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8.13-14 Experimental Physics I & II "Junior Lab"
Fall 2007 - Spring 2008

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Introduction to Junior Lab

Feb 6 2008

Getting started

- Find a good partner
 - 18 units: 6 in lab, 12 outside lab
 - this is not an overestimate
 - make sure you + your partner can coordinate schedule and work together efficiently

Experiments

- Major advances in science (e.g. Nobel prizes)
- Refine your skills in the art + science of experimental physics
 - How to obtain good data
 - How to document your work
 - How to estimate errors
 - How to present your results
- As close to real life as possible

Experiments

- 26 sessions total
 - attendance is required
- You will do
 - 4 out of 8 exp's (5 sessions each)
 - prep questions
 - oral presentation + paper for each
 - 2x30min per partnership for oral
 - Last 4 weeks: Possible “challenge” work on selected experiment

Ethics in Science

- Fabrication/Falsification of data
 - document everything as you go (Notebook)
 - complete record of everything you have done, including mistakes
- Plagiarism
 - never use other work without acknowledgement
 - mark quotes as quotes
 - do not import text (from web resources)
 - Comparison to known values is ok, but not substitution/modification of your data, unless clearly marked
- No tolerance in JLab

Safety

- Electrical safety
 - be careful
 - never work alone
- Laser safety
 - Wear goggles
 - Use common sense
- Cryo Safety
- Radiation Safety

Grading Scheme

- 10% attendance/lab performance
 - change 'lead' from exp to exp
- 8% Notebooks
 - graded by Scott Sewell (2x in semester)
- 10% prep problems
 - come prepared, you will need the time
- 40% orals
 - 20' are short
 - split topic between partners (but not along theory/experiment)
- 32% papers
 - < 4 pages, due morning after oral
 - both partners have to write their own paper

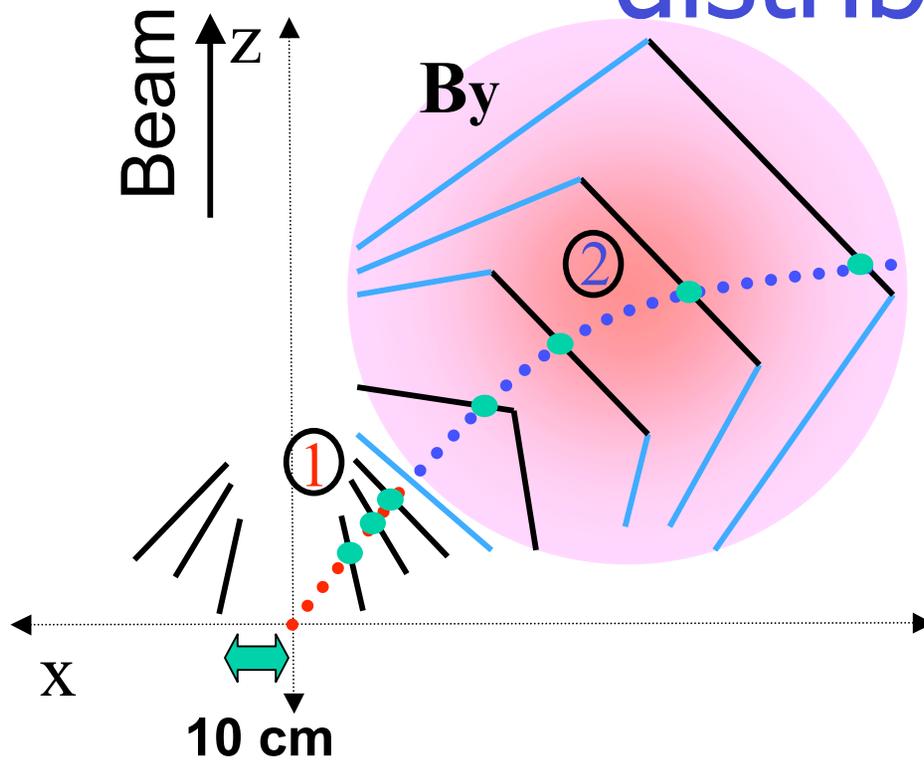
Introduction to Data Analysis

c.f. Bevington
Chapters 1-3

Data Reduction

- Translate measured data into one or more physical variables of interest
- Obtain
 - best estimate for physical variable
 - estimate for precision and accuracy of measurement (systematic and statistical uncertainty)
- Example (from my experiment):

EXAMPLE. DETERMINATION OF particle momentum distribution



$$R = p_{zx}/(q \cdot B_y)$$

Histograms

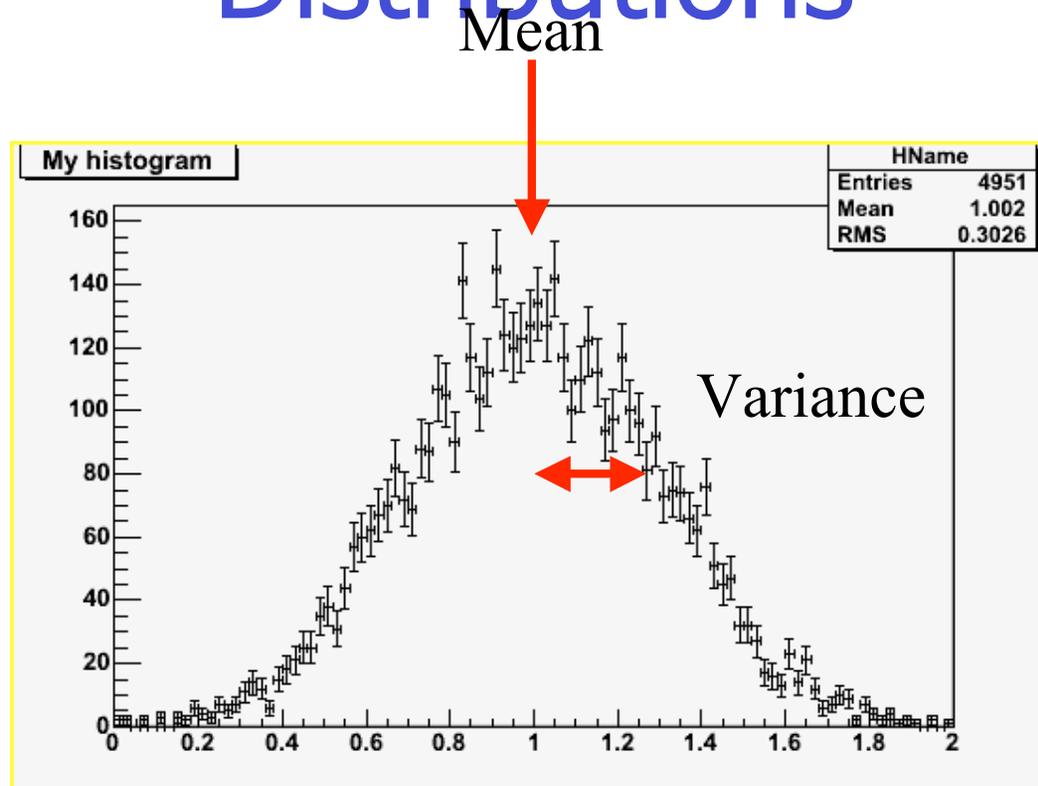
Binned representation of data in 1, 2 or 3 dimensional variable space

Statistical and Systematic Error

- Systematic error
 - inherent to measurement, apparatus, methods
 - estimate magnitude by comparing different approaches
 - limits accuracy
- Statistical error
 - Measurements jitter around truth
 - Average many measurements to improve estimate (if they are independent)
 - limits precision, but $\lim(N \rightarrow \infty) = \text{truth}$

Distributions

Sample
Distribution



$$\text{Mean } \langle x \rangle = \frac{1}{N} \sum_{i=1}^N x_i$$

$$\text{Variance } s_x^2 = \frac{1}{N-1} \sum_{i=1}^N (x_i - \langle x \rangle)^2$$

for $N \rightarrow \text{inf}$:
Parent distribution

Binominal Distribution

$$P(x : n, p) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}$$

$$\text{Mean } \mu = n \cdot p$$

$$\text{Std. Dev } \sigma = \sqrt{n \cdot p(1-p)}$$

$P(x:n,p)$: Probability to get 'yes' x times out of n tries, if probability of 'yes' for single try is p

Poisson Distribution

$$P(x, \mu) = \frac{\mu^x}{x!} e^{-\mu}$$

$$\text{Mean } \mu = \mu$$

$$\text{Std. Dev } \sigma = \sqrt{\mu}$$

Derives from binomial distribution for $p \ll 1$ with $n \cdot p = m$
Important for counting experiments with low count rate

Gaussian Distribution

$$P(x, \mu, \sigma) = \frac{1}{\sqrt{2\pi}\sigma} e^{-(x-\mu)^2 / (2\sigma^2)}$$

Mean $\mu = \mu$

Std. Dev $\sigma = \sigma = \sqrt{\mu}$

Derives from Poisson distribution for $n \cdot p \gg 1$

Seen everywhere b/c of Central Limit Theorem

Statistical Error on Mean

$$\langle x \rangle \pm \sigma_x / \sqrt{N} = \frac{1}{N} \sum_{i=1}^N x_i \pm \frac{1}{N-1} \sum_{i=1}^N (x_i - \langle x \rangle)^2 / \sqrt{N}$$

- Repeated measurements increase precision
 - but only as sqrt(N)
 - ultimate limit may come from systematic uncertainty

Error propagation

Interested in error on 'x', but measure 'u', with $x = f(u)$

$$\sigma_x^2 = \sigma_u^2 \left(\frac{\partial x}{\partial u} \right)^2$$

What if $x = f(u,v)$:

$$\sigma_x^2 = \sigma_u^2 \left(\frac{\partial x}{\partial u} \right)^2 + \sigma_v^2 \left(\frac{\partial x}{\partial v} \right)^2$$

Errors add in quadrature, provided error in u and v is independent

Fitting of Data

c.f. Bevington
Chapters 4-6

Fitting

- Fitting: Find a functional form that describes data within errors
- Why fit data?
 - extract physical parameters from data
 - test validity of data or model
 - interpolate/extrapolate data

How to fit data?

- Vary parameters of function until you find a global maximum of goodness-of-fit criterion
- Goodness-of-fit  most common
 - chi-square
 - likelihood
 - (Kolmogorov-Smirnov test)

Chi-square fit

Chi-square $\rightarrow \chi^2 = \sum_{i=1}^N \frac{(x_i - e_i)^2}{\sigma_i^2}$

Data points $\rightarrow x_i$

Model expectation $\rightarrow e_i$

Estimated uncertainty $\rightarrow \sigma_i^2$

Best Fit: Global minimum of Chi-square

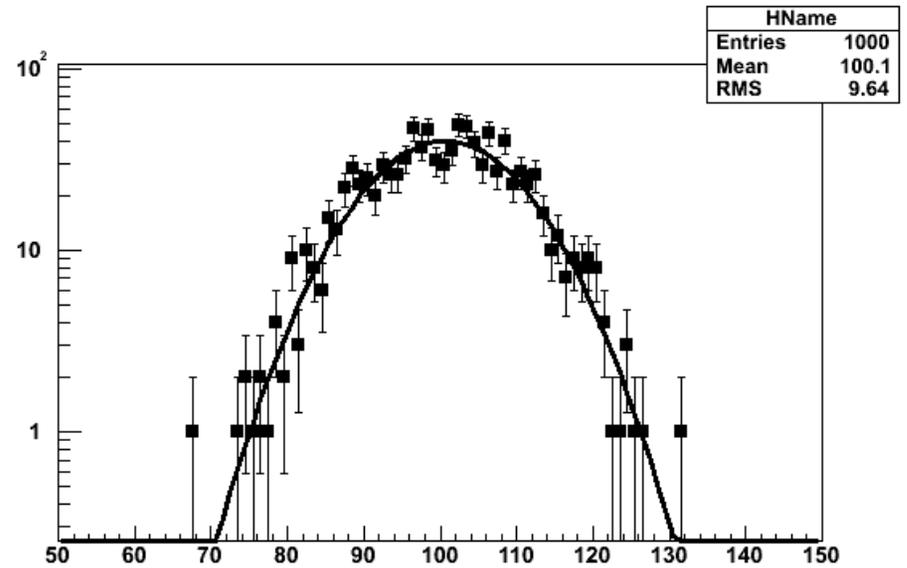
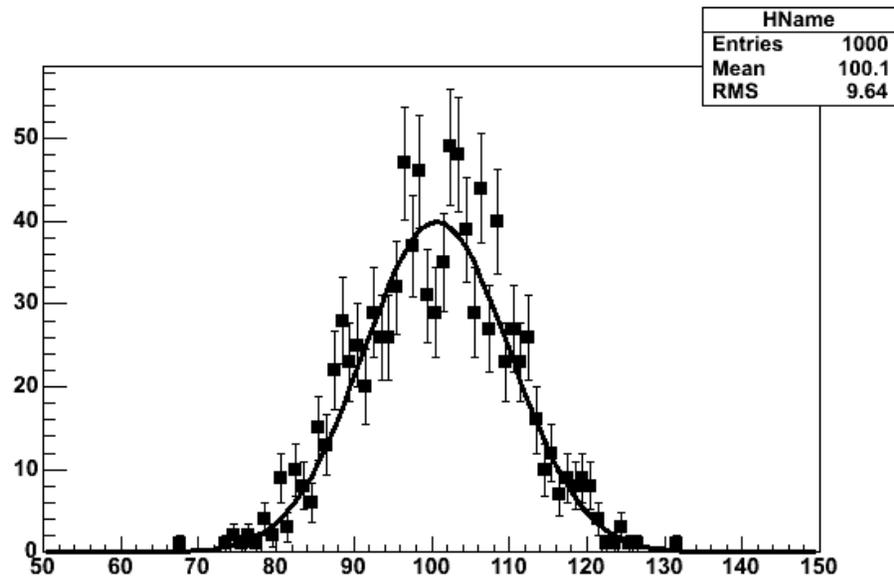
Goodness of fit

- You EXPECT data to fluctuate by \sim error
- For Gaussian errors, only 68% of your data points should agree with model/fitted function by better than 1σ
- Chi-square per Degree of Freedom (DoF) should be ~ 1
 - DoF: Number of data points - number of fitted parameters
- Chi-square allows you to test if model/fitted function is compatible with data

Example

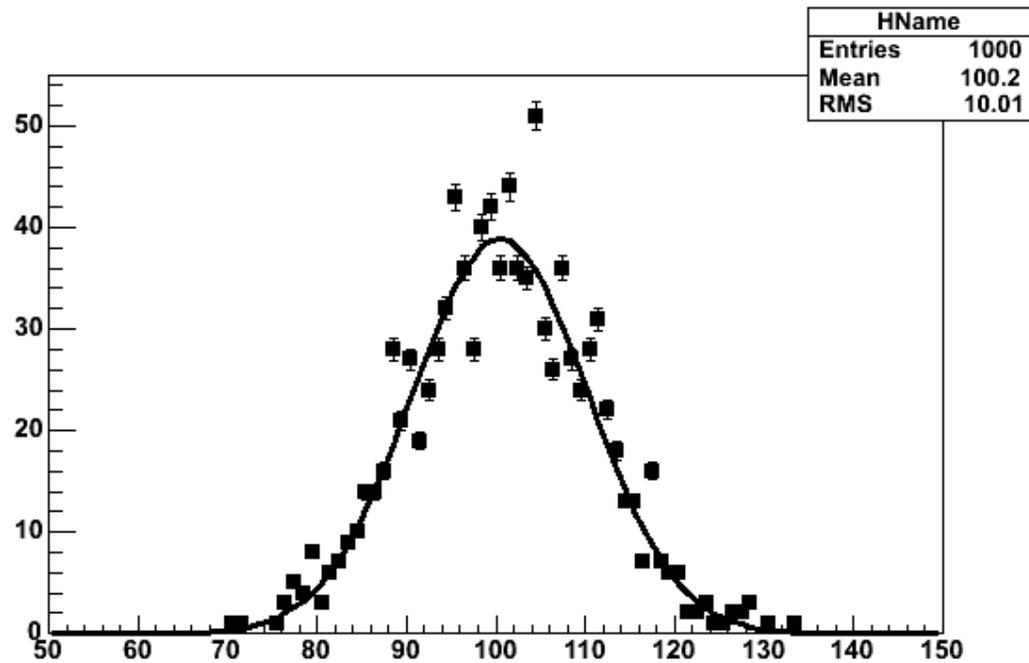
- Suppose you have 10^4 atoms
- Probability of decay is $10^{-2}/\text{sec}$
- Count #of decays/second
- Repeat experiment many times (always starting with 10^4 atoms)
- What distribution do you expect for the number of observed counts?

Example



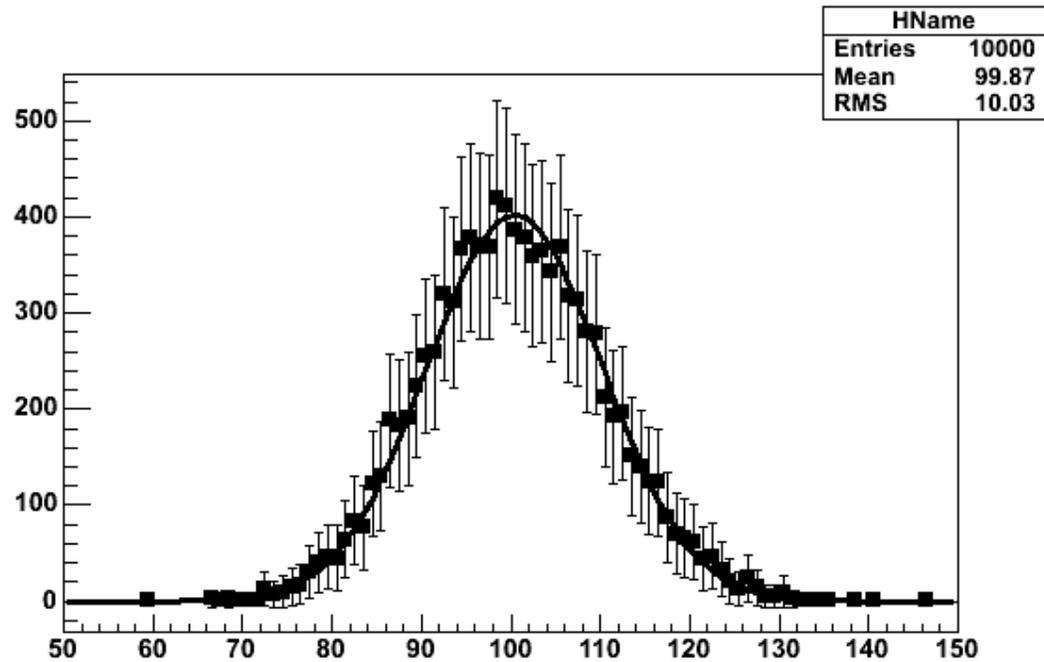
Chi-square/DoF = 60/53

Example



$$\text{Chi-square/DoF} = 1200/55$$

Example



Chi-square/DoF = 4.5/71

Chi-square distribution

- Even if your error estimate is correct and the model is correct
 - Chi-square will fluctuate from $\chi^2/\text{DoF} < 1$ to $\chi^2/\text{DoF} > 1$
 - The shape of the χ^2 -distribution depends only on the number of degrees of freedom
 - allows calculation of probability that data and model agree
- Chi-square probability
 - Percentage of all measurements that you expect to have a worse chi-square than what you see

Fitting

- Linear relationship: Analytical
- In general: Numerical minimization of chi-square
 - Many methods
 - simple grid-search
 - Non-linear Newton
 - Levenberg-Marquardt
 - Simulated annealing
 - Criteria
 - Dimension of parameter space
 - Data statistics
 - Challenge
 - Speed vs finding true global minimum