

# Progress on characterising PDF uncertainties

PDF4LHC Meeting, CERN

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Zahari Kassabov

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DAMTP, University of Cambridge



UNIVERSITY OF  
CAMBRIDGE



European Research Council





Progress in LHC phenomenology requires much improved understanding of uncertainties

- Improved understanding of experimental uncertainties
  - *Regularising experimental correlations in LHC data: theory and application to a global analysis of parton distributions* (ZK, Nocera, Wilson, arxiv:2207.00690)
- Improved understanding of theory uncertainties
  - *Parton distributions with scale uncertainties: a MonteCarlo sampling approach* (ZK, Ubiali, Voisey, arxiv:2207.07616)

# Covariance matrix regularization

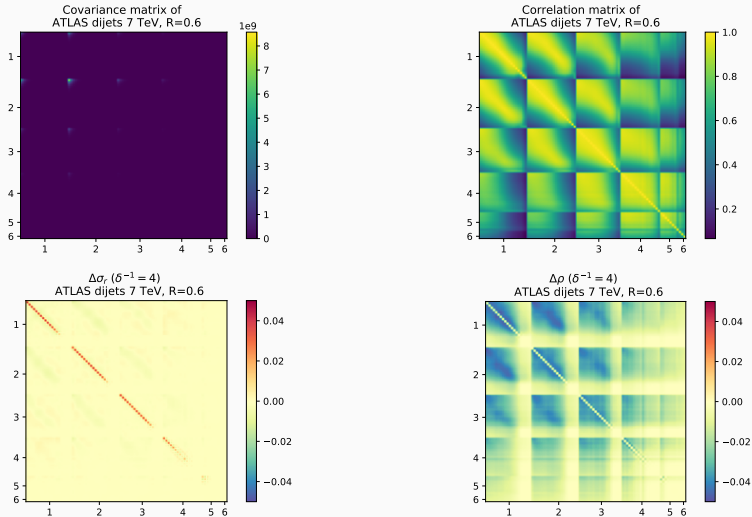
arxiv:2207.00690 studies the problem of covariance matrix stability

$$\chi^2 = (\text{data} - \text{theory})(\text{inverse covariance matrix})(\text{data} - \text{theory})$$

- If the covariance matrix is close to singular, then  $\chi^2$  is unstable
  - Small inaccuracies in the matrix make  $\chi^2$  appear **much larger**.
- Large correlations between systematics cause the matrix to be unstable
  - But correlation models usually difficult to determine, hence can be inaccurate

Derive a **regularization** procedure to make the covariance matrix stable.

## Example: ATLAS dijets at 7 TeV



- Original  $\chi^2/N_{\text{dat}} = 2.14$
- Regularized  $\chi^2/N_{\text{dat}} = 1.10$

- Consider matrix of uncertainties  $N_{\text{dat}} \times N_{\text{err}}$  such that the covariance is  $AA^t$ .
- Assuming the theory is known, fixed, and correct

$$\mathbf{d} - \mathbf{t} = A\mathbf{n}, \quad \mathbf{n} \sim \mathcal{N}(\mathbf{0}, I).$$

Then the expected value of the  $\chi^2$  is

$$\langle \chi^2 \rangle = \|A^+ A\|_F^2 = N_{\text{dat}},$$

- If the  $\chi^2$  is measured with a different matrix  $\bar{A}$  then the expected value is instead

$$\langle \bar{\chi}^2 \rangle = \|\bar{A}^+ A\|_F^2$$

Standard deviation of  $\chi^2$  distribution is  $\sqrt{2N_{\text{dat}}}$  hence we have stability if

$$\Delta\chi^2 = \|\bar{A}^+ A\|_F^2 - N_{\text{dat}} < \sqrt{2N_{\text{dat}}}$$

- No non trivial assumptions so far - Assumptions needed since we don't know  $A$ .

## A toy model with all the information

$$A(x) = \begin{pmatrix} \epsilon & 0 & 0 & 0 & 1 & 0 \\ 0 & \epsilon & 0 & 0 & 1 & 0 \\ 0 & 0 & \epsilon & 0 & 1 & 0 \\ 0 & 0 & 0 & \epsilon & 1-x & \sqrt{1-(1-x)^2} \end{pmatrix}$$

Assume  $\epsilon \ll 1$  and  $x \in [0, 1]$  unknown, sampled from

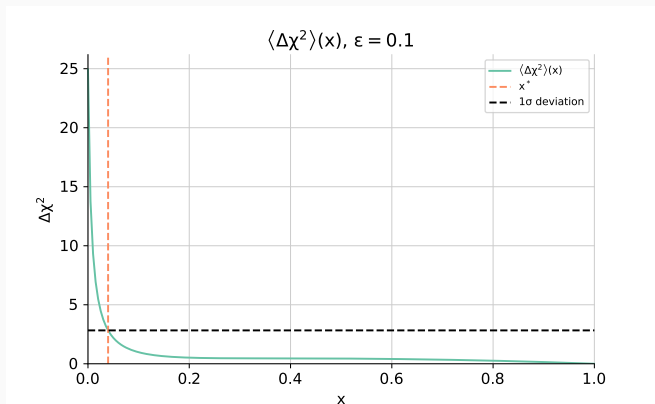
$$f_x(\xi) = 5(1 - \xi)^4$$

-  $x = 0$  is the most likely value. But it kills stability! - Measure with

$$\langle \Delta \chi^2 \rangle (x) = \int_0^1 \left| \|\bar{A}^+(\xi) A(x)\|_F^2 - N \right| f_x(\xi) d\xi$$



## Why not pick the highest correlation?



- $x = 0$  leads to an expected error of over 8 standard deviations
- $x = 0.04$  reduces the error to 1 standard deviation

# Regularization

Assuming that

- All inaccuracies are in correlations  $\bar{A} = D\bar{A}_{\text{corr}}$ 
  - $D$ : matrix of standard deviations
- Inaccuracies come from a small  $\mathcal{O}(1)$  number of systematics

The stability of  $A$  can be measured by the condition number

$$Z = \|\bar{A}_{\text{corr}}^+\|_2 = \|\bar{A}_{\text{corr}}\|_2^{-1}$$

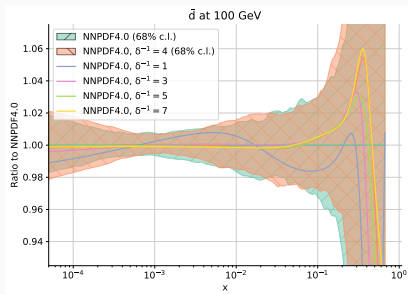
Closest matrix to  $A = DUSV^t$  with  $Z = \delta^{-1}$  for some acceptable  $\delta$

$$\bar{A}_{\text{reg}} = DUS_{\text{reg}}V^t$$

$$S_{\text{reg}(ii)} = \begin{cases} \delta & s_i < \delta \\ s_i & \text{otherwise} \end{cases}$$

Applying our preferred regularization to the full NNPDF dataset:

- Relative covariance differences smaller than 5%
- Correlation differences smaller than 0.05
- $\chi^2/N_{\text{dat}}$ :  $1.16 \rightarrow 1.11$
- Almost no effect on best fit PDFs



## Experimentalist (target audience)

- Measure stability of covmats
- Provide stable covmats
  - If sources of inaccuracy not known use the regularization procedure
  - If more information and resources available carry out detailed analysis (see Sect 3.2 of arxiv:2207.00690)

## Fitters (fallback)

- Measure stability of covmats
- Seek stable versions of the covmat
  - If not available, regularize them
- Discuss regularized  $\chi^2$  only

# What not to do

## Correlation models (so far)

### Advantages:

- Made by experimentalists using complete information

### Disadvantages:

- Appear much later than the original data, causing versioning confusion
- Enormously laborious to analyse (feedback loop with experimentalists)
- Stability of the covariance matrices not guaranteed

## Regularization procedure

### Advantages

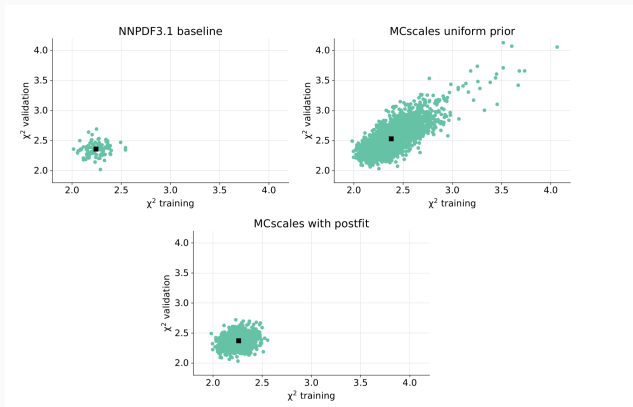
- Simple quick to apply formula
- Minimal modification of the covariance matrix
- Guaranteed stability
- Seems to yield similar results as correlation models

### Disadvantages

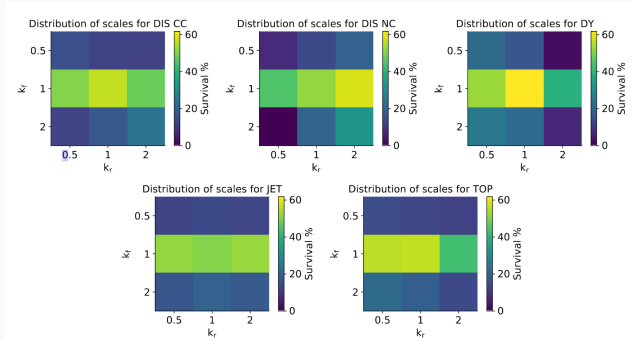
- Grounded on assumptions and incomplete information

- [arxiv:2207.07616](#) Studies the problem of matching PDF fits with scale variations
- Theory predictions require specifying factorisation and renormalisation scales:
  - Result depends on scale choice  $\rightarrow$  scale uncertainty.
- Idea: assign different scale multipliers to each NNPDF replica.
- Record the information so scales can be matched between the PDF and the partonic cross section.

# Fit quality allows assessing scale choices



# Survival fraction



- Statistical interpretation of scale variations
- Assessment of ranges of variation

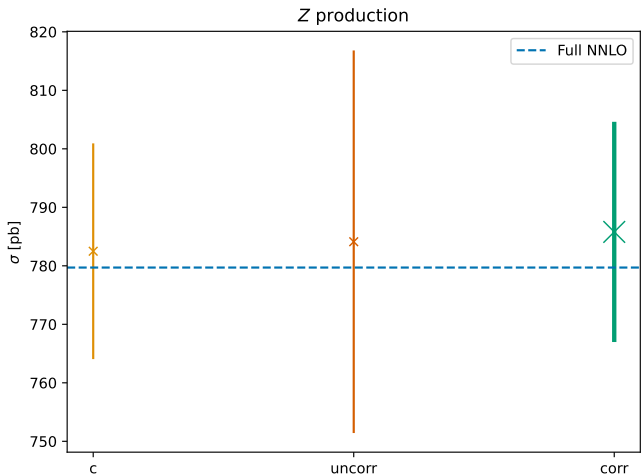


We record the scale multiplier choices for each fitted replica. This allows matching the partonic cross section with the scale choices within each replica

- Monte Carlo sample of  $N_{\text{rep}}$  **MCscales** prediction including correlated PDF and scale uncertainty

$$\left\{ \sigma_k = \hat{\sigma}_p(k_f^{(k)}, k_{r_p}^{(k)}) \otimes f_k(k_f^{(k)}, k_{r_p}^{(k)}) \quad \forall k \in 1 \dots N_{\text{rep}} \right\}$$

## Scales must be matched: Example $Z$ cross section



Treating scales as uncorrelated between PDF and partonic cross section largely overestimates the uncertainties

- Correlation between scale variations in PDFs and partonic cross sections is large.
  - MCscales allows for exact matching
- Transparent specification of scale uncertainties, with tools allowing users to manipulate it.
- Largest benchmark of effect of scale variations of fit quality.
- NNLO implementation on NNPDF4.0 expected.