



UNIVERSITÄT
HEIDELBERG
ZUKUNFT
SEIT 1386

IMPRS
for Precision Tests of
Fundamental Symmetries
INTERNATIONAL MAX PLANCK
RESEARCH SCHOOL



Generative Amplification with Surrogate Monte Carlo

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with Henning Bahl, Tilman Plehn

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Why amplitude surrogates?

Challenge

High-precision amplitudes are computationally expensive



The cost grows rapidly with final state multiplicity and perturbative order



Exact training data are limited

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Challenge

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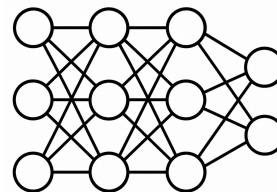
Exact training data are limited

Strategy

Exact amplitudes

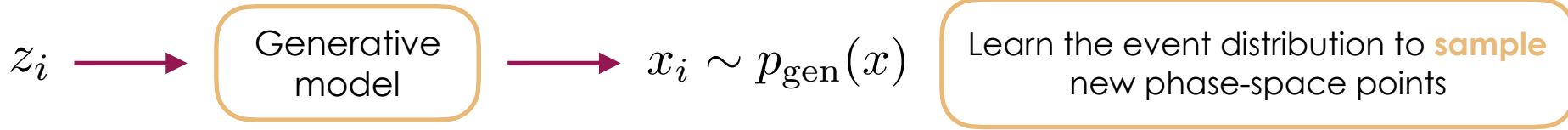


Surrogate

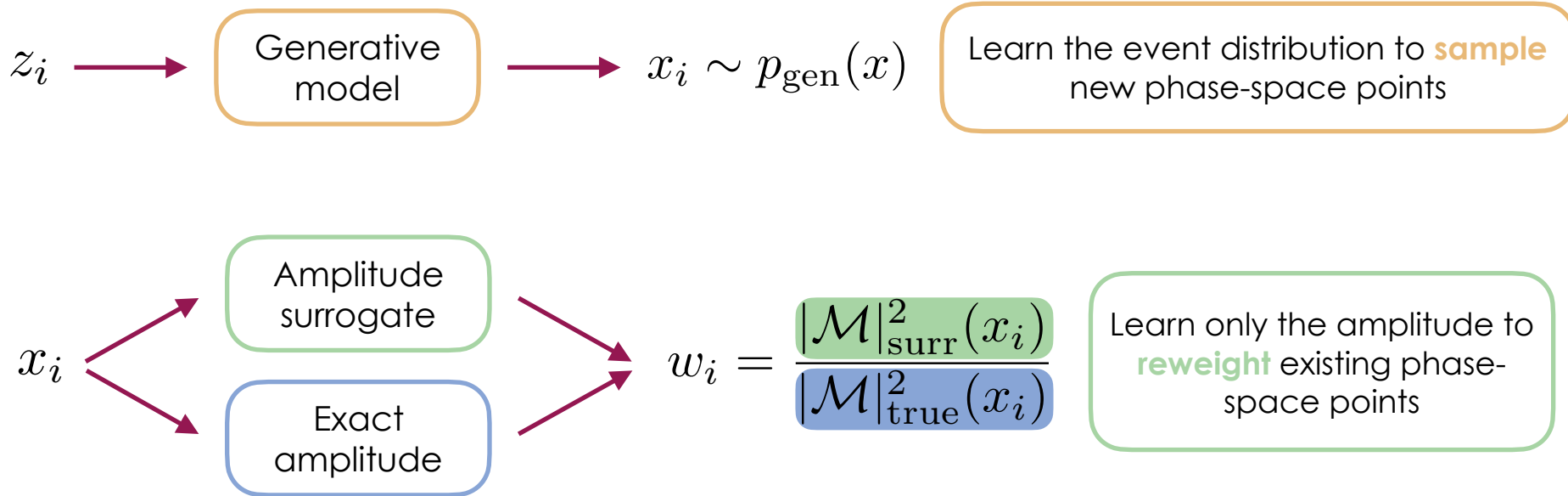


Fast evaluations

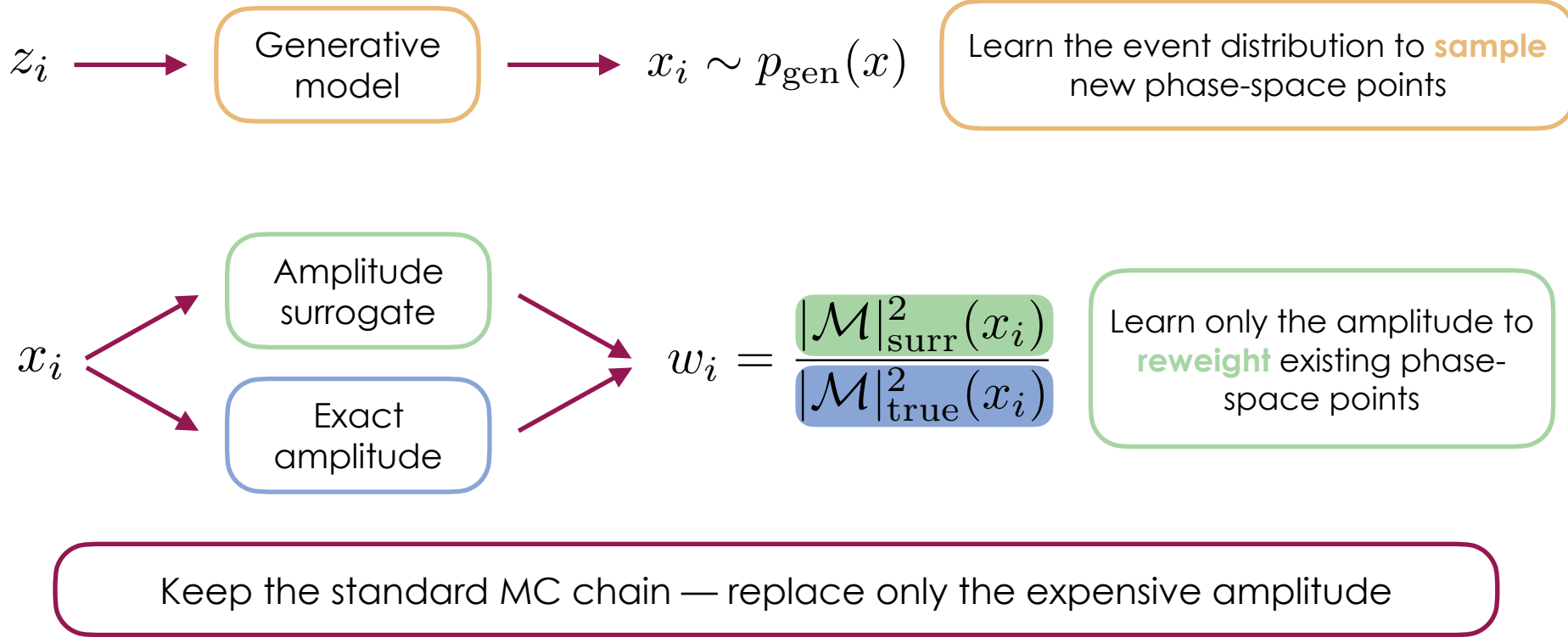
From generative models to Surrogate Monte Carlo



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From generative models to Surrogate Monte Carlo



What are we comparing?

$$q\bar{q} \rightarrow Z + n_g g, \quad n_g = 1, \dots, 4$$

Focus in this talk:

$$q\bar{q} \rightarrow Z + 4 g$$

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Training sample

- Exact amplitude evaluations
- Used to train the surrogate

$$n_{\text{train}} = 10^4, 10^5, 10^6$$

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Surrogate-weighted sample

- Independent truth sample
- Reweighted with the surrogate

$$n_{\text{surr}} = 10^6$$

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- Used as reference

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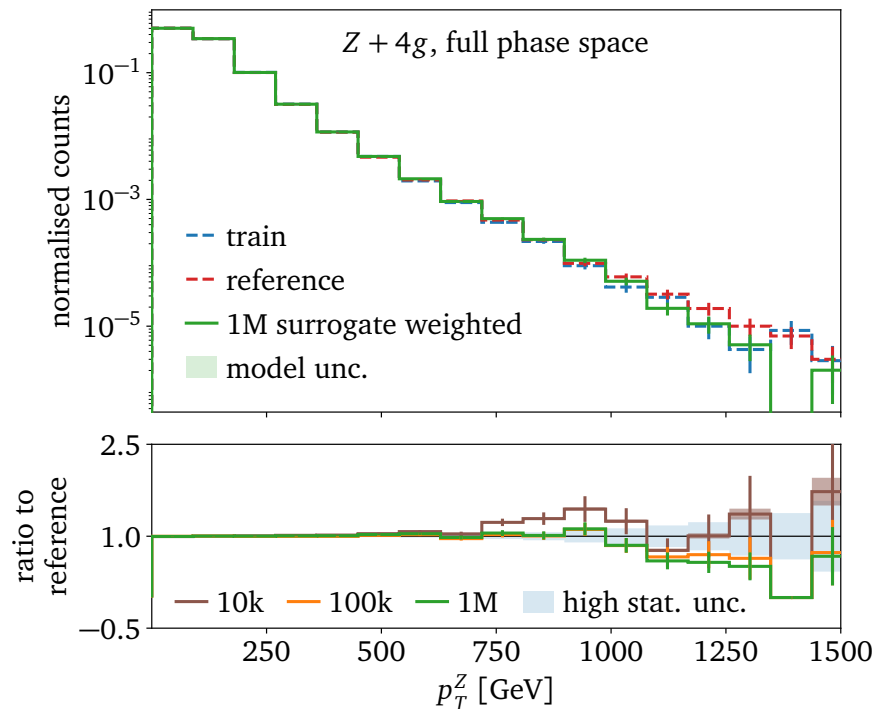
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Statistically independent samples

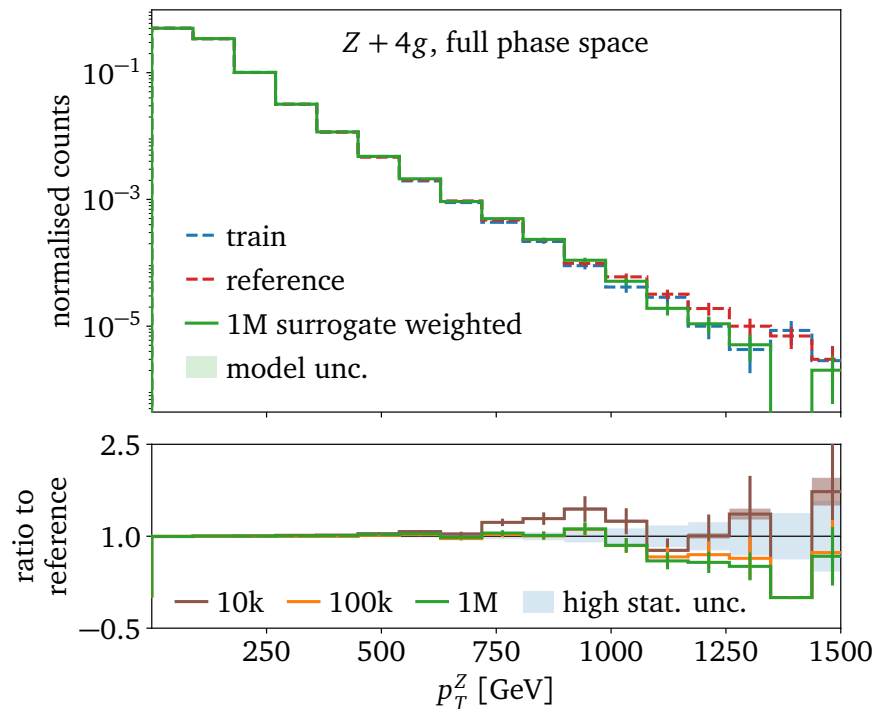
Surrogate accuracy even in the tails

Full phase space

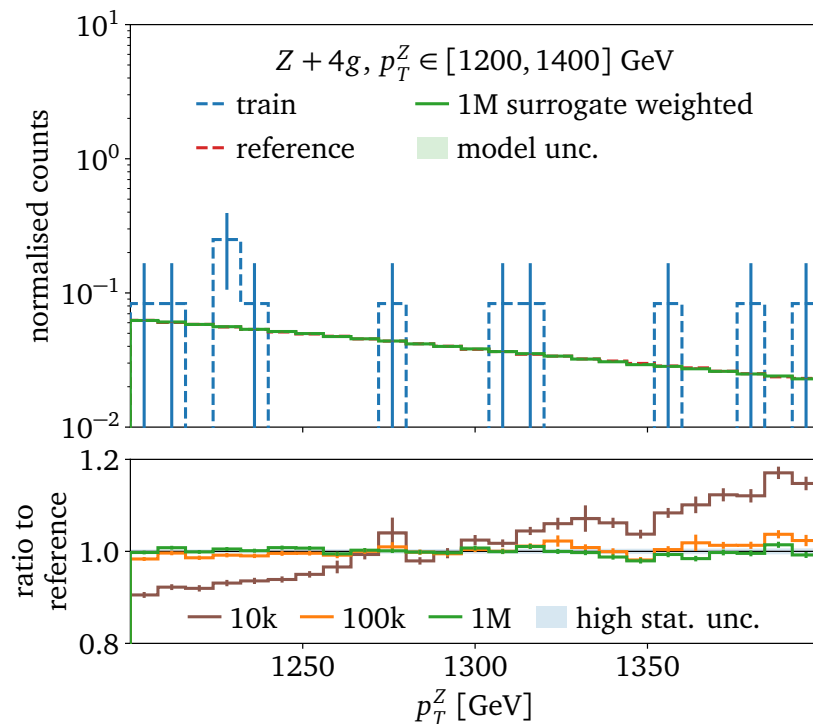


Surrogate accuracy even in the tails

Full phase space



Tail



Averaging amplification

Do we reproduce an **average** observable in V ?

True observable
in a region V

$$I(p_{\text{true}}) = \int_V dx p_{\text{true}}(x)$$

Estimated observable
from surrogate weights

$$\bar{I} = \frac{\sum_{i \in V} w_i}{\sum_i w_i}$$

Statistical
uncertainty

$$\sigma_{\text{stat}}^2(n_{\text{eff}}) \simeq \frac{\bar{I}(1 - \bar{I})}{n_{\text{eff}}}$$

Effective sample size

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Effective sample size

How many truth events are needed to
match the surrogate accuracy?

$$\longrightarrow \sigma_{\text{stat}}(n_{\text{equiv}}) = \sigma_{\text{model}}$$

$$G = \frac{n_{\text{equiv}}}{n_{\text{train}}}$$

$\longrightarrow$ statistical **amplification** over the training sample

Averaging amplification

$$M_I = \left\langle [I(p_{\text{true}}) - \bar{I}(D_{\text{surr}}^{n_{\text{gen}}})]^2 \right\rangle = \sigma_{\text{stat}}^2 + \sigma_{\text{model}}^2$$

- The statistical uncertainty decreases as

$$\sigma_{\text{stat}} \propto 1/\sqrt{n_{\text{eff}}}$$

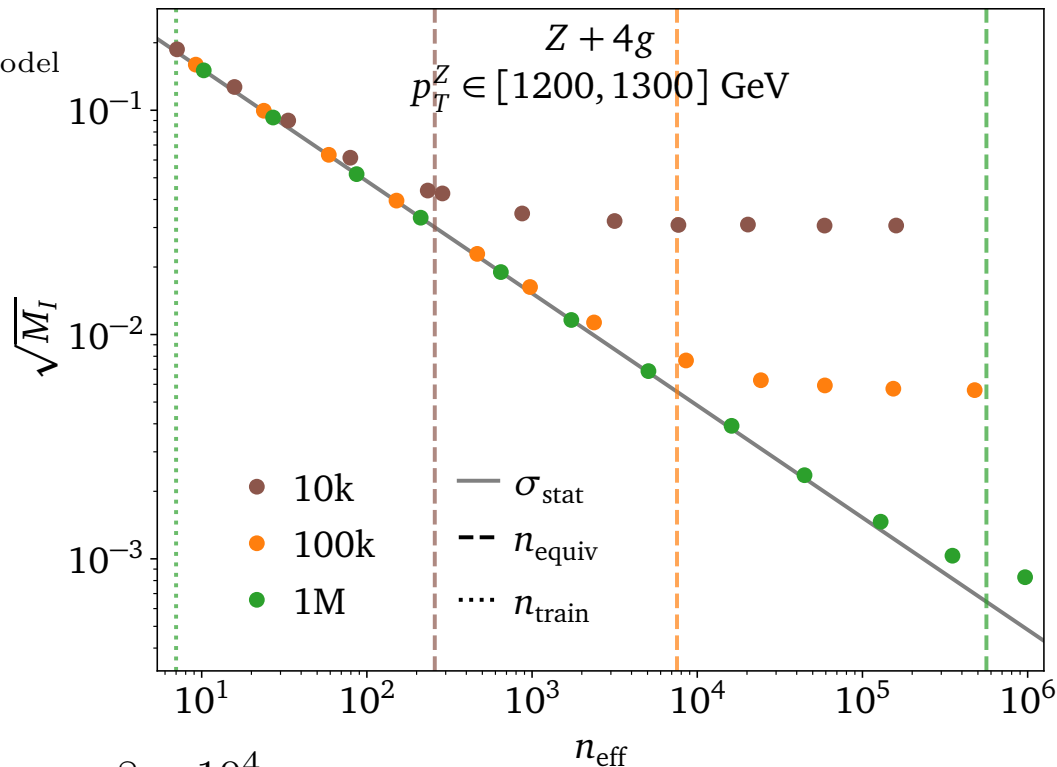
- The surrogate reaches a model-limited plateau

$$\sigma_{\text{stat}} \sim \sigma_{\text{model}}$$

- The intersection defines

$$n_{\text{equiv}}$$

$$n_{\text{train}} = 7 \longrightarrow n_{\text{equiv}} \simeq 5.6 \times 10^5 \longrightarrow G_{\text{avg}} \simeq 8 \times 10^4$$



Differential amplification

Do we reproduce the **distribution** within V ?

Optimal test statistic

$$\begin{aligned} T(x) &= \log \frac{p_{\text{surr}}(x)}{p_{\text{true}}(x)} \\ &= \log w(x) \end{aligned}$$

Cumulative distributions

$$\begin{aligned} F_{\text{surr}}(t) \\ F_{\text{true}}(t) \end{aligned}$$

Kolmogorov-Smirnov
distance

$$M_{\text{KS}} = \sup_t |F_{\text{surr}}(t) - F_{\text{true}}(t)|$$

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Differential amplification

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- Statistical fluctuations decrease as

$$M_{\text{KS}}^{\text{stat}} \propto 1/\sqrt{n_{\text{eff}}}$$

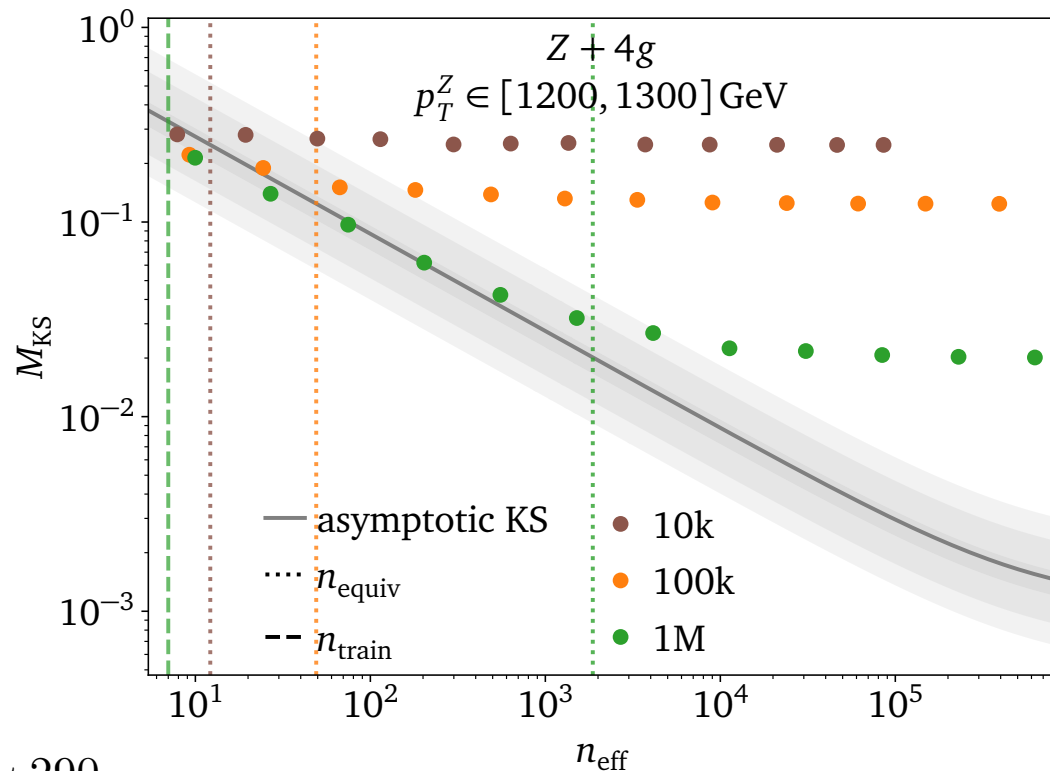
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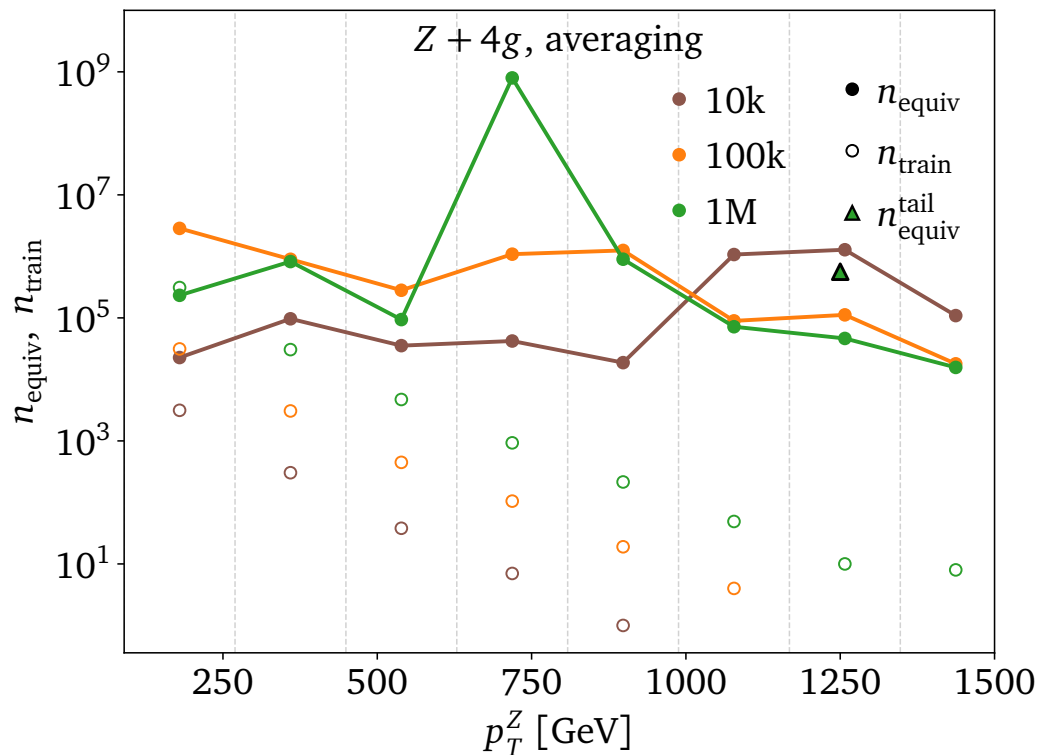
$$n_{\text{equiv}}$$

$$n_{\text{train}} = 7 \longrightarrow n_{\text{equiv}} \simeq 2000 \longrightarrow G_{\text{KS}} \simeq 290$$



Amplification across phase space

- Averaging amplification provides a robust measure
- Large equivalent statistics across the full spectrum
- Training statistics drop in sparsely populated regions
- Amplification increases where training data are scarce



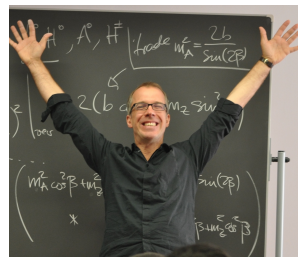
Take-home message

- Amplitude surrogates provide **statistical amplification** beyond their limited training statistics
- **Amplification** is strongest in sparsely populated **tails**
- Surrogate Monte Carlo preserves the **standard simulation** chain while replacing the **expensive amplitude**

Team



Henning

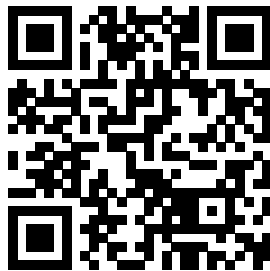


Tilman



Rebecca

Paper



Poster

