

# CY metrics for CICYs and Toric Varieties

---

FABIAN RUEHLE

Strings: Geometry and Symmetries for Phenomenology  
Oaxaca

November 11, 2021

Based on:

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

[Larfors, Lukas, FR, Schneider: 2111.01436]

[Ashmore, FR: 2103.07472]

[For a review see FR: Phys. Rept. 839 (2020)]





# Motivating Question

---

## Question:

Can string theory describe the universe we live in?  $\Rightarrow$  **Landscape**

## Question:

Are there phenomena that always/never come from string theory  
(or even any quantum theory of gravity)?  $\Rightarrow$  **Swampland**

# Swampland Distance Conjecture

---

- ▶ An intuiting conjecture about a property that any string theory satisfies is the **Swampland Distance Conjecture** [Ooguri, Vafa '06]

## Conjecture:

Compare a string theory compactified on a CY  $X$  at a point  $p_1$  in its moduli space with the theory at a point  $p_0$ . Denote the (geodesic) moduli space distance by  $d$ .

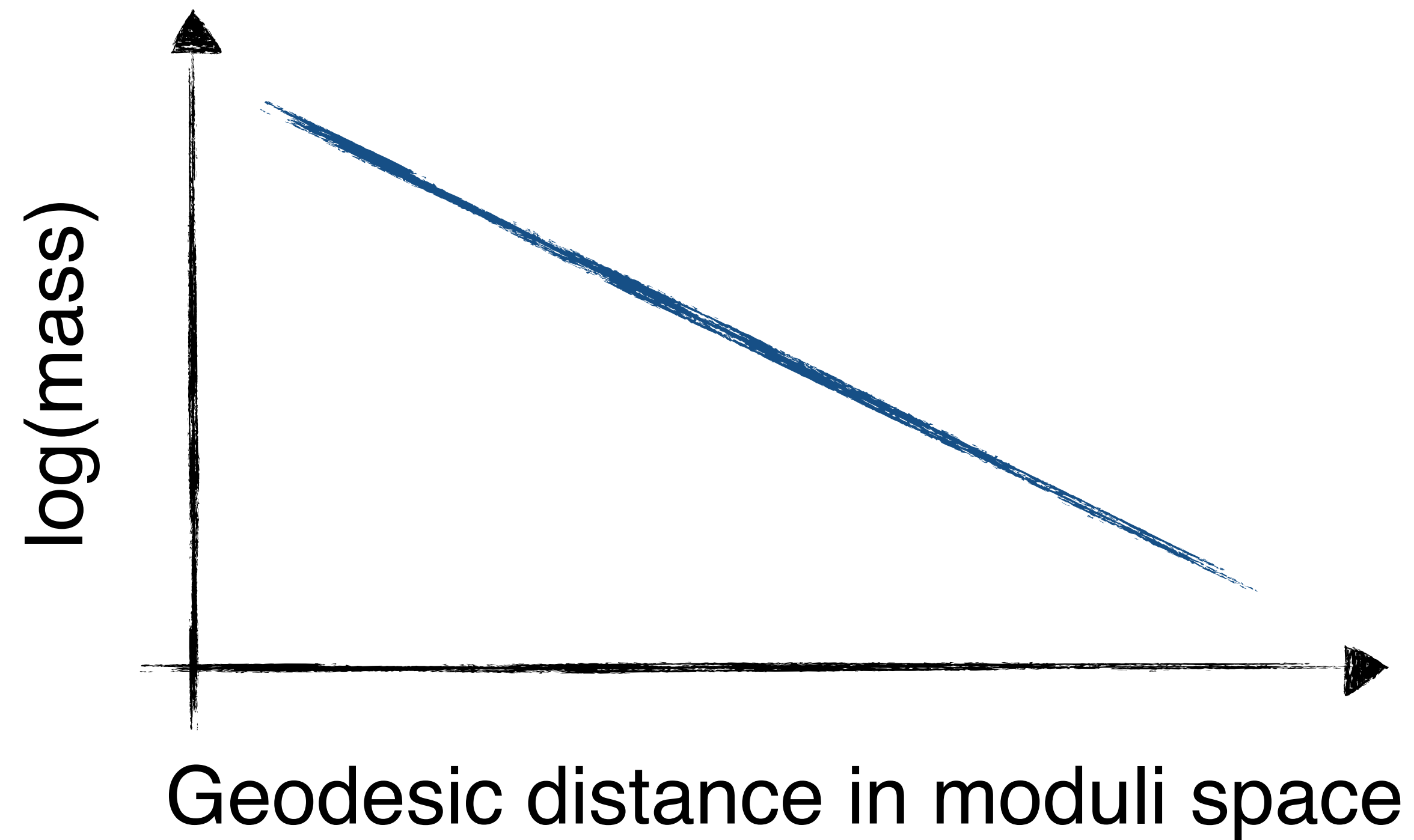
Then, the theory at  $p_1$  has an infinite number of light particles, with mass starting at the order  $m \sim e^{-\alpha d}$  with  $\alpha = \mathcal{O}(1)$  in Planck units.

# Swampland Distance Conjecture

---

We want to check that for a specific CY (the quintic):

$$m \sim e^{-\alpha d}$$



# Compute massive KK states (schematically!)

---

- ▶ Starting point: **10D Klein-Gordon equation**

$$\Delta_{10D} \Phi_{10D} = 0 \quad \Delta_{10D} \sim g^{MN} \partial_M \partial_N$$

- ▶ Now decompose

$$\Delta_{10D} = (\Delta_{4D}; \square_{6D}) \quad \Phi_{10D} = (\phi_{4D}; \varphi_{6D}) \quad g^{MN} = (g^{\mu\nu}; g^{ab})$$

- ▶ Use **6D eigenfunctions** of d'Alembertian

$$\square_{6D} \varphi_{6D} = \lambda \varphi_{6D} \quad \square = d\delta + \delta d = \sqrt{\det(g)}^{-1} \partial_a \sqrt{\det(g)} g^{ab} \partial_b$$

- ▶ Get **4D Klein-Gordon** field equation with a mass term

$$(\Delta_{4D} + \lambda) \phi_{4D} = 0 \quad m^2 \sim \lambda$$

To **compute** the tower of **masses**, we **need** the  
**CY metric**  $g^{ab}$

# Outline

---

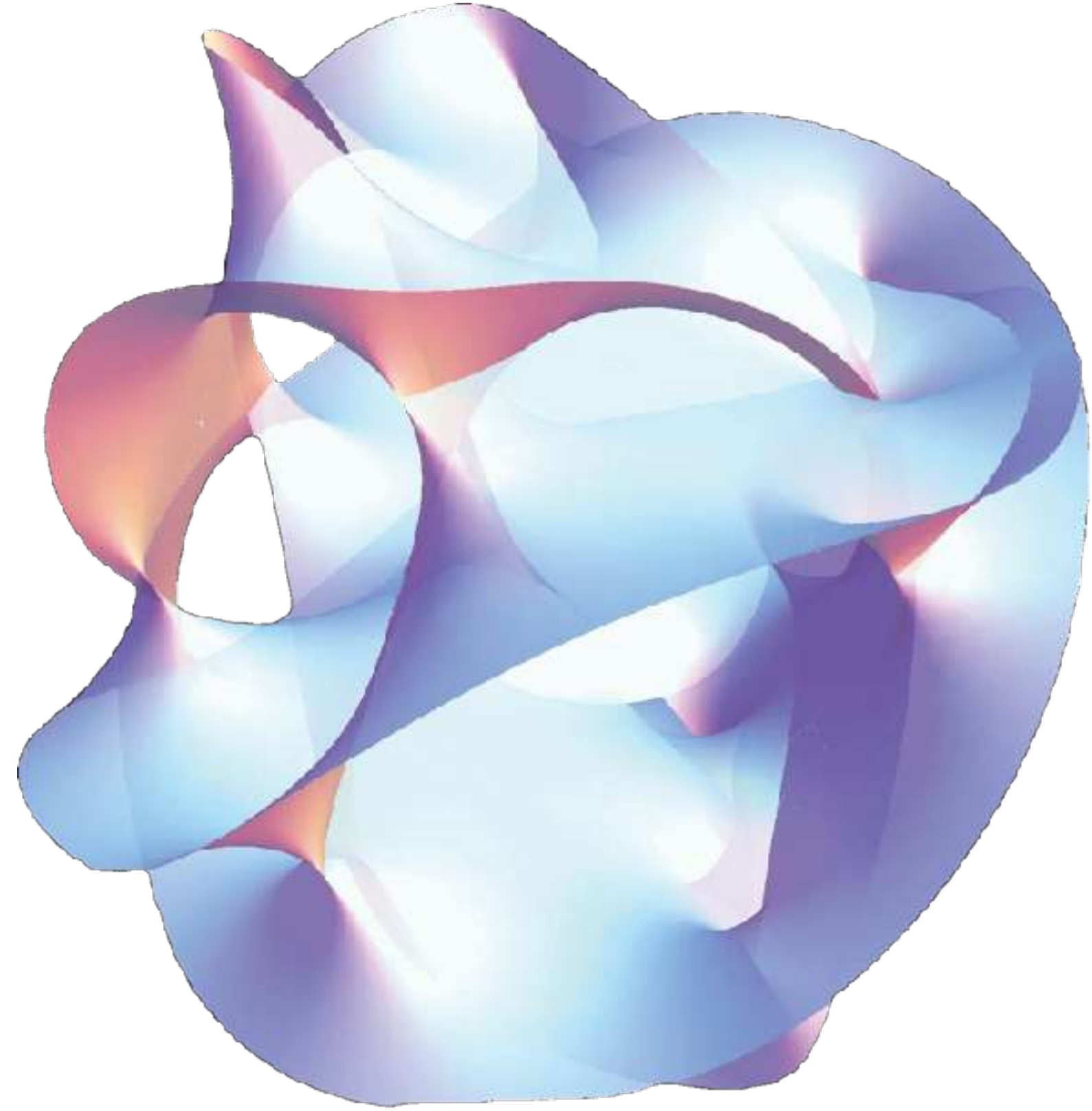
## ▶ Part I

- CY metrics
  - ◆ Introduction
  - ◆ CY metrics from machine learning

## ▶ Part II

- Compute massive string spectrum
- Compute geodesic distance

## ▶ Conclusion



# Calabi-Yau metrics

---



# Calabi-Yau manifolds

---



Calabi



Yau

=



Complex



Kähler



Vanishing first  
Chern class

## **Theorem:**

A complex Kähler manifold with vanishing first Chern class admits a Ricci flat Kähler metric (which is unique in its Kähler class)

## **Problem:**

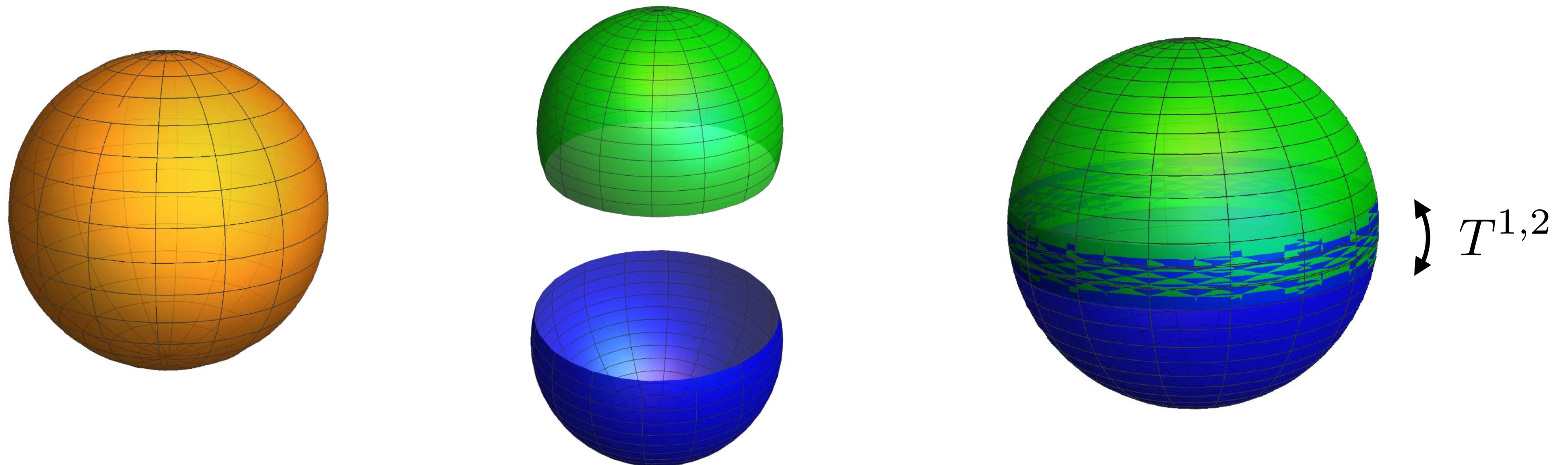
Yau's proof of this theorem is non-constructive, so we don't know what the metric looks like or how to construct it explicitly for CY 3-folds



# 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**  $K$

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

- ▶ 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):  $J^3 = \kappa |\Omega|^2$
- ▶ So instead of minimizing the Ricci tensor, we can minimize this surrogate loss, which is equivalent to Ricci flatness by the powerful theorem of Calabi-Yau, which ensures existence and uniqueness.

# 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 Kähler property and keep the volume moduli fixed if we write

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

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

- ▶ ...but actually we are in a peculiar situation here:  
We have a PDE in  $\Phi$ , but we not care about  $\Phi$  - rather, we care about  $\partial\bar{\partial}\Phi$



# CY metric ansatze

---

- ▶ ... so we can learn  $g_{\text{correction}} = \partial\bar{\partial}\Phi$  instead, turning the surrogate loss into an algebraic equation
- ▶ However, we still need to impose that the resulting metric is Kähler, which requires taking one derivative

Ricci flat (4 derivatives)  $\longrightarrow$  MA equation (2 derivatives)  $\longrightarrow$  algebraic equation (1 derivative)

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

# Reference metric

---

- ▶ So how do we find the reference metric?
- ▶ For  $\mathbb{P}^N$ , there is the Fubini-Study metric:
  - In general, the Kähler metric can be written in terms of the sections of the Kähler cone generators
  - For projective spaces, these are just the homogeneous coordinates  $z_i$
  - The FS metric in these homogeneous coordinates is then

$$g_{\text{FS}} = \partial\bar{\partial} \ln \sigma, \quad \sigma = \sum_i |z_i|^2$$

- We can use projective scalings to set one  $z_i=1$  and pull the metric back to the CY via the defining equations

# Reference metric

---

- ▶ For toric ambient spaces, there is a similar construction
  - Find the sections  $s_i$  of the Kähler cone generators of the toric variety
  - These will be expressions in the toric coordinates  $x_a$
  - The analog of the FS metric now looks exactly like before, just in terms of the  $s_i$  rather than the  $z_i$ , i.e. we get an FS metric “in projective section space”  $\mathbb{P}^{h^0(J)}$

$$g_{\text{FS}} = \partial\bar{\partial} \ln \sigma, \quad \sigma = \sum_i |s_i|^2$$

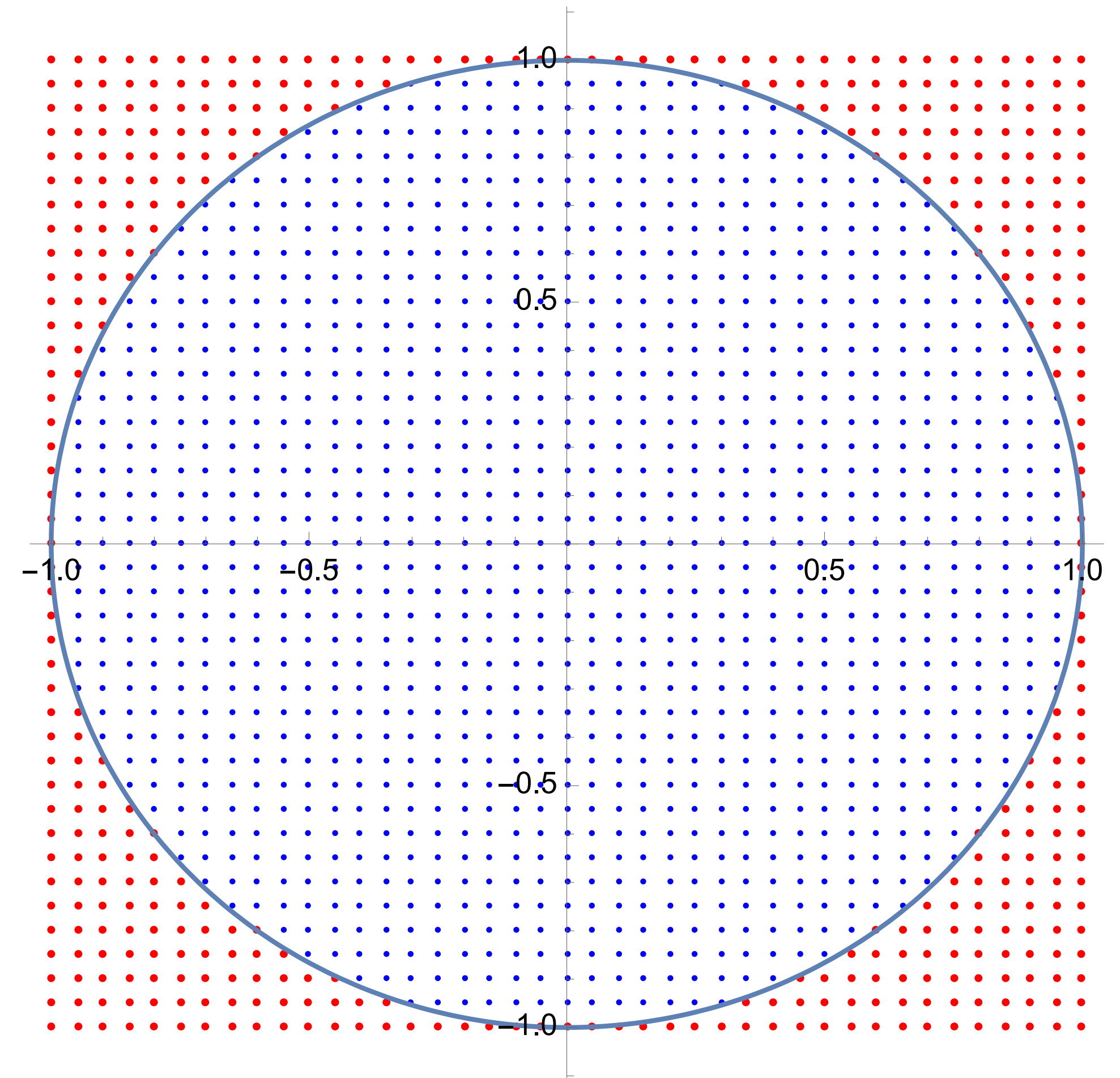
- It can be shown that the dependence of  $s_i$  on  $x_i$  is such that in each patch always one  $s_i=1$ . Moreover, not all  $s_i=0$  simultaneously.



# Numerical Integration

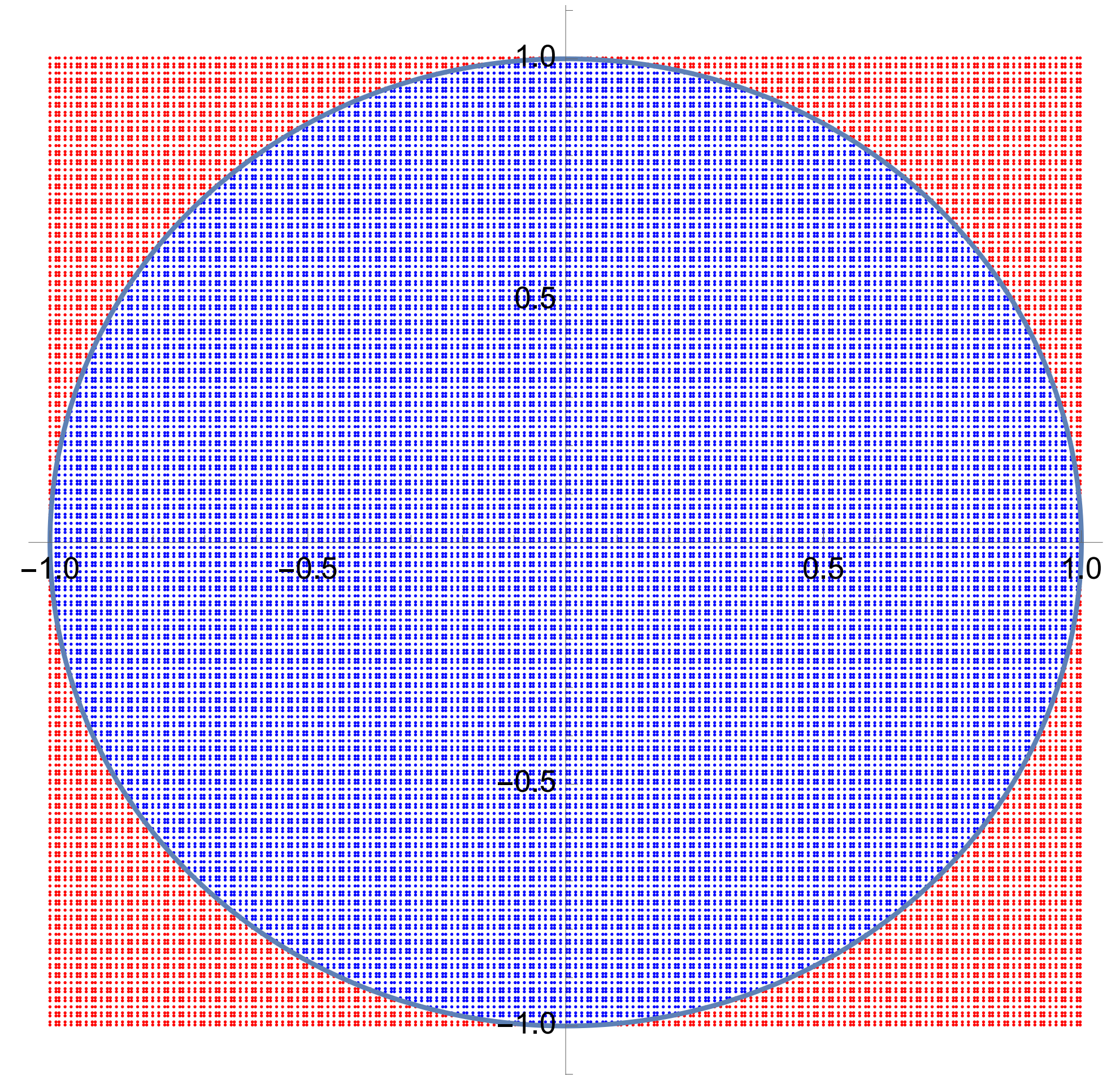
---

- ▶ To the area of a circular lake, you can throw equally spaced stones within a square of known area and count the fraction of times you hit the lake
- ▶ In the example,  $41 \times 41 = 1681$  throws
- ▶ Out of these, 1245 hit the lake
- ▶ Hence the area is  $2 \times 2 \times 1245/1681 = 2.96$



# Numerical Integration

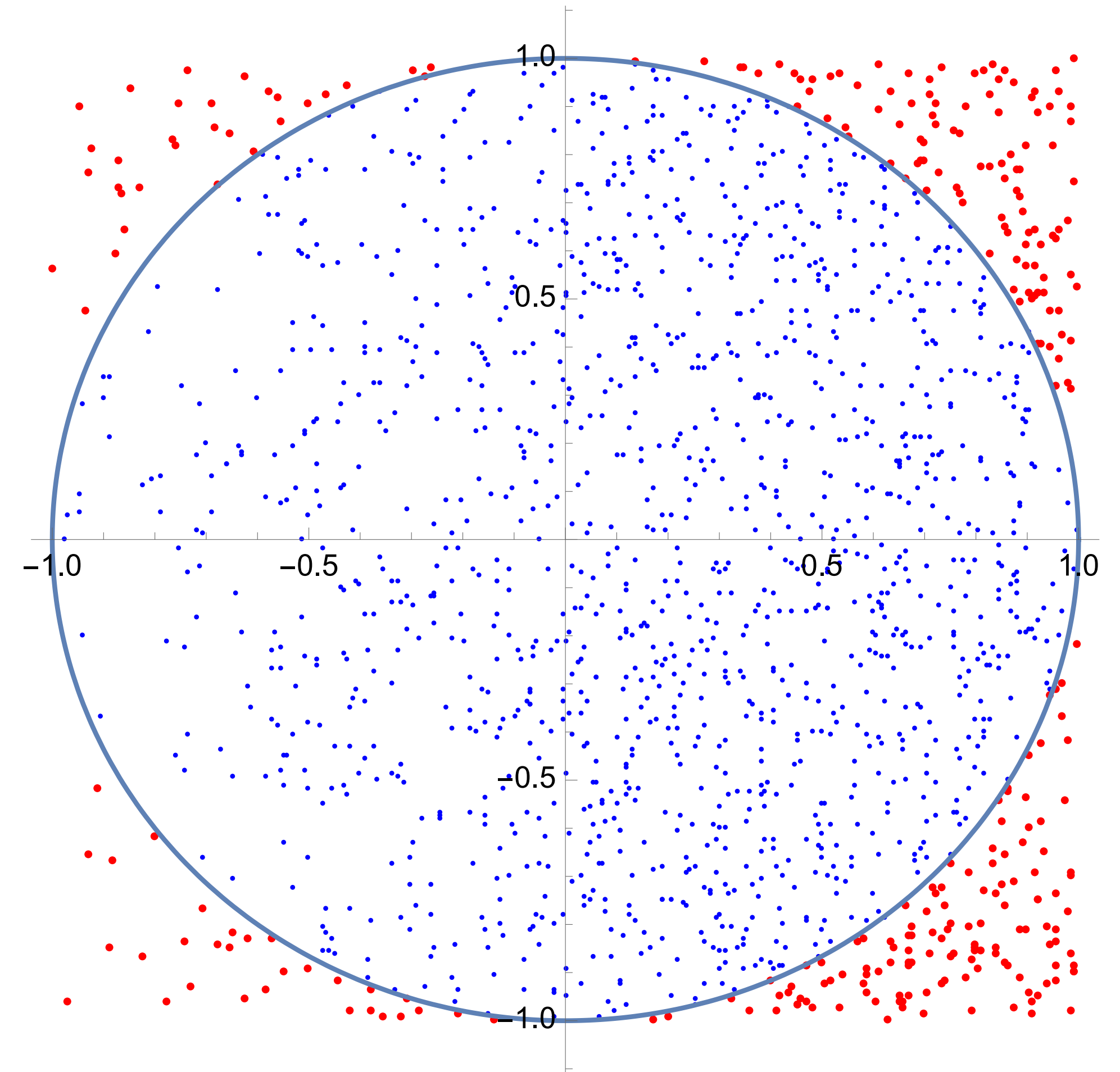
- ▶ To the area of a circular lake, you can throw equally spaced stones within a square of known area and count the fraction of times you hit the lake
- ▶ In the example,  $41 \times 41 = 1681$  throws
- ▶ Out of these, 1245 hit the lake
- ▶ Hence the area is  $2 \times 2 \times 1245/1681 = 2.96$
- ▶ For more throws, we approach  $\pi$  (for  $201 \times 201$  get 3.11)



# Numerical Integration

---

- ▶ However, if I actually throw stones into a lake, it looks more like this
- ▶ There is a bias towards the bottom right
- ▶ If you know the bias, you can correct of it by weighting points on the left stronger than the ones on the right





# Points on the CY

---

- ▶ In order to specify the metric on the CY numerically, we need to specify it at **points on the CY**
- ▶ In principle, we can just solve the CY equation numerically and generate points this way (this can be done using homotopy continuation)
- ▶ However, doing it this way we do not know how the points are distributed on the CY

Theorem: **[Shiffman, Zelditch `99]**

The zeros of random sections in  $\mathbb{P}^N$  are distributed according to the FS metric



# Points on the CY

---

- ▶ So to find points on the CY, we can follow this procedure
- ▶ For CICYs in products of projective ambient spaces, you use effectively the Kodaira embedding of  $\mathbb{P}^N$  into  $\mathbb{P}^N$  via the sections of the Kähler cone generator (which are the homogeneous coordinates themselves)
- ▶ Construct random sections in the embedded  $\mathbb{P}^N$ , intersect them to find a collection of lines, pull back to (intersect with) CY hypersurface
- ▶ Similarly for toric varieties, use map from toric coordinates into  $\mathbb{P}^{h^0(J)}$  via the sections of the Kähler cone generators

$$[z_0 : z_1 : \dots : z_N] \xrightarrow{\phi} [z_0 : z_1 : \dots : z_N]$$

$$[x_0 : x_1 : \dots : x_k] \xrightarrow{\phi} [s_0 : s_1 : \dots : s_{h^{1,1}(J)}]$$

# Steps to get point measure

---

- ▶ So to summarize you need to
  - Find the Kähler cone generators
  - Construct a map into the projectivization  $\mathbb{P}\Gamma(\mathcal{A}, J_i)$
  - Find the common zeros of random iid Gaussian sections
  - Intersect these with the CY hypersurface to get points on the CY
  - Construct reference metric on the ambient space from the FS metric on  $\mathbb{P}\Gamma(\mathcal{A}, J_i)$
  - Pull the metric back to the CY via the hypersurface equation
  - Construct volume measure on the CY from  $\text{vol}_{\text{CY}} = \Omega \wedge \bar{\Omega}$
  - Compute weights for the individual points w.r.t. known volume measure

# Library to do that

- ▶ Luckily this can be automated
- ▶ We provide a python code that integrates into SAGE or Mathematica to do that

```
Install the package

In [1]: pip install --user -e /home/ruehle/Github/su3-metric

In [2]: vertices = [[-1,0,0,0],[-1,0,0,1],[-1,0,1,0],[-1,1,0,0],[2,-1,-1,0],[2,0,0,-1]]
polytope = LatticePolytope(vertices)
p_config = PointConfiguration([list(x) for x in polytope.points()], stars=[0 for _ in range(len(vertices[0]))])

In [3]: triangs = p_config.restrict_to_connected_triangulations().restrict_to_line_triangulations().restrict_to_star_triangul
triang = triangs[0]
tv_fan = triang.fan()
lv = ToricVariety(lv_fan)

Compute sections, patches, etc. from toric variety

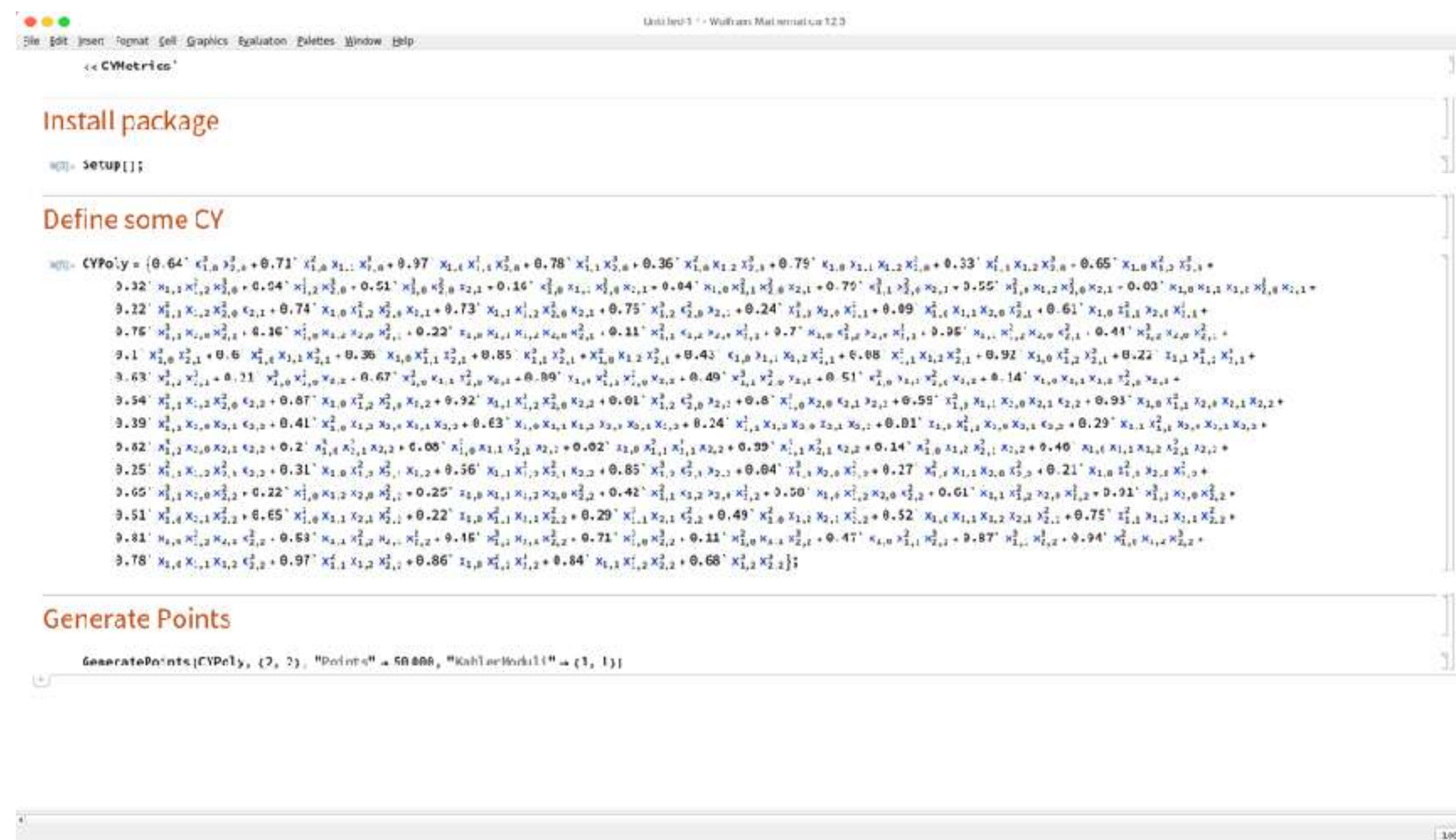
In [4]: from cymetric.sage.sagelib import prepare_toric_cy_data

work_dir = "/home/ruehle/toric_model"
toric_data = prepare_toric_cy_data(tv, os.path.join(work_dir, "toric_data.pickle"))

Compute points on CY, weights, etc.

In [5]: from cymetric.pointgen.pointgen.mathematica import PointGeneratorToricMathematica
from cymetric.pointgen.npholper import prepare_dataset, prepare_basis_pickle

point_generator = PointGeneratorToricMathematica(50000, toric_data, kmoduli=[1,1])
prepare_dataset(point_generator, num_pts, work_dir);
prepare_basis_pickle(point_generator, work_dir);
```



```
Install package

%%> SETUP[];

Define some CY

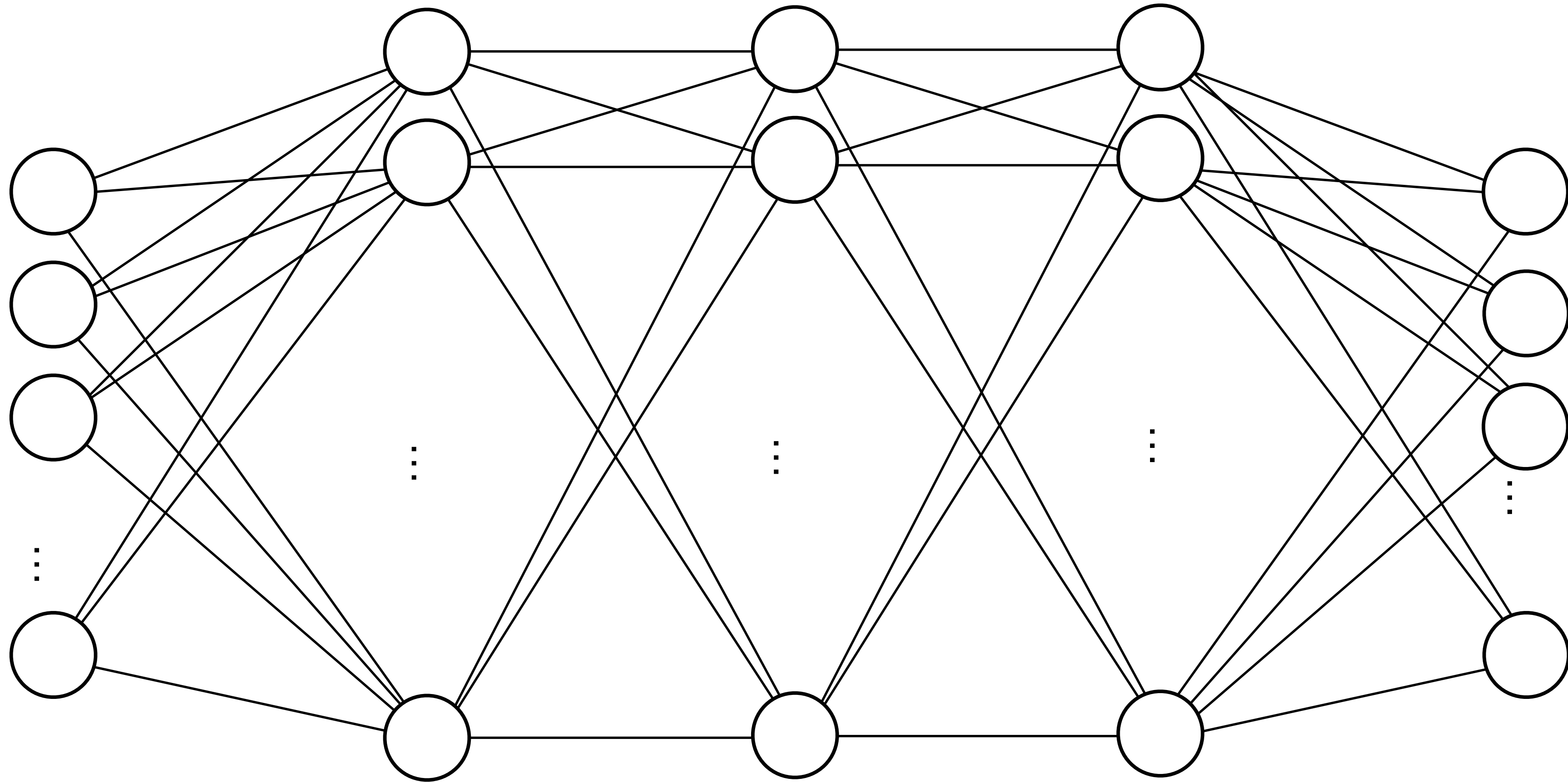
%%> CYPoly = {0.64 x1,0^2 x2,0^2 + 0.71 x1,0 x1,1 x2,0^2 + 0.97 x1,0 x1,1 x2,0^2 + 0.78 x1,1 x2,0^2 + 0.36 x1,0 x1,2 x2,0^2 + 0.79 x1,0 x1,1 x1,2 x2,0^2 + 0.33 x1,0 x1,2 x2,0^2 - 0.65 x1,0 x1,1 x2,1^2 +
0.32 x1,1 x1,2 x2,0^2 + 0.04 x1,2 x2,0^2 + 0.51 x1,0 x2,0^2 x2,1 + 0.16 x1,0 x1,1 x2,0 x2,1 + 0.04 x1,0 x1,1 x2,0 x2,1 + 0.79 x1,1 x2,0 x2,1 + 0.55 x1,0 x1,2 x2,0 x2,1 - 0.03 x1,0 x1,1 x1,2 x2,0 x2,1 +
0.22 x1,1 x1,2 x2,0 x2,1 + 0.74 x1,0 x1,2 x2,0 x2,1 + 0.73 x1,1 x1,2 x2,0 x2,1 + 0.75 x1,2 x2,0 x2,1 + 0.24 x1,1 x2,0 x2,1 + 0.09 x1,0 x1,1 x2,0 x2,1 + 0.61 x1,0 x1,1 x2,0 x2,1 +
0.76 x1,1 x2,0 x2,1 + 0.16 x1,0 x1,1 x2,0 x2,1 + 0.22 x1,0 x1,1 x2,0 x2,1 + 0.11 x1,0 x1,1 x2,0 x2,1 + 0.7 x1,0 x1,1 x2,0 x2,1 + 0.96 x1,0 x1,1 x2,0 x2,1 + 0.41 x1,0 x1,1 x2,0 x2,1 +
0.1 x1,0 x1,1 x2,1^2 + 0.6 x1,0 x1,1 x2,1^2 + 0.36 x1,0 x1,1 x2,1^2 + 0.85 x1,0 x1,1 x2,1^2 + 0.43 x1,0 x1,1 x2,1^2 + 0.08 x1,0 x1,1 x2,1^2 + 0.92 x1,0 x1,1 x2,1^2 + 0.22 x1,1 x2,1^2 +
0.63 x1,1 x2,1^2 + 0.21 x1,0 x1,1 x2,1^2 + 0.67 x1,0 x1,1 x2,1^2 + 0.89 x1,0 x1,1 x2,1^2 + 0.40 x1,0 x1,1 x2,1^2 + 0.51 x1,0 x1,1 x2,1^2 + 0.14 x1,0 x1,1 x2,1^2 +
0.54 x1,0 x1,1 x2,1^2 + 0.87 x1,0 x1,1 x2,1^2 + 0.92 x1,0 x1,1 x2,1^2 + 0.01 x1,0 x1,1 x2,1^2 + 0.8 x1,0 x1,1 x2,1^2 + 0.59 x1,0 x1,1 x2,1^2 + 0.93 x1,0 x1,1 x2,1^2 + 0.93 x1,0 x1,1 x2,1^2 +
0.39 x1,0 x1,1 x2,1^2 + 0.41 x1,0 x1,1 x2,1^2 + 0.63 x1,0 x1,1 x2,1^2 + 0.24 x1,0 x1,1 x2,1^2 + 0.81 x1,0 x1,1 x2,1^2 + 0.29 x1,0 x1,1 x2,1^2 + 0.81 x1,0 x1,1 x2,1^2 +
0.62 x1,0 x1,1 x2,1^2 + 0.2 x1,0 x1,1 x2,1^2 + 0.08 x1,0 x1,1 x2,1^2 + 0.02 x1,0 x1,1 x2,1^2 + 0.39 x1,0 x1,1 x2,1^2 + 0.14 x1,0 x1,1 x2,1^2 + 0.46 x1,0 x1,1 x2,1^2 +
0.25 x1,0 x1,1 x2,1^2 + 0.31 x1,0 x1,1 x2,1^2 + 0.56 x1,0 x1,1 x2,1^2 + 0.85 x1,0 x1,1 x2,1^2 + 0.04 x1,0 x1,1 x2,1^2 + 0.27 x1,0 x1,1 x2,1^2 + 0.21 x1,0 x1,1 x2,1^2 +
0.65 x1,0 x1,1 x2,1^2 + 0.22 x1,0 x1,1 x2,1^2 + 0.25 x1,0 x1,1 x2,1^2 + 0.42 x1,0 x1,1 x2,1^2 + 0.50 x1,0 x1,1 x2,1^2 + 0.61 x1,0 x1,1 x2,1^2 + 0.91 x1,0 x1,1 x2,1^2 +
0.51 x1,0 x1,1 x2,1^2 + 0.65 x1,0 x1,1 x2,1^2 + 0.22 x1,0 x1,1 x2,1^2 + 0.29 x1,0 x1,1 x2,1^2 + 0.49 x1,0 x1,1 x2,1^2 + 0.52 x1,0 x1,1 x2,1^2 + 0.75 x1,0 x1,1 x2,1^2 +
0.81 x1,0 x1,1 x2,1^2 + 0.69 x1,0 x1,1 x2,1^2 + 0.16 x1,0 x1,1 x2,1^2 + 0.71 x1,0 x1,1 x2,1^2 + 0.11 x1,0 x1,1 x2,1^2 + 0.47 x1,0 x1,1 x2,1^2 + 0.87 x1,0 x1,1 x2,1^2 + 0.94 x1,0 x1,1 x2,1^2 +
0.78 x1,0 x1,1 x2,1^2 + 0.97 x1,0 x1,1 x2,1^2 + 0.86 x1,0 x1,1 x2,1^2 + 0.84 x1,0 x1,1 x2,1^2 + 0.68 x1,0 x1,1 x2,1^2};

Generate Points

GeneratePoints[CYPoly, {2, 2}, {"Points" -> 50000, "KahlerModuli" -> {1, 1}}]
```

[<https://github.com/pythoncymetric/cymetric>]





# Calabi-Yau metrics from Neural Networks

---

# Calabi-Yau metrics

---

- ▶ The **metric** is some **unknown function**

$$g^{ab} : \begin{array}{c} \text{point in} \\ \text{moduli space} \end{array} \times \begin{array}{c} \text{point on} \\ \text{CY} \end{array} \longrightarrow \begin{array}{c} \text{complex hermitian} \\ \text{3x3 metric} \end{array}$$

- ▶ So we are looking for a **map** (for the example of the quintic)
  - **point** in **moduli space**: 1 complex number (2 real DOFs)
  - **point** on **CY**: 5 complex numbers (10 real DOFs)
  - **complex** hermitian 3x3 **matrix**: 3 real + 3 complex (9 real DOFs)

# Calabi-Yau metrics

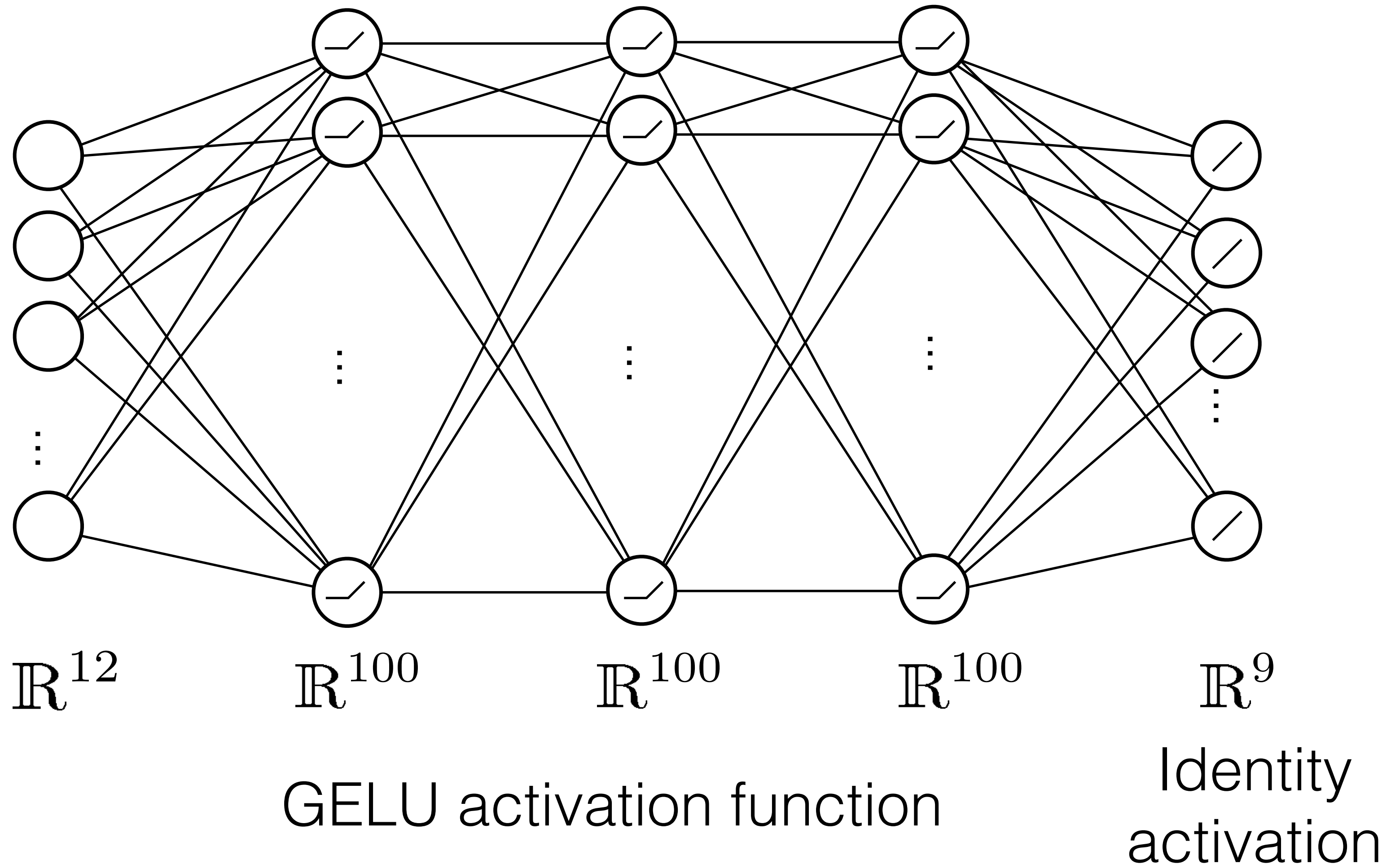
---

- ▶ Simplest possibility:
  - Take a  $12 \times 9$  matrix with  $12 \times 9 = 108$  parameters
  - Choose the parameters to produce a Ricci-flat Kahler metric
- ▶ ...but such linear maps are too simple
  - Take a  $12 \times 100$  matrix
  - Make it non-linear by taking the tanh of each of the 100 nodes
  - Take a  $100 \times 100$  matrix, and take again the tanh, ...
  - Take a  $100 \times 9$  matrix
  - Find the parameters in all these matrices such that the result is a Ricci-flat Kahler metric  
[Ashmore et al `19; Anderson, Gray, Gerdes, Krippendorf, Raghuram, FR `20; Douglas et al `20; Jejjala et al `20; Larfors, Lukas, FR, Schneider `21]
- ▶ What we have just described is a feed-forward NN! It is a universal approximator  
[Cybenko `89; ...; Kidger, Lyons `19]



# NNs for CY metrics

---



# Library to do that

- ▶ This can also be automated
- ▶ Again, call directly from SAGE or Mathematica

## Define the NN

```
In [ ]: import tensorflow as tf
tfk = tf.keras

nHiddens = [64, 64, 64]
acts = ['gelu', 'gelu', 'gelu']
model = tf.keras.Sequential()
model.add(tfk.Input(shape=(int(n_in))))
for nHidden, act in zip(nHiddens, acts):
    model.add(
        tfk.layers.Dense(
            nHidden,
            activation=act,
        )
    )
model.add(tfk.layers.Dense(n_out))
```

## Choose the CY metric ansatz

```
In [ ]: from cymetric.models.tfmodels import PhiFSModelToric
from cymetric.models.tfhelper import prepare_tf_basis
BASIS = prepare_tf_basis(np.load('basis.pickle'), allow_pickle=True)

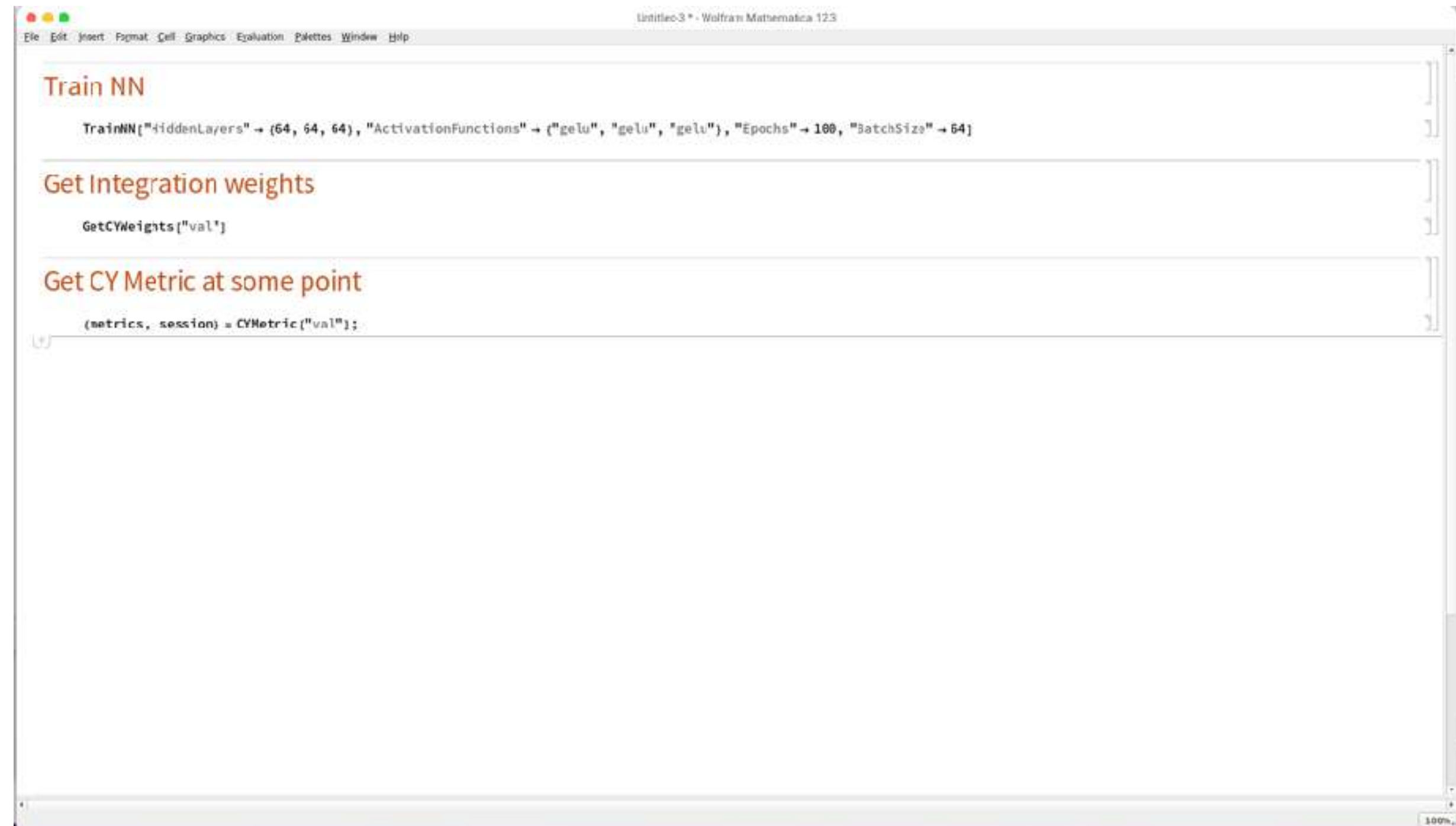
fs_model = PhiFSModelToric(model, BASIS, alpha=alpha, kappa=kappa, toric_data=toric_data)
fs_model.compile(custom_metrics=cmetrics, optimizer=tfk.optimizers.Adam())
```

## train the NN

```
In [ ]: fs_model.fit(data['X_train'], data['y_train'], epochs=100, batch_size=64)
```

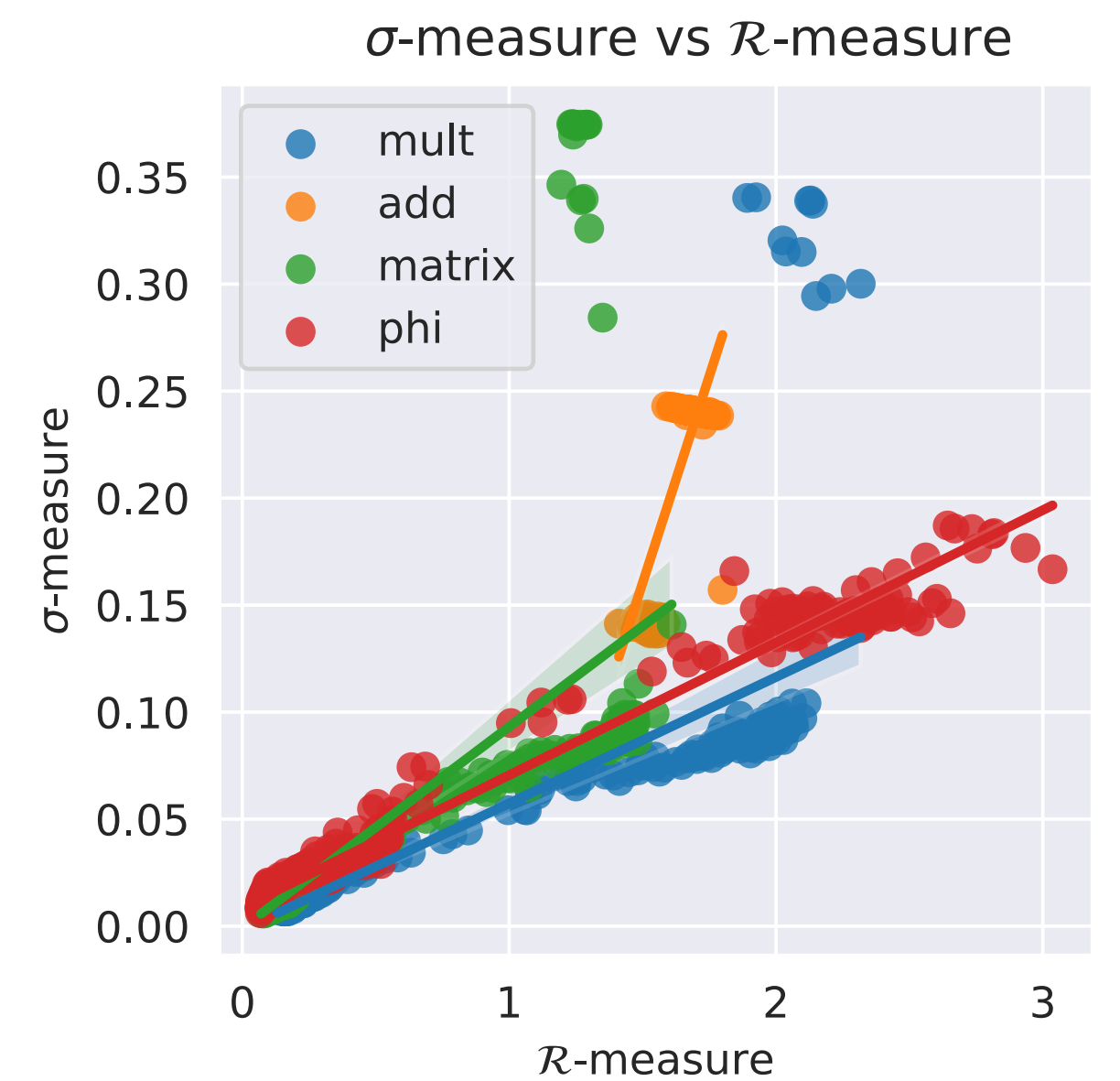
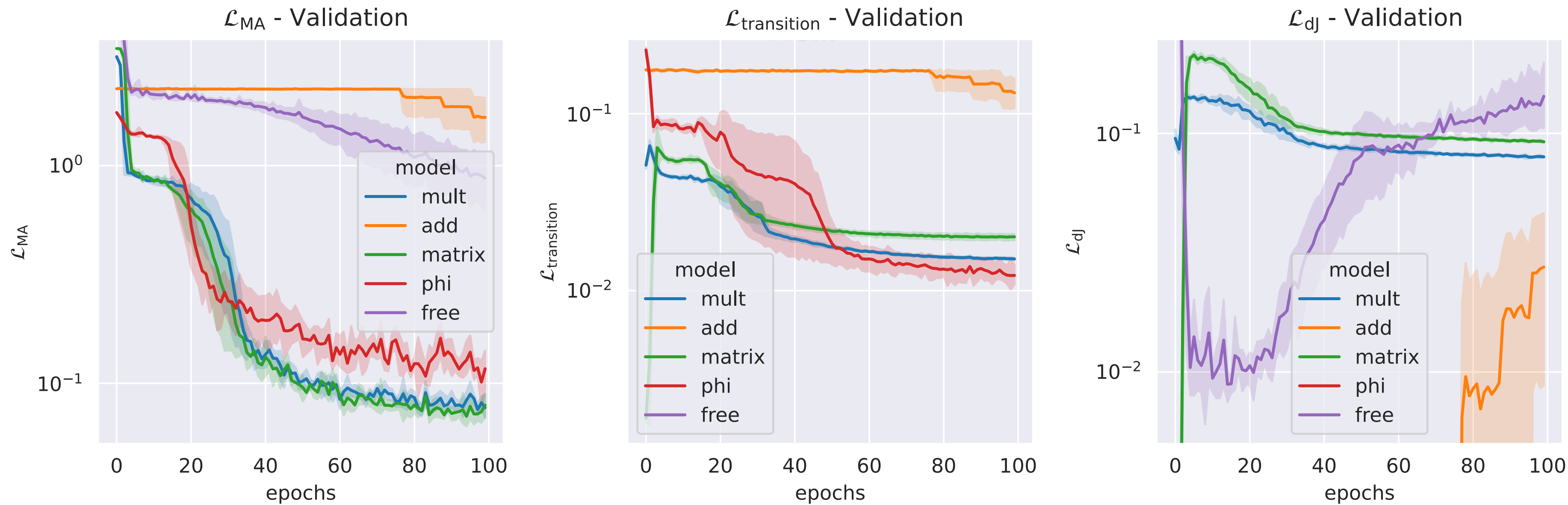
## CY metric at some points

```
In [ ]: cy_metrics = fs_model(data['X_val'])
```



[<https://github.com/pythoncymetric/cymetric>]

# CY metric results (for quintic)

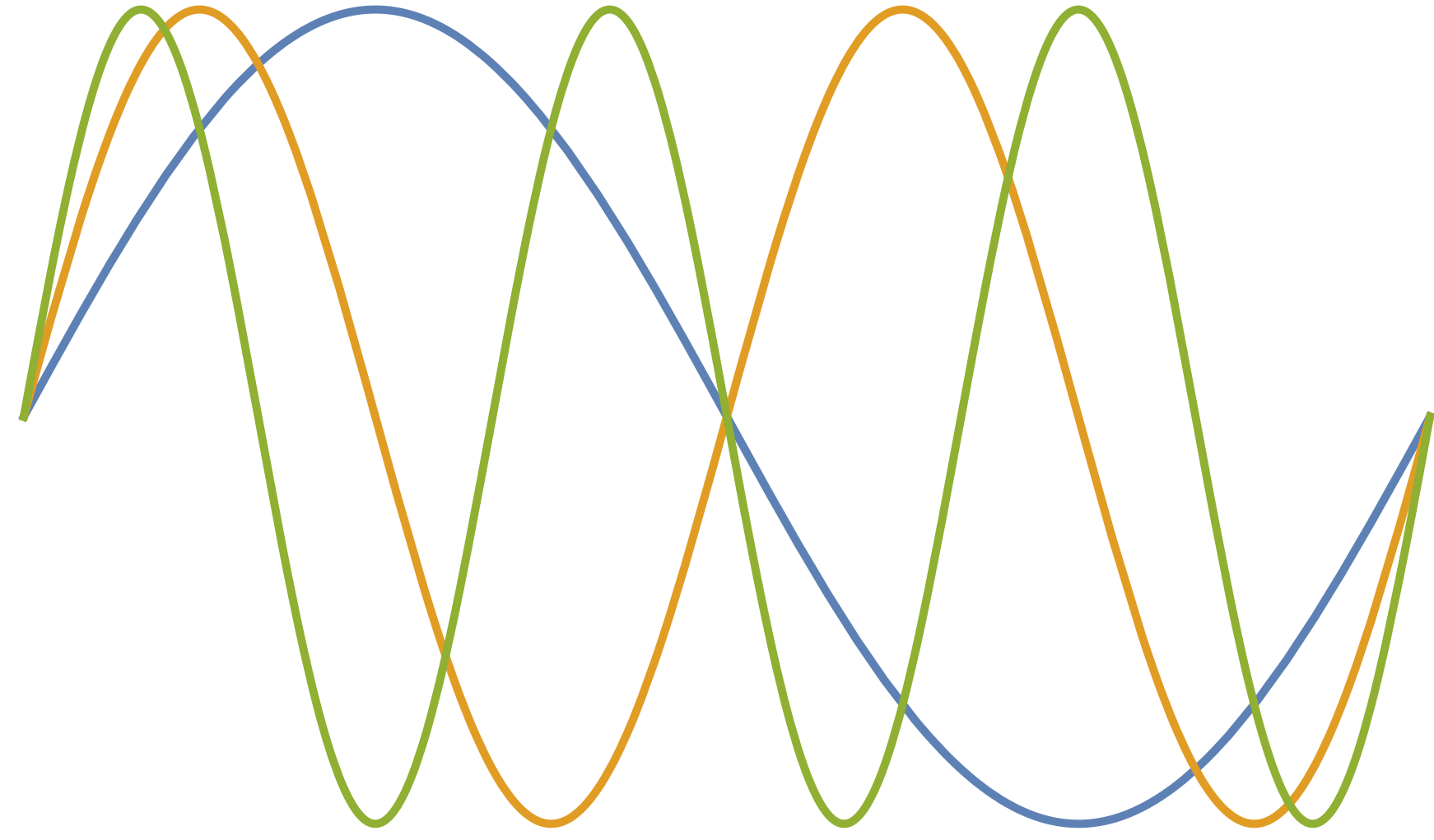


$$\mathcal{L} = \alpha_1 \mathcal{L}_{MA} + \alpha_2 \mathcal{L}_{dJ} + \alpha_3 \mathcal{L}_{\text{transition}} + \alpha_4 \mathcal{L}_{\text{Ricci}} + \alpha_5 \mathcal{L}_{\text{vol}}$$

$$\sigma = \int_X \left| 1 - \frac{J^3}{|\Omega|^2} \right|$$

$$\mathcal{R} = \int_X R^{ab} R_{ab}$$





Massive strings spectrum

---

# Compute massive string spectrum

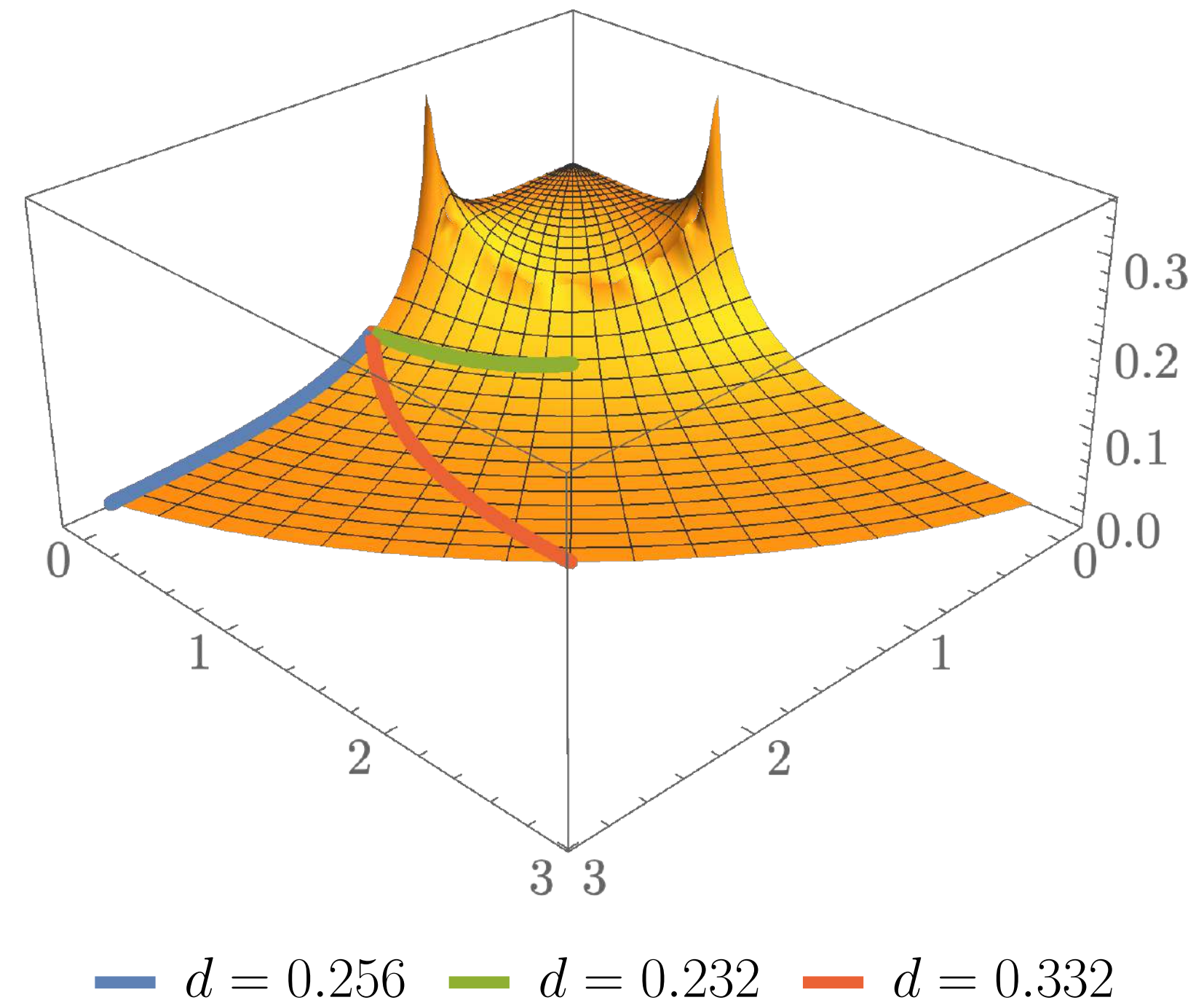
---

1. **Compute** the **CY metric** at **different** points in **moduli** space
2. **Choose** a (finite subset) of an (infinite) **function basis**  $\{\alpha_A\}$  on  $X$

$$\{\alpha_A\} = \frac{\{s_\alpha^{(k_\phi)} \bar{s}_{\bar{\beta}}^{(k_\phi)}\}}{(|z_0|^2 + \dots + |z_4|^2)^{k_\phi}} \quad \text{w/ } s_\alpha^{(k_\phi)} \text{ degree } k_\phi \text{ monomials in } z \text{'s}$$

3. Expand  $|\varphi\rangle = \varphi^A |\alpha_A\rangle$  and **compute**  $\Delta_{AB} \varphi^B = \lambda O_{AB} \varphi^B$  w/ **matrix elms**  
 $\Delta_{AB} = \langle \alpha_A, \Delta \alpha_B \rangle, \quad O_{AB} = \langle \alpha_A, \alpha_B \rangle$  (evaluated using MC integration)

- ▶ We compute at  $k_\phi = 3 \Rightarrow$  can compute the first 1,225 eigenvalues
- ▶ For higher eigenvalues, lose accuracy due to finite truncation of basis



# Geodesic Distance in Moduli Space

---



# Geodesics

---

- ▶ **Geodesics** give the **shortest distance** between 2 points on a manifold
- ▶ They are solution to the **geodesics equation**

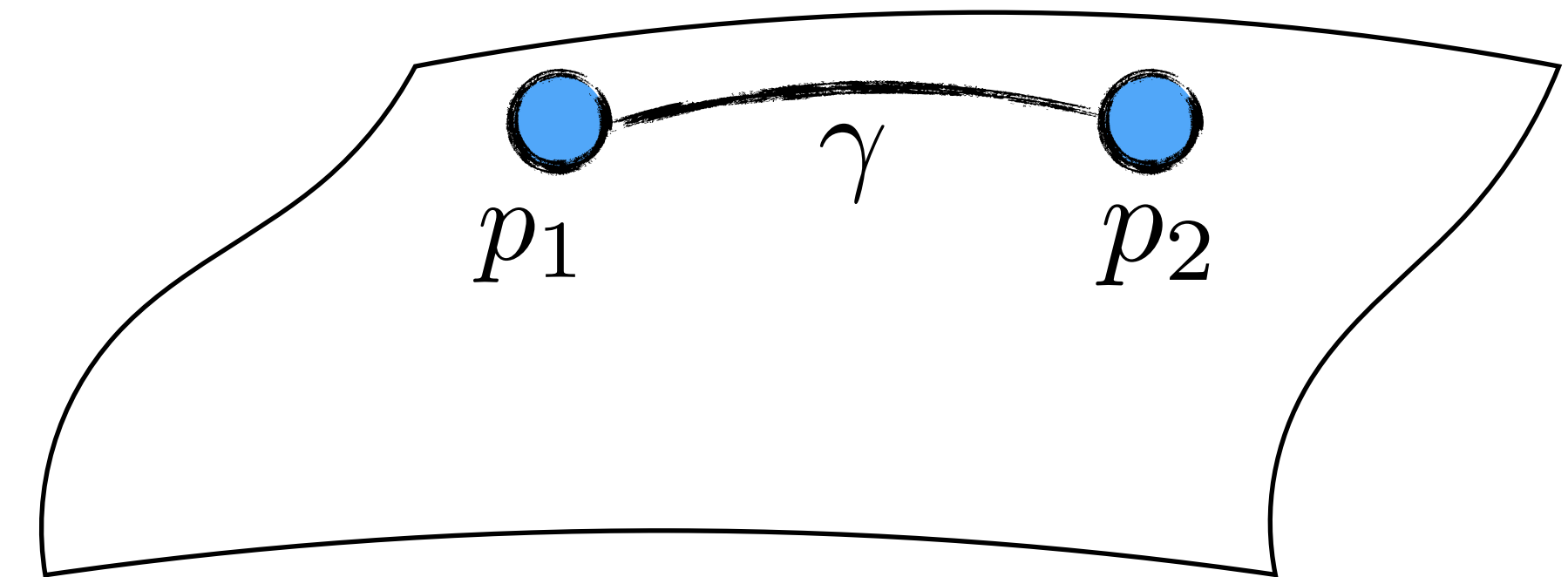
$$\ddot{\gamma}(\tau) + \Gamma_{ab}^c \dot{\gamma}^a(\tau) \dot{\gamma}^b(\tau) = 0$$

- ▶ The **geodesic distance** is then

$$d(p_1, p_2) = \int_{\tau_1}^{\tau_2} d\tau \sqrt{g_{a\bar{b}}(\gamma(\tau)) \dot{\gamma}^a(\tau) \dot{\gamma}^b(\tau)}$$

(evaluate using numerical integration)

- ▶ Usually, we specify boundary values rather than initial values  
⇒ **solve** equations **numerically** using RK/AB **shooting** methods



# Complex Structure moduli space metric

---

- ▶ The **CS moduli space metric** is also a **Kähler** metric

- ▶ It can be obtained from the **Kähler potential**  $K_{CS} = -\ln \left( i \int_X \Omega_\psi \wedge \overline{\Omega_\psi} \right)$

where  $\Omega$  is the holomorphic (3,0) form of the CY

- ▶ To express this, **choose** a **basis** of **3-cycles**

$$A^I \cap B_J = \int_X \alpha_J \wedge \beta^I = \int_{A^I} \alpha_J = \int_{B_J} \beta^I = \delta_J^I$$

- ▶ Then,  $\Omega \wedge \bar{\Omega} = z^I \bar{\mathcal{G}}_I - \bar{z}^I \mathcal{G}_I$  with  $\Pi = \begin{pmatrix} \mathcal{G}_I \\ z^I \end{pmatrix} = \begin{pmatrix} \int_{B_I} \Omega \\ \int_{A^I} \Omega \end{pmatrix}$

- ▶ The so-called **periods**  $\Pi_i$  are **solutions** to a hypergeometric Picard Fuchs **differential equation** [Candelas, De La Ossa, Green, Parkes '91]

# Complex structure moduli space metric

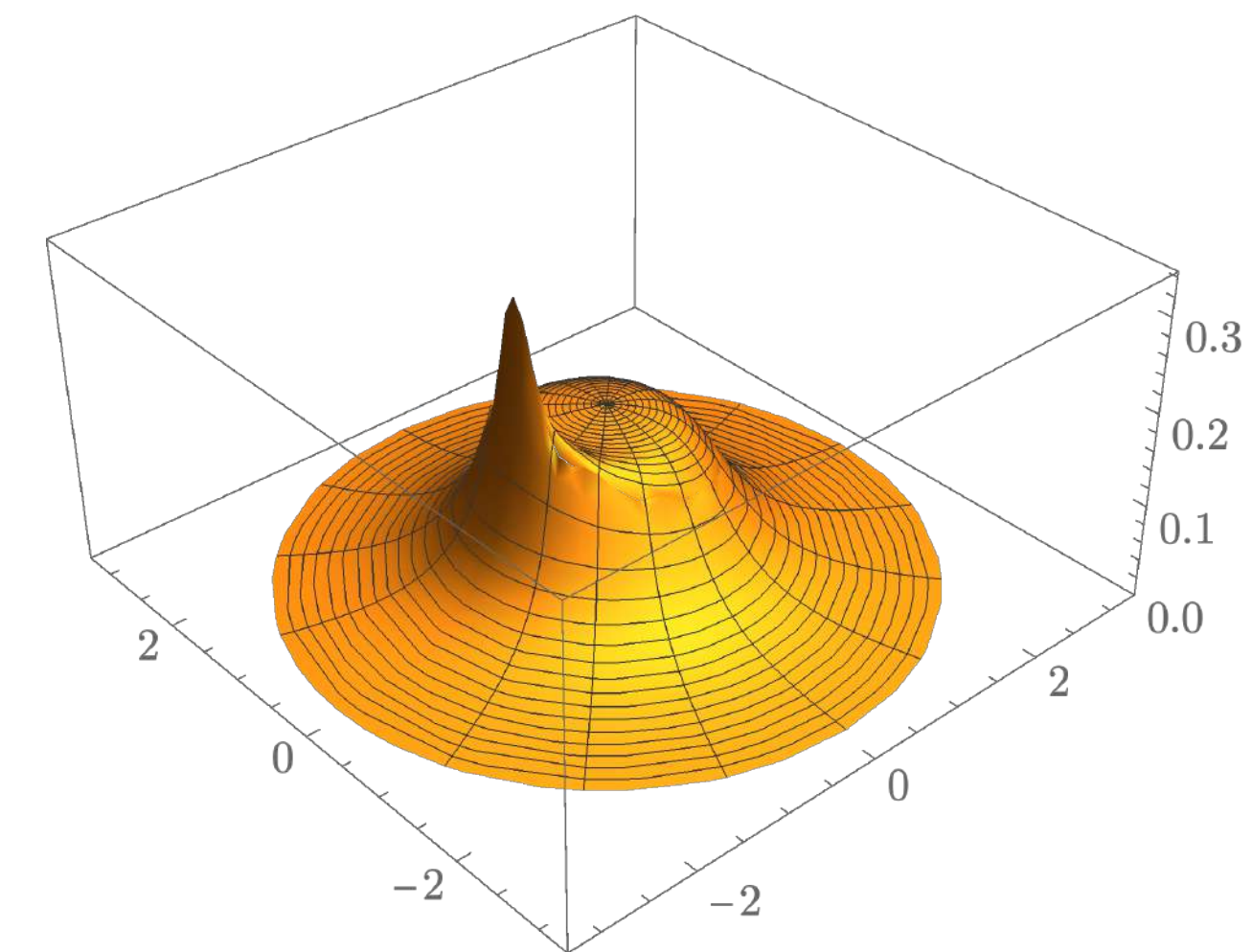
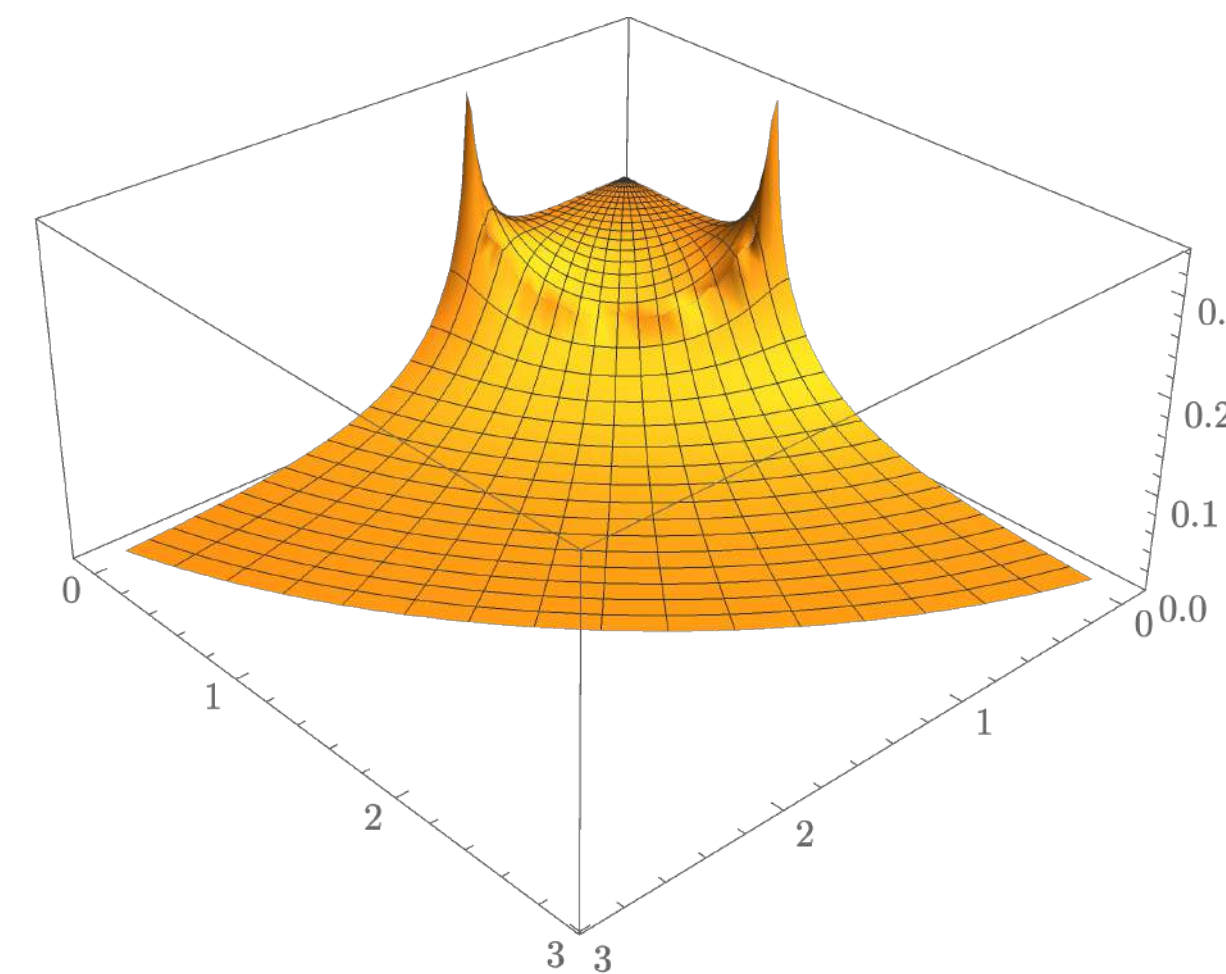
▶ In our case, we have a **single CS parameter**  $\psi : z_0^5 + z_1^5 + z_2^5 + z_3^5 + z_4^5 - 5\psi z_0 z_1 z_2 z_3 z_4 = 0$

▶ Note:

- The hypergeometric functions depend on  $\psi^5$  rather than  $\psi$ ; the trafo  $\psi \rightarrow e^{2\pi i/5}\psi$  can be undone by  $z_0 \rightarrow e^{2\pi i/5}z_0$
- The CY is singular at  $\psi = 1$ . This singularity is at finite distance (conifold)
- The CY is also singular at  $\psi \rightarrow \infty$ . This is at infinite distance.
- We can use the mirror map

$$t = \frac{z^1}{z^0} \sim -\frac{5}{2\pi i} \ln(5\psi) \Rightarrow g_{t\bar{t}} \sim \frac{3}{4\text{Im}(t)^2}$$

to turn CS into Kahler moduli statements





# Complex structure moduli space metric

---

- ▶ We **compare three approaches** to computing geodesics
  - Use the **exact analytic** (hypergeometric) function
  - Use the **large volume approximation**
  - Compute the Weil-Petersson **metric numerically** following: [Keller,Lukic `91]
    - ◆ Vary the hol. 3-form  $\Omega_{\psi+\varepsilon} = \Omega_{\psi} + \varepsilon^i \partial_{\psi_i} \Omega_{\psi} + \mathcal{O}(\varepsilon^2)$
    - ◆ Construct a basis of (non-holomorphic) 3-forms  $\partial_{\psi_i} \Omega_{\psi} \in H^{3,0}(X_{\psi}) \oplus H^{2,1}(X_{\psi})$
    - ◆ Construct a basis of vectors  $\varepsilon^i \in T_{\psi} \mathcal{M}_{CS}$
    - ◆ Then the inner product w.r.t. the moduli space metric is (integration over  $X$  is again done using MC)

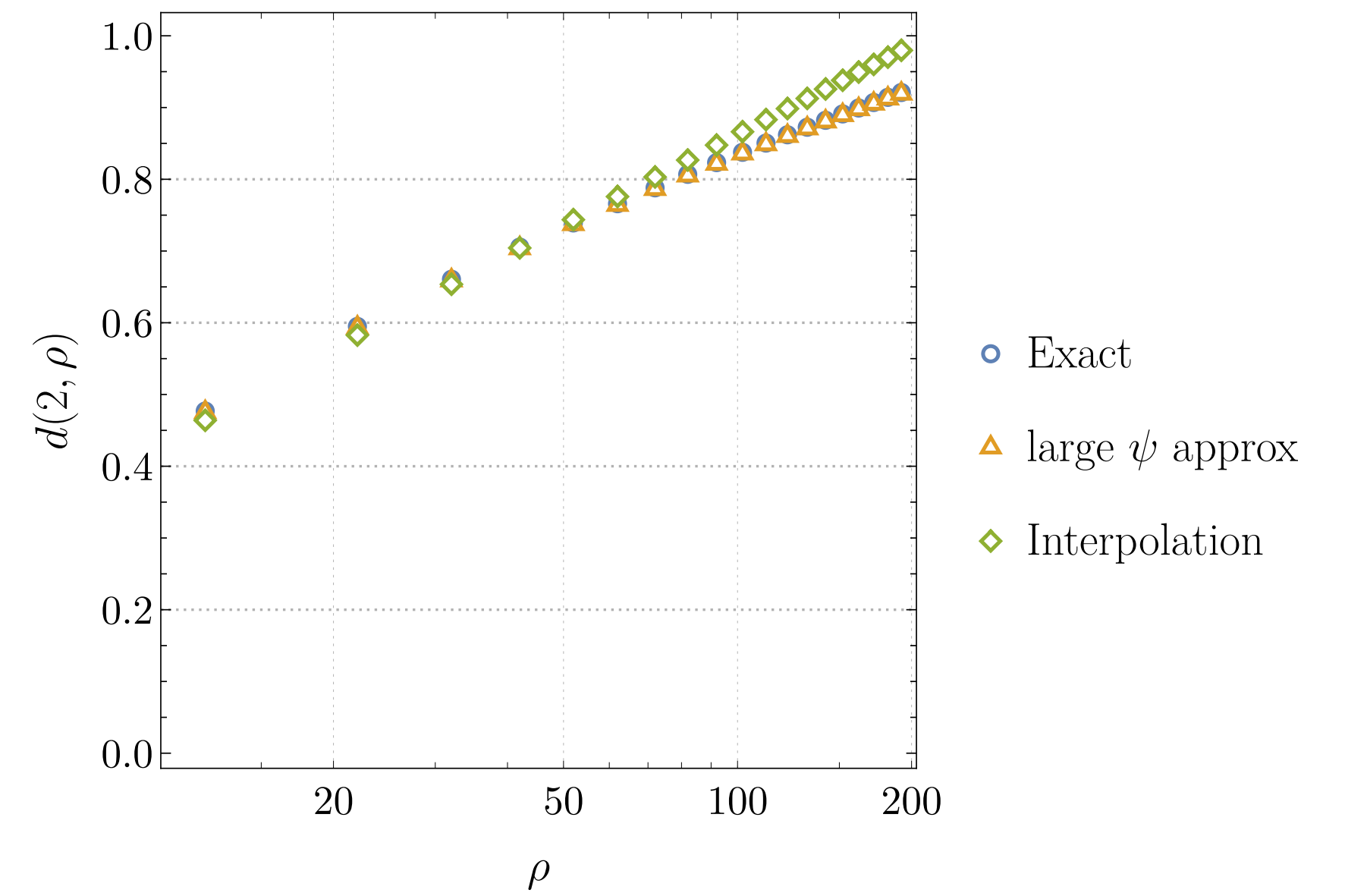
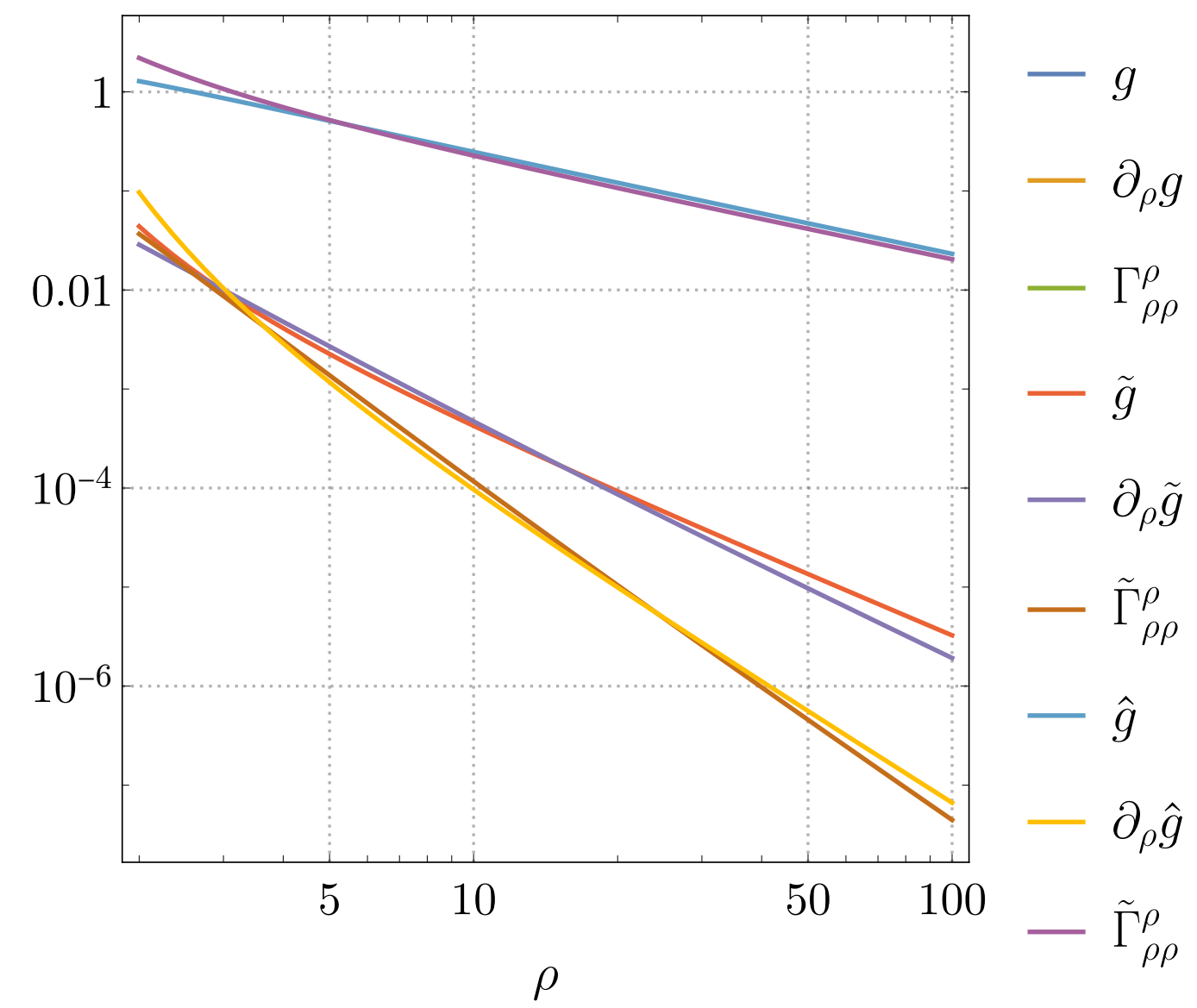
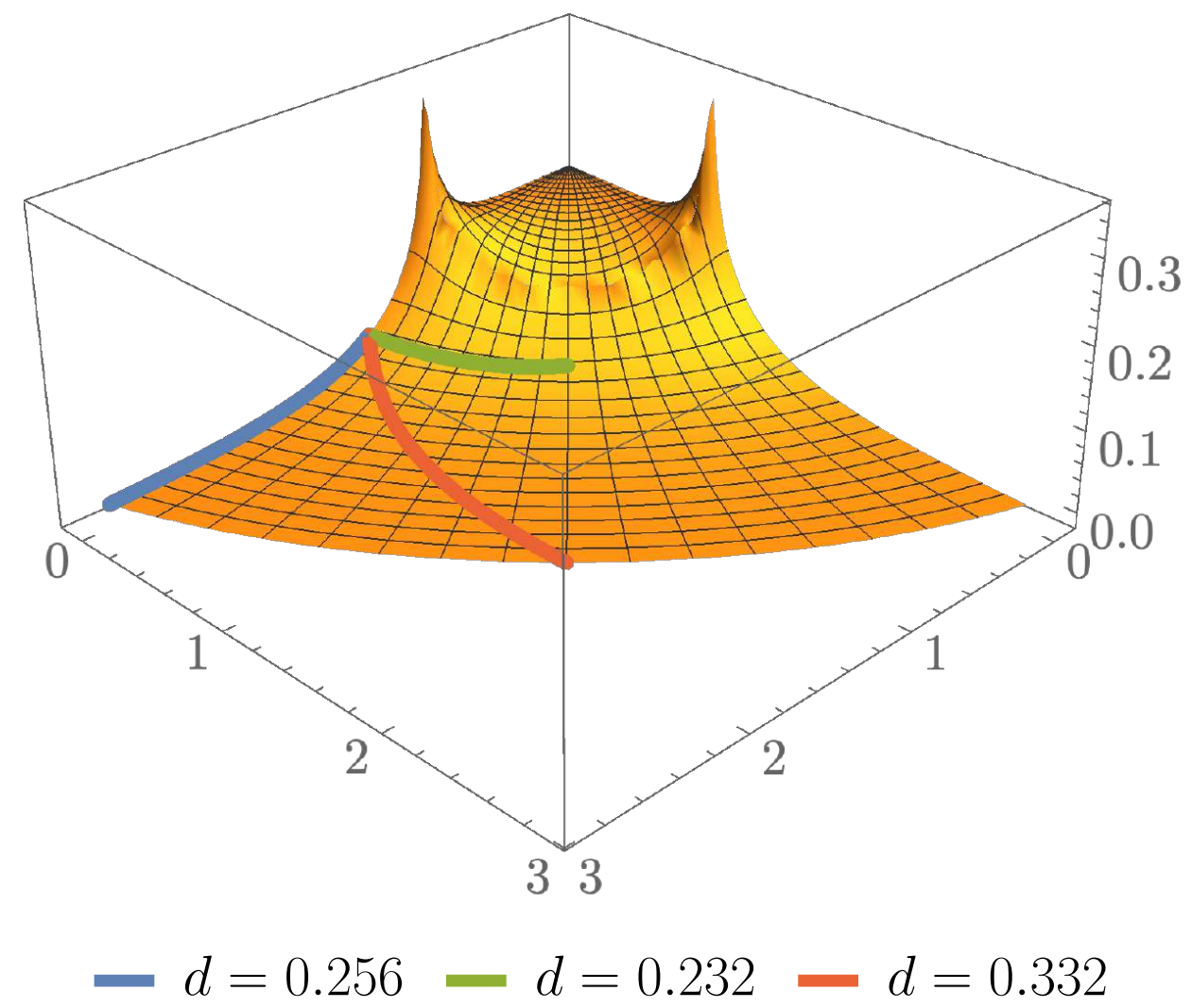
$$\langle \varepsilon_1, \varepsilon_2 \rangle = -\frac{\varepsilon_1^i \bar{\varepsilon}_2^{\bar{j}} \int_X \partial_{\psi_i} \Omega_{\psi} \overline{\partial_{\psi_j} \Omega_{\psi}}}{\int_X \Omega_{\psi} \wedge \bar{\Omega}_{\psi}} + \frac{\varepsilon_1^a \bar{\varepsilon}_2^{\bar{j}} \int_X \partial_{\psi_i} \Omega_{\psi} \wedge \bar{\Omega}_{\psi} \int_X \Omega_{\psi} \wedge \overline{\partial_{\psi_j} \Omega_{\psi}}}{\left(\int_X \Omega_{\psi} \wedge \bar{\Omega}_{\psi}\right)^2}$$

# Reconstructing the metric

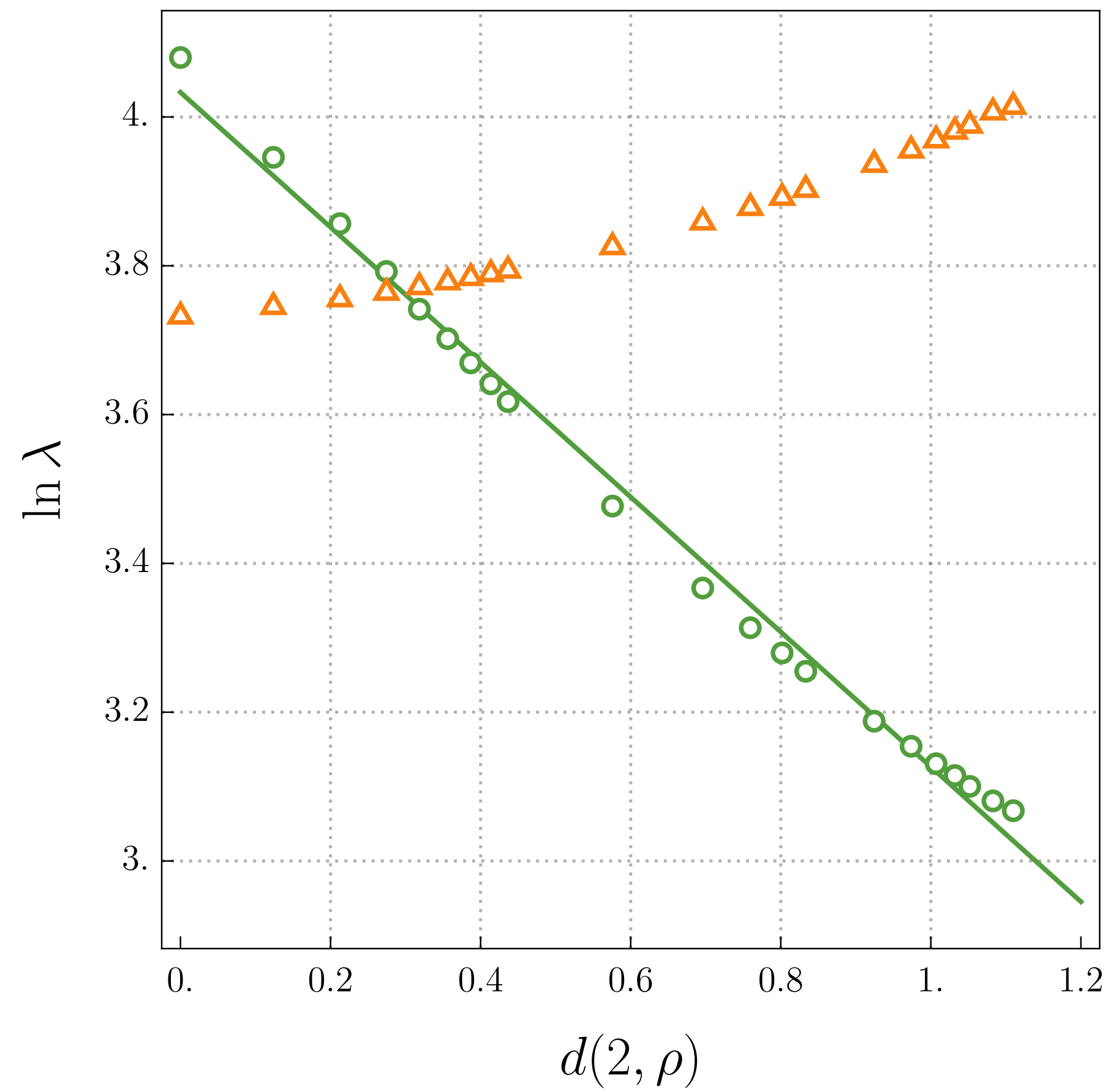
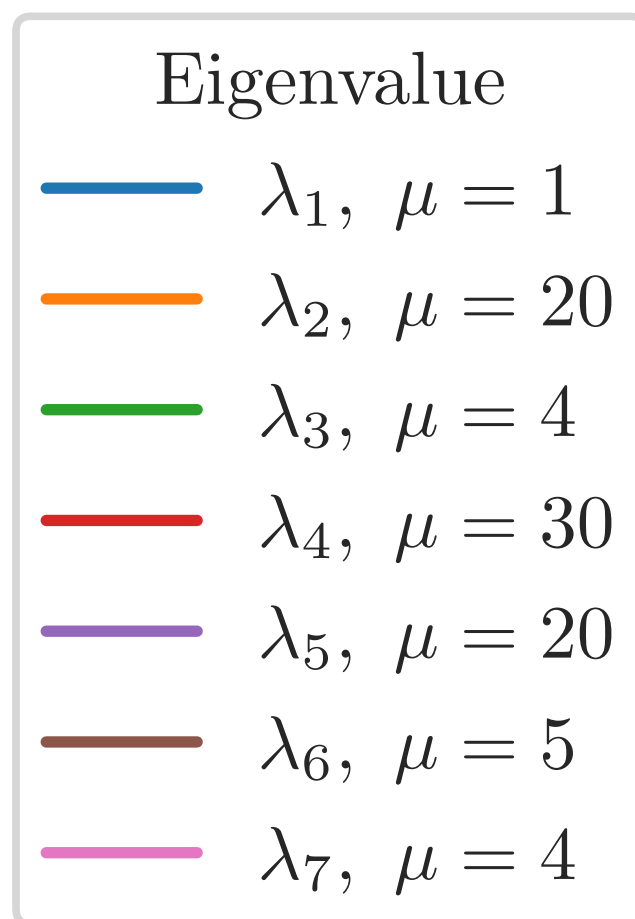
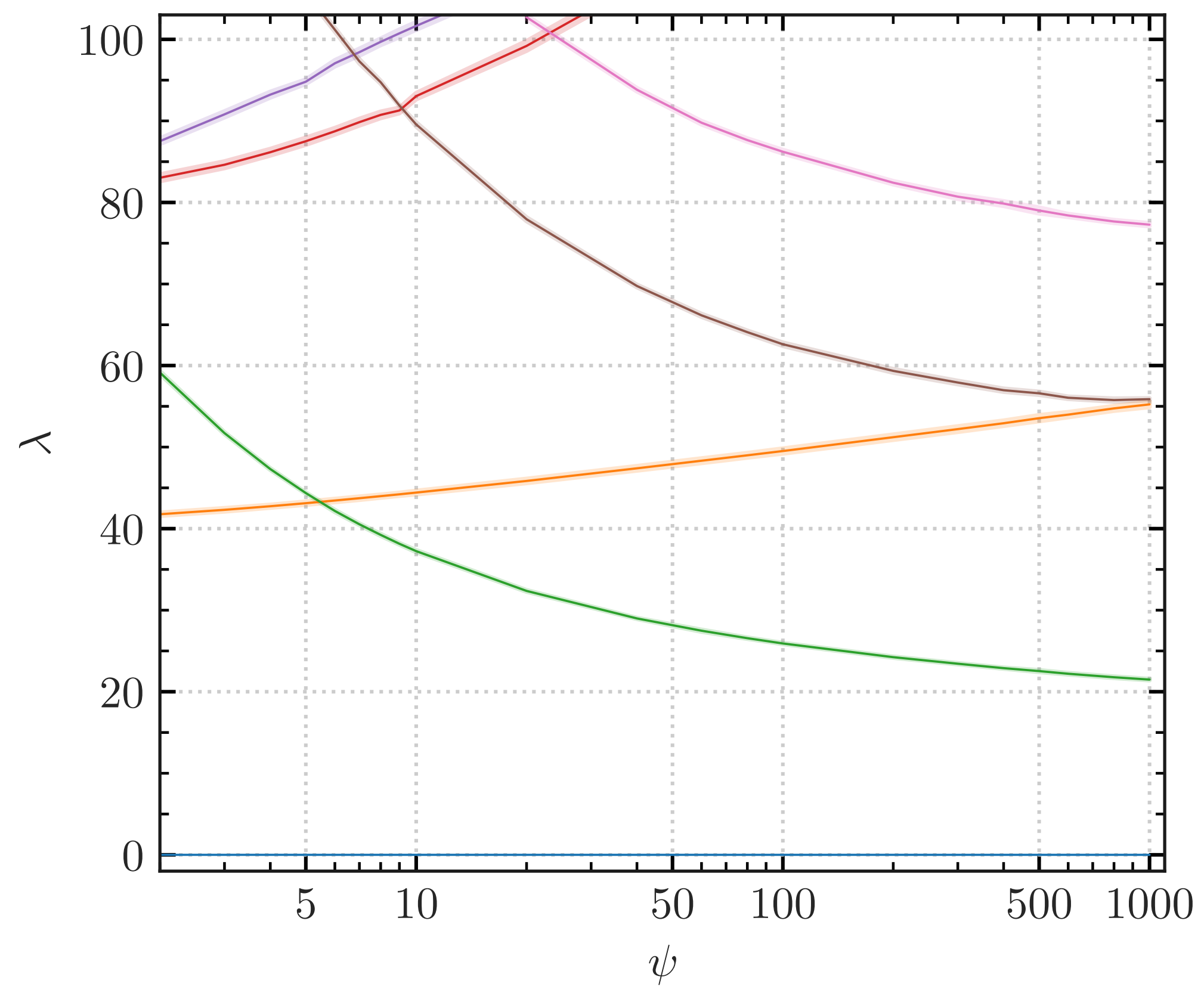
---

- ▶ The numeric computation allows us to get the moduli space metric at fixed points  $\psi$
- ▶ We **compute** the value of the **metric** in a box around the start and end point of the geodesic trajectory **and interpolate** the full metric
  - **Fitting polynomials** / splines
  - **Use** a **NN** similar to the one used for the CY metric
  - **Use symbolic regression** (FindFormula or AIFeynman) [Mathematica; Udrescu et.al. '20]
  - **Use domain knowledge** (cheating) of underlying Picard Fuchs equation to fit a function  $g(\psi) = \frac{a}{\psi^2} + \frac{b}{(\psi \ln \psi)^2}$  with L1 norm and Tikhonov regularization

# Geodesics



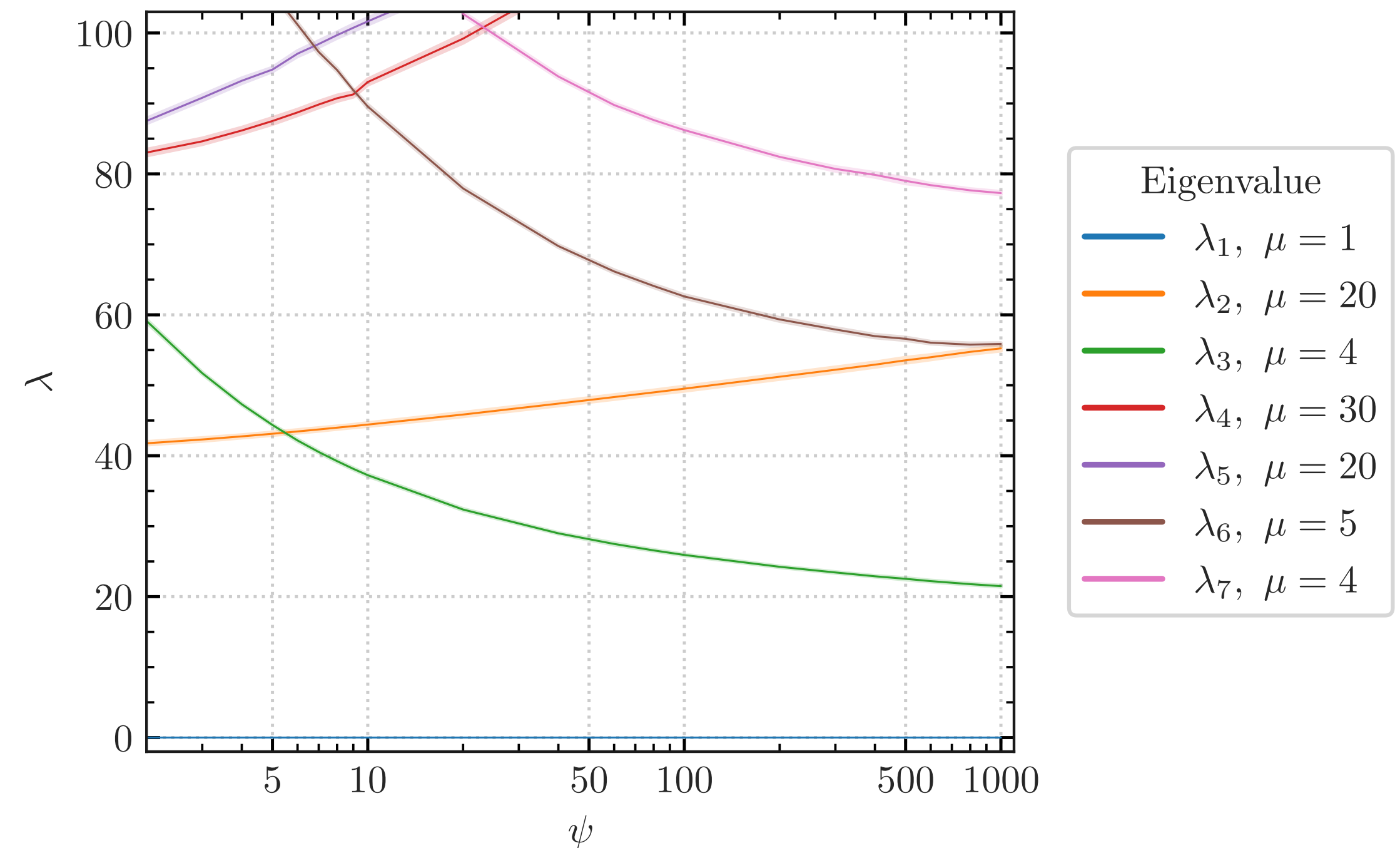




# Results

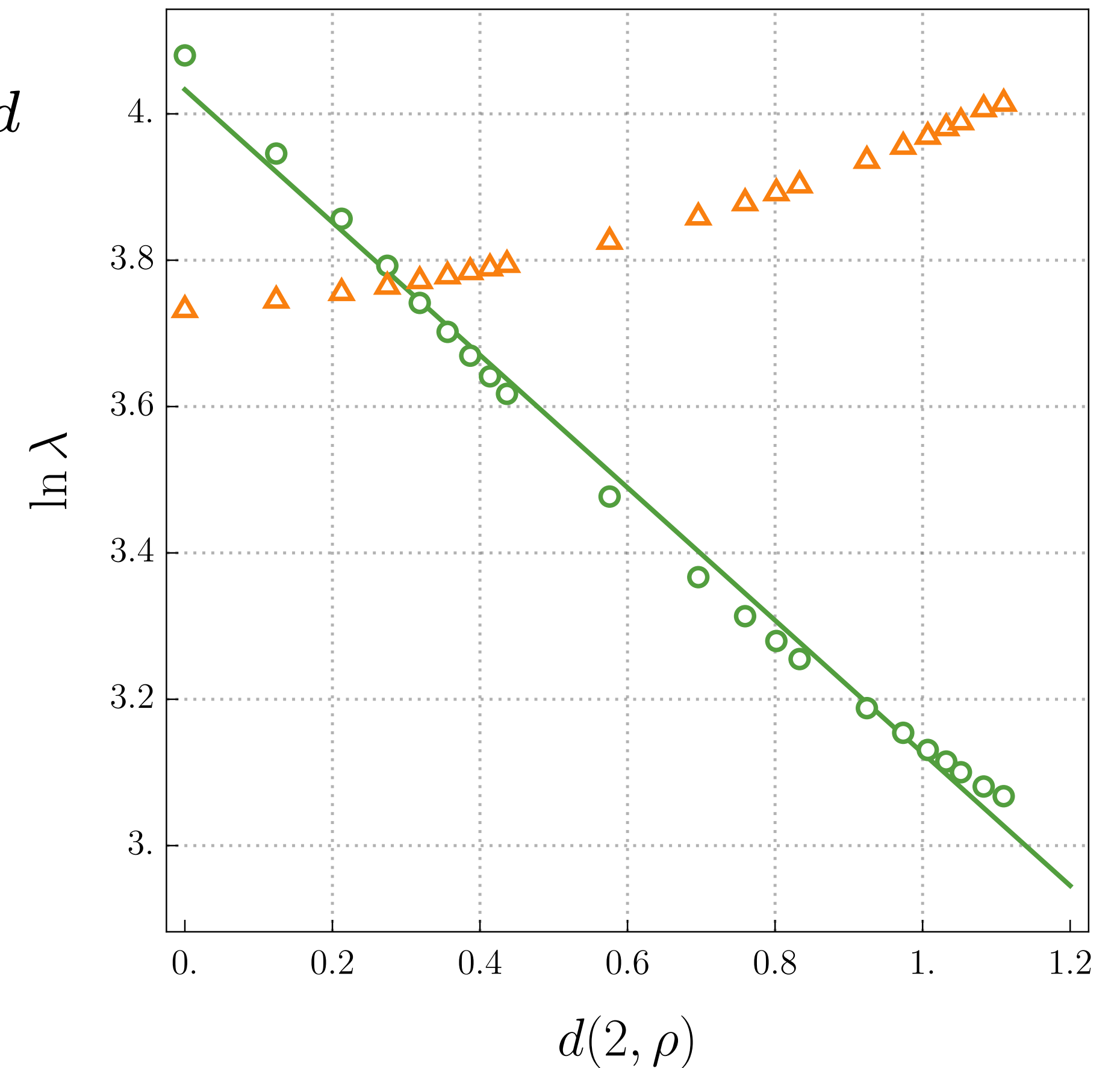
# Results - I

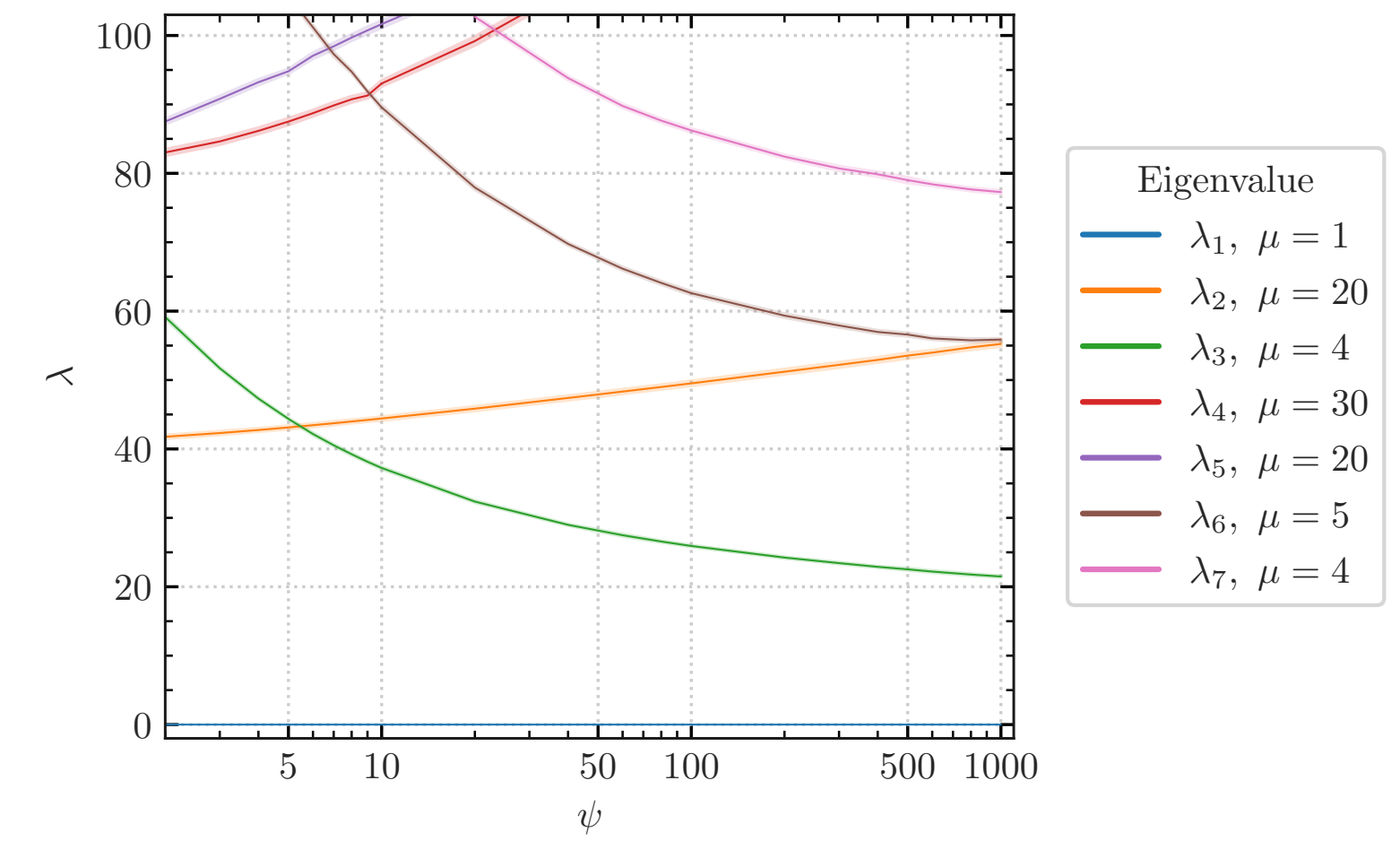
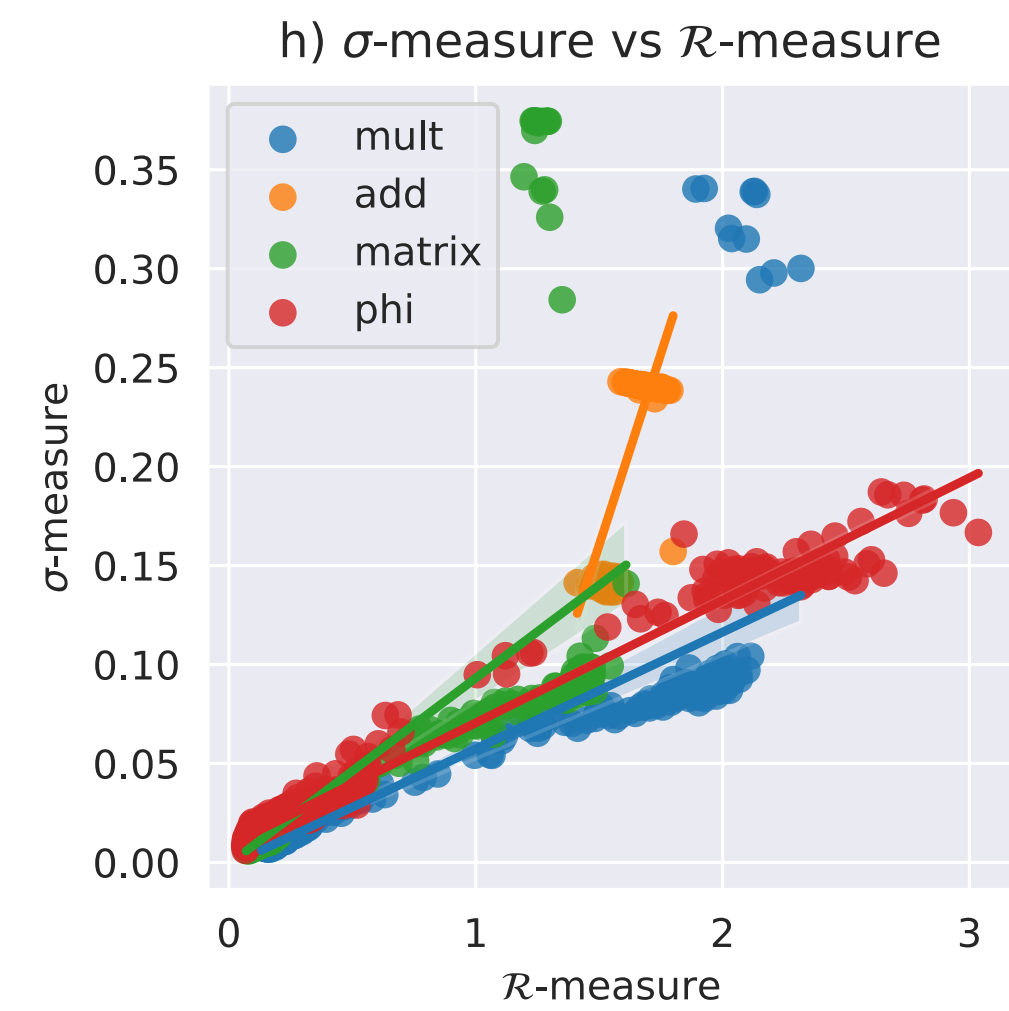
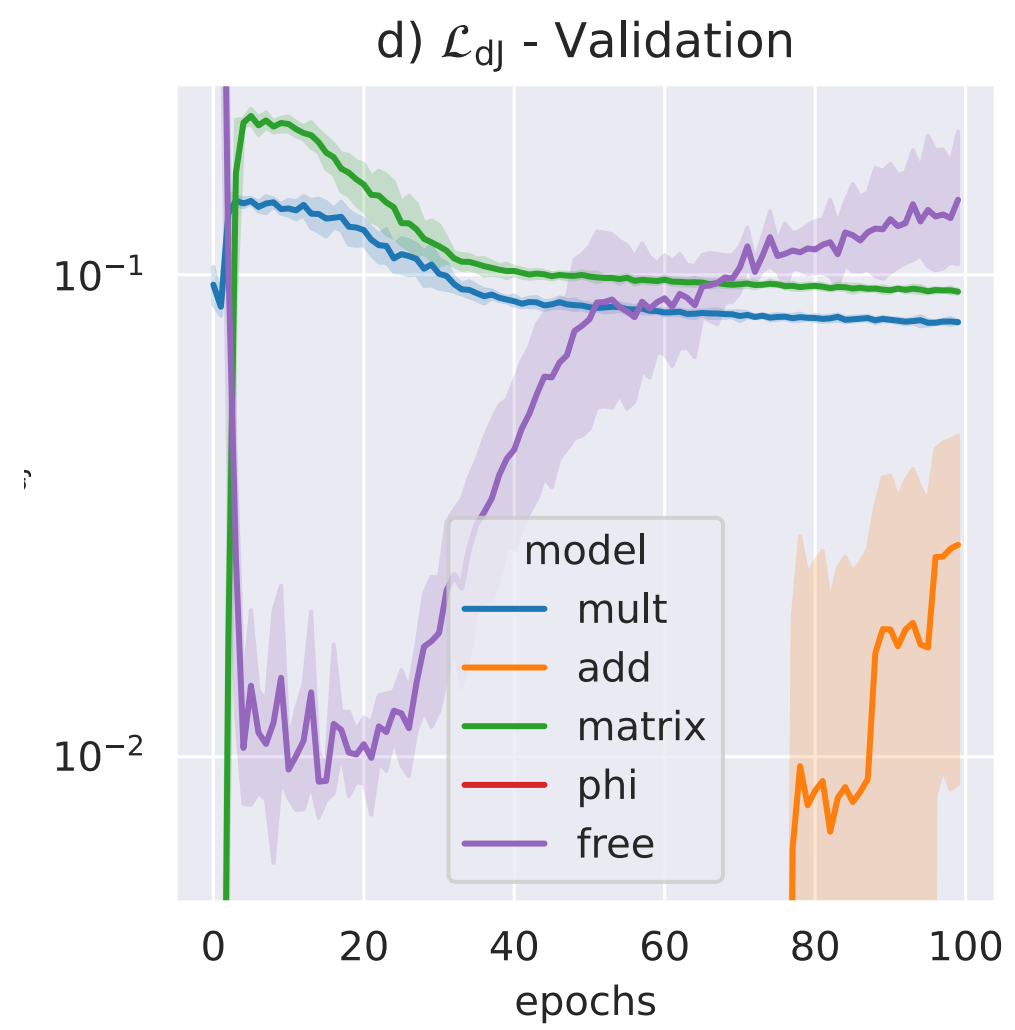
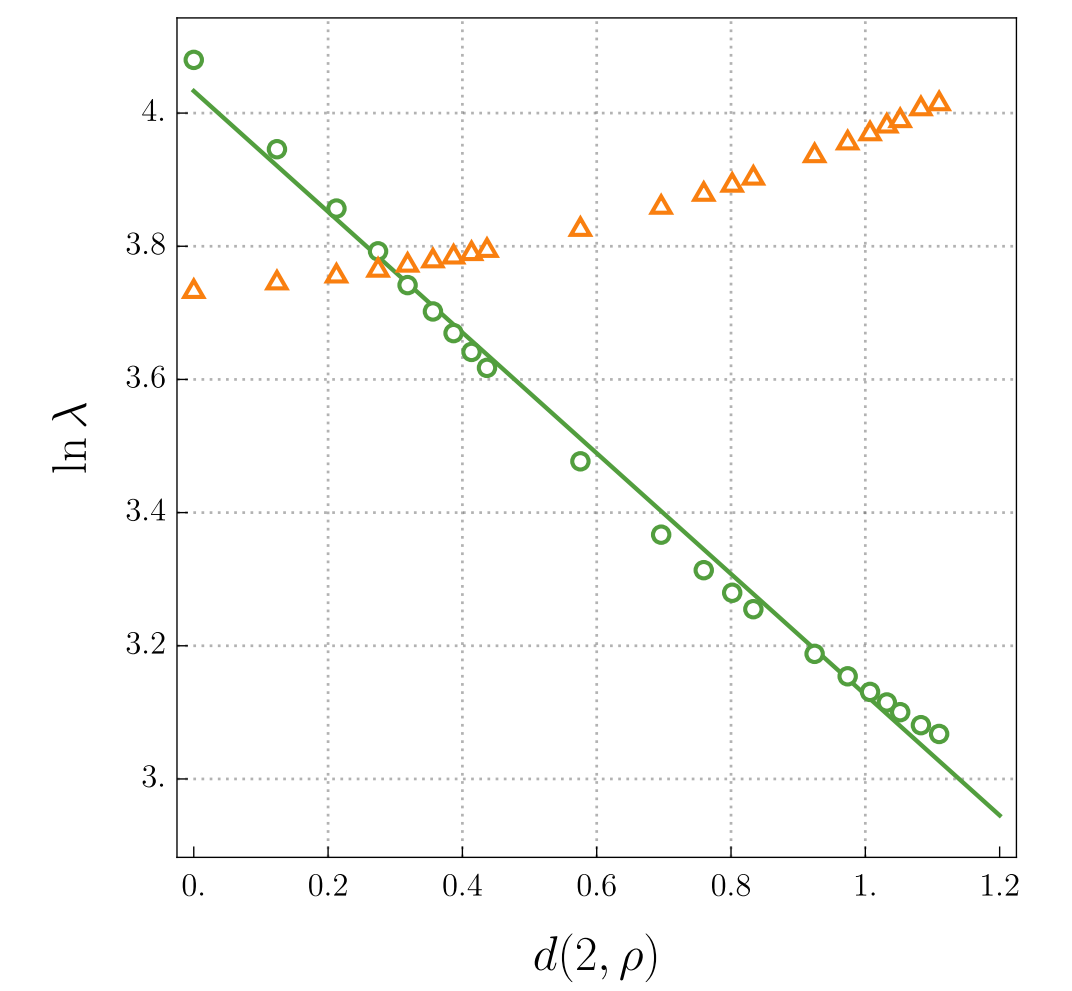
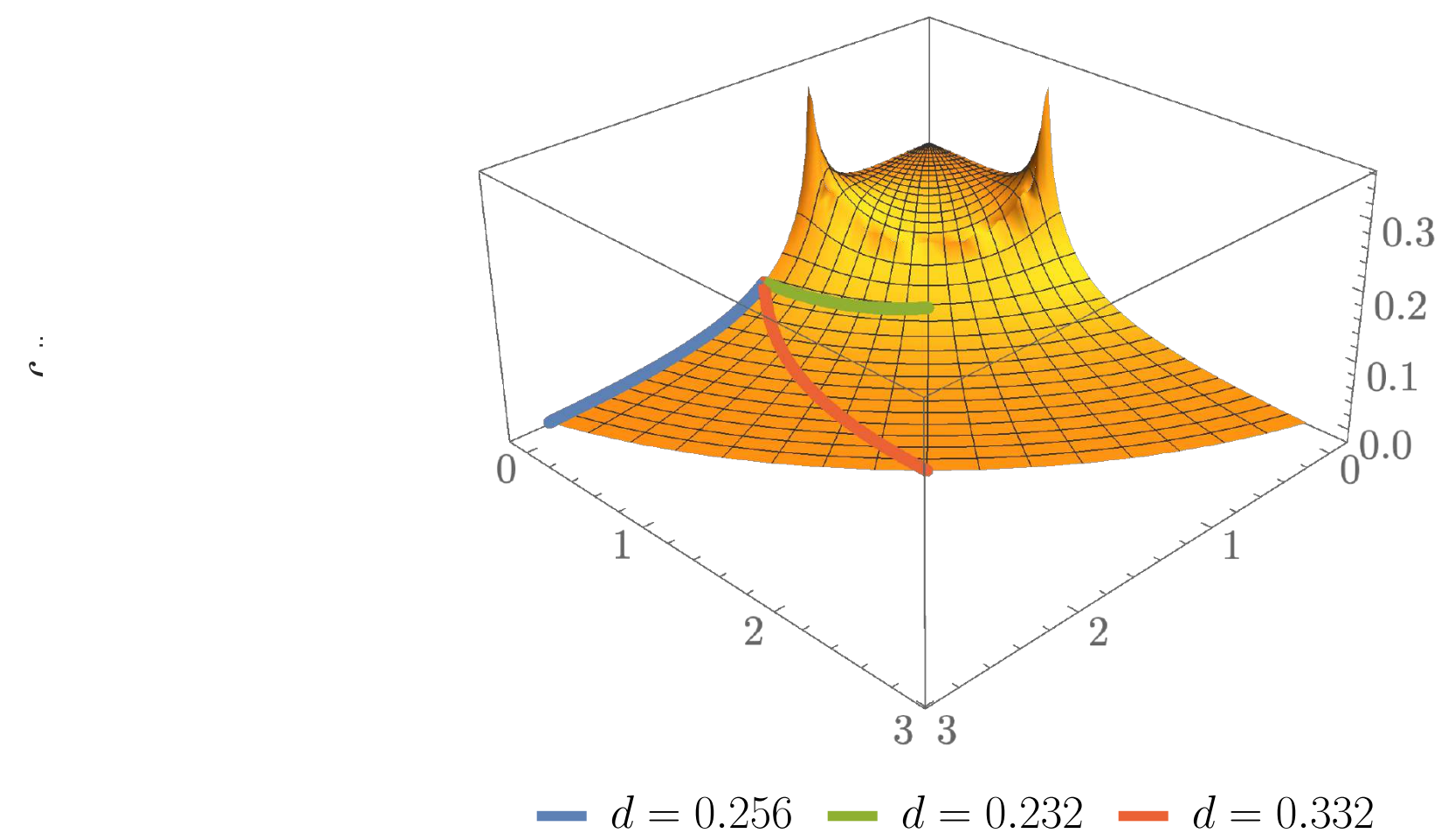
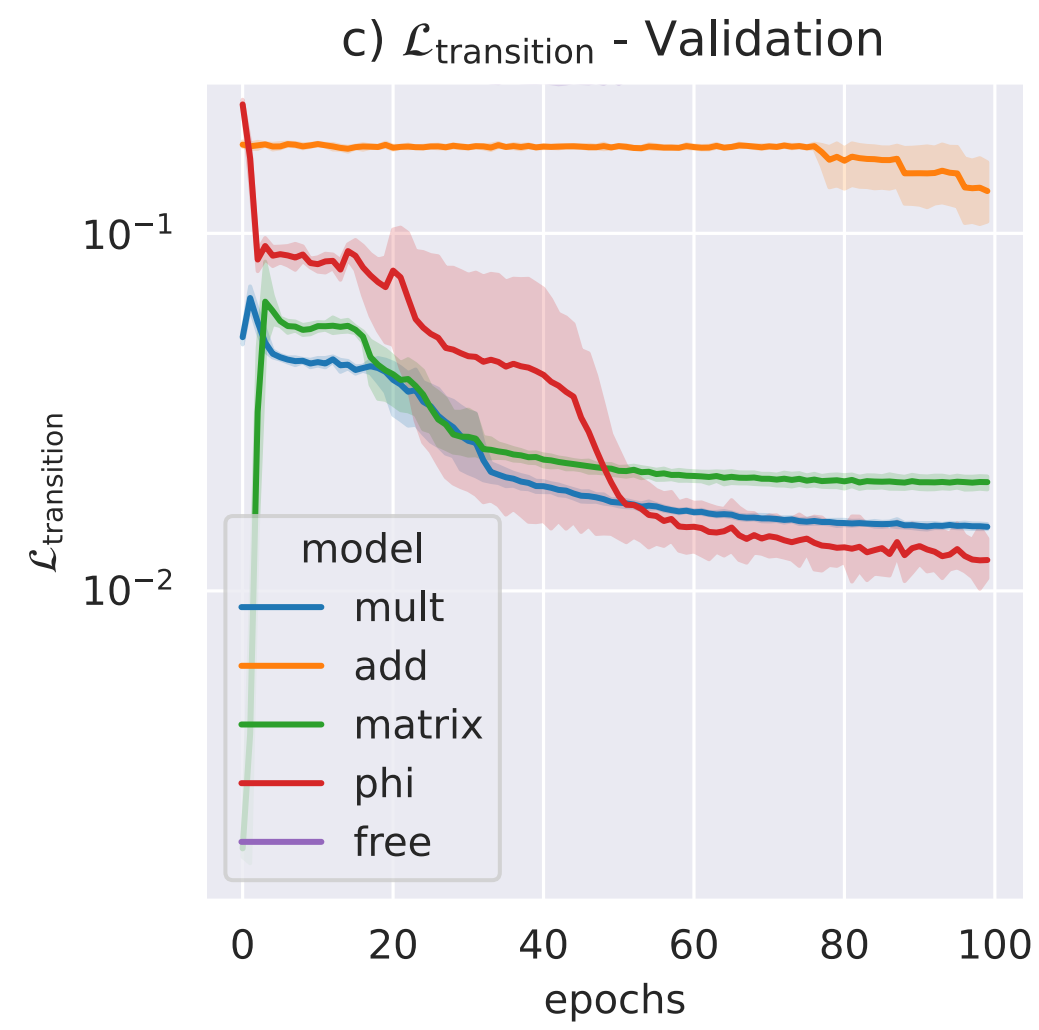
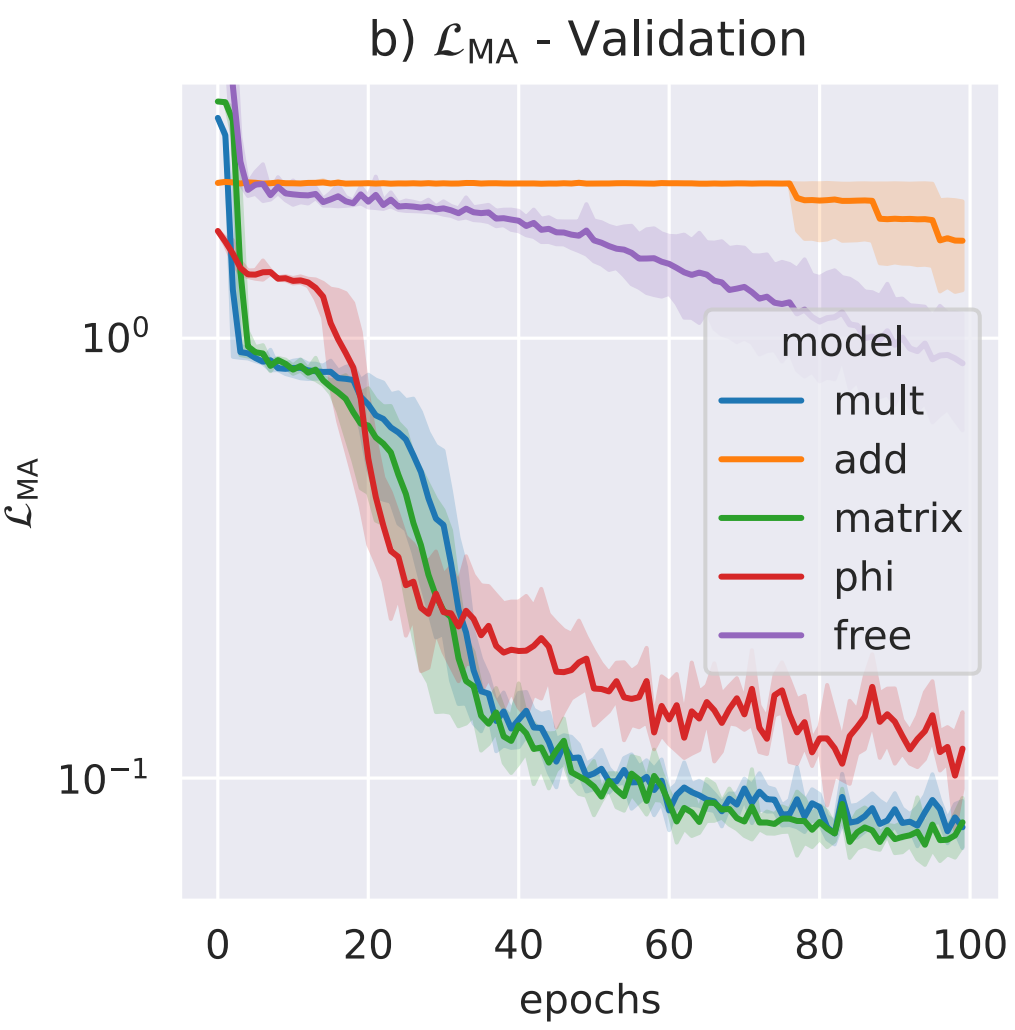
- ▶ One massless scalar ( $\psi$ )
- ▶ **Spectrum degenerate** w/ degeneracies given by **irreps of CY symmetry group**
- ▶ **Eigenvalues** (mass levels) **cross** in contrast (but not contradiction) to
  - No-crossing theorems in QM
  - Eigenvalue repulsion in RMT for hermitian matrices
- ▶ **States** with **large multiplicity** become **heavier**, states with **small multiplicity** become **lighter**



# Results - II

- ▶ **Scalars** become **exponentially light**
- ▶ Best fit gives  $m = \lambda^{1/2} = 7.5e^{-(0.45 \pm 0.02)d}$
- ▶ First numerical check of the SDC
- ▶ Value very close to conjectured lower bound for  $\alpha = 1/\sqrt{6}$  [Andriot,Cribiori,Erkinger `20]
- ▶ Interesting effect at sub-Planckian distances: mass of lightest state does not change for  $.3 M_P$



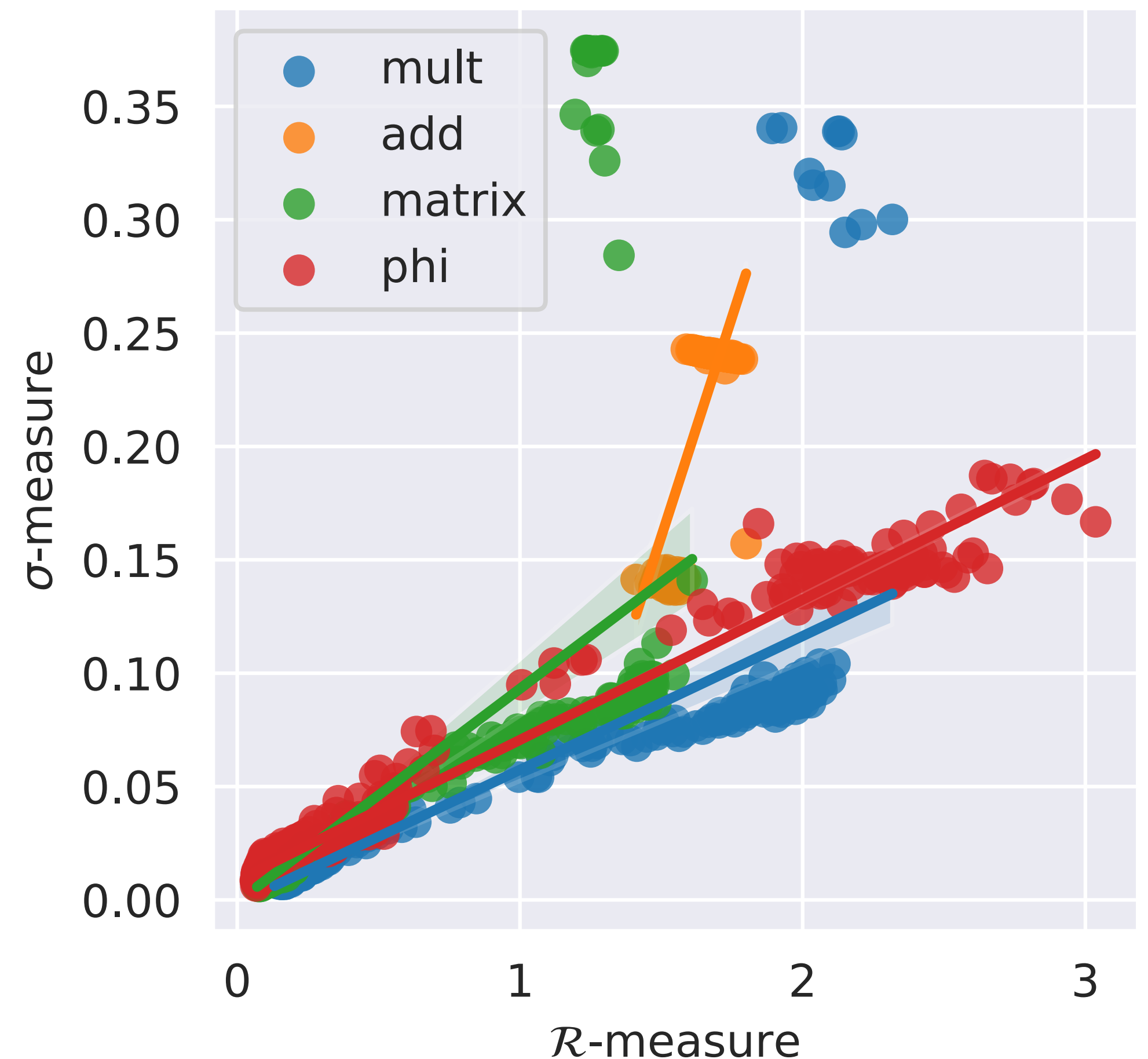


# Conclusions



# Conclusions - Part I

- ▶ Extended the techniques to compute CY metrics to the vast majority of known CY constructions (Kreuzer Skarke, CICYs)
- ▶ Computations possible for arbitrary number of Kähler moduli
- ▶ Develop MA NN that allows to fix the Kähler class
- ▶ Provide fast implementation that integrates into Mathematica and Sage
- ▶ Observe linear relation between MA equation



# Conclusions - Part II

- ▶ Extended techniques to compute CY metrics
- ▶ Verified the SDC
  1. Use NNs to approximate CY metric and MC integration to compute massive string spectrum
  2. Use periods + mirror symmetry or numerical methods + MC integration + interpolation to compute CS moduli space metric
  3. Use PDE shooting methods to solve geodesics equations
  4. Fit masses to distance
- ▶ Surprising level-crossing and dependence on space-time symmetry group observed

