

Working with Collie: A Confidence Level Limit Evaluator

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Outline

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- Collie inputs and Effects from Systematics
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What is Collie?

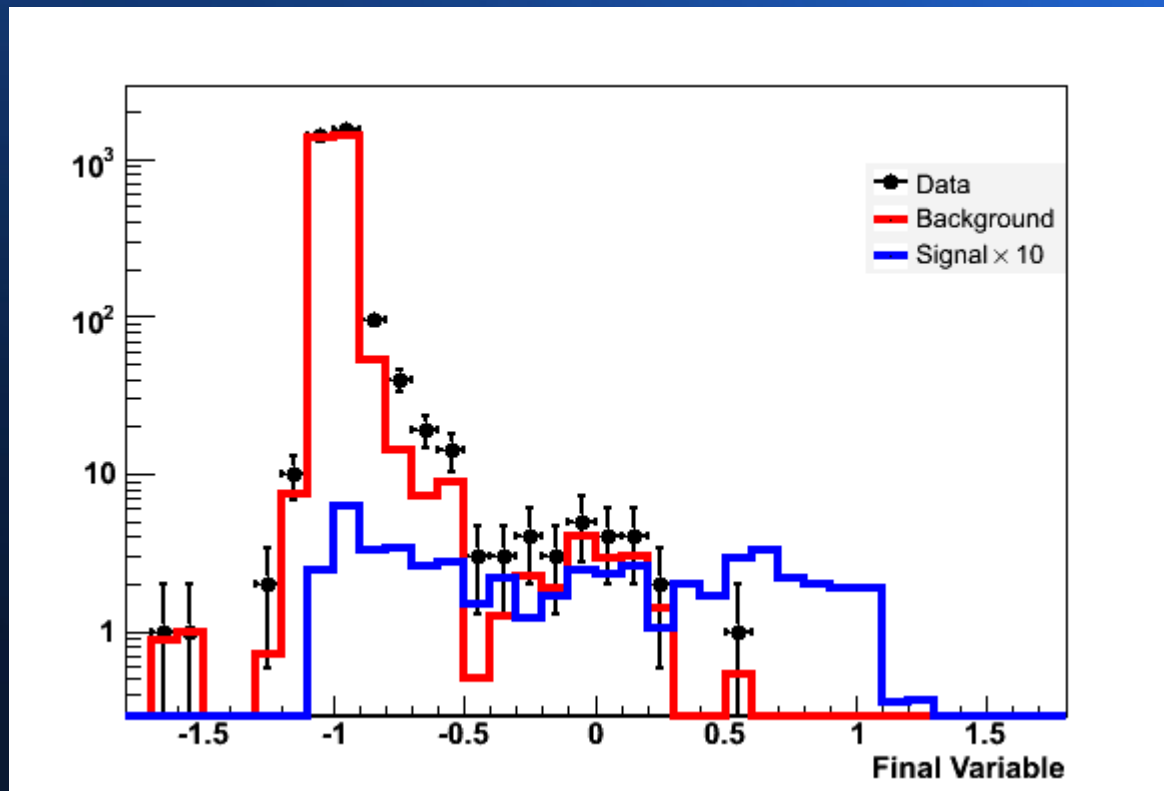
- Collie is a software suite designed to generate Confidence Levels (CL) which is used to calculate upper limits for TEST and NULL hypothesis.
 - TEST hypothesis: Signal + background hypothesis, i.e. new Physics.
 - NULL hypothesis: background only.
- This tool tries to separate TEST and NULL hypothesis in a way to get observed and expected limits with $\pm 1 / 2$ sigmas.
- χ^2 fits are applied to compare data/MC considering:
 - Gaussian statistics
 - Poisson likelihood statistics
- Limits are calculated for the minimum χ^2 value

Collie inputs and Effects from Systematics

- High Energy Physics search analysis often has a final variable chosen to be sensitive to a parameter of the search system.
- This final variable is composed by signal, background and data distributions.
- Measurements are affected by nuisance parameters.
 - Nuisance parameters are not parameters of interest, but somehow affect measurements.
 - Example of nuisance parameters: luminosity, efficiencies, background cross sections.
- Nuisance parameters have systematic uncertainties which can degrade the sensitive of the final variable.
- So Collie tries to reduce the effects from systematics.

Example of Collie input

- Final variable from TMVA analysis



Creating a Collie IO file

- `int main(int argc, char* argv[]) {`
- `CollieIOFile* cfile = new CollieIOFile();`
`//outputfile and channel name`
- `cfile->initFile("exampleCollieIOfile.root", "Test Channel");`
- `cfile->setInputHist(0.0,1.0,20);` `//Define your input histograms`
- `cfile->setSmooth(false);` `//Option to smooth histograms`
- `//Define backgrounds`
- `vector<string> bkgdNames;`
- `bkgdNames.push_back("Bkgd1");`
- `bkgdNames.push_back("Bkgd2");`
- `cfile-> createChannel(bkgdNames);`

...creating collie IO file

```
// Get inputs from a file:
```

- `TFile infile("myInputFile.root");`
- `TH1D* data = (TH1D*)infile.Get("data");`
- `TH1D* signal = (TH1D*)infile.Get("signal");`
- `TH1D* bkgd1 = (TH1D*)infile.Get("bkgd1");`
- `//Backgrounds are passed in via vector`
- `vector<TH1D*> vbkgd;`
- `vbkgd.push_back(bkgd1);`
- `vbkgd.push_back(bkgd2);`
- `// Mass points are passed here`
- `cfile->createMassPoint(100, data, sig, 1,vbkgd,valpha);`

...creating IO files

```
//Add systematics...either flat or by shape (ie, function of final variable)
```

```
// if by shape, must supply a histogram of the values in percent(%)  
fluctuations...
```

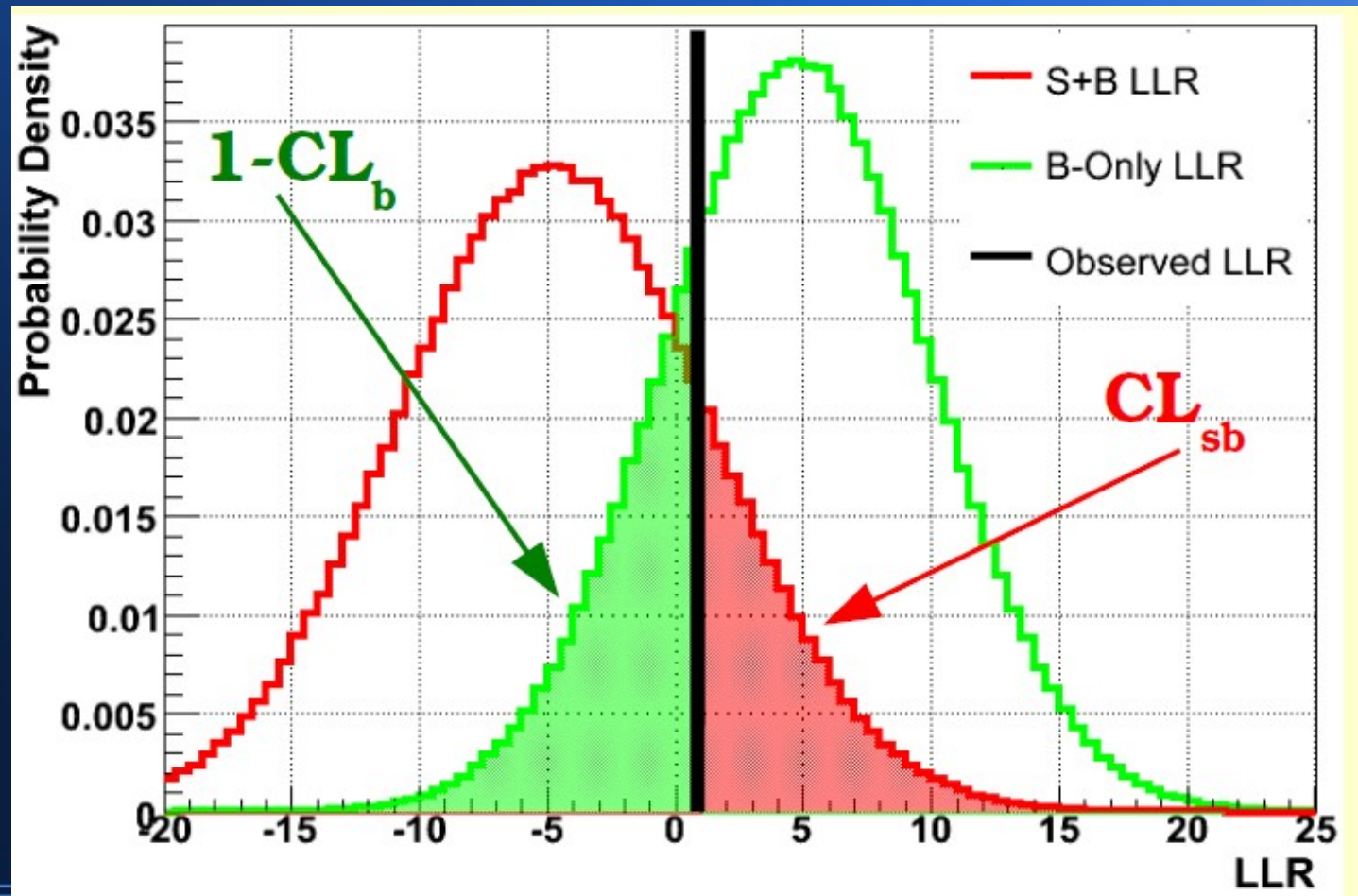
```
// Signal requires no index, but backgrounds must be specifically indexed  
(0>N bkgds)
```

- `cfile->createFlatSigSystematic("Lumi",0.06,0.06,100);`
- `cfile->createShapeSigSystematic("SigShape",sigSystP,sigSystN,100);`
- `cfile->createFlatBkgdSystematic(0,"Lumi",0.06,0.06,100);`
- `cfile->createFlatBkgdSystematic(1,"Lumi",0.06,0.06,100);`
- `///
store channel`
- `cfile->storeFile();`
- `}`

Log Likelihood Ratio distributions

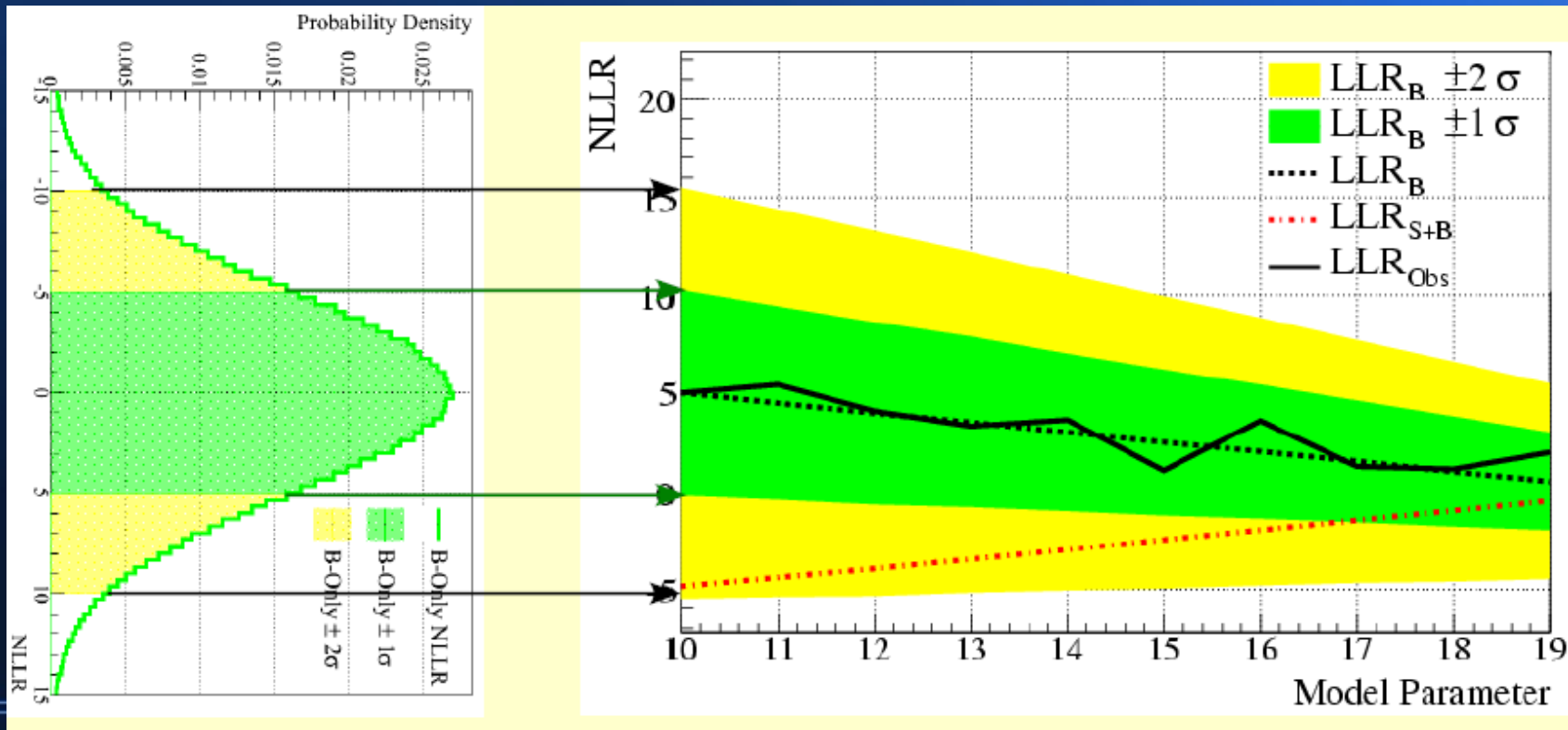
- $1 - CL_s = 1 - \alpha = \text{confidence level}$

$$CL_s = \frac{CL_{s+b}}{CL_b}$$



Confidence level calculations

- Confidence level can be evaluated by 4 different kinds of tools:
 - CLfast: fast tool, but ignores systematics
 - CLsyst: standard gaussian including systematics
 - CLfit: single likelihood fit, maximized over systematics
 - CLfit2: double likelihood fit, maximized over systematics



χ^2 formulas

- Gaussian statistics

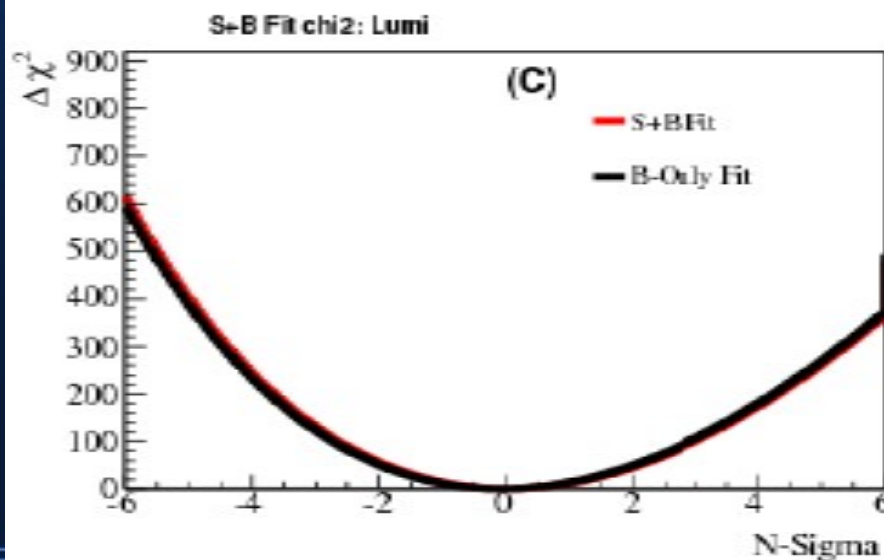
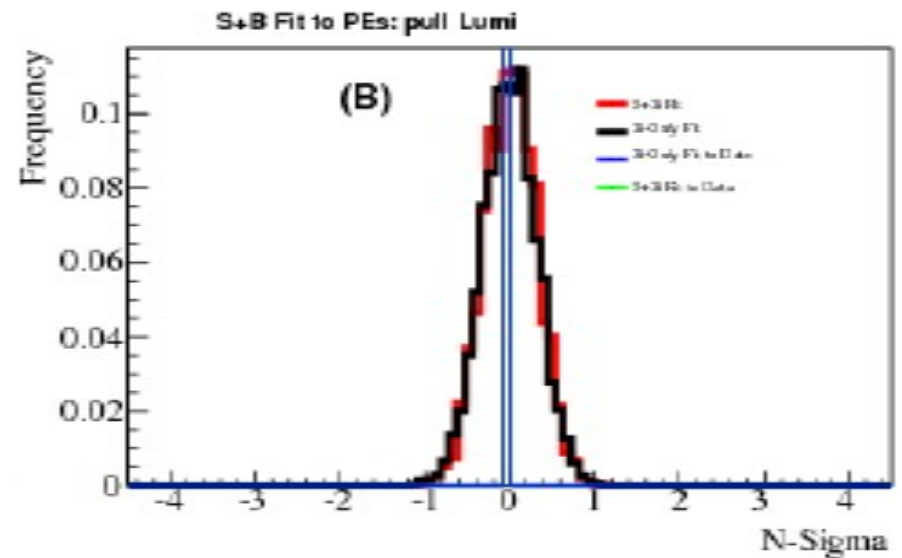
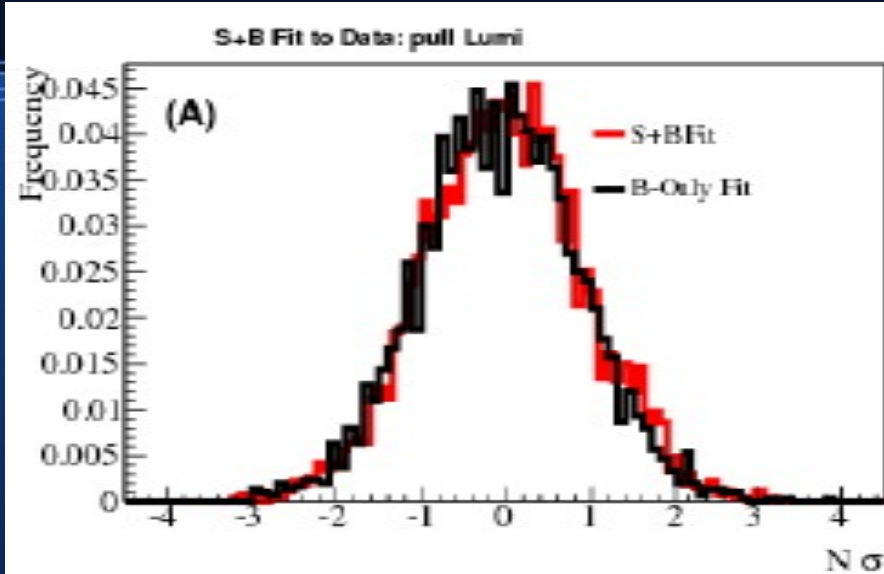
$$\chi^2 = \sum_i^{N^{\text{bins}}} \frac{(d_i - p_i)^2}{\sigma_i^2} = -2 \ln \left(\frac{\mathcal{L}}{\mathcal{L}_0} \right)$$
$$\mathcal{L} = \prod_i^{N^{\text{bins}}} \left(\frac{1}{\sigma_i \sqrt{2\pi}} \right) \exp \left(-\frac{(d_i - p_i)^2}{2\sigma_i^2} \right)$$
$$\mathcal{L}_0 = \prod_i^{N^{\text{bins}}} \left(\frac{1}{\sigma_i \sqrt{2\pi}} \right)$$

- Poisson statistics

$$\mathcal{L}^P = \prod_i^{N^{\text{bins}}} \frac{p_i^{d_i} \exp^{-p_i}}{d_i!}$$
$$\mathcal{L}_0^P = \prod_i^{N^{\text{bins}}} \frac{d_i^{d_i} \exp^{-d_i}}{d_i!}$$

$$\chi^2 = -2 \ln \left(\frac{\mathcal{L}^P}{\mathcal{L}_0^P} \right)$$
$$= -2 \ln \prod_i^{N^{\text{bins}}} \left(\frac{p_i}{d_i} \right)^{d_i} \exp^{(d_i - p_i)}$$
$$= 2 \sum_i^{N^{\text{bins}}} (p_i - d_i) - d_i \ln \left(\frac{p_i}{d_i} \right)$$

χ^2 fit results



[A] Smeared B-Only (S+B) Fit to Data

Mean: -0.077 (-0.034)

Width: 0.909 (0.932)

[B] B-Only (S+B) Fit to PseudoData

Mean: -0.008 (0.005)

Width: 0.319 (0.311)

[C] B-Only (S+B) χ^2 Response Function

Minimum: -0.09 (-0.01)

+1 σ : 0.280 (0.280)

+1 σ : 0.300 (0.280)

Matrix correlation between systematics

