

Exercise 6: Signal and Background

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Please send your solutions to nberger@physi.uni-heidelberg.de until 26. 11. 2012, 12:00. Put your answers (including plots, no (zip-) archives) in an email (subject line *SMIPP:Exercise06*). Test your programs before sending them off...

1. **Signal and background** Go back to exercise 5.2 and again simulate equal amounts of kaons (Landau distribution with mean 2 and "sigma" 0.4) and pions (1, 0.3) in a five layer energy loss particle identification detector. Form a truncated mean by throwing away the largest measurement each. Now make plots of the efficiency, purity, signal to background ratio, signal to $\sqrt{\text{background}}$ ratio for both pions and kaons as a function of the cut value. How does this change if only 10% of the particles are kaons? (Attach .C or .py files and suitable plots)
2. **Unbinned Likelihood - Searching for a Resonance** Assume we are studying an invariant mass distribution in some decay channel (e.g. the $\pi^+\pi^-$ mass in $J/\psi \rightarrow \gamma\pi^+\pi^-$ or the $t\bar{t}$ mass in $pp \rightarrow t\bar{t}X$ or whatever your favourite physics example is. Assume this mass distribution is dominated by a single resonance, which we can describe by a Breit-Wigner distribution (`TRandom3::BreitWigner()`) of mass 2 (GeV/ c^2 for the pions, TeV/ c^2 for the tops) and width 0.1 (same units). Generate 100 events (masses m_i) according to that distribution. Now construct the log likelihood function

$$\log L = \log \left(\prod_{i=1}^N BW(m_i, M, \Gamma) \right) = \sum_{i=1}^N \log (BW(m_i, M, \Gamma)), \quad (1)$$

where you can use `TMath::BreitWigner()` for the Breit-Wigner function. Assume that you know the width and scan the assumed resonance mass M and plot $\log L$ as a function of the mass. The maximum of that curve is your measured value, 1σ errors are where your likelihood changes by 0.5 on either side. What is your result? (Attach .C or .py files and suitable plots)

3. **Unbinned Likelihood - Searching for a Resonance II** Take the example from above, but now add in 2000 equidistributed background events in the mass range from 1 to 3 (units again up to you). Throw away Breit-Wigner events with masses outside that range. Now construct a background-only likelihood function (flat) and a signal + background likelihood function, where you leave the relative normalisation free. Assuming you know mass and width of the resonance, now perform a scan as above in that normalisation parameter. Then compare the $\log L$ for

the background only hypothesis and the (optimal) signal + background hypothesis.

(Attach .C or .py files and suitable plots)

4. **Unbinned Likelihood - Searching for a Resonance III** Again take the example from above and perform a toy MC test of your findings. Perform the experiment 1000 times, generating only background; how often do you find a likelihood ratio ($\log L$ difference) larger than in 3? Do the same whilst generating the resonance and the background, how often do you find a likelihood ratio smaller than in 3? If you had not generated it yourself, how sure would you be that there is a resonance?
(Attach .C or .py files and suitable plots)