

# MATRIX ELEMENT METHOD

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## *Definition*

- Commonly referred as the method that includes the matrix elements calculations for signal and background events to evaluate probability densities in an event-by-event basis
- Sometimes referred as any analysis that uses matrix element as variables, weights or other way
  - ideogram, neutrino weighting, etc. could fit in this category

## *Measurements and searches*

- Originally developed to measure the top quark mass and W helicity in top events ( $\alpha = m_t, f_{0,+,-}$ )

$$P(\alpha)$$

- Later adapted to searches
  - Main of the effort was made for observation of single top and Higgs

$$P$$

## Building $P$ in an ideal case

- Probability density function for ONE event is characterized by a set of measurements  $x$ , for a set of parameters  $\alpha$

$$P(x | \alpha)$$

for a perfect detector

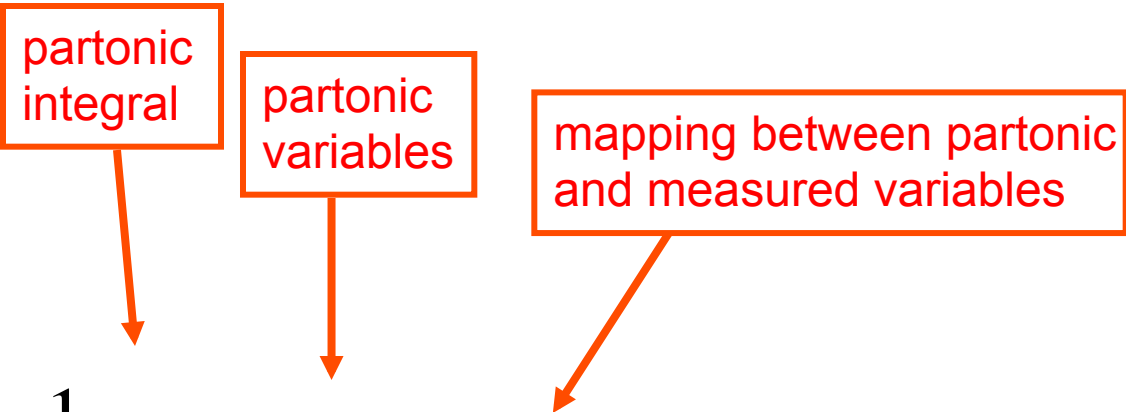
$$P(x | \alpha) dx = \frac{d\sigma(x | \alpha)}{\sigma(\alpha)}$$

measured  
variables

parameters

$$d\sigma \approx \frac{|M(\alpha)|^2}{\text{flux factor}} \times \text{phase space}$$

## *Building $P$ with detector resolutions*


$$P_W(x|\alpha) = \frac{1}{\sigma(\alpha)} \int d\sigma(y|\alpha) W(y,x)$$

Note that  $W(y,x)$  can be a function of parameters like JES

# Building $P$ - Acceptance

Acceptance limited by the physical properties of the detector and the event selection

$$P(x | \alpha) = \frac{A(x) \int d\sigma(y | \alpha) W(y, x)}{\sigma(\alpha) \langle A(\alpha) \rangle}$$

Mean acceptance

$$\langle A(\alpha) \rangle = \int A(x) P_W(x | \alpha) dx$$

# Background

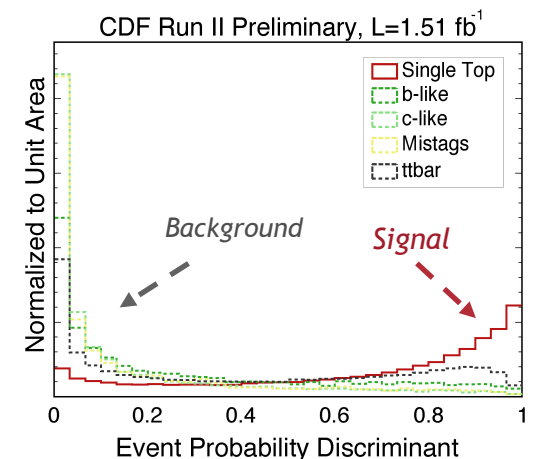
- Measurement: sum over all states that can lead to the set of measurements  $x$

$$P(x | \alpha) = \sum_{i=\text{states}} c_i P_i(x | \alpha)$$

- Search: method is turned into a single-variable template analysis

$$EPD = \frac{P_{\text{signal}}}{P_{\text{signal}} + P_{\text{background}}}$$

In principle the extraction of signal is not different than any other template analysis



# Likelihood for searches

Example single top at CDF

$$\mathcal{L}(\beta_1, \dots, \beta_5; \delta_1, \dots, \delta_{10}) = \underbrace{\prod_{k=1}^B \frac{e^{-\mu_k} \cdot \mu_k^{n_k}}{n_k!}}_{\text{Poisson term}} \cdot \underbrace{\prod_{j=2}^5 G(\beta_j | 1, \Delta_j)}_{\text{Gauss constraints}} \cdot \underbrace{\prod_{i=1}^{10} G(\delta_i, 0, 1)}_{\text{Systematics}}$$

Expected mean in bin k:

$$\mu_k = \sum_{j=1}^5 \beta_j \cdot \underbrace{\left\{ \prod_{i=1}^{10} [1 + |\delta_i| \cdot (\epsilon_{ji+} H(\delta_i) + \epsilon_{ji-} H(-\delta_i))] \right\}}_{\text{Normalization Uncertainty}}$$

$$\cdot \underbrace{\alpha_{jk}}_{\text{Shape } P.} \cdot \underbrace{\left\{ \prod_{i=1}^{10} (1 + |\delta_i| \cdot (\kappa_{jik+} H(\delta_i) + \kappa_{jik-} H(-\delta_i))) \right\}}_{\text{Shape Uncertainty}}$$

$\beta_j = \sigma_j / \sigma_{SM}$  parameter

single top (j=1)

W+bottom (j=2)

W+charm (j=3)

Mistags (j=4)

ttbar (j=5)

k = Bin index

i = Systematic effect

$\delta_i$  = Strength of effect

$\epsilon_{j\pm} = \pm 1\sigma$  norm. shifts

$\kappa_{jik\pm} = \pm 1\sigma$  shift in bin k

ⓂCorrelation between Shape/Normalization uncertainty included ( $\delta_i$ )

ⓂProfile Likelihood with respect to all nuisance parameters



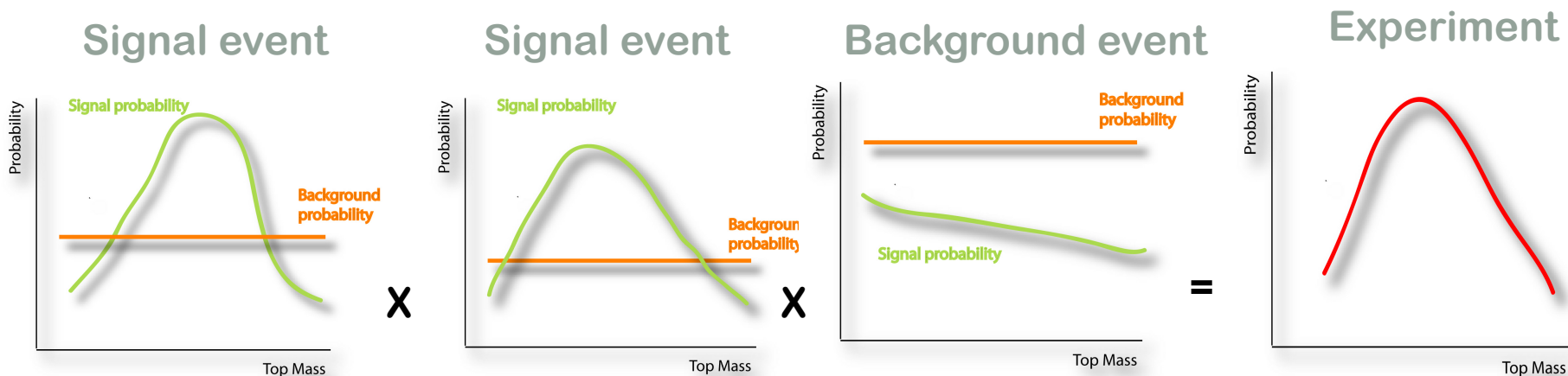
# Likelihood for measurements

- Extracting a set of parameters  $\alpha$  given  $N$  events is obtained by maximizing

$$L(\alpha, c_s) \propto \prod_{i=1}^N P(x_i | \alpha) \quad \left(\text{with } \int P(x | \alpha) dx = 1\right)$$

- Fraction of signal events is obtained simultaneously

$$P(x | \alpha) = c_s P_s(x | \alpha) + (1 - c_s) P_b(x)$$



# History

- Introduced in 2000 to measure  $m_t$ 
  - Very limited statistics, 2 to 1 background contamination
  - Until then  $m_t$  was measured using a template of the reconstructed top mass
  - D0 experiment had no b-tagging. CDF much better precision on  $m_t$
  - The goal of the MEM analysis was to use more information with less dependence on the MC
- Before 2000 there were similar ideas
  - K. Kondo: J. Phys. Soc. Japan, 57, 4126 (1988), J. Phys. Soc. Japan, 60, 836 (1991), J. Phys. Soc. Japan, 62, 1177 (1993)
  - R.H. Dalitz and G.R. Goldstein: Phys. Rev. D45, 1541 (1992), Phys. Lett. B287, 225 (1992), Phys. Rev. D47, 967 (1993) (w/ K. Sliwa), J. Mod. Phys. A9, 635 (1994), Proc. Roy. Soc. Lond. A455, 2803 (1999)

## History


- In 2004, D0 experiment Nature 429,638 (2004)
  - First complete measurement, including: 1) all detector effects (e.g. reconstruction efficiencies, cuts, trigger, ...), 2) correct normalization, 3) background probabilities, 4) MC tests of linearity, 5) pull calculations and 6) estimation of systematic effects.
- After 2004, it has been applied to all  $t\bar{t}$  channels, single top, WH, H to WW at the Tevatron and some LHC use (H to ZZ (4l and 2l2j), VH, H to WW)

## Technical details – Matrix element

- Matrix Element

- In ttbar LO (Mahlon-Parke)
- For searches Madgraph is usually used for single top and Higgs (HELAS)

$$P(x | \alpha) = \frac{1}{\sigma(\alpha)} \int d\sigma(y | \alpha) W(y, x | JES) f(x_{Bj}^1) f(x_{Bj}^2) dx_{Bj}^1 dx_{Bj}^2$$



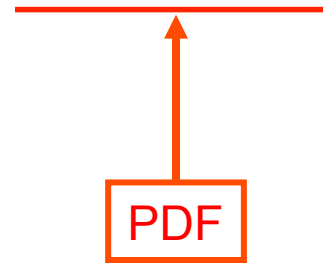
$$d\sigma \approx \frac{|M(\alpha)|^2}{\text{flux factor}} \times \text{phase space}$$

The ME @ LO limits the size of the sample that can be described well by P  
 For example: ttbar in l+jets has 4 jets at LO, we could calculate P for the 4 leading jets in the >4 sample, but not optimally

## *Technical details - Parton Distribution Functions*

- Use CTEQ routines (LO vs NLO)

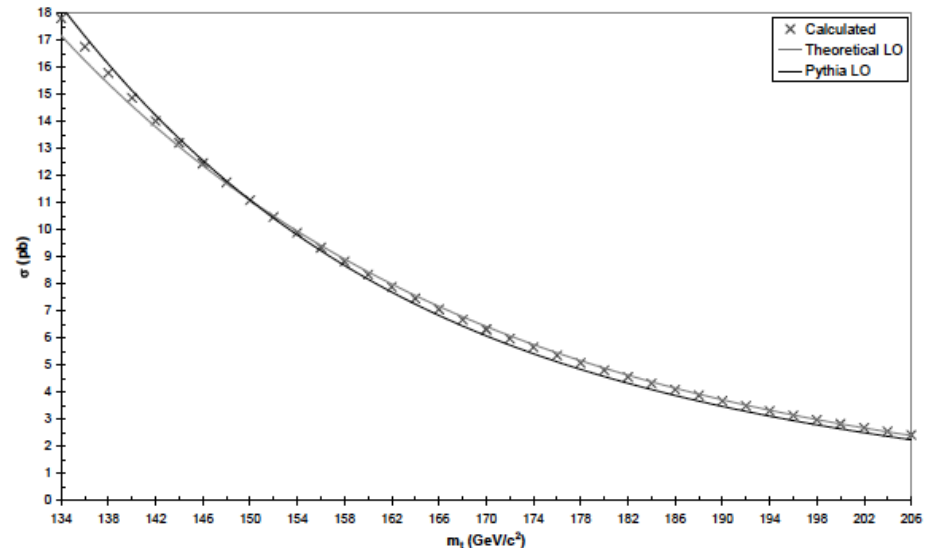
$$P(x|\alpha) = \frac{1}{\sigma(\alpha)} \int d\sigma(y|\alpha) W(y, x | JES) f(x_{Bj}^1) f(x_{Bj}^2) dx_{Bj}^1 dx_{Bj}^2$$



## Technical details - Normalization I

- Cumbersome and CPU intensive
  - Allows to test assumptions and to debug method
  - Sometimes used the  $\sigma(\alpha)$  or  $\sigma$  obtained from MC

$$P(x | \alpha) = \frac{1}{\sigma(\alpha)} \int d\sigma(y | \alpha) W(y, x | JES) f(x_{Bj}^1) f(x_{Bj}^2) dx_{Bj}^1 dx_{Bj}^2$$

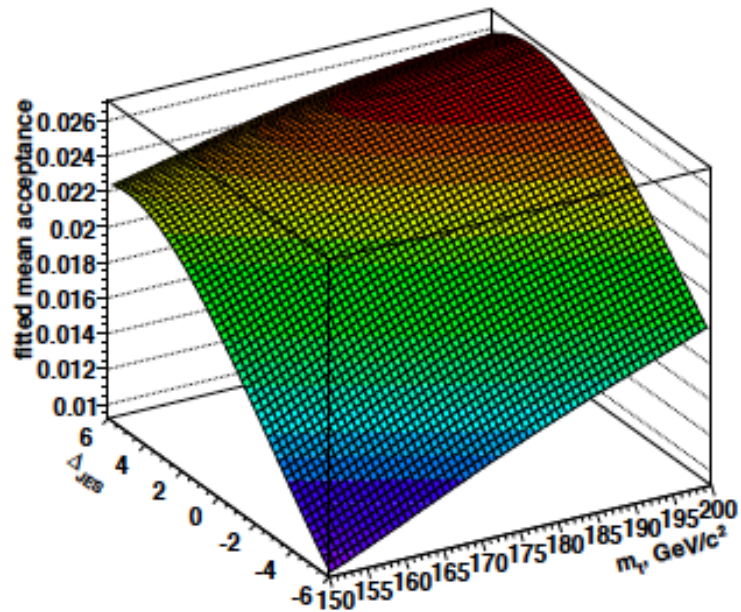
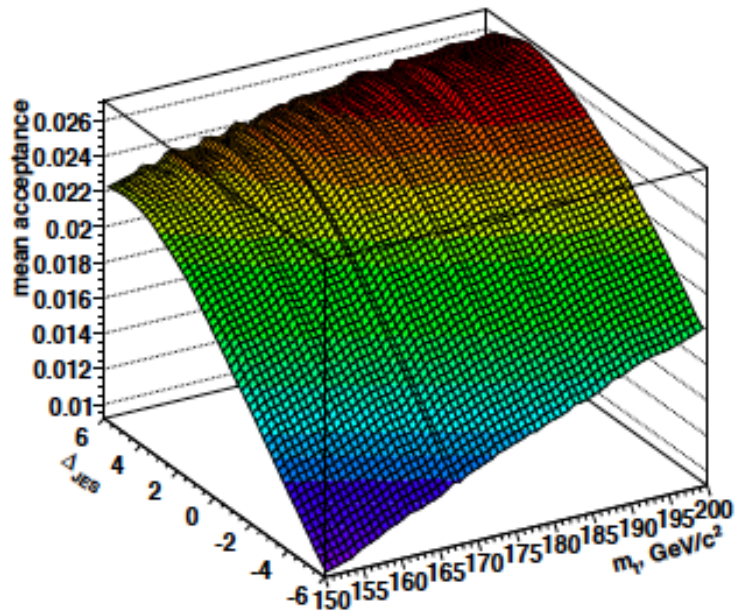


## Technical details - Normalization II

- Mean acceptance calculated as function of the parameters to measure

$$\langle A(\alpha) \rangle = \int A(x) P_W(x | \alpha) dx$$

$$\langle A(\alpha) \rangle = \frac{n_{\text{accepted}}(\alpha)}{n_{\text{produced}}(\alpha)}$$



## *Technical details - Background*

- Use MC matrix element generators

$$P_b(x) = \frac{1}{\sigma_b} \int d\sigma_b(y) W(y, x)$$

- Note that not all the background processes can be calculated in  $P_b$  (fakes, multijets)
  - Creative ways have been used to develop pdfs for these backgrounds



## *Technical details - Integration*

- Careful choice of integration variables
- Careful choice of assumptions in transfer functions
- Use narrow width approximation to smooth out integrand
- Convergence tests
  
- Methods
  - VEGAS most used for background (larger number of integrations)
  - Radmul (adaptive quadrature) (used for low number of integrations)
  - DIVONNE/CUBA
  
- CPU (2011) (depends on precision required) in a 2.0 GHz
  - Top mass with 5 integrations: 4 sec per event per point per jet parton-assignment (12 comb x 31 mass points x 17 JES points = 7 hours/event)
  - Single top (3 integrations): 1 s to 10 s per event all processes but ttbar in 3 jets (6 integrations) 5 mins
  - Both analyses were close to a million CPU-hours

## Transfer functions

- Describe hadronization, detector resolution and reconstruction effects, including the deposition of energy from a parton outside the corresponding jet algorithm and extra energy from underlying event

$$W_{jet}(E_{jet}, E_{parton}) = \frac{1}{\sqrt{2\pi}(p_1 + p_2 p_5)} \left[ \exp\left(-\frac{(\delta_E - p_1)^2}{2p_2^2}\right) + p_3 \exp\left(-\frac{(\delta_E - p_4)^2}{2p_5^2}\right) \right]$$

$$\text{where: } p_i = a_i + b_i E_{parton} \quad \delta E = (E_{parton} - E_{jet})$$

- Assume lepton well measured, angles and energy (delta functions)
- Jes, parameterized function for energies (and angles)
  - Use Pythia MC
  - Done for different jet flavors and different eta regions using the same event selection
- Use more information by including tracks, fraction of had and em energy,  $p_T$  of jet, etc. in NN to obtain a new “ $E_{jet}$ ”

## *Jet-parton permutations and neutrino solutions*

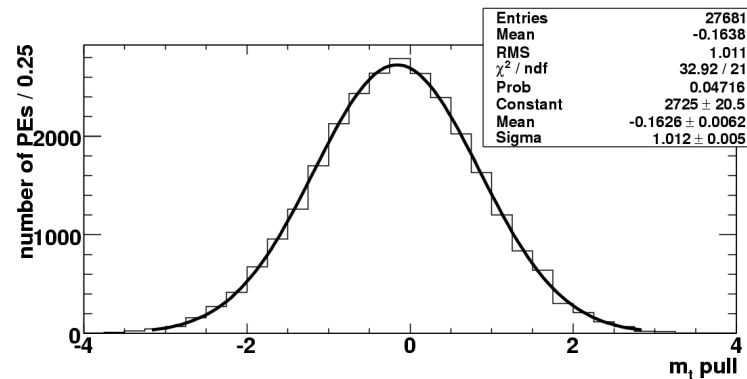
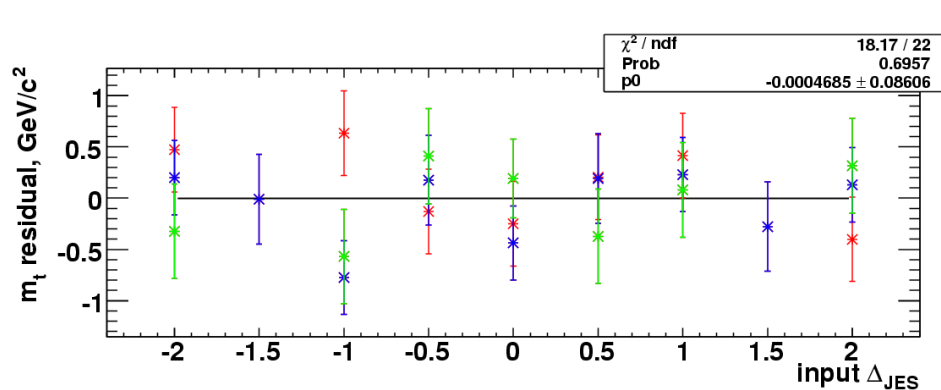
- One of the more appealing advantages of the MEM method is that each jet-parton permutation (or lepton) enters with a different weight

$$P(x|\alpha) = \frac{1}{\sigma(\alpha)\langle A(\alpha)\rangle} \sum_{jet-parton}^{2,6,12} \int |M(\alpha)|^2 W(y,x|JES) f(x_{Bj}^1) f(x_{Bj}^2) dx_{Bj}^1 dx_{Bj}^2 d\Phi$$

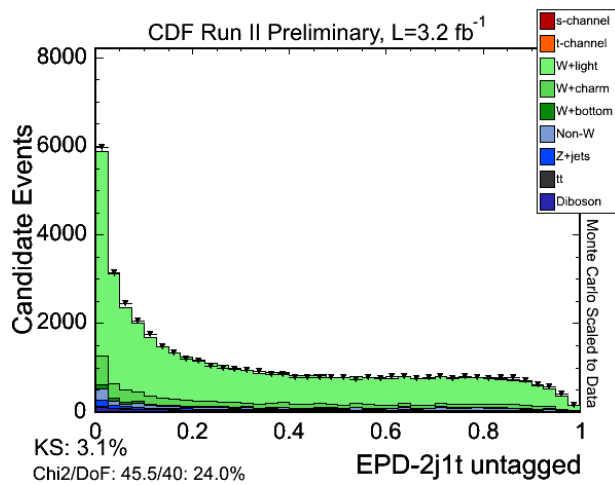
- For example in ttbar:
  - Most of the permutations do not contribute to P
  - The right permutation always contribute (among the first 3)
- The neutrino solutions are part of the integration

# Tests and checks

- Linearity and pull tested with MC



- EPD data/MC in different control regions



## *Is it a better method ?*

- In measurements with  $\delta_{\text{stat}} > \delta_{\text{sys}}$  benefit from MEM approach
  - In original mt analysis had a factor of 2 improvement using the same data !
  - If statistics is large, MEM or a simple Mt reco reach the similar sensitivity  
That being said, more statistical power can help decrease systematics
- The current case of the top mass
  - All uncertainties between 0.1-0.6 GeV
  - Total uncertainty in one analysis is about 1 GeV (0.5 stat + 0.8 syst)
  - Going below will require more understanding of color reconnection, initial/final state radiation, etc.
  - The statistical power of the MEM could help decrease systematics

## *Is it a better method ?*

- In searches the ME has performed between 0% and 10% gain over the other multivariate analysis (NN and BDT)
  - Found to be 50-70% correlated (later combined)
  - MEM as template ?
  - NLO information not totally used ?
  - More detector information used by NN/BDT ?
  - Analyzer dependence ?
- In general systematic uncertainties have been found to be smaller in MEM results than with other methods

## *Summary and conclusions*

- MEM has provided experiments a different way to analyze data
  - Beyond cut and count and one-variable templates (larger stat power)
  - With complete analyzers control (vs NN)
- Many improvements and optimizations could be envisioned
- Experimentally is a very expensive analysis
  - It is delicate and slow
  - The pace at the LHC might not be right for MEM
- A more collaborative effort among experimentalist and theorists on creating MEM tools would be very beneficial