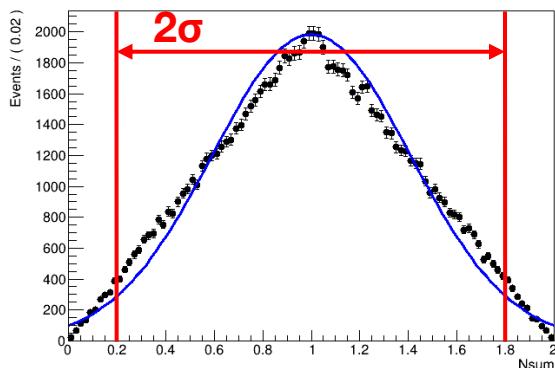


( Solutions - day 1 )

# Solution – Exercise 1

**N=2**

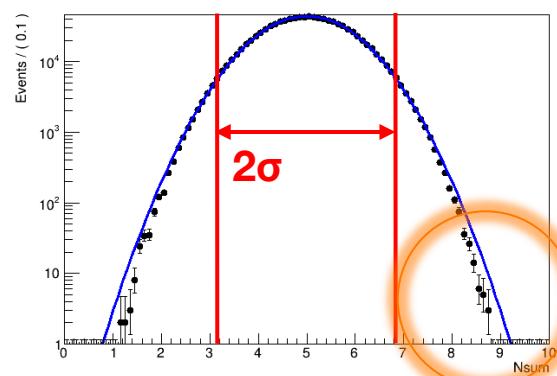
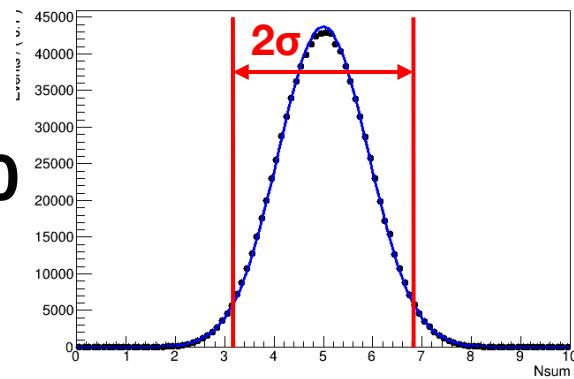


$$\begin{aligned} V(N=1) &= E[x^2] - E[x]^2 \\ &= \int_0^1 x^2 dx - \left( \int_0^1 x \right)^2 \\ &= \frac{1}{3} - \left( \frac{1}{2} \right)^2 = \frac{1}{12} \end{aligned}$$

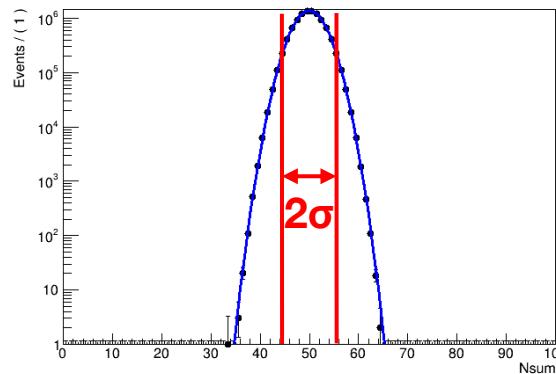
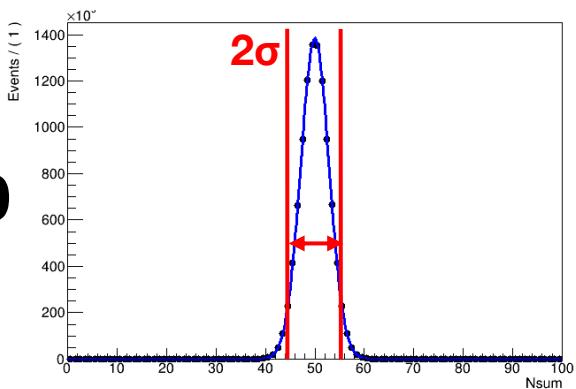


$$\begin{aligned} \sigma(N) &= \frac{\sigma(1)}{\sqrt{N}} N \\ &= \sqrt{\frac{N}{12}} \end{aligned}$$

**N=10**



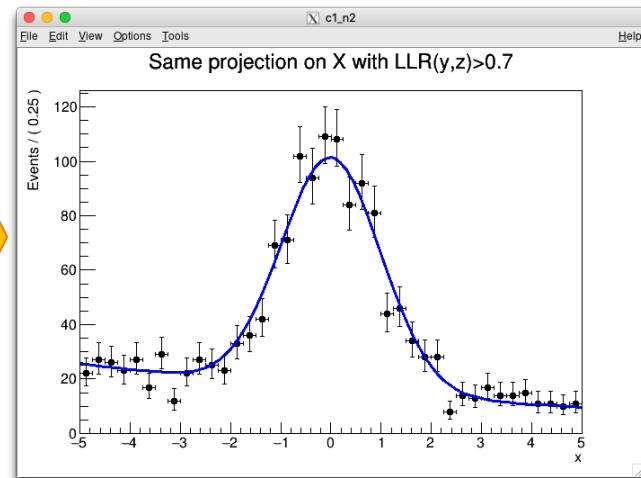
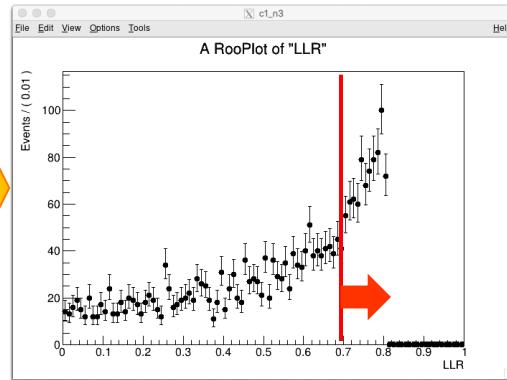
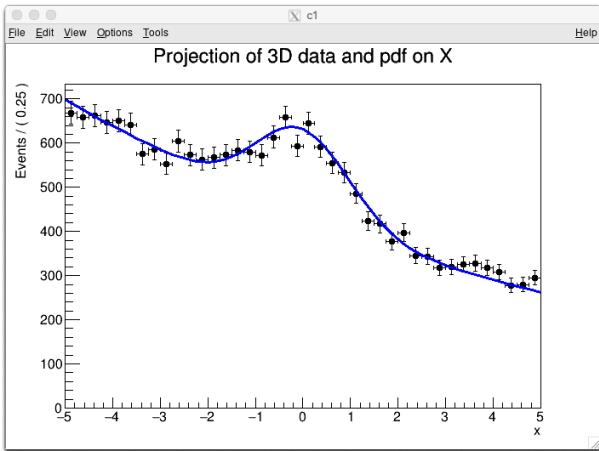
**N=100**



# Solution – Exercise 3

- Analytical application of Neyman-Pearson lemma

'per-event signal purity  
according to model(y,z)'



```
RooFitResult: minimized FCN value: 135474, estimated distance to minimum: 2.25574e-06
covariance matrix quality: Full, accurate covariance matrix
Status : MINIMIZE=0 HESSE=0
```

Floating Parameter	FinalValue +/- Error
fsig	9.7816e-02 +/- 2.78e-03

→  $35.2\sigma$

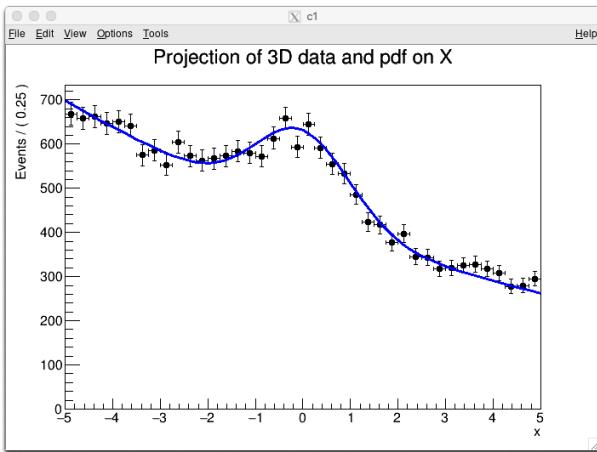
```
RooFitResult: minimized FCN value: 3096.86, estimated distance to minimum: 2.14652e-05
covariance matrix quality: Full, accurate covariance matrix
Status : MINIMIZE=0 HESSE=0
```

Floating Parameter	FinalValue +/- Error
fsig	5.6477e-01 +/- 1.92e-02

→  $29.4\sigma$

- Note design choice to cut *only* on model  $f(y,z)$  with NP-lemma
- Can use  $f(x)$  to *measure* remaining background

# Solution – Exercise 3



```
RooFitResult: minimized FCN value: 135474, estimated distance to minimum: 2.25574e-06
covariance matrix quality: Full, accurate covariance matrix
Status : MINIMIZE=0 HESSE=0
```

Floating Parameter	FinalValue +/- Error
fsig	9.7816e-02 +/- 2.78e-03

$\rightarrow 35.2\sigma$

```
RooFitResult: minimized FCN value: 3096.86, estimated distance to minimum: 2.14652e-05
covariance matrix quality: Full, accurate covariance matrix
Status : MINIMIZE=0 HESSE=0
```

Floating Parameter	FinalValue +/- Error
fsig	5.6477e-01 +/- 1.92e-02

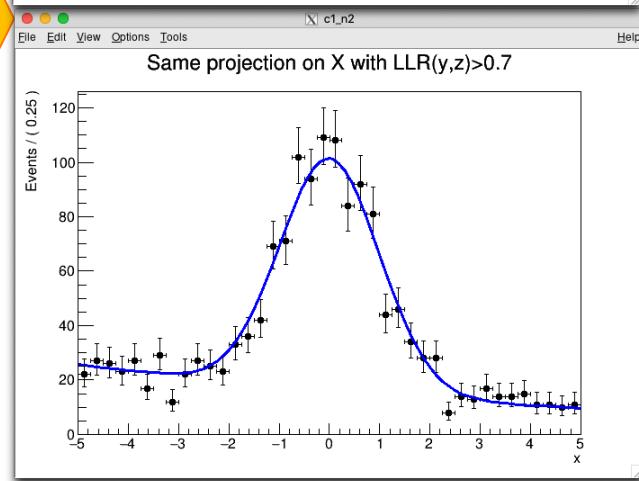
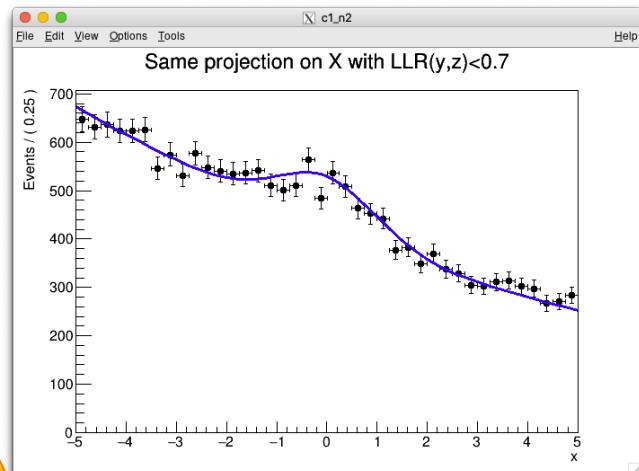
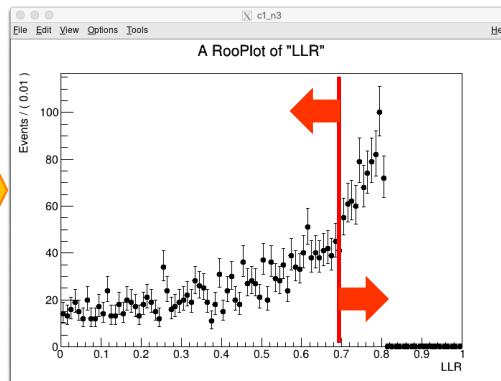
$\rightarrow 29.4\sigma$

```
RooFitResult: minimized FCN value: 41908.6, estimated distance to minimum: 5.35272e-05
covariance matrix quality: Full, accurate covariance matrix
Status : MINIMIZE=0 HESSE=0
```

Floating Parameter	FinalValue +/- Error
fsig	4.8966e-02 +/- 5.49e-03

$\rightarrow 8.9\sigma$

'per-event signal purity  
according to model(y,z)'



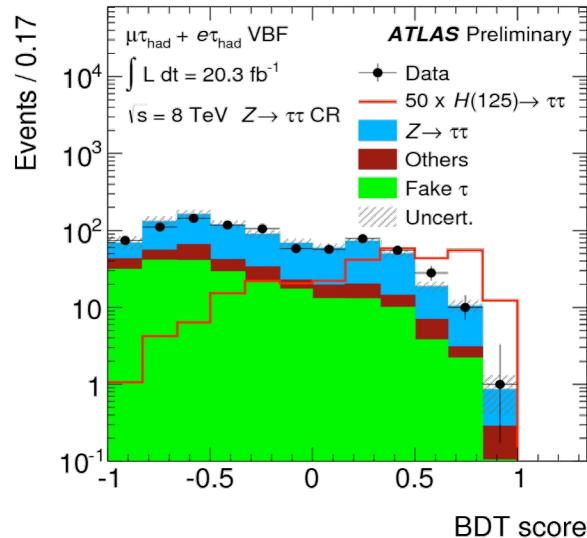
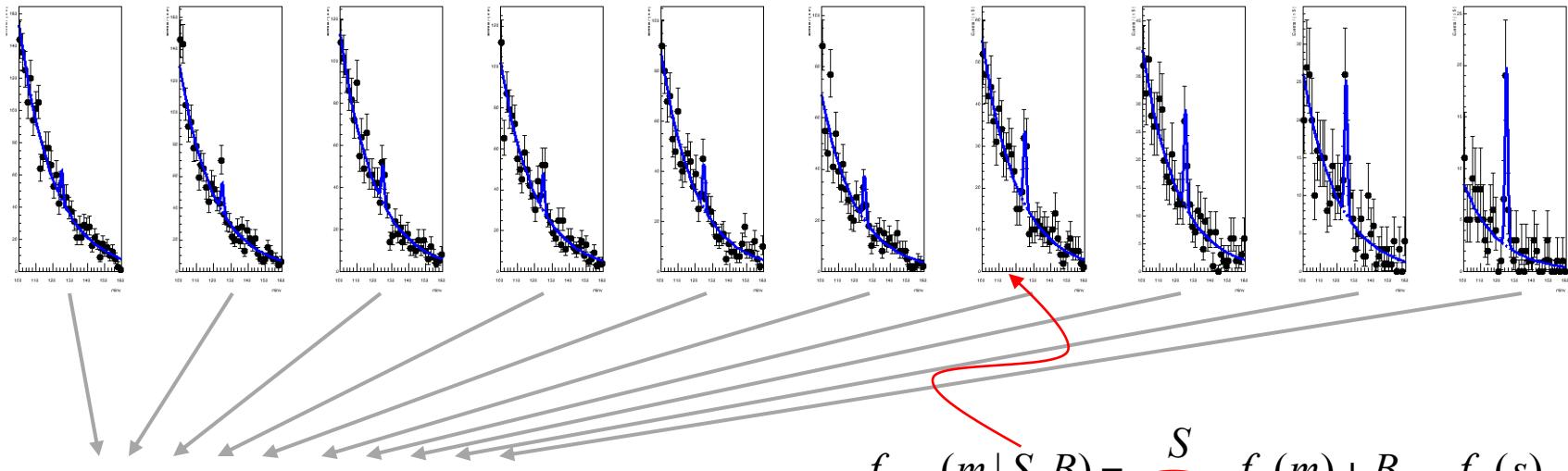
$30.7\sigma$

'not using prediction of  
event ratio over LLR cut'

NB: Feature!

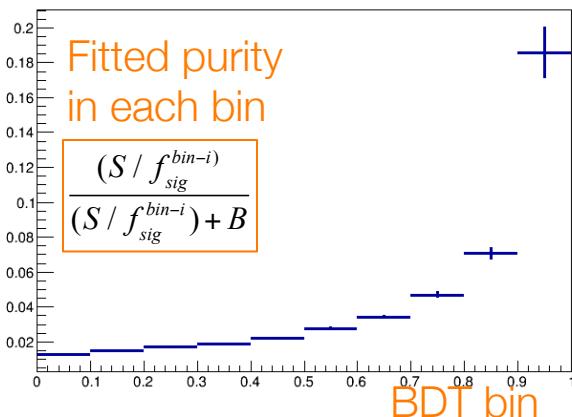
## (Preview to day 3) → Can also split in >>2 regions

- In that case only exploit variation in purity, not distribution of LLR/BDT  
 → Analysis insensitive to mismodelling of LLR/BDT distribution!



$$f_{bin-i}(m | S, B) = \frac{S}{f_{sig}^{bin-i}} f_S(m) + B_{bin-i} f_B(s)$$

Scale factor that ensures  
that every bin interprets  
S as the total signal yield



# (Preview to day 3) → Can also split in >>2 regions

- In that case only exploit variation in purity, not distribution of LLR/BDT  
 → Analysis insensitive to mismodelling of LLR/BDT distribution!

Joint PDF for this model

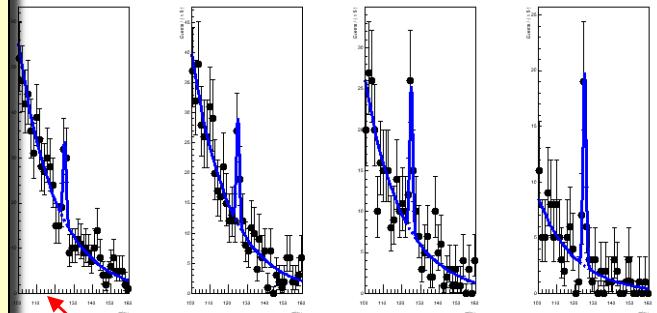
$$f(m, n_{BDT} | S, \vec{B}) = \text{lookup}(n_{BDT})$$

$$\left. \begin{aligned} f_{bin-0}(m | S, B_0) &= \frac{S}{f_{sig}^{bin-0}} f_S(m) + B_{bin-0} f_B(s) \\ f_{bin-1}(m | S, B_1) &= \frac{S}{f_{sig}^{bin-1}} f_S(m) + B_{bin-1} f_B(s) \\ f_{bin-2}(m | S, B_2) &= \frac{S}{f_{sig}^{bin-2}} f_S(m) + B_{bin-2} f_B(s) \\ f_{bin-3}(m | S, B_3) &= \frac{S}{f_{sig}^{bin-3}} f_S(m) + B_{bin-3} f_B(s) \\ &\vdots \\ f_{bin-N}(m | S, B_N) &= \frac{S}{f_{sig}^{bin-N}} f_S(m) + B_{bin-N} f_B(s) \end{aligned} \right\}$$

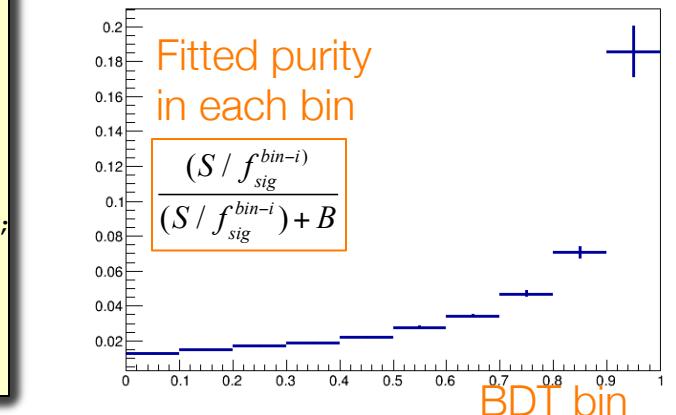
```
// Construct template model
w.factory("SUM::fit_template(prod(Nsig[30,0,100],frac[1])*sig1,
                           Nbkg[1000,0,10000]*bkg1)") ;

// Construct joint model from template clones
w.factory("SIMCLONE::fitmodel(fit_template,
                             $SplitParam({Nbkg,frac},bdtBin))") ;
```

BDT score

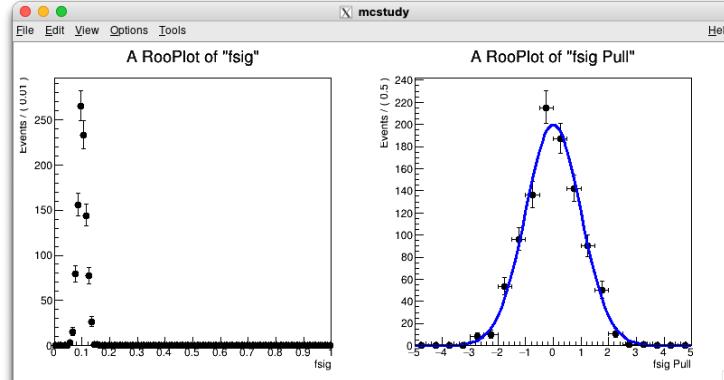
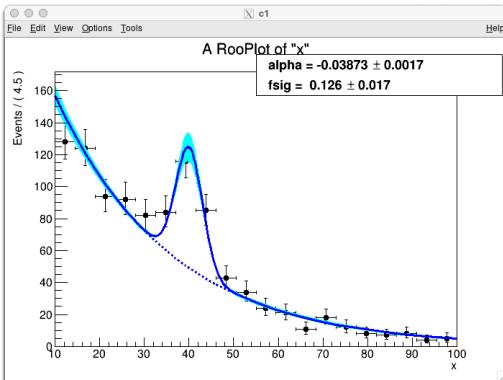


$$f_i(m | S, B) = \frac{S}{f_{sig}^{bin-i}} f_S(m) + B_{bin-i} f_B(s)$$

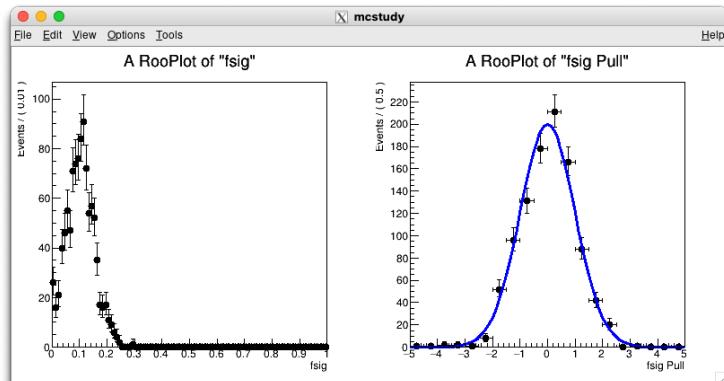
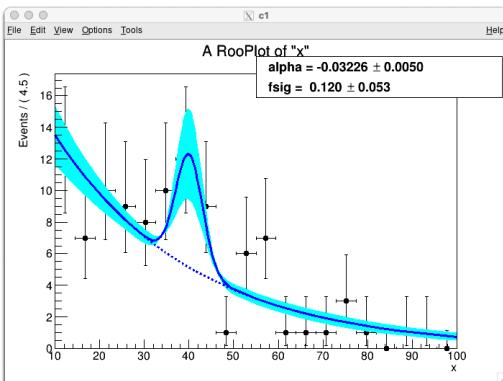


# Solution – Exercise 5

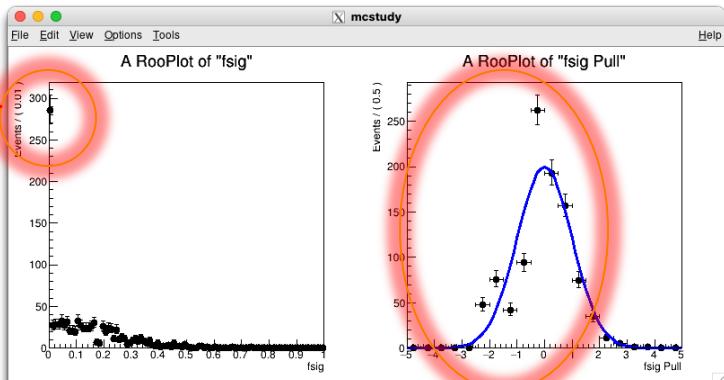
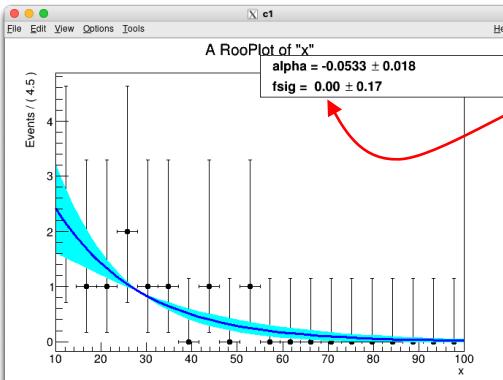
**N=1000**



**N=100**

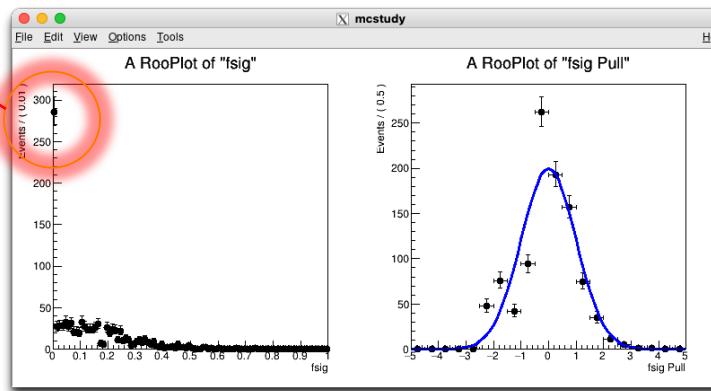
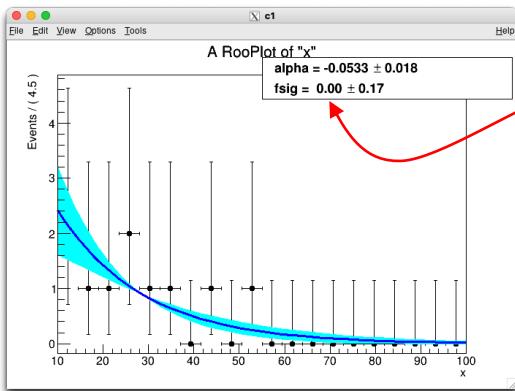


**N=10**



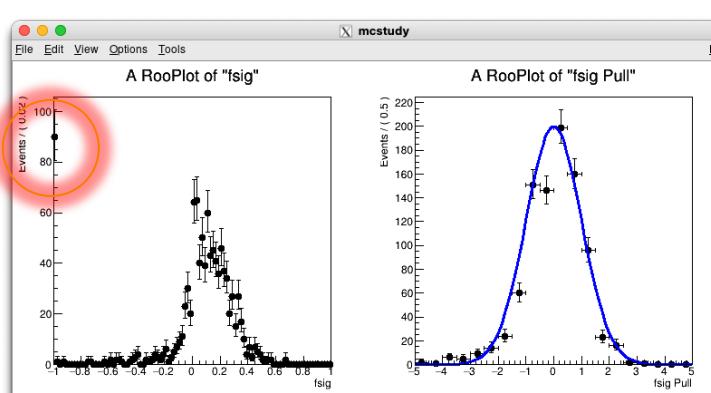
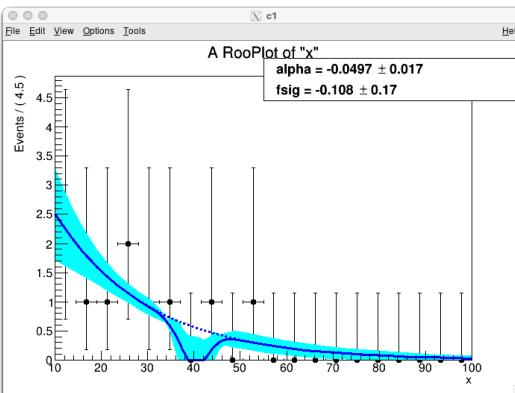
# Solution – Exercise 5

N=10



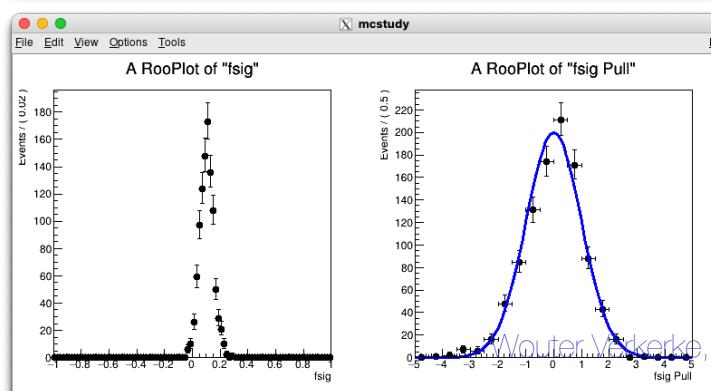
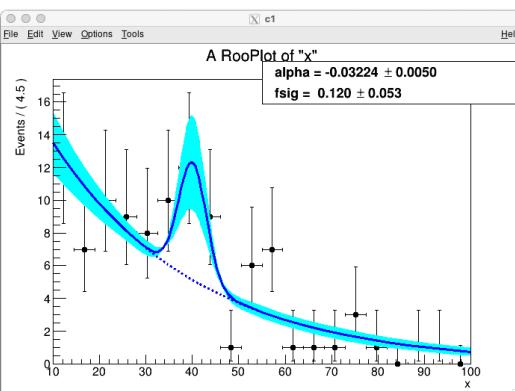
fsig[-1,1]

N=10



fsig[-1,1]

N=100



OuterVakkerke, NIKHEF

# Solution – Exercise 5

$f_{\text{sig}}[-1, 1]$

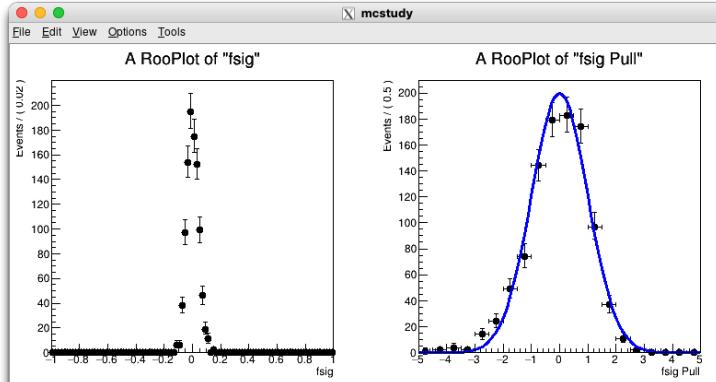
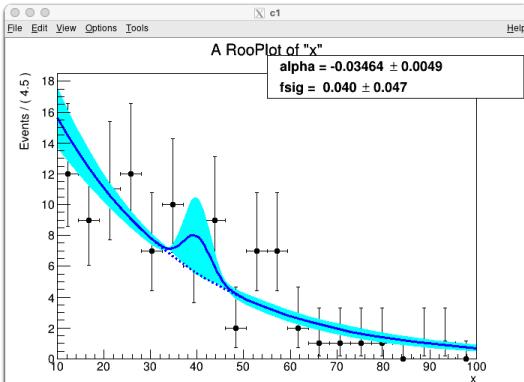
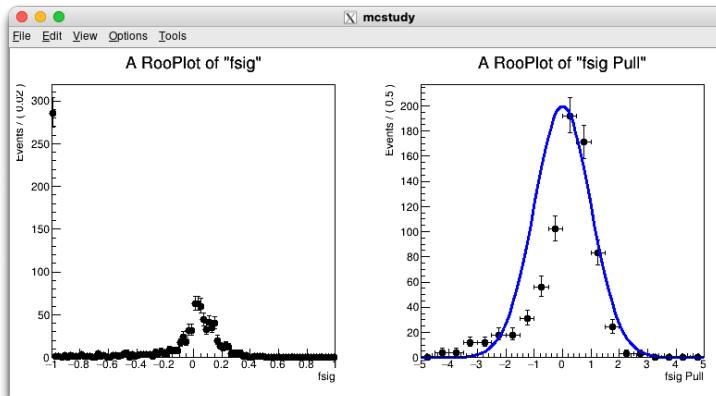
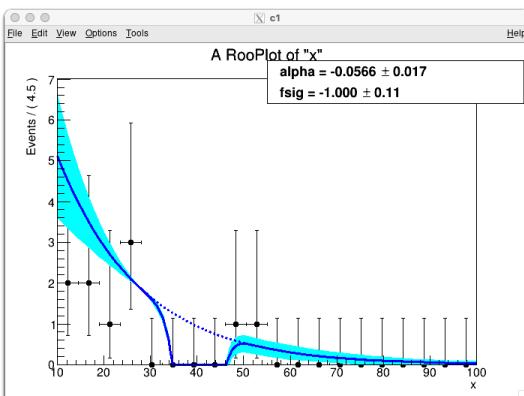
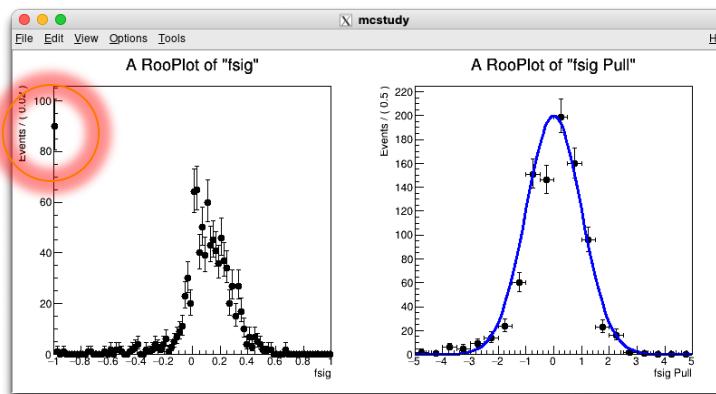
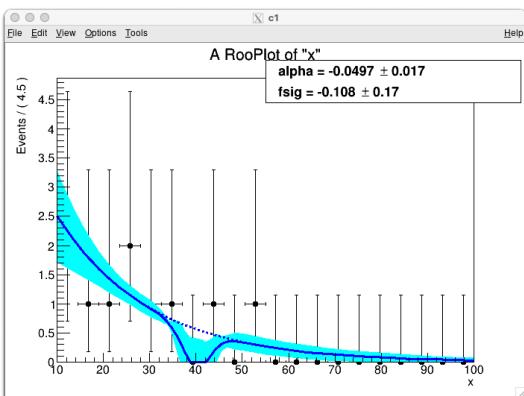
**N=10**

$f_{\text{sig}}=0$

**N=10**

$f_{\text{sig}}=0$

**N=100**

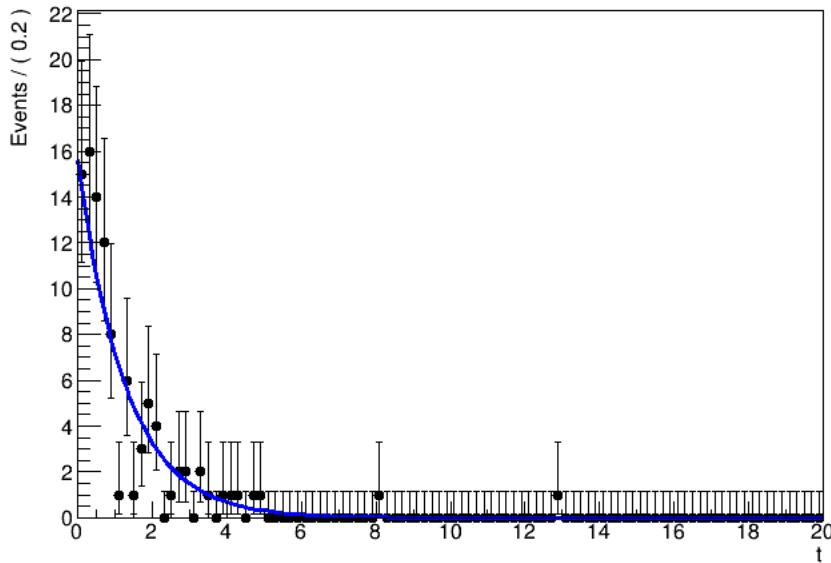


## Solution – Exercise 6

- Normalized Exponential p.d.f  $f(x) = \tau e^{-\tau x}$
- Negative log-Likelihood (1 event) 
$$\begin{aligned} -\log(f(x)) &= -\log(\tau) - \log(e^{-\tau x}) \\ &= -\log(\tau) + \tau x \end{aligned}$$
- Negative log-Likelihood (N events) 
$$\begin{aligned} -\log(L) &= \sum_{i=1}^N -\log(\tau) + \tau x_i \\ &= -N \log(\tau) + \tau \sum_{i=1}^n x_i \end{aligned}$$
- ML estimator for N events  $0 = \frac{d \log L}{d \tau}$  
$$\begin{aligned} 0 &= -N \frac{1}{\tau} + \sum_{i=1}^n x_i \\ \frac{N}{\tau} &= \sum_{i=1}^n x_i \quad \Rightarrow \quad \frac{1}{\tau} = \frac{\sum_{i=1}^n x_i}{N} = \langle x \rangle \end{aligned}$$

# Solution – Exercise 6

- Analytical vs numeric calculation



```
COVARIANCE MATRIX CALCULATED SUCCESSFULLY
FCN=124.735 FROM HESSE      STATUS=OK          5 CALLS      23 TOTAL
EDM=7.97722e-09   STRATEGY= 1    ERROR MATRIX ACCURATE
EXT PARAMETER                INTERNAL      INTERNAL
NO. NAME        VALUE       ERROR      STEP SIZE     VALUE
 1 tau         1.28062e+00 1.28048e-01 4.17606e-05 -1.07870e+00
                           ERR DEF= 0.5
EXTERNAL ERROR MATRIX.  NDIM= 25   NPAR= 1   ERR DEF=0.5
1.640e-02
```

$$\frac{1}{\tau} = \frac{\sum_{i=1}^n x_i}{N} = \langle x \rangle \quad \rightarrow \quad \tau = 1.28063$$

Agreement MINUIT/analytical within  $\sim 10^{-5}$  relative precision  
Default MINUIT numerical precision (EDM  $10^{-3}$ )  $\sim 4\%$  of error  $\rightarrow 5 \times 10^{-3}$