

$$A \in \mathbb{R}^{r \times n}, \quad A^T = A$$

Last time:

$$R_A(\vec{u}) = \frac{(A\vec{u}) \cdot (\vec{u})}{\vec{u} \cdot \vec{u}} \quad \text{Rayleigh quotient}$$

Rem: $(f/g)' = \frac{1}{g^2} (f'g - fg')$

Compare $\alpha_0, \beta_0 \neq 0$ (calculus)

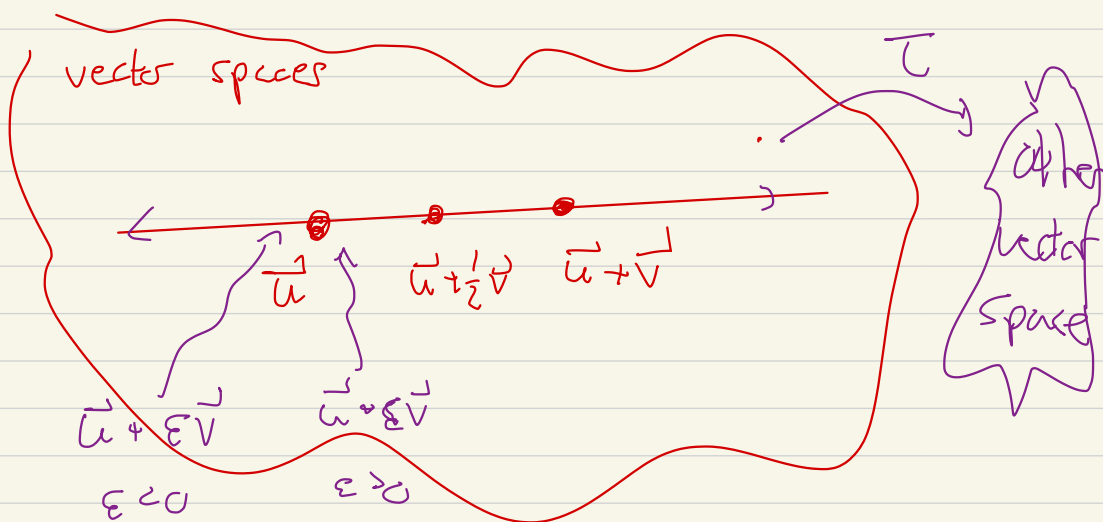
$$\frac{\alpha_0 + \varepsilon \alpha_1}{\beta_0 + \varepsilon \beta_1} = \alpha_0 \left(1 + \frac{\varepsilon \alpha_1}{\alpha_0} \right) \frac{1}{\beta_0} \left(\frac{1}{1 + \varepsilon \beta_1 / \beta_0} \right)$$

$$= \frac{\alpha_0}{\beta_0} \left(1 + \varepsilon \left(\frac{\alpha_1}{\alpha_0} - \frac{\beta_1}{\beta_0} \right) + O(\varepsilon^2) \right)$$

Andrew Gleason (never got a PhD —
super macho...)

Claim: When working with —
vector spaces, Banach spaces, etc
you only work with lines in
these spaces:

Variational methods



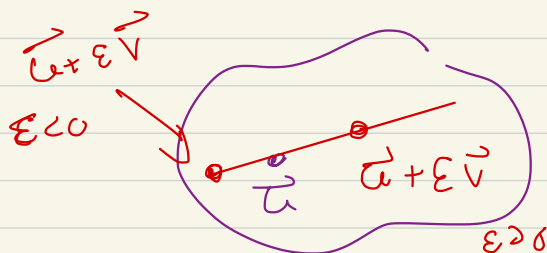
all you're doing

$$h(\epsilon) = \mathcal{T}(\vec{u} + \epsilon\vec{v})$$

~ Variational method ~

or $\begin{matrix} \text{global min} \\ \text{local min} \end{matrix}$

If $\vec{u}$ is a maximum, $\overbrace{\text{local maximum}}$



$\rightarrow \mathbb{R}$

OR
saddle

then $\forall \vec{v}$, $h(\epsilon) = \tau(\vec{u} + \epsilon \vec{v})$,

$h'(\epsilon)$ must be zero

Say:

$$\tau_{\mathbb{R}}(\vec{v} + \epsilon \vec{u}) = \frac{f(\epsilon)}{g(\epsilon)}$$

where $f(\epsilon) = \alpha_0 + \epsilon \alpha_1 + O(\epsilon^2)$

$g(\epsilon) = \beta_0 + \epsilon \beta_1 + O(\epsilon^2)$

Assume $\frac{f(\varepsilon)}{g(\varepsilon)}$ has a local

max at $\varepsilon = 0$ ---, $g(0) \neq 0$

$g(\varepsilon)$ differentiable near $\varepsilon = 0$

$\Rightarrow \left(\frac{f(\varepsilon)}{g(\varepsilon)} \right)' \Big|_{\varepsilon=0}$ must be 0

or ---

$$\nabla \left(\frac{f}{g} \right) = 0$$

↑
as appropriate

$$\mathcal{R}_A(\cdot) : \mathbb{R}^h \rightarrow \mathbb{R}$$

really $\mathbb{R}^n \setminus \{0\} \rightarrow \mathbb{R}$

only a

$$\sum_{i=1}^{n-1}$$

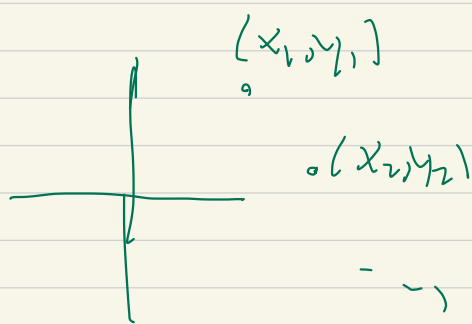
we should talk about

Solve $A\vec{u} = \vec{b}$

①

$$= \|A\vec{u} - \vec{b}\|^2$$

$$= (A\vec{u} - \vec{b}) \cdot (A\vec{u} - \vec{b})$$



Model

$$y = u_1 + u_2 x^2$$

$$+ u_3 \sin(x)$$

$$(x_m, y_m)$$

Want smallest value of

$$y_1 = u_1 + u_2 x_1^2 + u_3 \sin(x_1)$$

$$y_2 = u_1 + u_2 x_2^2 + u_3 \sin(x_2)$$

$\vdots$

$$y_m = u_1 + u_2 x_m^2 + u_3 \sin(x_m)$$

$\vec{b} = \begin{bmatrix} y_1 \\ \vdots \\ y_m \end{bmatrix}$ model $\approx \underbrace{\begin{bmatrix} 1 & x_1^2 & \sin(x_1) \\ 1 & x_2^2 & \sin(x_2) \\ \vdots & \vdots & \vdots \\ 1 & x_m^2 & \sin(x_m) \end{bmatrix}}_A \underbrace{\begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix}}_{\vec{u}}$

For the best solution $\vec{u}^*$

$$J(\vec{u}^*) = \left\| A \vec{u}^* - \vec{b} \right\|_{L^2}^2$$

we look at

$$J(\vec{u} + \varepsilon \vec{v})$$

$$= \left(A(\vec{u} + \varepsilon \vec{v}) - \vec{b} \right)$$

$$= \left(A(\vec{u} + \varepsilon \vec{v}) - \vec{b} \right)$$

$$\left(\text{Implicit in } \mathcal{P}_N \left(\vec{u} + \varepsilon \vec{v} \right) \right)$$

denominator

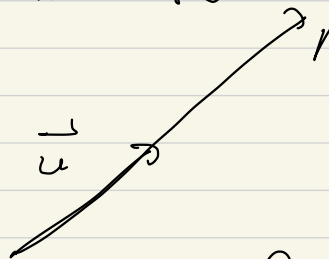
Our ultimate for eigenpairs of
a symmetric matrix A ,
is to show that if

$$\vec{u} \text{ maximizes } \mathcal{R}_A(\vec{u}) \stackrel{\text{def}}{=} \frac{(A\vec{u}) \cdot \vec{u}}{\vec{u} \cdot \vec{u}}$$

then $\forall \vec{v}$ s.t. $\vec{v} \perp \vec{u}$

we have also $\vec{v} \perp A\vec{u}$

and in this case


$$A\vec{u} = \lambda \vec{u}$$

for some λ

(λ can be 0)

Lemme 1: If $A \in \mathbb{R}^{m \times n}$
matrix, $\vec{u}, \vec{v} \in \mathbb{R}^n$, $\vec{b} \in \mathbb{R}^m$

$$J(\vec{u}) = \|A\vec{u} - \vec{b}\|_{L^2}^2$$

$$= (A\vec{u} - \vec{b}) \cdot (A\vec{u} - \vec{b}).$$

Then

$$J(\vec{u} + \varepsilon \vec{v}) =$$

$$\varepsilon^0 \text{ Term}_0 + \varepsilon^1 \text{ Term}_1$$

$$+ \underbrace{O(\varepsilon^2)}_{\text{order } \varepsilon^2}$$

where $\text{Term}_1 = ???$

$$\mathcal{L}(\vec{u} + \epsilon \vec{v})$$

$$= (A(\vec{u} + \epsilon \vec{v}) - \vec{b}) \cdot$$

$$(A(\vec{u} + \epsilon \vec{v}) + \vec{b})$$

$$= (A\vec{u} - \vec{b}) + \epsilon A(\vec{v}) \cdot$$

$$(A\vec{u} - \vec{b}) + \epsilon A(\vec{v})$$

$\text{Term}_0 = \text{whatever}$

$$\underline{\mathcal{E} \cdot \text{Term}_1} = \mathcal{E} \left(A \vec{v} \right) \cdot \left(A \vec{u} - \vec{b} \right)$$

$$+ \mathcal{E} \left(A \vec{u} - \vec{b} \right) \cdot \left(A \vec{v} \right)$$

$$= \underline{2\mathcal{E}} \left(A \vec{v} \right) \cdot \left(A \vec{u} - \vec{b} \right)$$

$$= 2\mathcal{E} \left(\vec{v} \cdot \underbrace{A^T (A \vec{u} - \vec{b})}_{\text{"normal equations"}} \right)$$

Lemma:

$$\frac{\alpha_0 + \varepsilon \alpha_1 + o(\varepsilon)}{\beta_0 + \varepsilon \beta_1 + o(\varepsilon)} \quad \text{or} \quad \frac{\alpha_0 + \varepsilon \alpha_1 + O(\varepsilon^2)}{\beta_0 + \varepsilon \beta_1 + O(\varepsilon^2)}$$

$$= \text{Term}_0 + \varepsilon \text{Term}_1 + \begin{cases} o(\varepsilon) \\ O(\varepsilon) \end{cases}$$

where

$$\text{Term}_1 = 0 \iff \frac{\alpha_1}{\alpha_0} = \frac{\beta_1}{\beta_0}$$
