Constructing the Standard Model from String Theory



UNIVERSITY OF

YTF 17/12/2021

Thomas Harvey In Collaboration with: Steve Abel, Andrei Constantin and Andre Lukas

2108.07316 and 2110.14029

Overview

- 1. Heterotic String Theory
- 2. Calabi-Yau Manifolds
- 3. Vector Bundles
- 4. Monad Bundles
- 5. Modern Computational Methods
- 6. Conclusion

I will try and be as pedagogical as possible



• Two of the five superstring theories



- Two of the five superstring theories
- Low energy: 10D N=1 SUGRA, coupled to SYM
 - Gauge Group **<u>E8 X E8</u>** or SO(32) both contain the standard model as subgroups

$$S = \frac{1}{2\kappa_{10}^2} \int d^{10}x \sqrt{-g} e^{-2\phi} \left(R - \frac{1}{2} |H_3|^2 + 4\partial\phi^2 - \frac{\alpha'}{8} \left[tr(F^2) - tr(R^2) \right] + (fermions) \right)$$

- Two of the five superstring theories
- Low energy: 10D N=1 SUGRA, coupled to SYM
 - Gauge Group **E8 X E8** or SO(32) both contain the standard model as subgroups

$$S = \frac{1}{2\kappa_{10}^2} \int d^{10}x \sqrt{-g} e^{-2\phi} \left(R - \frac{1}{2} |H_3|^2 + 4\partial\phi^2 - \frac{\alpha'}{8} \left[tr(F^2) - tr(R^2) \right] + (fermions) \right)$$

- Need to compactify, subject to two constraints $M_{10} = \mathbb{R}^{1,3} \times M_6$
 - Solves equations of motion of S
 - Looks like the standard model coupled to gravity



- Two of the five superstring theories
- Low energy: 10D N=1 SUGRA, coupled to SYM
 - Gauge Group **<u>E8 X E8</u>** or SO(32) both contain the standard model as subgroups

$$S = \frac{1}{2\kappa_{10}^2} \int d^{10}x \sqrt{-g} e^{-2\phi} \left(R - \frac{1}{2} |H_3|^2 + 4\partial\phi^2 - \frac{\alpha'}{8} \left[tr(F^2) - tr(R^2) \right] + (fermions) \right)$$

- Need to compactify, subject to two constraints $M_{10} = \mathbb{R}^{1,3} \times M_6$
 - Solves equations of motion of S
 - Looks like the standard model coupled to gravity
- First attempt: "Vacuum Configurations for Superstrings" Candelas, Horowitz, Strominger and Witten 1985
 - There are other approaches! F-Theory, intersecting D-Branes...

• Different choices of M_6 leads to different 4D physics



- Different choices of M_6 leads to different 4D physics
- Simplest choice for M_6 is T_6
 - \circ $\,$ Leads to N=4 SUSY in 4D $\,$



- Different choices of M_6 leads to different 4D physics
- Simplest choice for M_6 is T_6
 - \circ Leads to N=4 SUSY in 4D
 - Need a manifold with reduced holonomy





- Different choices of M_6 leads to different 4D physics
- Simplest choice for M_6 is T_6
 - Leads to N=4 SUSY in 4D
 - Need a manifold with reduced holonomy
- Calabi-Yau 3-folds
 - 3 complex dimensions, with SU(3) holonomy





- Different choices of M_6 leads to different 4D physics
- Simplest choice for M_6 is T_6
 - Leads to N=4 SUSY in 4D
 - Need a manifold with reduced holonomy
- Calabi-Yau 3-folds
 - 3 complex dimensions, with SU(3) holonomy
 - Leads to N=1 SUSY in 4D phenomenologically viable!
 - Gives access to powerful computational techniques (Algebraic Geometry)





- Different choices of M_6 leads to different 4D physics
- Simplest choice for M_6 is T_6
 - Leads to N=4 SUSY in 4D
 - Need a manifold with reduced holonomy
- Calabi-Yau 3-folds
 - 3 complex dimensions, with SU(3) holonomy
 - Leads to N=1 SUSY in 4D phenomenologically viable!
 - Gives access to powerful computational techniques (Algebraic Geometry)
- Note, we will have extra scalar fields in 4D





- Different choices of M_6 leads to different 4D physics
- Simplest choice for M_6 is T_6
 - Leads to N=4 SUSY in 4D
 - Need a manifold with reduced holonomy
- Calabi-Yau 3-folds
 - 3 complex dimensions, with SU(3) holonomy
 - Leads to N=1 SUSY in 4D phenomenologically viable!
 - Gives access to powerful computational techniques (Algebraic Geometry)
- Note, we will have extra scalar fields in 4D
 - Couplings appear as functions of these scalars
 - "Moduli Stabilisation"



• Vector bundles (collections of fields on a manifold)





- Vector bundles (collections of fields on a manifold)
 - Fibre-Bundle B over M, looks locally like M x F
 - Vector Bundle F is a vector space





- Vector bundles (collections of fields on a manifold)
 - Fibre-Bundle B over M, looks locally like M x F
 - Vector Bundle F is a vector space
- Sections of bundles are F-valued field configurations over M



- Vector bundles (collections of fields on a manifold)
 - Fibre-Bundle B over M, looks locally like M x F
 - Vector Bundle F is a vector space
- Sections of bundles are F-valued field configurations over M
- Can have non-trivial topology







• The gauge fields are forced to have a non-trivial value in the hidden dimensions - encoded by a bundle



• The gauge fields are forced to have a non-trivial value in the hidden dimensions - encoded by a bundle

$$H \neq dB_2$$

$$H = dB_2 - \alpha'/4[\omega_3(A) + \omega_3(R)]$$

$$\Rightarrow c_2(TX) - c_2(V) = 0$$

• This looked very restrictive -> Standard Embedding



• The gauge fields are forced to have a non-trivial value in the hidden dimensions - encoded by a bundle

$$H \neq dB_2$$

$$H = dB_2 - \alpha'/4[\omega_3(A) + \omega_3(R)]$$

$$\Rightarrow c_2(TX) - c_2(V) = 0$$

- This looked very restrictive -> Standard Embedding
- Can actually add non-perturbative physics, and is actually less restrictive

$$\Rightarrow c_2(TX) + c_2(V) = (NS5 - Branes)$$

• The gauge fields are forced to have a non-trivial value in the hidden dimensions - encoded by a bundle

$$H \neq dB_2$$

$$H = dB_2 - \alpha'/4[\omega_3(A) + \omega_3(R)]$$

$$\Rightarrow c_2(TX) - c_2(V) = 0$$

- This looked very restrictive -> Standard Embedding
- Can actually add non-perturbative physics, and is actually less restrictive

 $\Rightarrow c_2(TX) + c_2(V) = (NS5 - Branes)$

• Different choice of V, lead to different 4D physics

• Ignore one of our E8 factors (Hidden Bundle)



- Ignore one of our E8 factors (Hidden Bundle)
- Consider a 4D SO(10) GUT theory made from the other E8 factor
 - Hidden dimensions will need an SU(4) bundle $E_8 \supset SO(10) imes SU(4)$



- Ignore one of our E8 factors (Hidden Bundle)
- Consider a 4D SO(10) GUT theory made from the other E8 factor
 - Hidden dimensions will need an SU(4) bundle $E_8 \supset SO(10) imes SU(4)$
- Constructing Non-Abelian bundles on a Calabi-Yau is hard!



- Ignore one of our E8 factors (Hidden Bundle)
- Consider a 4D SO(10) GUT theory made from the other E8 factor
 - \circ Hidden dimensions will need an SU(4) bundle $E_8 \supset SO(10) imes SU(4)$
- Constructing Non-Abelian bundles on a Calabi-Yau is hard!
 - Our approach build it out of simpler objects



- Ignore one of our E8 factors (Hidden Bundle)
- Consider a 4D SO(10) GUT theory made from the other E8 factor
 - Hidden dimensions will need an SU(4) bundle $E_8 \supset SO(10) imes SU(4)$
- Constructing Non-Abelian bundles on a Calabi-Yau is hard!
 - Our approach build it out of simpler objects
- Line bundles are the simplest vector bundle
 - Look locally like C x M (or R x M for real)



• We can classify line bundles by a vector of integers



- We can classify line bundles by a vector of integers
- We can construct two sums of line bundles B and C
 - \circ rk(B) rk(C) = #LB(B) #LB(C) = 4 for SU(4) bundle



- We can classify line bundles by a vector of integers
- We can construct two sums of line bundles B and C
 - \circ rk(B) rk(C) = #LB(B) #LB(C) = 4 for SU(4) bundle
- Define a Monad bundle V by a short-exact sequence

 $0 \to V \to B \to C \to 0$



- We can classify line bundles by a vector of integers
- We can construct two sums of line bundles B and C
 - \circ rk(B) rk(C) = #LB(B) #LB(C) = 4 for SU(4) bundle
- Define a Monad bundle V by a short-exact sequence

$0 \to V \to B \to C \to 0$

- Intuitively: $V \sim B/C$
- V is now given by a large number of integers

Reinforcement Learning (RL): An agent is trained to find gain rewards by exploring some

large parameter space.



Reinforcement Learning (RL): An agent is trained to find gain rewards by exploring some

large parameter space.

Genetic Algorithms (GA): Survival of the fittest, starting with a random population from a large parameter space.



Reinforcement Learning (RL): An agent is trained to find gain rewards by exploring some

large parameter space.

Genetic Algorithms (GA): Survival of the fittest, starting with a random population from a large parameter space.

See "Data science applications to string theory" by Fabian Ruehle for a review

https://www.sciencedirect.com/science/article/pii/S0370157319303072

- Aim of RL Maximise Reward
- Aim of GA Maximise Fitness



- Aim of RL Maximise Reward
- Aim of GA Maximise Fitness
- String Model building often requires searching large parameter spaces
 - Such as the large spaces of integers to define a monad bundle



- Aim of RL Maximise Reward
- Aim of GA Maximise Fitness
- String Model building often requires searching large parameter spaces
 - Such as the large spaces of integers to define a monad bundle
- Define a reward/fitness that is maximised when a theory looks like the standard model



lodern C		property	term in $v(B,C)$	comment	
		index match	$-rac{2 \mathrm{ind}(V)-\tau }{hM^3}$	$\tau = -3 \Gamma $ is the target index,	
			10111	ind(V) computed from Eq. (2.20)	
I	Aim of RL - I	anomaly	$\frac{1}{hM^2} \sum_{i=1}^{h} \min\left(c_{2i}(TX) - c_{2i}(V), 0\right)$	no penalty if anomaly condition satisfied, $c_{2i}(V)$ computed from Eq. (2.20)	
	Aims of CA	bundleness	$-(d_{\text{deg}}+1)$	$d_{\text{deg}} = \text{dimension of degeneracy locus}$	
	AIM OF GA -			as discussed in Sec. 2.4; if the degeneracy locus is empty, d_{deg} is to be taken as -1	
	String Mode	split bundle	$-n_{ m split}$	$n_{\text{split}} = \text{number of splits in } V$	
	• Such as	equivariance	$-\sum_{U\subset B,C} \operatorname{mod}(\operatorname{ind}(U), \Gamma)$	U runs over all line bundles in B, C	
1	Define a rew			or blocks of same line bundles, as discussed in Sec. 2.4	ıda
		trivial bundle	$-n_{ m trivial}$	$n_{\rm trivial} = {\rm number of trivial line bundles}$	
		stability V	$-\frac{\max(0, h^0(X, B) - h^0(X, C))}{hM^3}$	tests Hoppe's criterion for V ,	
		stability V^*	$-\frac{\max(0, h^0(X, B^*) - h^0(X, C^*))}{hM^3}$	tests Hoppe's criterion for V^* , cohomologies from formulae in Sec. 2.3	

Table 2: Contributions to the intrinsic value for the monad environment. The intrinsic value v(B,C) is the sum of all eight terms and $M = \max(b_{\max}, c_{\max})$.

Idard model

- Aim of RL Maximise Reward
- Aim of GA Maximise Fitness
- String Model building often requires searching large parameter spaces
 - Such as the large spaces of integers to define a monad bundle
- Define a reward/fitness that is maximised when a theory looks like the standard model
 - \circ ~ Agent can move through the integer lattice of theories/monad bundles



- Aim of RL Maximise Reward
- Aim of GA Maximise Fitness
- String Model building often requires searching large parameter spaces
 - Such as the large spaces of integers to define a monad bundle
- Define a reward/fitness that is maximised when a theory looks like the standard model
 - Agent can move through the integer lattice of theories/monad bundles
- These were hugely successful, on sizes up to 10^22!
 - 28 unique models found for the rank-(6,2) bicubic
 - after removing redundancies and further checks
 - Only one was known before!



Reinforcement Learning Plots



(a) Loss vs batch number.



(d) TD return vs batch number.



(b) Policy loss vs batch number.



(e) Number of terminal states vs episode number.



(c) Value loss vs batch number.



(f) Terminal fraction vs episode number.

Figure 6: Training metrics for the bicubic monad environment with $(r_B, r_C) = (6, 2)$.

Genetic Algorithms Plots



(a) Fitness histogram: number of individuals as a function of generation and fitness.



Figure 1: Performance measures for a typical GA initialisation on the bicubic.

RL Vs GA









(b) Number of perfect states found as a function of the number of states visited.

• Can construct the SM from ST is by compactifying Heterotic ST



- Can construct the SM from ST is by compactifying Heterotic ST
- The low energy theory is specified by the choice of a vector bundle and manifold



- Can construct the SM from ST is by compactifying Heterotic ST
- The low energy theory is specified by the choice of a vector bundle and manifold
 - We chose a CY3 for N=1 SUSY in 4D and computational control
 - Monad bundles are specified by a large number of integers



- Can construct the SM from ST is by compactifying Heterotic ST
- The low energy theory is specified by the choice of a vector bundle and manifold
 - We chose a CY3 for N=1 SUSY in 4D and computational control
 - Monad bundles are specified by a large number of integers
- We can, and have, used RL and GA to find models on these spaces
 - 28 on the (6,2) bicubic, including the one known model



- Can construct the SM from ST is by compactifying Heterotic ST
- The low energy theory is specified by the choice of a vector bundle and manifold
 - We chose a CY3 for N=1 SUSY in 4D and computational control
 - Monad bundles are specified by a large number of integers
- We can, and have, used RL and GA to find models on these spaces
 - 28 on the (6,2) bicubic, including the one known model
- Thank you for listening!

