

QR Decomp.

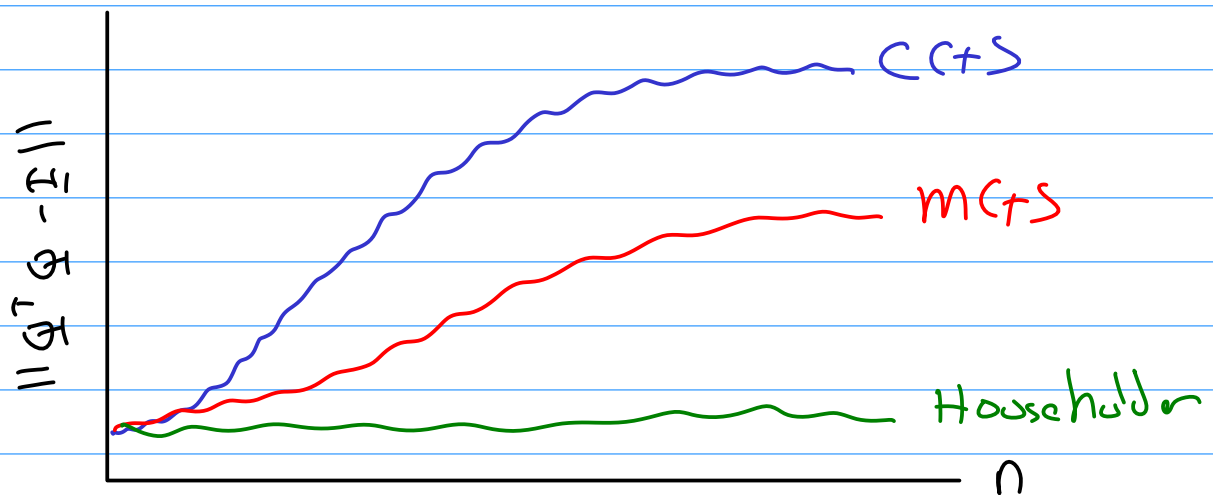
$$A = QR \quad \text{w/} \quad Q^{-1} = Q^T$$

$$Ax = b \Rightarrow QRx = b \Rightarrow x = R^{-1}Q^T b$$

Solving $Ax = b$ via QR returns the least squares solution.

Even if $A \in M_{mn}$ w/ $m \neq n$

3 methods: Classical Gram-Schmidt
Modified Gram-Schmidt
Householder Reflectors



Eigensystems

Motivation: look at $\frac{dy}{dt} = a y(t)$ $a \in \mathbb{R}$

Solution is $y(t) = C e^{at}$ $C \in \mathbb{R}$

$$\frac{d}{dt}(C e^{at}) = C a e^{at} = a(C e^{at}) = a y(t)$$

What about coupled ODEs?

$$\frac{dy_1}{dt} = a_{11} y_1(t) + a_{12} y_2(t)$$

$$\frac{dy_2}{dt} = a_{21} y_1(t) + a_{22} y_2(t)$$

Arise, in chemical reaction, springs, etc.

$$\text{let } \underline{A} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \quad \underline{y}(t) = \begin{bmatrix} y_1(t) \\ y_2(t) \end{bmatrix}$$

$$\frac{d\underline{y}}{dt} = \begin{bmatrix} \partial y_1 / \partial t \\ \partial y_2 / \partial t \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} y_1(t) \\ y_2(t) \end{bmatrix} = \underline{A} \underline{y}(t)$$

$$\Rightarrow \frac{d\underline{y}}{dt} = \underline{A} \underline{y}(t)$$

As $\frac{dy}{dt} = a y(t)$ has a solution of $y(t) = C e^{at}$,

make the ansatz that $\underline{y}(t) = e^{at} \underline{x}$ solve,

$$\frac{d\underline{y}}{dt} = \underline{A} \underline{y}(t)$$

$$\frac{d\underline{y}}{dt} = \frac{d}{dt} (e^{\lambda t} \underline{x}) = \lambda e^{\lambda t} \underline{x}$$

\Downarrow

$$A\underline{y} = A(e^{\lambda t} \underline{x}) = e^{\lambda t} A\underline{x}$$

$$\Rightarrow \cancel{e^{\lambda t}} \underline{x} = \cancel{e^{\lambda t}} A\underline{x}$$

$\Rightarrow A\underline{x} = \lambda \underline{x} \Rightarrow$ (Given A , if you can find a λ & \underline{x} such that $A\underline{x} = \lambda \underline{x}$ then $\underline{y}(t) = e^{\lambda t} \underline{x}$ solves $\frac{d\underline{y}}{dt} = A\underline{y}$)

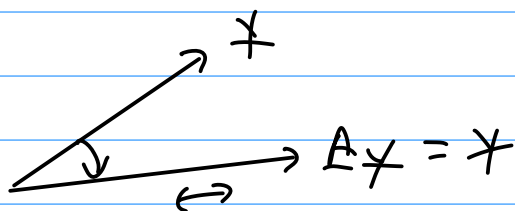
$A\underline{x} = \lambda \underline{x}$ is the **Eigensystem of A**

\underline{x} = **eigenvector**

λ = **eigenvalue**

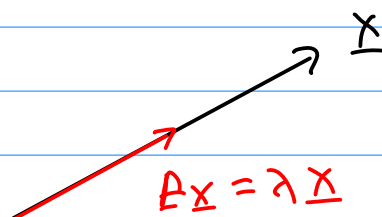
Why are λ & \underline{x} special?

Look at $\underline{y} \neq$ eigenvector



Now look at $A\underline{x} = \lambda \underline{x}$

\underline{x} just scales by λ



$$\text{ex.) } \underline{A} = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix} \quad \underline{x} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$\underline{A}\underline{x} = \begin{bmatrix} 2 \\ 1 \end{bmatrix} \neq \lambda \underline{x}$$

$$\text{try } \underline{x} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$\underline{A}\underline{x} = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 3 \\ 3 \end{bmatrix} = 3 \begin{bmatrix} 1 \\ 1 \end{bmatrix} = 3 \underline{x}$$

$\Rightarrow \lambda = 3$

Eigenvalues & eigenvectors always
come as a pair.

How to compute λ

let $\underline{A} \in M_{nn}$ (only square)

$$\begin{aligned} \underline{A}\underline{x} &= \lambda \underline{x} \\ (n \times p)(p \times 1) &= p \times 1 \\ n \times 1 &= p \times 1 \quad \Rightarrow n = p \end{aligned}$$

Given \underline{A} find λ

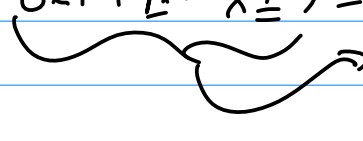
$$\underline{A}\underline{x} = \lambda \underline{x}$$

$$\underline{A}\underline{x} - \lambda \underline{x} = \underline{0}$$

$$(\underline{A} - \lambda \underline{I})\underline{x} = \underline{0}$$

1) \underline{x} must be in the Nullspace of $\underline{A} - \lambda \underline{I}$

2) $\det(\underline{A} - \lambda \underline{I}) = 0$

 Characteristic Eq. of \underline{A}
A polynomial

\Rightarrow Procedure :

1) Solve $\det(A - \lambda I) = 0$ for all λ
(roots)

2) for each λ find the nullspace of
 $A - \lambda I$

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

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

$$\begin{vmatrix} 1-\lambda & 2 \\ 2 & 1-\lambda \end{vmatrix} = (1-\lambda)(1-\lambda) - 4 = 0$$

$$= (1-\lambda)^2 - 4 = 0$$

$$\Rightarrow (1-\lambda)^2 = 4 \Rightarrow 1-\lambda = \pm 2$$

$$\Rightarrow \lambda_1 = -1 \quad \lambda_2 = +3$$

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

$$\text{ref}([A - \lambda I : 0]) = \begin{bmatrix} 1 & 1 & : & 0 \\ 0 & 0 & : & 0 \end{bmatrix}$$

Note: If $\text{ref}(A - \lambda I) = I$, then λ is
not an eigenvalue.

$$\text{Set } x_2 = 1 \Rightarrow x_1 = -1 \Rightarrow \underline{x}_1 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

$$\text{Check: } A \underline{x}_1 = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \underset{\lambda_1}{(-1)} \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

$$\lambda_2 = 3 : A - \lambda_2 I = \begin{bmatrix} -2 & 2 \\ 2 & -2 \end{bmatrix}$$

$$\text{rank}(A - \lambda_2 I) = \begin{bmatrix} 1 & -1 \\ 0 & 0 \end{bmatrix}$$

$$x_2 = +1 \Rightarrow x_1 = +1 \Rightarrow \underline{x}_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Note: Eigenvectors are unique up to a sign & scale.

If \underline{x}_1 is an eigenvector so is $\alpha \underline{x}_1$, $\alpha \in \mathbb{R}$, $\alpha \neq 0$

$$A(\alpha \underline{x}_1) = \lambda(\alpha \underline{x}_1) \Rightarrow \alpha(A\underline{x}_1) = \alpha(\lambda \underline{x}_1)$$

Comments

1) Eigenvalues of A^2 are the square of those of A ,

$$\text{let } A\underline{x} = \lambda \underline{x}$$

$$A^2 \underline{x} = A(A\underline{x}) = A(\lambda \underline{x}) = \lambda A\underline{x} = \lambda(\lambda \underline{x}) = \lambda^2 \underline{x}$$

In general, if $A\underline{x} = \lambda \underline{x}$ then

$$A^n \underline{x} = \lambda^n \underline{x}$$

2) Row operations on A do not preserve eigenvalues or eigenvectors

$$\begin{bmatrix} 4 & -1 & 0 \\ 0 & 1 & 0 \\ 2 & -3 & 6 \end{bmatrix}$$

$$\lambda = 4, 1, 6$$

$$\begin{bmatrix} 2 & -3 & 6 \\ 0 & 1 & 0 \\ 4 & -1 & 0 \end{bmatrix}$$

$$\lambda = 4, 1, 6$$

$$3) \operatorname{tr}(A) = \sum_i \lambda_i$$

$$4) \det(A) = \prod_i \lambda_i$$

Thus if $\det(A) = 0$ then at least one λ is zero.

$$\text{If } \lambda = 0 : (A - \lambda I)x = \underline{0}$$

$$Ax = \underline{0}$$

$\Rightarrow \det(A) = 0$ if x is not trivial

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

1) All roots of $\det(A - \lambda I)$ are real & unique
(Given n eigenvalues find n eigenvectors.)

2) Complex eigenvalues

$$\text{ex. 1 } \underline{A} = \begin{bmatrix} 6 & 0 & 0 \\ 0 & 2 & 2 \\ 4 & -1 & 0 \end{bmatrix}$$

$$|\underline{A} - \lambda \underline{I}| = \begin{vmatrix} 6-\lambda & 0 & 0 \\ 0 & 2-\lambda & 2 \\ 4 & -1 & -\lambda \end{vmatrix} = (6-\lambda)(2-\lambda)(-\lambda) - (-1)(2)(6-\lambda) = 0$$

$$(6-\lambda)(\lambda^2 - 2\lambda + 2) = 0$$

$\uparrow \qquad \qquad \uparrow$

$$\lambda_1 = 6 \qquad \lambda_{2,3} = 1 \pm i$$

As $\lambda_{2,3}$ are complex, $\underline{x}_{2,3}$ is complex.

$$\lambda_2 = 1+i : (\underline{A} - \lambda_2 \underline{I})\underline{x} = \begin{bmatrix} 6-(1+i) & 0 & 0 \\ 0 & 2-(1+i) & 0 \\ -4 & -1 & -(1+i) \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\text{rref}(\underline{A} - \lambda_2 \underline{I}) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1-i & 2 \\ 0 & 0 & 0 \end{bmatrix} \Rightarrow x_1 = 0$$

$\Rightarrow (1-i)x_2 + 2x_3 = 0$

$$\text{Set } x_2 = 1 \Rightarrow 1-i + 2x_3 = 0$$
$$\Rightarrow x_3 = -\frac{1}{2} + \frac{1}{2}i$$

$$\Rightarrow \lambda_2 = 1+i \quad \underline{x}_2 = \begin{bmatrix} 0 \\ 1 \\ -\frac{1}{2}(1+i) \end{bmatrix}$$

If A is real-valued, all eigenvectors, & eigenvalues come in complex conjugate pairs.

If $\lambda = a + ib$ w/ $A \underline{x} = \lambda \underline{x}$

$$\overline{(A \underline{x})} = \overline{(\lambda \underline{x})} \Rightarrow \underline{\bar{A}} \underline{\bar{x}} = \bar{\lambda} \underline{\bar{x}}$$

$$\underline{A} \underline{\bar{x}} = \bar{\lambda} \underline{\bar{x}}$$

$\Rightarrow \lambda = a - ib$ also eigenvalue

$$\Rightarrow \lambda_3 = 1 - i \quad \& \quad \underline{x}_3 = \begin{pmatrix} 0 \\ 1 \\ -1/2(1-i) \end{pmatrix}$$

3) Repeated Eigenvalues

$$\underline{A} = \begin{bmatrix} 3 & -1 & 2 \\ 3 & -1 & 6 \\ -2 & 2 & 2 \end{bmatrix} \quad \lambda = -4, 2, 2$$

Algebraic Multiplicity = # of times an eigenvalue shows up

$\lambda = -4$: Algebraic multiplicity of $\frac{1}{2}$

$\lambda = 2$: " " " $\frac{1}{2}$

$$\lambda = 2: \underline{A} - \lambda \underline{I} = \begin{bmatrix} 1 & -1 & 2 \\ 3 & -3 & 6 \\ -2 & 2 & 4 \end{bmatrix}$$

$$\text{rref}(A - \lambda I) = \begin{bmatrix} 1 & -1 & 2 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$\text{nullity}(A - \lambda I) = 2 \Rightarrow 2$ unique eigenvectors

$$x_1 - x_2 + 2x_3 = 0 \quad x_2 \text{ \& } x_3 \text{ are free}$$

$$x_2 = 1, x_3 = 0 \Rightarrow x_1 = 1$$

$$x_2 = 0, x_3 = 1 \Rightarrow x_1 = -2$$

$$\Rightarrow \underline{x}_2 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \quad \underline{x}_3 = \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix} \quad \lambda = 2, 2$$

Geometric Multiplicity: the # of
eigenvectors for an eigenvalue.
 $= \text{nullity}(A - \lambda I)$

In this case it's 2

If algebraic & geometric multiplicity
are equal that eigenvalue is **complete**

If not, then λ is **defective**

If any λ is defective, A is defective

$$\text{ex 1) } A = \begin{bmatrix} 3 & 1 \\ 0 & 3 \end{bmatrix} \quad \lambda = 3, 3 \quad A - \lambda I = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$$

$\text{nullity} = 1 \Rightarrow$ only 1 eigenvector \Rightarrow defective

All unique λ must be complete

Real, Symmetric Matrices

$$\underline{A}^T = \underline{A} \quad \overline{\underline{A}} = \underline{A}$$

1) A real, symmetric matrix only has
real eigenvalues,

let \underline{A} be real & symmetric

let λ be an eigenvalue | even complex.

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

\underline{A} is real $\Rightarrow \bar{\lambda} \neq \bar{\underline{x}}$ are an eigenpair

$$\underline{A} \underline{x} = \lambda \underline{x}$$
$$\underline{\bar{x}}^T \underline{A} \underline{x} = \lambda \underline{\bar{x}}^T \underline{x}$$

$$\underline{A} \underline{\bar{x}} = \bar{\lambda} \underline{\bar{x}}$$
$$(\underline{A} \underline{\bar{x}})^T = (\bar{\lambda} \underline{\bar{x}})^T$$
$$\underline{\bar{x}}^T \underline{A}^T = \bar{\lambda} \underline{\bar{x}}^T$$
$$\underline{\bar{x}}^T \underline{A} = \bar{\lambda} \underline{\bar{x}}^T \quad \downarrow$$
$$\underline{\bar{x}}^T \underline{A} \underline{x} = \bar{\lambda} \underline{\bar{x}}^T \underline{x}$$

$$\lambda \underline{\bar{x}}^T \underline{x} = \underline{\bar{x}}^T \underline{A} \underline{x} = \bar{\lambda} \underline{\bar{x}}^T \underline{x}$$
$$\lambda \underline{\bar{x}}^T \underline{x} = \bar{\lambda} \underline{\bar{x}}^T \underline{x}$$

$$\underline{\bar{x}}^T \underline{x} = |x_1|^2 + |x_2|^2 + \dots > 0$$

$\lambda = \bar{\lambda}$ < only true if λ is real

2) All eigenvectors of a real symmetric matrix are orthogonal for individual λ

$$\text{let } Ax = \lambda_1 x \quad Ay = \lambda_2 y \quad \lambda_1 \neq \lambda_2 \\ \lambda_1, \lambda_2 \text{ are real}$$

$$(\lambda_1 x)^T y = (Ax)^T y = x^T A^T y = x^T Ay = x^T (\lambda_2 y) = x^T (\lambda_2 y)$$

$$\Rightarrow \lambda_1 x^T y = \lambda_2 x^T y \\ \lambda_1 \neq \lambda_2$$

$$\Rightarrow x^T y = 0 \text{ must hold}$$

$$\Rightarrow x \perp y$$

might not hold for algebraic multiplicity > 1

Matrix Diagonalization

Given A find a P such that

$$P^{-1}AP = \Lambda \quad (\Lambda \text{ Diagonal matrix})$$

$$A = P \Lambda P^{-1}$$

$$\text{let } Ax = \lambda x$$

let λ_i be an eigenvalue of A

$$\Rightarrow Ax_1 = \lambda_1 x_1 \rightarrow Ax_p = \lambda_p x_p \text{ if } A \in M_{nn} \\ \text{max of } p = n$$

Introduce $\underline{S} = [\underline{x}_1 \quad \underline{x}_2 \quad \dots \quad \underline{x}_p]$

If \underline{A} is complete, then $n = p$
+ all eigenvectors are unique,
 $\Rightarrow \text{rank}(\underline{S}) = n \Rightarrow \underline{S}^{-1}$ exists

$$\underline{A} \underline{x}_1 = \lambda_1 \underline{x}_1 \rightarrow \underline{A} \underline{x}_n = \lambda_n \underline{x}_n$$

$$\underline{A} \underline{S} = \underline{A} \underline{S}$$

$$\underline{A} [\underline{x}_1 \quad \underline{x}_2 \quad \dots \quad \underline{x}_n] = \begin{bmatrix} \lambda_1 & & & 0 \\ & \lambda_2 & & \\ & & \ddots & \\ 0 & & & \lambda_n \end{bmatrix} [\underline{x}_1 \quad \underline{x}_2 \quad \dots \quad \underline{x}_n]$$

$$\Rightarrow \text{Crt } \underline{A} \underline{S} = \underline{A} \underline{S} \Rightarrow \underline{A} \underline{S} = \underline{S} \underline{\Lambda}$$

$$\Rightarrow \underline{A} = \underline{S} \underline{\Lambda} \underline{S}^{-1}$$

Called an **Eigendecomposition**

Need \underline{A} to be complete.

If \underline{A} is defective, then $\text{rank}(\underline{S}) < n$
 \underline{S}^{-1} does not exist \Rightarrow
No eigendecomposition

Markov Chains

let \underline{A} = transition matrix that is complete

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

\underline{P}_0 = initial probability vector

$$\underline{P}_1 = \underline{A} \underline{P}_0 \Rightarrow \underline{P}_n = \underline{A}^n \underline{P}_0$$

$$\begin{aligned} \underline{A}^2 &= \underline{A} \underline{A} = (\underline{S} \underline{\Lambda} \underline{S}^{-1})(\underline{S} \underline{\Lambda} \underline{S}^{-1}) \\ &= \underline{S} \underline{\Lambda}^2 \underline{S}^{-1} \end{aligned}$$

$$\underline{A}^2 = \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_n \end{bmatrix} \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_n \end{bmatrix} = \begin{bmatrix} \lambda_1^2 & 0 \\ 0 & \lambda_n^2 \end{bmatrix}$$

$$\underline{A}^n = \underline{S} \underline{\Lambda}^n \underline{S}^{-1}$$

$$\Rightarrow \underline{P}_n = \underline{S} \underline{\Lambda}^n \underline{S}^{-1}$$

If $|\lambda_i| < 1$ for all eigenvalues, $\underline{P}_\infty = \underline{0}$

A stochastic Markov process will have exactly one eigenvalue of 1 with all others $|\lambda_i| < 1$

$$\Rightarrow \underline{A}^\infty = \underline{S} \underline{A}^\infty \underline{S}^{-1} = \underline{S} \begin{bmatrix} 1 & & \\ & \lambda_2^\infty & \\ & 0 & \lambda_n^\infty \end{bmatrix} \underline{S}^{-1}$$

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

$\Rightarrow \underline{P}_\infty$ related to eigenvectors