

Computer Exercise: Toy-experiments of fits of a constant

Macro `p0toyf.C`, accessible at

[http : //www.desy.de/~obehnke/stat/school_apr14/p0toyf.C](http://www.desy.de/~obehnke/stat/school_apr14/p0toyf.C)

generates toy experiments for the fit of a constant (“p0-fit”) from several measurements of the same resolution. A physics example for this would be the determination of the (average) vertical position of a horizontal flying track in track-detectors. Additional Problem: the detectors are noisy, for each detector sometimes instead of a good signal hit a broadly distributed noise hit is observed.

- Steering parameters in the macro:

- *Ntra* = Number of repeated toy experiments (Default is 1000)
- *Ndet* = Number of measurements (Default is 10)
- *frac_noise* Average fraction of noise hits (Default is 0.)

- Output

- Histo *p0d* - Distribution of residual: fitted constant - true value
- Histo *chi2d* - χ^2 distribution from the *p0*-fits
- Histo *pchi2d* - χ^2 -probability distribution from the *p0*-fits
- Histo *a* - shows the *p0*-fit for the last toy experiment

Tasks:

- Take a deep breath and have a few minutes look in the code and try to understand what's going on
- a) Run the macro as it is with `.x p0toyf.C` and plot the histograms by `p0d → Draw(); chi2d → Draw(); pchi2d → Draw(); a → Draw();` Fill the RMS value of the *p0d* histogram in the table below (first row) and also the mean values of the *chi2d* and *pchi2d* distributions.
- b) Now edit the macro and set `frac_noise=0.1;`
 - run the macro again and fill the obtained values in the table (second row).
 - How much has the RMS of the residuals increased?
 - Can you identify the bad track-fits in the *pchi2d* distribution?
- c) Rejection of tracks with bad χ^2 -probability: Book another residual histogram *p0d_rej* in the macro:
 - fill it only for the case that the χ^2 -probability is not too small (see also hints in the code)
 - fill in the third row of the table how many tracks survive this selection and the resulting RMS of the residuals.
- d) Hit-outlier-rejection: Advanced method for super experts!
Book another three histograms *p0d_iter*, *chi2d_iter* and *pchi2d_iter* in the macro and
 - try hit outlier rejection and repeating the track-fit (see detailed instructions in the code);

- fill the resulting RMS of *p0d_iter* into the fourth row of the table.
- How does the RMS results of this method compare to the other cases a)-c)?
- Repeat the above exercises b)-d) for increased $frac_noise=0.5$ and fill the results in the table. How much are the RMS results worse in this case?
- If time permits repeat the above exercises a)-d) for different number of detectors $ndet = 5$ and study how the results change qualitatively.

| Task | N_{det} | $frac_noise$ | Outlier rejection | Resid.-distr. | #tracks | Resid. RMS | χ^2_{2d} mean | $p\chi^2_{2d}$ mean |
|------|-----------|---------------|--|----------------|---------|------------|--------------------|---------------------|
| a) | 10 | 0.0 | No | p_{0d} | 1000 | 0.0032 | 8.8 | 0.51 |
| b) | 10 | 0.1 | No | p_{0d} | 1000 | 0.0064 | 15.9 | 0.245 |
| c) | 10 | 0.1 | Reject tracks with bad $p\chi^2_{2d}$ | p_{0d_rej} | 596 | 0.0036 | — | — |
| d) | 10 | 0.1 | Outlier hit-rejection and repeated track-fit | p_{0d_iter} | 1000 | 0.00396 | — | — |
| b) | 10 | 0.5 | No | p_{0d} | 1000 | 0.0133 | 34 | 0.0053 |
| c) | 10 | 0.5 | Reject tracks with bad $p\chi^2_{2d}$ | p_{0d_rej} | 23 | 0.004 | — | — |
| d) | 10 | 0.5 | Outlier hit-rejection and repeated track-fit | p_{0d_iter} | 1000 | 0.0099 | — | — |