

MAT225 Section Summary: 5.1

Eigenvalues and Eigenvectors

Summary

We're considering the transformation $A_{n \times n} : \mathbf{x} \mapsto A\mathbf{x}$. Eigenvectors provide the ideal basis for \mathbf{R}^n when considering this transformation. Their images under the transformation are simple scalings.

Eigenstuff: An **eigenvector** of $A_{n \times n}$ is a nonzero vector \mathbf{x} such that $A\mathbf{x} = \lambda\mathbf{x}$. The scalar λ is called the **eigenvalue** of A corresponding to \mathbf{x} . There may be several eigenvectors corresponding to a given λ .

The idea is that an eigenvector is simply scaled by the transformation, so the actions of a transformation are easily understood for eigenvectors. If we could write a vector as a linear combination of eigenvectors, then it would be easy to calculate its image: if there are n eigenvectors \mathbf{v}_i , with n eigenvalues λ_i , then if

$$\mathbf{u} = c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_n\mathbf{v}_n$$

then

$$A\mathbf{u} = c_1\lambda_1\mathbf{v}_1 + c_2\lambda_2\mathbf{v}_2 + \dots + c_n\lambda_n\mathbf{v}_n$$

Nice, no?

http://klebanov.homeip.net/tildarooniepavel/fb//java/la_applets/Eigen/ Eigenvalue applet/A

If λ is an eigenvalue of matrix A corresponding to eigenvector \mathbf{v} , then

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

This means the

$$A\mathbf{v} - \lambda\mathbf{v} = \mathbf{0}$$

which is equivalent to

$$(A - \lambda I)\mathbf{v} = \mathbf{0}$$

So \mathbf{v} is in the null space of $A - \lambda I$. If the null space is trivial, then \mathbf{v} is the zero vector, and λ is not an eigenvalue. Alternatively, all vectors in the null space are eigenvectors corresponding to the eigenvalue λ .

*IF $\det(A - \lambda I) = 0$,
then there is a
non-trivial null-
space - \mathbf{v} is in
 \mathcal{N} !*

As for determining the eigenvectors and eigenvalues, there is some cases in which this is extremely easy:

The eigenvalues of a diagonal matrix are the entries on its diagonal. More generally,

Theorem 1: The eigenvalues of a triangular matrix are the entries on its main diagonal.

Theorem 2: If v_1, \dots, v_r are eigenvectors corresponding to distinct eigenvalues $\lambda_1, \dots, \lambda_r$ of an $n \times n$ matrix A , then the set $\{v_1, \dots, v_r\}$ is linearly independent. *n distinct eigenvalues of an n x n matrix =>*

The eigenvectors and difference equations portion of this section can be illustrated with the example of the Fibonacci numbers transformation: recall that the Fibonacci numbers are those obtained by the recurrence relation

$$F_n = F_{n-1} + F_{n-2}$$

and $F_0 = 1$ and $F_1 = 1$.

$$A = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \mathbf{x}_n = \mathbf{x}_{n+1}$$

where

$$\mathbf{x}_0 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$A \bar{\mathbf{x}}_0 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}; \quad A^2 \bar{\mathbf{x}}_0 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

$$A^3 \bar{\mathbf{x}}_0 = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$$

The eigenvalues of this matrix are approximately $\gamma = \frac{1+\sqrt{5}}{2} \approx 1.618033988749895$ and -0.618033988749894 . γ is the so-called "golden mean", which is a nearly sacred number in nature, well approximated by the ratio of consecutive Fibonacci numbers.

An eigenvector corresponding to the golden mean (normalized to have a norm of 1) is approximately

$$\begin{bmatrix} 0.5257311121191337 \\ 0.8506508083520401 \end{bmatrix}$$

so that

$$\begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 0.5257311121191337 \\ 0.8506508083520401 \end{bmatrix} = \gamma \begin{bmatrix} 0.5257311121191337 \\ 0.8506508083520401 \end{bmatrix}$$

Example: full matrix

$$A = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix}$$

Find eigenvalues: $A - \lambda I = \begin{bmatrix} 1-\lambda & 2 \\ 2 & 1-\lambda \end{bmatrix}$

Set $\det(A - \lambda I) = 0$

$$(1-\lambda)(1-\lambda) - 4 = 0$$

$$\lambda^2 - 2\lambda - 3 = 0$$

$$(\lambda - (-1))(\lambda - 3) = 0$$

$$\lambda_1 = -1 \text{ and } \lambda_2 = 3 \quad (\text{distinct!})$$

Now we need corresponding eigenvectors:

$$\lambda_1 = -1: \quad A - (-1)I = \begin{bmatrix} 2 & 2 \\ 2 & 2 \end{bmatrix}$$

Spot an eigenvector! $\bar{v}_1 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$

$$\lambda_2 = 3: \quad A - 3I = \begin{bmatrix} -2 & 2 \\ 2 & -2 \end{bmatrix} \quad \bar{v}_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

#2 $A - (-2)I = \begin{bmatrix} 7 & 3 \\ 3 & -1 \end{bmatrix} - \begin{bmatrix} -2 & 0 \\ 0 & -2 \end{bmatrix} = \begin{bmatrix} 9 & 3 \\ 3 & 1 \end{bmatrix}$

Columns are linearly dependent \Rightarrow non-trivial null space. $\det(A - (-2)I) = 9 - 9 = 0$.

$$\bar{v}_1 = \begin{bmatrix} 1 \\ -3 \end{bmatrix} \quad (A - (-2)I) \bar{v}_1 = \bar{0}$$

an eigenvector

$$\#9 \quad A = \begin{bmatrix} 5 & 0 \\ 2 & 1 \end{bmatrix} \quad \lambda_1 = 1, \lambda_2 = 5$$

$$(A - \lambda_1 I) \bar{v}_1 = \bar{0} \quad A - \lambda_1 I = \begin{bmatrix} 4 & 0 \\ 2 & 0 \end{bmatrix}$$

$$\bar{v}_1 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$(A - \lambda_2 I) \bar{v}_2 = \bar{0} \quad A - \lambda_2 I = \begin{bmatrix} 0 & 0 \\ 2 & -4 \end{bmatrix}$$

$$\bar{v}_2 = \begin{bmatrix} -2 \\ -1 \end{bmatrix}$$

So $\{\bar{v}_1, \bar{v}_2\}$ forms a basis for \mathbb{R}^2 ,
since they correspond to distinct eigenvalues.

$$A - \lambda I = \begin{bmatrix} 5 - \lambda & 0 \\ 2 & 1 - \lambda \end{bmatrix}$$

Demand $\det(A - \lambda I) = 0$ for eigenvalues

$$(5 - \lambda)(1 - \lambda) = 0$$

Back to #2:

$$A - \lambda I = \begin{bmatrix} 7 - \lambda & 3 \\ 3 & -1 - \lambda \end{bmatrix} \quad \det(A - \lambda I) = 0$$

$$(7 - \lambda)(-1 - \lambda) - 9 = \lambda^2 - 6\lambda - 14$$

$$= (\lambda + 2)(\lambda - 8)$$

$$\lambda = -2 \quad \lambda = 8$$