



PDF uncertainties using a Monte Carlo method

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In collaboration with A. Glazov

DESY

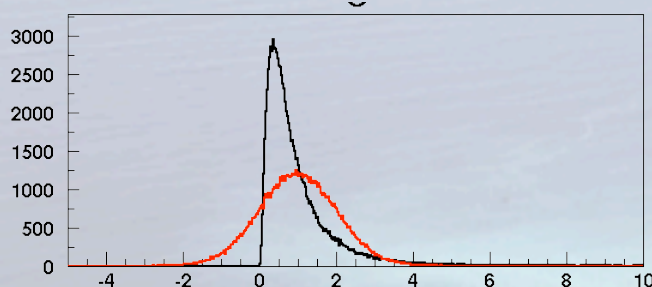
- Motivation
- Method
- Cross check of the method:
 - Comparison with the standard error estimation
- Test various assumptions for the error distributions:
 - Gauss, Log-normal, Uniform
- Summary

Motivation

- The idea is to use a Monte Carlo method to estimate PDF uncertainties under various assumptions for the error distributions.
- Standard error estimation of PDFs relies on the assumption that all errors follow Gauss statistics.
 - ✓ Monte Carlo method can provide an independent cross check of it.
- However, Gaussian assumption is not always correct:

$$\sigma \sim \frac{N}{\mathcal{L}A}$$

- Some systematic uncertainties follow Log-Normal Distribution:
 - lumi ,detector acceptance,etc.



Gauss and Log-normal distributions:

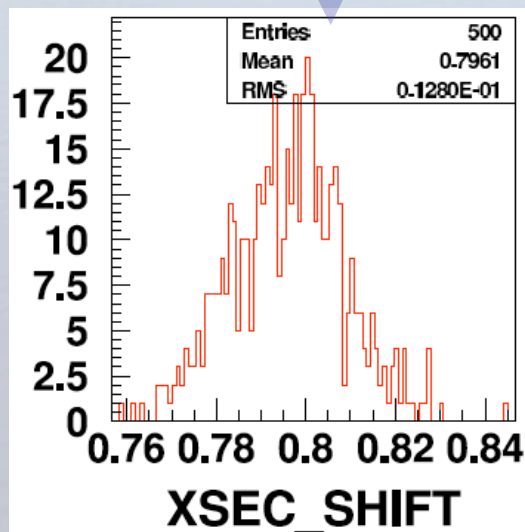
- same mean
- shifted peaks

How is that affecting PDF's errors?

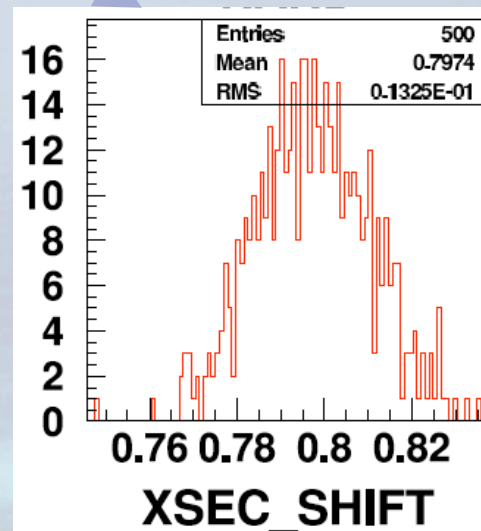
- Some systematic uncertainties follow a Uniform Distribution:
 - “upper” limit uncertainties
- ✓ Monte Carlo method allows to test the various assumptions.

Method (I)

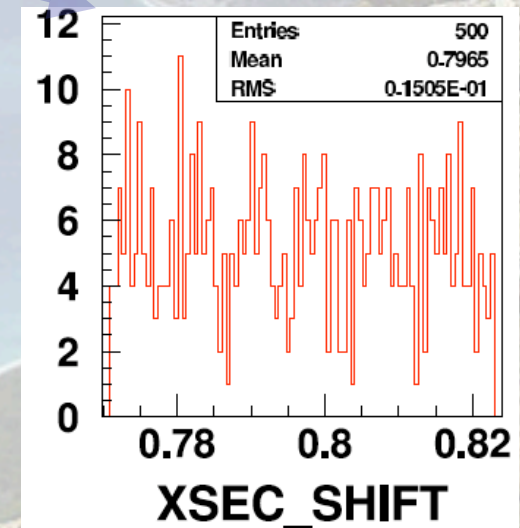
- Prepare “shifted data sets”:
 - Allow the central value of the cross sections (σ_i) to fluctuate within its systematic and statistical uncertainties taking into account all the correlations:
 - Various assumptions can be considered for the error distributions:
 - Gauss, Lognormal, Uniform



$$\sigma_i = \sigma_i (1 + \delta_i^{uncorr} RAND_i)$$



$$\sigma_i = \sigma_i R_i^{log} (1, \delta_i^{uncorr})$$



$$\sigma_i = \sigma_i - a + 2aR_j^{uni}$$

Method (II)



- **Shifts for Statistical errors:**
 - allow each data point to randomly fluctuate within its statistical uncertainty assuming either Gauss, Log-normal, or uniform distributions
- **Shifts for Systematic errors:**
 - For each systematic source j uniformly select “fluctuation probability” P_j
 - For each data point shift the central value of cross section such that probability of this shift for systematic source j is equal P_j (or $(1-P_j)$)

Method (III)

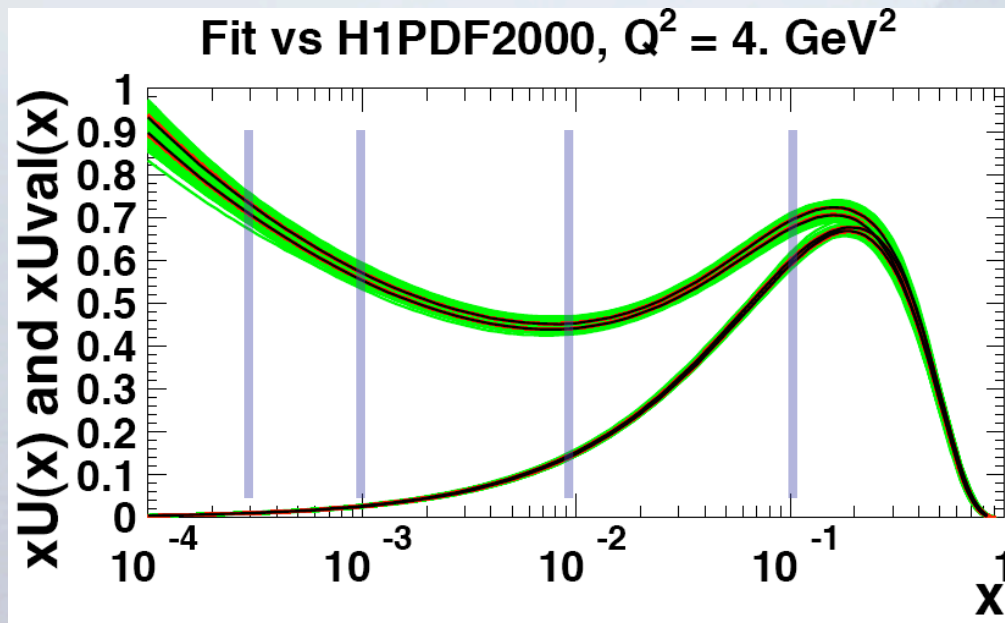
- Repeat the preparations for N times (here $N \geq 100$)
- Perform the NLO QCD fit N times to extract PDFs
- PDF uncertainties => from the RMS of the spread

- This study is performed using:
 - published H1-HERA I data of NC and CC $e^\pm p$ scattering cross sections [ref: [Eur. Phys. J. C 30, 1-32 \(2003\)](#)]
 - fit program H1 QCDNUM implementation at NLO:
 - $\overline{\text{MS}}$ renormalisation scheme, DGLAP evolution at NLO, massless quarks, polynomial form for PDF parametrisation a' la H1PDF2000

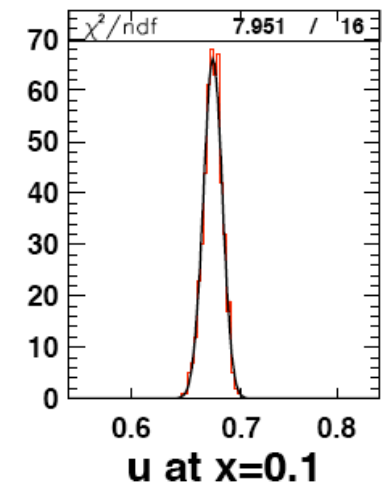
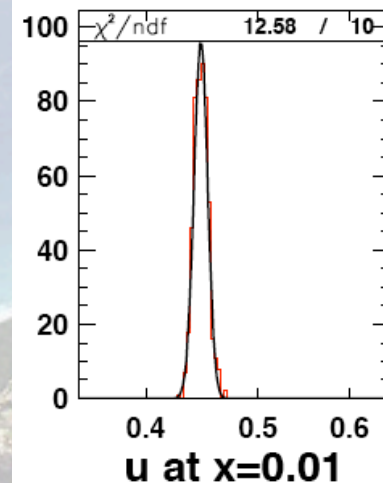
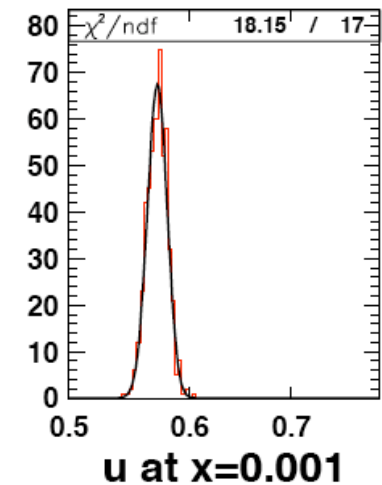
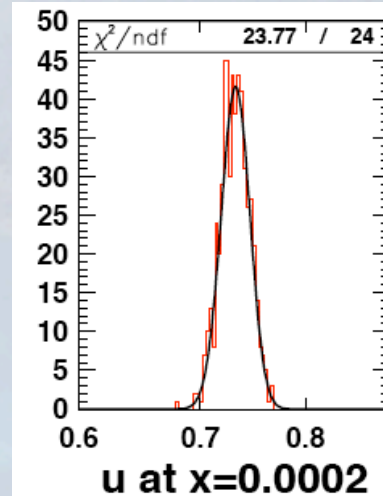
Test of the Method: Gauss Distribution (xU)



- To test the method, assume all errors follow Gauss distribution and compare the results to standard error estimation



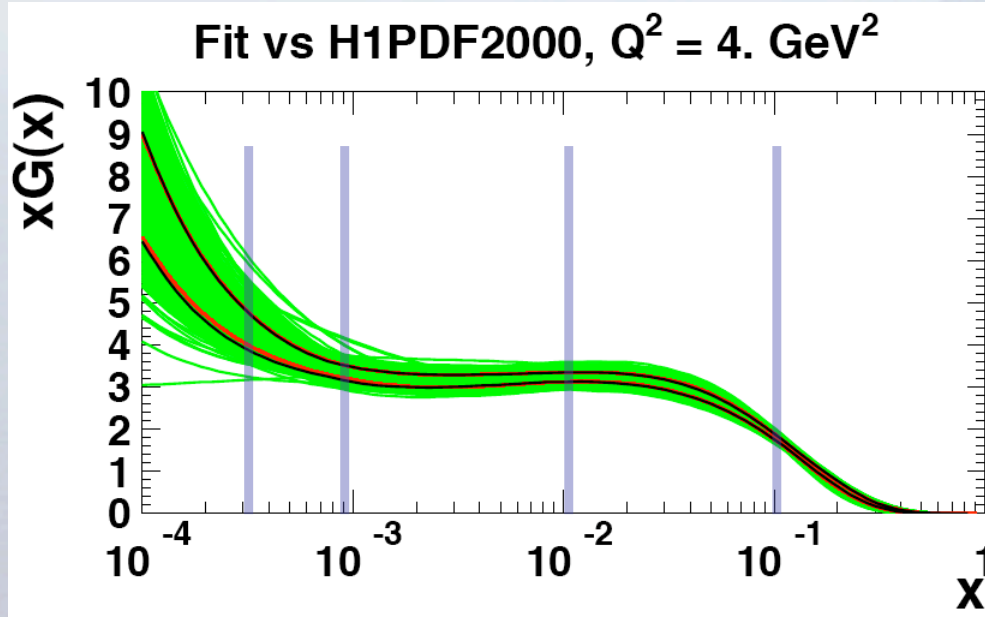
- 500 Green lines
- Red lines: PDF uncertainties from RMS
- Black lines: Hessian errors
- Projections at various x values are shown:
 - Good Gauss Fit



Test of the Method: Gauss Distribution (xG)

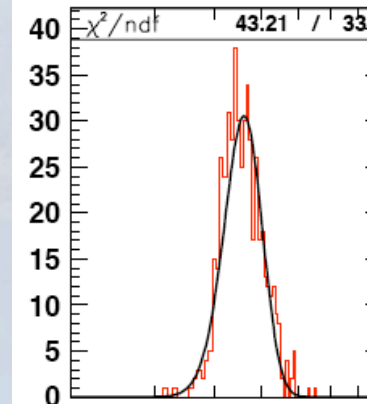


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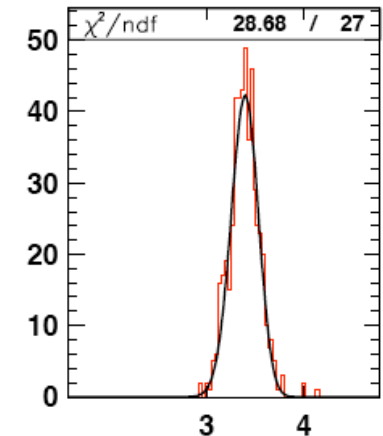


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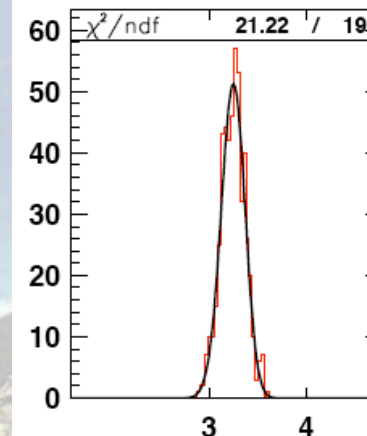
Good agreement with the standard error estimation!



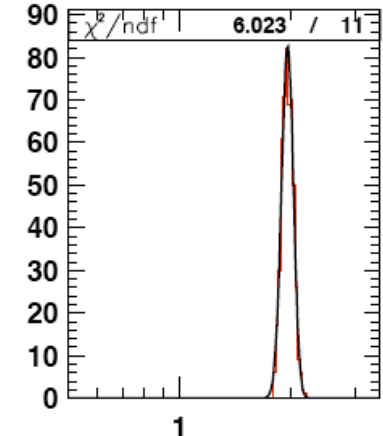
gluon at $x=0.0002$



gluon at $x=0.001$



gluon at $x=0.01$



gluon at $x=0.1$

Test various assumptions for the errors

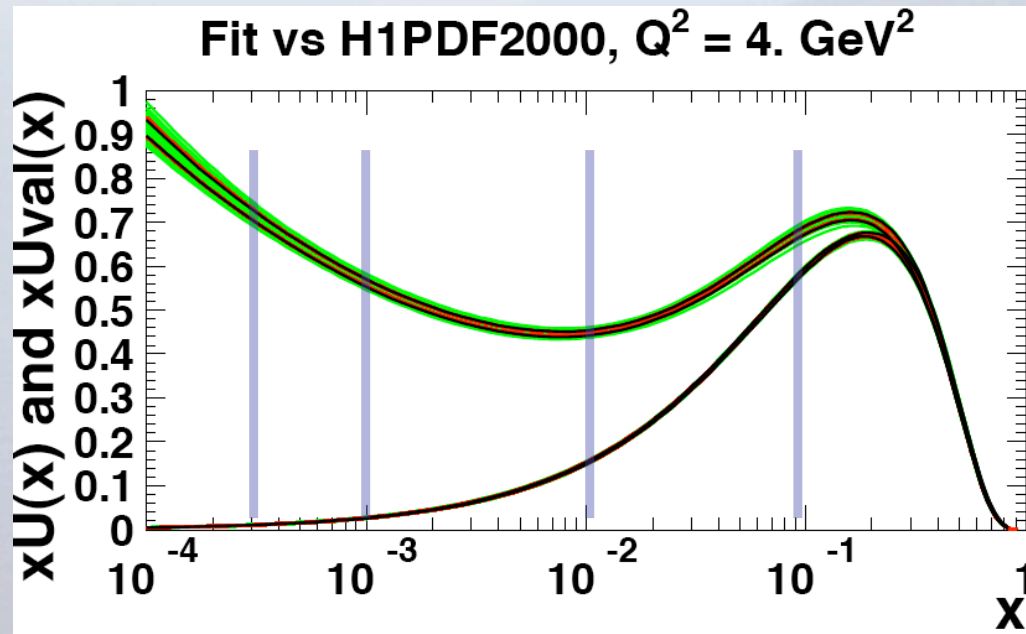


- Now that the method is cross checked, we are ready to test other assumptions:
 1. Log-normal for lumi, all the rest set to Gauss
 2. Log-normal for all systematic errors, Gauss for statistical uncertainty
 3. Uniform for all errors

1. Log-normal dist. for Lumi (xU)

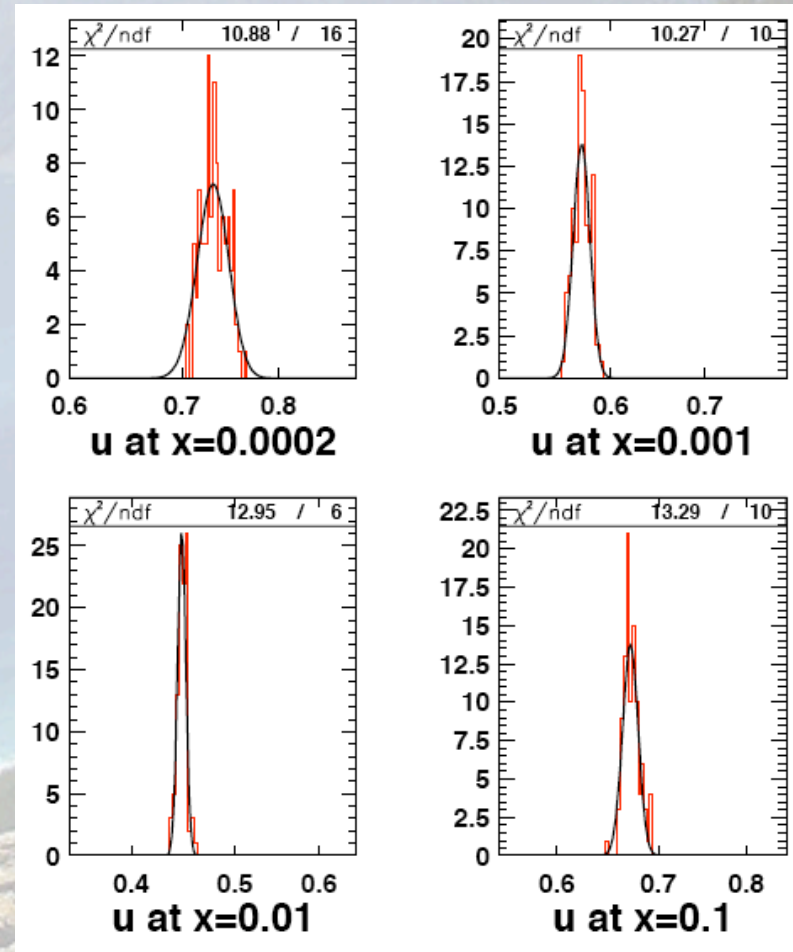


- Assume that all errors, apart from Lumi, follow Gauss
 - Test the effect of log-normal assumption for Lumi uncertainty



- 100 Green lines
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- Black lines: Hessian errors

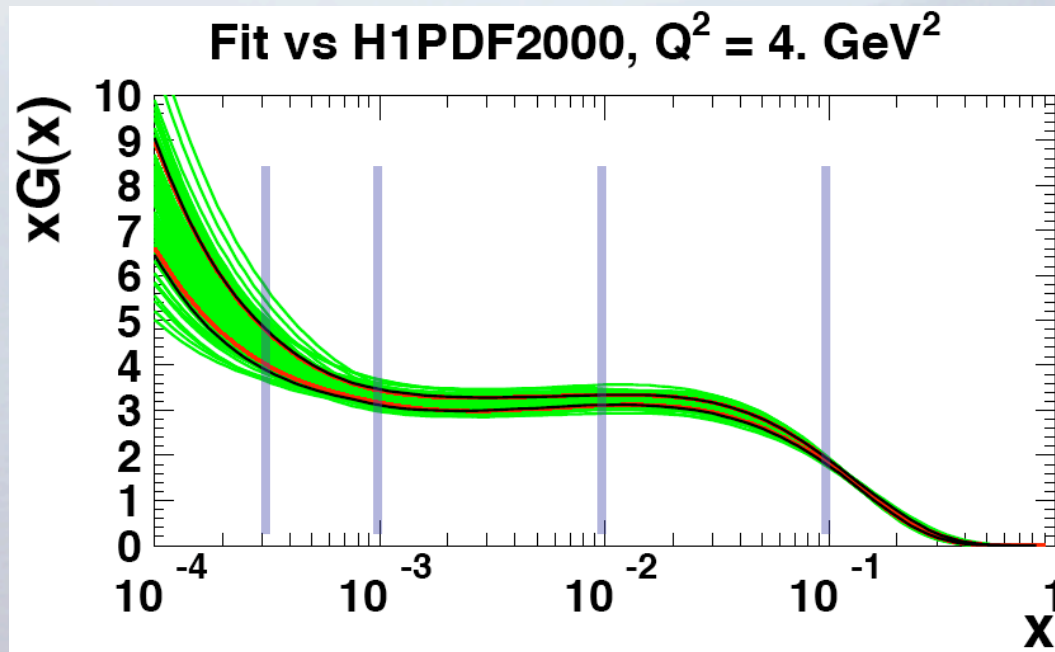
Similar effect to pure gaussian case!



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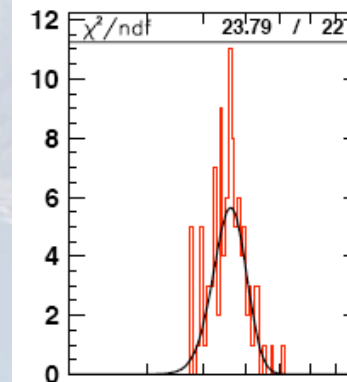


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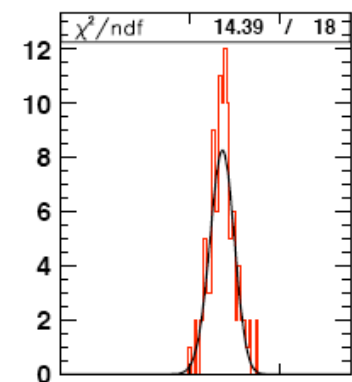


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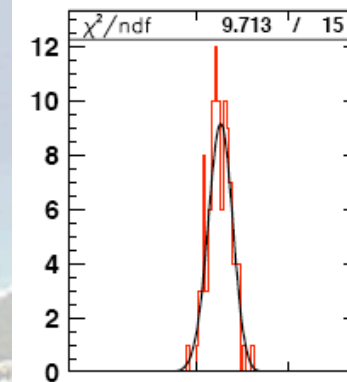
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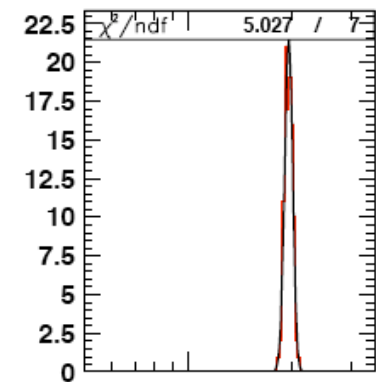
gluon at $x=0.0002$



gluon at $x=0.001$



gluon at $x=0.01$

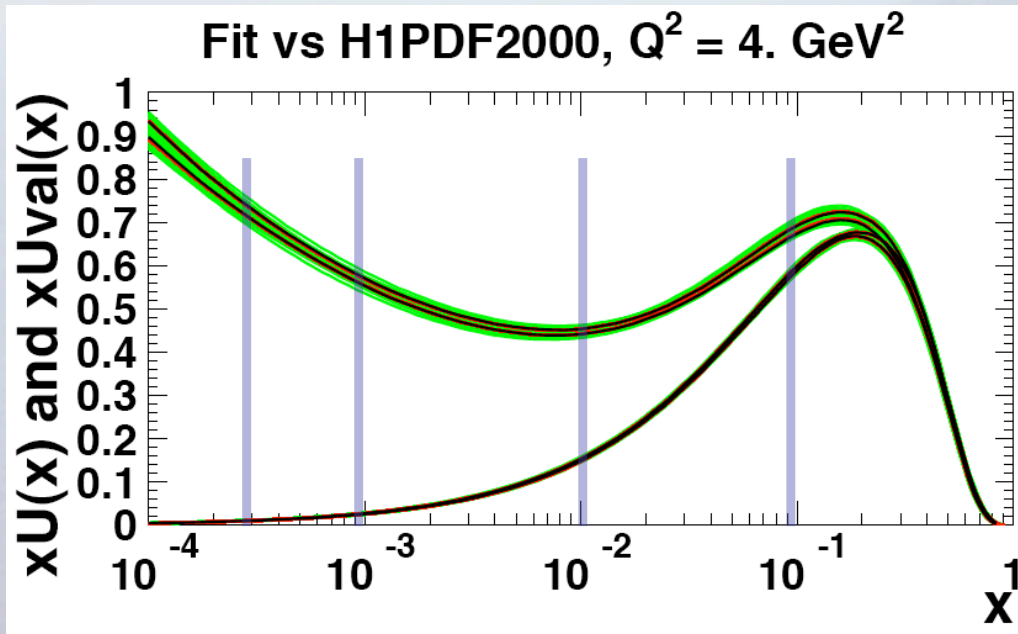


gluon at $x=0.1$

2. Log-normal dist. for all systematic (xU)

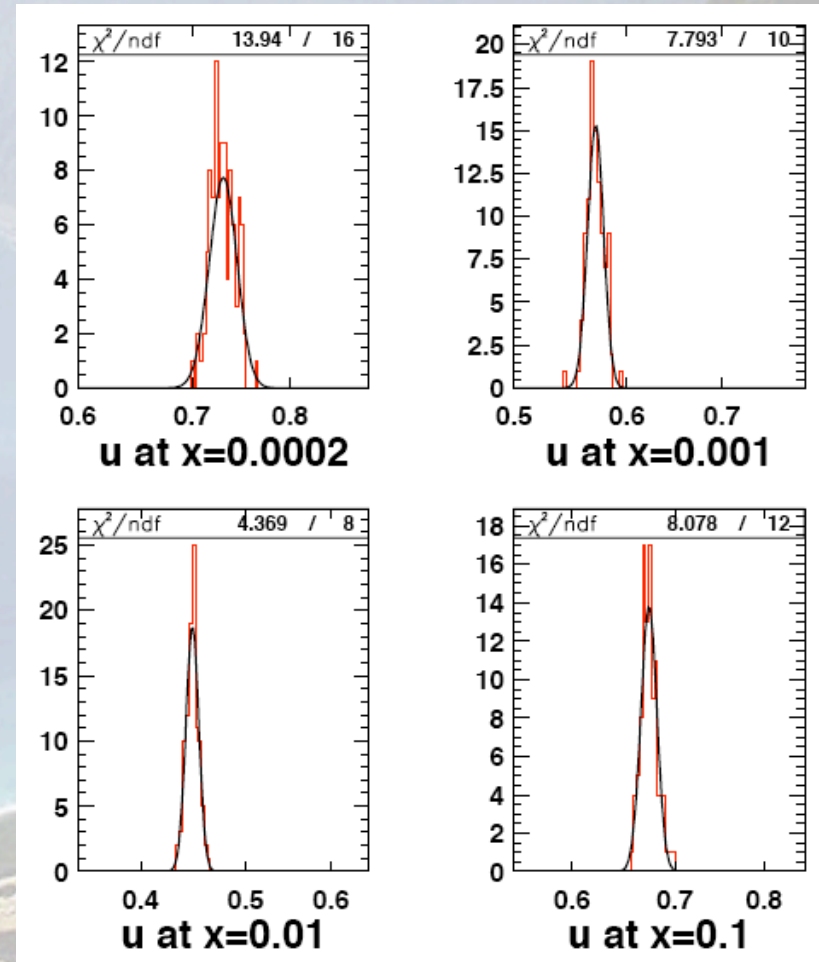


- Assume that statistical errors follow Gauss distribution and all systematic errors follow Log-normal distribution



- 100 Green lines
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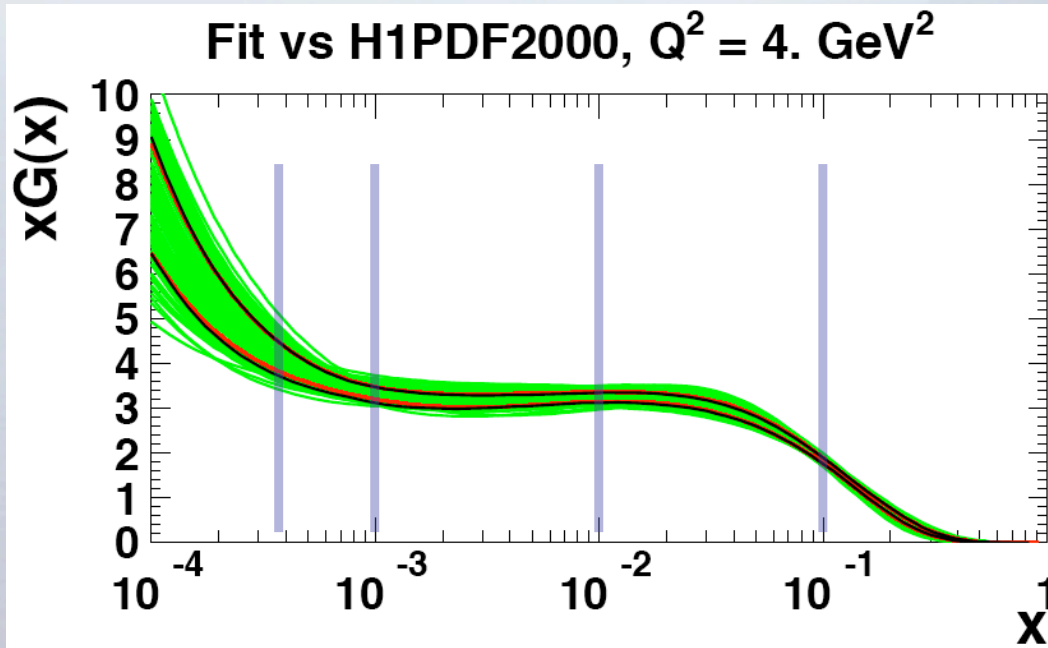
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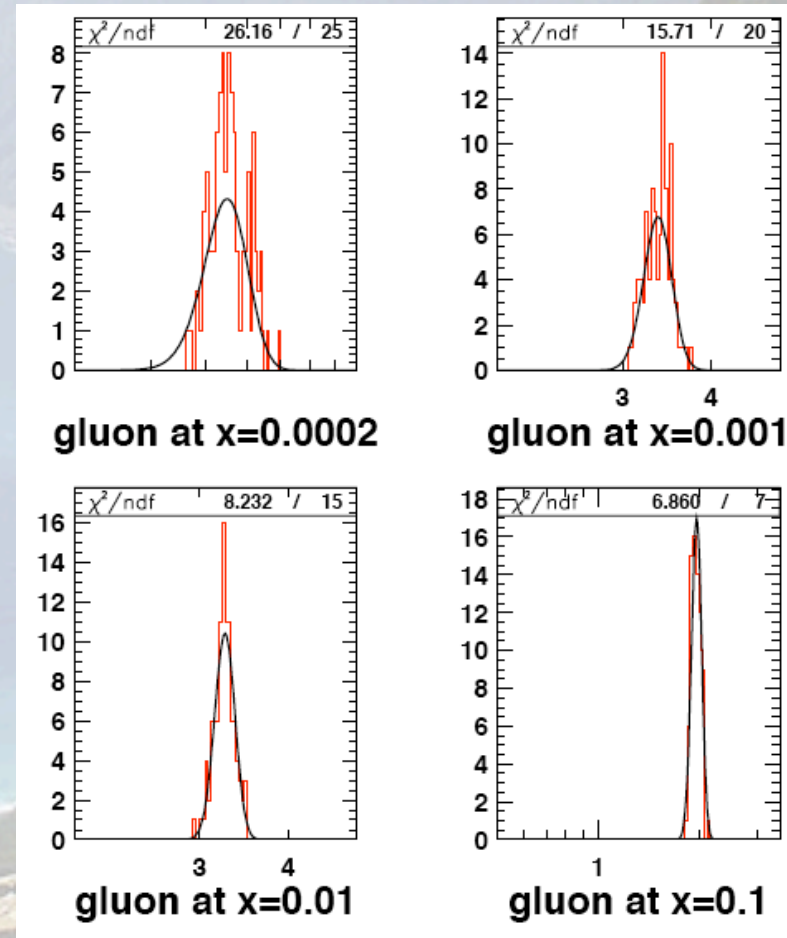


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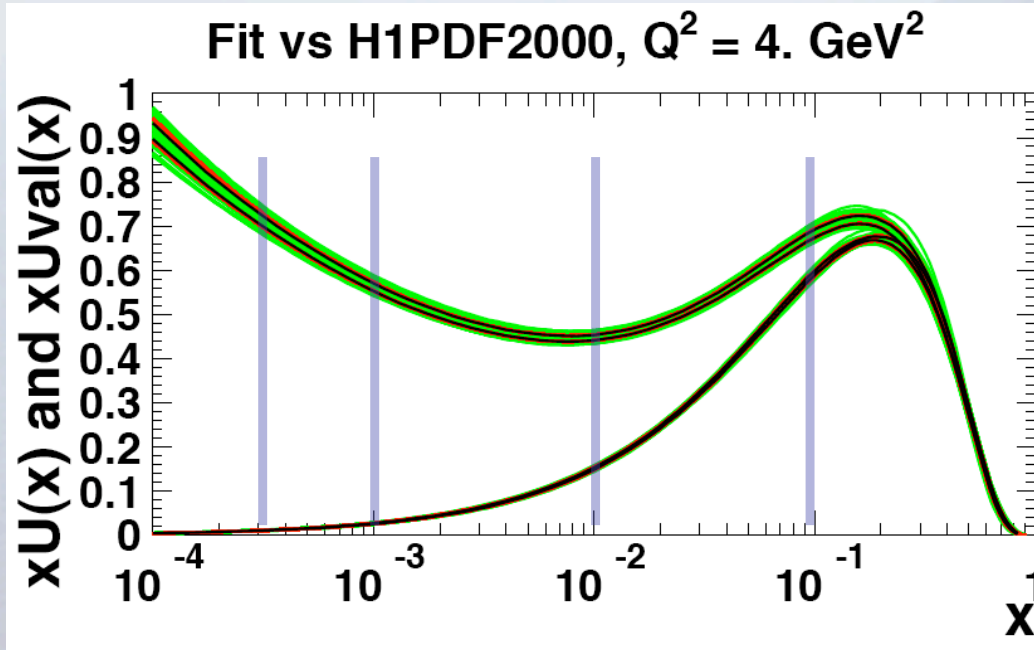
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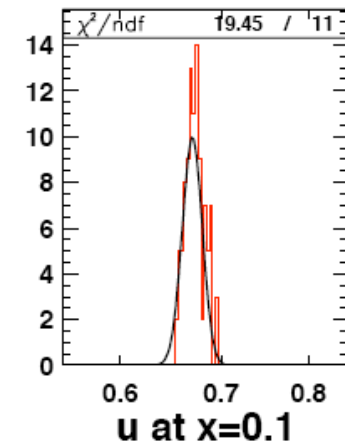
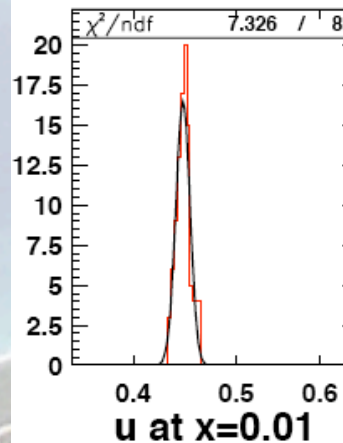
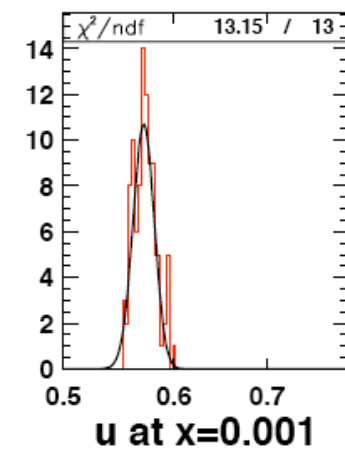
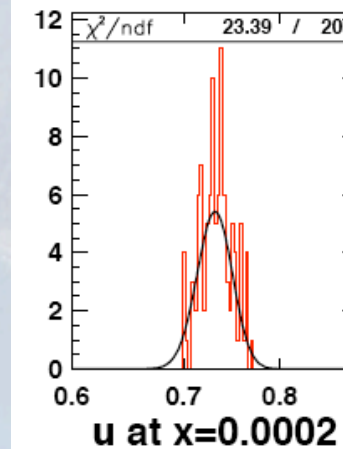
3. Uniform dist. for all errors (xU)



- Assume all errors follow uniform distribution.



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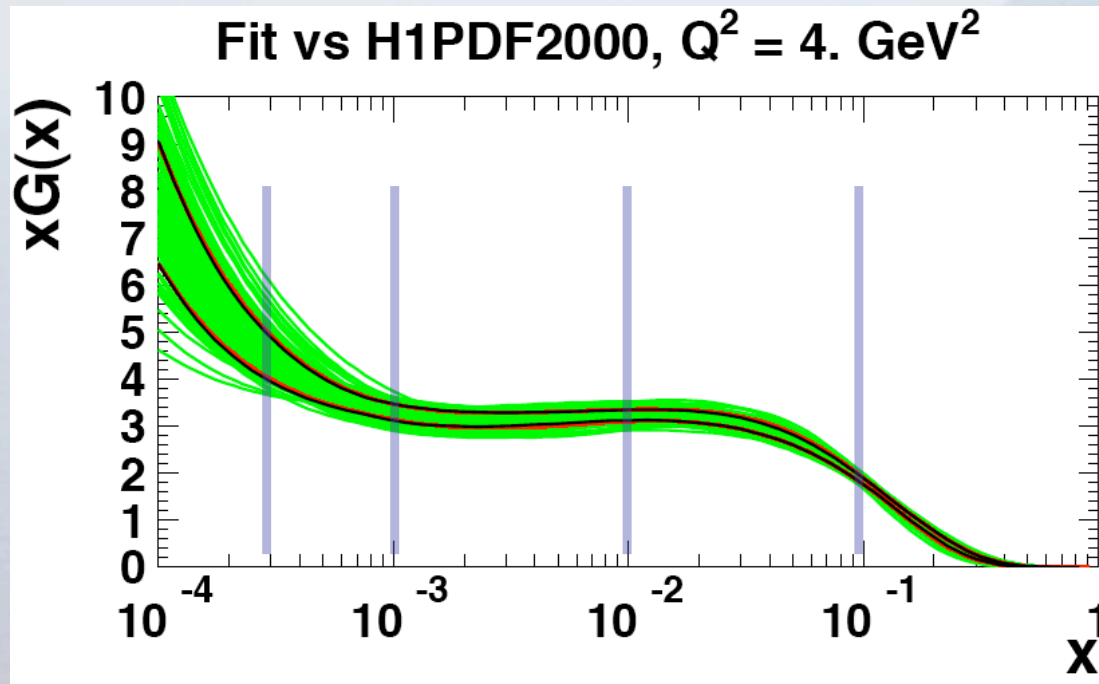


Similar effect to pure gaussian case!

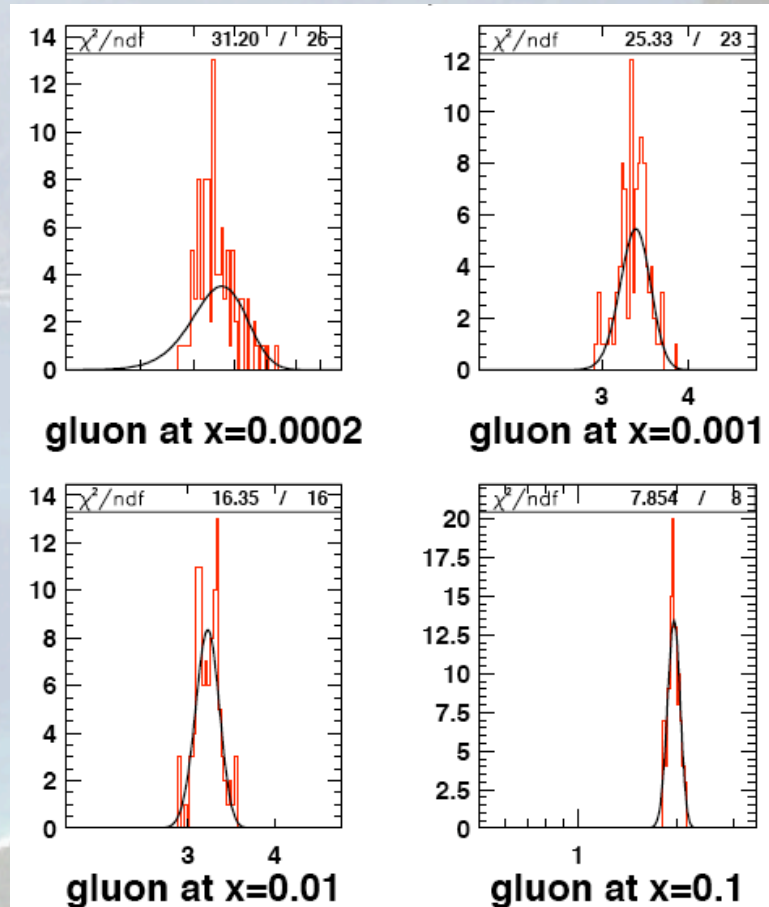
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Summary

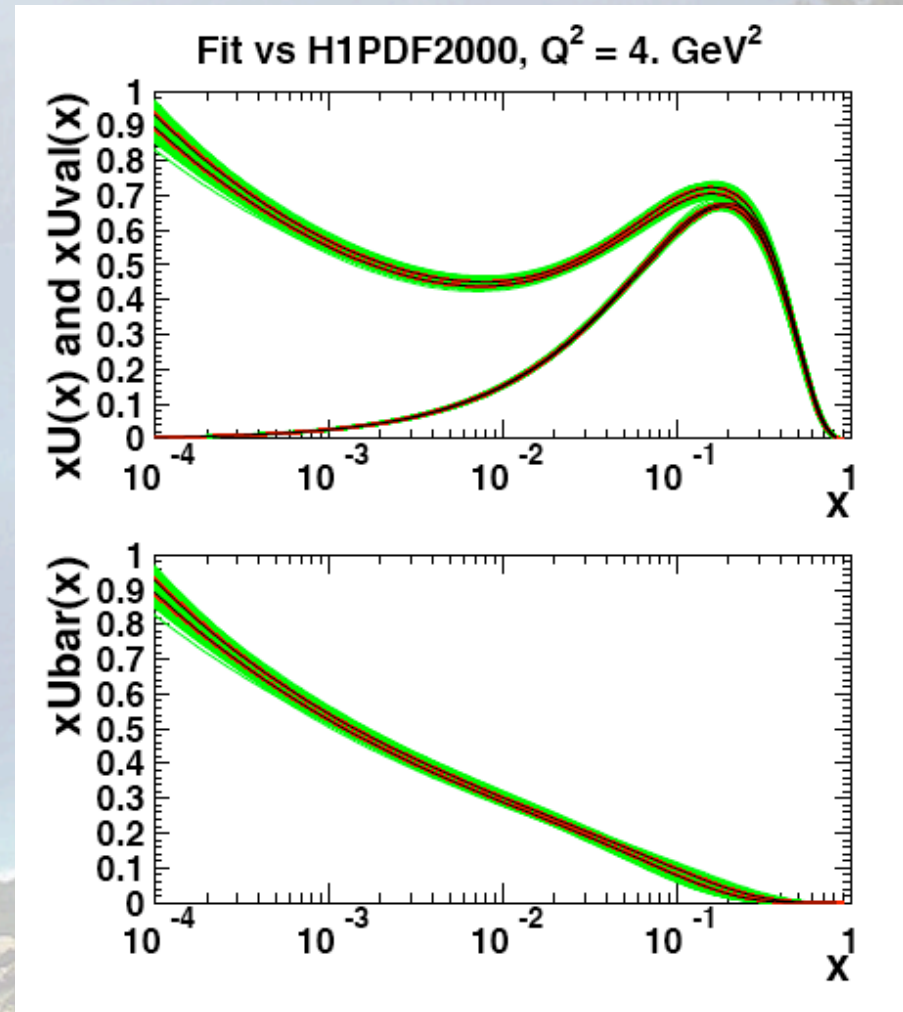
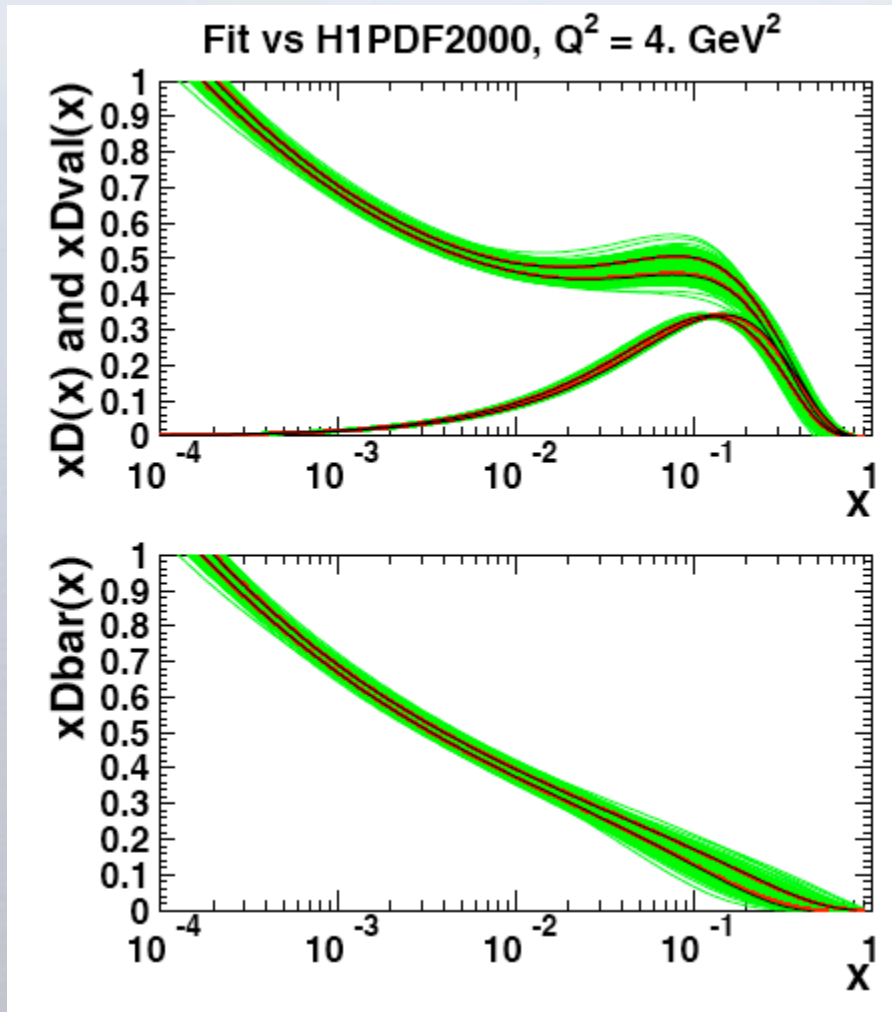


- A simple method to estimate PDF uncertainties has been built within QCD Fit framework:
 - Assuming only Gaussian distribution of all errors, the results agree well with the standard error estimation
- This method allows to check the effect of non-Gaussian assumptions for distributions of the experimental uncertainties:
 - For the H1 data, results are similar to the Gaussian case when using Log-normal and Uniform distributions of the uncertainties
- The method could be extended for other physical variables (i.e. cross sections) for cross checks with the standard error evaluation

Backup



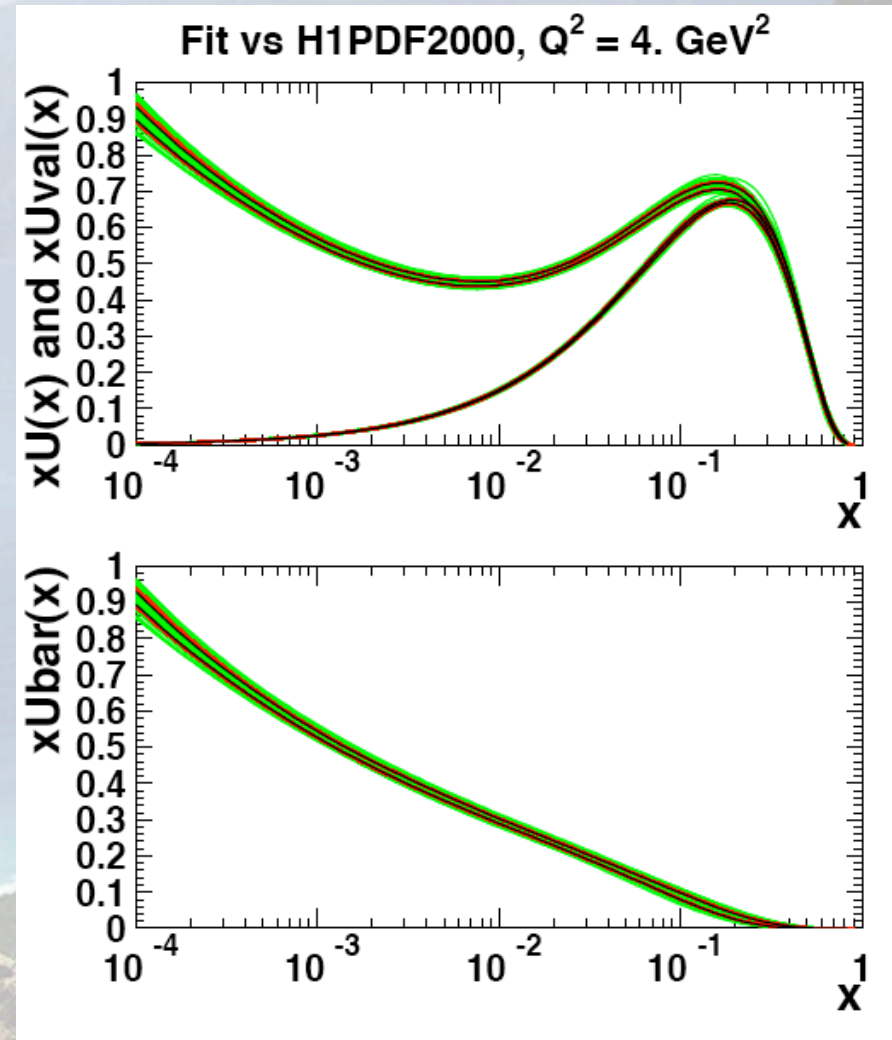
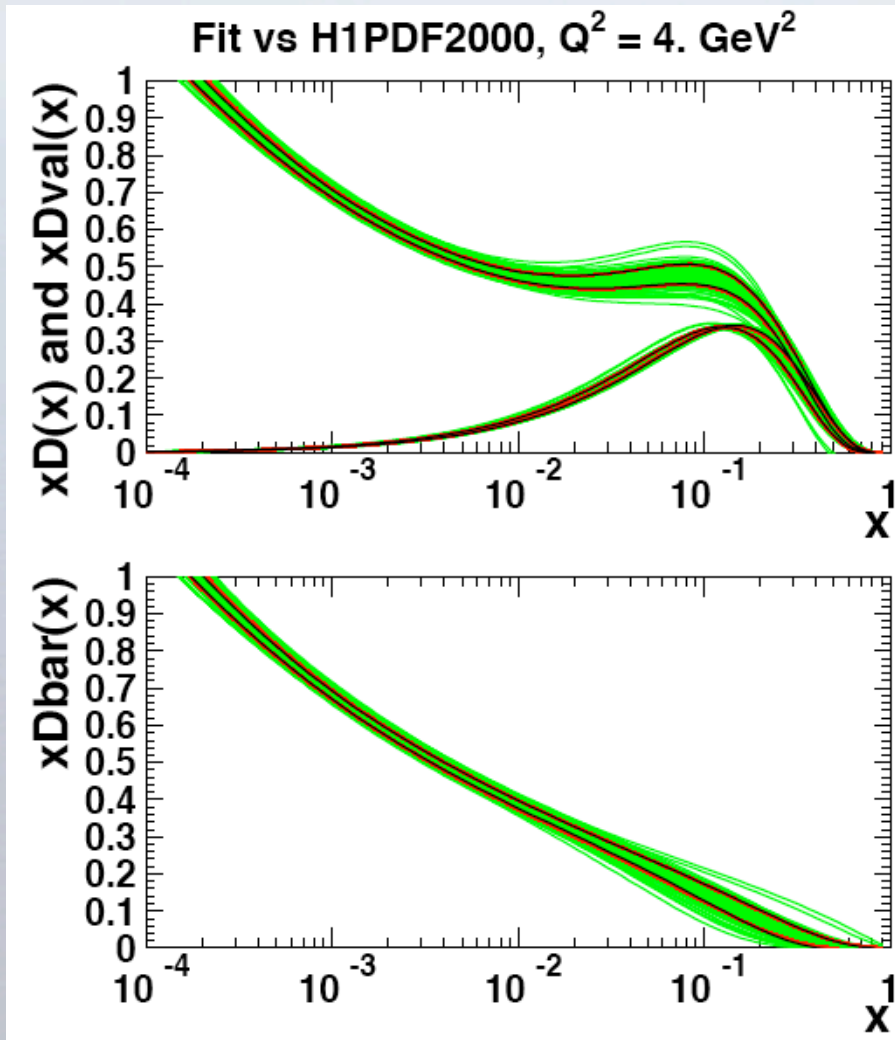
- Gauss Errors: more distributions



Backup



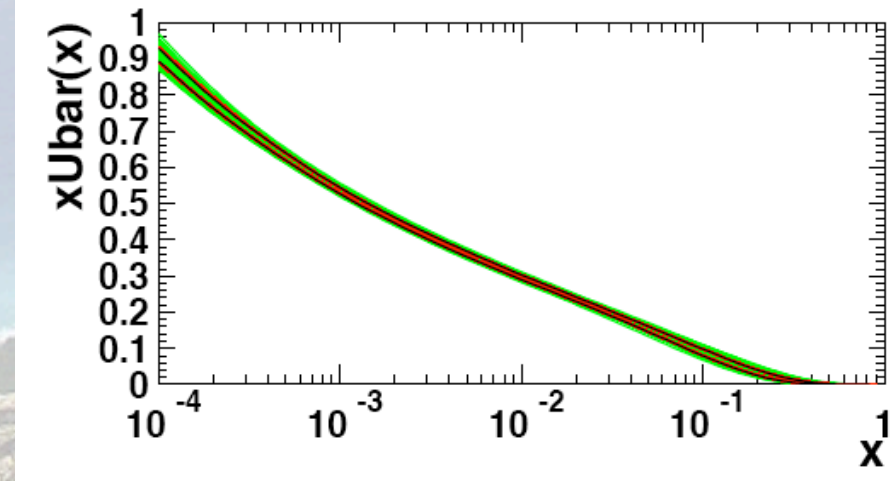
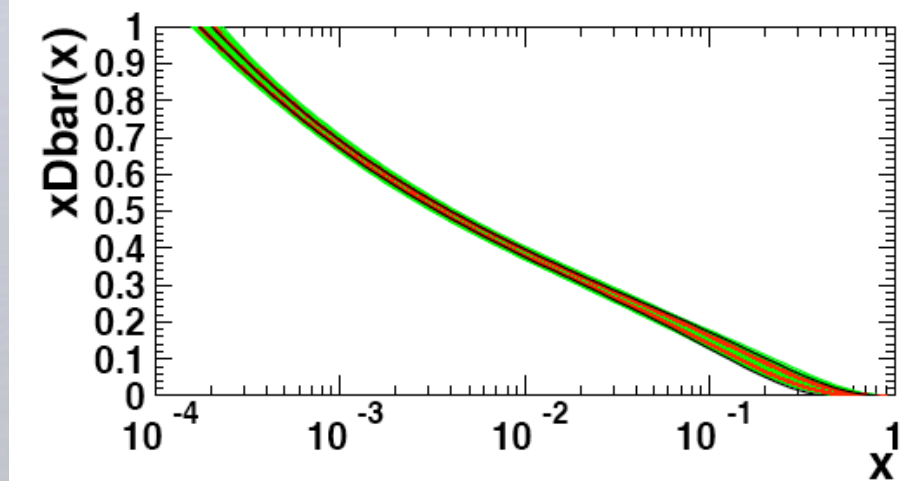
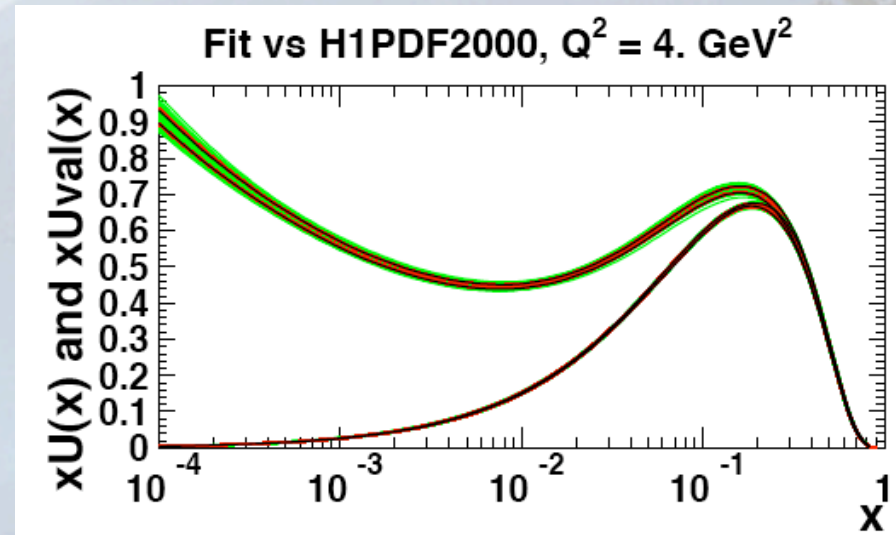
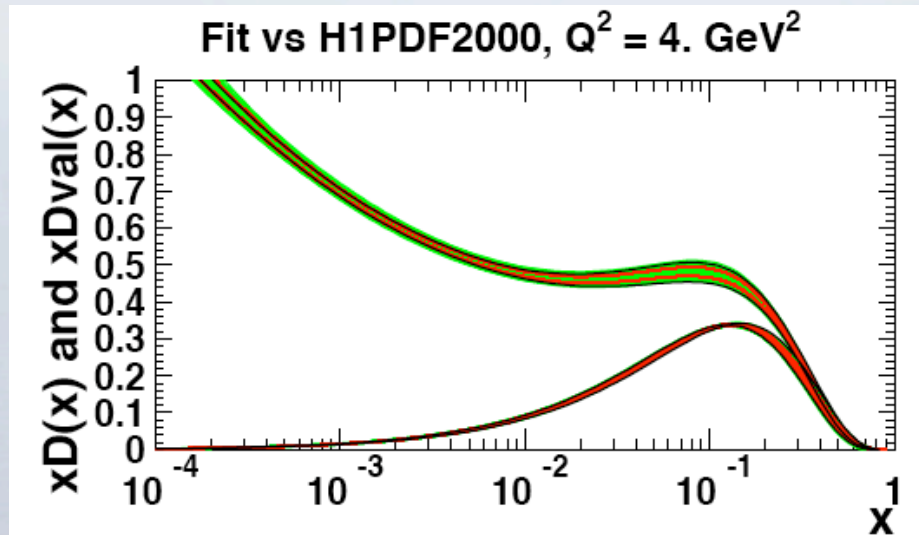
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Backup



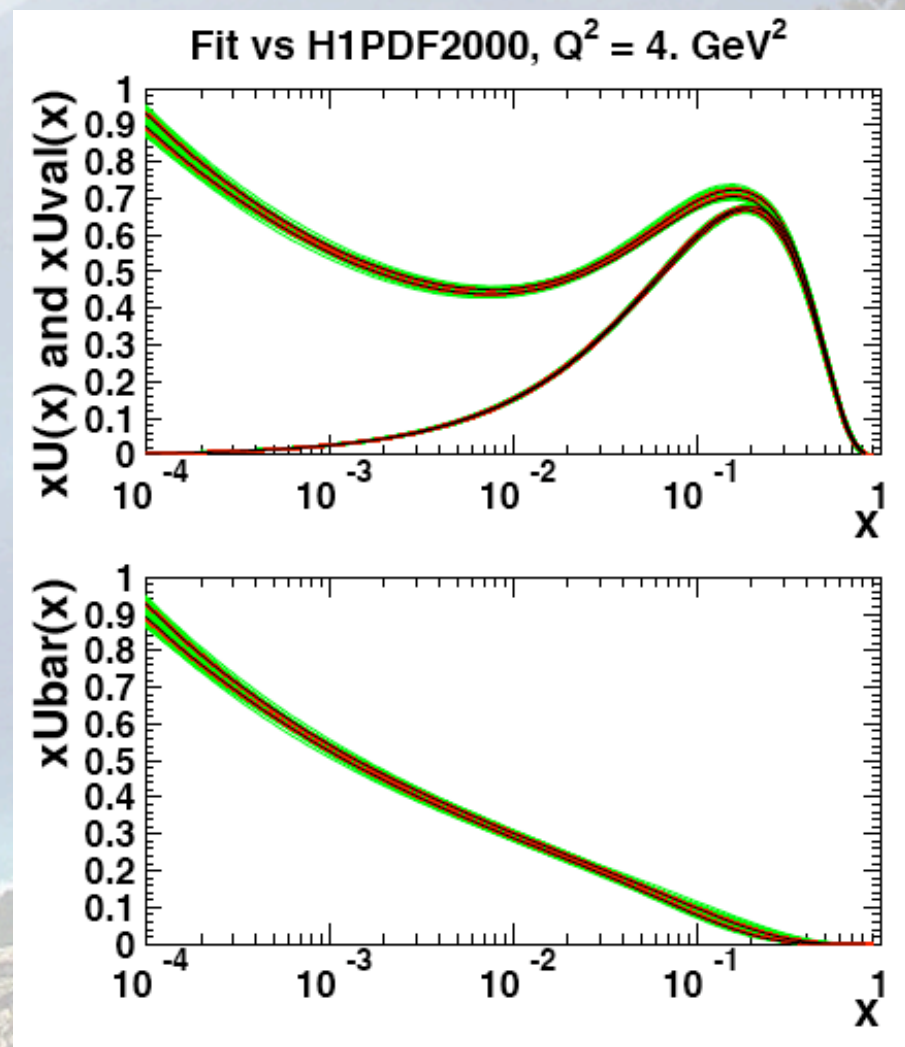
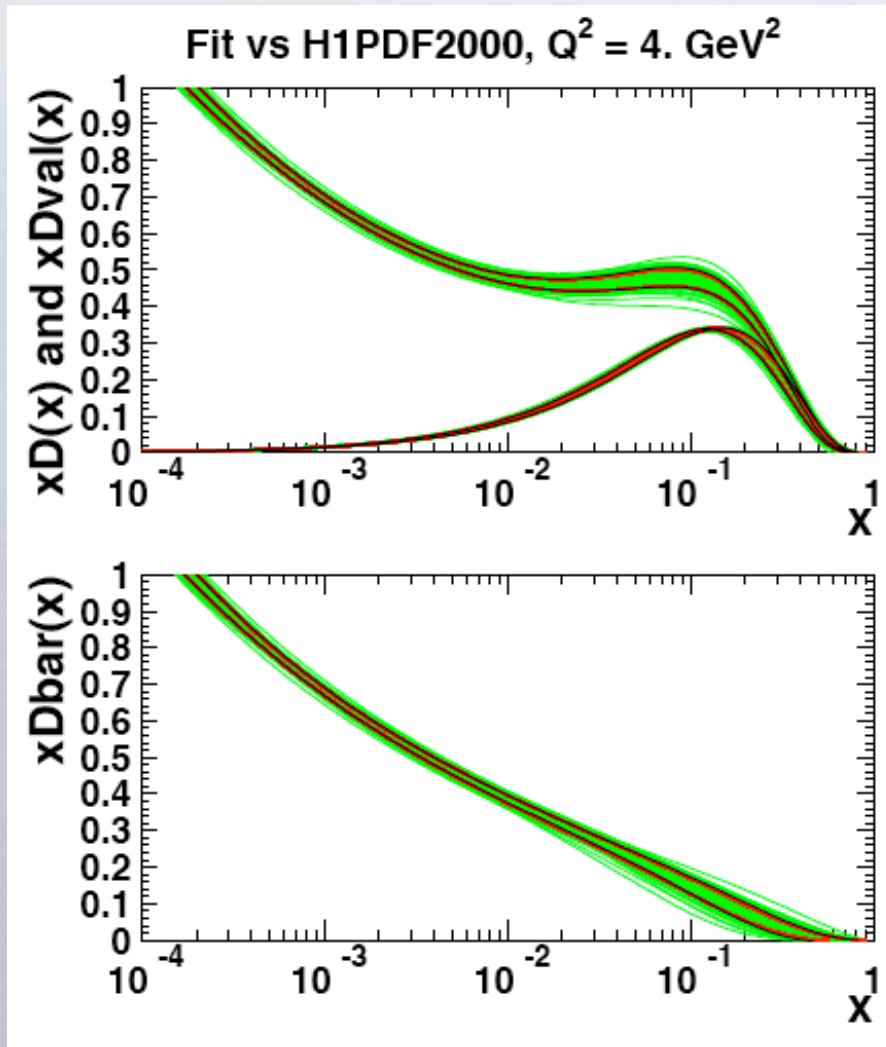
- All gauss, apart from Lumi (log-normal):



Backup



- Stat. gauss, sys log-normal:



Method (IV): more details ...

- Use uniform random number to select the “fluctuation probability” P_j
- Cross section point i has a sensitivity to systematic j δ_{ij}
 - If $\delta_{ij} > 0$, select cross section shifts $\delta\sigma_{ij}$ such that:

$$\int_{-\infty}^{\delta\sigma_{ij}} dx P(x, \delta_{ij}) = P_j$$

- If $\delta_{ij} < 0$, select cross section shifts $\delta\sigma_{ij}$ such that:

$$\int_{\delta\sigma_{ij}}^{+\infty} dx P(x, \delta_{ij}) = P_j$$

Where P is either Gauss, Log-normal, or Uniform distribution

