

# Statistical Methods for Data Analysis

# Parameter estimates with RooFit

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#### Fits with RooFit



- Get data sample (or generate it, for Toy Monte Carlo)
- Specify data model (PDF's)
- Fit specified model to data set with preferred technique (ML, Extended ML, ...)

### Example



```
RooRealVar x("x", "x", -10,10);
RooRealVar mean("mean", "mean of gaussian", 0, -10, 10);
RooRealVar sigma("sigma", "width of gaussian", 3);
   RooGaussian gauss("gauss", "gaussian PDF", x, mean, sigma);
RooDataSet* data = gauss.generate(x,10000);
                                                         Further drawing options:
// ML fit is the default
                                                         pdf.paramOn(xframe,data);
gauss.fitTo(*data);
                                   A RooPlot of "x"
                                                         data.statOn(xframe);
                                 % 300 <u></u>
mean.Print();
                                 Events / ( )
// RooRealVar::mean =
// 0.0172335 +/- 0.0299542
sigma.Print();
                                  200
// RooRealVar::sigma =
// 2.98094 +/- 0.0217306
                                  150
                                                      PDF
RooPlot* xframe = x.frame();
                                  100
                                                  automatically
data->plotOn(xframe);
                                                  normalized
                                   50
gauss.plotOn(xframe);
                                                   to dataset
xframe->Draw();
```

#### Extended ML fits



 Specify extended ML fit adding one extra parameter:

```
pdf.fitTo(*data, RooFit::Extended
  (kTRUE));
```

#### Import external data sets



Read a ROOT tree:

```
RooRealVar x("x","x",-10,10);
RooRealVar c("c","c",0,30);
RooDataSet data("data","data",inputTree,
RooArgSet(x,c));
```

- Automatic removal of entries out of variable range
- Read an ASCII file:

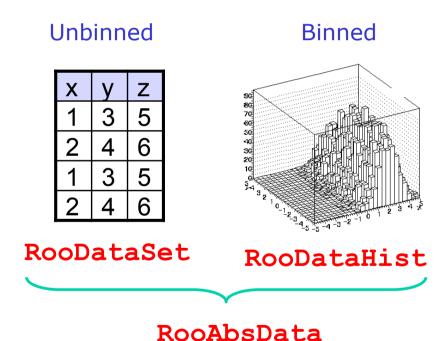
```
RooDataSet* data =
RooDataSet::read("ascii.file",
RooArgList(x,c));
```

Une line per entry; variable order given by argument list

### Histogram fits



- Use a binned data set:
  - RooDataHist instead of RooDataSet
- Fit with binned model



Statistical Methods for Data Analysis

#### Import external histograms



From ROOT TH1/TH2/TH3:

```
RooDataHist bdata1("bdata","bdata",RooArgList(x),histo1d);
RooDataHist bdata2("bdata","bdata",RooArgList
    (x,y),histo2d);
RooDataHist bdata3("bdata","bdata",RooArgList
    (x,y,z),histo3d);
```

• Binning an unbinned data set:

```
RooDataHist* binnedData = data->binnedClone();
```

Specifying binning:

```
x.setBins(50);
RooDataHist binnedData("binnedData", "data", RooArgList
(x), *data);
```

#### Discrete variables



- Define categories
- E.g.: b-tag:

```
RooCategory b0flav("b0flav", "B0 flavour");
b0flav.defineType("B0", -1);
b0flav.defineType("B0bar", 1);
Indices automatically assigned if omitted
```

- Several tools defined to combine categories (RooSuperCategory) and analyze data according to categories
  - See Root user manual for more details...
- Switch between PDF's based on a category can be implemented for simultaneous fits of multiple categories:

```
RooSimultaneous simPdf("simPdf","simPdf", categoryType);
simPdf.addPdf(pdfA,"A");
simPdf.addPdf(pdfB,"B");
```

## **Explicit Minuit minimization**



Build negative log-Likelihood finction (NLL)

```
// Construct function object representing -log(L)
RooNLLVar nll("nll", "nll", pdf, data);

// Minimize nll w.r.t its parameters
RooMinuit m(nll);
m.migrad();
m.hesse();
```

• Extra arguments: specify extended likelihood:

```
RooNLLVar nll("nll","nll",pdf,data,Extended());
```

Chi-squared functions (only accepts RooDataHist):

```
RooNLLVar chi2("chi2","chi2",pdf,data);
```





```
// Start Minuit session on above nll
RooMinuit m(nll);
// MIGRAD likelihood minimization
m.migrad();
// Run HESSE error analysis
m.hesse();
// Set sx to 3, keep fixed in fit
sx.setVal(3);
sx.setConstant(kTRUE);
// MIGRAD likelihood minimization
m.migrad();
// Run MINOS error analysis
m.minos();
// Draw 1,2,3 'sigma' contours in sx,sy
m.contour(sx, sy);
```

#### Minuit function MIGRAD



**Progress information,** 

Purpose: find minimum

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```
watch for errors here
******
                       1000
     13 **MIGRAD
*****
(some output omitted)
MIGRAD MINIMIZATION HAS CONVERGED.
MIGRAD WILL VERIFY CONVERGENCE AND ERROR MATRIX
COVARIANCE MATRIX CALCULATED SUCCESSFULLY
FCN=257.304 FROM MIGRAD STATUS=CONVERGED
                                                31 CALLS
                                                                  32 TOTAL
                   EDM=2.36773e-06
                                                       ERROR MATRIX ACCURATE
                                      STRATEGY= 1
 EXT PARAMETER
                                                STEP
                                                             FIRST
                                 ERROR
 NO.
      NAME
                VALUE
                                                SIZE
                                                         DERIVATIVE
                  8.84225e-02 3.23862e-01
                                             3.58344e-04 -2.24755e-02
    mean
                             2.39540e-01
    sigma
                  3.20763e+00
                                             2.78628e-04 -5.34724e-02
                             ERR DEF= 0.5
EXTERNAL ERROR MATRIX.
                                                  ERR DEF=0.5
                         NDIM = 25
                                      NPAR
 1.049e-01 3.338e-04
 3.338e-04 5.739e-02
                                       Parameter values and approximate
PARAMETER CORRELATION COEFFICIENTS
                                          errors reported by MINUIT
                      1
     NO. GLOBAL
       1 0.00430 1.000 0.004
                                       Error definition (in this case 0.5 for
          0.00430
                   0.004
                          1.000
                                                a likelihood fit)
```

Statistical Methods

#### Minuit function MIGRAD



Purpose: find minimum

```
Value of \chi^2 or likelihood at
*****
                       minimum
    13 **MIGR
*****
               (NB: \chi^2 values are not divided
(some output o
                        by N_{dof}
MIGRAD MINIMIZ
MIGRAD WILL VERIE
                      MOLNUL AND ERRUR MATRIX.
COVARIANCE MATRIX CALCULATED SUCCESSFULLY
FCN=257.304 FROM MIGRAD
                                              31 CALLS
                         STATUS=CONVERGED
                                                                32 TOTAL
                                     STRATEGY= 1
                   EDM=2.36773e-06
                                                     ERROR MATRIX ACCURATE
 EXT PARAMETER
                                               STEP
                                                           FIRST
                VALUE
                                ERROR
NO.
      NAME
                                               SIZE
                                                        DERIVATIVE
              8.84225e-02 3.23862e-01
                                            3.58344e-04 -2.24755e-02
    mean
                            2.39540e-01
                                            2.78628e-04 -5.34724e-02
    sigma
                 3.20763e+00
                            EPP DEF= 0.5
EXTERNAL ERROR MATRIX.
                        NDIM= 25
                                     NPAR= 2
                                                ERR DEF=0.5
1.049e-01 3.338e-04
 3.338e-04 5.739e-02
                                            Approximate
PARAMETER CORRELATION COEFFICIENTS
                                             Error matrix
                      1
     NO. GLOBAL
                             2
                                        And covariance matrix
      1 0.00430 1.000 0.004
         0.00430
                   0.004
                        1.000
```

#### Minuit function MICD AT

INFN

Purpose: find minimula

\*\*\*\*\*\*\*

\*\* 13 \*\*MIGRAD 1000

\*\*\*\*\*\*\*

(some output omitted)

MIGRAD MINIMIZATION HAS CONVERGED

MIGRAD WILL VERIFY CONVERGENCE AND

**Status:**Should be 'converged' but can be 'failed'

Estimated Distance to Minimum should be small O(10<sup>-6</sup>)

Error Matrix Quality should be 'accurate', but can be 'approximate' in case of trouble

MATRIX.

STRATEGY= 1

```
COVARIANCE MATRIX CALCULATED SUCCESSFULLY
```

FCN=257.304 FROM MIGRAD

STATUS=CONVERGED

31 CALLS

32 TOTAL

ERROR MATRIX ACCURATE

EXT PARAMETER STEP FIRST
NO. NAME VALUE ERROR SIZE DERIVATIVE

EDM=2.36773e-06

1 mean 8.84225e-02 2 sigma 3.20763e+00

3.23862e-01 2.39540e-01

3.58344e-04 -2.24755e-02

2.78628e-04 -5.34724e-02

ERR DEF= 0.5

EXTERNAL ERROR MATRIX. NDIM= 25 NPAR= 2 ERR DEF=0.5

1.049e-01 3.338e-04

3.338e-04 5.739e-02

PARAMETER CORRELATION COEFFICIENTS

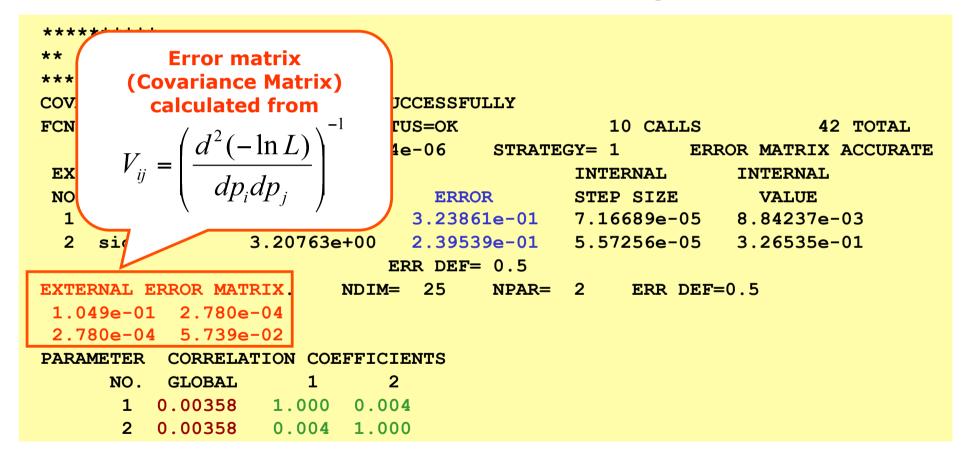
NO. GLOBAL 1 2 1 0.00430 1.000 0.004

2 0.00430 0.004 1.000



```
*****
     18 **HESSE
                       1000
                                                  Symmetric errors
*****
                                                 calculated from 2<sup>nd</sup>
COVARIANCE MATRIX CALCULATED SUCCESSFULLY
                                              derivative of -ln(L) or \chi^2
FCN=257.304 FROM HESSE
                           STATUS=OK
                                                                        OTAL
                    EDM=2.36534e-06
                                       STRAT
                                                                        CURATE
                                               TERNAL.
                                                             TNTERNAT.
 EXT PARAMETER
                                  ERROR
                                              STEP SIZE
                                                               VALUE
NO.
       NAME.
                 VALUE
               8.84225e-02
                                3.23861e-01
                                              7.16689e-05 8.84237e-03
    mean
                  3.20763e+00
                                2.39539e-01
                                              5.57256e-05 3.26535e-01
     sigma
                              ERR DEF= 0.5
                          NDIM = 25
                                                   ERR DEF=0.5
EXTERNAL ERROR MATRIX
                                              2
                                       NPAR=
1.049e-01 2.780e-04
 2.780e-04 5.739e-02
PARAMETER CORRELATION COEFFICIENTS
     NO. GLOBAL
                       1
       1 0.00358 1.000 0.004
       2 0.00358 0.004 1.000
```







```
*****
     18 **HESSE
                        1000
*****
COVARIANCE MATRIX CALCULATED SUCCESSFULLY
FCN=257.304 FROM HESSE
                            STATUS=OK
                                                   10 CALLS
                                                                      42 TOTAL
                    EDM=2.36534e-06
                                                          ERROR MATRIX ACCURATE
                                        STRATEGY= 1
EXT PARAMETER
                                                INTERNAL
                                                               INTERNAL
                                                CHED CIPE
                                                                VALUE
NO.
       NAME.
                 VALUE
                8.84225e-02
                                                              8.84237e-03
  1 mean
                                   Correlation matrix \rho_{ii}
     sigma
                  3.20763e+00
                                     calculated from
                                                              3.26535e-01
                                    V_{ij} = \sigma_i \sigma_j \rho_{ij}
                                                           F = 0.5
EXTERNAL ERROR MATRIX.
                           NDIN
1.049e-01 2.780e-04
 2.780e-04 5.739e-02
PARAMETER CORRELATION COEFFICIENT
      NO. GLOBAL
       1 0.00358
                    1.000 0.004
       2 0.00358
                   0.004 1.000
```

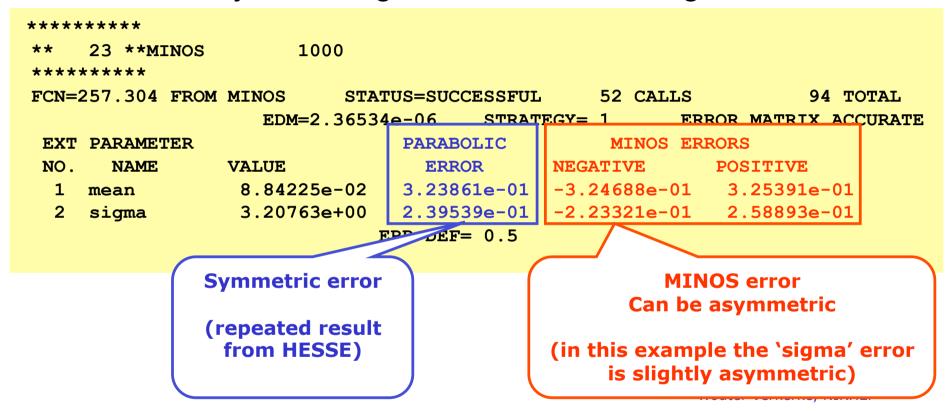


```
*****
    18 **HESSE
                       1000
*****
COVARIANCE MATRIX CALCULATED SUCCESSFULLY
FCN=257.304 FROM HESSE
                           STATUS=OK
                                                 10 CALLS
                                                                   42 TOTAL
                   EDM=2.36534e-06
                                                        ERROR MATRIX ACCURATE
                                       STRATEGY= 1
EXT PARAMETER
                                              INTERNAL
                                                            INTERNAL
                                              STEP SIZE
                                                              VALUE
NO.
      NAME
                                              7.16689e-05 8.84237e-03
                 Global correlation vector:
    mean
               correlation of each parameter
                                              5.57256e-05 3.26535e-01
    sigma
                 with all other parameters
EXTERNAL ERROR
                                              2
                                                   ERR DEF=0.5
1.049e-01
 2.780e-04
           5.739e
          CORRELA
                       COEFFICIENTS
PARAMETER
     NO. GLOBAL
                              2
         lo.00358|
                    1.000 0.004
         0.00358
                    0.004 1.000
```

#### Minuit function MINOS



Error analysis through ∆nll contour finding

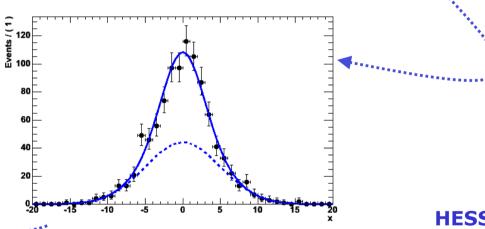


## Mitigating fit stability problems



- Strategy I More orthogonal choice of parameters
  - Example: fitting sum of 2 Gaussians of similar width

$$F(x; f, m, s_1, s_2) = f G(x; s_1, m) + (1 - f)G_2(x; s_2, m)$$



**HESSE** correlation matrix

#### Widths s<sub>1</sub>,s<sub>2</sub> strongly correlated fraction f

r**action f** Luca Lista

```
NO.
     GLOBAL
                                      [s2]
                       [ m]
                               [s1]
    0.96973
                             0.918
              1.000 -0.135
                                     0.915
             -0.135
                     1.000 -0.144 -0.114
    0.14407
             0.918 - 0.144
    0.92762
                             1.000
                                     0.786
             0.915 - 0.114
                             0.786
    0.92486
                                     1.000
```

CORRELATION COEFFICIENTS

## Mitigating fit stability problems



– Different parameterization:

$$f G(x; s_1, m_1) + (1 - f)G_2(x; \underline{s_1 \cdot s_2}, m_2)$$

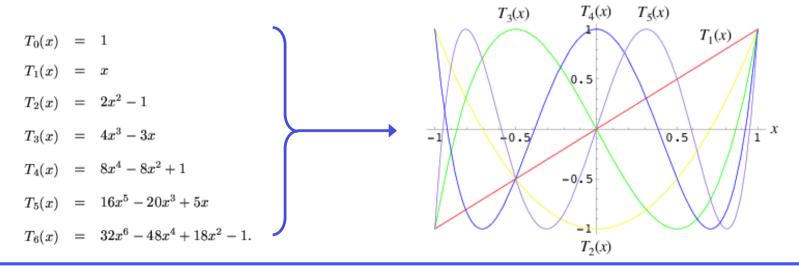
```
CORRELATION COEFFICIENTS
PARAMETER
                       [f]
                                    [s1]
                                            [s2]
      NO.
           GLOBAL
                              [m]
          0.96951 1.000 -0.134 0.917 -0.681
    [ f]
          0.14312 -0.134 1.000 -0.143
                                          0.127
    [ m]
          0.98879 0.917 - 0.143 1.000 - 0.895
    [s1]
    [s2]
          0.96156 - 0.681 \quad 0.127 - 0.895
                                          1.000
```

- Correlation of width s2 and fraction f reduced from 0.92 to 0.68
- Choice of parameterization matters!
- Strategy II Fix all but one of the correlated parameters
  - If floating parameters are highly correlated, some of them may be redundant and not contribute to additional degrees of freedom in your model

## Fit stability with polynomials



- Warning: Regular parameterization of polynomials  $a_0 + a_1 x$ + $a_2 x^2 + a_3 x^3$  nearly always results in strong correlations between the coefficients  $a_i$ .
  - Fit stability problems, inability to find right solution common at higher orders
- Solution: Use existing parameterizations of polynomials that have (mostly) uncorrelated variables
  - Example: Chebychev polynomials



## Browsing fit results



- As fits grow in complexity (e.g. 45 floating parameters), number of output variables increases
  - Need better way to navigate output that MINUIT screen dump
- RooFitResult holds complete snapshot of fit results
  - Constant parameters
  - Initial and final values of floating parameters

sigma

- Global correlations & full correlation matrix
- Returned from RooAbsPdf::fitTo() when "r" option is supplied

2.9803e-01 +/- 4.00e-03

Compact & verbose printing mode

Compact Mode

## Browsing fit results



Verbose printing mode

```
fitres->Print("v") ;
 RooFitResult: min. NLL value: 1.6e+04, est. distance to min: 1.2e-05
 Constant Parameter
                   Value
                                     Constant parameters
              cutoff 9.0000e+00
                                     listed separately
              glfrac 3.0000e-01
   Floating Parameter InitialValue
                                     FinalValue +/- Error
                                                             GblCorr.
              argpar -5.0000e-01 -4.6855e-01 +/- 7.11e-02 0.191895
              g2frac 3.0000e-01 3.0652e-01 +/- 5.10e-03 0.293455
               mean1 7.0000e+00 7.0022e+00 +/- 7.11e-03 0.113253
               mean2 2.0000e+00
                                    1.9971e+00 +/- 6.27e-03 0.100026
               sigma 3.0000e-01
                                    2.9803e-01 +/- 4.00e-03 0.276640
```

Initial, final value and global corr. listed side-by-side

#### Correlation matrix accessed separately

## Browsing fit results



- Easy navigation of correlation matrix
  - Select single element or complete row by parameter name

```
fitres->correlation("argpar","sigma")
(const Double_t)(-9.25606412005910845e-02)

fitres->correlation("mean1")->Print("v")
RooArgList::C[mean1,*]: (Owning contents)
   1) RooRealVar::C[mean1,argpar]: 0.11064 C
   2) RooRealVar::C[mean1,g2frac]: -0.0262487 C
   3) RooRealVar::C[mean1,mean1]: 1.0000 C
   4) RooRealVar::C[mean1,mean2]: -0.00632847 C
   5) RooRealVar::C[mean1,sigma]: -0.0339814 C
```

- RooFitResult persistable with ROOT I/O
  - Save your batch fit results in a ROOT file and navigate your results just as easy afterwards

#### References



- RooFit online tutorial
  - http://roofit.sourceforge.net/docs/tutorial/ index.html
- Credits:
  - RooFit slides and examples extracted, adapted and/or inspired by original presentations by Wouter Verkerke