

July 26 (Lecture 19)

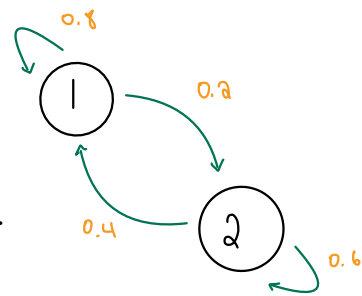
Overview: First, we'll look a bit more at diagonalization and what it can do for us. Then we'll backtrack a little bit to coordinates and start seeing how we can tie together coordinates, general linear transformations, and diagonalization!

Learning Goals:

- Correctly diagonalize matrices!
- Use diagonalization to solve certain linear algebra problems.
- Correctly define the matrix of a linear transformation from one basis to another.

As you're getting settled:

- Homework 10 is due Friday, at 11:30 pm!
- Please fill out the CES! I appreciate your feedback. I'll give you 10 minutes at the start of class on Thursday to complete it, if that makes it easier for you.
- You should have your Test 2 marks back by the end of today. Please have a look at the grading and let me know before August 2nd if you have any concerns.



Example. Markov Chains (not on the final).
Think: changing between two brand loyalties,
or passers-by flicking a light switch on and off.

Let $P = \begin{matrix} & \begin{matrix} 1 & 2 \end{matrix} \\ \begin{matrix} 1 \\ 2 \end{matrix} & \begin{bmatrix} 0.8 & 0.4 \\ 0.2 & 0.6 \end{bmatrix} \end{matrix}$ "transition probability matrix". Note: column sums = 1.

If $\vec{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$ w/ $x_i \geq 0$ and $x_1 + x_2 = 1$, then \vec{x} is a "probability vector". (Theorem: $P\vec{x}$ is also a "probability vector"!)

Let's diagonalize P !

1) $\det(P - \lambda I_2) = \begin{vmatrix} 0.8 - \lambda & 0.4 \\ 0.2 & 0.6 - \lambda \end{vmatrix} = \lambda^2 - 1.4\lambda + 0.44 - 0.08 = (\lambda - 1)(\lambda - 0.4) = 0$
 $\Rightarrow \lambda = 1, 0.4$ (P is diagonalizable)!

2) $\lambda = 1: \begin{bmatrix} -0.2 & 0.4 \\ 0.2 & -0.4 \end{bmatrix} \xrightarrow{\text{row ops}} \begin{bmatrix} 1 & -2 \\ 0 & 0 \end{bmatrix} \Rightarrow E_P(1) = \text{span} \left\{ \begin{bmatrix} 2 \\ 1 \end{bmatrix} \right\}$
 $= \text{span} \left\{ \begin{bmatrix} 2/3 \\ 1/3 \end{bmatrix} \right\}$.

$\begin{bmatrix} 2/3 \\ 1/3 \end{bmatrix}$ is invariant: $P \begin{bmatrix} 2/3 \\ 1/3 \end{bmatrix} = 1 \begin{bmatrix} 2/3 \\ 1/3 \end{bmatrix}$. "steady state" for the Markov Chains.

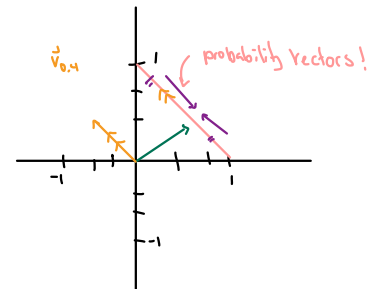
$\lambda = 0.4: \begin{bmatrix} 0.4 & 0.4 \\ 0.2 & 0.2 \end{bmatrix} \xrightarrow{\text{row ops}} \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} \Rightarrow E_P(0.4) = \text{span} \left\{ \begin{bmatrix} 1/2 \\ -1/2 \end{bmatrix} \right\}$.

All probability vectors look like $\vec{x} = \vec{v}_1 + c\vec{v}_{0.4}$

$\Rightarrow P\vec{x} = P\vec{v}_1 + cP\vec{v}_{0.4} = \vec{v}_1 + c(0.4)\vec{v}_{0.4}$

\vdots

$\Rightarrow P^k \vec{x} = \vec{v}_1 + \underbrace{c(0.4)^k}_{\rightarrow 0} \vec{v}_{0.4} \xrightarrow{k \rightarrow \infty} \vec{v}_1$.



Sect. 4.4, 6.6

Matrices of Linear Transformations

Let's return to general linear transformations. We saw that linear maps $L : \mathbb{R}^n \rightarrow \mathbb{R}^m$ were determined by what they did to the standard basis. Is that true in general? Can we extend the "standard matrix" idea to general linear maps?

[No p. ref !!] **Theorem.** Let V and W be vector spaces, let T and S be linear transformations from V to W , and let $B = \{\vec{v}_1, \dots, \vec{v}_n\}$ be a basis for V . If $T(\vec{v}_i) = S(\vec{v}_i)$ for all i , then $T=S$.

Proof. $T=S$ when $T(\vec{v}) = S(\vec{v})$ for all $\vec{v} \in V$. So, if $\vec{v} = \alpha_1 \vec{v}_1 + \dots + \alpha_n \vec{v}_n$, then we have:
 $T(\vec{v}) \stackrel{\text{by linearity}}{=} \alpha_1 T(\vec{v}_1) + \dots + \alpha_n T(\vec{v}_n)$
 $= \alpha_1 S(\vec{v}_1) + \dots + \alpha_n S(\vec{v}_n) \stackrel{\text{by linearity}}{=} S(\vec{v}).$



Now, suppose that $L : V \rightarrow W$ is a linear map, $B = \{\vec{v}_1, \dots, \vec{v}_n\}$ be a basis for V , and $C = \{\vec{w}_1, \dots, \vec{w}_m\}$ be a basis for W . What are the C -coordinates of $L(\vec{v})$? (First, a helping theorem.)

p. 266 **Theorem (4.4.1).** Let V be a vector space with basis $B = \{\vec{v}_1, \dots, \vec{v}_n\}$. Then for $\vec{x}, \vec{y} \in V$, $s, t \in \mathbb{R}$, we have $[s\vec{x} + t\vec{y}]_B = s[\vec{x}]_B + t[\vec{y}]_B$.
(i.e. $[\cdot]_B$ is a linear map from V to \mathbb{R}^n !)

Let's try to compute $[L(\vec{v})]_C$: if $[\vec{v}]_B = \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{bmatrix}$, then

$$[L(\vec{v})]_C = [\alpha_1 L(\vec{v}_1) + \dots + \alpha_n L(\vec{v}_n)]_C \stackrel{(4.4.1)}{=} \alpha_1 [L(\vec{v}_1)]_C + \dots + \alpha_n [L(\vec{v}_n)]_C$$

column vectors \rightarrow

$$= \begin{bmatrix} [L(\vec{v}_1)]_C & \cdots & [L(\vec{v}_n)]_C \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{bmatrix}$$

$m \times n$ $[\vec{v}]_B$

p. 288
(but for us it's more general!).

Definition. Let V and W be vector spaces, let $B = \{\vec{v}_1, \dots, \vec{v}_n\}$ and $C = \{\vec{w}_1, \dots, \vec{w}_m\}$ be bases for V and W respectively, and let $L : V \rightarrow W$ be linear. The *matrix of L from B to C* is the $m \times n$ matrix $[L]_{C \leftarrow B}$ defined by

$$[L]_{C \leftarrow B} = \begin{bmatrix} [L(\vec{v}_1)]_C & \cdots & [L(\vec{v}_n)]_C \end{bmatrix}.$$

Example. We've already looked at $V = \mathbb{R}^n$, $W = \mathbb{R}^m$, $B = S_n$, $C = S_m$. Standard matrix: $[L]_{S_m \leftarrow S_n}$!

$$[L]_{S_m \leftarrow S_n} = [L(\vec{e}_1) \dots L(\vec{e}_n)]_{S_m}, \text{ remembering that } L(\vec{e}_i) = [L(\vec{e}_i)]_{S_m} \text{ since } S_m \text{ is the standard basis.}$$

$L : P_2(\mathbb{R}) \rightarrow P_3(\mathbb{R})$, $B = \{1-x, 1+x, x^2\}$, $C = \{1, x, x^2, x^3\}$.

$L(a+bx+cx^2) = (a-c) + (b+2c)x + ax^2 + (3a-b+c)x^3$. (L is linear!)

Compute L of the vectors in B , & find C -coords:

$$[L(1-x)]_C = [1-x+x^2+4x^3]_C = \begin{bmatrix} -1 \\ 1 \\ 1 \\ 4 \end{bmatrix}.$$

$$[L(1+x)]_C = [1+x+x^2+2x^3]_C = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 2 \end{bmatrix}.$$

$$[L(x^2)]_C = [-1+2x+x^3]_C = \begin{bmatrix} -1 \\ 2 \\ 1 \\ 1 \end{bmatrix}.$$

If $[p(x)]_B = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$, then $[L(p(x))]_C = \begin{bmatrix} -1 & 1 & -1 \\ 1 & 1 & 2 \\ 1 & 1 & 1 \\ 4 & 2 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \\ 2 \\ 6 \end{bmatrix} \rightarrow L(p(x)) = 2+2x^2+6x^3. \checkmark$
($p(x)=2$).

From our previous computation, $[L(\vec{v})]_C = [L]_{C \leftarrow B} [\vec{v}]_B$.

What happens if we apply this definition to $L = \text{id} : V \rightarrow V$?

p. 267

Definition. Let V be a vector space with bases $B = \{\vec{v}_1, \dots, \vec{v}_n\}$ and C . Then the *change of coordinates* (or *change of basis*) matrix from B to C is $P_{C \leftarrow B} = [id]_{C \leftarrow B} = \left[\begin{array}{ccc} [\vec{v}_1]_C & | & - \\ \vdots & & \vdots \\ [\vec{v}_n]_C & | & - \end{array} \right]$.

Example. $B = \left\{ \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}, \begin{bmatrix} -3 \\ 2 \\ -3 \end{bmatrix} \right\}$, $C = \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \right\}$.

Let's find $P_{C \leftarrow B}$: SLEs: $c_1 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + c_3 \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}$, and $\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$, and $\begin{bmatrix} -3 \\ 2 \\ -3 \end{bmatrix}$.

Augment:

$$\left[\begin{array}{ccc|ccc} 1 & 1 & 1 & 1 & 1 & -3 \\ 0 & 1 & 1 & 0 & -1 & 2 \\ 0 & 0 & 1 & 2 & 0 & -3 \end{array} \right] \xrightarrow{\text{row ops}} \left[\begin{array}{ccc|ccc} 1 & 0 & 0 & 1 & 2 & -5 \\ 0 & 1 & 0 & -2 & -1 & 5 \\ 0 & 0 & 1 & 2 & 0 & -3 \end{array} \right] \Rightarrow P_{C \leftarrow B} = \begin{bmatrix} 1 & 2 & -5 \\ -2 & -1 & 5 \\ 2 & 0 & -3 \end{bmatrix}$$

$[\vec{v}_1]_C \quad [\vec{v}_2]_C \quad [\vec{v}_3]_C$

Example. $B = \{1-x, 1+x, x^2\}$, $C = \{1-x^2, x-1, 1+x+x^2\}$ bases for $P_2(\mathbb{R})$.

To compute $P_{C \leftarrow B}$: $c_1(1-x^2) + c_2(x-1) + c_3(1+x+x^2)$
 $= (c_1 - c_2 + c_3) + (c_2 + c_3)x + (c_3 - c_1)x^2 = 1-x$, and $1+x$, and x^2 !

$$\left[\begin{array}{ccc|ccc} 1 & -1 & 1 & 1 & 1 & 0 \\ 0 & 1 & 1 & -1 & 1 & 0 \\ -1 & 0 & 1 & 0 & 0 & 1 \end{array} \right] \xrightarrow{\text{row ops}} \left[\begin{array}{ccc|ccc} 1 & 0 & 0 & 0 & 2/3 & -2/3 \\ 0 & 1 & 0 & -1 & 1/3 & -1/3 \\ 0 & 0 & 1 & 0 & -2/3 & 1/3 \end{array} \right]$$

How do we convert from C to B , if we know how to convert from B to C ? Well... maybe $P_{C \leftarrow B}$ is invertible?