

# Model Unspecific Search in CMS: Results with latest data

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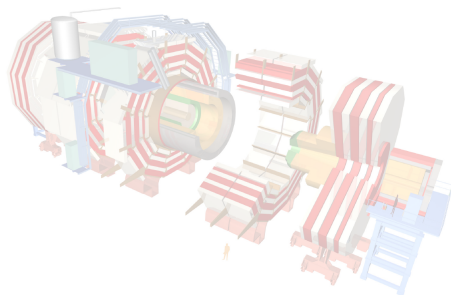
Terascale Alliance Workshop, Dresden 01.-03.12.2010



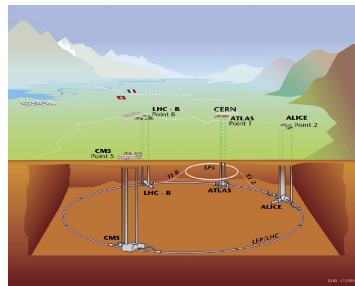
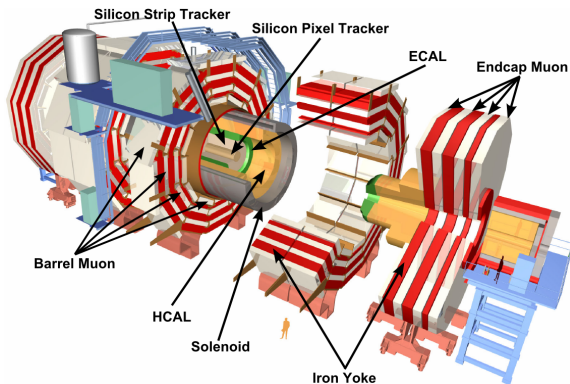
Bundesministerium  
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und Forschung

# Outline

- 1 Introduction
- 2 Implementation
- 3 Looking at Data
- 4 Summary/Outlook



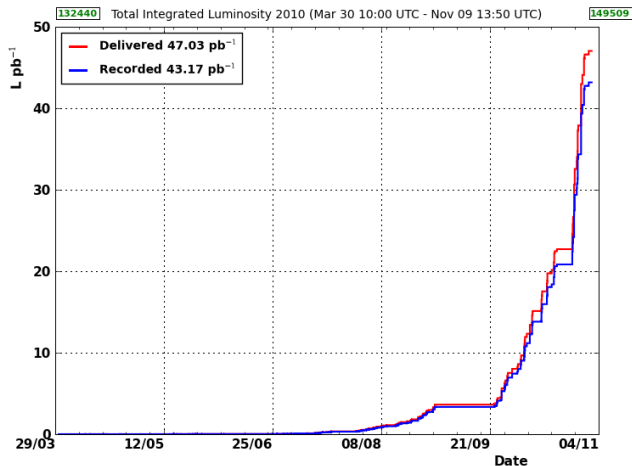
# CMS @ LHC



## CMS

- length  $\approx 21$  m
- diameter  $\approx 16$  m
- mass  $\approx 12500$  t
- solenoid up to  $\approx 4$  T

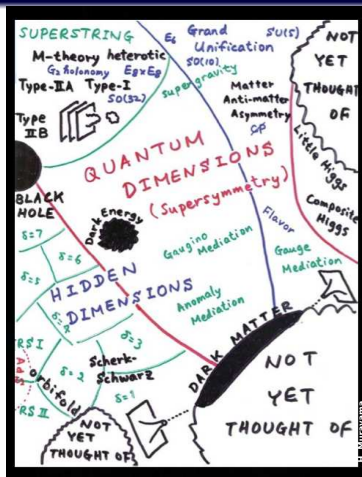
# Integrated Luminosity



## Data:

- Delivered: 47 /  $\text{pb}$
- Recorded: 43 /  $\text{pb}$
- Certified: 34 /  $\text{pb}$

# Motivation



## Challenge

Many competing models!

Can we have (dedicated) analyses for all of these?

What are the models no one has thought of yet?

## Example: SUSY

A whole “model framework” with many free parameters.

## Idea:

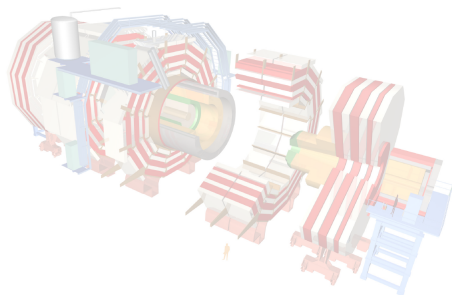
Model independent search!

## Minimise the theoretical bias:

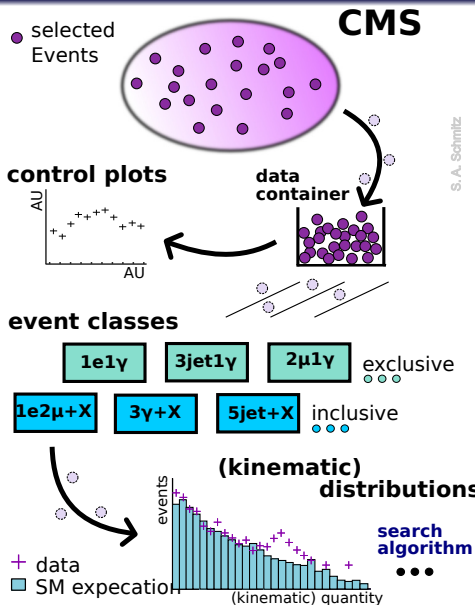
- Assume only one model: The **Standard Model**
- Search for deviations from the SM expectation in many final states

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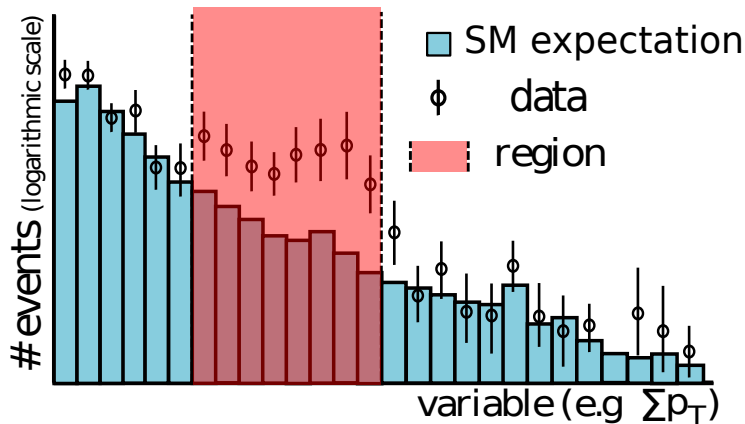
# MUSiC Concept



S.A. Schmitz

- **Select** good events and objects
- Sort events by their physics object content ( $\mu, e, \gamma, (b-)\text{jets}, \cancel{E}_T$ ) into **event classes**
- **Kinematic distributions** of interest:  $\sum p_T, M_{\text{inv}}, \cancel{E}_T$
- Run the **search algorithm** on these distributions
- Find the most significant **connected bin region** in every distribution

# Search algorithm



## Most significant region:

In **every distribution** of **every event class**: Find the region with the lowest probability of MC to deviate even more, i. e. the smallest ***p*-value**.



# MUSiC's $p$ -value

## Standard treatment:

Convolute a Poisson (statics) with a **Gaussian** (systematics) to model the uncertainties on the mean.

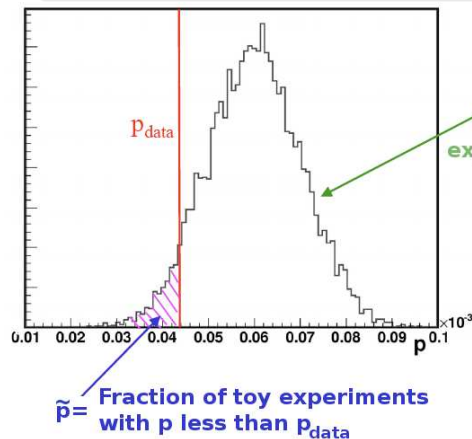
## Alternative approach:

Use a **Lognormal** prior.

- Better treatment of some uncertainties, worse for others.
- Good cross check.

# “Look-elsewhere effect”

Considering **many regions** in **many distributions** it becomes more probable to see a deviation by chance due to statistical fluctuations.

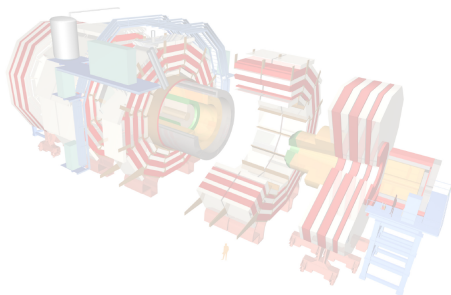


## Toy experiments:

- Randomise MC expectation bin by bin, taking all known uncertainties into account
- Scan for most significant region
- Count pseudo experiments with higher significance than data
- $\tilde{p}$  = Fraction of toy experiments with  $p_{\text{toy}} < p_{\text{data}}$

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# MUSiC on 7 TeV data

## Data:

Used SM backgrounds: Multijet, electroweak, top,  $\gamma$  + Jets, low mass resonances

## Major systematic uncertainties:

MC statistics, luminosity (11 %), cross section (5 %), PDF, JES (5 %)

## Cuts:

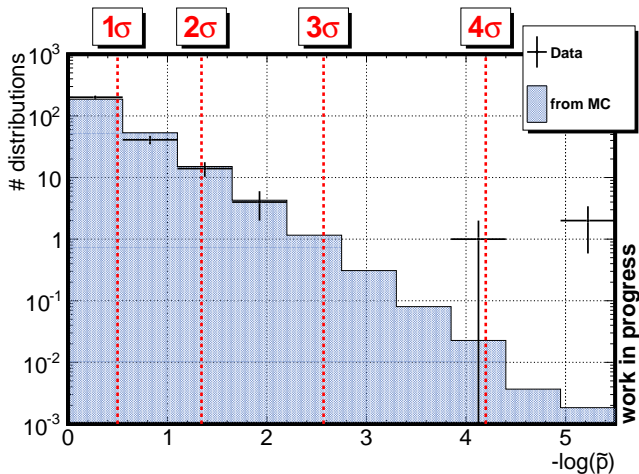
Require at least one e ( $p_T > 60$  GeV) **or**  $\mu$  ( $p_T > 25$  GeV)

Allows **loose**  $p_T$ -Selection for other objects

| Object   | $p_T^{\min}$ / GeV |
|----------|--------------------|
| e        | 25                 |
| $\gamma$ | 25                 |
| $\mu$    | 18                 |
| Jet      | 50                 |
| (MET     | 30 )               |

# $\tilde{p}$ distribution

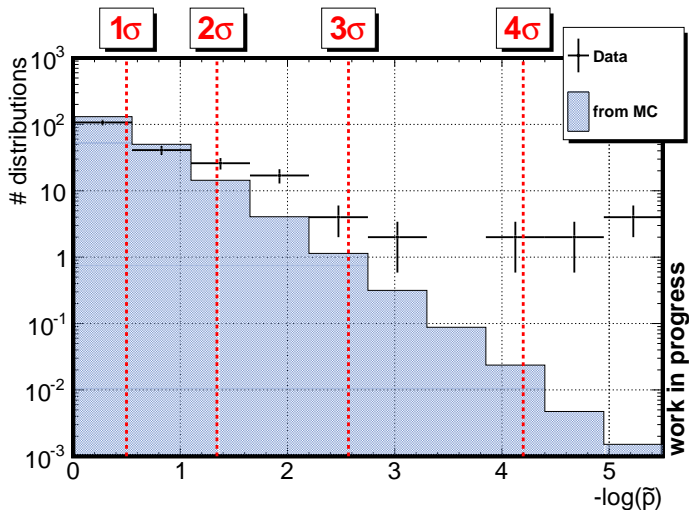
Without MET: Good agreement with MC prediction!



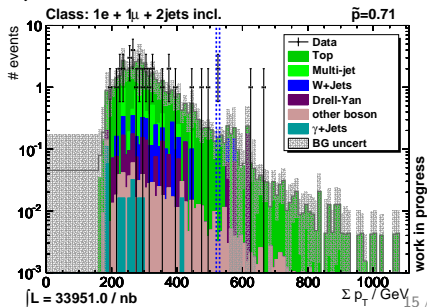
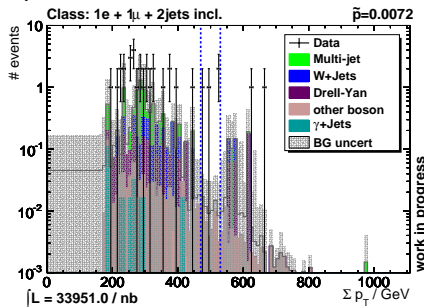
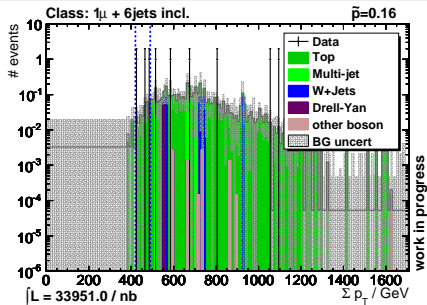
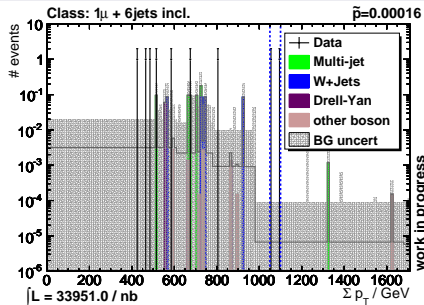
Few discrepant event classes can be explained by instrumental effects or simulation

# Cross check: Rediscovering the Standard Model

Removed the top quark MC samples:  
excess in a wide range of event classes.

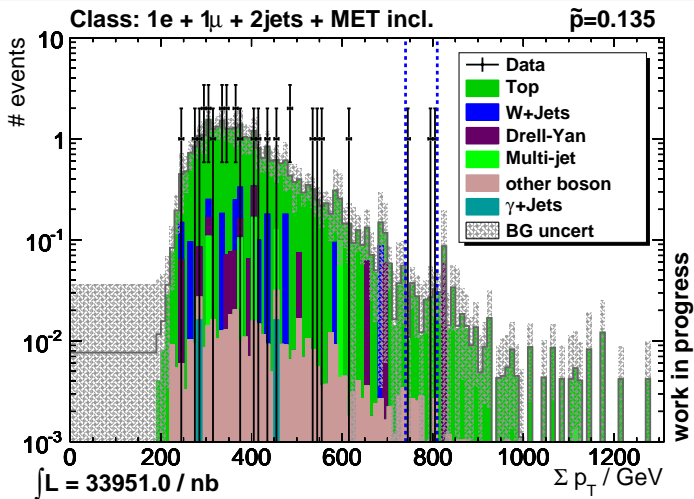


# $t\bar{t}$ “most significant”



# Preview: MET

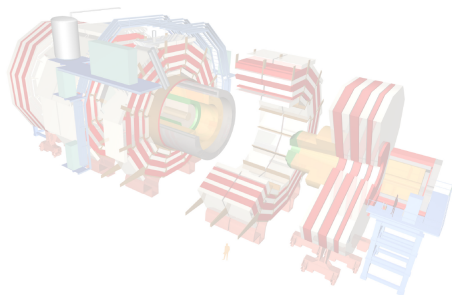
MET not yet fully implemented in MUSiC but:  
Promising first results





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# Summary and outlook

## Today:

- Model independent search that looks for deviations from the SM
- Complementary to dedicated analyses
- Possible deviations need detailed investigation

## Data results:

- Globally good agreement
- Few deviations left (under control)
- Able to rediscover SM physics (in non-standard final states)

## Future:

- Analyse more final states

Thank you for your attention!

# backup

# *p*-value

## Definition:

$$p \equiv \mathcal{P}(T(\mathbf{X}) \geq T(\mathbf{x}) \mid H_0)$$

With:

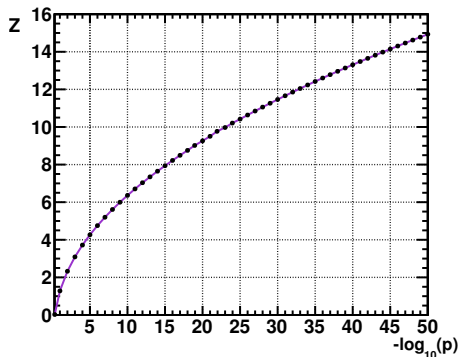
$\mathbf{X}$  = Possible data set

$\mathbf{x}$  = Observed data set

$T$  = Test statistic

= (number of entries in our case)

$H_0$  = Null hypothesis = SM



Often  $p$ -values are expressed in terms of **standard deviations**.

# MUSiC's $p$ -value

Use a **Gaussian prior** to model the uncertainties on the mean of a *Poisson counting experiment*  $\Rightarrow$  prior predictive  $p$ -value.

$$p^N = \begin{cases} \underbrace{\sum_{i=N_{\text{data}}}^{\infty} C}_{\text{normalisation}} \cdot \underbrace{\int_0^{\infty} d\lambda \exp\left(-\frac{(\lambda - N_{\text{SM}})^2}{2\sigma_{\text{SM}}^2}\right)}_{\text{systematics}} \cdot \underbrace{\frac{e^{-\lambda} \lambda^i}{i!}}_{\text{statistics}} & \text{if } N_{\text{data}} \geq N_{\text{SM}} \\ \underbrace{\sum_{i=0}^{N_{\text{data}}} C}_{\text{normalisation}} \cdot \underbrace{\int_0^{\infty} d\lambda \exp\left(-\frac{(\lambda - N_{\text{SM}})^2}{2\sigma_{\text{SM}}^2}\right)}_{\text{systematics}} \cdot \underbrace{\frac{e^{-\lambda} \lambda^i}{i!}}_{\text{statistics}} & \text{if } N_{\text{data}} < N_{\text{SM}} \end{cases}$$

with:

$N_{\text{SM}}$  = Pure SM (Monte Carlo) expectation in this region

$\sigma_{\text{SM}} = \sqrt{\sigma_{\text{stat}}^2 + \sum_i \sigma_{i,\text{syst}}^2}$  = Uncertainty on the MC prediction for the SM

Region with the smallest  $p$ -value: **Region of Interest**

Another approach: **Lognormal** prior

# The “look-elsewhere effect”

Considering **many regions** in **many distributions** it becomes more probable to see a deviation by chance due to statistical fluctuations.

The single region  $p$ -value alone is not a good significance estimator.

⇒ Compute “new” estimator  $\tilde{p}$ :

$$\tilde{p} = \frac{\text{number of } H_0 \text{ experiments with a region featuring } p < p_{\text{data}}}{\text{total number of } H_0 \text{ experiments}}$$

Easy (toy) example: all regions **statistically independent**

$$\tilde{p} = 1 - (1 - p_{\text{data}})^n ; \quad n = \text{number of regions}$$

In realistic cases regions are correlated due to shared bins and/or **correlated** systematic uncertainties!

## Solution:

Dice a sufficiently large number of pseudo experiments in order to determine  $\tilde{p}$ .

# Determination of $\tilde{P}$

- ① In every distribution the data gives **one most significant region**, i. e. smallest  $p$ -value ( $p_{\text{data}}$ ).
- ② Dice the MC expectation according to uncertainties many times (pseudo experiments) to get a distribution of  $p$ -values.
- ③ Get the “relative number of pseudo experiments with  $p < p_{\text{data}}$ ” =  $\tilde{P}$ .

