

# Untangling Knots Through Curve Repulsion

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# What the curious folks ponder about

1 Introduction

2 Tangent-Point Energy

3 Gradient Flow

# Introduction

# A Cool Knot

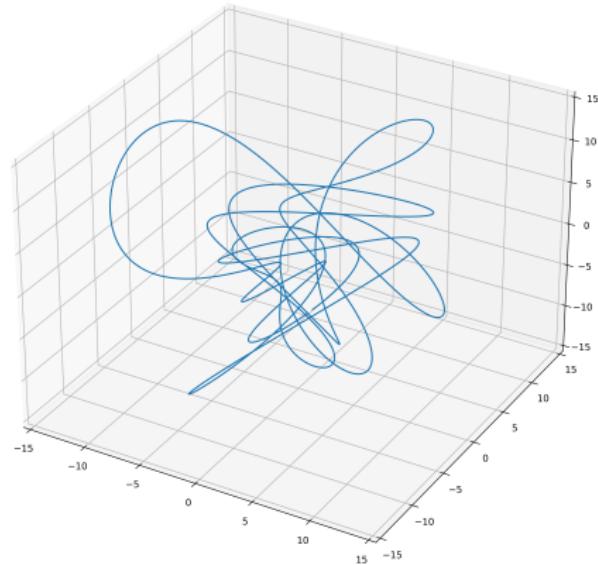


Figure: Imagine your earphones getting tangled like this...

# Aim

- Finding a “homotopy” from a knot to an unknot.

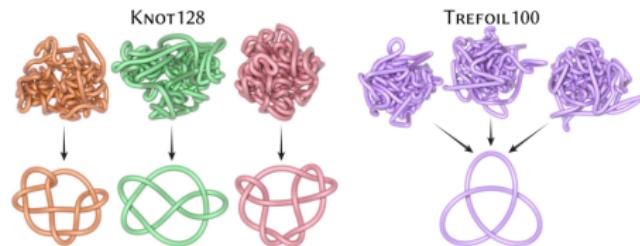


Figure: Unknots of test knots.[2]

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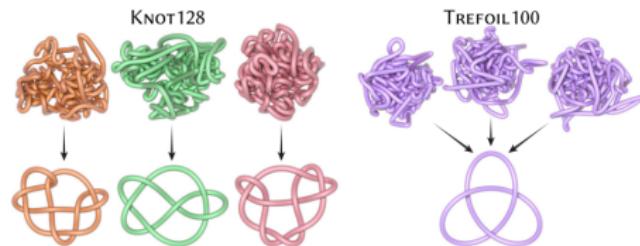


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- “Avoiding self-intersection”

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- ② Attempt to decrease the curve energy by continuously deforming the curve.
  - We evolve the curve according to the gradient flow equation.
  - There is a freedom in choosing the “gradient” here.
- ③ We expect the stationary state to be the “unknot”
  - Or at least a simpler state...

# Tangent-Point Energy

# Defining Curve Energy

Given an (arc-length parameterised) curve  $\gamma : M \rightarrow \mathbb{R}^3$ , we wish to assign energy of the form:

$$\mathcal{E}(\gamma) := \iint_{M^2} k(\gamma_x, \gamma_y) \, d\gamma_x \, d\gamma_y \quad (1)$$

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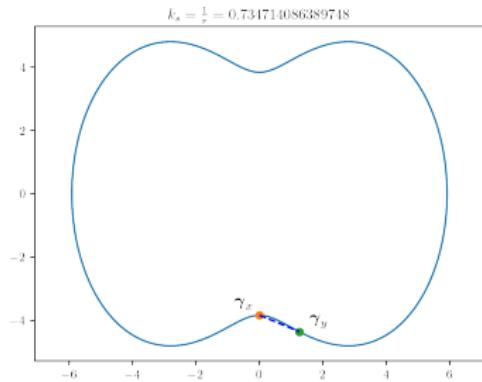
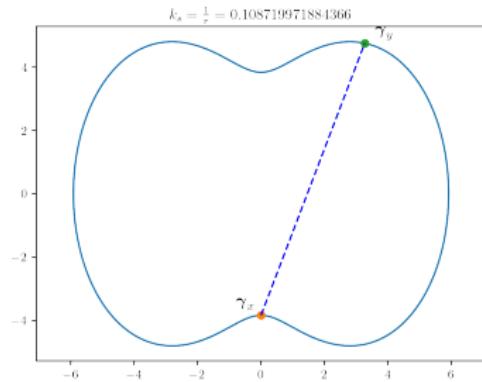
such that

- $\mathcal{E}$  is very high when two non-neighbouring points are very close.

A naïve choice is  $k(\gamma_x, \gamma_y) := \frac{1}{\|\gamma_x - \gamma_y\|}$

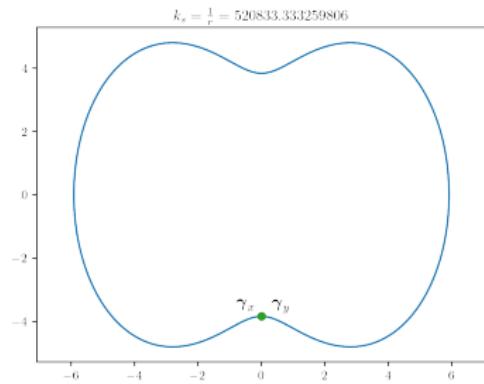
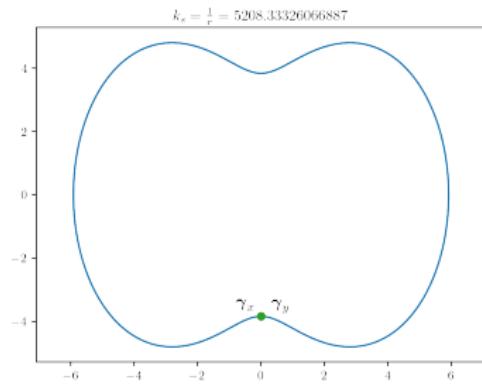
# Pitfall of the “Simple Energy”

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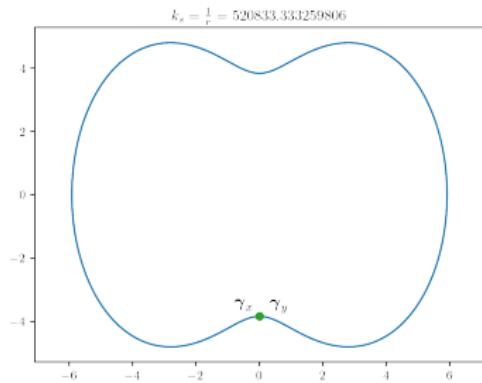
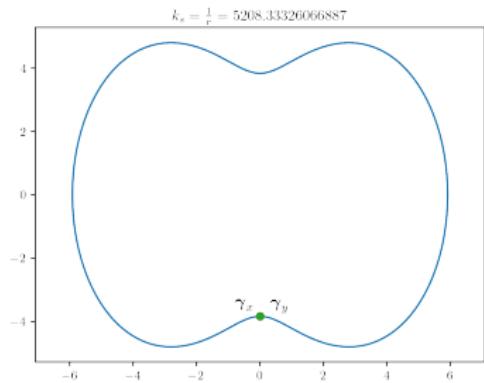
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This energy is not well-defined for a lot of curves!

# Buck-Orloff Tangent-Point Energy

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## Definition (Buck-Orloff Tangent-Point Energy)

For a smooth curve  $\gamma$ , define

$$\mathcal{E}(\gamma) := \iint_{M^2} k_4^2(\gamma_x, \gamma_y, \mathbf{T}_x) d\gamma_x d\gamma_y$$

where  $\mathbf{T}_x$  is the unit tangent vector at  $\gamma_x$ , with the kernel defined as

$$k_4^2(\mathbf{p}, \mathbf{q}, \mathbf{T}) := \frac{\|\mathbf{T} \wedge (\mathbf{p} - \mathbf{q})\|^2}{\|\mathbf{p} - \mathbf{q}\|^4}$$

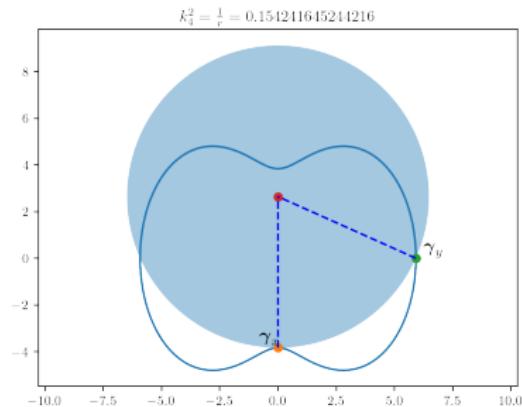
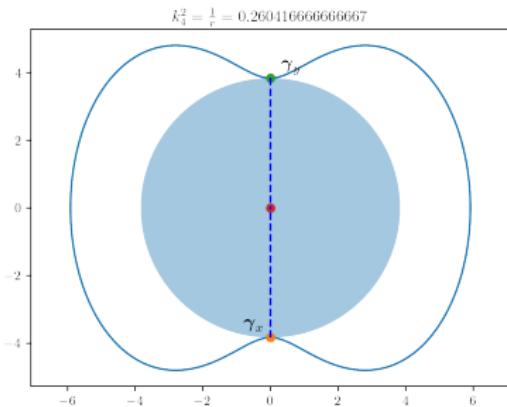
as **Buck-Orloff Tangent-Point Energy**.[1]

# Intuition

What is the intuition behind the kernel  $k_4^2(\mathbf{p}, \mathbf{q}, \mathbf{T}) := \frac{\|\mathbf{T} \wedge (\mathbf{p} - \mathbf{q})\|^2}{\|\mathbf{p} - \mathbf{q}\|^4}$ ?

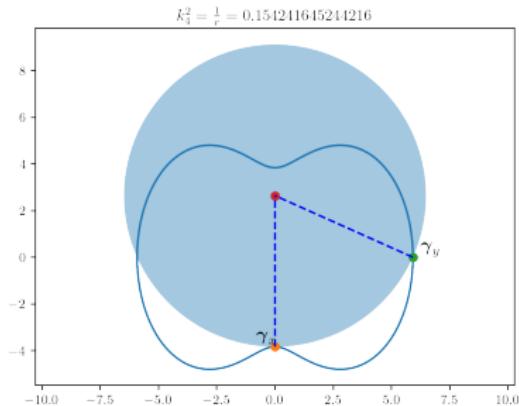
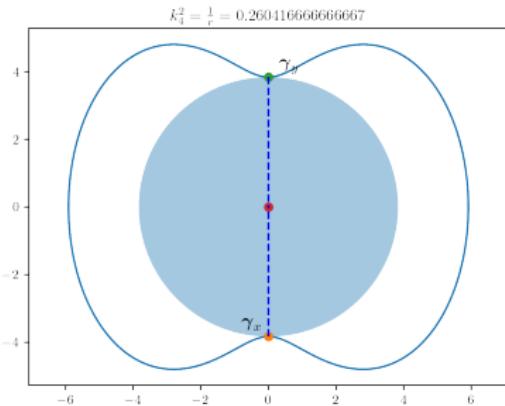
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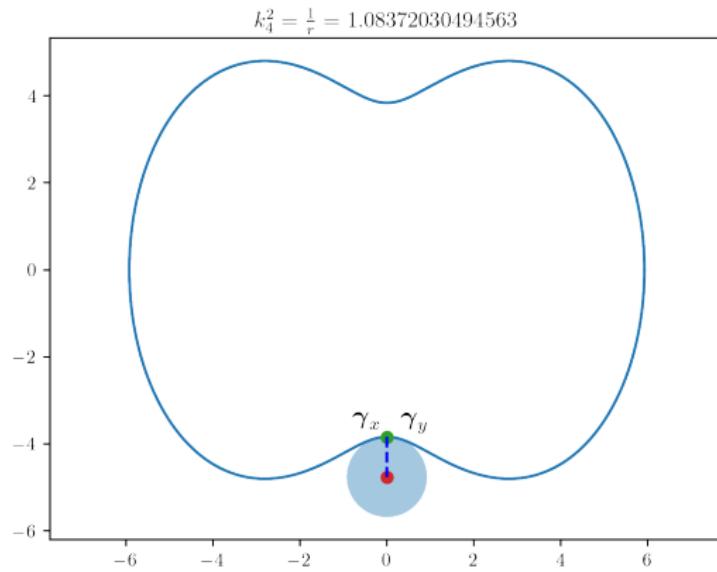
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## Remark

Note that closer does not necessarily mean the kernel is larger.

# Intuition



**Figure:** When two points are very close, the kernel converges to the curvature of the curve.

# Example: Buck-Orloff Tangent-Point Energy of a Circle

## Example

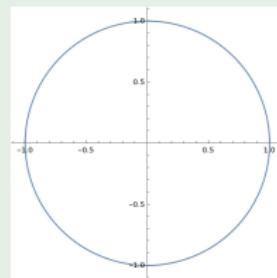
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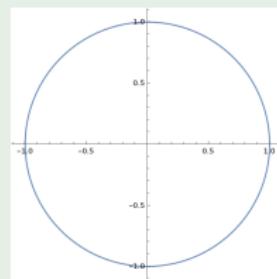


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Then write:

$$\begin{cases} \gamma_x(\theta) = (\cos \theta, \sin \theta, 0) \\ \gamma_y(\phi) = (\cos \phi, \sin \phi, 0) \\ \mathbf{T}_x(\theta) = (-\sin \theta, \cos \theta, 0) \end{cases}$$

## Example (Cont.)

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Substituting to Buck-Orloff Tangent-Point energy formula:

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Using a few identities:

$$\mathcal{E}(\gamma) = \int_{\theta=0}^{2\pi} \int_{\phi=0}^{2\pi} \frac{\|\mathbf{T}_x\|^2 \|\gamma_x - \gamma_y\|^2 - (\mathbf{T}_x \cdot (\gamma_x - \gamma_y))^2}{\|\gamma_x - \gamma_y\|^4} d\theta d\phi$$

## Example (Cont.)

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$$= \int_{\theta=0}^{2\pi} \int_{\phi=0}^{2\pi} \frac{\sin^2 \frac{\theta-\phi}{2}}{(-1 + \cos(\theta - \phi))^2} d\theta d\phi \quad (3)$$

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## Remark

Note that (4) suggests the order at “singularity” is inverse-square.

# General Tangent-Point Energy

A more general form of tangent-point energy comes from Yu, Schumacher, and Crane [2]:

## Definition (Generalised Tangent-Point Energy)

$$\mathcal{E}_\beta^\alpha(\gamma) := \iint_{M^2} \frac{||\mathbf{T}_x \wedge (\gamma_x - \gamma_y)||^\alpha}{||\gamma_x - \gamma_y||^\beta} d\gamma_x d\gamma_y$$

where  $\alpha > 1$  and  $\beta \in [\alpha + 2, 2\alpha + 1]$

## Remark

When  $\alpha = 2$  and  $\beta = 4$ , we are back to Buck-Orloff.

# Gradient Flow

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cf) For minimising a function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ , one may consider

$$\frac{\partial \mathbf{x}}{\partial t} = -\nabla f(\mathbf{x})$$

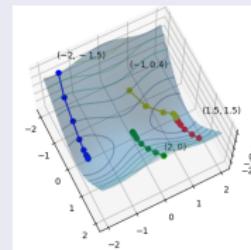


Figure: Steepest Descent

# Gradient in Space of Function?

(Gradient of Function)

$\nabla f(\mathbf{x})$  is such that for all  $\mathbf{y} \in \mathbb{R}^n$

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## Remark

For  $X = L^2$ ,  $(\text{grad}_X E)(f)$  is the first order variation from  $f$  with respect to  $\epsilon$ .

# Example: Dirichlet Energy in 1D

## Definition (1D Dirichlet Energy)

For a differentiable function  $f : \mathcal{I} \rightarrow \mathbb{R}$ , define **1D Dirichlet energy**

$$E_D(f) := \int_{\mathcal{I}} |\nabla f(x)|^2 \, dx = \int_{\mathcal{I}} |f'(x)|^2 \, dx$$

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## Remark

Dirichlet energy is high for function  $f$  that varies a lot, and minimised by any constant function.

## Example: $L^2$ Gradient Flow of Dirichlet Energy

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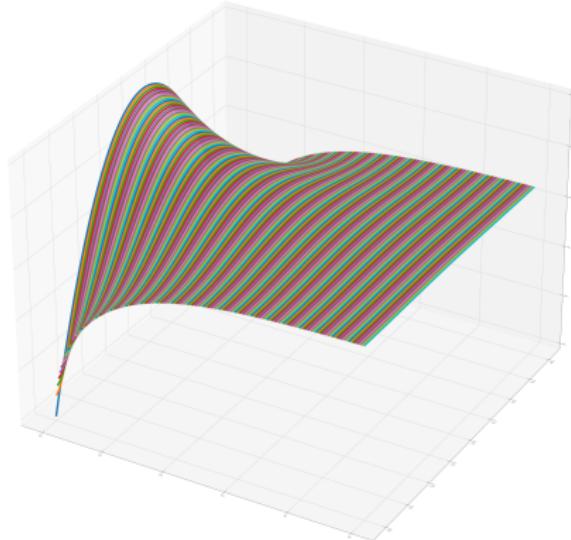
**Lemma ( $L^2$  Gradient of Dirichlet Energy)**

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Then the gradient flow equation for Dirichlet energy becomes:

$$\frac{\partial f}{\partial t} = \Delta f \tag{6}$$



**Figure:** Solution to  $L^2$  Gradient Flow Equation for Dirichlet Energy (AKA Heat Equation)

# Dirichlet Gradient Flow in Other Function Space

## Definition (Famous Inner Product Function Spaces)

- $L^p = \left\{ f \left| \left( \int |f|^p dx \right)^{1/p} < \infty \right. \right\}$ 
  - $\langle \langle f, g \rangle \rangle_{L^2} = \int fg dx$
- $p = 2$  Sobolev Space  $H^k = \left\{ f \left| \sum_{i=0}^k \int |\mathcal{D}^{(i)} f|^2 dx < \infty \right. \right\}$
- $L^2$ :  $\frac{\partial f}{\partial t} = -\Delta f$
- $H^1$ :  $\frac{\partial f}{\partial t} = -f$
- $H^{-1}$ :  $\frac{\partial f}{\partial t} = -\Delta^2 f$ 
  - $\langle \langle f, g \rangle \rangle_{H^{-1}} = \langle \langle \Delta^{-1} f, \Delta^{-1} g \rangle \rangle_{H^1}$
- $H^2$ :  $\frac{\partial f}{\partial t} = \Delta^{-1} f$

# Bibliography

- [1] Gregory Buck and Jeremey Orloff. “A simple energy function for knots”. In: *Topology and its Applications* 61.3 (Feb. 1995), pp. 205–214. doi: [10.1016/0166-8641\(94\)00024-w](https://doi.org/10.1016/0166-8641(94)00024-w).
- [2] Chris Yu, Henrik Schumacher, and Keenan Crane. “Repulsive Curves”. In: *ACM Transactions on Graphics* 40.2 (Apr. 2021), pp. 1–21. doi: [10.1145/3439429](https://doi.org/10.1145/3439429).