

Practical Linear Algebra: A GEOMETRY TOOLBOX

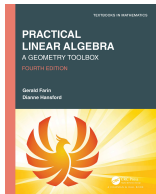
Fourth Edition

Chapter 5: 2×2 Linear Systems

Gerald Farin & Dianne Hansford

A K Peters/CRC Press
www.farinhanford.com/books/pla

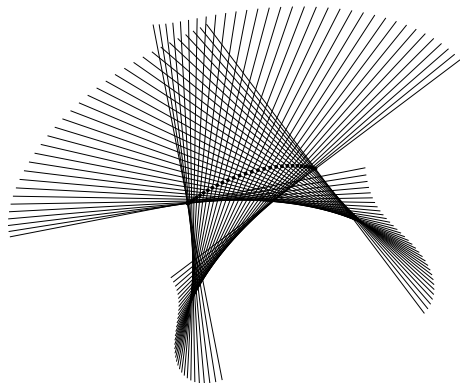
©2021



Outline

- 1 Introduction to 2×2 Linear Systems
- 2 Skew Target Boxes Revisited
- 3 The Matrix Form
- 4 A Direct Approach: Cramer's Rule
- 5 Gauss Elimination
- 6 Pivoting
- 7 Unsolvable Systems
- 8 Underdetermined Systems
- 9 Homogeneous Systems
- 10 Kernel
- 11 Undoing Maps: Inverse Matrices
- 12 Defining a Map
- 13 Change of Basis
- 14 Application: Intersecting Lines
- 15 WYSK

Introduction to 2×2 Linear Systems

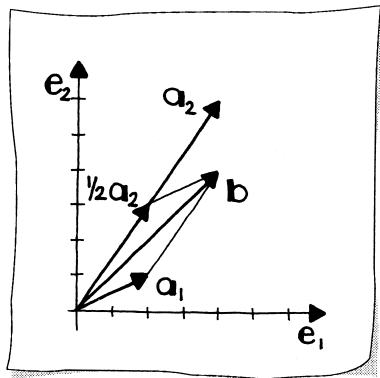


Two families of lines are shown
Intersections of corresponding line
pairs marked

For each intersection:
solve a 2×2 linear system

Skew Target Boxes Revisited

Geometry of a 2×2 system



\mathbf{a}_1 and \mathbf{a}_2 define a skew target box

Given \mathbf{b} with respect to the $[\mathbf{e}_1, \mathbf{e}_2]$ -system:

What are the components of \mathbf{b} with respect to the $[\mathbf{a}_1, \mathbf{a}_2]$ -system?

$$\mathbf{a}_1 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \quad \mathbf{a}_2 = \begin{bmatrix} 4 \\ 6 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 4 \\ 4 \end{bmatrix}$$

$$1 \times \begin{bmatrix} 2 \\ 1 \end{bmatrix} + \frac{1}{2} \times \begin{bmatrix} 4 \\ 6 \end{bmatrix} = \begin{bmatrix} 4 \\ 4 \end{bmatrix}$$

Skew Target Boxes Revisited

Two equations in the two unknowns u_1 and u_2

$$2u_1 + 4u_2 = 4$$

$$u_1 + 6u_2 = 4$$

Solution: $u_1 = 1$ and $u_2 = 1/2$

This chapter dedicated to solving these equations

The Matrix Form

Two equations

$$a_{1,1}u_1 + a_{1,2}u_2 = b_1$$

$$a_{2,1}u_1 + a_{2,2}u_2 = b_2$$

Also called a **linear system**

$$\begin{bmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$$

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

\mathbf{u} called the **solution** of linear system

Previous example:

$$\begin{bmatrix} 2 & 4 \\ 1 & 6 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 4 \\ 4 \end{bmatrix}$$

The Matrix Form

Recall geometric interpretation of $\mathbf{A}\mathbf{u} = \mathbf{b}$:

Express \mathbf{b} as a linear combination of \mathbf{a}_1 and \mathbf{a}_2

$$u_1\mathbf{a}_1 + u_2\mathbf{a}_2 = \mathbf{b}$$

At least one solution: linear system called **consistent**

Otherwise: called **inconsistent**

Three possibilities for **solution space**:

- 1 Exactly one solution vector \mathbf{u}

$$|A| \neq 0$$

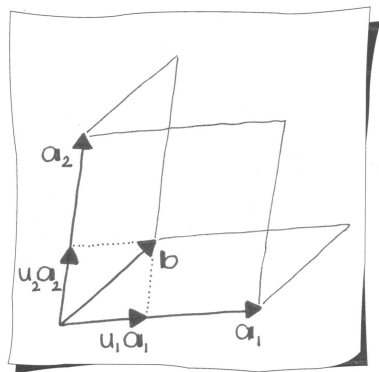
matrix has full rank and is non-singular

- 2 No solution (system is inconsistent)

- 3 Infinitely many solutions

(Sketches of each case to come)

A Direct Approach: Cramer's Rule



$$u_1 = \frac{\text{area}(\mathbf{b}, \mathbf{a}_2)}{\text{area}(\mathbf{a}_1, \mathbf{a}_2)} \quad u_2 = \frac{\text{area}(\mathbf{a}_1, \mathbf{b})}{\text{area}(\mathbf{a}_1, \mathbf{a}_2)}$$

Ratios of areas

Shear parallelogram formed by

— \mathbf{b}, \mathbf{a}_2 onto \mathbf{a}_1

— \mathbf{b}, \mathbf{a}_1 onto \mathbf{a}_2

(Shears preserve areas)

Signed area of a parallelogram given
by determinant

A Direct Approach: Cramer's Rule

Example:

$$\begin{bmatrix} 2 & 4 \\ 1 & 6 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 4 \\ 4 \end{bmatrix}$$

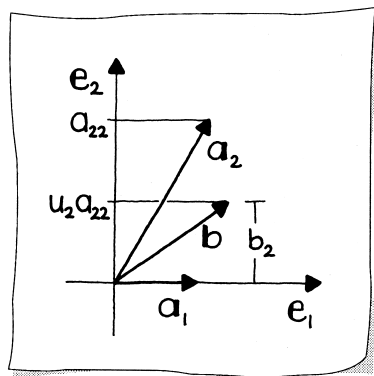
$$u_1 = \frac{\begin{vmatrix} 4 & 4 \\ 2 & 4 \end{vmatrix}}{\begin{vmatrix} 2 & 4 \\ 1 & 6 \end{vmatrix}} = \frac{8}{8} \quad u_2 = \frac{\begin{vmatrix} 2 & 4 \\ 1 & 4 \end{vmatrix}}{\begin{vmatrix} 2 & 4 \\ 1 & 6 \end{vmatrix}} = \frac{4}{8}$$

What if area spanned by \mathbf{a}_1 and \mathbf{a}_2 is zero?

Cramer's rule primarily of theoretical importance

For larger systems: expensive and numerically unstable

Gauss Elimination



Special 2×2 linear system:

$$\begin{bmatrix} a_{1,1} & a_{1,2} \\ 0 & a_{2,2} \end{bmatrix} \mathbf{u} = \mathbf{b}$$

Matrix is called *upper triangular*

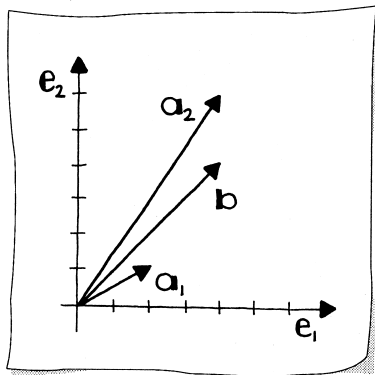
Solve with **back substitution**:

$$u_2 = b_2 / a_{2,2}$$

$$u_1 = \frac{1}{a_{1,1}} (b_1 - u_2 a_{1,2})$$

Diagonal elements key: called **pivots**

Gauss Elimination



Any linear system with non-singular matrix may be *transformed* to upper triangular via **forward elimination**

Process of forward elimination followed by back substitution is called **Gauss elimination**

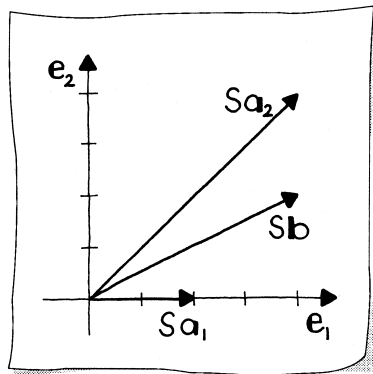
Example:

$$u_1 \begin{bmatrix} 2 \\ 1 \end{bmatrix} + u_2 \begin{bmatrix} 4 \\ 6 \end{bmatrix} = \begin{bmatrix} 4 \\ 4 \end{bmatrix}$$

Find u_1 and u_2

Key fact: *linear maps do not change linear combinations*

Gauss Elimination



Apply the same linear map to all vectors in system
then factors u_1 and u_2 won't change:

$$S \left(u_1 \begin{bmatrix} 2 \\ 1 \end{bmatrix} + u_2 \begin{bmatrix} 4 \\ 6 \end{bmatrix} \right) = S \begin{bmatrix} 4 \\ 4 \end{bmatrix}$$

Shear parallel to the e_2 -axis so that

$$\begin{bmatrix} 2 \\ 1 \end{bmatrix} \text{ is mapped to } \begin{bmatrix} 2 \\ 0 \end{bmatrix}$$

$$\text{Shear matrix: } S = \begin{bmatrix} 1 & 0 \\ -1/2 & 1 \end{bmatrix}$$

elementary matrix: applies one operation

Gauss Elimination

Transformed system:

$$\begin{bmatrix} 2 & 4 \\ 0 & 4 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 4 \\ 2 \end{bmatrix}$$

Next: **back substitution**

$$u_2 = 2/4 = 1/2,$$

$$u_1 = \frac{1}{2} \left(4 - 4 \times \frac{1}{2} \right) = 1$$

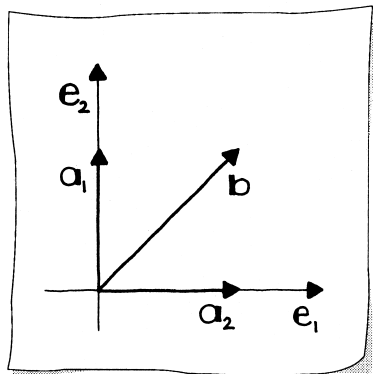
2×2 linear systems:

only one matrix entry to zero in the forward elimination procedure
(More algorithmic approach in Chapter 12 Gauss for Linear Systems)

Algebraically: transformed system by modifying the second row only

$$\text{row}_2 = -\frac{1}{2}\text{row}_1 + \text{row}_2.$$

Called an **elementary row operation**



Example:

$$\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Shearing \mathbf{a}_1 onto \mathbf{e}_1 -axis will not work

Solution: exchange two equations

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Exchanging equations (rows) so pivot is the largest in absolute value called **row or partial pivoting**

Used to improve numerical stability

Pivoting

Row exchange is a linear map

It is represented by the **permutation matrix**

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

Another example of an elementary matrix resulting in an elementary row operation

Pivoting

Example:

$$\begin{bmatrix} 0.0001 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

Shear \mathbf{a}_1 onto the \mathbf{e}_1 -axis

$$\begin{bmatrix} 0.0001 & 1 \\ 0 & -9999 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 1 \\ -9998 \end{bmatrix}$$

Performing back substitution

$$\mathbf{u}_t = \begin{bmatrix} 1.0001 \\ 0.9998\bar{9} \end{bmatrix} \quad (\text{"true" solution})$$

Suppose machine only stores three digits — system stored as

$$\begin{bmatrix} 0.0001 & 1 \\ 0 & -10000 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 1 \\ -10000 \end{bmatrix},$$

$$\mathbf{u}_r = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \quad (\text{"round-off" solution})$$

Not very close to the true solution \mathbf{u}_t : $\|\mathbf{u}_t - \mathbf{u}_r\| \approx 1.0001$

Pivoting

Pivoting dampens effects of round-off

$$\begin{bmatrix} 1 & 1 \\ 0.0001 & 1 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

Forward elimination

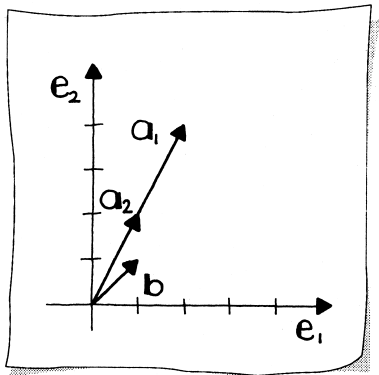
$$\begin{bmatrix} 1 & 1 \\ 0 & 0.9999 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 0.9998 \end{bmatrix}$$

$$\mathbf{u}_p = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad (\text{"pivoting" solution})$$

Closer to "true" solution: $\|\mathbf{u}_t - \mathbf{u}_p\| = 0.00014$

Unsolvable Systems

\mathbf{a}_1 and \mathbf{a}_2 are linearly dependent



$$\begin{bmatrix} 2 & 1 \\ 4 & 2 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Forward elimination
(shear \mathbf{a}_1 onto \mathbf{e}_1 -axis):

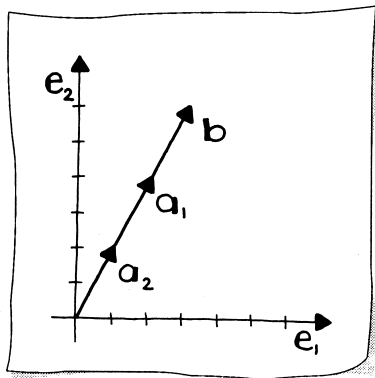
$$\begin{bmatrix} 2 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

Last equation: $0 = -1$
System is *inconsistent* \Rightarrow no solution

Approximate solution via least squares methods

Underdetermined Systems

b a multiple of \mathbf{a}_1 or \mathbf{a}_2



$$\begin{bmatrix} 2 & 1 \\ 4 & 2 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 3 \\ 6 \end{bmatrix}$$

Forward elimination
(shear \mathbf{a}_1 onto \mathbf{e}_1 -axis):

$$\begin{bmatrix} 2 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 3 \\ 0 \end{bmatrix}$$

Last equation: $0 = 0$
true, but a bit trivial

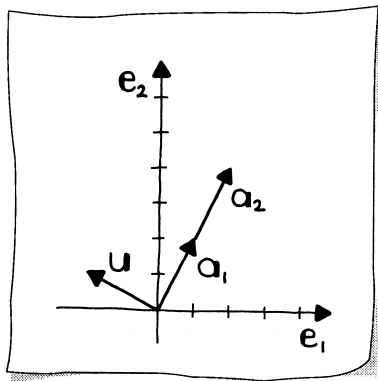
- This is one equation written twice
- System is **underdetermined**
- System **consistent**: at least one solution exists

Example: set $u_2 = 1$ then $u_1 = 1$

Homogeneous Systems

$$A\mathbf{u} = \mathbf{0}$$

Homogeneous: Right-hand side consists of zero vector



Trivial solution: $\mathbf{u} = \mathbf{0}$

— usually of little interest

If solution $\mathbf{u} \neq \mathbf{0}$ exists
then all $c\mathbf{u}$ are solutions

\Rightarrow *infinite number of solutions*

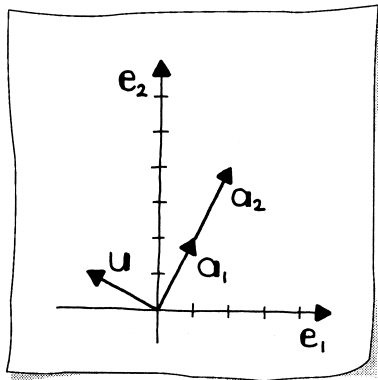
Vectors \mathbf{u} that satisfy the homogeneous system are orthogonal to the row vectors

Not all homogeneous systems have a non-trivial solution

— 2×2 matrices: only rank 1 maps
 $\Rightarrow \mathbf{a}_1$ and \mathbf{a}_2 linearly dependent

If only trivial solution exist $\Rightarrow A$ invertible

Homogeneous Systems



$$\begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix} \mathbf{u} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$\mathbf{a}_2 = 2\mathbf{a}_1 \Rightarrow A$ maps all vectors onto line defined by $\mathbf{0}, \mathbf{a}_1$

Forward elimination:

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

Pick $u_2 = 1$ then

back substitution gives $u_1 = -2$

Any $c\mathbf{u}$ perpendicular to \mathbf{a}_1 is a solution: $\mathbf{a}_1 \cdot \mathbf{u} = 0$

— all such \mathbf{u} make up **kernel** or **null space** of A

Homogeneous Systems

Example: only the trivial solution

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

Columns of A are linearly independent

A does not reduce dimensionality \Rightarrow cannot map $\mathbf{u} \neq \mathbf{0}$ to $\mathbf{0}$

Forward elimination:

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

Back substitution: $\mathbf{u} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$

Homogeneous Systems

Example: row pivoting not helpful – need column pivoting

$$\begin{bmatrix} 0 & 1/2 \\ 0 & 0 \end{bmatrix} \mathbf{u} = \mathbf{0}.$$

Column pivoting:

$$\begin{bmatrix} 1/2 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} u_2 \\ u_1 \end{bmatrix} = \mathbf{0}.$$

(Exchange unknowns too)

Set $u_1 = 1$ and back substitution results in $u_2 = 0$

Solutions: $\mathbf{u} = c \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

Kernel

All vectors \mathbf{u} that satisfy a homogeneous system

$$A\mathbf{u} = \mathbf{0}$$

make up the **kernel** or **null space** of the matrix

Vectors \mathbf{u} in the kernel are orthogonal to the row space of A

The dimension of the kernel is called the **nullity** of A

For 2×2 matrices:

$$\text{rank} + \text{nullity} = 2$$

Example: Homogeneous system with non-trivial solution
Rank = 1 Nullity = 1 Notice that $\mathbf{a}_2 = 2\mathbf{a}_1$

$$\begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix} \mathbf{u} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Example: Homogeneous system with only trivial solution
Rank = 2 Nullity = 0 \mathbf{a}_1 and \mathbf{a}_2 linearly independent

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

Kernel = the central, most important part of something

In linear algebra, it reveals information about a linear map (matrix) and also the solution to a linear system.

The kernel is a subspace of the domain with the following properties

- 1 Always contains the zero vector since $A\mathbf{0} = \mathbf{0}$
- 2 If \mathbf{u} is in the kernel, then $c\mathbf{u}$ is in the kernel:
 $c(A\mathbf{u}) = c\mathbf{0}$, thus $A(c\mathbf{u}) = \mathbf{0}$
- 3 If \mathbf{u} and \mathbf{v} are in the kernel, then $\mathbf{u} + \mathbf{v}$ is in the kernel
(distributive law)

Why is knowledge of the kernel useful?

Reveals the **existence and uniqueness of a solution**

If the nullity = 0, then the solution is unique.

If the nullity ≥ 1 , then the solution is not unique
and the nullity reveals the number of parameters available to specify a
solution

Kernel

Example: Rank 1, Nullity 1

$$\begin{bmatrix} 2 & 1 \\ 4 & 2 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 3 \\ 6 \end{bmatrix} \quad \text{with specific solution } \mathbf{u}_s = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Homogeneous linear system

$$\begin{bmatrix} 2 & 1 \\ 4 & 2 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad \text{with a kernel solution } \mathbf{u}_k = \begin{bmatrix} -1/2 \\ 1 \end{bmatrix}$$

Then all vectors

$$\mathbf{u} = \mathbf{u}_s + c\mathbf{u}_k$$

are solutions to the linear system since

$$A(\mathbf{u}_s + c\mathbf{u}_k) = A\mathbf{u}_s + Ac\mathbf{u}_k = A\mathbf{u}_s$$

\Rightarrow there is a one parameter family of solutions

Check an example with $c = 2$

Undoing Maps: Inverse Matrices

How to *undo* a linear map

Given $A\mathbf{u} = \mathbf{b}$

What matrix B maps \mathbf{b} back to \mathbf{u} : $\mathbf{u} = B\mathbf{b}$?

B is the **inverse map**

Recall: shears can be used to zero matrix elements

$$S_1 A \mathbf{u} = S_1 \mathbf{b}.$$

Example:

$$\begin{bmatrix} 2 & 4 \\ 1 & 6 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 4 \\ 4 \end{bmatrix} \Rightarrow \begin{bmatrix} 2 & 4 \\ 0 & 4 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 4 \\ 2 \end{bmatrix}$$

Undoing Maps: Inverse Matrices

Second shear: $S_2 S_1 A \mathbf{u} = S_2 S_1 \mathbf{b}$ (map new \mathbf{a}_2 to the \mathbf{e}_2 -axis)

$$S_2 = \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix} \quad \text{results in} \quad \begin{bmatrix} 2 & 0 \\ 0 & 4 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$$

Non-uniform scaling S_3 (map $\mathbf{a}_1, \mathbf{a}_2$ onto $\mathbf{e}_1, \mathbf{e}_2$)

$$S_3 = \begin{bmatrix} 1/2 & 0 \\ 0 & 1/4 \end{bmatrix} \quad \text{results in} \quad \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 1/2 \end{bmatrix}$$

All together:

$$S_3 S_2 S_1 A \mathbf{u} = S_3 S_2 S_1 \mathbf{b}$$
$$I \mathbf{u} = A^{-1} \mathbf{b}$$

I called the **identity matrix**

A^{-1} called the **inverse matrix** and A called **invertible**

Undoing Maps: Inverse Matrices

$$A^{-1}A = I \quad \text{and} \quad AA^{-1} = I \quad I^{-1} = I$$

Inverse of a scaling:

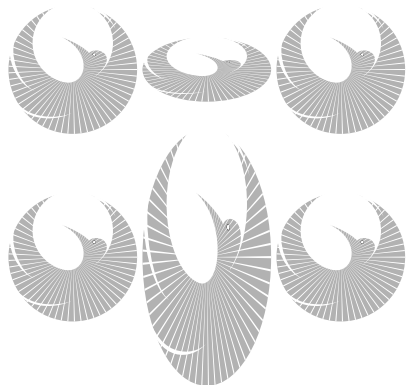
$$\begin{bmatrix} s & 0 \\ 0 & t \end{bmatrix}^{-1} = \begin{bmatrix} 1/s & 0 \\ 0 & 1/t \end{bmatrix}$$

Example:

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

Top: original Phoenix, scale, inverse scale

Bottom: original Phoenix, inverse scale, original scale



Undoing Maps: Inverse Matrices

Top: original Phoenix, shear, inverse shear

Bottom: original Phoenix, inverse shear, original shear



Undoing Maps: Inverse Matrices

Rotation matrices:

$$R_{-\alpha} = R_{\alpha}^{-1} = R_{\alpha}^T$$

Rotation matrix is an **orthogonal matrix**:

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

Column vectors satisfy $\|\mathbf{a}_1\| = 1$, $\|\mathbf{a}_2\| = 1$ and $\mathbf{a}_1 \cdot \mathbf{a}_2 = 0$

\Rightarrow vectors called **orthonormal**

\Rightarrow these linear maps called ***rigid body motions***

Characterized by determinant = ± 1

Undoing Maps: Inverse Matrices

$$A^{-1-1} = A \quad \text{and} \quad A^{-1T} = A^{T-1}$$



Example:

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

Top: I , A^{-1} , A^{-1T}

Bottom: I , A^T , A^{T-1}

Undoing Maps: Inverse Matrices

How to compute A 's inverse?

Start with

$$AA^{-1} = I$$

Denote two (unknown) columns of A^{-1} by $\bar{\mathbf{a}}_1$ and $\bar{\mathbf{a}}_2$

Denote columns of I by \mathbf{e}_1 and \mathbf{e}_2

$$A \begin{bmatrix} \bar{\mathbf{a}}_1 & \bar{\mathbf{a}}_2 \end{bmatrix} = \begin{bmatrix} \mathbf{e}_1 & \mathbf{e}_2 \end{bmatrix}$$

Short for two linear systems

$$A\bar{\mathbf{a}}_1 = \mathbf{e}_1 \quad \text{and} \quad A\bar{\mathbf{a}}_2 = \mathbf{e}_2$$

Both systems have the same matrix A

Defining a Map

Matrices map vectors to vectors

If \mathbf{v}_1 and \mathbf{v}_2 mapped to \mathbf{v}'_1 and \mathbf{v}'_2 , what matrix A did it?

$$A\mathbf{v}_1 = \mathbf{v}'_1 \quad \text{and} \quad A\mathbf{v}_2 = \mathbf{v}'_2$$

Combining into a matrix equation:

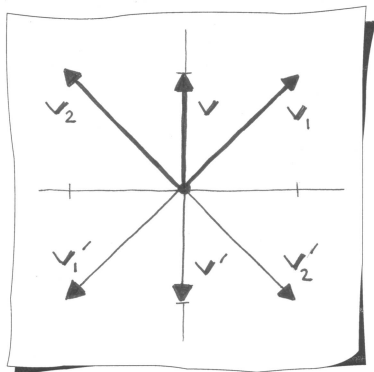
$$A[\mathbf{v}_1 \ \mathbf{v}_2] = [\mathbf{v}'_1 \ \mathbf{v}'_2] \quad \text{or} \quad AV = V'$$

Solution: find V^{-1} , then $A = V'V^{-1}$

\mathbf{v}_1 and \mathbf{v}_2 must be linearly independent for V^{-1} to exist

Defining a Map

Example:



$$\mathbf{v}_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad \text{and} \quad \mathbf{v}_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

$$\mathbf{v}'_1 = \begin{bmatrix} -1 \\ -1 \end{bmatrix} \quad \text{and} \quad \mathbf{v}'_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

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

$$\begin{aligned} A = V'V^{-1} &= \begin{bmatrix} -1 & 1 \\ -1 & -1 \end{bmatrix} \begin{bmatrix} 1/2 & 1/2 \\ -1/2 & 1/2 \end{bmatrix} \\ &= \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} \end{aligned}$$

Change of Basis

Basics of coordinate systems

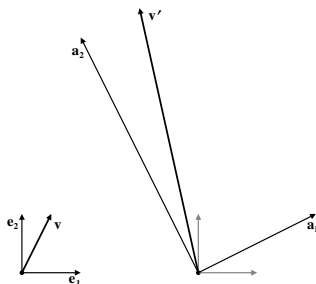


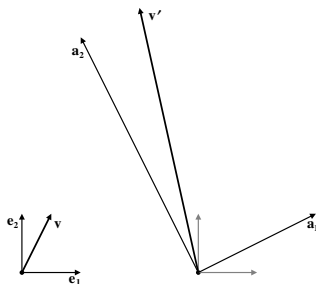
Figure (left): $\mathbf{v} = \begin{bmatrix} 1/2 \\ 1 \end{bmatrix} \Rightarrow \mathbf{v} = \frac{1}{2}\mathbf{e}_1 + 1\mathbf{e}_2$

Basis vectors and the origin establish a grid for navigating 2D space

Change of Basis

Choose any set of linearly independent vectors

Figure (right): $[\mathbf{a}_1, \mathbf{a}_2]$ -system



$$\mathbf{a}_1 = \begin{bmatrix} 2 \\ 1 \end{bmatrix} \quad \mathbf{a}_2 = \begin{bmatrix} -2 \\ 4 \end{bmatrix} \quad \mathbf{v}' = \frac{1}{2}\mathbf{a}_1 + \mathbf{a}_2$$

In the $[\mathbf{a}_1, \mathbf{a}_2]$ -system

$$\begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad \begin{bmatrix} 0 \\ 1 \end{bmatrix} \quad \text{defined as } \mathbf{a}_1 = 1\mathbf{a}_1 + 0\mathbf{a}_2 \quad \text{and} \quad \mathbf{a}_2 = 0\mathbf{a}_1 + 1\mathbf{a}_2$$

Change of Basis

\mathbf{a}_1 and \mathbf{a}_2 play the same role in the $[\mathbf{a}_1, \mathbf{a}_2]$ -system
as do \mathbf{e}_1 and \mathbf{e}_2 in the $[\mathbf{e}_1, \mathbf{e}_2]$ -system

Basis vector must not be orthogonal nor unit length but sometimes
desirable

Change of Basis

Change of basis problems

- 1 Given $\mathbf{v}'_{\mathbf{a}} = \begin{bmatrix} 1/2 \\ 1 \end{bmatrix}_{\mathbf{a}}$ in the $[\mathbf{a}_1, \mathbf{a}_2]$ -system, what are the components of this vector in the $[\mathbf{e}_1, \mathbf{e}_2]$ -system, referred to as $\mathbf{v}'_{\mathbf{e}}$?
- 2 Given $\mathbf{v}_{\mathbf{e}} = \begin{bmatrix} 1/2 \\ 1 \end{bmatrix}_{\mathbf{e}}$ in the $[\mathbf{e}_1, \mathbf{e}_2]$ -system, what are the components of this vector in the $[\mathbf{a}_1, \mathbf{a}_2]$ -system, referred to as $\mathbf{v}_{\mathbf{a}}$?

- Subscripts have been added to the vectors to make clear their defining basis vectors.
- Extra square brackets are added when needed to improve readability.

Change of Basis

Question 1:

Write $[\mathbf{a}_1, \mathbf{a}_2]$ -system vectors in terms of the $[\mathbf{e}_1, \mathbf{e}_2]$ -system vectors

$$[\mathbf{a}_1]_{\mathbf{e}} = 2\mathbf{e}_1 + 1\mathbf{e}_2 \quad \text{and} \quad [\mathbf{a}_2]_{\mathbf{e}} = -2\mathbf{e}_1 + 4\mathbf{e}_2$$

Vector \mathbf{v}' with respect to the $[\mathbf{e}_1, \mathbf{e}_2]$ -system

$$\mathbf{v}'_{\mathbf{e}} = \frac{1}{2}[\mathbf{a}_1]_{\mathbf{e}} + 1[\mathbf{a}_2]_{\mathbf{e}}$$

In matrix form

$$\mathbf{v}'_{\mathbf{e}} = A \begin{bmatrix} 1/2 \\ 1 \end{bmatrix}_{\mathbf{a}} = \begin{bmatrix} -1 \\ 9/2 \end{bmatrix}_{\mathbf{e}} \quad \text{where} \quad A = [[\mathbf{a}_1]_{\mathbf{e}} \quad [\mathbf{a}_2]_{\mathbf{e}}] = \begin{bmatrix} 2 & -2 \\ 1 & 4 \end{bmatrix}$$

A maps a vector in the $[\mathbf{a}_1, \mathbf{a}_2]$ -system to one in the $[\mathbf{e}_1, \mathbf{e}_2]$ -system

Called a **change of basis matrix**

Change of Basis

Question 2:

Apply the inverse map to the matrix in Question 1

$$A^{-1} \begin{bmatrix} -1 \\ 9/2 \end{bmatrix}_{\mathbf{e}} = \begin{bmatrix} 1/2 \\ 1 \end{bmatrix}_{\mathbf{a}} \quad A^{-1} = \begin{bmatrix} 2/5 & 1/5 \\ -1/10 & 1/5 \end{bmatrix}$$

Components of \mathbf{e}_1 and \mathbf{e}_2 in the $[\mathbf{a}_1, \mathbf{a}_2]$ -system are revealed in A^{-1}

$$\mathbf{e}_1 = \frac{2}{5}\mathbf{a}_1 - \frac{1}{10}\mathbf{a}_2 \quad \text{and} \quad \mathbf{e}_2 = \frac{1}{5}\mathbf{a}_1 + \frac{1}{5}\mathbf{a}_2.$$

We could have constructed A^{-1} by solving two 2×2 linear systems

$$A[\mathbf{e}_1]_{\mathbf{a}} = \mathbf{e}_1 \quad \text{and} \quad A[\mathbf{e}_2]_{\mathbf{a}} = \mathbf{e}_2 \quad \text{then} \quad A^{-1} = [[\mathbf{e}_1]_{\mathbf{a}} \quad [\mathbf{e}_2]_{\mathbf{a}}]$$

The answer to Question 2:

$$\mathbf{v}_{\mathbf{a}} = A^{-1}\mathbf{v}_{\mathbf{e}} = \begin{bmatrix} 4/10 \\ 3/20 \end{bmatrix}$$

Change of Basis

Application: coordinate system axes rotation

Given: \mathbf{u}_a

Find: the matrix for a 90° in the $[\mathbf{a}_1, \mathbf{a}_2]$ -system

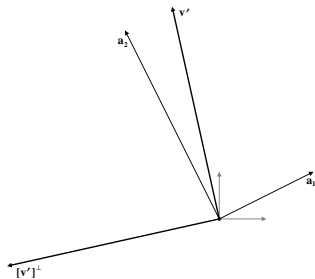
Solution:

- 1 construct change of basis matrix A : map \mathbf{u}_a into the $[\mathbf{e}_1, \mathbf{e}_2]$ -system
- 2 apply the “usual” rotation R
- 3 transform back to the $[\mathbf{a}_1, \mathbf{a}_2]$ -system with A^{-1}

$$\mathbf{u}'_a = A^{-1}R\mathbf{A}\mathbf{u}_a$$

Change of Basis

Application: coordinate system axes rotation



90° rotation of \mathbf{v}'_a into $[\mathbf{v}'_a]^\perp$

Example:

Rotate \mathbf{v}'_a by 90° in the $[\mathbf{a}_1, \mathbf{a}_2]$ -system

$$R' = A^{-1}RA = \begin{bmatrix} 0 & -2 \\ 1/2 & 0 \end{bmatrix}$$

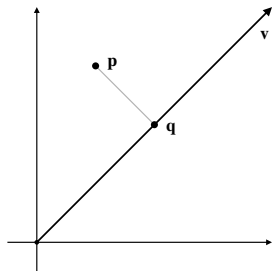
$$[\mathbf{v}'_a]^\perp = R'\mathbf{v}'_a = \begin{bmatrix} -2 \\ 1/4 \end{bmatrix}$$

R and R' describe the same linear map, but with respect to different bases

They are **similar matrices**

Change of Basis

Application: projecting a point onto a line



Project point \mathbf{p} onto line defined by \mathbf{v}
resulting in \mathbf{q} closest to \mathbf{p}

Let \mathbf{v} form angle θ with \mathbf{e}_1

$$M = R_\theta P R_{-\theta}$$

where

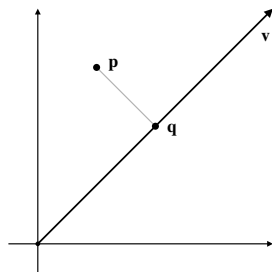
$$R = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

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

P and M are similar matrices

Change of Basis

Application: projecting a point onto a line



\mathbf{v} is not drawn normalized

Example:

$$\theta = 45^\circ \quad \mathbf{v} = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} \quad \mathbf{p} = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$$

$$M = \begin{bmatrix} 0.5 & 0.5 \\ 0.5 & 0.5 \end{bmatrix}$$

Projection of \mathbf{p} onto \mathbf{v} :

$$\mathbf{q} = M\mathbf{p} = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$$

Application: Intersecting Lines

Two interpretations of of a linear system

- 1 “column view”: coordinate system or linear combination approach
- 2 “row view”: focus on the row equations

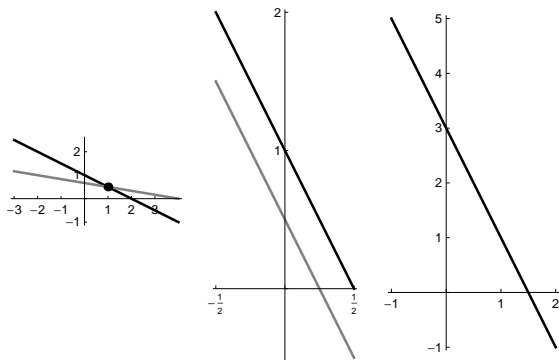
Line intersection problems provide examples of both:

- parametric/parametric line \Rightarrow column view
- implicit/implicit line \Rightarrow row view

Choose the view that best suits given information

Application: Intersecting Lines

Linear systems from this chapter interpreted as line intersection



Left to right: unique solution, inconsistent, underdetermined

- linear system
- solution spaces
- consistent linear system
- Cramer's rule
- upper triangular
- Gauss elimination
- forward elimination
- back substitution
- linear combination
- inverse matrix
- orthogonal matrix
- orthonormal
- rigid body motion
- inconsistent system of equations
- underdetermined system of equations
- homogeneous system
- kernel
- null space
- row pivoting
- column pivoting
- complete pivoting
- change of basis
- column and row views of linear systems