

Theoretical physics, Machine Learning and Bioinformatics

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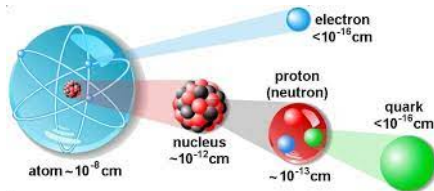
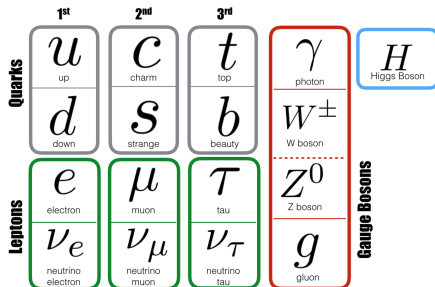
Outline

- ① QCD in a nutshell
 - The fundamental interactions
 - Exploring matter at the small scales
 - Hadronic physics and the LHC
- ② Machine Learning for particle physics
 - The N3PDF project
 - The HTurbo project
- ③ Bioinformatics
 - Applying data sciences to life sciences
- ④ Summary

Quantum Chromodynamics in a nutshell

QCD in a nutshell

The Standard Model



Quantum Field Theory describing physics at the TeV scale \rightarrow **less than a fermi!**

- 1 Fermions (quarks and leptons) composing matter
- 2 Bosons mediating interactions
- 3 Scalar Higgs field generating mass

Quantum Chromodynamics is the theory describing the strong interactions

QCD in a nutshell

Explore the strong interactions

How to explore proton's inner structure?

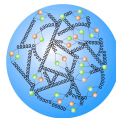


- Point-like projectile on the object \rightarrow DIS
- Smash the two objects \rightarrow LHC physics

"A way to analyze high energy collisions is to consider any hadron as a composition of point-like constituents \rightarrow **partons**" R.Feynman, 1969

QCD in a nutshell

Parton Distribution Functions



- Hadrons made of constantly interacting partons \longrightarrow non perturbative physics
- Parton distribution functions (PDFs) are required for the precision era of the LHC
- PDFs can not be predicted yet not measured \longrightarrow extracted from data via Machine Learning algorithms

Machine Learning for particle physics

Machine learning

What is Machine Learning?

- 1 Machine Learning algorithms are a subset of Artificial Intelligence (AI) algorithms
- 2 Used to solve *complex* tasks like classification, regression and pattern recognition
- 3 Rely on comparison with data → Learning



The N3PDF project

General structure of n3fit

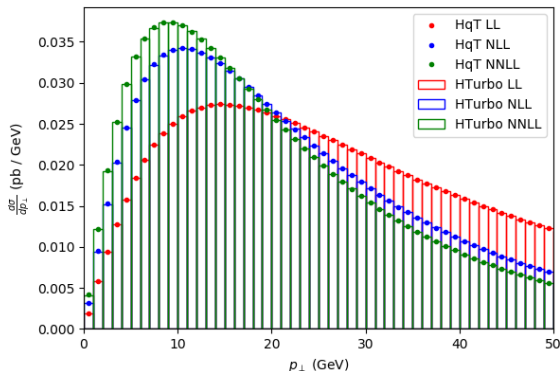
Parton Distribution Functions (PDFs) can not be predicted or measured
PDFs need to be extracted from data!



- Use TensorFlow and Keras to determine the PDFs using neural networks
- Use Stochastic Gradient Descent **n3fit** replacing primitive fitting algorithms
- See paper by S.Carraza - J.Cruz-Martinez
"Towards a new generation of parton densities with deep learning models",
<https://arxiv.org/abs/1907.05075>
- Operator Implementation in TF - Urtasun-Elizari et al.
"Towards hardware acceleration for parton densities estimation",
<https://arxiv.org/abs/1909.10547>

The HTurbo project

Comparison HRes and HqT - all orders



- Older codes (**HRes**, **HqT**) need 3 days to produce NNLL distribution
- 3 minutes with **HTurbo**! ✓
- Agreement up to NNLL → ready for N³LL

Bioinformatics and data science for life sciences

Bioinformatics

Comoputer sciences

- Python, C++, R, Machine Learning
- [<https://github.com/JesusUrtasun/CppCourse>]
- [<https://github.com/JesusUrtasun/MLcourse>]
- [<https://github.com/JesusUrtasun/Bioinformatics>]

Bioinformatics

DNA and RNA sequencing data

- Processing and data grooming of DNA and RNA sequencing data
- Statistics and data analysis of BAM files with R
- **Rsubred** and **Dseq2** packages
- Fast learner!

Summary & Conclusions

- 1 Precise knowledge of sub-nuclear interactions are required towards the precision era of the LHC
- 2 Machine Learning models provide a robust way for PDFs determination optimized through [operator implementation in TF](#)
- 3 We develop a numerical code **HTurbo**, implementing q_\perp resummation for Higgs boson production, which is [faster than any of the existing codes](#)
- 4 Experience with Python and R for NGS data, and still looking to improve!

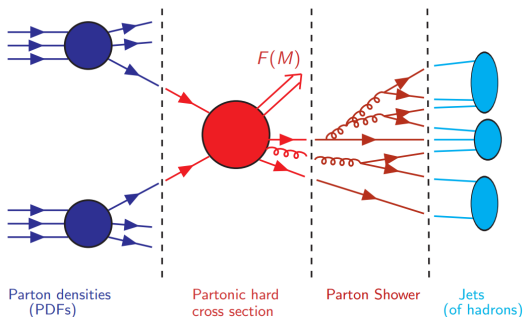
Thank you!



Back up

Hadronic collisions

Hadronic Physics $h_1(p_1) + h_2(p_2) \rightarrow F + X$

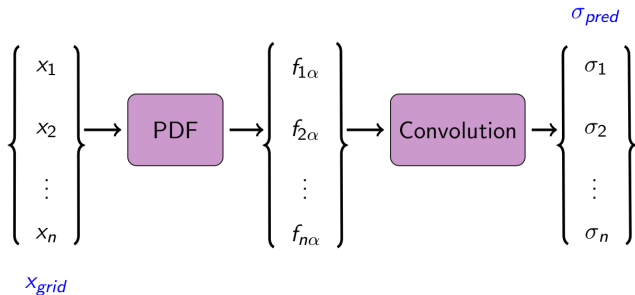


Factorize process as **PDFs** and **partonic (hard) interaction**

$$\sigma^F(p_1, p_2) = \sum_{\alpha, \beta} \int_0^1 dx_1 dx_2 f_{\alpha/h_1}(x_1, \mu_F^2) * f_{\beta/h_2}(x_2, \mu_F^2) * \hat{\sigma}_{\alpha\beta}^F(x_1 p_1, x_2 p_2, \alpha_s(\mu_R^2), \mu_F^2)$$

Back up

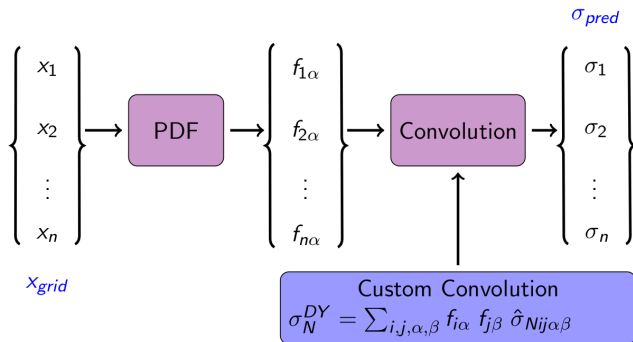
Operator implementation in TF



- Build a NN model to compute σ_{pred} observables from a grid x_i
- Perform χ^2 minimization comparing with data
- Update values of PDF \rightarrow Fit

The N3PDF project

Operator implementation in TF



- 1 TF relies in symbolic computation \rightarrow High memory usage
- 2 Implement c++ operator replacing the convolution
- 3 Further details in Urtasun-Elizari et al.

"Towards hardware acceleration for parton densities estimation",
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