

# ML applications to string theory

---

FABIAN RUEHLE  
AITP 2021

September 5th, 2021

Based on:

[Anderson, Gray, Gerdes, Krippendorf, Raghuram, FR: 2012.04656]

[Gukov, Halverson, FR, Sułkowski: 2010.16263]

[Halverson, Nelson, FR: 1903.11616]

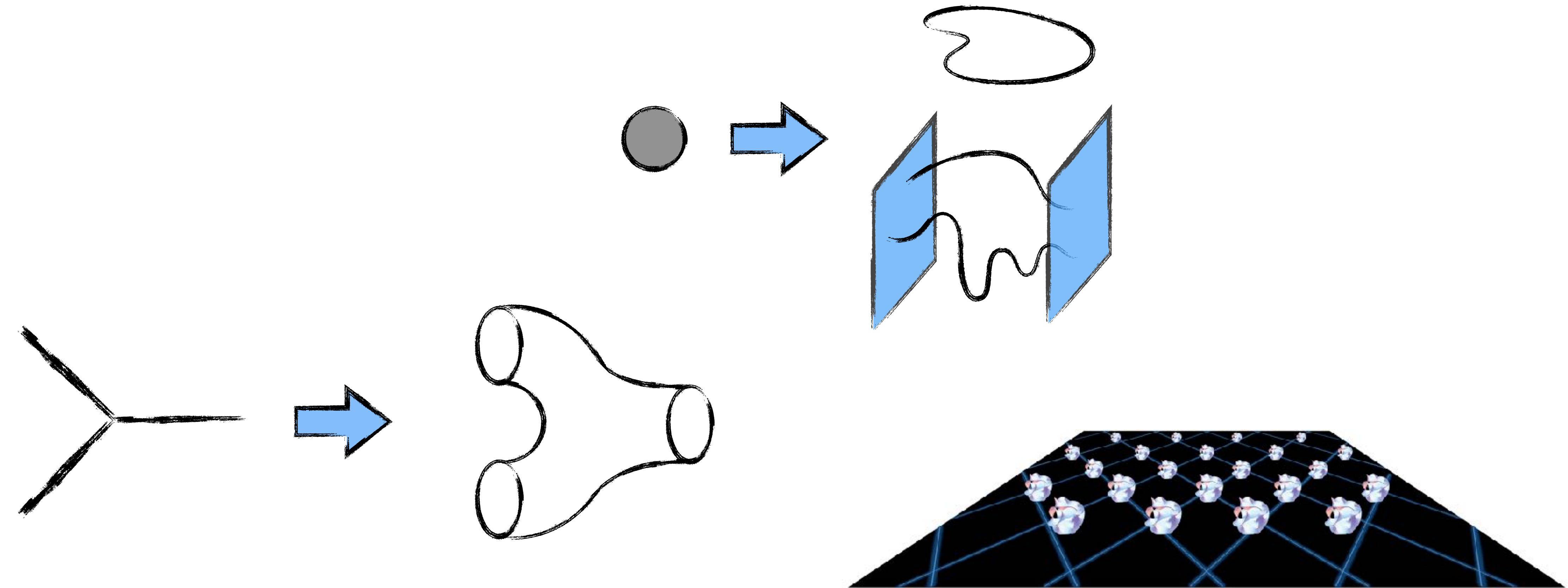
[FR: Physics Reports `20]



# Outline

---

- ▶ Introduction to String Theory
- ▶ Example applications of ML in String Theory
  - Find solutions to Diophantine equations
  - Find the Unknot
  - Find the Calabi-Yau metric
- ▶ Conclusions



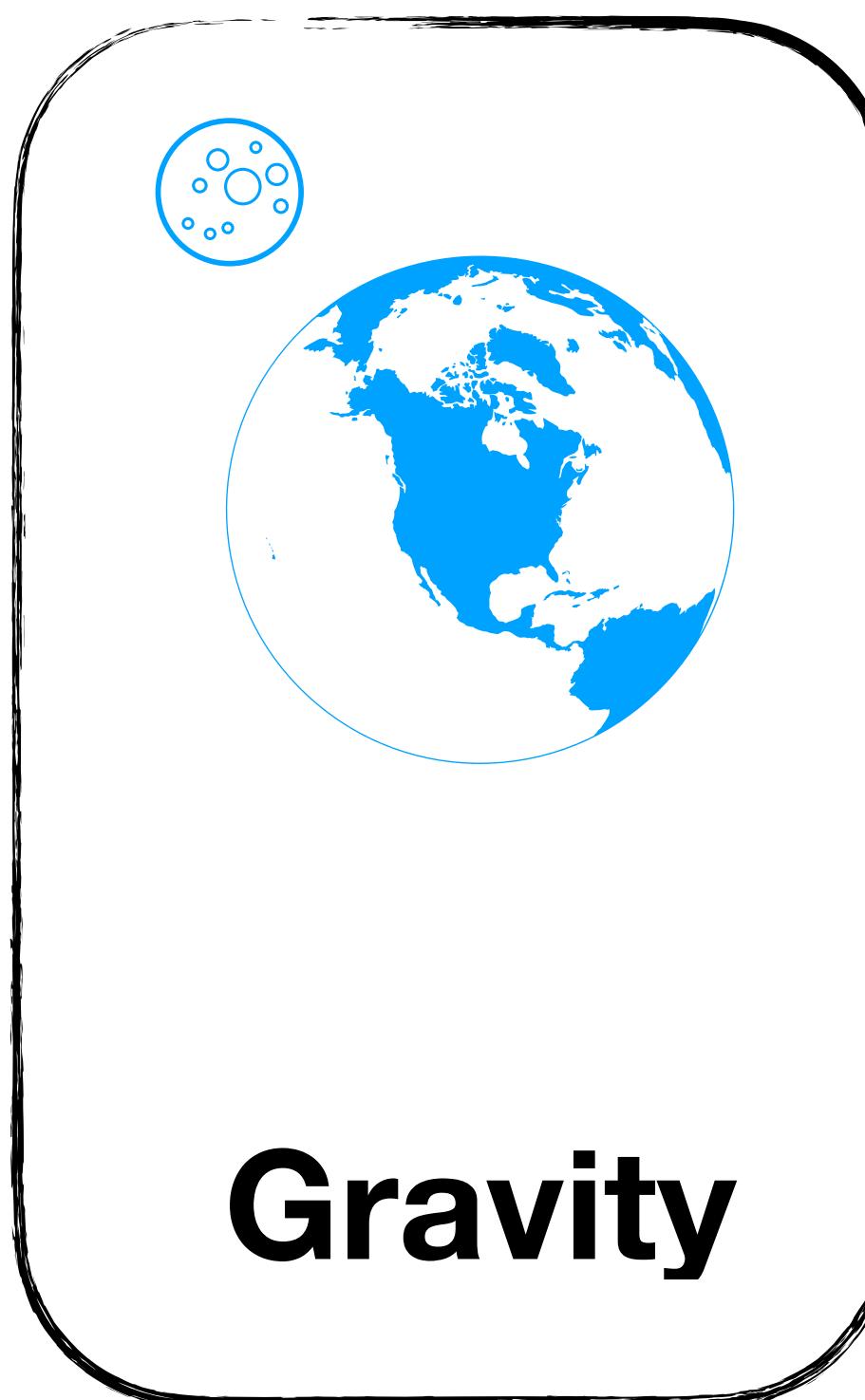
# Introduction to String Theory

---

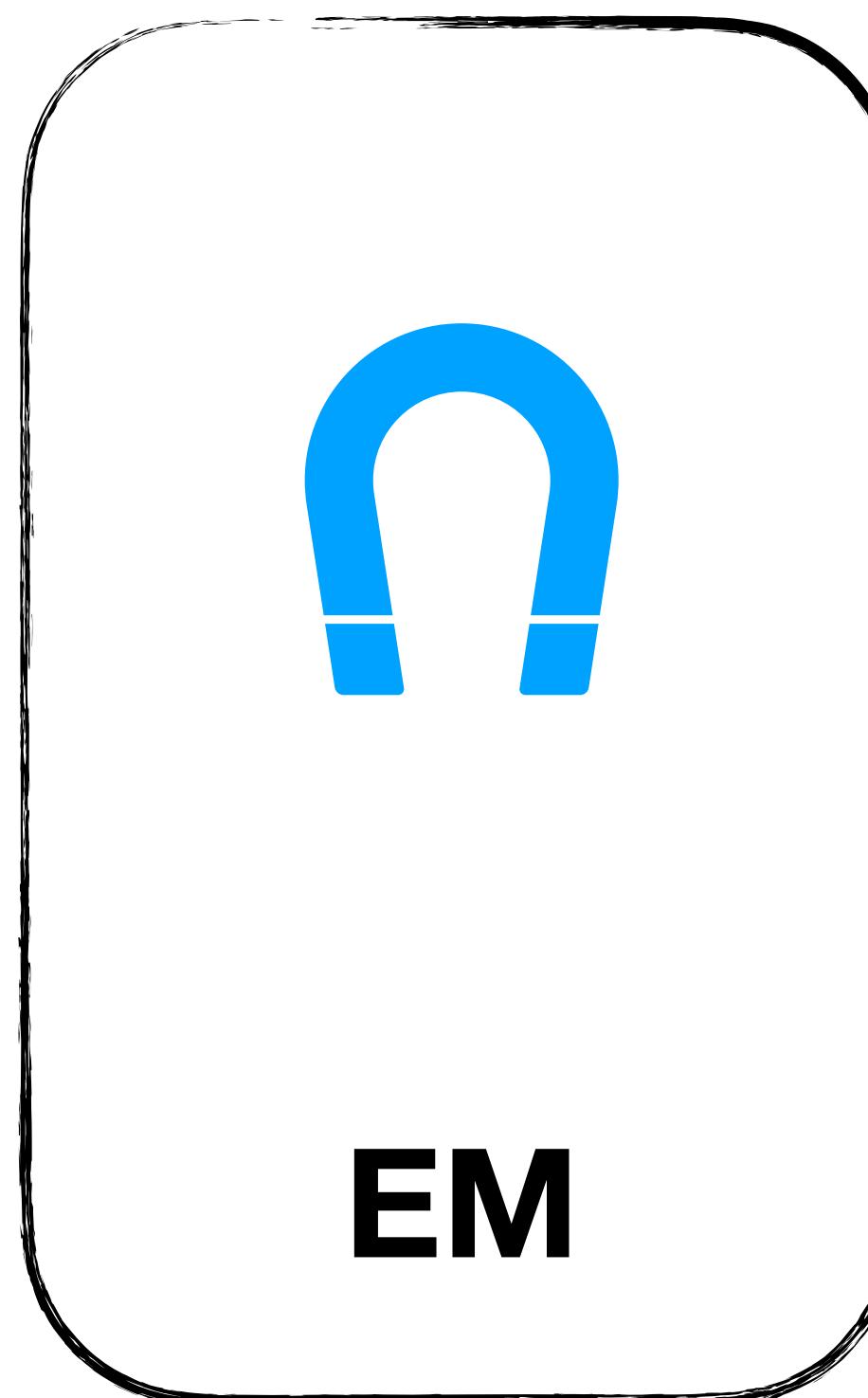
# Physics motivation

---

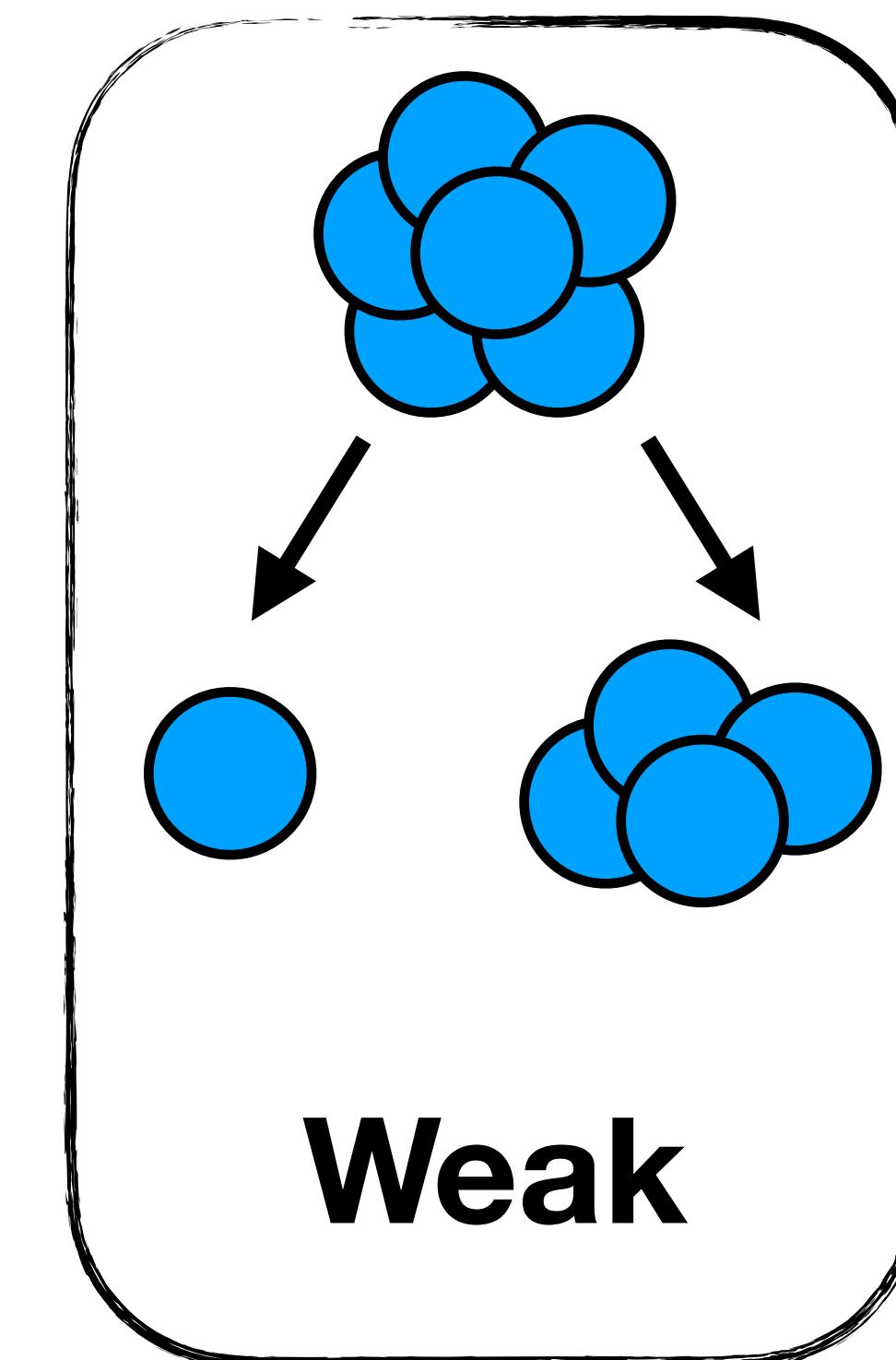
Many observations in our Universe can be explained with just four fundamental forces



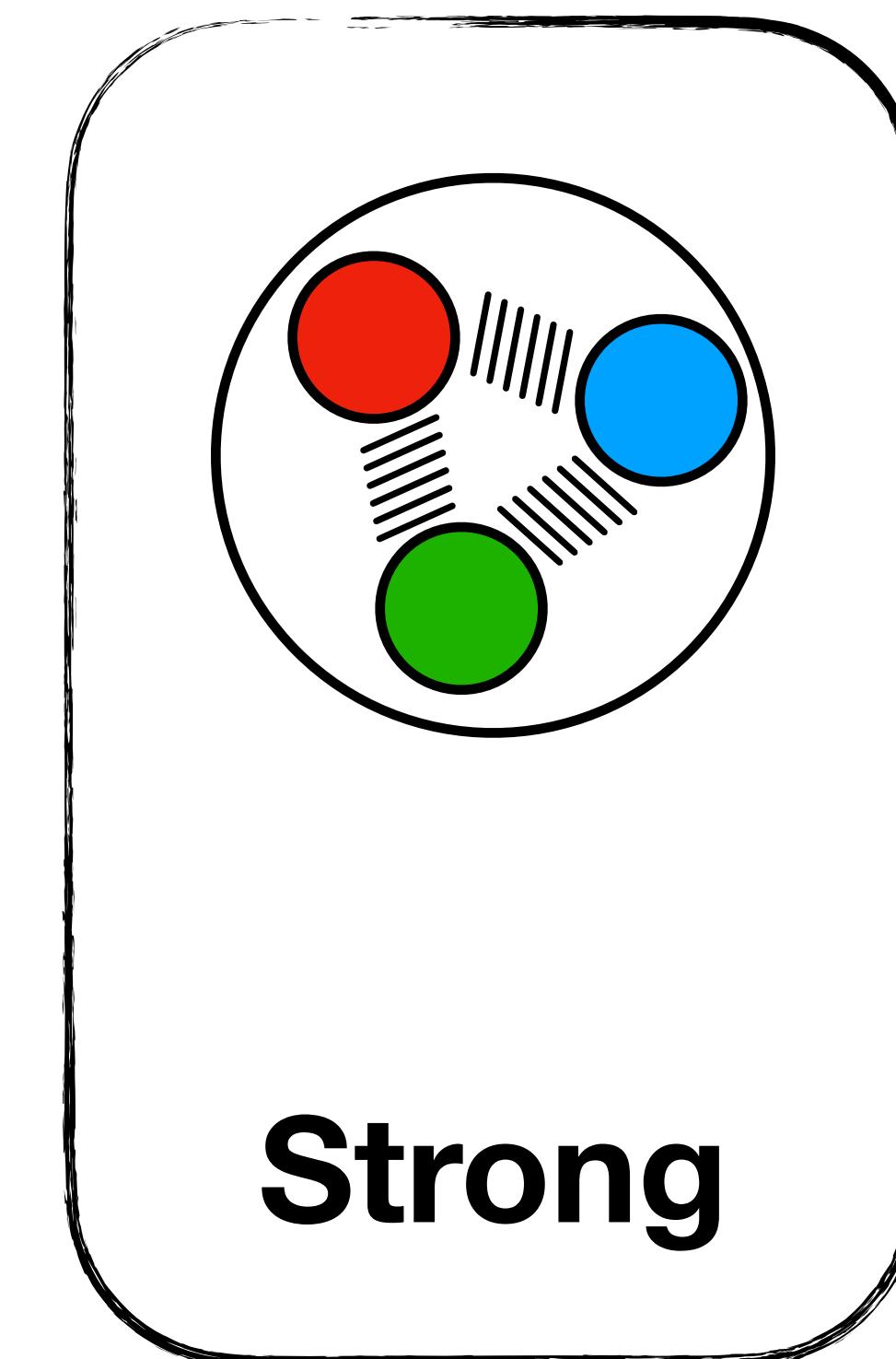
**Gravity**



**EM**



**Weak**

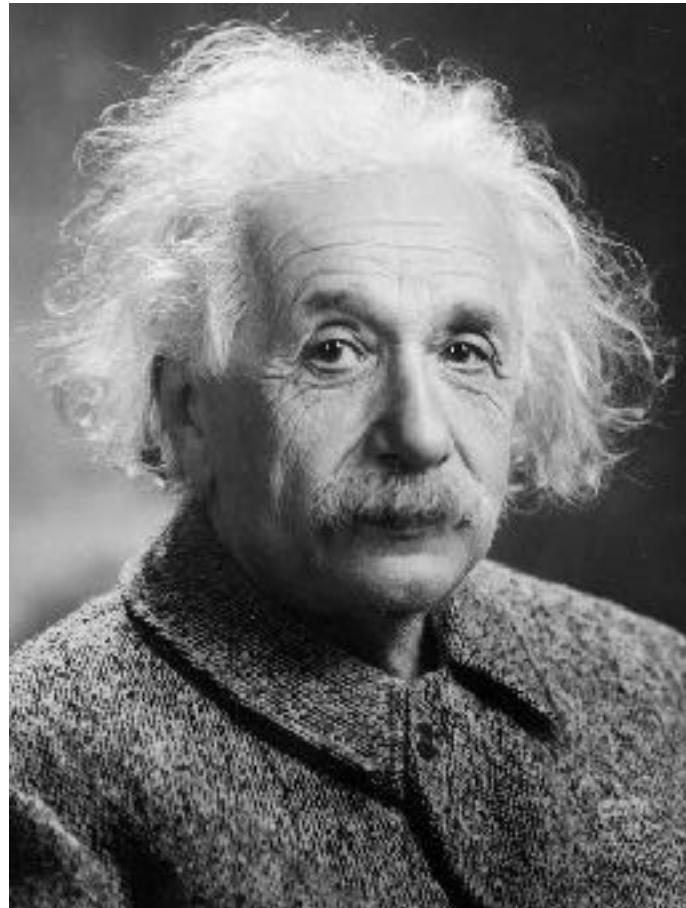


**Strong**

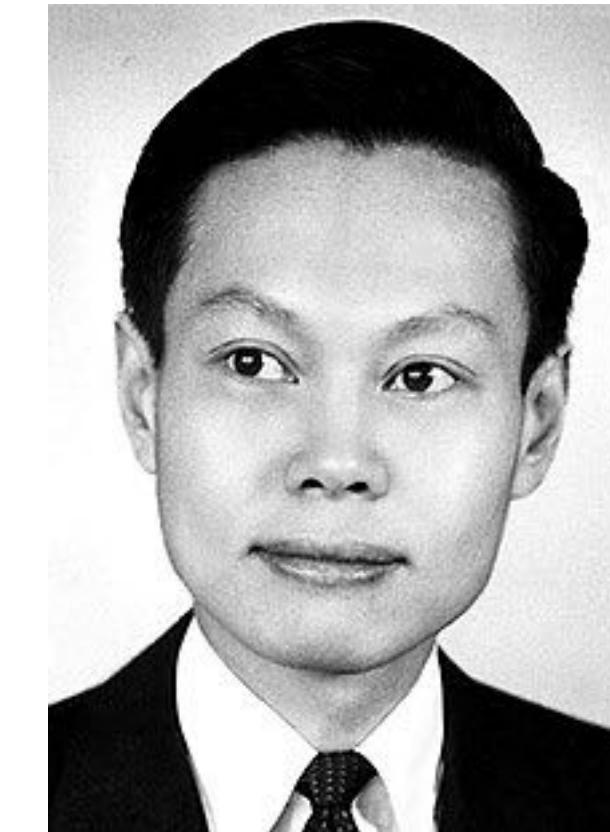
# Physics motivation

---

Two theories to describe these four forces.



**General  
Relativity**

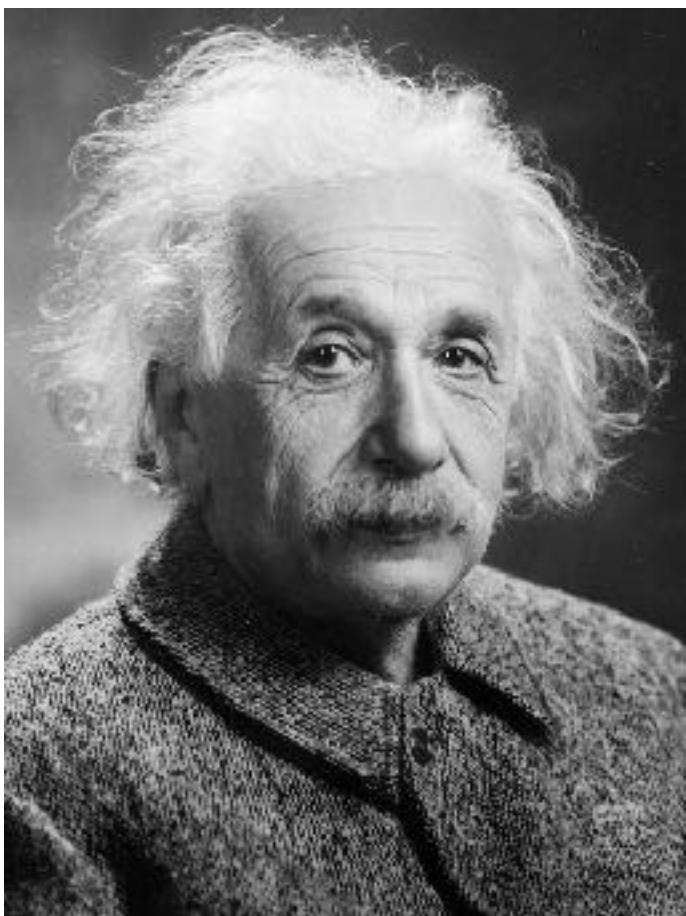


**Quantum Field Theory  
(Yang-Mills Theory)**



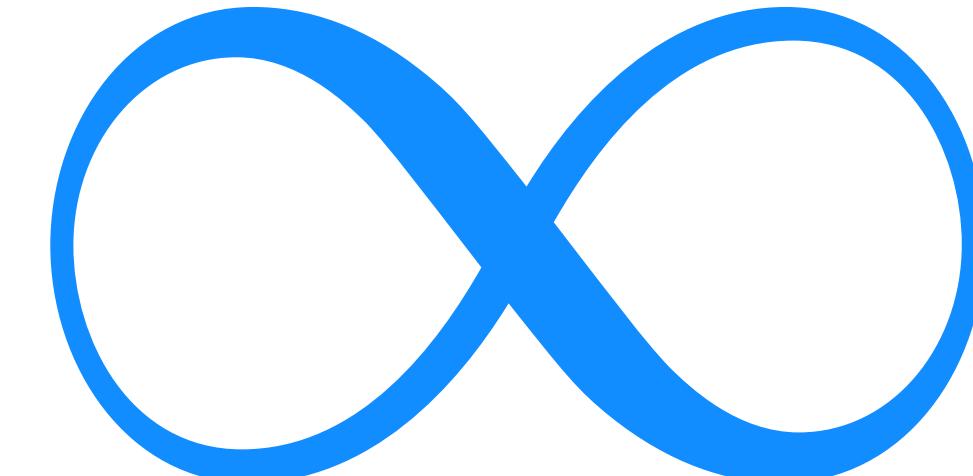
# Physics motivation

---



+

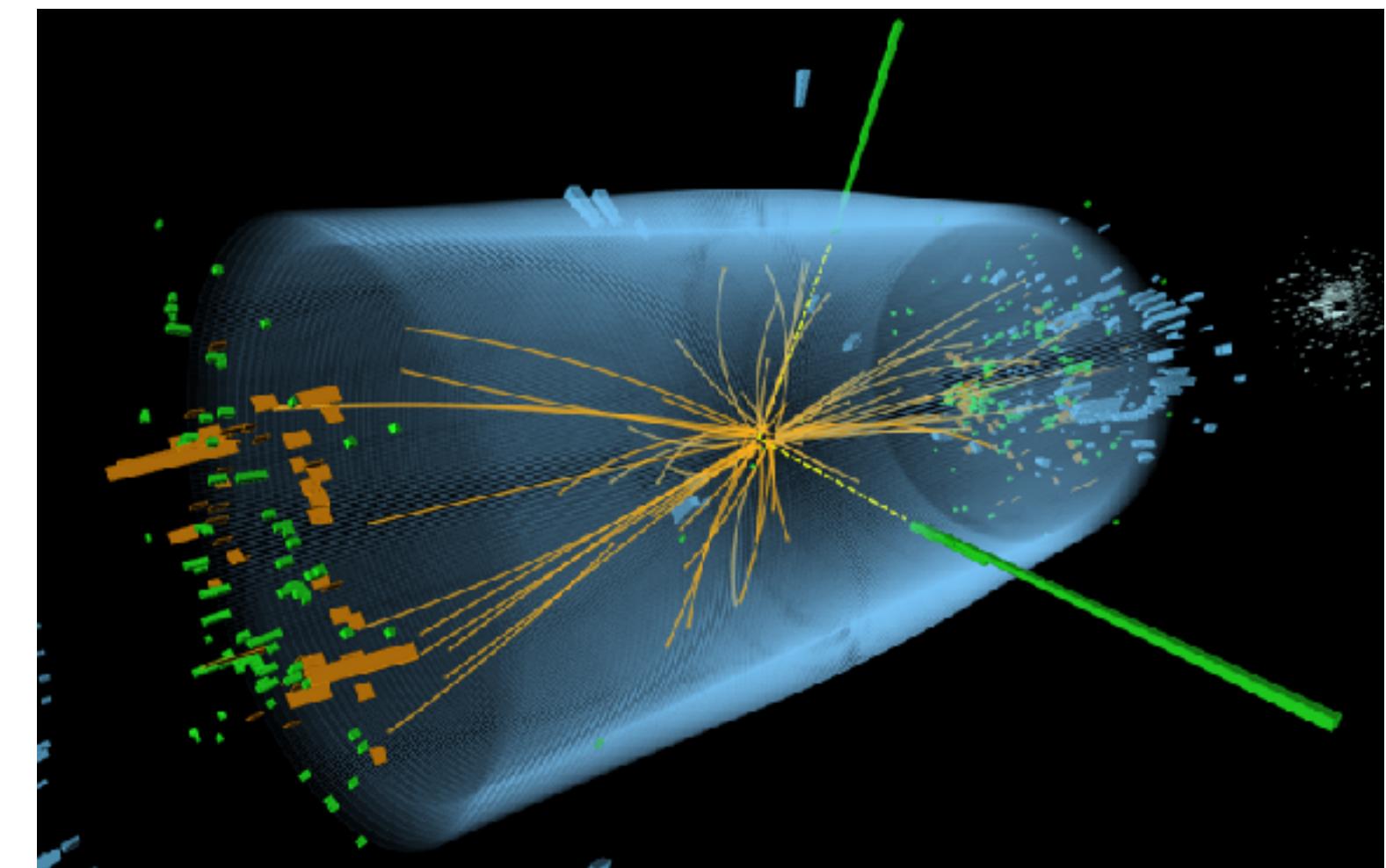
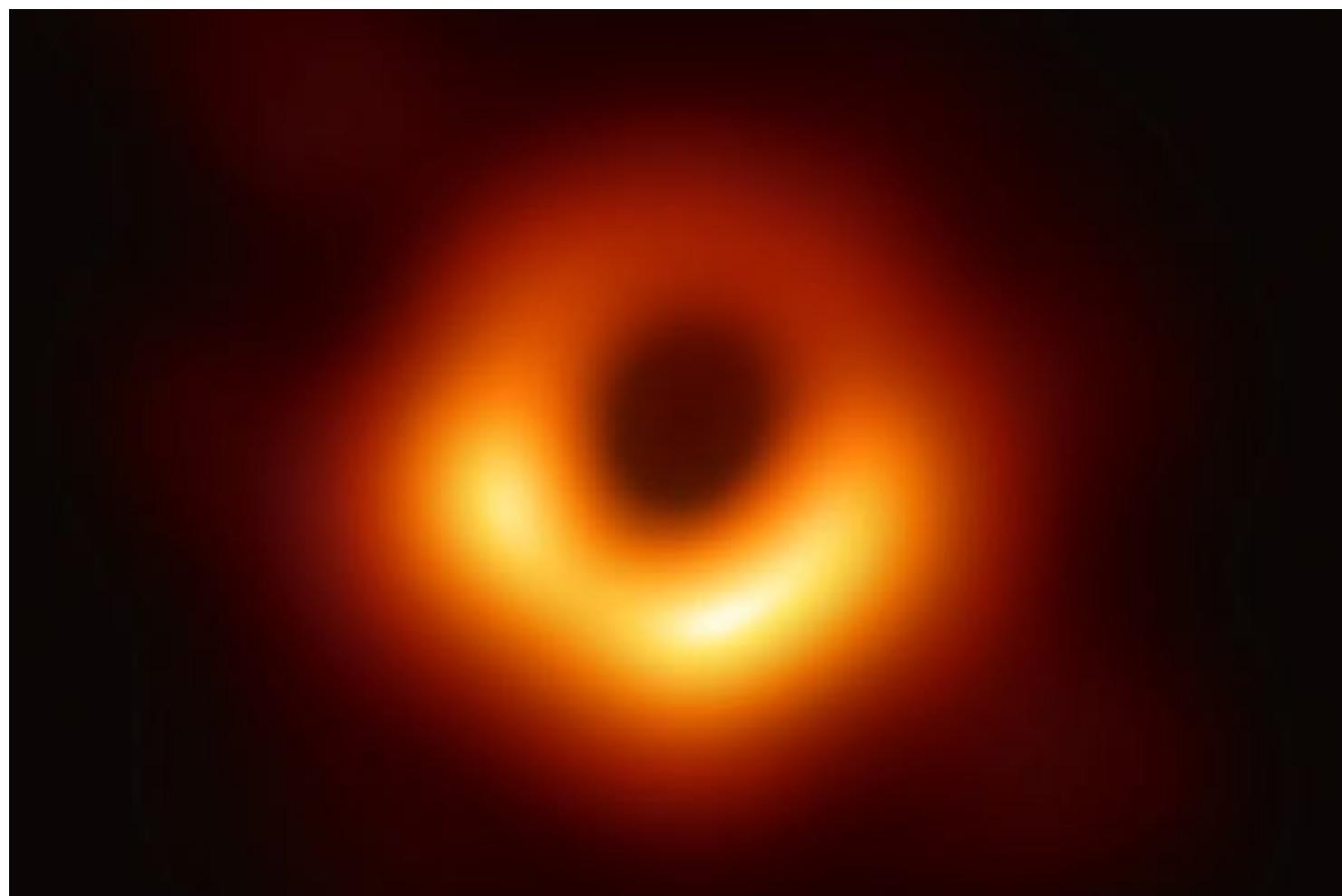
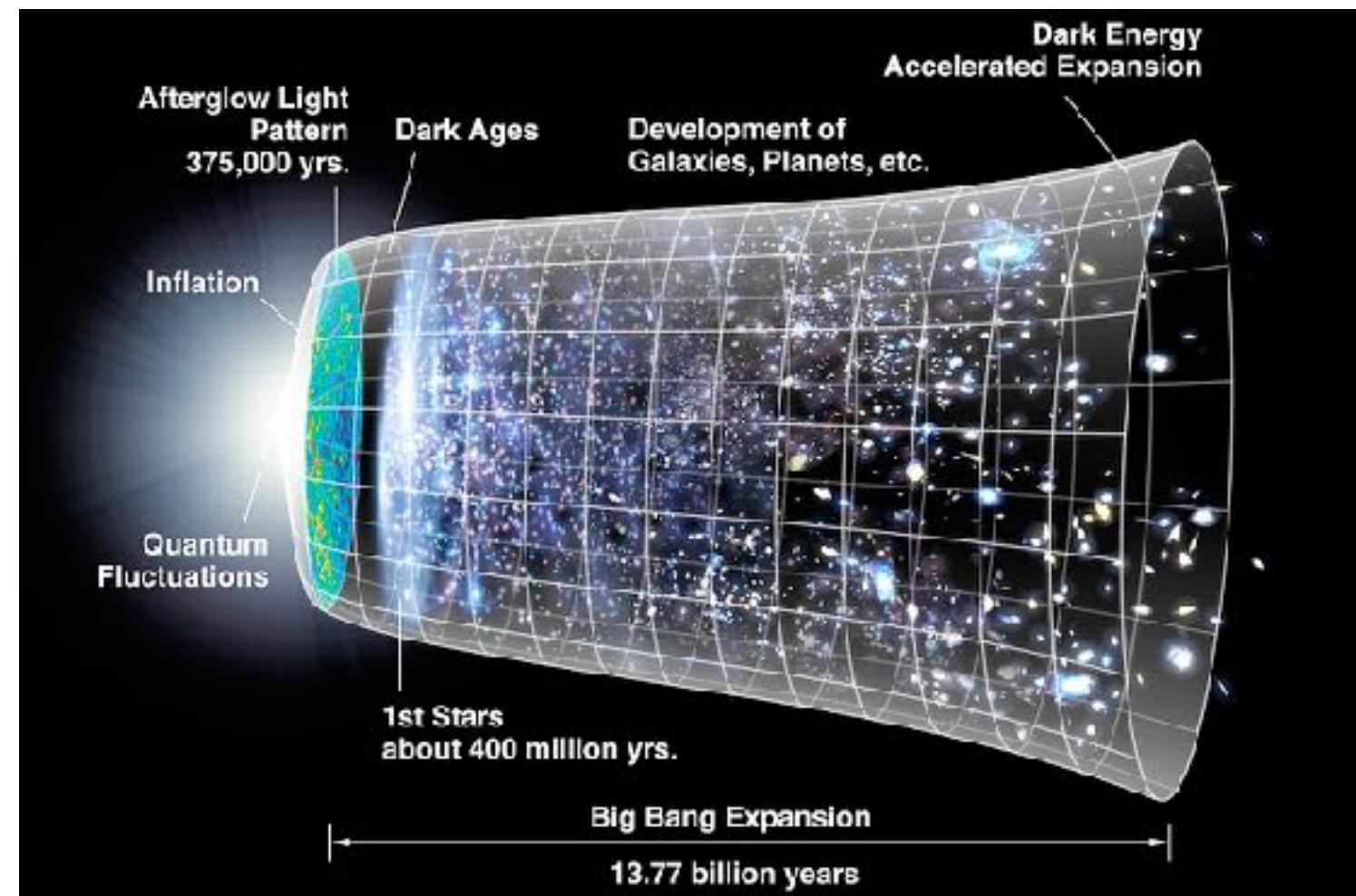


= 

# Physics motivation

---

...however, we need a unified description to study physics at high energies



Big Bang, Dark Energy, Inflation

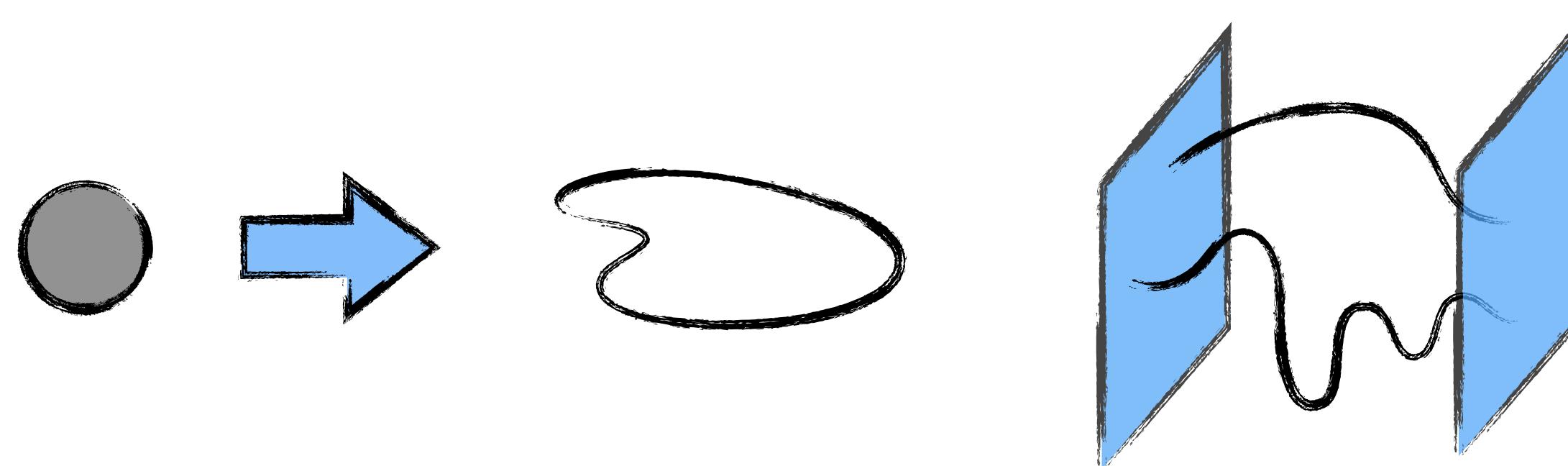
Black Hole Entropy and Information

GUT theories and Physics beyond SM

# String Theory

---

- ▶ One promising candidate for a unified description of General Relativity and Quantum Field Theory: **String Theory**
- ▶ Basic assumption: Fundamental constituents of the particles that mediate the four forces and of all matter are not-point-like, but one-dimensional, extended strings



# String Theory - Compactifications

---

This has far-reaching consequences!

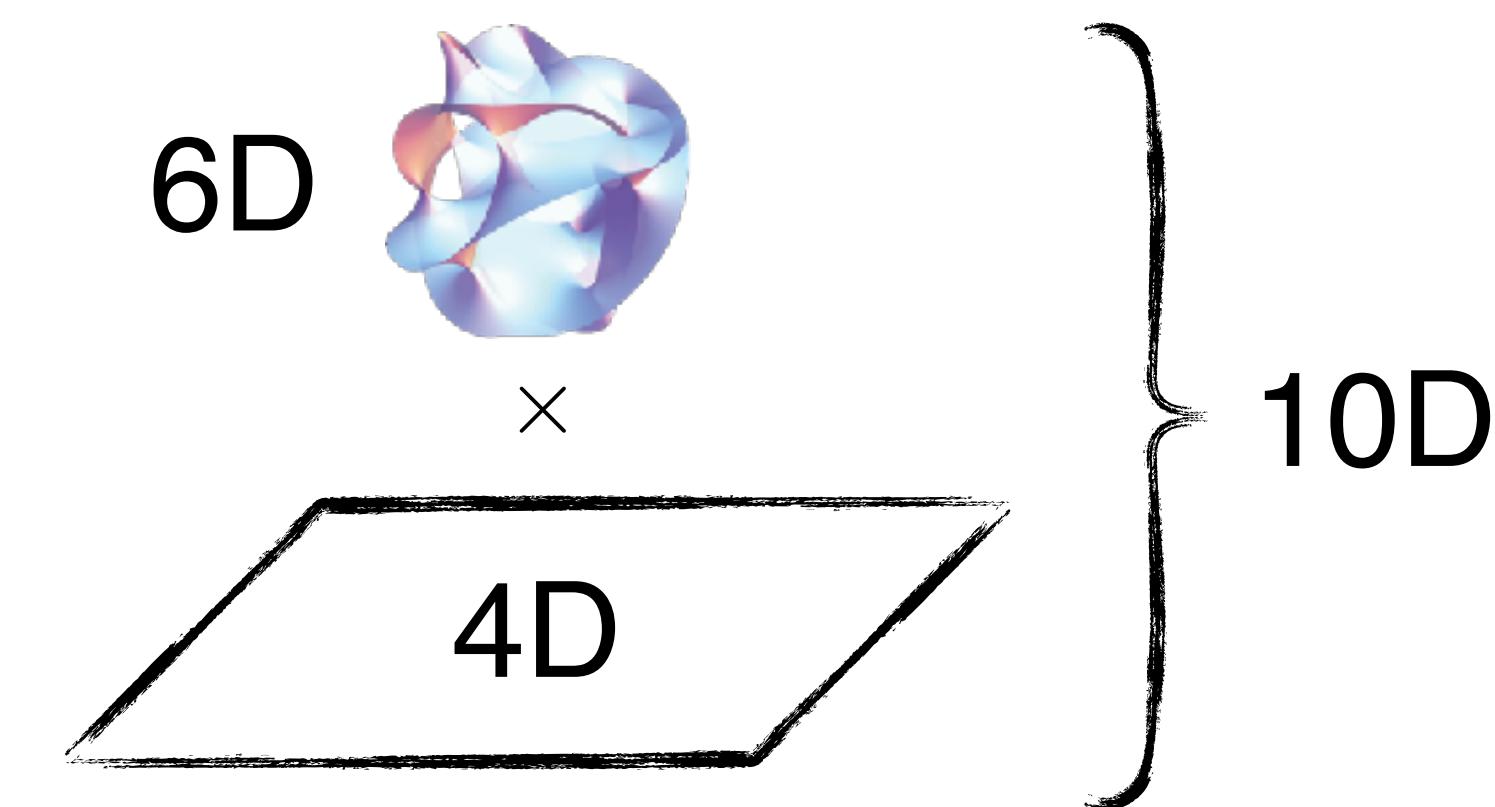
- ▶ Stringent constraints:
  - Consistency (mathematical)
  - Match with observed universe (physical)

# String Theory - Compactifications

---

This has far-reaching consequences!

- ▶ Stringent constraints:
  - Consistency (mathematical)
  - Match with observed universe (physical)
- ▶ Requires ten space-time dimensions
- ▶ 10D description essentially unique
- ▶ We only observe 3, so 6 have to be small to evade detection  $\Rightarrow$  compactifications
- ▶ All of the observable physics is encoded in the 6 compact dimensions (Calabi-Yau manifolds)

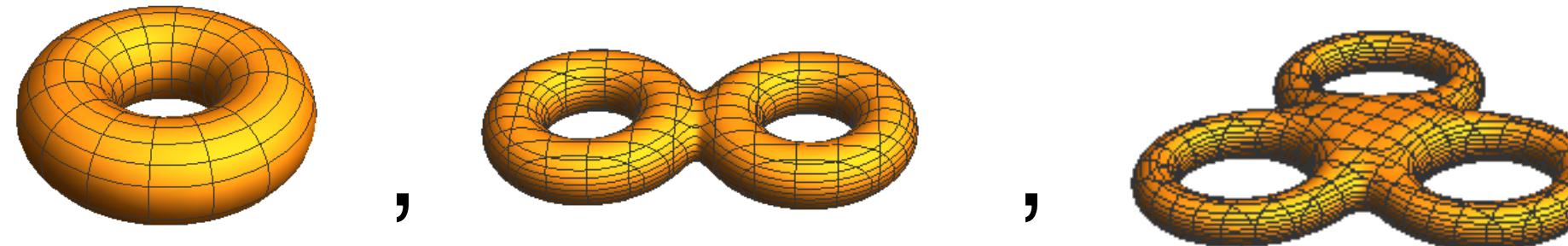


# String Theory - Compactifications

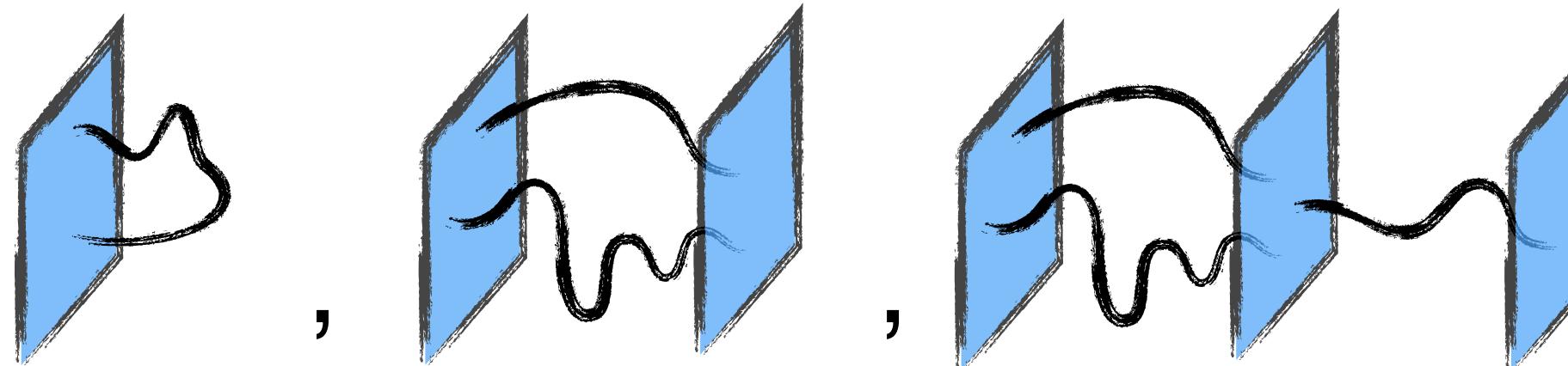
---

- ▶ Discrete data

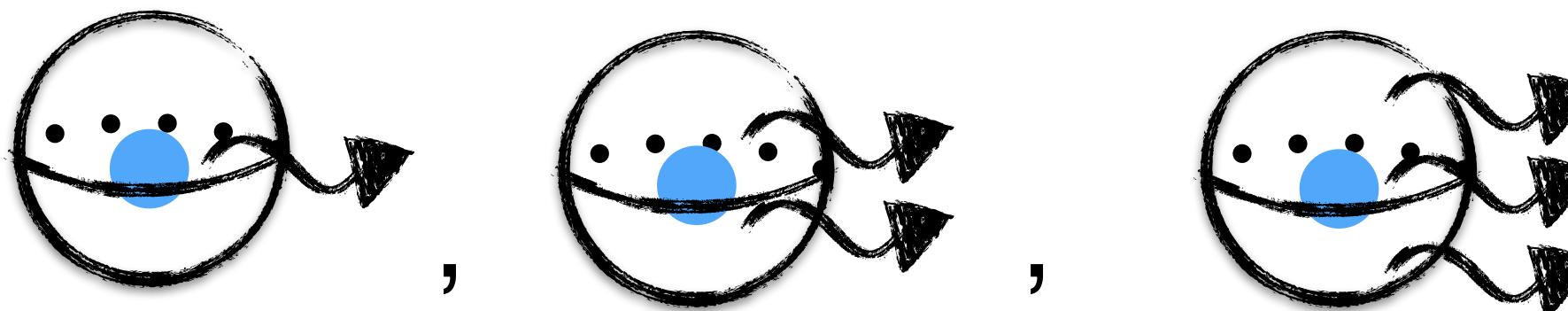
- Topology of CY



- Number of branes



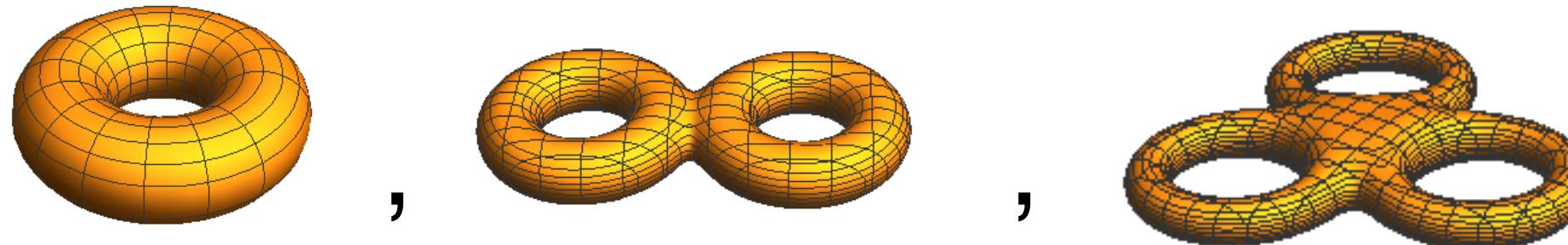
- Number of fluxes



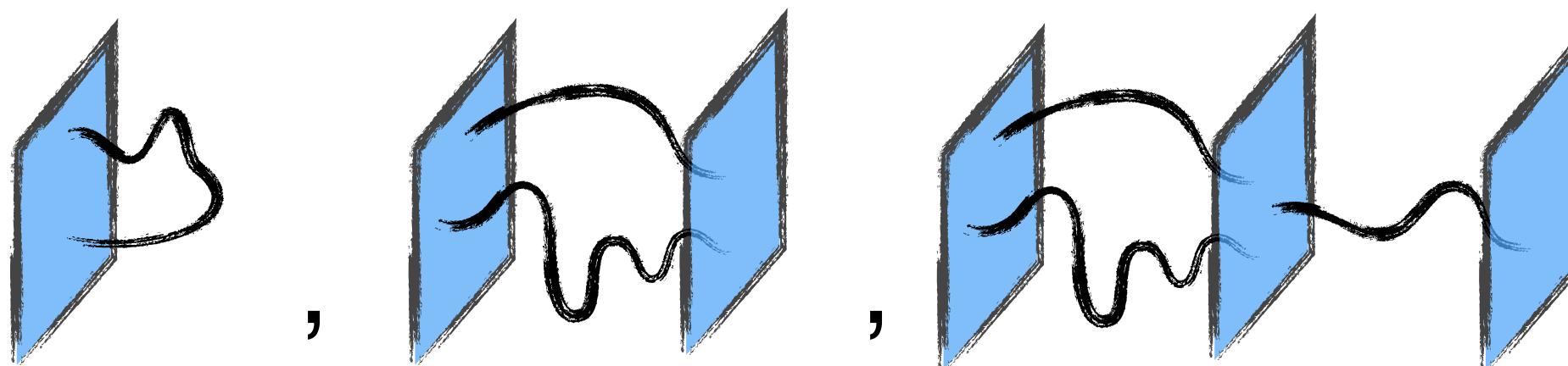
# String Theory - Compactifications

- Discrete data

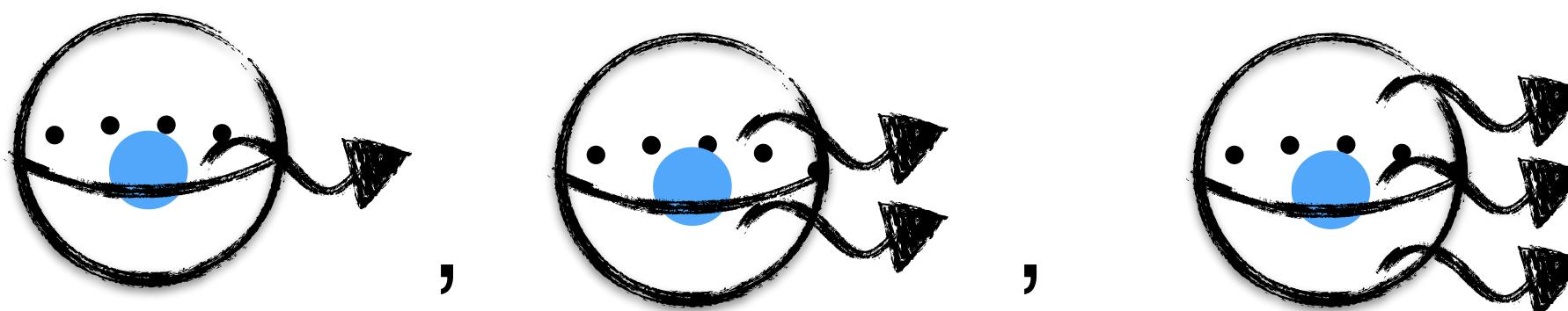
- Topology of CY



- Number of branes

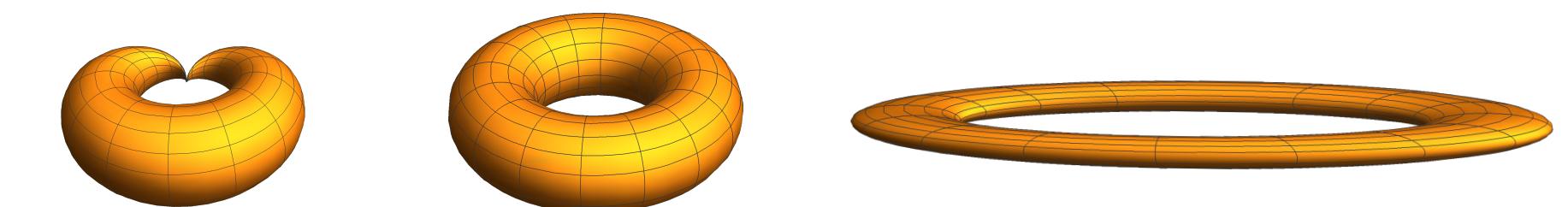


- Number of fluxes

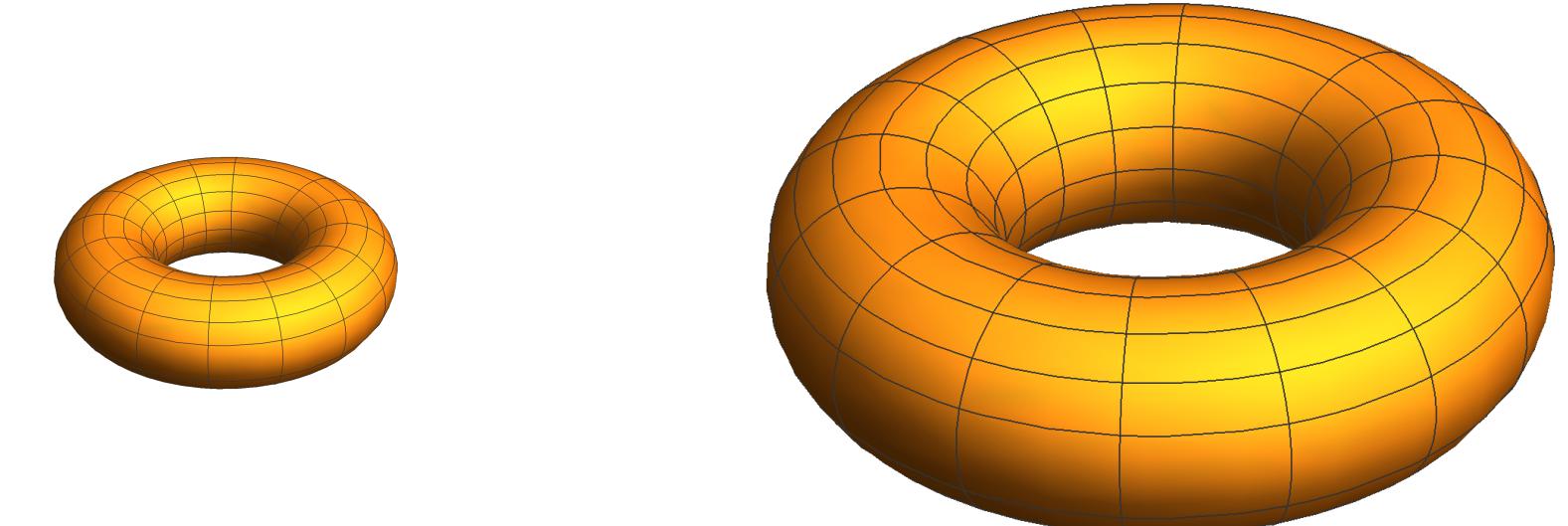


- Continuous data (moduli)

- Shape of CY



- Size of CY



# ML Applications

---

## Conjecture Generation

Try to learn a map between quantities with no previously known relation, formulate (and hopefully prove) conjecture

- Knot theory  
[Hughes `16; Jejjala,Kar,Parrikar `19; Gukov, Halverson,FR,Sulkowski `20; Craven,Jejjala,Kar `20]
- Toric geometry  
[Krefl,Seong`17;Carifio,Cunningham,Halverson, Krioukov,Long`17]
- Line bundle cohomology, Brill-Noether theory  
[FR `17; Klaewer,Schlechter `18; Brodie,Constantin, Deen,Lukas `18-20; Bies,Cvetič,Donagi,Lin,Liu,FR `20]
- Many more... [especially He et.al.]

# ML Applications

---

## Conjecture Generation

Try to learn a map between quantities with no previously known relation, formulate (and hopefully prove) conjecture

- Knot theory  
[Hughes `16; Jejjala,Kar,Parrikar `19; Gukov, Halverson,FR,Sulkowski `20; Craven,Jejjala,Kar `20]
- Toric geometry  
[Krefl,Seong `17; Carifio,Cunningham,Halverson, Krioukov,Long `17]
- Line bundle cohomology, Brill-Noether theory  
[FR `17; Klaewer,Schlechter `18; Brodie,Constantin, Deen,Lukas `18-20; Bies,Cvetič,Donagi,Lin,Liu,FR `20]
- Many more... [especially He et.al.]

## Optimization and Regression

Find solutions to a system of equations

- Searches for string vacua (discrete)  
[FR`17; Wang,Zhang `18; Mutter,Parr,Vaudrevange `18; Halverson,Nelson,FR `19; Brodie,Constantin,Deen, Lukas`19; Larfors,Schneider `20; Deen,He,Lee,Lukas `20; Otsuka,Takemoto `20; Cabo Bizet,Damian,Loaiza-Brito,Mayorga,Montañez-Barrera `20, Constantin, Harvey,Lukas `21]
- CY metrics (continuous)  
[Ashmore,He,Ovrut `19; Anderson,Gray,Gerdes, Krippendorf,Raghuram,FR `20; Douglas, Lakshminarasimhan,Qi `20; Jejjala,Mayorga,Mishra `20]

# ML Applications

## Conjecture Generation

Try to learn a map between quantities with no previously known relation, formulate (and hopefully prove) conjecture

- Knot theory  
[Hughes `16; Jejjala,Kar,Parrikar `19; Gukov, Halverson,FR,Sulkowski `20; Craven,Jejjala,Kar `20]
- Toric geometry  
[Krefl,Seong `17; Carifio,Cunningham,Halverson, Krioukov,Long `17]
- Line bundle cohomology, Brill-Noether theory  
[FR `17; Klaewer,Schlechter `18; Brodie,Constantin, Deen,Lukas `18-20; Bies,Cvetič,Donagi,Lin,Liu,FR `20]
- Many more... [especially He et.al.]

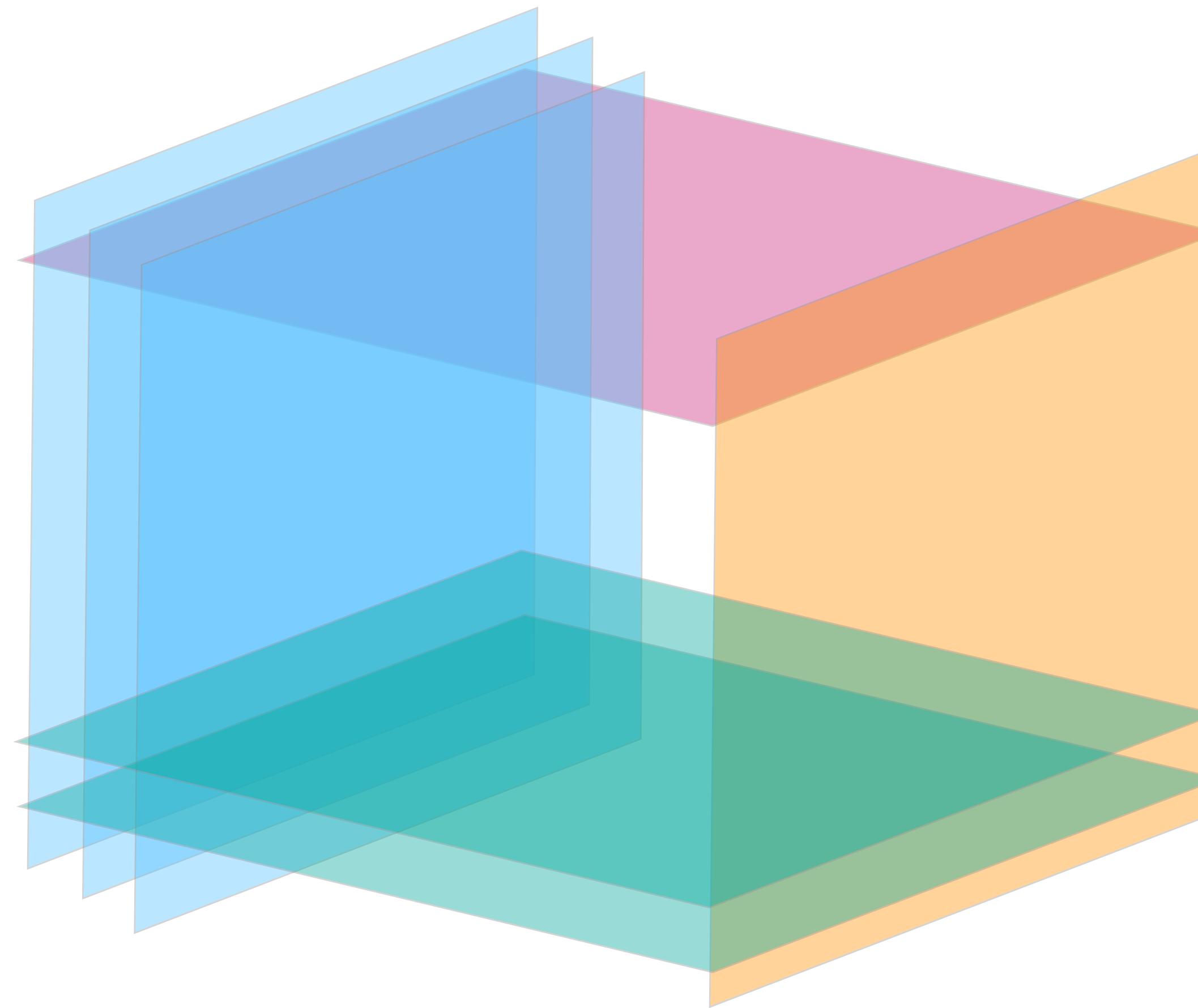
## Optimization and Regression

Find solutions to a system of equations

- Searches for string vacua (discrete)  
[FR`17; Wang,Zhang `18; Mutter,Parr,Vaudrevange `18; Halverson,Nelson,FR `19; Brodie,Constantin,Deen, Lukas`19; Larfors,Schneider `20; Deen,He,Lee,Lukas `20; Otsuka,Takemoto `20; Cabo Bizet,Damian,Loaiza-Brito,Mayorga,Montañez-Barrera `20, Constantin, Harvey,Lukas `21]

- CY metrics (continuous)

[Ashmore,He,Ovrut `19; Anderson,Gray,Gerdes, Krippendorf,Raghuram,FR `20; Douglas, Lakshminarasimhan,Qi `20; Jejjala,Mayorga,Mishra `20]



# Example I: Solving Diophantine equations

---

# Example I - Solving Diophantine equations

---

- ▶ **Background:**

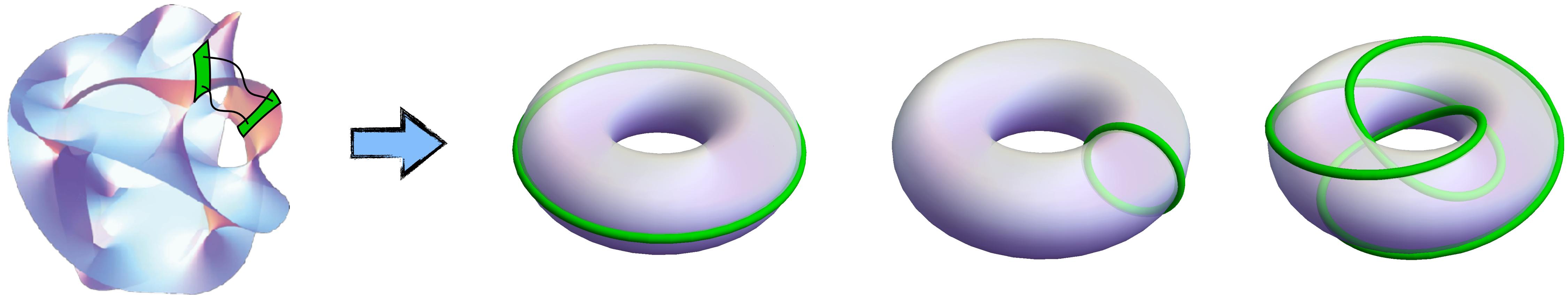
- Diophantine equations ubiquitous in ST (topological data, quantization conditions)
- Asking whether an arbitrary Diophantine equation has a solution (let alone finding one) is undecidable
- However, Diophantine equations in string theory are not arbitrary but inherit structure from consistency conditions, ...

- ▶ **Idea to solve the problem:** [Halverson, Nelson, FR `19]

- Set up a “game” in RL to solve a particular set of coupled Diophantine equations related to flux vacua of type II orientifolds. This showed that the NN
  - ◆ ... can rediscover human-derived solution strategy that leads to partial decoupling
  - ◆ ... can find new, more efficient strategies

# First Example - Finding string solutions

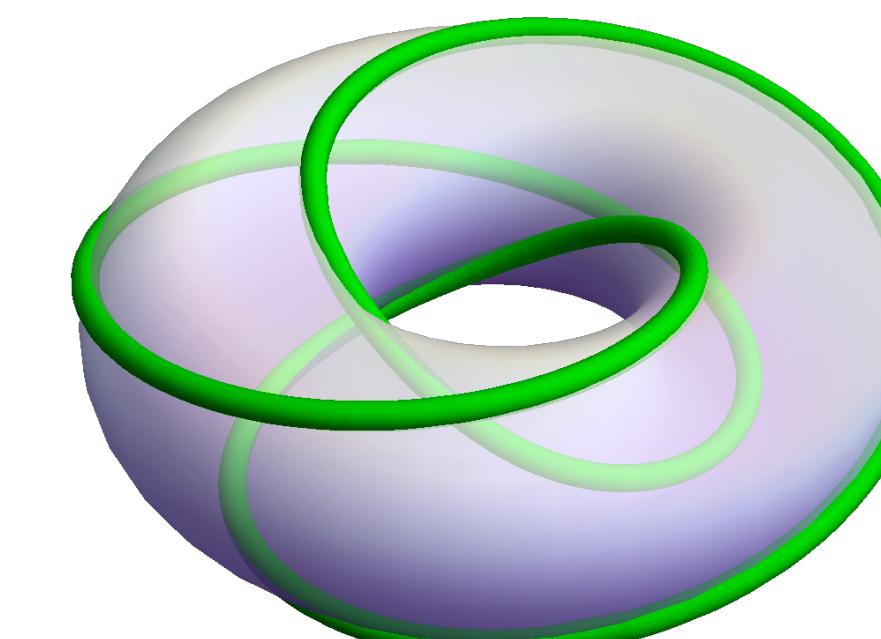
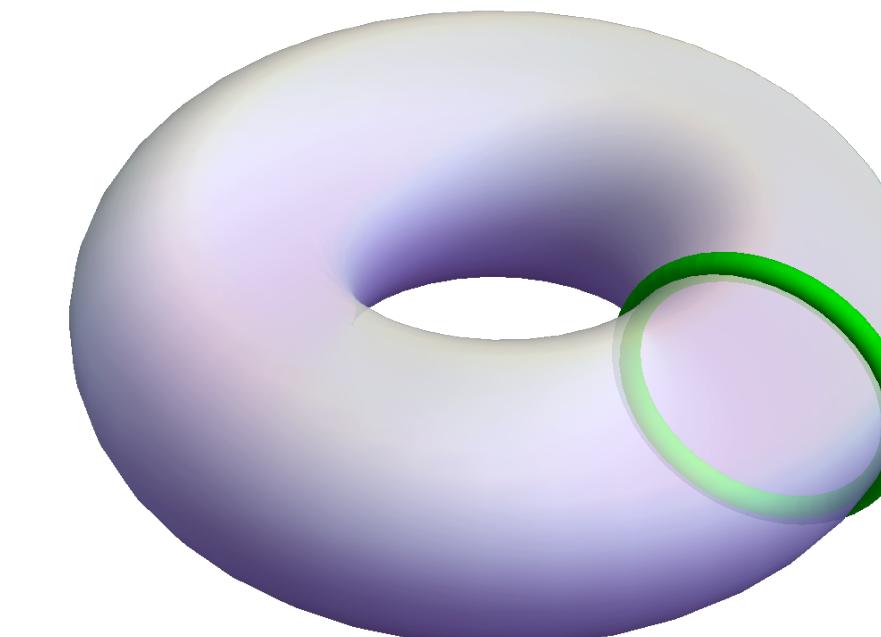
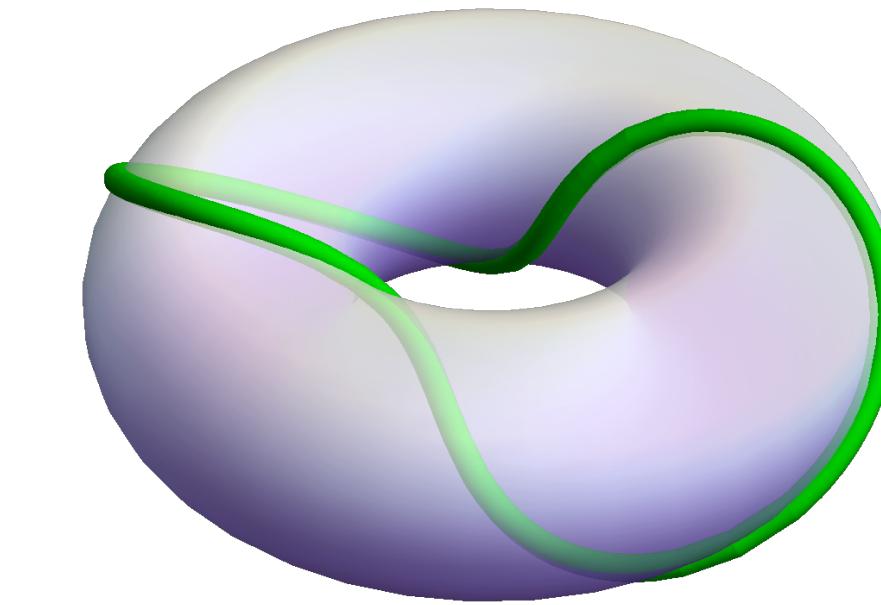
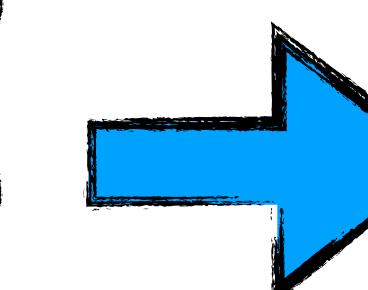
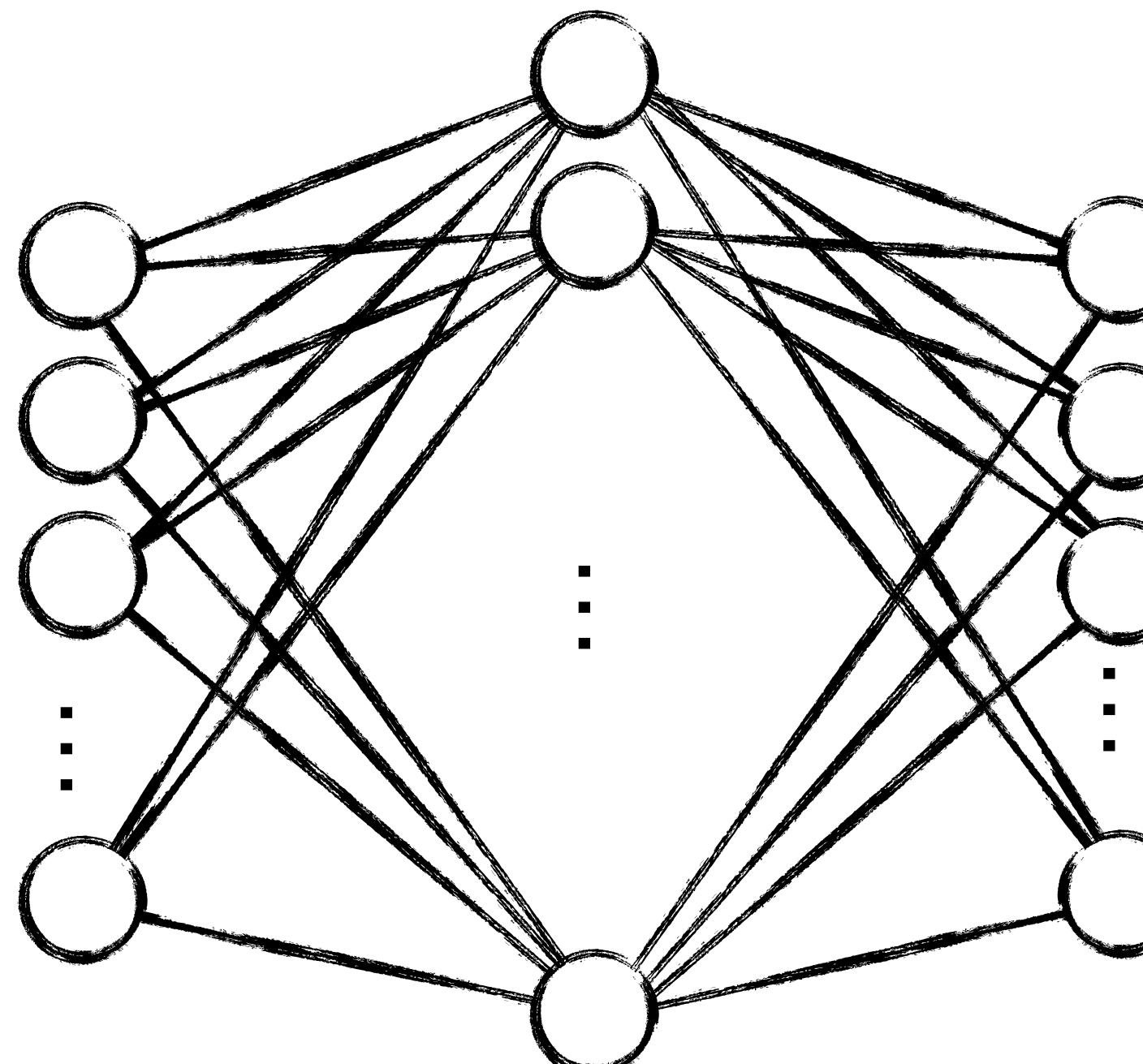
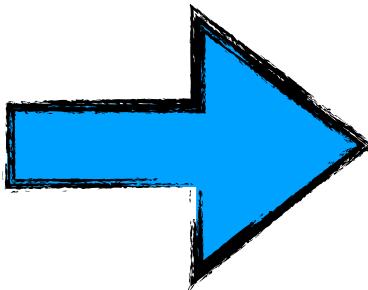
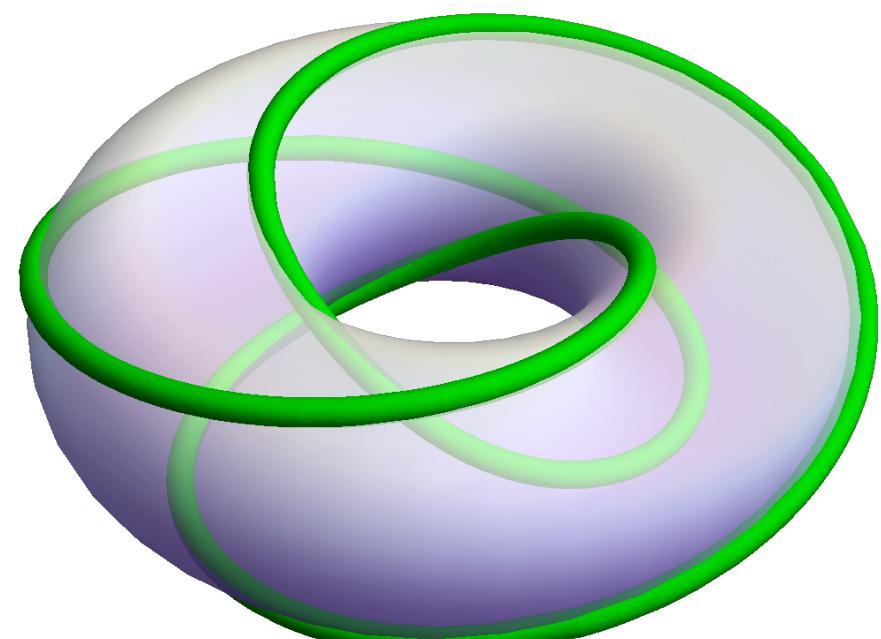
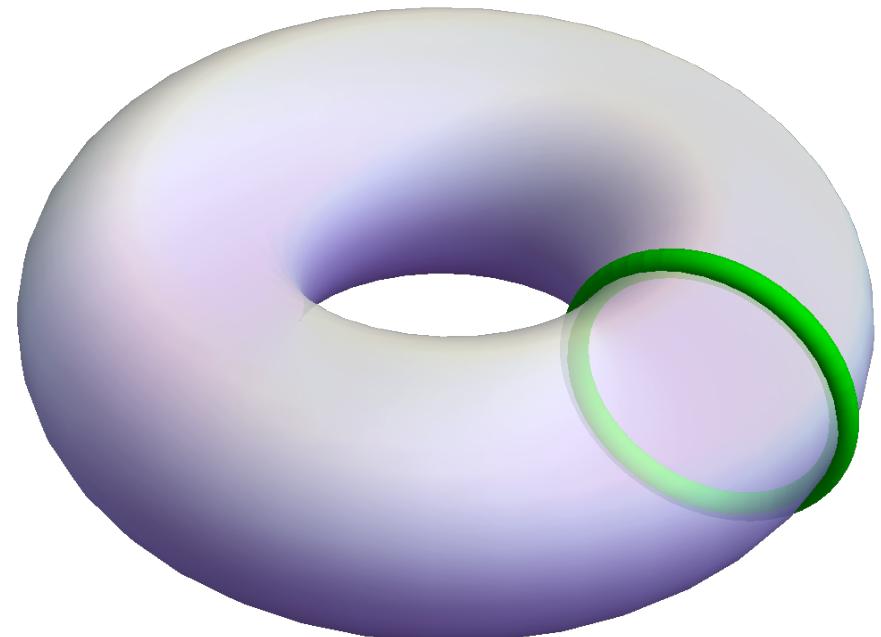
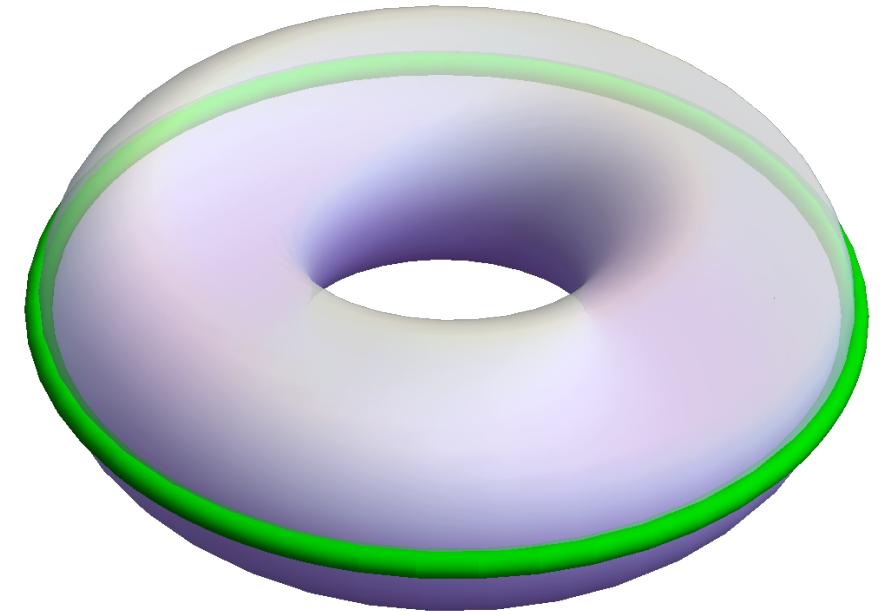
---



- Wrap branes around torus cycles and stack multiple branes on top of each other
- Brane stacks  $\Leftrightarrow$  Tuple:  $(N, n_1, m_1, n_2, m_2, n_3, m_3)$
- There is a finite (but huge) number of inequivalent configurations

# First Example - Finding string solutions

---



# First Example - Finding string solutions

---

Condition TC:

$$\#_{\text{stacks}} \begin{pmatrix} N^a n_1^a n_2^a n_3^a \\ -N^a n_1^a m_2^a m_3^a \\ -N^a m_1^a n_2^a m_3^a \\ -N^a m_1^a m_2^a n_3^a \end{pmatrix} = \begin{pmatrix} 8 \\ 4 \\ 4 \\ 8 \end{pmatrix}$$

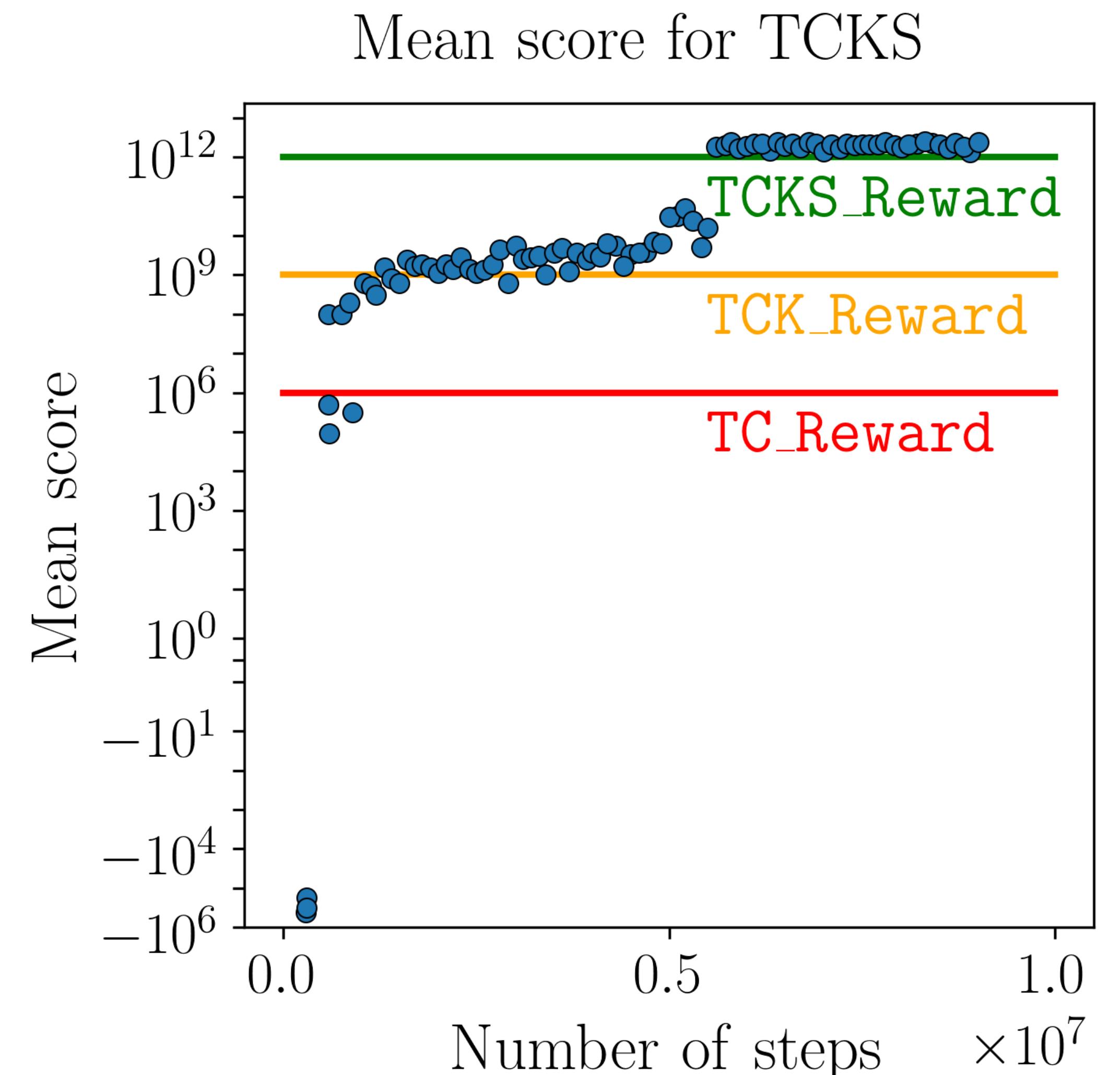
Condition K:

$$\#_{\text{stacks}} \begin{pmatrix} 2N^a m_1^a m_2^a m_3^a \\ -N^a m_1^a n_2^a n_3^a \\ -N^a n_1^a m_2^a n_3^a \\ -2N^a n_1^a n_2^a m_3^a \end{pmatrix} \bmod \begin{pmatrix} 2 \\ 2 \\ 2 \\ 2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

Condition S:

$$m_1^a m_2^a m_3^a - j m_1^a n_2^a n_3^a - k n_1^a m_2^a n_3^a - \ell n_1^a n_2^a m_3^a = 0$$

$$n_1^a n_2^a n_3^a - j n_1^a m_2^a m_3^a - k m_1^a n_2^a m_3^a - \ell m_1^a m_2^a n_3^a > 0$$



# First Example - Finding string solutions

---

**Condition K:**

$$\#_{\text{stacks}} \sum_{a=1} \begin{pmatrix} 2N^a m_1^a m_2^a m_3^a \\ -N^a m_1^a n_2^a n_3^a \\ -N^a n_1^a m_2^a n_3^a \\ -2N^a n_1^a n_2^a m_3^a \end{pmatrix} \text{ mod } \begin{pmatrix} 2 \\ 2 \\ 2 \\ 2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

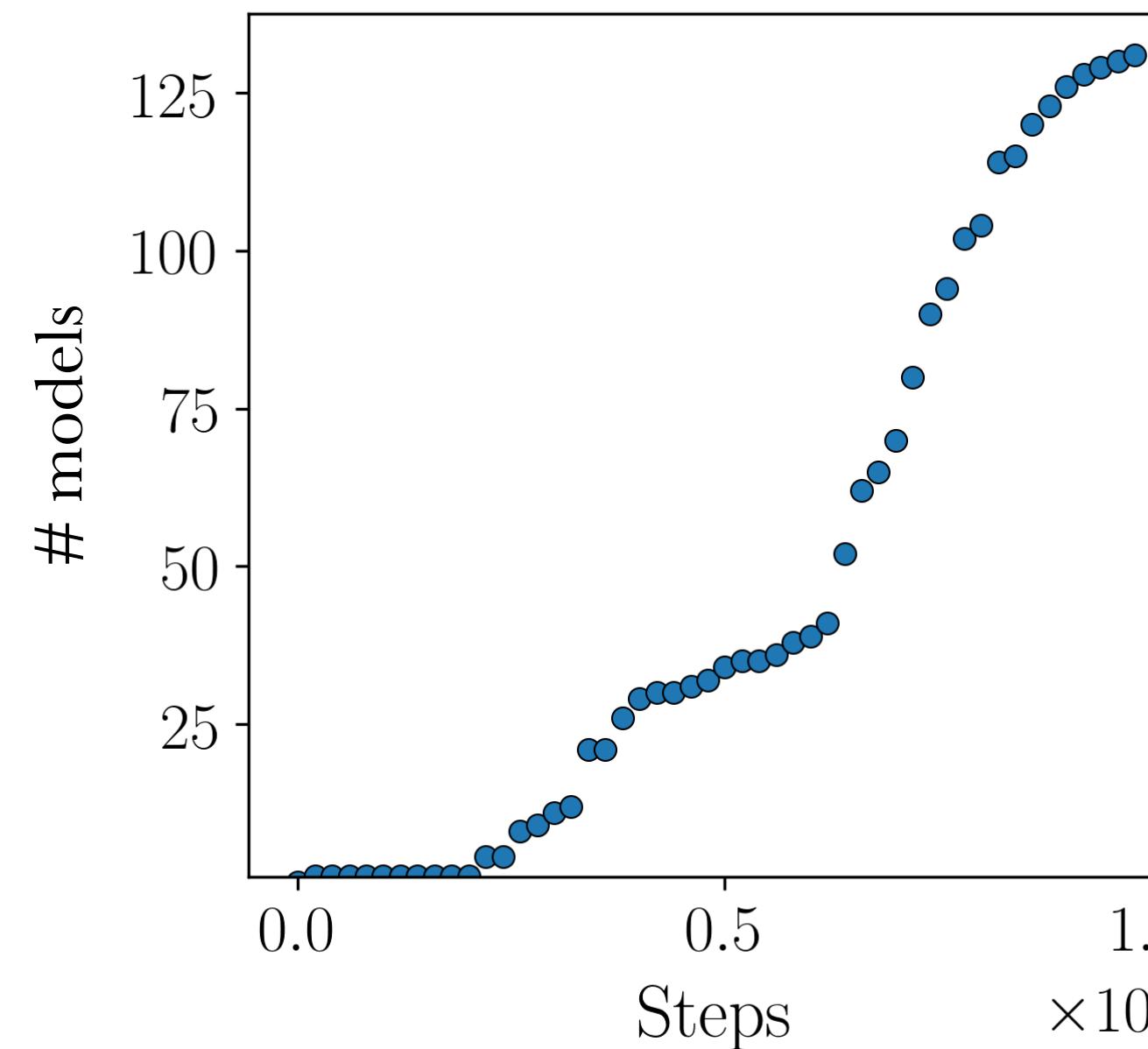
**Condition S:**

$$m_1^a m_2^a m_3^a - j m_1^a n_2^a n_3^a - k n_1^a m_2^a n_3^a - \ell n_1^a n_2^a m_3^a = 0$$

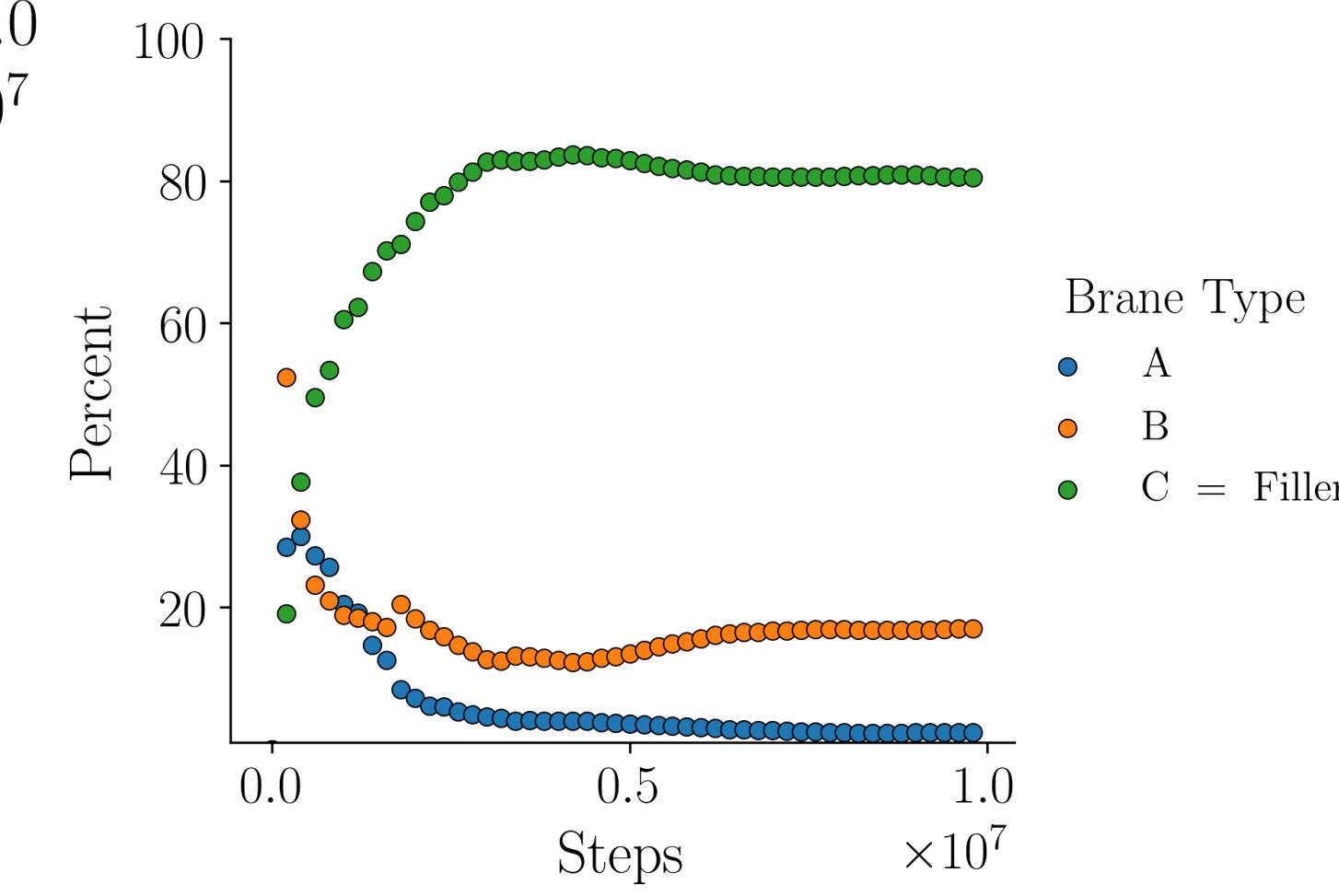
$$n_1^a n_2^a n_3^a - j n_1^a m_2^a m_3^a - k m_1^a n_2^a m_3^a - \ell m_1^a m_2^a n_3^a > 0$$

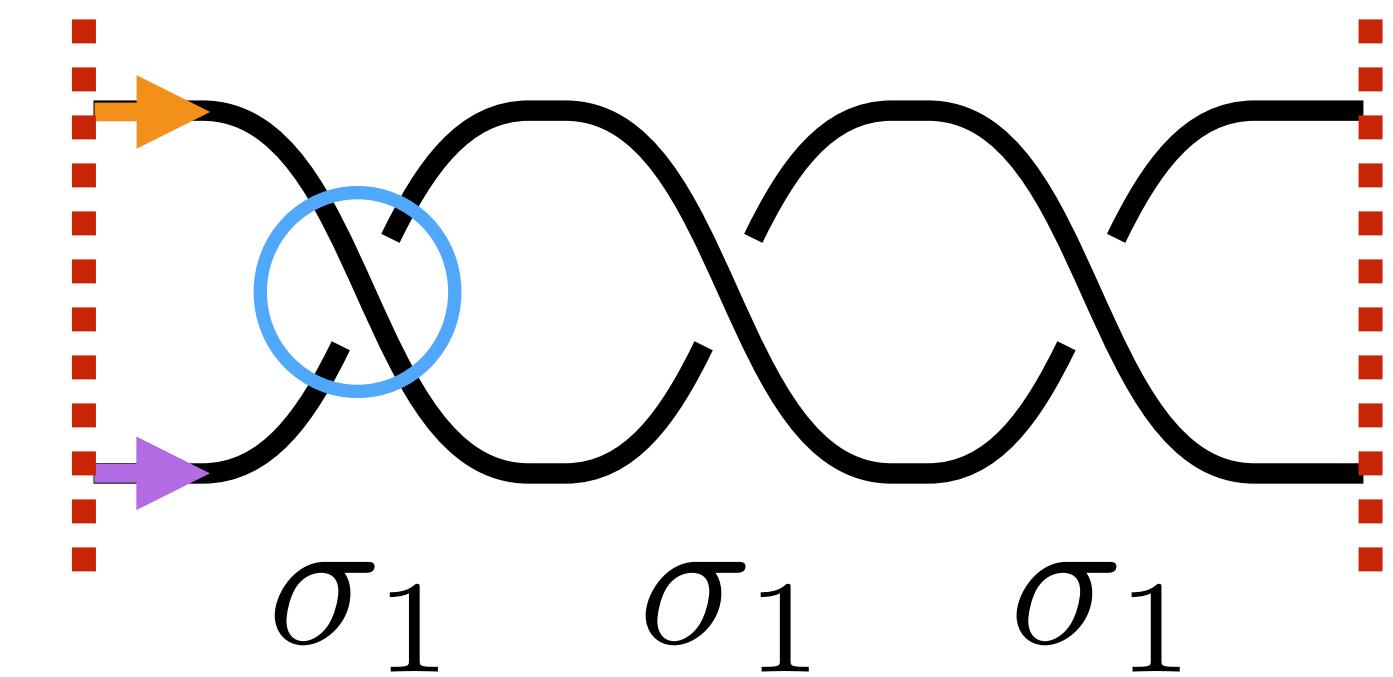
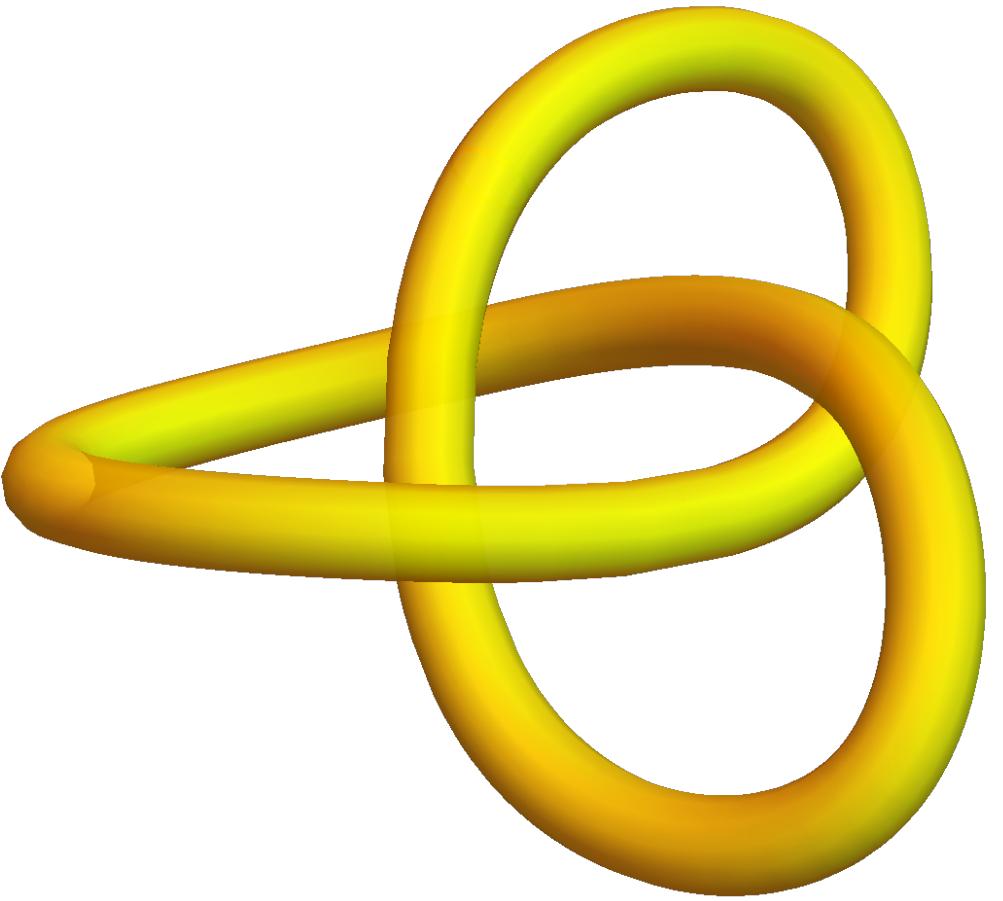
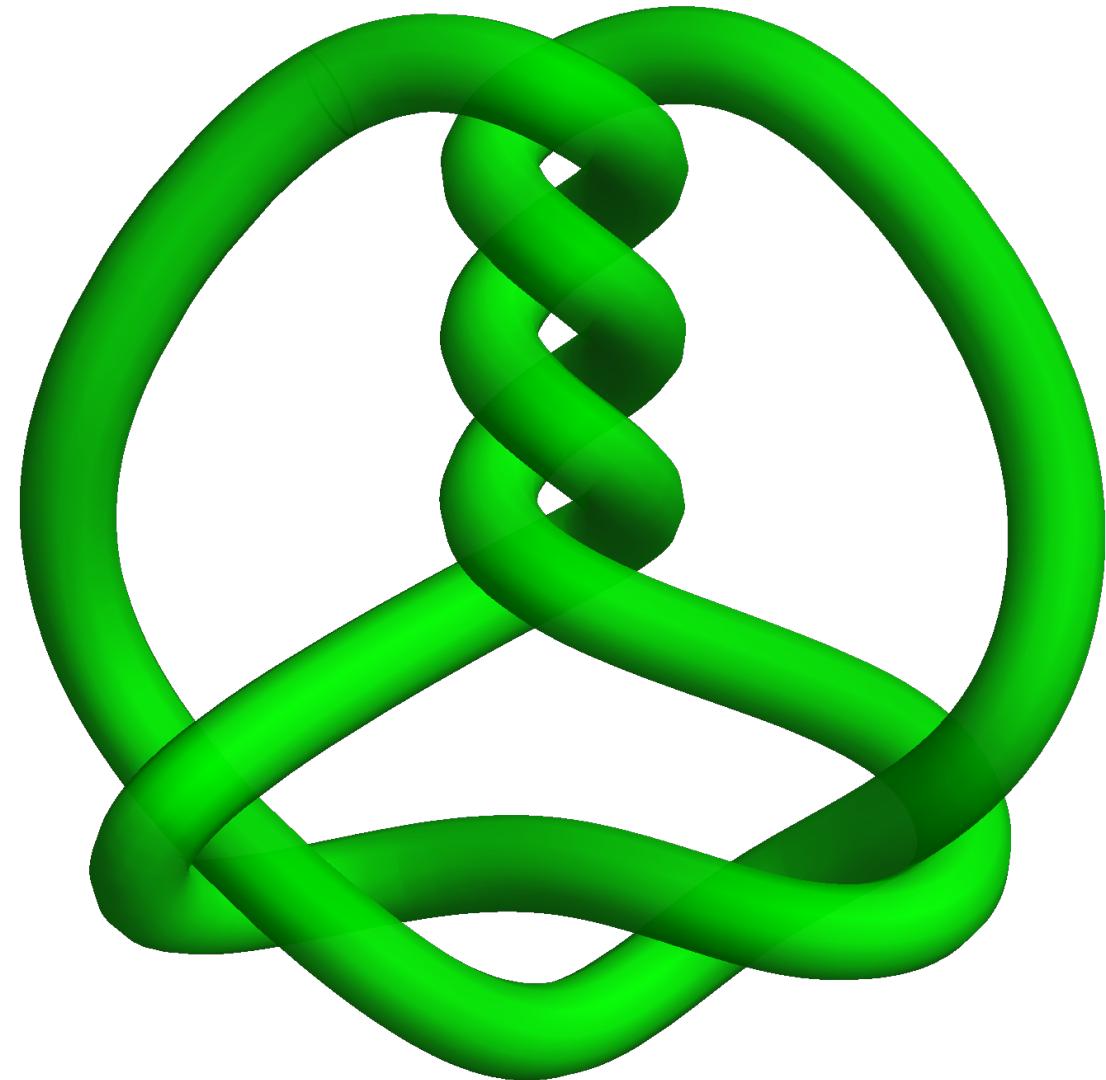
**Condition TC:**

$$\#_{\text{stacks}} \sum_{a=1} \begin{pmatrix} N^a n_1^a n_2^a n_3^a \\ -N^a n_1^a m_2^a m_3^a \\ -N^a m_1^a n_2^a m_3^a \\ -N^a m_1^a m_2^a n_3^a \end{pmatrix} = \begin{pmatrix} 8 \\ 4 \\ 4 \\ 8 \end{pmatrix}$$



Learning Filler Brane Strategy





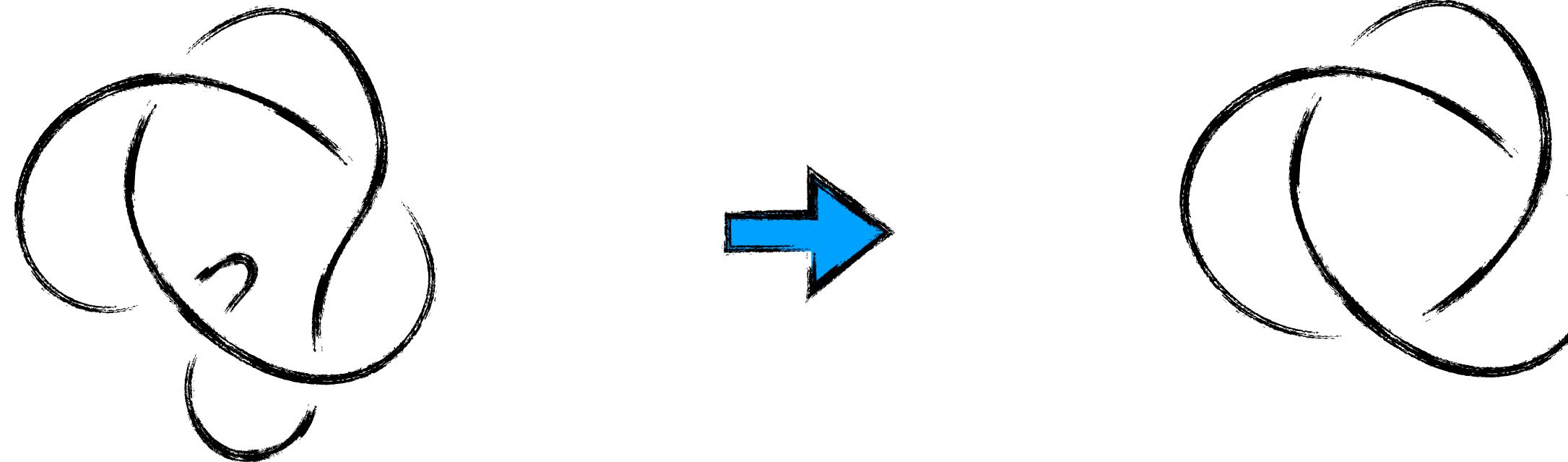
## Example II: Knot theory

---

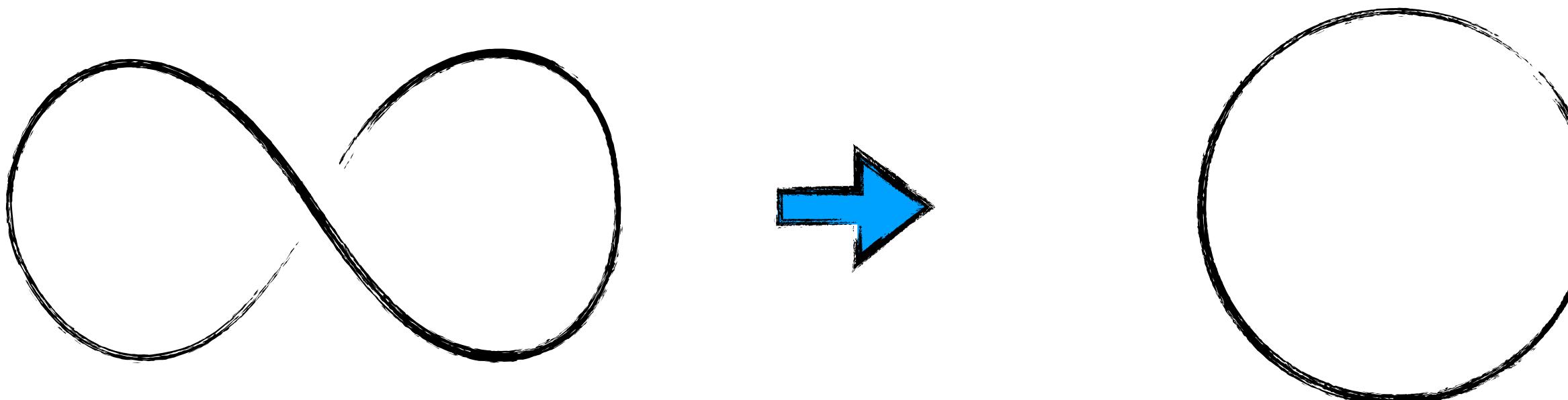
# The unknot problem

---

- ▶ Simplify a knot as much as possible

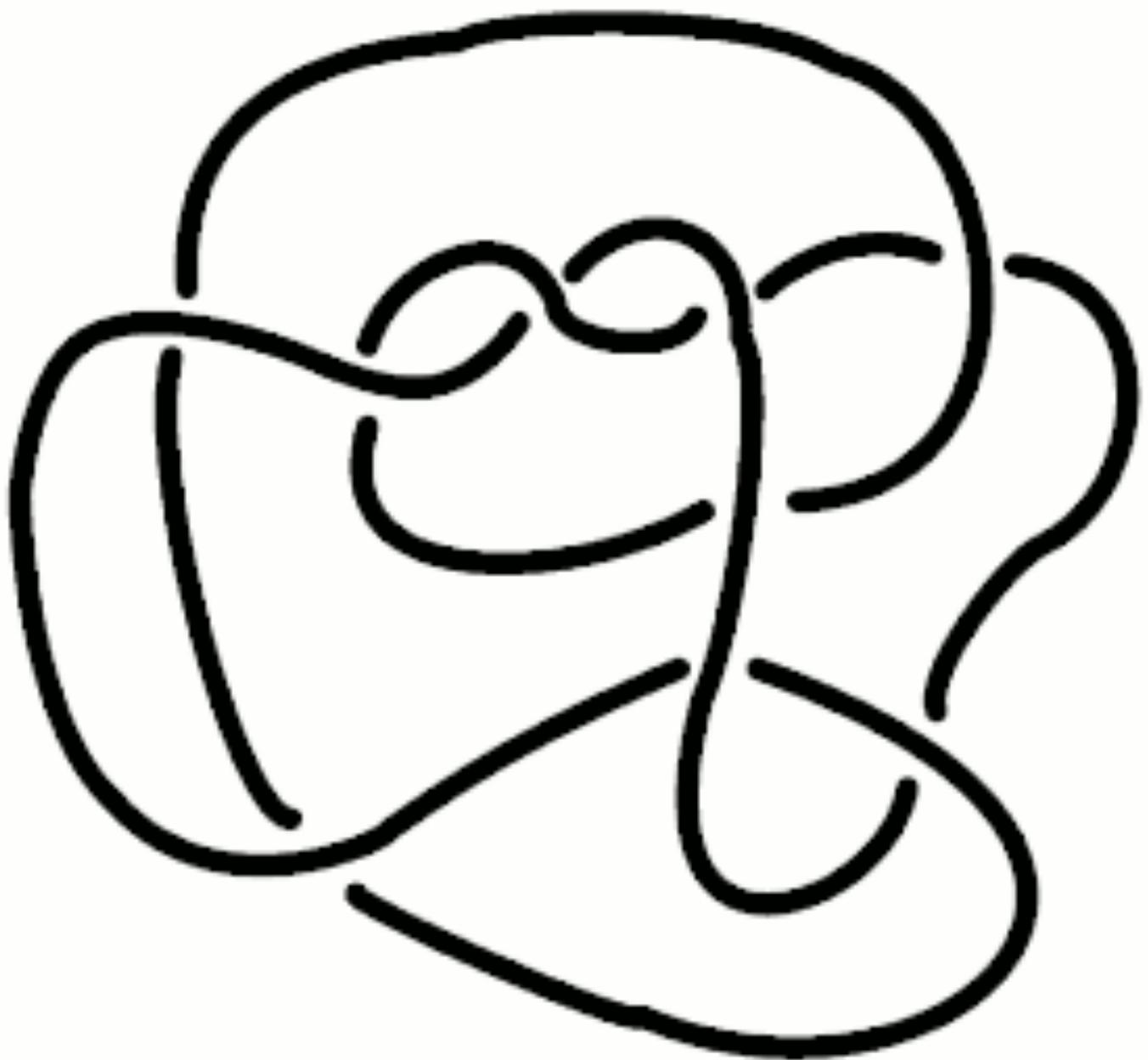


- ▶ If the knot can be made trivial (i.e. a circle), it is the unknot



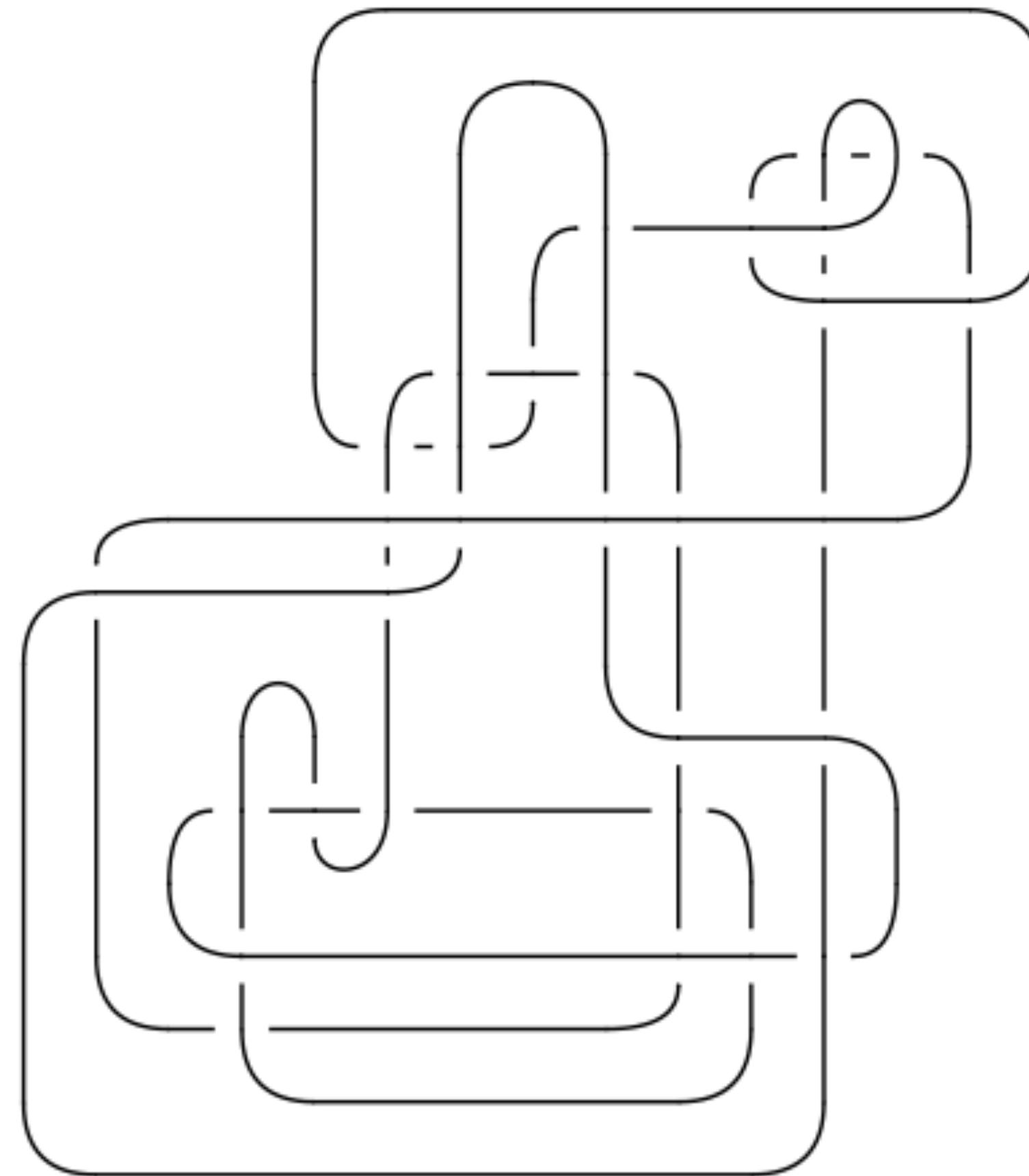
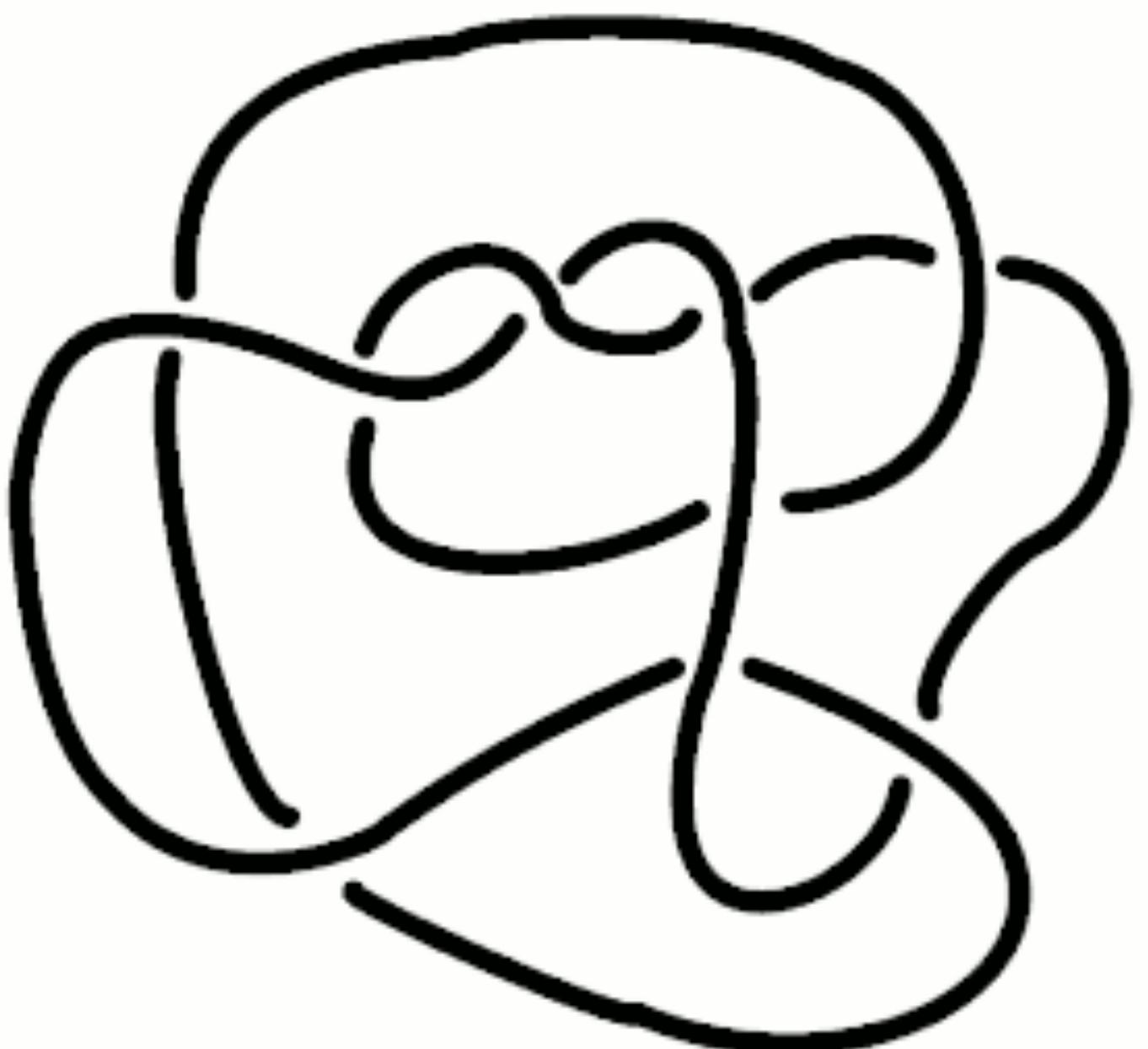
# Unknot Problem

---



# Unknot Problem

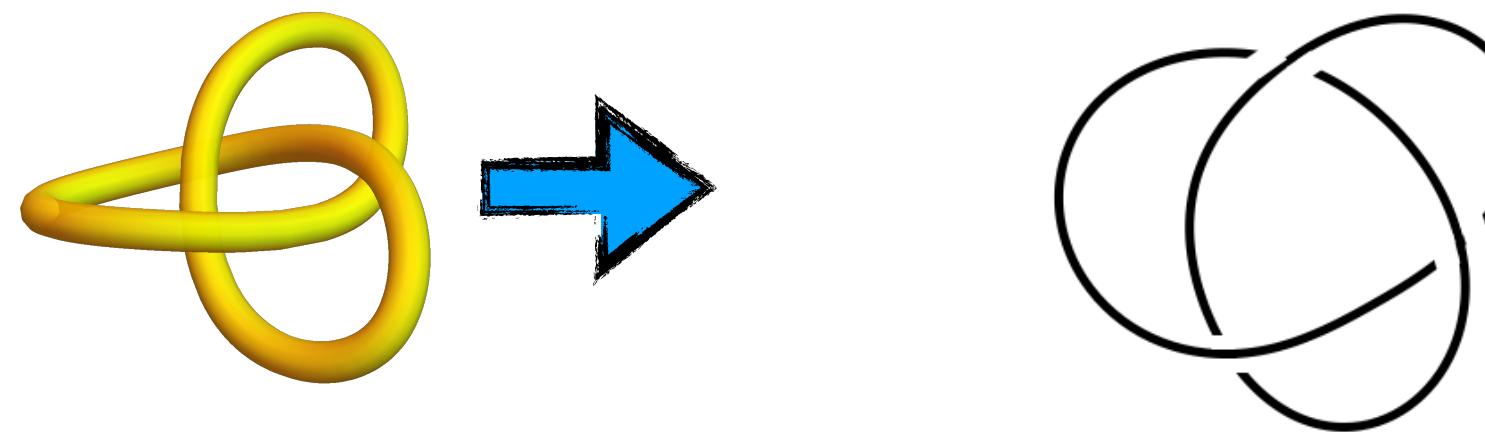
---



# Knot Theory as a natural language problem

---

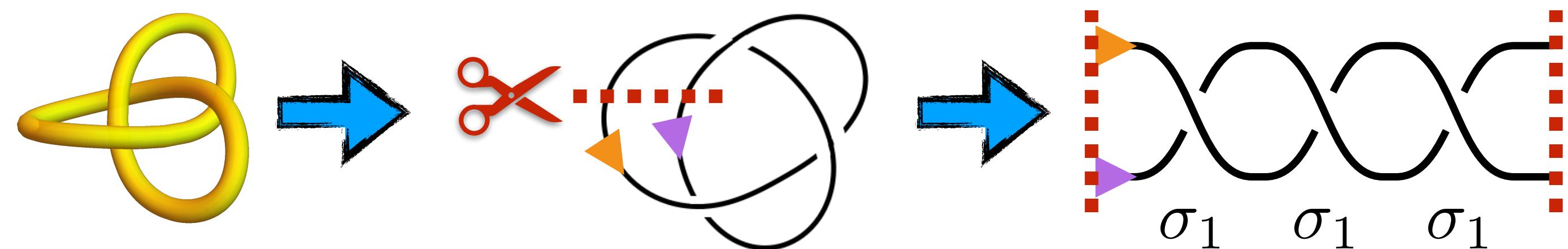
- ▶ Knots can be represented as words over some alphabet



# Knot Theory as a natural language problem

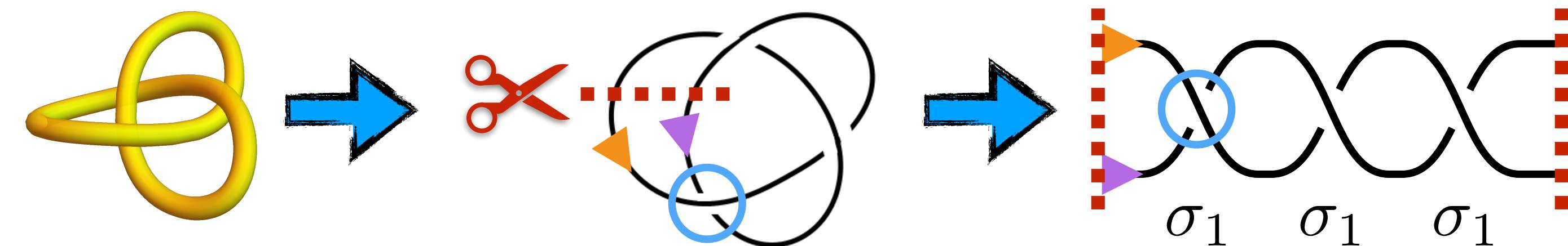
---

- ▶ Knots can be represented as words over some alphabet

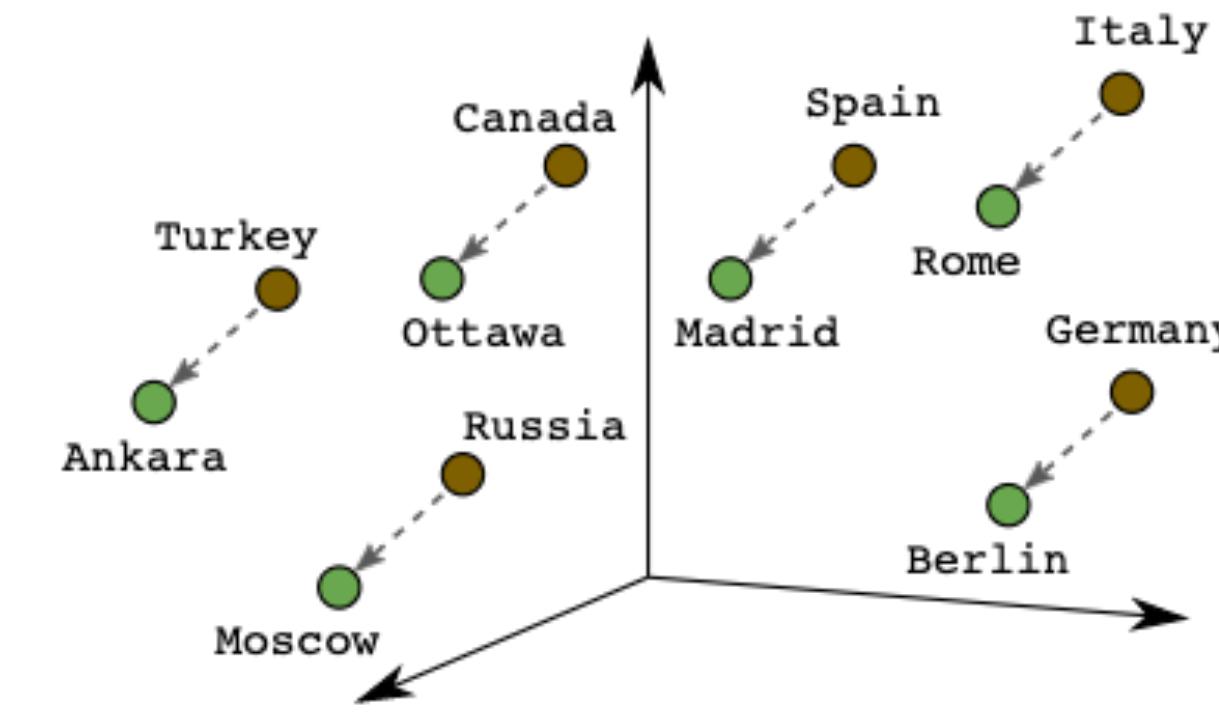


# Knot Theory as a natural language problem

- ▶ Knots can be represented as words over some alphabet

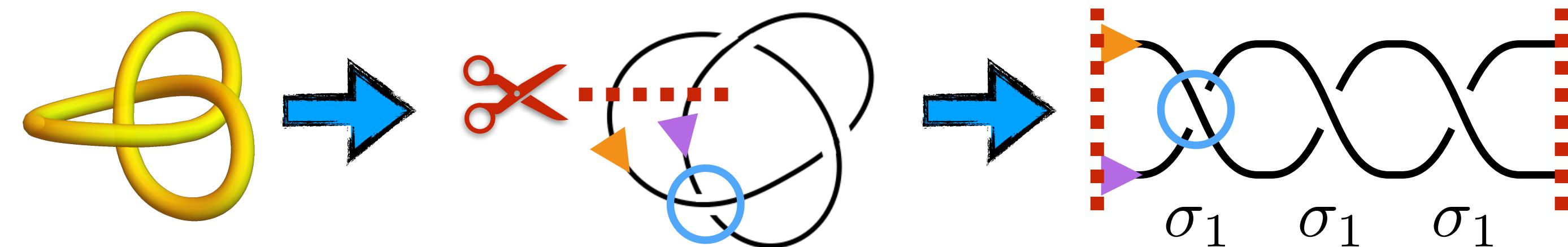


- ▶ NLP is something that ML is very good at

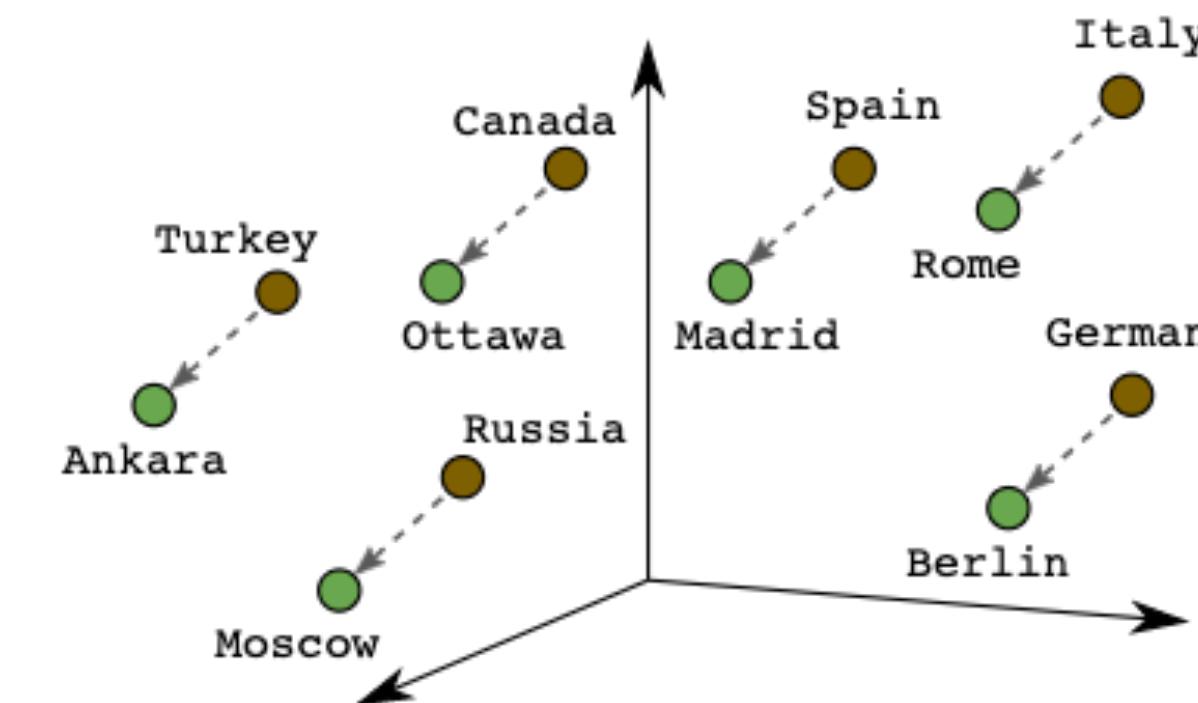


# Knot Theory as a natural language problem

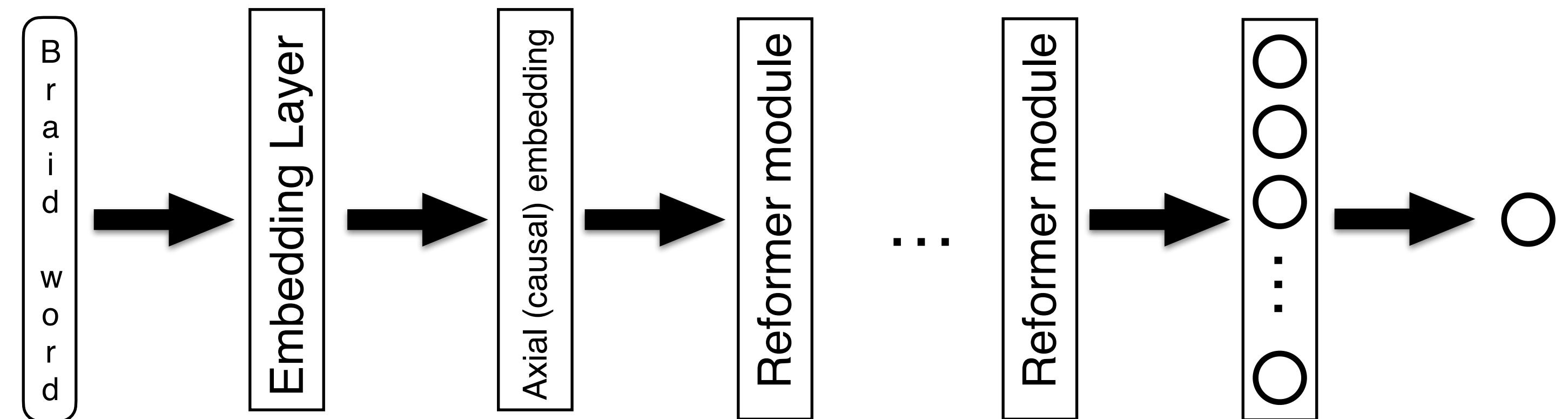
- ▶ Knots can be represented as words over some alphabet



- ▶ NLP is something that ML is very good at

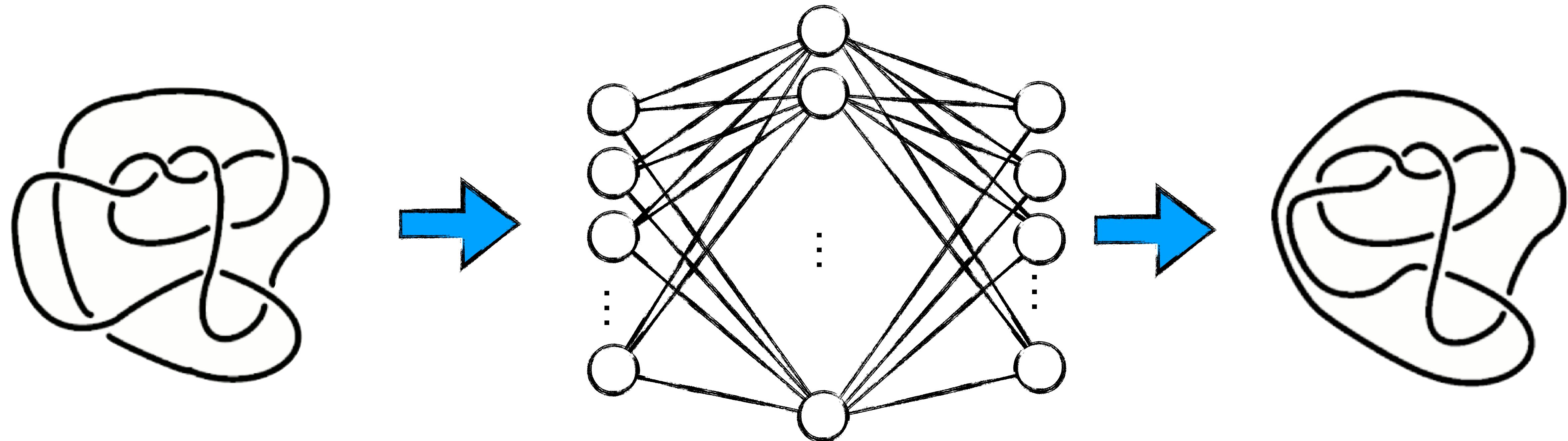


- ▶ We use a transformer to tackle the unknot problem



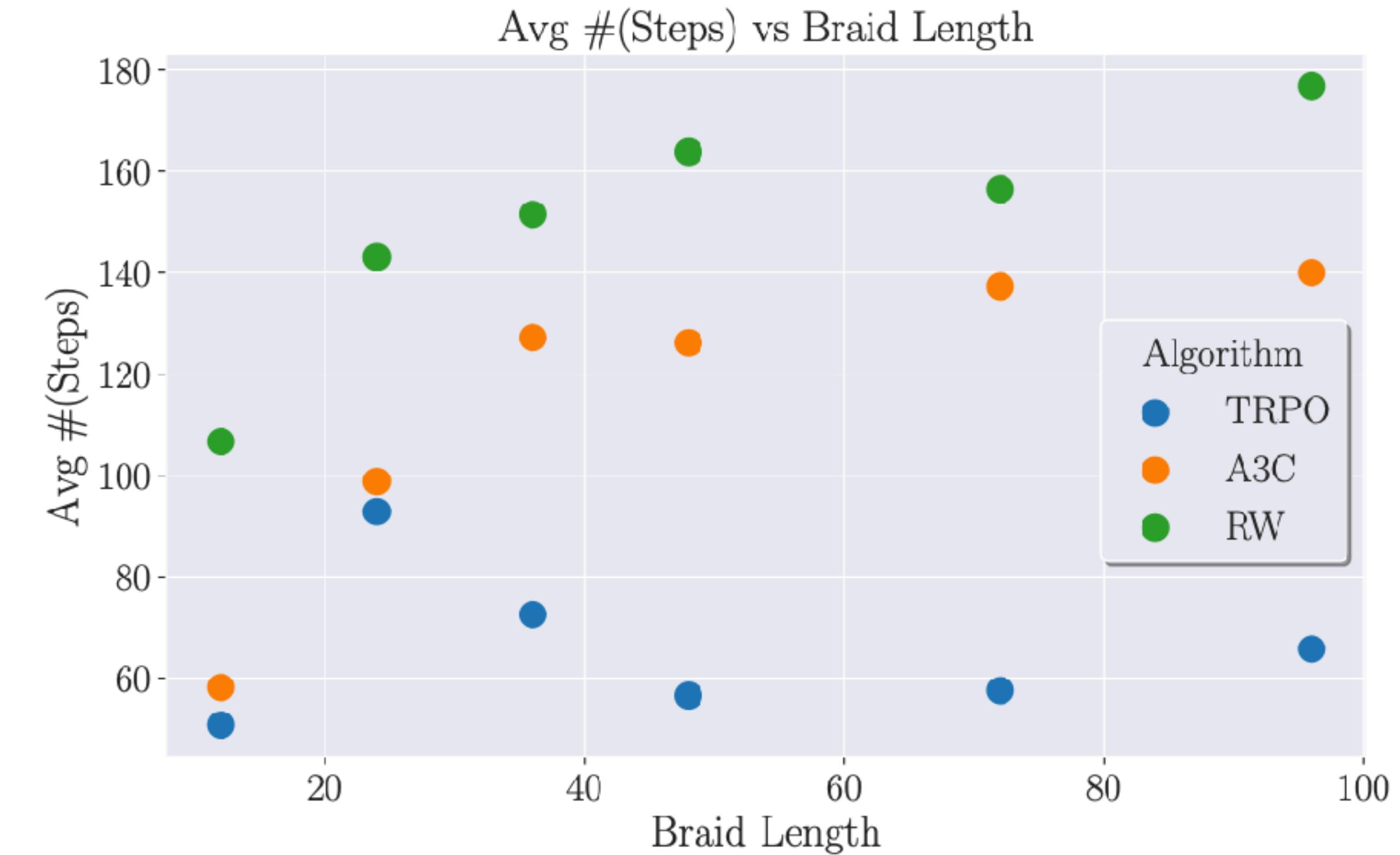
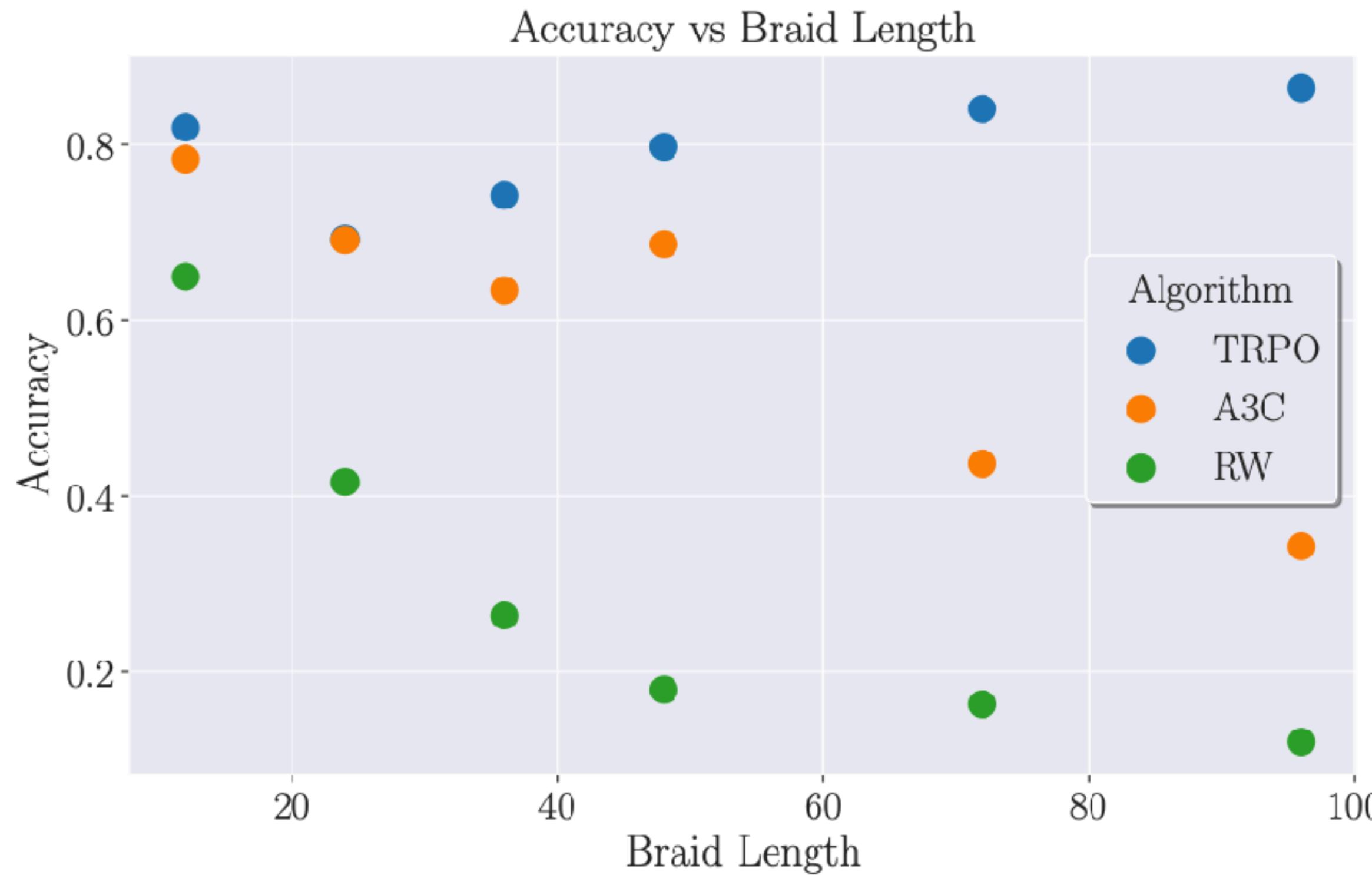
# RL for Knot Theory

---



# Unknot Problem

---



# Knot theory

---

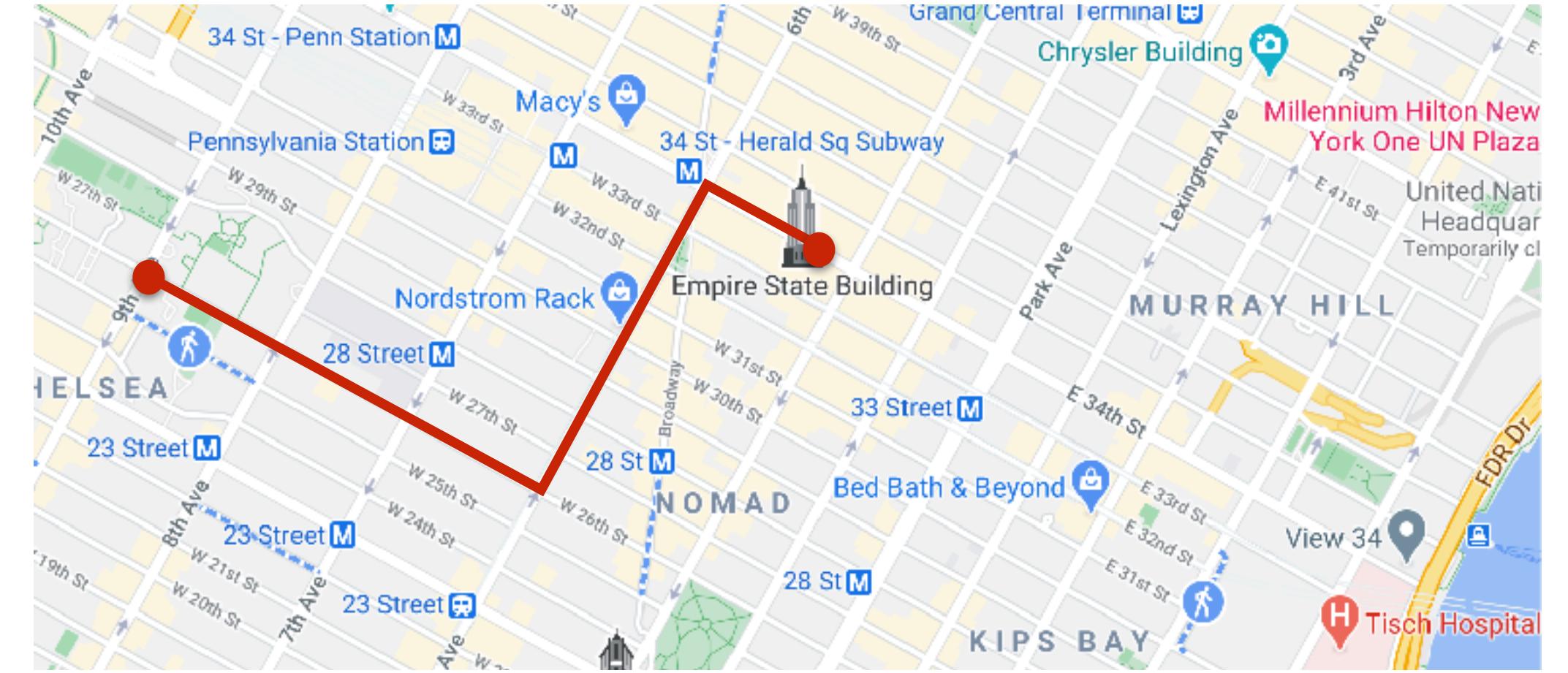
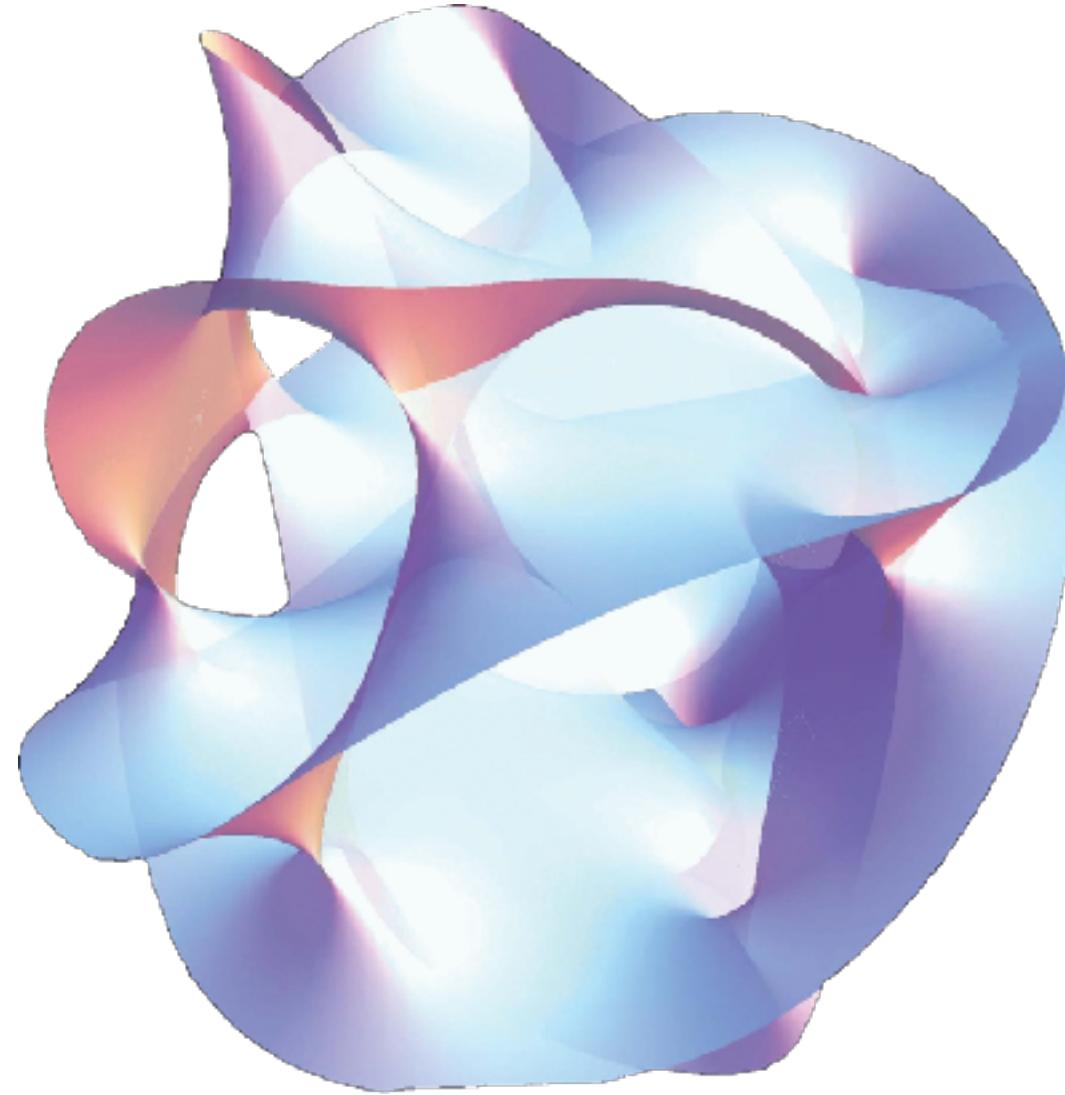
- **Generating Conjectures:**

Train NNs on some knot representation/invariant. If they learn to predict something, hints at a (previously unknown) connection

- Basic knot invariants  $\Leftrightarrow$  quasi-positivity, slice genus, OS  $\tau$ -invariance [Hughes '16]
- Jones Polynomial  $\rightarrow$  hyperbolic knot volume [Jejjala,Kar,Parrikar '19; Craven,Jejjala,Kar '20]

- **Checking conjectures and mining counter-examples:**

- sliceness [Hughes '16]
- smooth Poincare conjecture?



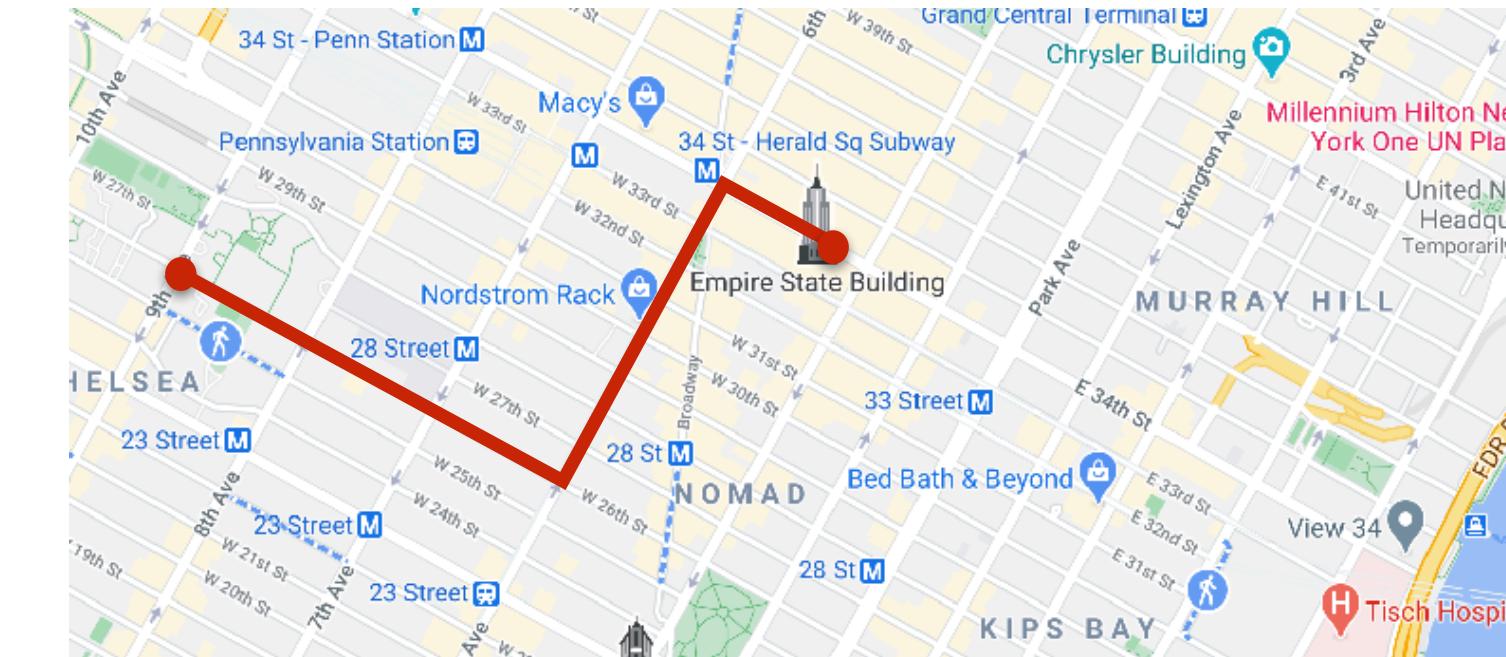
## Example III: Calabi-Yau metrics

# Metrics

- Metrics measure distances, but the choice is not unique

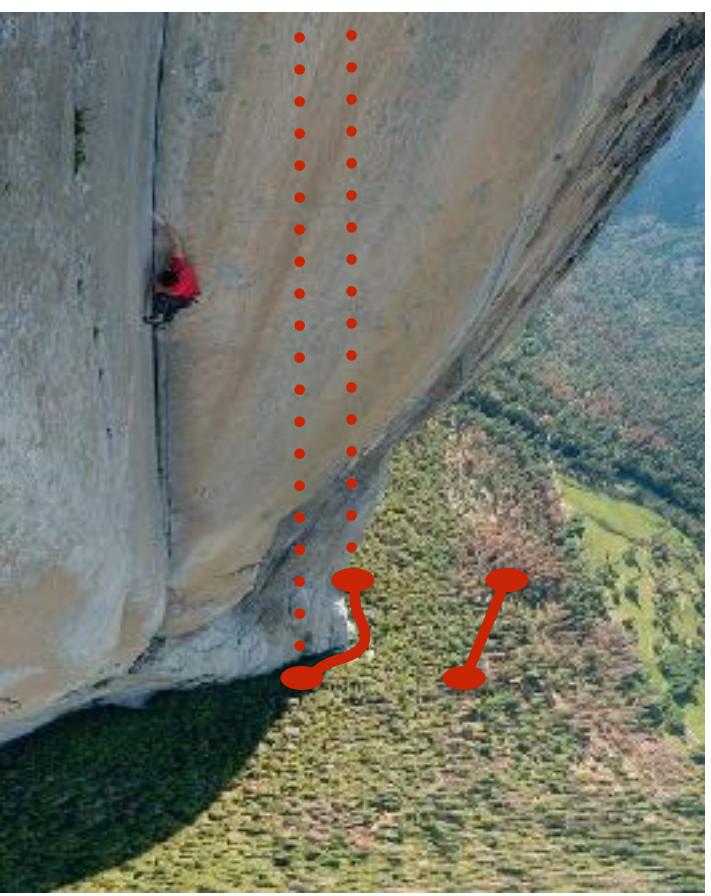


[Source: wikipedia]

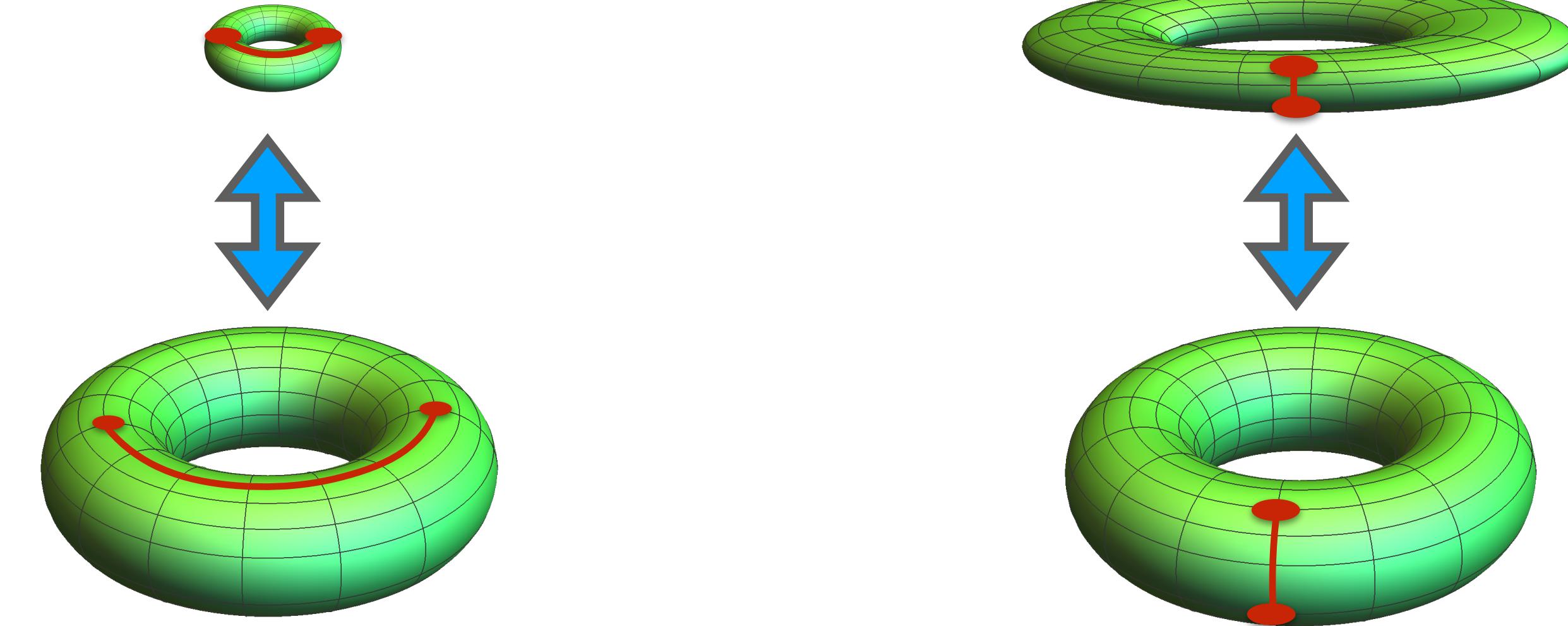


[Source: google maps]

- If space is curved, metric depends on the point you are at. It also depends on volume/shape



[Photo Credit: Jimmy Chin]



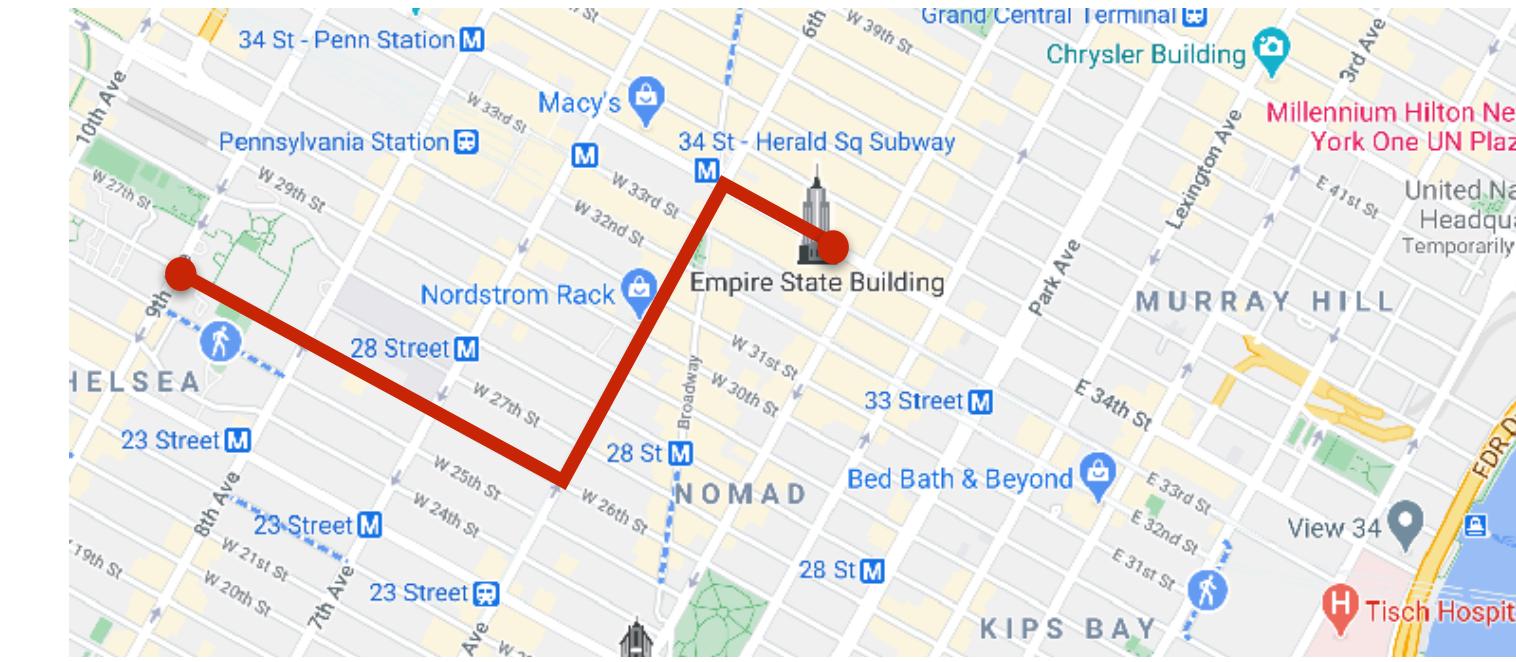
# Metrics

---

- ▶ Metrics measure distances, but the choice is not unique



[Source: wikipedia]



[Source: google maps]

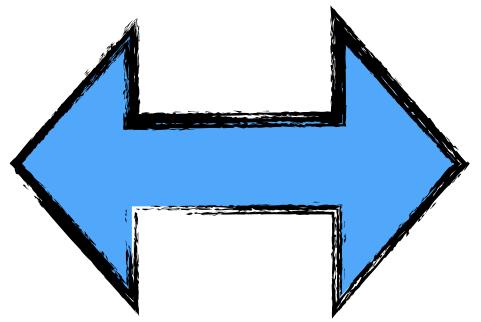
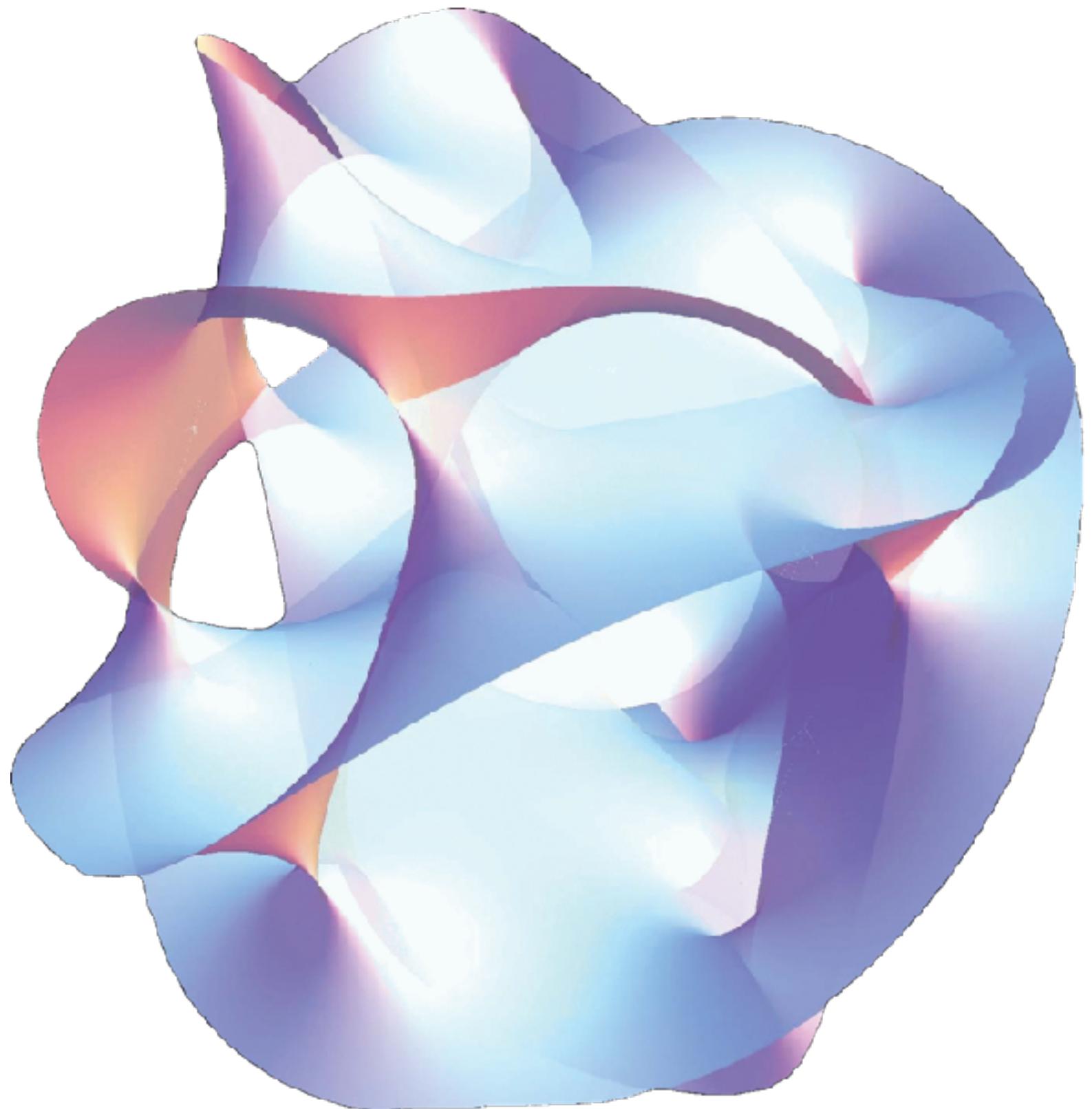
Think of a metric  $g$  as a function

$$g : \text{position} \times \text{volume} \times \text{shape} \rightarrow \mathbb{R}^{d \times d}$$

and optimize a NN to represent this function  
subject to the consistency conditions  
imposed by string theory

# Calabi-Yau manifolds - Properties

---

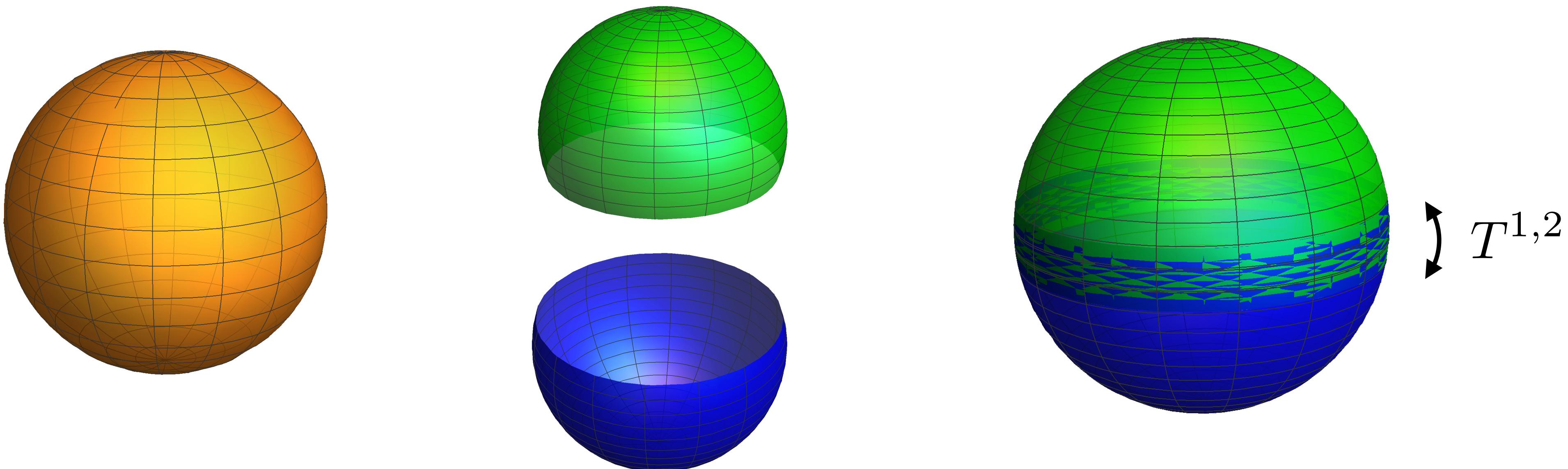


- ▶ Complex
- ▶ Kähler
- ▶ Ricci-flat

# CY Property 1 - Complex

---

- ▶ In general manifolds cannot be covered by a single patch
- ▶ On each patch, one can choose a local description, coordinate system, etc. But one must make sure that the descriptions can be matched on the overlap and everything can be patched to a complex manifold globally (e.g. choice  $i = \sqrt{-1}$  vs  $i = -\sqrt{-1}$  , ...)

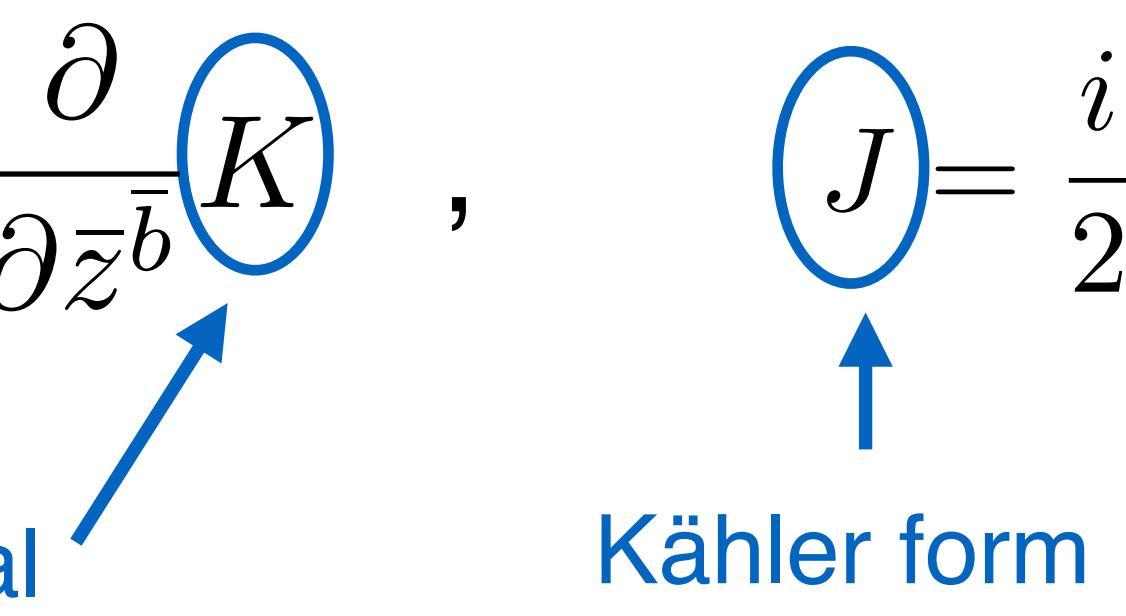


# CY Property 2 - Kähler

---

- ▶ The space must be Kähler
- ▶ This means that the metric can be written in terms of derivatives of a real, scalar function called the **Kähler potential**  $K$

$$g_{a\bar{b}} = \frac{\partial}{\partial z^a} \frac{\partial}{\partial \bar{z}^b} K , \quad J = \frac{i}{2} \sum_{a < b} g_{a\bar{b}} \varepsilon^{a\bar{b}} dz^a d\bar{z}^b , \quad z = x + iy , \quad \bar{z} = x - iy$$



**Kähler potential**                            **Kähler form**

- ▶ In general, integrating the metric to find the Kähler potential is hard. So one can either start with a Kähler potential and derive the metric, or one has to solve the differential equation  $\frac{\partial J}{\partial z^a} = \frac{\partial J}{\partial \bar{z}^b} = 0$ .

# CY Property 3 - Ricci-flat

---

- Calabi-Yau spaces are spaces on which a metric exists that is “flat enough”, i.e. their Ricci tensor vanishes

$$\begin{aligned} R_{ij} = & -\frac{1}{2} \sum_{a,b=1}^n \left( \frac{\partial^2 g_{ij}}{\partial x^a \partial x^b} + \frac{\partial^2 g_{ab}}{\partial x^i \partial x^j} - \frac{\partial^2 g_{ib}}{\partial x^j \partial x^a} - \frac{\partial^2 g_{jb}}{\partial x^i \partial x^a} \right) g^{ab} \\ & + \frac{1}{2} \sum_{a,b,c,d=1}^n \left( \frac{1}{2} \frac{\partial g_{ac}}{\partial x^i} \frac{\partial g_{bd}}{\partial x^j} + \frac{\partial g_{ic}}{\partial x^a} \frac{\partial g_{jd}}{\partial x^b} - \frac{\partial g_{ic}}{\partial x^a} \frac{\partial g_{jb}}{\partial x^d} \right) g^{ab} g^{cd} \\ & - \frac{1}{4} \sum_{a,b,c,d=1}^n \left( \frac{\partial g_{jc}}{\partial x^i} + \frac{\partial g_{ic}}{\partial x^j} - \frac{\partial g_{ij}}{\partial x^c} \right) \left( 2 \frac{\partial g_{bd}}{\partial x^a} - \frac{\partial g_{ab}}{\partial x^d} \right) g^{ab} g^{cd} \\ & = 0 \end{aligned}$$

- Note that ensuring  $g$  is Kähler introduces 2 more derivatives since  $g_{a\bar{b}} = \frac{\partial}{\partial z^a} \frac{\partial}{\partial \bar{z}^b} K$

# CY Property 3 - Ricci-flat

---

- ▶ This fourth-order partial differential equation is extremely hard to solve
- ▶ We can improve on this. On a CY, one can write down

$$J = \frac{i}{2} \sum_{a < b} g_{a\bar{b}} \varepsilon^{a\bar{b}} dz^a d\bar{z}^{\bar{b}} \quad \Rightarrow \quad J^3 = -\frac{i}{8} \sqrt{\det g} dz_1 d\bar{z}_1 dz_2 d\bar{z}_2 dz_3 d\bar{z}_3$$

$$\Omega = \left( \frac{\partial p}{\partial z_4} \right)^{-1} dz_1 dz_2 dz_3 \quad \Rightarrow \quad |\Omega|^2 = \left| \frac{\partial p}{\partial z_4} \right|^{-2} dz_1 dz_2 dz_3 d\bar{z}_1 d\bar{z}_2 d\bar{z}_3$$

- ▶ Since the volume form is unique (up to a constant):

$$J^3 = \kappa |\Omega|^2$$

# CY metric ansatze

---

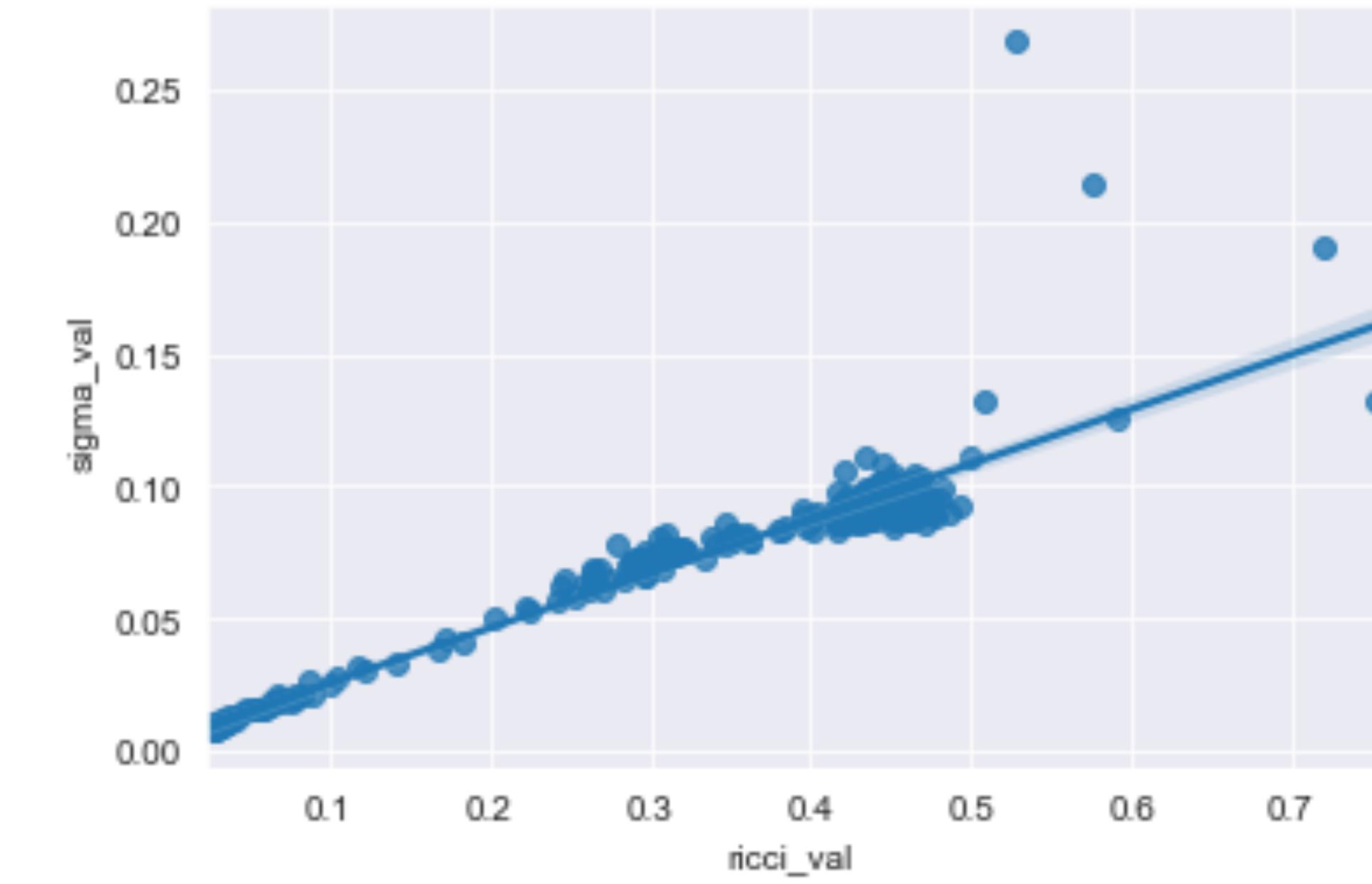
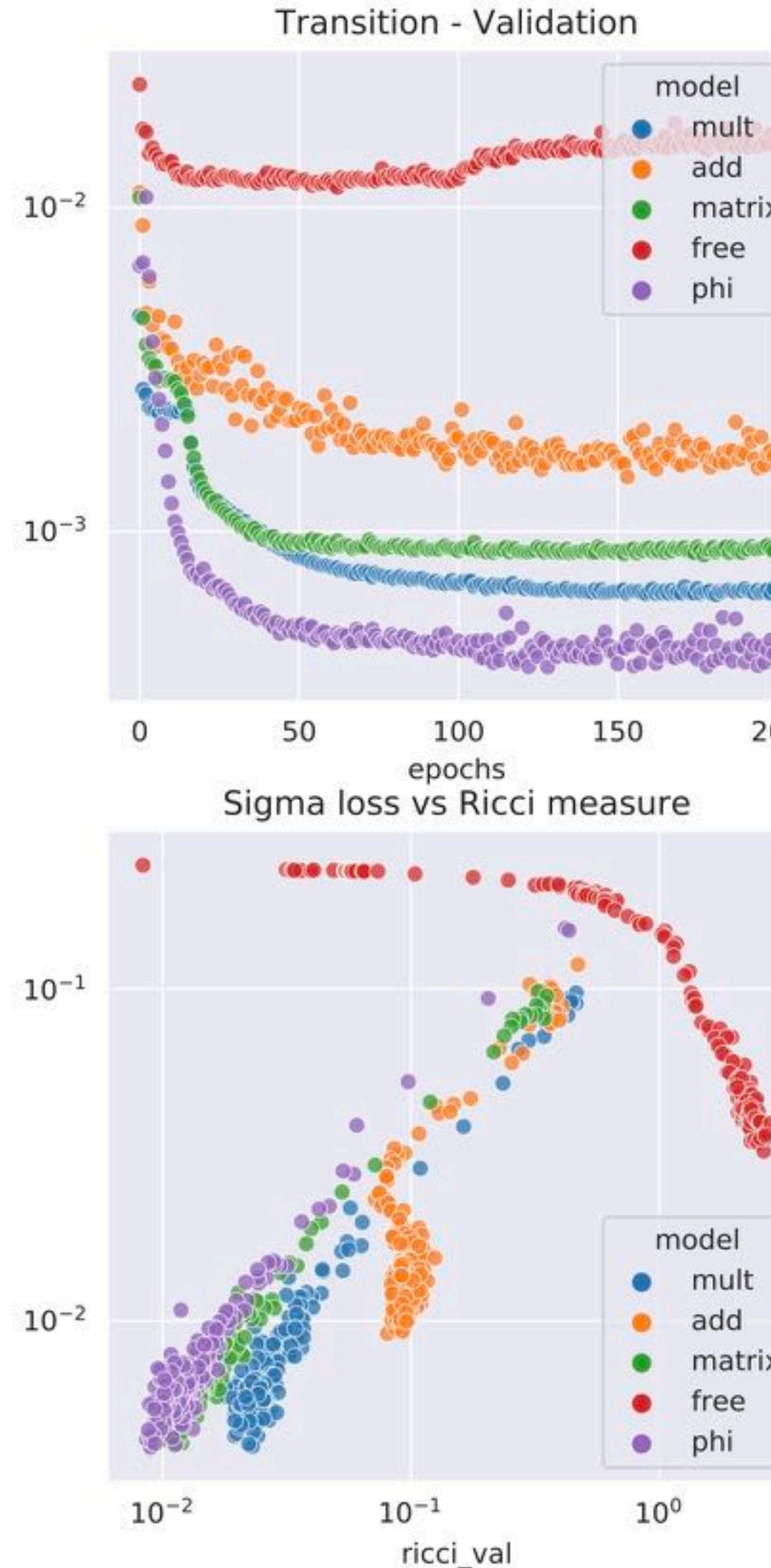
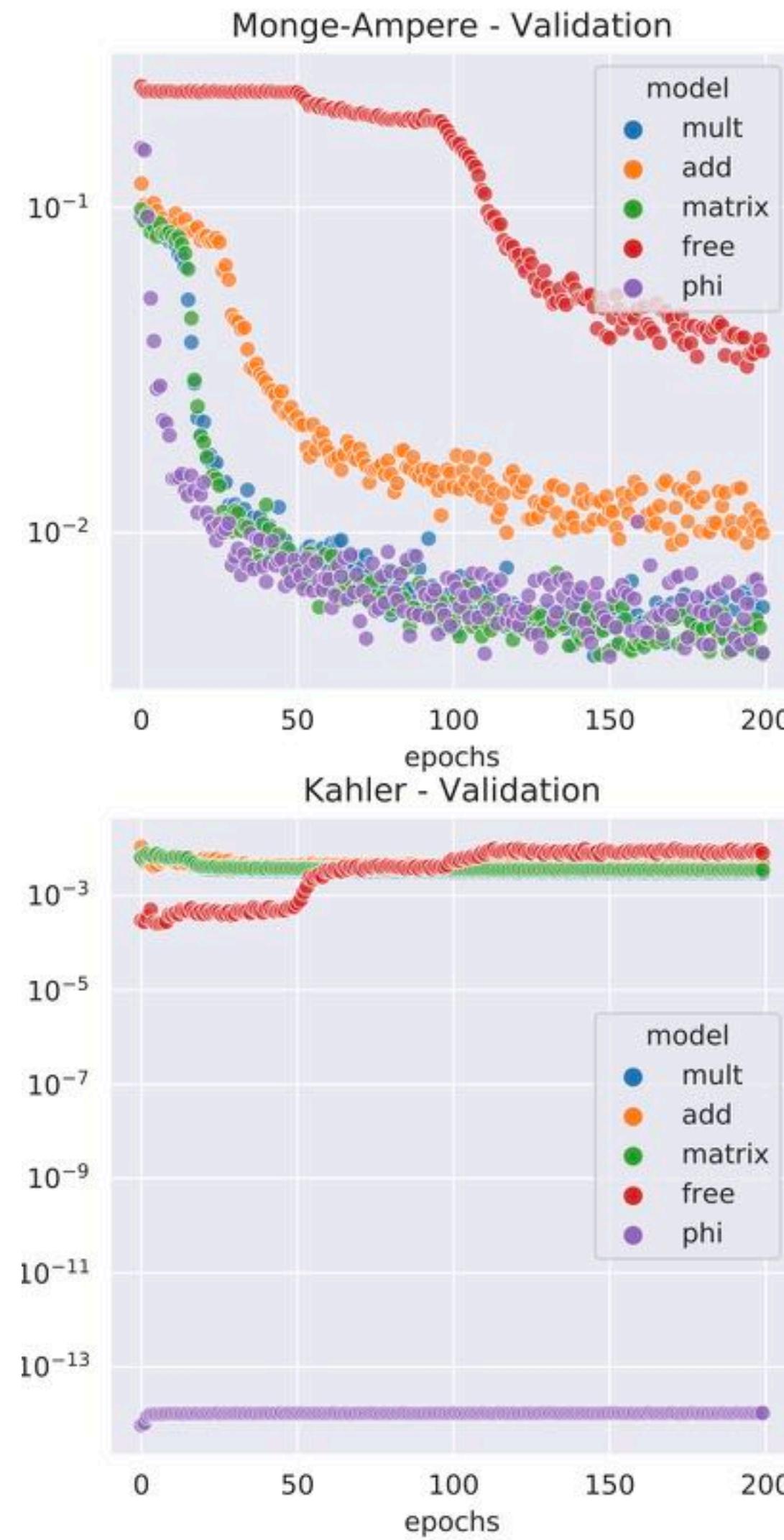
- ▶ The condition  $J^3 = \kappa |\Omega|^2$  can be turned into a (Monge-Ampere) PDE
- ▶ As it turns out, we can ensure the complex and Kahler property and keep the volume moduli fixed if we write

$$g_{\text{CY}} = g_{\text{reference}} + \partial\bar{\partial}\Phi$$

and approximate the (scalar) function  $\Phi = \Phi(\text{position, shape})$  with a NN

- ▶ Other possibilities (can depart from Kahler and fixed volume):
  - $g_{\text{CY}} = g_{\text{NN}}$  (works the least well)
  - $g_{\text{CY}} = g_{\text{reference}} + g_{\text{NN}}$  (works better)
  - $g_{\text{CY}} = g_{\text{reference}}(1 + g_{\text{NN}})$  (works best; as well as the  $\partial\bar{\partial}\Phi$  approach)

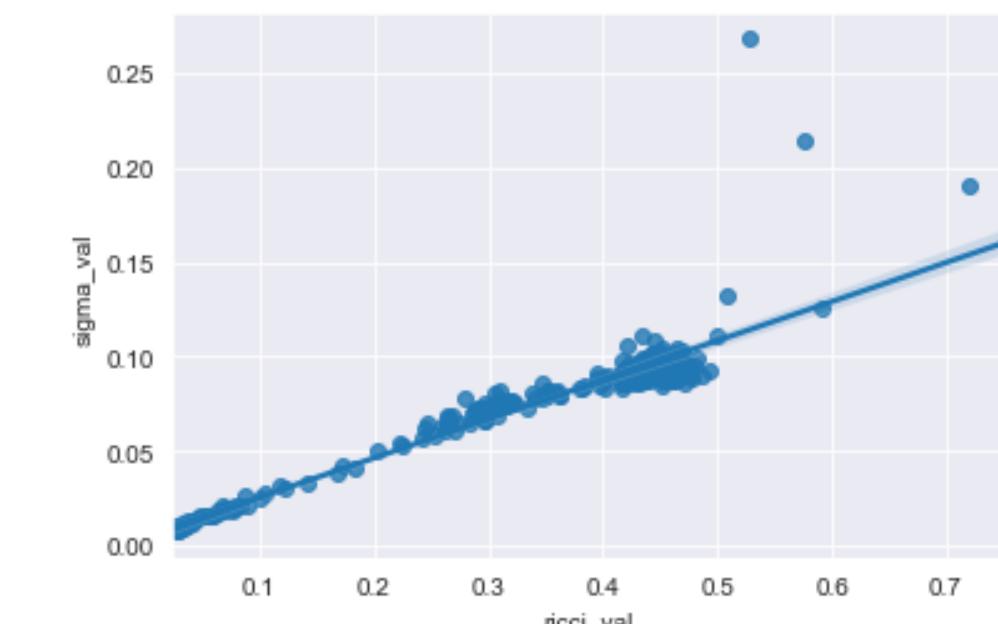
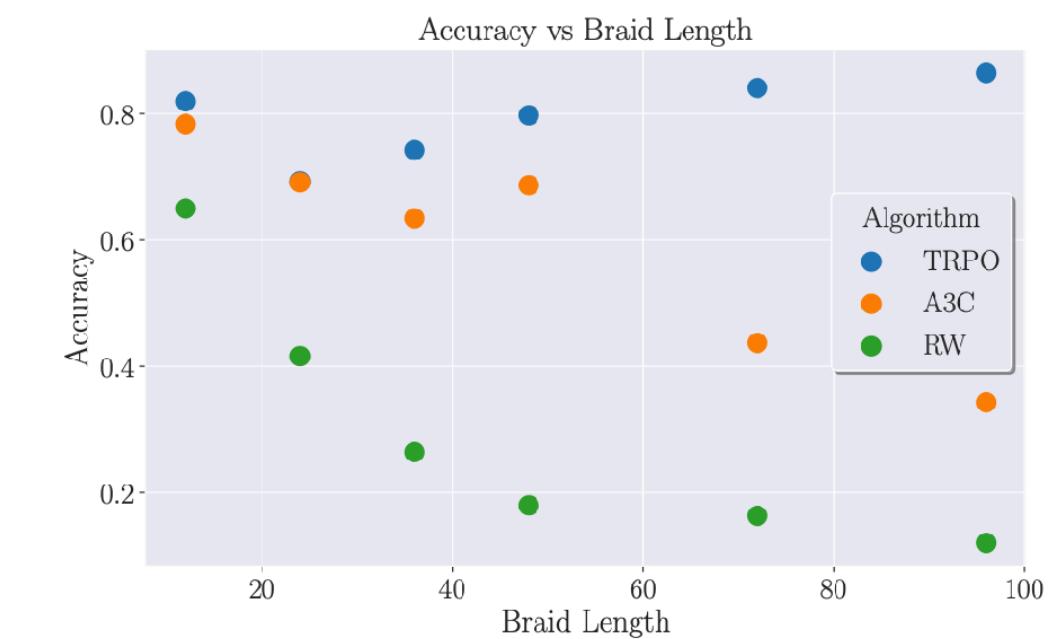
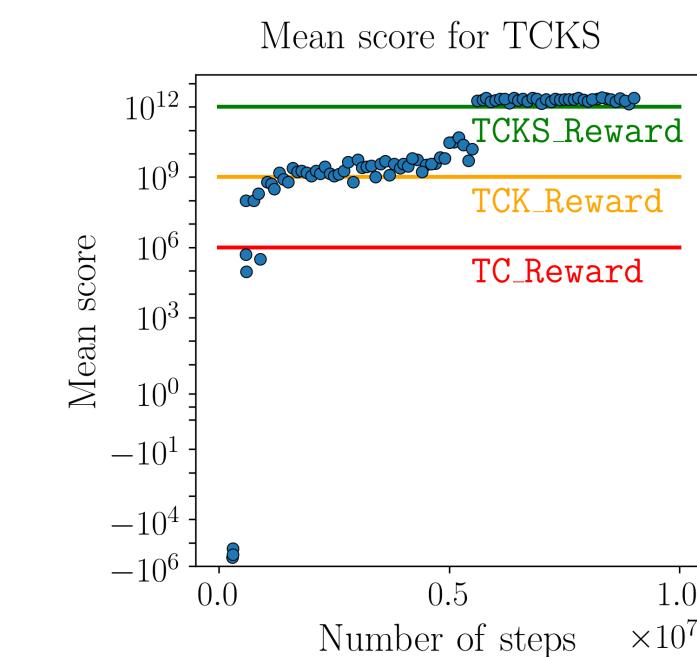
# CY metric results



# Conclusions

---

- ▶ String theory comes with discrete, hard combinatorial problems that seem amendable to RL
  - Solve Diophantine equations
  - Find the unknot
- ▶ ML techniques from other areas can be imported and successfully applied
  - Mapping knot theory to NLP
- ▶ String theory's continuous problems can be solved with fast optimization
  - PDE for CY metrics



# Conclusions

---

- String theory comes with discrete, hard combinatorial problems that seem amendable to RL
  - Solve Diophantine equations
  - Find the unknot
- ML techniques are important
  - Map string theory to ML
- String theory can be solved with fast optimization
  - PDE for CY metrics

Thank you for your  
attention!

