FastSim Parametrization of Beam Dump using Generative Adversarial Network (GAN)

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source-like particles -> computation will be faster by at least 1 order of magnitude. At the tracker last layer

 \star 100% of the neutron come from dump \star >10% of the photons come from dump \star ~1% of the electrons and positrons from the dump

Look at the dummy volume (sampling plane) located just outside of the surface of the dump

• plot
$$(\frac{dN}{dE} \text{ and } \frac{dN}{dt})$$
 or $\frac{d^2N}{dEdt}$, $\frac{d^2N}{drd\theta_p}$, $\frac{d^2N}{d\phi_p d\phi_{pos}}$, ...

• Here r is just
$$\sqrt{x^2 + y^2}$$
, position parameter.

- θ_p, ϕ_p is the polar angle and azimuthal angle ۲ of the momentum \rightarrow representative of direction of the particle.
- $\phi_{\rm pos}$ is position azimuthal angle, and t is time.
- Later: generate from the sampling plane according to these plots
 - ***** Use Generative Adversarial Network.

From Noam Tal Hod, WIS



Collaboration meeting <u>slides</u>



z=4125mm

z=5450.25mm

z=6621.91mm



Generative Adversarial Network (GAN)

Generative Adversarial Network architecture Steps to achieve our goal:

- Step 1: Quantile transformation: This is done to "Gaussianize" the Geant4 dump distributions.
 - Each variable distribution has its own quantile transformation.
 - This is to smoothen sharp features in distributions, making it easier for the network to learn.



Transformed r

Generative Adversarial Network architecture Steps to achieve our goal:

- Step 2: <u>Generator</u>: tries to generate meaningful distributions from random noise.
 - Consists of 4 hidden layers, LeakyReLU as the activation function.
 - Batch Normalization in each layer.
- Step 3: Discriminator: given the original distribution, it tells us how good the generation in Step 2 was.
 - Consists of 4 hidden layers, LeakyReLU as the activation function.
 - Instance Normalization in each layer.
- Step 4: the Loss function: $\mathscr{L}(p_r, p(z)) = \max_{w \in W, \theta \in \Theta} \left[\mathbb{E}_{x \sim p_r}[f_w(x)] - \mathbb{E}_{x \sim p(z)}[f_w(g_\theta(z))] \right] \longrightarrow \text{Wasserstein distance}$

original distribution • Step 5: Inverse quantile transformation to get physical distribution.

Expectation value from the generated distribution

Generating background using WGAN

Generative Adversarial Network (GAN) with Wasserstein loss function

Geant4

WGAN

neutron at sampling surface

Plots prepared by Alon Levi

All variables correlation plots: From Geant4

Neutron distributions

All variables correlation plots: From GAN distributions

Neutron distributions

Comparison with original (Geant4) distributions

Plot labels:

- In this talk there will be two types of plots compared:
 - 1. FullSim Distributions from full Geant4 processing of the dump
 - 2. FastSim Geant4 processing where dump is replaced by particles following GAN neural network.
 - (i) Plots made at test surfaces.

Neutron at test surface 1: r at very low value (Looking at the shape of the distribution)

The distance of scintillator screen from the beam pipe is ~76 mm.

FastSim was generated from 2D distributions using ROOT::TH2D::GetR andom2()

FastSim r matches with FullSim after the application of the scale factor

With scale factor on the sampling plot, FastSim has an ad-hoc weight to match the FullSim

Shown at the July 10 SAS meeting

FastSim generated by GAN, no ad-hoc weight applied

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Neutron at test surface 2: r at very low value (Looking at the shape of the distribution)

The distance of tracker inner layer from the beam pipe is ~52 mm.

FastSim was generated from 2D distributions using ROOT::TH2D::GetR andom2()

FastSim r matches with FullSim after the application of the scale factor

With scale factor on the sampling plot, FastSim has an adhoc weight to match the FullSim

Shown at the July 10 SAS meeting

FastSim generated by GAN, no ad-hoc weight applied

Other distributions from FastSim (GAN)

A few 1D distributions between FastSim (GAN) and FullSim: neutron at the test surface 1

 \bigstar Distributions are looked at **z=5450.25mm**. ★FullSim and FastSim distributions are matching quite well.

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A few 1D distributions between FastSim (GAN) and FullSim: neutron at the test surface 2

 \bigstar Distributions are looked at **z=4125mm**. ★FullSim and FastSim distributions are matching quite well.

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Summary and Outlook

Summary and next steps:

- Mis-modeling in very backward particles
 - The problem is being solved by using Generative Adversarial Network (GAN).
- Neutrons:
 - surface 1 and test surface 2.
- Looking into GANs
 - More natural to parameterize the correlations.
 - correlation at a time.
 - FullSim distribution.
- Worked only with neutron, but photon distributions will be looked at next.

GAN FastSim sample gives similar distribution shapes of neutrons as FullSim in test

• Consider all variable correlations unlike 2D plots which take only 2 variable

• The result is not perfect (yet), but trends going in the right direction in mimicking the

Thank you!

Baseline distribution plots for LUXE geometry at the sampling surface (z=6621.91mm)

 \star Plots used for sampling.

Baseline distributions from FullSim in LUXE: neutron at sampling surface

dump_plane_bkg_track_phi_pos_phi_neutron_cut

dump_plane_bkg_track_rDn_track_theta_neutron_cut

 \bigstar For E vs t plot of neutron, we only go up to 100 eV of neutron.

 \bigstar Neutron less energetic than that are not interesting.

Scale factor:

- Neutron and photon need different scale factor.
- Played with many different values and checked the FastSim distributions.
- Settled with scale factor depending only on θ_p :

Neutron scale factor

Photon scale factor

