Diagonalization

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Linear independence of eigenvectors

If A is a $n \times n$ matrix and v_1, \ldots, v_k are eigenvectors for distinct eigenvalues $\lambda_1, \ldots, \lambda_k$, then v_1, \ldots, v_k are linearly independent.

Proof

If there is exactly one eigenvector and one eigenspace, the eigenvectors are trivially linearly independent. Assume that $\{v_1, \ldots, v_i\}$ is linearly independent.

$$c_1 v_1 + \dots + c_i v_i + c_{i+1} v_{i+i} = \vec{0}$$

$$A(c_1 v_1 + \dots + c_i v_i + c_{i+1} v_{i+i}) = A\vec{0}$$

$$c_1 \lambda_1 v_1 + \dots + c_i \lambda_i v_i + c_{i+1} \lambda_{i+1} v_{i+1} = \vec{0}$$
(2)

Multiply (1) by λ_{i+1}

$$c_1 \lambda_{i+1} v_1 + \dots + c_i \lambda_{i+1} v_i + c_{i+1} \lambda_{i+1} v_{i+1} = 0$$
(3)

Subtract (3) from (2)

$$(c_1\lambda_1v_1 + \dots + c_i\lambda_iv_i + c_{i+1}\lambda_{i+1}v_{i+1}) - (c_1\lambda_{i+1}v_1 + \dots + c_i\lambda_{i+1}v_i + c_{i+1}\lambda_{i+1}v_{i+1}) = \vec{0} - \vec{0}$$

$$c_1(\lambda_1 - \lambda_{i+1})v_1 + \dots + c_i(\lambda_i - \lambda_{i+1})v_i + c_{i+1}(\lambda_{i+1} - \lambda_{i+1})v_{i+1} = \vec{0}$$

$$c_1(\lambda_1 - \lambda_{i+1})v_1 + \dots + c_i(\lambda_i - \lambda_{i+1})v_i = \vec{0}$$

Since $\{v_1, \ldots, v_i\}$ is linearly independent, then for all n, $c_n(\lambda_n - \lambda_{i+1}) = 0$. Using the zero product property, either $c_n = 0$ or $\lambda_n - \lambda_{i+1} = 0$, but since the eigenvalues are distinct, c_n must be equal to zero. Thus,

$$c_1, \dots, c_i = 0$$

$$c_1v_1 + \dots + c_iv_i + c_{i+1}v_{i+i} = \vec{0}$$

$$0v_1 + \dots + 0v_i + c_{i+1}v_{i+i} = \vec{0}$$

$$c_{i+1}v_{i+i} = \vec{0}$$

$$c_{1+i} = 0$$

Therefore $\{v_1, \ldots, v_i, v_{i+1}\}$ is linearly independent. Using the inductive hypothesis, every eigenvector belonging to distinct eigenvalues are linearly independent.

Similarity

For two $n \times n$ matrices A and B, we sat that A is similar to B if there exists a invertible matrix P such that $P^{-1}AP = B$. If A and B are similar then they have the same characteristic polynomial and thus, have the same eigenvalues.

$$\begin{split} \det(B-\lambda I) &= \det(P^{-1}AP - \lambda I) \\ &= \det(P^{-1}(A-\lambda I)P) \\ &= \det(P^{-1})\det((A-\lambda I)\det(P) \\ &= \det(A-\lambda I) \end{split}$$

Diagonalization

If a $n \times n$ matrix A have n different eigenvalues then A is diagonalizable. A is similar to a diagonal matrix iff there are n linearly independent eigenvectors for A. In $P^{-1}AP = D$, the columns of P are the eigenvectors and the diagonals of D are the eigenvalues.

$$P = \begin{bmatrix} v_1 & v_2 & \dots & v_n \end{bmatrix}$$

$$D = \begin{bmatrix} \lambda_1 & 0 & & 0 \\ 0 & \lambda_2 & & 0 \\ & & \ddots & \\ 0 & 0 & & \lambda_n \end{bmatrix}$$

From the construction of D, we can show that P contains the eigenvectors.

$$P^{-1}AP = D$$

$$AP = PD$$

$$AP = \begin{bmatrix} Av_1 & Av_2 & \dots & Av_n \end{bmatrix}$$

$$PD = \begin{bmatrix} \lambda_q v_1 & \lambda_2 v_2 & \dots & \lambda_n v_n \end{bmatrix}$$

Algorithm for Diagonal

- 1. Suppose $\lambda_1, \ldots, \lambda_k$ are eigenvalues of A.
- 2. For each eigenvalue, determine the basis for the eigenspace. If any eigenspace has a dimension less than it's algebraic multiplicity, it is not diagonalizable.
- 3. Otherwise, collect all the bases and put them in the columns of a matrix P.
- 4. P will be invertible and AP = PD where D is a diagonal matrix that has eigenvalue corresponding with the columns of P.

Powers of Diagonal Matrices

The powers of a diagonal matrix is the power of the elements in the diagonal matrix. If a matrix is diagonalizable, then the powers of that matrix can be easily computed by first diagonalizing

it and raising the diagonal matrix to the power.

$$D = \begin{bmatrix} d_1 & 0 \\ 0 & d_2 \end{bmatrix}$$

$$D^k = \begin{bmatrix} d_1^k & 0 \\ 0 & d_2^k \end{bmatrix}$$

$$A^{k} = (PDP^{-1})(PDP^{-1})\dots(PDP^{-1})$$

$$= PD(P^{-1}P)D(P^{-1}P)\dots(P^{-1}P)DP^{-1}$$

$$= PDD\dots DP^{-1})$$

$$= PD^{k}P^{-1})$$

Eigenvector Basis

Let B be a basis constructed from the eigenvectors of a transformation T.

$$[T]_B = \begin{bmatrix} [T(v_1)]_B & [T(v_1)]_B & \dots & [T(v_n)]_B \end{bmatrix}$$
$$= \begin{bmatrix} \lambda_1 e_1 & \lambda_2 e_2 & \dots & \lambda_n e_n \end{bmatrix}$$

The standard matrix for a linear transformation with respect to the basis of eigenvectors is a diagonal matrix.

Example 1

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

$$x_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, Ax_1 = 3x_1$$

$$x_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}, Ax_2 = 1x_2$$

$$P = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$$

$$P^{-1} = \frac{1}{2} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}$$

$$P^{-1}AP = \begin{bmatrix} 3 & 0 \\ 0 & 1 \end{bmatrix}$$

Example 2

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

$$T_A : \mathbb{R}^3 \mapsto \mathbb{R}^3$$

$$A = [T_A]_{\{e1,e2,e3\}}$$

$$B = \left\{ \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix} \right\}$$

$$[T]_B = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

Example 3

$$V = span\{\sin x, \cos x\}$$

$$dim(V) = 2$$

$$f \in V$$

$$T(f) = \frac{df}{dx}$$

$$T: V \mapsto V$$

$$B = \{\sin x, \cos x\}$$

$$[T]_B = [[T(v_1)]_B \quad [T(v_2)]_B]$$

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