

Linear algebra for AI and ML

(November 11)



Given $A \in \mathbb{R}^{n \times n}$

eigenvalues

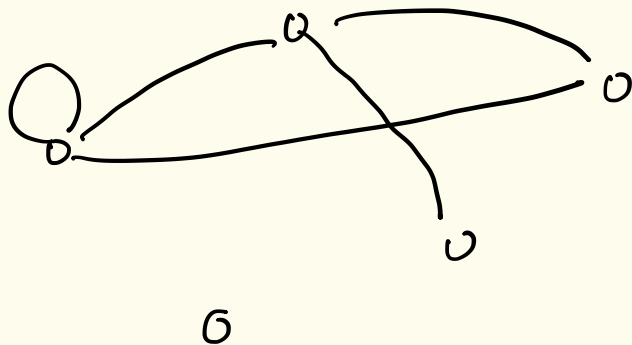
scaling

and

eigenvectors

invariant
directions

Markov chain: with finite states



P = Probability
transition
matrix

Eigenvector corresponding
to the eigenvalue
1 is called
stationary distⁿ.

$$\frac{d}{dt} x = Ax$$

$$x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$$

$$e^{At}$$

$$x = \alpha_1 e^{\lambda_1 t} v_1 + \dots$$

Given a matrix A , compute its eigenvalues and corresponding eigenvectors.

[QR - algorithm] \leftarrow iterative process

[Krylov subspace]

David Watkins / Golub

[Computational linear algebra]

Q: $A \in \mathbb{R}^{n \times n}$ \rightarrow compute eigenvalues (eigenvectors)

Q: Given a set of numbers (eigenvalues) and a set of vectors (corresponding eigenvectors), construct a matrix A such that this matrix has these numbers as its eigenvalues and these vectors as corresponding eigenvectors.

(Inverse eigenvalue problem)

Observe: If (λ, x) is an eigenpair of A

then

$$Ax = \lambda x$$

unknown

- i) unknown variables are n^2 . (entries in the matrix A)
- ii) if ' m ' pairs of eigenvalues and eigenvectors are given ($1 \leq m \leq n$), then we have mn number of relationships. (equations in unknowns)
- iii) Relationships (eq^m) are linear.

Ex: let $\left(2, \begin{pmatrix} -1 \\ 1 \end{pmatrix} \right)$ be given.

Compute $A \in \mathbb{R}^{2 \times 2}$ s.t. A has $\left(2, \begin{pmatrix} -1 \\ 1 \end{pmatrix} \right)$ as its eigen pair.

→ Eigenvalue - eigenvector relationship

$$\underbrace{\begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix}}$$

$$\Rightarrow Ax = \lambda x$$

$$\Rightarrow \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} -2 \\ 2 \end{bmatrix}$$

$$\Rightarrow -a_{11} + a_{12} = -2$$

$$-a_{21} + a_{22} = 2$$

$$\Rightarrow \begin{bmatrix} -1 & 1 & 0 & 0 \\ 0 & 0 & -1 & 1 \end{bmatrix} \begin{bmatrix} a_{11} \\ a_{12} \\ a_{21} \\ a_{22} \end{bmatrix} = \begin{bmatrix} -2 \\ 2 \end{bmatrix}$$

$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} -2 \\ 2 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} -1 & 0 & 1 & 0 \\ 0 & -1 & 0 & 1 \end{bmatrix} \begin{bmatrix} a_{11} \\ a_{21} \\ a_{12} \\ -a_{22} \end{bmatrix} = \begin{bmatrix} -2 \\ 2 \end{bmatrix}$$

↑
vectorization of a matrix

observation: In general, solving the inverse eigenvalue problem is equivalent to solving an underdetermined (partial) or square (complete) system of linear eq^{ns}.
 $m < n$

Ex: P : probability transition matrix. (n states)

$$x \in \mathbb{R}^n \text{ s.t.}$$

$$P x = x$$

(x is a stationary distribution)

Bernouli ($p = 1/2$)

$$x = \begin{pmatrix} 1/2 \\ 1/2 \end{pmatrix}$$

Uniform (6)

$$x = \begin{pmatrix} 1/6 \\ 1/6 \\ \vdots \\ 1/6 \end{pmatrix}$$

$N(0,1)$,

$$x \in \mathbb{R}^n \text{ where } n = 2^5$$

$$P_{ij} \geq 0$$

M.T. Chu

Boyd

(SIAM Review)