

Section 7.1

Diagonalization of symmetric matrices

Motivation: Diagonalization

How did we **recognize** diagonalizable matrices?

- ▶ They are already diagonal
- ▶ They have n distinct eigenvalues
Quick to check: only if matrix is triangular
- ▶ The algebraic and geometric multiplicities are equal for all eigenvalues and they sum up to n .

New criterion: Verify if matrix is *symmetric*!

- ▶ Symmetric, e.g. $\begin{pmatrix} 3 & 1 \\ 1 & 2 \end{pmatrix}$
- ▶ **Not** symmetric, e.g. $\begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}, \begin{pmatrix} 1 & -4 & 0 \\ 6 & 1 & -4 \\ 0 & 6 & 1 \end{pmatrix}$

Warm up: $u^T u$ vs uu^T

If v is a vector in \mathbf{R}^n with entries u_1, u_2, \dots, u_n , then

▶ $u^T u = u_1^2 + u_2^2 + \dots + u_n^2$ is a *scalar*.

▶ uu^T is an $n \times n$ **matrix**:

$$uu^T = \begin{pmatrix} u_1 \cdot u_1 & u_1 \cdot u_2 & \cdots & u_1 \cdot u_n \\ u_2 \cdot u_1 & u_2 \cdot u_2 & \cdots & u_2 \cdot u_n \\ \vdots & \vdots & \ddots & \vdots \\ u_n \cdot u_1 & u_n \cdot u_2 & \cdots & u_n \cdot u_n \end{pmatrix}$$

A projection matrix!

In fact, uu^T is the *standard matrix* for the transformation $T: \mathbf{R}^n \rightarrow \mathbf{R}^n$ that projects onto the line spanned by u .

Warm up: Inverse of an orthonormal matrix

For *orthogonal matrices* Q we already know that

$$Q^T Q = \begin{pmatrix} u_1 \cdot u_1 & 0 & \cdots & 0 \\ 0 & u_2 \cdot u_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ \vdots & \vdots & \vdots & u_n \cdot u_n \end{pmatrix}$$

so for **orthonormal matrices** Q

$$Q^T Q = \begin{pmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ \vdots & \vdots & \vdots & 1 \end{pmatrix}$$

What is *the inverse of* Q ?

Orthogonally diagonalizable

Definition

An $n \times n$ matrix A is **orthogonally diagonalizable** if $A = PDP^{-1}$ with D *diagonal* matrix and P an *orthonormal* matrix.

To stress the orthogonality of P we write $A = PDP^T$.

Avoiding errors

Computations using orthogonal matrices usually prevents numerical errors from accumulating.

Collection of eigenvalues = 'Spectral'

Spectral decomposition

If D has diagonal entries $\lambda_1, \dots, \lambda_n$ and P has columns u_1, \dots, u_n then

$$A = \lambda_1 u_1 u_1^T + \lambda_2 u_2 u_2^T + \dots + \lambda_n u_n u_n^T$$

- ▶ Fancy way of expressing the change of variables and
- ▶ the fact that principal axes are only stretch/contracted
- ▶ Each of $u_i u_i^T$ is a projection matrix

Why?

Example: Orthogonally diagonalizable

Example

Orthogonally diagonalize the matrix $A = \begin{pmatrix} 3 & -2 & 4 \\ -2 & 6 & 2 \\ 4 & 2 & 3 \end{pmatrix}$

its *characteristic equation* is $-(\lambda - 7)^2(\lambda + 2) = 0$.

Find a basis for each λ -eigenspace:

$$\text{For } \lambda = 7: \left\{ \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}, \begin{pmatrix} -1/2 \\ 1 \\ 0 \end{pmatrix} \right\} \quad \text{For } \lambda = -2: \left\{ \begin{pmatrix} -1 \\ -1/2 \\ 1 \end{pmatrix} \right\}$$

A suitable P

Is the set of eigenvectors above already orthogonal?
orthonormal?

$$A = P \begin{pmatrix} 7 & 0 & 0 \\ 0 & 7 & 0 \\ 0 & 0 & -2 \end{pmatrix} P^{-1}$$

Example: Orthogonally diagonalizable

continued

Verify:

- ▶ $v_3 = \begin{pmatrix} -1 \\ -1/2 \\ 1 \end{pmatrix}$ is *already orthogonal* to $v_1 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$ and $v_2 = \begin{pmatrix} -1/2 \\ 1 \\ 0 \end{pmatrix}$
- ▶ but $v_1 \cdot v_2 \neq 0$.

Tackle this: Use Gram-Schmidt

$$u_1 = v_1$$

$$u_2 = v_2 - \frac{v_2 \cdot v_1}{v_1 \cdot v_1} v_1 = \begin{pmatrix} -1/2 \\ 1 \\ 0 \end{pmatrix} - \frac{-1/2}{2} \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} -1/4 \\ 1 \\ 1/4 \end{pmatrix}$$

And $u_3 = v_3$. *Then normalize!*

$$P = \begin{pmatrix} 1/\sqrt{2} & -1\sqrt{18} & -2/3 \\ 0 & 4\sqrt{18} & -1/3 \\ 1/\sqrt{2} & 1\sqrt{18} & 2/3 \end{pmatrix}, \quad D = \begin{pmatrix} 7 & 0 & 0 \\ 0 & 7 & 0 \\ 0 & 0 & -2 \end{pmatrix}$$

Example: Spectral Decomposition

Example

Construct a *spectral decomposition* of the matrix A with orthogonal diagonalization

$$A = \begin{pmatrix} 7 & 2 \\ 2 & 4 \end{pmatrix} = \begin{pmatrix} 2/\sqrt{5} & -1/\sqrt{5} \\ 1/\sqrt{5} & 2/\sqrt{5} \end{pmatrix} \begin{pmatrix} 8 & 0 \\ 0 & 3 \end{pmatrix} \begin{pmatrix} 2/\sqrt{5} & 1/\sqrt{5} \\ -1/\sqrt{5} & 2/\sqrt{5} \end{pmatrix}$$

Solution: Then $A = 8u_1u_1^T + 3u_2u_2^T$, each matrix is

$$u_1u_1^T = \begin{pmatrix} 4/5 & 2/5 \\ 2/5 & 1/5 \end{pmatrix}$$

$$u_2u_2^T = \begin{pmatrix} 1/5 & -2/5 \\ -2/5 & 4/5 \end{pmatrix}$$

Check: $8u_1u_1^T + 3u_2u_2^T = \begin{pmatrix} 32/5 & 16/5 \\ 16/5 & 8/5 \end{pmatrix} + \begin{pmatrix} 3/5 & -6/5 \\ -6/5 & 12/5 \end{pmatrix} = A$

Symmetric matrices

Definition

An $n \times n$ matrix is **symmetric** if $A = A^T$.

Theorem

An $n \times n$ matrix A is *orthogonally diagonalizable* if and only if A is *symmetric*.

The **easy observation**: Let $A = PDP^T$ with D diagonal and P orthonormal.

Just check A is symmetric, that is $A = A^T$:

$$\underbrace{(PDP^T)}_A = (P^T)^T D^T P^T = \underbrace{PDP^T}_A$$

The **difficult part** (omitted here) is: if $A = A^T$ then

an *orthogonal diagonalization do exists*.

Spectral Theorem for Symmetric matrices

An $n \times n$ *symmetric* matrix A has the **following properties**.

- ▶ A has n *real eigenvalues*, counting multiplicities
- ▶ For each eigenvalue, the dimension of the λ -eigenspaces equal the algebraic multiplicity.
- ▶ The eigenspaces are *mutually orthogonal!* eigenvectors corresponding to different eigenvalues are orthogonal.
- ▶ A is *orthogonally diagonalizable*.

Extra: Eigenspaces are mutually orthogonal

Symmetric matrices only

Eigenspaces are mutually orthogonal

What does it mean?

If v_1 and v_2 are eigenvectors that correspond to distinct eigenvalues λ_1 and λ_2
then $v_1 \cdot v_2 = 0$.

Trick to see this: Find a way to show that $(\lambda_1 - \lambda_2)v_1 \cdot v_2 = 0$.

Why? We assumed that $\lambda_1 \neq \lambda_2$ so necessarily $v_1 \cdot v_2 = 0$.

Hint: Compute $v_1^T A v_2$ in two different 'orders'

- ▶ Symmetry is *important*: You'll have to substitute $A = A^T$ at some point.