

# MATH 200C: Linear Algebra



Class 36: Monday, May 11, 2026



- ▶ Notes on Assignment 32
- ▶ Course Response Forms

**Project 2**  
**Age – Class Population Models: DUE: Today**

**Course Response Forms**  
**In Class Today**

**Additional Office Hours This Week**  
**Wednesday, 9 – 11 AM**

**Final Exam**  
**Thursday, May 14**  
**9 AM – Noon**  
**Warner 100**

**Theorem 11:** If  $P$  is an  $n \times n$  regular stochastic matrix, then  $P$  has a unique steady-state vector  $\mathbf{q}$ .

Further, if  $\mathbf{x}_0$  is any initial state and  $\mathbf{x}_{k+1} = P\mathbf{x}_k$  for  $k = 0, 1, 2, \dots$ , then the Markov chain  $\{\mathbf{x}_k\}$  converges to  $\mathbf{q}$  as  $k \rightarrow \infty$

**Theorem:** If  $P$  is a regular  $n \times n$  transition matrix with  $n \geq 2$ , then the following are all true:

- ▶ There is a stochastic matrix  $\Pi = \lim_{m \rightarrow \infty} P^m$
- ▶ Each column of  $\Pi$  is the same probability vector  $\mathbf{q}$ .
- ▶ For any initial probability vector  $\mathbf{x}_0$ , we have  $\lim_{m \rightarrow \infty} P^m \mathbf{x}_0 = \mathbf{q}$ .
- ▶ The vector  $\mathbf{q}$  is the unique probability vector that is an eigenvector of  $P$  associated with the eigenvalue 1.
- ▶ All other eigenvalues  $\lambda$  of  $P$  have  $|\lambda| < 1$ .

## What Have We Proved So Far?

If  $P$  is a regular  $n \times n$  stochastic matrix, then

- ▶ There will always be an eigenvalue  $\lambda_1$  with value 1
- ▶ If  $P$  has  $n$  distinct eigenvalues  $\lambda_1 = 1, \lambda_2, \dots, \lambda_n$  with corresponding eigenvectors  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$ , then the set of eigenvectors is linearly independent and spans  $\mathbb{R}^n$
- ▶ For each possible starting vector  $\mathbf{x}_0$ , there are constants  $c_1, c_2, \dots, c_n$  such that  $c_1\mathbf{x}_1 + c_2\mathbf{x}_2 + c_3\mathbf{x}_3 + \dots + c_n\mathbf{x}_n$
- ▶  $\mathbf{x}_{k+1} = P\mathbf{x}_k =$   
 $c_1(\lambda_1)^k\mathbf{x}_1 + c_2(\lambda_2)^k\mathbf{x}_2 + c_3(\lambda_3)^k\mathbf{x}_3 + \dots + c_n(\lambda_n)^k\mathbf{x}_n$   
 $= c_1\mathbf{x}_1 + c_2(\lambda_2)^k\mathbf{x}_2 + c_3(\lambda_3)^k\mathbf{x}_3 + \dots + c_n(\lambda_n)^k\mathbf{x}_n$
- ▶ If  $\lambda \neq 1$  is an eigenvalue of  $P$  with corresponding eigenvector  $\mathbf{x}$ , then the sum of the coordinates of that eigenvector will be 0,
- ▶ If  $\lambda \neq 1$  is an eigenvalue of  $P$ , then  $|\lambda| < 1$
- ▶ **The coordinates of any eigenvalue associated with  $\lambda = 1$  all have the same sign**

## Our Goal Today

- ▶ The space of eigenvectors associated with eigenvalue 1 for a stochastic matrix has dimension 1.

**Theorem C:** If  $\{\mathbf{v}, \mathbf{w}\}$  is a linearly independent set of vectors in  $\mathbb{R}^n$ , then we can create a vector  $\mathbf{x} = s\mathbf{v} + t\mathbf{w}$  (a linear combination of  $\mathbf{v}$  and  $\mathbf{w}$  where  $s$  and  $t$  are not both 0 so that  $\mathbf{x}$  has both positive and negative components).

**Proof:** Let  $d$  be the sum of the components of  $\mathbf{v}$ .  
We need to consider two cases:

**Case 1:** If  $d = 0$ , then  $\mathbf{v}$  must have some positive and some negative components. We can let  $s = 1$  and  $t = 0$ .

**Case 2:** If  $d \neq 0$ , set  $s = \frac{\sum_i w_i}{d}$  and  $t = 1$ . Now neither  $s$  nor  $t$  is 0 so  $\mathbf{x} = s\mathbf{v} + t\mathbf{w}$  is not the zero vector as  $\{\mathbf{v}, \mathbf{w}\}$  is linearly independent. But the sum of the components of  $\mathbf{x}$  is 0 and thus it must contain both positive and negative components.

**Theorem D :** The space of eigenvectors associated with eigenvalue 1 for a stochastic matrix has dimension 1.

**Proof:** Suppose, to the contrary, that this space contains a pair  $\{\mathbf{v}, \mathbf{w}\}$  of linearly independent vectors.

Use Theorem C to create a vector  $\mathbf{x} = s\mathbf{v} + t\mathbf{w}$  which has both positive and negative components.

But  $\mathbf{x} = s\mathbf{v} + t\mathbf{w}$  is a linear combination of two eigenvectors so it is also an eigenvector associated with eigenvalue 1

$$( P\mathbf{x} = P(s\mathbf{v} + t\mathbf{w}) = sP\mathbf{v} + tP\mathbf{w} = s\mathbf{v} + t\mathbf{w} = \mathbf{x} )$$

This contradicts Theorem B so we can not have such a pair of linearly independent eigenvectors.

Let's Do Course Response Forms

<https://go.middlebury.edu/crf>

# The Final Exam

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Date/Time	Thursday, 9 AM to Noon
Location	Warner 100
Calculators/Computers/Text Book?	No
"Cheat Sheet? "	Yes
Format	Answer 4 of 5 Questions

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