

Extra page with
A few
REMARKS on spectral thm.

① Restated: $A = A^T$ real symm.

$$\iff A = P^T \Delta P \text{ for } P \text{ orthogonal } (P^{-1} = P^T)$$

and Δ real diagonal

\implies
Thm
proven

$$\begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} \quad (\text{so } \Delta = \Delta^T)$$

\longleftarrow
note

$$(P^T A P)^T = P^T \Delta^T (P^T)^T \\ = P^T \Delta P$$

② Similar proof strategy via induction on n would have proven ...

THM
(Hoffman
& Kunze
§ 8.5)

A $n \times n$ \mathbb{C} -matrix is normal, meaning $A(\bar{A}^T) = (\bar{A}^T)A$,

$$\iff A = U^T \Delta U \text{ for } U \text{ unitary } (U^{-1} = \bar{U}^T)$$

and Δ complex diagonal

$$\begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} \quad \lambda_i \in \mathbb{C}$$

③ (Fairly) Similar strategy would prove ...

THM Every A $n \times n$ \mathbb{C} -matrix can be triangularized by

an invertible matrix P over \mathbb{C} , i.e. $P^{-1}AP = \begin{bmatrix} \lambda_1 & * & \dots & * \\ & \lambda_2 & & * \\ & & \ddots & * \\ 0 & & & \lambda_n \end{bmatrix}$

... but in fact it's worth looking up a statement and/or quick proof of the Jordan canonical form for A , which is a much more precise triangular form for A .

④ Spectral thm is closely related to singular value decomposition (SVD)
(important for principal component analysis - PCA)

of a rectangular real $X = P \Sigma Q$

$\begin{matrix} m \times n & m \times m & m \times n & n \times n \\ \text{orthogonal} & \text{diagonal} & \text{orthogonal} & \\ P^T = P^{-1} & & Q^T = Q^{-1} & \end{matrix}$

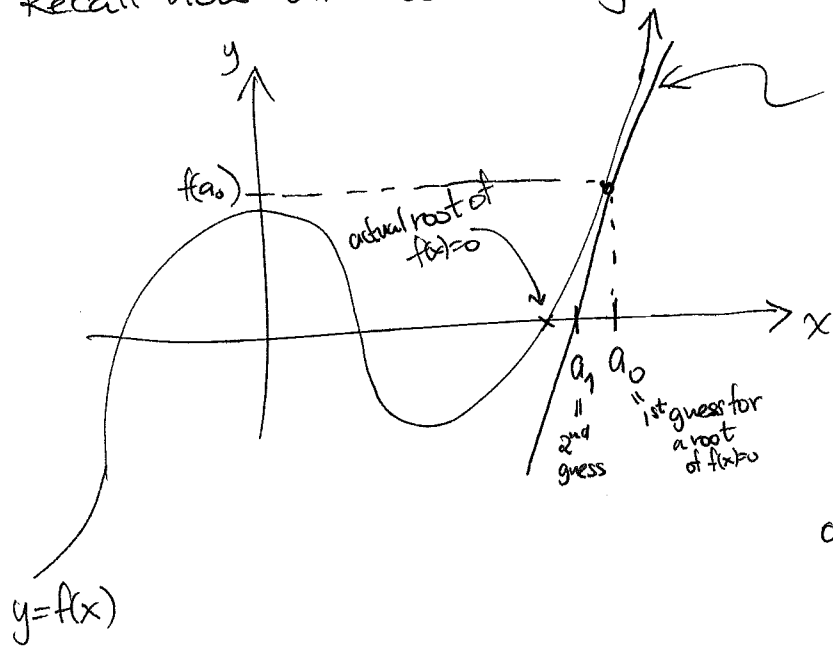
$$= \begin{bmatrix} \sigma_1 & & & 0 \\ & \ddots & & \\ & & \sigma_r & \\ 0 & & & 0 \end{bmatrix}$$

by applying spectral thm to
 $A_1 = X^T X = Q \Sigma^T \Sigma Q$
or $A_2 = X X^T = P \Sigma \Sigma^T P$

1/28/2016 → Back to curvy (non-linear) things...

(108) §2.8 Newton's method

Recall how this root-finding method works in one variable:



linear approximation to $y=f(x)$ at $x=a_0$ has equation
 $y - f(a_0) = f'(a_0)(x - a_0)$
 so solve for $y=0$ here to get the x -value a_1 for the approximate root:

$$0 - f(a_0) = f'(a_0)(x - a_0)$$

$$x - a_0 = -f'(a_0)^{-1} f(a_0)$$

$$x = \underbrace{-f'(a_0)^{-1} f(a_0)} + a_0$$

i.e. let a_1 be this.

Now repeat to get a_2, a_3, \dots

Note that we needed $f'(a_0) \neq 0$ so that we could divide by it. The multivariate version is analogous.

DEFIN 2.8.1 (multivariate Newton's method)

When looking for a solution to n equations in n unknowns $\bar{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$

$$\begin{cases} f_1(\bar{x}) = 0 \\ f_2(\bar{x}) = 0 \\ \vdots \\ f_n(\bar{x}) = 0 \end{cases}$$

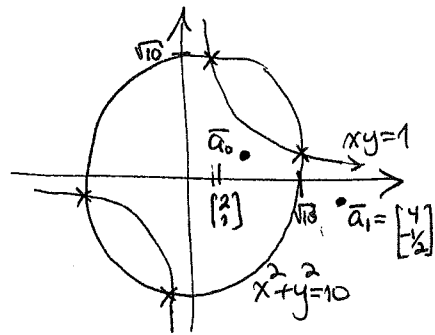
if we regard $\bar{f} = \begin{bmatrix} f_1 \\ \vdots \\ f_n \end{bmatrix}$ as a map $f: \bigcap_{\mathbb{R}^n} \mathbb{R}^n \rightarrow \mathbb{R}^n$, and it is differentiable

at some $\bar{a}_0 \in U$ with $D\bar{f}(\bar{a}_0)$ invertible, we can try to approximate a solution by instead solving the linear system $\bar{y} - \bar{f}(\bar{a}_0) = D\bar{f}(\bar{a}_0)(\bar{x} - \bar{a}_0)$

for $\bar{y} = \bar{0}$, i.e. find \bar{x} such that $D\bar{f}(\bar{a}_0)(\bar{x} - \bar{a}_0) = -\bar{f}(\bar{a}_0)$

(or equivalently $\bar{a}_1 = -D\bar{f}(\bar{a}_0)^{-1} \bar{f}(\bar{a}_0) + \bar{a}_0$)

(109) EXAMPLE: Where do the circle $x^2 + y^2 = 10$ and hyperbola $xy = 1$ intersect?



(We could solve this directly via $y = \frac{1}{x}$ and substituting $x^2 + (\frac{1}{x})^2 = 10$

$$x^4 + 1 = 10x^2 \quad \leftarrow \text{mult. by } x^2$$

$$x^4 - 10x^2 + 1 = 0$$

$$x^2 = \frac{10 \pm \sqrt{10^2 - 4}}{2}$$

$$x = \pm \sqrt{\frac{10 \pm \sqrt{96}}{2}}$$

from computer

$$\approx \pm (3.14626, 0.317837), \pm (0.317837, 3.14626)$$

Now let's try Newton's method for solving

$$\begin{cases} 0 = f_1(x,y) = x^2 + y^2 - 10 \\ 0 = f_2(x,y) = xy - 1 \end{cases}$$

so we have $\mathbb{R}^2 \xrightarrow{F} \mathbb{R}^2$

$$\bar{x} = \begin{pmatrix} x \\ y \end{pmatrix} \mapsto F(\bar{x}) = \begin{pmatrix} f_1(\bar{x}) \\ f_2(\bar{x}) \end{pmatrix} = \begin{pmatrix} x^2 + y^2 - 10 \\ xy - 1 \end{pmatrix}$$

with derivative $\mathbb{R}^2 \xrightarrow{DF(\bar{a})} \mathbb{R}^2$

$$\begin{pmatrix} x \\ y \end{pmatrix} \mapsto \underbrace{\begin{pmatrix} 2x & 2y \\ y & x \end{pmatrix}}_{\text{Jacobian } JF(\bar{a})} \begin{pmatrix} x \\ y \end{pmatrix} \quad \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} a_1 \\ a_2 \end{pmatrix}$$

As an initial guess, if we try $\bar{a}_0 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$, we have $DF(\bar{a}_0) = \begin{bmatrix} 2 & 2 \\ 1 & 1 \end{bmatrix}$ not invertible, can't use Newton!

If we try $\bar{a}_0 = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$, we have $DF(\bar{a}_0) = \begin{bmatrix} 4 & 2 \\ 1 & 2 \end{bmatrix}$ invertible,

and next guess $\bar{a}_1 = -DF(\bar{a}_0)^{-1} F(\bar{a}_0) + \bar{a}_0 = -\begin{bmatrix} 4 & 2 \\ 1 & 2 \end{bmatrix}^{-1} \begin{bmatrix} 2^2 + 1^2 - 10 \\ 2 \cdot 1 - 1 \end{bmatrix} + \begin{pmatrix} 2 \\ 1 \end{pmatrix} = \begin{pmatrix} 4 \\ -1/2 \end{pmatrix}$

then $\bar{a}_2 \approx \begin{pmatrix} 3.3 \\ 0.16 \end{pmatrix}$, $\bar{a}_3 \approx \begin{pmatrix} 3.154 \\ 0.31 \end{pmatrix}$, $\bar{a}_4 \approx \begin{pmatrix} 3.14629 \\ 0.317817 \end{pmatrix}$, $\bar{a}_5 \approx \begin{pmatrix} 3.14626 \\ 0.317837 \end{pmatrix}$ (same as above!)