

Calculus of Variations

$$f(x) \quad f(x, y) \quad f(\underline{x}) \quad x \in \mathbb{R}^d$$

plug \underline{x} get a real number

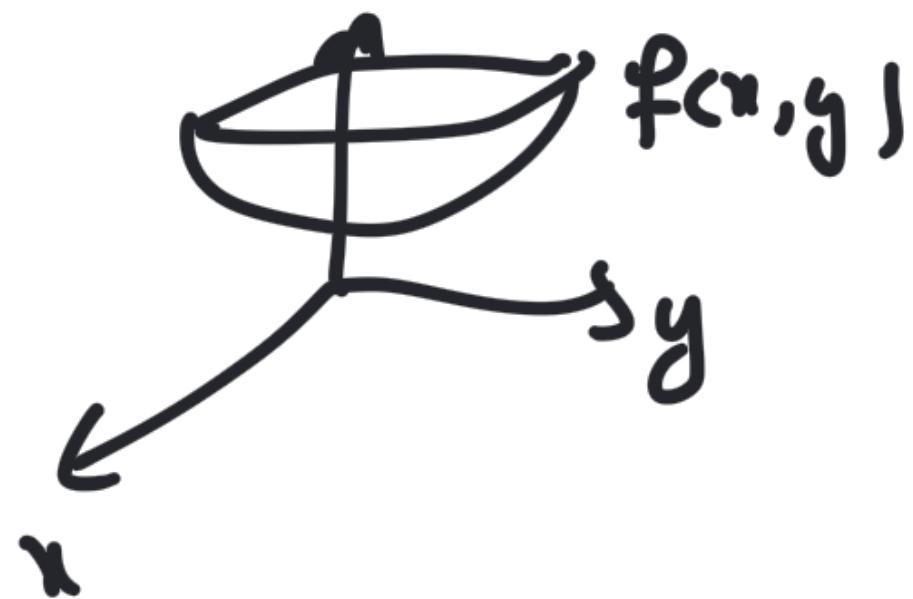
e.g. $f(5, 7, 2) = 25$

$$f: \mathbb{R}^d \rightarrow \mathbb{R}$$

graph: $f(x)$



$$f(x, y)$$



limitation in graphing.

$$\cdot f: \mathbb{R}^d \rightarrow \mathbb{R} \quad f(x_1, x_2, \dots, x_d)$$

$$\nabla f = \left(\frac{\partial f}{\partial x_1}, \frac{\partial f}{\partial x_2}, \dots, \frac{\partial f}{\partial x_d} \right)$$

$$\frac{\partial f}{\partial n} = \nabla f \cdot \underline{n} = \lim_{h \rightarrow 0} \frac{f(\underline{x} + h\underline{n}) - f(\underline{x})}{h}$$

directional derivative

Calculus on functionals

functional: function of a function

$E(u)$

\uparrow
fn that lives in some fn space

e.g. $u \in$ Cont. fns on \mathbb{R} $C(\mathbb{R})$

\in diff. fns on \mathbb{R} $C'(\mathbb{R})$

\in smooth fns on \mathbb{R} $C^\infty(\mathbb{R})$

$E: \text{fn space} \rightarrow \mathbb{R}$

Compare with a fn:

$f: \mathbb{R}^d \rightarrow \mathbb{R}$

Example: • $f: \mathbb{R}^2 \rightarrow \mathbb{R}$

$$f(x,y) = x^2 + y^2 \quad \text{so} \quad f(0,1) = 0^2 + 1^2 = 1$$

• $\Sigma: C^1(\mathbb{R}) \rightarrow \mathbb{R}$

$$\Sigma(u) = \int_0^1 u'^2 + u \, dx$$

Calculate $\Sigma(x^2)$

$$\begin{aligned} \Sigma(x^2) &= \int_0^1 (2x)^2 + x^2 \, dx = \int_0^1 4x^2 + x^2 \, dx \\ &= \int_0^1 5x^2 \, dx = 5 \frac{x^3}{3} \Big|_0^1 = \frac{5}{3} \end{aligned}$$

Let's do Calculus on functionals:

Derivatives and finding critical pts

and finding minima and maxima

optimization

This is called Calculus of Variations

Calculus on fn'l's instead of Calculus
on fns!!

Given a fn'l $\Sigma(u)$, we want its minimizer

$$\min_{u \in \text{some fn space}} \Sigma(u)$$

then we must find its critical pts!!

Critical pts happen when equivalent of
"derivative of $\Sigma(u)$ " in the fn space = 0

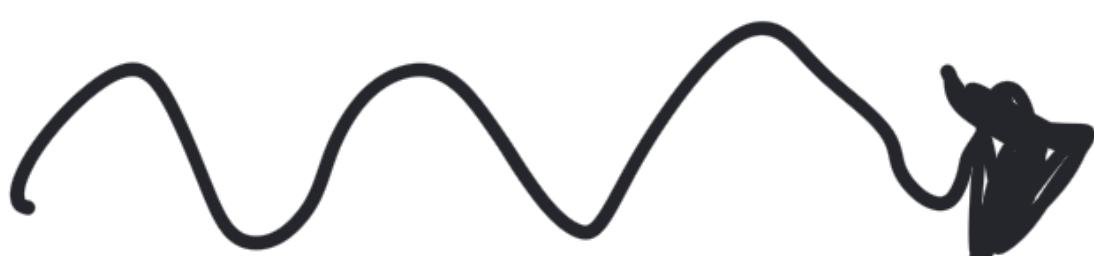
Euler-Lagrange eqn

This give us

ordinary differential eqns

and

partial differential eqns !!

Calculus  PDEs
and ODEs

1) f_n $f: \mathbb{R}^d \rightarrow \mathbb{R}$

$$\nabla f = (f_{x_1}, f_{x_2}, \dots, f_{x_d})$$

$$f = f(x_1, \dots, x_d)$$

let x_i depend on time

$$f = f(x_1(t), x_2(t), \dots, x_d(t))$$

$$\frac{df}{dt} = \frac{\partial f}{\partial x_1} \frac{dx_1}{dt} + \dots + \frac{\partial f}{\partial x_d} \frac{dx_d}{dt}$$

$$= \nabla f \cdot \frac{d}{dt} (x_1(t), x_2(t), \dots, x_d(t))$$

$$= \nabla f \cdot \dot{x}(t)$$

$$\dot{f} = \nabla f \cdot \dot{x}(t)$$

2) f_n' $E: f_n \text{ space} \rightarrow \mathbb{R}$

usually called an energy

Prb: find $\min_{u \in f_n \text{ space}} E(u)$

on some domain

If there is a minimizer, it will happen at the critical pt of $E(u)$ in the f_n space.

Critical pts happen at pts where

$$\nabla E = 0$$

fn space

have to define $\nabla_{fn \text{ space}}$

new: gradient
in a fn space

We know ∇f when $f: \mathbb{R}^d \rightarrow \mathbb{R}$

We also know $\frac{\partial f}{\partial n} = \nabla f \cdot \underline{n}$

↑
directional derivative

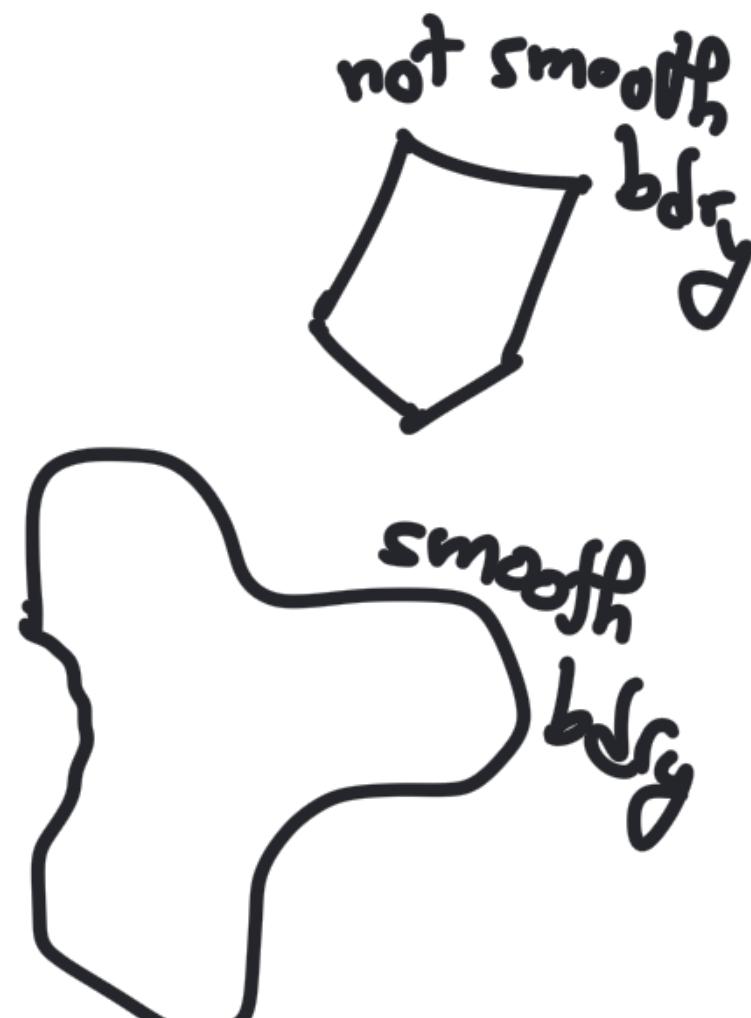
Example: Consider (famous) the Dirichlet

energy fn'l:

$$E(u) = \int_D \frac{1}{2} |\nabla u|^2 dx$$

D is a domain in \mathbb{R}^d

∂D is the bdry of D

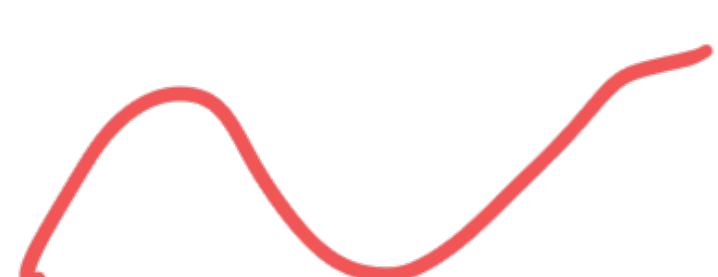


$u \in H^1(D)$ which means that

u has finite $\int_D (u^2 + |\nabla u|^2) dx$

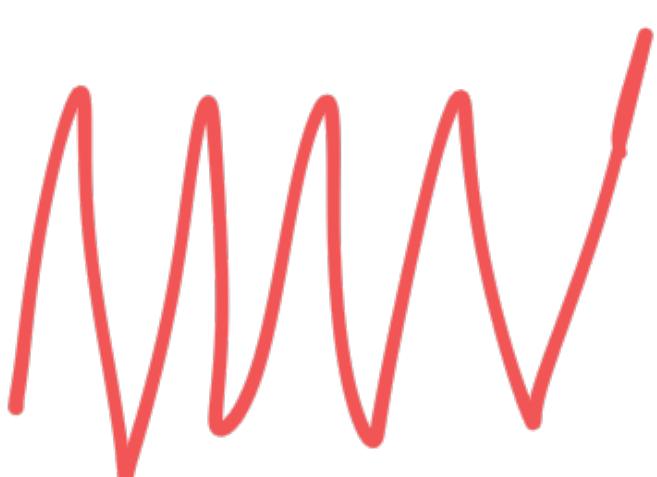
$H^1(D)$ is called a Sobolov Space.

Sobolov spaces allow a wide range of fns that are not so nice



nice fn

lives in
cont. and
continuously diff.
fn spaces



not nice fn



not nice fn

to be solns of important PDE that represent real life situations.

minimizer of the Dirichlet energy

satisfies

$$\nabla E(u) = 0$$

fn space

Euler-Lagrange eqn associated with
the energy fn'l $E(u)$
also called variational principle.

Soln: $E(u) = \int \frac{1}{2} |\nabla u|^2 dx$

$$E(u) = \int_D \frac{1}{2} (\nabla u)^2 dx$$

$$= \int_D \frac{1}{2} (\nabla u \cdot \nabla u) dx$$

$$|\nabla u|^2 = [u_1^2 + \dots + u_n^2] \\ = \nabla u \cdot \nabla u$$

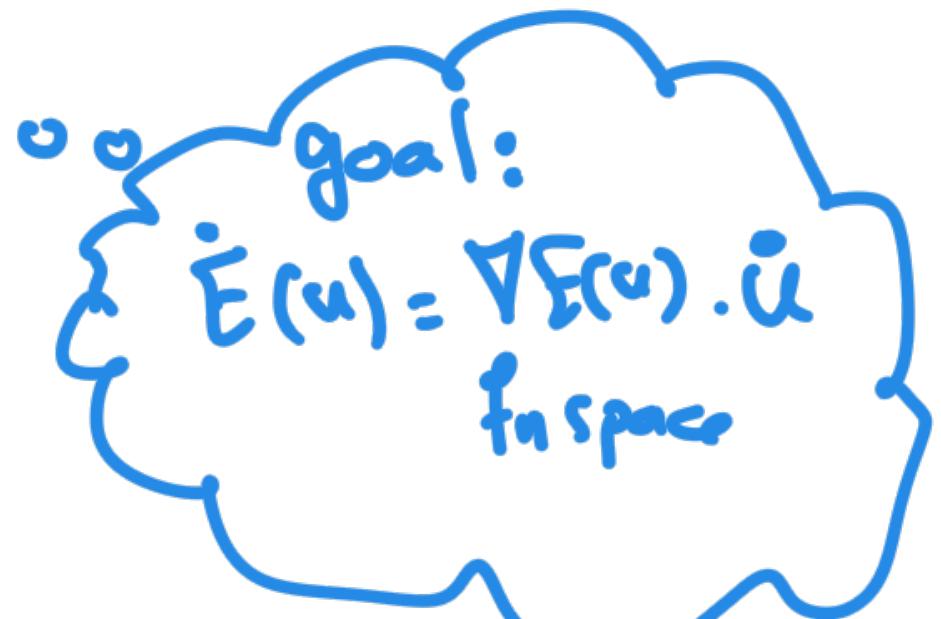
$$= \int_D \frac{1}{2} ((\nabla u) \cdot \nabla u + (\nabla u) \cdot (\nabla u)) dx$$

$$= \int_D \nabla u \cdot (\nabla u) dx$$

$$= \int_D \nabla u \cdot \nabla \bar{u} dx$$

most imp. trick in Calculus

Integration by parts!



$$\text{Calc I: } \int_a^b u v_x dx = - \int_a^b u_x v dx + uv \Big|_a^b$$

Calc III: Int. by parts Green's thm.

 follows from div. thm

$$\int_D u \nabla v dx = - \int_D \nabla u v dx + \int_{\partial D} u v \underline{n} ds$$

$$\text{Divergence thm: } \int_D \nabla \cdot \underline{f} dx = \int_{\partial D} \underline{f} \cdot \underline{n} ds$$

$$\begin{aligned} \Rightarrow \dot{E}(u) &= - \int_D \nabla \cdot (\nabla u) \bar{u} dx + \int_{\partial D} \nabla u \bar{u} \underline{n} ds \\ &= - \int_D \Delta u \bar{u} dx + \int_{\partial D} \bar{u} \nabla u \cdot \underline{n} ds \end{aligned}$$

$$\dot{E}(u) = - \int_D \Delta u \dot{u} dx + \int_{\partial D} \dot{u} \frac{\partial u}{\partial n} ds$$

true for any \dot{u} in my allowed fn space.

In particular, let $\dot{u}=0$ on ∂D

$$\Rightarrow \int_{\partial D} \dot{u} \frac{\partial u}{\partial n} ds = 0$$

and $\dot{E}(u) = - \int_D \Delta u \dot{u} dx$

$$= \langle -\Delta u, \dot{u} \rangle_{L^2(D)}$$

$$= \nabla_{L^2(D)} E(u) \cdot \dot{u}$$

$$\Rightarrow \nabla_{L^2(D)} E(u) = -\Delta u$$

minimizer satisfies F-L eqn: $\nabla_{L^2(D)} E(u) = 0$
 $\Rightarrow -\Delta u = 0$ or $\Delta u = 0$ Laplace eqn
 PDE

I defined inner product in a fn space

called L^2 :

$$\langle f, g \rangle_{L^2(D)} = \int_D f(x)g(x)dx$$

Therefore fns that minimize the
Dirichlet energy

$$E(u) = \int_D \frac{1}{2} \|u\|^2 dx$$

are harmonic fns (fns that
satisfy Laplace eqn on D)

$$\Delta u = 0 \text{ on } D \quad \leftarrow \begin{matrix} \text{EL eqn for} \\ E(u) \text{ is a} \\ \text{PDE} \end{matrix}$$

say in 2D: $u_{xx} + u_{yy} = 0$

Look up examples of harmonic fns.

Recall: Calc III

$$\nabla f = \left(\frac{\partial f}{\partial x_1}, \dots, \frac{\partial f}{\partial x_d} \right) : \text{vector}$$

$$\text{div } \underline{v} = \nabla \cdot \underline{v} = \frac{\partial v_1}{\partial x_1} + \frac{\partial v_2}{\partial x_2} + \dots + \frac{\partial v_d}{\partial x_d} : \text{scalar}$$

$$\nabla = \left(\frac{\partial}{\partial x_1}, \frac{\partial}{\partial x_2}, \dots, \frac{\partial}{\partial x_d} \right)$$

∇
as an operator

$$\underline{v} = (v_1, v_2, \dots, v_n) : \text{random vector}$$

$\nabla \cdot \underline{v}$ = dot product between operator
 ∇ and components of v

$$\nabla \cdot (\nabla u) = \frac{\partial^2}{\partial x_1^2} u_1 + \frac{\partial^2}{\partial x_2^2} u_2 + \dots + \frac{\partial^2}{\partial x_d^2} u_d \\ = \Delta u$$

In 2D:

$$\Delta u = u_{xx} + u_{yy} \quad \text{scalars}$$

$$\text{in 3D } \Delta u = u_{xx} + u_{yy} + u_{zz}$$

We proved that if

$$E(u) = \int_D \frac{1}{2} |\nabla u|^2 dx$$

then $\nabla_{L^2(D)} E(u) = -\Delta u$

Now: allow u to depend on t as well

$$u = u(x, t)$$

$$\dot{E}(u) = \left\langle \nabla_{L^2} E(u), \dot{u} \right\rangle_{L^2} = \langle -\Delta u, \dot{u} \rangle_{L^2}$$

Recall: Similar to: $f: \mathbb{R}^d \rightarrow \mathbb{R}$

$f(x)$, allow x to depend on t

$$f(x(t))$$

$$\dot{f}(x(t)) = \nabla f \cdot \dot{x}(t)$$

Don't have to take $\lim_{h \rightarrow 0} \frac{E(u+h\eta) - E(u)}{h}$

Goal: find u in appropriate fn space
with bdry cond. $u = h(x)$ on ∂D

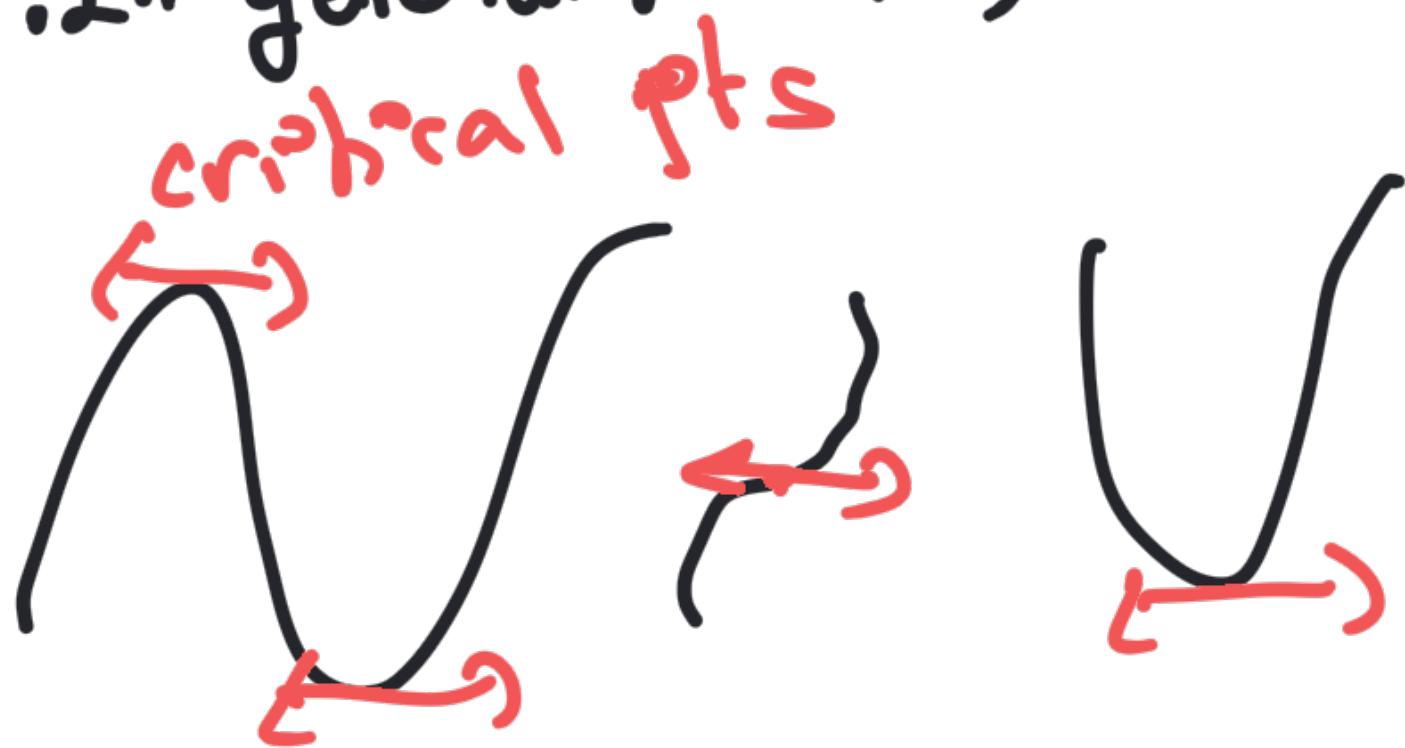
that minimizes

$$E(u) = \int_D \frac{1}{2} |\nabla u(x)|^2 dx$$

We found that a critical pt of $E(u)$
must satisfy $\begin{cases} \Delta u = 0 \text{ in } D \\ u = h(x) \text{ on } \partial D \end{cases}$ ($\nabla_{L^2} E(u) = 0$)

Harmonic fn with $h(x)$ on ∂D .

In general: $E(u)$ could look like

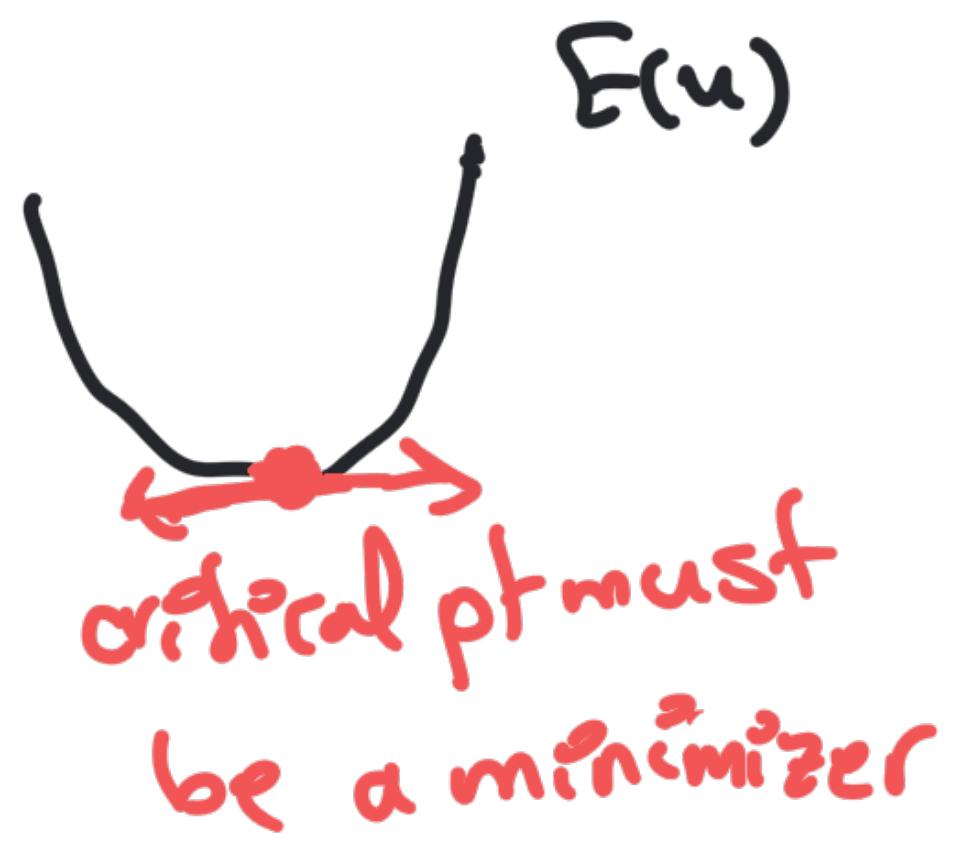


general landscapes can have many
peaks and valleys, mountain passes (saddles)
don't have to be smooth etc.

How do I know that the harmonic fn on D with $u = h(x)$ on ∂D is a minimizer of $E(u)$ not only a critical pt?

Answer: Shape of $E(u)$ forces that:

$E(u)$ is convex and $E(u)$ is bounded below by zero



Recall: $f(x) = x^2$

Convex, bd below by zero

Critical pt:

$$f'(x) = 0 \Rightarrow 2x = 0 \Rightarrow x = 0$$

$(0,0)$ is definitely a minimizer.

How does the heat eqn come into play??

Heat eqn: $u(x, t)$: usually temp. or conc. of a solute in a liquid, or conc. of gas in a room

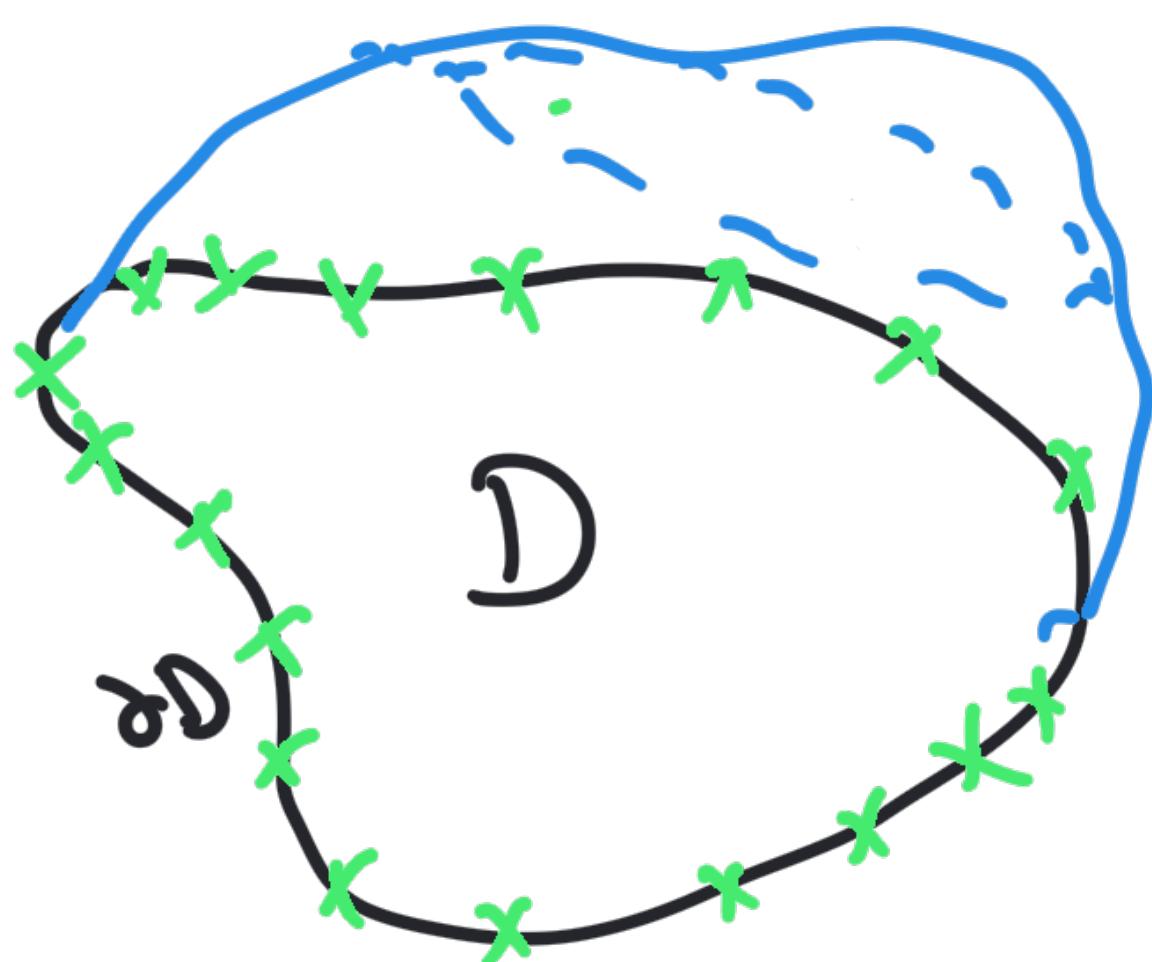
$$\left\{ \begin{array}{l} u_t = \Delta u \quad x \in D, t > 0 \\ u(x, 0) = g(x) \quad x \in D \\ u(x, t) = 0 \quad x \in \partial D \end{array} \right.$$

zero bdry cond.

Recall:

$$\Delta u(x, t) = u_{xx}(x, t) + u_{yy}(x, t)$$

in 2D.



or any quantity that diffuses

chemicals in a city (chemical weapons)

Note: So far from PDEs we have seen

1) Laplace eqn

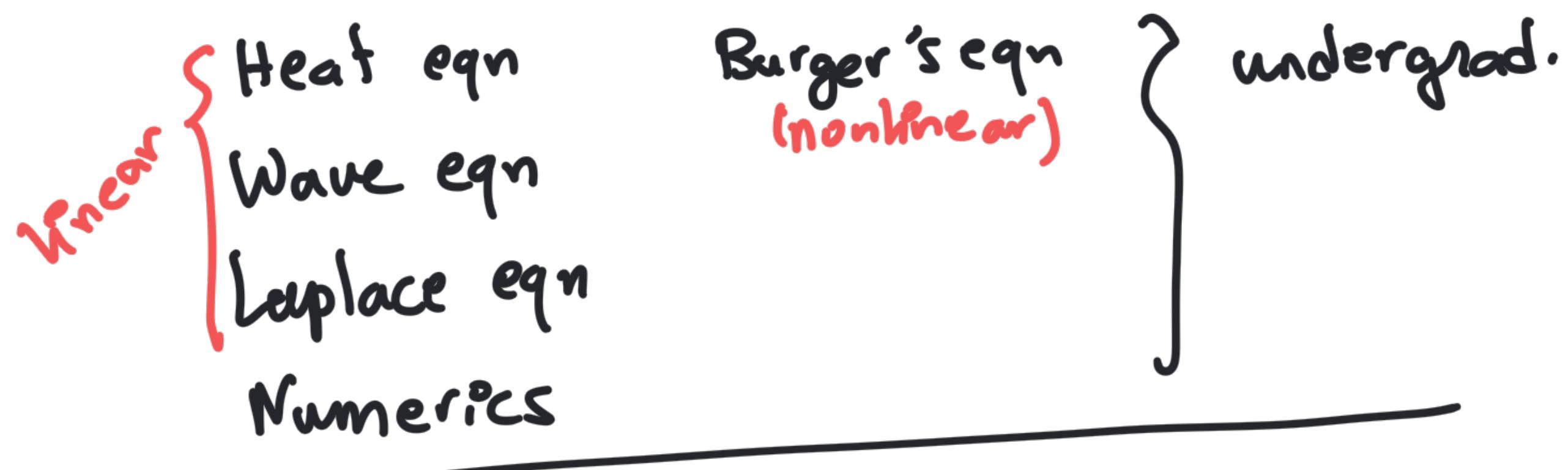
$$\left\{ \begin{array}{l} \Delta u = 0 \quad x \in D \\ u = g \quad x \in \partial D \end{array} \right.$$

2) Heat eqn

$$\left\{ \begin{array}{l} u_t = \Delta u \quad x \in D, t > 0 \\ u(x, 0) = \dots \text{Initial Cond.} \\ u(x, t) = \dots \text{B.C.} \end{array} \right.$$

Put things in perspective:

A whole PDE course is



Recall: $\Delta u = -\nabla_{L^2} E(u)$

so heat eqn:

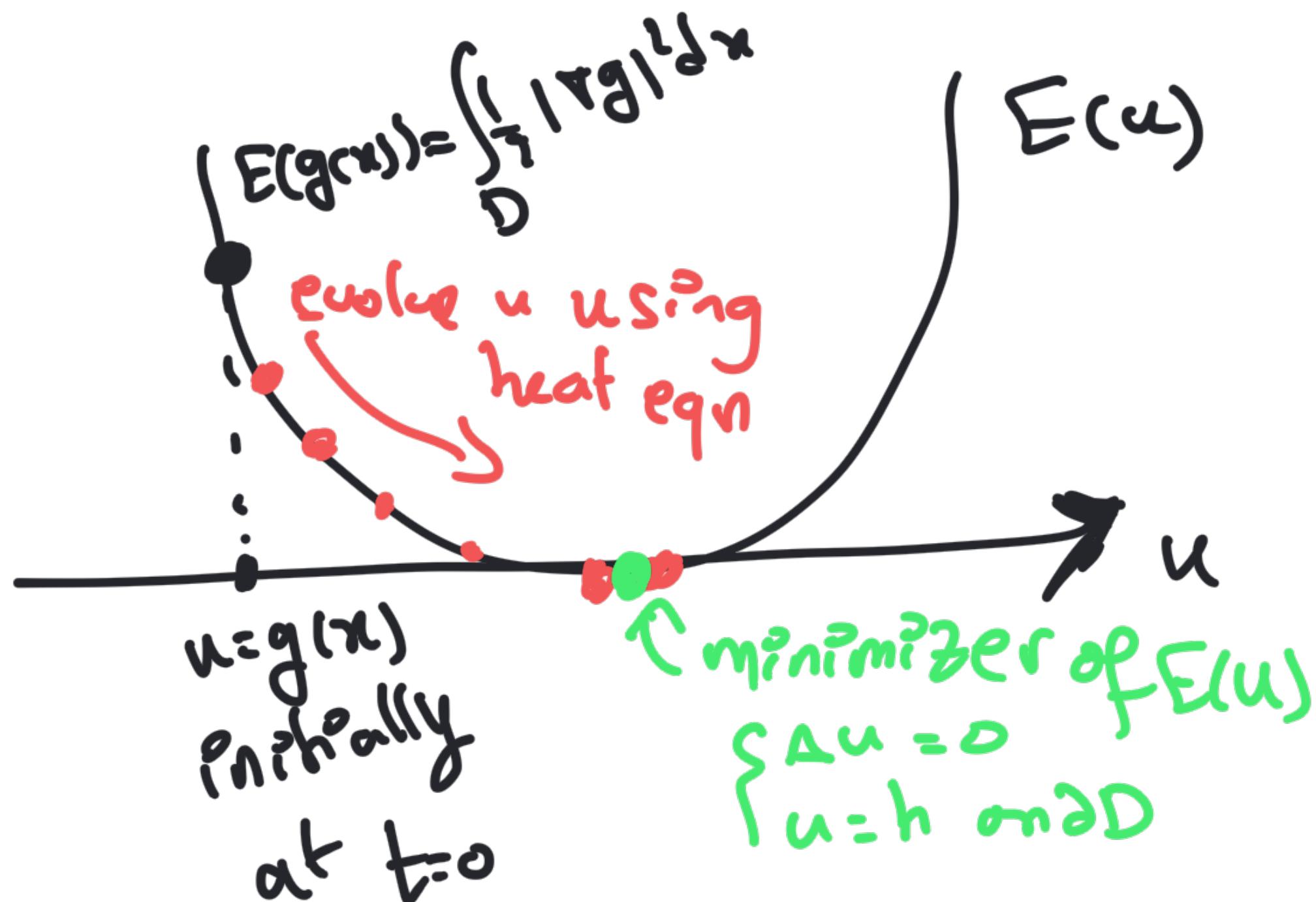
$$u_t = \Delta u = -\nabla_{L^2} E(u)$$

$$\Rightarrow u_t = -\nabla_{L^2} E(u)$$

So how does u evolve in time according to the heat eqn?

answer: $u(x,t)$ starts at $g(x)$ and then evolves in a way that decreases the Dir. Energy $E(u)$ the fastest: steepest descent direction

Concl: The heat eqn does L^2 -steepest descent for the Dir. Energy.



Happens in nature: Many systems evolve

 (you think evolution is random)
 but in reality they are decreasing
 some higher energy in the most
 efficient way!!

The heat eqn provides a minimizing scheme for the Dirichlet energy.

Start somewhere: a random fn $g(x)$, then evolve the heat eqn

(numerically if you don't have an explicit analytical soln)

So next time step you're at a new fn $g_1(x)$, keep going:

You know:

$$\rightarrow E(g_1(x)) \leq E(g(x)) \text{ & so on}$$

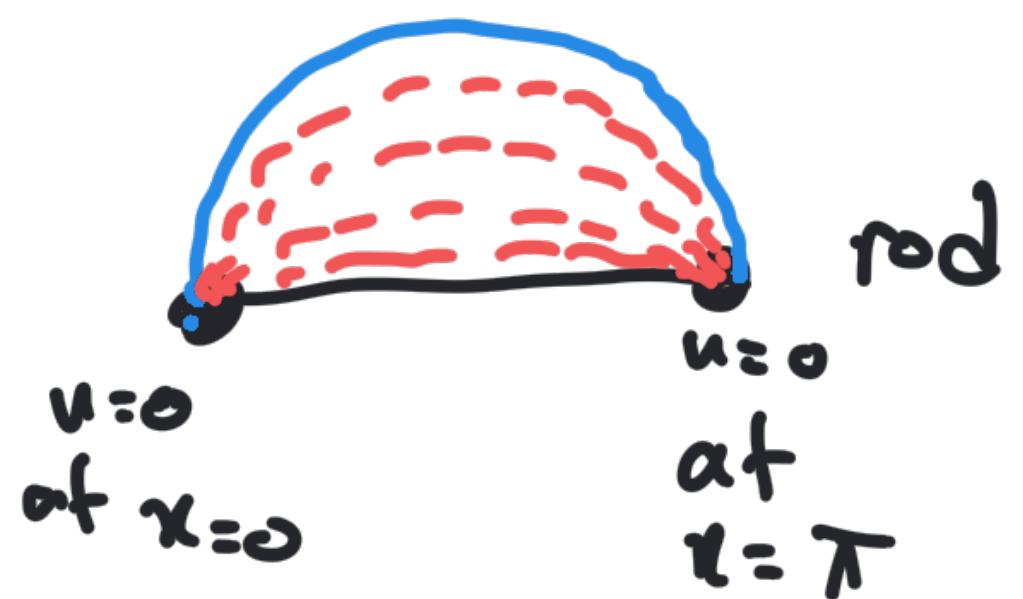
\rightarrow fastest way to decrease this energy

\rightarrow When you settle down (stationary soln), then you arrived at the minimizer of the Dir. energy in the fastest possible way.

Example Showing an analytical soln of
heat eqn on a simple domain: (1 dim.)

$$\begin{cases} u_t = u_{xx} & x \in [0, \pi], t > 0 \\ u(x, 0) = \sin x & I.C. \\ u(0) = 0, u(\pi) = 0 & B.C. \end{cases}$$

$$u(x, 0) = \sin x$$



sln: (from any PDE
course using
sep. of vars)

rod of length π
zero bdry cond's

$$u(x, t) = e^{-t} \sin x \quad ; \text{ initial } \sin x \text{ is scaled by } e^{-t}$$

(let's verify this is the soln: as $t \rightarrow \infty$ soln $\rightarrow 0$)

$$u_t = -e^{-t} \sin x$$

$$u_x = e^{-t} \cos x \quad u_{xx} = -e^{-t} \sin x$$

$$\Rightarrow u_t = u_{xx} \quad \checkmark$$

$$I.C.: u(x, 0) = e^0 \sin x = \sin x \quad \checkmark$$

$$B.C.: u(0, t) = e^{-t} \sin 0 = 0 \quad \checkmark$$

$$u(\pi, t) = e^{-t} \sin \pi = 0 \quad \checkmark$$

Note that as $t \rightarrow \infty$,

$$u(x, t) = e^{-t} \sin x \rightarrow 0 \quad (\text{as expected})$$

minimizer of Dir. energy

$$\left. \begin{aligned} &\text{satisfies } \Delta u = u_{xx} = 0 \\ &\text{and } u(0) = 0 \leq u(\pi) \end{aligned} \right\}$$

Soln of Laplace eqn with zero bdry cond's

is the zero fn

so All consistent.

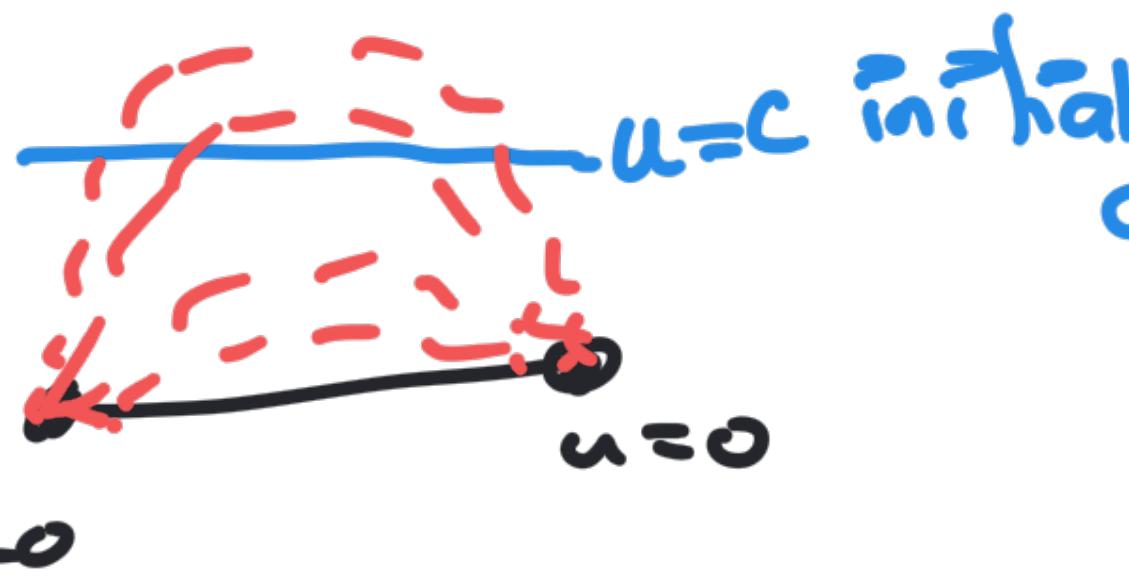
Soln of heat eqn indeed converges to the minimizer of Dir. energy as $t \rightarrow \infty$.

What happens if we start with a const. temp. throughout the rod and not the $\sin x$ temp.

from PDE class,
again from sep.
of variables,

the soln:

$$u(x,t) = \sum_{n=1}^{\infty} B_n e^{-n^2 t} \sin nx$$



exp decay of
the temp.

where the coef's B_n

$$\begin{aligned} B_n &= \frac{2}{\pi} \int_0^{\pi} c \sin nx dx \\ &= \frac{2}{\pi} c \left[-\frac{\cos nx}{n} \right]_0^{\pi} \\ &= \frac{2c}{n\pi} (1 - (-1)^n) \end{aligned}$$

Note: In general it's not easy to find energy fnl(s) that decrease with the evolution of a certain PDE (system) and when/if you find these energy fnl(s) it is very helpful for the community working on that PDE (Helps understand the evolution of the system "from above". Also provides bounded quantities that are needed for analysis).

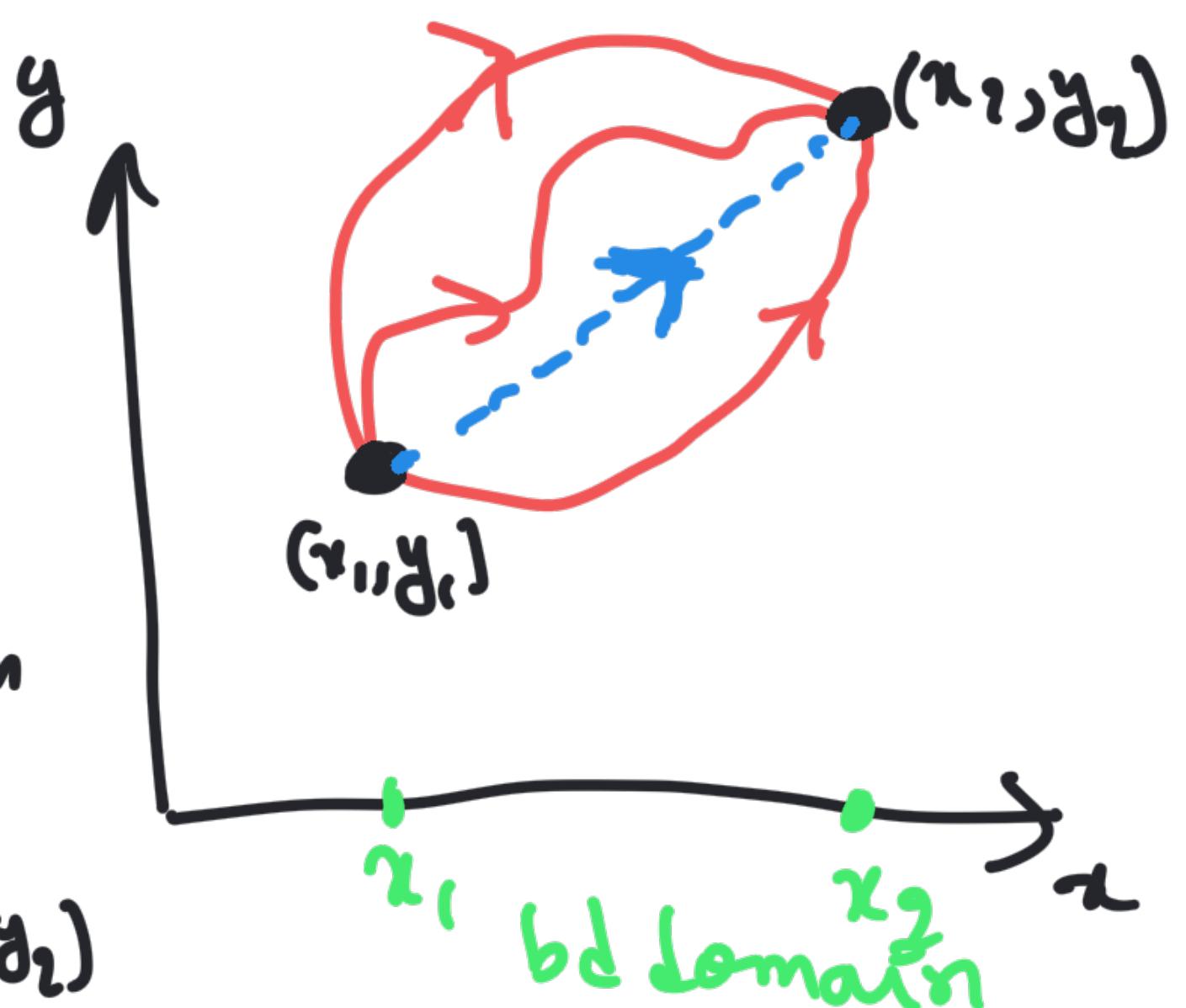
Note: Integration by parts helps find many energy fnl's!

Simplest examples from Calculus of Variations:

1) Shortest path prb:

prb:

\min length of a fn
 $y(x)$ between
 two pts
 (x_1, y_1) and (x_2, y_2)



Let $y(x)$ be a fn that passes through (x_1, y_1) and (x_2, y_2) . formula for arc length of a fn

$$\min_{y(x_1)=y_1, y(x_2)=y_2} \int_{x_1}^{x_2} \sqrt{1+y'(x)^2} dx$$

$y(x)$
 cont., one
 time
 cont. diff.

arc length

why is this
 the formula
 for an arc
 length?

$$E(y) = \int_{x_1}^{x_2} \sqrt{1 + y'(x)^2} dx$$

Want $\nabla E(y) = 0$ at a minimizer
some space candidate.

Euler-Lagrange eqn

Assume also dependence on time

$$\dot{E}(y) = \langle \nabla_{\dot{y}} E(y), \dot{y} \rangle_{\dot{y} \text{ in space}}$$

$$E(y) = \int_{x_1}^{x_2} (1 + y'^2)^{1/2} dx$$

$$\dot{E}(y) = \int_{x_1}^{x_2} \left[(1 + y'^2)^{1/2} \right]^\bullet dx$$

$$= \int_{x_1}^{x_2} \frac{1}{2} (1 + y'^2)^{\frac{1}{2}-1} (1 + y'^2)^\bullet dx$$

$$= \int_{x_1}^{x_2} \frac{1}{2 \sqrt{1+y'^2}} (0 + 2y' \dot{y}') dx$$

$$= \int_{x_1}^{x_2} \frac{1}{\sqrt{1+y'^2}} y' \dot{y}' dx$$

Recall:
 $f = f(x(t))$
 $\frac{df}{dt} = \dot{f} = \nabla f \cdot \dot{x}$
chain rule

Integration by parts so I can get \dot{y} alone inside
the integral

$$= - \int_{x_1}^{x_2} \left(\frac{\dot{y}'}{\sqrt{1+y'^2}} \right)' \dot{y} dx + \left(\frac{y}{\sqrt{1+y'^2}} \right) \dot{y} \Big|_{x_1}^{x_2}$$

Valid for all \dot{y} . Let $\dot{y} = 0$ on x_1 and x_2

This kills the boundary term.

Left with:

$$\begin{aligned} E(y) &= \int_{x_1}^{x_2} - \left(\frac{y'}{\sqrt{1+y'^2}} \right)' \dot{y} dx \\ &= \left\langle - \left(\frac{y'}{\sqrt{1+y'^2}} \right)', \dot{y} \right\rangle_{L^2([x_1, x_2])} \end{aligned}$$

$$\rightarrow \nabla_{L^2([x_1, x_2])} E(y) = - \left(\frac{y'}{\sqrt{1+y'^2}} \right)'$$

Now at a minimizer, $\nabla E(y) = 0$
 $L^2([x_1, x_2])$

$$\rightarrow \left(\frac{y'}{\sqrt{1+y'^2}} \right)' = 0 \quad \text{E-L eqn}$$

$$\Rightarrow \frac{y'}{\sqrt{1+y'^2}} = c \quad (\text{since its derivative is zero})$$

$$y' = c \sqrt{1+y'^2}$$

Square both sides

$$y'^2 = c^2(1+y'^2)$$

$$y'^2 = c^2 + c^2 y'^2$$

$$(1-c^2) y'^2 = c^2$$

$$y'^2 = \frac{c^2}{1-c^2}$$

$$y'(x) = \pm \sqrt{\frac{c^2}{1-c^2}}$$

constant

$$y'(x) = a$$

$$\Rightarrow y(x) = ax + b$$



st. line.

satisfying b.c. $y(x_1) = y_1$ and $y(x_2) = y_2$

and we just proved that shortest path is indeed

a st. line. What's the exact eqn of this minimize?

$$y - y_1 = \text{slope}(x - x_1)$$

$$y = \text{slope}(x - x_1) + y_1$$

$$y = \frac{y_2 - y_1}{x_2 - x_1} (x - x_1) + y_1$$

Note: $E(y) = \int_{x_1}^{x_2} \sqrt{1+y'^2} dx$

is actually convex and bd below by zero.

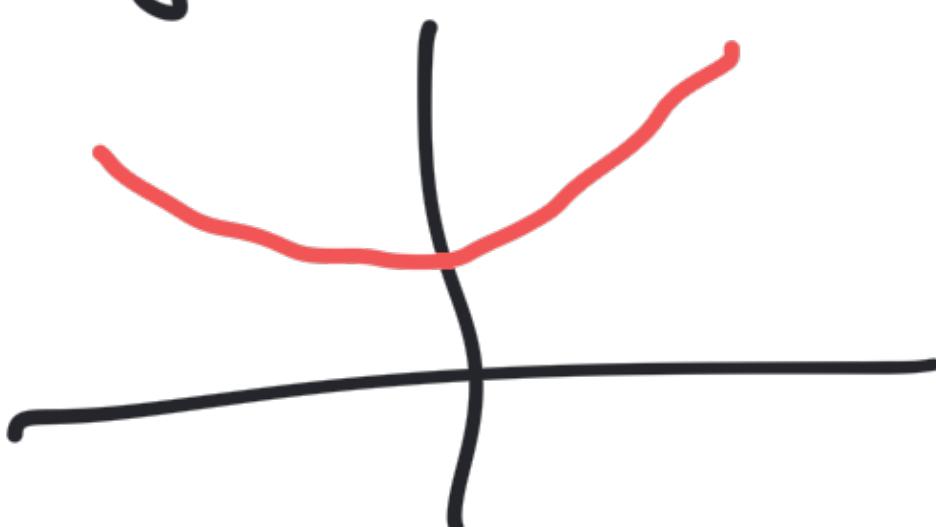
This means that the critical pt that we found: $f_n y(x)$ satisfying

$$\nabla_{L^2} E(y) = 0 \quad (\text{E-L eqn})$$

is indeed a minimizer.

Convexity of $E(y)$ follows from the convexity of the calculus fn that's in the integrand:

$$g(s) = \sqrt{1+s^2}$$



Good news

Review from Calc III:

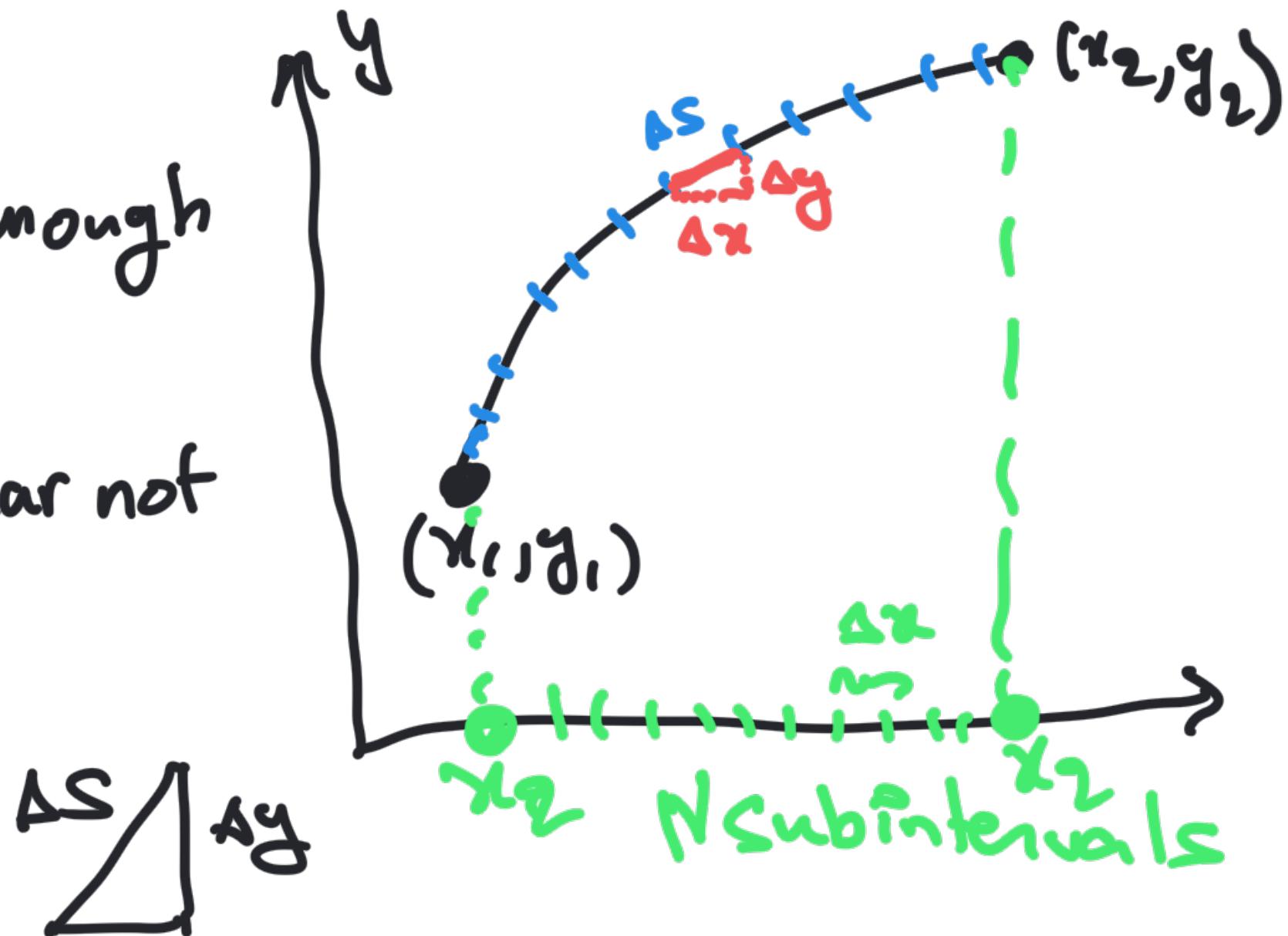
Why is the arc length of a fn between (x_1, y_1) and (x_2, y_2) equal to

$$\int_{x_1}^{x_2} \sqrt{1 + (y'(x))^2} dx ??$$

Answer:

At a small enough scale:

Δs as linear not curved



$$\text{length of } \Delta s = \sqrt{\Delta x^2 + \Delta y^2} = \sqrt{\Delta x^2 \left(1 + \frac{\Delta y^2}{\Delta x^2}\right)}$$

$$= \sqrt{1 + \left(\frac{\Delta y}{\Delta x}\right)^2} \Delta x$$

$$\text{Now length of the arc of } y(x) = \lim_{N \rightarrow \infty} \sum_{i=1}^N \sqrt{1 + \left(\frac{\Delta y}{\Delta x}\right)^2} \Delta x$$

$$\Rightarrow \text{Arc length} = \int_{x_1}^{x_2} \sqrt{1 + y'^2} dx$$

• Second Simplest prob in Calculus of Variations

minimal surface prob

Intuitively

fn with $f=0$ on D

with minimal

surface area is

the flat surface (minimal curvature)

Write a minimization prob (formula for surface area of a fn)

then E-L. eqn : $\underset{\text{some fn space}}{\text{TE}(u)} = 0$

End up getting

$$\left\{ \begin{array}{l} \nabla \cdot \frac{\nabla u}{(1 + |\nabla u|^2)^{1/2}} = 0 \quad \text{in } D \\ \text{on } \partial D \end{array} \right.$$

$$u = c$$

Don't solve this PDE. Just set the min. prob up (var'l principle) and get to this E-L



think of
a soap
bubble
maker.
Pull
out
the tip
get flat
soap
film.

• Third prob in Calculus of var's.

Isoperimetric prob. Look it up.

This will be on the midterm.

NEXT WEEK NEW TOPIC

