

Kirchhoff  
Institut  
für Physik

# Mitigating detector effects in measurements with the ATLAS Experiment

Thomas Spieker  
Kirchhoff Institute for Physics, Heidelberg  
May 2, 2017



Bundesministerium  
für Bildung  
und Forschung



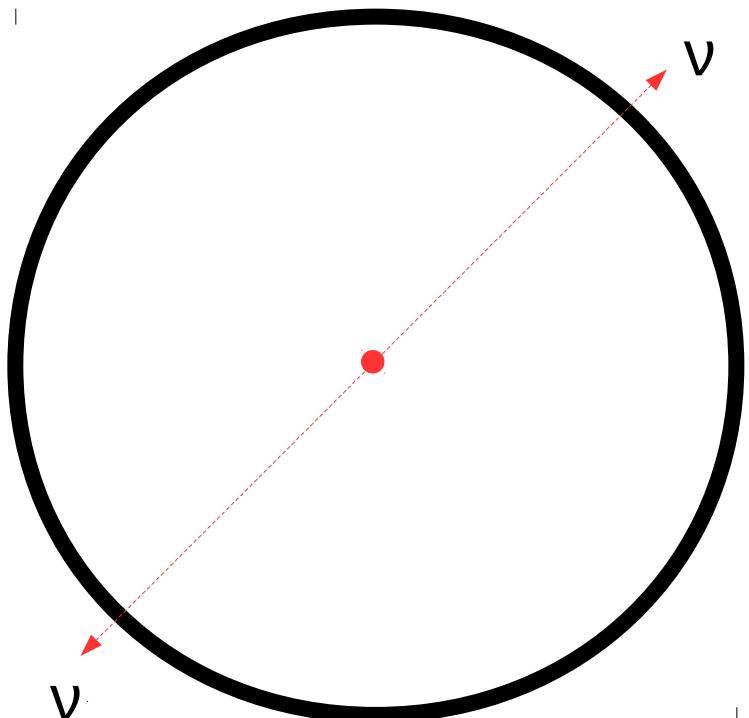
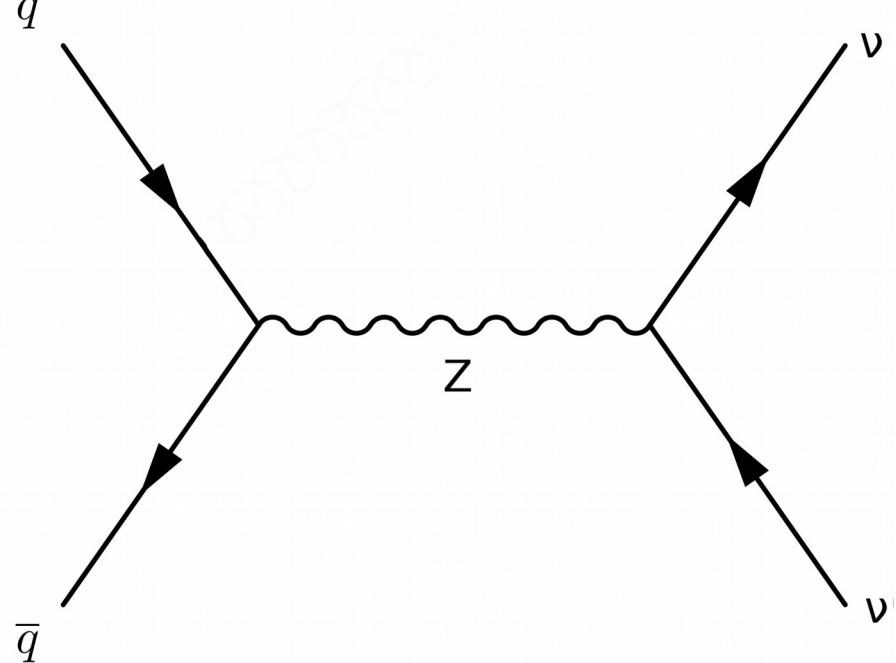
INTERNATIONAL  
MAX PLANCK  
RESEARCH SCHOOL

PT  
FS

FOR PRECISION TESTS  
OF FUNDAMENTAL  
SYMMETRIES

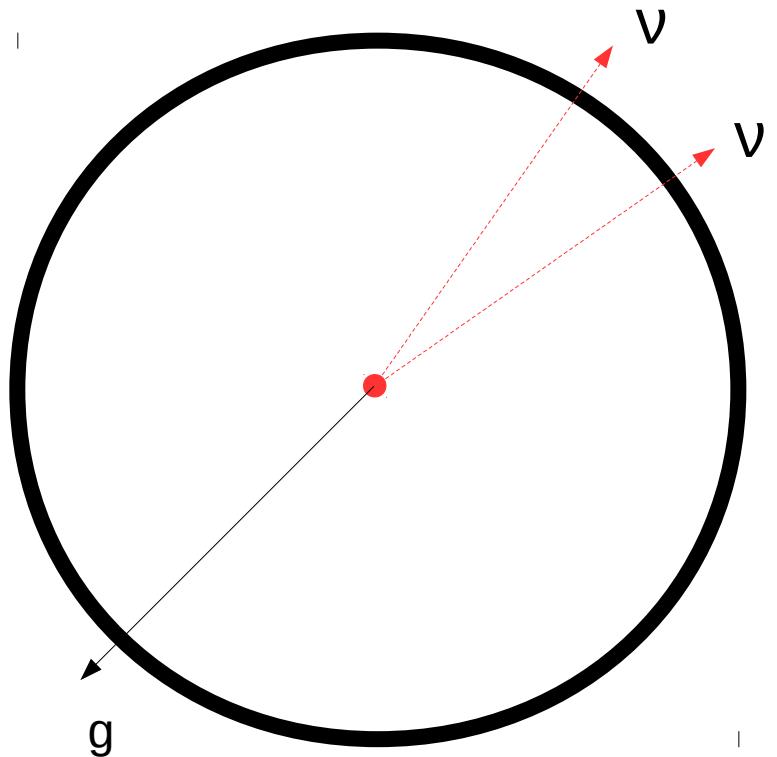
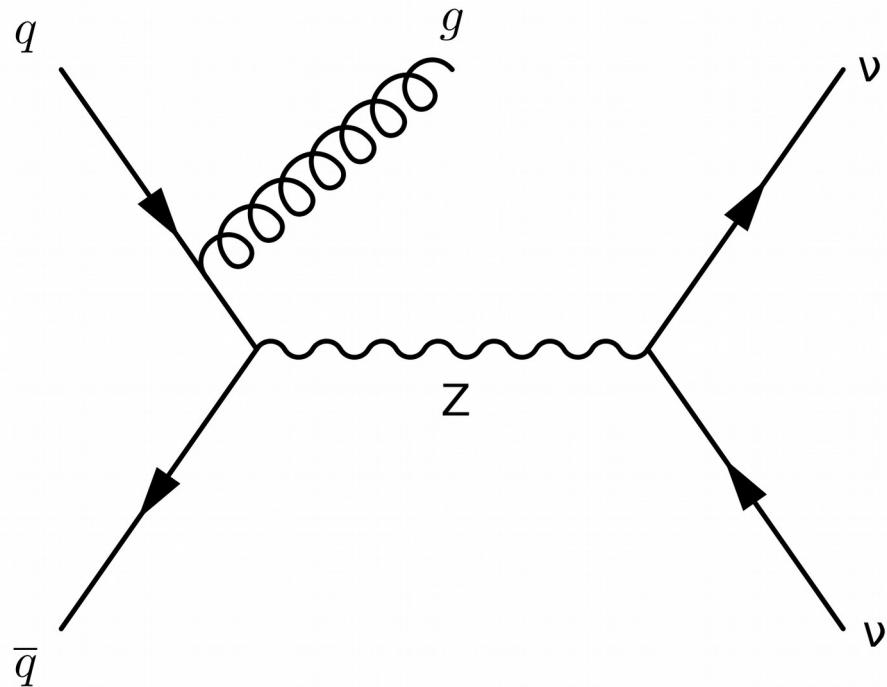
# The measurement: $Z \rightarrow \nu\nu$

Want to measure  $Z \rightarrow \nu\nu$



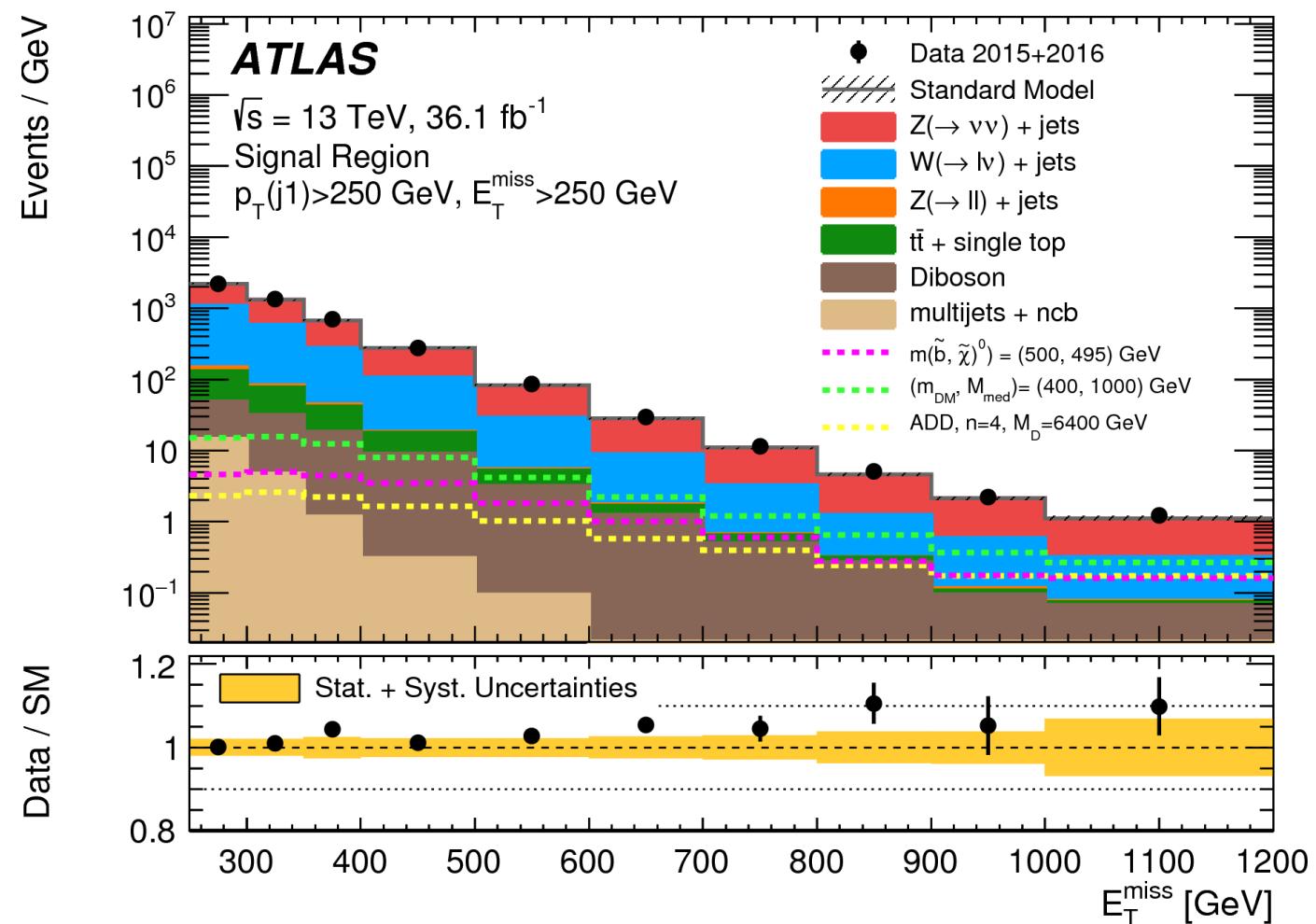
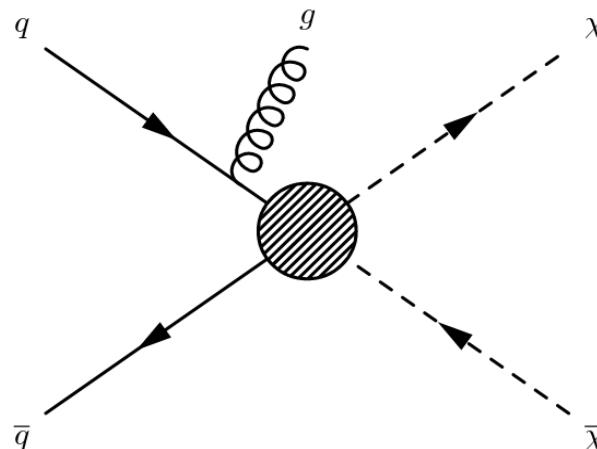
# The measurement: $Z \rightarrow \nu\nu + \text{jets}$

Really: Want to measure  $Z \rightarrow \nu\nu + \text{jets}$



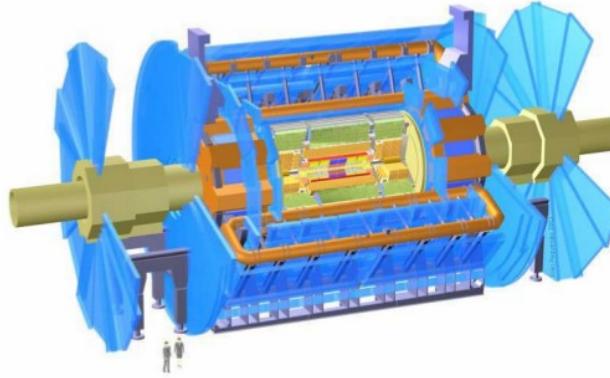
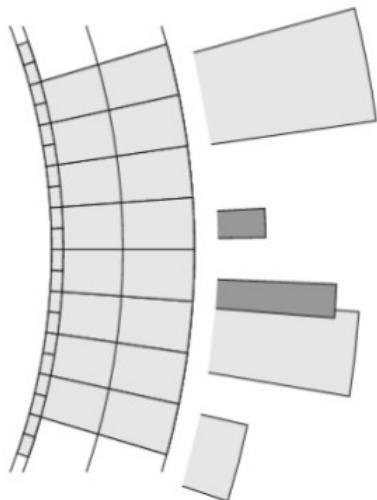
# Could measure this for a search

JHEP 1801 (2018) 126



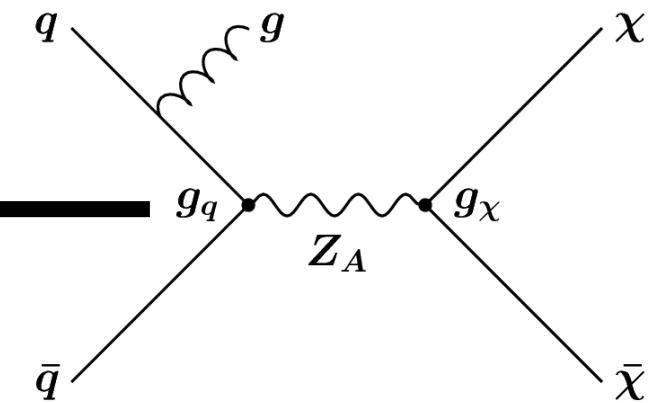
# Search has limitation: choice of model

Data @ detector level



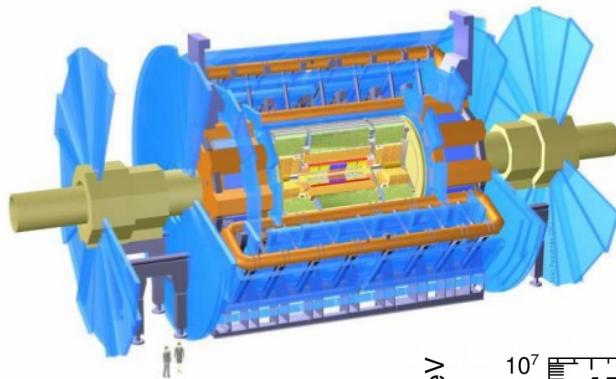
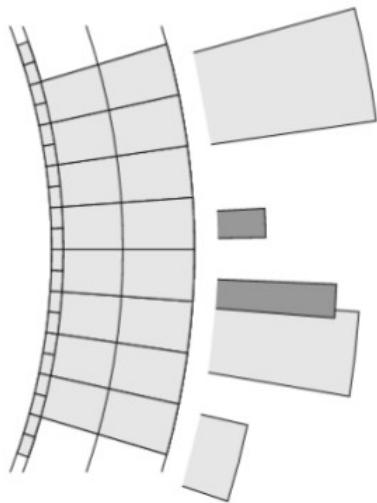
Full ATLAS simulation

theory

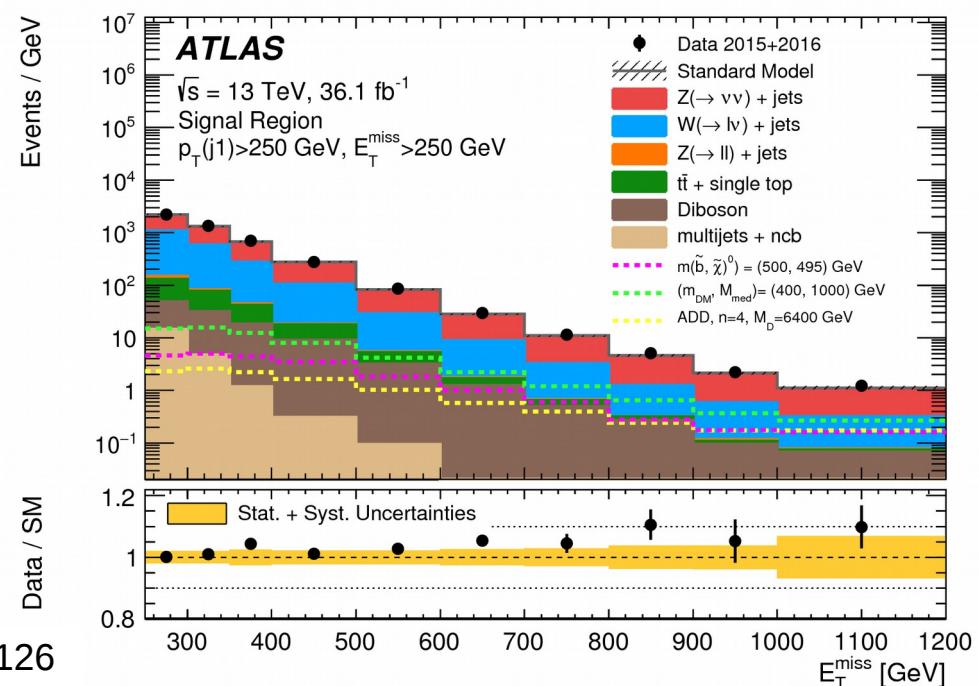
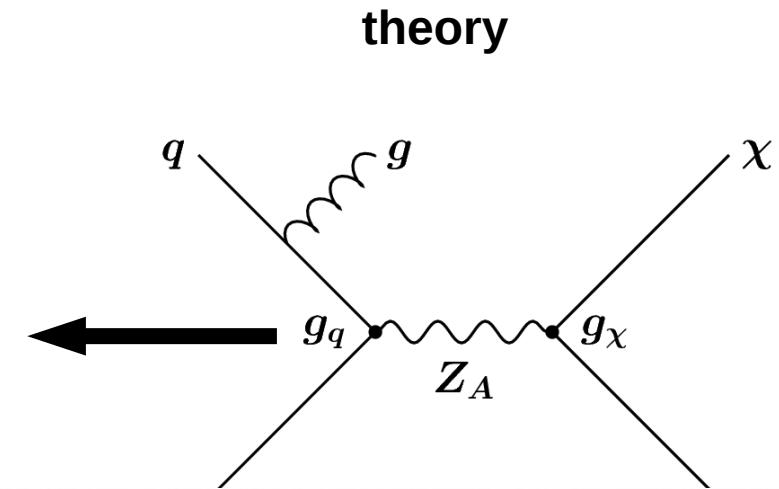


# Search has limitation: choice of model

Data @ detector level



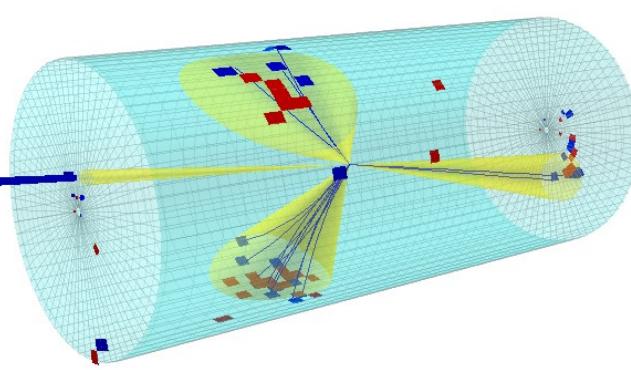
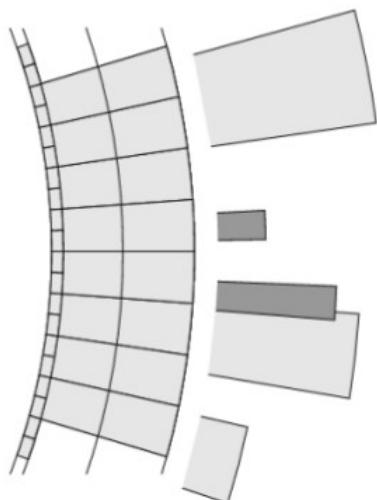
Full ATLAS sim



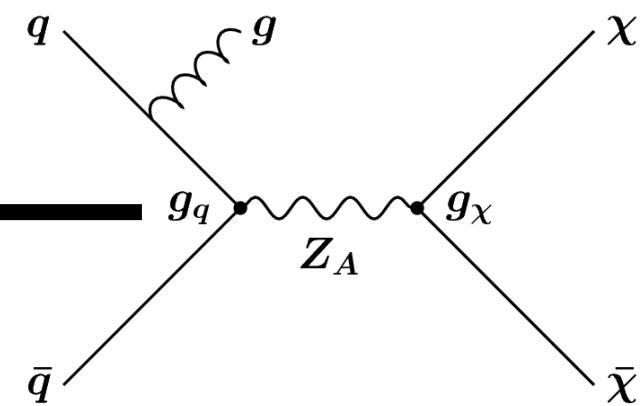
# New theories difficult to constrain

JHEP 1402 (2014) 057

Data @ detector level



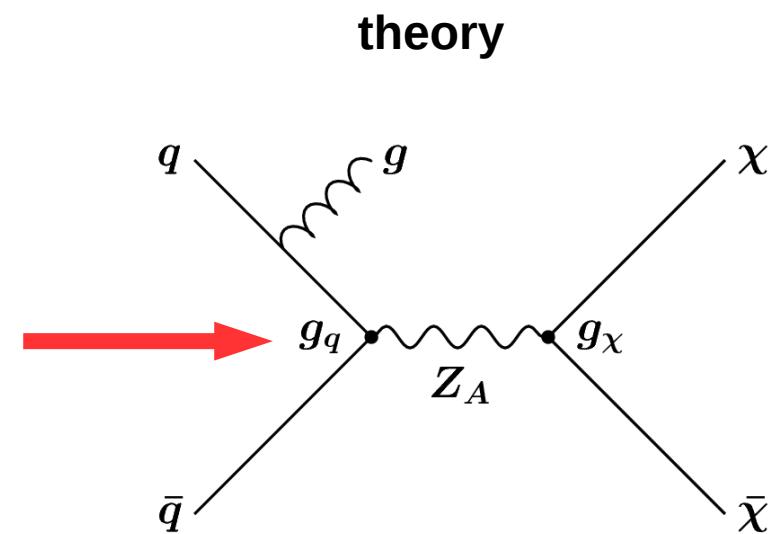
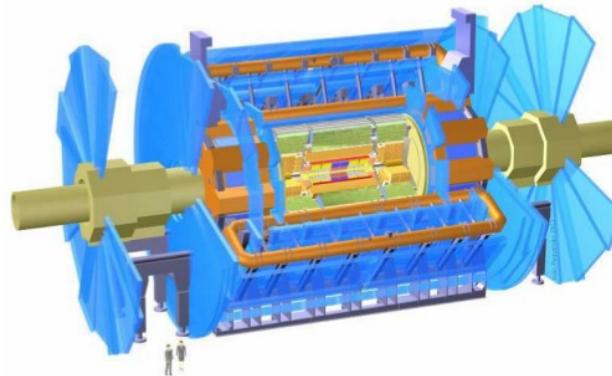
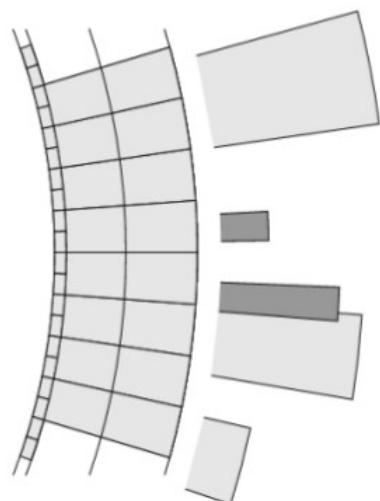
theory



Delphes (parametrized  
detector simulation)

# Bring data to theories!

Data @ detector level



Full ATLAS simulation

Unfolding

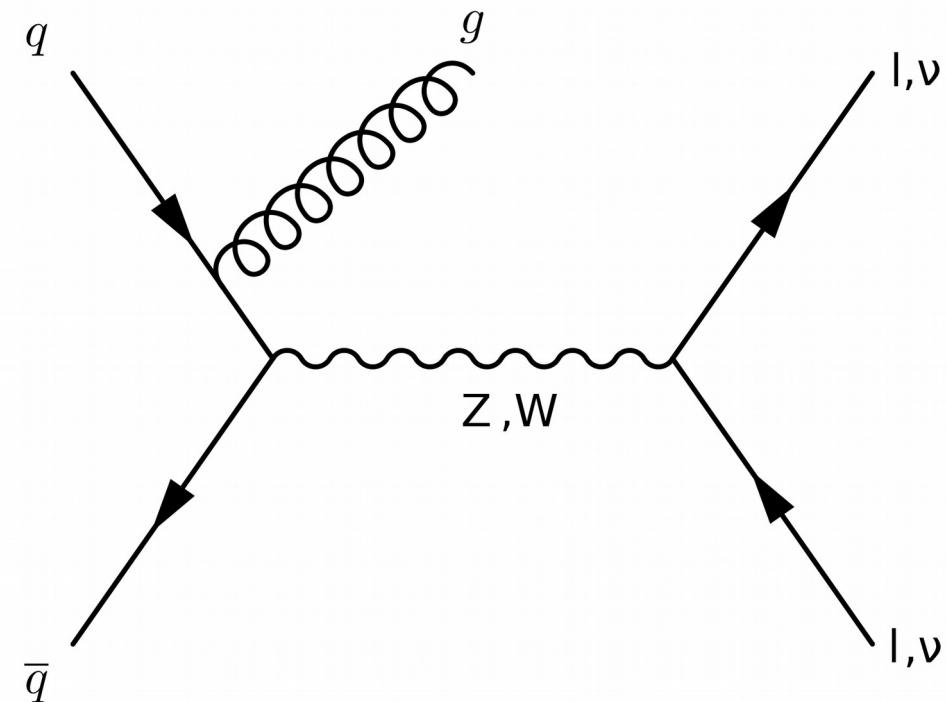
# Detector simulation

- Differences between detector- and particle level

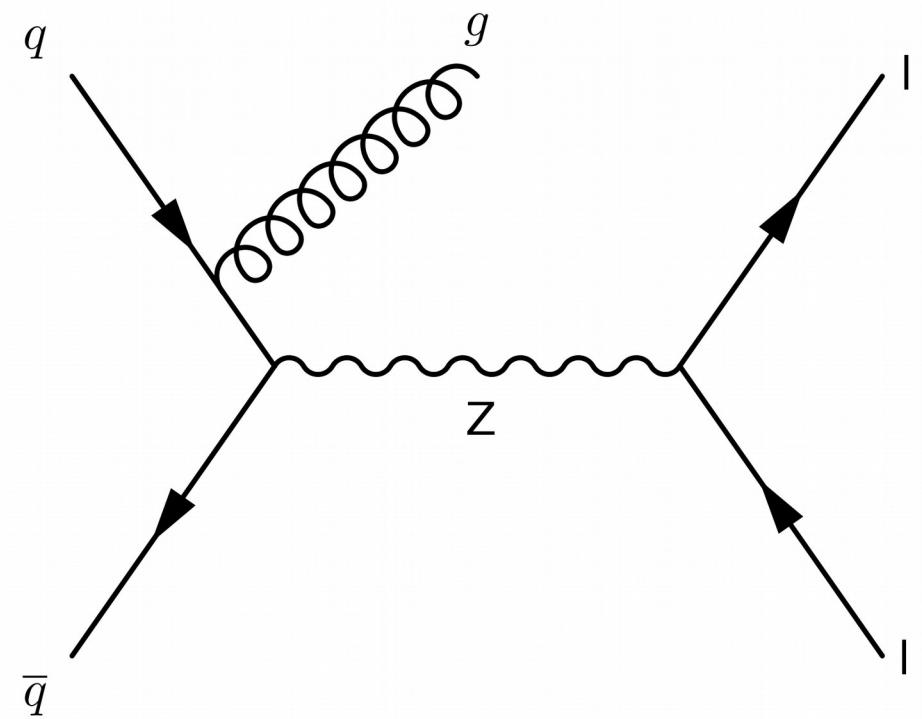
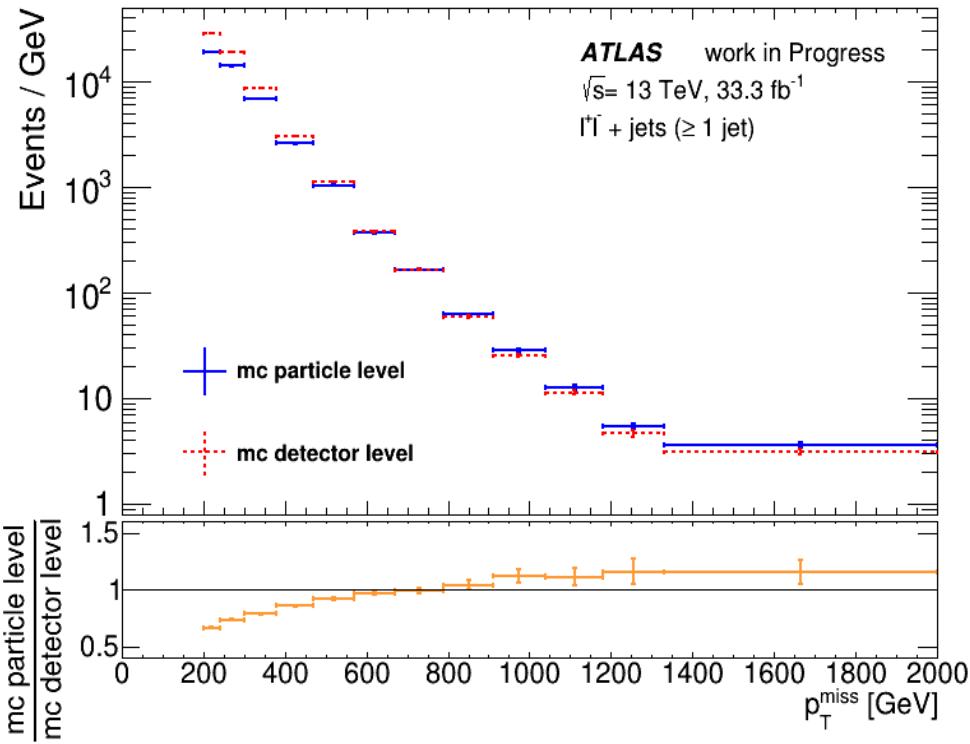
- Fakes
    - Found @ detector level
    - not present @ particle level

- Misses
    - Not found @ detector level
    - present @ particle level

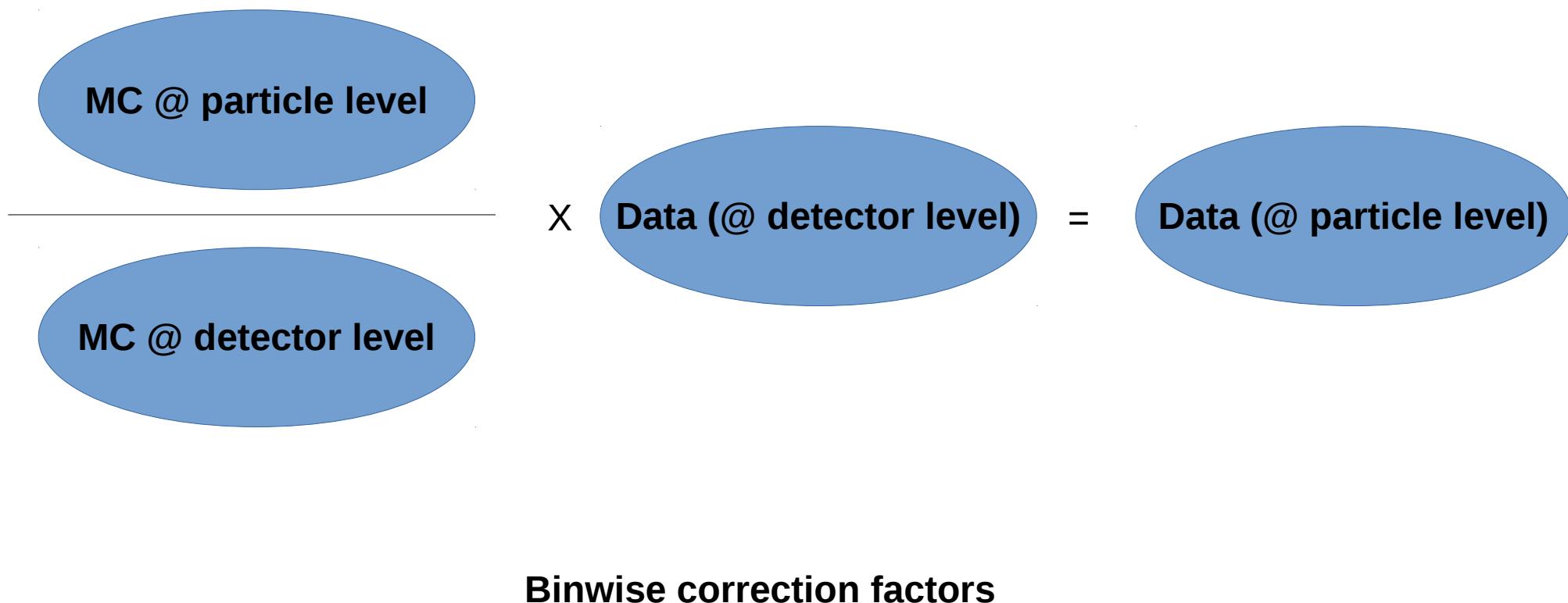
- Migrations
    - Present in both, but in different bins



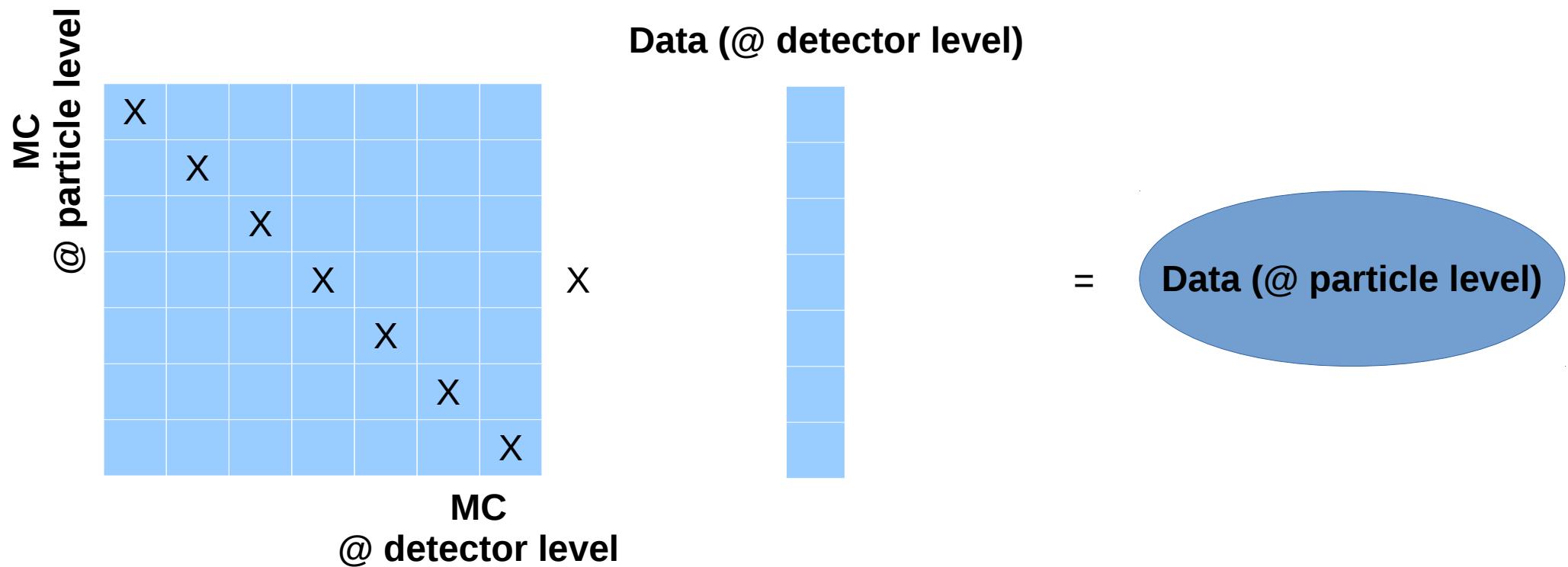
# Monte Carlo knows



# Bin-by-Bin Unfolding

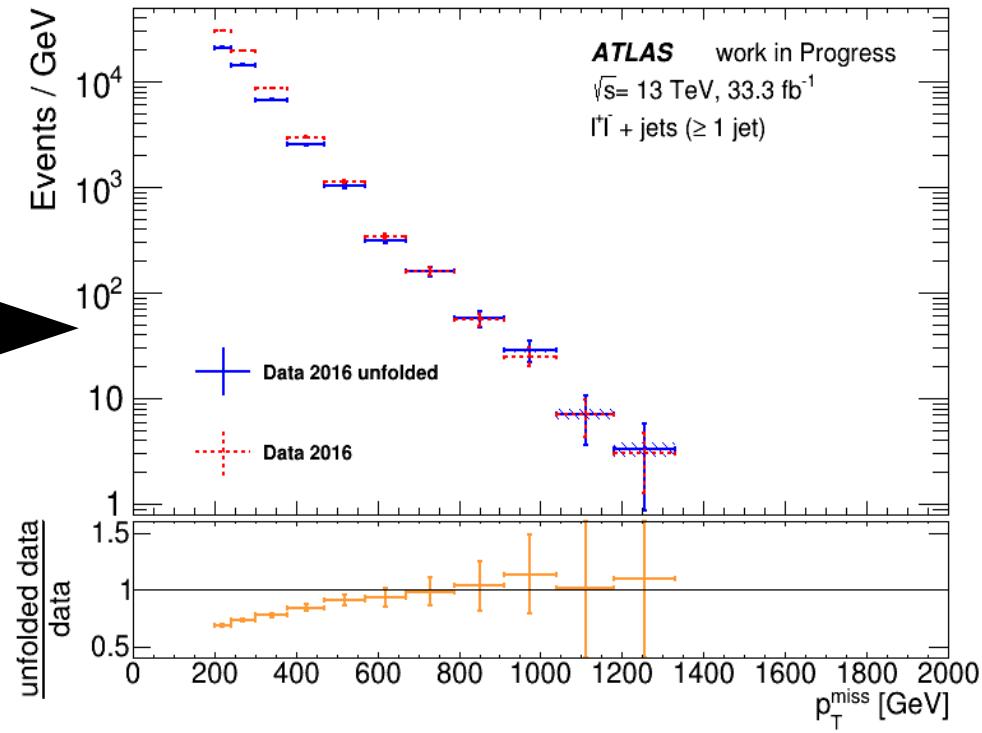
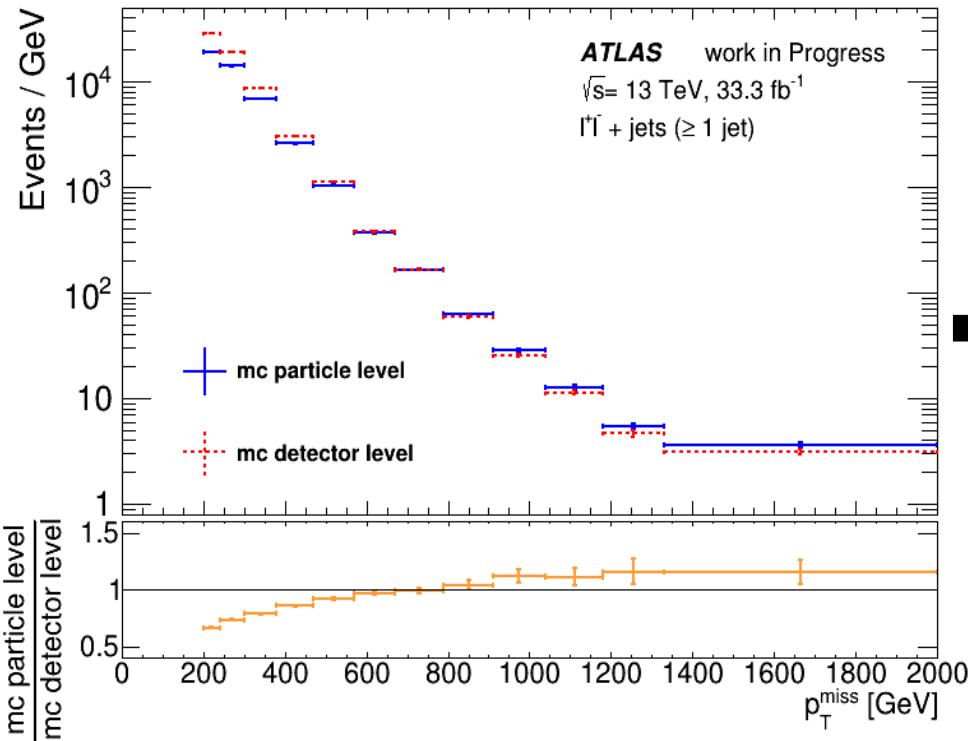


# Bin-by-Bin Unfolding



**Migration matrix diagonal if resolution is negligible**

# Monte Carlo knows



Drawback: relies entirely on MC prediction

# Iterative Dynamically Stabilized Unfolding

arXiv:1106.3107

Bayes Theorem:  $P(A | B) = \frac{P(B | A)P(A)}{P(B)}$

B = detector level  
A = particle level

# Iterative Dynamically Stabilized Unfolding

arXiv:1106.3107

Bayes

Theorem:

$$P(A | B) = \frac{P(B | A)P(A)}{P(B)}$$

B = detector level

A = particle level

P(A)

Monte Carlo Prediction

P(B)

Data

# Iterative Dynamically Stabilized Unfolding

arXiv:1106.3107

Bayes

Theorem:

$$P(A | B) = \frac{P(B | A)P(A)}{P(B)}$$

B = detector level

A = particle level

P(A)

Monte Carlo Prediction

P(B)

Data

Data unfolded (Prediction 1)

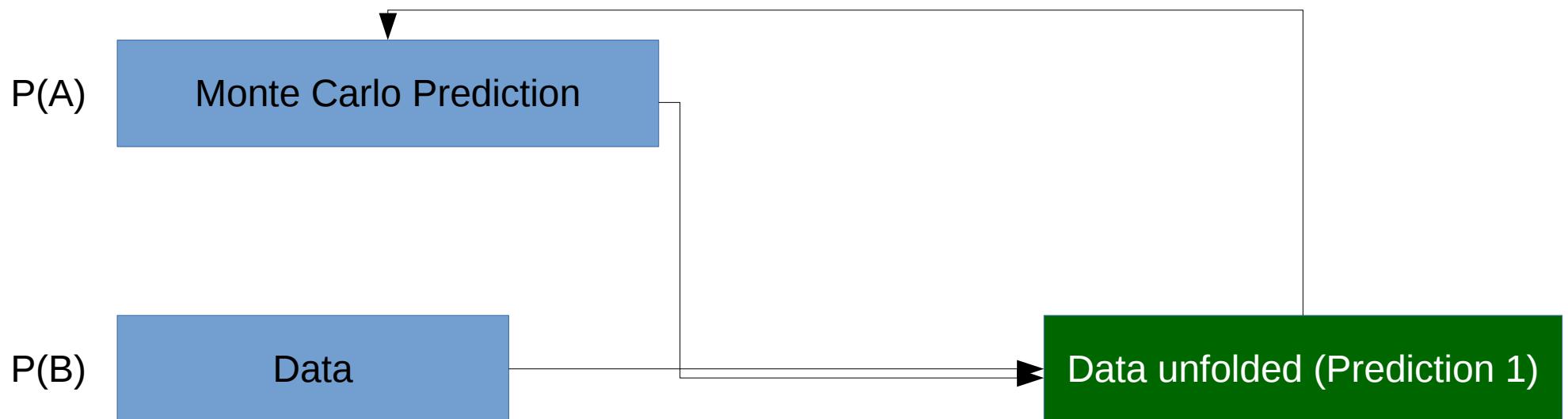
# Iterative Dynamically Stabilized Unfolding

arXiv:1106.3107

Bayes Theorem:

$$P(A | B) = \frac{P(B | A)P(A)}{P(B)}$$

B = detector level  
A = particle level



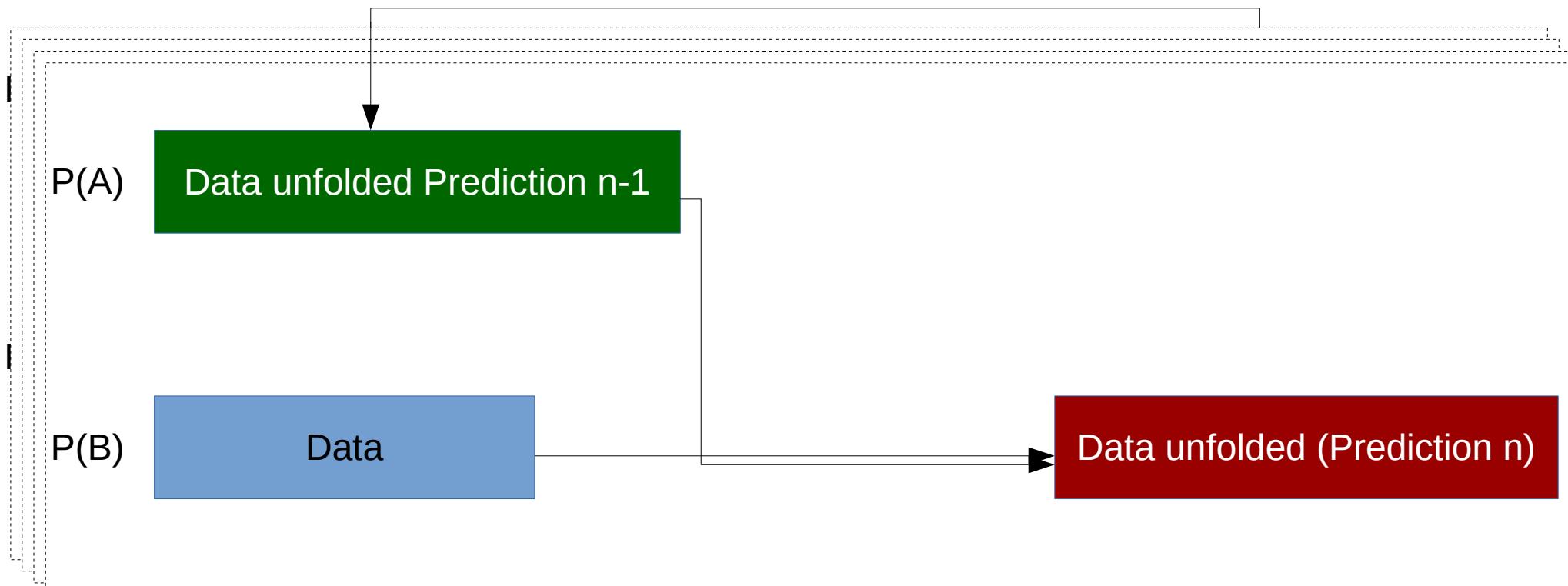
# Iterative Dynamically Stabilized Unfolding

arXiv:1106.3107

Bayes Theorem:

$$P(A | B) = \frac{P(B | A)P(A)}{P(B)}$$

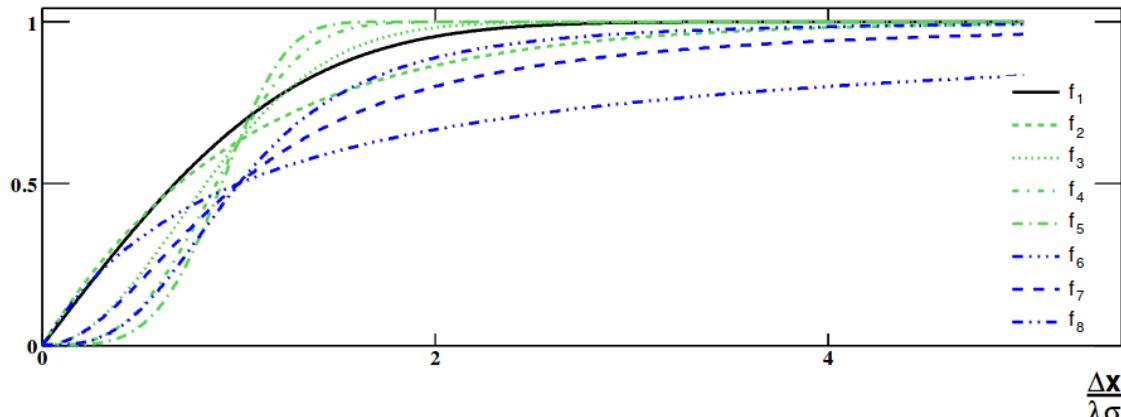
B = detector level  
A = particle level



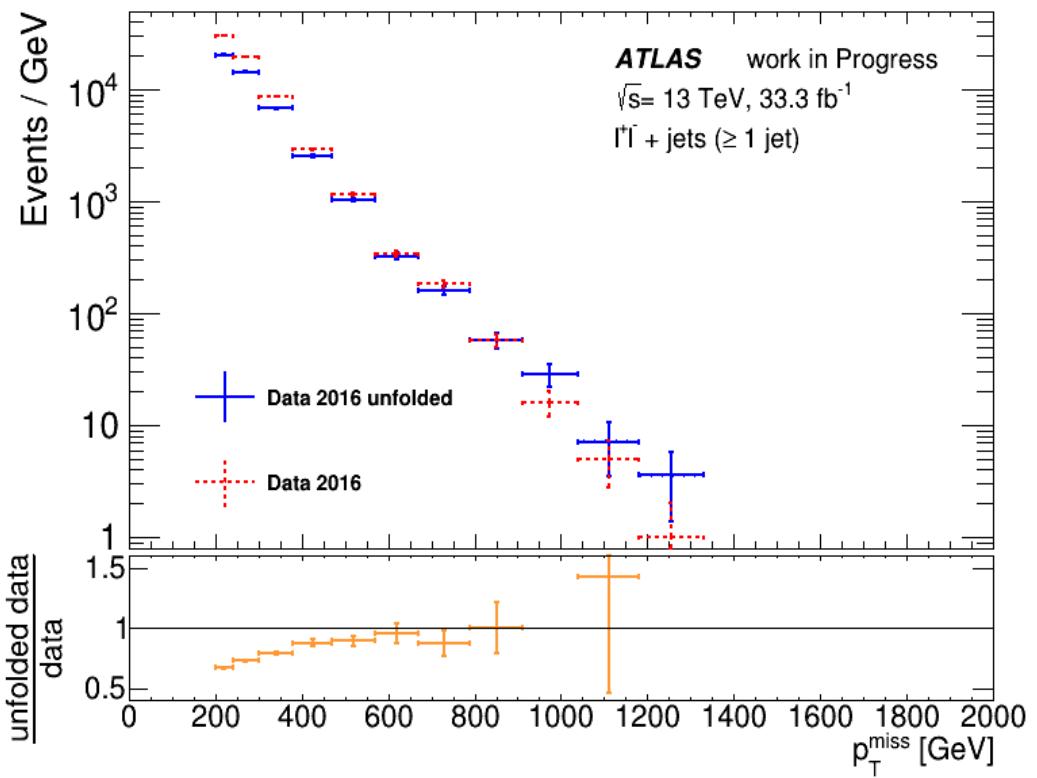
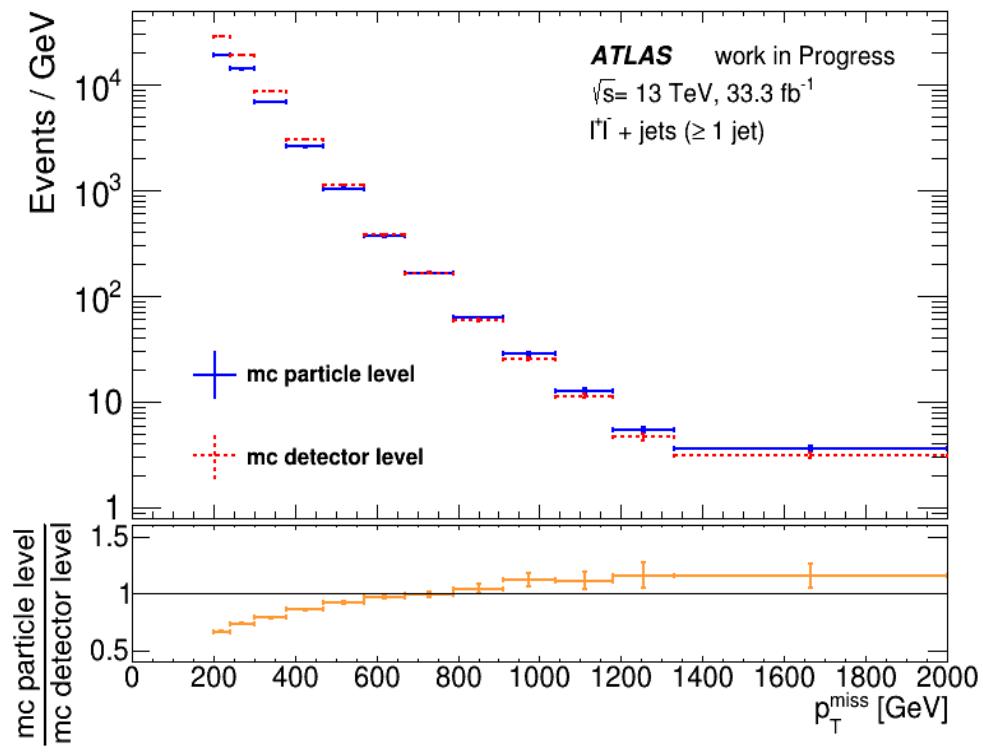
# IDS regularization

- Bayes unfolding: Regulatized by number of iterations
- IDS unfolding: Regularized by change in interation
  - If small changes remain, unfolding stops

$$u_j = t_j \cdot \frac{N_d^{MC}}{N_{MC}} + B_j^u + \sum_{k=1}^{n_d} f(|\Delta d_k|, \tilde{\sigma} d_k, \lambda) \Delta d_k \tilde{P}_{kj} + (1 - f(|\Delta d_k|, \tilde{\sigma} d_k, \lambda)) \Delta d_k \delta_{kj},$$

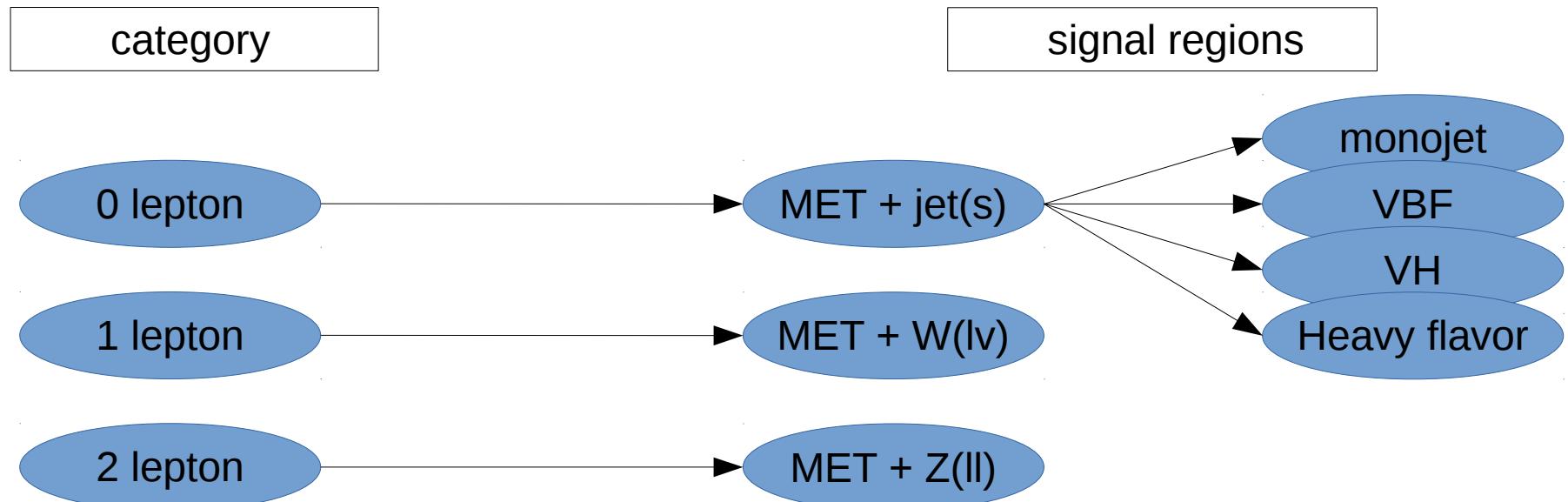


# IDS unfolding used in analysis



# Outlook

- Measurement in several lepton categories:



- Unfold and combine them all at particle level

# Summary

- Unfolding has many advantages
  - Increases longevity
  - Simplifies combination of results
- Detector-corrected measurement sensitive to anomalous MET+jet production published recently  
(Eur. Phys. J. C 77 (2017) 765)
- Next iteration extends this approach
  - More sophisticated unfolding
  - Combination of several topologies at particle level

# Summary

- Unfolding has many advantages
  - Increases longevity
  - Simplifies combination of results
- Detector-corrected measurement sensitive to anomalous MET+jet production published recently  
(Eur. Phys. J. C 77 (2017) 765)
- Next iteration extends this approach
  - More sophisticated unfolding
  - Combination of several topologies at particle level

**Thanks for the attention!**

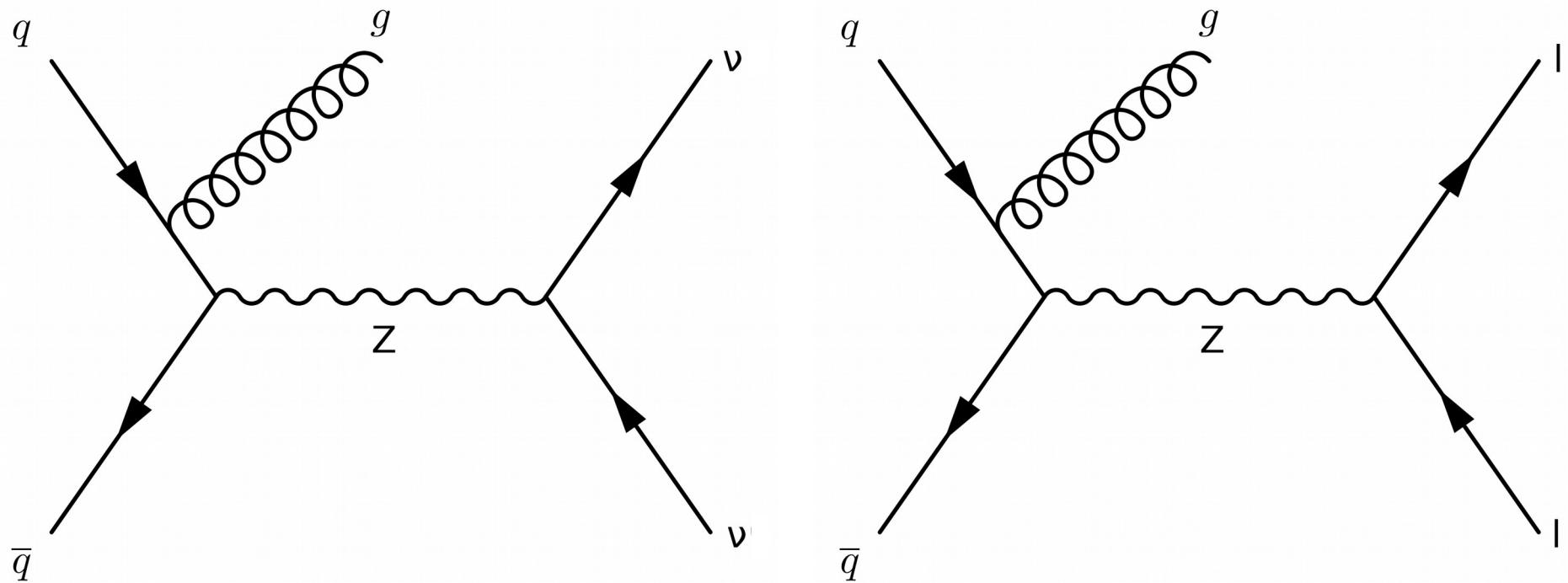
# BACKUP

# Detailed event selection

Numerator and denominator	$\geq 1$ jet	VBF
$p_T^{\text{miss}}$		$> 200 \text{ GeV}$
(Additional) lepton veto		No $e, \mu$ with $p_T > 7 \text{ GeV}$ , $ \eta  < 2.5$
Jet $ y $		$< 4.4$
Jet $p_T$		$> 25 \text{ GeV}$
$\Delta\phi_{\text{jet}_i, p_T^{\text{miss}}}$	$> 0.4$ , for the four leading jets with $p_T > 30 \text{ GeV}$	
Leading jet $p_T$	$> 120 \text{ GeV}$	$> 80 \text{ GeV}$
Subleading jet $p_T$	—	$> 50 \text{ GeV}$
Leading jet $ \eta $	$< 2.4$	—
$m_{jj}$	—	$> 200 \text{ GeV}$
Central-jet veto	—	No jets with $p_T > 25 \text{ GeV}$
Denominator only	$\geq 1$ jet and VBF	
Leading lepton $p_T$		$> 80 \text{ GeV}$
Subleading lepton $p_T$		$> 7 \text{ GeV}$
Lepton $ \eta $		$< 2.5$
$m_{\ell\ell}$		66–116 GeV
$\Delta R$ (jet, lepton)		$> 0.5$ , otherwise jet is removed

# Can also measure a ratio

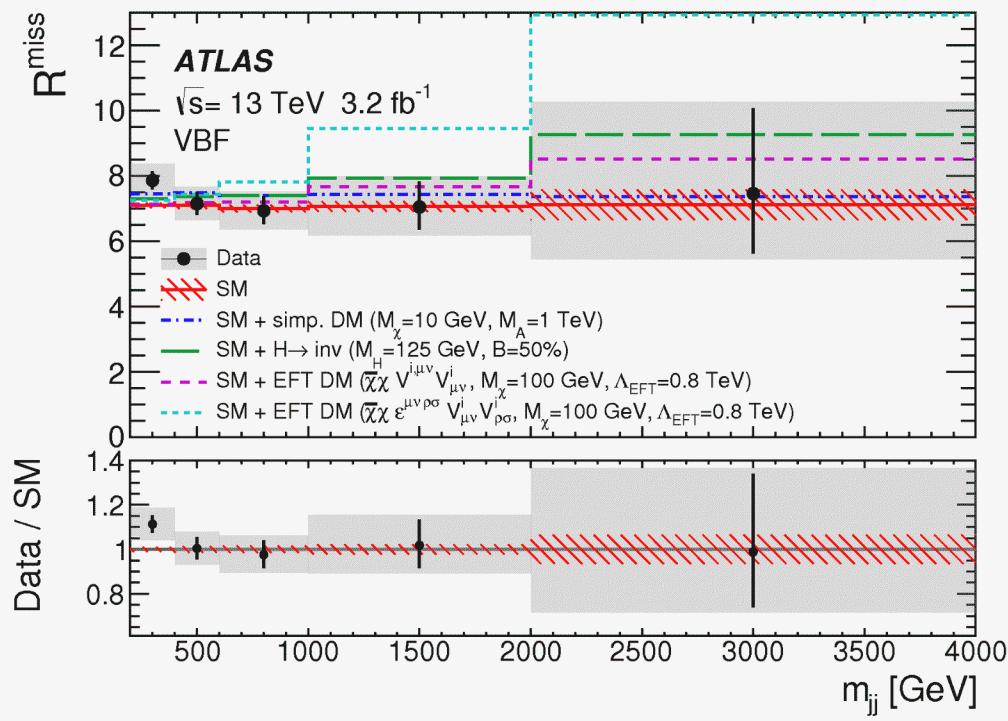
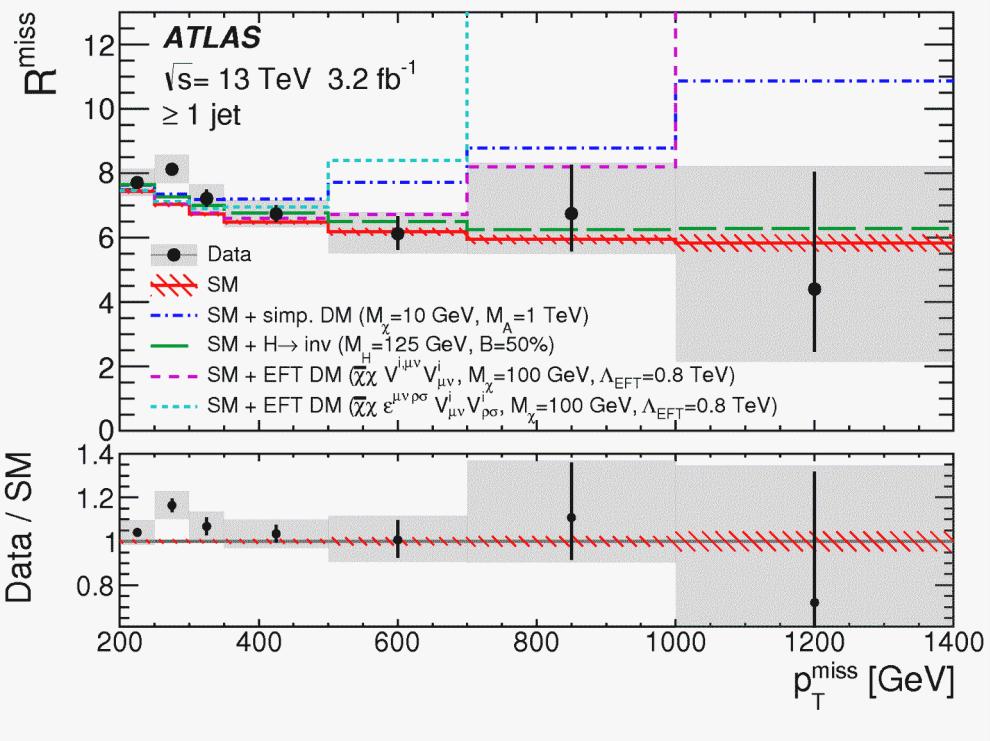
Measure ration of  $Z \rightarrow \nu\nu + \text{jets}$  and  $Z \rightarrow \ell\ell + \text{jets}$



$$R_{\text{miss}} = \frac{\sigma(\cancel{p}_T + \text{jets})}{\sigma(Z \rightarrow \ell^+ \ell^- + \text{jets})}$$

# Can also measure a ratio

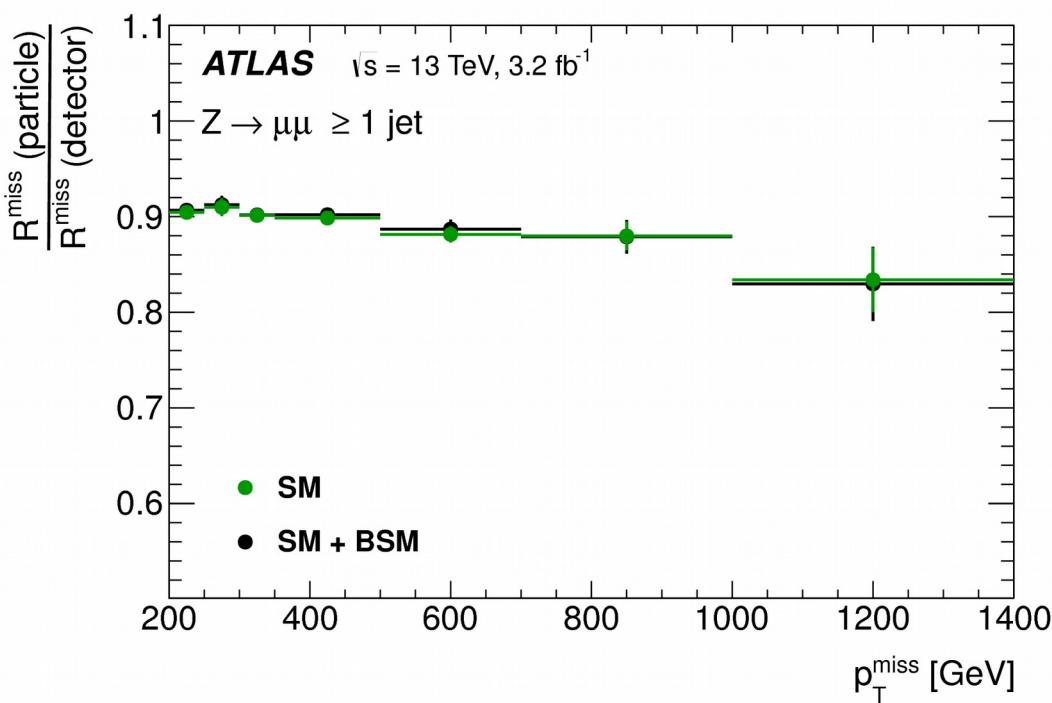
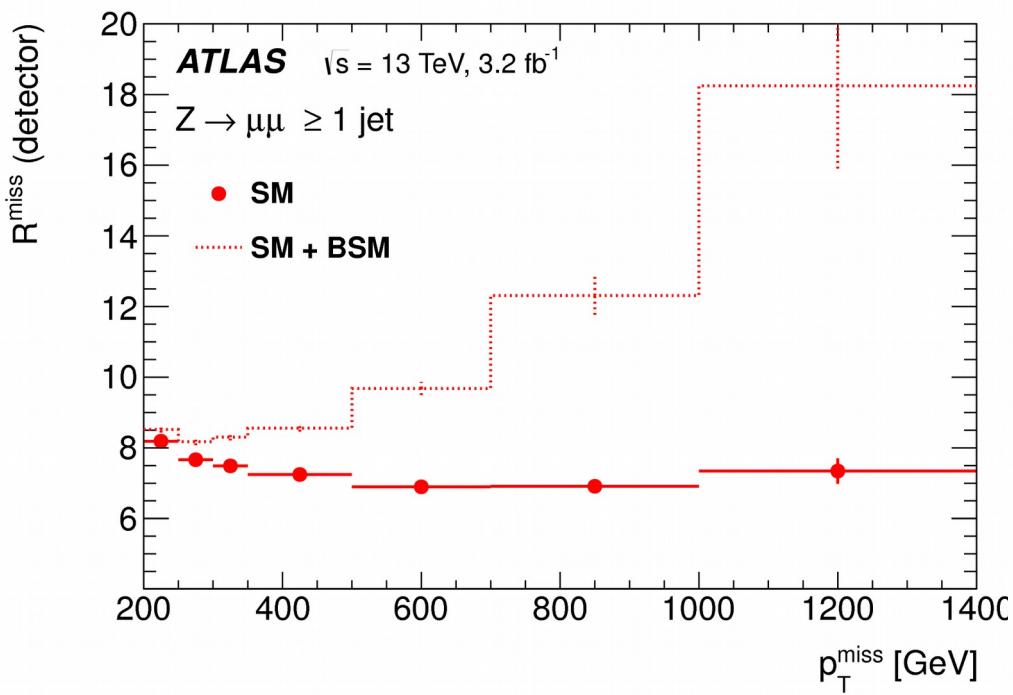
$$R_{\text{miss}} = \frac{\sigma(p_T + \text{jets})}{\sigma(Z \rightarrow \ell^+ \ell^- + \text{jets})}$$



# Bin-by-Bin Unfolding in searches

Eur. Phys. J. C 77 (2017) 765

Relies 100% on Monte Carlo modeling of the events!



Can still be perfectly applicable if “mismodeling” does not affect the correction factors

Here:

BSM signal large enough to modify correction factors, would show large deviations in data