



# Data-Driven Background Estimates at the LHC

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4<sup>th</sup> November 2010, Brussel

On Behalf of ATLAS and CMS Collaboration



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# Outline

**LHC is starting the searches**

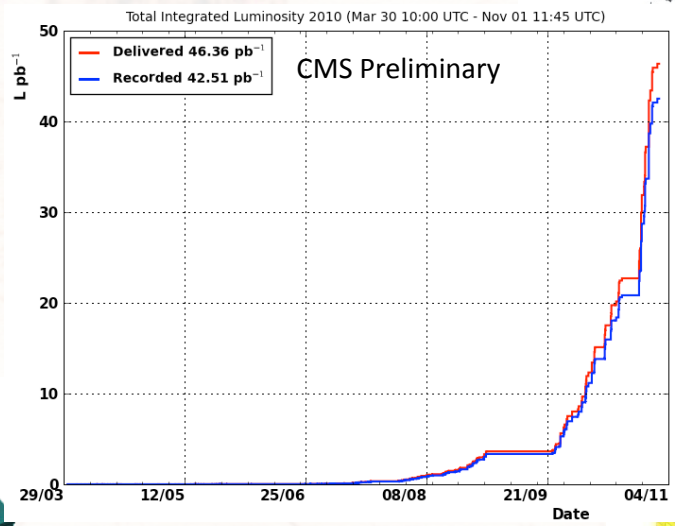
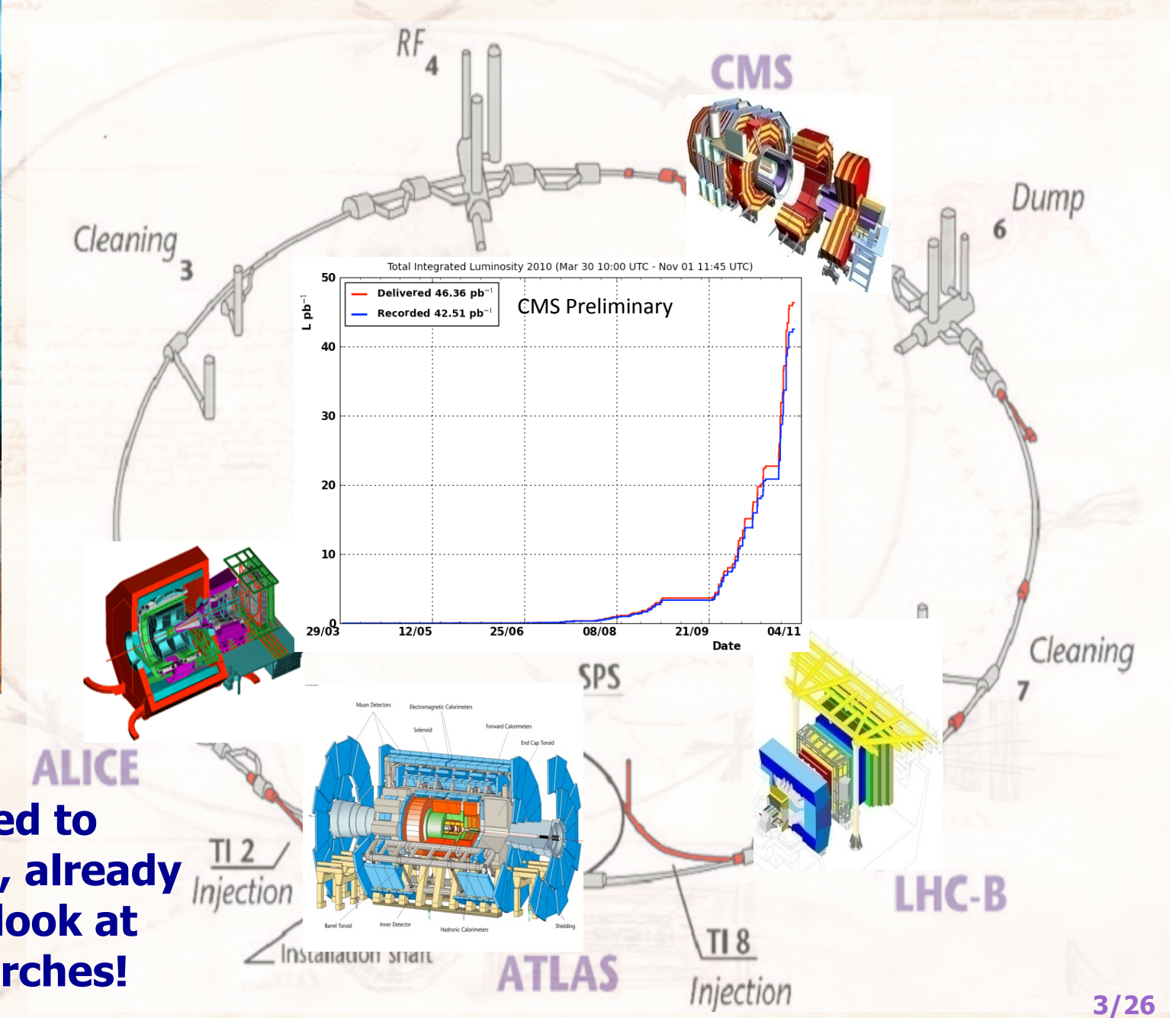
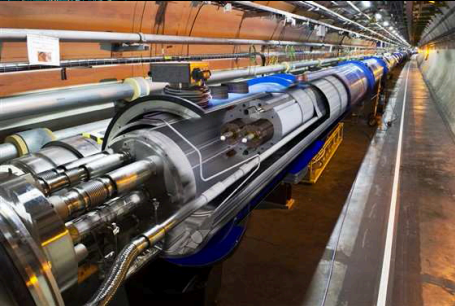
**Main methods on the market:**

- **Fit**
- **Scaling**
- **Templates**
- **Replacement Method**
- **Matrix Methods**

**Conclusion**



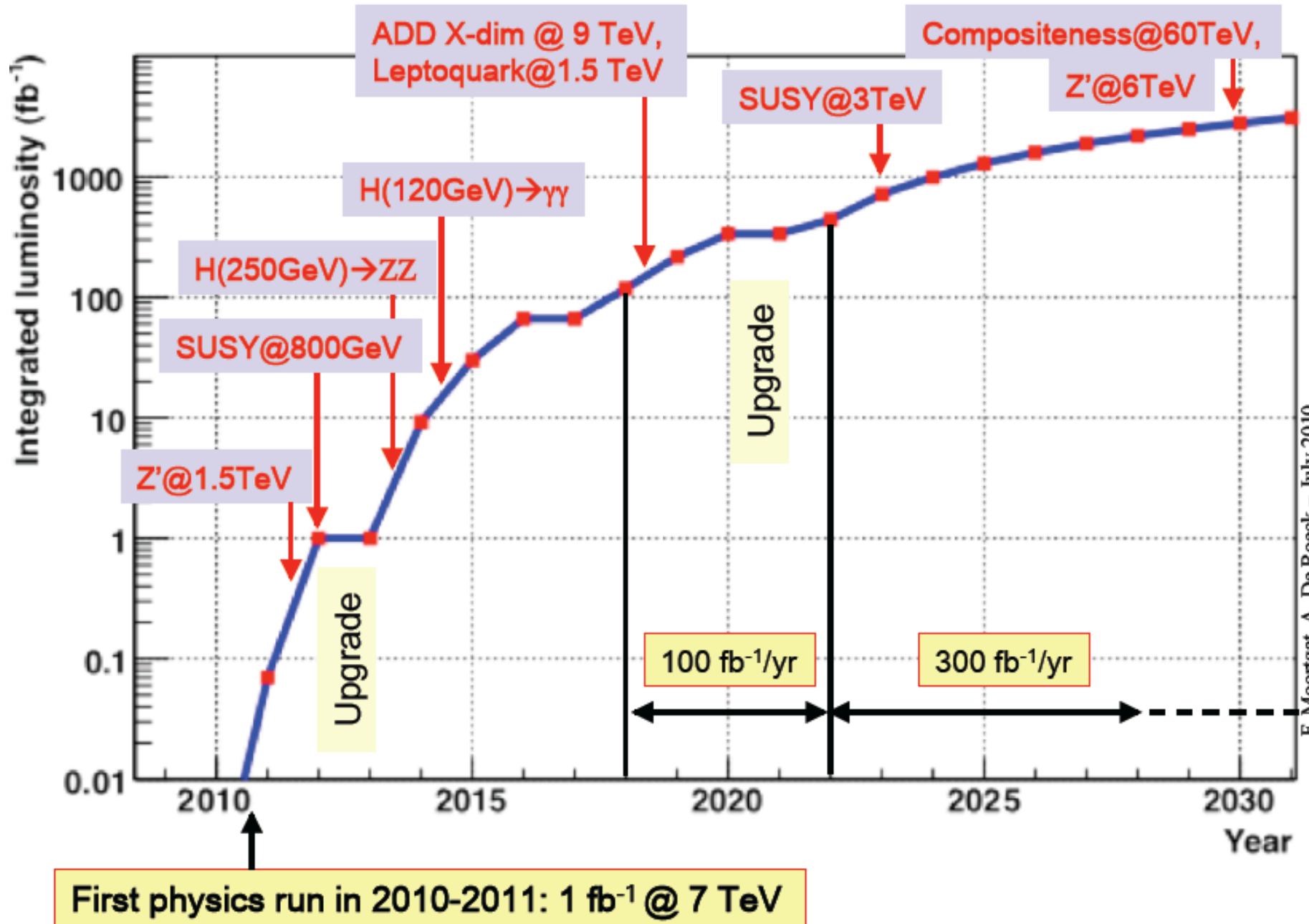
# LHC Opens Window to Searches



**More than  
40 pb-1 delivered to  
ATLAS and CMS, already  
a lot of data to look at  
and to start searches!**



# Road Map to Discoveries





# Commissioning Detectors: Understanding the Variables...

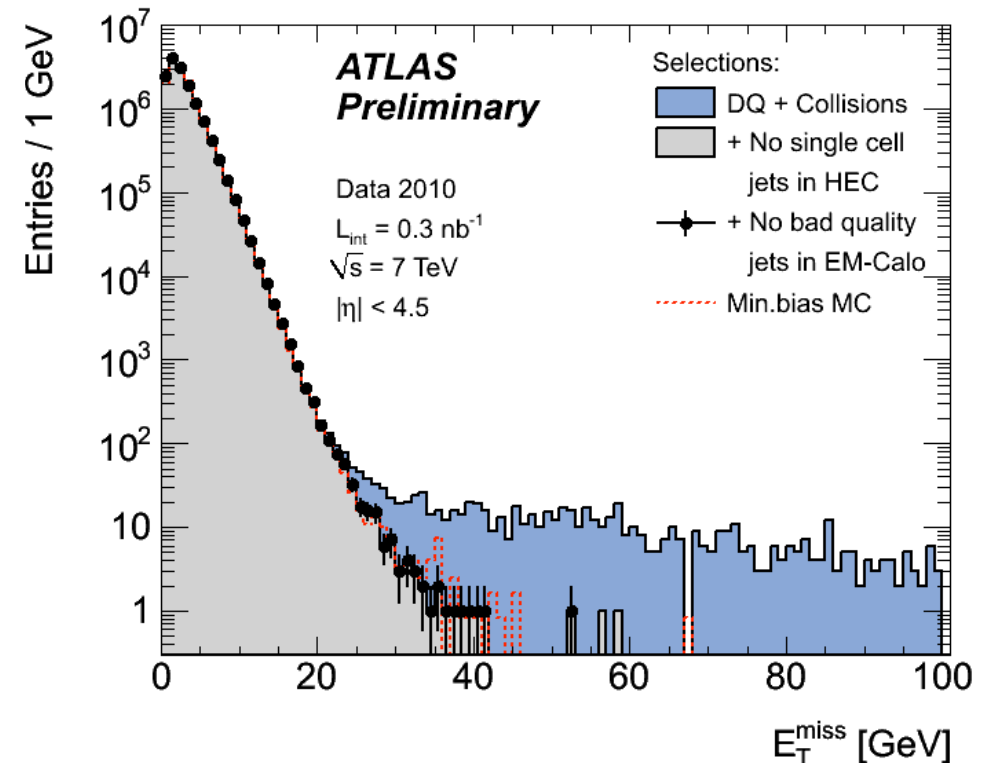


From the first collisions day, a lot of results have been appearing very quickly

→ Understanding and commissioning of the detector is in well advanced stage

→ Mandatory before exploring new territories...

Standard Model signals are becoming background of searches, need to have a proper evaluation of their contamination in signal area (too large to number of events to be simulated).





# Which Methods to Use...



**Depending of the signal studied, different kind of background:**

**- Resonance like signal:**

→ Propagation Fit and subtract background from the fit

→ Factorization cuts

**- Looking in tail of distributions (on top of previous):**

→ Templates

→ Replacement Method

→ Various Matrix Method

→ Various techniques can be used for cross check, some time mandatory to do them in sequence

# Fit Propagation

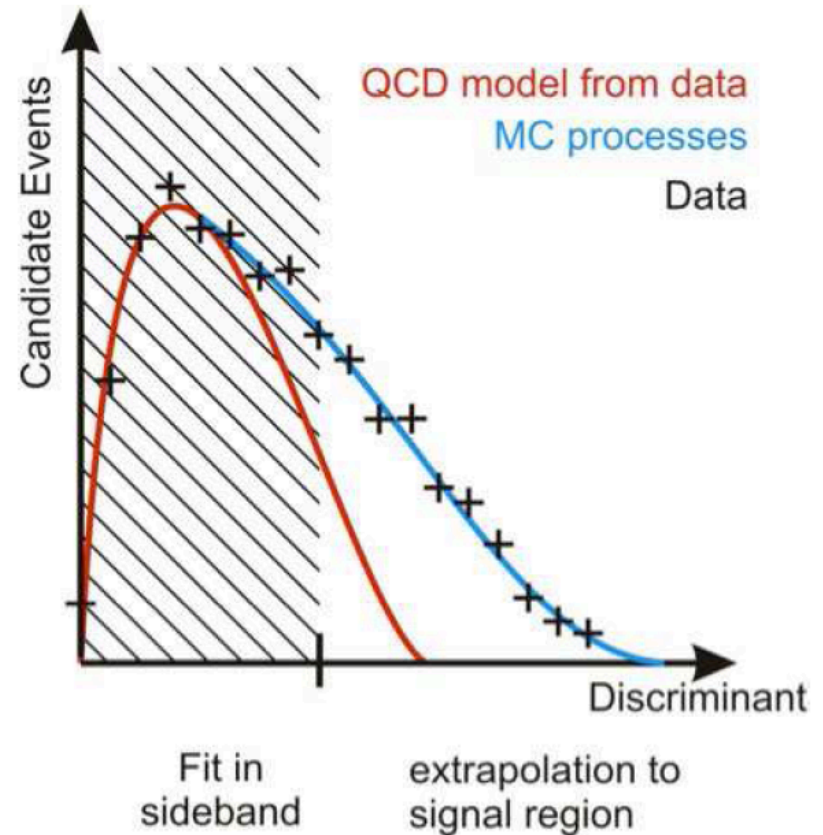
**Find a control region in phase space where SM background dominates.**

**Use measurements in this region to infer SM background in signal region.**

**Should ensure the fit function is valid in the signal area.**

**Ex: Searches with isolated leptons to determine contamination from non isolated leptons.**

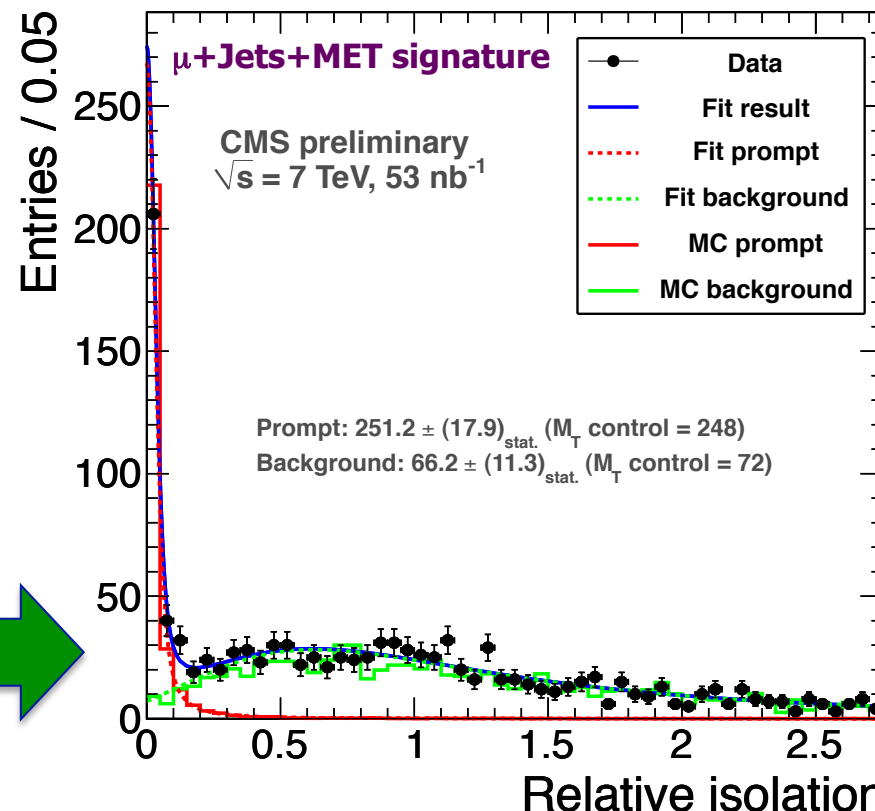
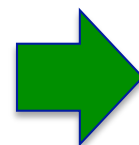
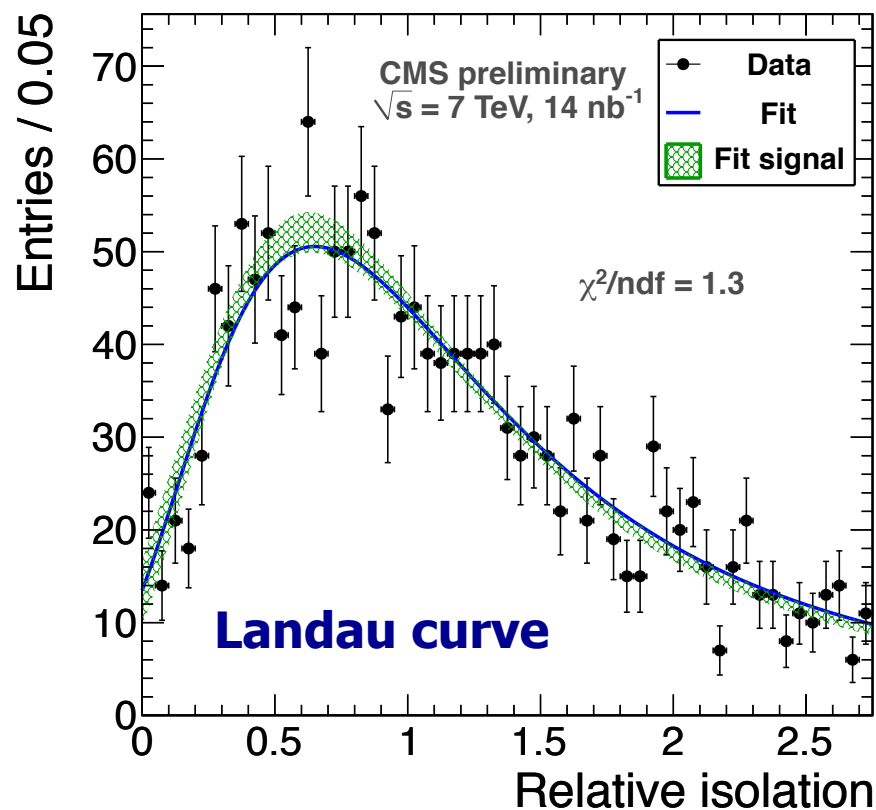
**Variation: Fit of multiple contributions**



# $\mu + \text{Jets} + \text{ME}_T$ Signature

Looking at samples after full selection except isolation.  
Determine the shape of the function to fit in a background like sample.

CMS PAS SUS-10-001



Fit of signal can also be done using simulation.

→ Good agreement between fit estimation, data and simulation control



# e+Jets+ME<sub>T</sub> Signature

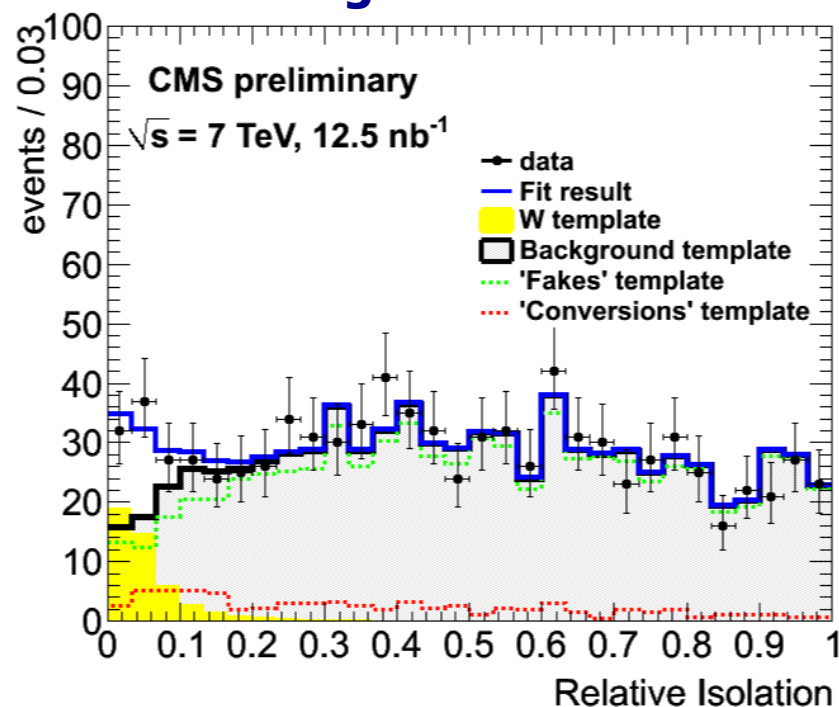
Two kinds of background:

- heavy-flavor decays and jets mis-identified as electron
- electrons due to photon conversion

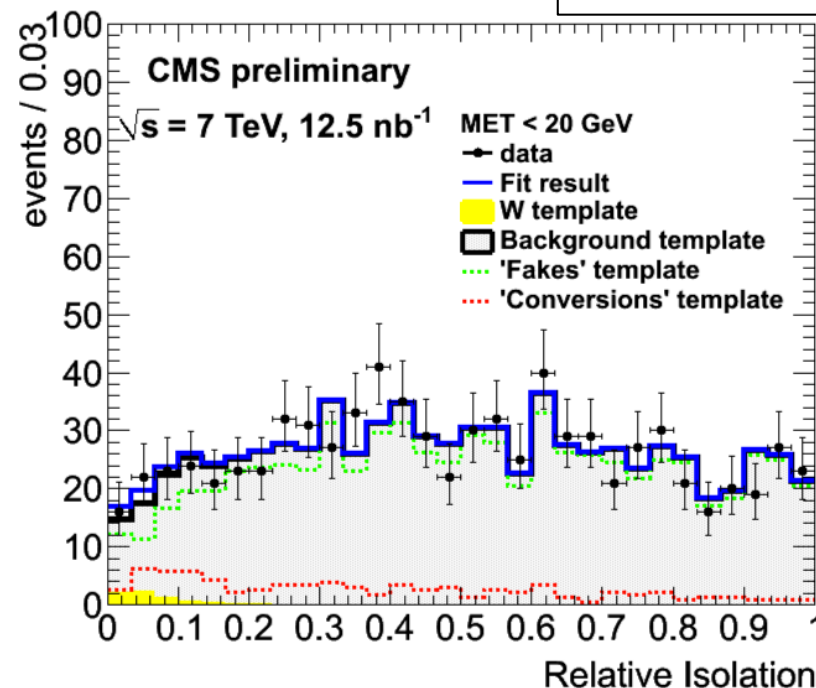
Select control samples dominated by each of above sources by inverting selection cuts

Perform fit using Relative Isolation ( $\text{RelIso} = p_T(e)/\Sigma E_{T R<0.3}$ ) distributions for each background.

CMS PAS SUS-10-001



After:  $\text{RelIso} < 0.3$   
 Predicted :  $224 \pm 13$   
 Observed : 263



After ( $\text{RelIso} < 0.3$ )  
 Predicted :  $215 \pm 13$   
 Observed : 215



# Factorization Cuts/Scaling



**Determine all efficiencies of the cuts selection and weight a background like sample by all efficiencies.**

**Mainly to ensure that a given SM background can be neglected in the final selection, or using higher statistics sample:**

- **Berends-Giele scaling method:**  $W^{\geq 4\text{jets}} = W^{2\text{jets}} \cdot \sum_{i=2}^{\infty} (Z^{2\text{jets}} / Z^{1\text{jet}})^i$
- **Scaling distribution according to resolution etc**

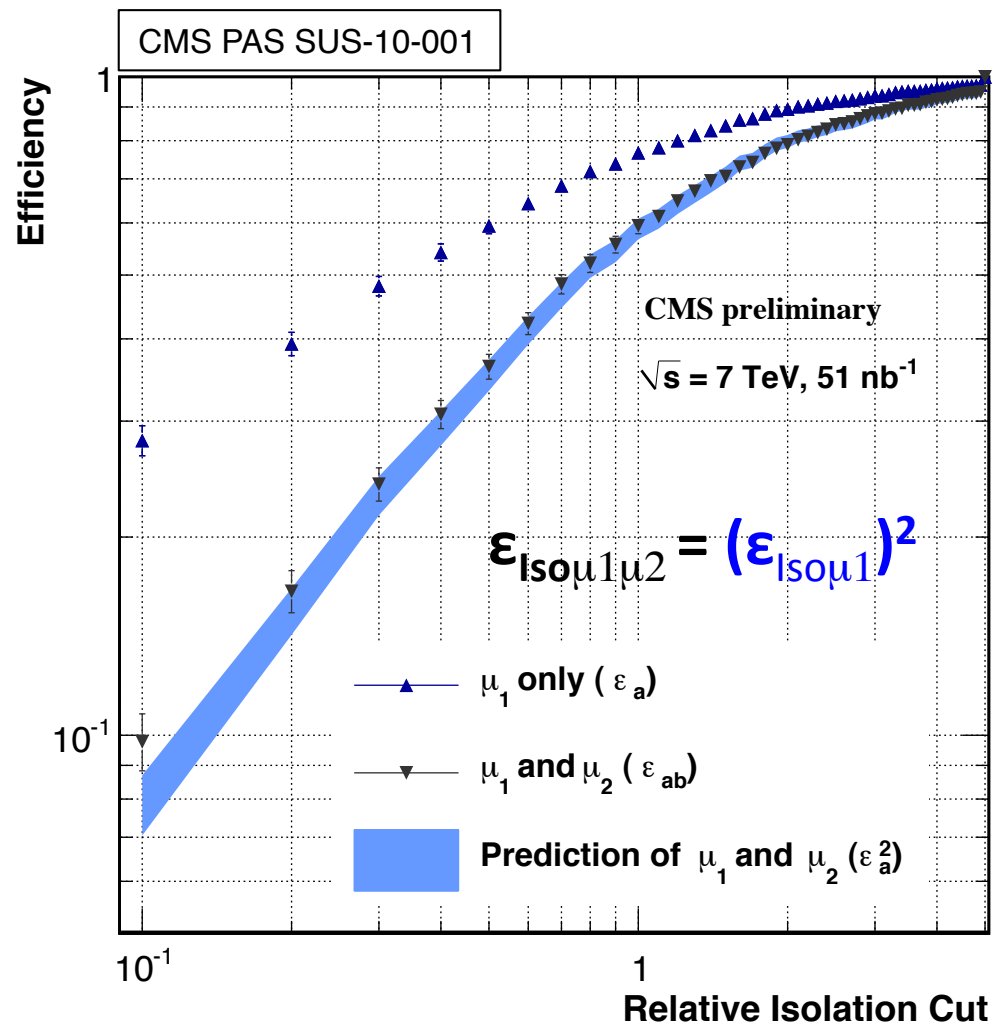
**Need to control the correlation between cuts and/or ensure that selection do not bias scaling.**

# Same Sign di-Muons Searches



Selection cuts are uncorrelated

→ selection efficiency for each cut measured in control samples



Di-Muons samples before isolation (dominated by multijet events)

Isolation of  $\mu_1 = \epsilon_{\text{Iso}\mu_1}$

Isolation of  $\mu_2 = \epsilon_{\text{Iso}\mu_2}$

$$\epsilon_{\text{AllCuts}} = \epsilon_{\text{Iso}\mu_1} \cdot \epsilon_{\text{Iso}\mu_2}$$

Good agreement between prediction and observed  
 → multijet background can be scaled down by  $(\epsilon_{\text{Iso}\mu_1})^2$



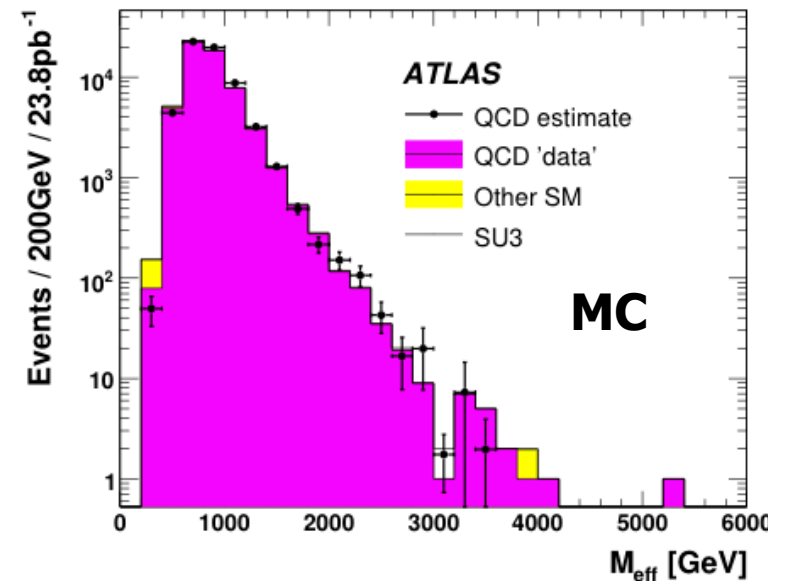
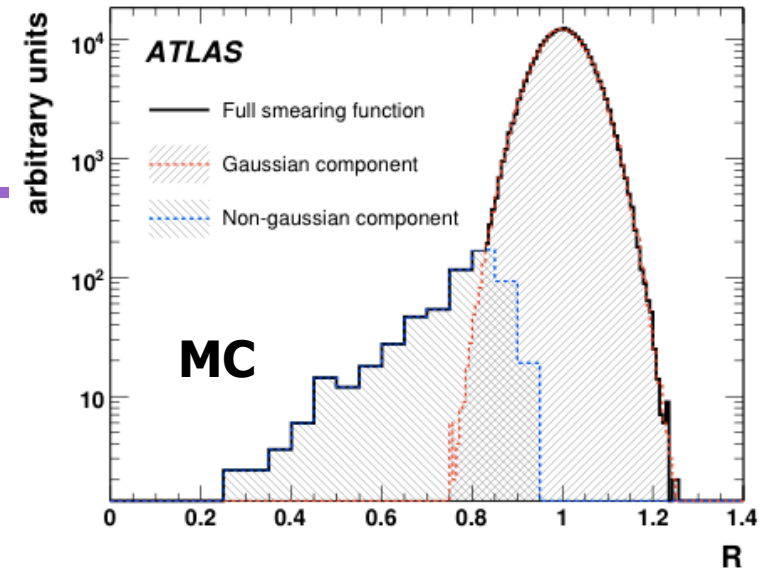
# Smearing

Modify Monte Carlo samples to mimic the data:

Mostly used for QCD events to introduce Jet Resolution and its effect on missing ET.

- Derive Gaussian part of smearing function from  $\gamma + \text{jet}$  control sample
- Derive non-Gaussian part from Mercedes events ( $\gamma + \text{jet}$ ), requiring that the MET is co-linear with one of the jets
- Combine smearing functions, normalising with di-jet sample
- Apply smearing function to low MET events to predict the tail in the high MET signal region.

arXiv:0901.0512 (2009)

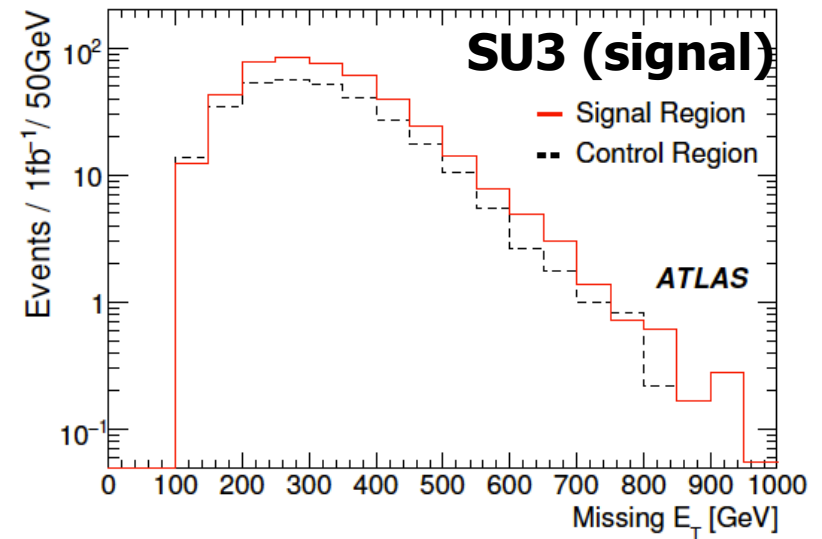
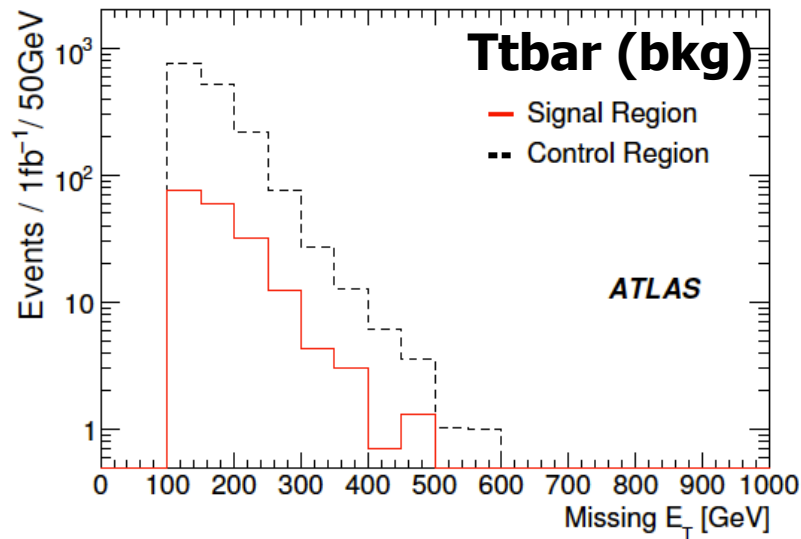




# Templates



- Define a signal-depleted control sample
- Determine the shape of background in this region
- Propagate the shape of the background in a signal like region.
- Need to understand the variables shape in control region to port it in signal region.

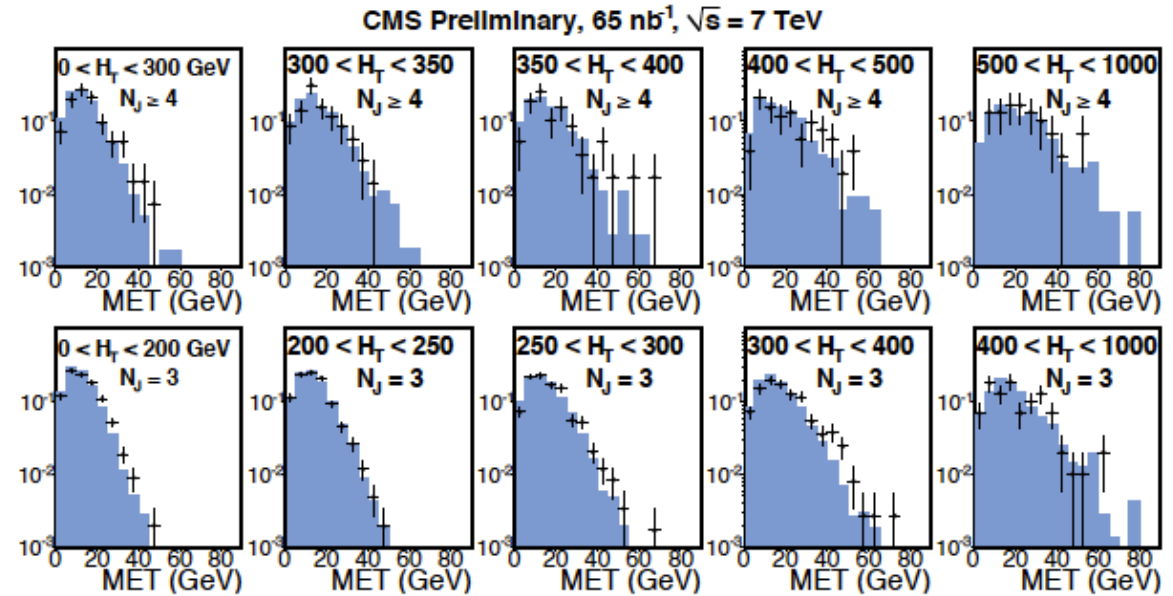


# Lepton+jets+MET Signatures

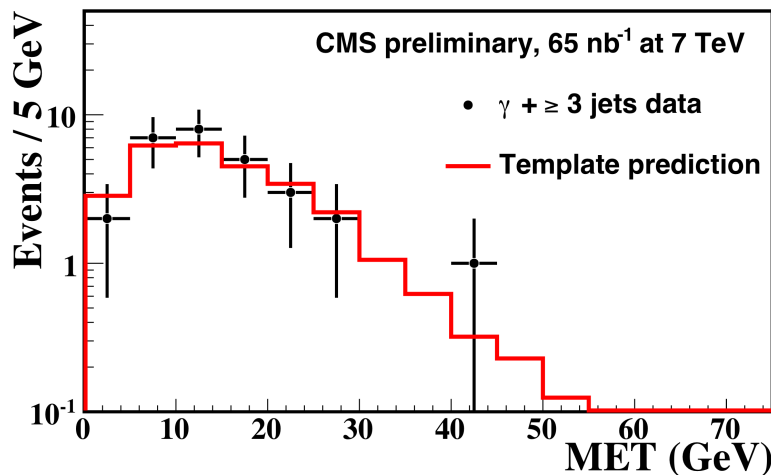


CMS PAS SUS-10-001

- MET background from real MET (e.g. in W/Z) and MET due to mis-measurements
- Use MET templates from multi-jet events to predict MET for g+jets events



MET templates from multi-jet events



**Good agreement between predicted and observed distributions:**  
**for MET > 15 GeV**  
**predicted = 12.5**  
**observed = 11**

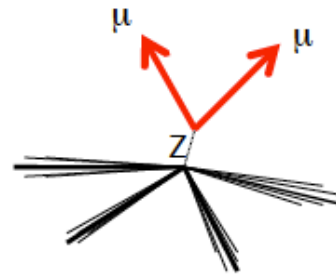
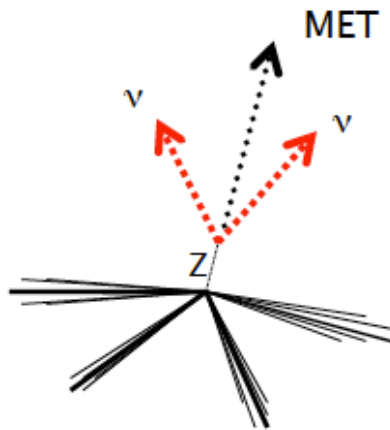
# Replacement Method

Use a none standard model process identified from data and “modify” it in order to simulate another standard model process.

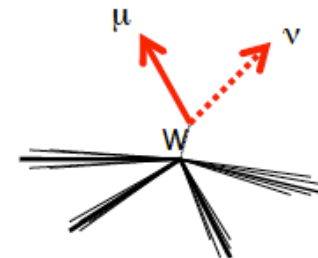
Example:

Large missing  $E_T$  searches + jets:

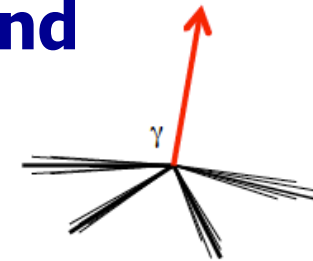
$Z + \text{jets} \rightarrow \nu\nu + \text{jets} \rightarrow$  irreducible background



**Z → ll + jets**  
 Strength: very clean  
 Weakness: low statistics



**W → lv + jets**  
 Strength: larger statistics  
 Weakness: background from SM and SUSY

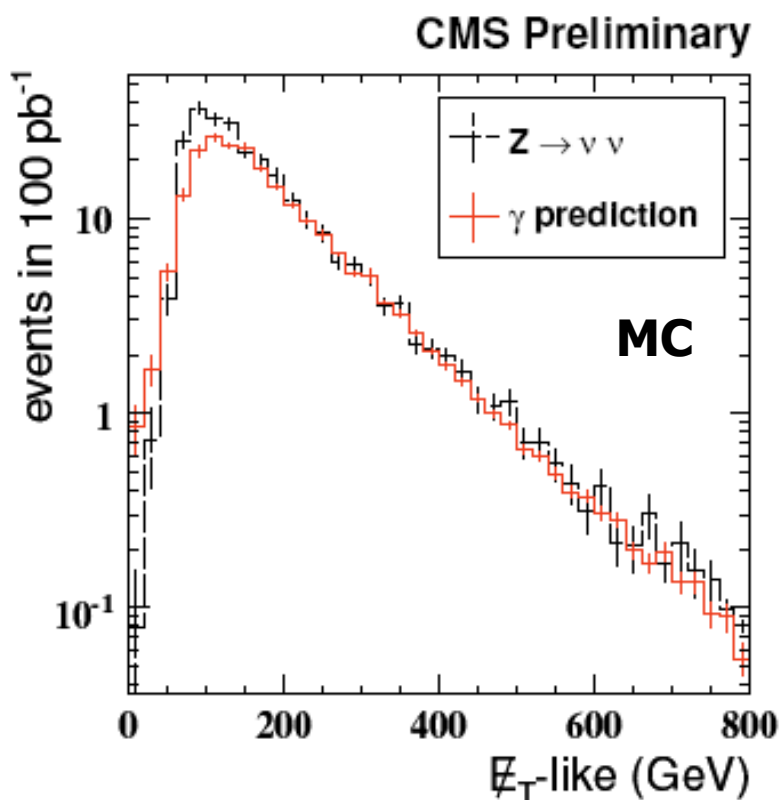
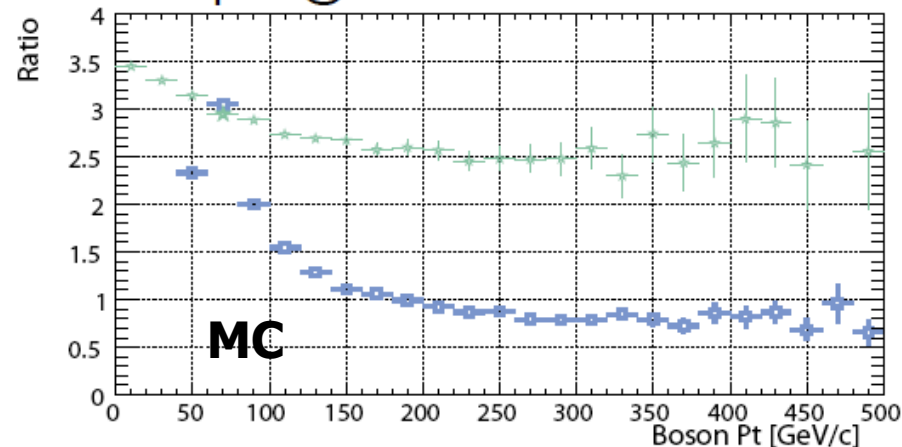
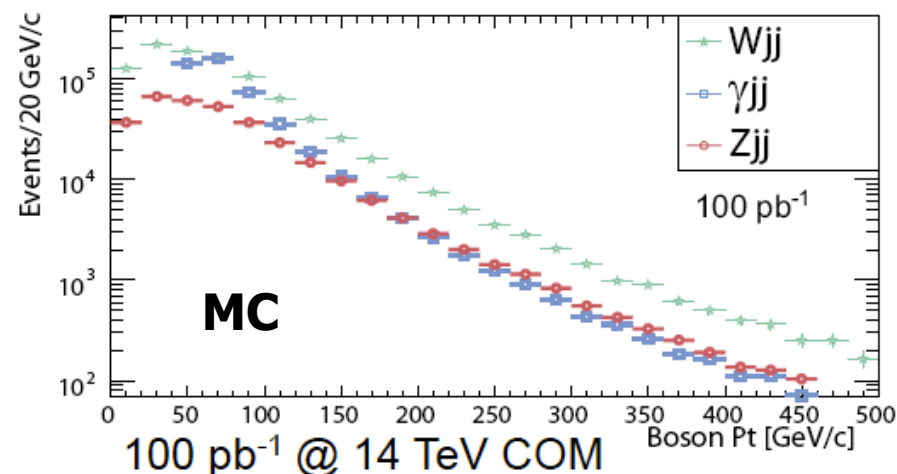


**γ + jets**  
 Strength: large statistics and clean at high  $E_T$   
 Weakness: background at low  $E_T$ , theoretical errors

# Z+jets $\rightarrow$ $\nu\nu$ + jets

CMS-PAS-SUS-08-002

- Select  $\gamma$  +  $\geq 3$  jets with  $E(\gamma) > 150$  GeV
- Remove photon from the event
- Recalculate MET
- Normalise with  $\sigma(\text{Z+jets})/\sigma(\gamma\text{+jets})$  from MC or measurements



**Good agreement between prediction and estimation.**





# Matrix Method "à la DØ"

(or Tight/Loose Ratio)



An initial sample containing  $N_{\text{loose}}$  events  
→ Applying an additional cut to reach a second sample containing  $N_{\text{tight}}$  events which is a subset of the initial sample

Each sample contains a given number of signal ( $N_{\text{real}}$ ) like and background ( $N_{\text{fake}}$ ) like. Fraction are changing as follow:

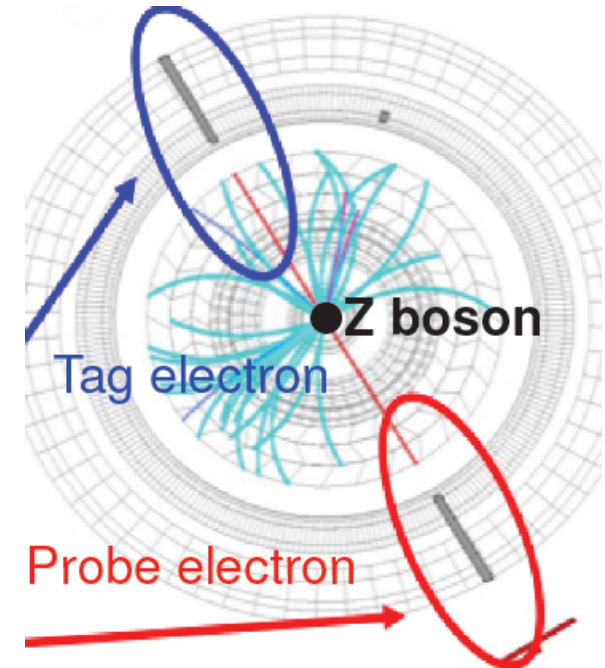
$$\begin{array}{l} \text{Cut} \left\{ \begin{array}{l} N^{\text{loose}} \\ N^{\text{tight}} \end{array} \right. \end{array} = \begin{array}{l} N_{\text{real}}^{\text{loose}} \\ \epsilon_{\text{real}} N_{\text{real}}^{\text{loose}} \end{array} + \begin{array}{l} N_{\text{fake}}^{\text{loose}} \\ \epsilon_{\text{fake}} N_{\text{fake}}^{\text{loose}} \end{array}$$

Challenge: calculating  $\epsilon_{\text{real}}$  and  $\epsilon_{\text{fake}}$

Mainly used to determine multi jets background in analysis selecting on leptons.

**When using leptons, use Tag and Probe to compute  $\epsilon_{\text{tight}}$ :**

- require a the  $l^+l^-$  pair to be within a  $m_Z$  window
- high lepton purity can be reached with tight ID cuts on the “tag” and the  $m_Z$  window

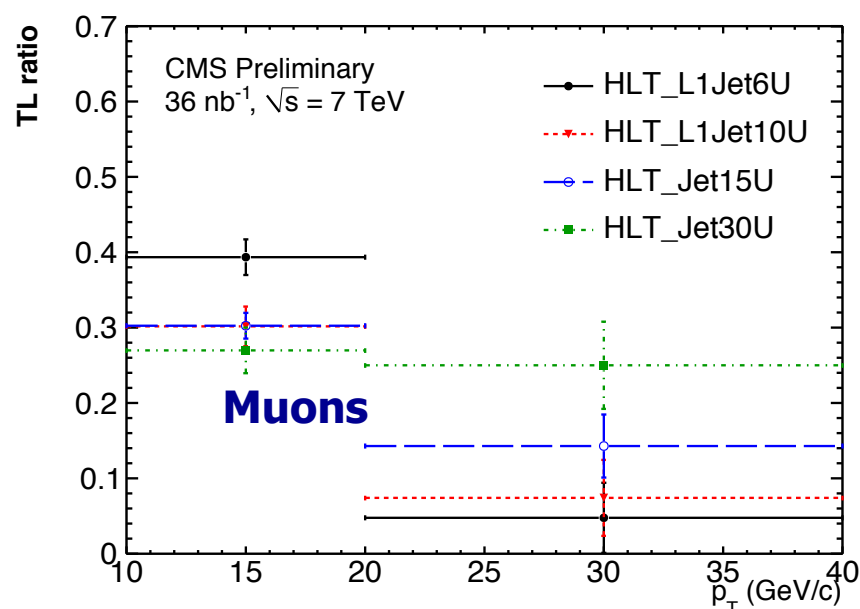
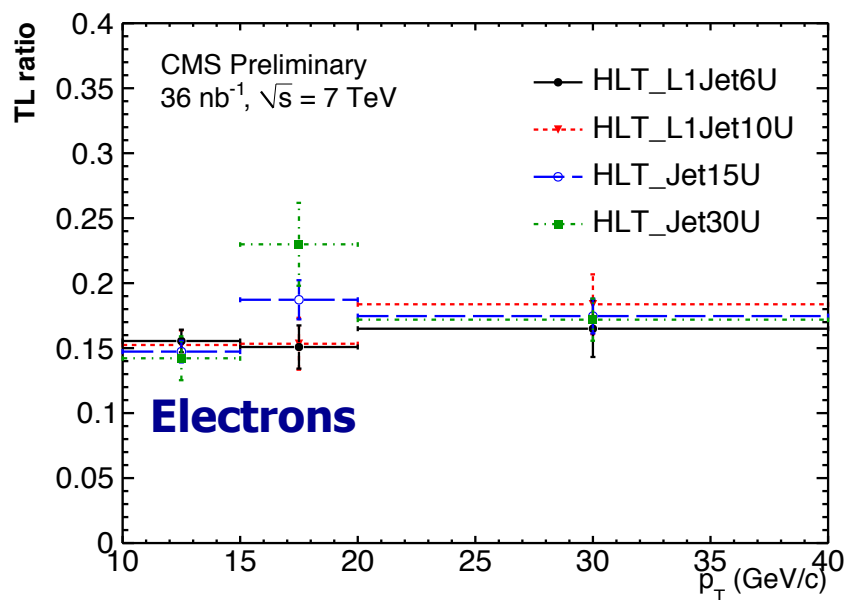


**For  $\epsilon_{\text{fake}}$ , look for background dominated samples (jets dominated samples, lepton-jets back to back or W+jets with W in the other lepton flavor)**

# Same Sign Searches

Use a jets dominated control sample (loose lepton-id & isolation) to measure  $\epsilon_{\text{fake}}$  (= "TL ratio") as function of kinematics variables

Tight-to-Loose-Ratios using different jet-triggered samples



Consistency in predicted & observed number of events.

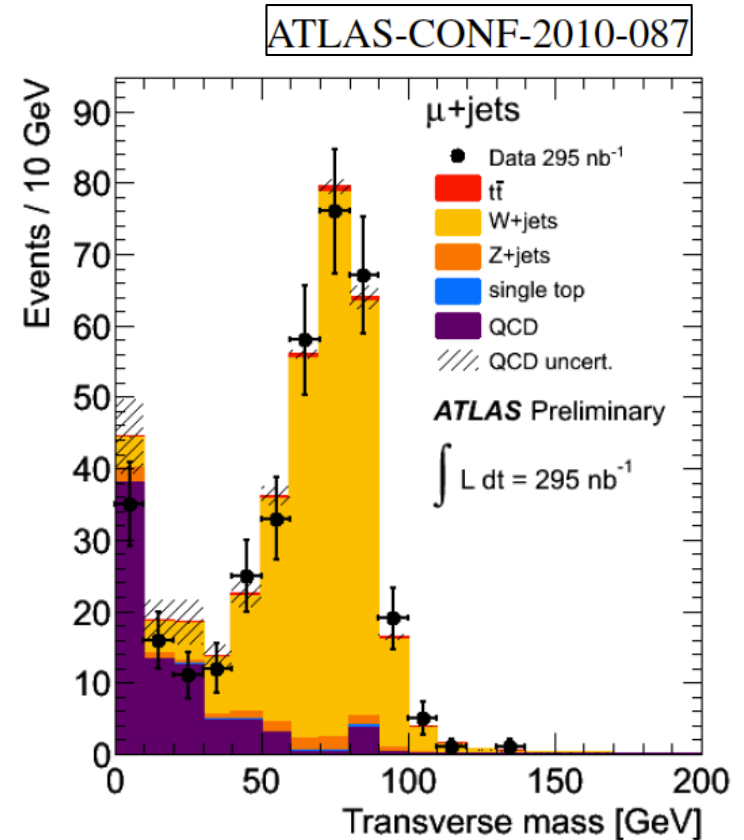
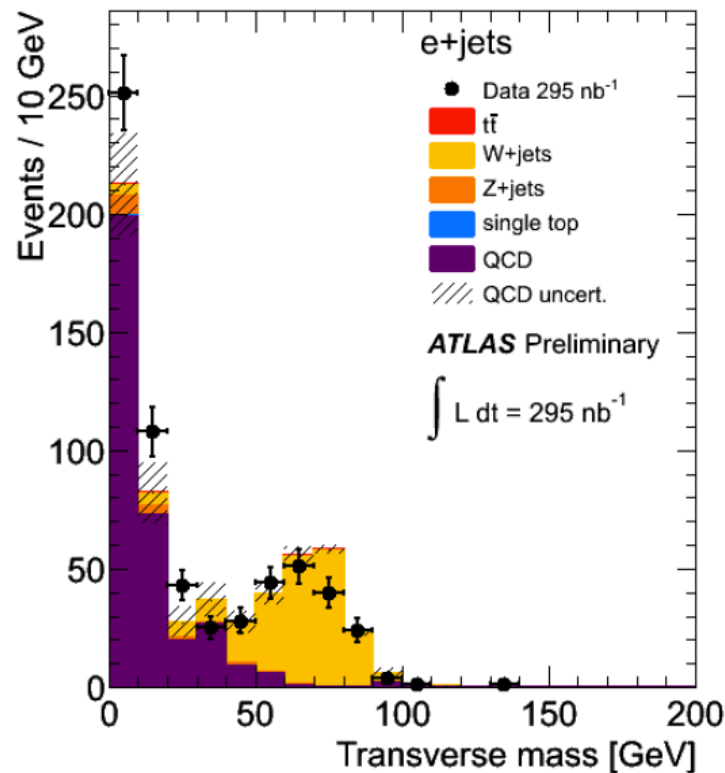
Channel	Predicted	Observed
$ee$	$0.43^{+0.18}_{-0.14}$	0
$e\mu$	$0.14^{+0.18}_{-0.09}$	1
$\mu\mu$	$0.22^{+0.51}_{-0.18}$	0



# Top Rediscovery

In sample for  $\epsilon_{\text{fake}}$ , contamination of signal can appear.  
Equation of  $N_{\text{loose}}$  and  $N_{\text{tight}}$  can be rewritten and by iteration,  
bias on  $\epsilon_{\text{fake}}$  can be removed.

QCD is estimated by this method in each of the bin of the distribution for semi-leptonic top searches.



ATLAS-CONF-2010-087

Fair agreement between data and the sum of MC samples and multijets estimation.



# Extension to Di-Leptons



The system of equation can be written for Di-Lepton final states searches:

$$\begin{bmatrix} N_{TT} \\ N_{TL} \\ N_{LT} \\ N_{LL} \end{bmatrix} = \begin{bmatrix} rr & rf & fr & ff \\ r(1-r) & r(1-f) & f(1-r) & f(1-f) \\ (1-r)r & (1-r)f & (1-f)r & (1-f)f \\ (1-r)(1-r) & (1-r)(1-f) & (1-f)(1-r) & (1-f)(1-f) \end{bmatrix} \begin{bmatrix} N_{RR} \\ N_{RF} \\ N_{FR} \\ N_{FF} \end{bmatrix}$$

**With:**

$$\mathbf{f} = \epsilon_{\text{fake}}$$

$$\mathbf{r} = \epsilon_{\text{real}}$$

$N_{TT}$  = Number of events in Tight-Tight

$N_{LL}$  = Number of events in Loose-Loose

By solving the equation, each sample composition ( $N_{RR}$  = Number of events containing two real leptons) can be found.



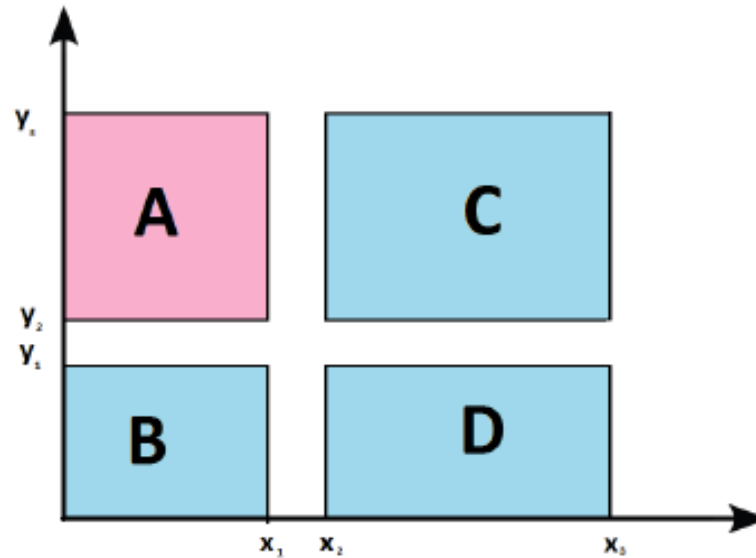
# Matrix Method "à la CDF"



(ABCD method/ $M_T$ /Tiles)

Simplified version of the matrix method "à la DØ".

Splitting a 2D phase space by 2 criteria to obtain a signal like area and background like area:



**Hypothesis:**

- Neglecting signal contribution in regions B and D
  - X variables has no effect on studied background
  - Assuming that variables x and y are uncorrelated
- Number of background events in signal region A can be evaluated as  $N_A = N_B \times N_C / N_D$ .

**Main issue: find uncorrelated variables**



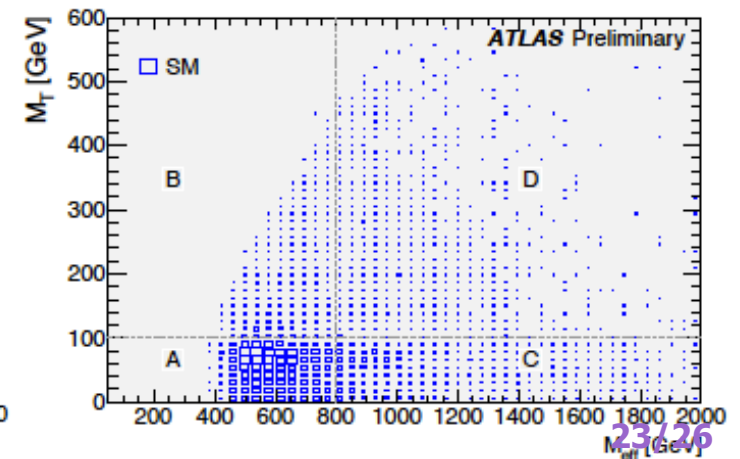
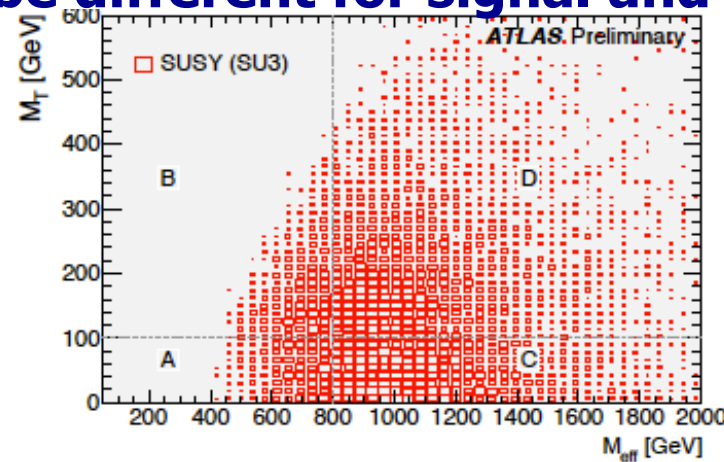
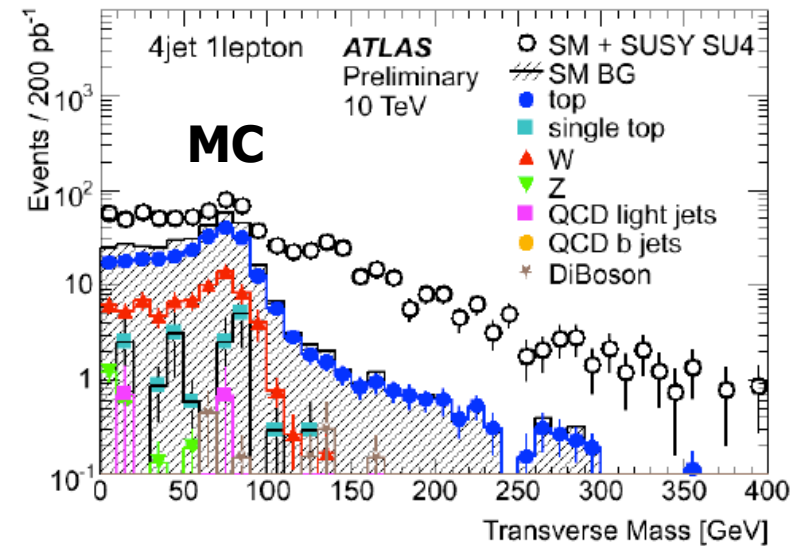
# Tiles Method

ATL-PHYS-PUB-2009-077

Variation of Matrix Method "à la CDF":  
Use  $M_T$  and  $M_{\text{eff}} (= \sum E_T \text{ of ALL objects})$  as the two variables  
( $M_T > 100 \text{ GeV}$ ,  $W$  decay is background).  
Each quadrant is named tiles.

Hypothesis:

- Relative inclusive fractions of SM background events in each tile are predicted by MC simulation.
- Discriminating variables are mutually independent for signal events.
- In presence of signal, the distributions of events among the tiles need to be different for signal and background.





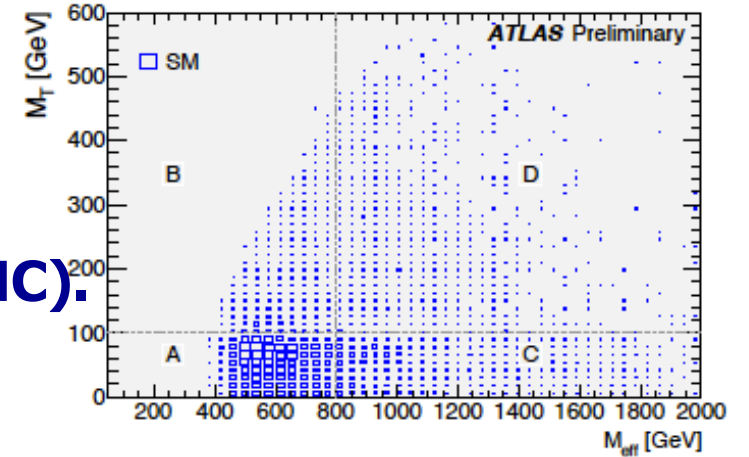
# 2x2 Tiles Method

In each of the tile:

$$\begin{aligned}\bar{N}_A &= f_A^{\text{SM}} \bar{N}^{\text{SM}} + f_A^{\text{S}} \bar{N}^{\text{S}}, & \bar{N}_B &= f_B^{\text{SM}} \bar{N}^{\text{SM}} + f_B^{\text{S}} \bar{N}^{\text{S}} \\ \bar{N}_C &= f_C^{\text{SM}} \bar{N}^{\text{SM}} + f_C^{\text{S}} \bar{N}^{\text{S}}, & \bar{N}_D &= f_D^{\text{SM}} \bar{N}^{\text{SM}} + f_D^{\text{S}} \bar{N}^{\text{S}}\end{aligned}$$

Where the  $f$  represents respectively the fraction of SM/Signal in a given tile (from MC).  
Requiring further that the signal variables be independent:

$$\begin{aligned}f_A^{\text{S}} &= (1 - f_{M_{\text{eff}}}^{\text{S}})(1 - f_{M_T}^{\text{S}}), & f_B^{\text{S}} &= (1 - f_{M_{\text{eff}}}^{\text{S}})f_{M_T}^{\text{S}}, \\ f_C^{\text{S}} &= f_{M_{\text{eff}}}^{\text{S}}(1 - f_{M_T}^{\text{S}}), & f_D^{\text{S}} &= f_{M_{\text{eff}}}^{\text{S}}f_{M_T}^{\text{S}},\end{aligned}$$



→ System can be solved:

$$\begin{aligned}N^{\text{SM}} &= \frac{1}{2(f_A f_D - f_B f_C)} \left\{ f_D N_A - f_C N_B - f_B N_C + f_A N_D \right. \\ &\quad - \left[ \left( -(f_C N_B) - f_D (N_A + 2N_B) + f_B N_C + f_A N_D + 2f_B N_D \right)^2 \right. \\ &\quad \left. \left. - 4(f_D N_B - f_B N_D) \left( (f_C + f_D) (N_A + N_B) - (f_A + f_B) (N_C + N_D) \right) \right]^{1/2} \right\}\end{aligned}$$

→ And signal:  $N^{\text{S}} = N_A + N_B + N_C + N_D - N^{\text{SM}}$





# NxN Tiles Method

ATL-PHYS-PUB-2009-077

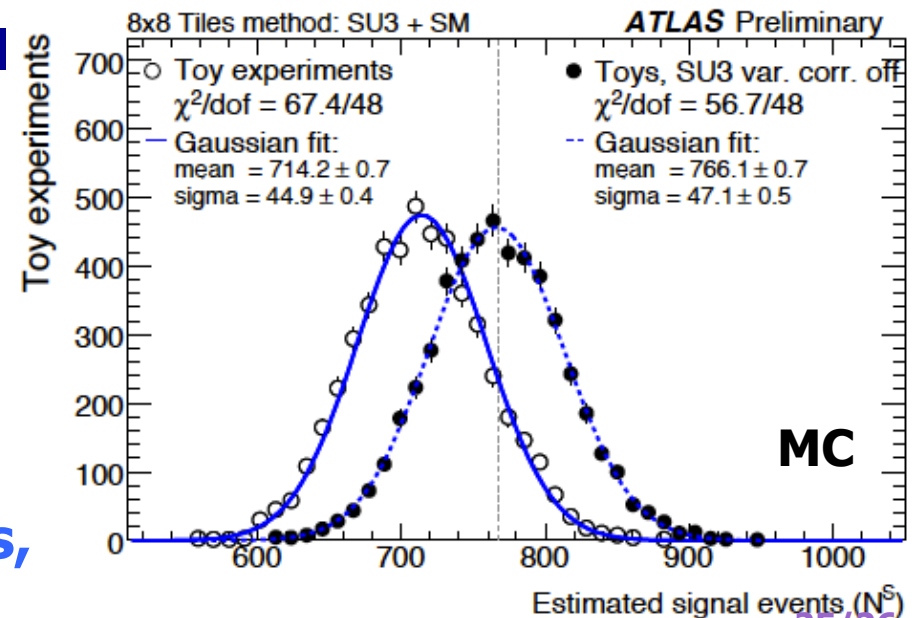
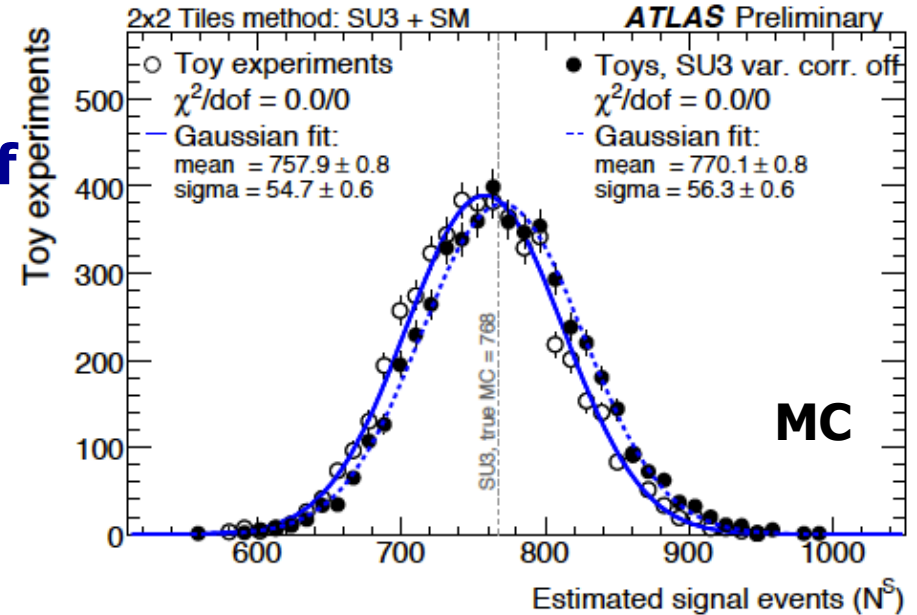
Split the phase space in N tiles,  $N^2$  equations can be written.

Ignoring signal correlation in each of the tiles, the problems is over constraint

→ Define extended negative log-likelihood: 
$$-\ln \mathcal{L} = \sum_{i,j=1}^n (\bar{N}_{ij} - N_{ij} \ln \bar{N}_{ij})$$

Minimizing  $-\ln \mathcal{L}$  ⇔ solving an unbinned maximum-likelihood (ML) fit, where the background and signal probability density functions (PDF) are one two-dimensional and two one-dimensional binned histograms.

- Improve information content of the fit (more precise determination)
- Probes the signal shape in 2D
- But signal correlation in each tiles, induce a bias...





# Conclusion

- **LHC is delivering a huge chunk of data that experiments are currently using for commissioning and looking for new physics.**
  - **A large variety of method to estimate SM process from data have been looked at over MC to understand the bias and are currently exercised on data.**
  - **The variety of methods allows cross check and combination of them to reduce systematic/bias.**
- Moriond results will integrate all this and perhaps we will see some signal above the SM background...**

# Charge Asymmetry

**In case of dilepton searches, use the symmetry in the charge of multijet background to determine it.**

**Same sign searches:**

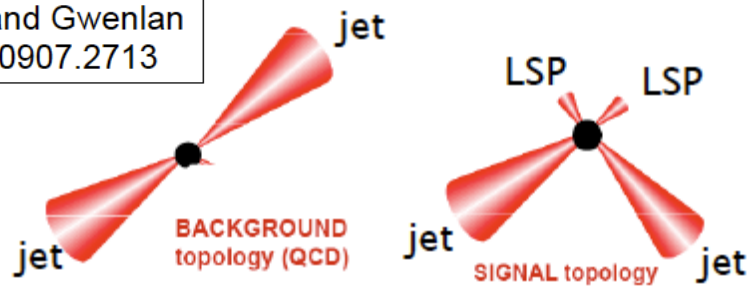
- **Very low Standard Model background rate**
- **Backgrounds from charge mis-identified**

**Opposite sign searches:**

- **Use opposite-sign, opposite-flavor sample to subtract SM background**

# New Variables: All Hadronic Searches

Barr and Gwenlan  
arXiv:0907.2713



$$\alpha_T = \frac{E_{Tj2}}{M_{Tj1j2}} = \frac{\sqrt{E_{Tj2}/E_{Tj1}}}{\sqrt{2(1-\cos\Delta\phi)}}$$

**A new variable combining angular and energy measurements ( $\alpha_T$ )**  
**No dependence on MET  $\rightarrow$  robust**  
**Originally proposed for di-jet events**  
 **$\rightarrow$  generalised up to 6 jets**  
**Perfectly balanced events have  $\alpha_T = 0.5$**   
**Mis-measurement of either jet leads to lower values**  
**Studies the variation of the variable as function of others**

PRL101:221803 (2008) & CMS-PAS-SUS-09-001

