Determination of the Top-Quark Mass from the  $m_{lb}$  Distribution

in Dileptonic Top-Quark Pair Events at  $\sqrt{s}$ = 8 TeV

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- Motivation -

- Event Selection & m<sub>lb</sub> Definition-
  - Extraction Method & Results-







- Inclusive tt cross sections used to extract  $m_{t,pole}$  at NNLO
- Taking advantage of different slope of  $\sigma_{pred}(m_t)$  and  $\sigma_{meas}(m_t)$



### Next step:

(PLB 728 (2014) 496)

- Extract m<sub>t</sub> from differential cross sections
- Possibly more sensitive to m<sub>t</sub>, more clear theoretical interpretation
- Step to simultaneous  $m_t$ , pdf,  $\alpha_s$  extraction
- $\rightarrow$  Observable sensitive to  $m_t$ , measurement + prediction

#### Reference: CMS PAS-TOP-14-014





- Full 2012, 8 TeV dataset (19.7 fb<sup>-1</sup>)
- Signal: MadGraph+Pythia (7 mass-points)
- Selection
  - ► ≥2 opposite charged isolated leptons **e**µ
  - Remove low  $m_{ll} < 20 \text{ GeV}$
  - ▶ ≥2 high p<sub>T</sub> jets
  - ▶ ≥1 b-tagged jet











### Definition on **generator** and **reconstruction** level



 $m_{lb,min}$ : smaller  $m_{lb}$  of 2 combinations

- Leading b-jet: smaller JES uncertainties
- Lepton kinematics known to high precision
- ➡ Visible phase space (vis. PS)

- leading b-quark from top / leading b-jet
  |η| < 2.4, p<sub>T</sub> > 30 GeV
  - leptons from top (e/µ/tau(→e,µ)) / selected oppositely charged lepton pair
  - |η| < 2.4, p<sub>T</sub> > 20 GeV





Observable  $m_{lb,min}(2)$ 



- Good data/MC agreement
- Low background
- Only small contribution of migrations from outside visible PS





 τ decays, efficiencies included in response matrix

$$\vec{x}_{reco} = \mathcal{L} \cdot M^{resp} \quad \vec{x}_{pred}$$

• Strong correlation between reconstructed and generated m<sub>lb,min</sub>







- Compare reconstruction level distributions (MadGraph + detector sim.)
- Compare other prediction to data on reconstruction level?
- Fold theory:

$$M_{ij}^{resp} = N_{ij} / \sum_{j=0}^{n} N_{ij}$$
  
 $\vec{x}_{reco} = \mathcal{L} \cdot M^{resp} \quad \vec{x}_{pred}$ 

- Separate matrix for each systematic variation and m<sub>t</sub>
  - Detector modeling uncertainties
  - Signal modeling uncertainties (reconstruction efficiencies)



- No additional unfolding/regularization uncertainties
- No stat. correlations between bins of reconstructed distribution
- Here: Fold prediction within visible phase space
  - Reconstruction level prediction can be directly compared to measured points



## Method Overview





prediction: either or + data



- For each m<sub>lb</sub> bin:
  - Derive cont. mt dependence of pred (fit)
  - Construct  $\chi^2$  (m<sub>t</sub>) per bin
- Combine bins
  - No stat correlations  $\rightarrow \text{sum } \chi^2 \text{ over bins}$
- Determine m<sub>t</sub>
  - Repeat for each syst. variation







Likelihood

rate shape & rate shape

 $m_t$ 





- Fit mt-dependence of prediction, N<sub>pred</sub>. with 2<sup>nd</sup> order polynomial
  - Includes backgrounds
  - Use statistical uncertainty (mostly background) from central point  $\rightarrow \sigma_{\text{pred}}$
  - Parametrization uncertainties not included here  $(\rightarrow \text{see systematics})$
- Data yield N<sub>data</sub>: constant with  $\sigma_{data}$ ۲





• Define estimator  $\chi^2 = \frac{(N_{\text{pred}} - N_{\text{data}})^2}{\sigma_{\text{pred}}^2 + \sigma_{\text{data}}^2}$ 

- Extract top mass from minimum of  $\chi^2(m_t)$
- For total rate or in bins of any ۲ distribution  $(m_{lb})$







- MadGraph+Pythia + background processes
- Compare data to different m<sub>t</sub> hypothesis in prediction
  - for shape only, shape & rate and rate-only
- Repeated for each systematic uncertainty





- Use MCFM to simulate m<sub>lb,min</sub> distribution in. vis PS
- Fold to detector level and extract top mass
  - normalized (rate underestimated)
- Pure MCFM NLO / LO
- NLO production + decay 1
  - Real and virt. corrections in decay & production

## NLO production + decay 2

- Real and virt. corrections in production only (Closest to MadGraph ME level)
- Plain LO (no FSR)





• Uncertainties similar to MadGraph+Pythia





- $M_{lb}$  shows low  $m_t$ -sensitivity wrt. higher orders in production
- Overall consistent results from "shape", "shape & rate" and "rate" based extraction

# Dominant uncertainties (experimental/theory):

- Rate:
  - normalization uncertainties, JES
  - signal normalization
- Shape & rate:
  - partial cancellation of JES uncertainties
  - signal normalization
- Shape  $\rightarrow$  smallest total uncertainty
  - JES
  - Q<sup>2</sup> scale

	Fitted $m_t$ [GeV]
Fit method	from m <sup>min</sup> lb
shape+rate	$173.1 \substack{+1.9 \\ -1.8}$
rate	$173.7 \substack{+3.5 \\ -3.4}$
shape	$172.3 \substack{+1.3 \\ -1.3}$
shape	$171.5 \substack{+1.1 \\ -1.1}$
shape	$171.4 \substack{+1.0 \\ -1.1}$
	Fit method shape+rate rate shape shape shape shape





## Summary



- Extracted top mass from reco-level m<sub>lb</sub> distributions & total rate with MadGraph+Pythia
  - From shape: mt = 172.3 ±1.3 GeV
  - All "MadGraph+Pythia mass"-values consistent
  - Partial cancellation of uncertainties in shape+rate extraction
- Extracted using folded MCFM LO/NLO
  - No unfolding/regularization effects
  - Low sensitivity of m<sub>lb</sub> to production mechanism









## BACKUP





- Based on fitted  $N_{pred}(m_t)$  (per bin)
- Data toys:
  - Random yield (Poisson) around N<sub>pred</sub>
  - Same statistics as data
- Prediction toys:
  - Signal
    - Random yield (Poisson) with tt MC statistics
    - Uncorrelated for each mass point
  - Background
    - Random yield (Poisson) with background MC statistics
    - Correlated for each mass point
  - Scaled back to luminosity and added
- Extract  $m_t$  and evaluate  $m_{t,in}$  vs.  $m_{t,out}$  (x 10<sup>5</sup>)
- ⇒No bias
- Stat uncertainties estimated correctly



