

# Summary of the timebased EMC simulation and reconstruction chain in PandaRoot

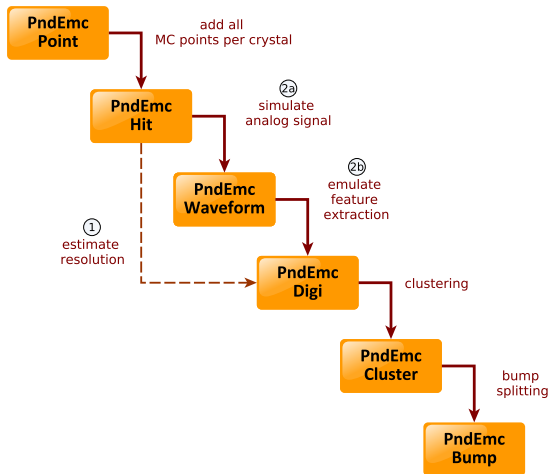
Philipp Mahlberg



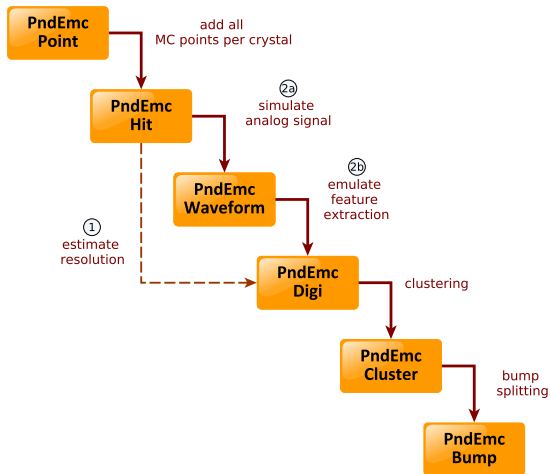
Helmholtz-Institut für Strahlen- und Kernphysik  
Universität Bonn

17.03.2015

# EMC simulation and reconstruction chain



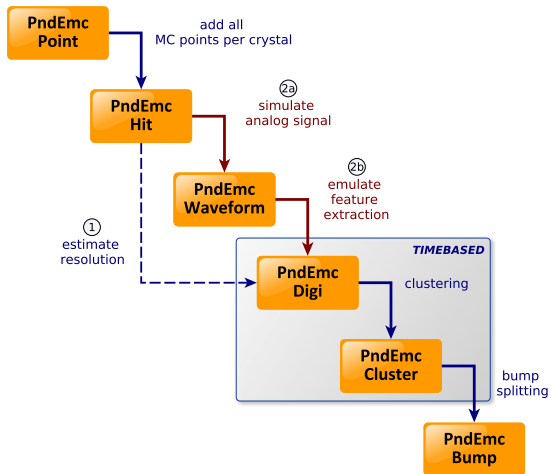
# EMC simulation and reconstruction chain



## two ways of generating EmcDigs:

- 1 directly out of corresponding EmcHit
- 2 impose intermediate step:  
waveform simulation  
→ “hardware-based” description

# EMC simulation and reconstruction chain

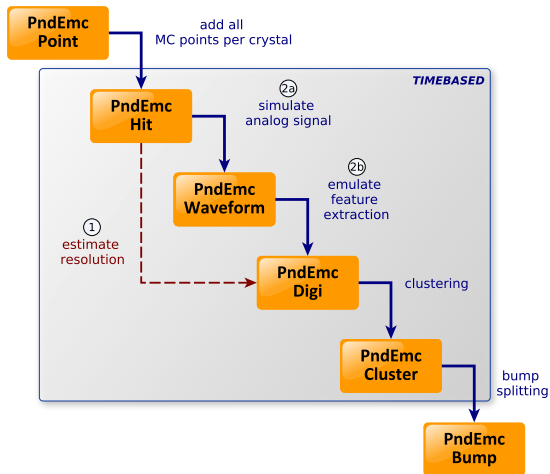


... and making things timebased

“classical” timebased approach:

- rearrange and sort digis in time
- use smeared (or even exact) information of EmcHits ①

# EMC simulation and reconstruction chain



... and making things timebased

include waveform simulation:

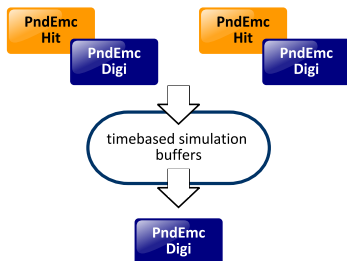
- simulate important parts of continuous waveform stream (2a)
- to be processed by free-running ADC (2b)

→ preserve realistic signal signature

# Timebased waveform simulation

Decouple hit grouping and waveform simulation:

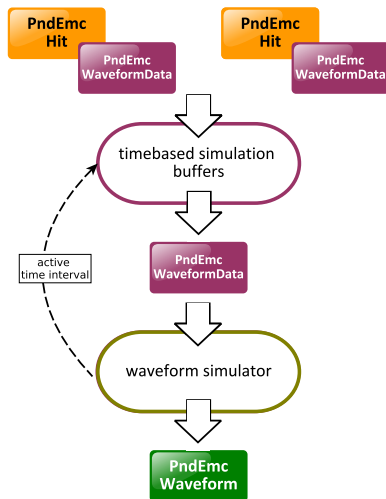
- PndEmcWaveformData stores hit(s) information (energy, time)
- closer to digi object than waveform
- PndEmcAbsWaveformSimulator simulates PndEmcWaveformData based on PndEmcWaveformData  
(1 data object yields 1 waveform, including  $n$  hits)
  - PndEmcAbsPSA scans single waveforms, producing PndEmcDigi objects  
( $n$  simulated hits,  $m$  recovered digis)
  - Call PndEmcDigiRingSorter ...



# Timebased waveform simulation

Decouple hit grouping and waveform simulation:

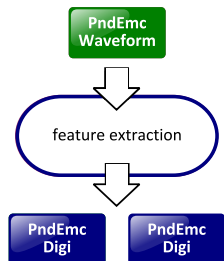
- PndEmcWaveformData stores hit(s) information (energy, time)  
→ closer to digi object than waveform
- PndEmcAbsWaveformSimulator simulates PndEmcWaveform based on PndEmcWaveformData  
(**1 data object yields 1 waveform, including  $n$  hits**)
- PndEmcAbsPSA scans single waveforms, producing PndEmcDigi objects  
( **$n$  simulated hits,  $m$  recovered digis**)
- Call PndEmcDigiRingSorter ...



# Timebased waveform simulation

Decouple hit grouping and waveform simulation:

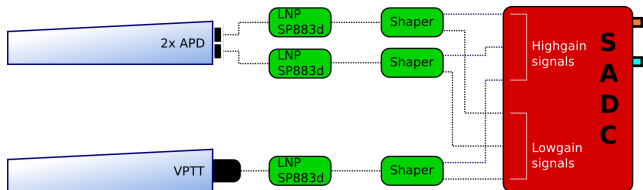
- PndEmcWaveformData stores hit(s) information (energy, time)
- closer to digi object than waveform
- PndEmcAbsWaveformSimulator simulates PndEmcWaveformData based on PndEmcWaveformData  
(*1 data object yields 1 waveform, including  $n$  hits*)
- PndEmcAbsPSA scans single waveforms, producing PndEmcDigi objects  
( *$n$  simulated hits,  $m$  recovered digis*)
- Call PndEmcDigiRingSorter ...





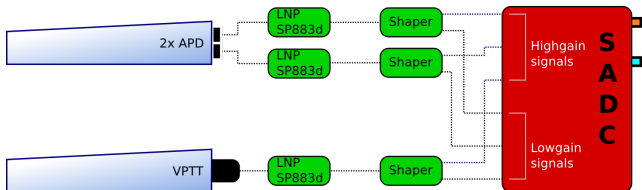
# Match experimental conditions

readout chain in the Forward Endcap:



# Match experimental conditions

readout chain in the Forward Endcap:



classes in PandaRoot:

WaveformSimulator



ShapingNoiseAdder



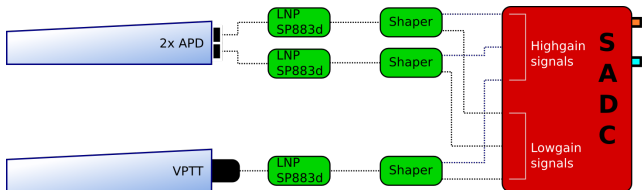
WaveformDigitizer



FeatureExtraction

# Match experimental conditions

readout chain in the Forward Endcap:



classes in PandaRoot:

WaveformSimulator



ShapingNoiseAdder

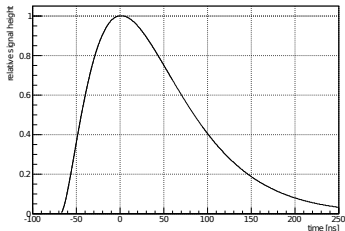


WaveformDigitizer



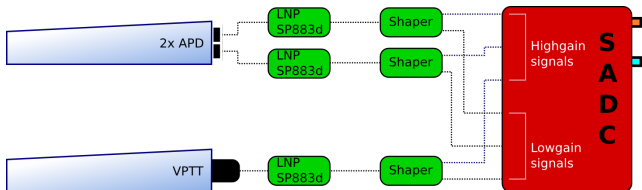
FeatureExtraction

fitted pulseshape:



# Match experimental conditions

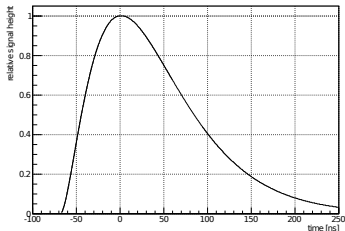
## readout chain in the Forward Endcap:



## classes in PandaRoot:

```
WaveformSimulator
  ↓
ShapingNoiseAdder
  ↓
WaveformDigitizer
  ↓
FeatureExtraction
```

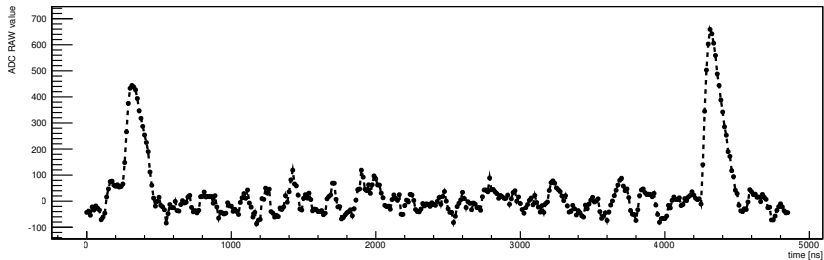
## fitted pulseshape:



## readout parameters for APD channel:

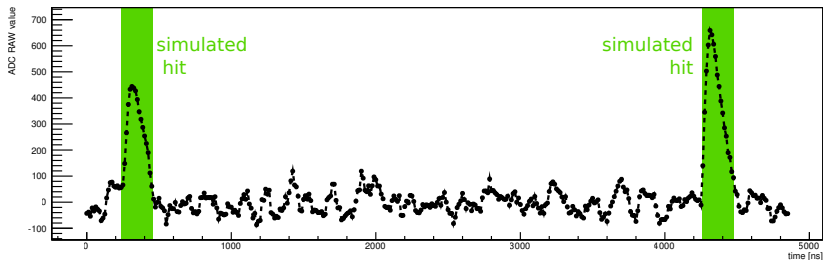
	high gain	low gain
multiplicity	2×	2×
sampling rate	80 MHz	80 MHz
channels	2 <sup>14</sup>	2 <sup>14</sup>
energy range	1 GeV	15 GeV
noise width	2.3 MeV	3.5 MeV

# Invoke feature extraction



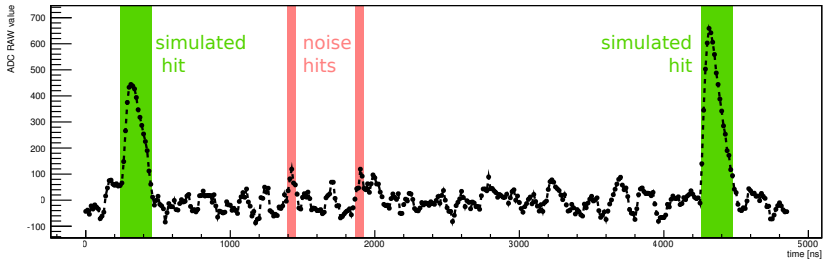
- resulting waveforms → close to hardware signal

# Invoke feature extraction



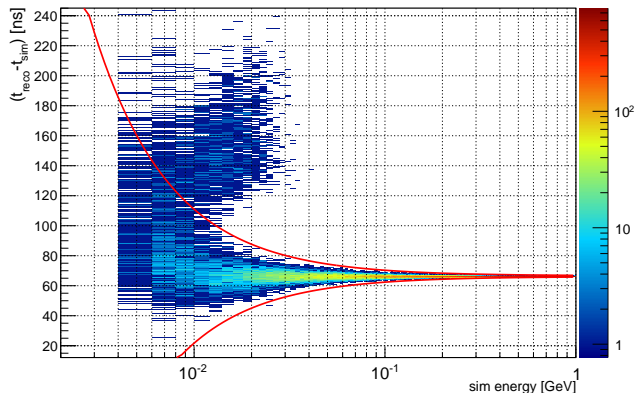
- resulting waveforms → close to hardware signal
- next step: process them by feature extraction
  - as intermediate step in simulation chain
  - to study its performance (testing, optimization)

# Invoke feature extraction



- resulting waveforms → close to hardware signal
- next step: process them by feature extraction
  - as intermediate step in simulation chain
  - to study its performance (testing, optimization)
- include relevant noise (false positive) hits in simulation: enlarge simulation window

# Timing of reconstructed hits



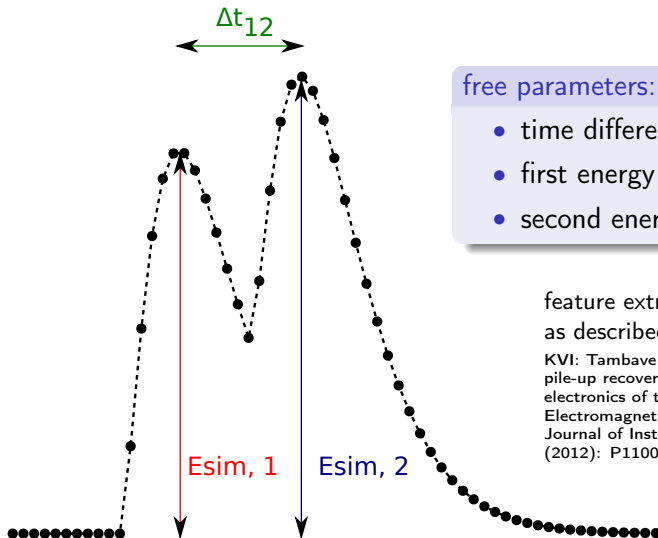
- $1 \times 10^6$  simulated pulses
- single hit with  $E_{sim} \sim \mathcal{U}(5 \text{ MeV}, 1 \text{ GeV})$
- only waveforms with 1 reconstructed hit selected

Triggerless concept demands proper hit timing:

- $\lesssim 20 \text{ MeV}$ : hit detection uncertain
- $\lesssim 10 \text{ MeV}$ : hit detection fails



# Testing pileup recovery capabilities



free parameters:

- time difference  $\Delta t_{12}$
- first energy  $E_{sim, 1}$
- second energy  $E_{sim, 1}$

feature extraction  
as described in:

KVI: Tambave, G., et al. "Pulse pile-up recovery for the front-end electronics of the PANDA Electromagnetic Calorimeter." *Journal of Instrumentation* 7.11 (2012): P11001.

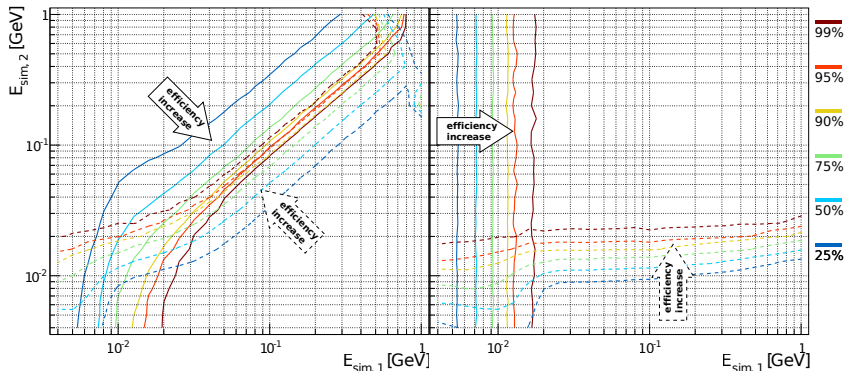
# Pulse detection efficiency

- define successful detection:  $\pm 15$  ns around expected time

— : recovery efficiency of 1. pulse      - - - : recovery efficiency of 2. pulse

$\Delta t_{12} = 75$  ns

$\Delta t_{12} = 150$  ns



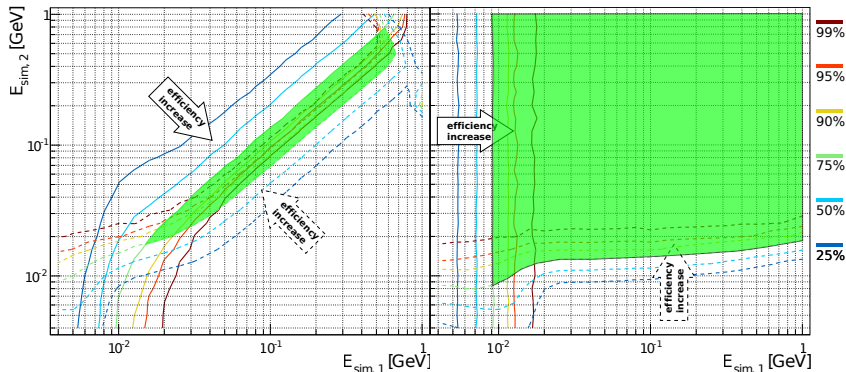
# Pulse detection efficiency

- define successful detection:  $\pm 15$  ns around expected time
- : recovery efficiency of 1. pulse      - - - : recovery efficiency of 2. pulse

$\Delta t_{12} = 75$  ns

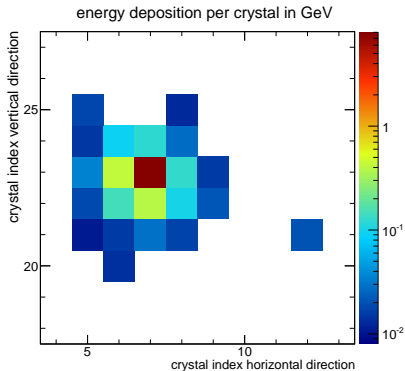
detection efficiency > 75 %

$\Delta t_{12} = 150$  ns



- $\Delta t_{12} > 100$  ns for almost always successful detection

# Eventbased clustering algorithm in PandaRoot ...

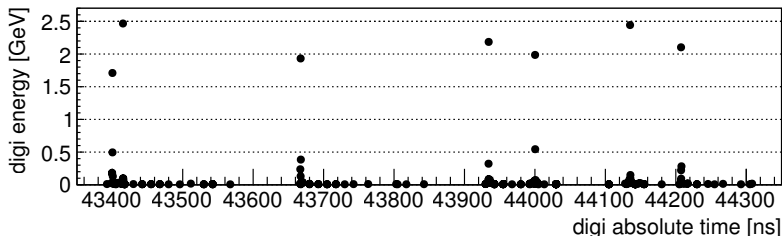


simulation of a 10 GeV photon cluster in the Forward Endcap with split-off

- 1 main clustering:
  - group all fired neighboring crystals into one cluster
  - ⇔ two distinct clusters are separated by inactive crystals
- 2 subsequent tasks:
  - detect local maxima within clusters
  - bump splitting

## ... and its timebased continuation

- 1 group incoming digi stream into *pseudo events*  
→ search for gaps  $\geq t_{\text{active}}$



- 2 perform spacial clustering on pseudo events
- 3 reimpose timing condition (*defined by*  $t_{\text{active}}$ )

⇒ overall clustering conditions:

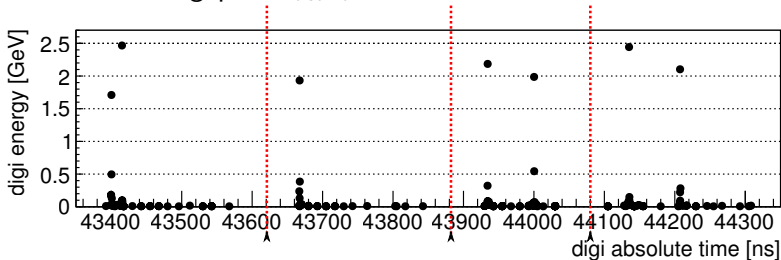
$$\left\{ \begin{array}{l} \text{space : } \text{direct neighbor} \\ \text{time : } \Delta t < t_{\text{active}} \end{array} \right. \quad \text{to any crystal already in cluster}$$

- still use of solely energy based bumpsplitting methods

## ... and its timebased continuation

- 1 group incoming digi stream into *pseudo events*

→ search for gaps  $\geq t_{\text{active}}$



- 2 perform spacial clustering on pseudo events

- 3 reimpose timing condition (*defined by*  $t_{\text{active}}$ )

⇒ overall clustering conditions:

$$\left\{ \begin{array}{l} \text{space : direct neighbor} \\ \text{time : } \Delta t < t_{\text{active}} \end{array} \right. \quad \text{to any crystal already in cluster}$$

- still use of solely energy based bumpsplitting methods

# The PndEmcClusterObject

```
class PndEmcCluster {  
    ...  
    std::vector<Int_t> fDigiList;  
    std::map<Int_t, Int_t> fMemberDigiMap;  
    std::map<Int_t, Int_t> fMcMap;  
    ...  
};
```

**digi's index in  
TClonesArray**

**digi's detectorID  
(crystal index)**

**MC trackID**

**count**

- 
- cluster stores reference to incorporated digis
  - objects rely on (eventbased) digi grouping
  - same for MC track information handling
  - detectorID serves as key
  - detectorID must be unique
  - ⇒ **conflicts with idea of timebased reconstruction**

## Timebased clustering in PandaRoot

the best future solution: ...

cleanest solution

Rewrite cluster class, make heavy use of FairLinks

→ *still reasonably performant?*



# Timebased clustering in PandaRoot

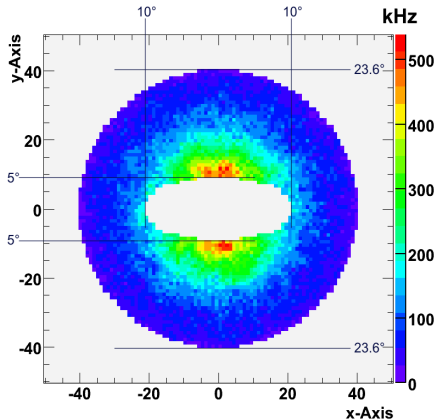
needed to do full physics simulations timebased:

## MC backpropagation

- available backpropagation broken in timebased runs
- chain of FairLinks considered to be continuous
- but PndEmcMatchTasks did not work either??

probably fixed with new version of FairLinks

# Results



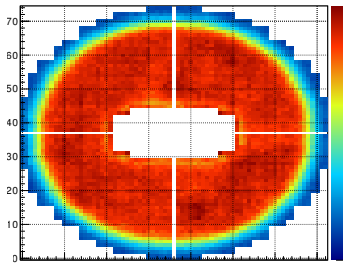
TDR:

expected hit rate on Forward  
Endcap

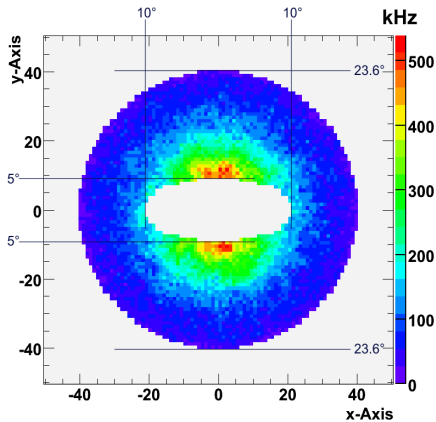
( $E_{\text{hit}} \geq 3 \text{ MeV}$ ) @  $p_{\bar{p}} = 14 \text{ GeV}$

simulation:

- photons with fixed energy  
(*no need for  
backpropagation*)
- time between events:  
Poisson-distributed,  
characterized by  $\tau$



# Results



TDR:

expected hit rate on Forward  
Endcap

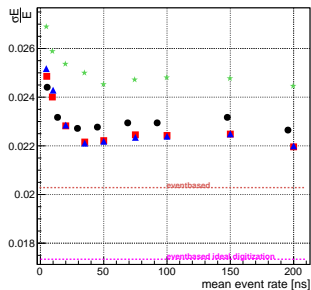
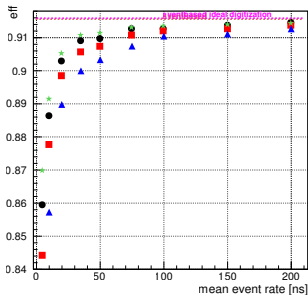
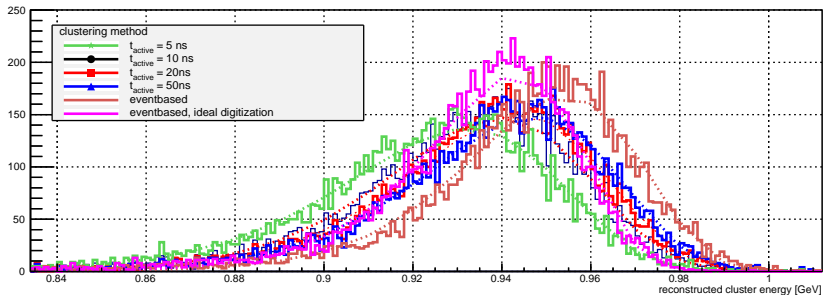
( $E_{\text{hit}} \geq 3 \text{ MeV}$ ) @  $p_{\bar{p}} = 14 \text{ GeV}$

simulation:

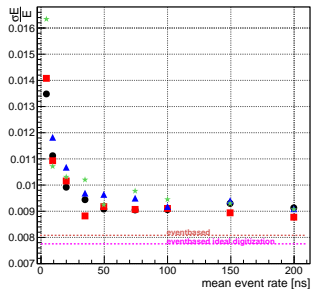
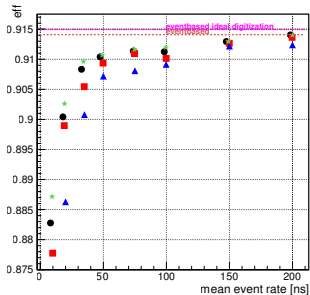
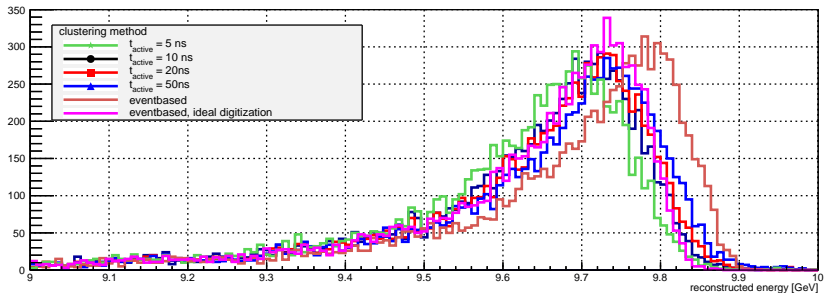
- photons with fixed energy  
(*no need for backpropagation*)
- time between events:  
Poisson-distributed,  
characterized by  $\tau$

$\tau$	$E_{\gamma}$ 1 GeV	$E_{\gamma}$ 10 GeV
hit rate 100 kHz	30 ns	100 ns
hit rate 500 kHz	6 ns	20 ns

# $\gamma$ with 1 GeV on Forward Endcap ( $N=10^4$ )



# $\gamma$ with 10 GeV on Forward Endcap ( $N=10^4$ )



# Timebased clustering: $\gamma$ on DPM background

scenario:

- DPM generator background  
 $p_{\bar{p}} = 14 \text{ GeV}, \tau = 50 \text{ ns}$
- mixed with single photons:  
ratio Signal/BG: 1/10

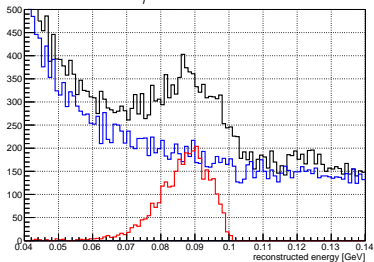
— event mixing

— only DPM background

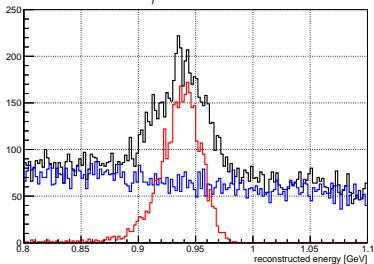
— only photon events

(eventbased, ideal digitization)

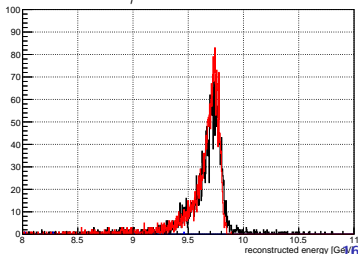
$E_{\gamma} = 0.1 \text{ GeV}$



$E_{\gamma} = 1 \text{ GeV}$



$E_{\gamma} = 10 \text{ GeV}$



# Conclusion

- EMC simulation and reconstruction chain has been adapted to allow for timebased simulations of:
  - `PndEmcDigis` via generation of timebased `PndEmcWaveforms` and usage of experimental feature extraction
  - `PndEmcClusters` by applying extended clustering methods: considering energy **and** time
- tasks to do:
  - review simulation parameters
  - conform and improve clustering stage towards timebased needs
  - make MC backpropagation possible

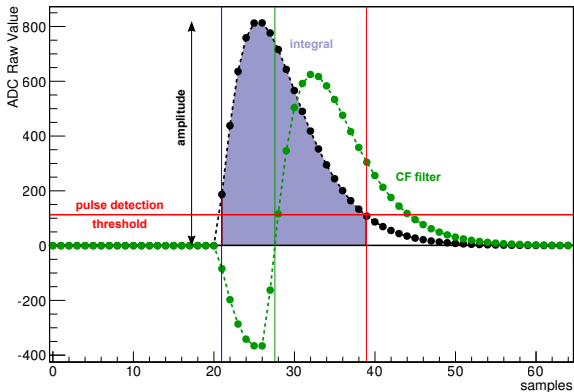
# Backup slides



# Feature extraction

- experiment: FPGA based feature extraction (designed by KVI, Groningen)

→ reproduce its behavior in simulation



pulse detection:

- rigid threshold

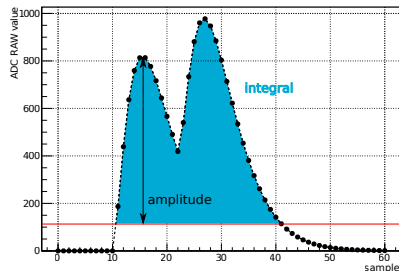
energy features:

- amplitude
- integral

timing features:

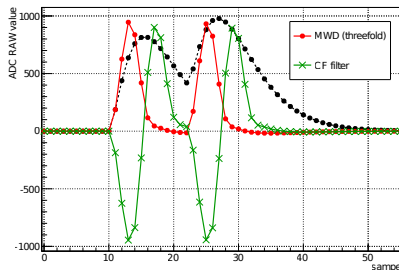
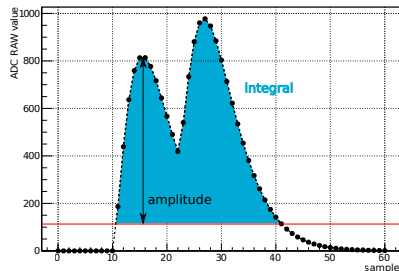
- zero crossing of CF-filter value

## Foreseen pileup treatment of the feature extraction:

Pileup detection & recovery of  $E_2$ :

- define ratio  $R = \frac{integral}{amplitude}$
- assumptions made:
  - amplitude not affected  $\Rightarrow E_1$
  - $R = const$  for single pulses
- pileup detected:
  - $R > R_{thres}$
- recover  $E_2$  as a function of:
  - amplitude, integral,  $R_{mean}$

# Foreseen pileup treatment of the feature extraction:



## Pileup detection & recovery of $E_2$ :

- define ratio  $R = \frac{integral}{amplitude}$
- assumptions made:
  - amplitude not affected  $\Rightarrow E_1$
  - $R = const$  for single pulses
- pileup detected:
  - $R > R_{thres}$
- recover  $E_2$  as a function of:
  - amplitude, integral,  $R_{mean}$

## Recovery of $T_1, T_2$ :

- shorten pulses by means of **threefold MWD filter**