

- Determine Errors on Parameters
 - Random, or Statistical - Precision
 - Systematic – Accuracy
- Consider radioactive decay
 - Determine decay constant
 - Measure decay rate
 - Mass of the sample
- Innate randomness gives a random error (Poisson distributed)

dots

- Systematic Error
 - Measured counting rate is lower
 - Efficiency x Acceptance
 - Estimate the correction
 - Uncertainty in this estimate contributes to systematic error
 - Another Example: Charged Particle Multiplicity in each rapidity bin
 - Acceptance determined using angular distribution, and hence event generators
 - Uncertainty in correction factor is systematic error
 - Sometimes cancels out

$$\sigma_i = \frac{n_i}{Bt}$$

Error Propagation

- $a = b \pm c$

- Error on 'a'

$$\sigma_a^2 = \sigma_b^2 + \sigma_c^2$$

- $a = b^r c^s$

- Fractional error on a

$$\left(\frac{\sigma_a}{a}\right)^2 = r^2 \left(\frac{\sigma_b}{b}\right)^2 + s^2 \left(\frac{\sigma_c}{c}\right)^2 + \frac{2rs \operatorname{cov}(b,c)}{bc}$$

- Last term is zero if b and c are independent

- Errors on Scaled Factorial Moments:

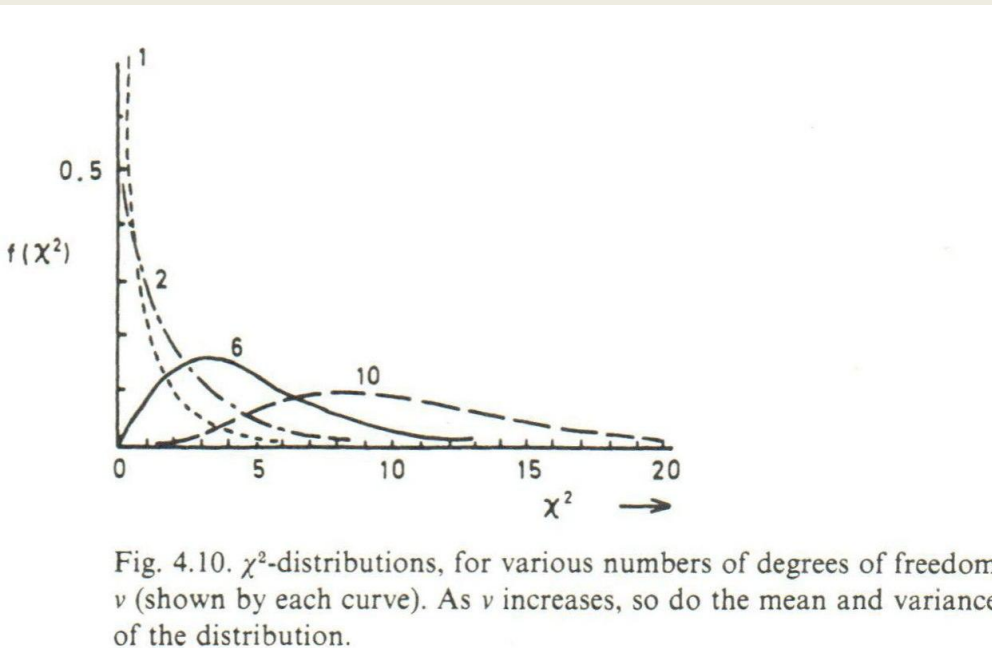
- What happens when we fit a distribution?
 - Goodness of fit statistic χ^2

$$\frac{\overline{\chi^2}}{dof} = 1 \quad \text{Is this value sacred? (Mean, not MP)}$$

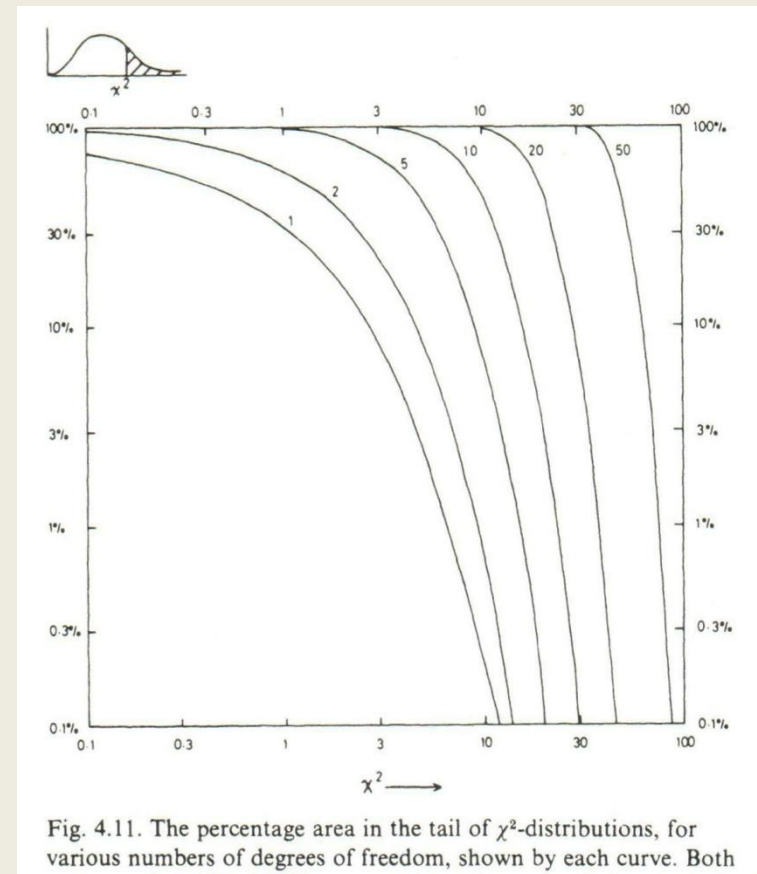
For a fixed ‘ p ’ value, χ^2/dof is different for different no. of degrees of freedom

‘p-value’ is the area under tail of the χ^2 distribution

- The χ^2 distribution and p-values



Most probable values do not correspond to $\chi^2/\text{dof} = 1.0$



How do we decide on rejecting the null hypothesis

For 5 degrees of freedom, if χ^2 is 3.0, then the probability for hypothesis to be correct is 70%.

However, if it is ~ 11 , then the probability is 5%

The χ^2 values for 10 degrees of freedom, for the same probability are ~ 7.3 and ~ 18.3

For what value of χ^2 can we say 100%?

Based upon p-values, for $\chi^2 = 0$!

Gaussian peaked at x_m , width $\sigma = 1$

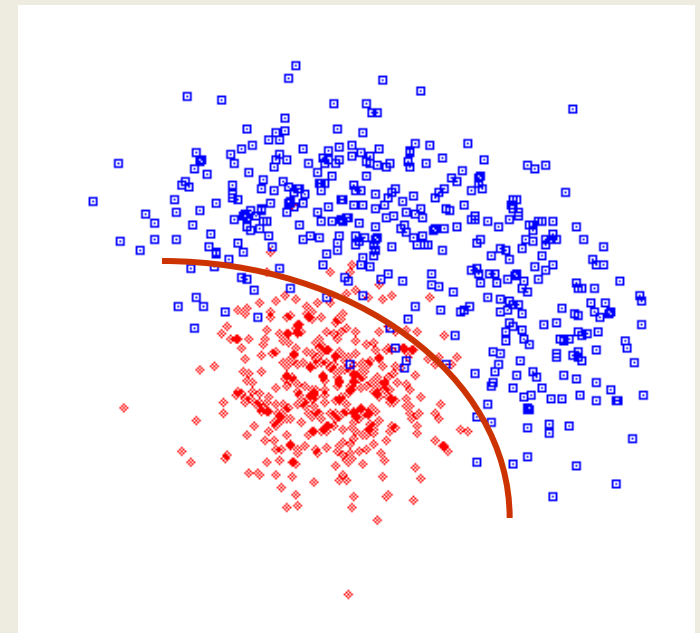
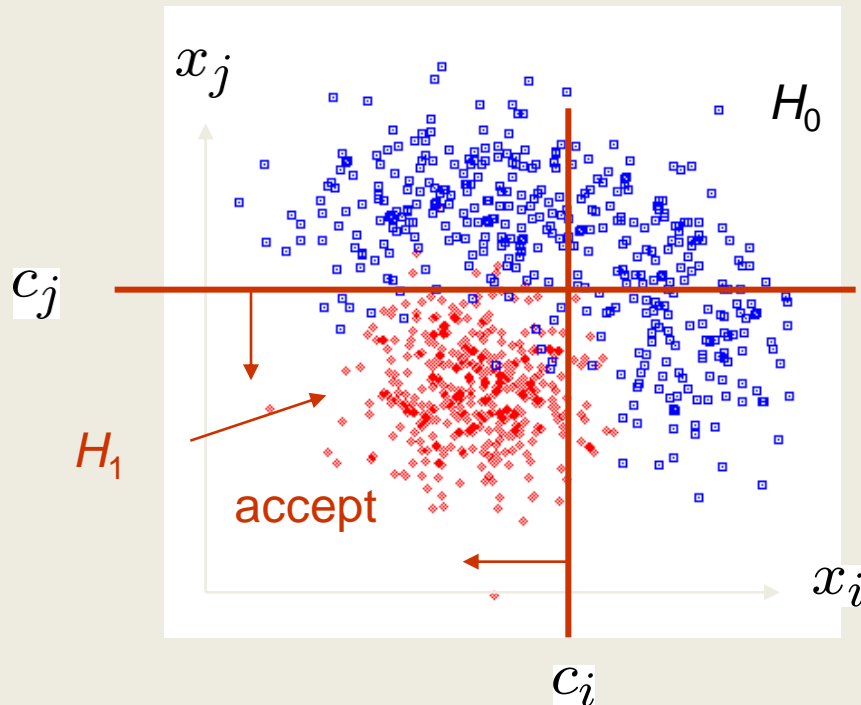
$$\frac{\int_0^{x_l} f(x) dx}{\int_0^{\infty} f(x) dx} = 0.9 \quad (1)$$

$$x_l = x_m + 1.28\sigma \quad (2)$$

	90% CL upper limits : x_l	
x_m	Method 1	Method 2
5	6.3	6.3
3	4.3	4.3
1	2.4	2.3
0.5	2.0	1.8
0	1.6	1.3
-0.5	1.4	0.8
-1	1.2	0.3
-3	0.6	(-1.7)
-5	0.5	(-3.7)

EVENT SELECTION (G.Cowan's)

- Event characterised by X (multidimensional)
 - $P(X|H_0)$ corresponds to background
 - $P(X|H_1)$ corresponds to signal
- Need a decision boundary
 - PMD: charged particle and photon separation



- For the decision boundary

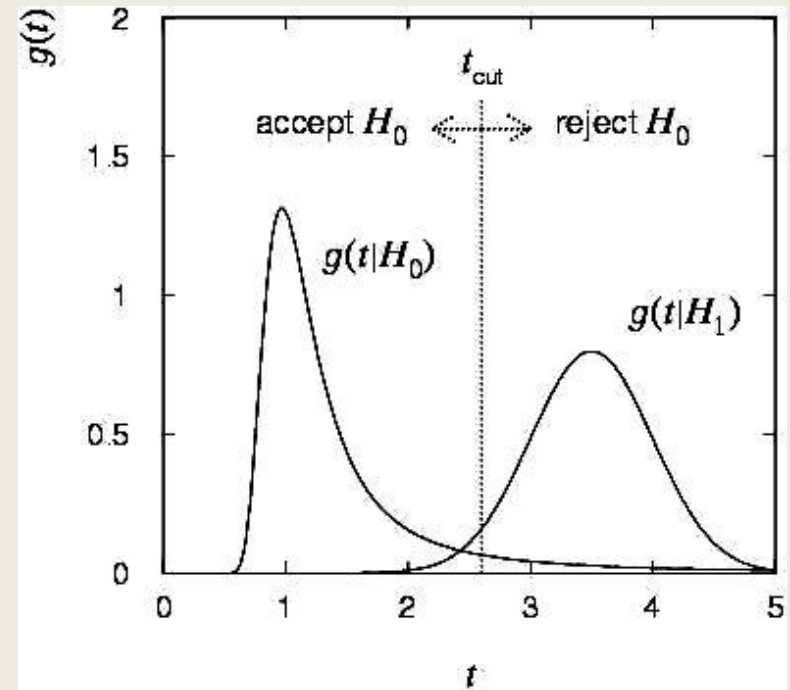
- Make a test statistic $t(X)$, boundary defined by t_{cut}
- $g(t|H_0)$, $g(t|H_1)$

Probability to reject H_0 if it is the correct hypothesis (Error of Type-I)

$$\alpha = \int_{t_{\text{cut}}}^{\infty} g(t|H_0) dt$$

Probability to accept H_0 if H_1 is the correct hypothesis (Type-II)

$$\beta = \int_{-\infty}^{t_{\text{cut}}} g(t|H_1) dt$$



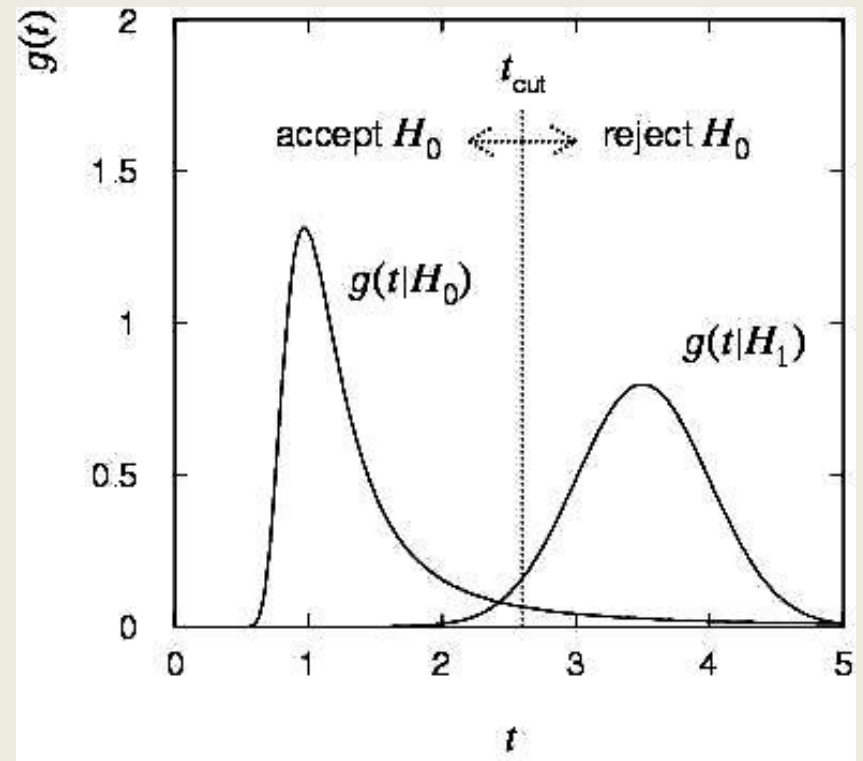
Signal/Background Efficiency

Assume it is a background event. We may (mis)identify it as a signal event. The probability of ‘background efficiency’ is

$$\varepsilon_b = \int_{t_{\text{cut}}}^{\infty} g(t|b) dt = \alpha$$

Assume a signal event. The probability to identify it correctly is ‘signal efficiency’ and is

$$\varepsilon_s = \int_{t_{\text{cut}}}^{\infty} g(t|s) dt = 1 - \beta$$



Purity of the Sample

- Assume fractions of signal and background events are π_s and π_b . Then, the purity of signal is

$$\begin{aligned} P(s|t > t_{\text{cut}}) &= \frac{P(t > t_{\text{cut}}|s)\pi_s}{P(t > t_{\text{cut}}|s)\pi_s + P(t > t_{\text{cut}}|b)\pi_b} \\ &= \frac{\varepsilon_s \pi_s}{\varepsilon_s \pi_s + \varepsilon_b \pi_b} \end{aligned}$$

Purity depends upon prior probabilities ! Uncertainty in the prior probabilities contributes to systematic error.

- More-than-one parameter
 - Correlations and error bars
 - X_j are the set of parameters
 - Maximise the probability

$$\left. \frac{\partial P}{\partial X_i} \right|_{\{X_{oj}\}} = 0$$

$$L = L(X_0, Y_0) + \frac{1}{2} \left[\begin{aligned} & \frac{\partial^2 L}{\partial X^2} \Big|_{X_0, Y_0} (X - X_0) + \frac{\partial^2 L}{\partial Y^2} \Big|_{X_0, Y_0} (Y - Y_0) \\ & + 2 \frac{\partial^2 L}{\partial X \partial Y} \Big|_{X_0, Y_0} (X - X_0)(Y - Y_0) \end{aligned} \right] + \dots$$

$$Q = (X - X_0 \quad Y - Y_0) \begin{pmatrix} A & C \\ C & B \end{pmatrix} \begin{pmatrix} X - X_0 \\ Y - Y_0 \end{pmatrix}$$

$$A = \frac{\partial^2 L}{\partial X^2} \Big|_{X_0, Y_0}, \quad B = \frac{\partial^2 L}{\partial Y^2} \Big|_{X_0, Y_0}, \quad C = 2 \frac{\partial^2 L}{\partial X \partial Y} \Big|_{X_0, Y_0}$$

$$\sigma_X = \sqrt{\frac{-B}{AB - C^2}}; \quad \sigma_Y = \sqrt{\frac{-A}{AB - C^2}}; \quad \sigma_{XY}^2 = \frac{C}{AB - C^2}$$

$$\begin{pmatrix} \sigma_X^2 & \sigma_{XY}^2 \\ \sigma_{XY}^2 & \sigma_Y^2 \end{pmatrix} = \frac{1}{AB - C^2} \begin{pmatrix} -B & C \\ C & -A \end{pmatrix} = - \begin{pmatrix} A & C \\ C & B \end{pmatrix}^{-1}$$

As magnitude of C increases, skewed contours

Same thing....differently.

$$L = \ln p = \text{const} - \frac{1}{2} \left(\frac{X^2}{\sigma_X^2} + \frac{Y^2}{\sigma_Y^2} \right)$$

Equations above give $A = -1/\sigma_x^2$; $B = -1/\sigma_y^2$ and $C = 0$

$X_0=Y_0=0$; $\sigma_x = \sqrt{2}/4$; $\sigma_y = \sqrt{2}/2$ gives $8x^2 + 2y^2 = 1$

Make the transformation and choose $\theta = 30^\circ$

$x' = x \cos \theta - y \sin \theta$ and $y' = y \cos \theta + x \sin \theta$

- This gives

$$\frac{1}{2} [13x'^2 + 6\sqrt{3}x'y' + 7y'^2] = 1$$

$$(x' \ y') \begin{pmatrix} \frac{13}{2} & \frac{3\sqrt{3}}{2} \\ \frac{3\sqrt{3}}{2} & \frac{7}{2} \end{pmatrix} \begin{pmatrix} x' \\ y' \end{pmatrix} = 1$$

Inverting gives the error matrix

$$\frac{2}{64} \begin{pmatrix} 7 & -3\sqrt{3} \\ -3\sqrt{3} & 13 \end{pmatrix}$$

Yielding $\sigma_{x'}^2 = \frac{14}{64} = (0.468)^2$

$$\sigma_{y'}^2 = \frac{26}{64} = (0.637)^2$$

$$\text{cov}(x', y') = -\frac{6\sqrt{3}}{64} = -(0.403)^2$$

Negative sign

- Simple Examples

- Function of variables $f = f(x,y)$

- Given the errors on x and y , find the error on f

- $$\overline{\delta f^2} = \left(\frac{\partial f}{\partial x}\right)^2 \overline{\delta x^2} + \left(\frac{\partial f}{\partial y}\right)^2 \overline{\delta y^2} + 2\left(\frac{\partial f}{\partial x} \frac{\partial f}{\partial y}\right) \overline{\delta x \delta y}$$

$$\overline{\delta f^2} = \begin{pmatrix} \frac{\partial f}{\partial x} & \frac{\partial f}{\partial y} \end{pmatrix} \begin{pmatrix} \overline{\delta x^2} & \overline{\delta x \delta y} \\ \overline{\delta x \delta y} & \overline{\delta y^2} \end{pmatrix} \begin{pmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{pmatrix}$$

$$\sigma_f^2 = \overline{DMD}$$

Change of variables $p = p(x,y)$ and $q = q(x,y)$

- Examples

- $y=x+2x$

- Asymmetry $A = \frac{F - B}{F + B}$

- F and B independent ($N = F + B$)
 - Error (Poissonian on F and B)

$$\sigma_a = \frac{1 - A^2}{2} \sqrt{\left(\frac{1}{F} + \frac{1}{B}\right)}$$

- N is F+B is a constant (completely correlated)
 - Error

$$\sigma_a = \frac{2}{N} \sqrt{\frac{FB}{N}}$$

Various Distributions Used in Particle Physics

Binomial

Multinomial

Poisson

Gaussian

Cauchy

(Breit-Wigner)

Chi-square

Branching Ratio

Histogram

Counting Rate

Measurement Error

Resonance Formation

Goodness of Fit Estimate

Summarise

- Bayesian methods
- Simple examples of hypothesis testing and parameter determination, fitting distributions
- Rules about error propagation
- Meaning of errors and confidence limits
- Event Selection and Decision Boundary
- Correlated errors and error matrix