# Math 415 - Lecture 32

Complex numbers and eigenvectors

## Wednesday November 11th 2015

Textbook reading: first part of Chapter 5.5

Suggested practice exercises: 5.5 1, 2, 3

Khan Academy video: Complex Numbers (part 1)

Strang lecture: Lecture 21: Eigenvalues and eigenvectors

### SOME DATES.

- \* Friday November 13th: No class.
- \* Next week there will be discussion sections, but no quiz. Prepare for the midterm!
- \* Thursday November 19th, 7-8:15PM: Midterm 3.
- \* Friday November 20th: No class.
- \* November 23-27th Thanksgiving break: no class.
- \* Wednesday December 9th: last day of class
- \* Thursday December 17th: Final Exam.

# 1 Review

# 1.1 Properties of eigenvectors and eigenvalues

- If  $A\mathbf{x} = \lambda \mathbf{x}$  then  $\mathbf{x}$  is an **eigenvector** of A with **eigenvalue**  $\lambda$ .
- $\lambda$  is an eigenvalue of  $A \iff \det(A \lambda I) = 0$ .

**Definition 1.** An  $n \times n$  matrix A is a **Markov matrix** if has non negative entries, and the entries in each column add to 1.

Theorem 1. Let A be a Markov matrix. Then

- (i) 1 is an eigenvalue of A and any other eigenvalue  $\lambda$  satisfies  $|\lambda| \leq 1$ .
- (ii) If A has only positive entries, then any other eigenvalue satisfies  $|\lambda| < 1$ .

**Theorem 2.** Let A be an  $n \times n$ -Markov matrix with only positive entries and let  $\mathbf{v} \in \mathbb{R}^n$ . Then

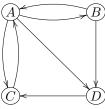
$$\boldsymbol{v}_{\infty} := \lim_{k \to \infty} A^k \boldsymbol{v} \text{ exists},$$

and  $Av_{\infty} = v_{\infty}$ . In this case  $v_{\infty}$  is often called the **steady state**.

# 2 Page rank (or: the 25000000000 \$ eigenvector)

Google's success is based on an algorithm to rank webpages, the **Page rank**, named after Google founder Larry Page. The idea is to determine how likely it is that a web user randomly gets to a given webpage. The webpages are ranked by these probabilities.

Suppose the internet consisted of the only four webpages A, B, C, D linked as in the following graph.



Imagine a surfer following these links at random. For the probability  $PR_n(A)$  that she is at A (after n steps), we need to think about how she could have reached A. We add:

- the probability that she was at B (at exactly one step before), and left for A,(that's  $PR_{n-1}(B) \cdot \frac{1}{2}$ )
- the probability that she was at C, and left for A,
- the probability that she was at D, and left for A.

Hence: 
$$PR_n(A) = PR_{n-1}(B) \cdot \frac{1}{2} + PR_{n-1}(C) \cdot \frac{1}{1} + PR_{n-1}(D) \cdot \frac{0}{1}$$
.  
Encode the probabilties at step  $n$  in a state vector with four entries. 
$$\begin{bmatrix} PR_n(A) \\ PR_n(B) \\ PR_n(C) \\ PR_n(D) \end{bmatrix} = \frac{1}{2} \cdot \frac{1}{1} \cdot \frac{1}{1$$

$$\begin{bmatrix} 0 & \frac{1}{2} & 1 & 0 \\ \frac{1}{3} & 0 & 0 & 0 \\ \frac{1}{3} & 0 & 0 & 1 \\ \frac{1}{3} & \frac{1}{2} & 0 & 0 \end{bmatrix} \begin{bmatrix} PR_{n-1}(A) \\ PR_{n-1}(B) \\ PR_{n-1}(C) \\ PR_{n-1}(D) \end{bmatrix}$$

**Definition 2.** The **PageRank vector** is the long-term equilibrium. It is an eigenvector of the Markov matrix with eigenvalue 1.

Let's call the Markov matrix with the probabilities T:

$$\bullet \ T - 1I = \begin{bmatrix} -1 & \frac{1}{2} & 1 & 0 \\ \frac{1}{3} & -1 & 0 & 0 \\ \frac{1}{3} & 0 & -1 & 1 \\ \frac{1}{3} & \frac{1}{2} & 0 & -1 \end{bmatrix} \xrightarrow{\text{RREF}} \begin{bmatrix} 1 & 0 & 0 & -2 \\ 0 & 1 & 0 & -\frac{2}{3} \\ 0 & 0 & 1 & -\frac{5}{3} \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$\implies$$
 eigenspace of  $\lambda = 1$  is spanned by  $\begin{bmatrix} 2 \\ \frac{2}{3} \\ \frac{3}{3} \end{bmatrix}$ .

• Now we need to make the entries add up to 1.

$$\begin{bmatrix} PR(A) \\ PR(B) \\ PR(C) \\ PR(D) \end{bmatrix} = \frac{3}{16} \begin{bmatrix} \frac{2}{2} \\ \frac{5}{3} \\ \frac{1}{1} \end{bmatrix} = \begin{bmatrix} 0.375 \\ 0.125 \\ 0.313 \\ 0.188 \end{bmatrix}.$$

This is the PageRank vector.

• The corresponding ranking of the webpages is A, C, D, B.

**Remark.** In practical situations the system might be too large for finding the eigenvalues by row operations.

- Google reports having met 60 trillion webpages. Google's search index is over 100,000,000 gigabytes. Number of Google's servers is secret: about 2,500,000 More than 1,000,000,000 websites (i.e. hostnames; about 75% not active)
- Thus we have a gigantic but very sparse matrix.

An alternative to row operations is the **power method** (see Theorem 2):

#### Power method

If T is an (acyclic and irreducible) Markov matrix, then for any  $\mathbf{v}_0$  the vectors  $T^n\mathbf{v}_0$  converge to an eigenvector with eigenvalue 1

Here:

$$T = \begin{bmatrix} 0 & \frac{1}{2} & 1 & 0 \\ \frac{1}{3} & 0 & 0 & 0 \\ \frac{1}{3} & 0 & 0 & 1 \\ \frac{1}{3} & \frac{1}{2} & 0 & 0 \end{bmatrix}.$$

Start with an arbitrary state vector, hit it with powers of T.

$$\begin{pmatrix}
\begin{bmatrix} PR(A) \\ PR(B) \\ PR(C) \\ PR(D) \end{bmatrix} = \begin{bmatrix} 0.375 \\ 0.125 \\ 0.313 \\ 0.188 \end{bmatrix}, T \begin{bmatrix} 1/4 \\ 1/4 \\ 1/4 \\ 1/4 \end{bmatrix} = \begin{bmatrix} 0.375 \\ 0.083 \\ 0.333 \\ 0.208 \end{bmatrix}$$

. Note that the ranking of the webpages is already A, C, D, B if we stop here.

$$T\begin{bmatrix} 1/4\\1/4\\1/4\\1/4 \end{bmatrix} = \begin{bmatrix} 0.375\\0.083\\0.333\\0.208 \end{bmatrix}, \quad T^2\begin{bmatrix} 1/4\\1/4\\1/4\\1/4 \end{bmatrix} = \begin{bmatrix} 0.375\\0.125\\0.333\\0.167 \end{bmatrix}, \quad T^3\begin{bmatrix} 1/4\\1/4\\1/4\\1/4 \end{bmatrix} = \begin{bmatrix} 0.396\\0.125\\0.292\\0.188 \end{bmatrix}$$

**Remark.** • If all entries of T are positive (no zero entries!), then the power method is guaranteed to work.

 $\bullet$  In the context of PageRank, we can make sure that this is the case by replacing T with

$$(1-p) \cdot \begin{bmatrix} 0 & \frac{1}{2} & 1 & 0 \\ \frac{1}{3} & 0 & 0 & 0 \\ \frac{1}{3} & 0 & 0 & 1 \\ \frac{1}{3} & \frac{1}{2} & 0 & 0 \end{bmatrix} + p \cdot \begin{bmatrix} \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \end{bmatrix}.$$

Just to make sure: still a Markov matrix, now with positive entries Google used to use p=0.15.

# 3 Eigenbasis?

# 3.1 Number of (independent) eigenvectors

An  $n \times n$  matrix A has up to n different eigenvalues. Namely, the roots of degree n characteristic polynomial  $\det(A - \lambda I)$ .

- For each eigenvalue  $\lambda$ , A has at least one eigenvector. That is because  $Nul(A \lambda I)$  has dimension at least one.
- If  $\lambda$  has multiplicity m, then A has up to m (independent) eigenvectors for  $\lambda$ .

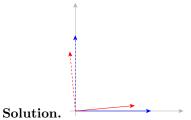
Ideally, we would like to find a total of n (independent) eigenvectors for A. This would give an **EIGENBASIS**. Why can there be no more than n independent eigenvectors?!

Two sources of trouble: eigenvalues can be

- complex numbers (that is, not enough real roots), or
- repeated roots of the characteristic polynomial.

#### 3.2 Trouble I: complex eigenvalues

Example 3. Find the eigenvectors and eigenvalues of  $A = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$ . Geometrically, what is the trouble?



$$A \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} -y \\ x \end{bmatrix}$$

 $A\begin{bmatrix}x\\y\end{bmatrix} = \begin{bmatrix}-y\\x\end{bmatrix}$  i.e. multiplication by A is a rotation by  $90^\circ$  (counter-

Which vector is parallel after rotation by 90°? Trouble.

# Complex numbers review

**Definition.**  $\mathbb{C} = \{x + iy \mid x, y \in \mathbb{R}\}\$ 

- $i = \sqrt{-1}$ , or  $i^2 = -1$ .
- Any point in  $\mathbb{R}^2$  can be viewed as a complex number:  $\binom{x}{y} \leftrightarrow x + iy$

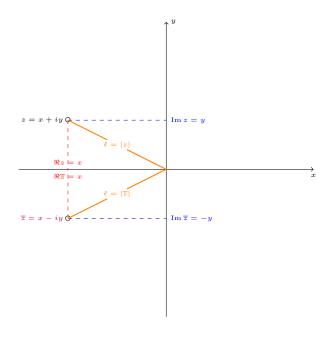
**Definition.** Let z = x + iy be a complex number

**Real part** The real part of z, denoted  $\Re(z)$  is defined by  $\Re(z) = x$ .

**Imaginary part** The imaginary part of z, denoted Im(z) is defined by Im(z) =

Complex conjugate The complex conjugate of z, denoted  $\overline{z}$ , is defined by  $\overline{z} = x - iy$ .

**Absolute value** The absolute value, or magnitude of z, denoted |z| or ||z||, is given by  $|z| = \sqrt{x^2 + y^2}$ .



## Adding complex numbers

**Definition.** Given z = x + iy, w = u + iv, we define

$$z + w = (x + u) + i(y + v)$$

**Remark.** This corresponds exactly to addition of vectors in  $\mathbb{R}^2$ .

### Multiplying complex numbers

**Definition.** Given z = x + iy, w = u + iv, we define

$$zw = (x + iy)(u + iv)$$

$$= xu + x(iv) + (iy)u + (iy)(iv)$$

$$= (xu - yv) + i(xv + yu)$$

### Absolute value and complex conjugate

Remark.  $\bullet \ \overline{\overline{z}} = z$ 

- $\bullet |z|^2 = z\overline{z}$
- $|z| = |\overline{z}|$

Proof.

$$z\overline{z} = (x+iy)(x-iy)$$

$$= x^2 - x(iy) + (iy)x - (iy)(iy)$$

$$= x^2 + y^2$$

# 4.1 Complex Linear Algebra

Until now we took as our scalars the real numbers. In particular we used the vector space  $\mathbb{R}^n$  of column vectors

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}.$$

If c is a real number (a scalar) we defined

$$c\mathbf{x} = \begin{bmatrix} cx_1 \\ cx_2 \\ \vdots \\ cx_n \end{bmatrix}.$$

Now we want to use **COMPLEX** scalars. We need a new context to make sense of this.

**Definition.**  $\mathbb{C}^n$  is the (complex) vector space of *complex* column vectors  $\mathbf{z} = \mathbf{z}$ 

$$\begin{bmatrix} z_1 \\ z_2 \\ \vdots \\ z_n \end{bmatrix}$$
, where  $z_1, z_2, \dots z_n$  are complex numbers.

- Now multiplication by a complex scalar makes sense.
- We can define subspaces, Span, independence, basis, dimension for  $\mathbb{C}^n$  in the usual way.
- We can multiply complex vectors by complex matrices. Column space and Null space still make sense.
- The only difference is the dot product, you need to use the complex conjugate to get a good notion of length. (Later more.)

# 5 Back to eigenvectors

Example 4. Find the eigenvectors and eigenvalues of  $A = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$ . Now, we can use complex numbers!

**Solution** (continued). • 
$$\det(A - \lambda I) = \begin{vmatrix} -\lambda & -1 \\ 1 & -\lambda \end{vmatrix} = \lambda^2 + 1$$
 So the eigenvalues are  $\lambda_1 = i$  and  $\lambda_2 = -i$ .

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• 
$$\lambda_1 = i$$
:  $\begin{bmatrix} -i & -1 \\ 1 & -i \end{bmatrix} \mathbf{x} = \mathbf{0} \implies \mathbf{x}_1 = \begin{bmatrix} i \\ 1 \end{bmatrix}$  Let us check  $\begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} i \\ 1 \end{bmatrix} = \begin{bmatrix} -1 \\ i \end{bmatrix} = i \begin{bmatrix} i \\ 1 \end{bmatrix}$ 

• 
$$\lambda_2 = -i$$
:  $\begin{bmatrix} i & -1 \\ 1 & i \end{bmatrix}$   $\mathbf{x} = \mathbf{0} \implies \mathbf{x}_2 = \begin{bmatrix} -i \\ 1 \end{bmatrix}$ 

Summary: We had  $A = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$ .

- $\bullet$  Eigenvalues: i, -i These are conjugates!
- Eigenvectors:  $\mathbf{x}_1 = \begin{bmatrix} i \\ 1 \end{bmatrix}$  and  $\mathbf{x}_2 = \begin{bmatrix} -i \\ 1 \end{bmatrix}$  These are also conjugates!

**Theorem 3.** If A is a matrix with real entries and  $\lambda$  is a **complex eigenvalue**, then  $\bar{\lambda}$  is also a complex eigenvalue. Furthermore, if  $\mathbf{x}$  is an eigenvector with eigenvalue  $\lambda$ , then  $\bar{\mathbf{x}}$  is an eigenvector with eigenvalue  $\bar{\lambda}$ .

**Remark.** Note that we are using vectors in  $\mathbb{C}^2$ , instead of vectors in  $\mathbb{R}^2$ . Works pretty much the same!

# 5.1 Trouble II: generalized eigenvectors

Example 5. Find the eigenvectors and eigenvalues of  $A = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$ . What is the trouble?

**Solution.** •  $\det(A - \lambda I) = \begin{vmatrix} 1 - \lambda & 1 \\ 0 & 1 - \lambda \end{vmatrix} = (1 - \lambda)^2$  So:  $\lambda = 1$  is the only eigenvalue (it has multiplicity 2).

- $\lambda = 1 : \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \mathbf{x} = \mathbf{0} \implies \mathbf{x}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$  So the eigenspace is Span  $\left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\}$ . Only dimension 1!
- Trouble: We can not find an **Eigenbasis** for this matrix. This kind of problem cannot really be fixed. We have to lower our expectations and look for generalized eigenvectors. These are solutions to  $(A \lambda I)^2 \mathbf{x} = \mathbf{0}, (A \lambda I)^3 \mathbf{x} = \mathbf{0}, \dots$

# 6 Practice problems

Example 6. Find the eigenvectors and eigenvalues of  $A = \begin{bmatrix} 1 & 2 & 1 \\ 0 & -5 & 0 \\ 1 & 8 & 1 \end{bmatrix}$ .