

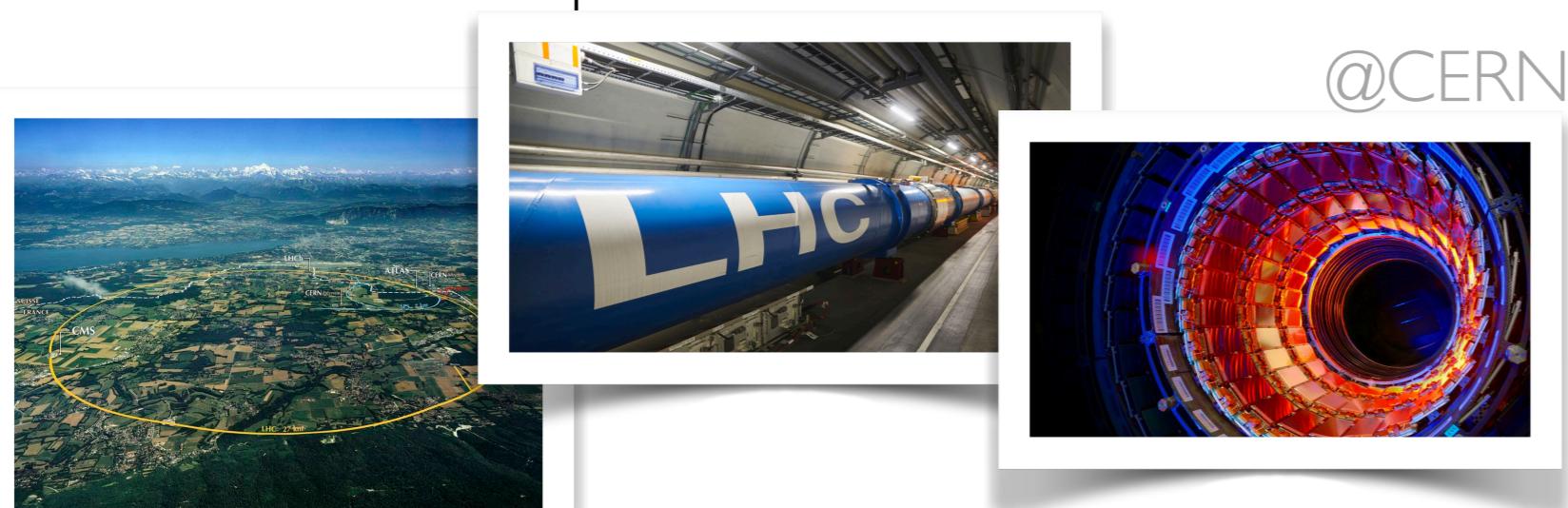
Symbolic Regression for Precision LHC Physics

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Introduction

- In high energy physics (HEP), fundamental interactions can be described by the **Standard Model** (SM) of particle physics to a good extent.
- The SM is tested at the world's most powerful microscope: the **Large Hadron Collider** (LHC). Theoretical predictions and observables are compared.



- Formulas** are the actual language of theoretical predictions.
- The question:** Is there a way to find simple, accurate, and closed analytical formulas from noisy, potentially high dimensional datasets?

Methodology

- We use symbolic regression (SR) with the PySR [1] library.
- Formulas are described as expression trees, as shown in Fig. 1. An evolutionary algorithm implements mutations/crossover between trees, as shown in Fig. 2, and new trees arise.

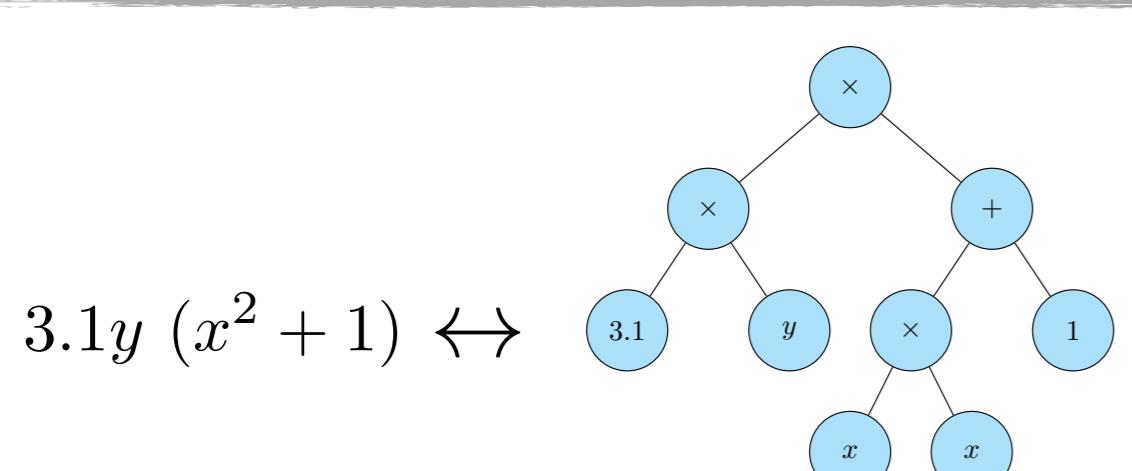


Fig. 1: Expression tree

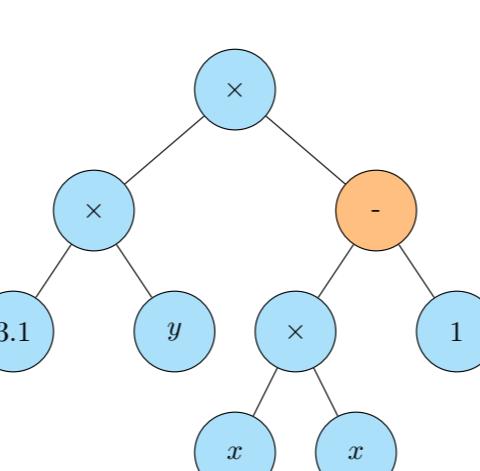
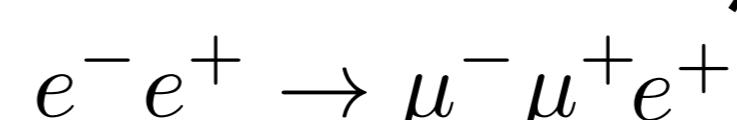


Fig. 2: Mutation

- Each tree has a complexity c , related to its number of nodes.
- Fittest trees optimise a certain metric

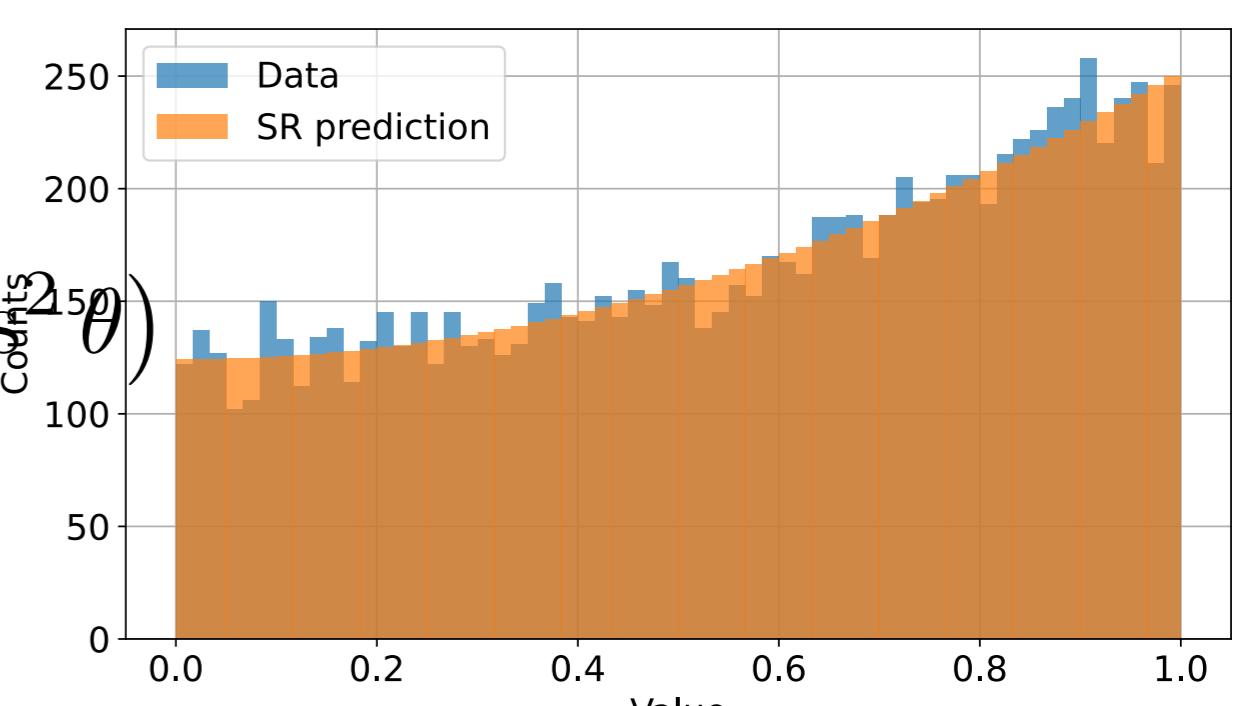
Accuracy:	Score:	Best:
Minimise	Maximise	Highest score with
$L = \frac{1}{n} \sum_{i=1}^n (Y_i - \hat{Y}_i)^2$	$-\frac{\partial \log(L)}{\partial c}$	$L \leq 1.5 \times L_{\min}$

3



Results

Distribution of $|\cos \theta|$

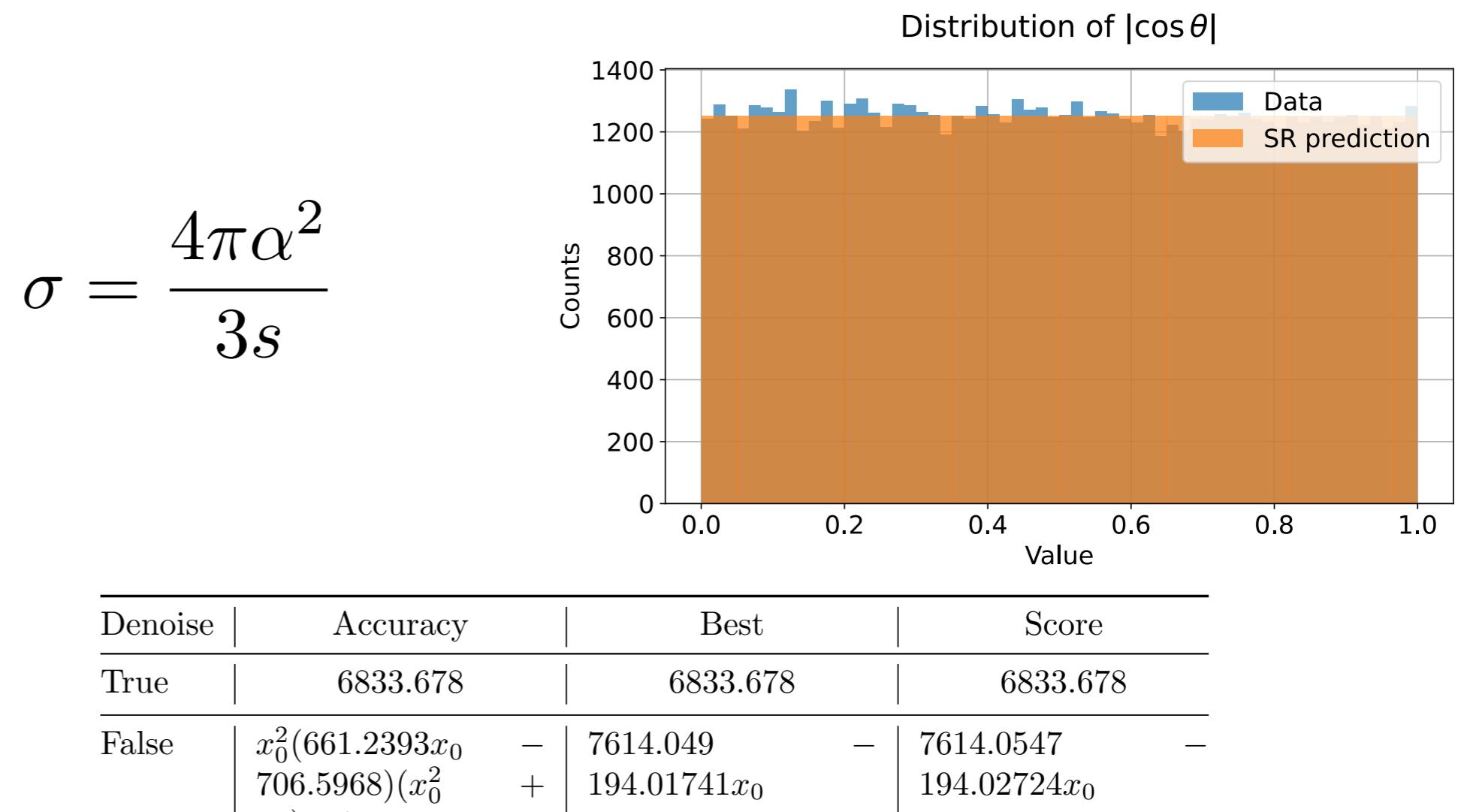


Bins	Accuracy	Best		Score
		q	\bar{l}	
10	$(296.52358194355 \cdot x_0^2 + 7046.0674)$	7250.1396	$x_0^2 + 7589.319$	7250.1396 $\cdot x_0^2 + 7589.319$
30	$7613.42 \cdot x_0^2 (123.43398x_0^4 + 2326.98053420264)$	2417.7627 x_0^2	$+ 2415.3643x_0 + 2125.6453$	Z/γ^*
100	$x_0(207.340216x_0 + 428.81232) + 109.830989048$	726.08685 x_0^2	$+ 725.2477x_0 + 637.3749$	750.30175

$x_0 = \cos \theta$

Fig. 1: SR equations for l^-

Reweighted distribution (constant)



$$\frac{d^5 \sigma}{dp_T d\eta dm d \cos \theta d\phi} = \frac{3}{16\pi} \frac{d^3 \sigma^{U+L}}{dp_T d\eta dm} \left[(1 + \cos^2 \theta) + \sum_{i=0}^7 P_i(\theta, \phi) A_i \right]$$

$$A_i = A_i(p_T, \eta, m)$$

References