

Flow-based generative models for particle calorimeter simulation

— EuCAIFCon, Amsterdam, NL —

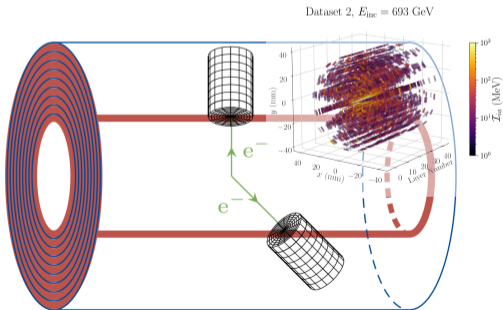
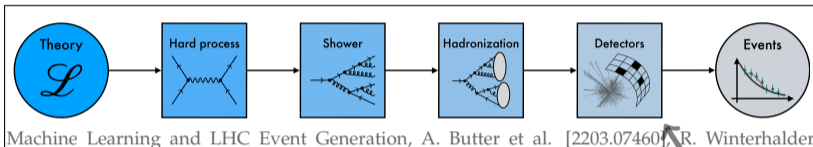
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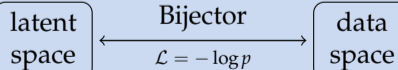


Our computational resources are limited!



We need fast and faithful surrogates.

⇒ Normalizing Flows are both.



Flow-based generative models for particle calorimeter simulation

- 1 Learn how total energy is distributed across layers: $p_1(E_1, E_2, \dots, E_n | E_{inc})$
- 2 Learn normalized shower:
 - ▶ direct: learn $p_2(\hat{\mathcal{I}}_{1:n} | E_{1:n}, E_{inc})$
 - ▶ autoregressive: learn first layer $p_2(\hat{\mathcal{I}}_1 | E_1, E_{inc})$ and step from $(n - 1)$ to n : $p_3(\hat{\mathcal{I}}_n | \hat{\mathcal{I}}_{n-1}, n, E_n, E_{n-1}, E_{inc})$

CaloChallenge datasets

Dataset		Method	generation time per shower [ms] ↓	AUC on voxels ↓
1: γ 368-dim		GEANT4	$\mathcal{O}(10^4)$	0.499(2)
	d	CALOFLOW IAF	0.79 ± 0.01	0.761(2)
	d	CALOINN	0.51 ± 0.03	0.626(4)
1: π^+ 533-dim		GEANT4	$\mathcal{O}(10^4)$	0.609(4)
	d	CALOFLOW IAF	1.00 ± 0.02	0.884(2)
	d	CALOINN	0.44 ± 0.01	0.784(2)
2: e^- 6480-dim	a	GEANT4	$\mathcal{O}(10^5)$	0.500(2)
		iCALOFLOW IAF	13.2 ± 0.5	0.819(4)
	d	CALOINN	1.18 ± 0.03	0.743(2)
3: e^- 40500-dim		GEANT4	$\mathcal{O}(10^5)$	0.498(2)
	a	iCALOFLOW IAF	16.7 ± 0.5	0.891(3)

Classify Showers vs. GEANT4 as metric.