



Workflow developments for interpolation grids with NNLOJET and APPLfast

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(APPLfast team)**



Outline



- **Workflow overview**
- **CPU time & statistical precision**
- **Interpolation in x (Master student Johannes Gäßler)**
- **Workflow on Ixplus (Summer student Christiane Mayer)**



The Components



➔ NNLOJET:

Theory program providing NNLO predictions

➔ APPLfast:

Common interface of APPLgrid & fastNLO to pQCD theory programs

➔ APPLgrid, fastNLO:

Creation of fast interpolation grids

NNLOJET

NNLOJET, A. Gehrmann-de Ridder et al., Z+jet: JHEP07 (2016) 133,
Framework: T. Gehrmann et al., PoS RADCOR2017 (2018) 074.

APPLfast, D. Britzger et al., ch. I.3 in arXiv:1803.07977.



APPLgrid, T. Carli et al., EPJC66 (2010) 503,
FastNLO, D. Britzger et al., arXiv:1208.3641.

➔ Luigi & Luigi Analysis Workflow (LAW):

Design of distributed, pipelined analysis workflow with inter-dependencies

➔ NNLO LAW Analysis & Website:

Example application to NNLO calculations with integrated plot server



Luigi (Spotify): <https://github.com/spotify/luigi>
LAW: <https://github.com/riga/law>

LAW workflow with NNLOJET & APPLfast
<https://gitlab.etp.kit.edu/qcd-public/law-analysis>

M. Santos Correa, Master Thesis, EKP-2019-00003.pdf.



Luigi and Spotify



Python package to build complex pipelines of batch jobs:

- **Features:**

- ➔ **Dependency resolution**
- ➔ **Workflow management**
- ➔ **Visualisation**
- ➔ **Handling failures**
- ➔ **Command line integration**
- ➔ **...**



Luigi analysis workflow



Built on top of Luigi, adds abstraction of run locations, storage locations, and software environments:

• Features:

- ➔ **Remote targets with automatic retries and local caching**
(WebDAV, HTTP, Dropbox, WLCG protocols: srm, xrootd, dcap ...)
- ➔ **Automatic submission to batch systems within tasks**
(HTCondor, LSF, gLite, **Slurm (not tested yet)**, ...)
- ➔ **Environment sandboxing, configurable on task level**
(Docker, Singularity, ...)



A typical workflow



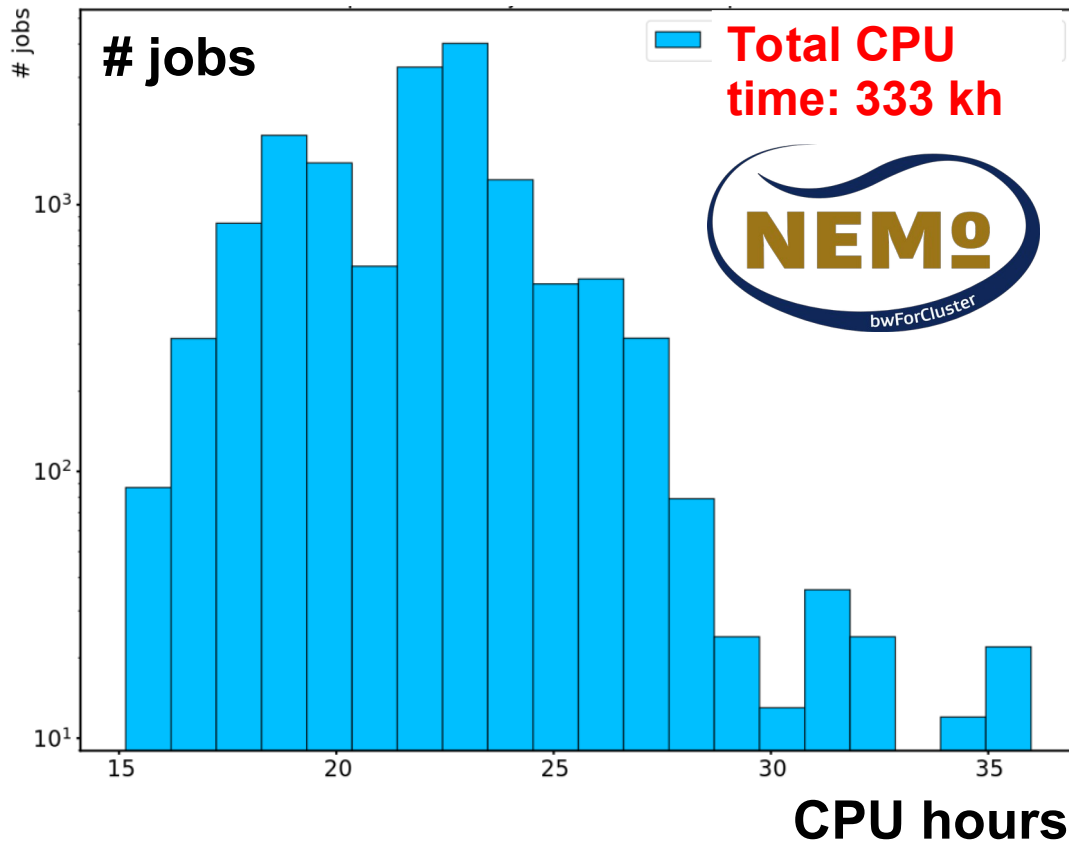
- **1. Preprocessing: Check of interpolation quality** FTE time
 - ➔ Short test jobs to check interpolation settings (& optimise if necessary)
- **2. NNLOJET Warm-up: Vegas integration optimisation** CPU & FTE time
 - ➔ 1 long (multi-core) job per process
- **3. APPLgrid/fastNLO Warm-up: Adapt x- and scale-grids to accessed phase space (exact strategy differs between APPLgrid & fastNLO)** FTE time
 - ➔ Only phase space provided from NNLOJET → significant speed-up
- **4. Interpolation grid production:** A lot of CPU time!
 - ➔ Thousands of parallel jobs
- **5. Postprocessing: Statistical evaluation and combination of all produced grids ...** CPU time
 - ➔ Job to combine all grids and estimate statistical uncertainty
- **6. Provide final closure & result plots** No time
 - ➔ Automated plotting scripts



The investment



Typical total runtime of a grid production



Total # of ~24h jobs: $O(15000)$

Non-CMS CPU resources from:
NEMO Freiburg, MPI Munich, CERN Theory

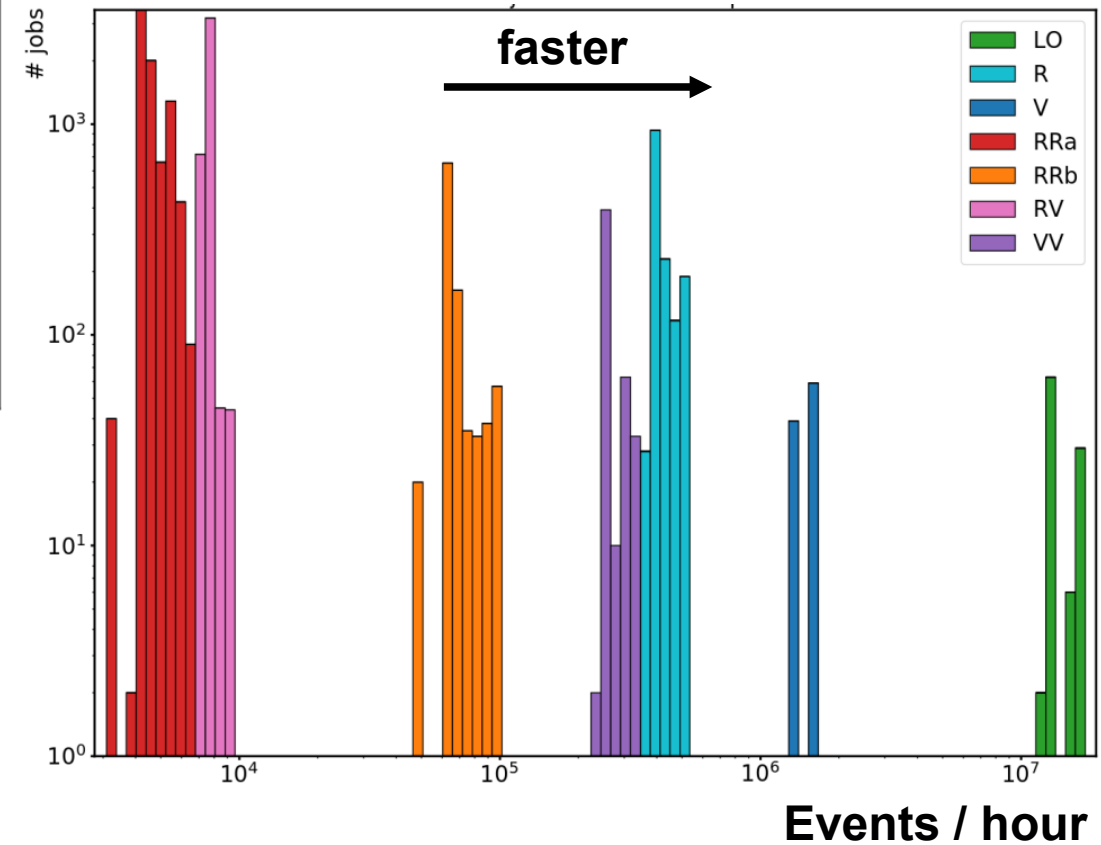
“Events” / hour

RRa, RV:
most expensive

LO, NLO, VV:
easy

$O(10^3 - 10^4)$

$O(10^5 - 10^7)$





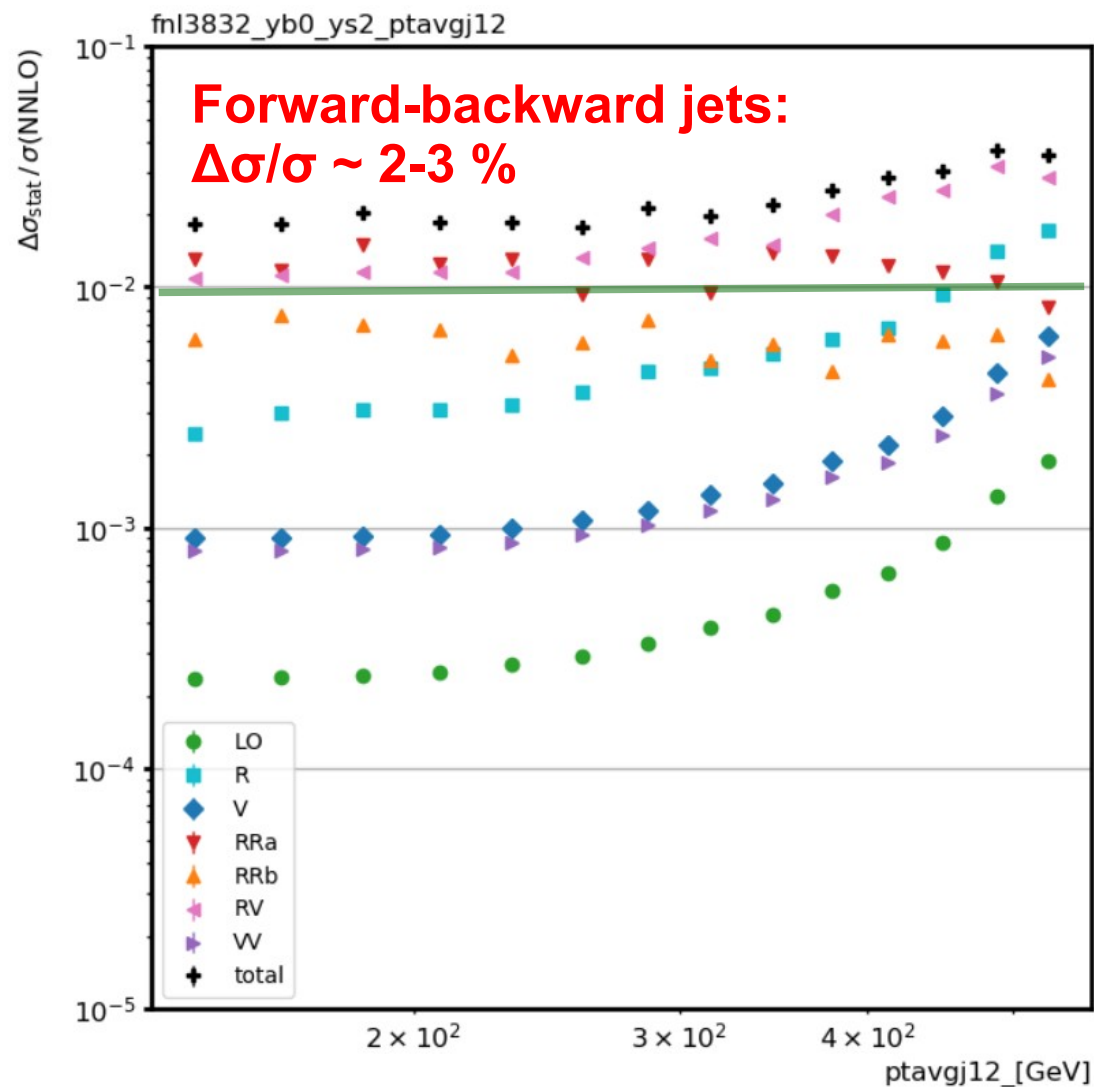
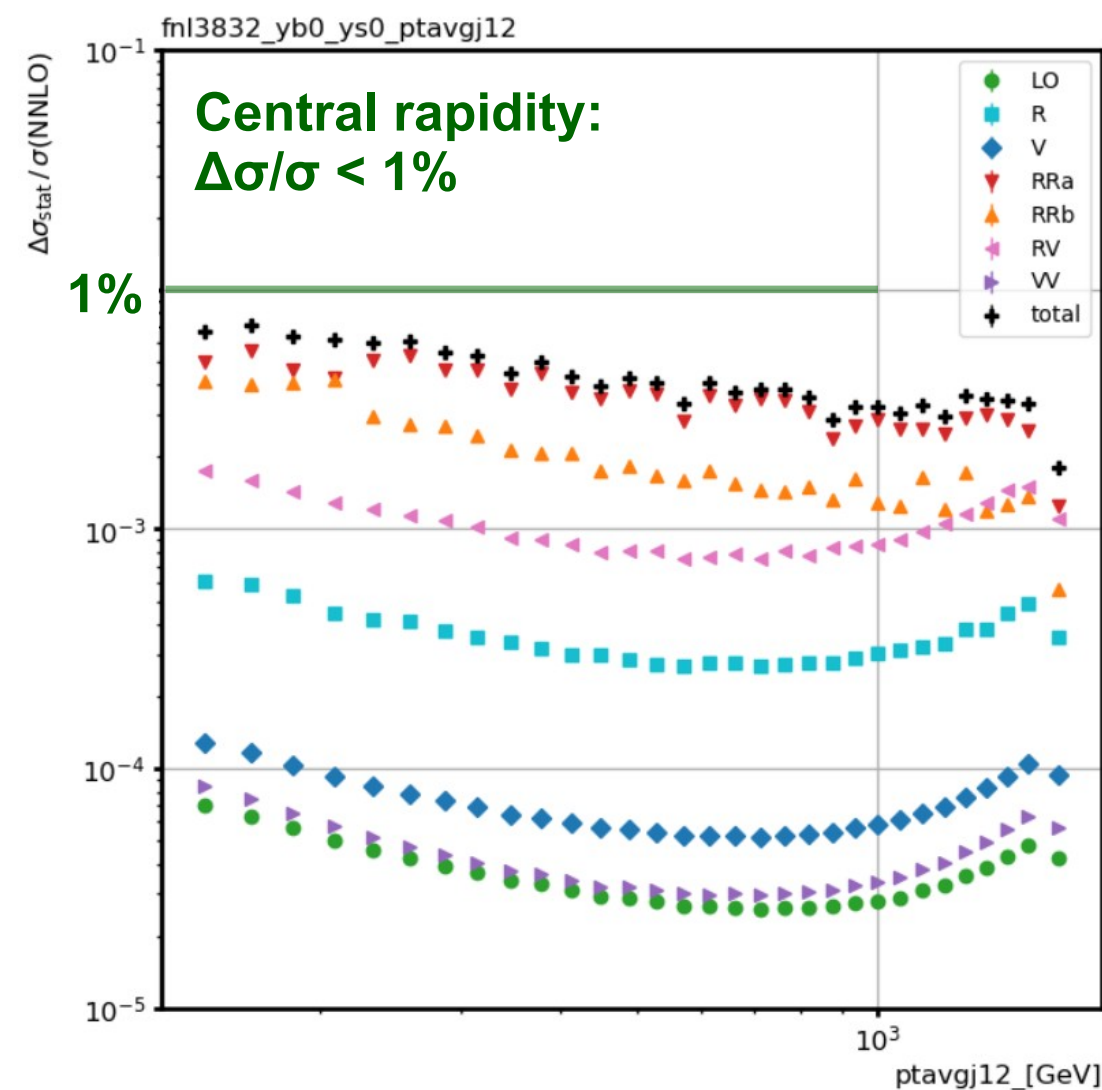
Return on investment



Relative numerical uncertainty of NNLO 3D dijet cross section

Dominated by RRa and RV channels

Numerical uncertainty provided inside grids

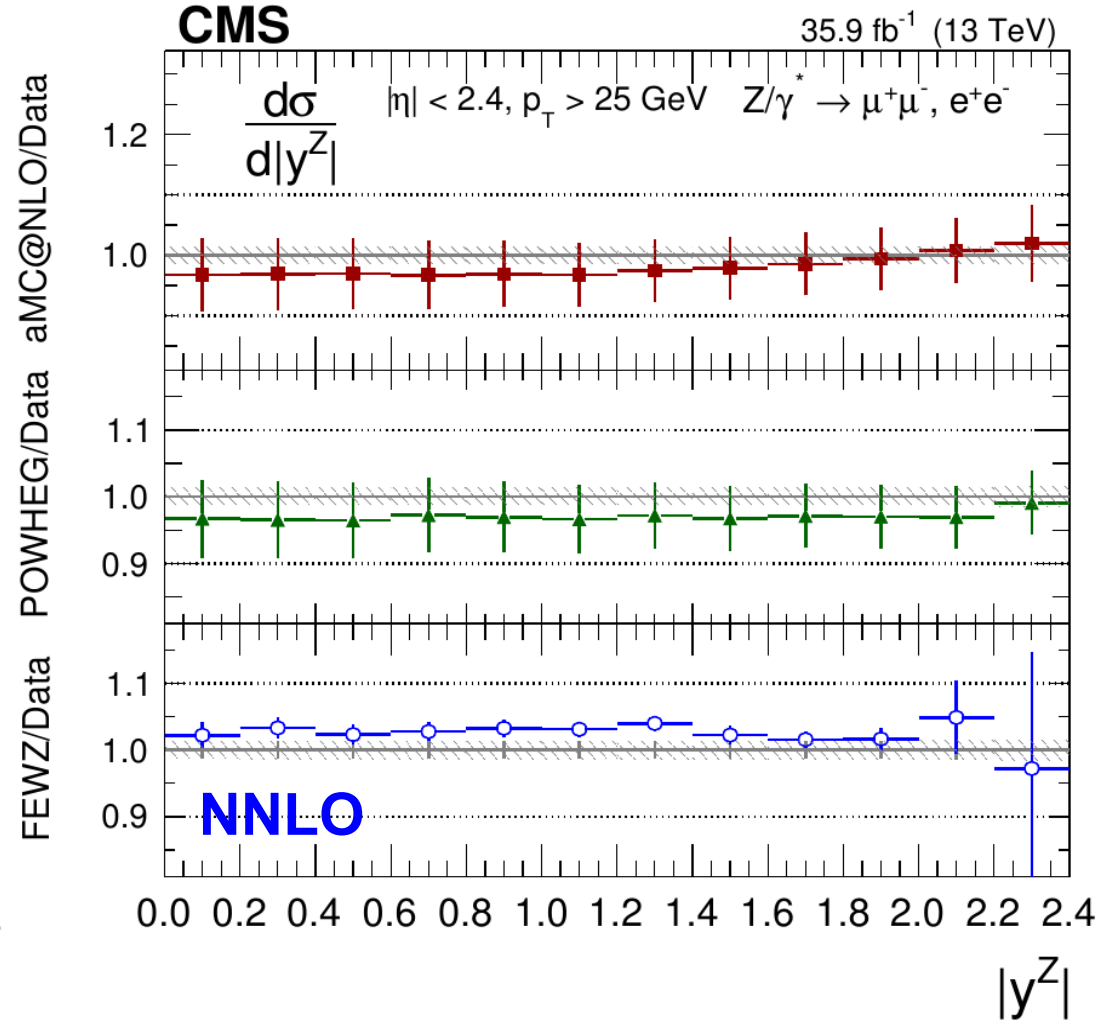
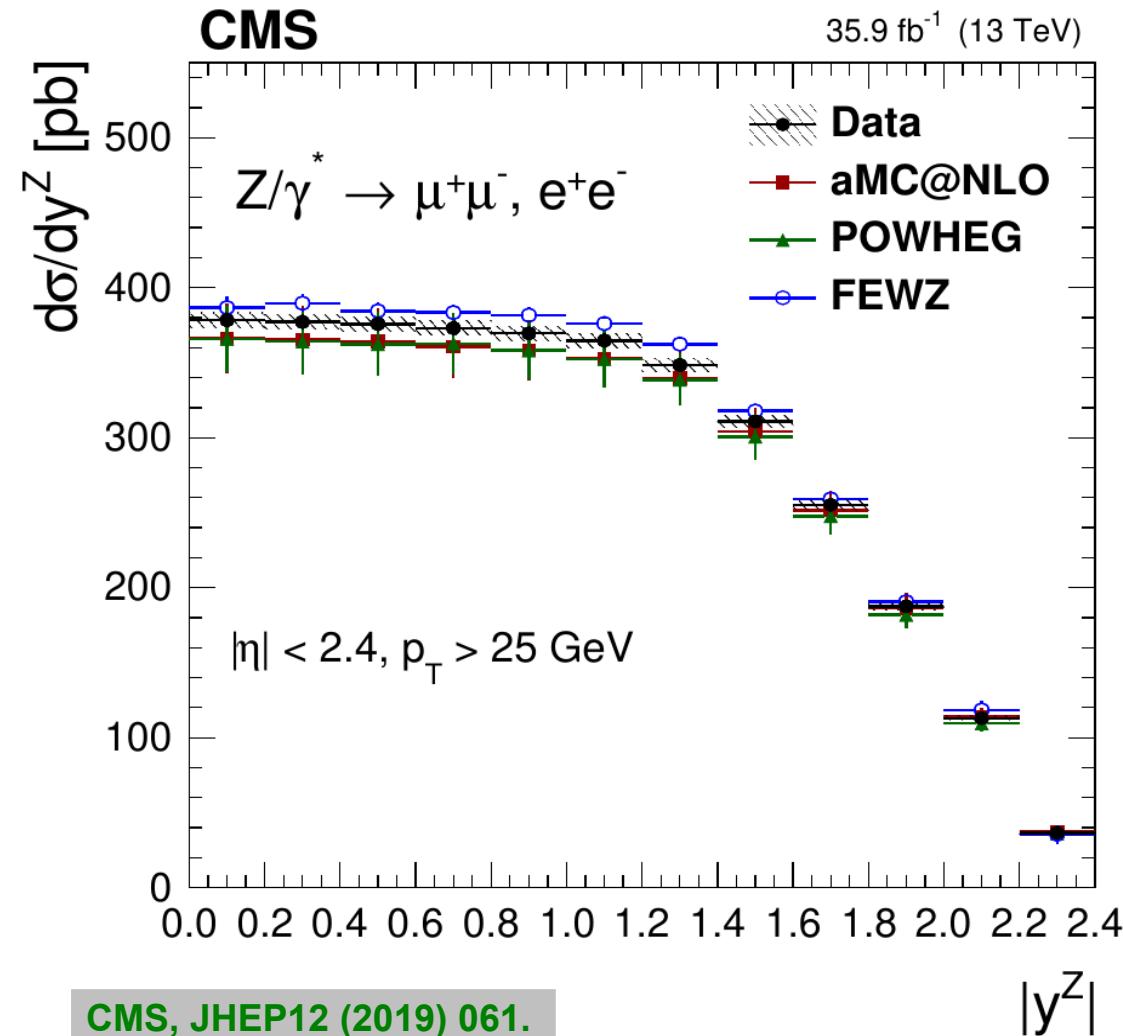




Summer student project



Christiane Mayer: - Establish grid production workflow (LAW) on **Ixplus** (with Alexander) - Test automated **VEGAS** grid optimisation



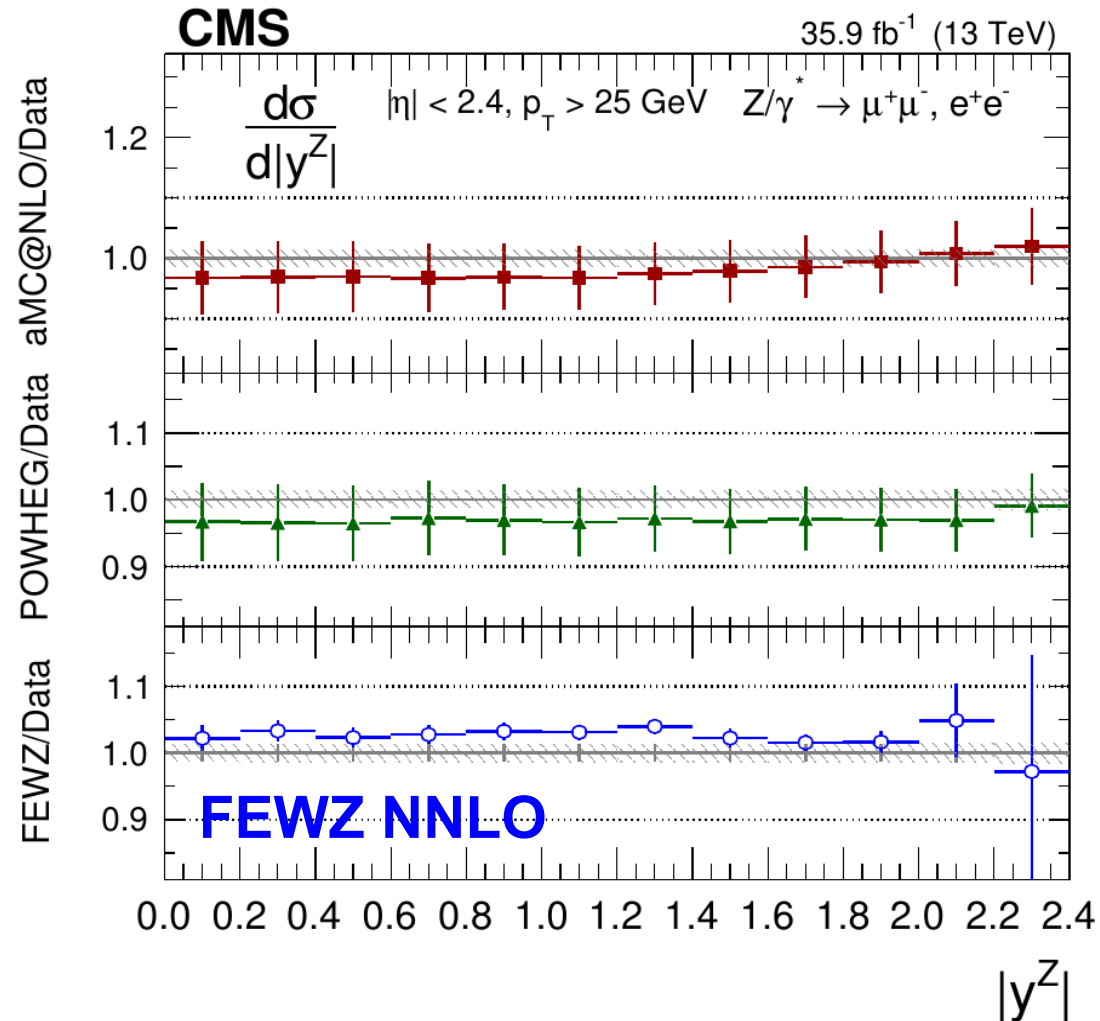
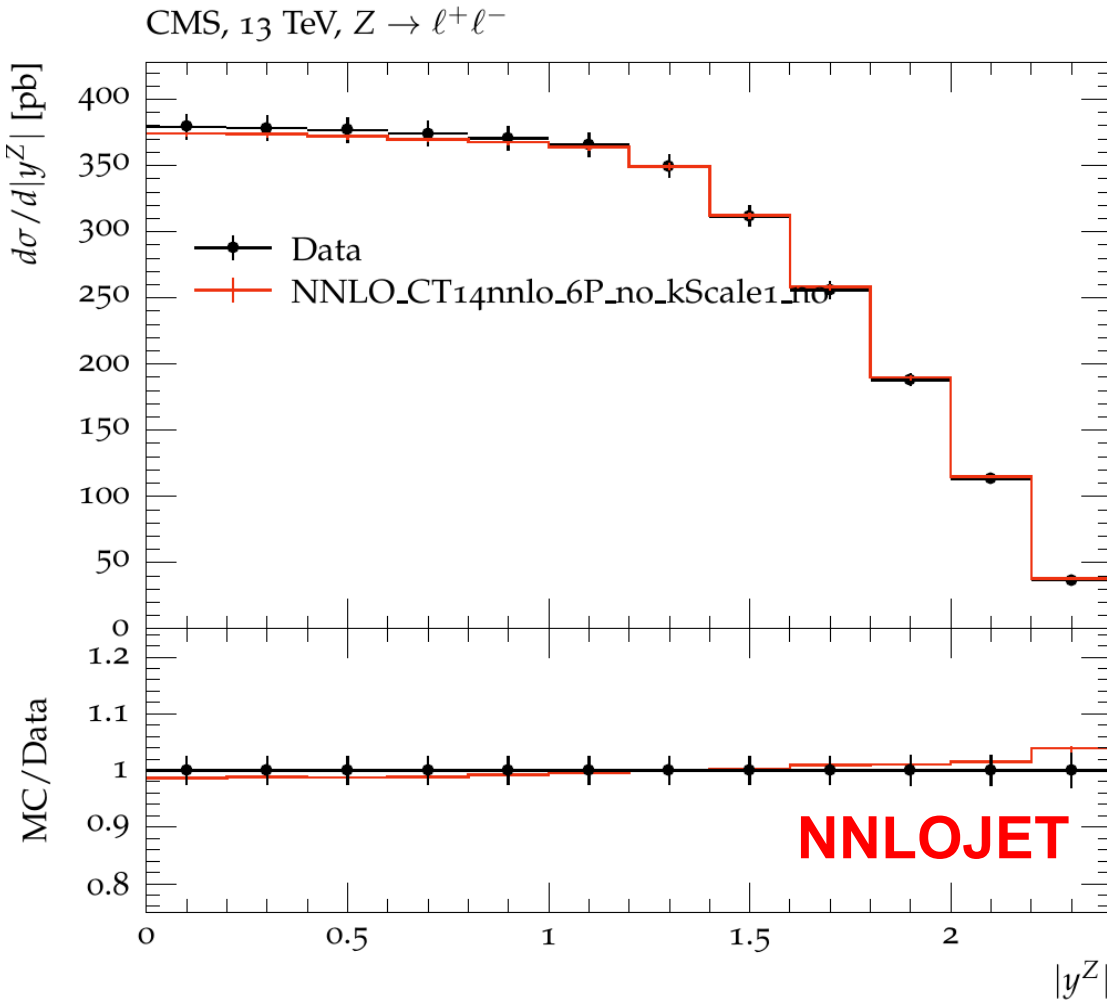
CMS, JHEP12 (2019) 061.



Data comparison



Christiane Mayer: - Ran over night
 (with A. Huss, TH) - Nicely agrees with CMS data
 - Slightly lower than FEWZ



Christiane Mayer



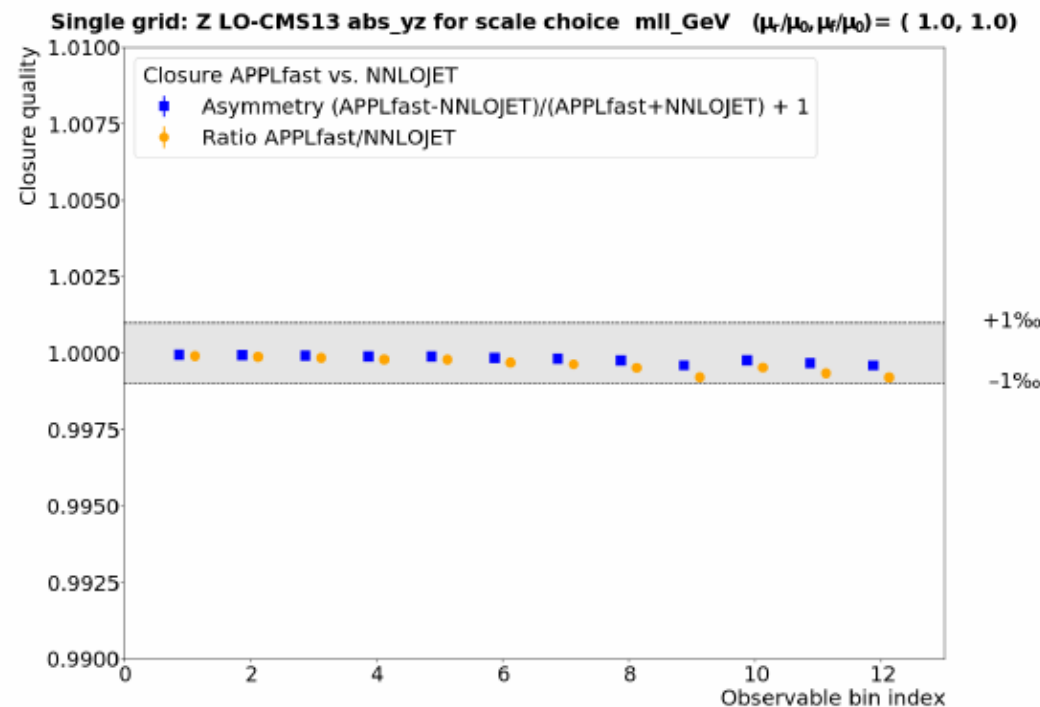
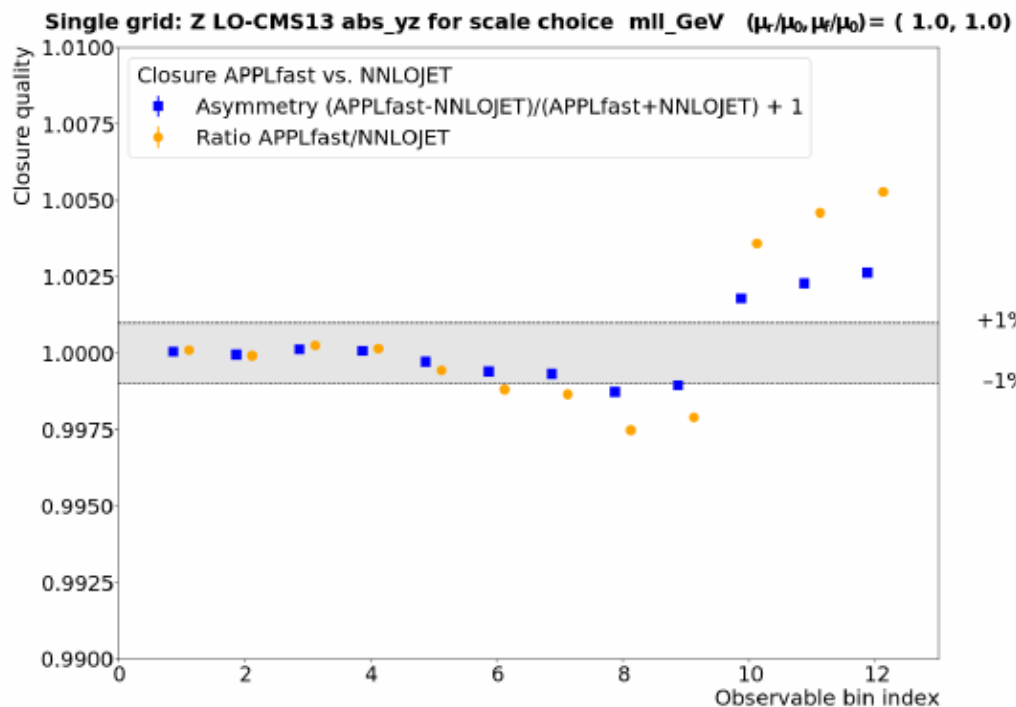
Grid closure vs. NNLOJET



- No. of nodes vs. grid size
- Ex. DY varying x node distribution
- ➔ **Left: Equidistant in $-\sqrt{-\log(x)}$**
- ➔ **Right: Equidistant in $\log(x)$**

support nodes per magnitude	table size [MB]
8	50
12	100
16	180

Table 1: Increasing Table Size



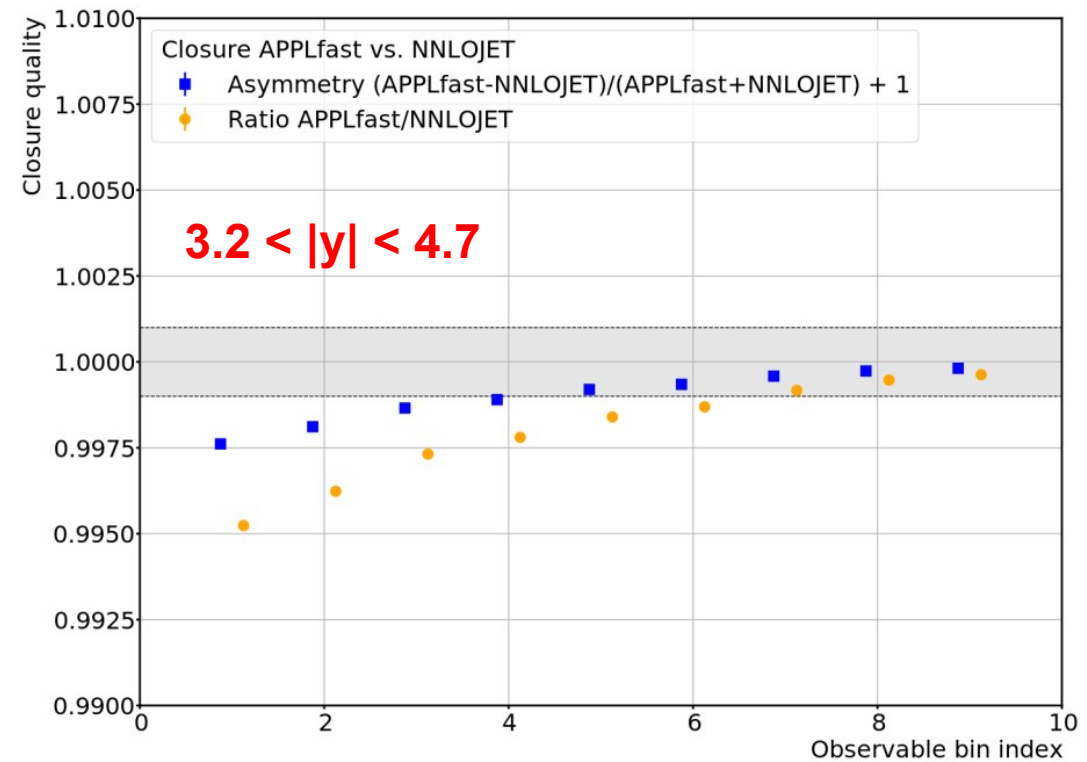
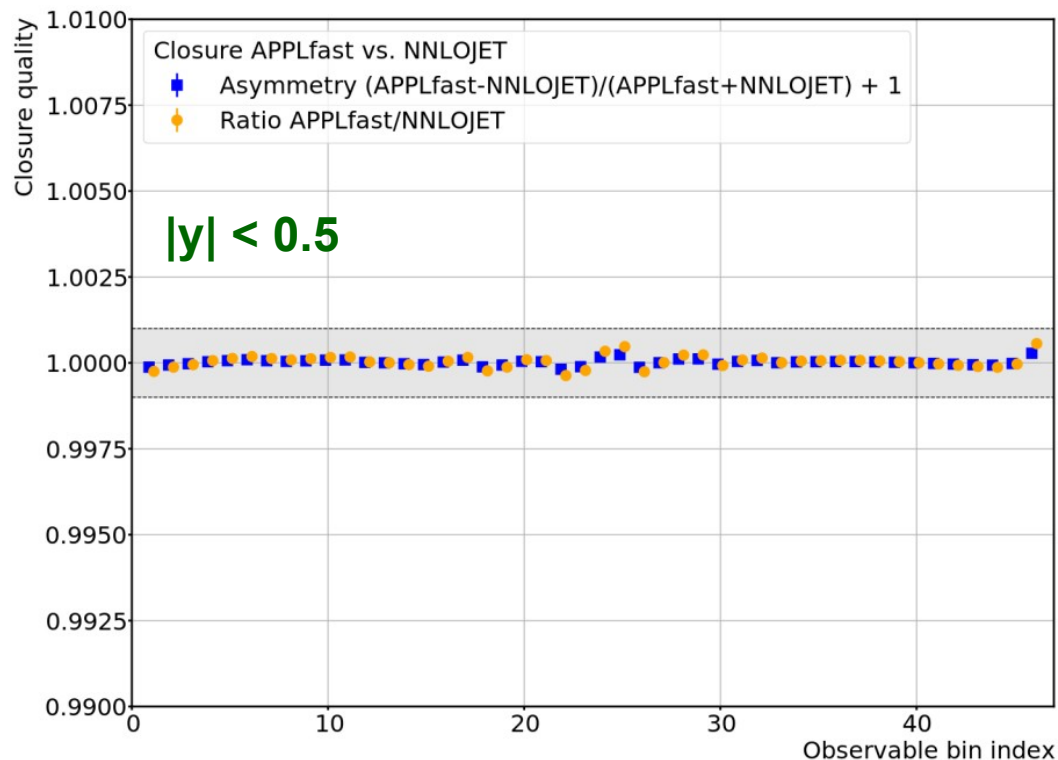
Christiane Mayer



Grid closure vs. NNLOJET



- Example of CMS inclusive jets at 8 TeV cms energy
 - ➔ Interpolation in x fine for $|y| < 3$; setting not optimal for larger rapidity
 - ➔ Use x node density setting instead of fixed no. of nodes (works already)
 - ➔ Automated extension of x range on-the-fly (→ next slides)

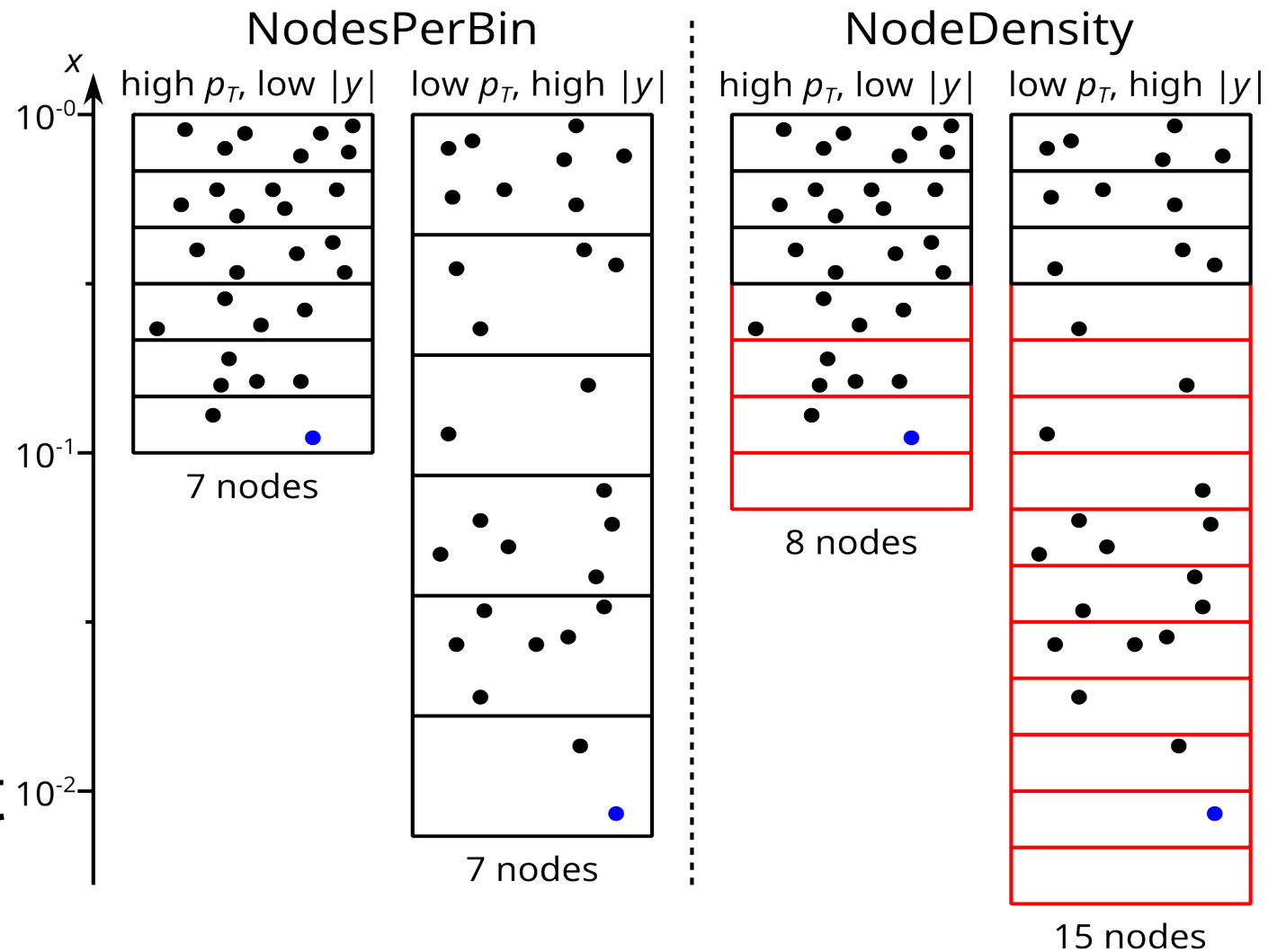




Fixed # Nodes vs. Node Density



- x : fixed node density instead of fixed node number
- Natural upper x bound at 1
- Start with 4 nodes, **expand** dynamically to fit **minimal** x



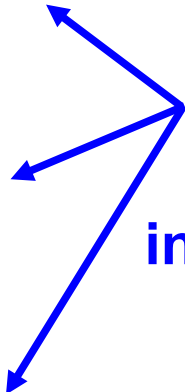


x-node extension



- Adding more *x* nodes to storage array a little tricky ...

- Deep inelastic scattering: $(1 \ 2 \ 3) \rightarrow (0 \ 1 \ 2 \ 3)$

- Proton-proton: $\begin{pmatrix} 1 & & \\ 2 & 3 & \\ 4 & 5 & 6 \end{pmatrix} \rightarrow \begin{pmatrix} 0 & & & \\ 0 & 1 & & \\ 0 & 2 & 3 & \\ 0 & 4 & 5 & 6 \end{pmatrix}$ 

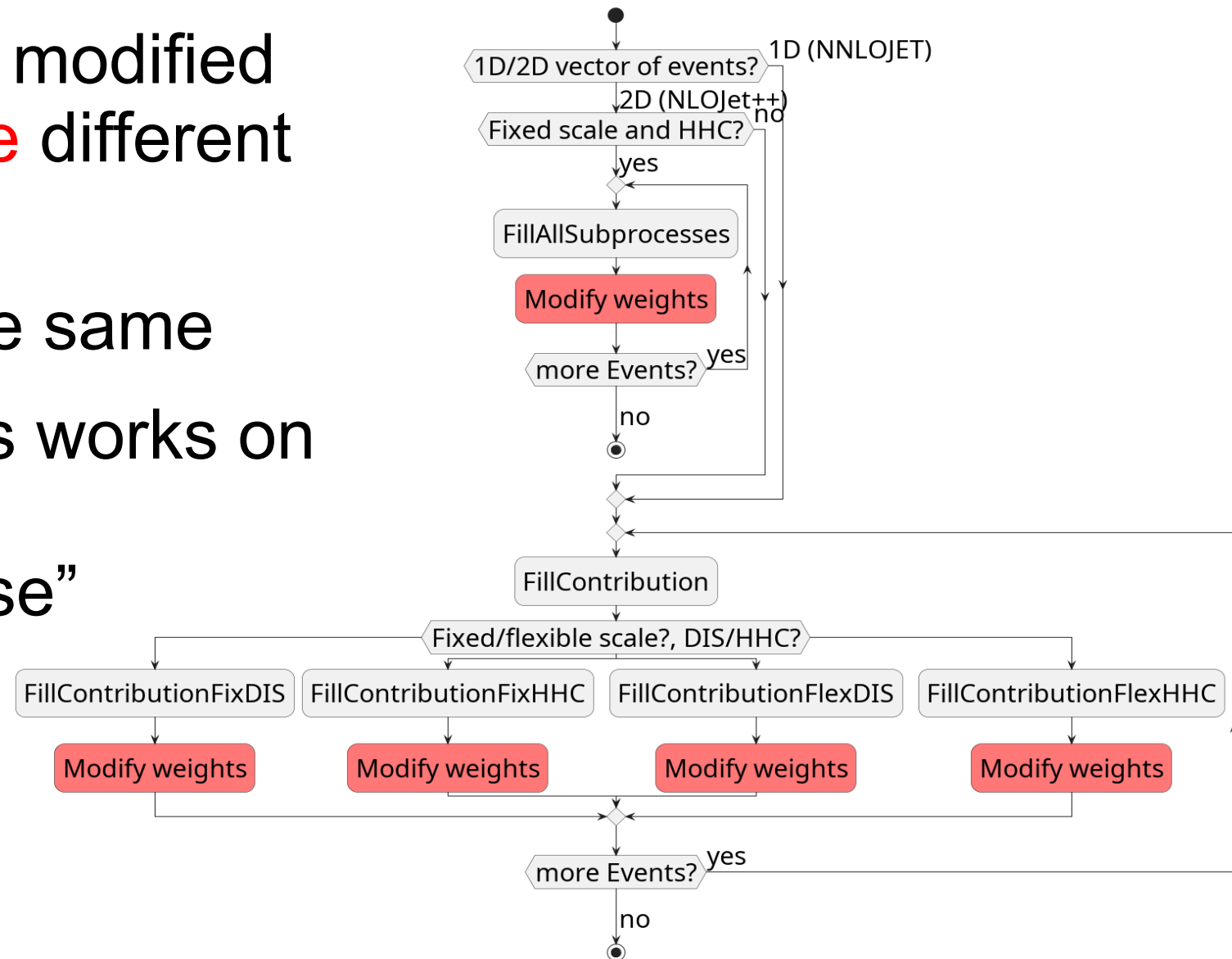
- Proton-antiproton: $\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 2 & 3 \\ 0 & 4 & 5 & 6 \\ 0 & 7 & 8 & 9 \end{pmatrix}$



Implementation



- Original code modified weights in **five** different places
- Code 80% the same
- “Create” class works on attributes of “CoeffAddBase” classes

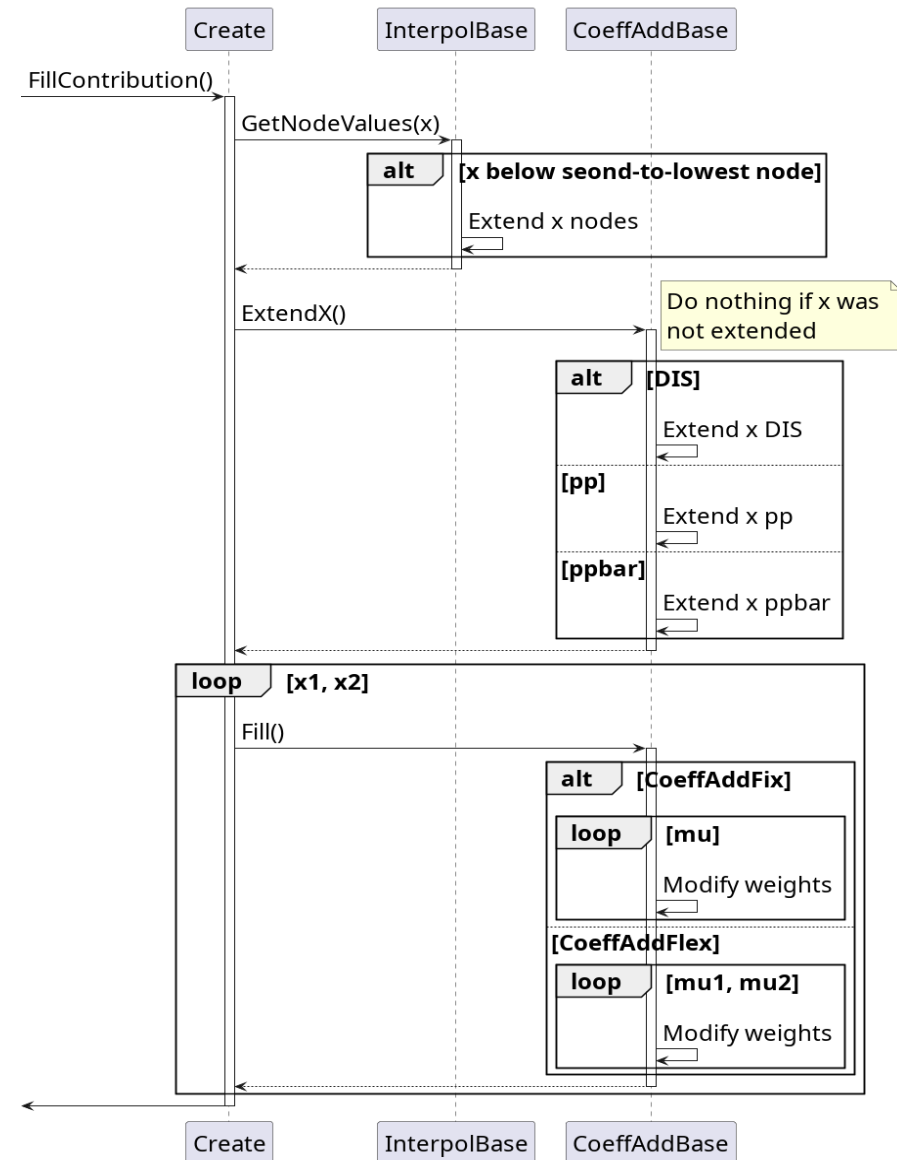
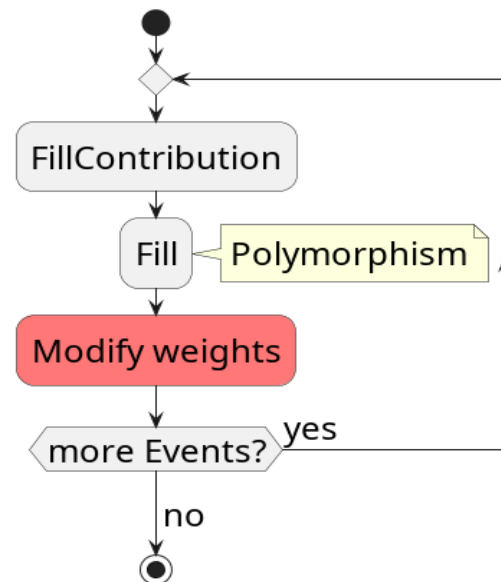




Implementation



- Deduplicated code as part of density implementation
- Object-oriented design: polymorphism, encapsulation
- Also some speedup for flexible scale





Performance optimizations



- fastNLO code deduplicated and refactored
 - ➔ In addition some code speedup
- Test at NLO with NLOJet++
 - ➔ Helps significantly since ~90% of runtime spent in fastNLO
- First test with NNLOJET & APPLfast2 for modules 2
 - ➔ RRa channel <10% of runtime spent in APPLfast
- More checks to follow ...
- Would also be interesting to see gain when not accessing LHAPDF anymore (after closure established)

Valgrind





Summary & Outlook



- **Grid production workflow successfully used for most of the published pp grids**
- **Summer student project:**
 - **Adaptation of LAW workflow to Ixplus and production of grids for Drell-Yan**
 - **Test automisation of Vegas grid production**
- **Master student work:**
 - **Workflow simplifications (x-interpolation) and performance improvements in progress**
- **Outlook**
 - **Adapt workflow at KIT and Ixplus to NNLOJET modules 2 & APPLfast2**
 - **Test and run workflow on acquired HPC resources (HoreKa at KIT), also with Slurm**
 - **Include additional workflow & performance improvements**
 - **Provide documentation for more user friendliness ...**




Backup Slides





Ingredients



- **Theory: Today interface to**  (others NLOJet++, MCFM, ...)
 - ➔ **NNLOJET:** T. Gehrmann et al., RADCOR2017 PoS (2018) 074, arXiv:1801.06415.
 - ➔ **Inclusive jets:** J. Currie et al, PRL 118 (2017) 072002; JHEP 10 (2018) 155.
 - ➔ **Dijets:** J. Currie et al., PRL 119 (2017) 152001; A. Gehrmann-de Ridder et al., PRL 123 (2019) 102001.
 - ➔ **Full-colour NNLO:** X. Chen et al., arXiv:2204.10173. (Not yet included in grids!)

- **Tools:**



- ➔ **APPLfast interface:** D. Britzger et al., EPJC 79 (2019) 845, arXiv:1906.05303.
- ➔ **fastNLO:** D. Britzger et al., Proc. DIS2012 (2012) 217, arXiv:1208.3641.
- ➔ **APPLgrid:** T. Carli et al., EPJC 66 (2010) 503, arXiv:0911.2985.
- ➔ **xfitter:** S. Alekhin et al., EPJC 75 (2015) 304, arXiv:1410.4412.





Interpolation concept



Implemented in APPLgrid & fastNLO

Use interpolation kernel

- Introduce set of n discrete **x-nodes**, x_i 's being equidistant in a function $f(x)$
- Take set of **Eigenfunctions** $E_i(x)$ around nodes x_i

→ Interpolation kernels

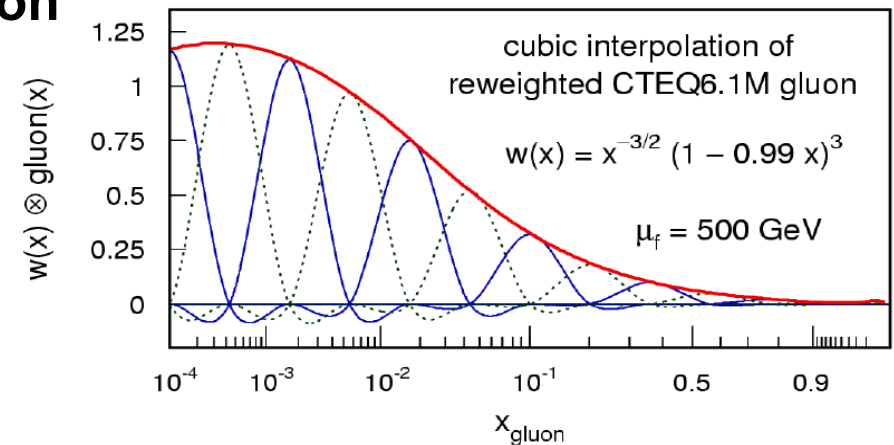
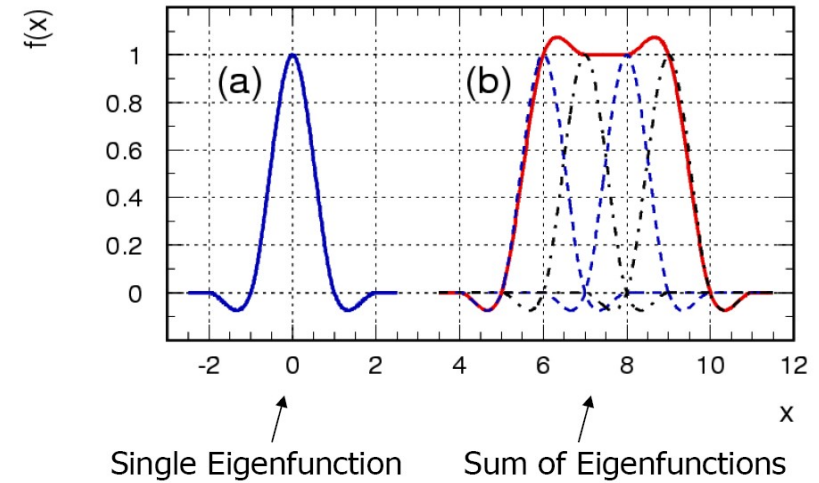
- Actually a rather old idea, see e.g.

C. Pascaud, F. Zomer (Orsay, LAL), LAL-94-42

→ Single PDF is replaced by a linear combination of interpolation kernels

$$f_a(x) \cong \sum_i f_a(x_i) \cdot E^{(i)}(x)$$

- Then the integrals are done only once
- Afterwards only summation required to change PDF



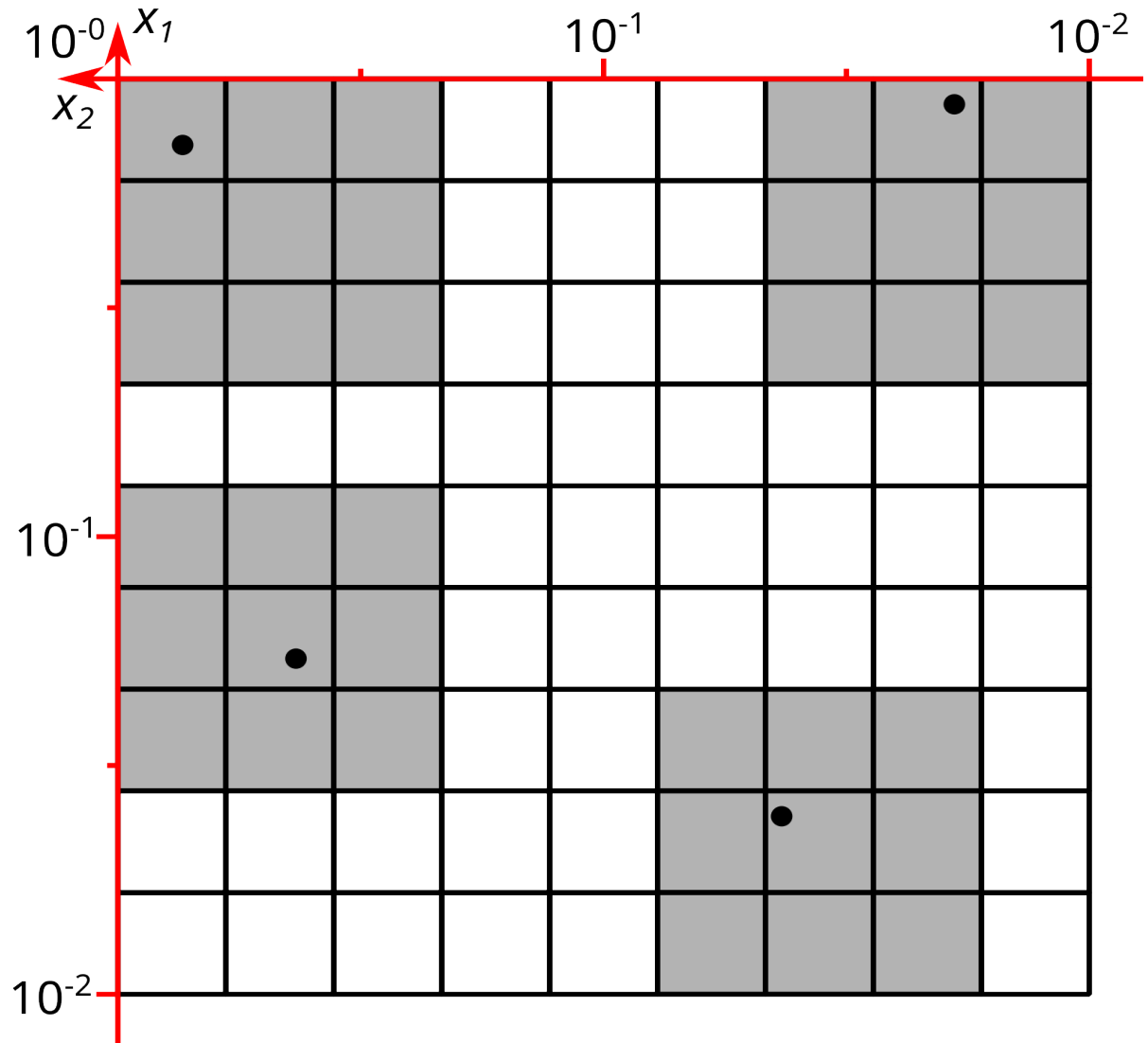
Tabulate the convolution of the perturbative coefficients with the interpolation kernel



Node Weight Spread



- Weights spread over 16 (x_1, x_2) nodes
- Special treatment of **boundaries**
- No lower boundary for node density
- Need to add one more x node

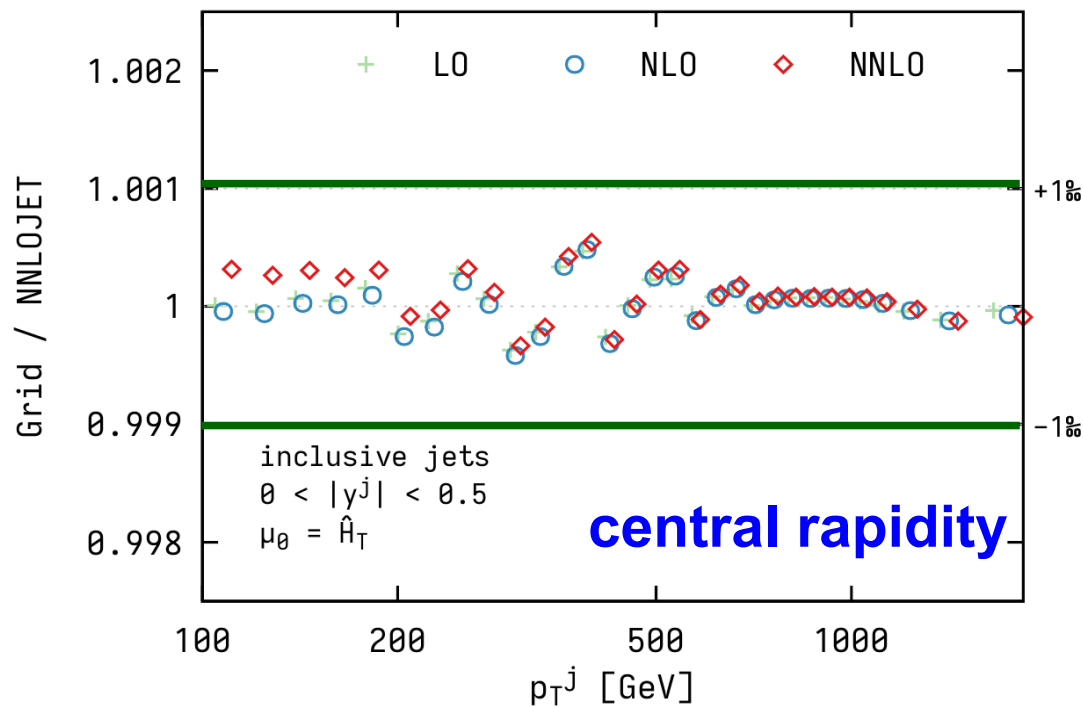




Grid closure vs. NNLOJET



APPLfast + NNLOJET ATLAS $\sqrt{s} = 7$ TeV

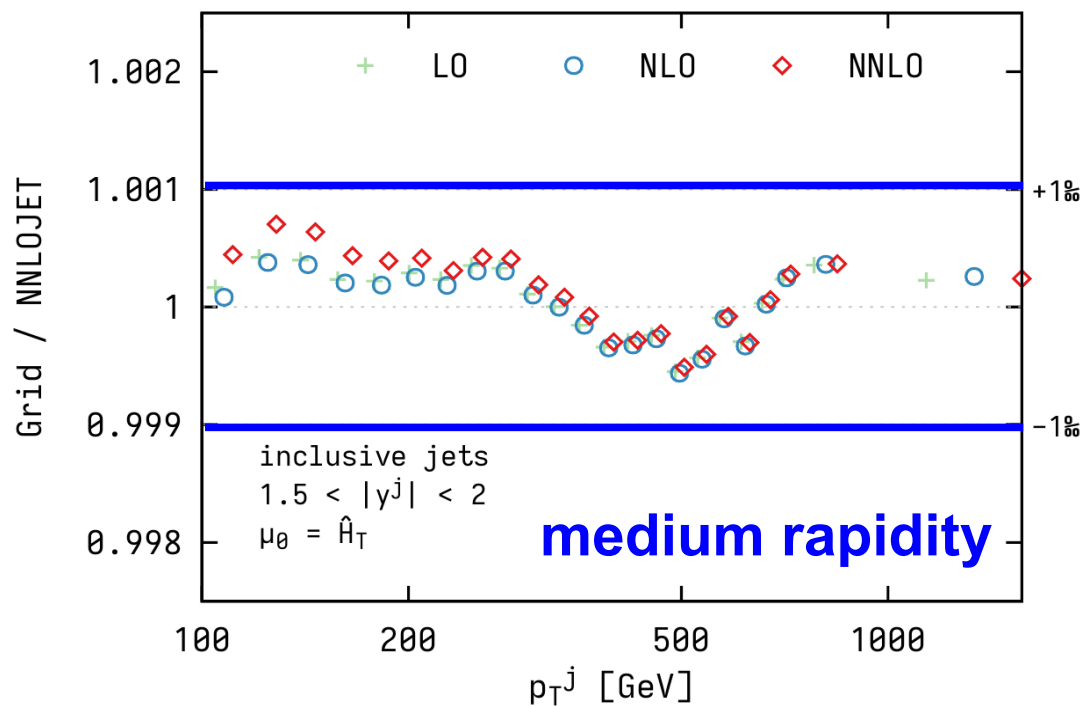


Closure deteriorates somewhat towards phase space limits; exceptionally may exceed 1 ‰ at phase space edges

ATLAS inclusive jets at 7 TeV

Generally aim at closure better than 1 ‰ at each level, LO, NLO, NNLO

APPLfast + NNLOJET ATLAS $\sqrt{s} = 7$ TeV



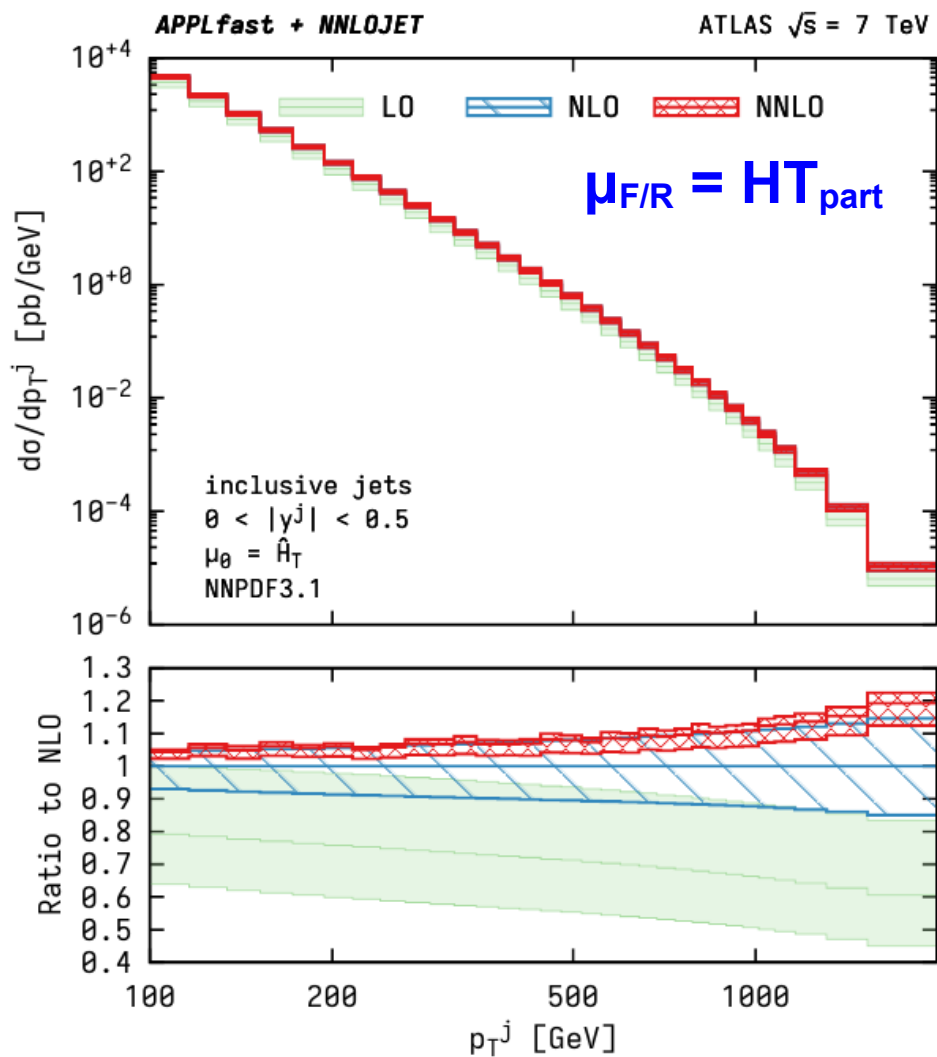


Scale dependence

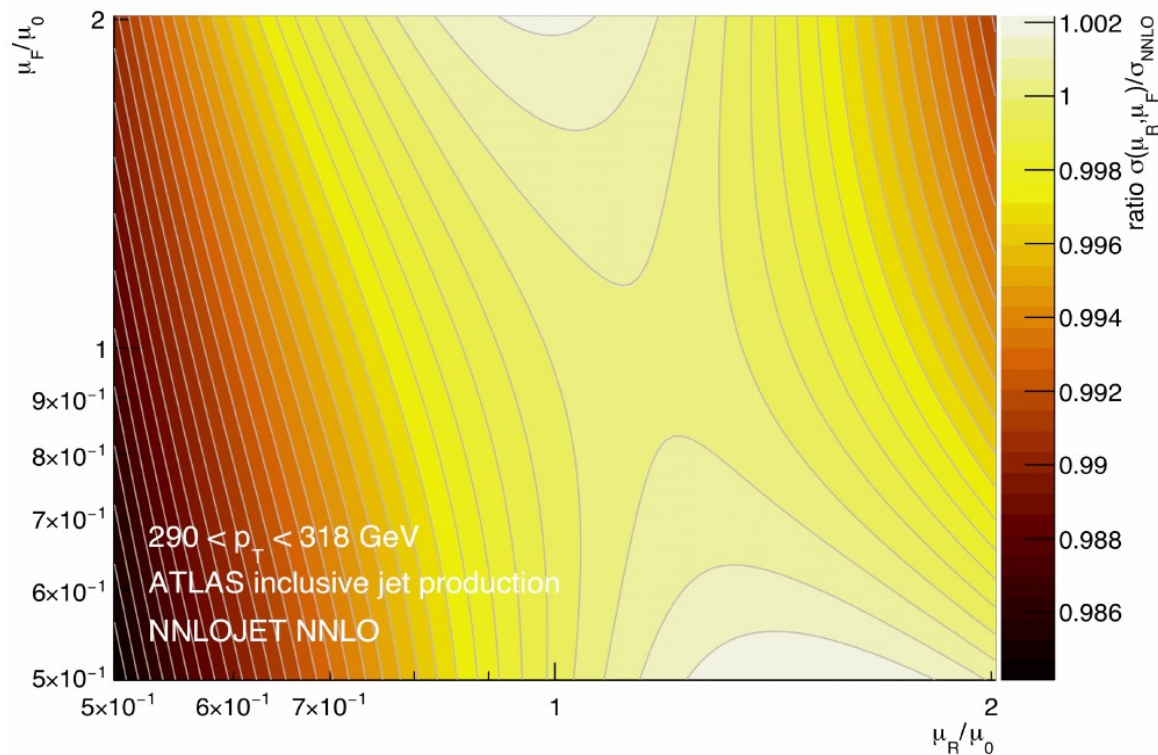


ATLAS inclusive jets at 7 TeV

Scale uncertainty bands: LO, NLO, NNLO



Full 2-dimensional scale dependence in $\mu_{F/R}$ for each bin



Alternative scale possible: p_{Tjet} instead of HT_{part}

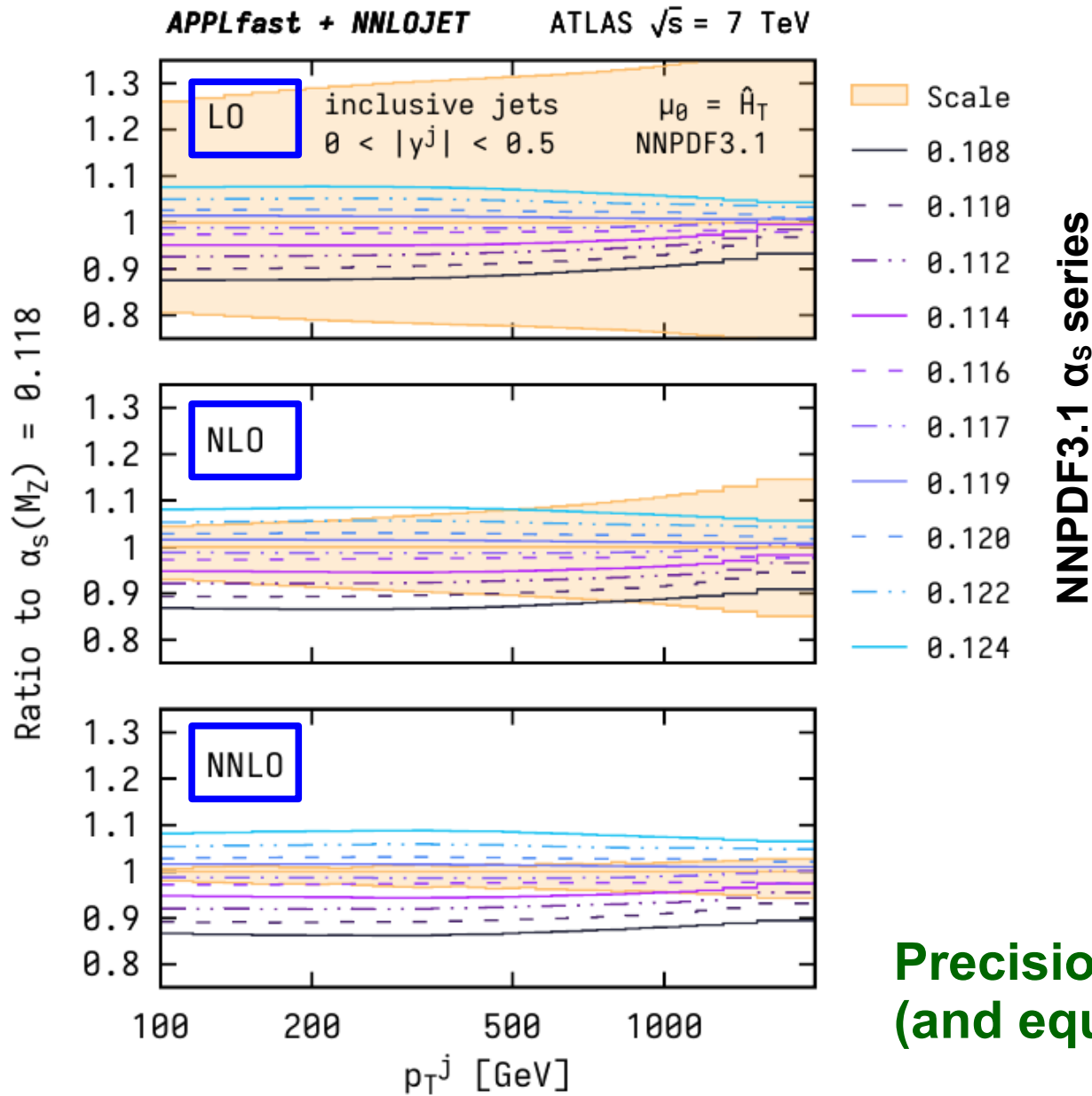
From Snowmass report: arXiv:2203.13923



α_s dependence



ATLAS inclusive jets at 7 TeV



Scale uncertainty versus α_s dependence
(0.108 – 0.124)
 at each order

Precision determinations require NNLO
 (and equally accurate data!)



PDF dependence



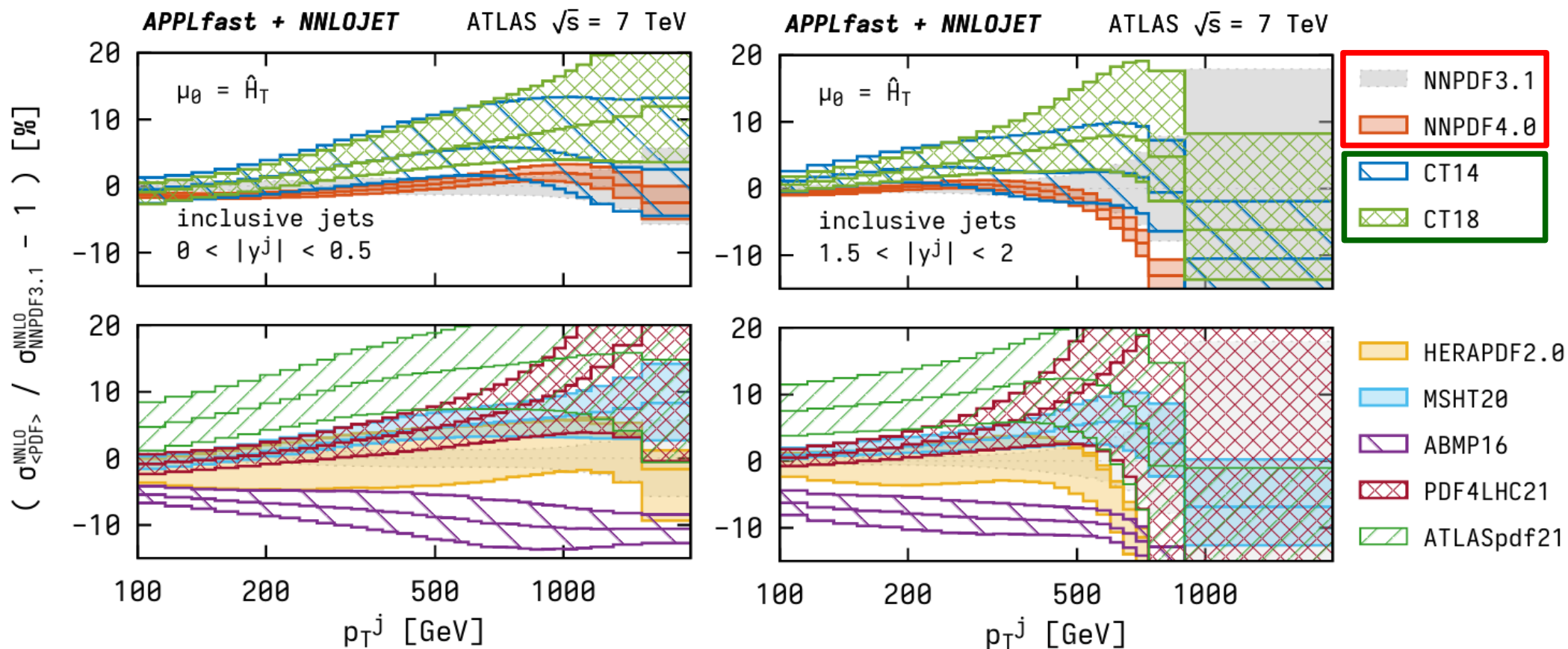
ATLAS inclusive jets at 7 TeV

PDF uncertainty bands for selection of 9 PDF sets

central rapidity

“Central” PDF: NNPDF3.1

medium rapidity



Interesting feature for Δ PDF: CT14 \rightarrow CT18 larger

NNPDF 3.1 \rightarrow 4.0 much smaller



Inclusive jet datasets



- Seven inclusive jet datasets from ATLAS & CMS, 2D in p_T and y
 - Four centre-of-mass energies, three jet radii R
 - Two central scales for $\mu_{R/F}$

**Sample plots
in this talk**

Data	\sqrt{s} [TeV]	\mathcal{L} [fb $^{-1}$]	no. of points	anti- k_T R	kinematic range [GeV]	fiducial cuts	$\mu_{R/F}$ -choice
CMS [30]	2.76	0.00543	81	0.7	$p_T^{\text{jet}} \in [74, 592]$	$ y < 3.0$	$p_T^{\text{jet}}, \hat{H}_T$
ATLAS [28]	7.0	4.5	140	0.6	$p_T^{\text{jet}} \in [100, 1992]$	$ y < 3.0$	$p_T^{\text{jet}}, \hat{H}_T$
CMS [31]	7.0	5.0	133	0.7	$p_T^{\text{jet}} \in [114, 2116]$	$ y < 3.0$	$p_T^{\text{jet}}, \hat{H}_T$
ATLAS [32]	8.0	20.3	171	0.6	$p_T^{\text{jet}} \in [70, 2500]$	$ y < 3.0$	$p_T^{\text{jet}}, \hat{H}_T$
CMS [33]	8.0	5.6 19.7	248	0.7	$p_T^{\text{jet}} \in [21, 74]$ $p_T^{\text{jet}} \in [74, 2500]$	$ y < 4.7$	$p_T^{\text{jet}}, \hat{H}_T$
ATLAS [34]	13.0	3.2	177	0.4	$p_T^{\text{jet}} \in [100, 3937]$	$ y < 3.0$	$p_T^{\text{jet}}, \hat{H}_T$
CMS [35]	13.0	36.3 33.5	2×78	0.4 0.7	$p_T^{\text{jet}} \in [97, 3103]$	$ y < 2.0$	$p_T^{\text{jet}}, \hat{H}_T$



Dijet datasets



- Four dijet datasets from ATLAS & CMS, 2D in m_{12} and y^* or y_{\max} , or 3D in $\langle p_{T12} \rangle$, y^* , y_b
- Three centre-of-mass energies, three jet radii R
- One central scale for $\mu_{R/F}$, except for 3D data with two

**Sample plots
in this talk**

Data	\sqrt{s} [TeV]	\mathcal{L} [fb ⁻¹]	no. of points	anti- k_T R	kinematic range [GeV]	fiducial cuts	$\mu_{R/F}$ -choice
ATLAS [55]	7.0	4.5	90	0.6	$m_{12} \in [260, 5040]$	$ y_1 , y_2 < 3.0$ $[p_{T,1}, p_{T,2}] > [100, 50]$ GeV $y^* < 3.0$	m_{12}
CMS [31]	7.0	5.0	54	0.7	$m_{12} \in [197, 5058]$	$ y < 5.0$ $[p_{T,1}, p_{T,2}] > [60, 30]$ GeV $ y_{\max} < 2.5$	m_{12}
CMS [49]	8.0	19.7	122	0.7	$\langle p_{T1,2} \rangle \in [133, 1784]$	$ y < 5.0$ $p_{T,1}, p_{T,2} > 50$ GeV $ y_1 , y_2 < 3.0$	$p_{T,1} \exp(0.3 y^*)$ m_{12}
ATLAS [34]	13.0	3.2	136	0.4	$m_{12} \in [260, 9066]$	$ y_1 , y_2 < 3.0$ $p_{T,1}, p_{T,2} > 75$ GeV $\langle p_{T1,2} \rangle > 100$ GeV $y^* < 3.0$	m_{12}