

Simple Analysis Optimization

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26 January, 2006

Abstract: A frequently asked question concerning both measurements and searches in particle physics is how to optimize the event selection and the analysis of the selected events. It is well-known by now that using likelihoods for measurements (parameter estimation) or searches (hypothesis testing) increases the sensitivity of the analysis. What is not so well-known is how, apart from brute-force methods, to optimize a measurement or search except in the case of a simple counting experiment where the criteria S/\sqrt{B} and $S/\sqrt{S+B}$ are often used. A derivation will be shown of a general formula (but still fairly simple) for binned pdf's which reproduces the criteria given above for the limiting case of 1 bin (a counting experiment).

Outline

- Event selection
- Likelihood and likelihood ratio
- Significance and power
- Optimization
- Summary

Selecting events

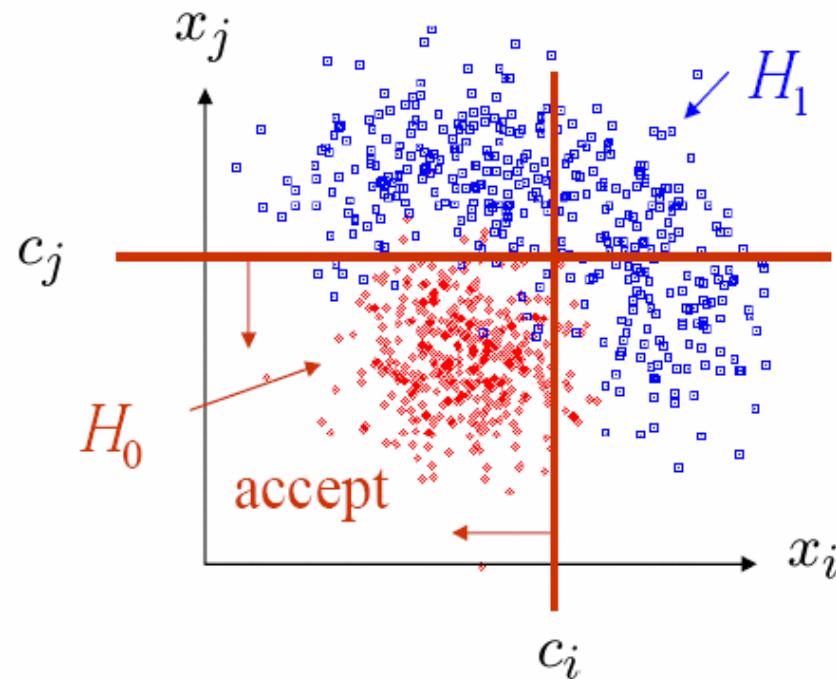
Suppose we have a data sample with two kinds of events, corresponding to hypotheses H_0 and H_1 and we want to select those of type H_0 .

Each event is a point in \vec{x} space. What decision boundary should we use to accept/reject events as belonging to event type H_0 ?

Probably start with cuts:

$$x_i < c_i$$

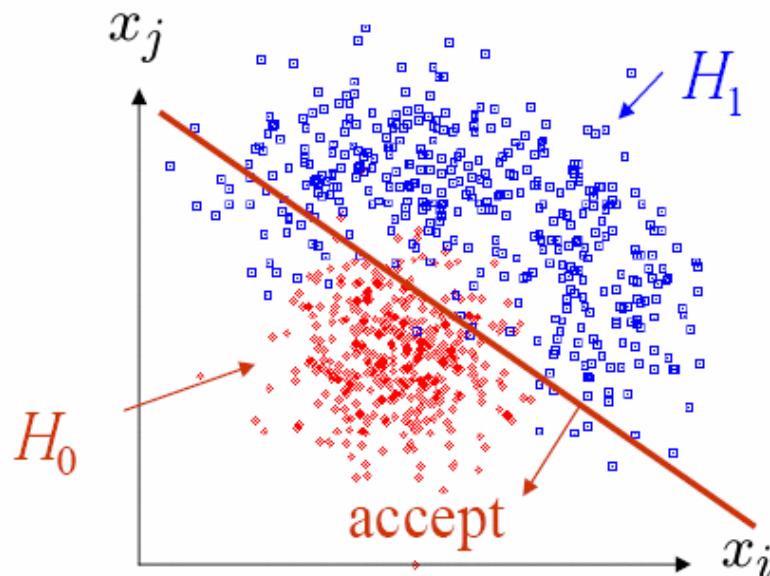
$$x_j < c_j$$



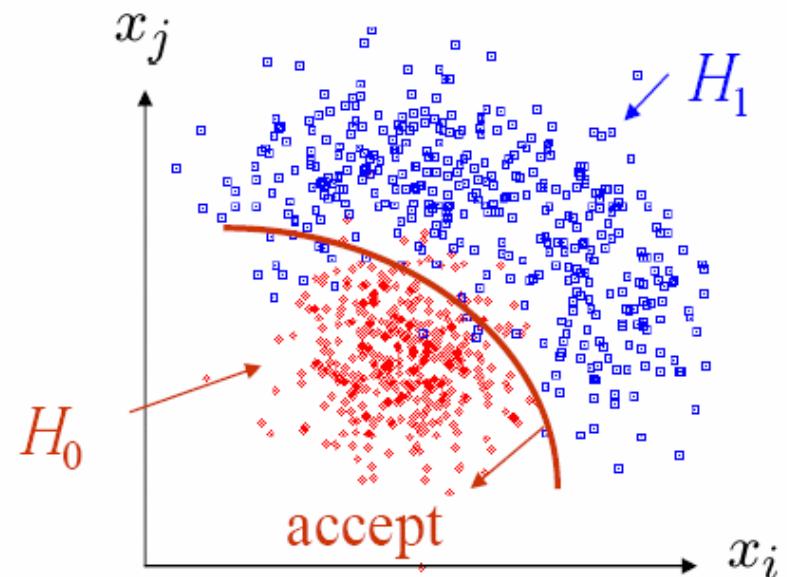
Other ways to select events

Or maybe use some other sort of decision boundary:

linear



or nonlinear



How can we do this in an 'optimal' way?

Likelihood ratio

$$Q = L(H_0)/L(H_1)$$

**or approximation in
case of complexity**

page 7

Neyman-Pearson lemma

"Exclude"

Perturb
contour

X_{s+b}

"Discover"

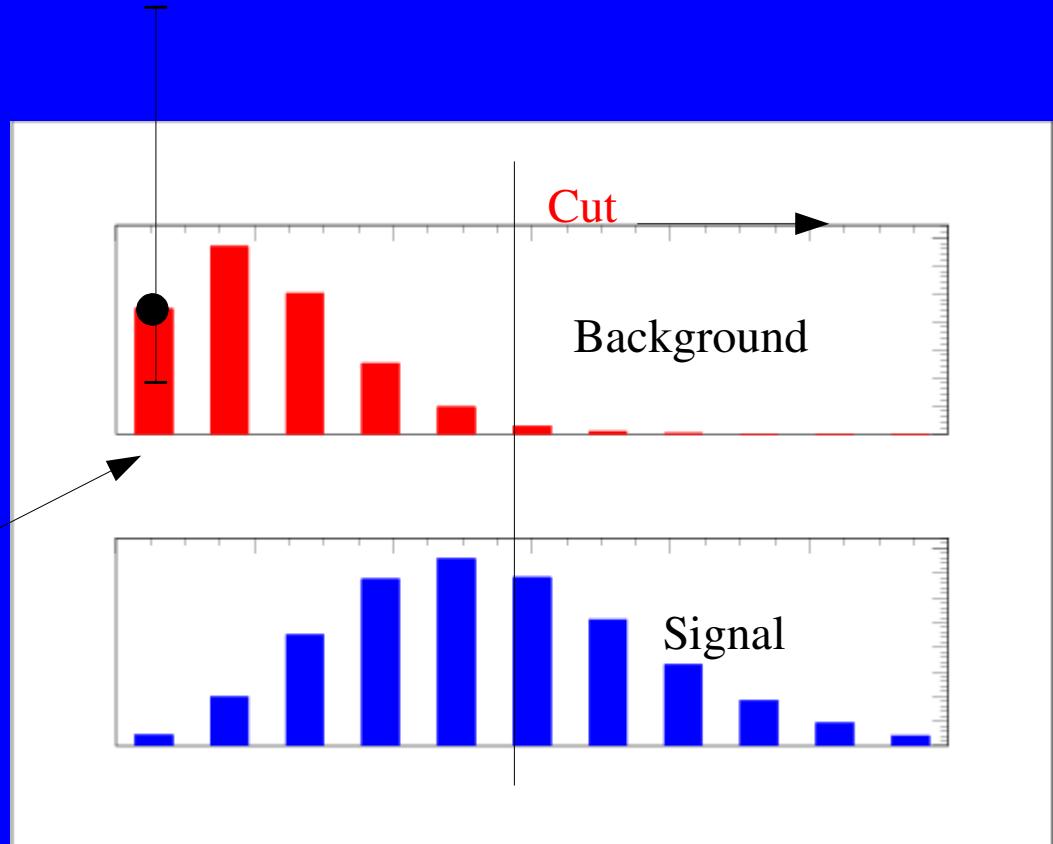
$Q > Q_0$

$Q = Q_0$

$Q < Q_0$

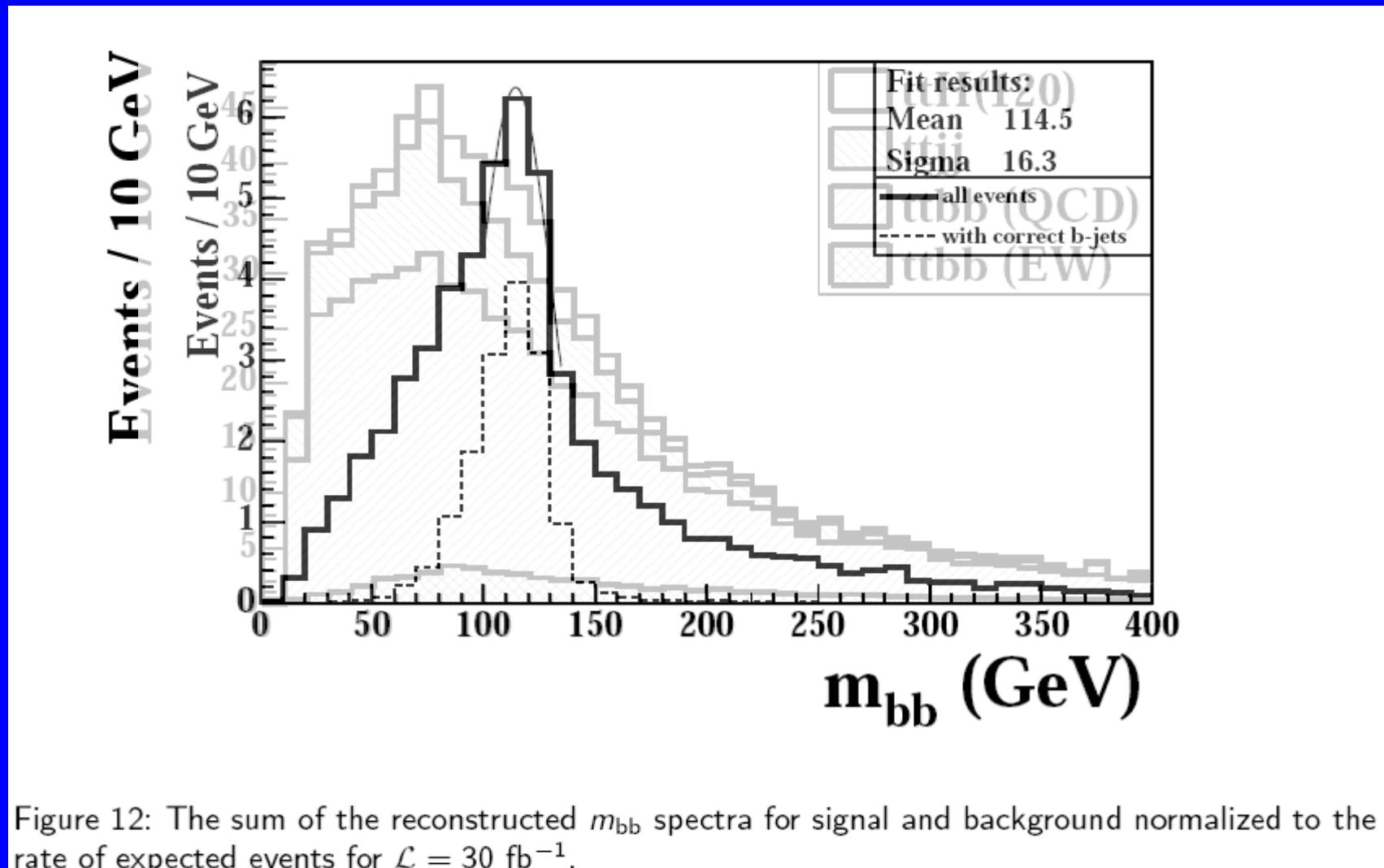
Event selection

- Remove background
 - which contains ~no signal
 - which is highly uncertain (sometimes)
- Usually better not to make the last cut
 - loss of sensitivity
 - single events around the cut unpleasant



Example: ttH in ATLAS

<http://documents.cern.ch/cgi-bin/setlink?base=atlnot&categ=Note&id=phys-2003-024>



Likelihood ratio for binned counts

$$Q_i = \frac{\frac{e^{-(s_i+b_i)}(s_i+b_i)^{n_i^{cand}}}{n_i^{cand}!}}{e^{-b_i} \frac{b_i^{n_i^{cand}}}{n_i^{cand}!}}$$

$$-2 \ln Q_i = 2 s_i - 2 n_i \ln \left(1 + \frac{s_i}{b_i} \right)$$

$$Q = \prod Q_i$$

$$-2 \ln Q = \sum -2 \ln Q_i$$

- Likelihood ratio for bin i
 - Poisson statistics

- ...in useful form

- Likelihood ratio for entire distribution (or combination of channels, experiments, etc)
- ...in useful form

Means and RMS's of $-2\ln Q_i$

$$w_i = \ln \left(1 + \frac{s_i}{b_i} \right)$$

$$\langle -2 \ln Q_i \rangle_{s+b} = 2 s_i - 2 (s_i + b_i) w_i$$

$$\langle -2 \ln Q_i \rangle_b = 2 s_i - 2 (b_i) w_i$$

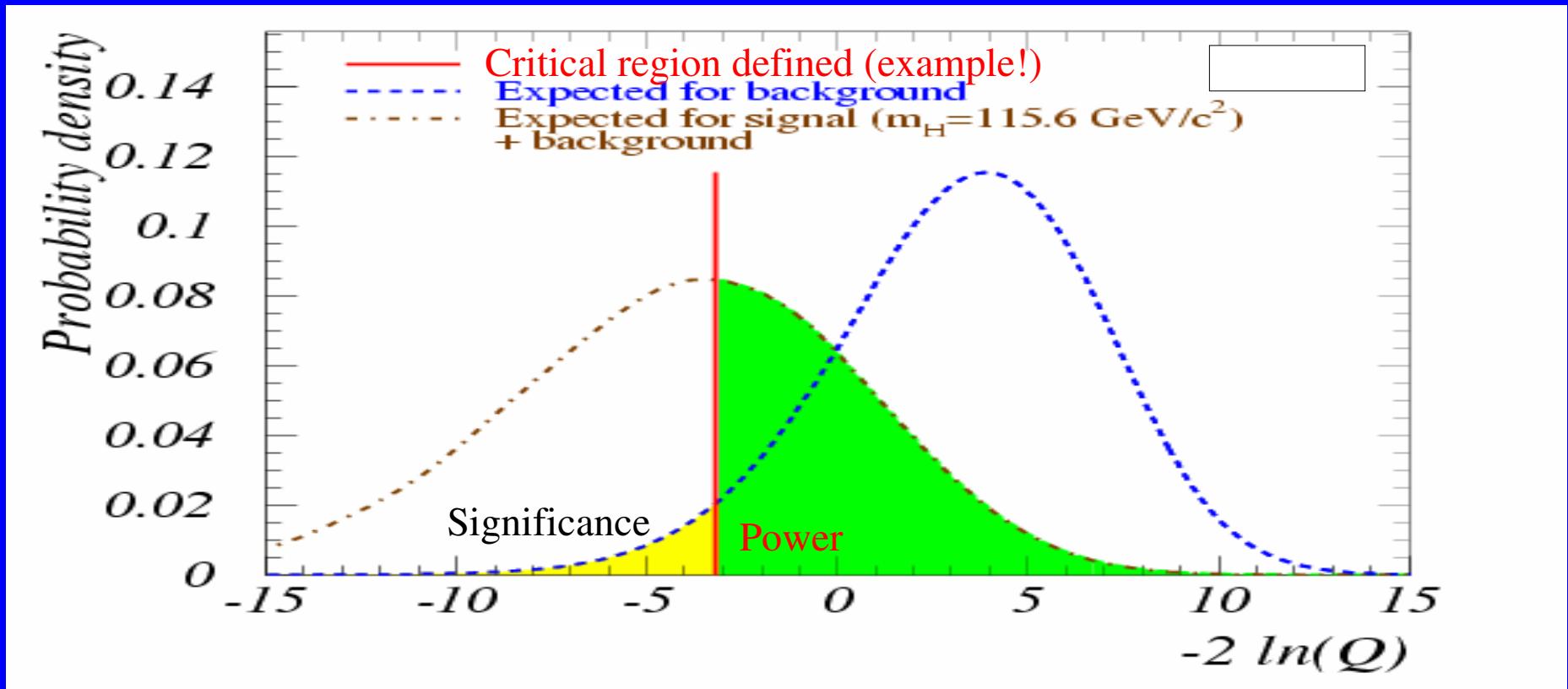
$$\sigma_{n_i}^2 = n_i$$

$$\sigma_{s_i+b_i}^2 = 4 (s_i + b_i) w_i^2$$

$$\sigma_{s_i+b_i}^2 = 4 (b_i) w_i^2$$

- LR weight (optimal)
- Mean $-2\ln Q$ for “s+b” expts.
- Mean $-2\ln Q$ for “b” expts.
- RMS for Poisson pdf.
- RMS for $-2\ln Q$ for “s+b”
- RMS of $-2\ln Q$ for “b”

Distributions of $-2\ln Q$



- These are pdf's of $-2\ln Q$ for 2 hypotheses
 - “s+b” or signal plus background
 - “b” or background-only

Optimize significance of signal

$$\lambda = \frac{\langle -2 \ln Q \rangle_b - \langle -2 \ln Q \rangle_{s+b}}{\sigma_b}$$

$$\lambda = \frac{\sum 2(s_i - b_i w_i) - 2(s_i - (s_i + b_i) w_i)}{\sqrt{\sum 4 b_i w_i^2}}$$

$$\lambda = \frac{\sum s_i w_i}{\sqrt{\sum b_i w_i^2}}$$

$$\lambda = \frac{S w}{\sqrt{B w}} = \frac{S}{\sqrt{B}}$$

- Number of background sigmas between average signal and average background experiment

- For a single bin (counting)

Optimize power of analysis

$$\kappa = \frac{\langle -2 \ln Q \rangle_b - N \sigma_b - \langle -2 \ln Q \rangle_{s+b}}{\sigma_{s+b}}$$

$$\kappa = \frac{\sum s_i w_i - N \sqrt{\sum b_i w_i^2}}{\sqrt{\sum (s_i + b_i) w_i^2}}$$

$$\kappa = \frac{S - N \sqrt{B}}{\sqrt{(S+B)}}$$

$$\kappa = \frac{S}{\sqrt{(S+B)}}$$

- Number of signal + background sigmas included in **critical region** (discovery)

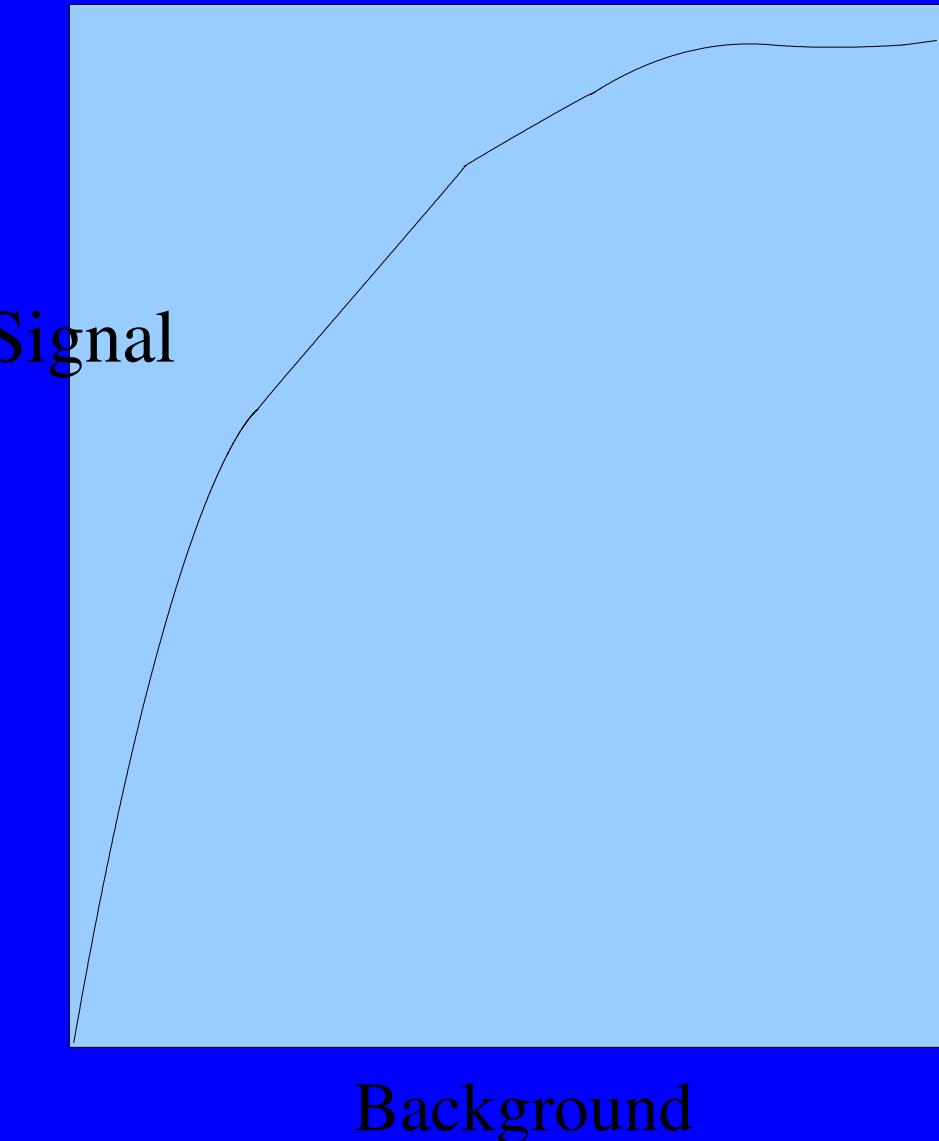
- For a single bin (counting)

- For strong signals (measurement)

High statistics

- Correspondence of integrals in tails to Gaussian or normal distribution depends on statistics
 - If not $> \sim 10$ events in every bins the correspondence may be poor – and more sophisticated or heavy methods needed
 - alrmc (alrmc++) developed for LEP/DELPHI, but heavy to use
 - For high statistics (like the ttH example) we know the tails quite well and can use tables or lookup functions.

Signal efficiency vs. background



- Efficiency vs. background depends on event selection!

Optimizing

- Since both shapes and rates may depend on event selection there **is** an optimal working point
- The optimizations (exclusion, observation, discovery, measurement) are not identical
 - If not similar consider 2-d likelihood functions and don't make the previous cut

Summary

- Event selection and analysis optimization is a nontrivial exercise but standard statistics practice gives us solid guidelines
- Formulae for significance of average expectations S/\sqrt{B} and power of search or measurement $S/\sqrt{(S+B)}$ are easily generalized for binned likelihood ratio
 - in this case and for high statistics it is relatively easy to make statements about significance and power