

# Systematic study of the spectral shape dependence on neutral meson reconstruction in the ALICE EMCal

Alena Lösle

Supervisors: Dr. Jason Kamin

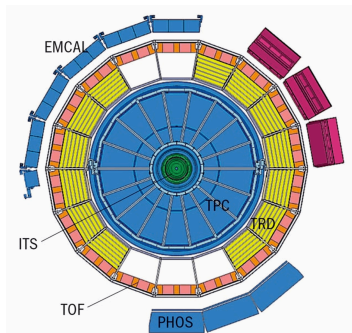
Dr. Constantin Loizides

Student Session

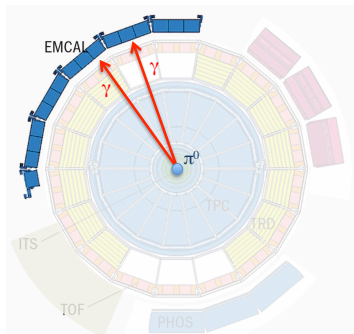


- $\pi^0$  reconstruction in ALICE EMCal
- our ToyMC analysis
- spectral shape dependence of reconstructed  $\pi^0$
- summary

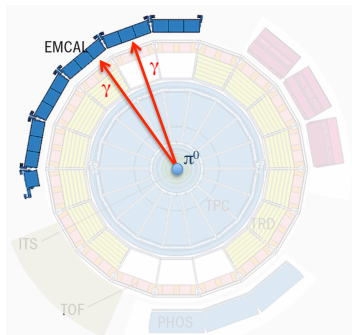
# $\pi^0$ reconstruction in ALICE EMcal



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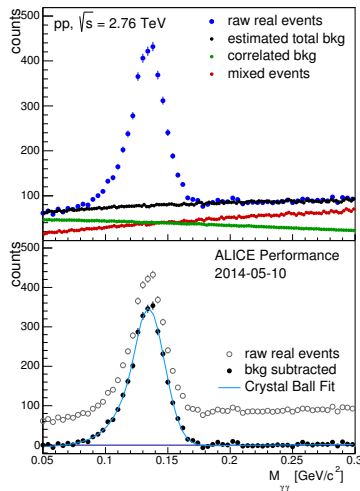
# $\pi^0$ reconstruction in ALICE EMcal



$\pi^0$  reconstruction:

- combine all  $\gamma\gamma$  pairs in EMCal
- estimate and subtract background
- fit mass peak

$p_T$  range: 1.8 - 2.0 GeV



# Our ToyMC analysis



$\pi^0 \rightarrow \gamma\gamma$  using TGenPhaseSpace

- **energy smearing:**                      **position smearing:**

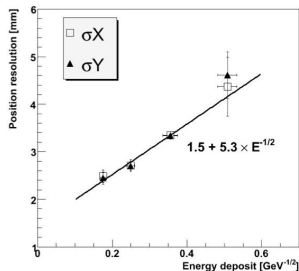
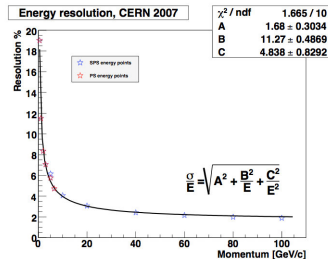
$$\frac{\sigma_E}{E} = A \oplus \frac{B}{\sqrt{E}} \oplus \frac{C}{E}$$

$$\sigma_P = a + \frac{b}{\sqrt{E}}$$

**A constant term:** detector geometry

**B sampling term:** counting statistics  $\propto$  signal

**C noise term:** pedestal due to electronics



<http://arxiv.org/abs/1008.0413>

# Our ToyMC analysis



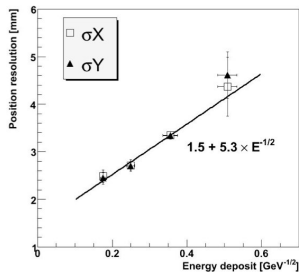
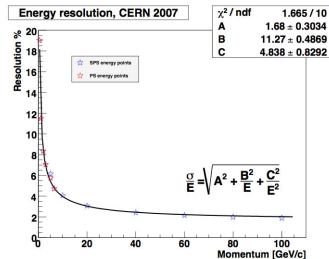
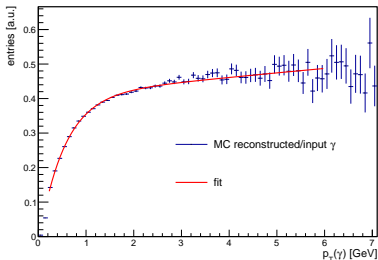
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- kinematic cut  $p_T^\gamma > 0.2$  GeV
- apply single photon efficiency



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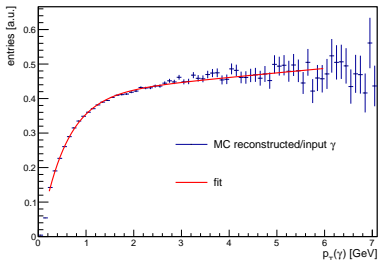
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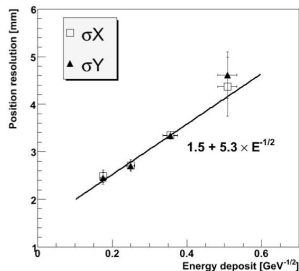
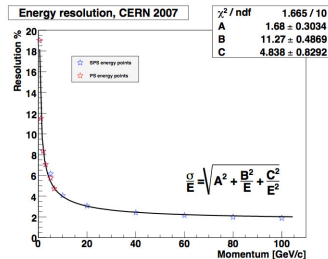
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- reconstruct  $\pi^0$  by adding  $\gamma\gamma$  pairs



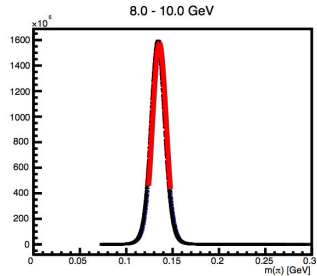
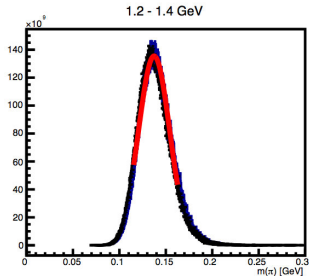
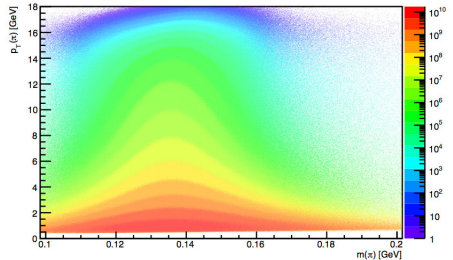
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# Mass distribution of reconstructed $\pi^0$



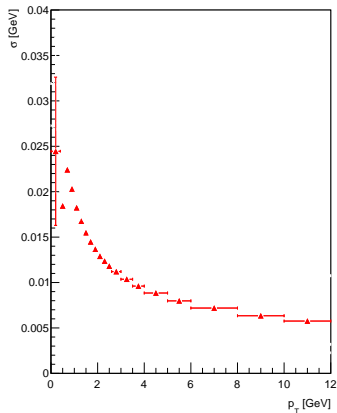
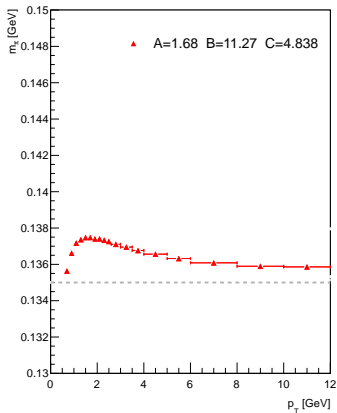
- mass distribution as function of pion  $p_T$   
→ project different  $p_T$  slices
- fitting of mass peaks for different  $p_T$  slices  
with gaussian
- get  $\mu$  and  $\sigma$  as function of pion  $p_T$



# Effect of energy smearing on reconstructed $\pi^0$



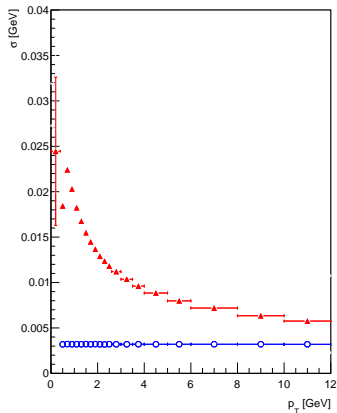
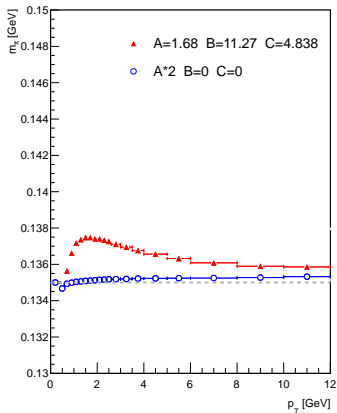
using EMCal parametrization for mass distribution:  $\frac{\sigma_E}{E} = A \oplus \frac{B}{\sqrt{E}} \oplus \frac{C}{E}$



# Effect of energy smearing on reconstructed $\pi^0$



using EMCal parametrization for mass distribution:  $\frac{\sigma_E}{E} = A \oplus \frac{B}{\sqrt{E}} \oplus \frac{C}{E}$   
constant term

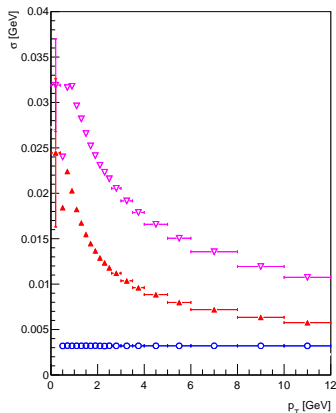
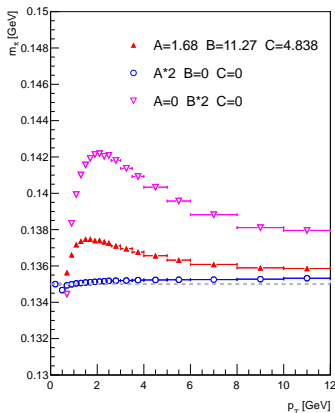


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using EMCal parametrization for mass distribution:  $\frac{\sigma_E}{E} = A \oplus \frac{B}{\sqrt{E}} \oplus \frac{C}{E}$

constant term      sampling term

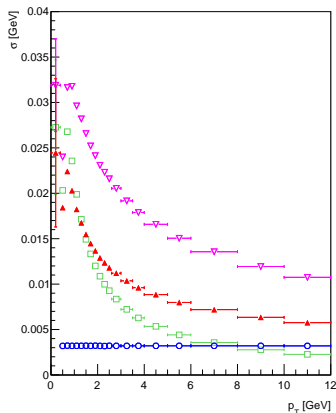
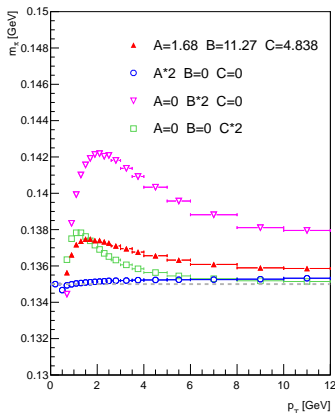


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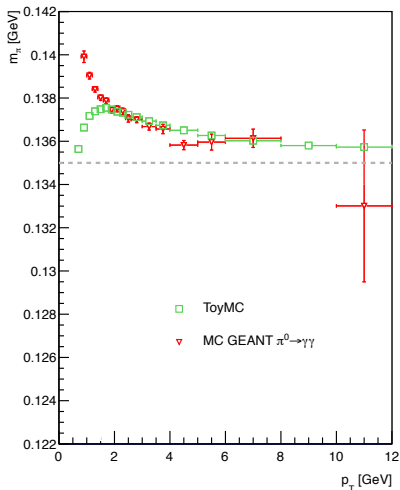
constant term      sampling term      noise term



# Compare fitted $\pi^0$ mass distribution



compare to GEANT simulation:

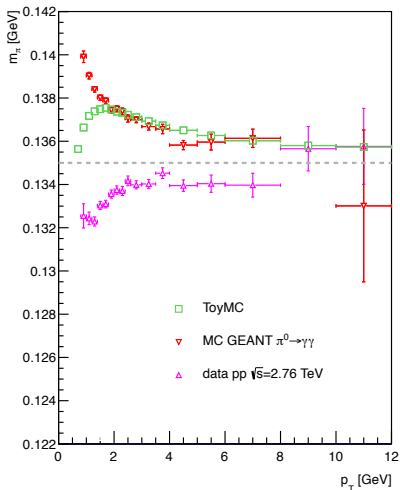


→ ToyMC and MC GEANT fit reasonably well

# Compare fitted $\pi^0$ mass distribution



compare to GEANT simulation and data:



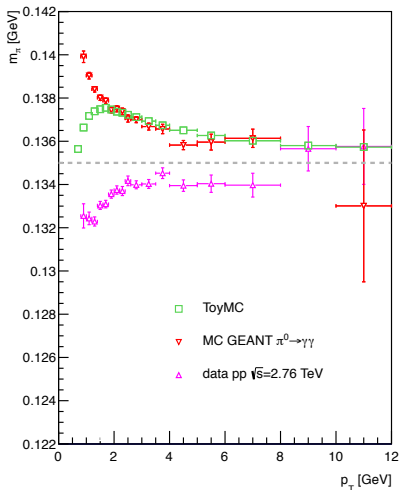
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→ different shapes for ToyMC and data

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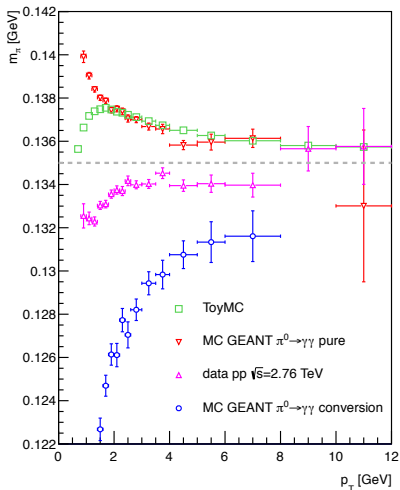
→ merged conversion photons!



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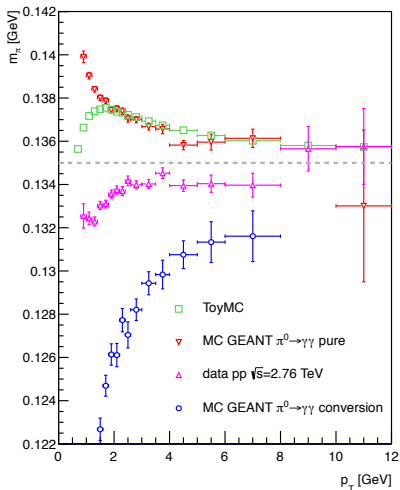
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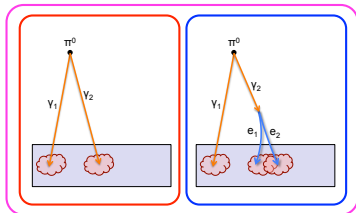
compare to GEANT simulation and data:



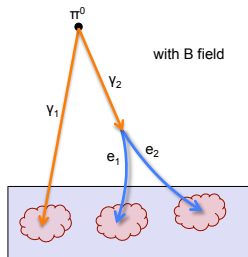
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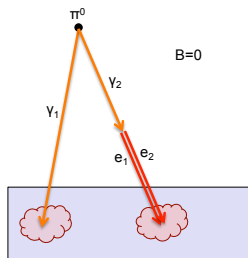
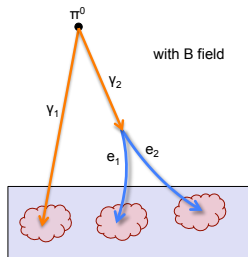
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# How to deal with non merging $e^+e^-$

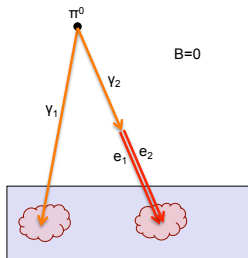
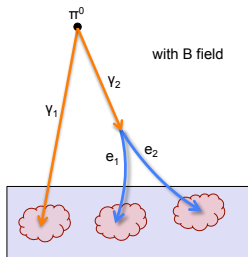


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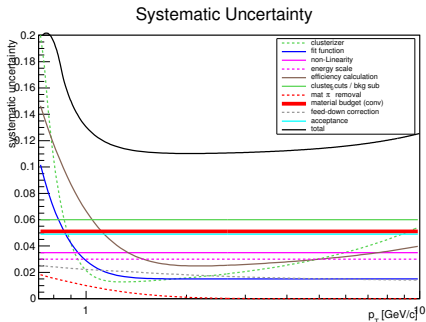


- conversion decreases reconstructed  $\pi^0$  yield
- without magnetic field: conversion  $e^+e^-$  merge in one EMCal cluster
- compare yield with and without magnetic field  
→ estimate material budget

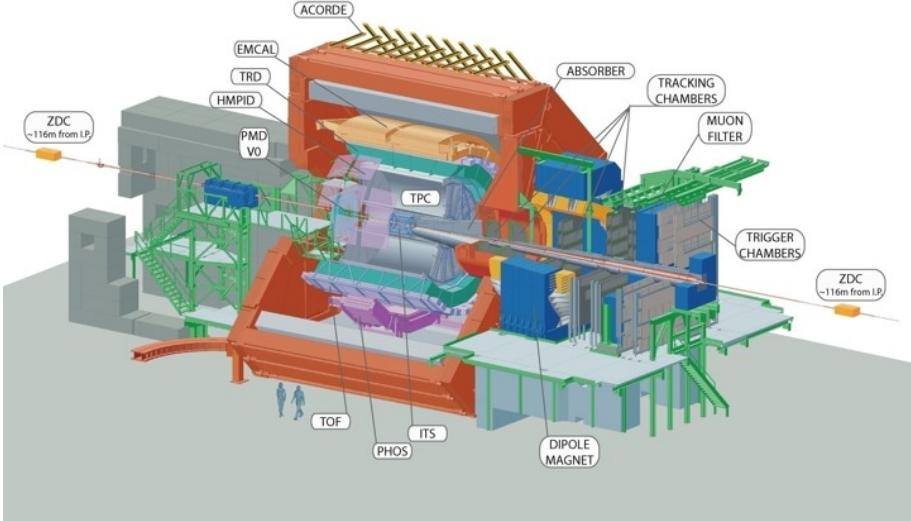
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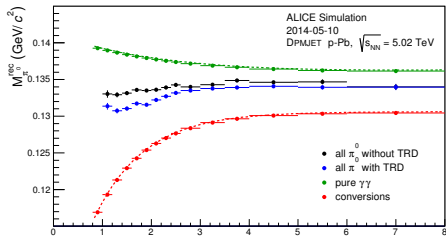
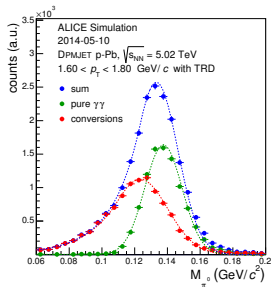
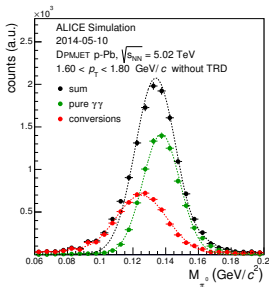
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- ToyMC looking at  $\pi^0 \rightarrow \gamma\gamma$  (no photon conversion)
- comparison of  $\pi^0$  mass position from ToyMC to data and GEANT
  - take conversion photons into account
- to deal with conversion  $e^+e^-$  that don't overlap in EMCal:
  - compare  $\pi^0$  yield with and without magnetic field
  - material budget estimation in front of EMCal



## Example: Impact of TRD on $\pi^0$ reconstruction





## Crystal Ball function:

Crystal Ball

```

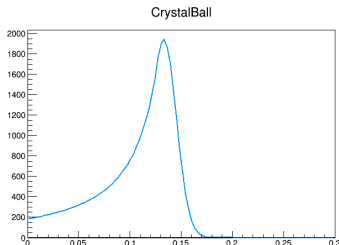
Double_t CrystalBall(Double_t *x, Double_t *par) {

  // The Crystal Ball shape is a Gaussian that is 'connected' to an exponential tail at
  // 'alpha' sigma of the Gaussian. The sign determines if it happens on the left or
  // right side. The 'n' parameter controls the slope of the exponential part.
  // typical par limits:  1.0 < alpha < 5.0   and   0.5 < n < 100.0

  Double_t alpha = par[0];
  Double_t n     = par[1];
  Double_t meanx = par[2];
  Double_t sigma = par[3];
  Double_t nn    = par[4];
  Double_t a = TMath::Power((n/TMath::Abs(alpha)), n) * TMath::Exp(-0.5*alpha*alpha);
  Double_t b = n/TMath::Abs(alpha) - TMath::Abs(alpha);
  Double_t arg = (x[0] - meanx)/sigma;
  Double_t fitval = 0;
  if (arg > -1.0*alpha) {
    fitval = nn * TMath::Exp(-0.5*arg*arg);
  } else {
    fitval = nn * a * TMath::Power((b-arg), (-1*n));
  }
  return fitval;
}

// here's just the lefthand part:
TF1 *f_cr = new TF1("f_cr",
  "[4]*TMath::Power((1/TMath::Abs([0])), [1])*
  TMath::Exp(-0.5*[0]*[0])*TMath::Power(((1/TMath::Abs([0]) -
  TMath::Abs([0]))-(x - [2])/[3]),(-1*[1]))", 0.01,0.13)

```



remaining issue: understanding the width distributionn

