

# Diagonalization

**Geometric Algorithms**

**Lecture 20**

# 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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We'll also use

$$\text{numpy.linalg.eig}(A)$$

# 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

$$\lambda^1 (\lambda - 1)^2 (\lambda - 4)^1 \text{ multiplicities}$$

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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  $\lambda$  is at most the multiplicity of  $\lambda$  in  $\det(A - \lambda I)$  (and at least 1)

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

# Example

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

*» What is  $\text{rank}(A)$ ?*

*» 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**

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



# Motivating Diagonalization via Linear Dynamical Systems

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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

$$\mathbf{v}_1 = A\mathbf{v}_0$$

$$\mathbf{v}_2 = A\mathbf{v}_1 = A^2\mathbf{v}_0$$

$$\mathbf{v}_3 = A\mathbf{v}_2 = A^3\mathbf{v}_0$$

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

⋮

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

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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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closed-form solution

Verify:

# Application: Eigenbases and Limiting Behavior

**Theorem.** If  $A$  has an eigenbasis with eigenvalues

$$\lambda_1 \geq \lambda_2 \dots \geq \lambda_k$$

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

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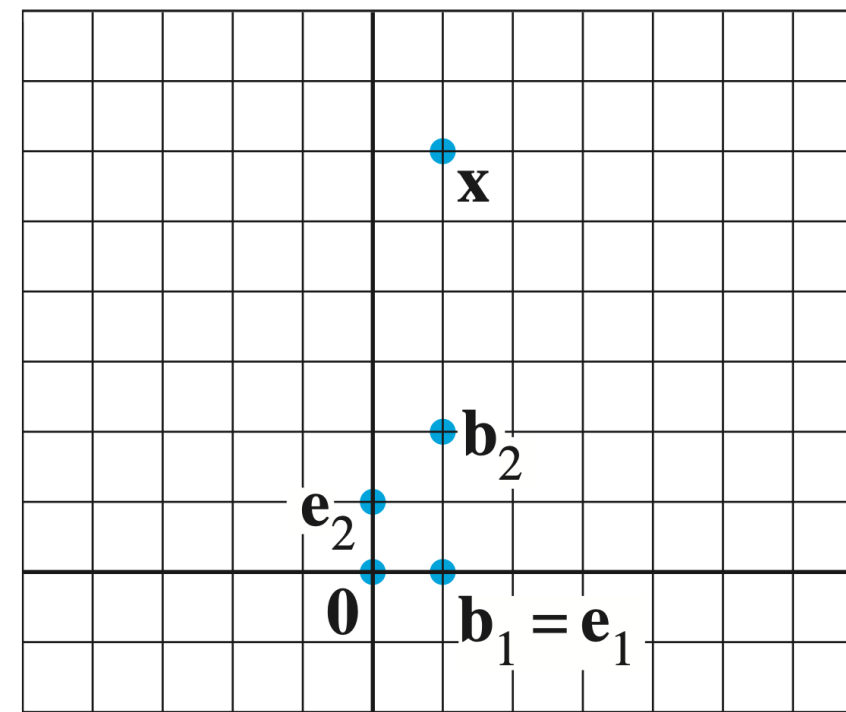
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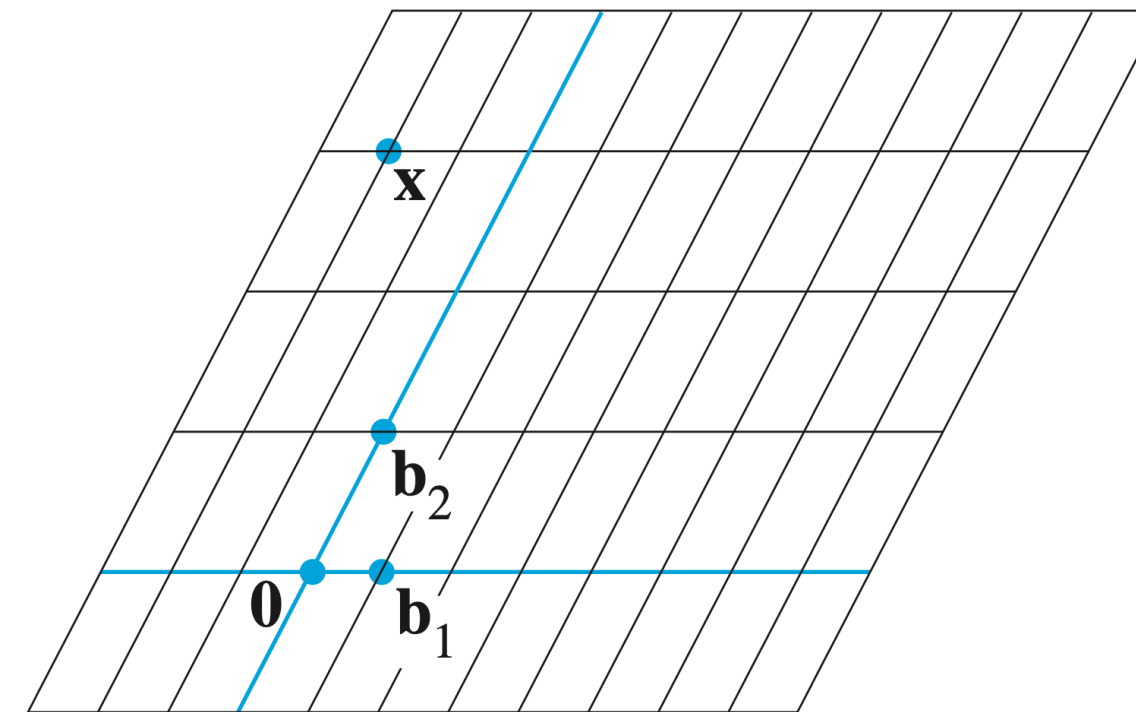
**What we're really doing is changing our coordinate system to expose a behavior of  $A$ .**

# Recap: Change of Basis

# Recall: Bases define Coordinate Systems



**FIGURE 1** Standard graph paper.



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

# Recall: Bases define Coordinate Systems

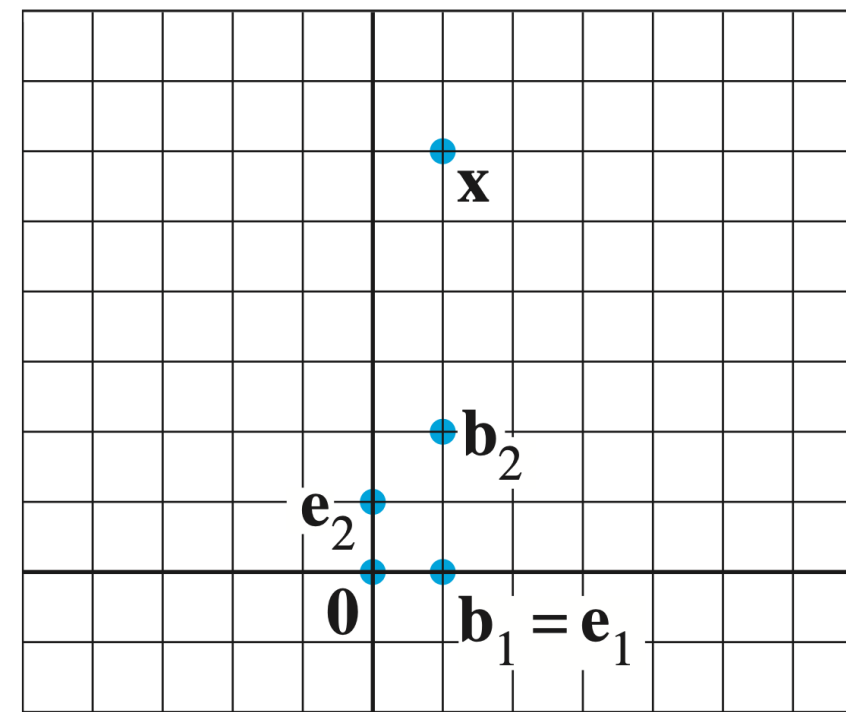


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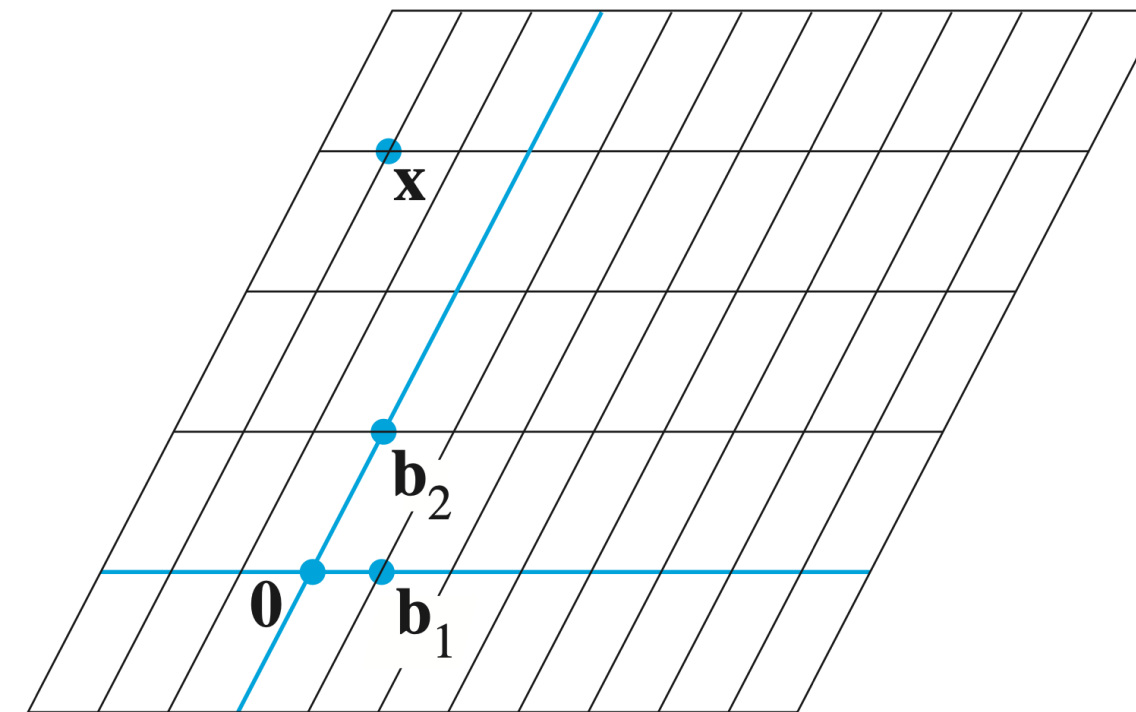


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Given a basis  $\mathcal{B}$  of  $\mathbb{R}^n$ , there is **exactly one way** to write every vector as a linear combination of vectors in  $\mathcal{B}$

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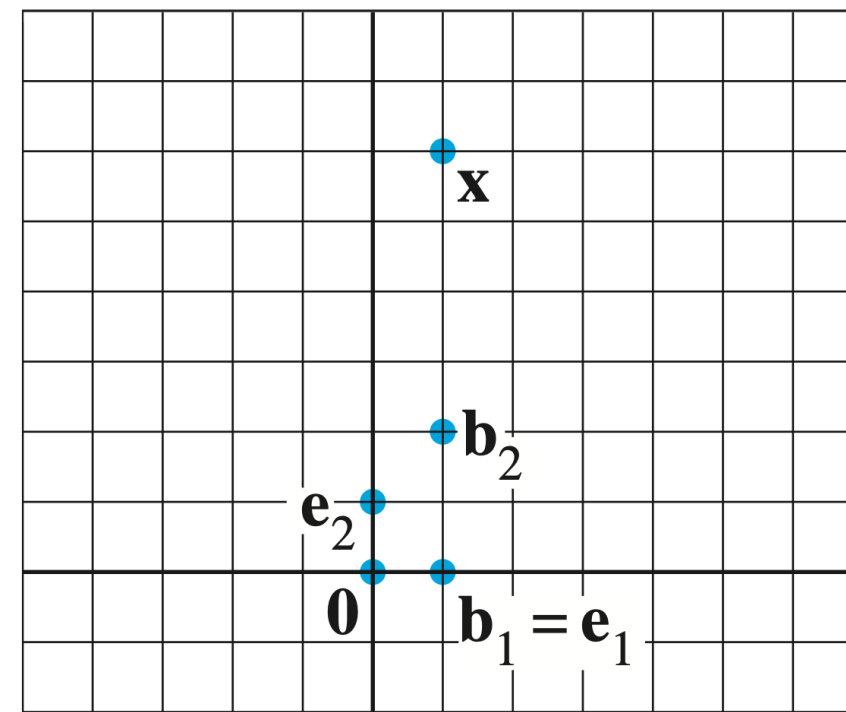


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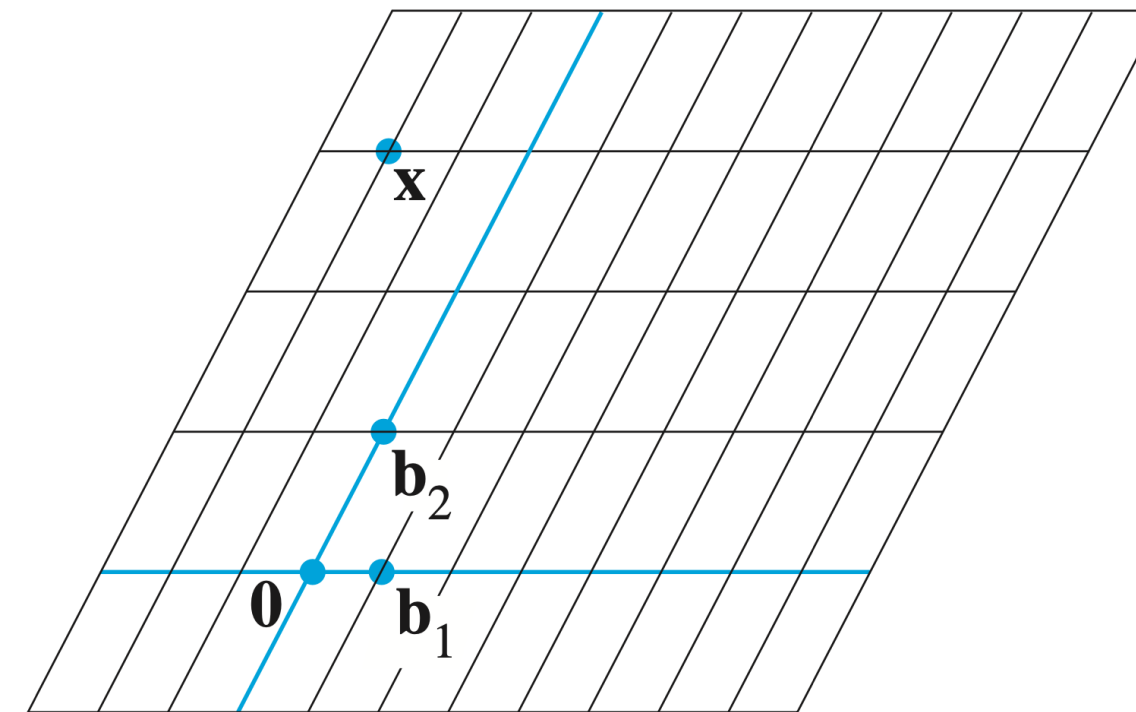


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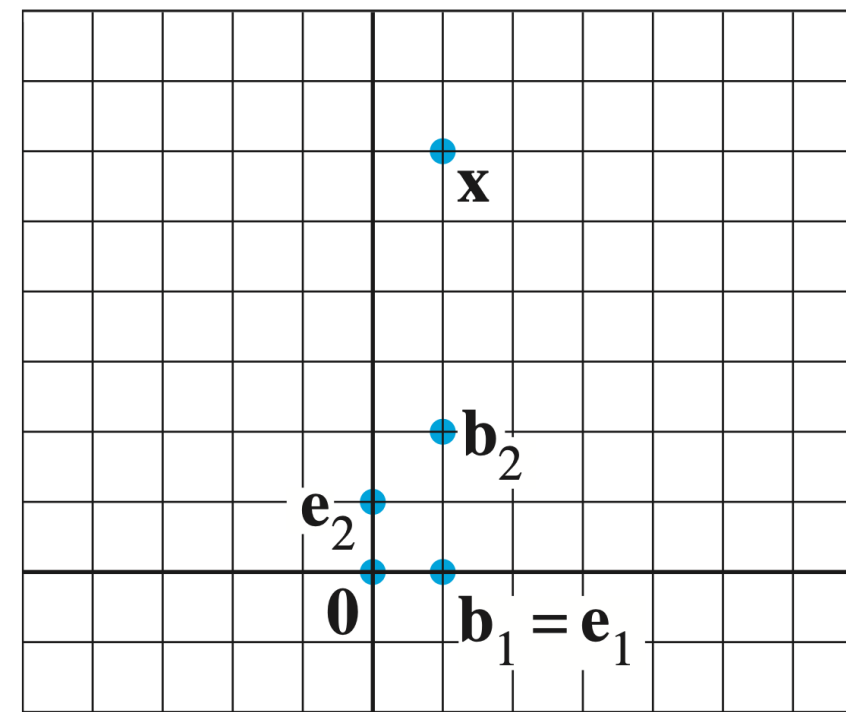


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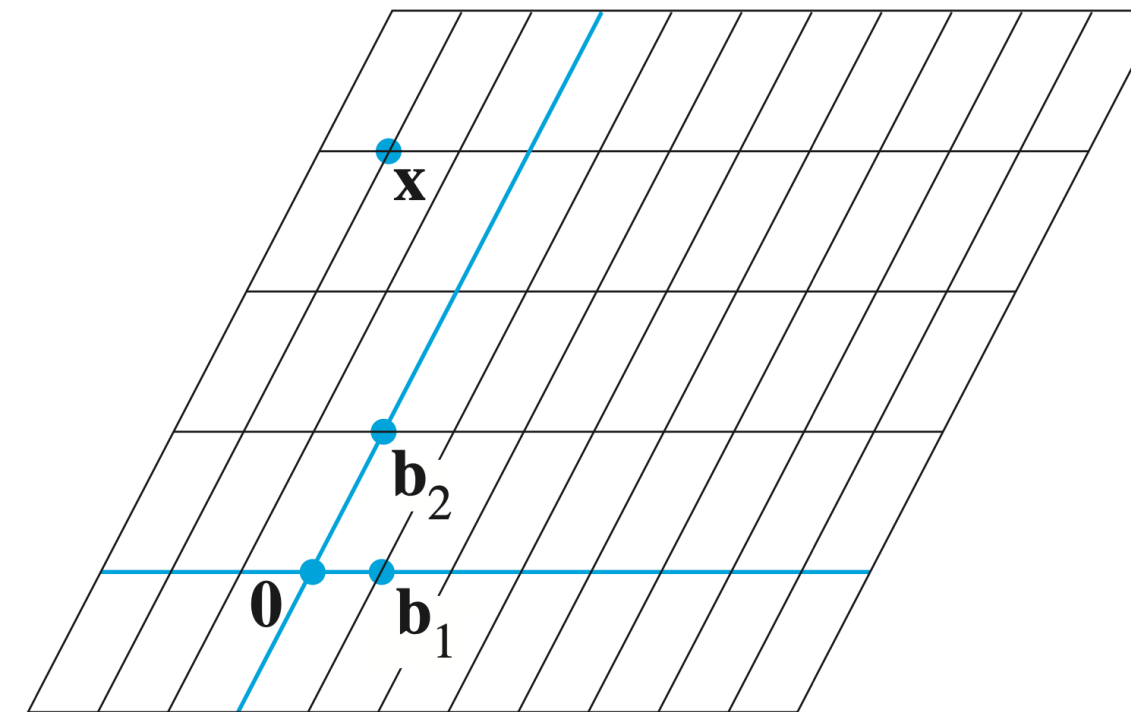


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$\mathcal{B}$  defines a "different grid for our graph paper"

# Recall: Coordinate Vectors

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

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

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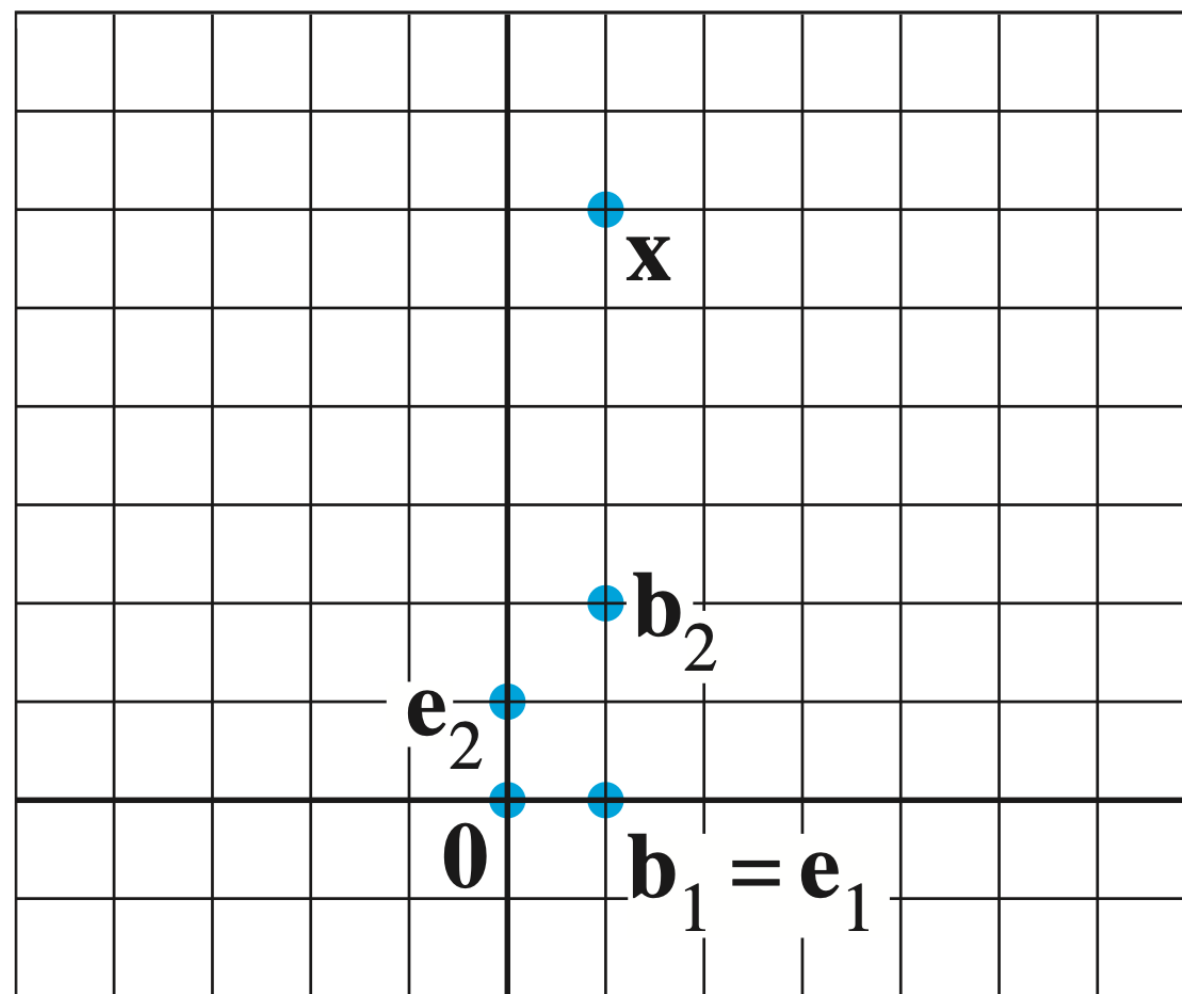
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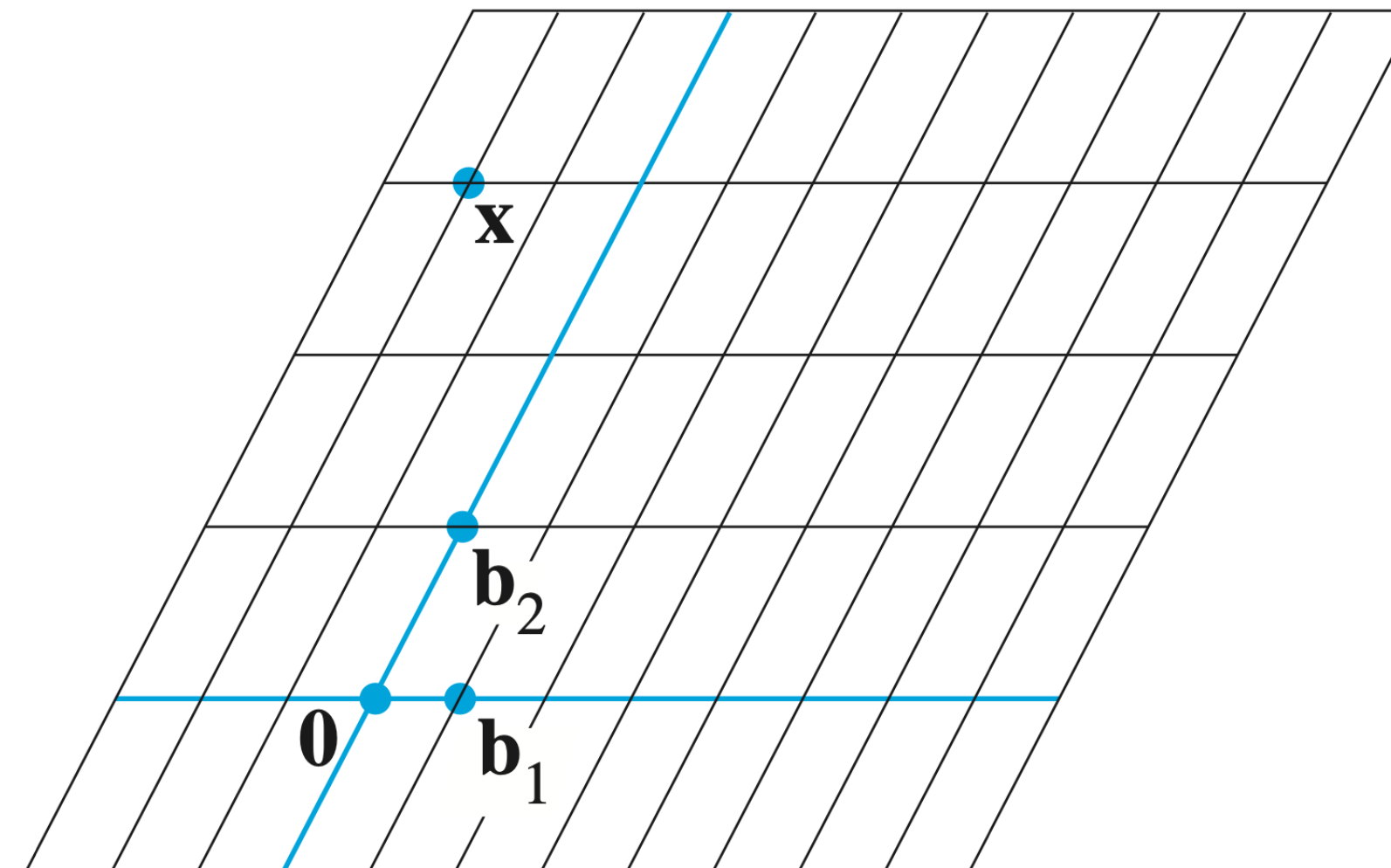
**Definition.** The coordinate vector of  $\mathbf{v}$  relative to  $\mathcal{B}$  is

$$[\mathbf{v}]_{\mathcal{B}} = \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix}$$

# Recall: Coordinate Vectors



**FIGURE 1** Standard graph paper.



**FIGURE 2**  $B$ -graph paper.

# Question (Conceptual)

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

*What is the matrix that implements the transformation*

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

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

# Change of Basis Matrix

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

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

**Matrix inverses perform changes of bases.**

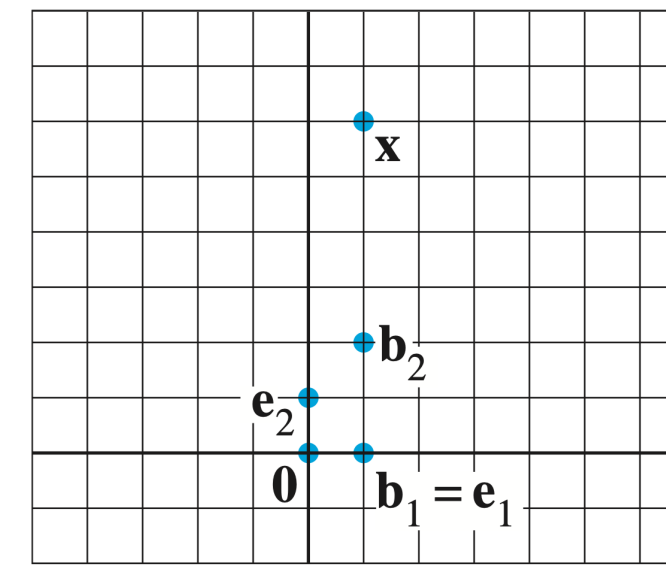


# How To: Change of Basis

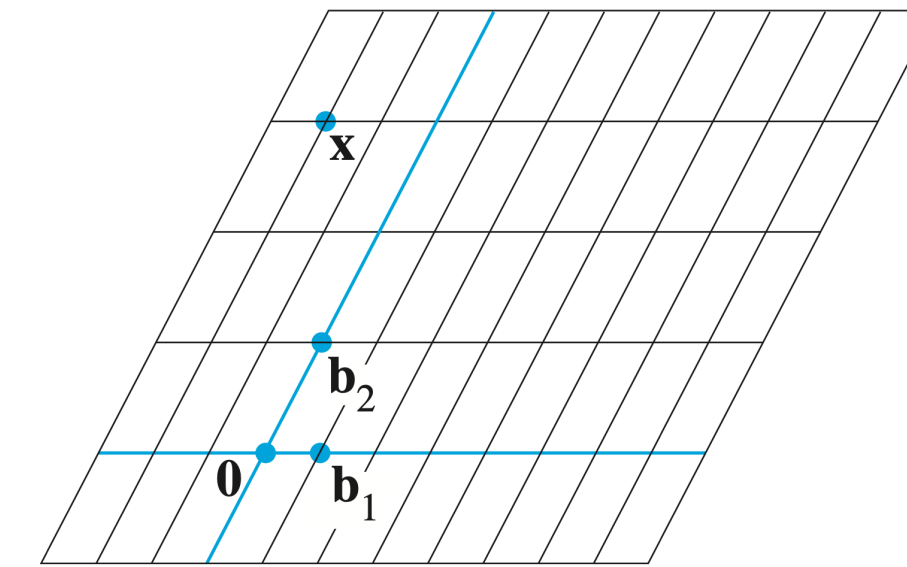
**Question.** Given a basis  $\mathcal{B} = (\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_n)$  of  $\mathbb{R}^n$ , find the matrix which implements  $\mathbf{x} \mapsto [\mathbf{x}]_{\mathcal{B}}$ .

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

# Example



**FIGURE 1** Standard graph paper.



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

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

# Diagonalization

# Diagonal Matrices

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

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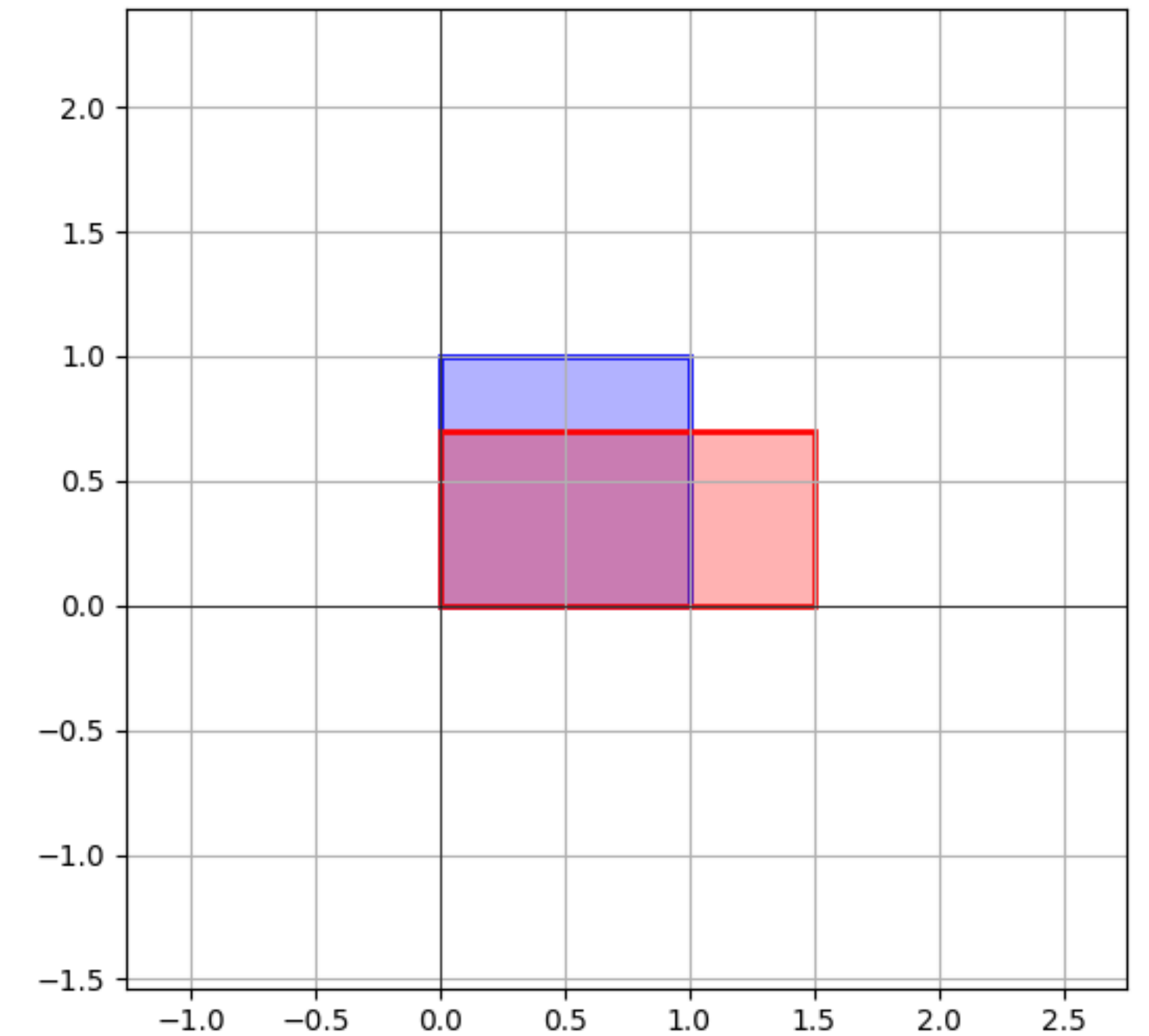
**Diagonal matrices are scaling matrices**

# Recall: Unequal Scaling

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

The diagonal entries scale 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:**

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

# Scaling and Eigenvectors

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

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

# 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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Factorizations can:

- » 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 same eigenvalues.

Verify:

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

# Important: Not all Matrices are Diagonalizable

**This is very different from the LU factorization**

We will need to figure out which matrices are diagonalizable

Question. *Is the zero matrix diagonalizable?*



# 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:

# 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

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

Verify:

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

# Columns of $P$ are eigenvectors

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

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**A diagonalization exposes a lot of information about  $A$**

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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

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# High Level

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

The columns of  $P$  form an eigenbasis 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.*

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

## Step 2: Eigenvectors

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

$$\lambda_1 = 1$$

Find **bases** of the corresponding eigenspaces  $\lambda_2 = -2$

*e.g.*

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

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

# Step 3: Construct P

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

Build the matrix with these vectors as the columns

*e.g.*

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

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

$$\lambda_1 = 1$$

$$\lambda_2 = -2$$

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

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

## Step 5: Construct D

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.*

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

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

$$\lambda_1 = 1$$

$$\lambda_2 = -2$$

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

## 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

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

# 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

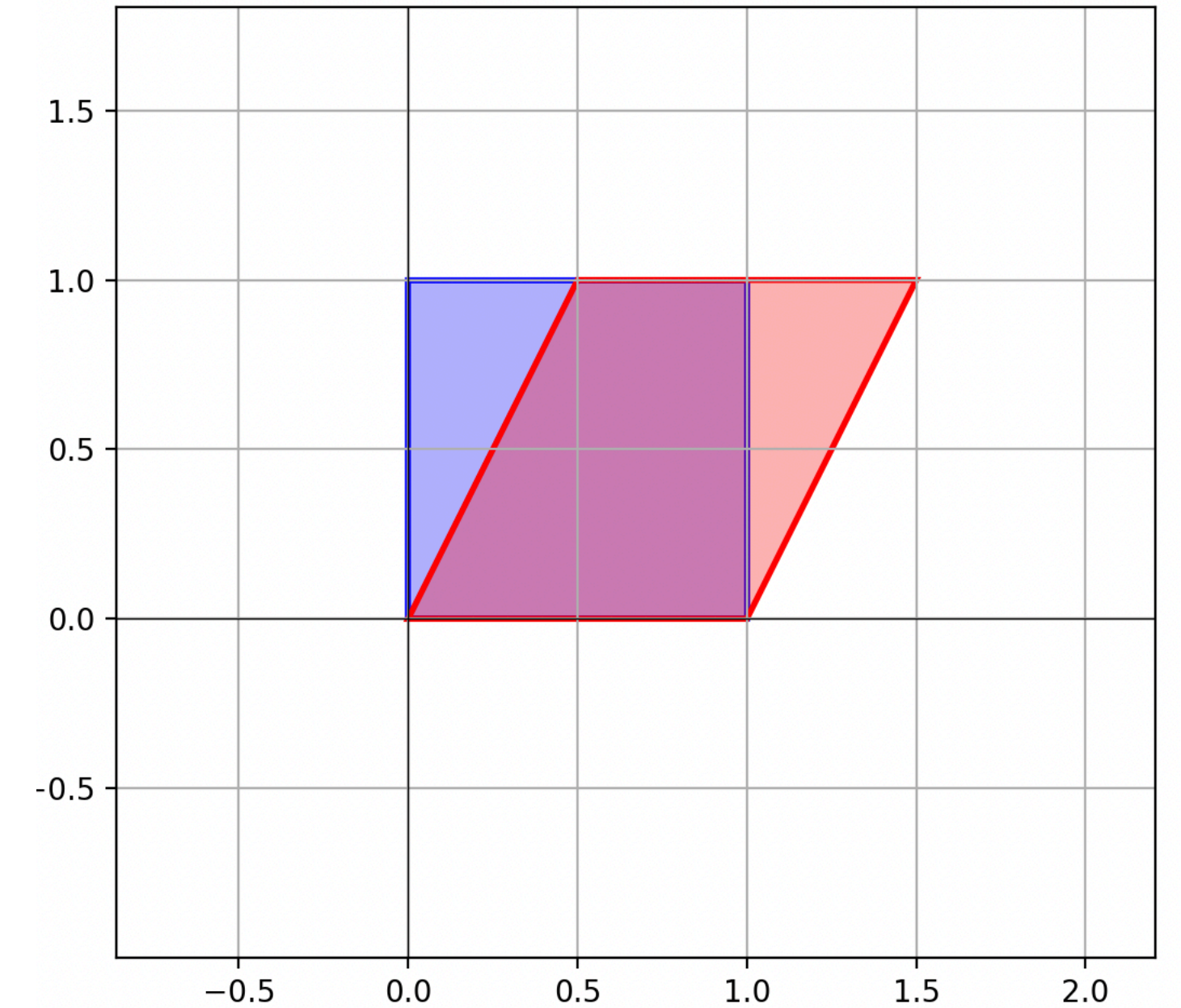
# Example of Failure: Shearing

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

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



## 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  $n$  distinct eigenvalues, then it is diagonalizable

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

# Example

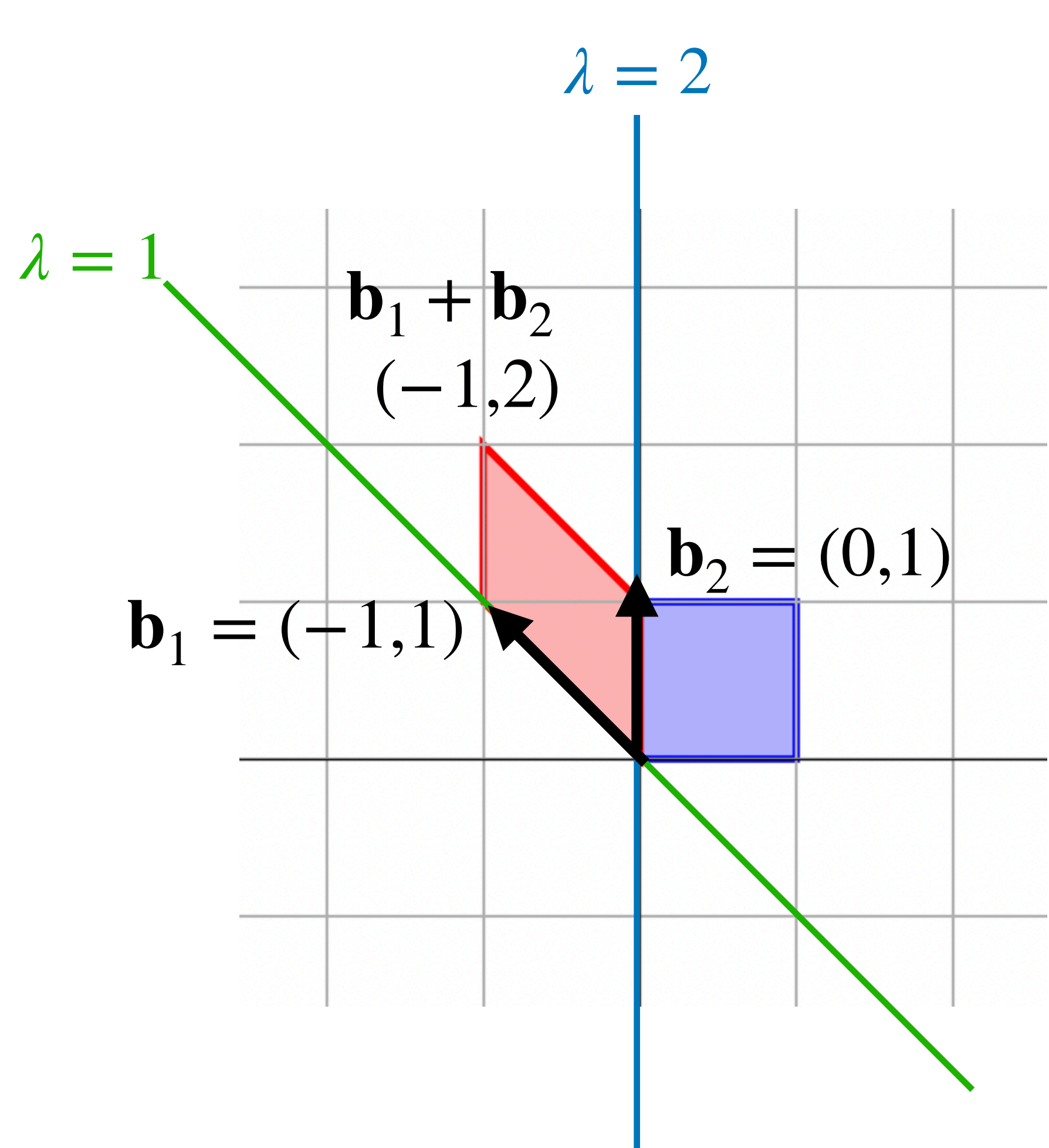
$$\begin{bmatrix} 2 & 0 \\ -1 & 1 \end{bmatrix}$$

*Find a diagonalization of the above matrix*

# The Picture

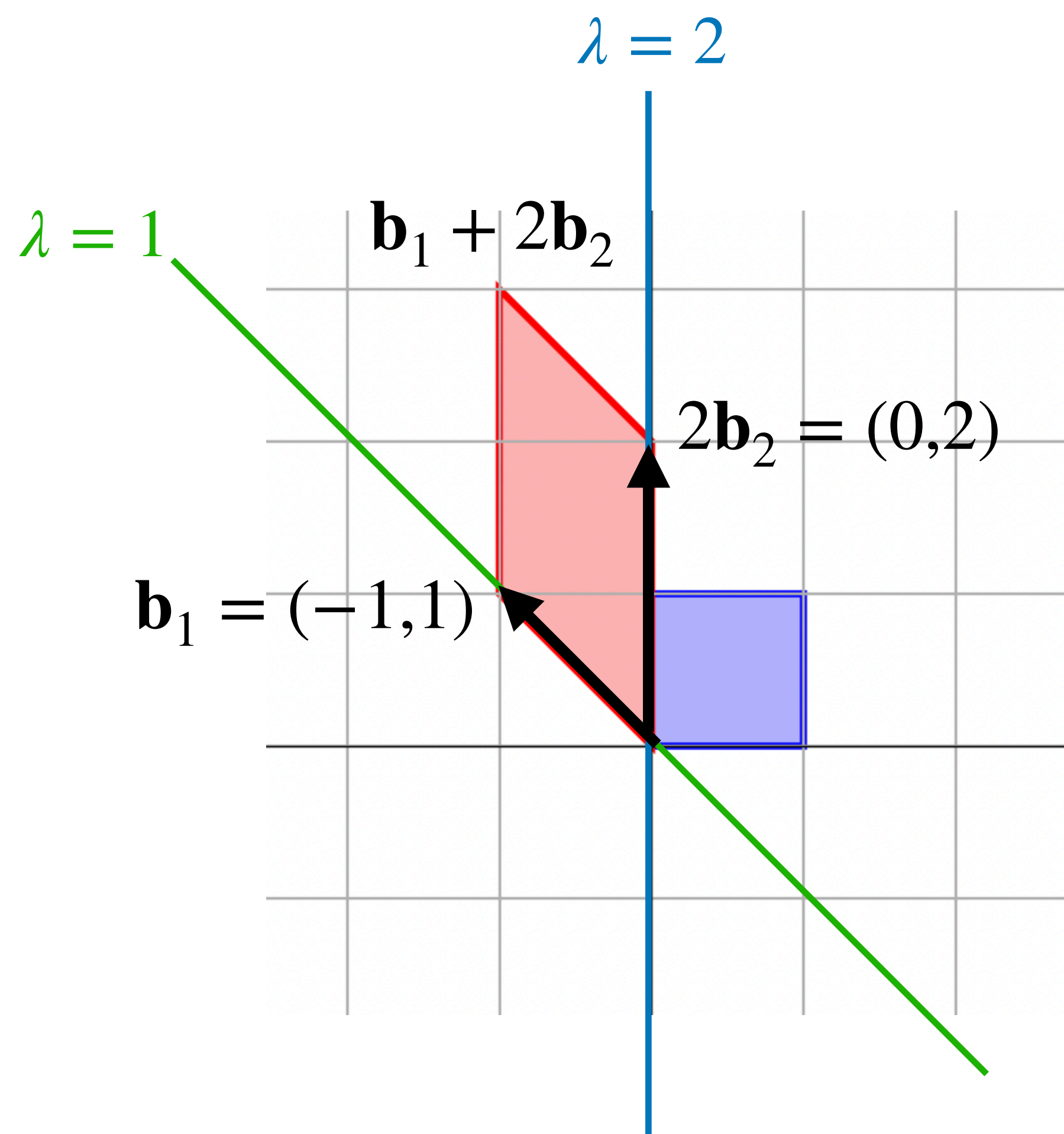
# Example (Geometric)

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



$A$

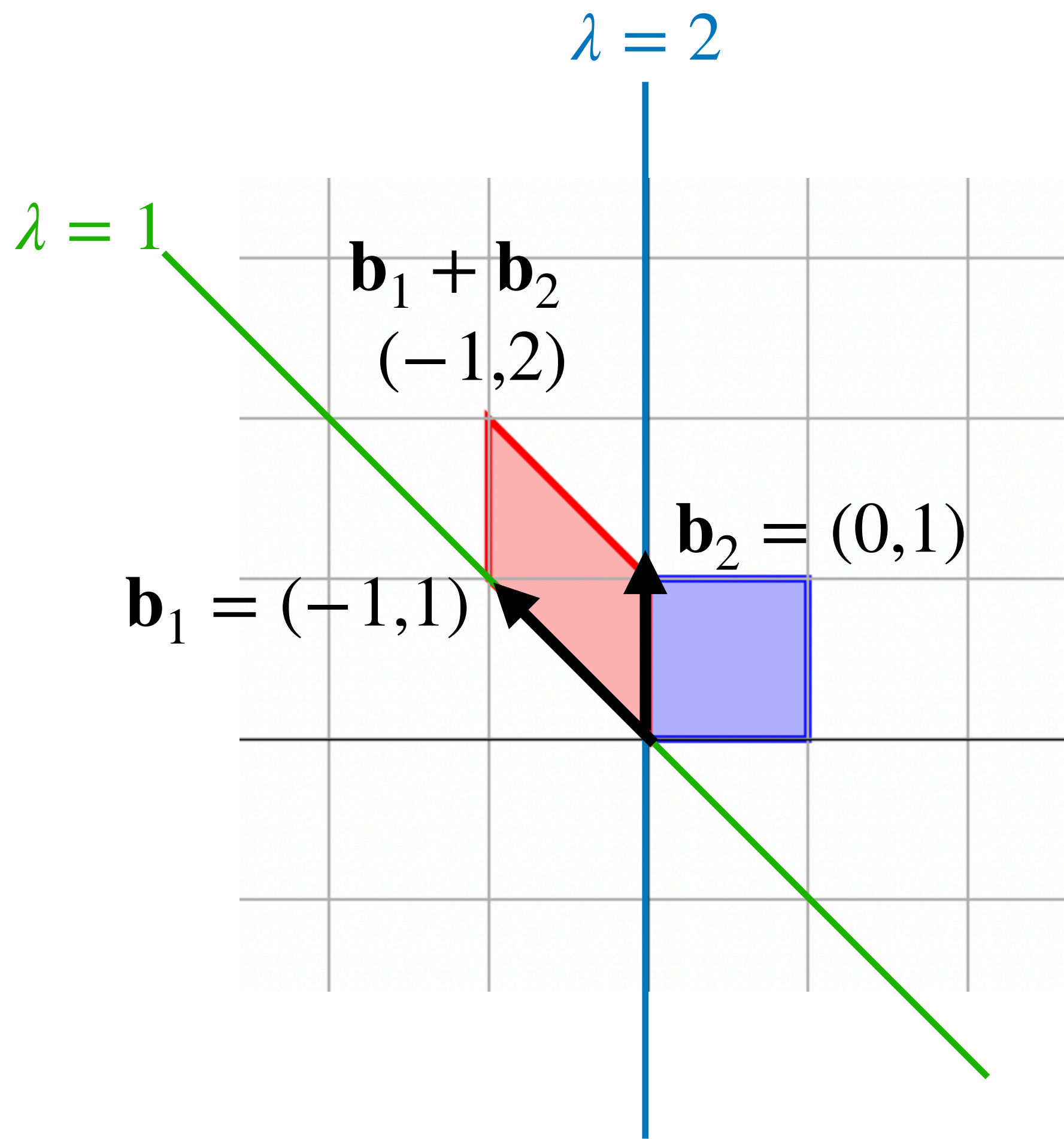
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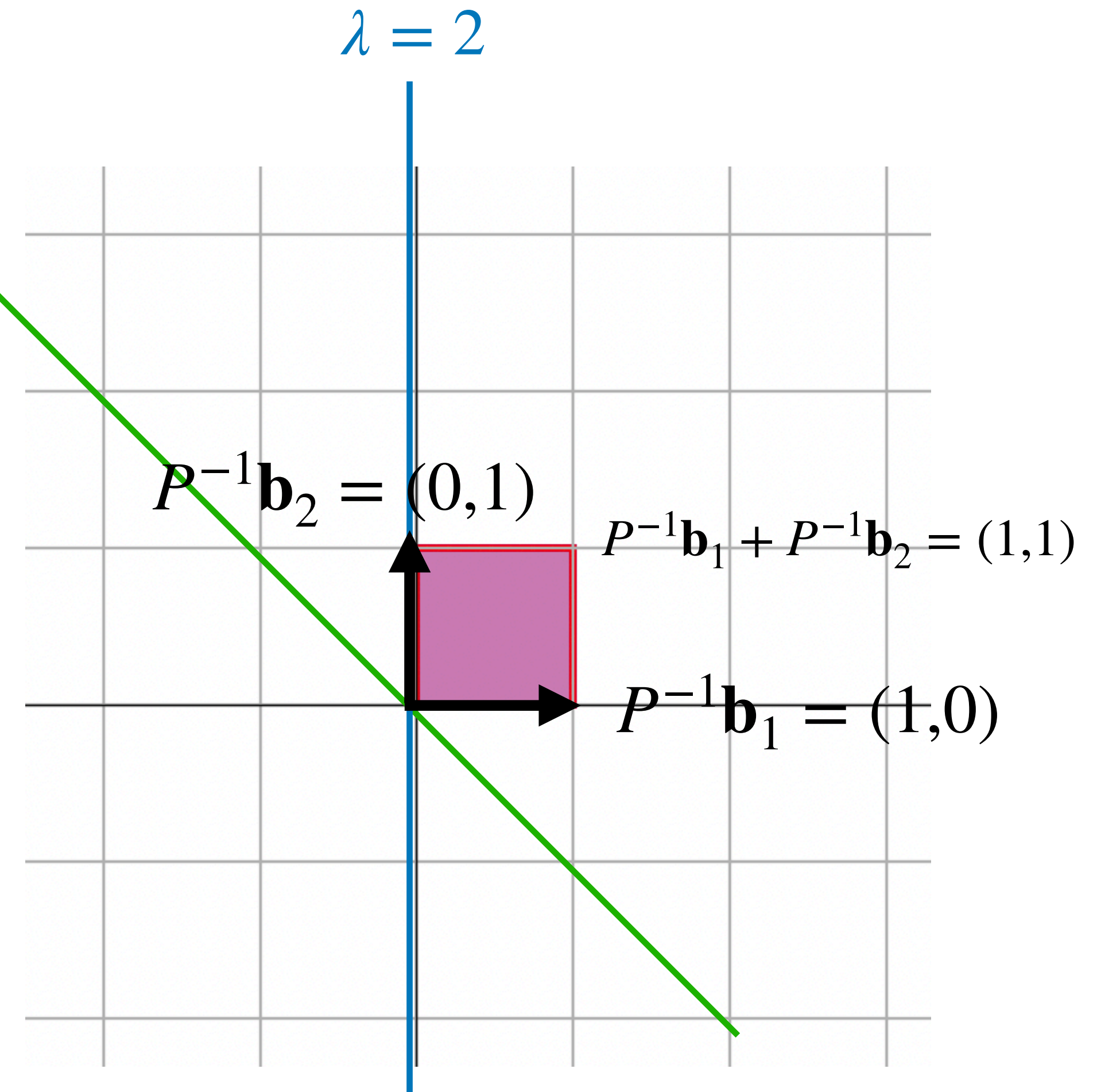


# Example (Geometric)

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

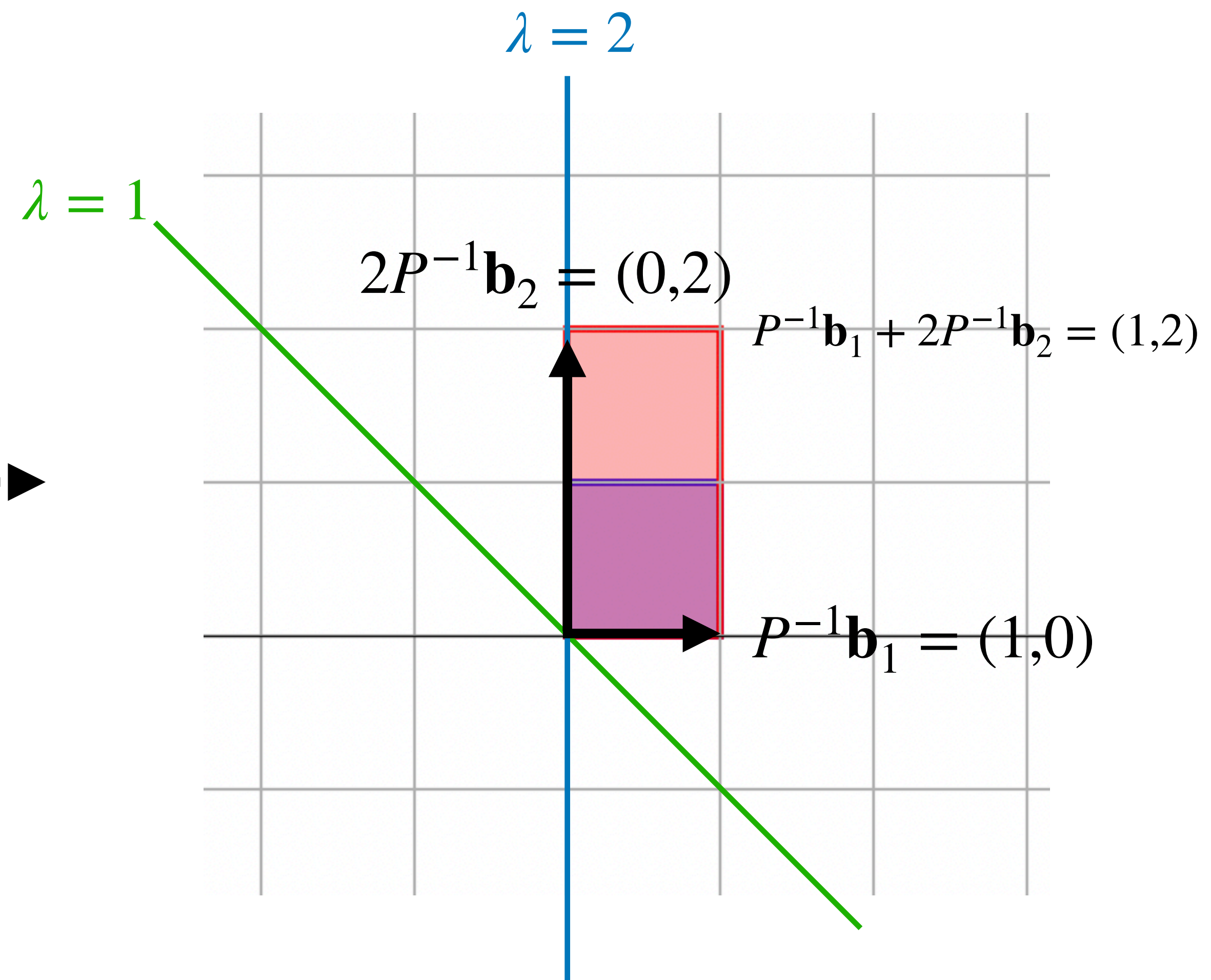
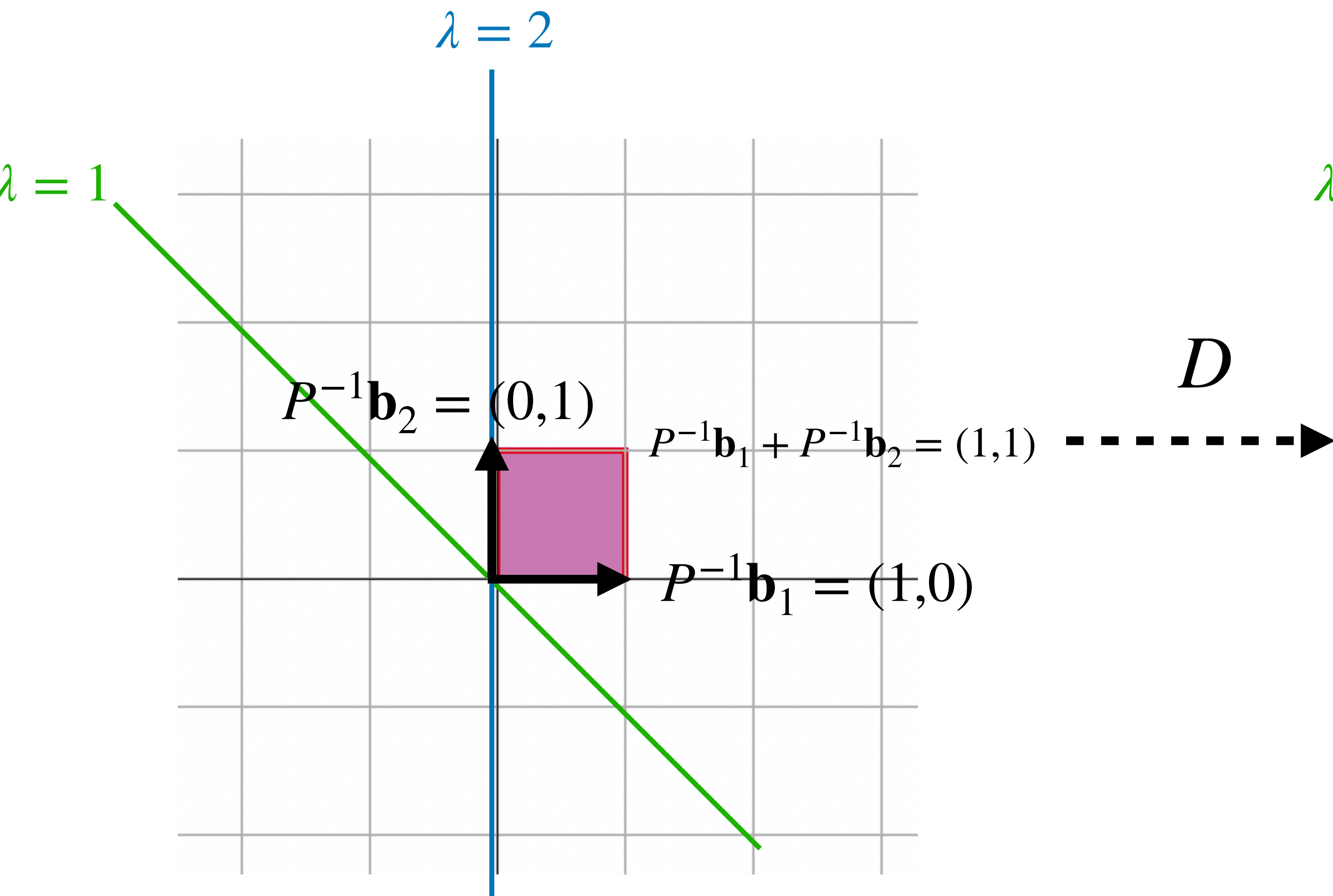


$P^{-1}$



# Example (Geometric)

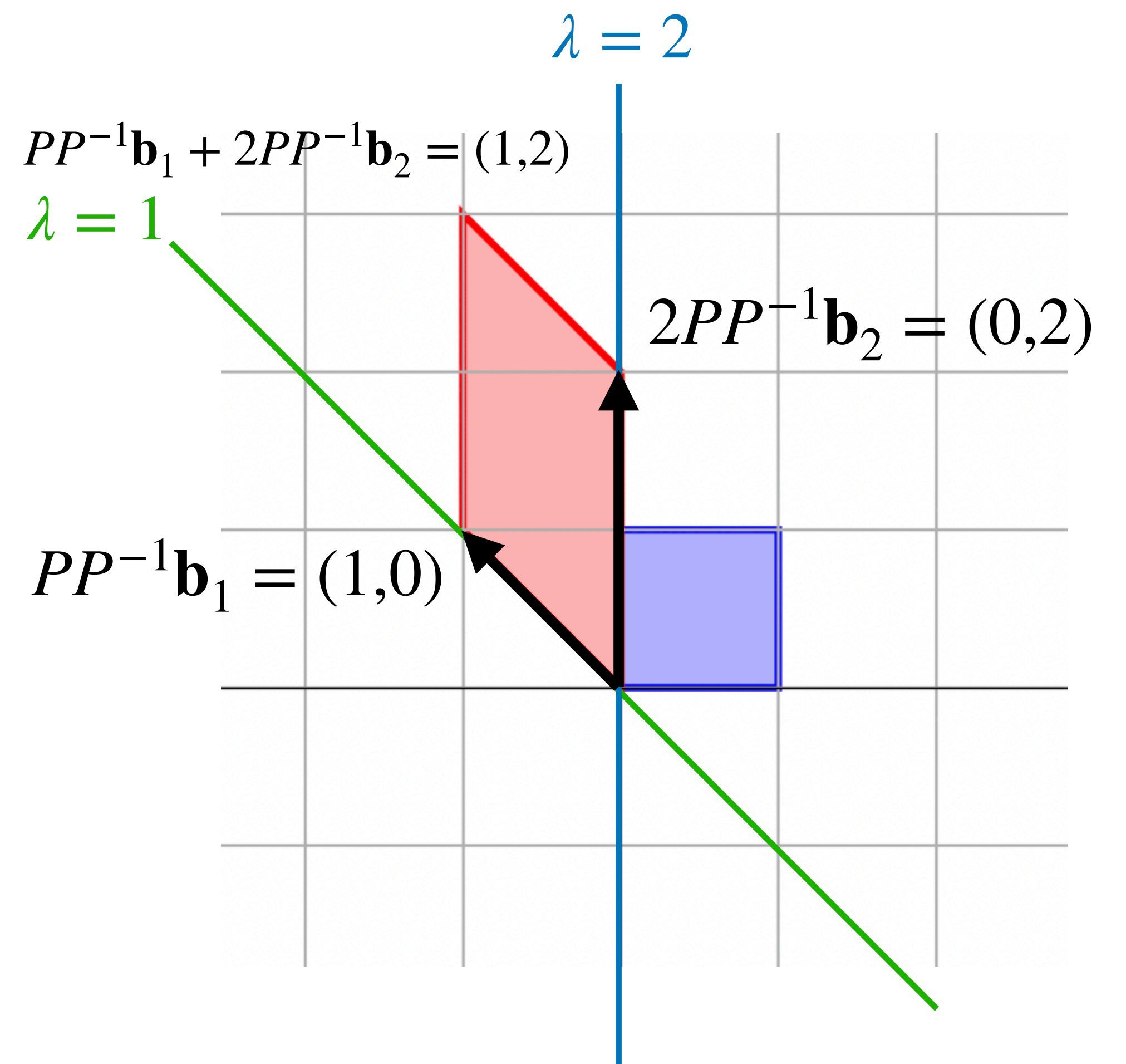
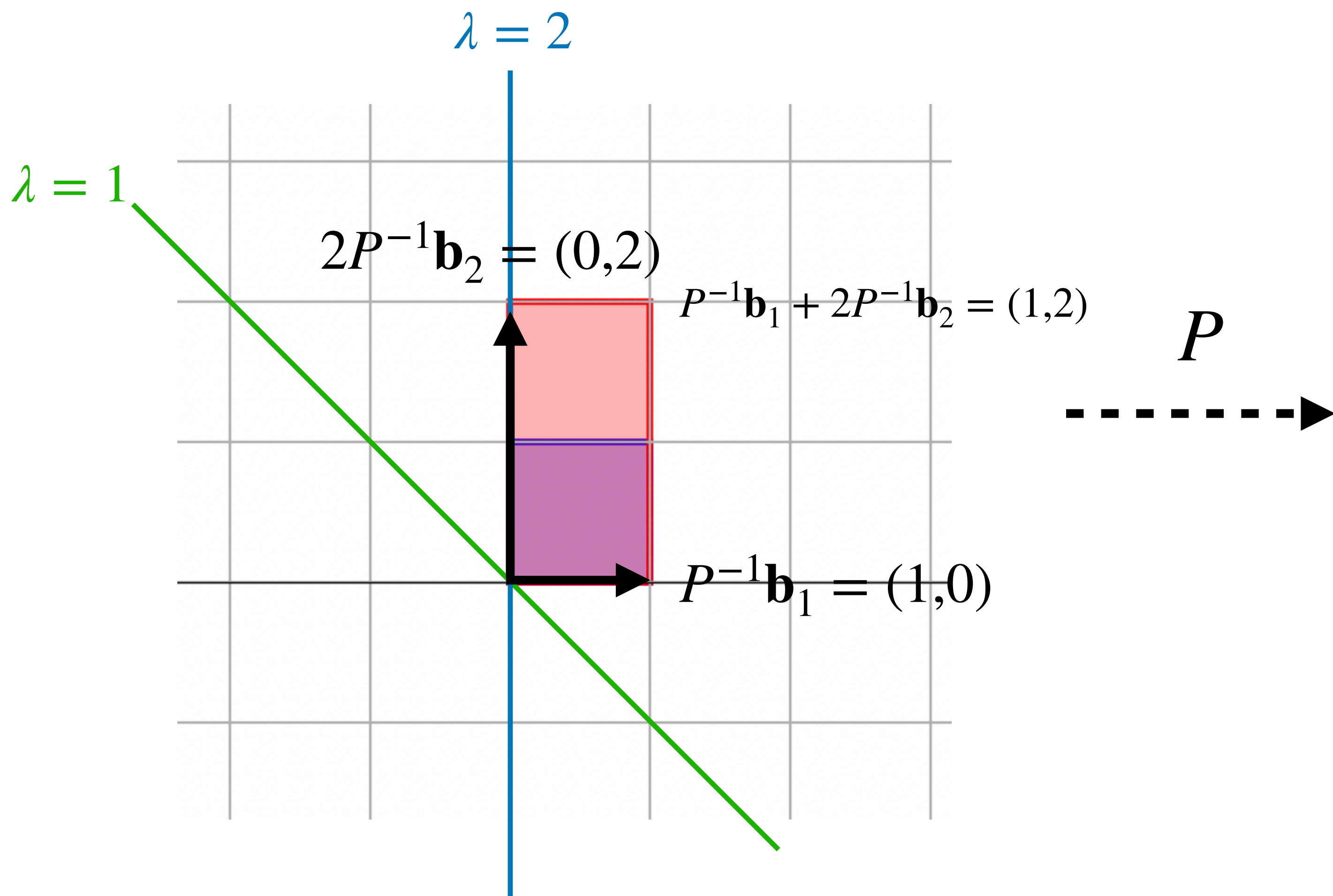
$$D = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}$$





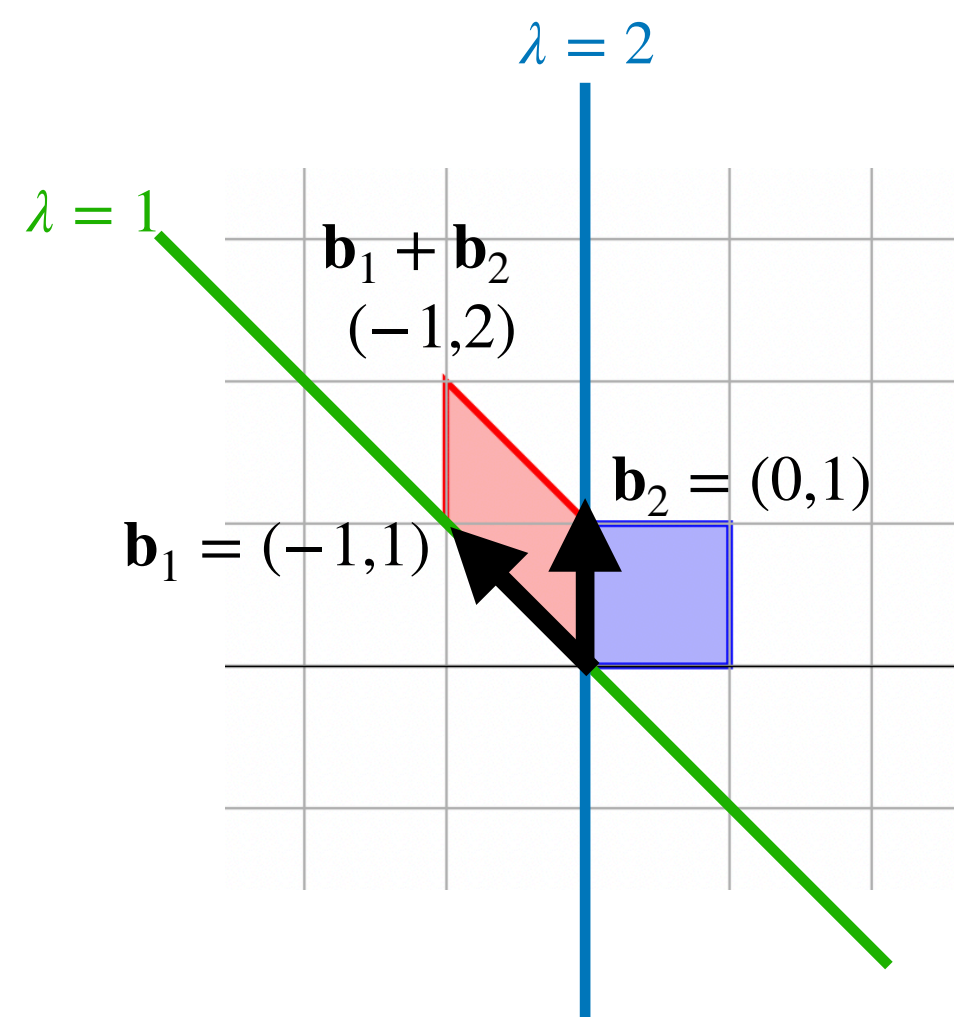
# Example (Geometric)

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





# Example (Geometric)



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

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

