



Machine Learning • PDFs • QCD



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# Generative Adversarial Neural Networks in Parton Distribution Functions (PDFs)

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UNIVERSITÀ DEGLI STUDI DI MILANO

DIPARTIMENTO DI FISICA



Istituto Nazionale di Fisica Nucleare

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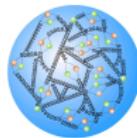
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# What are PDFs

## Definition of Parton Distribution Functions (PDFs):

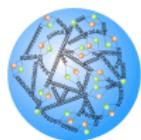
The **probability** of finding a constituent of the proton, commonly known as **parton** (gluons, quarks, and anti-quarks), with a momentum fraction  $x$  and momentum transfer  $Q^2$ .



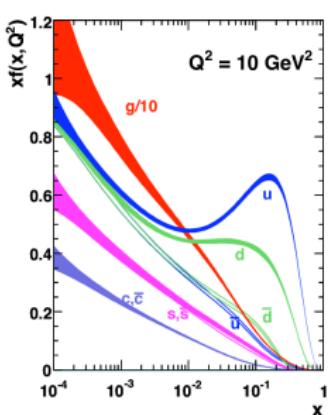
# What are PDFs

## Definition of Parton Distribution Functions (PDFs):

The **probability** of finding a constituent of the proton, commonly known as **parton** (gluons, quarks, and anti-quarks), with a momentum fraction  $x$  and momentum transfer  $Q^2$ .



## PDFs Snapshot



$g(x, Q^2)$ : Probability of finding a gluon, carrying a fraction  $x$  of the proton momentum when probed with energy  $Q^2$ .

$x$ : Fraction of the proton momentum carried out by the gluon.

$Q^2$ : Energy of the hard-scattering reaction (inverse of the resolution length)

# What do we need PDFs for?

PDFs are crucial for Theoretical Predictions!

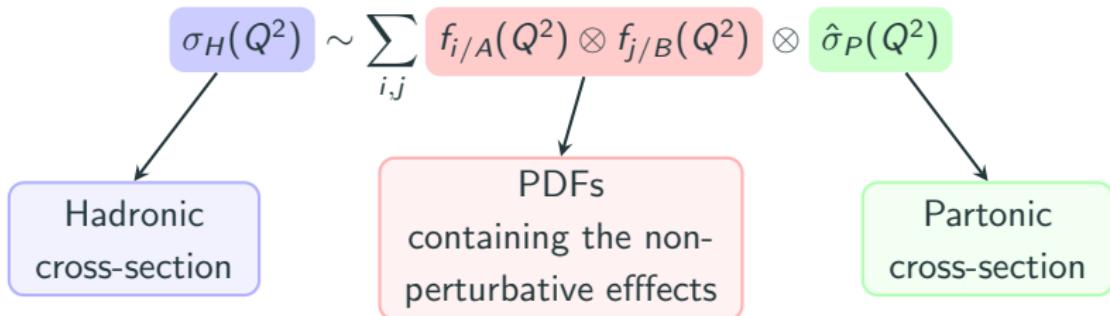
The **Factorization Theorem** in QCD is a main guiding principle that studies the **interplay** between the **perturbative** and **non-perturbative** regime in order to make meaningful theoretical predictions.

$$\sigma_H(Q^2) \sim \sum_{i,j} f_{i/A}(Q^2) \otimes f_{j/B}(Q^2) \otimes \hat{\sigma}_P(Q^2)$$

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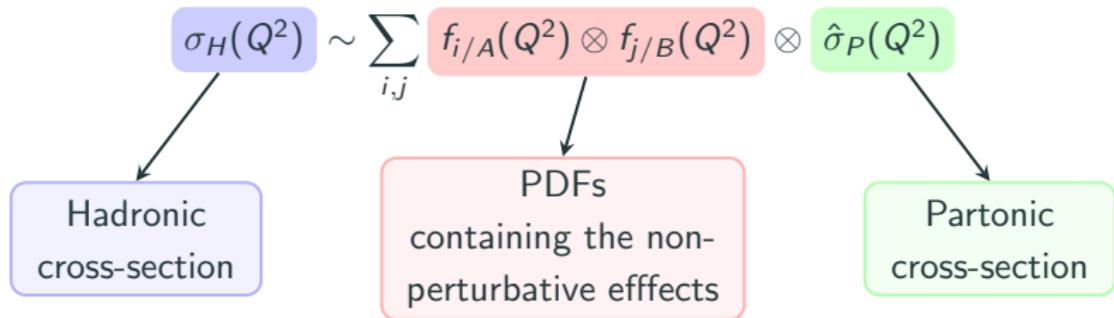
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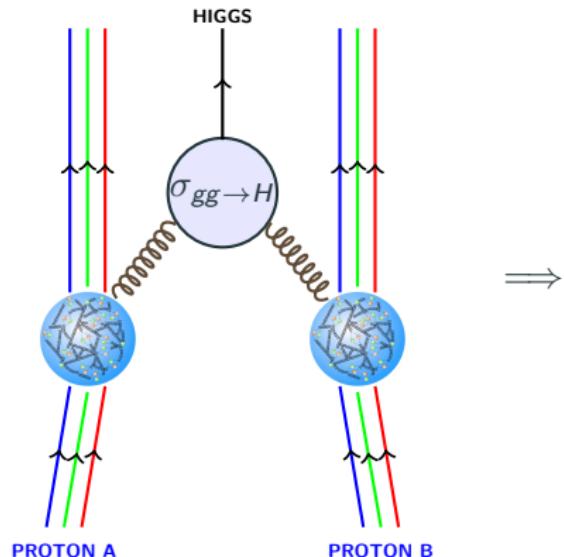


Warning

PDF **cannot** be computed from 1st principle due to **confinement**. It has to be extracted from experimental data using a global QCD analysis.

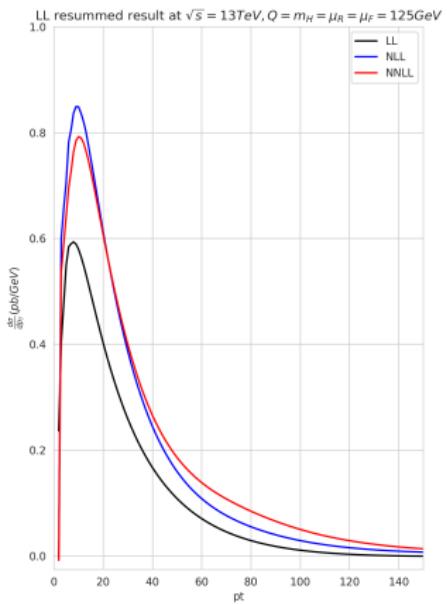
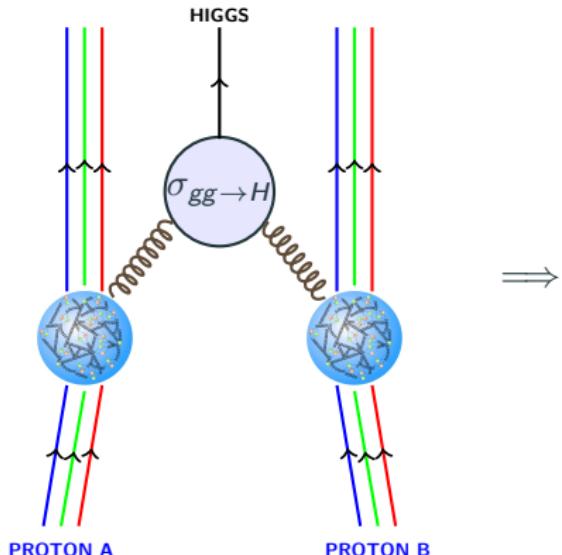
# What do we need PDFs for?

Example: Higgs Production from gluon-gluon fusion



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Example: Higgs Production from gluon-gluon fusion



# PDFs and New Physics

Can new physics hide inside a proton?

$\sigma$	$\delta(\text{PDF})$	$\delta(\alpha_s)$	$\delta(\text{scale})$	$\delta(\text{trunc})$	$\delta(\text{PDF-TH})$	$\delta(\text{EW})$	$\delta(tbc)$	$\delta(1/m_t)$	
48.58	$\pm 0.90$	$^{+1.27}_{-1.25}$	$^{+0.10}_{-1.15}$	$\pm 0.18$	$\pm 0.56$	$\pm 0.49$	$\pm 0.40$	$\pm 0.49$	pb
	$\pm 1.86$	$^{+2.61}_{-2.58}$	$^{+0.21}_{-2.37}$	$\pm 0.37$	$\pm 1.16$	$\pm 1$	$\pm 0.83$	$\pm 1$	%

$$\sigma = 48.58 \text{ pb} {}^{+2.22 \text{ pb} (+4.56\%)}_{-3.27 \text{ pb} (-6.72\%)} (\text{theory}) \pm 1.56 \text{ pb} (3.20\%) (\text{PDF} + \alpha_s)$$

REDUCE PDF **UNCERTAINTIES**  $\Rightarrow$  **PRECISION PHYSICS**

**Solutions:**

- Adding more data: for example, top quark pairs production constrains the gluon PDF
- Improving the methodology: robust and flexible framework, future-proof for future data

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# NNPDF status & Achievements

## PDF Fitter Groups:

- Functional form approach: uses specific polynomials to fit experimental data (**How to define the polynomials?**)

CTEQ

MMHT

MSTW

HERAPDF

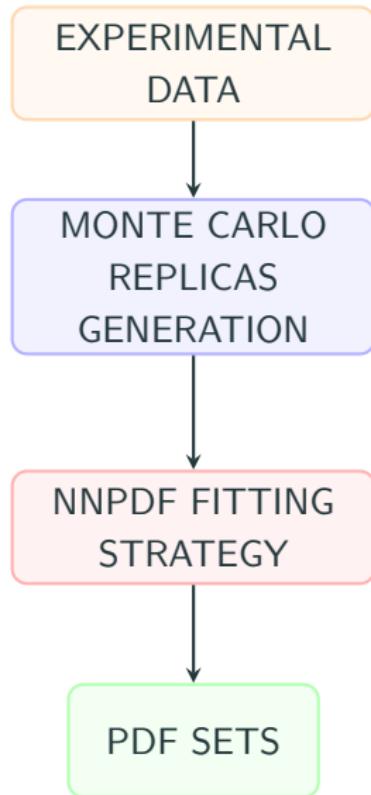
- Neural Network approach: uses techniques from machine learning and Artificial Intelligence to reduce all sources of theoretical biases

NNPDF

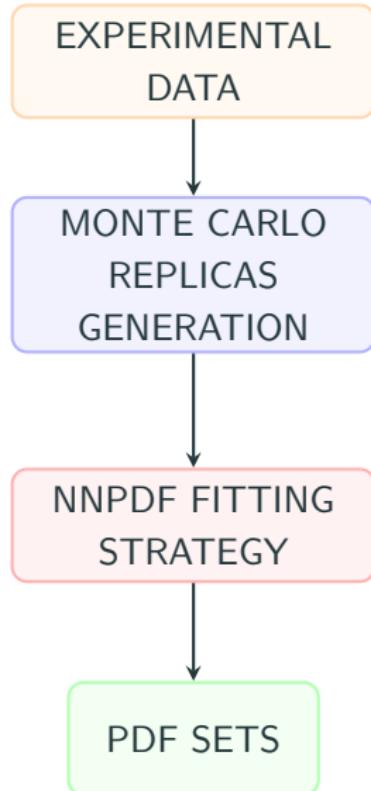
## NNPDF's Features:

- Robust and reliable methodology (under upgrade)
- Only fitter group to include Theory Uncertainties in the PDF  $\square_1 \square_2$

# NNPDF Workflow



# NNPDF Workflow



The total number of replicas is chosen in such a way that it is **large enough** to produce the statistical properties of the original data to the desired accuracy. Generating Replicas is time consuming.  
↔ 1 Replicas takes about 40 Hours

Fits each generated replicas with a PDF parametrized by an Artificial Neural Network. The training is controlled by a Genetic Algorithm (GA).

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# Overview of GANs

## Definition

GANs are deep neural network architectures comprised of two networks **competing** against each other: a **generative** model **G** that creates object instances (e.g., images, sentences) while capturing the data distribution, and a **discriminative** model **D** that estimates the probability that a sample came from the training data rather than **G**. [Ian Goodfellow, 2014]

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THE FOLLOWING PERSONS DO NOT ACTUALLY EXIST!



# Overview of GANs

Examples of generative model applications from [OpenAI](#) ↗ :

## Transformer from GPT-2 (Text Generator) ↗

**Doing a PhD in physics** isn't easy because you're always trying to do too much. For people who have a strong passion for physics, it'll be really helpful to be able to think in a more abstract mathematical fashion. Then it will be a lot easier to work on papers that make sense.



## Musenet ↗

Musenet is a deep neural network that can generate 4-minute musical compositions with 10 different instruments, and can combine styles from country to Mozart to the Beatles. The following composition ↗ improvises Chopin from Mozarts Rondo alla Turca.

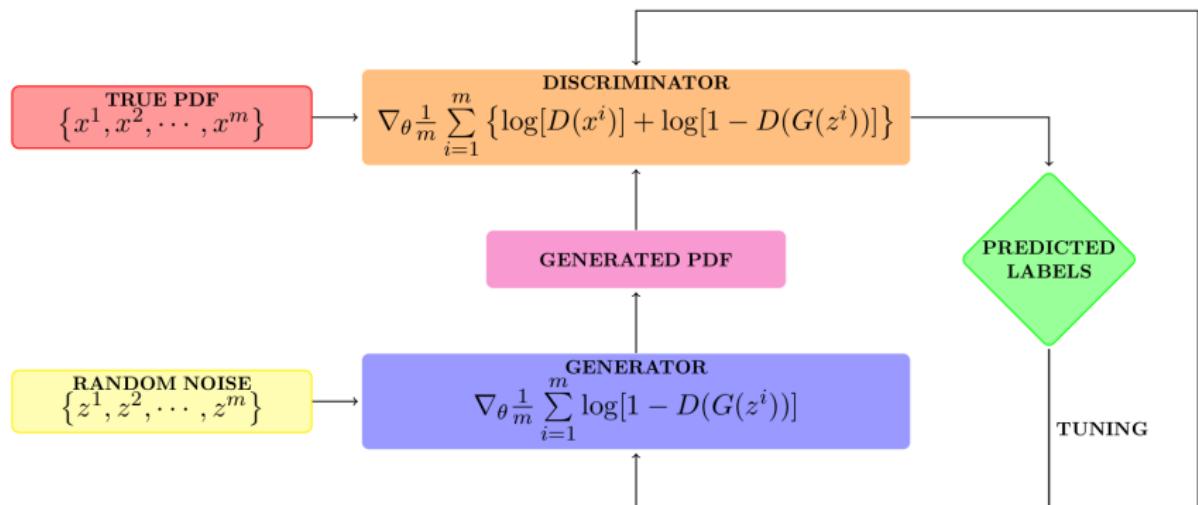


# GANs for Replicas Generation

## Why the need of GANs?

By-pass the need of generating large sample of Monte Carlo Replicas

## Workflow layout of our GAN model:



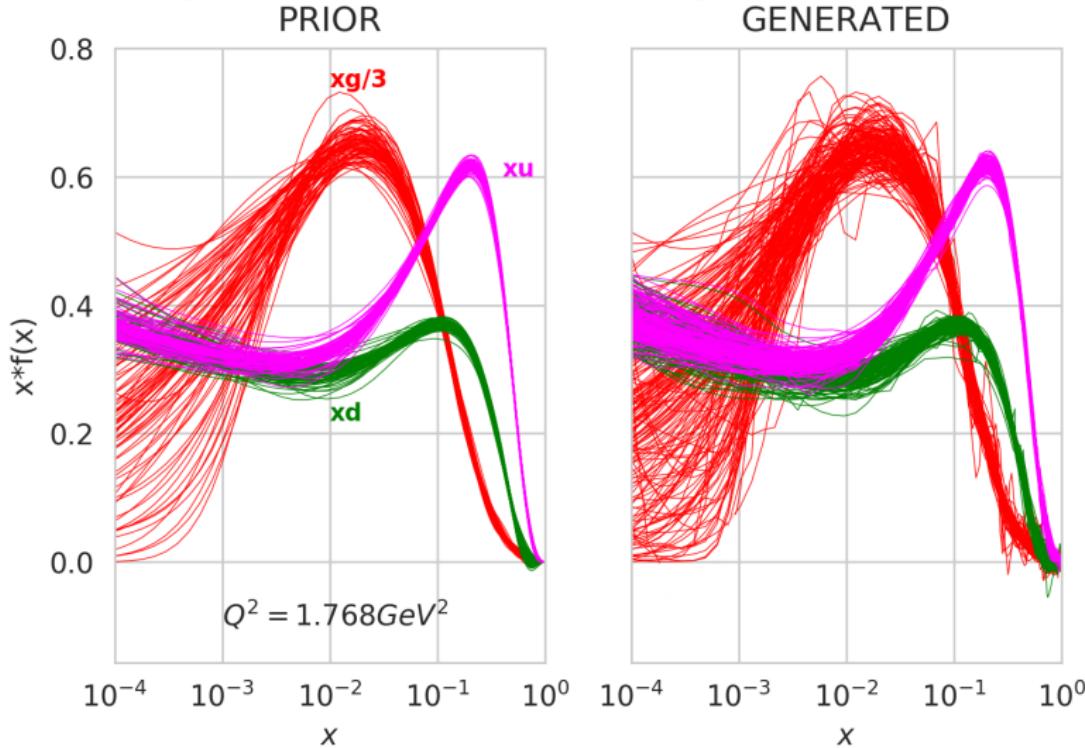
## Preliminary Results

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- We chose PN3\_GLOBAL\_NNPDF31\_nnlo\_as\_0118\_070219-001 from N3FIT with 80 replicas as the prior and generated 160 output replicas
- We compare the probability distribution of the real and generated data for a given value of  $x$

## Preliminary Results

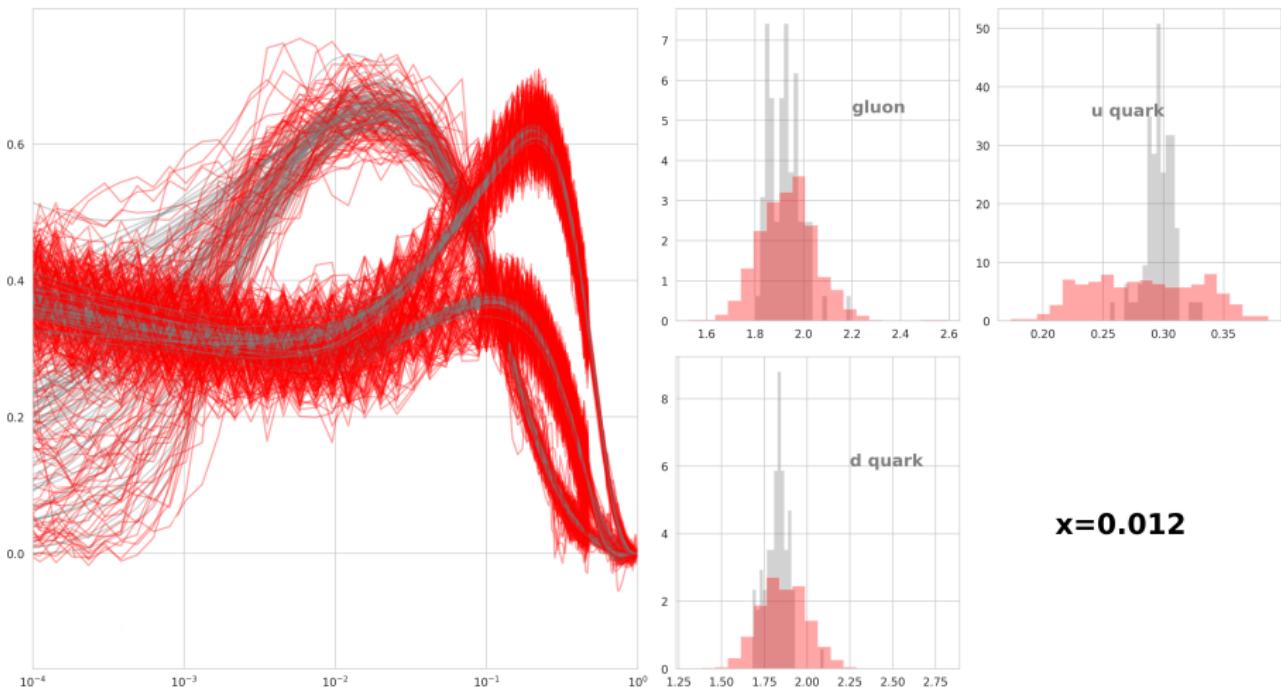
## Comparison between true and generated replicas



# Preliminary Results

Evolution of the training ...

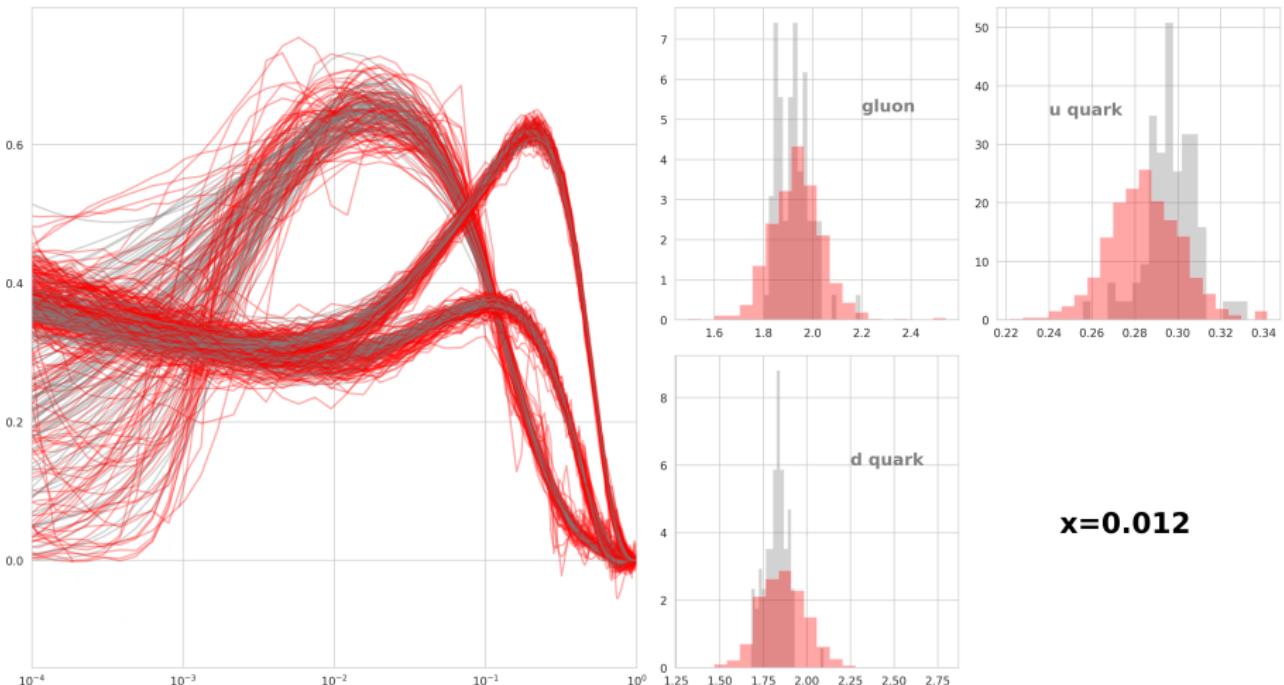
generated\_replicas\_at\_1000\_iteration.png



# Preliminary Results

Evolution of the training ...

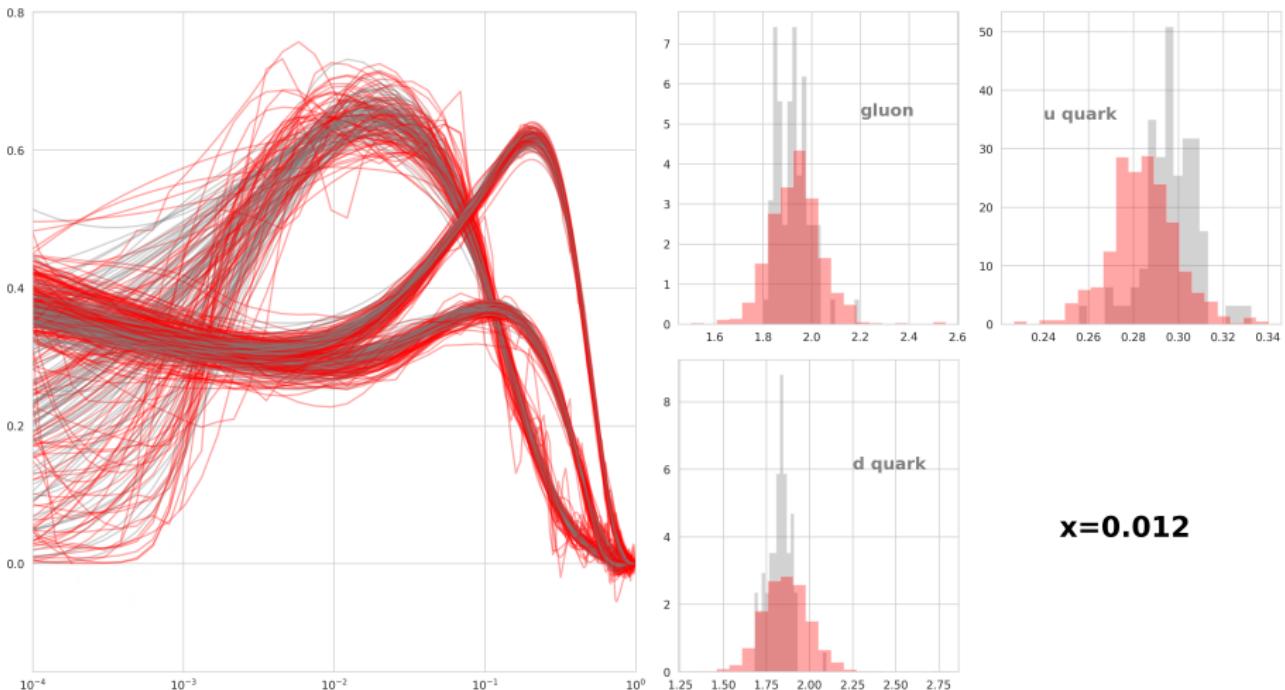
generated\_replicas\_at\_7000\_iteration.png



# Preliminary Results

Evolution of the training ...

generated\_replicas\_at\_10000\_iteration.png



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# Summary and Outlook

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- **What we have achieved so far with GANs:**
  - ✓ Custom and flexible framework
  - ✓ Model that is able to learn the underlying distribution of the original replicas
  - ✓ Generate a large (as much as one want) sample of PDF replicas
- **Ongoing and future works:**
  - ✗ Perform statistical tests to assess performance & efficiency (compare the real and generated distribution)
  - ✗ Extend the model to include all flavors
  - ✗ Modify the model to take into account the momentum sum rules

**THANK YOU**

# BACKUP SLIDES

Here are the backups