



RooFit

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\$ROOTSYS/tutorials/roofit

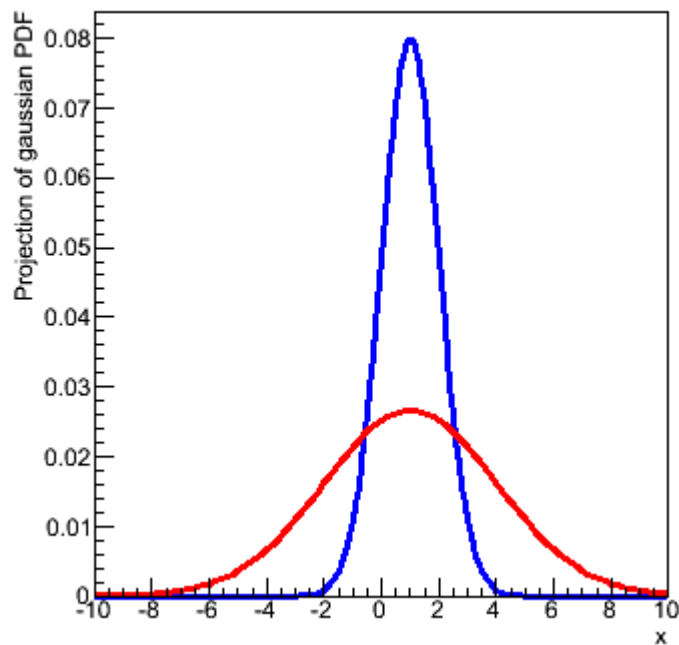
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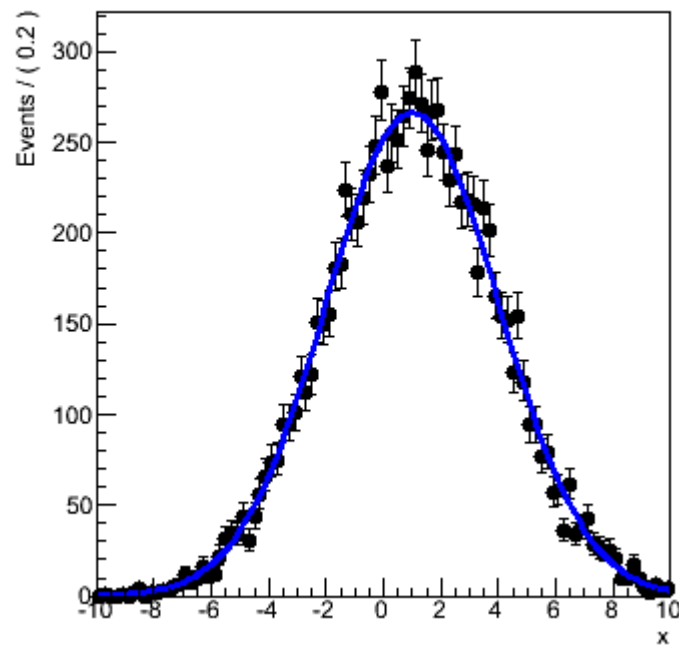


rf101_basics.C

Gaussian p.d.f.



Gaussian p.d.f. with data





```
//  
// 'BASIC FUNCTIONALITY' RooFit tutorial macro #101  
//  
// Fitting, plotting, toy data generation on one-dimensional p.d.f  
//  
// pdf = gauss(x,m,s)  
//  
//  
// 07/2008 - Wouter Verkerke  
//  
////////////////////////////////////  
  
#ifndef __CINT__  
#include "RooGlobalFunc.h"  
#endif  
#include "RooRealVar.h"  
#include "RooDataSet.h"  
#include "RooGaussian.h"  
#include "TCanvas.h"  
#include "RooPlot.h"  
#include "TAxis.h"  
using namespace RooFit ;  
  
void rf101_basics()  
{  
  // S e t u p   m o d e l  
  // -----  
  
  // Declare variables x,mean,sigma with associated name, title, initial value and allowed range  
  RooRealVar x("x","x",-10,10) ;  
  RooRealVar mean("mean","mean of gaussian",1,-10,10) ;  
  RooRealVar sigma("sigma","width of gaussian",1,0.1,10) ;  
}
```



```
// Build gaussian p.d.f in terms of x,mean and sigma
RooGaussian gauss("gauss","gaussian PDF",x,mean,sigma) ;

// Construct plot frame in 'x'
RooPlot* xframe = x.frame(Title("Gaussian p.d.f.")) ;

// Plot model and change parameter values
// -----

// Plot gauss in frame (i.e. in x)
gauss.plotOn(xframe) ;

// Change the value of sigma to 3
sigma.setVal(3) ;

// Plot gauss in frame (i.e. in x) and draw frame on canvas
gauss.plotOn(xframe,LineColor(kRed)) ;

// Generate events
// -----

// Generate a dataset of 1000 events in x from gauss
RooDataSet* data = gauss.generate(x,10000) ;

// Make a second plot frame in x and draw both the
// data and the p.d.f in the frame
RooPlot* xframe2 = x.frame(Title("Gaussian p.d.f. with data")) ;
data->plotOn(xframe2) ;
```



```
// Fit model to data
// -----
```

```
// Fit pdf to data
gauss.fitTo(*data) ;
```

```
// Print values of mean and sigma (that now reflect fitted values and errors)
mean.Print() ;
sigma.Print() ;
```

```
// Draw all frames on a canvas
```

```
TCanvas* c = new TCanvas("rf101_basics", "rf101_basics", 800, 400) ;
```

```
c->Divide(2) ;
```

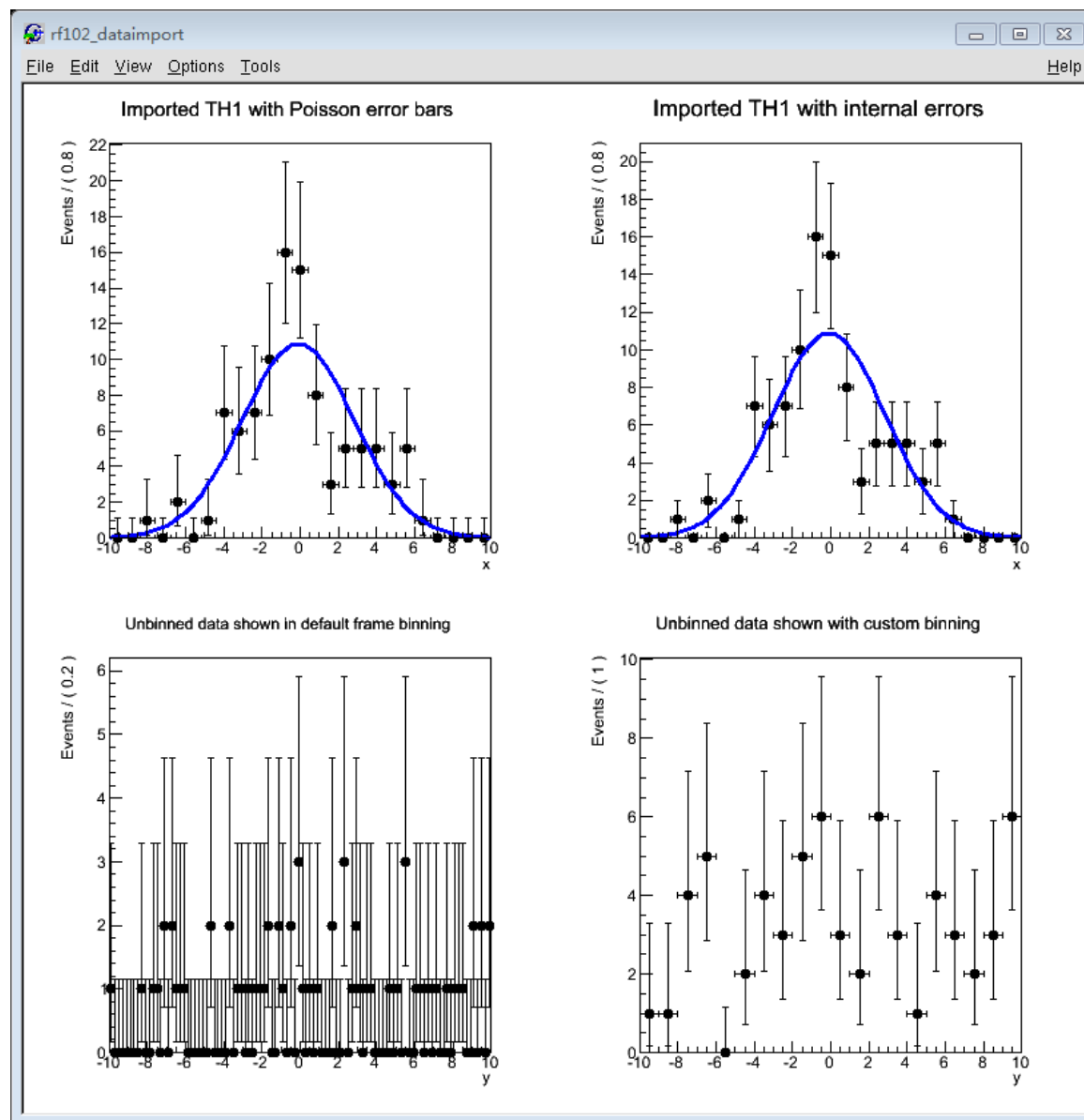
```
c->cd(1) ; gPad->SetLeftMargin(0.15) ; xframe->GetYaxis()->SetTitleOffset(1.6) ; xframe->Draw() ;
```

```
c->cd(2) ; gPad->SetLeftMargin(0.15) ; xframe2->GetYaxis()->SetTitleOffset(1.6) ; xframe2->Draw() ;
```

```
} :
```



rf102_dataimport.C





```
////////////////////////////////////  
//  
// 'BASIC FUNCTIONALITY' RooFit tutorial macro #102  
//  
// Importing data from ROOT TTrees and THx histograms  
//  
//  
// 07/2008 - Wouter Verkerke  
//  
////////////////////////////////////  
  
#ifndef __CINT__  
#include "RooGlobalFunc.h"  
#endif  
#include "RooRealVar.h"  
#include "RooDataSet.h"  
#include "RooDataHist.h"  
#include "RooGaussian.h"  
#include "TCanvas.h"  
#include "RooPlot.h"  
#include "TTree.h"  
#include "TH1D.h"  
#include "TRandom.h"  
using namespace RooFit ;  
  
TH1* makeTH1() ;  
TTree* makeTTree() ;
```




```
void rf102_dataimport()
{
    //////////////////////////////////////
    // Importing ROOT histograms //
    //////////////////////////////////////

    // Import TH1 into a RooDataHist
    // -----

    // Create a ROOT TH1 histogram
    TH1* hh = makeTH1() ;

    // Declare observable x
    RooRealVar x("x","x",-10,10) ;

    // Create a binned dataset that imports contents of TH1 and associates its contents to observable 'x'
    RooDataHist dh("dh","dh",x,Import(*hh)) ;

    // Plot and fit a RooDataHist
    // -----

    // Make plot of binned dataset showing Poisson error bars (RooFit default)
    RooPlot* frame = x.frame(Title("Imported TH1 with Poisson error bars")) ;
    dh.plotOn(frame) ;

    // Fit a Gaussian p.d.f to the data
    RooRealVar mean("mean","mean",0,-10,10) ;
    RooRealVar sigma("sigma","sigma",3,0.1,10) ;
    RooGaussian gauss("gauss","gauss",x,mean,sigma) ;
    gauss.fitTo(dh) ;
    gauss.plotOn(frame) ;
}
```



```
// Plot and fit a RooDataHist with internal errors
// -----

// If histogram has custom error (i.e. its contents is does not originate from a Poisson process
// but e.g. is a sum of weighted events) you can data with symmetric 'sum-of-weights' error instead
// (same error bars as shown by ROOT)
RooPlot* frame2 = x.frame(Title("Imported TH1 with internal errors")) ;
dh.plotOn(frame2,DataError(RooAbsData::SumW2)) ;
gauss.plotOn(frame2) ;

// Please note that error bars shown (Poisson or SumW2) are for visualization only, the are NOT used
// in a maximum likelihood fit
//
// A (binned) ML fit will ALWAYS assume the Poisson error interpretation of data (the mathematical definition :
// of likelihood does not take any external definition of errors). Data with non-unit weights can only be corr:
// ectly
// fitted with a chi^2 fit (see rf602_chi2fit.C)
```



```
// Import TTree into a RooDataSet
// -----

TTree* tree = makeTTree() ;

// Define 2nd observable y
RooRealVar y("Y","Y",-10,10) ;

// Construct unbinned dataset importing tree branches x and y matching between branches and RooRealVars
// is done by name of the branch/RRV
//
// Note that ONLY entries for which x,y have values within their allowed ranges as defined in
// RooRealVar x and y are imported. Since the y values in the import tree are in the range [-15,15]
// and RRV y defines a range [-10,10] this means that the RooDataSet below will have less entries than the TTree 'tree'

RooDataSet ds("ds","ds",RooArgSet(x,y),Import(*tree)) ;

// Plot dataset with multiple binning choices
// -----

// Print number of events in dataset
ds.Print() ;

// Print unbinned dataset with default frame binning (100 bins)
RooPlot* frame3 = y.frame(Title("Unbinned data shown in default frame binning")) ;
ds.plotOn(frame3) ;

// Print unbinned dataset with custom binning choice (20 bins)
RooPlot* frame4 = y.frame(Title("Unbinned data shown with custom binning")) ;
ds.plotOn(frame4,Binning(20)) ;

// Draw all frames on a canvas
TCanvas* c = new TCanvas("rf102_dataimport","rf102_dataimport",800,800) ;
c->Divide(2,2) ;
c->cd(1) ; gPad->SetLeftMargin(0.15) ; frame->GetYaxis()->SetTitleOffset(1.4) ; frame->Draw() ;
c->cd(2) ; gPad->SetLeftMargin(0.15) ; frame2->GetYaxis()->SetTitleOffset(1.4) ; frame2->Draw() ;
c->cd(3) ; gPad->SetLeftMargin(0.15) ; frame3->GetYaxis()->SetTitleOffset(1.4) ; frame3->Draw() ;
c->cd(4) ; gPad->SetLeftMargin(0.15) ; frame4->GetYaxis()->SetTitleOffset(1.4) ; frame4->Draw() ;

}
```



```
TH1* makeTH1()
{
    // Create ROOT TH1 filled with a Gaussian distribution

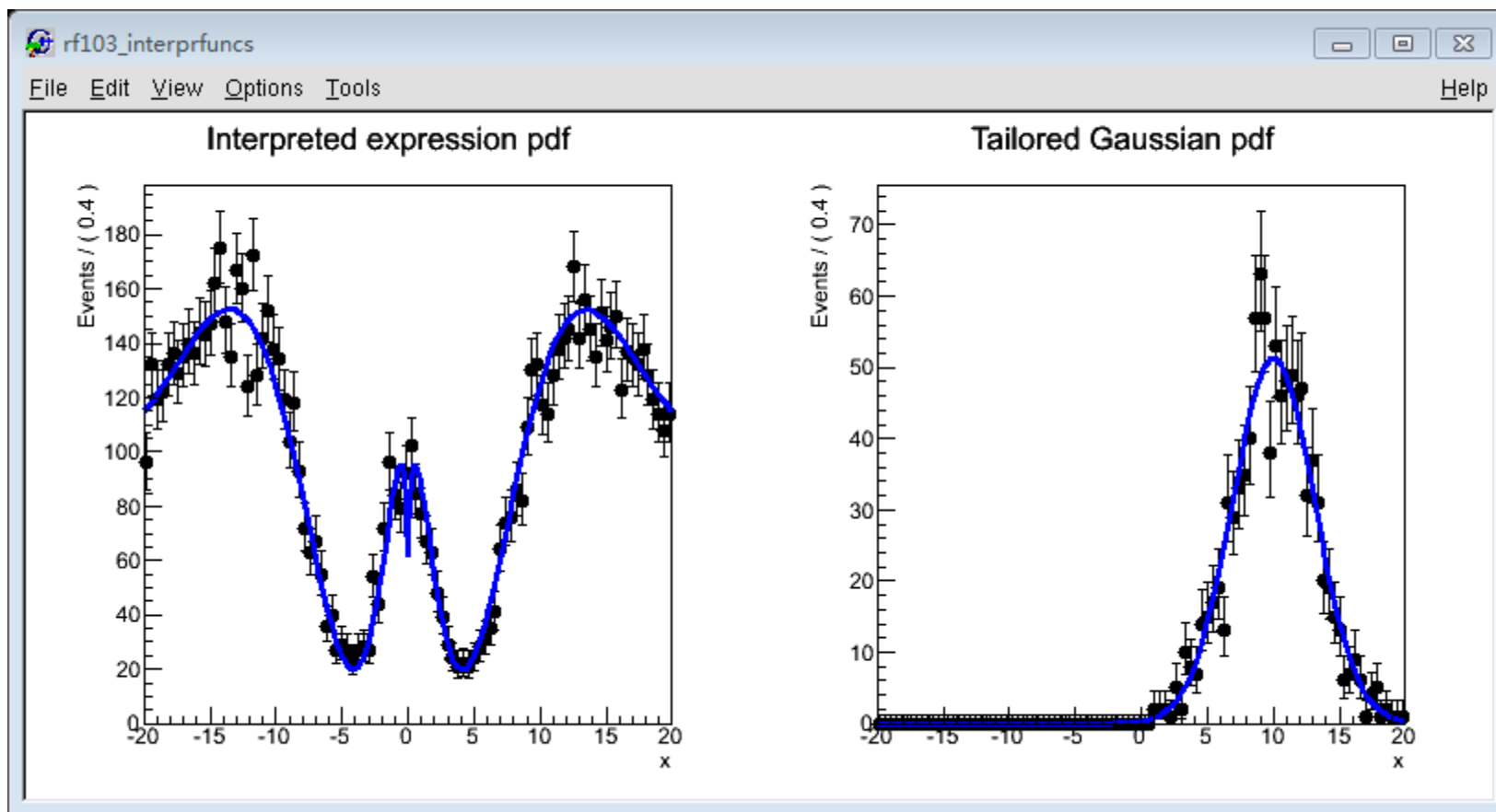
    TH1D* hh = new TH1D("hh", "hh", 25, -10, 10) ;
    for (int i=0 ; i<100 ; i++) {
        hh->Fill(gRandom->Gaus(0,3)) ;
    }
    return hh ;
}

TTree* makeTTree()
{
    // Create ROOT TTree filled with a Gaussian distribution in x and a uniform distribution in y

    TTree* tree = new TTree("tree", "tree") ;
    Double_t* px = new Double_t ;
    Double_t* py = new Double_t ;
    tree->Branch("x", px, "x/D") ;
    tree->Branch("y", py, "y/D") ;
    for (int i=0 ; i<100 ; i++) {
        *px = gRandom->Gaus(0,3) ;
        *py = gRandom->Uniform()*30 - 15 ;
        tree->Fill() ;
    }
    return tree ;
}
```



rf103_interprfuncs.C





```
////////////////////////////////////  
//  
// 'BASIC FUNCTIONALITY' RooFit tutorial macro #103  
//  
// Interpreted functions and p.d.f.s  
//  
//  
//  
// 07/2008 - Wouter Verkerke  
//  
////////////////////////////////////  
  
#ifndef __CINT__  
#include "RooGlobalFunc.h"  
#endif  
#include "RooRealVar.h"  
#include "RooDataSet.h"  
#include "RooGaussian.h"  
#include "TCanvas.h"  
#include "TAxis.h"  
#include "RooPlot.h"  
#include "RooFitResult.h"  
#include "RooGenericPdf.h"  
#include "RooConstVar.h"  
  
using namespace RooFit ;
```



```
void rfl03_interprfuncs()
{
    //////////////////////////////////////
    // Generic interpreted p.d.f. //
    //////////////////////////////////////

    // Declare observable x
    RooRealVar x("x","x",-20,20) ;

    // Construct generic pdf from interpreted expression
    // -----

    // To construct a proper p.d.f, the formula expression is explicitly normalized internally by dividing
    // it by a numeric integral of the expression over x in the range [-20,20]
    //
    RooRealVar alpha("alpha","alpha",5,0.1,10) ;
    RooGenericPdf genpdf("genpdf","genpdf","(1+0.1*abs(x)+sin(sqrt(abs(x*alpha+0.1))))",RooArgSet(x,alpha)) ;

    // Sample, fit and plot generic pdf
    // -----

    // Generate a toy dataset from the interpreted p.d.f
    RooDataSet* data = genpdf.generate(x,10000) ;

    // Fit the interpreted p.d.f to the generated data
    genpdf.fitTo(*data) ;

    // Make a plot of the data and the p.d.f overlaid
    RooPlot* xframe = x.frame(Title("Interpreted expression pdf")) ;
    data->plotOn(xframe) ;
    genpdf.plotOn(xframe) ;
}
```



```
// Construct parameter mean2 and sigma
RooRealVar mean2("mean2","mean^2",10,0,200) ;
RooRealVar sigma("sigma","sigma",3,0.1,10) ;

// Construct interpreted function mean = sqrt(mean^2)
RooFormulaVar mean("mean","mean","sqrt(mean2)",mean2) ;

// Construct a gaussian g2(x,sqrt(mean2),sigma) ;
RooGaussian g2("g2","h2",x,mean,sigma) ;

// Generate toy data
// -----

// Construct a separate gaussian g1(x,10,3) to generate a toy Gaussian dataset with mean 10 and width 3
RooGaussian g1("g1","g1",x,RooConst(10),RooConst(3)) ;
RooDataSet* data2 = g1.generate(x,1000) ;

// Fit and plot tailored standard pdf
// -----

// Fit g2 to data from g1
RooFitResult* r = g2.fitTo(*data2,Save()) ;
r->Print() ;

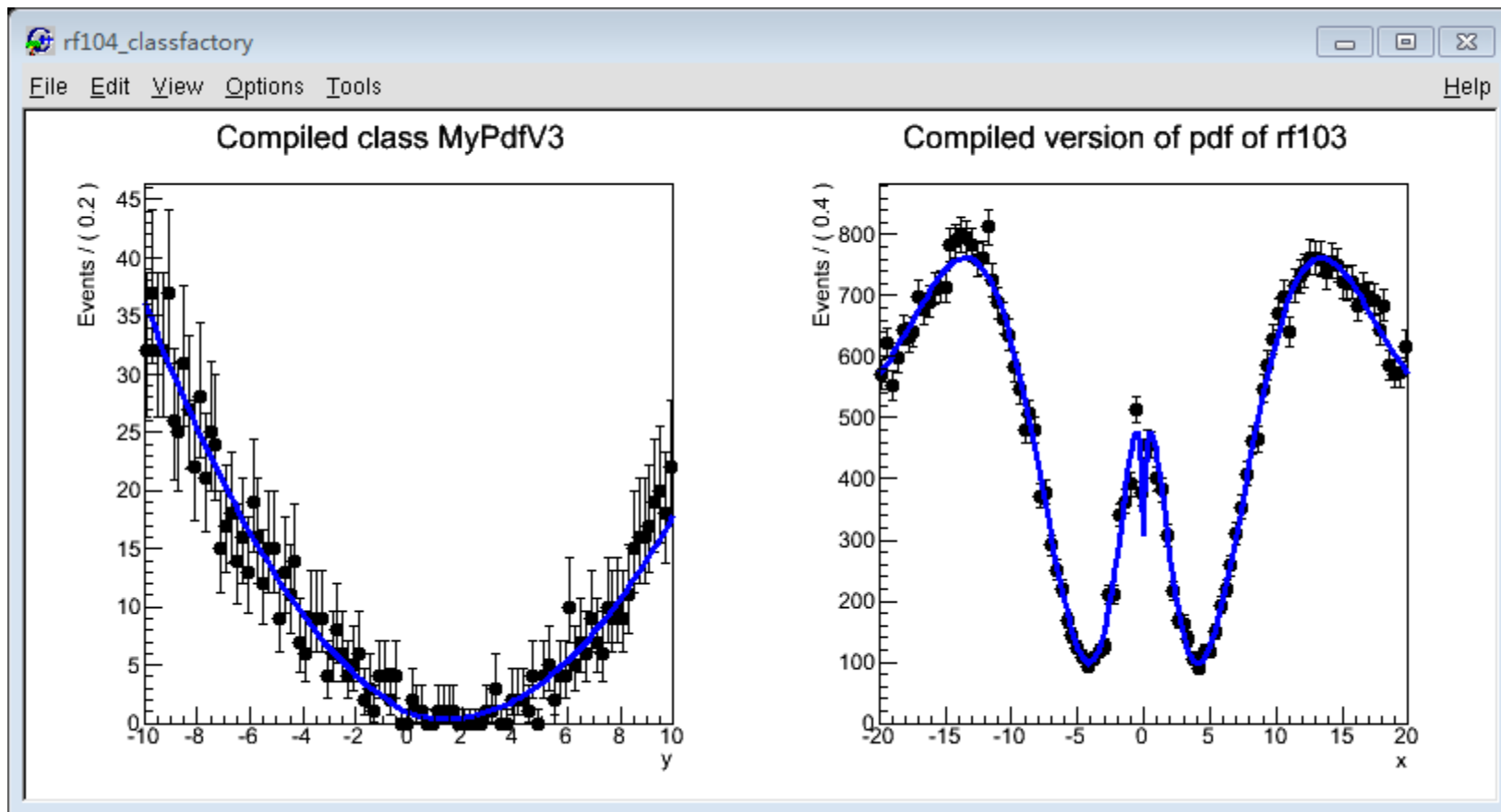
// Plot data on frame and overlay projection of g2
RooPlot* xframe2 = x.frame(Title("Tailored Gaussian pdf")) ;
data2->plotOn(xframe2) ;
g2.plotOn(xframe2) ;

// Draw all frames on a canvas
TCanvas* c = new TCanvas("rf103_interprfuncs","rf103_interprfuncs",800,400) ;
c->Divide(2) ;
c->cd(1) ; gPad->SetLeftMargin(0.15) ; xframe->GetYaxis()->SetTitleOffset(1.4) ; xframe->Draw() ;
c->cd(2) ; gPad->SetLeftMargin(0.15) ; xframe2->GetYaxis()->SetTitleOffset(1.4) ; xframe2->Draw() ;

}
```




rf104_classfactory.C





```
// NOTE: This demo uses code that is generated by the macro,  
//         therefore it cannot be compiled in one step by ACliC.  
//         To run this macro compiled with ACliC do  
//  
//         root>.x rf104_classfactory.C // run interpreted to generate code  
//         root>.L MyPdfV3.cxx+        // Compile and load created classs  
//         root>.x rf104_classfactory.C+ // run compiled code  
//  
//  
// 07/2008 - Wouter Verkerke  
//  
////////////////////////////////////  
  
#ifndef __CINT__  
#include "RooGlobalFunc.h"  
#endif  
#include "RooRealVar.h"  
#include "RooDataSet.h"  
#include "RooGaussian.h"  
#include "TCanvas.h"  
#include "TAxis.h"  
#include "RooPlot.h"  
#include "RooClassFactory.h"  
#include "TROOT.h"  
  
#ifndef __CINT__  
#include "MyPdfV3.h"  
#endif  
  
using namespace RooFit ;
```



```
void rf104_classfactory()
{
// Write class skeleton code
// -----

// Write skeleton p.d.f class with variable x,a,b
// To use this class,
// - Edit the file MyPdfV1.cxx and implement the evaluate() method in terms of x,a and b
// - Compile and link class with '.x MyPdfV1.cxx+'
//
RooClassFactory::makePdf("MyPdfV1","x,A,B") ;

// With added initial value expression
// -----

// Write skeleton p.d.f class with variable x,a,b and given formula expression
// To use this class,
// - Compile and link class with '.x MyPdfV2.cxx+'
//
RooClassFactory::makePdf("MyPdfV2","x,A,B","", "A*fabs(x)+pow(x-B,2)") ;

// With added analytical integral expression
// -----

// Write skeleton p.d.f class with variable x,a,b, given formula expression _and_
// given expression for analytical integral over x
// To use this class,
// - Compile and link class with '.x MyPdfV3.cxx+'
//
RooClassFactory::makePdf("MyPdfV3","x,A,B","", "A*fabs(x)+pow(x-B,2)", kTRUE, kFALSE,
                        "x: (A/2)*(pow(x.max(rangeName),2)+pow(x.min(rangeName),2))+(1./3)*(pow(x.max(rangeName),3)-B,3)-pow(x.min(rangeName)-B,3))") ;
}
```

Bool_t makePdf(const char* name, const char* realArgNames = 0, const char* catArgNames = 0, const char* expression = "1.0", Bool_t hasAnaInt = kFALSE, Bool_t hasIntGen = kFALSE, const char* intExpression = 0)



```
// Use instance of created class
// -----

// Compile MyPdfV3 class (only when running in CINT)
#ifdef __CINT__
gROOT->ProcessLineSync(".x MyPdfV3.cxx+") ;
#endif

// Create instance of MyPdfV3 class
RooRealVar a("a","a",1) ;
RooRealVar b("b","b",2,-10,10) ;
RooRealVar y("y","y",-10,10);
MyPdfV3 pdf("pdf","pdf",y,a,b) ;

// Generate toy data from pdf and plot data and p.d.f on frame
RooPlot* frame1 = y.frame(Title("Compiled class MyPdfV3")) ;
RooDataSet* data = pdf.generate(y,1000) ;
pdf.fitTo(*data) ;
data->plotOn(frame1) ;
pdf.plotOn(frame1) ;
```



一定让同学们实现自己做拟合的Class

用rootlogon.C函数内写编译后的.so文件

注：用gSystem->Load(".....so");

修改rf104_classfactory.C,把自己编译的文件当成
RooFit自带的文件进行参加拟合



```
// Compiled version of example rf103 //
////////////////////////////////////

// Declare observable x
RooRealVar x("x","x",-20,20) ;

// The RooClassFactory::makePdfInstance() function performs code writing, compiling, linking
// and object instantiation in one go and can serve as a straight replacement of RooGenericPdf

RooRealVar alpha("alpha","alpha",5,0.1,10) ;
RooAbsPdf* genpdf = RooClassFactory::makePdfInstance("GenPdf","(1+0.1*fabs(x)+sin(sqrt(fabs(x*alpha+0.1))))",RooArgSet(x,alpha)) ;

// Generate a toy dataset from the interpreted p.d.f
RooDataSet* data2 = genpdf->generate(x,50000) ;

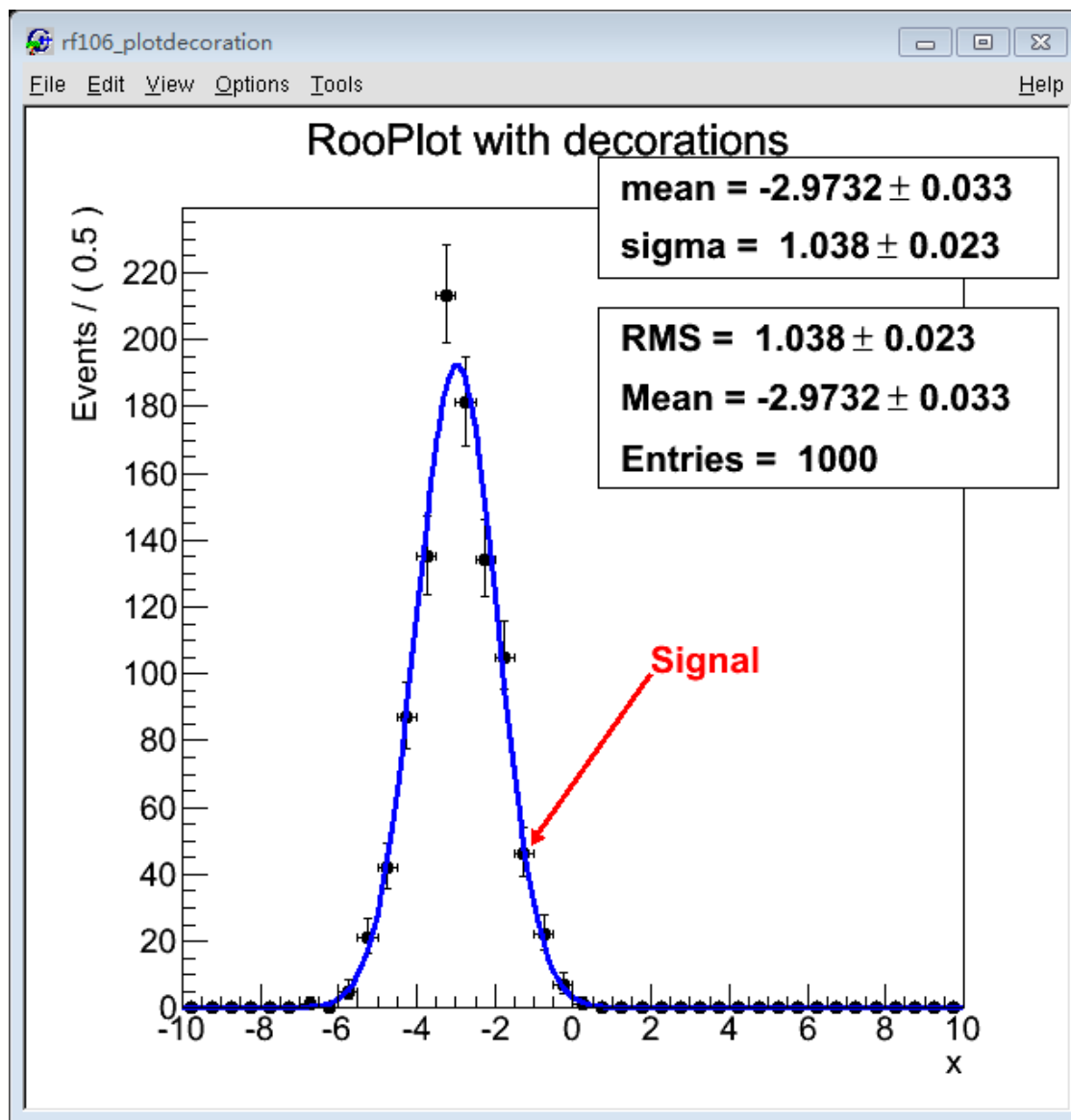
// Fit the interpreted p.d.f to the generated data
genpdf->fitTo(*data2) ;

// Make a plot of the data and the p.d.f overlaid
RooPlot* frame2 = x.frame(Title("Compiled version of pdf of rf103")) ;
data2->plotOn(frame2) ;
genpdf->plotOn(frame2) ;

// Draw all frames on a canvas
TCanvas* c = new TCanvas("rf104_classfactory","rf104_classfactory",800,400) ;
c->Divide(2) ;
c->cd(1) ; gPad->SetLeftMargin(0.15) ; frame1->GetYaxis()->SetTitleOffset(1.4) ; frame1->Draw() ;
c->cd(2) ; gPad->SetLeftMargin(0.15) ; frame2->GetYaxis()->SetTitleOffset(1.4) ; frame2->Draw() ;
}
```



rf106_plotdecoration.C





```
//  
// 'BASIC FUNCTIONALITY' RooFit tutorial macro #106  
//  
// Adding boxes with parameters, statistics to RooPlots.  
// Decorating RooPlots with arrows, text etc...  
//  
//  
// 07/2008 - Wouter Verkerke  
//  
////////////////////////////////////  
  
#ifndef __CINT__  
#include "RooGlobalFunc.h"  
#endif  
#include "RooRealVar.h"  
#include "RooDataSet.h"  
#include "RooGaussian.h"  
#include "TCanvas.h"  
#include "TAxis.h"  
#include "RooPlot.h"  
#include "TText.h"  
#include "TArrow.h"  
#include "TFile.h"  
using namespace RooFit ;
```




```
void rf106_plotdecoration()
{
    // s e t u p   m o d e l
    // -----

    // Create observables
    RooRealVar x("x", "x", -10, 10) ;

    // Create Gaussian
    RooRealVar sigma("sigma", "sigma", 1, 0.1, 10) ;
    RooRealVar mean("mean", "mean", -3, -10, 10) ;
    RooGaussian gauss("gauss", "gauss", x, mean, sigma) ;

    // Generate a sample of 1000 events with sigma=3
    RooDataSet* data = gauss.generate(x, 1000) ;

    // Fit pdf to data
    gauss.fitTo(*data) ;

    // P l o t   p . d . f   a n d   d a t a
    // -----

    // Overlay projection of gauss on data
    RooPlot* frame = x.frame(Name("xframe"), Title("RooPlot with decorations"), Bins(40)) ;
    data->plotOn(frame) ;
    gauss.plotOn(frame) ;
```

```
data->plotOn(frame, Binning(40));
```



```
// Add box with pdf parameters
// -----

// Left edge of box starts at 55% of Xaxis)
gauss.paramOn(frame,Layout(0.55)) ;

// Add box with data statistics
// -----

// X size of box is from 55% to 99% of Xaxis range, top of box is at 80% of Yaxis range)
data->statOn(frame,Layout(0.55,0.99,0.8)) ;

// Add text and arrow
// -----

// Add text to frame
TText* txt = new TText(2,100,"Signal") ;
txt->SetTextSize(0.04) ;
txt->SetTextColor(kRed) ;
frame->addObject(txt) ;

// Add arrow to frame
TArrow* arrow = new TArrow(2,100,-1,50,0.01,"|>") ;
arrow->SetLineColor(kRed) ;
arrow->SetFillColor(kRed) ;
arrow->SetLineWidth(3) ;
frame->addObject(arrow) ;
```



```
// Persist frame with all decorations in ROOT file  
// -----
```

```
TFile f("rf106_plotdecoration.root","RECREATE") ;  
frame->Write() ;  
f.Close() ;
```

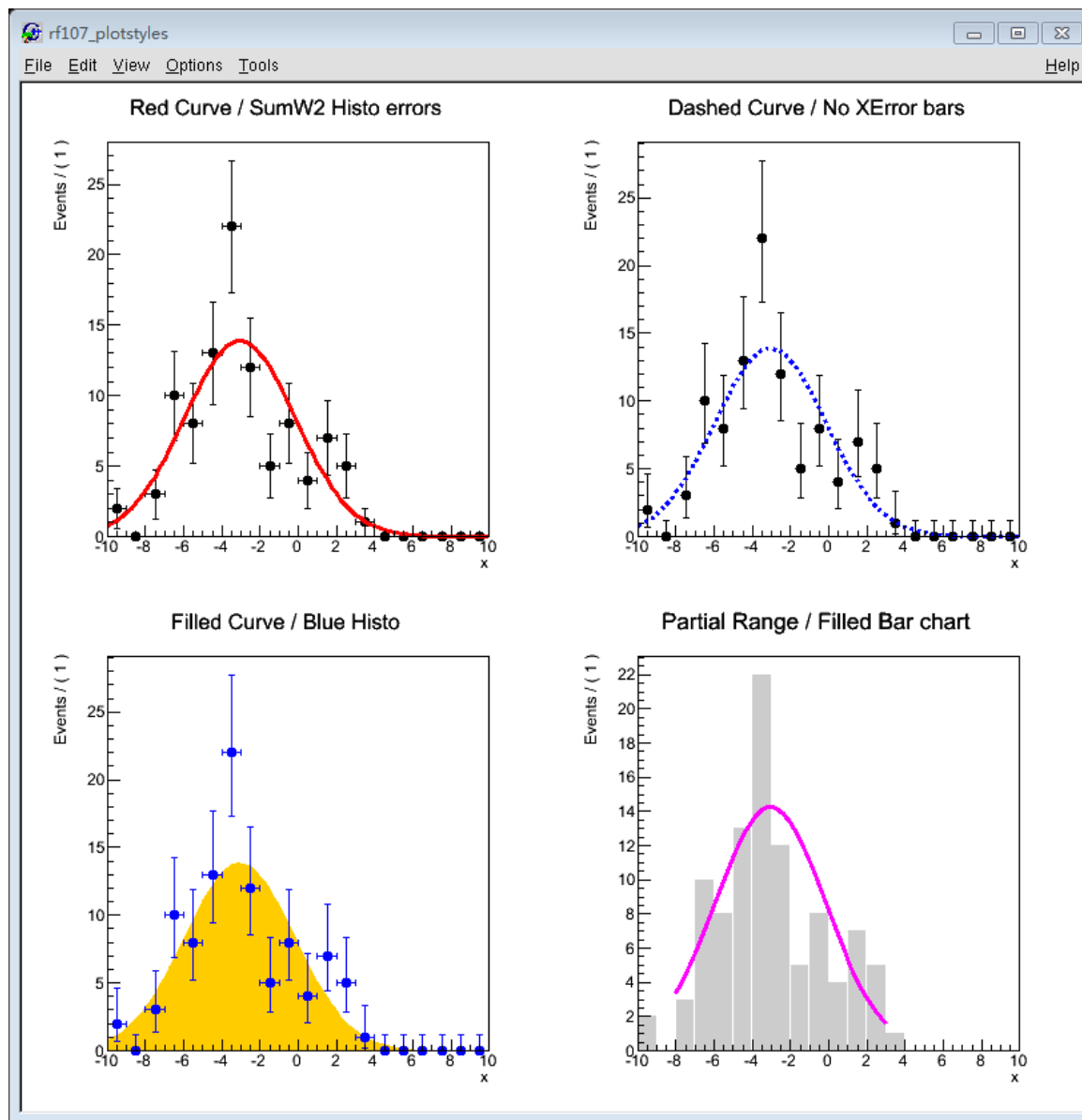
```
// To read back and plot frame with all decorations in clean root session do  
// root> TFile f("rf106_plotdecoration.root") ;  
// root> xframe->Draw() ;
```

```
new TCanvas("rf106_plotdecoration","rf106_plotdecoration",600,600) ;  
gPad->SetLeftMargin(0.15) ; frame->GetYaxis()->SetTitleOffset(1.6) ; frame->Draw() ;
```

```
}
```



rf107_plotstyles.C





```
// 'BASIC FUNCTIONALITY' RooFit tutorial macro #107
//
// Demonstration of various plotting styles of data, functions
// in a RooPlot
//
// 07/2008 - Wouter Verkerke
//
////////////////////////////////////
#ifndef __CINT__
#include "RooGlobalFunc.h"
#endif
#include "RooRealVar.h"
#include "RooDataSet.h"
#include "RooGaussian.h"
#include "TCanvas.h"
#include "TAxis.h"
#include "RooPlot.h"
using namespace RooFit ;
```



```
void rf107_plotstyles()
{
    // S e t u p   m o d e l
    // -----

    // Create observables
    RooRealVar x("x", "x", -10, 10) ;

    // Create Gaussian
    RooRealVar sigma("sigma", "sigma", 3, 0.1, 10) ;
    RooRealVar mean("mean", "mean", -3, -10, 10) ;
    RooGaussian gauss("gauss", "gauss", x, mean, sigma) ;

    // Generate a sample of 100 events with sigma=3
    RooDataSet* data = gauss.generate(x, 100) ;

    // Fit pdf to data
    gauss.fitTo(*data) ;

    // M a k e   p l o t   f r a m e s
    // -----

    // Make four plot frames to demonstrate various plotting features
    RooPlot* frame1 = x.frame(Name("xframe"), Title("Red Curve / SumW2 Histo errors"), Bins(20)) ;
    RooPlot* frame2 = x.frame(Name("xframe"), Title("Dashed Curve / No XError bars"), Bins(20)) ;
    RooPlot* frame3 = x.frame(Name("xframe"), Title("Filled Curve / Blue Histo"), Bins(20)) ;
    RooPlot* frame4 = x.frame(Name("xframe"), Title("Partial Range / Filled Bar chart"), Bins(20)) ;
}
```



```
// Data plotting styles
// -----

// Use sqrt(sum(weights^2)) error instead of Poisson errors
data->plotOn(frame1,DataError(RooAbsData::SumW2)) ;

// Remove horizontal error bars
data->plotOn(frame2,XErrorSize(0)) ;

// Blue markers and error bors
data->plotOn(frame3,MarkerColor(kBlue),LineColor(kBlue)) ;

// Filled bar chart
data->plotOn(frame4,DrawOption("B"),DataError(RooAbsData::None),XErrorSize(0),FillColor(kGray)) ;

// Function plotting styles
// -----

// Change line color to red
gauss.plotOn(frame1,LineColor(kRed)) ;

// Change line style to dashed
gauss.plotOn(frame2,LineStyle(kDashed)) ;

// Filled shapes in green color
gauss.plotOn(frame3,DrawOption("F"),FillColor(kOrange),MoveToBack()) ;

//
gauss.plotOn(frame4,Range(-8,3),LineColor(kMagenta)) ;
```



```
TCanvas* c = new TCanvas("rf107_plotstyles","rf107_plotstyles",800,800) ;
c->Divide(2,2) ;
c->cd(1) ; gPad->SetLeftMargin(0.15) ; frame1->GetYaxis()->SetTitleOffset(1.6) ; frame1->Draw() ;
c->cd(2) ; gPad->SetLeftMargin(0.15) ; frame2->GetYaxis()->SetTitleOffset(1.6) ; frame2->Draw() ;
c->cd(3) ; gPad->SetLeftMargin(0.15) ; frame3->GetYaxis()->SetTitleOffset(1.6) ; frame3->Draw() ;
c->cd(4) ; gPad->SetLeftMargin(0.15) ; frame4->GetYaxis()->SetTitleOffset(1.6) ; frame4->Draw() ;
}
```




Binned data (histograms)

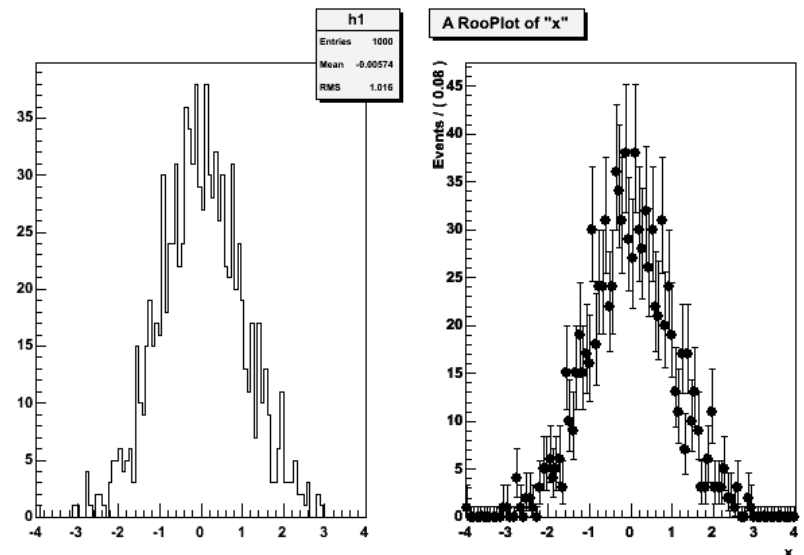
```
void ImportData(){  
    gSystem->Load("libRooFit");  
    using namespace RooFit;
```

```
    TH1F *h1 = new TH1F("h1","",100,-4,4);  
    h1 -> FillRandom("gaus",1000);  
    RooRealVar x("x","x",-4,4);
```

```
    RooDataHist data("data", "Dataset With X",x,h1);
```

```
    RooPlot* frame = x.frame();  
    data.plotOn(frame);
```

```
    TCanvas *c1 = new TCanvas("c1","");  
    c1 -> SetFillColor(10);  
    c1->Divide(2,1);  
    c1->cd(1);  
    h1->Draw();  
    c1->cd(2);  
    frame->Draw();  
    c1->cd();  
}
```





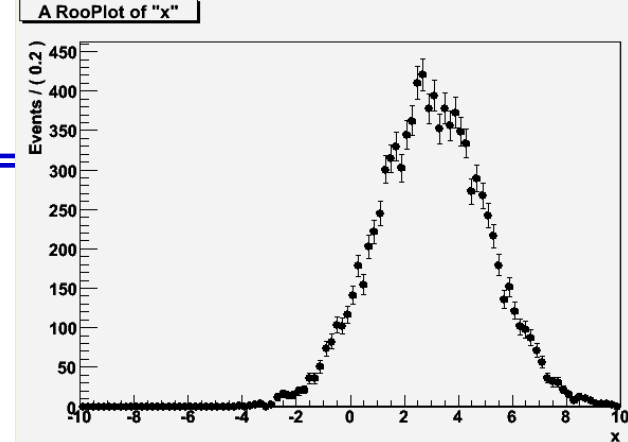
Unbinned data (add)

```
void UnbinnedData_add(){
    gSystem->Load("libRooFit");
    using namespace RooFit;

    RooRealVar x("x","x",-10,10);
    RooDataSet data("data","Dataset With X",x);

    for(Int_t j=0; j<10000; j++){
        x = gRandom->Gaus(3.0,2); //mean=-1, sigma =2
        data.add(x);
    }

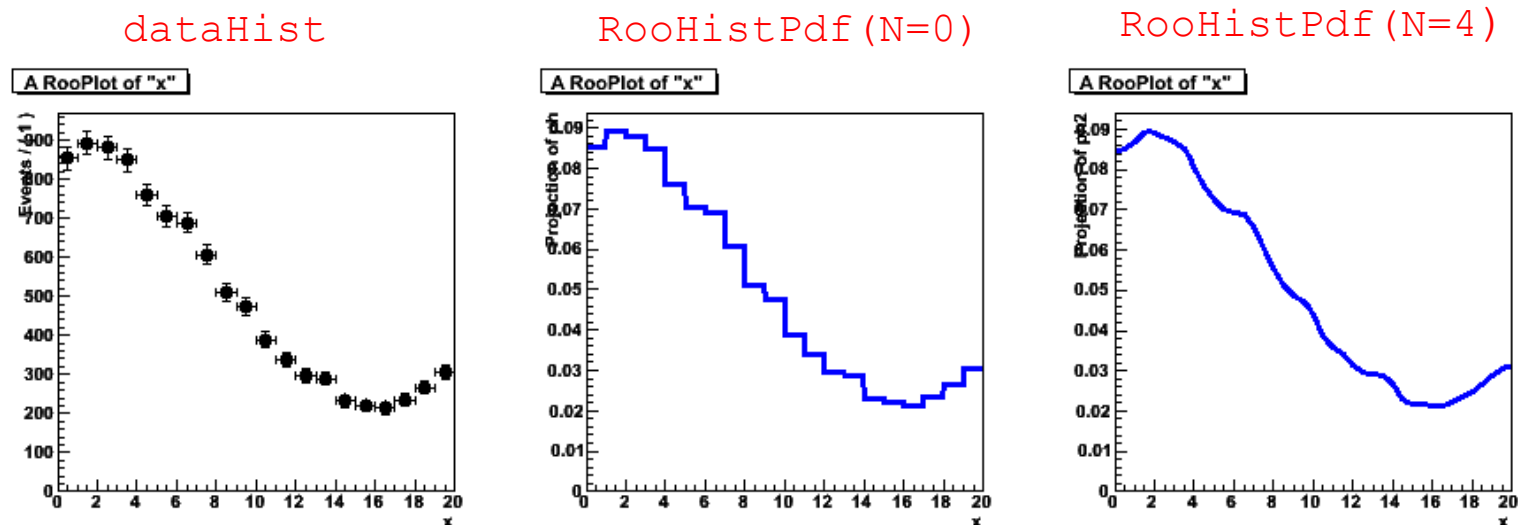
    RooPlot* frame = x.frame();
    data.plotOn(frame,Binning(100));
    frame->Draw();
}
```





Highlight of non-parametric shapes - histograms

- Will highlight two types of non-parametric p.d.f.s
- Class **RooHistPdf** – a p.d.f. described by a histogram

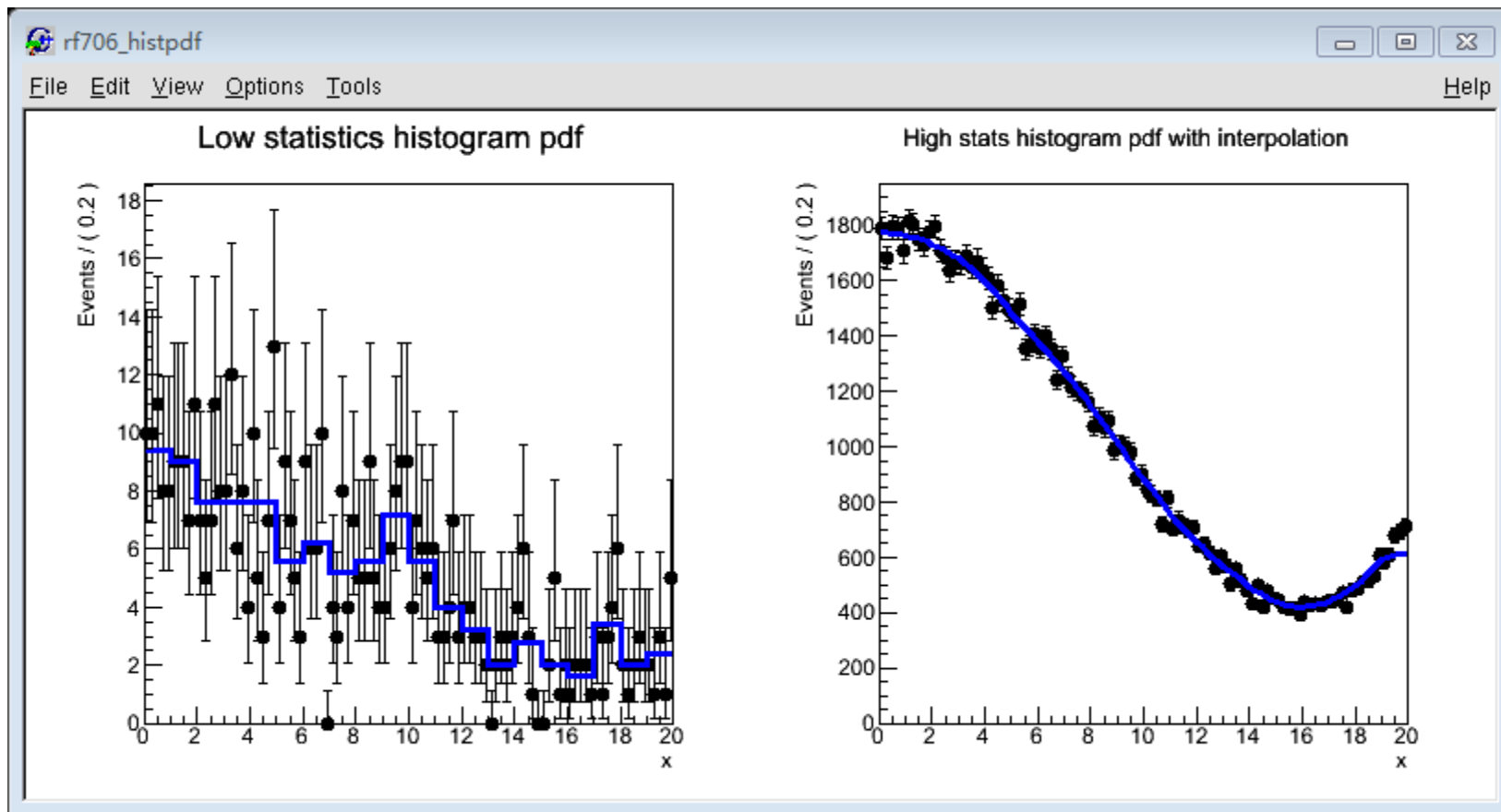


```
// Histogram based p.d.f with N-th order interpolation  
RooHistPdf ph("ph","ph",x,*dataHist,N) ;
```

- Not so great at low statistics (especially problematic in >1 dim)



rf706_histpdf.C

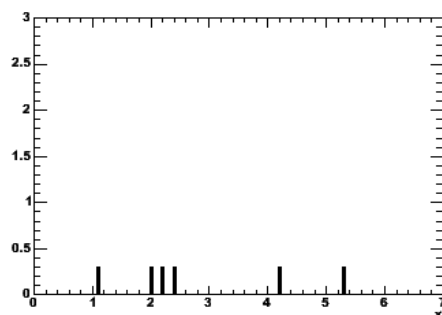




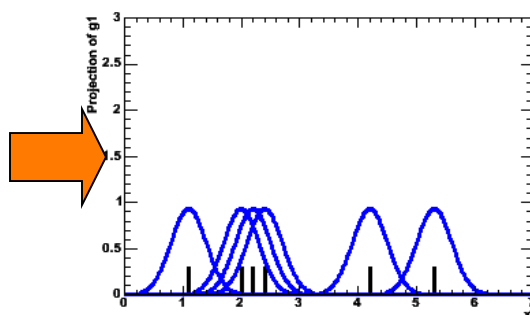
● Class **RooKeysPdf** – A kernel estimation p.d.f.

- Uses *unbinned* data
- Idea represent each event of your MC sample as a Gaussian probability distribution
- Add probability distributions from all events in sample

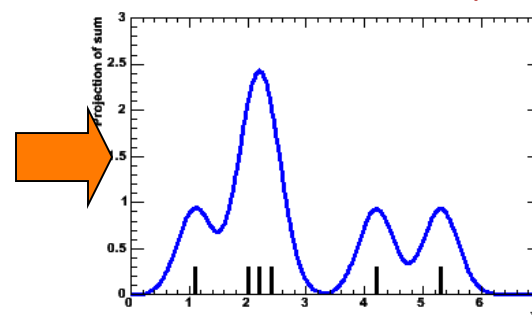
Sample of events



Gaussian
probability distributions
for each event



Summed
probability distribution
for all events in sample

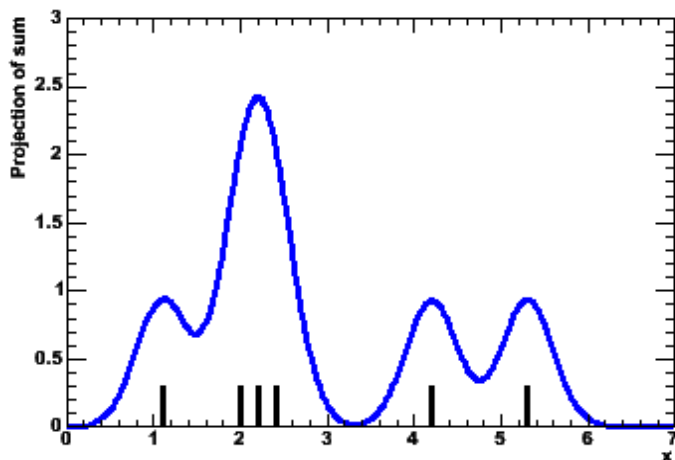




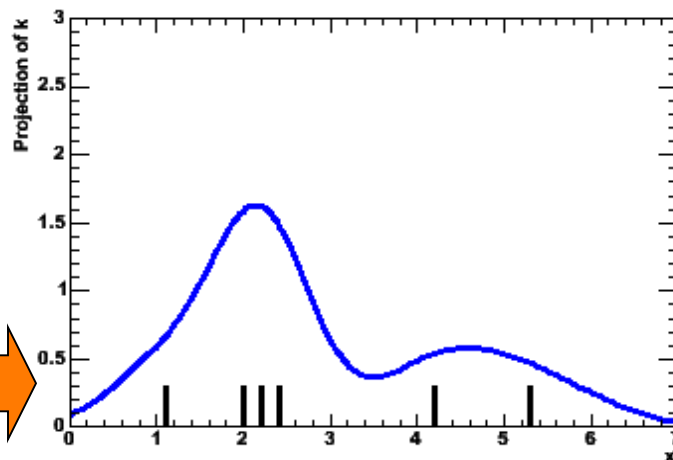
Highlight of non-parametric shapes – kernel estimation

- Width of Gaussian kernels need not be the same for all events
 - As long as each event contributes $1/N$ to the integral
- Idea: 'Adaptive kernel' technique
 - Choose wide Gaussian if local density of events is low
 - Choose narrow Gaussian if local density of events is high
 - Preserves small features in high statistics areas, minimize jitter in low statistics areas
 - Automatically calculated

Static Kernel
(with of all Gaussian identical)



Adaptive Kernel
(width of all Gaussian depends on local density of events)



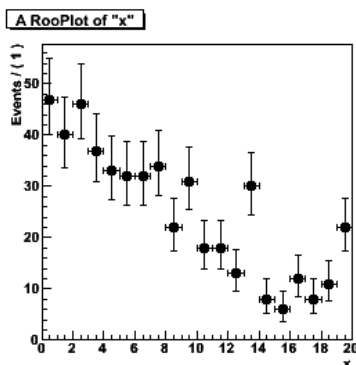


Highlight of non-parametric shapes – kernel estimation

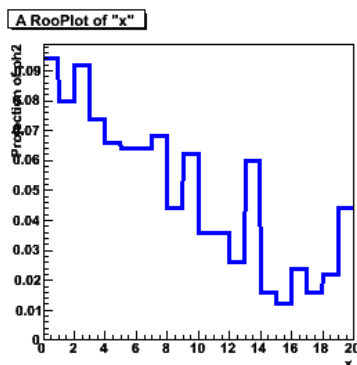
```
// Adaptive kernel estimation p.d.f  
RooKeysPdf k("k","k",x,*d,RooKeysPdf::MirrorBoth) ;
```

- Example with comparison to histogram based p.d.f
 - Superior performance at low statistics
 - Can mirror input data over boundaries to reduce ‘edge leakage’
 - Works also in >1 dimensions (class **RooNDKeysPdf**)

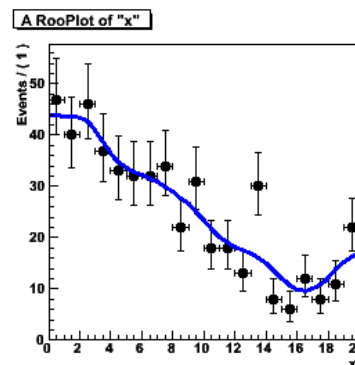
Data (N=500)



RooHistPdf(data)



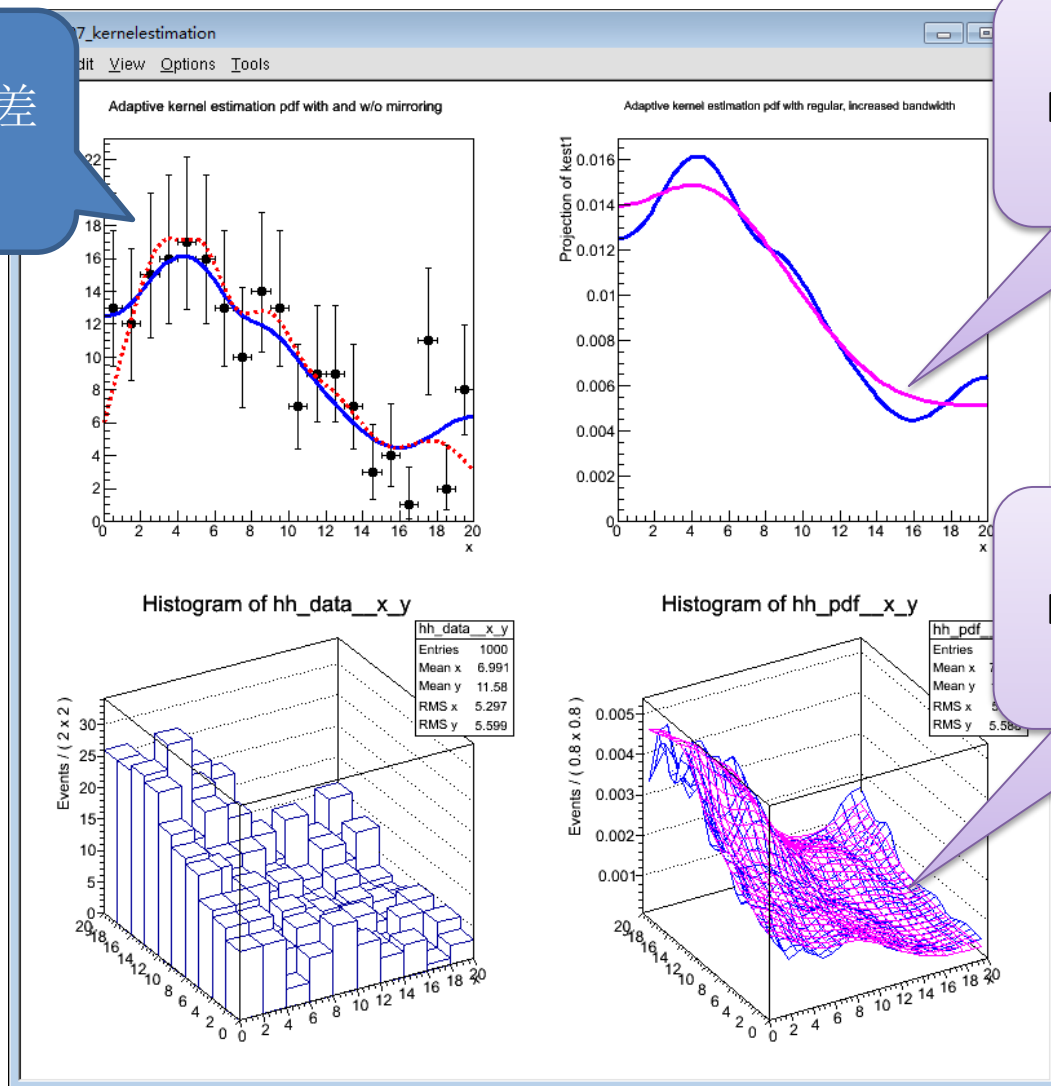
RooKeysPdf(data)





rf707_kernelestimation.C

是否镜像的差别





rf707_kernelestimation.C

```
void rf707_kernelestimation()
{
    // Create low stats 1-D dataset
    // -----

    // Create a toy pdf for sampling
    RooRealVar x("x","x",0,20) ;
    RooPolynomial p("p","p",x,RooArgList(RooConst(0.01),RooConst(-0.01),RooConst(0.0004))) ;

    // Sample 500 events from p
    RooDataSet* data1 = p.generate(x,200) ;


    // Create 1-D kernel estimation pdf
    // -----

    // Create adaptive kernel estimation pdf. In this configuration the input data
    // is mirrored over the boundaries to minimize edge effects in distribution
    // that do not fall to zero towards the edges
    RooKeysPdf kest1("kest1","kest1",x,*data1,RooKeysPdf::MirrorBoth) ;

    // An adaptive kernel estimation pdf on the same data without mirroring option
    // for comparison
    RooKeysPdf kest2("kest2","kest2",x,*data1,RooKeysPdf::NoMirror) ;
```



rf707_kernelestimatation.C

2! 比缺省光滑!

```
// Adaptive kernel estimation pdf with increased bandwidth scale factor
// (promotes smoothness over detail preservation)
RooKeysPdf kest3("kest1","kest1",x,*data1,RooKeysPdf::MirrorBoth,2) ;

// Plot kernel estimation pdfs with and without mirroring over data
RooPlot* frame = x.frame(Title("Adaptive kernel estimation pdf with and w/o mirroring"),Bins(20)) ;
data1->plot0n(frame) ;
kest1.plot0n(frame) ;
kest2.plot0n(frame,LineStyle(kDashed),LineColor(kRed)) ;

// Plot kernel estimation pdfs with regular and increased bandwidth
RooPlot* frame2 = x.frame(Title("Adaptive kernel estimation pdf with regular, increased bandwidth"))
kest1.plot0n(frame2) ;
kest3.plot0n(frame2,LineColor(kMagenta)) ;
```



rf707_kernelestimation.C

```
// Construct a 2D toy pdf for sampling
RooRealVar y("y","y",0,20) ;
RooPolynomial py("py","py",y,RooArgList(RooConst(0.01),RooConst(0.01),RooConst(-0.0004))) ;
RooProdPdf pxy("pxy","pxy",RooArgSet(p,py)) ;
RooDataSet* data2 = pxy.generate(RooArgSet(x,y),1000) ;

// Create 2 - D kernel estimation pdf
// -----

// Create 2D adaptive kernel estimation pdf with mirroring
RooNDKeysPdf kest4("kest4","kest4",RooArgSet(x,y),*data2,"am") ;

// Create 2D adaptive kernel estimation pdf with mirroring and double bandwidth
RooNDKeysPdf kest5("kest5","kest5",RooArgSet(x,y),*data2,"am",2) ;
```



rf707_kernelestimation.C

```
// Create a histogram of the data
```

```
TH1* hh_data = data2->createHistogram("hh_data",x,Binning(10),YVar(y,Binning(10))) ;
```

```
// Create histogram of the 2d kernel estimation pdfs
```

```
TH1* hh_pdf = kest4.createHistogram("hh_pdf",x,Binning(25),YVar(y,Binning(25))) ;
```

```
TH1* hh_pdf2 = kest5.createHistogram("hh_pdf2",x,Binning(25),YVar(y,Binning(25))) ;
```

```
hh_pdf->SetLineColor(kBlue) ;
```

```
hh_pdf2->SetLineColor(kMagenta) ;
```

```
TCanvas* c = new TCanvas("rf707_kernelestimation","rf707_kernelestimation",800,800) ;
```

```
c->Divide(2,2) ;
```

```
c->cd(1) ; gPad->SetLeftMargin(0.15) ; frame->GetYaxis()->SetTitleOffset(1.4) ; frame->Draw() ;
```

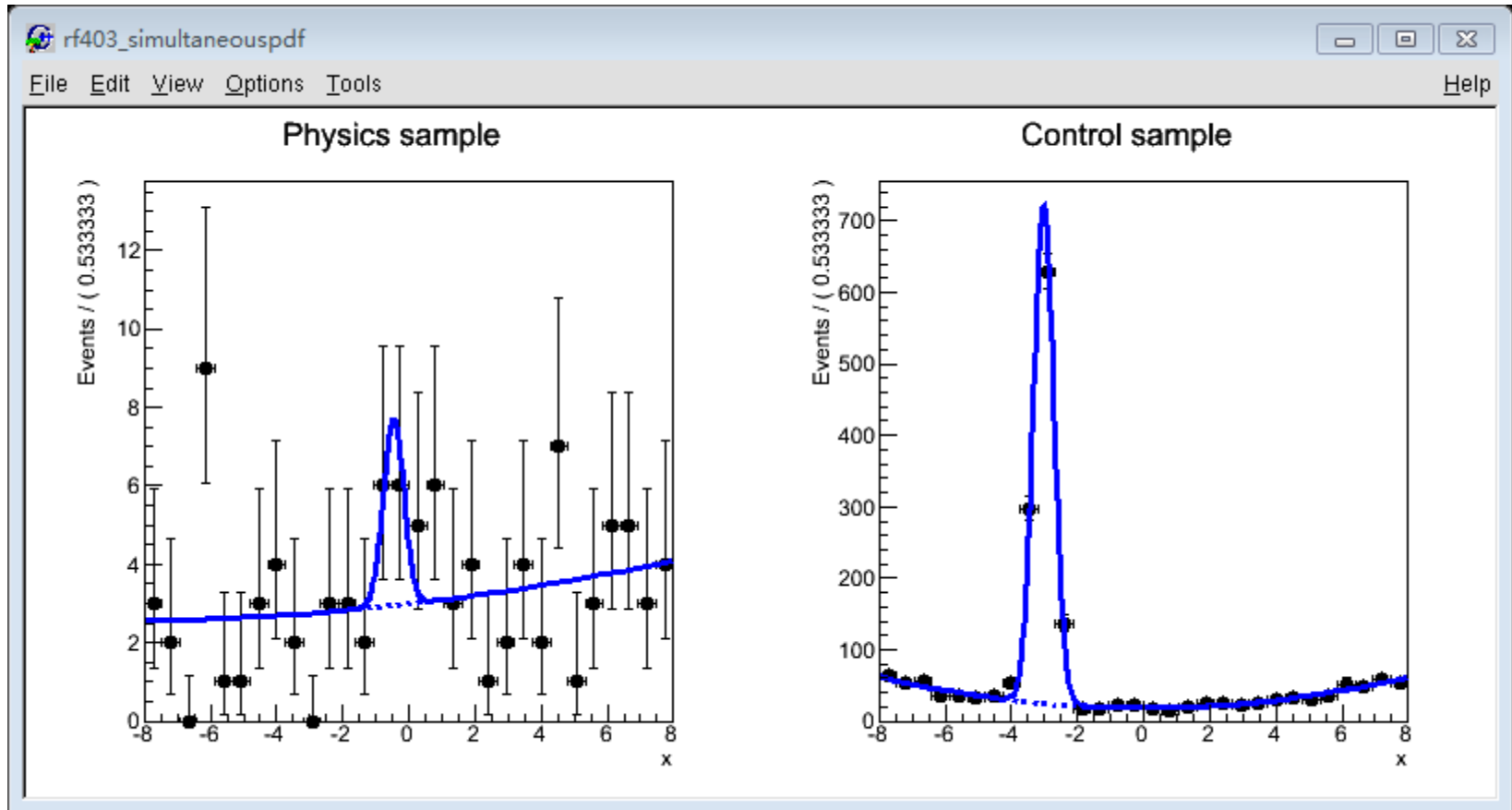
```
c->cd(2) ; gPad->SetLeftMargin(0.15) ; frame2->GetYaxis()->SetTitleOffset(1.8) ; frame2->Draw() ;
```

```
c->cd(3) ; gPad->SetLeftMargin(0.15) ; hh_data->GetXaxis()->SetTitleOffset(1.4) ; hh_data->Draw("lego") ;
```

```
c->cd(4) ; gPad->SetLeftMargin(0.20) ; hh_pdf->GetXaxis()->SetTitleOffset(2.4) ; hh_pdf->Draw("surf") ; hh_pdf2->Draw("surfsame") ;
```



rf501_simultaneouspdf.C





rf501_simultaneouspdf.C

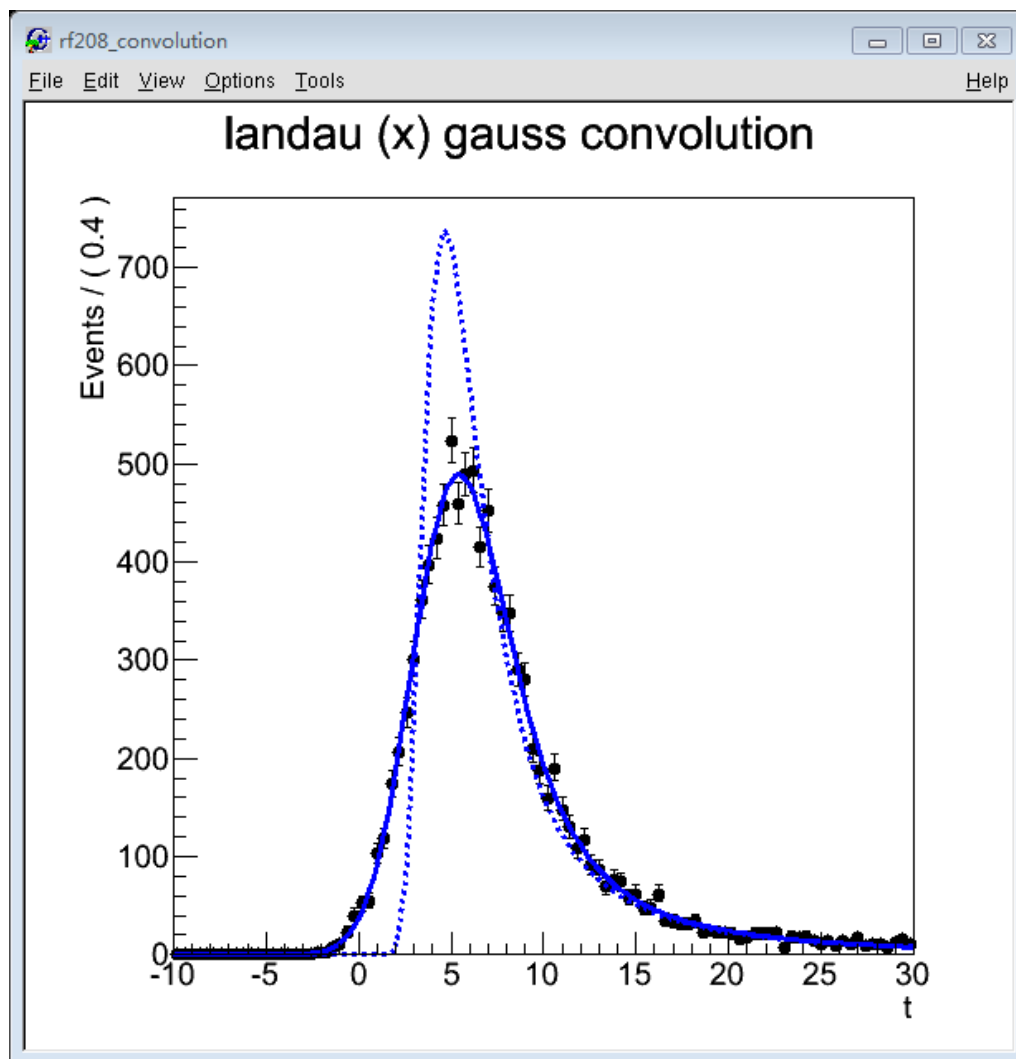
```
#ifndef __CINT__
#include "RooGlobalFunc.h"
#endif
#include "RooRealVar.h"
#include "RooDataSet.h"
#include "RooGaussian.h"
#include "RooConstVar.h"
#include "RooChebychev.h"
#include "RooAddPdf.h"
#include "RooSimultaneous.h"
#include "RooCategory.h"
#include "TCanvas.h"
#include "TAxis.h"
#include "RooPlot.h"
using namespace RooFit ;

void rf501_simultaneouspdf()
{
    // Create model for physics sample
    // -----

    // Create observables
    RooRealVar x("x", "x", -8, 8) ;
```



rf208_convolution.C





rf208_convolution.C

```
void rf208_convolution()
{
    // S e t u p   c o m p o n e n t   p d f s
    // -----

    // Construct observable
    RooRealVar t("t","t",-10,30) ;

    // Construct landau(t,ml,sl) ;
    RooRealVar ml("ml","mean landau",5.,-20,20) ;
    RooRealVar sl("sl","sigma landau",1,0.1,10) ;
    RooLandau landau("lx","lx",t,ml,sl) ;

    // Construct gauss(t,mg,sg)
    RooRealVar mg("mg","mg",0) ;
    RooRealVar sg("sg","sg",2,0.1,10) ;
    RooGaussian gauss("gauss","gauss",t,mg,sg) ;

    // Set #bins to be used for FFT sampling to 10000
    t.setBins(10000,"cache") ;

    // Construct landau (x) gauss
    RooFFTConvPdf lxxg("lxxg","landau (X) gauss",t,landau,gauss) ;
```




rf208_convolution.C

```
// Sample 1000 events in x from gxlx
RooDataSet* data = lxg.generate(t,10000) ;

// Fit gxlx to data
lxg.fitTo(*data) ;

// Plot data, landau pdf, landau (X) gauss pdf
RooPlot* frame = t.frame(Title("landau (x) gauss convolution")) ;
data->plotOn(frame) ;
lxg.plotOn(frame) ;
landau.plotOn(frame,LineStyle(kDashed)) ;

// Draw frame on canvas
new TCanvas("rf208_convolution","rf208_convolution",600,600) ;
gPad->SetLeftMargin(0.15) ; frame->GetYaxis()->SetTitleOffset(1.4) ; frame->Draw() ;
}
```



Install Root with FFTW3

- `tar -zxvf fftw-3.2.2.tar.gz`
- `cd fftw-3.2.2`
- `fftw-3.2.2 > ./configure --prefix=/ClusterDisks/HDN13/WorkSpace/testroot/root/fftw`
- `make -j7`
- `make install`

```
[hepfarm02] /ClusterDisks/HDN13/WorkSpace/testroot/root/fftw > ll  
total 32
```

```
drwxr-xr-x  2 testroot testroot 4096 Nov 19  2013 bin  
drwxr-xr-x  2 testroot testroot 4096 Nov 19  2013 include  
drwxr-xr-x  3 testroot testroot 4096 Nov 19  2013 lib  
drwxr-xr-x  4 testroot testroot 4096 Nov 19  2013 share
```



Install Root with FFTW3

```
#!/bin/bash
```

```
export ROOTSYS=/ClusterDisks/HDN13/WorkSpace/testroot/root/528
```

```
#if you have not un-zipped the file, remove the # at the head of next line
```

```
#tar -zxvf root_v5.28.00.source.tar.gz
```

```
cd root
```

```
./configure --enable-roofit --enable-fftw3 --with-fftw3-
```

```
incdir=/ClusterDisks/HDN13/WorkSpace/testroot/root/fftw/include --with-fftw3-
```

```
libdir=/ClusterDisks/HDN13/WorkSpace/testroot/root/fftw/lib
```

```
make -j7
```

```
make install
```

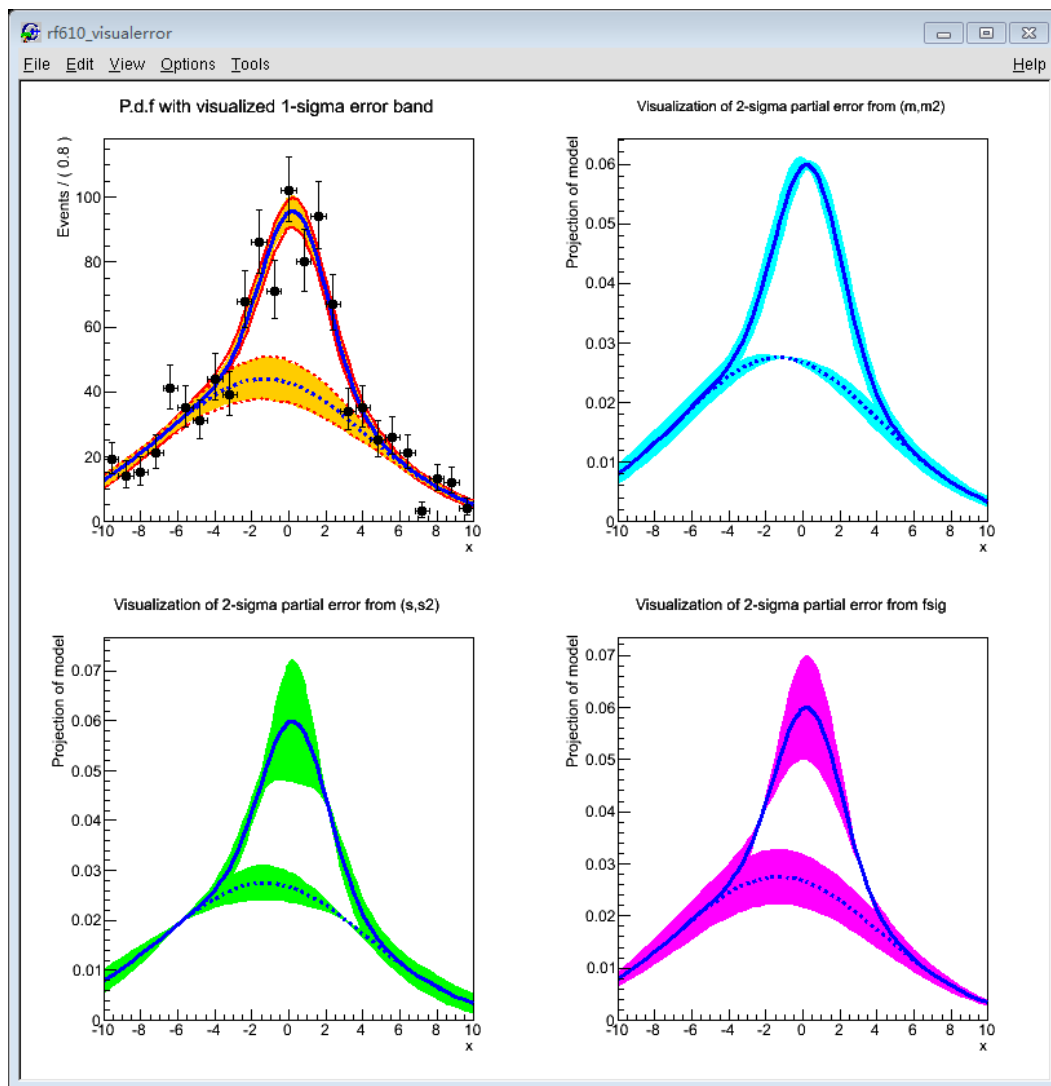


RooVoigtian

RooVoigtian(const char* name, const char*
title, RooAbsReal& _x, RooAbsReal&
_mean, RooAbsReal& _width, RooAbsReal&
_sigma, Bool_t doFast = kFALSE)



rf610_visualerror.C





rf610_visualerror.C

```
// Create sum of two Gaussians p.d.f. with factory
```

```
RooRealVar x("x","x",-10,10) ;
```

```
RooRealVar m("m","m",0,-10,10) ;
```

```
RooRealVar s("s","s",2,1,50) ;
```

```
RooGaussian sig("sig","sig",x,m,s) ;
```

```
RooRealVar m2("m2","m2",-1,-10,10) ;
```

```
RooRealVar s2("s2","s2",6,1,50) ;
```

```
RooGaussian bkg("bkg","bkg",x,m2,s2) ;
```

```
RooRealVar fsig("fsig","fsig",0.33,0,1) ;
```

```
RooAddPdf model("model","model",RooArgList(sig,bkg),fsig) ;
```

```
// Create binned dataset
```

```
x.setBins(25) ;
```

```
RooAbsData* d = model.generateBinned(x,1000) ;
```

```
// Perform fit and save fit result
```

```
RooFitResult* r = model.fitTo(*d,Save()) ;
```

```
// Make plot frame
```

```
RooPlot* frame = x.frame(Bins(40),Title("P.d.f with visualized 1-sigma error band")  
d->plotOn(frame) ;
```



rf610_visualerror.C

```
// Visualize 1-sigma error encoded in fit result 'r' as orange band using linear error propagation
// This results in an error band that is by construction symmetric
//
// The linear error is calculated as
//  $\text{error}(x) = Z * F\_a(x) * \text{Corr}(a,a') * F\_a'(x)$ 
//
// where  $F\_a(x) = [ f(x,a+da) - f(x,a-da) ] / 2$ ,
//
// with  $f(x)$  = the plotted curve
//      'da' = error taken from the fit result
//       $\text{Corr}(a,a')$  = the correlation matrix from the fit result
//      Z = requested significance 'Z sigma band'
//
// The linear method is fast (required  $2*N$  evaluations of the curve, where N is the number of parameters),
// but may not be accurate in the presence of strong correlations ( $\sim 0.9$ ) and at  $Z > 2$  due to linear and
// Gaussian approximations made
//
model.plotOn(frame, VisualizeError(*r, 1), FillColor(kOrange)) ;
```



rf610_visualerror.C

```
// Calculate error using sampling method and visualize as dashed red line.  
//  
// In this method a number of curves is calculated with variations of the parameter values, as sampled  
// from a multi-variate Gaussian p.d.f. that is constructed from the fit results covariance matrix.  
// The error(x) is determined by calculating a central interval that capture N% of the variations  
// for each value of x, where N% is controlled by Z (i.e. Z=1 gives N=68%). The number of sampling curves  
// is chosen to be such that at least 100 curves are expected to be outside the N% interval, and is minimally  
// 100 (e.g. Z=1->Ncurve=356, Z=2->Ncurve=2156)) Intervals from the sampling method can be asymmetric,  
// and may perform better in the presence of strong correlations, but may take (much) longer to calculate  
model.plotOn(frame,VisualizeError(*r,1,kFALSE),DrawOption("L"),LineWidth(2),LineColor(kRed)) ;
```




rf610_visualerror.C

```
// Perform the same type of error visualization on the background component only.  
// The VisualizeError() option can generally applied to _any_ kind of plot (components, asymmetries, efficiencies etc.  
.)  
model.plotOn(frame, VisualizeError(*r, 1), FillColor(kOrange), Components("bkg")) ;  
model.plotOn(frame, VisualizeError(*r, 1, kFALSE), DrawOption("L"), LineWidth(2), LineColor(kRed), Components("bkg"), LineStyle(kDashed)) ;  
  
// Overlay central value  
model.plotOn(frame) ;  
model.plotOn(frame, Components("bkg"), LineStyle(kDashed)) ;  
d->plotOn(frame) ;  
frame->SetMinimum(0) ;
```



rf610_visualerror.C

```
RooPlot* frame2 = x.frame(Bins(40),Title("Visualization of 2-sigma partial error from (m,m2)")) ;

// Visualize partial error. For partial error visualization the covariance matrix is first reduced as follows
//
// 
$$\overline{V_{red}} = V_{11} - V_{12} * V_{22}^{-1} * V_{21}$$

//
// Where V11,V12,V21,V22 represent a block decomposition of the covariance matrix into observables that
// are propagated (labeled by index '1') and that are not propagated (labeled by index '2'), and V22bar
// is the Shur complement of V22, calculated as shown above
//
// (Note that Vred is _not_ a simple sub-matrix of V)

// Propagate partial error due to shape parameters (m,m2) using linear and sampling method
model.plotOn(frame2,VisualizeError(*r,RooArgSet(m,m2),2),FillColor(kCyan)) ;
model.plotOn(frame2,Components("bkg"),VisualizeError(*r,RooArgSet(m,m2),2),FillColor(kCyan)) ;

model.plotOn(frame2) ;
model.plotOn(frame2,Components("bkg"),LineStyle(kDashed)) ;
frame2->SetMinimum(0) ;
```



rf610_visualerror.C

```
// Make plot frame
RooPlot* frame3 = x.frame(Bins(40),Title("Visualization of 2-sigma partial error from (s,s2)")) ;

// Propagate partial error due to yield parameter using linear and sampling method
model.plotOn(frame3,VisualizeError(*r,RooArgSet(s,s2),2),FillColor(kGreen)) ;
model.plotOn(frame3,Components("bkg"),VisualizeError(*r,RooArgSet(s,s2),2),FillColor(kGreen)) ;

model.plotOn(frame3) ;
model.plotOn(frame3,Components("bkg"),LineStyle(kDashed)) ;
frame3->SetMinimum(0) ;
```



rf610_visualerror.C

```
// Make plot frame
RooPlot* frame4 = x.frame(Bins(40),Title("Visualization of 2-sigma partial error from fsig")) ;

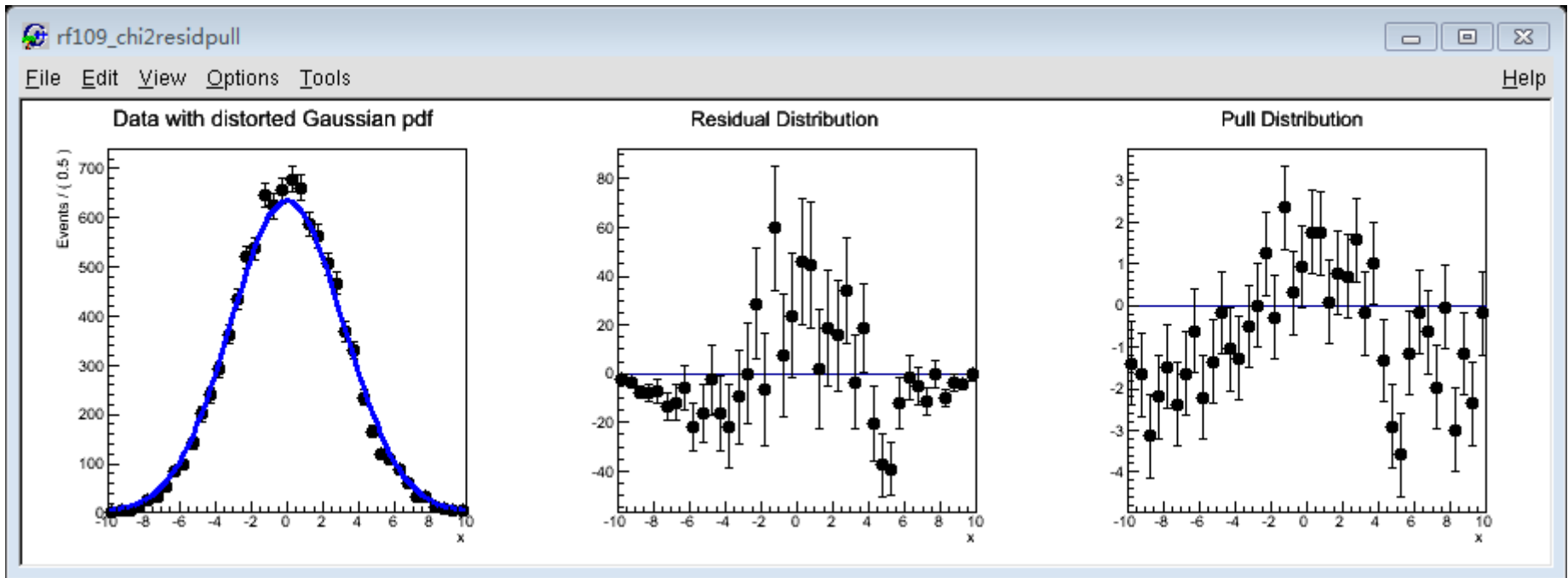
// Propagate partial error due to yield parameter using linear and sampling method
model.plotOn(frame4,VisualizeError(*r,RooArgSet(fsig),2),FillColor(kMagenta)) ;
model.plotOn(frame4,Components("bkg"),VisualizeError(*r,RooArgSet(fsig),2),FillColor(kMagenta)) ;

model.plotOn(frame4) ;
model.plotOn(frame4,Components("bkg"),LineStyle(kDashed)) ;
frame4->SetMinimum(0) ;

TCanvas* c = new TCanvas("rf610_visualerror","rf610_visualerror",800,800) ;
c->Divide(2,2) ;
c->cd(1) ; gPad->SetLeftMargin(0.15) ; frame->GetYaxis()->SetTitleOffset(1.4) ; frame->Draw() ;
c->cd(2) ; gPad->SetLeftMargin(0.15) ; frame2->GetYaxis()->SetTitleOffset(1.6) ; frame2->Draw() ;
c->cd(3) ; gPad->SetLeftMargin(0.15) ; frame3->GetYaxis()->SetTitleOffset(1.6) ; frame3->Draw() ;
c->cd(4) ; gPad->SetLeftMargin(0.15) ; frame4->GetYaxis()->SetTitleOffset(1.6) ; frame4->Draw() ;
}
```



rf109_chi2residpull.C



$$\text{pull}(N_{\text{sig}}) = \frac{N_{\text{sig}}^{\text{fit}} - N_{\text{sig}}^{\text{true}}}{\sigma_N^{\text{fit}}}$$

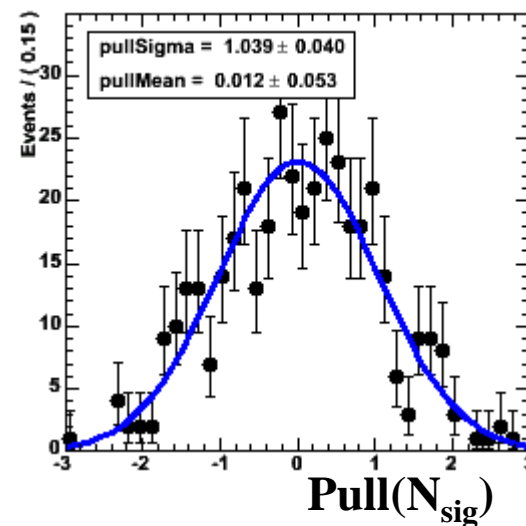


Pull分析

检查“Pull”分布,看误差是否对称

$$Pull(N_{sig}) = \frac{N_{sig}^{fit} - N_{sig}^{true}}{\sigma_{sig+bg}^{fit}}$$

- a) 如果拟合结果无偏则均值为0;
- b) 如果拟合误差正确则宽度为1。





```
// 'BASIC FUNCTIONALITY' RooFit tutorial macro #109
//
// Calculating  $\chi^2$  from histograms and curves in RooPlots,
// making histogram of residual and pull distributions
//
//
// 07/2008 - Wouter Verkerke
//
////////////////////////////////////

#ifdef __CINT__
#include "RooGlobalFunc.h"
#endif
#include "RooRealVar.h"
#include "RooDataSet.h"
#include "RooGaussian.h"
#include "RooConstVar.h"
#include "TCanvas.h"
#include "TAxis.h"
#include "RooPlot.h"
#include "RooHist.h"
using namespace RooFit ;
```



```
void rf109_chi2residpull()
{
    // s e t u p   m o d e l
    // -----

    // Create observables
    RooRealVar x("x", "x", -10, 10) ;

    // Create Gaussian
    RooRealVar sigma("sigma", "sigma", 3, 0.1, 10) ;
    RooRealVar mean("mean", "mean", 0, -10, 10) ;
    RooGaussian gauss("gauss", "gauss", x, RooConst(0), sigma) ;

    // Generate a sample of 1000 events with sigma=3
    RooDataSet* data = gauss.generate(x, 10000) ;

    // Change sigma to 3.15
    sigma=3.15 ;

    // P l o t   d a t a   a n d   s l i g h t l y   d i s t o r t e d   m o d e l
    // -----

    // Overlay projection of gauss with sigma=3.15 on data with sigma=3.0
    RooPlot* frame1 = x.frame(Title("Data with distorted Gaussian pdf"), Bins(40)) ;
    data->plotOn(frame1, DataError(RooAbsData::SumW2)) ;
    gauss.plotOn(frame1) ;
}
```




```
// Calculate chi^2
// -----

// Show the chi^2 of the curve w.r.t. the histogram
// If multiple curves or datasets live in the frame you can specify
// the name of the relevant curve and/or dataset in chiSquare()
cout << "chi^2 = " << frame1->chiSquare() << endl ;

// Show residual and pull dists
// -----

// Construct a histogram with the residuals of the data w.r.t. the curve
RooHist* hresid = frame1->residHist() ;

// Construct a histogram with the pulls of the data w.r.t the curve
RooHist* hpull = frame1->pullHist() ;

// Create a new frame to draw the residual distribution and add the distribution to the frame
RooPlot* frame2 = x.frame(Title("Residual Distribution")) ;
frame2->addPlotable(hresid,"P") ;

// Create a new frame to draw the pull distribution and add the distribution to the frame
RooPlot* frame3 = x.frame(Title("Pull Distribution")) ;
frame3->addPlotable(hpull,"P") ;

TCanvas* c = new TCanvas("rf109_chi2residpull","rf109_chi2residpull",900,300) ;
c->Divide(3) ;
c->cd(1) ; gPad->SetLeftMargin(0.15) ; frame1->GetYaxis()->SetTitleOffset(1.6) ; frame1->Draw() ;
c->cd(2) ; gPad->SetLeftMargin(0.15) ; frame2->GetYaxis()->SetTitleOffset(1.6) ; frame2->Draw() ;
c->cd(3) ; gPad->SetLeftMargin(0.15) ; frame3->GetYaxis()->SetTitleOffset(1.6) ; frame3->Draw() ;
}
```



To get the Value from Curve

```
RooHist* histo = (RooHist*) frame->findObject("data") ;  
RooCurve* func = (RooCurve*) frame->findObject("model") ;  
for (Int_t i=0 ; i<histo->GetN() ; i++) {  
    Double_t xdata,ydata;  
    histo->GetPoint(i,xdata,ydata) ;  
    Double_t yfunc = curve->interpolate(xdata) ;  
    // Use xdata,ydata,yfunc here  
}
```



SetSeed for RooFit Function



```
#include <RooRandom.h>
```

```
Int_t idx=1;
```

```
void gen_fitBK(){
```

```
    RooRandom::randomGenerator()->SetSeed(idx);
```

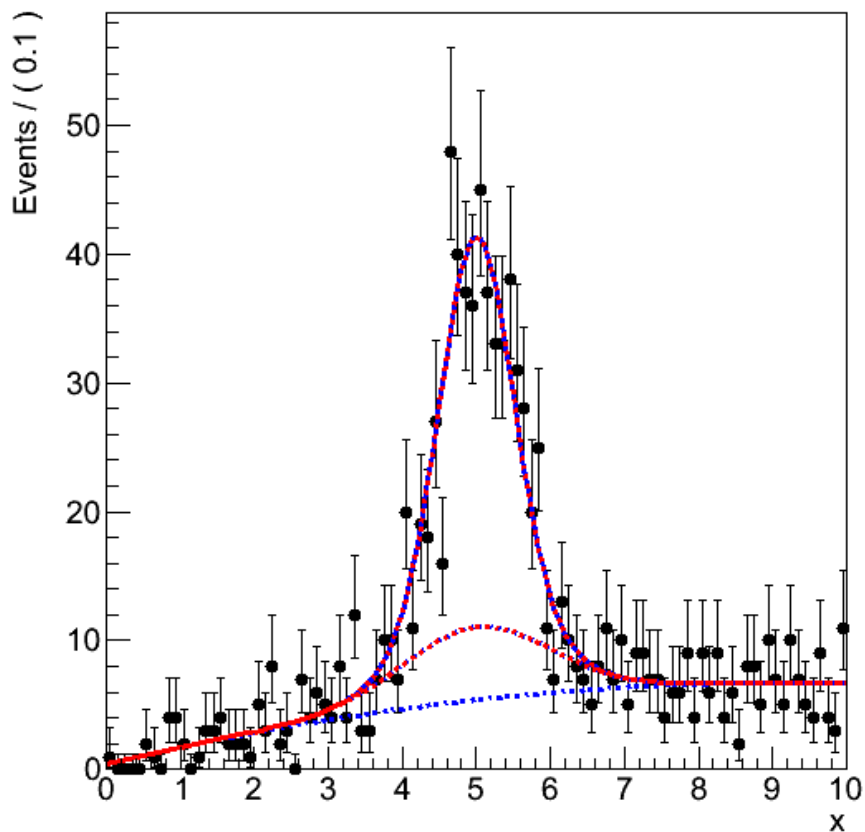
```
....
```



Useful examples

rf201_composite.C

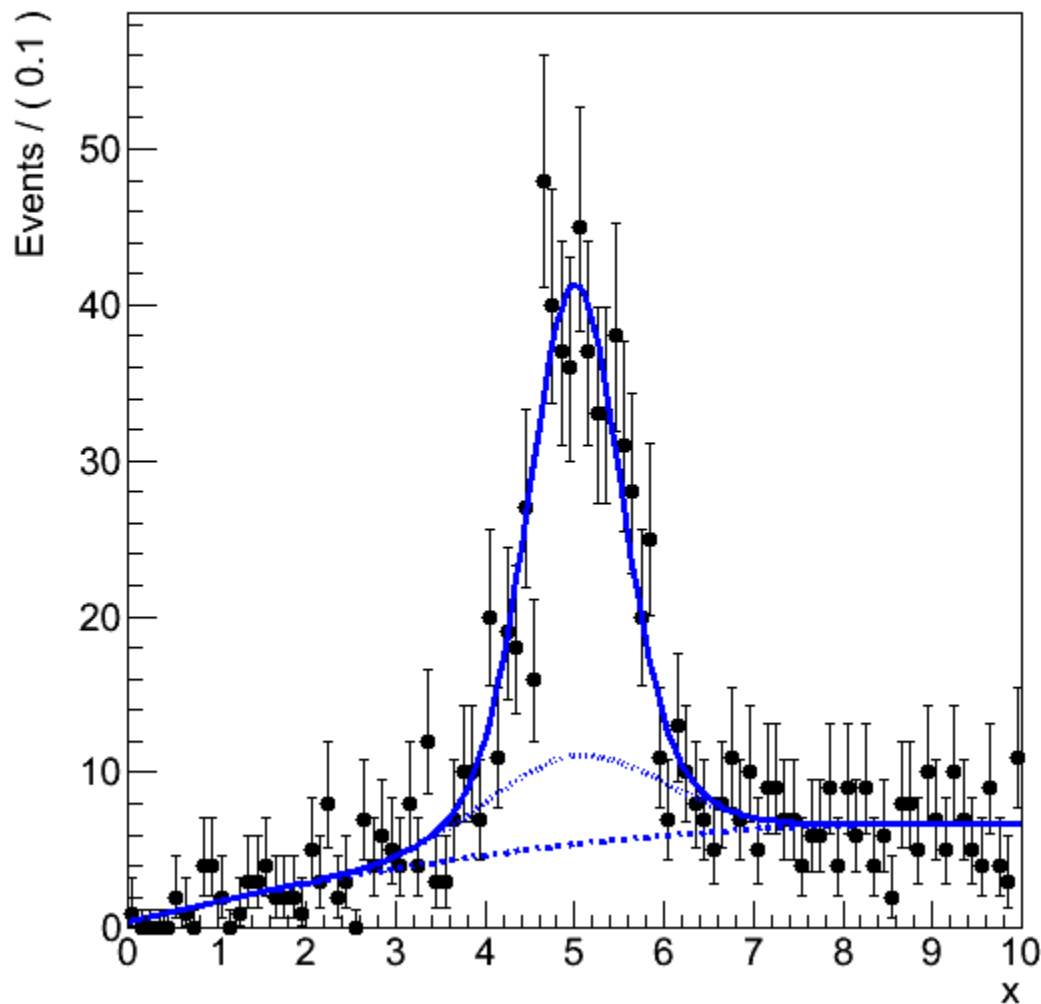
Example of composite pdf=(sig1+sig2)+bkg





root rf202_extendedmlfit.C

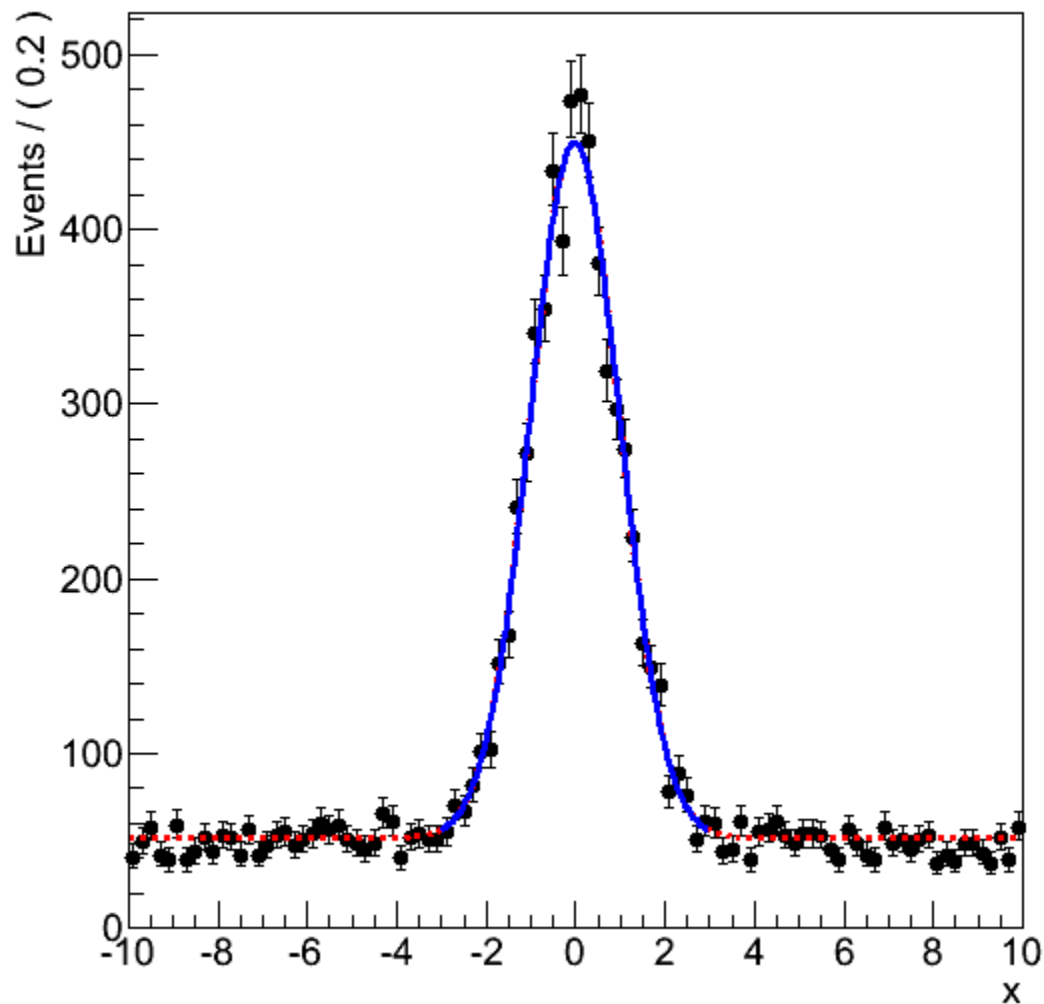
extended ML fit example





root rf203_ranges.C

Fitting a sub range





rf205_compplot.C

