

# Determination of $\alpha_s(M_Z)$

## in a fit to inclusive jet data from multiple experiments

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D. Britzger, K. Rabbertz, **D. Savoiu**, G. Sieber, M. Wobisch

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INSTITUT FÜR EXPERIMENTELLE KERNPHYSIK



## Why $\alpha_s(M_Z)$ ?

- $\alpha_s(M_Z)$  among least well known fundamental physical parameters
  - increased knowledge of  $\alpha_s(M_Z)$  needed for precision QCD

## How?

- use **inclusive jet cross sections** in hadron-induced collisions
  - ← jets abundantly produced at hadron colliders
  - ← directly sensitive to  $\alpha_s(M_Z)$
  - ← well defined observable for any process
- theory calculations standard: NLO
  - NNLO just around the corner

## Objective

- develop a robust procedure to estimate  $\alpha_s(M_Z)$ 
  - inclusion of more than one dataset
  - flexible and consistent treatment of theory (preparation for NNLO)

# Main “ingredients”

## Data

- abundance of inclusive jet data from collider experiments:

ATLAS, CDF, CMS, DØ, H1, STAR, ZEUS...

→ available for data/theory comparisons and  $\alpha_s$  determinations

- inclusive jet measurement

→ phase space, experimental uncertainties

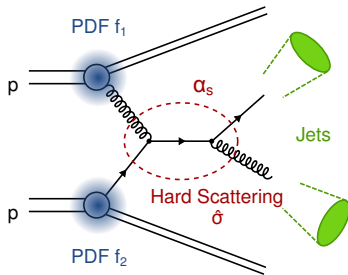
## Theory

- hard matrix element  $\hat{\sigma}$

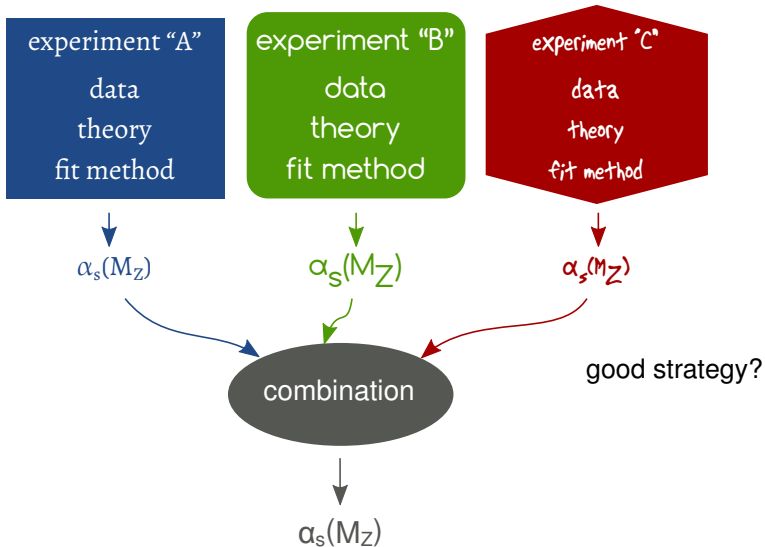
→ sensitive to  $\alpha_s(M_Z)$

- convolution with PDFs

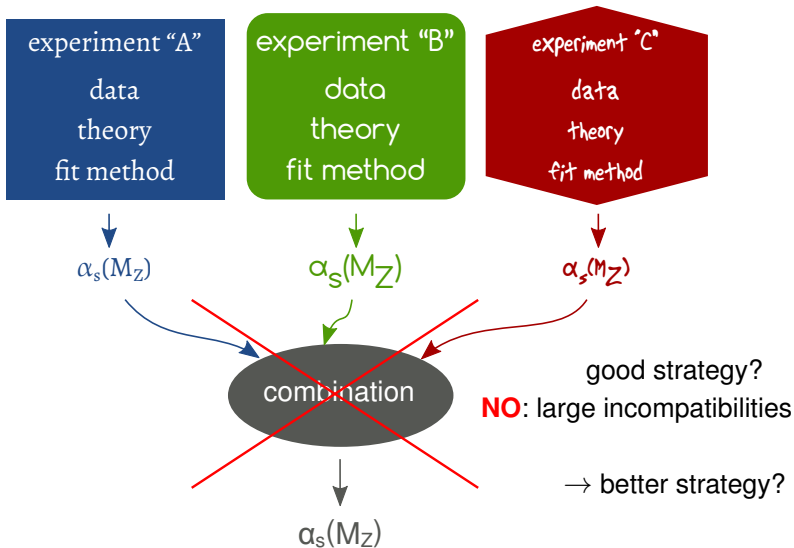
→  $\alpha_s(M_Z)$  dependence

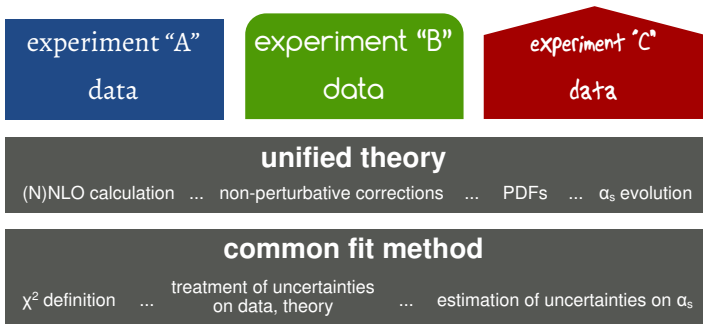


# Strategy?



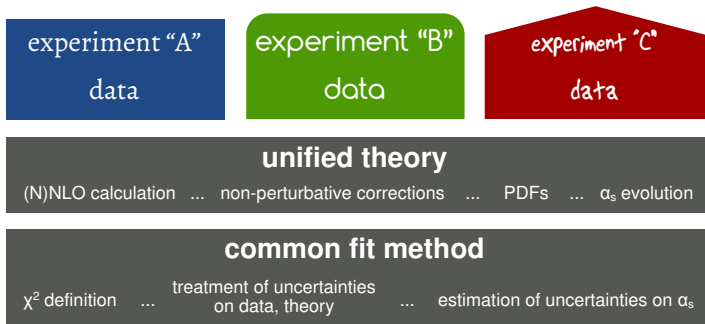
# Strategy?





↓

$$\alpha_s(M_Z)$$



$$\alpha_s(M_Z)$$

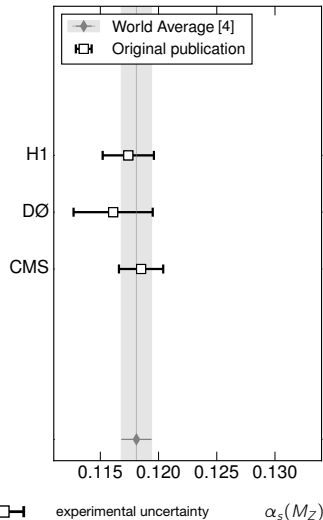
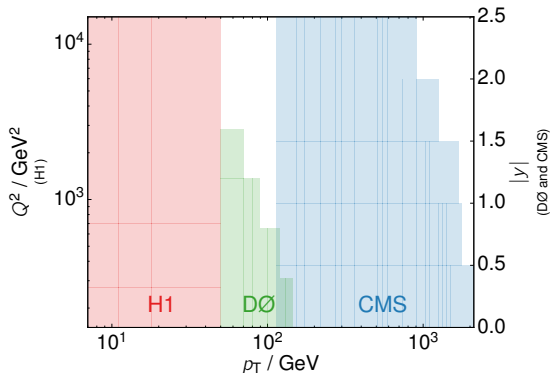
✓ clean

✓ consistent

✓ convincing

# First look: $\alpha_s(M_Z)$ at CMS, DØ and H1

- recent  $\alpha_s$  extractions by CMS [3], DØ [1] and H1 [2]
- data from cover large and complementary phase space





# Comparison of fit setups

	H1	DØ	CMS
theory predictions	NLO	approximate NNLO	NLO
$\alpha_s(M_Z)$ extraction procedure	direct $\chi^2$ minimization	direct $\chi^2$ minimization	“indirect” $\chi^2$ minimization (fit of parabola to discrete $\chi^2$ points)
$\chi^2$ definition	conventional $\chi^2$ (log data – log theory) + relative uncertainties	modified $\chi^2$ + nuisance parameters	conventional $\chi^2$ (data – theory) + absolute uncertainties
uncertainty estimation	linear error propagation	nuisance parameters	<ul style="list-style-type: none"><li>• “<math>\Delta\chi^2 = +1</math>”</li><li>• subtraction in quadrature</li><li>• “offset” method</li></ul>

■ ■ ■

## Fit methods **differ significantly!**

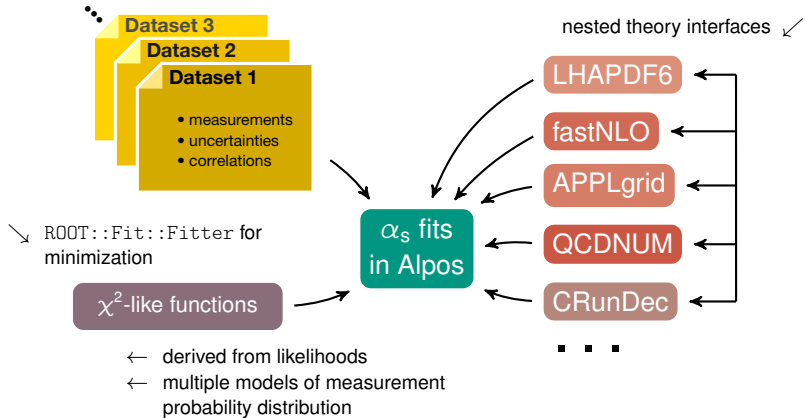
- “naive” combination of results (weighted average) not conclusive
- need to extract  $\alpha_s(M_Z)$  using measurements from all experiments in a **unified** fit procedure

# New fitting tool – Alpos

- **Alpos** → new modular C++ based fitting framework

- used within H1 and CMS for PDF, electroweak and  $\alpha_s(M_Z)$  fits

↘ input format: experience with xFitter/HERAFitter

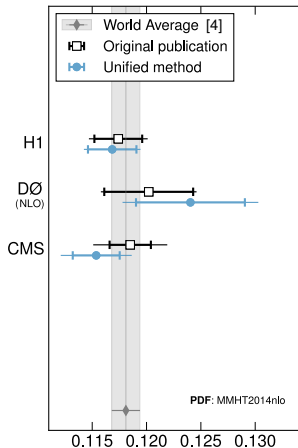




[<https://ekptrac.ekp.kit.edu/svn/Alpos>]

# Common fit method

theory predictions	consistent (N)NLO
$\alpha_s(M_Z)$ extraction procedure	direct $\chi^2$ minimization
$\chi^2$ definition	conventional $\chi^2$ ( $\log$ data – $\log$ theory) + relative uncertainties
PDF and non-perturbative uncertainties	included in $\chi^2$ definition
PDF $\alpha_s(M_Z)$ dependence	additional uncertainty on $\alpha_s(M_Z)$

- applying common method yields different  $\alpha_s(M_Z)$  for the same dataset
  - results remain in agreement with original publication
- $\alpha_s(M_Z)$  values consistent due to common method



 experimental uncertainty  $\alpha_s(M_Z)$   
 total uncertainty (except scale)

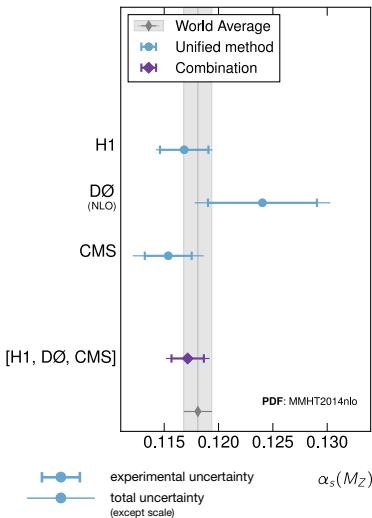
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## Result

$$\alpha_s(M_Z) = 0.1172 (15)_{\text{exp}} (14)_{\text{theo}} (\text{except scale}) (50)_{\text{scale}}$$

$$[\chi^2_{\text{min}}/\text{ndf} = 152.2/178 = 0.855]$$



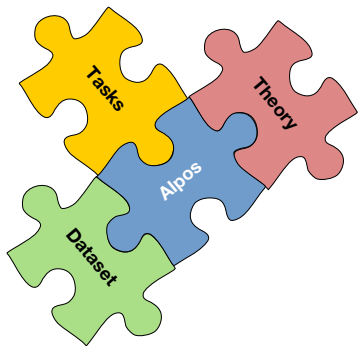
- developed method of extracting  $\alpha_s(M_Z)$  from multiple inclusive jet data sets
- **Alpos**: recently developed, public fitting code
- applied to data from the H1, CMS and DØ experiments
  - consistent results
  - agreement with world average for  $\alpha_s(M_Z)$
  - reduced experimental uncertainties
- scale uncertainties remain the largest single contribution to the total uncertainty

## Outlook

- switch to NNLO accuracy
- include data from ATLAS, CDF, STAR, ZEUS. . .
- use insights gained for more complex QCD studies with data from multiple experiments

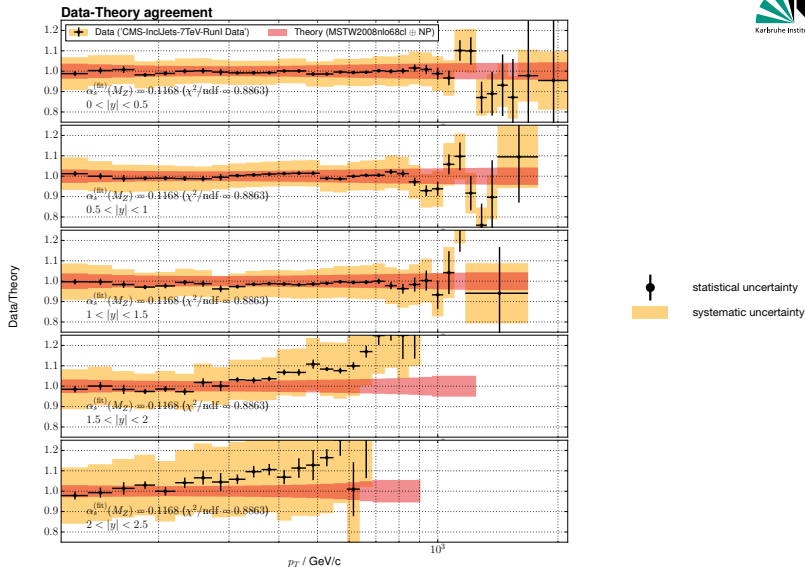
- [1] V.M. Abazov et al. “Determination of the strong coupling constant from the inclusive jet cross section in  $p\bar{p}$  collisions at  $\sqrt{s} = 1.96$  TeV”. In: *Phys. Rev. D* 80 (2009), p. 111107. DOI: 10.1103/PhysRevD.80.111107. arXiv: 0911.2710 [hep-ex].
- [2] V. Andreev et al. “Measurement of multijet production in  $ep$  collisions at high  $Q^2$  and determination of the strong coupling  $\alpha_s$ ”. In: *Eur. Phys. J. C* 75 (2015), p. 65. DOI: 10.1140/epjc/s10052-014-3223-6. arXiv: 1406.4709 [hep-ex].
- [3] Vardan Khachatryan et al. “Constraints on parton distribution functions and extraction of the strong coupling constant from the inclusive jet cross section in pp collisions at  $\sqrt{s} = 7$  TeV”. In: *Eur. Phys. J. C* 75 (2015), p. 288. DOI: 10.1140/epjc/s10052-015-3499-1. arXiv: 1410.6765 [hep-ex].
- [4] K.A. Olive and others (Particle Data Group). “Review of Particle Physics”. In: *Chin. Phys. C* 38 (2014). (2015 update), p. 090001. DOI: 10.1088/1674-1137/38/9/090001.

Many thanks to G. Flouris and P. Kokkas for their insights, as well as for providing the original CMS fitting code



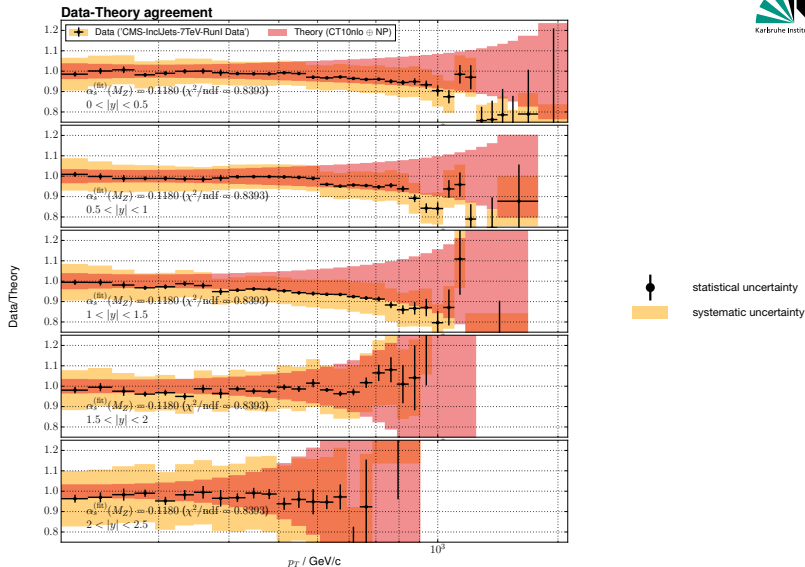
# Backup

## References

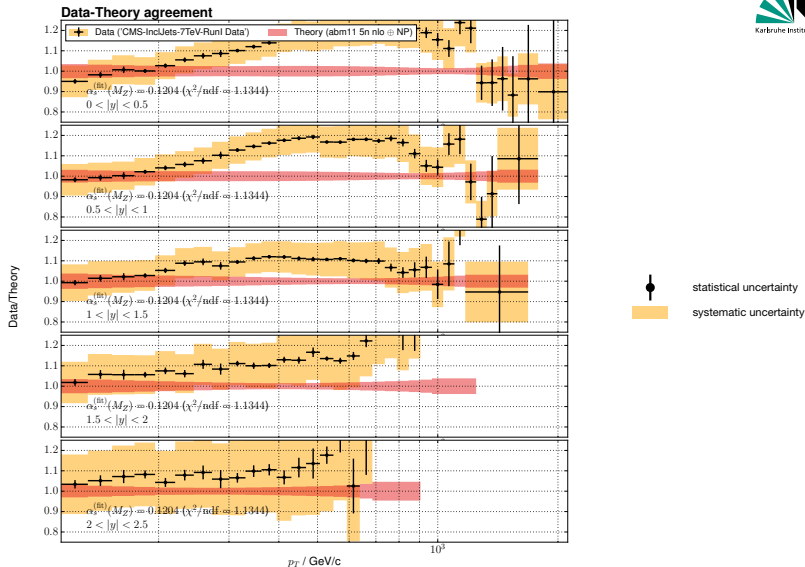


#### References

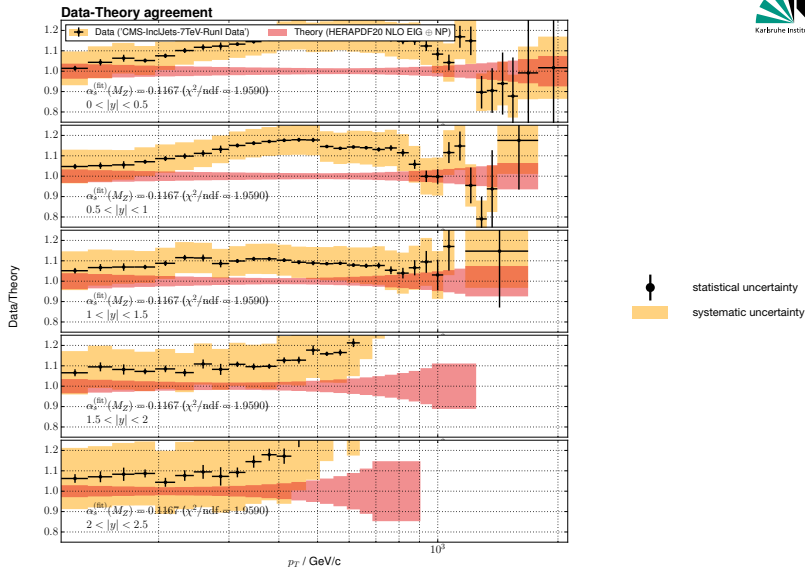




#### References



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# Comparison of $\alpha_s(M_Z)$ extraction methods

## H1 fit methodology

$$\chi_{H1}^2 \rightarrow \sum_{ij} (\ln m_i - \ln t_i) [\mathbf{V}_{(rel)}^{-1}]_{ij} (\ln m_j - \ln t_j)$$

- iterative  $\chi^2$  minimization (*MINUIT*)
  - determine central value with **experimental uncertainties** only
  - assume PDF without  $\alpha_s(M_Z)$  dependence; use MSTW2008nlo with  $\alpha_s(M_Z) = 0.118$
- additional theory uncertainties: **NP corr.**, **PDF**, PDF  $\alpha_s(M_Z)$ , PDF set,  $\mu_r$ ,  $\mu_f$ 
  - obtained through additional fits / linear error propagation

## DØ fit methodology

$$\chi_{DØ}^2 \rightarrow \sum_i \left[ \frac{m_i - t_i \frac{1 + \sum_k \delta_{ik}^{(NP)} (\alpha_k^{(NP)}) + \sum_l \delta_{il}^{(PDF)} (\alpha_l^{(PDF)})}{1 + \sum_j \delta_{ij} (\epsilon_j)}}{\sigma_{i,stat}^2 + \sigma_{i,uncorr}^2} \right]^2$$

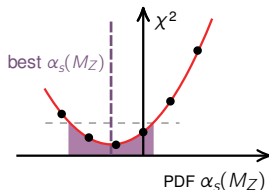
- iterative  $\chi^2$  minimization (*MINUIT*)
  - one nuisance parameter for **each PDF eigenvector** and **each NP correction factor**
  - interpolate cross section predictions obtained for PDFs assuming different values of  $\alpha_s(M_Z)$
- aNNLO (NLO predictions with threshold corrections + NNLO PDFs)
- 88 out of 110 data points excluded  $\leftarrow$  correlations with MSTW2008 PDFs

# Comparison of $\alpha_s(M_Z)$ extraction methods

## CMS fit methodology

$$\chi_{\text{CMS}}^2 \rightarrow \sum_{ij} (m_i - t_i) [(V_{\text{exp}} + V_{\text{PDF}})^{-1}]_{ij} (m_j - t_j)$$

- $\chi^2$  is evaluated for each PDF in an  $\alpha_s(M_Z)$  series
  - resulting  $(\chi^2, \alpha_s(M_Z))$  points are assumed to lie on a parabola
  - fit of second-degree polynomial function  $\rightarrow$  central value and uncertainty on  $\alpha_s(M_Z)$
- PDF: CT10nlo (results are also provided for MSTW2008 and NNPDF21)
- NP uncertainties obtained by performing additional fits with correlated variation of theory



## Fit methods **differ significantly!**

- $\rightarrow$  “naive” combination of results (weighted average) not very conclusive
- $\rightarrow$  need to extract  $\alpha_s(M_Z)$  using measurements from all experiments in a **unified** fit procedure

# PDF $\alpha_s(M_Z)$ dependence

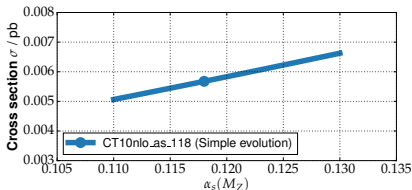
How to express the  $\alpha_s(M_Z)$  dependence of the cross section?

- ❓ how to account for  $\alpha_s(M_Z)$  dependence in PDFs?

Two methods are studied:

## “Fixed PDF” (○)

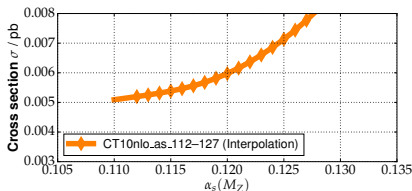
(method used in H1 publication)



- calculate cross section using PDF for **one** chosen  $\alpha_s(M_Z)$
- prediction  $\sigma(\alpha_s(M_Z))$  directly from fastNLO

## “PDF Interpolation” (◇)

(method used in DØ publication)



- calculate cross section using PDF for **each** available  $\alpha_s(M_Z)$
- prediction  $\sigma(\alpha_s(M_Z))$  from **interpolation** between the points

## “Fixed PDF”

- +  $\alpha_s(M_Z)$  dependence is **quadratic**, as expected for inclusive jet cross sections
- + well-defined theory
- + clear breakdown of PDF uncertainties
- introduces an additional procedural uncertainty due to  $\alpha_s(M_Z)$  used in PDF fit
- possible bias towards assumed PDF  $\alpha_s(M_Z)$

## “PDF Interpolation”

- + provides a way to include the uncertainty due to the choice of PDF  $\alpha_s(M_Z)$  **in the fit**
- o interpolation method needs to be defined (e.g. fit or splines)
  - ▲ spline interpolation not well suited for some PDFs (e.g. NNPDF)
- spline extrapolation may give unphysical results
- ? does procedure reproduce PDF  $\alpha_s$  dependence?

In most cases, both methods yield comparable results

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- possible bias towards assumed PDF  $\alpha_s(M_Z)$

choose this as main method

## “PDF Interpolation”

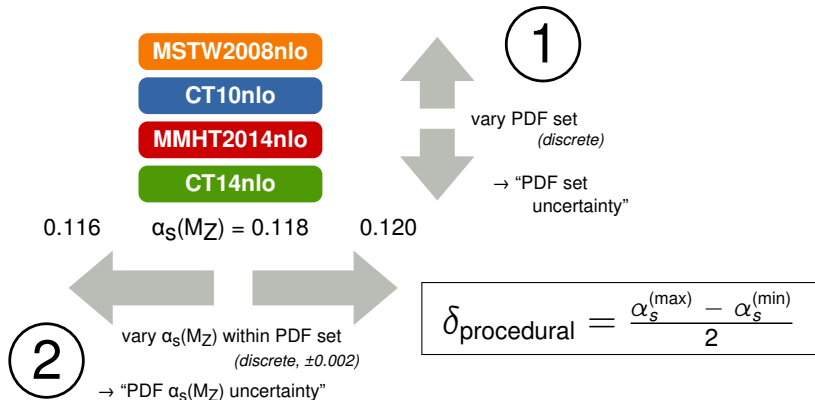
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- ? does procedure reproduce PDF  $\alpha_s$  dependence?

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# Procedural PDF uncertainties

- additional PDF-related procedural uncertainties arise in addition to “PDF uncertainties” themselves:
  - ① choice of PDF set
  - ② choice of  $\alpha_s(M_Z)$  assumed when fitting PDF



# Refits with unified procedure

## Refit results

- all refits have reasonable  $\chi^2/\text{ndf}$  and compatible with each other

- $\chi^2/\text{ndf}$  values at minimum:

$$\text{H1} \quad \chi^2/\text{ndf} = 23.1/23 = 1.004$$

$$\text{D}\emptyset \quad \chi^2/\text{ndf} = 17.2/21 = 0.819$$

$$\text{CMS} \quad \chi^2/\text{ndf} = 110/132 = 0.832$$

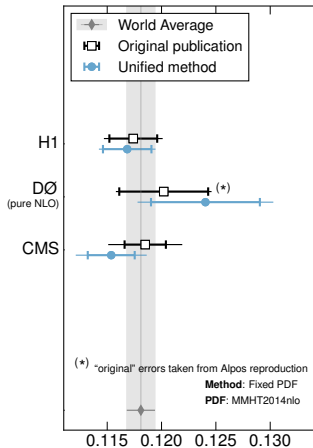
## Comparison to published values

- good agreement

H1 method similar, but  $V_{\text{PDF}}$  and  $V_{\text{NP}}$  in  $\chi^2$

CMS, D $\emptyset$  change of fitting method  $\rightarrow$  changes of fit values and uncertainties

Final results remain comparable with each other and with original publication  
 $\rightarrow$  Can proceed with combination



—•— experimental uncertainty  $\alpha_s(M_Z)$   
—•— total uncertainty (except scale)