

# The NNPDF open source code

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Zahari Kassabov

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DAMTP, University of Cambridge



UNIVERSITY OF  
CAMBRIDGE



European Research Council

**NNPDF**

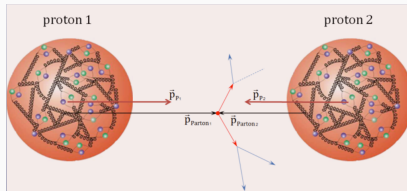
# An open-source machine learning framework for global analyses of parton distributions



- <https://github.com/NNPDF/nnpdf>
- <https://docs.nnpdf.science/>
- arxiv:2109.02671, Eur.Phys.J.C 81 (2021)

# Parton distribution functions (PDFs)

- Large Hadron Collider collides protons.
- Fundamental particle interaction theory given in terms of *partons* (quarks, gluons).
- PDFs describe the constituents of the proton.
- Every collider analysis needs to know about them.



Cannot be determined from first principles.

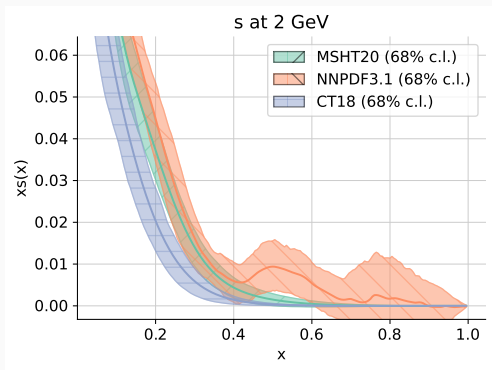
Collider data should be described from physical law  $\rightarrow$  obtain physical law from data.

The NNPDF collaboration has been solving the problem using neural networks since the early 2000s.

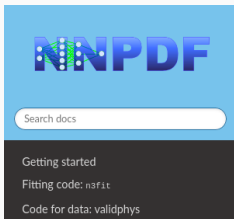
Characteristics:

- “Low” data.
- Complicated data processing.
  - Need to understand both experiment and theory well.
  - Data science problem
- Uncertainty estimation is crucial.

- Critical of LHC analysis necessitates from hard-to-reproduce analyses by experts.
- Differences between analyses are hard to understand



- Ability to look into the code provides reassurance.



🏠 » [Tutorials](#) » [How to reproduce an NNP4.0 fit](#)

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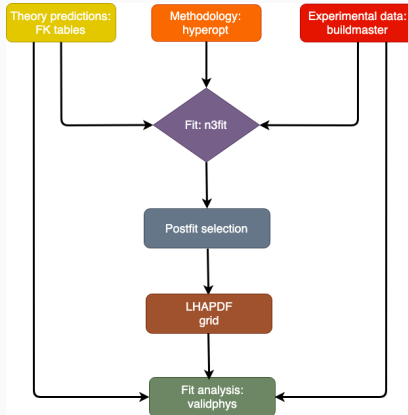
## How to reproduce an NNP4.0 fit

Here we describe how to reproduce the NNP4.0 PDF sets that are publicly available through LHAPDF as

- Fitting code
- Data science framework
- All of the data
- Documentation
- Python code base (with some, mostly legacy C++)

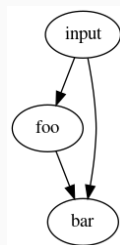
# Fitting code

- The `n3fit` code, originally presented in [Carrazza, Cruz, 2019]
- Based on Tensorflow, with access to all its algorithms and target processors.
  - Order of magnitude faster than previous genetic algorithm based implementation.
- Hyperoptimization framework that helps select a good algorithm.



# Analysis code

- A **reportengine** application [ZK, 2019]
  - Functional code structure
  - Runcard driven
  - Graph checks and execution
- Core data structures
- Analysis code and plots



```
# runcard.yaml
```

```
input: value
```

```
actions_:
```

```
- bar
```

```
# mymodule.py
```

```
def foo(input):
```

```
    return ...
```

```
@check_input_is_right_for_bar
```

```
def bar(foo, input):
```

```
    return ...
```

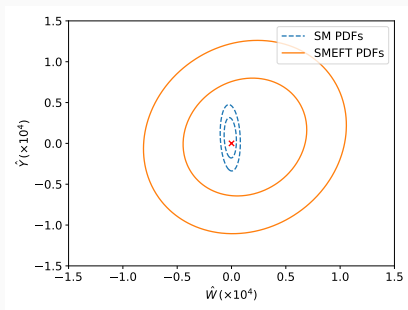


More complex features in **reportengine** (namespaces, dependency induction) allow building fairly complex data science pipelines, still reproducible with a declarative runcard.

- Typically the final output is an HTML report.
- E.g. **vp-comparefits** (included in the code) produces hundreds of plots and tables comparing two NNPDF fits.
- Comprehensive closure test framework.
- Dozens of specialized plotting tools.
- The fitting code itself.

# Extensions

- The code structure opens the door for interesting avenues.
  - E.g. extend to nuclear fits
- Precise enough analyses *need* simultaneous fits [Forte, ZK, 2020]
  - Including of EFTs



Greljo et al, 2021

- Watch out the PBSP group in Cambridge in this space!