

# Statistics and Data Analysis

## Evaluation Exercise for Unfolding Part

Simulate and solve an unfolding problem using the slides shown during the lectures (available at [Unfolding.pdf](#)) and the following steps:

1. Generate the input elements for the unfolding problem and compute non-regularized solutions:
  - (a) Consider equal range and number of bins ( $M = N$ ) for the true ( $\boldsymbol{\mu}$ ) and measured distributions ( $\boldsymbol{\nu}$ ), respectively. Construct the (square) migration matrix  $R$  by assuming a (e.g.) Gaussian estimator with no bias and a resolution of 2 times the width of the bins of  $\boldsymbol{\mu}$ .
  - (b) Construct the distribution  $\boldsymbol{\mu}$  using a true p.d.f. ( $f_{\text{true}}$ ) given by (e.g.) the sum of two Gaussian functions.
  - (c) Construct the distribution  $\boldsymbol{\nu}$  by multiplying the migration  $R$  matrix to the vector  $\boldsymbol{\mu}$ .
  - (d) Generate randomly the distribution  $\boldsymbol{n}$  by assuming uncorrelated Poisson distributed measurements and no background.
  - (e) Compute the exact solution  $\boldsymbol{\mu}_0$  using the inverse of the migration matrix.
  - (f) Compute (numerically) the maximum likelihood or minimum least-squares solution and compare it to the exact solution.
  - (g) Compute a solution using the correction factors method.
  - (h) Compute the solution for  $f_{\text{true}}$  using the forward unfolding method (e.g. for the double-Gaussian case, the means, variances and normalizations are free parameters).
2. Using the elements obtained in Exercise 1, compute a regularized solution:
  - (a) Consider the migration matrix  $R$  and the distribution of measured values  $\boldsymbol{n}$  produced in Exercise 1 (as in a real experiment, we assume we do not know  $\boldsymbol{\mu}$  or  $\boldsymbol{\nu}$ , but we are going to estimate them).
  - (b) Chose a regularization method (Tykhonov, maximum entropy,...).
  - (c) Maximize Equation 7 of the slides systematically for a set of values of  $\alpha$  (in practice we maximize Equation 5 keeping one  $\boldsymbol{\mu}$  component dependent to the rest to keep condition  $\nu_{\text{tot}}(\boldsymbol{\mu}) = n_{\text{tot}}$ ).
  - (d) For each considered  $\alpha$  value, compute the values of the four quantities used to chose the optimal  $\alpha$  (MSE, MSE',  $\chi_{\text{eff}}^2$  and  $\chi_b^2$ ). Plot them as a function of  $\Delta \log L$ .
  - (e) For each optimal  $\alpha$  value, plot the estimated true distribution  $\hat{\boldsymbol{\mu}}$  (with errors), together with the true distribution  $\boldsymbol{\mu}$  and the measured distribution  $\boldsymbol{n}$ .
  - (f) For each optimal  $\alpha$  value, plot the estimated biases for the  $\hat{\boldsymbol{\mu}}$  solution (with errors).