

# Studying Annihilation Distributions

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## Objective

A procedure to disentangle the contribution of annihilations of antihydrogen on the trap walls and on the residual gas has been developed and tested. It is based on a Maximum Likelihood fit of the radial distribution of the annihilations that has been tuned using an “ad-hoc” Monte Carlo simulation toy.

- The model/template for annihilations on the trap walls has been extracted from the 2-second mixing window.
- The model/template for annihilations of the residual gas has been obtained using the losses when antihydrogen is hold in the trap (microwaves - UW - losses).
- A model/template for the cosmic background is also used.

The PDFs used for the fit procedure and the data generation the data are listed below:

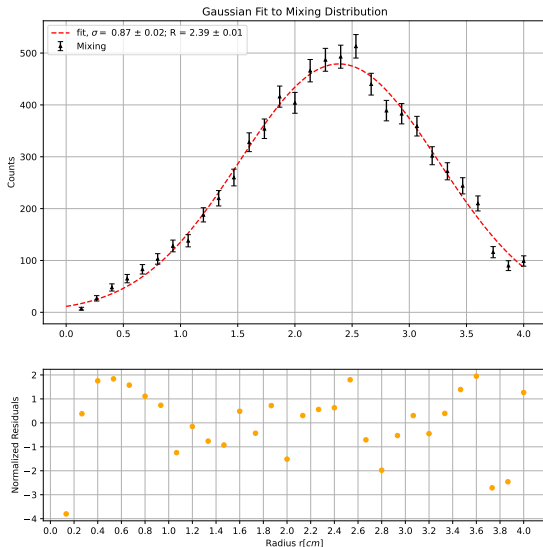
- PDF Mixing: the Normal distribution
- PDF Residual gas: Rayleigh distribution  $\frac{r}{\sigma^2} e^{-\frac{r^2}{2\sigma^2}}$
- PDF Cosmic:  $k \cdot x$

The factor  $k$  is the normalization constant. The Mixing, residual gas and cosmic data are fitted and the result are shown in the following slides.



# MIXING

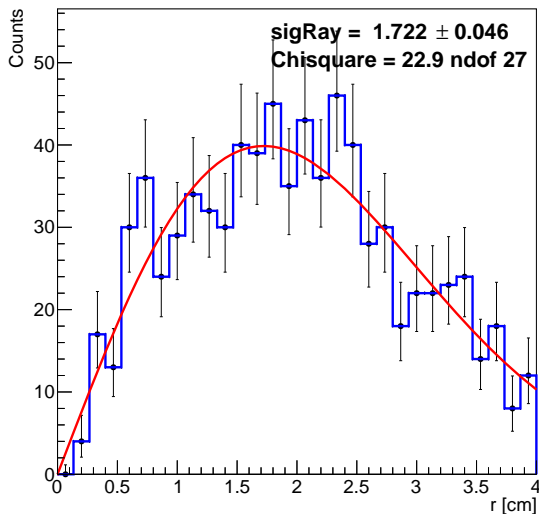
Mixing dataset represents almost pure data of anti-hydrogen annihilation on the walls. The radius distribution is fitted with a Gaussian.



## Residual Gas

The microwave losses represent an almost pure anti-hydrogen sample of annihilation events due to residual gas inside the trap.

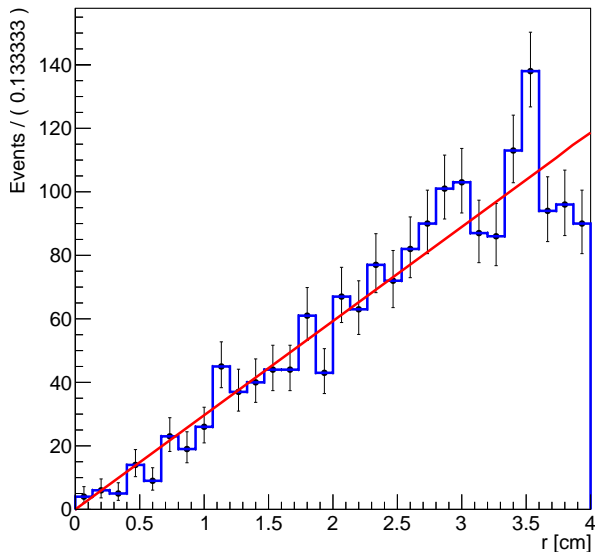
### UW losses PDF



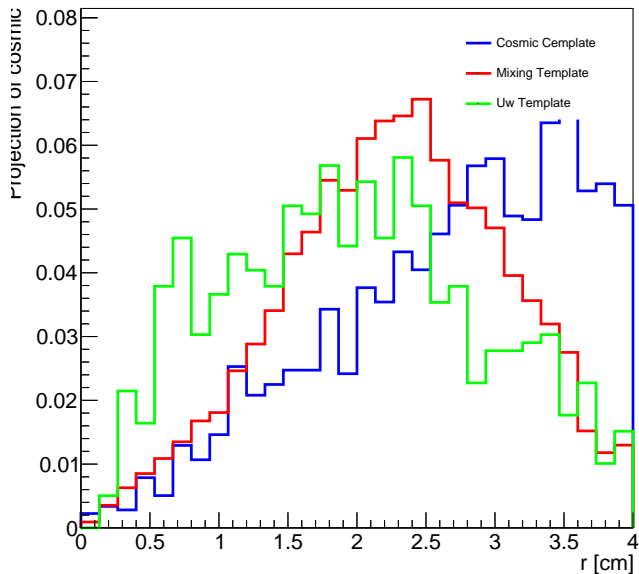
## Cosmic

The cosmic distribution is obtained from dataset without anti-hydrogen.

### Cosmic PDF



## All Pdf



## Radial Density

The histogram in  $r$  variable doesn't account for the different area of the bin which is  $2\pi r \cdot dr$ . So it is useful to divide per  $2\pi r$  to obtain the radial density of the events.

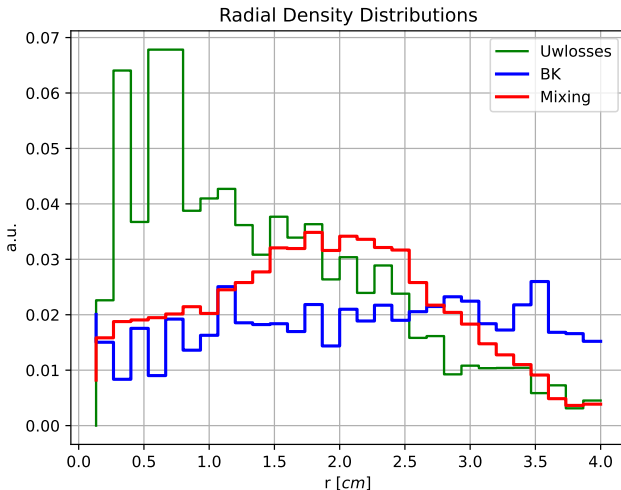


Figure: Radial density for r68465\_uw\_exp\_freq4.vertex.csv dataset

## Monte Carlo Simulation Toy

To study the accuracy of the algorithm to reconstruct the various parameter, we have developed a "toy" simulation tool. The model to generate the data is:

$$F_{gen}(r) = N_{sample} \cdot (a \cdot PDF_{mix} + b \cdot PDF_{gas} + c \cdot PDF_{cosmic}) \quad (1)$$

where  $a, b, c$  represent the "weights" of the various contributions to the PDF used to generate the data. The number of annihilation is indicated as  $N_{sample}$ . Once the data are generated, they are fitted with the model:

$$Nfit_{mix} \cdot PDF_{mix} + Nfit_{uw} \cdot PDF_{gas} + Nfit_{bk} \cdot PDF_{cosmic} \quad (2)$$

The parameters of the fit are  $Nfit_{mix}$ ,  $Nfit_{uw}$  and  $Nfit_{bk}$ . The "true value" are defined as:

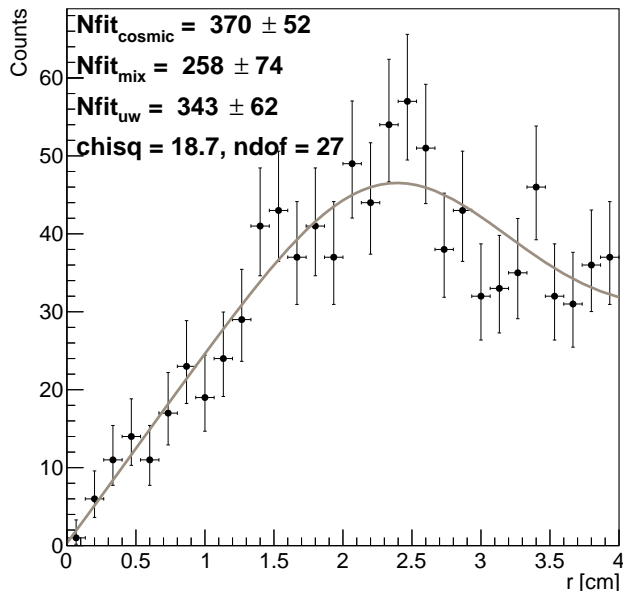
- $Ngen_{mix} = a \cdot N_{sample}$
- $Ngen_{gas} = b \cdot N_{sample}$
- $Ngen_{cosmic} = c \cdot N_{sample}$

In generation  $Ngen_{mix}, Ngen_{gas}, Ngen_{cosmic}$  are varied according to a Poissonian distribution.



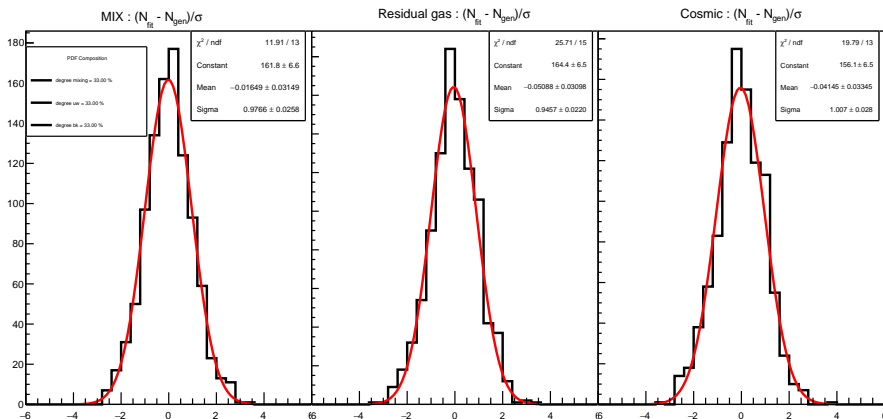
Example of fit, Toy:  $N_{\text{sample}} = 1000$ ,  $a = 33\%$ ,  $b = 33\%$ ,  $c = 33\%$

## Toy Model Fit



Toy:  $N_{\text{sample}} = 1000$ ,  $a = 33\%$ ,  $b = 33\%$ ,  $c = 33\%$

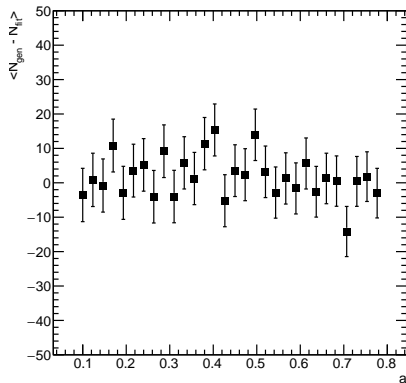
In this plot we have fixed the weight of each distribution to 33%, with  $N_{\text{sample}} = 1000$  and  $N_{\text{trials}} = 1000$ , to ensure that the algorithm is able to reconstruct the parameters, and check the presence of a bias. The variable of the histograms are:  $\frac{N_{\text{fit}} - N_{\text{gen}}}{\sigma_{\text{fit}}}$ . The distributions are normal and the fit procedure is behaving as expected.



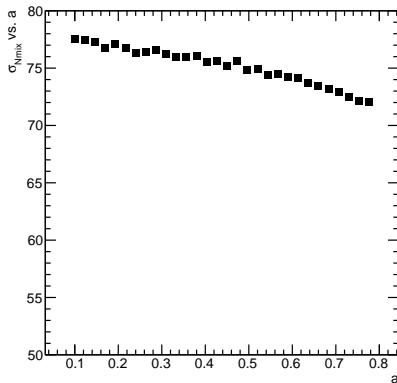
## weight variation $a$ for mix.

Now we study how the coefficients of the fit  $Nfit_{mix}$ ,  $Nfit_{uw}$  and  $Nfit_{bk}$  vary with the increment of the weight  $a$ . At fixed  $c = 10\%$ ,  $a$  is raised from 10% to 80% and  $b$  is decreased accordingly.

MIX:  $N_{gen} - N_{fit}$  averaged over 100 trials



MIX:  $\sigma$  vs weight



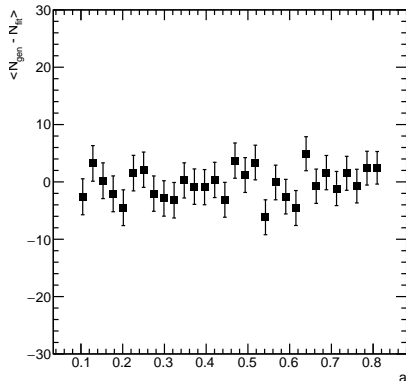
The number events is always  $N_{sample} = 1000$ . For each value of the weight  $a$  we iterate 100 times ( $N_{trials} = 100$ ) to study the reconstructed coefficients with the variation of the weights.



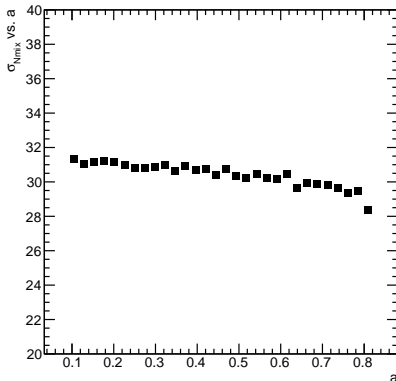
$$N_{\text{sample}} = 165.$$

Now we have done the same plot with  $N = 165$ , the same amount of data in `r68465_uw_exp_freq4.vertex.csv` after applying `cut1`. The value of  $c$  is fixed to reproduce the number of expected events from background ( $c = 6\%$ ).

MIX:  $N_{\text{gen}} - N_{\text{fit}}$  averaged over 100 trials



MIX:  $\sigma$  vs weight

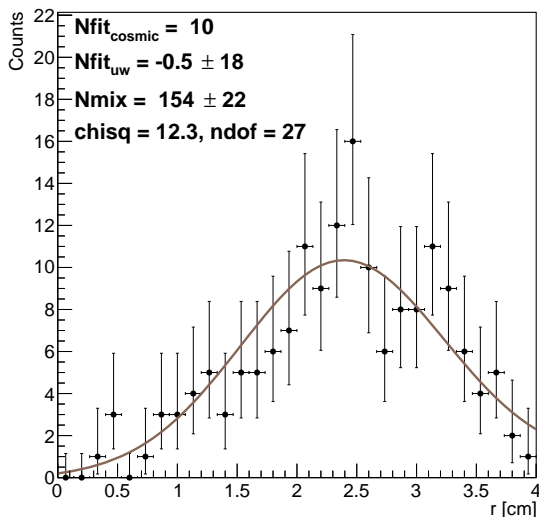


## Fit to Data

PDF = Gaussian (Mixing) + Rayleigh (Residual gas) + linear model (cosmic fixed).

Data taken from: r68465\_uw\_exp\_freq4.vertex.csv

r68465\_uw\_exp\_freq4.vertex

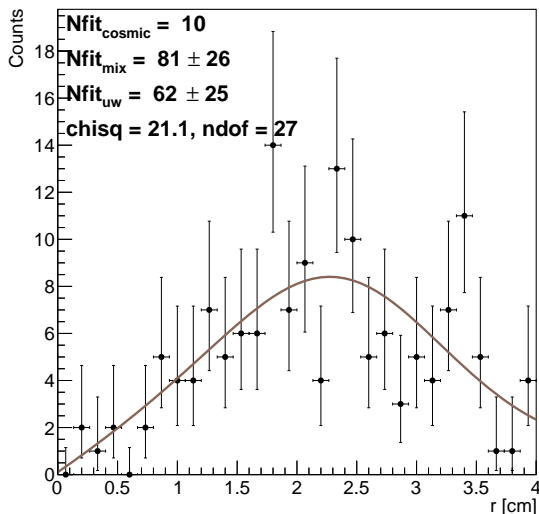


## Fit to Data

PDF = Gaussian (Mixing) + Rayleigh (Residual gas) + linear model (cosmic fixed).

Data taken from: r68465\_uw\_exp\_freq5.vertex.csv

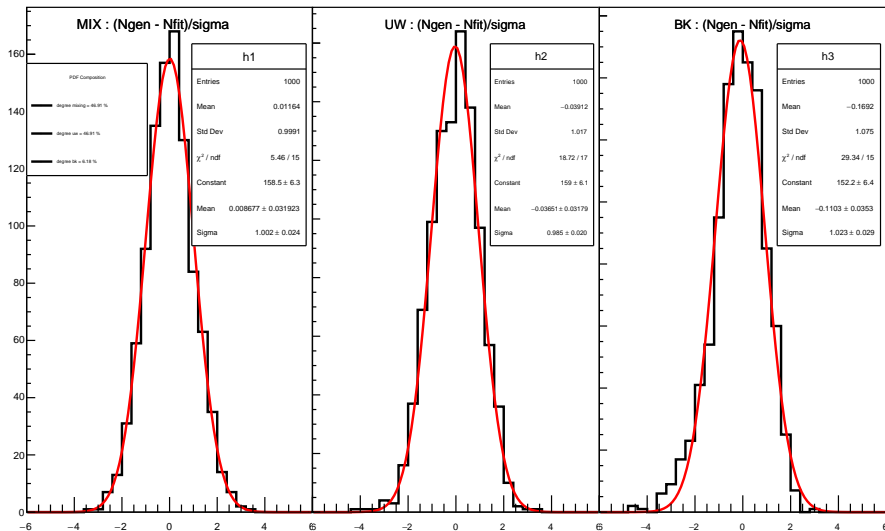
r68465\_uw\_exp\_freq5.vertex



## ADDITIONAL MATERIAL



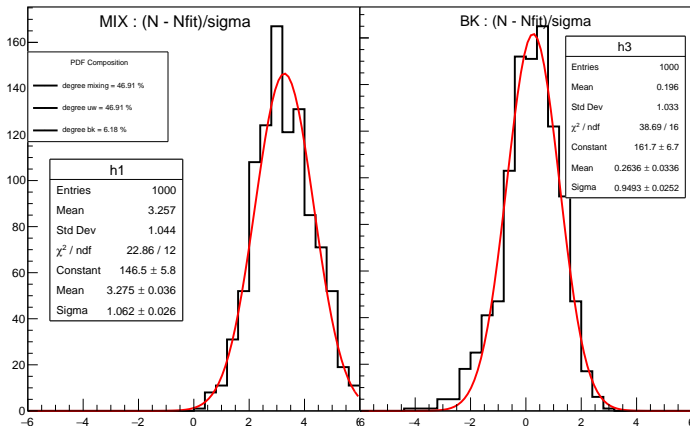
$N_{mix} - N_{fit}$  for  $a = 46\%$ ,  $b = 46\%$ ,  $c = 6\%$ .





## $N_{uw}$ parameter of the fit model fixed

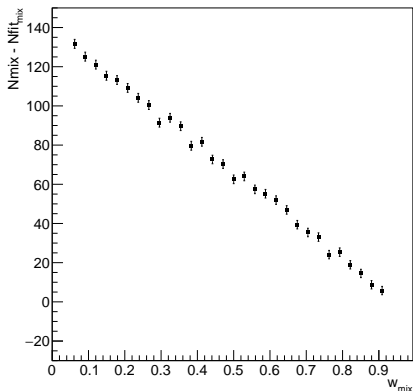
The Toy simulation is tested fixing the  $N_{uw}$  parameter of the fit model to 0. In the following plot the weight are  $a = 46\%$ ,  $b = 46\%$ ,  $c = 6\%$ , where  $c$  is fixed in such a way to reproduce the number of expected background events in dataset: r68465\_uw\_exp\_freq4.vertex.csv. Considering the 200 seconds of time length of r68465 we have fixed  $c$  to 6%, which corresponds to 10.2 events.



## $N_{uw}$ parameter of the fit model fixed

We study the bias  $N_{mix} - N_{reconstructed}$  with the parameter  $N_{uw}$  of the fit model fixed to 0. For small value of  $w_{mix}$ , corresponding to small contribution of *Mixing* (and, conversely, a significant contribution of *Residual gas* PDF) we observe a large bias.

$N_{mix}$  Generated - reconstructed averaged over 100 trials



Graph

