



The Bias of the Unbiased Estimator

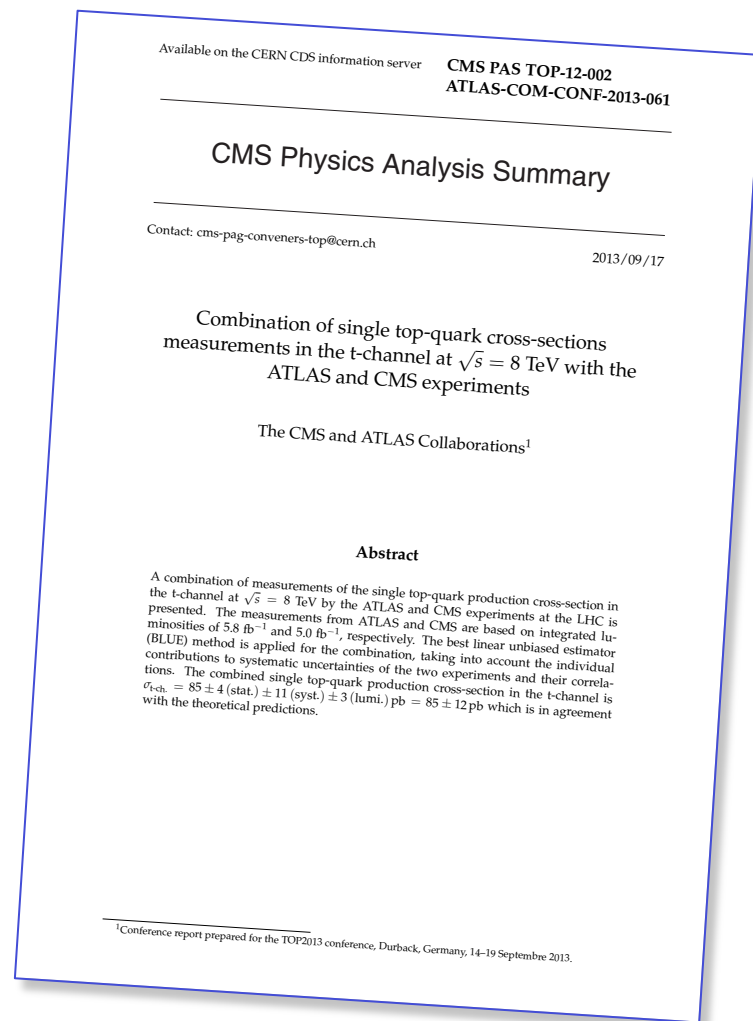
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Introduction



- The ATLAS+CMS combination of single-top production cross-section measurements in the t channel was performed using the BLUE (Best Linear Unbiased Estimator)^[*] method, as many other similar combinations
 - ^[*] L. Lyons et al. NIMA270 (1988) 110
- Compared to its original formulation, relevant contributions to the total uncertainty are known as relative uncertainties
 - Typically: systematic uncertainties
- This spoils the Gaussian assumption of the original formulation and may introduce a bias in BLUE estimate in case of sizable uncertainties (~10% or more)
- The iterative application of the BLUE method was proposed^[**] in the case of a lifetime measurement to reduce such bias
 - ^[*] L. Lyons et al. PRD41(1990)982985
 - For each iteration, uncertainties known as relative contributions are rescaled to the combined value
- We performed tests to motivate this choice for the single-top combination quantifying the bias in the two cases of the 'plain' vs iterative BLUE implementations





BLUE reminder



- Find linear combination of available measurements: $x = \sum w_i x_i$
 - No bias implies $\sum w_i = 1$
- Choose weights to minimize the variance of estimator
 - Take properly into account **correlations** between measurements!
- Equivalent to χ^2 minimization or maximum likelihood for Gaussian uncertainties

Simple example:

L.Lyons et al. NIM A270 (1988) 110

- Two measurements: $x_1 \pm \sigma_1$, $x_2 \pm \sigma_2$ with correlation ρ
- The weights that minimize the χ^2 :

$$\chi^2 = \begin{pmatrix} x_1 - x & x_2 - x \end{pmatrix} \begin{pmatrix} \sigma_1^2 & \rho\sigma_1\sigma_2 \\ \rho\sigma_1\sigma_2 & \sigma_2^2 \end{pmatrix}^{-1} \begin{pmatrix} x_1 - x \\ x_2 - x \end{pmatrix}$$

Cov. matrix

are:

$$w_1 = \frac{\sigma_2^2 - \rho\sigma_1\sigma_2}{\sigma_1^2 - 2\rho\sigma_1\sigma_2 + \sigma_2^2} \quad w_2 = \frac{\sigma_1^2 - \rho\sigma_1\sigma_2}{\sigma_1^2 - 2\rho\sigma_1\sigma_2 + \sigma_2^2} \quad (w_1 + w_2 = 1)$$



- The uncertainty of the combined value is:

$$\sigma_x = \sqrt{\frac{\sigma_1^2 \sigma_2^2 (1 - \rho^2)}{\sigma_1^2 - 2\rho\sigma_1\sigma_2 + \sigma_2^2}}$$

- Basic idea

- Consider simple weighted average
- Errors (σ_i) are supposed to be **true** errors
- But what are available are **estimated** errors
... that may vary with estimated **central value** (τ_i)
- Violates ‘combination principle’: combination of partial combinations differs from combination of all results

$$\hat{\tau} = \frac{\sum \tau_i / \sigma_i^2}{\sum 1 / \sigma_i^2}$$

$$\hat{\tau} = \frac{\sum N_i / \tau_i}{\sum N_i / \tau_i^2}$$

- Solution proposed

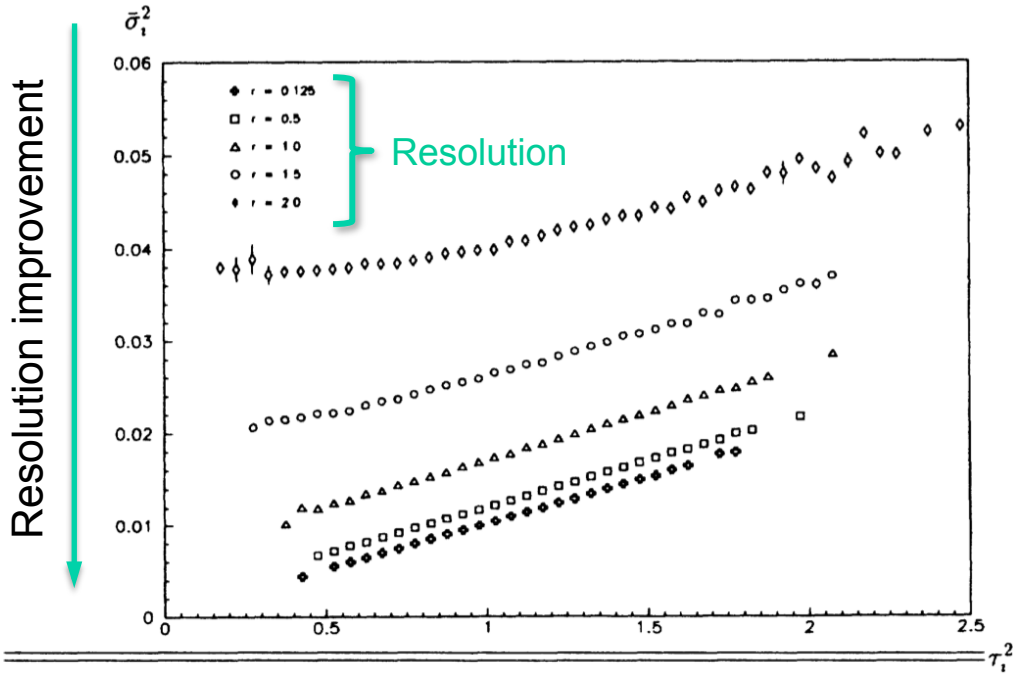
L. Lyons et al., Phys. Rev. D41 (1990) 982985

- LMS: *Locally Matched Solution* (we prefer **iterative BLUE**)
- Bias reduced if covariance matrix determined as if the central value is the one obtained from combination:
 - Rescale uncertainties to combined value
e.g.: if we know a relative uncertainty:

$$\sigma^{\text{rescaled}} = \sigma \cdot x_{\text{blue}} / x_1$$
 - Iterate until central value converges to stable value




Application to B lifetime (Lyons)



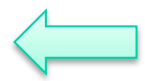
- Dependence of average estimated statistical uncertainties σ_i on lifetime estimate τ_i (squared) determined using toy experiments (100 decays each) and parameterized as:

$$\sigma_i^2(\hat{\tau}) = \frac{k_i \hat{\tau}^2}{N_i} + \frac{r_i^2}{N_i}$$

Resolution 

r		$\langle \hat{\tau} \rangle$	$\langle \Delta \rangle$	$\langle \Delta^2 \rangle^{1/2}$
0.50	(I)	0.991 ± 0.002	0.010	0.014
	(II)	1.000 ± 0.002	0.001	0.001
1.00	(I)	0.989 ± 0.003	0.010	0.014
	(II)	0.997 ± 0.003	0.002	0.003
0.5-1.5	(I)	0.991 ± 0.003	0.011	0.015
	(II)	1.000 ± 0.003	0.002	0.003
2.00	(I)	0.990 ± 0.002	0.011	0.015
	(II)	0.999 ± 0.002	0.003	0.004

Comparison of LMS method (II) with least square procedure (I).



$\Delta = \tau_{\text{fit}} - \tau$ with τ_{fit} the result from one single combined experiment ('combination principle')



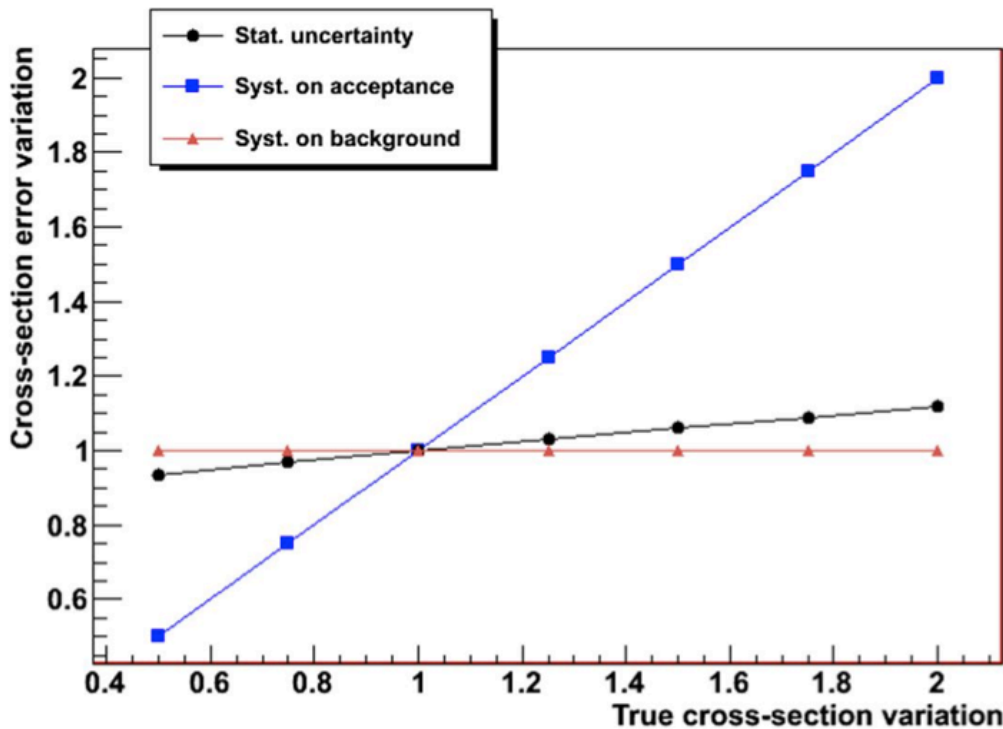
Application to single-top comb.



- Cross section measurements given by:

$$\sigma = \frac{N - B}{\mathcal{L} \cdot A}$$

sel. events \rightarrow $N - B$ \leftarrow bkg. estimate
Int. lumi. \rightarrow $\mathcal{L} \cdot A$ \leftarrow acceptance



σ varied from: 40 \rightarrow 160 pb
 $B = 3000$,
 $A = 0.25\%$
 $L = 5 \text{ fb}^{-1}$
 $\delta A = 5\%$, $\delta B = 5\%$



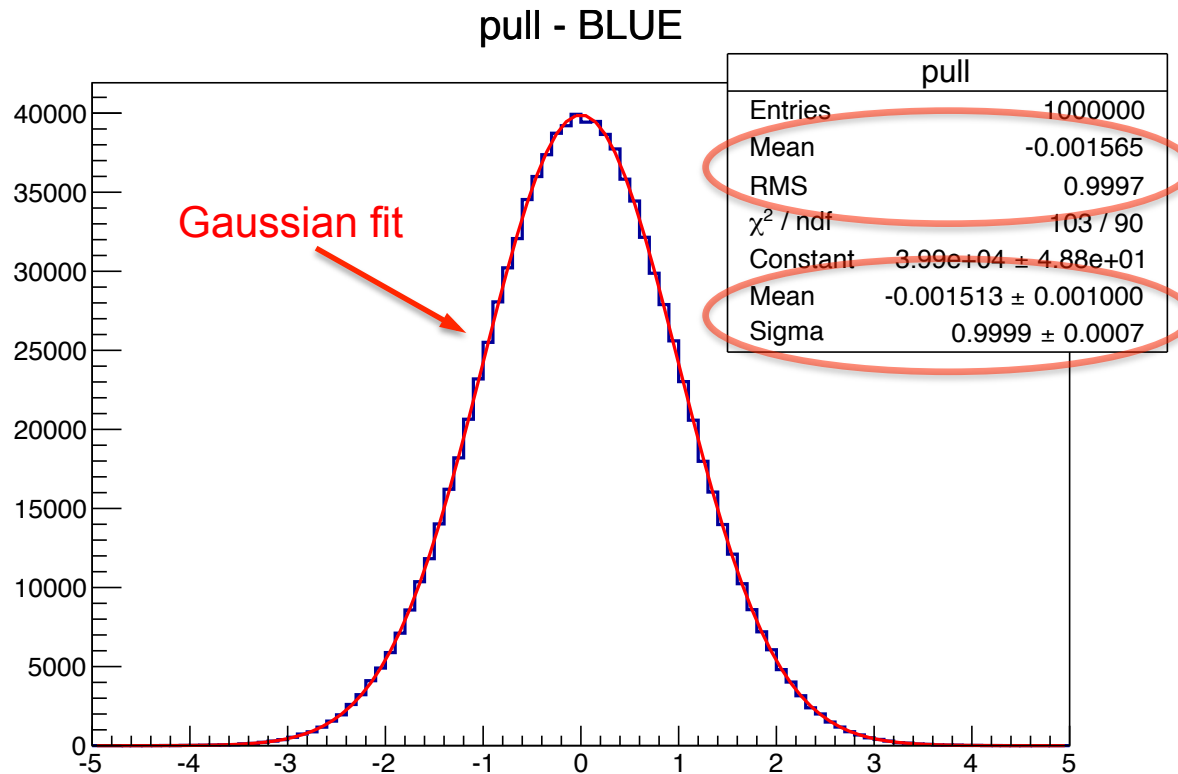
ATLAS+CMS combination



- **ATLAS** (ATLAS-CONF-2013-098):
 - $\sigma_{t\text{-ch.}} = 95.1 \pm 2.4$ (stat.) ± 18.0 (syst.) pb = 95.1 ± 18.1 pb
 - stat: 2.5% (fixed), syst.: 20% (relative uncert.)
- **CMS** (CMS-PAS-TOP-12-002):
 - $\sigma_{t\text{-ch.}} = 80.1 \pm 5.7$ (stat.) ± 11.0 (syst.) ± 4.0 (lumi.) pb = 80.1 ± 13.0 pb
 - stat: 7.1% (fixed), syst.: 15% (relative uncert.)
- Correlation: $\rho = 0.38$ (total uncert.), $\rho = 0.42$ (syst. only)
- Plain BLUE combination:
 - $\sigma_{t\text{-ch.}} = 83.6 \pm 12.1$ pb
- Iterative BLUE combination:
 - $\sigma_{t\text{-ch.}} = 85.3 \pm 12.2$ pb

Plain BLUE method

- Assuming as true uncertainties ATLAS and CMS estimates, with their correlation
- Generate toys according including correlation and apply BLUE (plain)





Test with Toy Monte Carlo

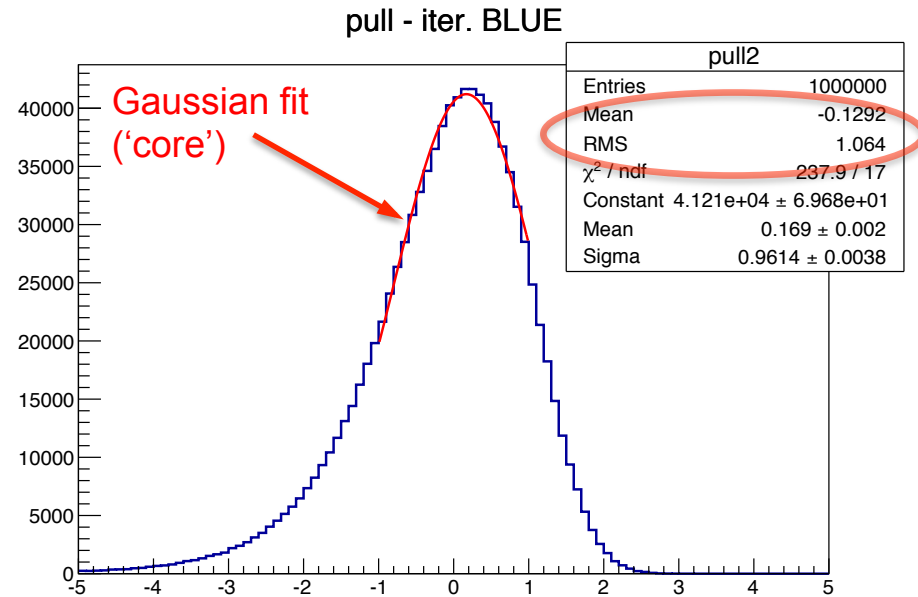
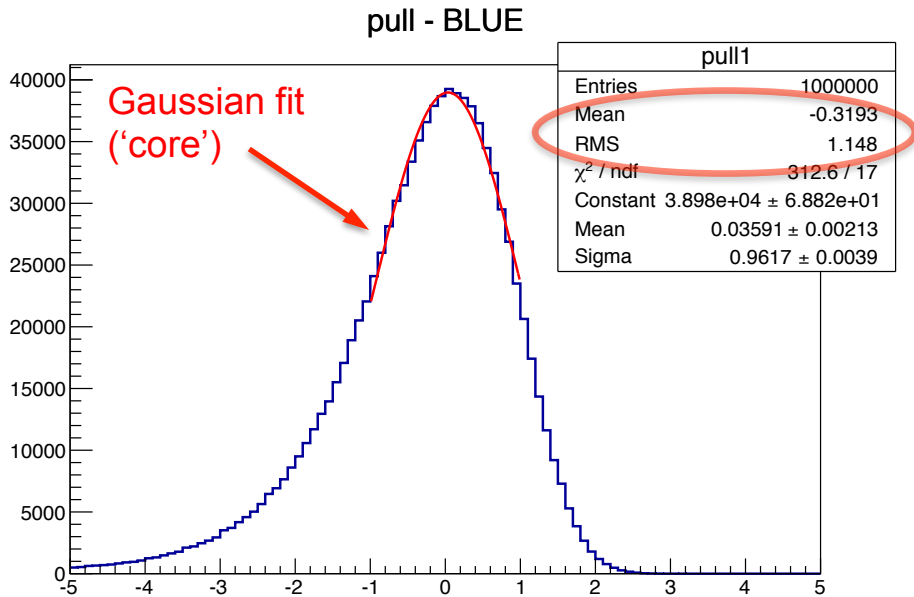


- To assess the possible bias using the two estimators, a simple toy generator was used:
- Assume as **true** value the combined one: **85 pb**
- Assume as **true** fluctuations the **nominal uncertainties** of the two experiments with their **correlation**
 - **Nominal uncertainty: stat + syst added in quadrature of each experiment**
- Two central values are extracted for ATLAS and CMS according to the know uncertainties, their correlation and the true central value
 - **Fixed stat errors, rel. syst errors**
- Relative (systematic) uncertainties are **rescaled** to the extracted central values for ATLAS and CMS



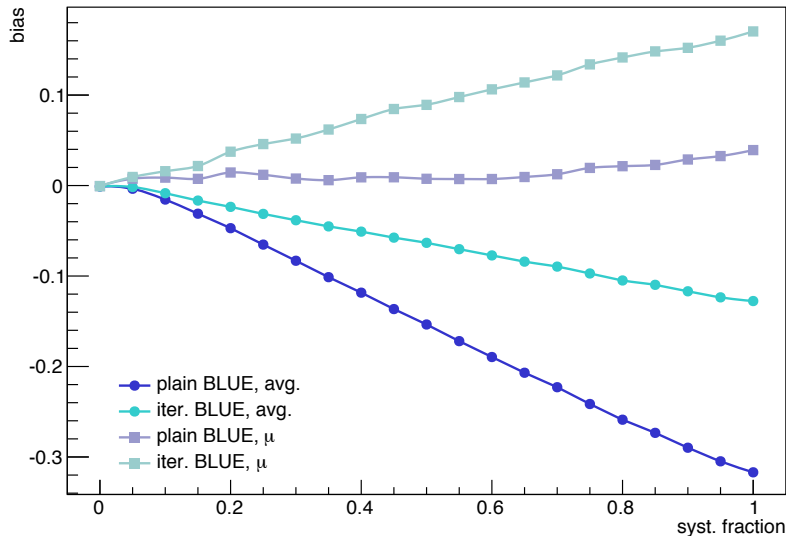
Iterative vs plain BLUE

- Bias: $-0.32\sigma \sim 3.9\text{pb}$ (plain) $\rightarrow -0.13\sigma \sim 1.6\text{pb}$ (iterative)
- RMS: 1.15 (plain) \rightarrow 1.06 (iterative): **uncertainty underestimate**
- Gaussian μ : 0.04 (plain) \rightarrow 0.169 (iterative)
- The iterative BLUE method reduces the bias by **shifting the 'core' of the distribution** and by **shrinking the non-Gaussian tails**

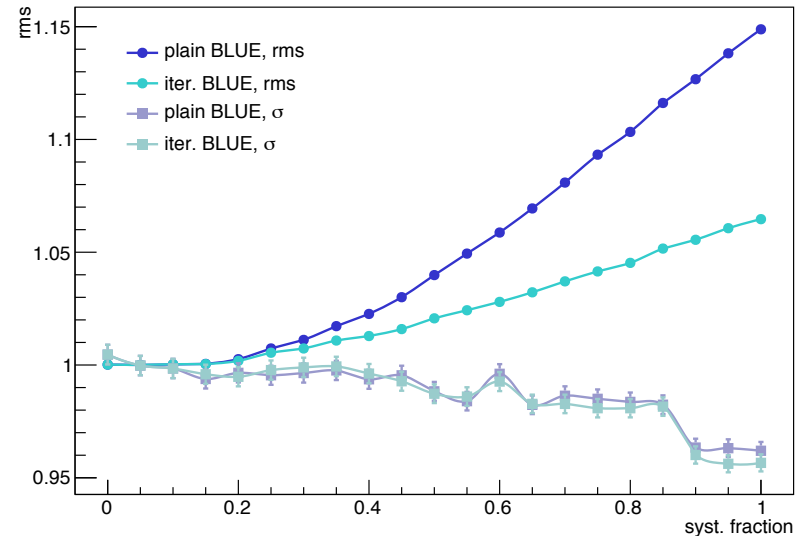


- The systematic contribution was varied from the nominal values to zero rescaling it by a fraction f going from 0 to 1
- Bias was measured varying the fraction f of the actual uncertainty
- As expected, the bias goes to zero as f goes to zero
 - Iterative BLUE is not an issue for precision measurements (e.g.: m_t)
- The iterative BLUE has in general smaller RMS and less bias

BLUE pull: mean



BLUE pull: rms



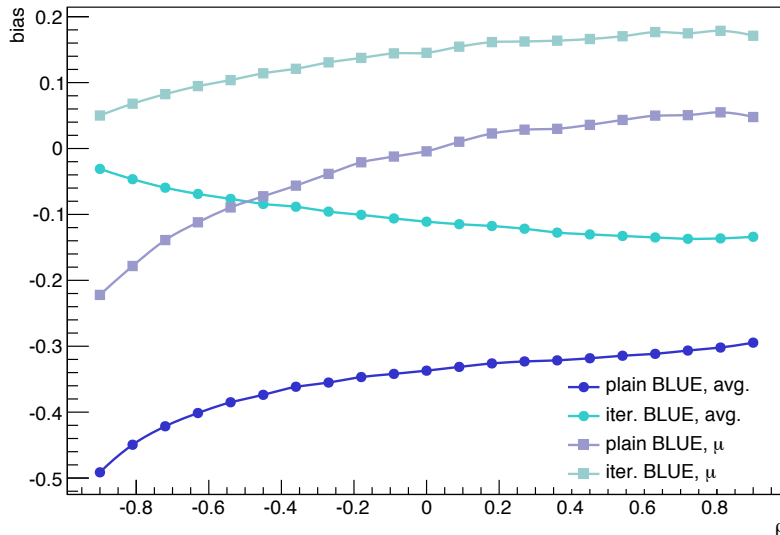


Dependence on correlation

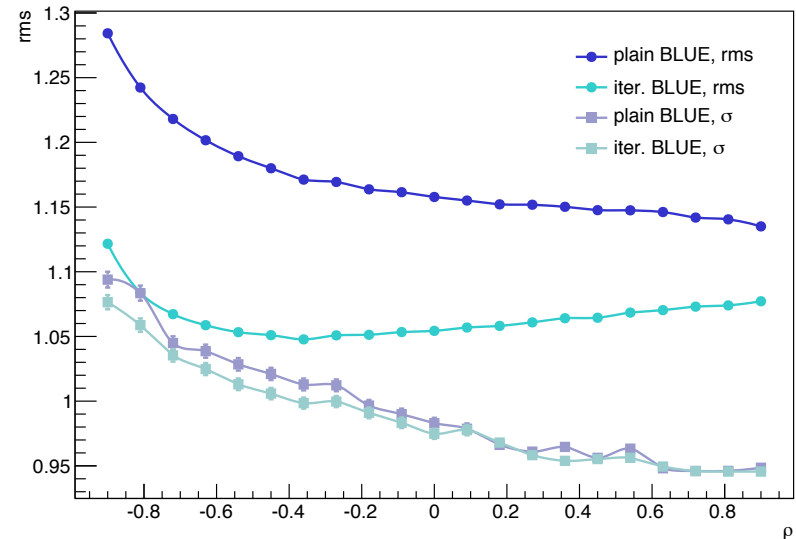


- Correlation between systematic uncertainties was varied from $\rho=0$ to $\rho=1$
- Bias was measured varying the correlation ρ
- The iterative BLUE has in general smaller RMS and less bias

BLUE pull: mean



BLUE pull: rms



What we have learned

- Iterative BLUE method in general reduce biases
 - Better behaved than standard BLUE method, but not perfect either
 - In case of single top combination: **difference is minor**
- Not all uncertainties should be rescaled
 - Data-driven bkg. uncertainties
 - Stat uncertainty (at 1st order)
 - Others ?
- More investigations ongoing...

