

# Uncertainty of the Uncertainty of $\alpha_s$

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Kirtimaan Mohan @Alphas-2025,  
Savoie, France Dec 2025

## CTEQ-TEA

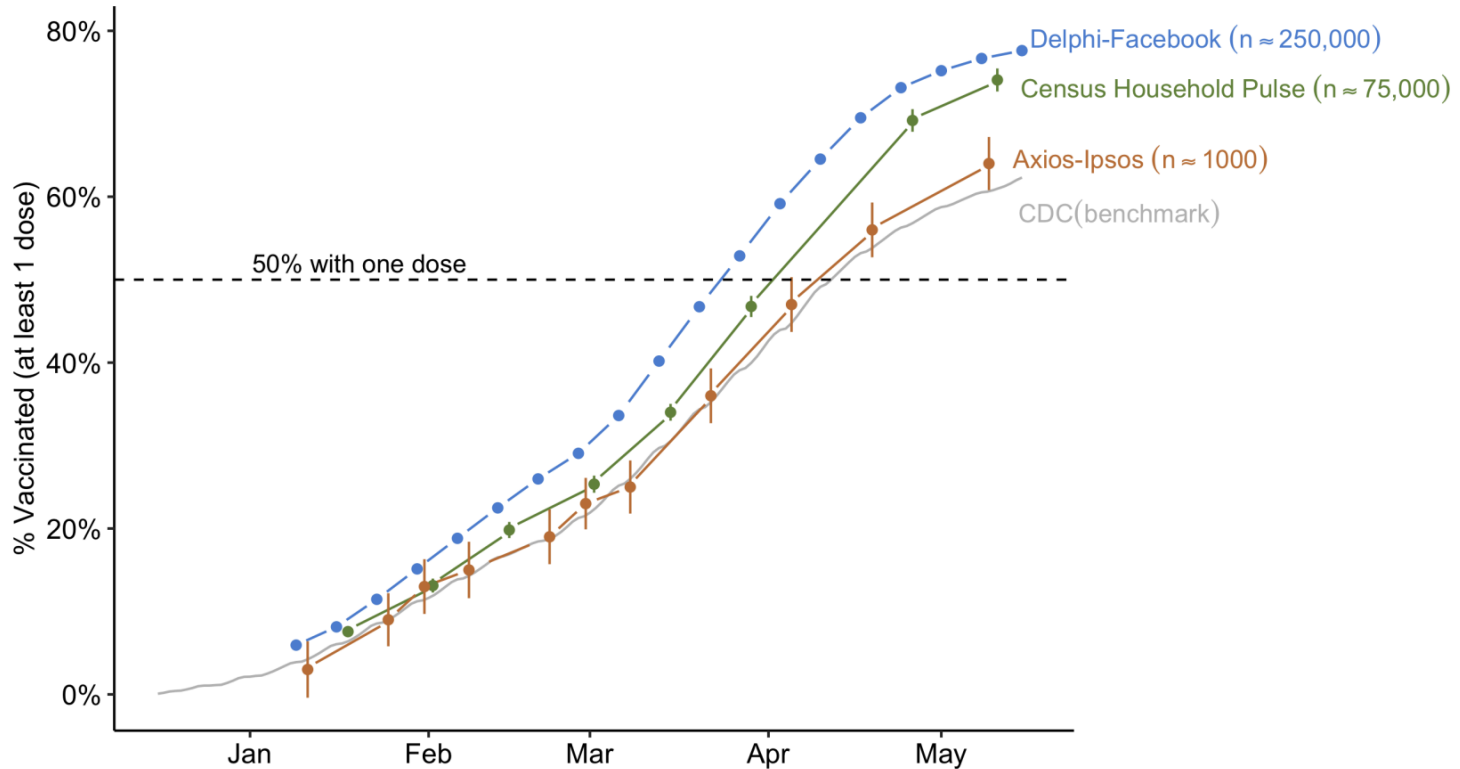
**Asia:** A. Ablat, S. Dulat, T.-J. Hou, I. Sitiwaldi

**North America:** A. Courtoy, Y. Fu, M. Guzzi,  
T.J. Hobbs, J. Huston, K. Mohan, H.-W. Lin, P.  
Nadolsky, M. Ponce-Chavez, K. Xie, C.-P.  
Yuan



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# Big Data Paradox: *The bigger the data, the surer we fool ourselves, if we do pay attention to data quality*



[Slide from Xiao Li Meng, PHYSTAT 2025](#)

# Uncertainty Quantification



- **Aleatoric** (“dicey”) Uncertainty:
  - Statistical uncertainty in data that is reduced by improving data quantity and quality
- **Epistemic** (“knowledge”) Uncertainty:
  - Due to lack of knowledge, which can introduce bias. Improved through better (or at least) representative modelling. Aka Systematic uncertainty.

[Hullermeier & Waegeman \(2021\)](#)

# This Talk

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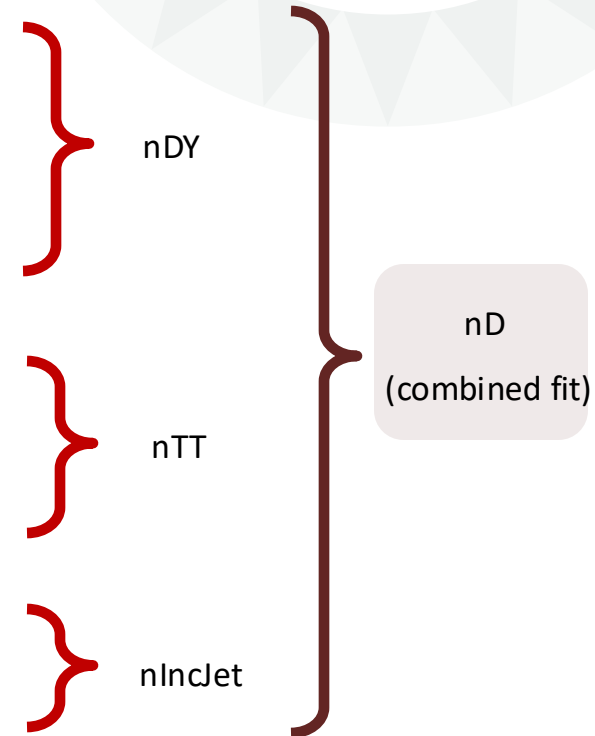
Present challenges of quantifying uncertainty  
preliminary results of the CTEQ-TEA's  
simultaneous determination of the strong coupling  
and PDFs using updated data sets.

# NNLO fits with new data (nD) from LHC at 8 and 13 TeV

$\chi^2/N_{pt}$  for CT18A + new data (vs. CT18A at right) NNLO fits; 68% CL

ID	Experimental data set	$N_{pt}$	CT25prel	CT18A
<b>Drell-Yan pair production</b>				
211	ATLAS 8 TeV W	22	$2.37^{+1.30}_{-0.68}$	$3.56^{+2.26}_{-1.65}$
212	CMS 13 TeV Z	12	$2.10^{+2.20}_{-0.38}$	$2.13^{+3.46}_{-0.30}$
214	ATLAS 8 TeV Z 3D	188	$1.14^{+0.10}_{-0.04}$	$1.21^{+0.32}_{-0.15}$
215	ATLAS 5.02 TeV W,Z	27	$0.70^{+0.27}_{-0.06}$	$0.74^{+0.31}_{-0.08}$
217	LHCb 8 TeV W	14	$1.36^{+0.37}_{-0.34}$	$1.47^{+0.43}_{-0.38}$
218	LHCb 13 TeV Z	16	$1.06^{+0.76}_{-0.38}$	$1.29^{+0.95}_{-0.44}$
<b>13 TeV <math>t\bar{t}</math> production</b>				
521	ATLAS all-hadronic $y_{t\bar{t}}$	12	$1.07^{+0.08}_{-0.05}$	$1.07^{+0.12}_{-0.07}$
528	CMS dilepton $y_{t\bar{t}}$	10	$1.10^{+0.56}_{-0.40}$	$1.13^{+0.85}_{-0.53}$
581	CMS lepton+jet $m_{t\bar{t}}$	15	$1.38^{+0.65}_{-0.40}$	$1.44^{+0.89}_{-0.56}$
587	ATLAS lepton+jet $m_{t\bar{t}} + y_{t\bar{t}} + y_{t\bar{t}}^B + H_T^{t\bar{t}}$	34	$0.94^{+0.13}_{-0.11}$	$0.94^{+0.28}_{-0.09}$
<b>Inclusive jet production</b>				
553	ATLAS 8 IncJet	171	$1.54^{+0.09}_{-0.06}$	$1.57^{+0.12}_{-0.07}$
554	ATLAS 13 IncJet	177	$1.25^{+0.07}_{-0.03}$	$1.26^{+0.08}_{-0.04}$
555	CMS 13 IncJet	78	$1.11^{+0.13}_{-0.09}$	$1.10^{+0.21}_{-0.10}$

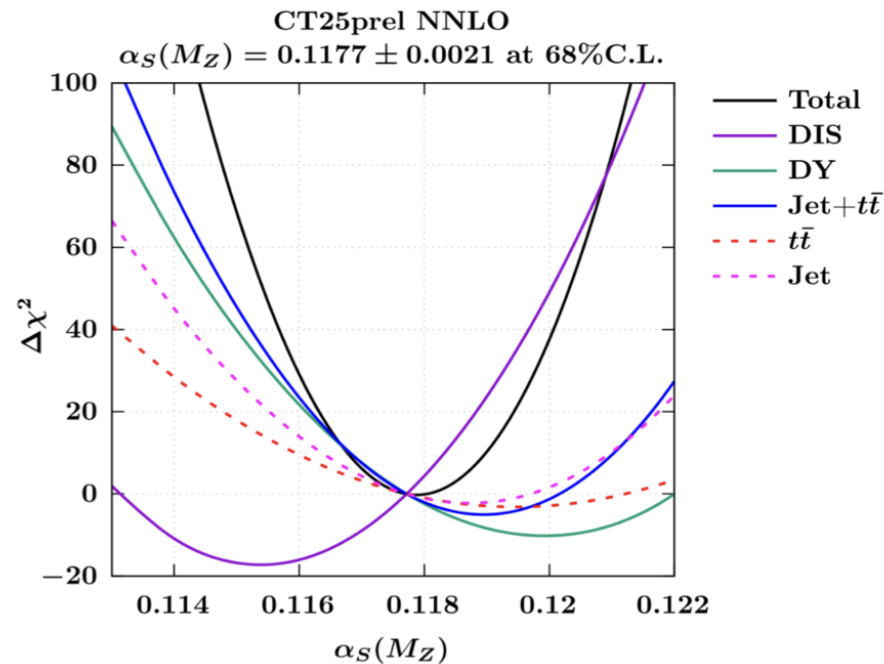
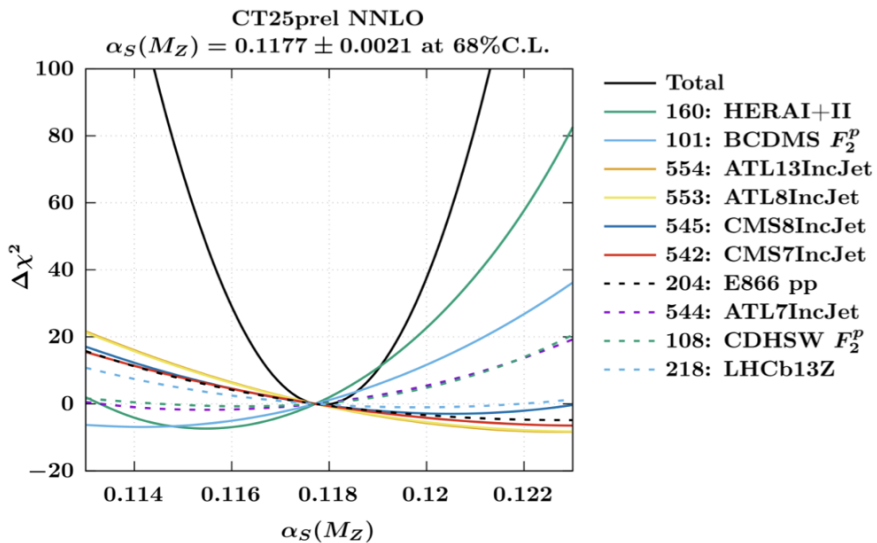
(fits with 1 new process, 'nProces')



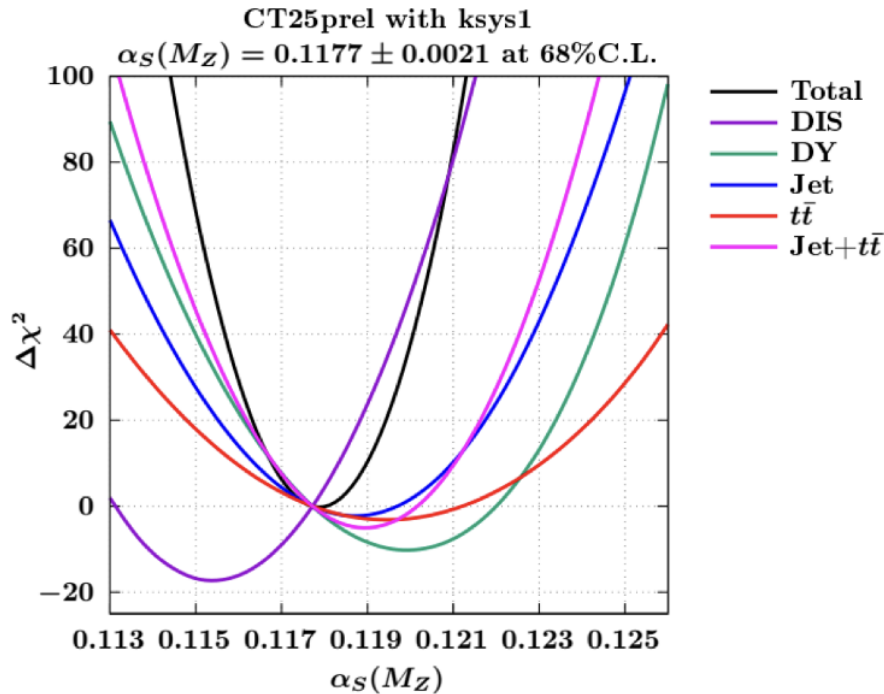
# Impact of new data on fits

Updates from the upcoming CT25 NNLO fits

- Significant pulls on  $\alpha_s$  from ATLAS Incl. Jets [553, 554] and 13 TeV LHCb Z data [218] and  $t\bar{t}$  production data
- Large tension between DIS, DY and Jet+  $t\bar{t}$



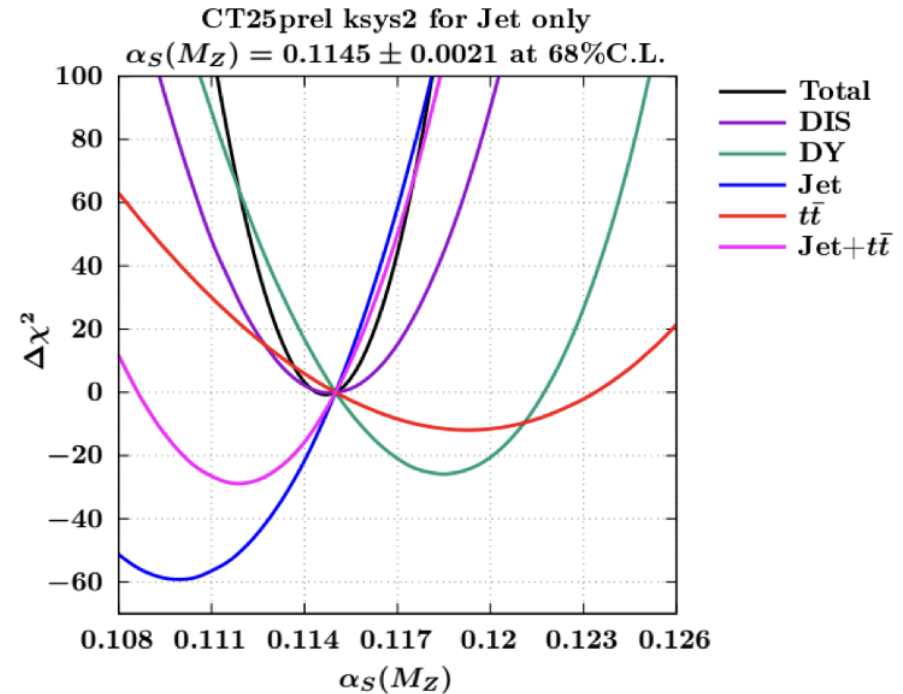
# Sensitivity to treatment of systematics



Left: **multiplicative errors** for all data sets

$$\alpha_S(M_Z) \approx 0.118$$

... but possible theory bias in syst. effects



Right: **additive errors** for jet data sets

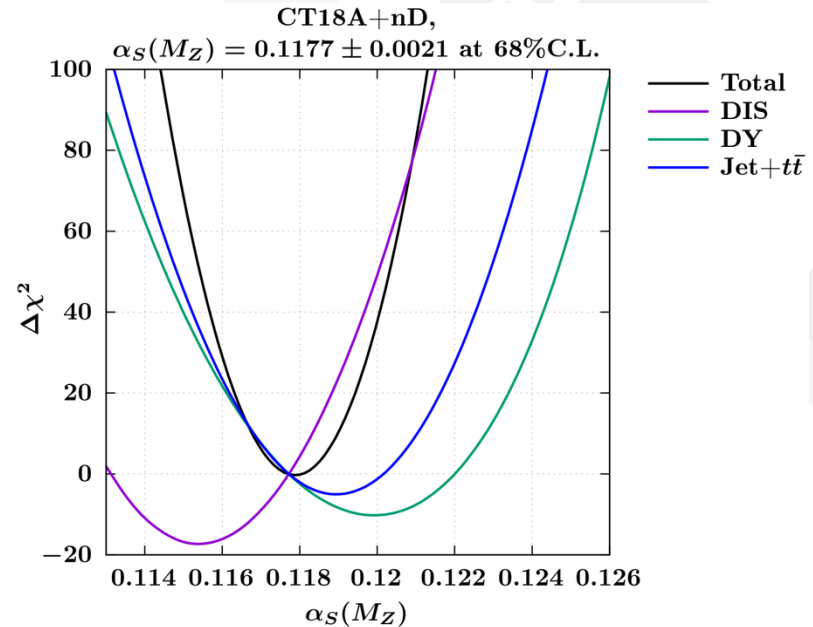
$$\alpha_S(M_Z) \approx 0.115$$

... likely reflects D'Agostini's bias – cf. also a similar shift in the 2025 NNPDF4.0  $\alpha_S$  study

The truth is within a range between the extremes

# Tolerance for Tolerating Tension

- It is easy to see that there is a large tension between DIS and other data sets.
- Treat each of the sets (DIS, DY, Jet +  $t\bar{t}$ ) as independent and identically distributed measurements of  $\alpha_s$
- $\chi_{tot}^2 = \sum_i \chi_i^2$
- Mean and Error given by minimizing
- $\tilde{\chi}^2 = \sum_i \frac{(\alpha_{s_i} - \bar{\alpha}_s)^2}{\sigma_i^2}$
- $\bar{\alpha}_s = \sigma_{tot}^2 \sum_i \frac{\alpha_{s_i}}{\sigma_i^2}$ ,  $\frac{1}{\sigma_{tot}^2} = \sum_i \frac{1}{\sigma_i^2}$
- **Large Tension:**  $\frac{\tilde{\chi}^2}{dof} \simeq 17$
- Yet small uncertainty:
  - $\alpha_s = 0.1179 \pm 0.000329$
  - (Profile Likelihood)



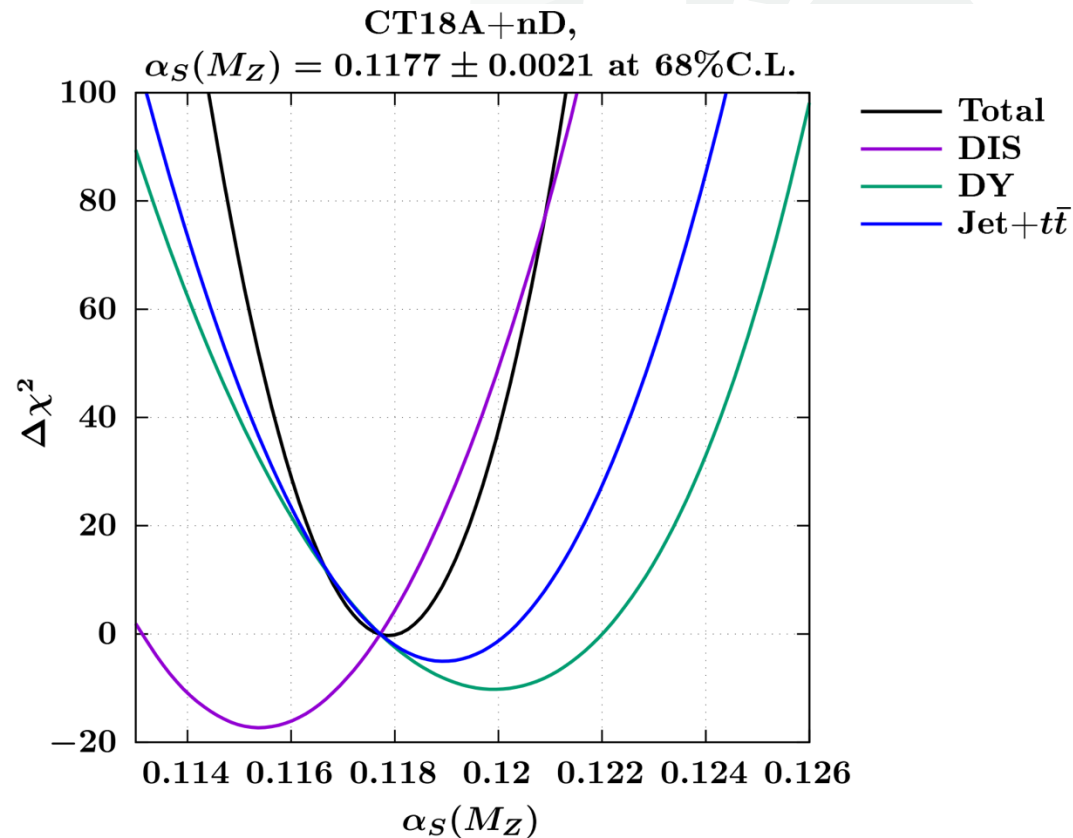
	DIS	DY	Jet + $t\bar{t}$	$\chi_{tot}^2$
$\alpha_s \times 10^3$ at min of $\chi_i^2$	115.88	119.91	118.94	<b>117.88</b>
Error ( $\sigma_i \times 10^3$ ) ( $\Delta\chi_i^2 = 1$ )	0.553	0.655	0.539	<b>0.329</b>

# How to proceed?

What should we do when we don't know how to proceed?

Form committees!

# Committee #1: Global Tolerance



Use global Tolerance:

$$\Delta\chi^2 = T^2$$

$$\alpha_s(M_Z) = 0.1177 \pm 0.0021 \text{ at } 68\% \text{ CL}$$

How should we justify the choice of Tolerance here?

# Committee #2: Global Tolerance through Effective Gaussian variable

$$S_n = \sqrt{2\chi^2(N_n)} - \sqrt{N_n - 1}$$

Criteria determined by values of  $S_n \approx 0.468$   
for 68% CL for the total  $\chi^2$   
Note: Related to Quantile of  $\chi^2$

$$\alpha_s(M_Z) = 0.1179 + 0.0024 - 0.0025$$

# Committee #3: Dynamical Tolerance

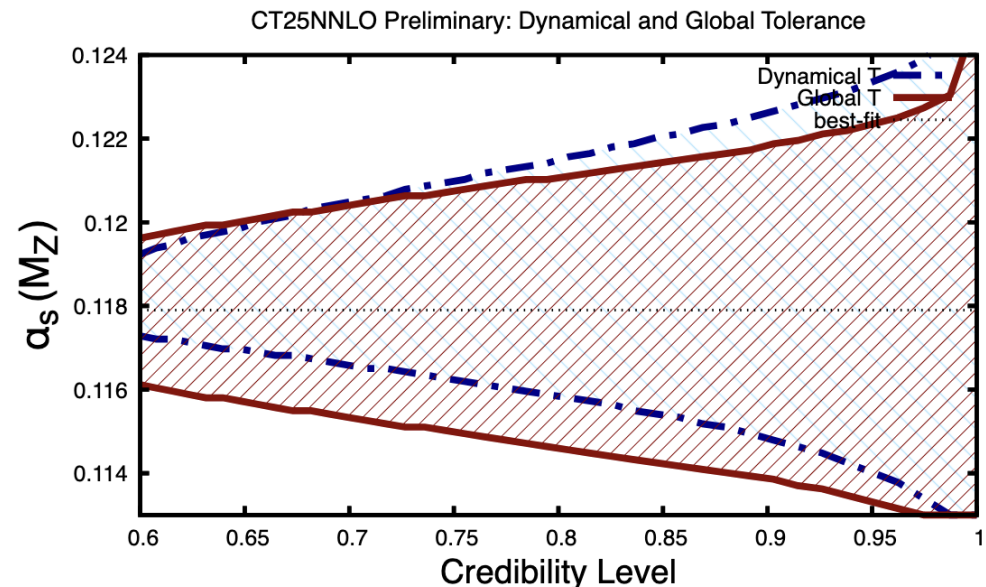
$$\chi_E^2(a, \lambda) = \sum_{k=1}^{N_{pt}} \frac{1}{s_k^2} \left( D_k - T_k(a) - \sum_{\alpha=1}^{N_\lambda} \beta_{k\alpha} \lambda_\alpha \right)^2 - \sum_{\alpha=1}^{N_\lambda} \lambda_\alpha^2,$$

Dynamic Tolerance  
determine by using Lewis  
formula for 52 individual  
experiments

$$\alpha_s(M_Z) = 0.1179 + 0.0024 - 0.0012$$

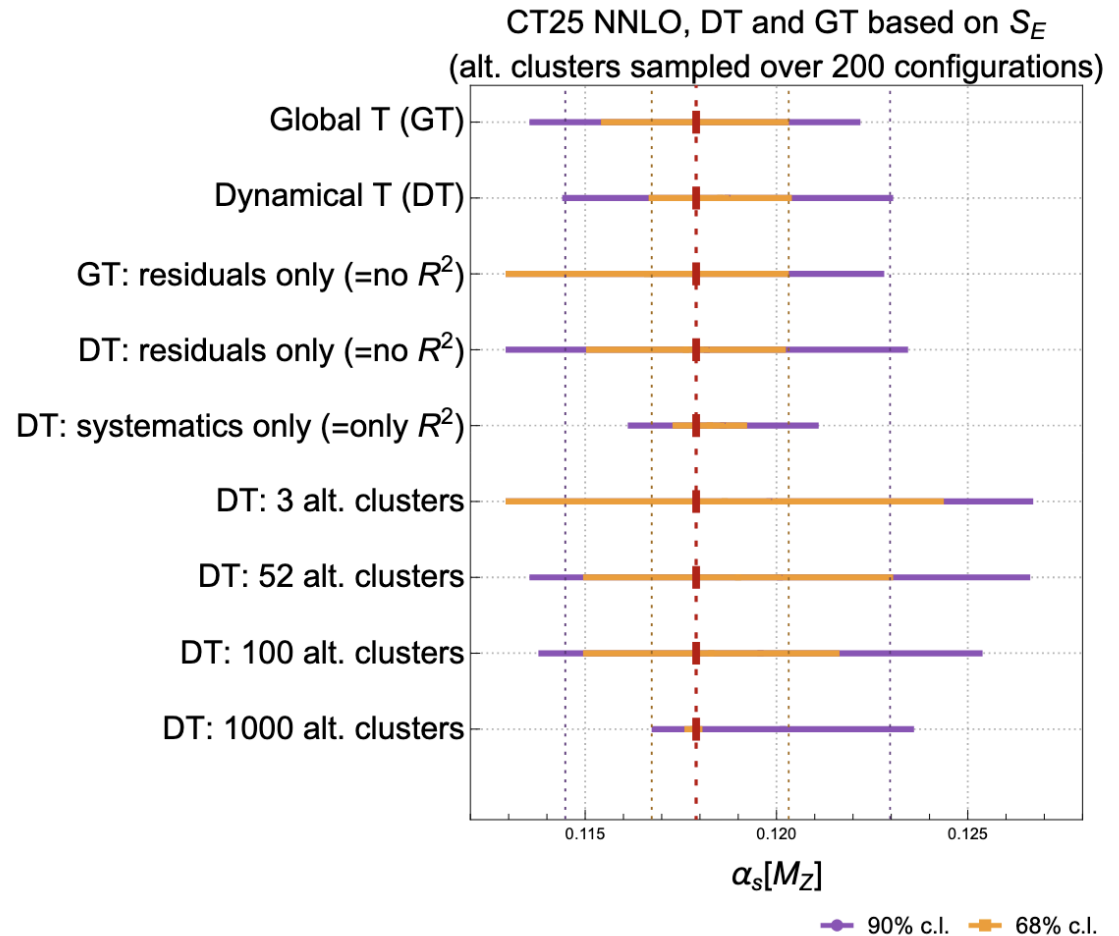
Comparison of  
Credibility level for  
Global Tolerance and  
Dynamic Tolerance

Dynamic Tolerance  
shows dependence  
one how data is  
partitioned



# Sensitivity to partitioning of data sets

In an ideal scenario, in which all residuals are normally distributed according to  $N(0,1)$ , alternative clusters would produce approximately the same estimate for the uncertainty.



# Committee #4: PDG Prescription: Scaling Errors

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- **PDG proposal:** scale errors by a factor  $e_s$  to make fits more consistent, i.e. each  $\sigma_i \rightarrow e_s \sigma_i$
- $e_{SPDG} = \sqrt{\frac{\tilde{\chi}^2}{dof}} \simeq 4.1$  so that each  $\sigma_i \rightarrow 4.1 \times \sigma_i$  and  $\frac{\tilde{\chi}^2}{dof} \rightarrow 1$
- **Caveat:** For very large  $\sqrt{\frac{\tilde{\chi}^2}{dof}}$ , PDG recommends making an educated guess of the uncertainty rather than scaling the errors.
- $\bar{\alpha}_s = 0.11795 \pm 0.00135$

# Sensitivity to partitioning of data sets

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- $N = 3$
- $\alpha_s = 0.1179 \pm 0.000329.$
- $\frac{\tilde{\chi}^2}{dof} \simeq 17$
- $\bar{\alpha}_s = 0.11795 \pm 0.00135$
- $N = 52$
- $\alpha_s = 0.1181 \pm 0.000334.$
- $\frac{\tilde{\chi}^2}{dof} \simeq 3.1$
- $\bar{\alpha}_s = 0.1181 \pm 0.0006$

# Committee #5: The “Error on the error”

$$\ln L(\boldsymbol{\mu}, \boldsymbol{\theta}, \sigma_{\mathbf{u}}^2) = \ln P(\mathbf{y}|\boldsymbol{\mu}, \boldsymbol{\theta})$$

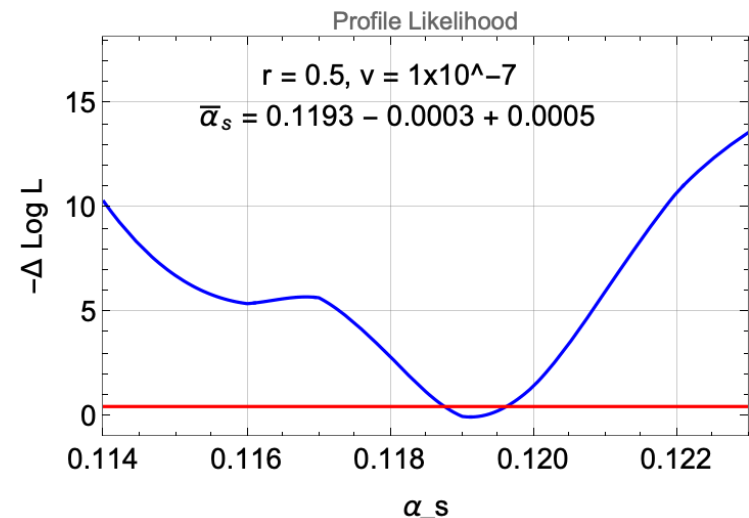
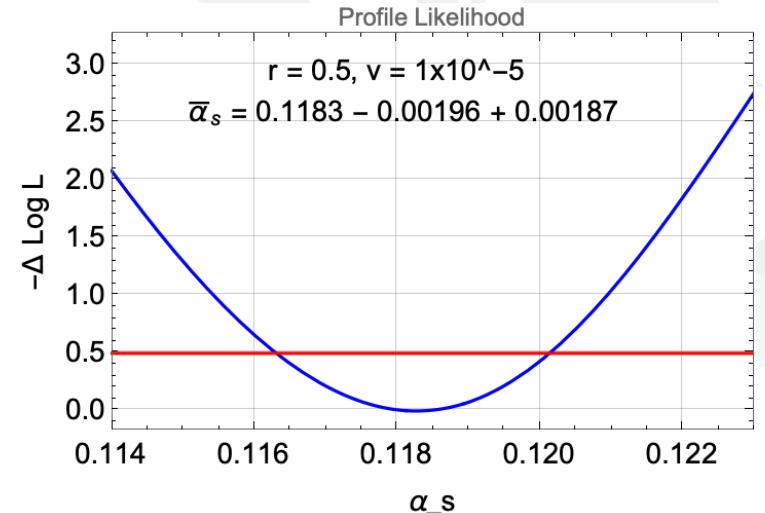
$$-\frac{1}{2} \sum_{i=1}^N \left[ \frac{(u_i - \theta_i)^2}{\sigma_{u_i}^2} + \left( 1 + \frac{1}{2r_i^2} \right) \ln \sigma_{u_i}^2 + \frac{v_i}{2r_i^2 \sigma_{u_i}^2} \right]$$

$$P(\mathbf{y}|\boldsymbol{\mu}, \boldsymbol{\theta}) = \prod_{i=1}^N \frac{1}{\sqrt{2\pi}\sigma_{y_i}} e^{-(y_i - \varphi(x_i; \boldsymbol{\mu}) - \theta_i)^2 / 2\sigma_{y_i}^2}$$

Needs as input an estimate of the uncertainty on the systematic uncertainty characterized by  $r$ .

Sensible values of  $r$  need to be chosen by analyst, making an educated guess.

[G. Cowan arXiv:1809.05778](https://arxiv.org/abs/1809.05778), see also  
[M. Reader arxiv.org:2408.12922](https://arxiv.org/abs/2408.12922)



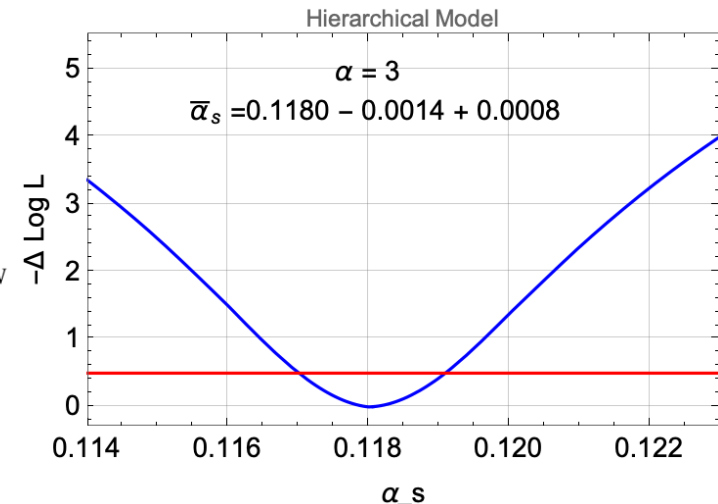
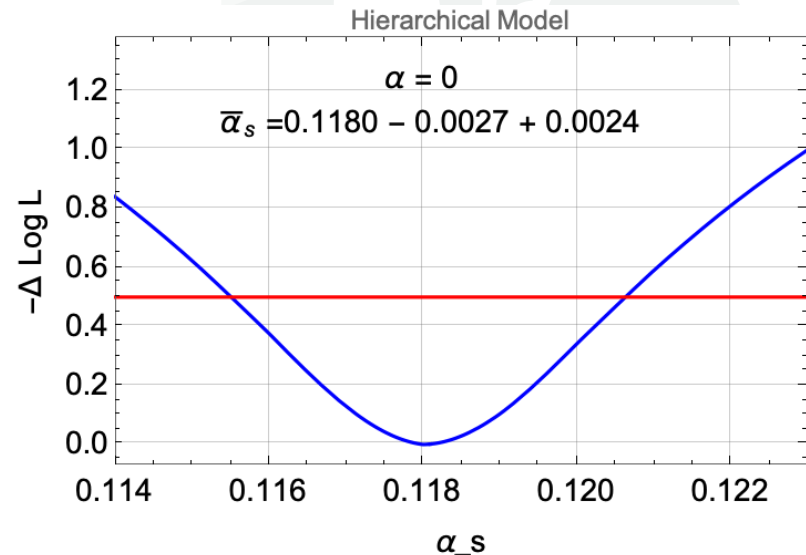
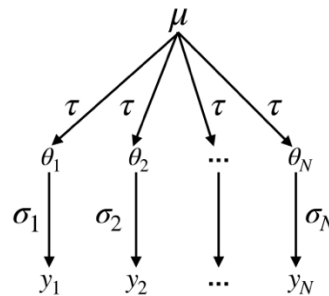
# Committee #6: Bayesian Hierarchical Model

$$p(\mu|y_i) \propto \int_0^\infty \prod_{i=1}^N (\sigma_i^2 + \tau^2)^{-\frac{1}{2}(1+\frac{\alpha}{N})} e^{-\frac{(\mu-y_i)^2}{2(\sigma_i^2+\tau^2)}} d\tau^2.$$

Depends on choice of  $\alpha$ :

$0 \leq \alpha \leq 3$  for large unknown systematics.

Use large values of  $\alpha$  if quoted uncertainties are trusted.

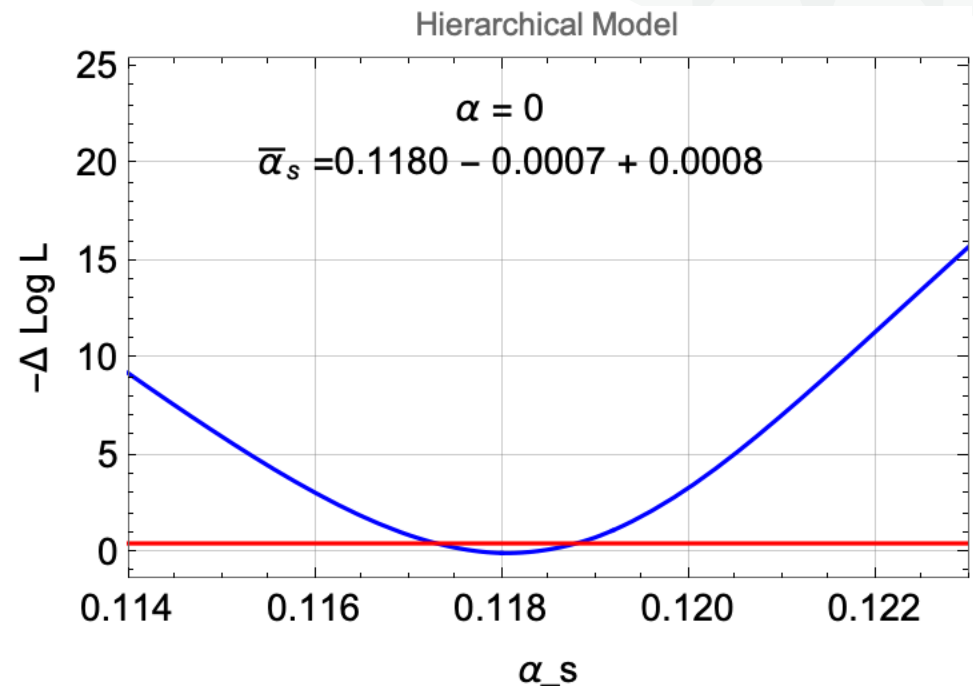


# Impact of Clustering

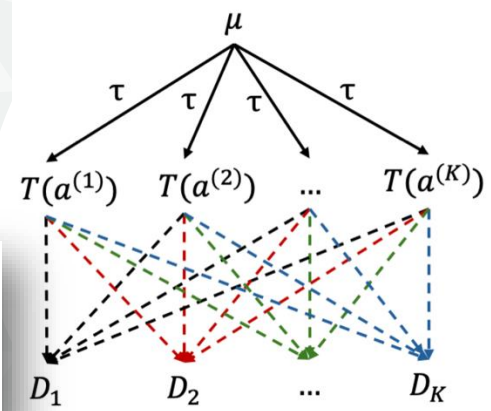
$N=52$  ,  $e_{SPDG} \approx 1.8$

More consistent fit?

Tracks PDG error  
scaling factor



# Gaussian Mixture Model



1. Modified Posterior

$$\prod_{i=1}^{N_D} \left( \sum_{k=1}^K P(T(a^{(k)})|D_i) \right) \propto \prod_{i=1}^{N_D} \left( \sum_{k=1}^K \omega_k \mathcal{N}(D_i|T(a^{(k)}), \sigma_i) \right)$$

2. Implementation via MLE of Mixture of Gaussians

$$\pi(Y|\vec{\theta}) = \prod_{j=1}^{N_{pt}} \pi(y_j, \Delta y_j|\vec{\theta}) = \prod_{j=1}^{N_{pt}} \sum_{i=1}^K \omega_i \mathcal{N}(y_j, \Delta y_j|\theta_i),$$

$$0 \leq \omega_k \leq 1 \quad \text{and} \quad \sum_k \omega_k = 1,$$

3. Calculate Mean

$$\mathbb{E}[\theta] = \sum_{i=1}^K \omega_i \hat{\theta}_i.$$

4. Estimate uncertainty via observed Fisher Information Matrix

$$\text{cov}_{\text{GMM}} = \sum_{i=1}^K \omega_i \text{cov}_{\text{GMM},i} + \sum_{i=1}^K \omega_i (\mathbb{E}[\theta] - \hat{\theta}_i)^2$$

$$= \sum_{i=1}^K \omega_i \left( \sum_{j=1}^{N_{pt}} \frac{1}{\Delta y_j^2} \left( \frac{\partial y_j(\theta_i)}{\partial \theta_i} \right)^2 \frac{\mathcal{N}(y_j, \Delta y_j|\theta_i)}{\pi(y_j, \Delta y_j|\vec{\theta})} \right)^{-1} + \sum_{i=1}^K \omega_i (\mathbb{E}[\theta] - \hat{\theta}_i)^2.$$

5. Use Information Criteria (AIC/BIC) to determine the number of Gaussians

$$\text{AIC} = N_{\text{parm}} \log N_{\text{pt}} - 2 \log L|_{\theta=\hat{\theta}},$$

$$\text{BIC} = 2 N_{\text{parm}} - 2 \log L|_{\theta=\hat{\theta}}.$$

$$N_{\text{parm}} = 2K + (K - 1).$$

M.Yan, T.-J. Hou, Z. Li, KM, C.-P. Yuan, arxiv: 2406.01664



# Committee #7: $\alpha_s$ Uncertainty with GMM

GMM (K=2) (Yellow shaded)

$$\bar{\alpha}_s = 0.11801 \pm 0.00192$$

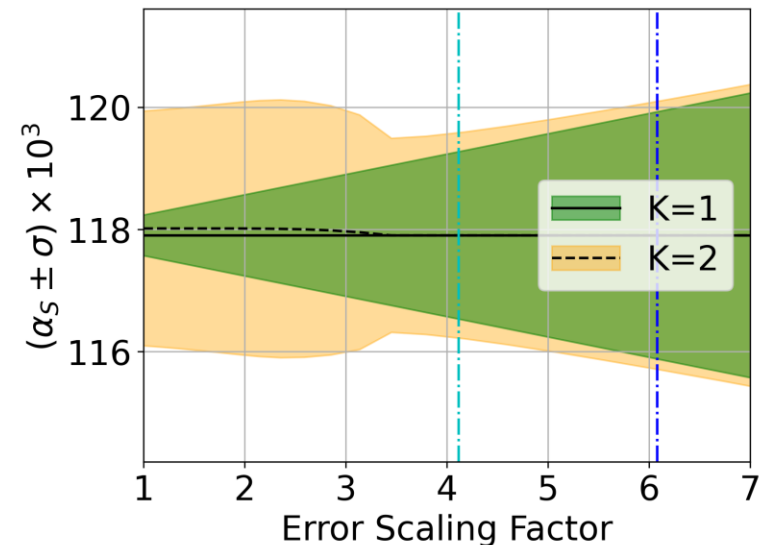
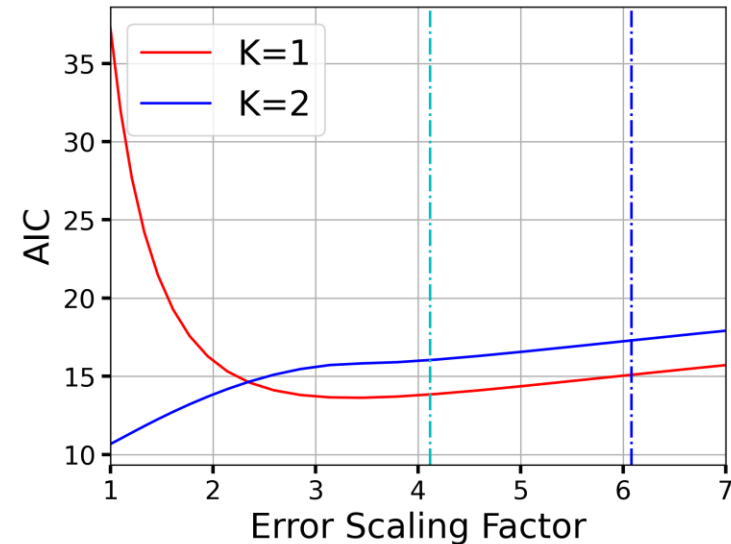
$e_{SPDG} \sim 4.1$  (Green shaded, Cyan line)

$$\bar{\alpha}_s = 0.11795 \pm 0.00135$$

$e_s \sim 6.1$  (Green shaded, Blue line)

$$\bar{\alpha}_s = 0.11795 \pm 0.0020$$

Caveat: How we partition the data sets does have an impact on uncertainty determination. More complete study is underway



# Impact of clustering

Using 52 data sets.

Caveat: This is a simplified application of the GMM on the profile likelihood of the strong coupling.

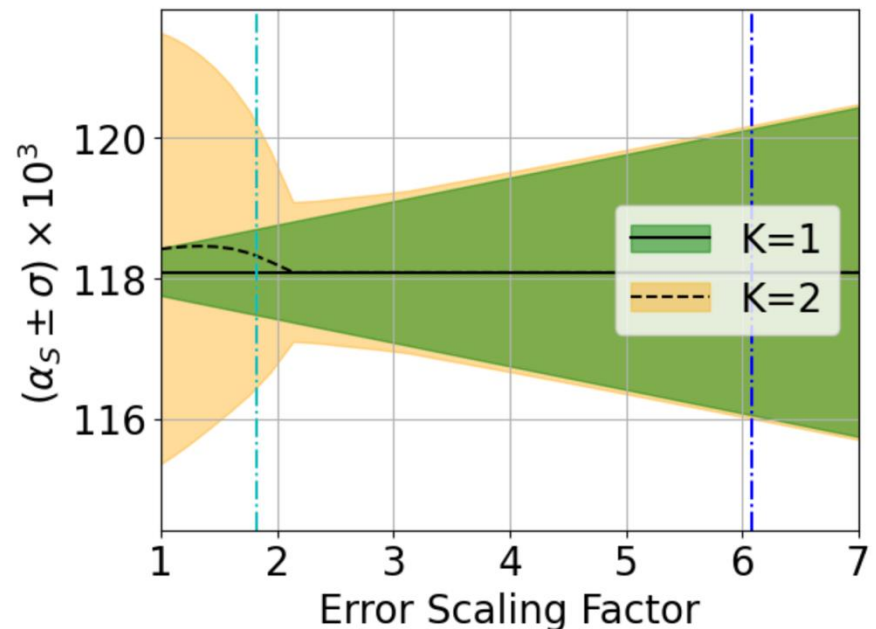
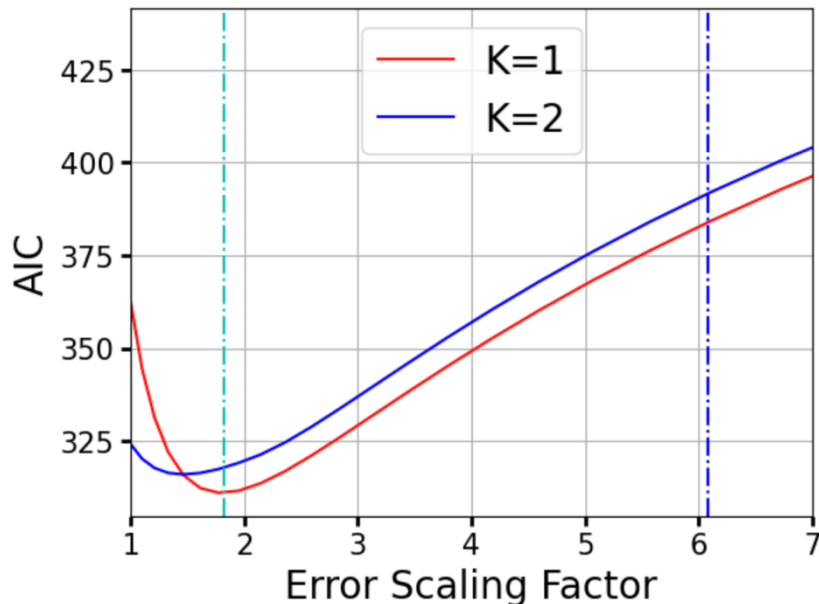
Full application needs fits over GMM likelihood

GMM (K=2, N=3)

$$\bar{\alpha}_s = 0.11801 \pm 0.00192$$

GMM (K=2, N=52) (Yellow shaded)

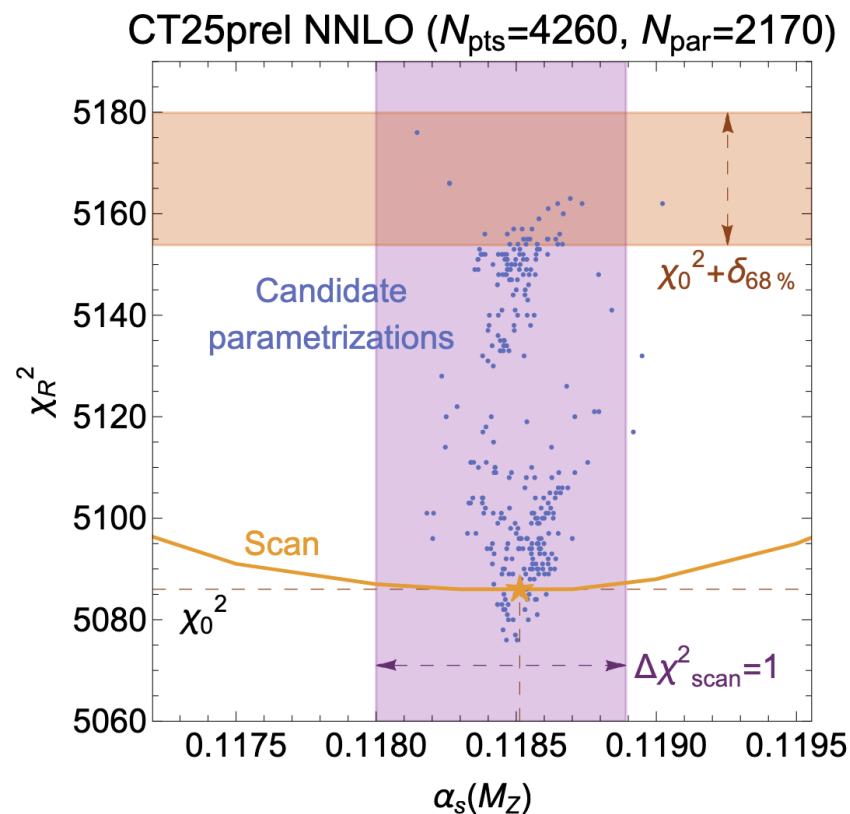
$$\bar{\alpha}_s = 0.1184 \pm 0.003$$



# Committee #8: PDF Parametrization Uncertainty for $\alpha_s$

$\alpha_s$  has low correlation with PDF parametrizations.

Epistemic uncertainty from PDF parametrization is small

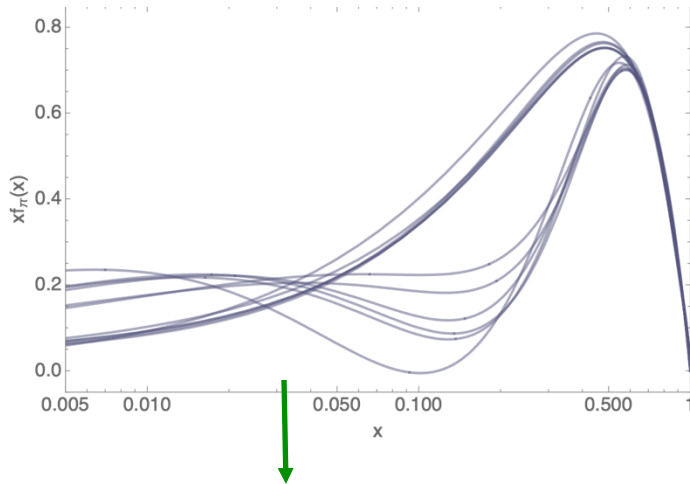


# No unique solution to PDF's inverse problem

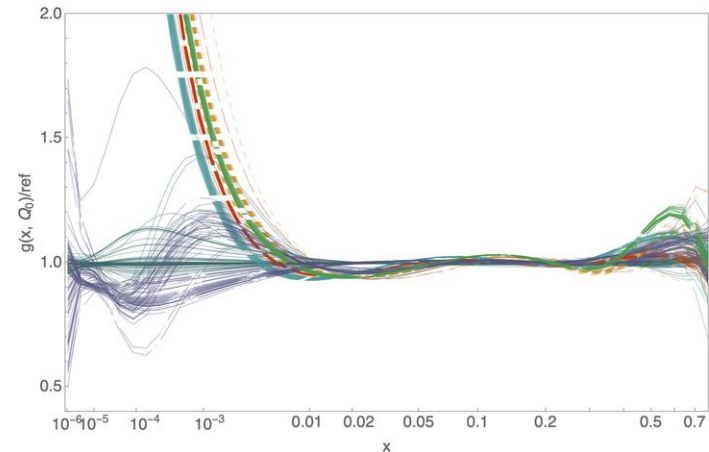
CT error bands quantify parametrization choice as a crucial source of epistemic uncertainty

- representative sampling over PDF models=study of parameter dependence  
(Bernstein basis ad-hoc forms or adaptable Bernstein basis—Fantômas )
- fitting methodologies also contribute to epistemic uncertainties

Pion PDF with adaptable Bernstein basis —a sandbox



Proton PDF with Bernstein basis ad-hoc forms



Fantômas:

*flexible basis with hyperparameters on top of trainable parameters,  $c_i$*

$$xf^q(x, Q_0^2) = A_q x^{\alpha_q} (1-x)^{\beta_q} \times \sum_i^N c_i \phi_i(g(x))$$

[Kotz et al, Phys.Rev.D109, 2507.22969]

To be applied to proton PDF soon!

# Combination of solutions for epistemic PDF uncertainty

CT error bands quantify parametrization choice as a crucial source of epistemic uncertainty

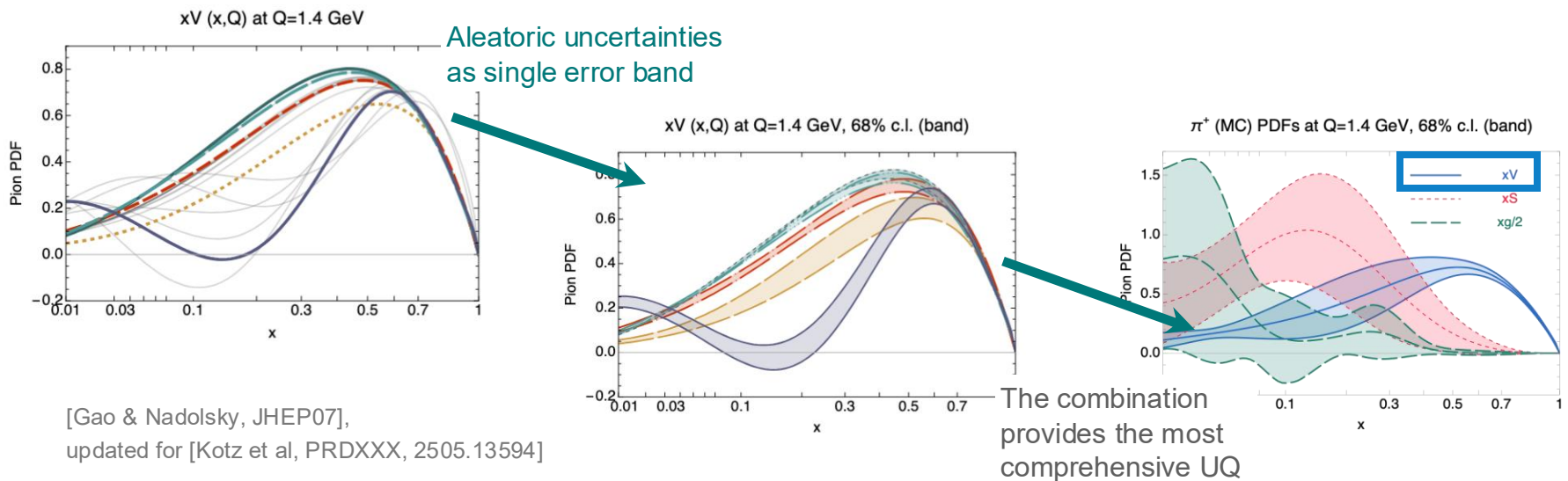
Statistical meaning of the sampling over parametrization:

*Unknown truth: all acceptable solutions stem from an unknown latent distribution of shapes*

— no probability density interpretation in the space of data (aleatoric uncertainty accounted for the Hessian way)

— sufficient to select a few solutions with most diverse shapes in  $N_{\text{flavor}}$ -dimensional space for  $x \in [0,1]$

A convex log-likelihood justifies the combination à la METAPDF



Next:

— information criteria to systematize the selection of solutions

# Central Committee (Preliminary)

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Statistical Method	Eq.	$\alpha_s(M_Z)$
Global Tolerance	4.8	$0.1179 + 0.0024 - 0.0025$
Dynamical Tolerance	4.10	$0.1179 + 0.0024 - 0.0012$
BHM	4.13	$0.118 + 0.0024 - 0.0027$
GMM	4.15	$0.11801 \pm 0.00192$
Average		$0.11795 + 0.00225 - 0.00182$

# Outlook

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- Presented (Preliminary) updated CTEQ-TEA fits to  $\alpha_s$
- Significant upward pulls from Jet and ttbar data from LHC
- Statistical Modeling is challenging!
- Set up committees to determine the uncertainty on the strong coupling.
- Each committee has its own estimate on the uncertainty – represents uncertainty on the uncertainty
- Central committee to build consensus between committees.