

March 16, 2021 CPSC 531 F

- Today:

Variational Methods:

① Rayleigh quotients to diagonalize self-adjoint operators

② SVD (singular-value decomposition)

③ Calculus of variations and other variational methods

Not part of course, "∞-dimensional" vector spaces

Our context!

- \mathbb{R}^n , $(\vec{u}, \vec{v}) = u_1 v_1 + u_2 v_2 + \dots + u_n v_n$

- More generally

this is mostly enough $\left\{ \begin{array}{l} (\vec{u}, \vec{v})_{\vec{w}} = u_1 v_1 w_1 + \dots + u_n v_n w_n \\ \vec{w} - \text{weighted dot product} \end{array} \right.$

More generally yet,

enough \rightarrow any inner product $(,)$ on \mathbb{R}^h

- More generally yet

any inner product $(,)$ on

finite dim

vector space

- All this works in inner products \langle , \rangle

over \mathbb{P} .

\equiv

Weighted inner product!

$$\mathcal{L}: \mathbb{R}^3 \longrightarrow \mathbb{R}^3$$

$$\vec{v} \longmapsto A\vec{v}, \quad A \in M_3(\mathbb{R})$$

IF $\mathbb{R}^3 =$ usual space in which we live,

$$\vec{v} = \begin{bmatrix} 3 \text{ cm} \\ 4 \text{ cm} \\ 2 \text{ cm} \end{bmatrix}, \quad \vec{v} = \begin{bmatrix} 3 \text{ ft} \\ 4 \text{ ft} \\ 5 \text{ ft} \end{bmatrix}$$

but imagine

$$\vec{v} = \begin{bmatrix} 3 \text{ cm} \\ 2 \text{ km} \\ 5 \text{ ft} \end{bmatrix} \dots \text{ (sad face) }$$

For use, weighted inner products

come from $(,)_{\pi}$ } π is
 $(,)_{1/\pi}$ } stationary
dist
of MC

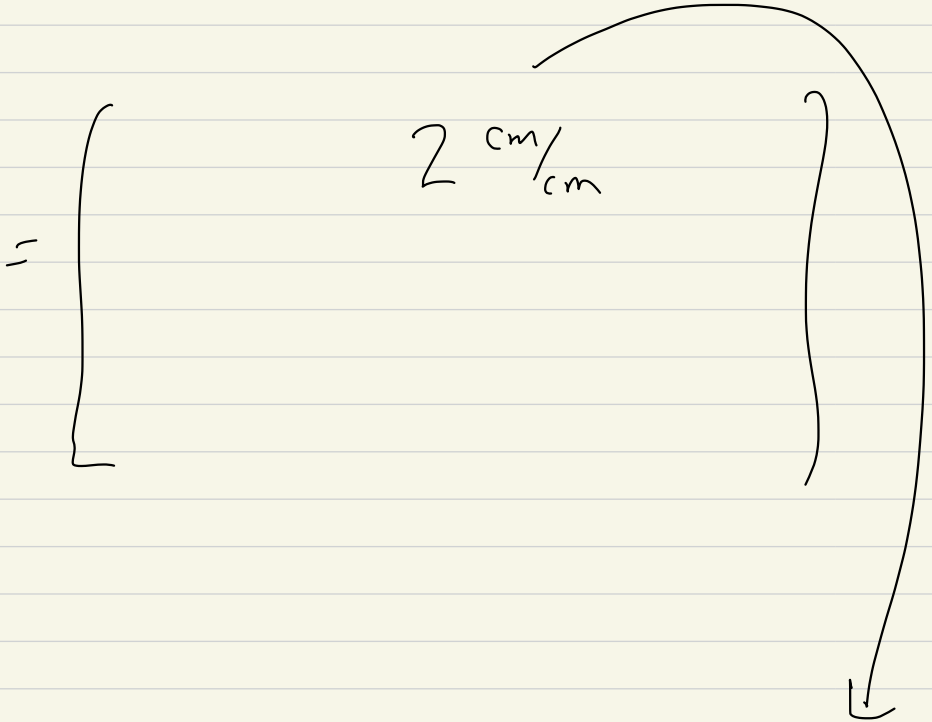
=

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 5 & 6 \\ 3 & 6 & 9 \end{bmatrix},$$

$$L \begin{bmatrix} 3 \text{ cm} \\ 2 \text{ km} \\ 5 \text{ ft} \end{bmatrix} = \begin{bmatrix} ? \text{ cm} \\ ? \text{ light years} \\ ? \text{ } \mu\text{m} \end{bmatrix}$$



$$A = \begin{bmatrix} 1 \text{ cm/cm} & 2 \text{ cm/cm} & \dots \\ & - & \vdots \\ & - & - \end{bmatrix} \dots$$



$$2 \frac{\text{cm}}{\text{cm}} = \frac{.02}{1000} \frac{\text{cm}}{\text{km}}$$



$$L \begin{bmatrix} 3 \text{ cm} \\ 2 \text{ km} \\ 5 \text{ ft} \end{bmatrix} = \begin{bmatrix} ? \text{ cm} \\ ? \text{ km} \\ ? \text{ ft} \end{bmatrix}$$



$$\begin{bmatrix} d_1 \frac{\mu\text{m}}{\text{cm}} & 0 & 0 \\ 0 & d_2 \frac{\mu\text{m}}{\text{km}} & 0 \\ 0 & 0 & d_3 \frac{\mu\text{m}}{\text{ft}} \end{bmatrix} \begin{bmatrix} 3 \text{ cm} \\ 2 \text{ km} \\ 5 \text{ ft} \end{bmatrix}$$

then \rightarrow

then \rightarrow invert



Simpler: Inner Product!

$$\left(\begin{array}{l} \begin{bmatrix} 3 \text{ cm} \\ 2 \text{ km} \\ 5 \text{ ft} \end{bmatrix}, \begin{bmatrix} -2 \text{ cm} \\ 4 \text{ km} \\ 19 \text{ ft} \end{bmatrix} \end{array} \right)$$

= gives everything in same fixed units

→ weighted dot product

$$\begin{bmatrix} x_1 \text{ m} \\ x_2 \text{ km} \end{bmatrix} \cdot \begin{bmatrix} y_1 \text{ m} \\ y_2 \text{ km} \end{bmatrix} = \left(x_1 y_1 + x_2 y_2 \cdot 10^6 \right) \text{ m}$$

↑
weights

① $L: \mathbb{R}^n \rightarrow \mathbb{R}^n$, (\cdot, \cdot)
inner product

Assume L is self-adjoint:

$$\forall \vec{x}, \vec{y} \in \mathbb{R}^n \quad (L\vec{x}, \vec{y}) = (\vec{x}, L\vec{y})$$

(intuition comes from special case:
 $(\vec{x}, \vec{y}) = \vec{x} \cdot \vec{y}$, $L(\vec{v}) \mapsto A\vec{v}$
with $A = A^T$ symmetric)

Thm: Assuming the above, L has
an orthonormal basis of eigenvectors
 $\vec{v}_1, \dots, \vec{v}_n$, $L(\vec{v}_i) = \lambda_i \vec{v}_i$, and
 $\lambda_i \in \mathbb{R}$.

Proof!

$$R_L(\vec{v}) = \frac{(L\vec{v}, \vec{v})}{(\vec{v}, \vec{v})}$$

- defined for $\vec{v} \neq \vec{0}$

- $R_L(\vec{v}) = R_L(\alpha\vec{v})$ for any
 $\vec{v} \neq \vec{0}$ and $\alpha \neq 0$.

Claim: $R_L(\vec{v})$ has its maximum

value somewhere. Idea: suffices to

consider

$$S' = \left\{ \vec{v} \mid \|\vec{v}\|_2 = \sqrt{v_1^2 + \dots + v_n^2} = 1 \right\}$$

(the unit sphere). Then using (Math 320)

sequential compactness to prove this.

{ $\geq 100-150$ years ago, did this }
without proof

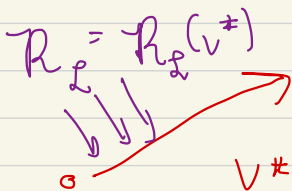
Assume this: we have \vec{v}^* s.t.,

$$\max_{\vec{v} \neq 0} R_L(\vec{v}) = R_L(\vec{v}^*).$$

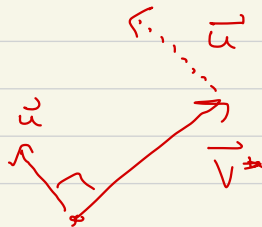
Claim! \vec{v}^* is an eigenvector

$$\text{of } L! \quad L(\vec{v}^*) = \lambda \vec{v}^*.$$

$R_L = R_L(\vec{v}^*)$



v^*



Take any $\vec{u} \perp \vec{v}^*$

$$(\vec{u}, \vec{v}^*) = 0$$

Look at

$$\vec{v}_\varepsilon = \vec{v}^* + \varepsilon \vec{u} \quad \left. \vphantom{\vec{v}_\varepsilon} \right\} \text{"variations"}$$

think of ε near 0:

$$R_L(\vec{v}^*) \geq R_L(\vec{v}_\varepsilon) \quad \begin{array}{l} \text{for} \\ \text{all} \\ \varepsilon \end{array}$$

We consider !

$$g(\varepsilon) = \mathcal{R}_\varepsilon(\vec{v}_\varepsilon) = \frac{(\mathcal{L}\vec{v}_\varepsilon, \vec{v}_\varepsilon)}{(\vec{v}_\varepsilon, \vec{v}_\varepsilon)}$$

$$g(\varepsilon) \leq g(0) = \mathcal{R}_\varepsilon(\vec{v}^*) \text{ for all } \varepsilon.$$


hence

$$g'(0) = 0$$

now

$$\frac{d}{d\varepsilon} g(\varepsilon) = \frac{d}{d\varepsilon}$$

$$= \frac{(\vec{v}_\varepsilon, \vec{v}_\varepsilon) \left(\frac{d}{d\varepsilon} (\mathcal{L}\vec{v}_\varepsilon, \vec{v}_\varepsilon) \right) - \text{other way}}{(\vec{v}_\varepsilon, \vec{v}_\varepsilon)^2}$$


$$(\vec{V}_\epsilon, \vec{V}_\epsilon) =$$

$$(\vec{V}^* + \epsilon \vec{u}, \vec{V}^* + \epsilon \vec{u}) =$$

$$(\vec{V}^*, \vec{V}^*) + 2\epsilon (\vec{V}^*, \vec{u}) + \epsilon^2 (\vec{u}, \vec{u})$$

0

$$= \underbrace{(\vec{V}^*, \vec{V}^*)}_{\text{non-zero}} + \text{order } \epsilon^2 \text{ term}$$

$$\frac{1}{(\vec{V}_\epsilon, \vec{V}_\epsilon)} = \frac{1}{\underbrace{(\vec{V}^*, \vec{V}^*)}_{\gamma_0} + \text{order}(\epsilon^2) + \gamma_1 \epsilon^2}$$

$$= \frac{1}{(\vec{V}^*, \vec{V}^*)} + \text{order}(\varepsilon^2)$$

really $\frac{1}{\gamma_0 + \gamma_1 \varepsilon^2} = \frac{1}{\gamma_0} \frac{1}{1 + (\gamma_1 \varepsilon^2 / \gamma_0)}$

$$= \frac{1}{\gamma_0} \left(1 - \frac{\gamma_1}{\gamma_0} \varepsilon^2 + \left(\frac{\gamma_1}{\gamma_0} \varepsilon^2 \right)^2 - \dots \right)$$

ε small,

bounded by $C \varepsilon^2$

$$\left(\mathcal{L} \vec{V}_\varepsilon, \vec{V}_\varepsilon \right) =$$

$$\left(\mathcal{L}(\vec{V}^* + \varepsilon \vec{u}), \vec{V}^* + \varepsilon \vec{u} \right)$$

$$= (\mathcal{L} \vec{v}^*, \vec{v}^*) + 2\varepsilon (\mathcal{L} \vec{v}^*, \vec{u}^*)$$

$$+ O(\varepsilon^2)$$

Something bounded
by $\text{const} \cdot \varepsilon^2$

\Rightarrow

$$R_{\mathcal{L}}(\vec{v}_{\varepsilon}) = \frac{(\mathcal{L} \vec{v}_{\varepsilon}, \vec{v}_{\varepsilon})}{(\vec{v}_{\varepsilon}, \vec{v}_{\varepsilon})}$$

$$= \frac{1}{(\vec{v}_{\varepsilon}, \vec{v}_{\varepsilon})} (\mathcal{L}(\vec{v}_{\varepsilon}), \vec{v}_{\varepsilon})$$

$$= \left(\frac{1}{(\vec{v}^*, \vec{v}^*)} + O(\varepsilon^2) \right)$$

$$\left(\mathcal{L}(\vec{v}^*, \vec{v}^*) + 2\varepsilon \mathcal{L}(\vec{v}^*, \vec{u}^*) + O(\varepsilon^2) \right)$$

$$= \frac{\mathcal{L}(\vec{v}^*), \vec{v}^*)}{(\vec{v}^*, \vec{v}^*)} + 2\varepsilon \frac{\mathcal{L}(\vec{v}^*, \vec{u}^*)}{(\vec{v}^*, \vec{v}^*)} + O(\varepsilon^2)$$

In terms of

$$g(\varepsilon) = \mathcal{R}_{\mathcal{L}}(\vec{v}_\varepsilon), \quad g'(0)$$

$$= 2 \frac{\mathcal{L}(\vec{v}^*), \vec{u}^*)}{(\vec{v}^*, \vec{v}^*)}$$

Something positive

Hence, for any $\vec{u} \perp \vec{v}^*$

$(\vec{u}, \vec{v}^*) = 0$ we have

$$\mathcal{L}(\vec{v}^*) \perp \vec{u}$$

$$(\mathcal{L}(\vec{v}^*), \vec{u}) = 0.$$

Hence

$\mathcal{L}(\vec{v}^*)$ is proportional to
 \vec{v}^*

$$\mathcal{L}(\vec{v}^*) = \text{Const} \cdot \vec{v}^* = \lambda \vec{v}^*$$

Why? --

Next steps:

We can use the same idea to

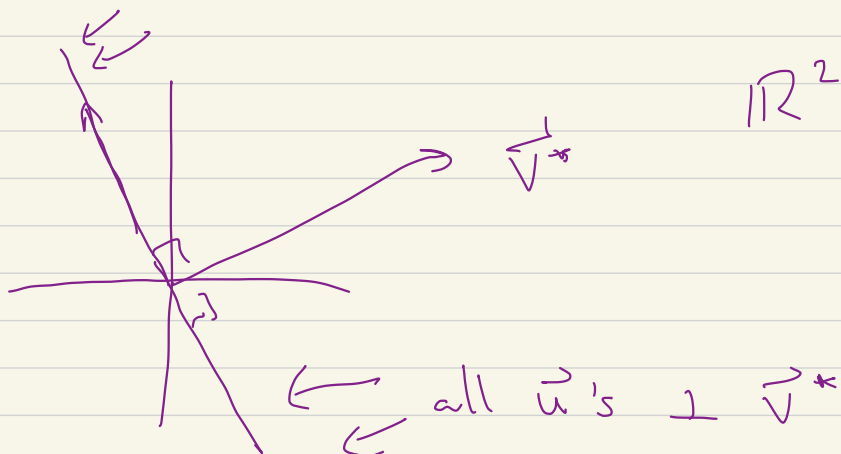
(1) get an entire orthonormal
eigenbasis for L

(2) Same set of ideas
give SVD.

4 minute break

Exercise...

Intuition for dot products



if $\vec{w} \perp \left(\begin{array}{l} \text{all } \vec{u}'\text{s} \\ \perp \text{ to } \vec{v}^* \end{array} \right)$

$$\Rightarrow \vec{w} = c \cdot \vec{v}^*$$



Everything with \circ products goes through for $(,)$ inner products.

Rem! $\mathcal{L}(\vec{u}) = \mu \vec{u}$ $\left\{ \begin{array}{l} \vec{u} \neq \vec{0} \\ \vec{u} \in \mathbb{R}^n, \\ \mu \in \mathbb{R} \end{array} \right.$

then

$$\eta_{\mathcal{L}}(\vec{u}) = \frac{(\mathcal{L}(\vec{u}), \vec{u})}{(\vec{u}, \vec{u})}$$

$$= \frac{(\mu \vec{u}, \vec{u})}{(\vec{u}, \vec{u})} = \mu$$

$$\mu \text{ eigenvalue} = \eta_{\mathcal{L}}(\vec{u})$$

We know

$$R_{\mathcal{L}}(\vec{v}^*) = \lambda \quad \text{with}$$

$$\mathcal{L}(\vec{v}^*) = \lambda \vec{v}^*$$

and

$$\lambda = R_{\mathcal{L}}(\vec{v}^*) = \max R_{\mathcal{L}}$$

at least as big
as a

$\Rightarrow \lambda$ is the
largest eigenvalue

of \mathcal{L}

(s.t.
 $\mathcal{L}(\vec{u}) = \mu \vec{u}$
for some $\vec{u} \neq 0$)

New set

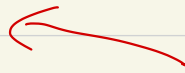
$$\vec{v}_1 = \vec{v}^*, \quad \lambda_1 = \lambda$$

$$= \mathcal{R}_L(\vec{v}^*)$$

Consider

$$\max \mathcal{R}_L(\vec{v})$$

$$\left\{ \begin{array}{l} \vec{v} \neq \vec{0} \text{ and} \\ \vec{v} \perp \vec{v}_1 \end{array} \right\}$$



clever
trick

If this max is attained at

$$\vec{v}^{**} = \vec{v}_2, \text{ i.e.}$$

$$\vec{v}_2 \perp \vec{v}_1, \quad \vec{v}_2 \neq 0, \text{ and}$$

$$R_L(\vec{v}_2) \geq R_L(\vec{v}) \text{ for}$$

$$\text{any } \vec{v} \perp \vec{v}_1,$$

Same variational argument shows that if $\vec{u} \perp$ both \vec{v}_1, \vec{v}_2 , then

$$\vec{u} \perp R(\vec{v}_2)$$

$$\Rightarrow \mathcal{L}(\vec{v}_2) = c_1 \vec{v}_1 + c_2 \vec{v}_2$$

$$\vec{v}_2 \perp \vec{v}_1 = c_2 \vec{v}_2$$

we write

$$\lambda_2 \vec{v}_2$$

So

$$\mathcal{L}(\vec{v}_1) = \lambda_1 \vec{v}_1$$

$$\mathcal{L}(\vec{v}_2) = \lambda_2 \vec{v}_2$$

$\vec{v}_2 \perp \vec{v}_1$. Now let \vec{v}_3 be

the vector $\perp \vec{v}_1, \vec{v}_2$ where

$$\max \quad R_L(\vec{v})$$

$$\vec{v} \neq \vec{0}$$

$$\vec{v} \perp \vec{v}_1, \vec{v}_2$$

is attained ...

Claim!

SVD comes from almost

the same argument

Class ends

Idea!

$$f(x_1, x_2) = x_1^2 + x_2^2, \quad x_1, x_2 \in \mathbb{R}$$

where is the min

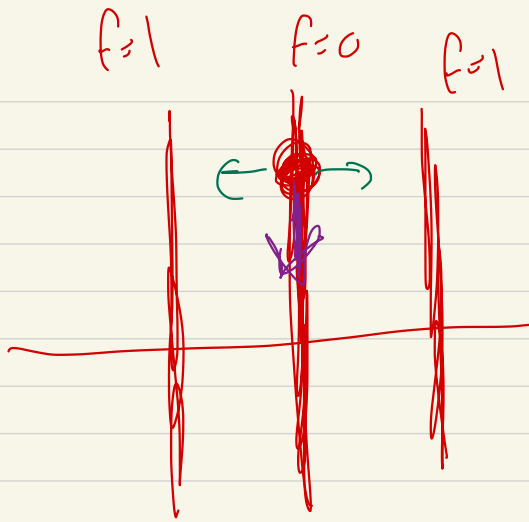
$$\nabla f = (2x_1, 2x_2)$$

$$\nabla f = \vec{0} \Rightarrow x_1 = x_2 = 0$$

$$f(x_1, x_2) = x_1^2$$

$$\nabla f = (2x_1, 0)$$

$$\nabla f = \vec{0} \Rightarrow x_1 = 0, \quad x_2 \text{ anything}$$

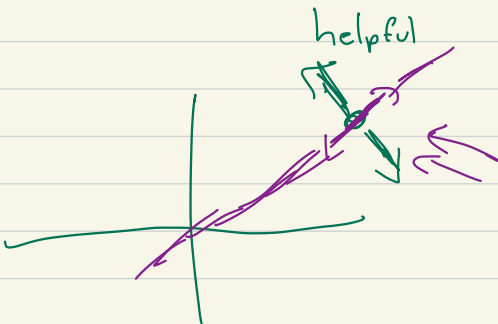


$$f(x_1, x_2) = x_1^2$$

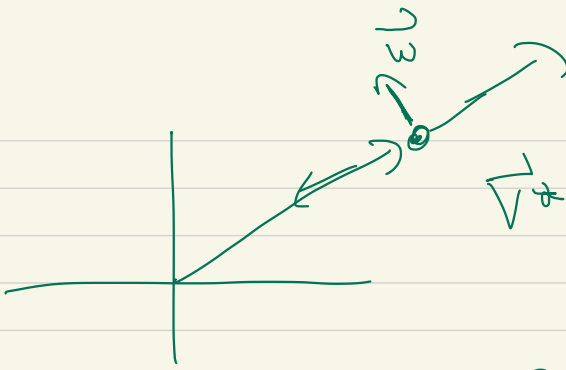
↓ not helpful

↔ helpful

$$R_{\begin{bmatrix} 2 \\ 1 \end{bmatrix}}(x_1, x_2) = \frac{2x_1^2 + x_2^2}{x_1^2 + x_2^2}$$



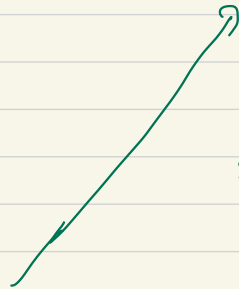
$R = \text{constant}$



$$R_{\epsilon}(v^*)$$

$$= \cancel{R_{\epsilon}}(\alpha v^*)$$

$$v \approx v^* + \epsilon w$$



$$= \epsilon w$$

$$v \approx v^* + \epsilon w$$

$$\epsilon w \perp v^*$$

$$\left(\vec{v}_\varepsilon, \vec{v}_\varepsilon \right)$$

$$= \left(\vec{v}^*, \vec{v}^* \right) + 2\varepsilon \underbrace{\left(\vec{v}^*, \vec{u} \right)} + \left(\vec{u}, \vec{u} \right) \varepsilon^2$$

$$\text{if } \vec{u} = 3\vec{v}^* + \vec{w}$$

$\vec{w} \perp \vec{v}^*$

$$\left(\vec{v}^*, \vec{u} \right) = \left(\vec{v}^*, 3\vec{v}^* + \vec{w} \right)$$

$$= 3 \left(\vec{v}^*, \vec{v}^* \right) + 0$$

$$\neq 0$$

EXERCISE

Solve Heat Eq $\frac{\partial}{\partial t} u + \Delta u = 0$

Wave Eq

$$\left(\frac{\partial}{\partial t}\right)^2 u = \Delta u$$

Lapl Eq

$$\Delta u = 0, \text{ no time}$$

To solve $\vec{u}: \mathbb{R} \rightarrow \mathbb{R}^n, \vec{u} = \vec{u}(t)$

$$\frac{d}{dt} u = Au \Rightarrow$$

want to

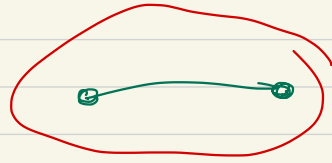
$$\vec{u}(t) = e^{At} \vec{u}(0)$$

find eigenpairs of A

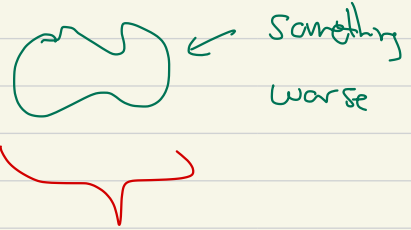
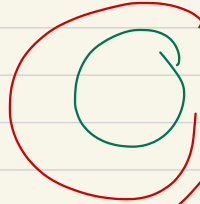
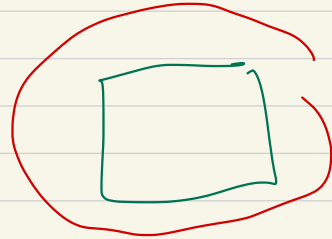
Want to eigenpairs of Δu

Wave
Heat

1-dim



2-dim



Can
explicitly
find eigenpairs

Can't
generally

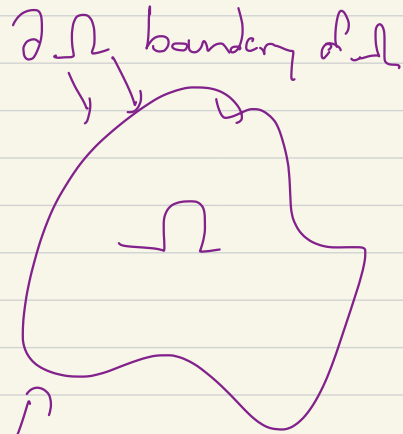
but eigenpairs of Δ can be found by $\left\{ \begin{array}{l} \mathbb{R} = (Df, f) \\ \rightarrow \Delta (f, f) \end{array} \right.$

$$\geq \frac{(\nabla f, \nabla f)}{(f, f)}$$

minimize

$$\frac{(\nabla f, \nabla f)}{(f, f)}$$

over



$$f \text{ s.t. } f|_{\partial\Omega} = 0$$

gives lowest

eigenfunction } of $-\Delta$
eigenvalue }

wave equation \Rightarrow becoming all the eigenvalues of $-\Delta$