# Learning Random Matrix Approximations of String Theory: Kähler Metrics

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#### Abstract

We do lots of interesting things.

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### 1 Introduction

## 2 Methods

#### 2.1 Optimization of Bergman Metrics

#### 2.1.1 A brief review of Bergman metrics

In the same way that many functions can be well-approximated by a Taylor series, one might hope to find a universal class of metrics that converge to the metrics of interest. For Kähler metrics, the class of Bergman metrics serve this purpose. It was shown by (Tian-Yau-Zelditch) that a Bergman metric can provide an arbitrarily good approximation to a given Kähler metric, in a fixed Kähler class. To review, let us consider a compact Kähler manifold M with an effective line bundle L, such that  $L^k$  is ample for some k > 0. We consider a basis of sections of  $L^k$  of the form  $s^{\alpha}(z^i)$ , where the  $z^i$  are projective coordinates on M. The Kähler potential for the Bergman metric takes the form

$$k = \log(s_{\alpha} P_{\alpha\bar{\beta}} \bar{s}_{\bar{\beta}}), \qquad (1)$$

where the matrix P is positive-definite, and therefore can be written as  $P = A^{\dagger}A$ .

- 2.2 New Ensembles from Generative Adversarial Networks
- 3 Kähler metrics at fixed Picard Number
- **3.1**  $h^{11} = 10$
- **3.2**  $h^{11} = 16$
- 4 Interpolation and Extrapolation in Picard Number
- 5 Discussion