

Computing Yukawa Couplings from String Theory



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University of Oxford

EuroStrings 2025, Nordita, Stockholm, August 2025

based on: [arXiv:2402.01615](#), [arXiv:2507.03076](#), [arXiv:2508.nnnn](#)

in collaboration with: Andrei Constantin, Kit Fraser-Taliente, Thomas Harvey,
Lucas Leung, Luca Nutricati, Burt Ovrut

Outline

- Introduction: Yukawa couplings – why are they hard to compute?
- The tool: learning solutions to differential equations
- A heterotic line bundle example model
- Results
- Heterotic Froggatt–Nielsen models
- Conclusion

Introduction

Standard models from string theory

- It is now possible to engineer (many) string models with the exact (MS)SM spectrum.
- Models are based on heterotic CY compactifications with general vector bundles.
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Low-energy Yukawa couplings

Matter fields: C^I

Moduli fields: ϕ^A

Superpotential:

Kahler potential:

$$W = Y_{IJK}(\phi) C^I C^J C^K + \dots$$

$$K = K_{IJ}^{\text{matter}}(\phi, \bar{\phi}) C^I \bar{C}^J + \dots$$

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For example, up-quark Yukawa couplings:

$$W = Y_{ij}^u(\phi) H^u Q^i U^j + \dots$$

$$K = K_{ij}^Q(\phi, \bar{\phi}) Q^i \bar{Q}^j + K_{ij}^u(\phi, \bar{\phi}) U^i \bar{U}^j \\ + k(\phi, \bar{\phi}) H^u \bar{H}^u + \dots$$

Heterotic CY models

Compactification data:

- CY three-fold X with freely acting symmetry group Γ
- ‘Observable’ vector bundle $V \rightarrow X$, Γ - equivariant, poly-stable, with structure group $SU(5), S(U(1)^5), \dots$
- ‘Hidden’ vector bundle $\tilde{V} \rightarrow X$, Γ - equivariant, poly-stable, with structure group that embeds into E_8

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Outcome (for appropriate choices): MSSM + moduli + hidden sectors

Specific model classes

- ‘Standard embedding’ models: $V = TX$, $\tilde{V}$ trivial, no 5-branes
-> analytic formulae for physical Yukawa coupling,
limited model building options

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matter fields $\leftrightarrow$ bundle-valued harmonic forms 1-forms

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can be computed analytically
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Computing K_{IJ}^{matter} has been the main problem since Ricci-flat metric, HYM connection and harmonic forms are not known analytically.

Need to solve complicated diff. eqs. on CY manifolds numerically.

Machine-learning solutions to differential equations

Neural networks

Function families $F_\theta : \mathbb{R}^{d_{\text{in}}} \rightarrow \mathbb{R}^{d_{\text{out}}}$ constructed by function composition:

$$F_\theta = f_{\theta_\delta}^{(\delta)} \circ \dots \circ f_{\theta_2}^{(2)} \circ f_{\theta_1}^{(1)} : \mathbb{R}^{d_0} \rightarrow \mathbb{R}^{d_\delta}$$

where $\theta = (\theta_1, \dots, \theta_\delta)$, $d_{\text{in}} = d_0$, $d_{\text{out}} = d_\delta$.

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Function families $F_\theta : \mathbb{R}^{d_{\text{in}}} \rightarrow \mathbb{R}^{d_{\text{out}}}$ constructed by function composition:

The diagram shows the function composition $F_\theta = f_{\theta_\delta}^{(\delta)} \circ \dots \circ f_{\theta_2}^{(2)} \circ f_{\theta_1}^{(1)} : \mathbb{R}^{d_0} \rightarrow \mathbb{R}^{d_\delta}$ enclosed in a red rectangular box. Three red annotations with arrows point to parts of the expression: 'depth' points to the composition operator $\circ$, 'width' points to the input and output spaces $\mathbb{R}^{d_0}$ and $\mathbb{R}^{d_\delta}$, and 'layers' points to the individual functions $f_{\theta_i}^{(i)}$ with a dashed arrow indicating the sequence.

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$$\sigma(z) = \theta(z)z$$

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Red arrows point from the labels "activation fct.", "weights", and "biases" to the corresponding parts of the function: σ , W , and b respectively.

$$\sigma(z) = \theta(z)z \quad (\text{ReLU})$$

Solving differential eqs. with neural networks

Differential eq. $p(D^2g(x), Dg(x), g(x), x) = 0 \quad x \in X$

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- Leads to minimum θ_0 of loss function $\mathcal{L}$

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- The neural net F_{θ_0} with those parameters is an approximate solution

Solving diff. eqs. on CYs

M. Larfors, A. Lukas, F. Ruehle, R. Schneider, arXiv:2111.01436,
cymetric package

- Ricci-flat metric on CY X

diff. eq. (Monge-Ampere)

Ansatz

$$J^3 = \kappa \Omega \wedge \bar{\Omega}$$

$$J = J^{(\text{ref})} + \partial\bar{\partial}\phi$$

loss

$$\mathcal{L}_{\text{metric}} = \left\| 1 - \frac{1}{\kappa} \frac{J(\phi) \wedge J(\phi) \wedge J(\phi)}{\hat{\Omega} \wedge \bar{\hat{\Omega}}} \right\|_1 + \dots$$

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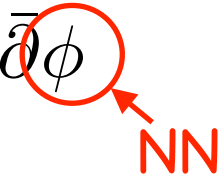
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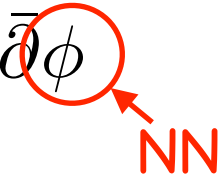
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- Hermitian YM (line) bundle metric on $L \rightarrow X$

diff. eq. (HYM)

Ansatz

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$$J^2 \wedge \partial \bar{\partial} \ln(H) = 0$$

$$H = e^\beta H^{(\text{ref})}$$

$$\mathcal{L}_{\text{HYM}} = ||\Delta \beta - \rho_\beta||_1 + \dots$$

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$$\mathcal{L}_{\text{harmonic}} = \|\Delta_L \sigma - \rho_{\sigma}\|_1 + \dots$$

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Reference quantities are taken to be of 'Fubini-Study' type.

A heterotic example model

Compactification data and low-energy spectrum

- Tetra-quadric CY $X \in \left[\begin{array}{c|c} \mathbb{P}^1 & 2 \\ \mathbb{P}^1 & 2 \\ \mathbb{P}^1 & 2 \\ \mathbb{P}^1 & 2 \end{array} \right]$ with symmetry $\Gamma = \mathbb{Z}_2 \times \mathbb{Z}_2$

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- Line bundle sum $X \leftarrow V = \mathcal{O}_X \begin{array}{c} L_1 \ L_2 \ L_3 \ L_4 \ L_5 \\ \left[\begin{array}{ccccc} -1 & -1 & 0 & 1 & 1 \\ 0 & -3 & 1 & 1 & 1 \\ 0 & 2 & -1 & -1 & 0 \\ 1 & 2 & 0 & -1 & -2 \end{array} \right] \end{array}$

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particle	associated cohomology
$Q^{1,2}, U^{1,2}, E^{1,2}$	$H^1(X, L_2)$
Q^3, U^3, E^3	$H^1(X, L_5)$
$D^{1,2}, L^{1,2}$	$H^1(X, L_2 \otimes L_5)$
D^3, L^3	$H^1(X, L_2 \otimes L_4)$
H^d	$H^1(X, L_2 \otimes L_5)$
H^u	$H^1(X, L_2^* \otimes L_5^*)$

plus bundle, Kahler and complex structure moduli (all SM singlets)

Up-quark Yukawa couplings

Structure of up-quark Yukawas (due to $S(U(1)^5)$ symmetry):

$$Y^u = \begin{pmatrix} 0 & 0 & \lambda_1 \\ 0 & 0 & \lambda_2 \\ \lambda_3 & \lambda_4 & 0 \end{pmatrix}$$

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Up-quark Yukawa couplings

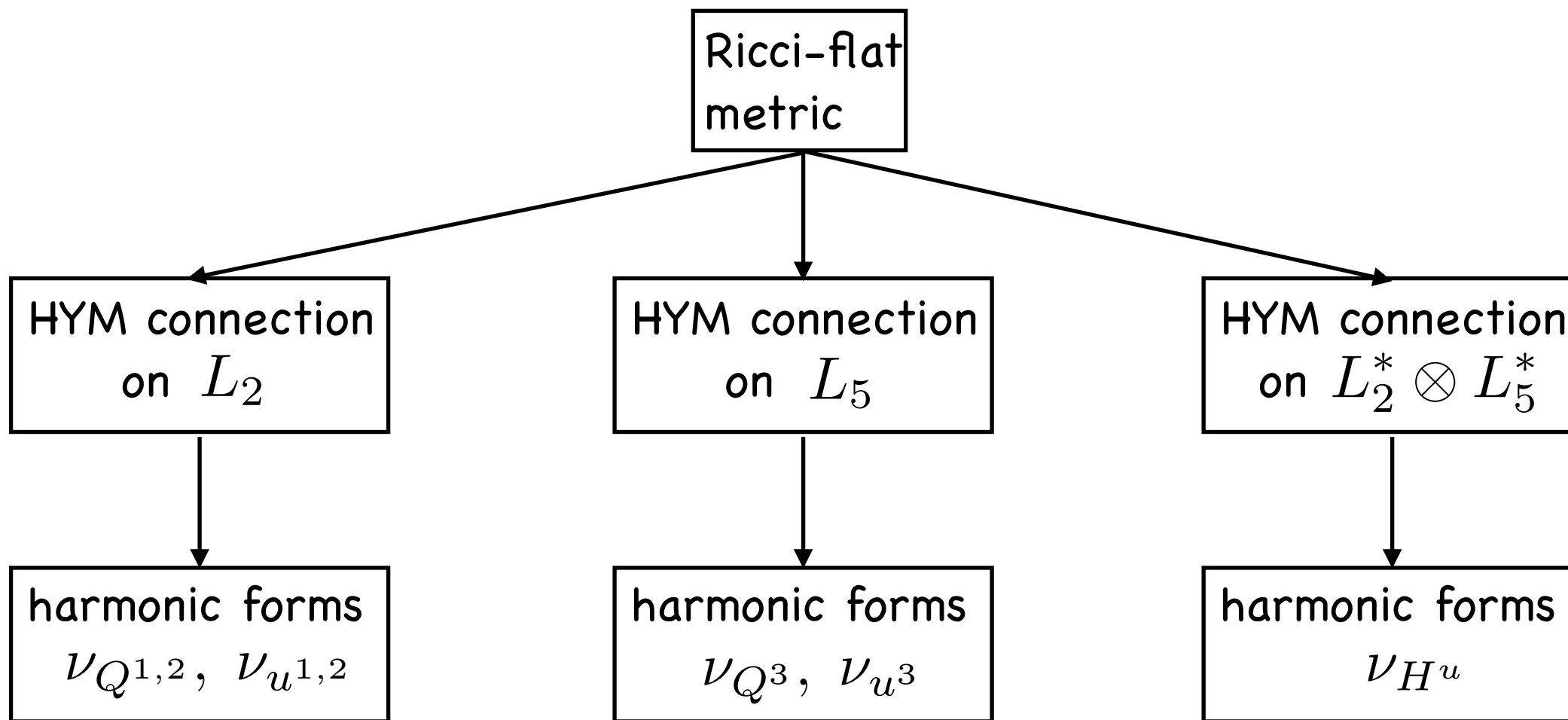
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The calculation, schematically:



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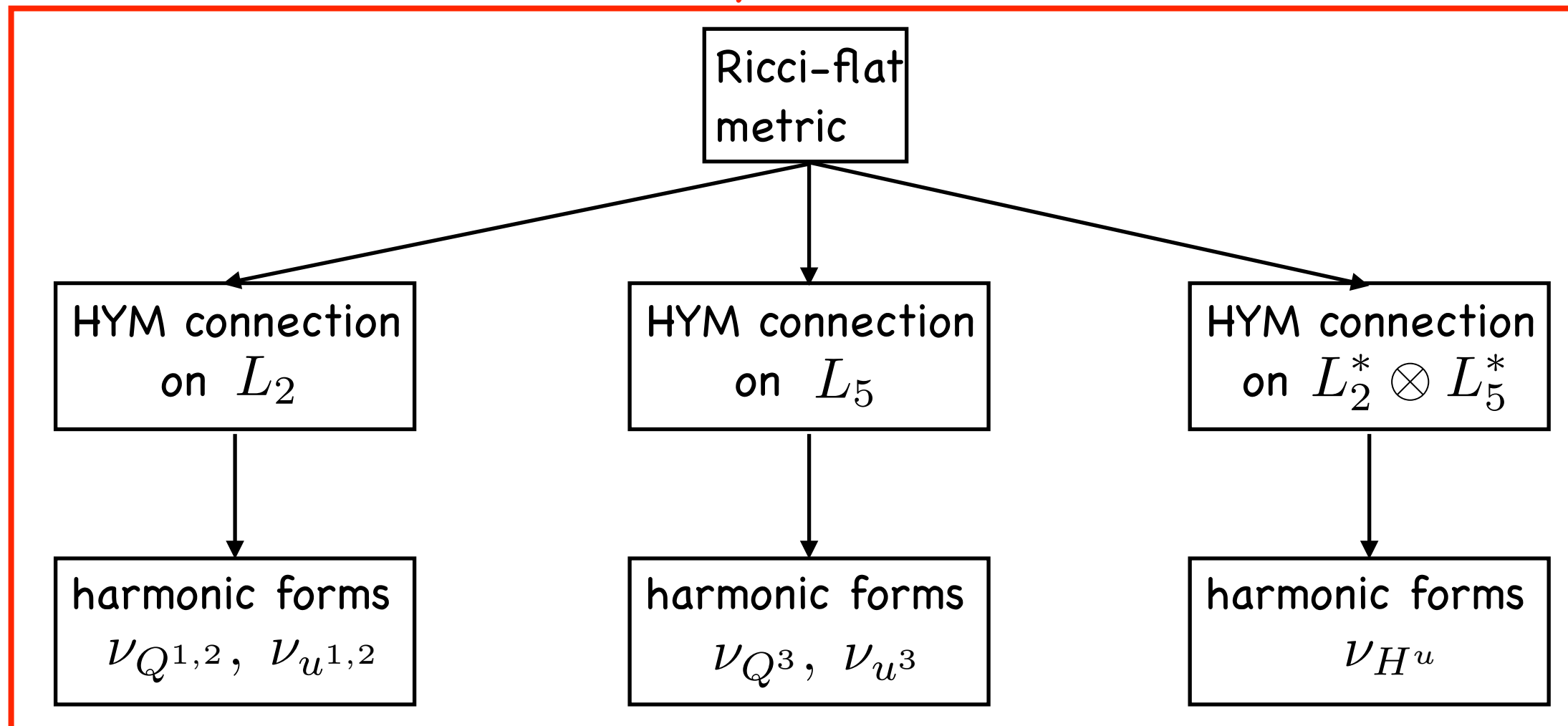
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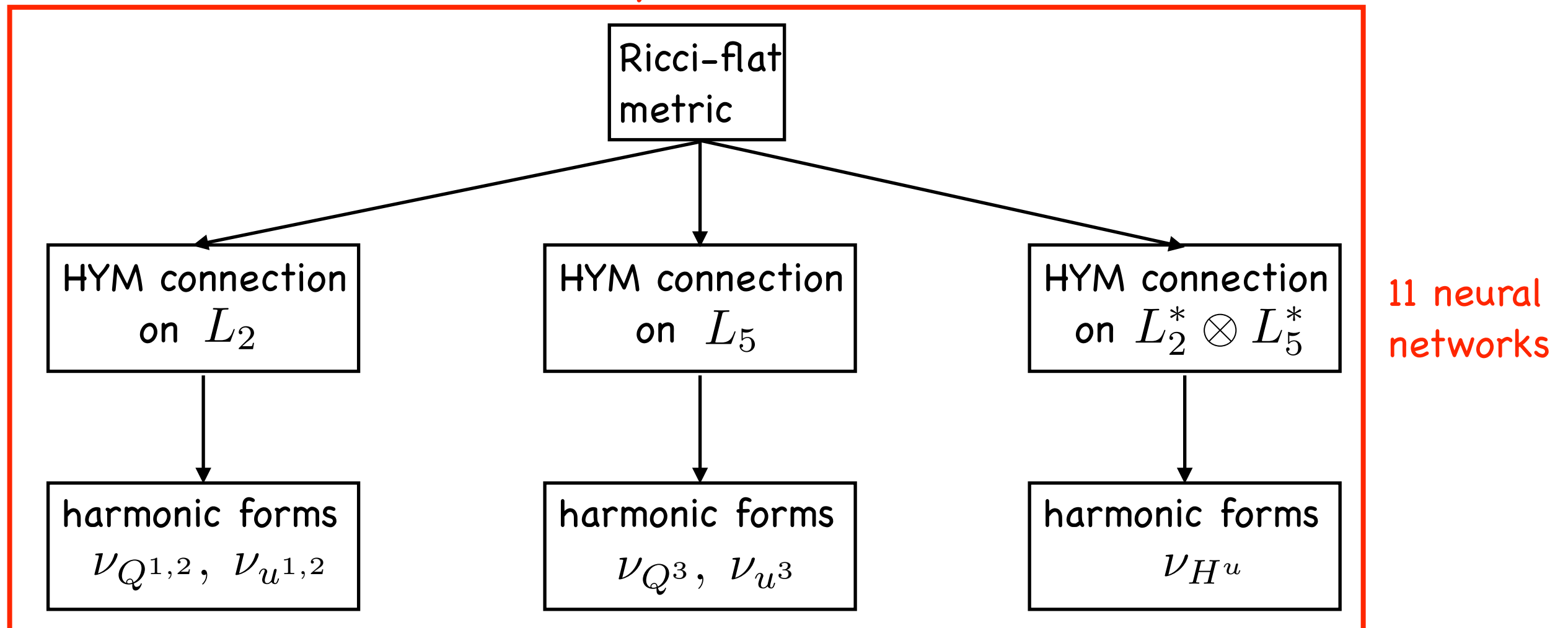
11 neural networks

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The calculation, schematically:



Then use Monte-Carlo integration on X to evaluate

$$Y_{IJK} \sim \int_X \nu_I \wedge \nu_J \wedge \nu_K \wedge \Omega \quad K_{IJ}^{\text{matter}} \sim \int_X \nu_I \wedge (\bar{\star}_U \nu_J)$$

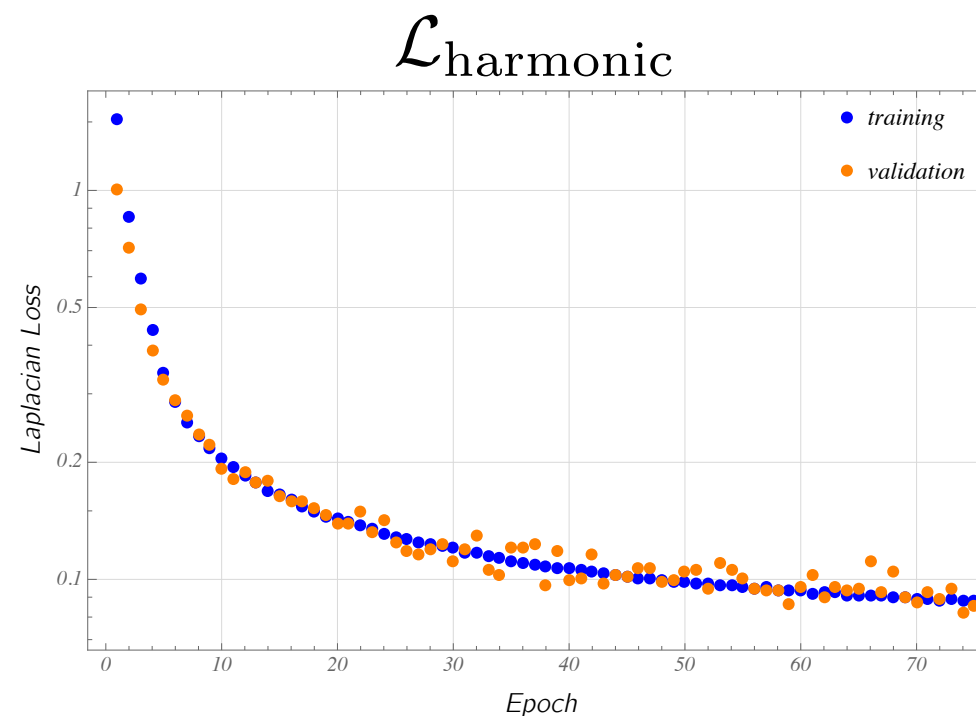
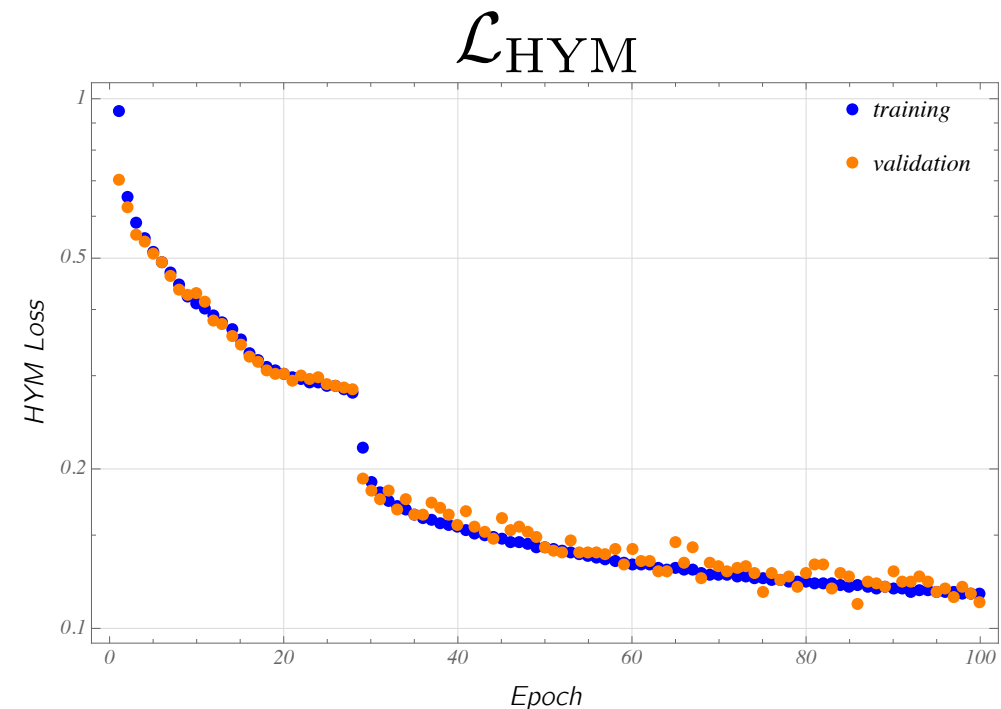
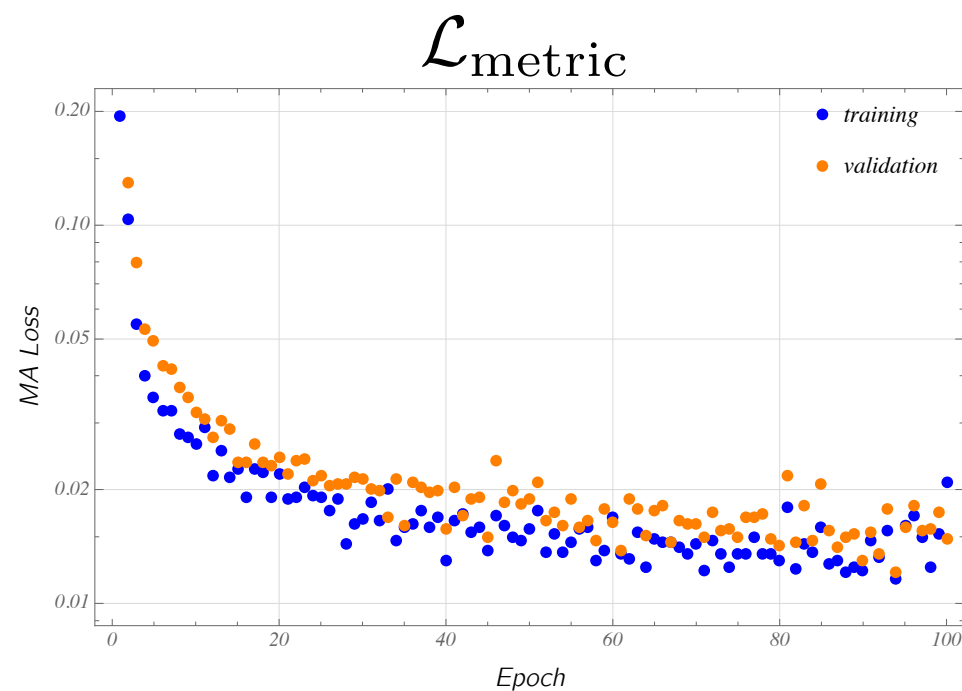
Results

Training

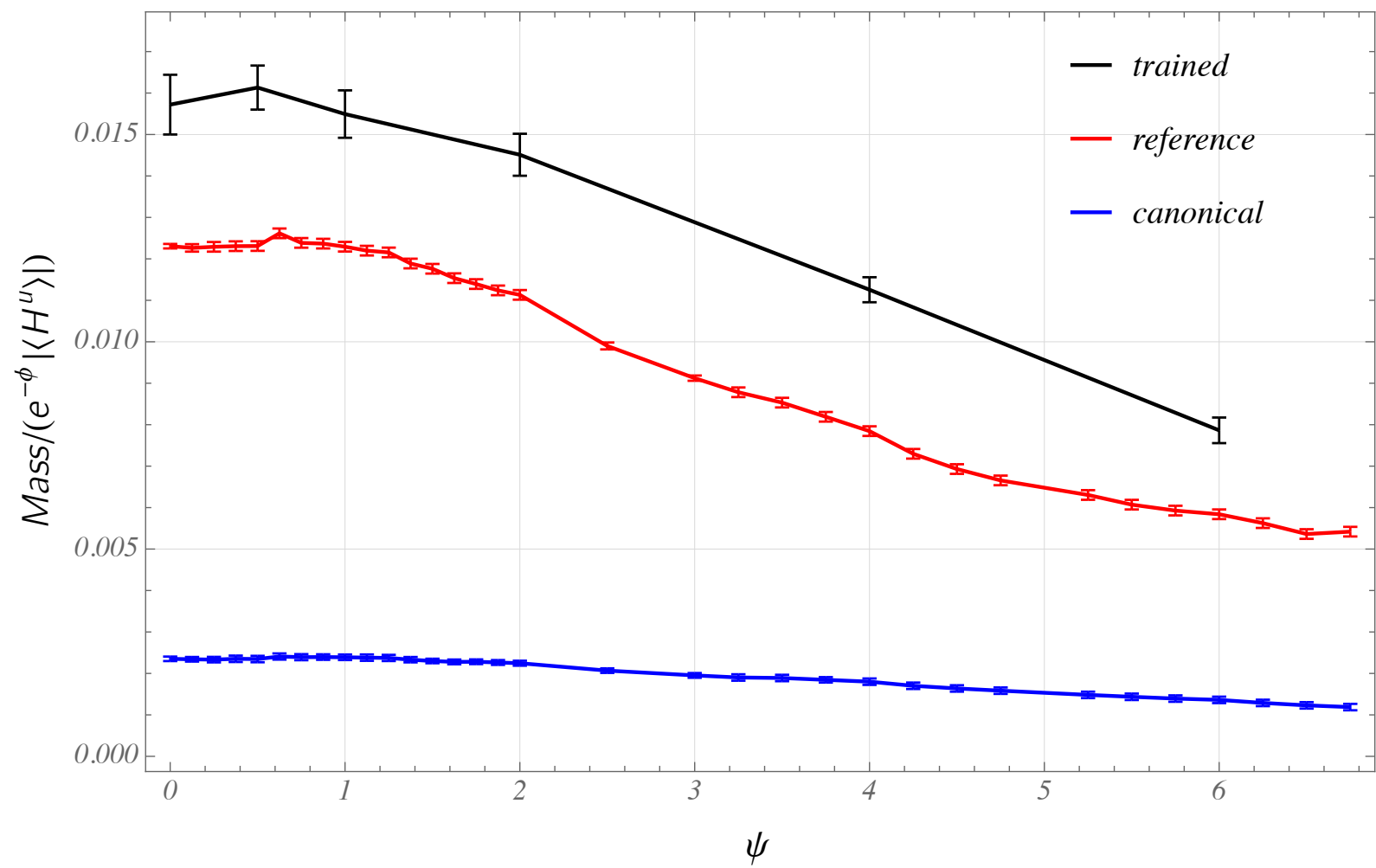
- training set: 300000 points $x_i \in X$ (Fubini-Study distribution)
- NN: fully-connected, width 128, depth 4, GeLU activation, 100 epochs.
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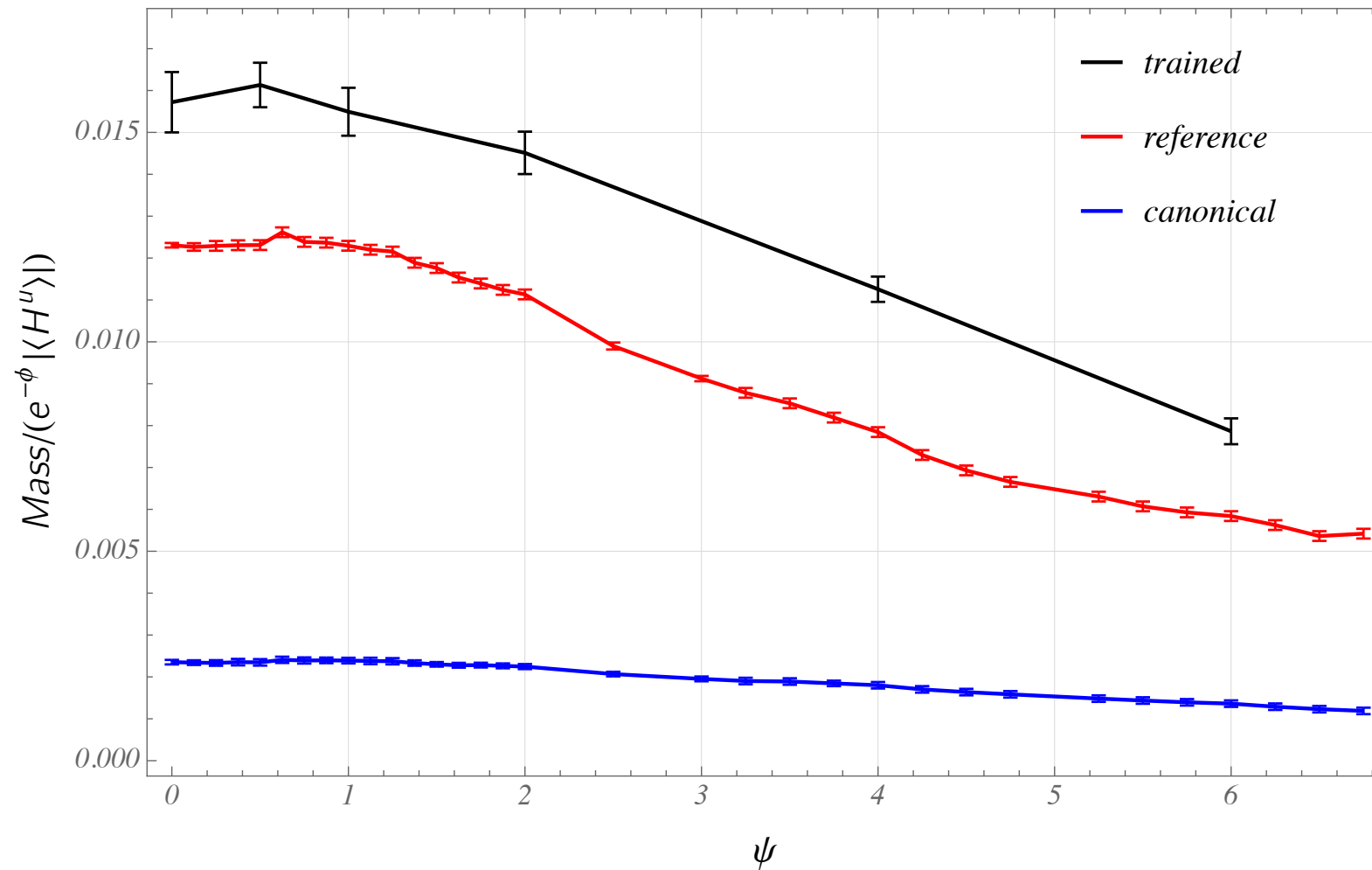
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Numerical Yukawa couplings



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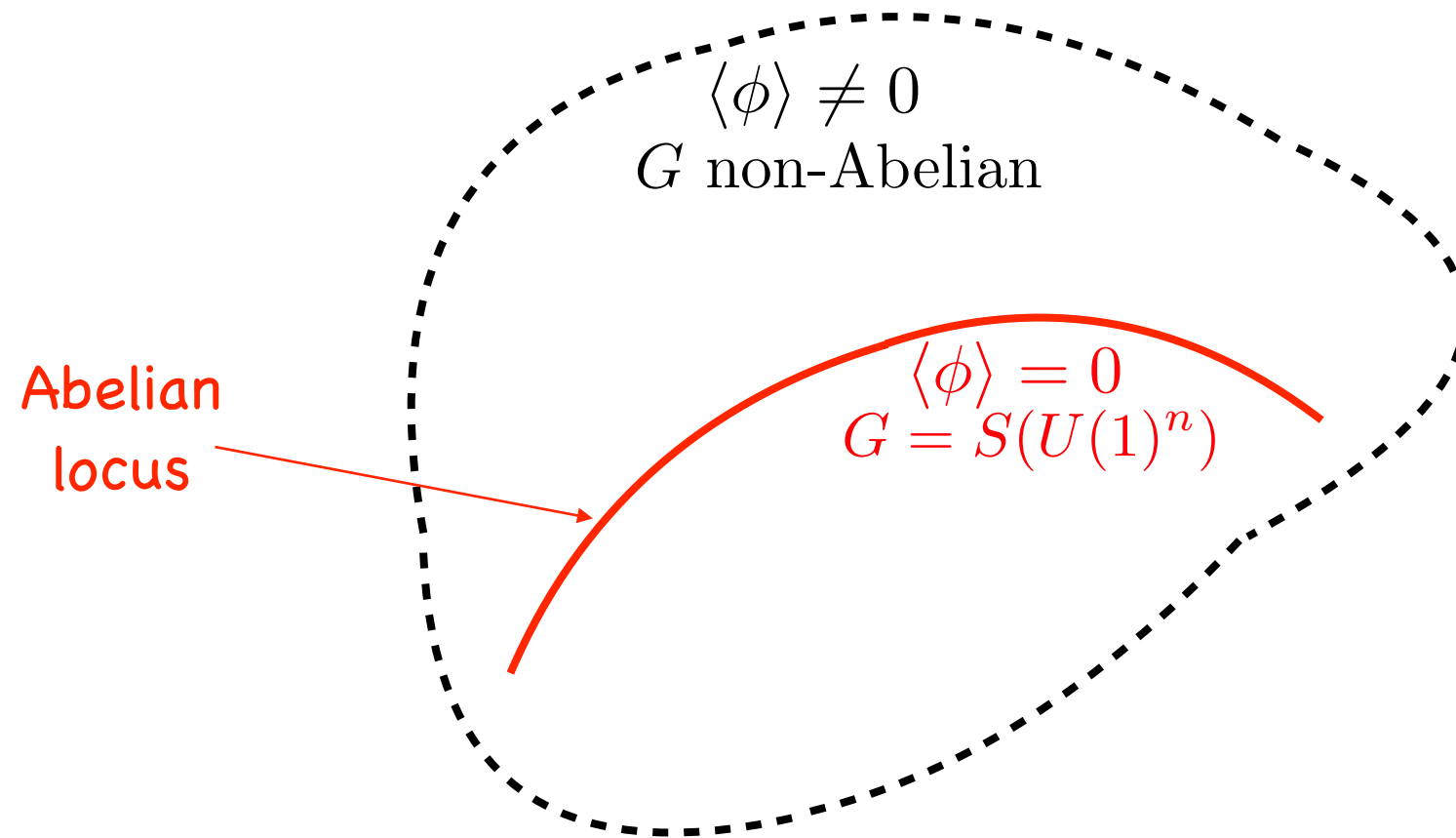
Remarks:

- Additional symmetry means two massive families are degenerate.
- By exploring compl. str. moduli space we can generate a hierarchy.
- Calculation with reference quantities is 'not bad'.
- Assuming canonical normalisation gives result far off.

Heterotic Froggatt Nielsen models

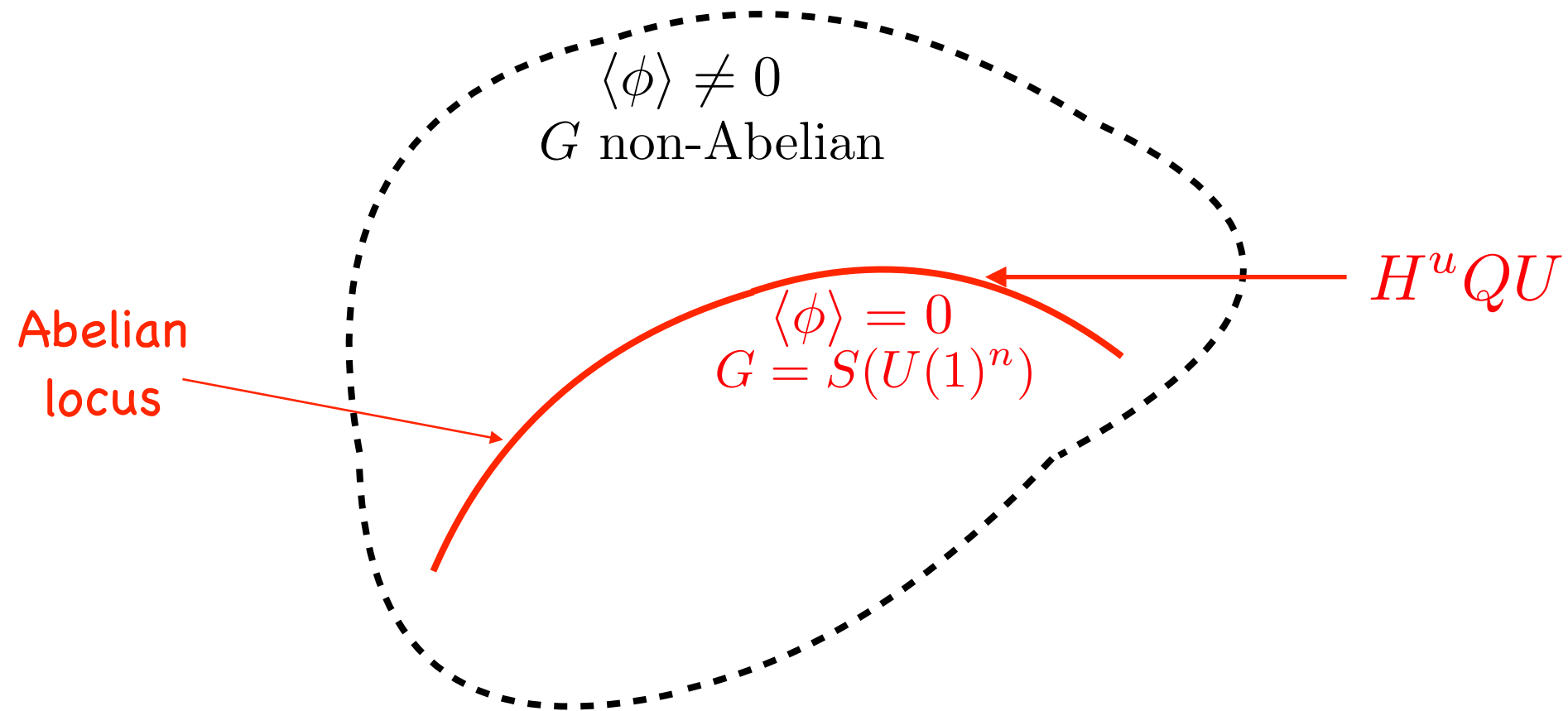
Non-Abelian models from line bundle models

sketch of bundle moduli space $\{\langle\phi\rangle\}$:



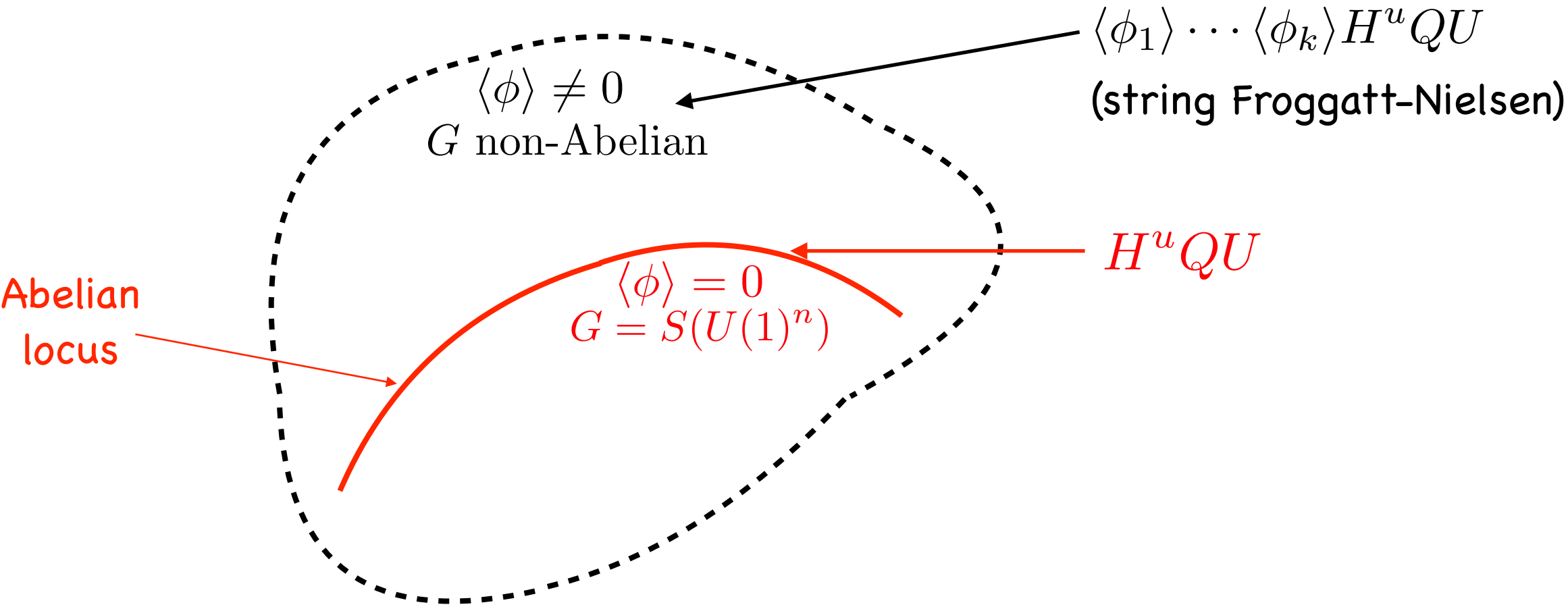
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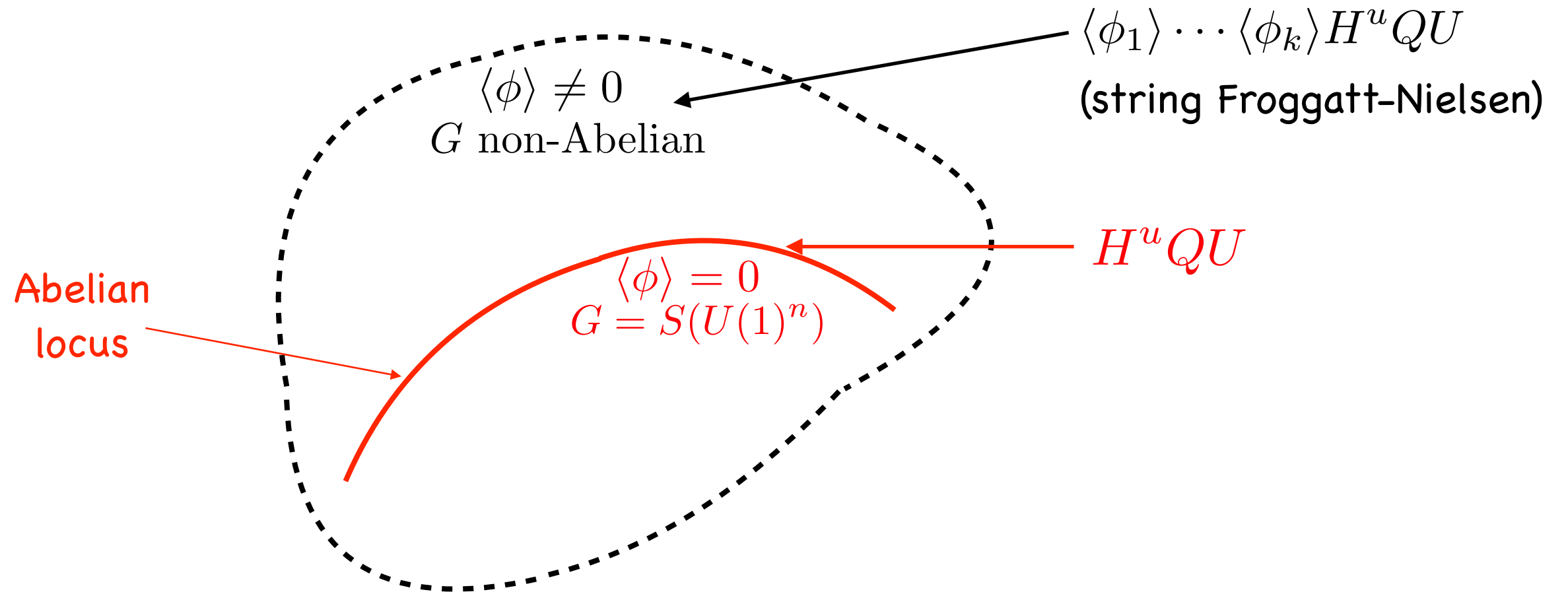
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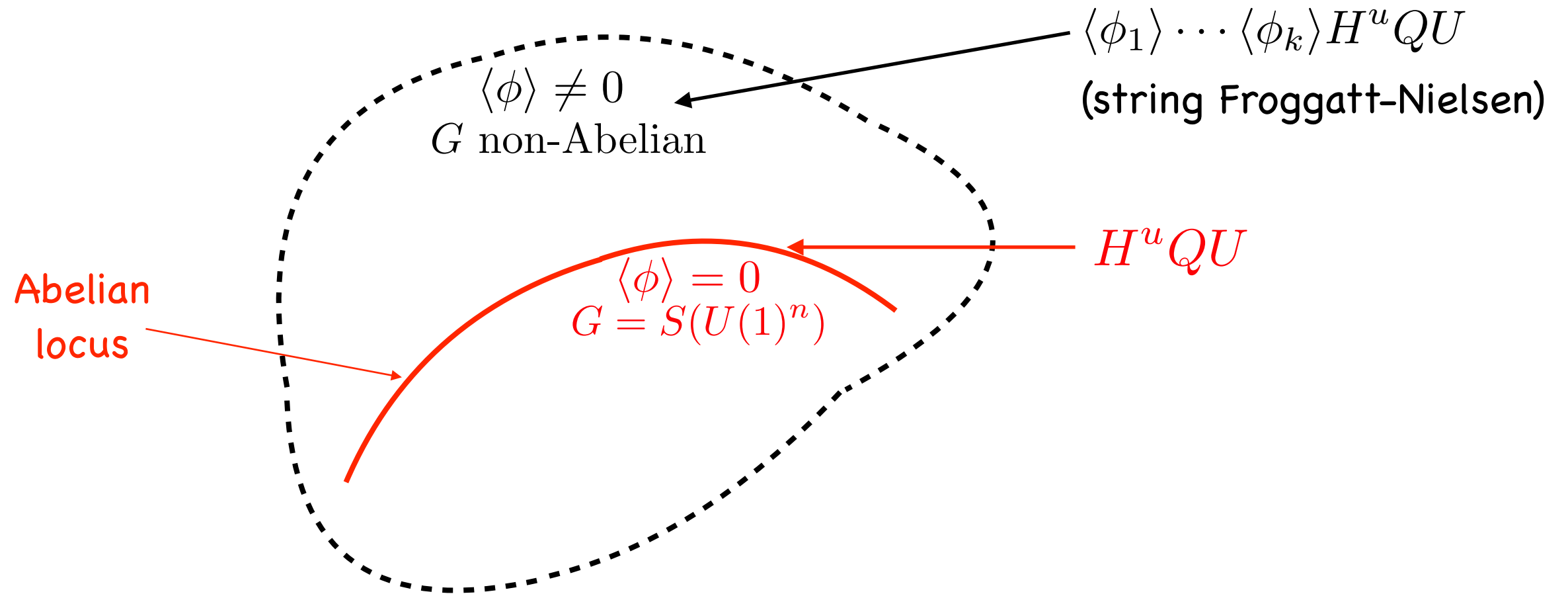
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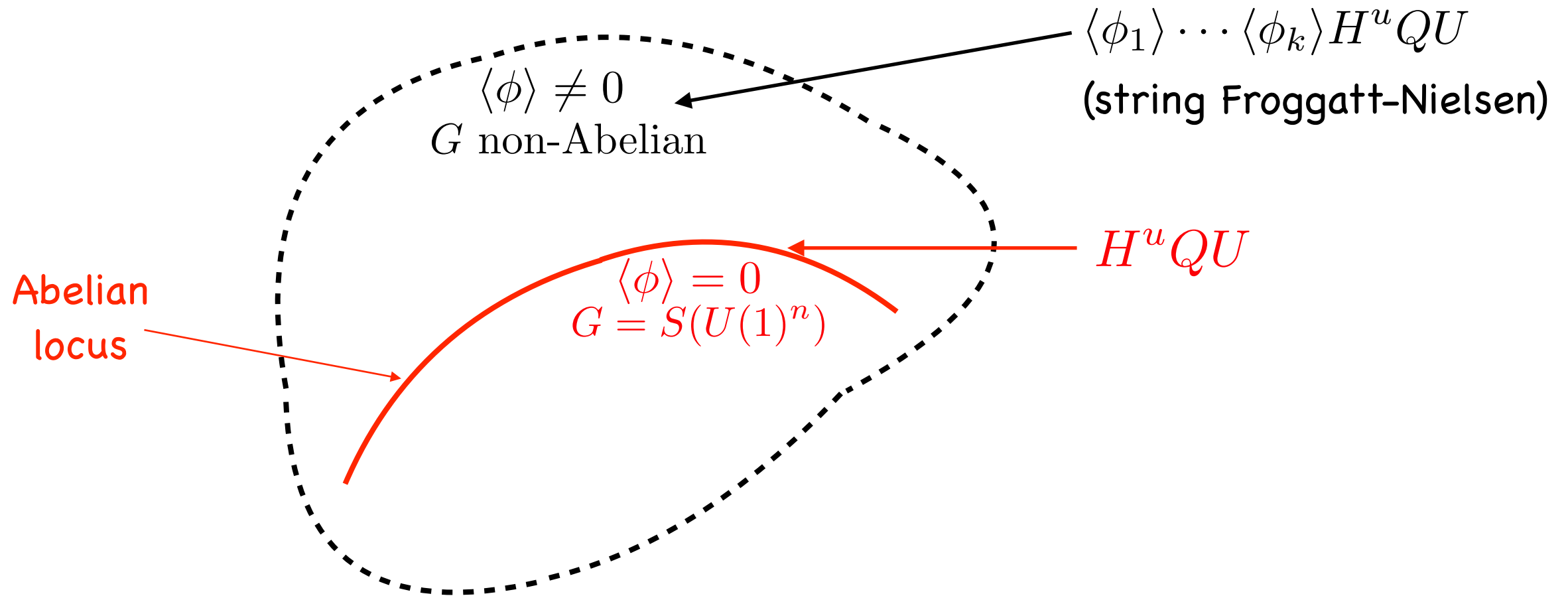


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- Two options:
- 1) Construct non-Abelian bundles and generalise Yukawa calculations to this case.
 - 2) Deform line bundle models by switching on VEVs for bundle moduli ϕ .

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Non-Abelian models from line bundle models

at Abelian locus

complex
structure



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away from Abelian locus

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$S(U(1)^5)$ charges determined by topology of compactification.

For example: $U \leftrightarrow \nu_u \in H^1(X, L_a)$ means $q(U) = e_a$, $q(\Phi^i) = (k_1^i, \dots, k_5^i)$

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e^{-T} points to Φ

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→ Heterotic Froggatt Nielsen models

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This model allows for viable quark masses, charged lepton masses and CKM matrix for suitable VEVs $\langle \phi \rangle$ and $\langle \Phi \rangle$.

Conclusions

- Calculating Yukawa couplings in geometric string models is hard – it requires the Ricci-flat CY metric and related quantities.
- For heterotic line bundle models, we have shown that these quantities and the resulting Yukawa couplings can be calculated numerically, using ML techniques.
- The resulting Yukawa couplings are not yet realistic. This requires moving away from the line bundle locus to non-Abelian bundles.
- We have taken some steps in this direction by studying heterotic Froggatt–Nielsen models.
- We have identified a few viable models of this kind.

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- We need to understand how to perform the numerical computation for non-Abelian bundles.
- We need to consider moduli stabilisation!

Thanks