

§7.2-7.3: Eigenvalues and eigenvectors

Recall from the coyote-roadrunner discrete dynamical system example the importance of:

If $A_{n \times n}$ and $\vec{v} \neq \vec{0}$ with $A\vec{v} = \lambda\vec{v}$
then \vec{v} is an eigenvector of A with eigenvalue λ

How to find eigenvalues, then eigenvectors:

$$A\vec{v} = \lambda\vec{v}$$

iff $A\vec{v} - \lambda\vec{v} = \vec{0}$

iff $A\vec{v} - \lambda I\vec{v} = \vec{0}$

iff $(A - \lambda I)\vec{v} = \vec{0}$. For $\vec{v} \neq \vec{0}$ this can happen iff $A - \lambda I$ is not invertible
i.e. $\det(A - \lambda I) = 0$

step 1: Compute $\det(A - \lambda I) = \det \begin{bmatrix} a_{11} - \lambda & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} - \lambda & \dots & \dots \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & \dots & \dots & a_{nn} - \lambda \end{bmatrix}$
// $p(\lambda)$

called the
characteristic
polynomial

this is a polynomial in λ of degree n , (prove by induction!)
 $p(\lambda) = (-1)^n \lambda^n + \dots + \det A$

Its roots are eigenvalues.

step 2: For each eigenvalue λ the

$\{\vec{v} \mid (A - \lambda I)\vec{v} = \vec{0}\}$ is the kernel of $A - \lambda I$
so is a subspace, and has a basis of eigenvectors,
which we find in the usual way by
computing $\text{rref}(A - \lambda I \mid \vec{0})$
and backsolving.

step 3: In applications we often hope there is a basis of \mathbb{R}^n made
out of eigenvectors of A (like in the coyote-roadrunner problem).
This will not always be the case, although under certain
conditions on A such a result is guaranteed.

Example: $A = \begin{bmatrix} 1 & 2 \\ 4 & 3 \end{bmatrix}$

- Find the eigenvalues λ of A

- For each eigenvalue find an eigenbasis for the corresponding eigenspace.

- Is there a basis of \mathbb{R}^2 made out of eigenvectors of A ?
What is the matrix of $f(x) = \begin{bmatrix} 1 & 2 \\ 4 & 3 \end{bmatrix}x$ with respect to this basis?
Find this matrix B two ways!

$$2 \times 2 \text{ case: } A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$$

$$A - \lambda I = \begin{bmatrix} a_{11} - \lambda & a_{12} \\ a_{21} & a_{22} - \lambda \end{bmatrix}$$

$$\begin{aligned} \det(A - \lambda I) &= (a_{11} - \lambda)(a_{22} - \lambda) - a_{12}a_{21} \\ &= \lambda^2 - (a_{11} + a_{22})\lambda + (a_{11}a_{22} - a_{12}a_{21}) \\ &= \lambda^2 - \text{trace}(A)\lambda + \det A \end{aligned}$$

\uparrow
 sum of
 diagonal
 elts (also for $n \times n$)

e.g. page 2:
 $\lambda^2 - 4\lambda - 5$

$$3 \times 3 \text{ case: } A - \lambda I = \begin{bmatrix} a_{11} - \lambda & a_{12} & a_{13} \\ a_{21} & a_{22} - \lambda & a_{23} \\ a_{31} & a_{32} & a_{33} - \lambda \end{bmatrix}$$

Can you show

$$\det(A - \lambda I) = -\lambda^3 + (\text{trace}(A))\lambda^2 + (\det(A_{1,1}) + \det(A_{2,2}) + \det(A_{3,3}))\lambda + \det A$$

\downarrow \downarrow \downarrow
 $\begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} + \begin{vmatrix} a_{11} & a_{13} \\ a_{31} & a_{33} \end{vmatrix} + \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix}$

Any guesses on how this generalizes to the $n \times n$ case?

Example

$$\text{Let } A = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 2 & 2 \\ 0 & 0 & 3 \end{bmatrix}$$

- What are the eigenvalues of A ?

- Find eigenbases for each eigenvalue

- Is A similar to a diagonal matrix?
(See them on next page to save work in checking whether you get an \mathbb{R}^3 basis of A -eigenvectors.)

Theorem: Let $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k\}$ be eigenvectors of $A_{n \times n}$, with distinct eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_k$. (5)

↑
no two
the same.

Then $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k\}$ are linearly independent.

proof: let

$$c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots + c_k \vec{v}_k = \vec{0}$$

apply A : $\Rightarrow c_1 \lambda_1 \vec{v}_1 + c_2 \lambda_2 \vec{v}_2 + \dots + c_k \lambda_k \vec{v}_k = \vec{0}$

A^2 $c_1 \lambda_1^2 \vec{v}_1 + c_2 \lambda_2^2 \vec{v}_2 + \dots + c_k \lambda_k^2 \vec{v}_k = \vec{0}$

A^{k-1} $c_1 \lambda_1^{k-1} \vec{v}_1 + \dots + c_k \lambda_k^{k-1} \vec{v}_k = \vec{0}$

in matrix form:

$$\begin{bmatrix} | & | & | & | \\ \vec{v}_1 & \vec{v}_2 & \dots & \vec{v}_k \\ | & | & | & | \end{bmatrix} \begin{bmatrix} c_1 & c_2 & \dots & c_k \\ c_1 \lambda_1 & c_2 \lambda_2 & \dots & c_k \lambda_k \\ c_1 \lambda_1^2 & c_2 \lambda_2^2 & \dots & c_k \lambda_k^2 \\ \vdots & \vdots & \ddots & \vdots \\ c_1 \lambda_1^{k-1} & c_2 \lambda_2^{k-1} & \dots & c_k \lambda_k^{k-1} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} | & | & | & | \\ \vec{v}_1 & \vec{v}_2 & \dots & \vec{v}_k \\ | & | & | & | \end{bmatrix} \begin{bmatrix} c_1 & & & \\ & c_2 & & \\ & & \ddots & \\ & & & c_k \end{bmatrix} \begin{bmatrix} 1 & \lambda_1 & \lambda_1^2 & \dots & \lambda_1^{k-1} \\ 1 & \lambda_2 & \lambda_2^2 & \dots & \lambda_2^{k-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \lambda_k & \lambda_k^2 & \dots & \lambda_k^{k-1} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

↑
transpose of Vandermonde matrix,
its $\det = \prod_{i>j} (\lambda_i - \lambda_j) \neq 0$

So is invertible.

Multiply by its inverse (on the right)

$$\Rightarrow \begin{bmatrix} | & | & | & | \\ \vec{v}_1 & \vec{v}_2 & \dots & \vec{v}_k \\ | & | & | & | \end{bmatrix} \begin{bmatrix} c_1 & & & \\ & c_2 & & \\ & & \ddots & \\ & & & c_k \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} | & | & | & | \\ c_1 \vec{v}_1 & c_2 \vec{v}_2 & \dots & c_k \vec{v}_k \\ | & | & | & | \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \Rightarrow c_1 = c_2 = \dots = c_k = 0 \quad \square$$

Corollary: If $A_{n \times n}$ has n distinct eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$, then there is an \mathbb{R}^n basis of eigenvectors of A .

Proof: Each eigenvalue λ_i has at least one non-zero eigenvector \vec{v}_i . (Since the λ_i -eigenspace is at least 1-dim)

By theorem above, $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n\}$ are linearly independent.

since $\dim(\mathbb{R}^n) = n$ they also span \mathbb{R}^n and are a basis.

If A doesn't have n distinct eigenvalues, harder to tell without working:

(6)

example

$$A = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix}$$

$$p(\lambda) = (2-\lambda)^3$$

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

$$p(\lambda) = (2-\lambda)^3$$