



12th International Workshop on Top Quark Physics
24 September 2019 - Beijing

Running of the top quark mass from pp collisions at $\sqrt{s} = 13$ TeV

results from [arXiv:1909.09193](https://arxiv.org/abs/1909.09193)
(submitted to Phys. Lett. B)

Matteo Defranchis (DESY) - on behalf of the CMS Collaboration

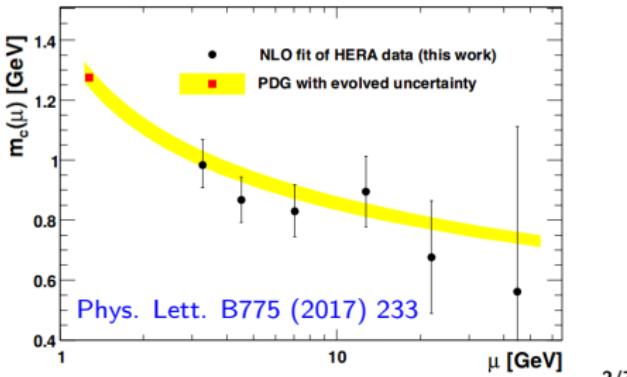
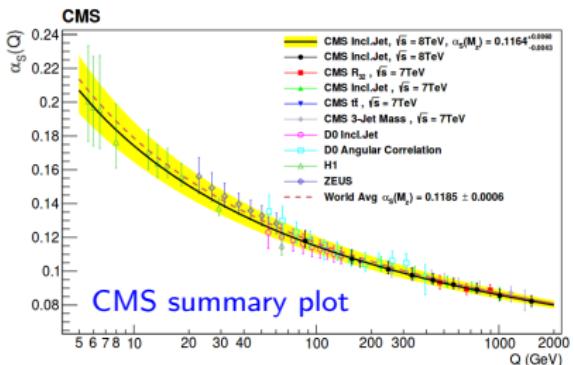
introduction and motivation

in $\overline{\text{MS}}$ scheme, running of QCD parameters (α_S , m_q) described by a set of RGEs

$$\text{For } m_q: \quad \mu^2 \frac{dm_q}{d\mu^2} = -\gamma(\alpha_S) m_q$$

anomalous dimension γ calculated pQCD, and can be modified by BSM physics

- running of α_S experimentally verified on a wide range of scales
 - running of m_c and m_b investigated at HERA and LEP experiments
- running of m_t experimentally investigated for the first time



starting point and final goal

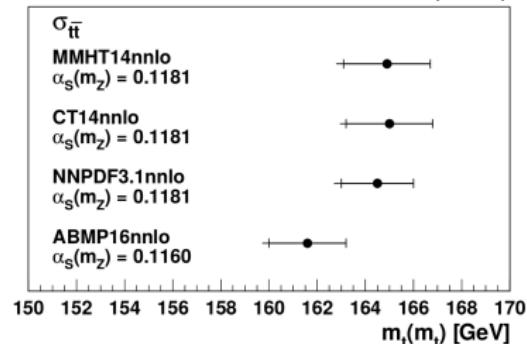
Eur. Phys. J. C79 (2019) 368

CMS

35.9 fb⁻¹ (13 TeV)

inclusive analysis (presented last year)

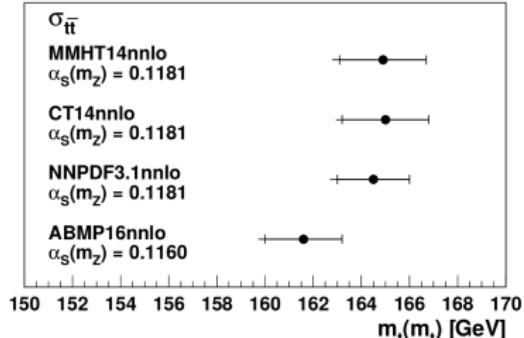
- simultaneous measurement of inclusive $\sigma_{t\bar{t}}$ and m_t^{MC} from likelihood template fit
- $m_t(m_t)$ extracted in $\overline{\text{MS}}$ scheme @NNLO from measured $\sigma_{t\bar{t}}$



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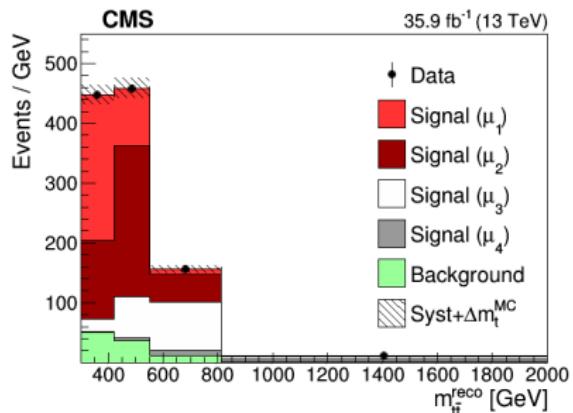
running: measure $m_t(\mu)$ as a function of the scale $\mu = m_{t\bar{t}}$

- perform precise measurement of $d\sigma_{t\bar{t}}/dm_{t\bar{t}}$
- extract running by comparing to differential theory predictions in $\overline{\text{MS}}$ scheme

2016 data: 35.9 fb^{-1} (13 TeV)

offline selection

- $e^\mp\mu^\pm$ with $p_{T1(2)} > 25(20) \text{ GeV}$
- jets with $p_T > 30 \text{ GeV}$ considered
- b-tagging used to classify events
- kinematic reconstruction of $t\bar{t}$ system
in events with ≥ 2 jets $\rightarrow m_{t\bar{t}}^{\text{reco}}$



signal definition and scale choice

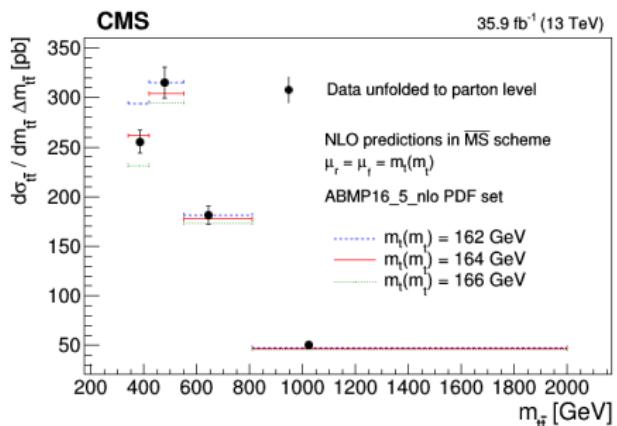
- $t\bar{t}$ signal split into 4 subsamples in bins of parton-level $m_{t\bar{t}}$
- each subsample treated as independent signal, and corresponds to a bin in $d\sigma_{t\bar{t}}/dm_{t\bar{t}}$
- representative scale μ_k assigned to each signal:
 $\mu_k = \text{centre-of-gravity of parton-level } m_{t\bar{t}}$

bin	range [GeV]	μ_k [GeV]
1	< 420	384
2	420-550	476
3	550-810	644
4	> 810	1024

- fit performed in categories of b-jet multiplicity and bins of $m_{t\bar{t}}^{\text{reco}}$
- systematic uncert. constrained within visible phase space
- dependence on m_t^{MC} fully incorporated in the fit

response matrix embedded in the likelihood \Rightarrow maximum likelihood unfolding to parton-level

- $m_t(m_t)$ extracted in each bin of $m_{t\bar{t}}$ independently via χ^2 fit of theory predictions to data
- $m_t(m_t)$ converted to $m_t(\mu_k)$ using one-loop RGE solutions ($n_f = 5$)



NLO differential calculations obtained with version of MCFM where m_t is treated in $\overline{\text{MS}}$ scheme ([EPJ C74 \(2014\) 3167](#))

extraction of the running

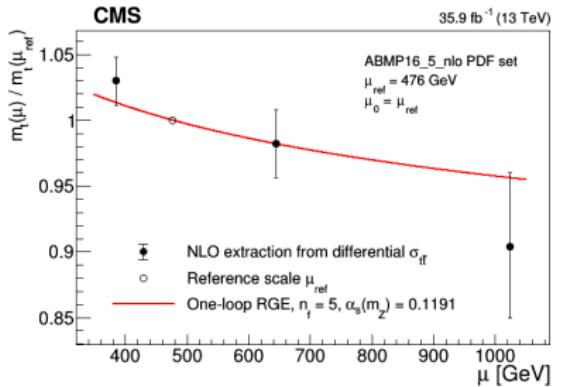
running $r(\mu)$ is defined as ratio of $m_t(\mu)$ to reference mass $m_t(\mu_{\text{ref}})$

$$\text{th: } r(\mu) = m_t(\mu)/m_t(\mu_{\text{ref}})$$

$$\text{exp: } r_k = m_t(\mu_k)/m_t(\mu_{\text{ref}})$$

- $r(\mu)$ depends solely on RGE
- r_k benefits from cancellation of correlated uncertainties

→ choice: $\mu_{\text{ref}} = \mu_2 = 476 \text{ GeV}$



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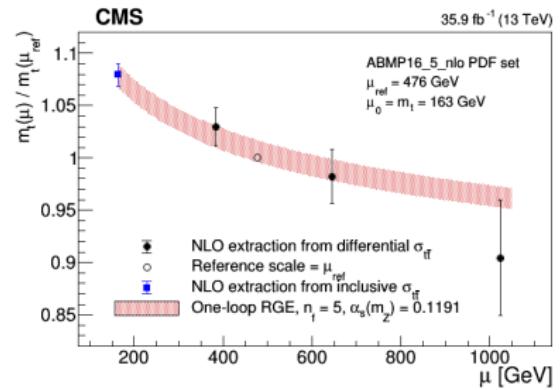
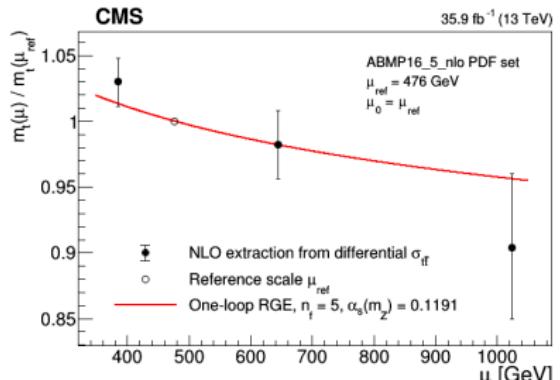
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- $r(\mu)$ depends solely on RGE
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→ choice: $\mu_{\text{ref}} = \mu_2 = 476 \text{ GeV}$

- result compared to value of $m_t(m_t)$ extracted at NLO from inclusive $\sigma_{t\bar{t}}$
- good agreement with RGE on a wide range of scales, up to $\mu > 1 \text{ TeV}$

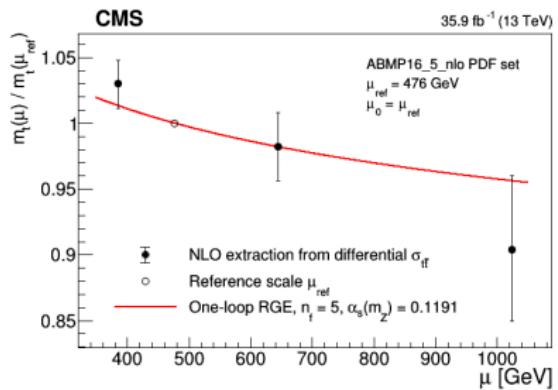
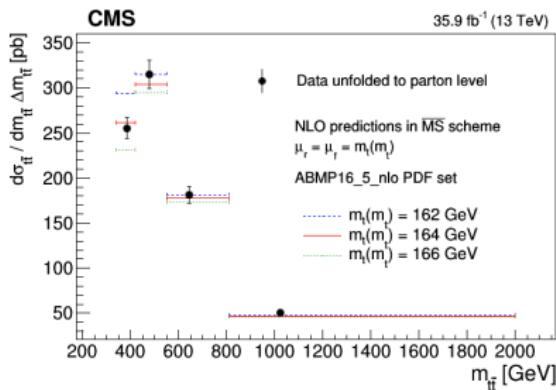


- first experimental investigation of running of the top quark mass
- good agreement with RGE, up to $\mu > 1$ TeV
- looking forward to NNLO calculations in the $\overline{\text{MS}}$ scheme to probe the running at two-loops precision

summary and outlook in a nutshell

- first experimental investigation of running of the top quark mass
- good agreement with RGE, up to $\mu > 1$ TeV
- looking forward to NNLO calculations in the $\overline{\text{MS}}$ scheme to probe the running at two-loops precision

Thank you for your attention!



BACKUP



b-tagging efficiencies

b-tagging efficiencies are determined *in situ* by exploiting the $t\bar{t}$ topology, separately in each bin of $m_{t\bar{t}}$

$$\begin{aligned} S_{1b}^k &= \mathcal{L}\sigma_{t\bar{t}}^{(\mu_k)} A_{\text{sel}}^k \epsilon_{\text{sel}}^k 2\epsilon_b^k (1 - C_b^k \epsilon_b^k) \\ S_{2b}^k &= \mathcal{L}\sigma_{t\bar{t}}^{(\mu_k)} A_{\text{sel}}^k \epsilon_{\text{sel}}^k C_b^k (\epsilon_b^k)^2 \\ S_{\text{other}}^k &= \mathcal{L}\sigma_{t\bar{t}}^{(\mu_k)} A_{\text{sel}}^k \epsilon_{\text{sel}}^k [1 - 2\epsilon_b^k (1 - C_b^k \epsilon_b^k) - C_b^k (\epsilon_b^k)^2] \end{aligned}$$

- ϵ_{sel}^k is the efficiency of the full selection in $m_{t\bar{t}}$ bin k
 - ϵ_b^k is the b-tagging efficiency in $m_{t\bar{t}}$ bin k
 - C_b^k represents the residual correlation of tagging the two b-jets
- all parameters are derived by the simulation and depend on the systematic uncertainties

binned Poisson Likelihood

$$\begin{aligned}L &= \prod_i \frac{e^{-\nu_i} \nu_i^{n_i}}{n_i!} \prod_j \pi(\omega_j) \prod_m \pi(\lambda_m) \\ \nu_i &= \sum_{k=1}^4 s_i^k(\sigma_{t\bar{t}}^{(\mu_k)}, \vec{\lambda}, m_t^{\text{MC}}) + \sum_j b_i^j(\omega_j, \vec{\lambda})\end{aligned}$$

- $\vec{\lambda}$ is the set of nuisance parameters
- ω_j is the normalization of background source j
- $\pi(\lambda_m)$ and $\pi(\omega_j)$ parametrize the prior knowledge of m^{th} nuisance parameter and j^{th} background normalization

fit performed in categories of b-tagged jet multiplicity and bins of $m_{t\bar{t}}^{\text{reco}}$

- b-tag categories constrain b-tagging efficiencies
- $m_{t\bar{t}}$ categories sensitive to different signals → constrain $\sigma_{t\bar{t}}^{(\mu_k)}$
- categories with < 2 jets, where kin reco cannot be performed, included in fit
 - increases visible phase space ⇒ **reduces extrapolation uncertainties**

for each category and sub-category, suitable differential distribution fitted

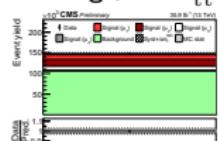
- $m_{\ell b}^{\min}$ distribution used to constrain m_t^{MC}
- p_T of softest jet in event used to constrain JES
- systematic uncertainties profiled in Poisson likelihood and constrained in the visible phase space
- additional uncertainties assigned to extrapolation to full phase space
(constraints in modelling uncertainties NOT considered in extrapolation)

→ this procedure yields results that are **unfolded to the parton level with maximum likelihood method**

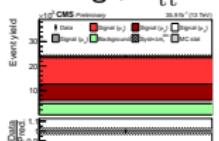
post-fit distributions of fit inputs

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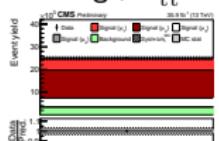
0 b-tags, no $m_{\text{tt}}^{\text{reco}}$



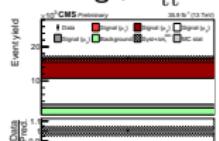
0 b-tags, $m_{\text{tt}}^{\text{reco}}$ 1



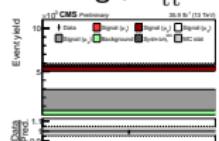
0 b-tags, $m_{\text{tt}}^{\text{reco}}$ 2



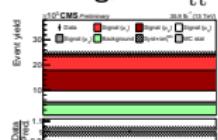
0 b-tags, $m_{\text{tt}}^{\text{reco}}$ 3



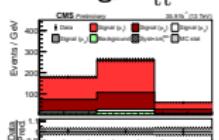
0 b-tags, $m_{\text{tt}}^{\text{reco}}$ 4



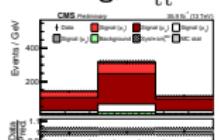
1 b-tag, no $m_{\text{tt}}^{\text{reco}}$



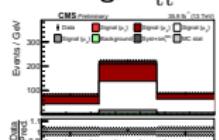
1 b-tag, $m_{\text{tt}}^{\text{reco}}$ 1



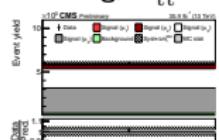
1 b-tag, $m_{\text{tt}}^{\text{reco}}$ 2



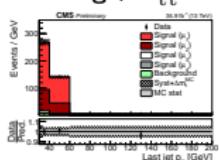
1 b-tag, $m_{\text{tt}}^{\text{reco}}$ 3



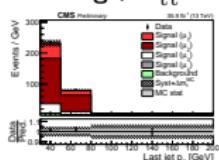
1 b-tag, $m_{\text{tt}}^{\text{reco}}$ 4



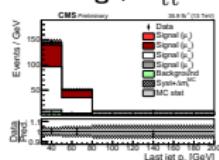
2 b-tags, $m_{\text{tt}}^{\text{reco}}$ 1



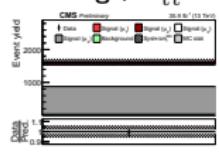
2 b-tags, $m_{\text{tt}}^{\text{reco}}$ 2



2 b-tags, $m_{\text{tt}}^{\text{reco}}$ 3



2 b-tags, $m_{\text{tt}}^{\text{reco}}$ 4



"no $m_{\text{tt}}^{\text{reco}}$ " = events with less than 2 jets

- differential predictions @NLO obtained with version of MCFM where m_t treated in $\overline{\text{MS}}$ scheme ([Eur. Phys. J. C74 \(2014\) 3167](#))
- only theory calculation available with top mass in $\overline{\text{MS}}$ scheme
- evolution of QCD couplings at 1-loop, 5 flavours
- scale choice: $\mu_r = \mu_f = m_t(m_t)$
- interfaced with ABMP16_5_nlo PDF set: only available PDF set with m_t in $\overline{\text{MS}}$ scheme, consistently with calculation

measurement of the slope

essence of this measurement: extract slope of NLO running, taking $m_t(\mu_2) = m_t(\mu_{\text{ref}})$ as reference

- $r(\mu) = m_t(\mu)/m_t(\mu_2)$
- $r_{k2} = m_t(\mu_k)/m_t(\mu_2), k = 1, 3, 4$

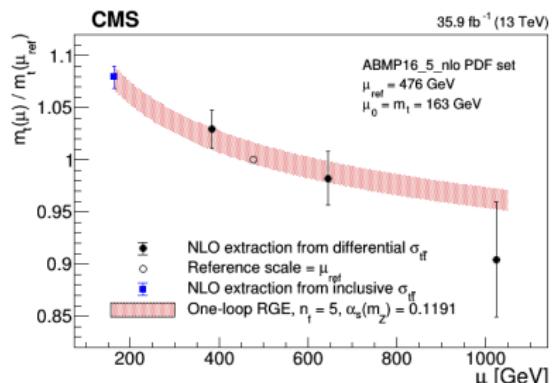
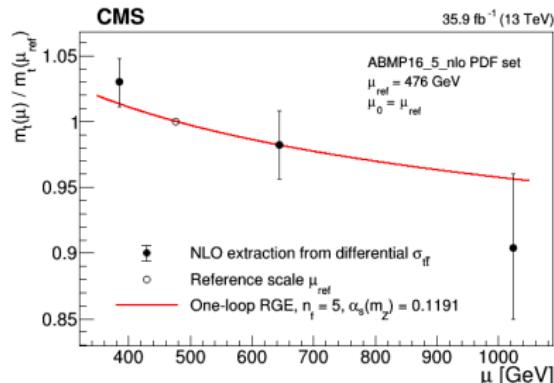
advantages

- slope $r(\mu)$ directly related to RGE prediction
- ratios r_{k2} benefit from partial cancellation of correlated uncertainties
- $\mu_{\text{ref}} = \mu_2$ to minimize correlations between extracted ratios

uncertainties in the ratios

- fit and extrapolation
- PDF and α_S from ABMP eigenvectors
- scale variations in MCFM not meaningful here, as scale dependence is being investigated
- all correlations properly taken into account

running of m_t : results



→ observed running consistent with RGE

$$r_{12} = m_t(\mu_1)/m_t(\mu_2) = 1.030 \pm 0.018 \text{ (fit)} \quad {}^{+0.003}_{-0.006} \text{ (PDF + } \alpha_S \text{)} \quad {}^{+0.003}_{-0.002} \text{ (extr)}$$

$$r_{32} = m_t(\mu_3)/m_t(\mu_2) = 0.982 \pm 0.025 \text{ (fit)} \quad {}^{+0.006}_{-0.005} \text{ (PDF + } \alpha_S \text{)} \quad {}^{+0.004}_{-0.004} \text{ (extr)}$$

$$r_{42} = m_t(\mu_4)/m_t(\mu_2) = 0.904 \pm 0.050 \text{ (fit)} \quad {}^{+0.019}_{-0.017} \text{ (PDF + } \alpha_S \text{)} \quad {}^{+0.017}_{-0.013} \text{ (extr)}$$

observed running parametrized as

$$f(x, \mu) = x [r(\mu) - 1] + 1$$

such that

- $f(1, \mu) = r(\mu) \rightarrow \text{RGE running}$
- $f(0, \mu) = 1 \rightarrow \text{no running}$

x_{\min} extracted from χ^2 fit to r_{k2} :

- correlations in extracted ratios studied with toy experiment procedure
- correlations fully taken into account in estimate of x_{\min}

$$x_{\min} = 2.05 \pm 0.61 \text{ (fit)} \quad {}^{+0.31}_{-0.55} \text{ (PDF + } \alpha_S \text{)} \quad {}^{+0.24}_{-0.49} \text{ (extr)}$$

→ compatible with RGE within 1.1σ