



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

$$\underline{A} \underline{0} = \lambda \underline{0} \quad \forall \lambda$$

$$\underline{0} = \underline{0}$$

$\underline{0}$ is not an eigenvector

Stability of linear syst. §18.3.2

$\frac{dx}{dt} = \underline{A}x$ is stable iff all eigenvalues of \underline{A} are negative, or at least not positive,

$$c \vee e^{\lambda t}$$

Positive Def §18.3.3.

$$\begin{array}{llll} \underline{x}^T \underline{A} \underline{x} > 0 & \text{P.D.} & \text{All eigenvalues} > 0 \\ \underline{x}^T \underline{A} \underline{x} \geq 0 & \text{P.S.D.} & \text{" " " } & \geq 0 \end{array}$$

Determinant

$\det(A) = \text{product of eigenvalues of } \underline{A}.$

Pg 165: Properties of determinant.

$$\underline{v}_1, \underline{v}_2, \dots, \underline{v}_n$$

Vectors span a space if for any $\underline{x} \in \text{space}$,

$$\underline{x} = a_1 \underline{v}_1 + a_2 \underline{v}_2 + \dots + a_n \underline{v}_n$$

Vectors are a basis if the decomposition is unique.

1. Span the space.
2. Linearly independent
3. # of vectors = dimension of space.

$$c_1 \underline{v}_1 + c_2 \underline{v}_2 + \dots + c_n \underline{v}_n = \underline{0} \Leftrightarrow c_1 = c_2 = \dots = c_n = 0$$

Lin ind.

$$\underline{V} = (\underline{v}_1 \ \underline{v}_2 \ \dots \ \underline{v}_n), \quad \text{Rank}(\underline{V}) = n$$

Orthonormal basis: basis mutually orthogonal & all vectors are unit vectors.

Eigendecomp.

$$\underline{A} = \underline{V} \underline{\Lambda} \underline{V}^{-1}$$

Columns equal
to eigenvectors

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

SVD

$$\underline{A} = \underline{U} \underline{\Sigma} \underline{V}^T$$

\underline{V} is orthogonal : $\underline{V}^{-1} = \underline{V}^T$
(orthonormal columns)

~~Singular~~

All eigenvectors for a matrix are unique.

But, eigenvalues are not necessarily unique.

Basis $\underline{v}_1, \dots, \underline{v}_n$

To decompose \underline{x} onto it.

$$\underline{V} = (\underline{v}_1 \ \underline{v}_2 \ \dots \ \underline{v}_n)$$

Coefficients \underline{a} satisfy

$$\underline{V}\underline{a} = \underline{x}$$

$$\underline{a} = \underline{V}^{-1}\underline{x}$$

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Shortcut for orthonormal basis

$$a_i = \underline{x} \cdot \underline{v}_i$$

$$\underline{y} = \underline{\sigma}(\underline{W}_d \dots \underline{\sigma}(\underline{W}_2(\underline{\sigma}(\underline{W}_1 \underline{x}))))$$

Multi
Layer
Perceptron

$$\underline{W}_{i+1}(\underline{\sigma}(\underline{W}_i \dots))$$

without $\underline{\sigma}$,

$$\underbrace{\underline{W}_{i+1} \underline{W}_i}_{\underline{W}}$$