with eigenvalues as diagonal entries

Note the order. It should be the same as the order of columns of P

e.g.

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

Step 6: Invert P

Find the inverse of P (we know how to do this)

$$A = \begin{bmatrix} 1 & 3 & 3 \\ -3 & -5 & 3 \\ 3 & 3 & 1 \end{bmatrix}$$

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

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

Putting it Together

$$\underset{A}{\begin{bmatrix} 1 & 3 & 3 \\ -3 & -5 & 3 \\ 3 & 3 & 1 \end{bmatrix}} = \underset{P}{\begin{bmatrix} 1 & -1 & -1 \\ -1 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix}} \underset{D}{\begin{bmatrix} 1 & 0 & 0 \\ 0 & -2 & 0 \\ 0 & 0 & -2 \end{bmatrix}} \underset{P^{-1}}{\begin{bmatrix} 1 & -1 & -1 \\ -1 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix}^{-1}}$$

How to: Diagonalizing a Matrix

Question. Find a diagonalization of $A \in \mathbb{R}^n$, or determine that A is not diagonalizable

Solution.

1. Find the eigenvalues of A, and bases for their eigenspaces. If these eigenvectors don't form a basis of $\mathbb{R}^n$, then A is **not diagonalizable**
2. Otherwise, build a matrix P whose columns are the eigenvectors of A
3. Then build a diagonal matrix D whose entries are the eigenvalues of A *in the same order*
4. Invert P
5. The diagonalization of A is PDP^{-1}

We know how to do every step, its a matter of putting it all together

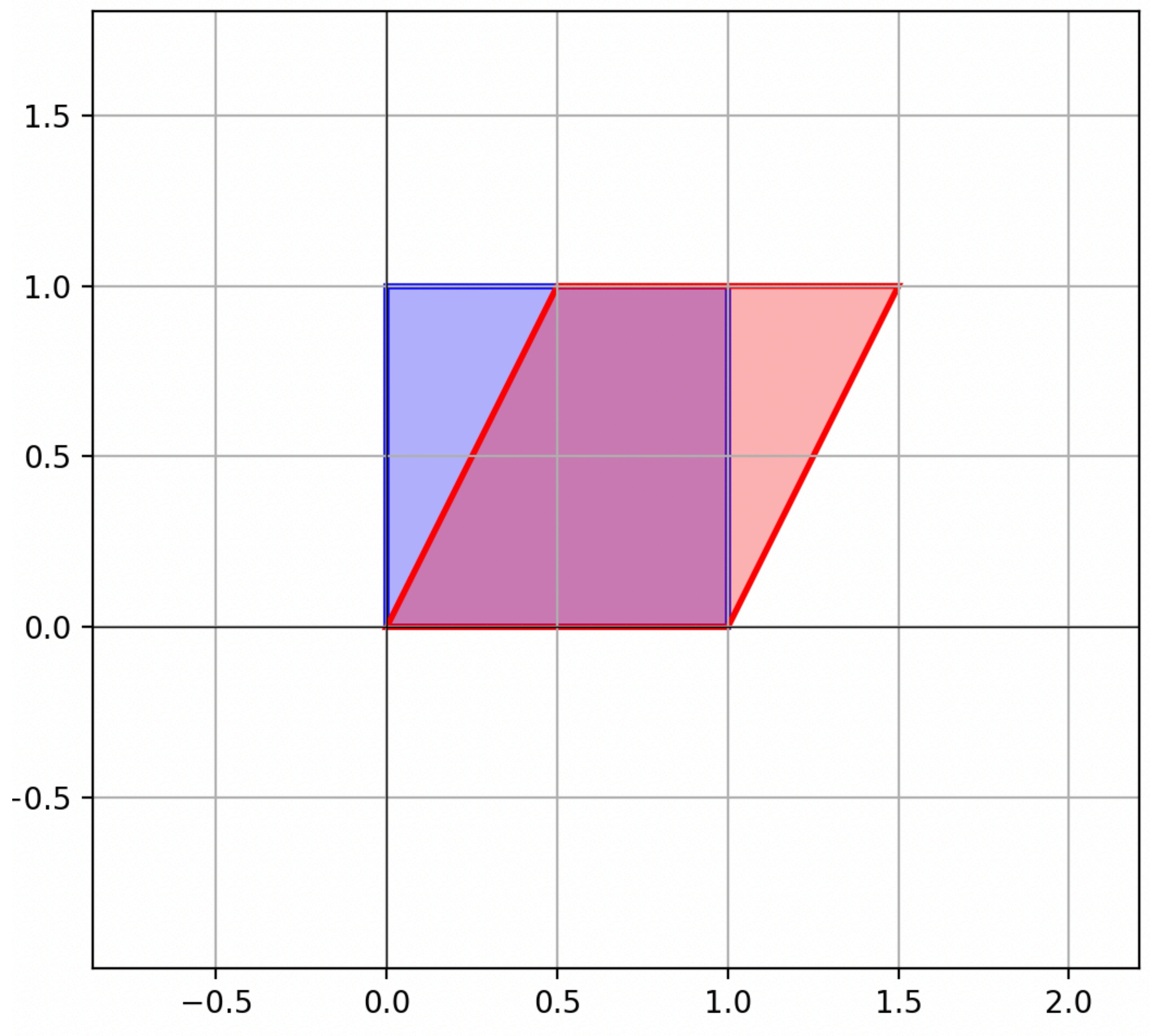
Example of Failure: Shearing

$$A = \begin{bmatrix} 1 & 0.5 \\ 0 & 1 \end{bmatrix}$$

The shearing matrix has a single eigenvalue with an eigenspace of dimension 1

We can't build an eigenbasis of $\mathbb{R}^2$ for A

In other words, A is not diagonalizable



$$\det(A - \lambda I) = (\lambda - 1)^2$$

$$\lambda = 1$$

$$A - I = \begin{bmatrix} 0 & 0.5 \\ 0 & 0 \end{bmatrix} \sim \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \Rightarrow \vec{x} = x_1 \begin{bmatrix} 1 \\ 0 \end{bmatrix} \Rightarrow \text{no eigen basis}$$

Important case: Distinct Eigenvalues

ex. $\begin{bmatrix} 1 & -3 & 4 & 2 \\ 0 & -2 & 3 & -1 \\ 0 & 0 & 10 & 5 \\ 0 & 0 & 0 & 6 \end{bmatrix}$

Theorem. If an $n \times n$ matrix has has n distinct eigenvalues, then it is diagonalizable

This is because eigenvectors with distinct eigenvalues are *linearly independent*

Example

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

Find a diagonalization of the above matrix

$\lambda_1 = 1$, $\lambda_2 = 2$

$$A - I = \begin{bmatrix} 1 & 0 \\ -1 & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \Rightarrow \vec{x} = x_2 \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$A - 2I = \begin{bmatrix} 0 & 0 \\ -1 & -1 \end{bmatrix} \sim \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} \Rightarrow \vec{x} = x_2 \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

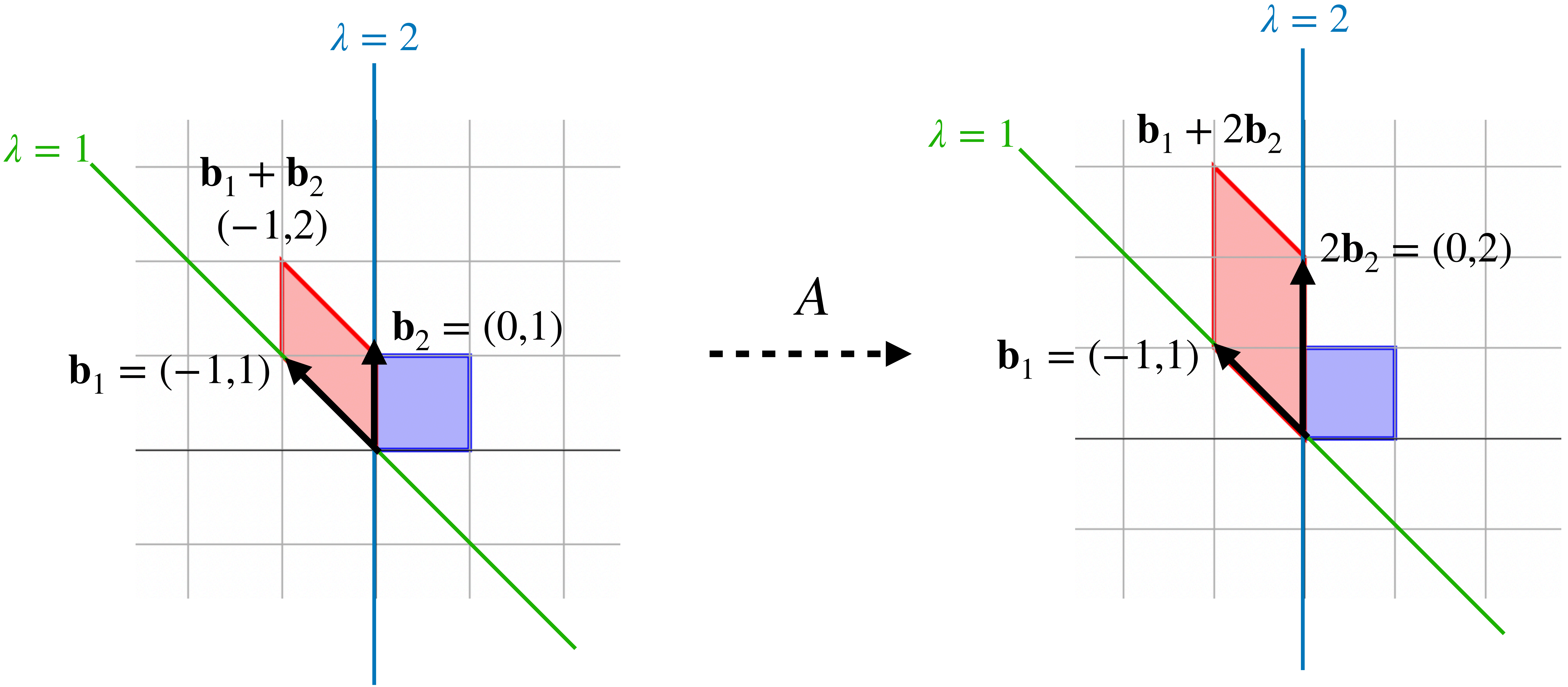
$$A = \begin{bmatrix} 0 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} 0 & -1 \\ 1 & 1 \end{bmatrix}^{-1}$$

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

The Picture

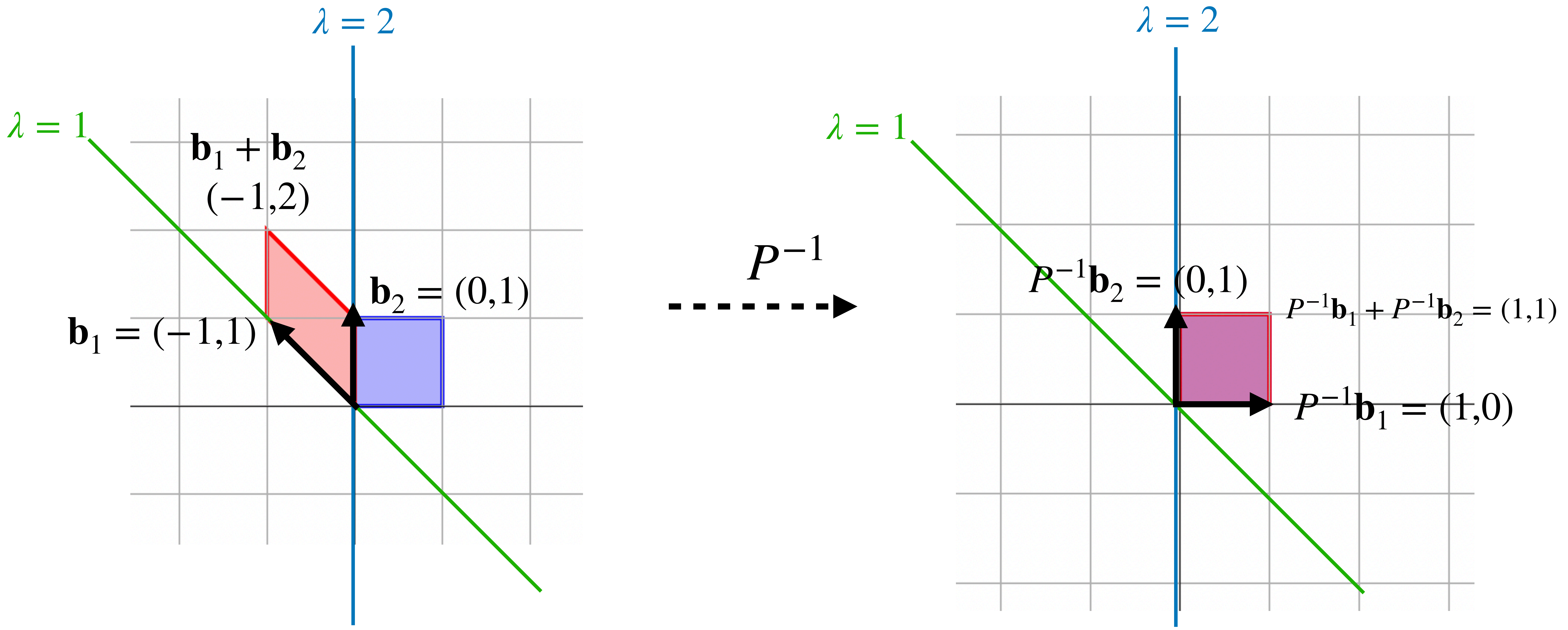
Example (Geometric)

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



Example (Geometric)

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

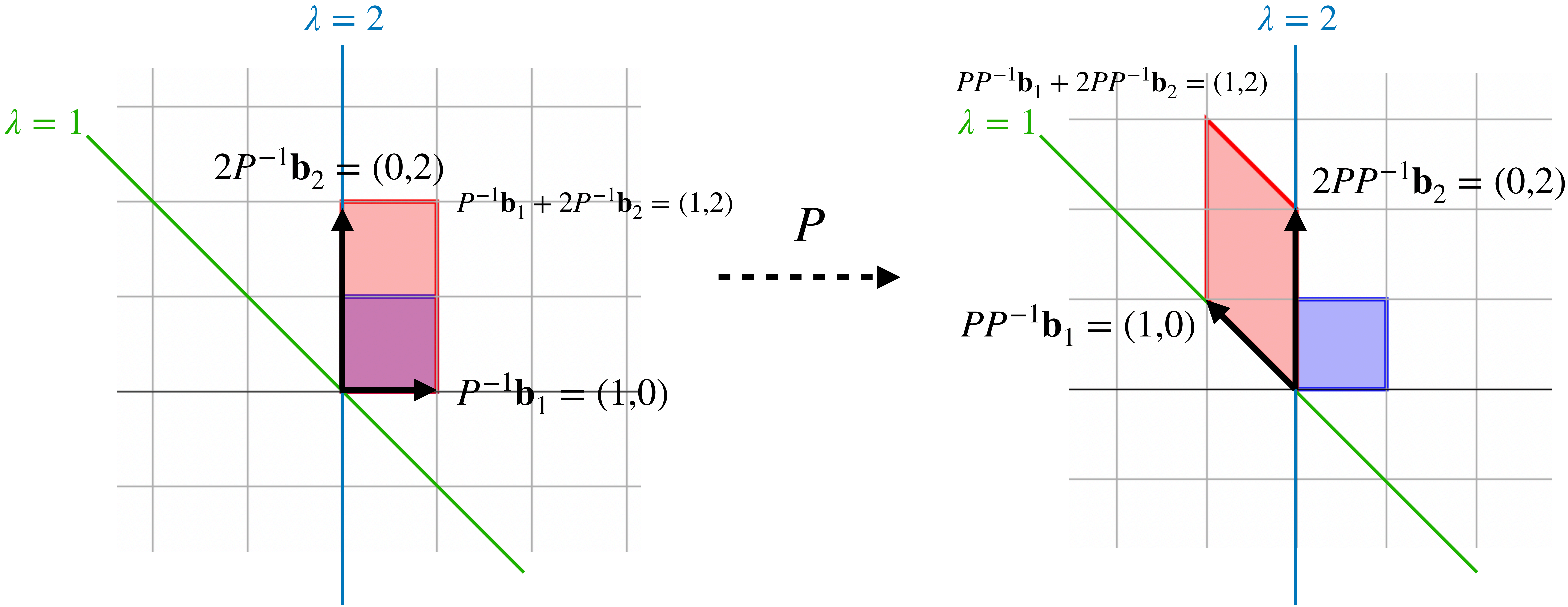


Example (Geometric)

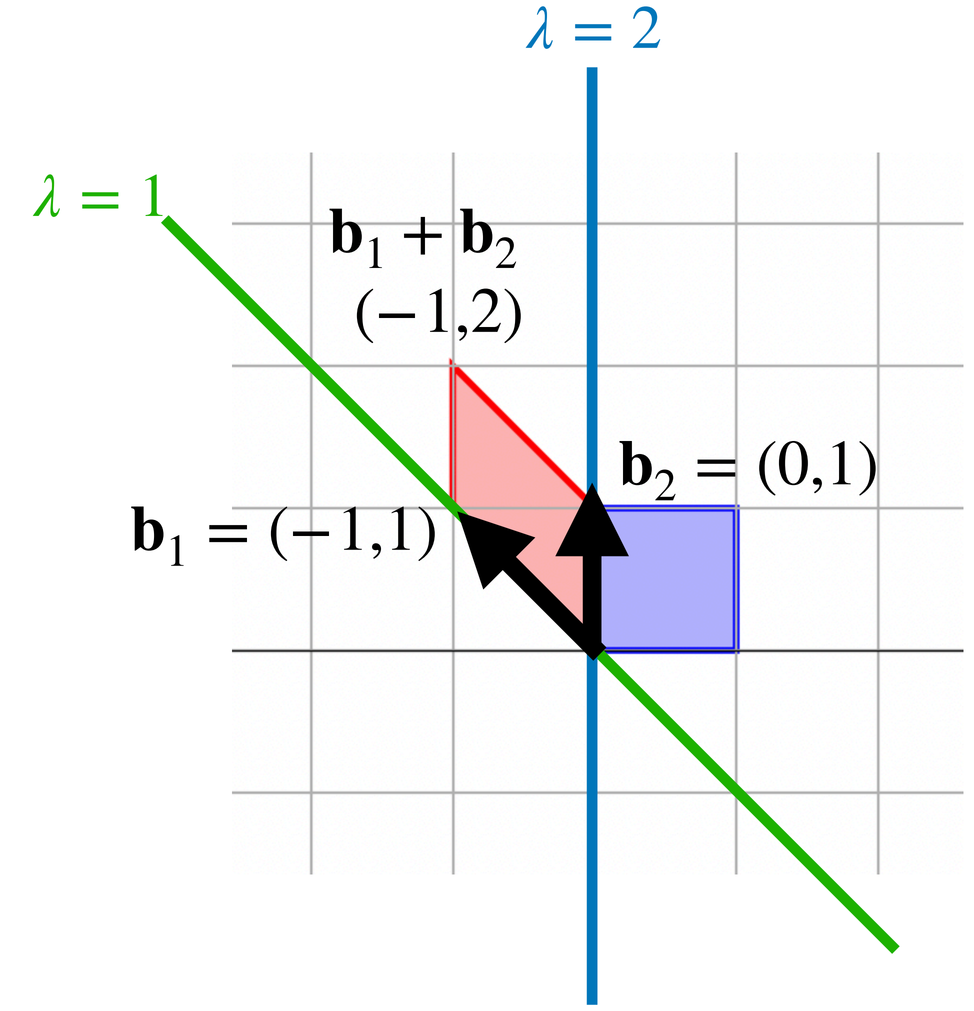
$$D = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}$$

Example (Geometric)

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



Example (Geometric)



$$A = PDP^{-1}$$

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