Announcements Wednesday, November 01

- ▶ WeBWorK 3.1, 3.2 are due today at 11:59pm.
- ▶ The quiz on Friday covers §§3.1, 3.2.
- My office is Skiles 244. Rabinoffice hours are Monday, 1–3pm and Tuesday, 9–11am.

Section 5.2

The Characteristic Equation

We have a couple of new ways of saying "A is invertible" now:

The Invertible Matrix Theorem

Let A be a square $n \times n$ matrix, and let $T \colon \mathbf{R}^n \to \mathbf{R}^n$ be the linear transformation T(x) = Ax. The following statements are equivalent.

- 1. A is invertible.
 - 2. T is invertible.
 - 3. A is row equivalent to I_n .
 - 4. A has n pivots.
 - 5. Ax = 0 has only the trivial solution.
 - 6. The columns of A are linearly independent.
 - 7. T is one-to-one.
 - 8. Ax = b is consistent for all b in \mathbb{R}^n .
 - 9. The columns of A span \mathbb{R}^n .
 - T is onto.

- 11. A has a left inverse (there exists B such that $BA = I_n$).
- 12. A has a right inverse (there exists B such that $AB = I_D$).
- 13. A^T is invertible.
- 14 The columns of A form a basis for \mathbb{R}^n
- 15. Col $A = \mathbb{R}^n$.
- 16. $\dim \operatorname{Col} A = n$.
- 17. rank A = n.
- 18. Nul $A = \{0\}$.
- 19. $\dim \text{Nul } A = 0$.
- 19. The determinant of A is *not* equal to zero.
- 20. The number 0 is *not* an eigenvalue of *A*.

The Characteristic Polynomial

Let A be a square matrix.

$$\lambda$$
 is an eigenvalue of $A \iff Ax = \lambda x$ has a nontrivial solution $\iff (A - \lambda I)x = 0$ has a nontrivial solution $\iff A - \lambda I$ is not invertible $\iff \det(A - \lambda I) = 0$.

This gives us a way to compute the eigenvalues of A.

Definition

Let A be a square matrix. The characteristic polynomial of A is

$$f(\lambda) = \det(A - \lambda I).$$

The characteristic equation of A is the equation

$$f(\lambda) = \det(A - \lambda I) = 0.$$

Important

The eigenvalues of A are the roots of the characteristic polynomial $f(\lambda) = \det(A - \lambda I)$.

Question: What are the eigenvalues of

$$A = \begin{pmatrix} 5 & 2 \\ 2 & 1 \end{pmatrix}$$
?

Answer: First we find the characteristic polynomial:

$$f(\lambda) = \det(A - \lambda I) = \det\begin{bmatrix} 5 & 2 \\ 2 & 1 \end{bmatrix} - \begin{pmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} = \det\begin{pmatrix} 5 - \lambda & 2 \\ 2 & 1 - \lambda \end{pmatrix}$$
$$= (5 - \lambda)(1 - \lambda) - 2 \cdot 2$$
$$= \lambda^2 - 6\lambda + 1.$$

The eigenvalues are the roots of the characteristic polynomial, which we can find using the quadratic formula:

$$\lambda = \frac{6 \pm \sqrt{36 - 4}}{2} = 3 \pm 2\sqrt{2}.$$

The Characteristic Polynomial Example

Question: What is the characteristic polynomial of

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}?$$

Answer:

$$f(\lambda) = \det(A - \lambda I) = \det\begin{pmatrix} a - \lambda & b \\ c & d - \lambda \end{pmatrix} = (a - \lambda)(d - \lambda) - bc$$
$$= \lambda^2 - (a + d)\lambda + (ad - bc)$$

What do you notice about $f(\lambda)$?

- ▶ The constant term is det(A), which is zero if and only if $\lambda = 0$ is a root.
- ▶ The linear term -(a+d) is the negative of the sum of the diagonal entries of A

Definition

The trace of a square matrix A is Tr(A) = sum of the diagonal entries of A.

Shortcut

The characteristic polynomial of a 2×2 matrix A is $f(\lambda) = \lambda^2 - \text{Tr}(A) \, \lambda + \text{det}(A).$

$$f(\lambda) = \lambda^2 - \operatorname{Tr}(A) \lambda + \det(A)$$

Question: What are the eigenvalues of the rabbit population matrix

$$A = \begin{pmatrix} 0 & 6 & 8 \\ \frac{1}{2} & 0 & 0 \\ 0 & \frac{1}{2} & 0 \end{pmatrix}?$$

Answer: First we find the characteristic polynomial:

$$f(\lambda) = \det(A - \lambda I) = \det\begin{pmatrix} -\lambda & 6 & 8\\ \frac{1}{2} & -\lambda & 0\\ 0 & \frac{1}{2} & -\lambda \end{pmatrix}$$
$$= 8\left(\frac{1}{4} - 0 \cdot -\lambda\right) - \lambda\left(\lambda^2 - 6 \cdot \frac{1}{2}\right)$$
$$= -\lambda^3 + 3\lambda + 2.$$

We know from before that one eigenvalue is $\lambda = 2$: indeed, f(2) = -8 + 6 + 2 = 0. Doing polynomial long division, we get:

$$\frac{-\lambda^3+3\lambda+2}{\lambda-2}=-\lambda^2-2\lambda-1=-(\lambda+1)^2.$$

Hence $\lambda = -1$ is also an eigenvalue.

Algebraic Multiplicity

Definition

The (algebraic) multiplicity of an eigenvalue λ is its multiplicity as a root of the characteristic polynomial.

This is not a very interesting notion *yet*. It will become interesting when we also define *geometric* multiplicity later.

Example

In the rabbit population matrix, $f(\lambda) = -(\lambda - 2)(\lambda + 1)^2$, so the algebraic multiplicity of the eigenvalue 2 is 1, and the algebraic multiplicity of the eigenvalue -1 is 2.

Example

In the matrix
$$\begin{pmatrix} 5 & 2 \\ 2 & 1 \end{pmatrix}$$
, $f(\lambda) = (\lambda - (3 - 2\sqrt{2}))(\lambda - (3 + 2\sqrt{2}))$, so the algebraic multiplicity of $3 + 2\sqrt{2}$ is 1, and the algebraic multiplicity of $3 - 2\sqrt{2}$ is 1.

Fact: If A is an $n \times n$ matrix, the characteristic polynomial

$$f(\lambda) = \det(A - \lambda I)$$

turns out to be a polynomial of degree n, and its roots are the eigenvalues of A:

$$f(\lambda) = (-1)^n \lambda^n + a_{n-1} \lambda^{n-1} + a_{n-2} \lambda^{n-2} + \dots + a_1 \lambda + a_0.$$

Poll

If you count the eigenvalues of A, with their algebraic multiplicities, you will get:

- A. Always n.
- B. Always at most n, but sometimes less.
- C. Always at least n, but sometimes more.
- D. None of the above.

The answer depends on whether you allow *complex* eigenvalues. If you only allow real eigenvalues, the answer is B. Otherwise it is A, because any degree-n polynomial has exactly n complex roots, counted with multiplicity. Stay tuned.

The \mathcal{B} -basis

Recall: If $\{v_1, v_2, ..., v_m\}$ is a basis for a subspace V and x is in V, then the \mathcal{B} -coordinates of x are the (unique) coefficients $c_1, c_2, ..., c_m$ such that

$$x = c_1v_1 + c_2v_2 + \cdots + c_mv_m.$$

In this case, the \mathcal{B} -coordinate vector of x is

$$[x]_{\mathcal{B}} = \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_m \end{pmatrix}.$$

Example: The vectors

$$v_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \qquad v_2 = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

form a basis for \mathbf{R}^2 because they are not collinear.

[interactive]

Coordinate Systems on \mathbb{R}^n

Recall: A set of n vectors $\{v_1, v_2, \dots, v_n\}$ form a basis for \mathbf{R}^n if and only if the matrix C with columns v_1, v_2, \dots, v_n is invertible.

If $x = c_1v_1 + c_2v_2 + \cdots + c_nv_n$ then

$$[x]_{\mathcal{B}} = \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{pmatrix} \implies x = c_1 v_1 + c_2 v_2 + c_n v_n = C[x]_{\mathcal{B}}.$$

Since $x = C[x]_{\mathcal{B}}$ we have $[x]_{\mathcal{B}} = C^{-1}x$.

Translation: Let \mathcal{B} be the basis of columns of C. Multiplying by C changes from the \mathcal{B} -coordinates to the usual coordinates, and multiplying by C^{-1} changes from the usual coordinates to the \mathcal{B} -coordinates:

$$[x]_{\mathcal{B}} = C^{-1}x$$
 $x = C[x]_{\mathcal{B}}.$

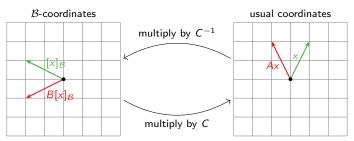
Similarity

Definition

Two $n \times n$ matrices A and B are **similar** if there is an invertible $n \times n$ matrix C such that

$$A=CBC^{-1}.$$

What does this mean? This gives you a different way of thinking about multiplication by A. Let \mathcal{B} be the basis of columns of C.



To compute Ax, you:

- 1. multiply x by C^{-1} to change to the \mathcal{B} -coordinates: $[x]_{\mathcal{B}} = C^{-1}x$
- 2. multiply this by B: $B[x]_{\mathcal{B}} = BC^{-1}x$
- 3. multiply this by C to change to usual coordinates: $Ax = CBC^{-1}x = CB[x]_{\mathcal{B}}$.

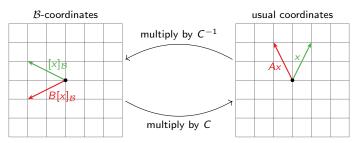
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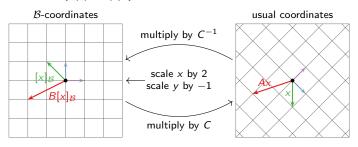


If $A = CBC^{-1}$, then A and B do the same thing, but B operates on the \mathcal{B} -coordinates, where \mathcal{B} is the basis of columns of C.

$$A = \begin{pmatrix} 1/2 & 3/2 \\ 3/2 & 1/2 \end{pmatrix}$$
 $B = \begin{pmatrix} 2 & 0 \\ 0 & -1 \end{pmatrix}$ $C = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$ $A = CBC^{-1}$.

It scales the x-direction by 2 and the y-direction by -1.

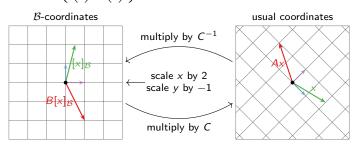
$$\mathcal{B} = \left\{ \begin{pmatrix} 2 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\} = \left\{ v_1, v_2 \right\}$$
 (the columns of C).



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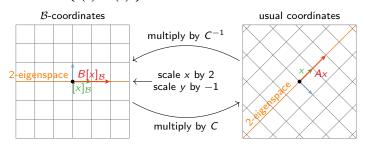
$$\mathcal{B} = \left\{ \begin{pmatrix} 2 \\ 1 \end{pmatrix}, \, \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\} = \left\{ \textit{v}_1, \, \textit{v}_2 \right\} \qquad \text{(the columns of \textit{C})}.$$



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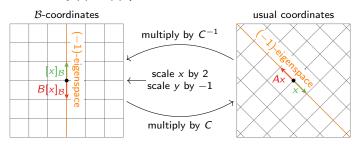
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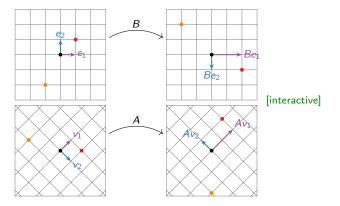
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 (the columns of C).



Similarity Example

What does A do geometrically?

- ▶ B scales the e_1 -direction by 2 and the e_2 -direction by -1.
- A scales the v_1 -direction by 2 and the v_2 -direction by -1.



Since B is simpler than A, this makes it easier to understand A. Note the relationship between the eigenvalues/eigenvectors of A and B.

Similarity Example (3×3)

$$A = \begin{pmatrix} -3 & -5 & -3 \\ 2 & 4 & 3 \\ -3 & -5 & -2 \end{pmatrix} \quad B = \begin{pmatrix} 2 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \quad C = \begin{pmatrix} -1 & 1 & 0 \\ 1 & -1 & 1 \\ -1 & 0 & 1 \end{pmatrix}$$
$$\implies A = CBC^{-1}.$$

What do A and B do geometrically?

- ▶ B scales the e_1 -direction by 2, the e_2 -direction by -1, and fixes e_3 .
- ▶ A scales the v_1 -direction by 2, the v_2 -direction by -1, and fixes v_3 .

Here v_1, v_2, v_3 are the columns of C.

[interactive]

Similar Matrices Have the Same Characteristic Polynomial

Fact: If A and B are similar, then they have the same characteristic polynomial.

Why? Suppose $A = CBC^{-1}$.

$$A - \lambda I = CBC^{-1} - \lambda I$$

= $CBC^{-1} - C(\lambda I)C^{-1}$
= $C(B - \lambda I)C^{-1}$.

Therefore,

$$det(A - \lambda I) = det(C(B - \lambda I)C^{-1})$$

$$= det(C) det(B - \lambda I) det(C^{-1})$$

$$= det(B - \lambda I),$$

because $det(C^{-1}) = det(C)^{-1}$.

Consequence: similar matrices have the same eigenvalues! (But different eigenvectors in general.)

Similarity Caveats

Warning

1. Matrices with the same eigenvalues need not be similar. For instance,

$$\begin{pmatrix} 2 & 1 \\ 0 & 2 \end{pmatrix} \quad \text{and} \quad \begin{pmatrix} 2 & 0 \\ 0 & 2 \end{pmatrix}$$

both only have the eigenvalue 2, but they are not similar.

2. Similarity has nothing to do with row equivalence. For instance,

$$\begin{pmatrix} 2 & 1 \\ 0 & 2 \end{pmatrix} \quad \text{ and } \quad \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

are row equivalent, but they have different eigenvalues.

Summary

We did two different things today.

First we talked about characteristic polynomials:

- We learned to find the eigenvalues of a matrix by computing the roots of the characteristic polynomial $p(\lambda) = \det(A \lambda I)$.
- ▶ For a 2×2 matrix A, the characteristic polynomial is just

$$p(\lambda) = \lambda^2 - \text{Tr}(A)\lambda + \text{det}(A).$$

The algebraic multiplicity of an eigenvalue is its multiplicity as a root of the characteristic polynomial.

Then we talked about similar matrices:

- ▶ Two square matrices A, B of the same size are **similar** if there is an invertible matrix C such that $A = CBC^{-1}$.
- Geometrically, similar matrices A and B do the same thing, except B operates on the coordinate system B defined by the columns of C:

$$B[x]_{\mathcal{B}} = [Ax]_{\mathcal{B}}.$$

► This is useful when we can find a similar matrix B which is simpler than A (e.g., a diagonal matrix).