

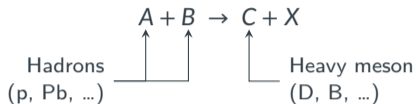
PineAPPL Grids of Open Heavy-Flavor Production in the GM-VFNS

CTEQ 2024 Spring Meeting

Jan Wissmann

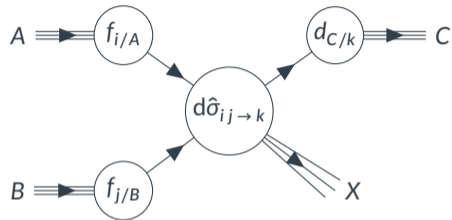
Introduction

- ▶ Process: Open heavy-quark hadroproduction



in collinear factorization:

$$d\sigma = f_{i/A} \otimes f_{j/B} \otimes d\hat{\sigma}_{ij \rightarrow k} \otimes d_{C/k}$$



- ▶ Light quarks q: u, d, s, heavy quarks Q: c, b
→ heavy on the absolute QCD scale: $m_Q \gg \Lambda_{\text{QCD}}$ so that the process is calculable perturbatively, i.e. $\alpha_s(m_Q) \ll 1$
- ▶ Importance of heavy-quark production: data goes to small momentum-fraction $x \approx \frac{p_T}{\sqrt{s}} e^y \sim 10^{-5}$
→ e.g. constrain gluon PDF in low-x region
- ▶ Mass effects non-negligible for $p_T \sim m_Q$
- ▶ Theory predictions for this process: GM-VFNS (NLO)

Outline

Part 1: General-mass variable-flavor-number schemes (GM-VFNS)

Part 2: Gridding with PineAPPL

Part 3: Results

Part 1: General-Mass Variable-Flavor-Number Schemes (GM-VFNS)

Flavor-number schemes

- ▶ In all processes with heavy quarks: new scale m_Q

threshold region

$$p_T \lesssim m_Q$$

FFNS

Fixed flavor-number scheme

- ▶ Heavy quark treated as massive particle, lighter quarks as massless partons
- ▶ Fixed number of light flavors
- ▶ Heavy-quark mass acts as a regulator

asymptotic region

$$p_T \gg m_Q$$

p_T

ZM-VFNS

Zero-mass variable-flavor-number scheme

- ▶ Heavy quarks treated as massless partons
- ▶ Number of light flavors is scale-dependent: Contributions from new flavors activate dynamically at their respective mass thresholds
- ▶ Collinear singularities due to massless quarks renormalized in the usual \overline{MS}

The General-Mass Variable-Flavor-Number Scheme

Expectation:

$$d\sigma_{\text{FFNS}} \xrightarrow{p_T \gg m_Q} d\sigma_{\text{ZM-VFNS}} \quad ?$$

→ $p_T \gg m_Q$ (i.e. $m_Q \rightarrow 0$) limit and subtraction of collinear singularities are not exchangeable

Solution: GM-VFNS:

$$d\sigma_{\text{FFNS}} \xleftarrow{m_Q \leftarrow p_T} d\sigma_{\text{GM-VFNS}} \xrightarrow{p_T \gg m_Q} d\sigma_{\text{ZM-VFNS}}$$

For intermediate p_T , the GM-VFNS interpolates between the ZM-VFNS and the FFNS.

Part 2: Gridding with PineAPPL

Gridding with PineAPPL

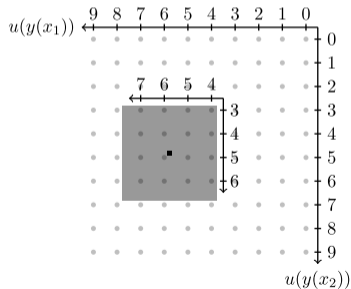
For theoretical predictions obtained with a Monte Carlo (MC) generator:

Problem:

- ▶ A-posteriori variation of α_s , scales and PDFs requires running the MC generator again each time (usually multiple hours per run)
- ▶ Same calculation of the hard-scattering matrix elements is performed every time

Solution:

- ▶ Pre-calculate the MC weights and store them in an interpolation “grid” independent of the PDFs, α_s and possibly scales
- Done by libraries such as
 - ▶ FastNLO [hep-ph/0609285]
 - ▶ APPLgrid [0911.2985]
 - ▶ PineAPPL [2009.03987]



[2009.03987]

Gridding with PineAPPL

QCD factorization (conceptually):

$$d\sigma = \int dx f(x) d\hat{\sigma}(x)$$

Gridding libraries store the MC weights of fixed-order calculations by interpolating the PDFs:

$$f(x) = \sum_i f_i L_i(x) \quad \Rightarrow \quad d\sigma = \sum_i f_i \underbrace{\int dx L_i(x) d\hat{\sigma}(x)}_{\substack{\text{can be precomputed} \\ \text{and stored}}} =: \sum_i f_i d\sigma_i$$

↑ Lagrange
basis functions

(analogous for α_s and scales)

Using a Monte-Carlo integrator:

$$d\sigma_i = \int dx L_i(x) d\hat{\sigma}(x) \stackrel{\text{MC}}{=} \frac{1}{N} \sum_{j=1}^N L_i(x^{(j)}) d\hat{\sigma}(x^{(j)})$$

Gridding with PineAPPL

1. Gridding Stage

- ▶ Build grid by integrating over the basis functions:

$$d\sigma_i = \int dx L_i(x) d\hat{\sigma}(x)$$

- ▶ In practice: Fill the grid with the MC weights, obtain grid file
- Takes multiple CPU hours, but only done once

2. Convolution Stage

- ▶ Obtain predictions by performing the convolution with the PDF:

$$d\sigma = \sum_i f_i d\sigma_i$$

- ~Instantaneous, can be done multiple times for different PDFs and scales

Gridding GM-VFNS heavy-quark hadroproduction

- ▶ The NLO GM-VFNS calculation in heavy-quark hadroproduction exists as Fortran code by B. A. Kniehl, G. Kramer, I. Schienbein, H. Spiesberger [hep-ph/0410289] [hep-ph/0502194] [hep-ph/0508129]
- ▶ **Our work:** Extending the existing code to produce PineAPPL grids and writing a Python interface to the code to make it publication-ready

NEW: produced NLO GM-VFNS predictions as PineAPPL grids

One grid, corresponding to one experimental dataset, includes cross-sections...

- ▶ double-differential in (p_T, y) corresponding to the bins of the experimental data
 - ▶ at LO (α_s^2) and NLO (α_s^3)
 - ▶ with the FF baked-in (Since PineAPPL allows up to two different convolutions at the moment)
- both PDFs and α_s can be varied a-posteriori, e.g. for PDF uncertainties or fits

PLANNED:

- ▶ Using these grids for nCTEQ PDF analyses in the future
- ▶ Publication of the grids and this version of the GM-VFNS code

Data taken into account so far – ALICE

Predictions and grids already produced for:

Experiment	arXiv	Initial State	Meson
ALICE	1111.1553	p + p	D^0
	1405.3452	p + Pb	D^0
	1605.07569	p + p	D^0
		p + Pb	D^+
	1702.00766	p + Pb	D^0
	1901.07979	p + p	D^0
	1906.03425	p + Pb	D^+
			D_s^+
			D^{*+}
2106.08278	p + p	D^0	

Fragmentation functions:

D^0, D^+, D^{*+} : KKKS08 [0712.0481]
 D_s^+ : BKK06_D [hep-ph/0607306]
 B^+ : BKK06_B [0705.4392]

Data taken into account so far – LHCb & CMS

Predictions and grids already produced for:

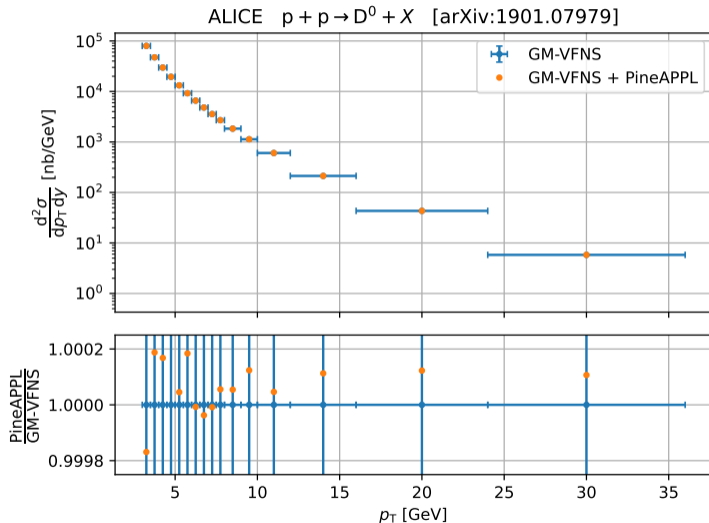
Experiment	arXiv	Initial State	Meson
CMS	1508.06678	p + Pb	B^+
LHCb	1302.2864	p + p	D^0
	1510.01707	p + p	D^0
	1610.02230	p + p	D^0
	1707.02750	p + Pb	D^0
	2205.03936	p + Pb	D^0

Fragmentation functions:

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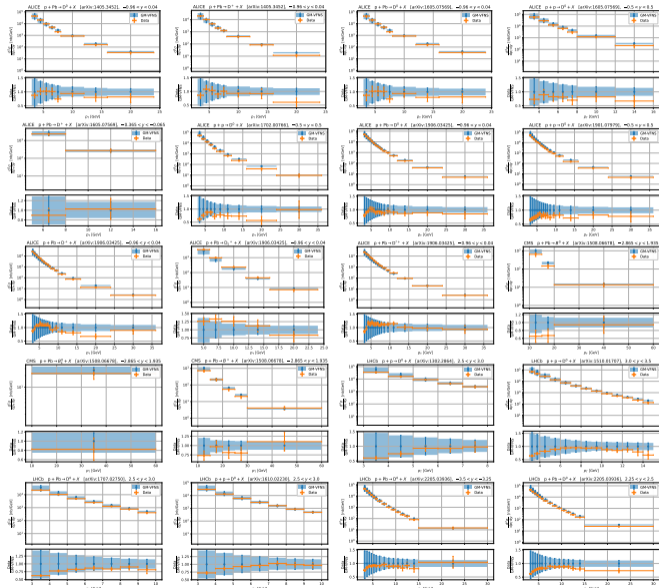
Part 3: Results

Results I – Prediction vs. Grid



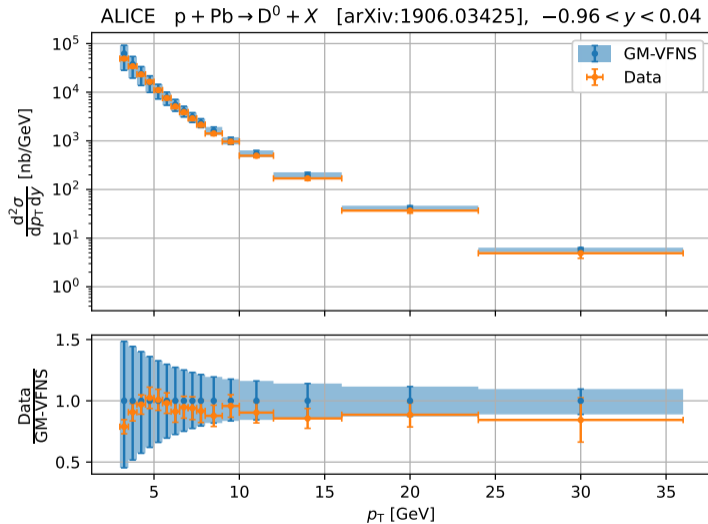
- ▶ Sub-permille agreement
- ▶ Shown here: Statistical (MC) errors
- ▶ Grid precision independent of run settings and phase space region

Results II – Prediction vs. Data



► Next slides: Some examples of the predictions

Results II – Prediction vs. Data



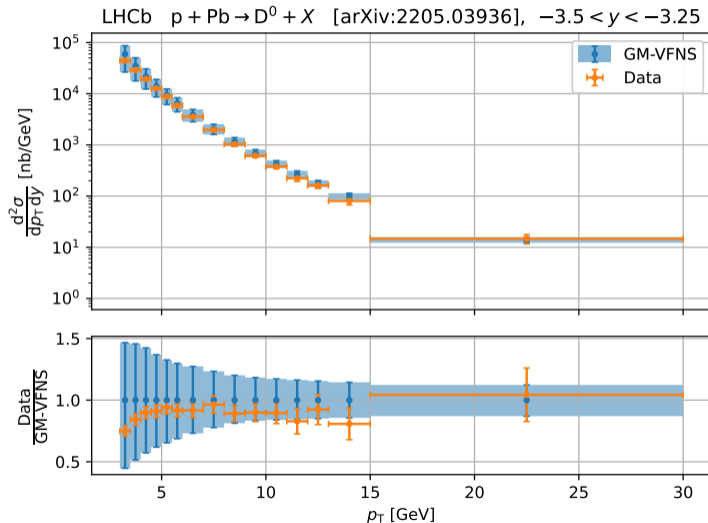
- ▶ 7-point (ξ_r, ξ_f) scale-variation

where $\mu_i = \xi_i \sqrt{p_T^2 + 4m_Q^2}$
and $\xi_i \in \{0.5, 1, 2\}$

- ▶ Previous work:
Scale choice improves agreement and enables meaningful predictions at lower p_T

[1907.12456]

Results II – Prediction vs. Data



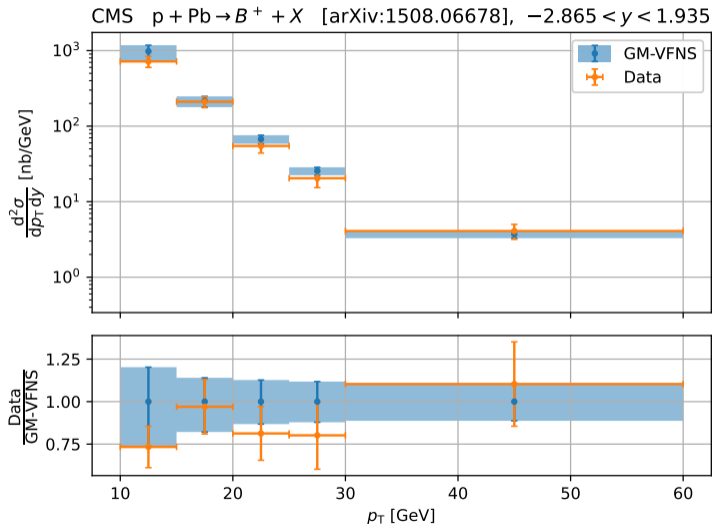
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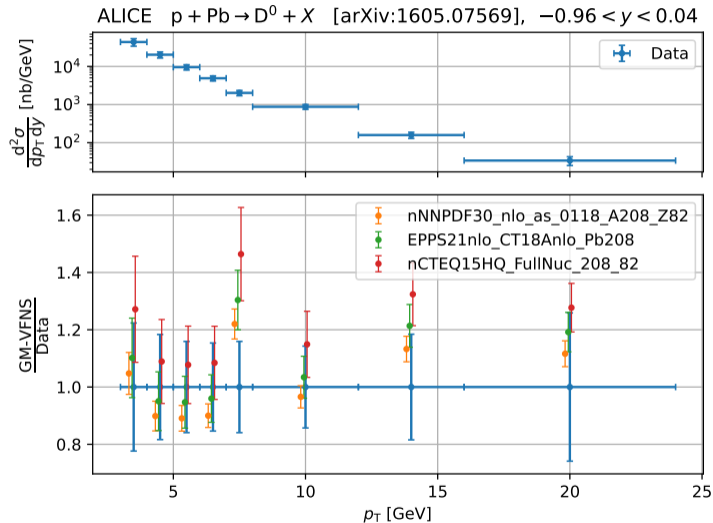
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Results III – PDF Uncertainties

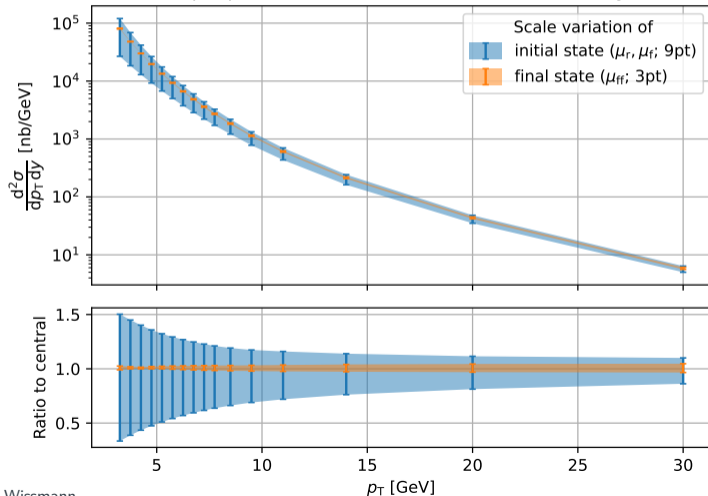


- ▶ Total number of PDF members: 347
- ▶ Here: Gridding reduces execution time by factor 347 → impossible without gridding

Results IV – Fragmentation Scale

$$d\sigma = f_{i/A}(\mu_f) \otimes f_{j/B}(\mu_f) \otimes d\hat{\sigma}_{i_j \rightarrow k}(\mu_r) \otimes d_{C/k}(\mu_{ff})$$

ALICE $p + p \rightarrow D^0 + X$ [arXiv:1901.07979], $-0.5 < y < 0.5$



► 7-point (ξ_r, ξ_f) scale-variation

where $\mu_i = \xi_i \sqrt{p_T^2 + 4m_Q^2}$
and $\xi_i \in \{0.5, 1, 2\}$

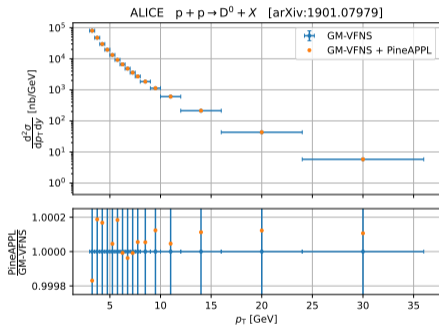
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[1907.12456]

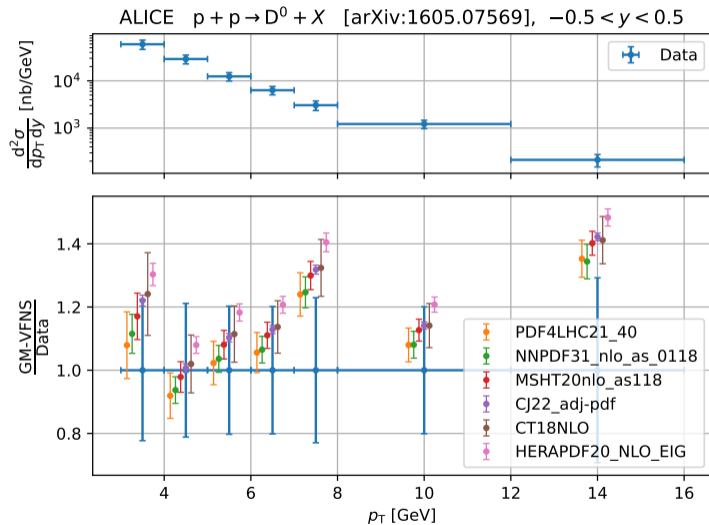
Conclusion

- ▶ The GM-VFNS gives the heavy-quark production prediction for a bigger kinematic range
- ▶ Gridding libraries like PineAPPL allow varying the PDFs and scales a-posteriori in a very efficient way
- ▶ The GM-VFNS prediction and the produced grids agree by less than one per-mille
- ▶ This version of the GM-VFNS code and the PineAPPL grids will be published



Backup

Results III – PDF Uncertainties



- ▶ Total number of PDF members: 342
- ▶ Here: Gridding reduces execution time by factor 342 → impossible without gridding