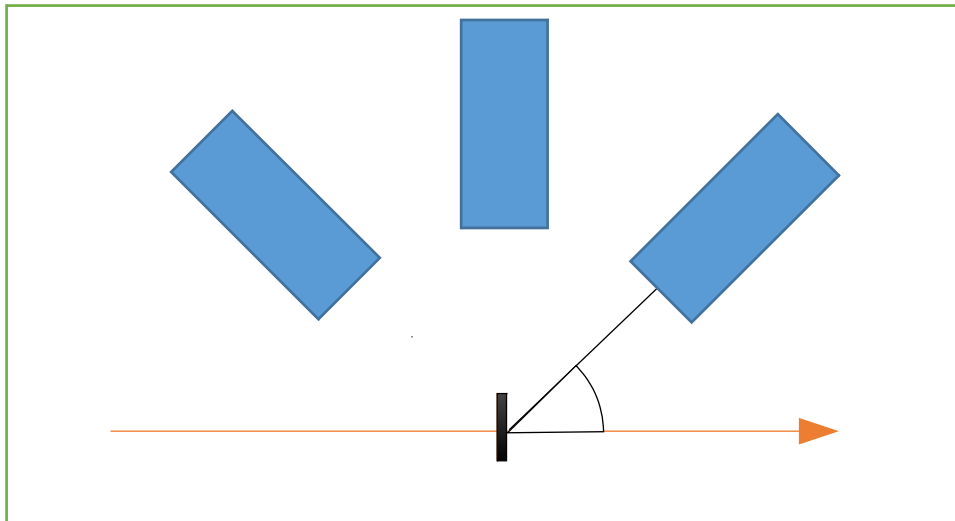


## Data Analysis Tutorials - Session 4

### The Experiment

Data are taken with an array consisting of 3 gamma-ray detectors positioned around a target position on an accelerator beam line. The 3 detectors are positioned, with respect to the target and beam, at angles of 45,90,135 degrees.



Data are stored in root “trees”, effectively long lists of individual events; an event being a single moment in time, in which the data acquisition computer detected an input and recorded all data from detectors during a short time window.

A series of .root data files are provided from the experiment:

Run1 – A source of  $^{60}\text{Co}$  at the target position

Run2 – A source of  $^{152}\text{Eu}$  at the target position

Run3 – A source of  $^{133}\text{Ba}$  at the target position

Run4 – An  $A=36$  beam with an energy of 250 MeV impinging on a thin  $A=40$  target.

Run5 – A decaying source feeding the nucleus observed in Run4

### Exercise 1 – Coincidences

As multiple hits may occur approximately simultaneously, such as in a cascade, it is advantageous to look not only at the singles spectrum (the 1D histogram of all gamma-rays) but also at coincident hits in multiple detectors.

Sort4.C includes a loop to produce a 2 dimensional histogram (sometimes referred to as a matrix) of coincidence events.

**Run Sort4.C on Run4.root and look at the resultant coincidence matrix.**

The matrix has been diagonalised, so  $M_{ij} = M_{ji}$  so that all coincidences across the matrix can be viewed without bias due to data ordering.

Now by projecting along one axis of this histogram at a given energy, you can see all the gamma-rays that were coincident with a gamma-ray of that energy.

Use the TH2 functions: SetShowProjectionX (from the right-click menu) to look at different “slices” of the matrix in the GUI by moving the mouse.

**A) Look at the coincidences of the largest peak. Which coincidence do you observe that is surprising?**

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Sort4.C also produces a histogram of the time difference between gamma-rays hits.

**B) Explain the features of the time spectrum?**

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From the timing spectrum determine a suitable timing coincidence requirement (a time gate) and input this using TgateLower and TgateUpper at the top Sort4.C

**C) Calculate the fraction of events selected by your requirement that are true coincidence events?**

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By making the reasonable assumption that false coincidences are randomly distributed with time, one can determine a time-random contribution that can be subtracted can subtract.

**Set your answer to question C as the TrueCoincidenceFraction at the top of Sort4.C and resort Run4.root.**

Now the background is subtracted use SetShowProjectionX again and compare the coincidences observed with what you observed before. Only the true coincidences should remain.

**D) What is the nucleus?**

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(Hint: If you are struggling with question D : Set gBeta=0, TrueCoincidenceFraction=1, and Sort Run5.root)

**EXTRA QUESTION ) Using your efficiency curve, explain the features at ~2000 keV in the singles sum spectra?**

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## Exercise 7 - Lifetimes

TimeExp.root contains a histograms sorted from Run5.root. It is the time difference between gamma-rays populating and depopulating the 1600 keV state.

By fitting the isomeric tail of the timing distribution determine the lifetime of the state.

***E) What is the lifetime of the 1600 keV state?***

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**Try fitting with the “L” binned likelihood fitting option rather than the chi-squared default. Do you notice a difference?** At very low counts the Pearson’s chi squared test becomes unreliable and binned maximum likelihood method is preferable for an accurate result.

**Modify Sort4.C to contain a new 1D histograms that will be filled with the time difference between any coincident 1600 keV and 439 keV gamma-rays.** (Set gBeta=0, TrueCoincidenceFraction=1 if you have not done so already)

**Sort Run5.root.** There are too few counts for a reliable exponential fit.

**Plot the same time spectra but in  $\log(t)$  (with histograms bins uniformly spaced in  $\log(t)$  instead of  $t$ ) a peak should form with the maximum at the state lifetime.** You can fit this with a Gaussian and see how it compares to the exponential method. This method (Often called the Schmidt method) is extremely useful in low statistics work where a reliable exponential fit is not viable.