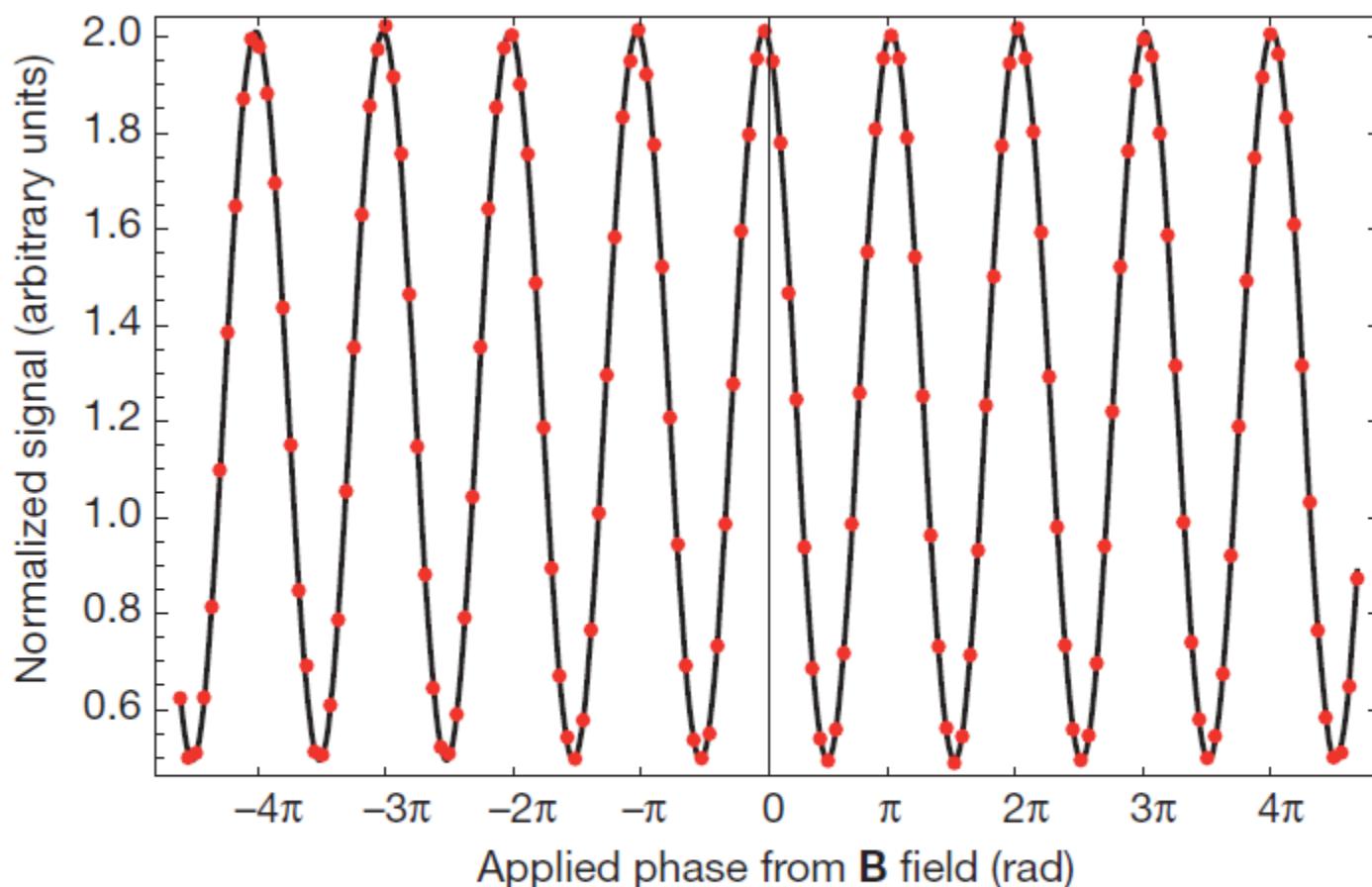


# **ERROR PROPAGATION**

**And an introduction to the Method of Least  
Squares**

One case study of data analysis for a single variable

# Electron EDM measurement



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**Figure 3 | Interferometer fringes produced by magnetic field scan.** Dots indicate the probe fluorescence normalized to the pump fluorescence. The line is the fit to the cosine-squared model.

15th October 2019

f YbF molecule in the molecular state. The pump laser creates a coherent superposition between the two hyperfine levels, where  $\hat{z}$  is the quantization axis. The effective magnetic dipole moment  $m_F = \pm 1$  is given by  $d_e E_{eff} T/h$ , where  $E_{eff}$  is the effective energy shift and  $T$  is the temperature. The effective magnetic dipole moment is then proportional to  $\cos^2 \phi$ , where  $\phi$  is the angle between the magnetic field and the z-axis. For the case of the pump laser, the effective magnetic dipole moment is shown to be

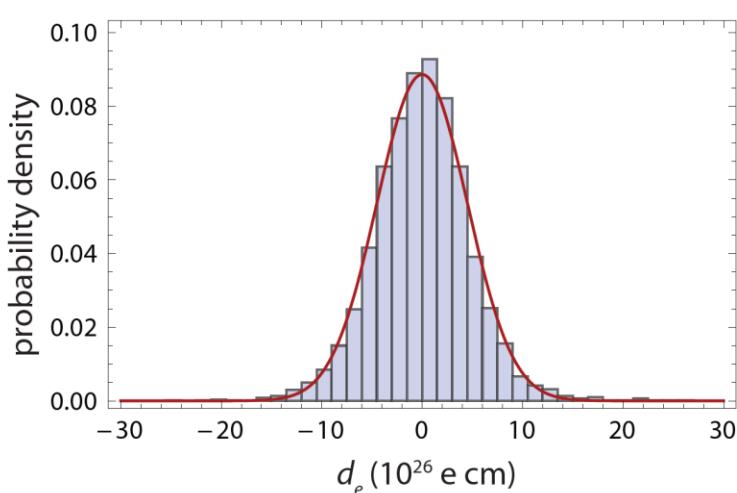
# Electron EDM measurement

Our measurement is derived from 6,194 blocks of data taken in 2010, comprising 25 million molecular beam pulses, together with many subsidiary measurements used to search for systematic errors.

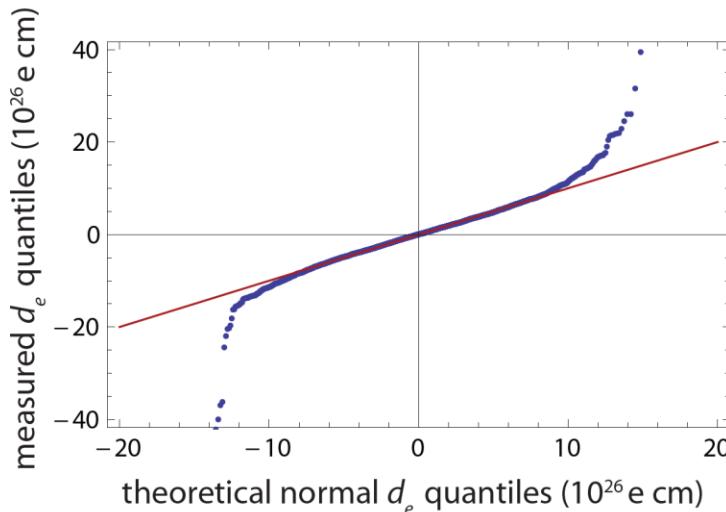
The EDM values obtained from the set of blocks are almost normally distributed but there tend to be a few more points in the wings of the distribution than in a normal distribution. The same is true of other quantities of interest that we extract from the data. For all these quantities, we calculate the 5% trimmed mean<sup>19</sup>, a simple robust statistic that drops the largest and smallest 5% of the data. We use the bootstrap method<sup>20</sup> to determine the associated statistical uncertainty. For non-normal distributions, these methods give more reliable measures than the mean and standard error.

After the offset, which was only removed once the data collection and analysis were complete.

## Robust statistics and bootstrap sampling methods used for non-normal distributions



(a)



(b)

FIGURE 3.6: The distribution of 6194  $d_e$  values that make up the 2011 dataset. For the most part, the distribution is fit well by a Gaussian distribution (red line) with a standard deviation given by the 68.3% statistical uncertainty of the bootstrapped trimmed mean multiplied by  $\sqrt{6194}$ . The distribution does, however, deviate from the Gaussian fit out in the wings, as demonstrated by the quantile-quantile plot on the right.

**Brute-force numerical techniques** are becoming more prevalent in data analysis

In total, we collected 1024 blocks (360.3 hours) of

PRL 119, 153001 (2017)

Selected for a *Viewpoint* in *Physics*  
PHYSICAL REVIEW LETTERS

week ending  
13 OCTOBER 2017



## Precision Measurement of the Electron's Electric Dipole Moment Using Trapped Molecular Ions

William B. Cairncross,<sup>\*</sup> Daniel N. Gresh, Matt Grau,<sup>†</sup> Kevin C. Cossel,<sup>‡</sup> Tanya S. Roussy,  
Yiqi Ni,<sup>§</sup> Yan Zhou, Jun Ye, and Eric A. Cornell

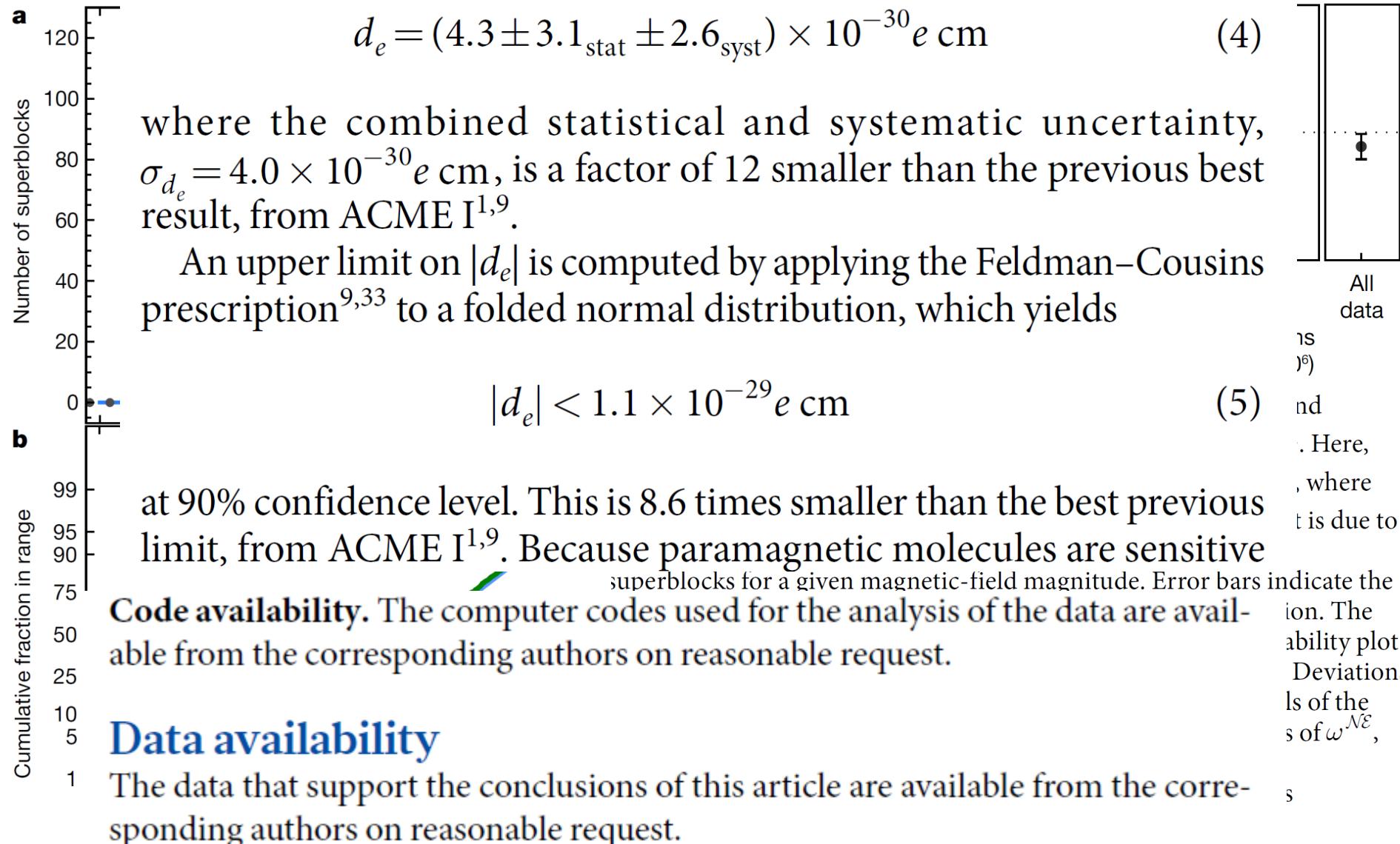
JILA, NIST and University of Colorado, Boulder, Colorado 80309-0440, USA  
and Department of Physics, University of Colorado, Boulder, Colorado 80309-0440, USA

(Received 21 April 2017; published 9 October 2017)

We describe the first precision measurement of the electron's electric dipole moment ( $d_e$ ) using trapped molecular ions, demonstrating the application of spin interrogation times over 700 ms to achieve high sensitivity and stringent rejection of systematic errors. Through electron spin resonance spectroscopy on  $^{180}\text{Hf}^{19}\text{F}^+$  in its metastable  ${}^3\Delta_1$  electronic state, we obtain  $d_e = (0.9 \pm 7.7_{\text{stat}} \pm 1.7_{\text{syst}}) \times 10^{-29} \text{ e cm}$ , resulting in an upper bound of  $|d_e| < 1.3 \times 10^{-28} \text{ e cm}$  (90% confidence). Our result provides independent confirmation of the current upper bound of  $|d_e| < 9.4 \times 10^{-29} \text{ e cm}$  [J. Baron *et al.*, *New J. Phys.* **19**, 073029 (2017)], and offers the potential to improve on this limit in the near future.

DOI: 10.1103/PhysRevLett.119.153001

The reduced chi-squared statistic for fitting a weighted mean to the eEDM data set is  $\chi_r^2 = 1.22(5)$ . This overscatter is present in all frequency channels and is attributable to



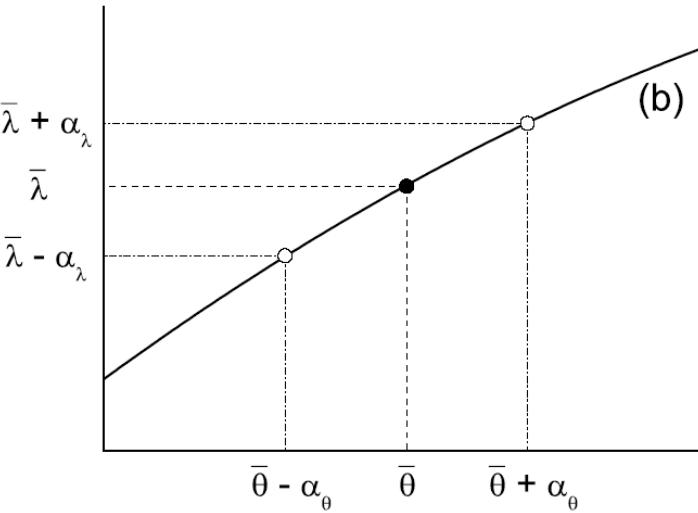
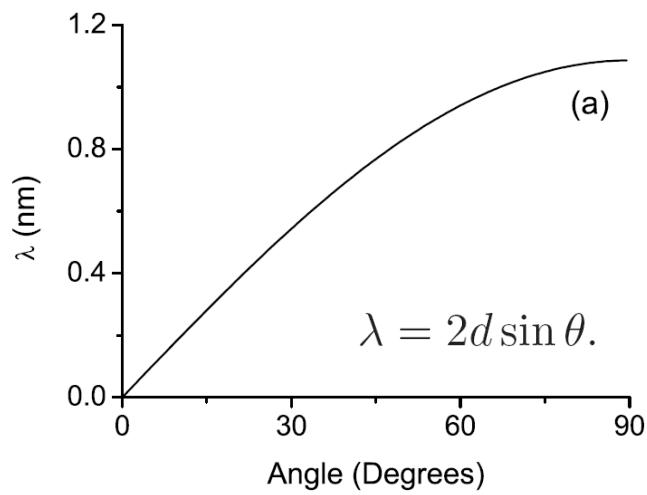
## Data availability

<sup>1</sup> The data that support the conclusions of this article are available from the corresponding authors on reasonable request.

In most experiments we have to combine measurements of several variables into a single quantity

Understanding how to propagate the error is a vital part of **data analysis and reduction**

Understanding which factors contribute to the **limiting error** is a vital part of **experimental design**



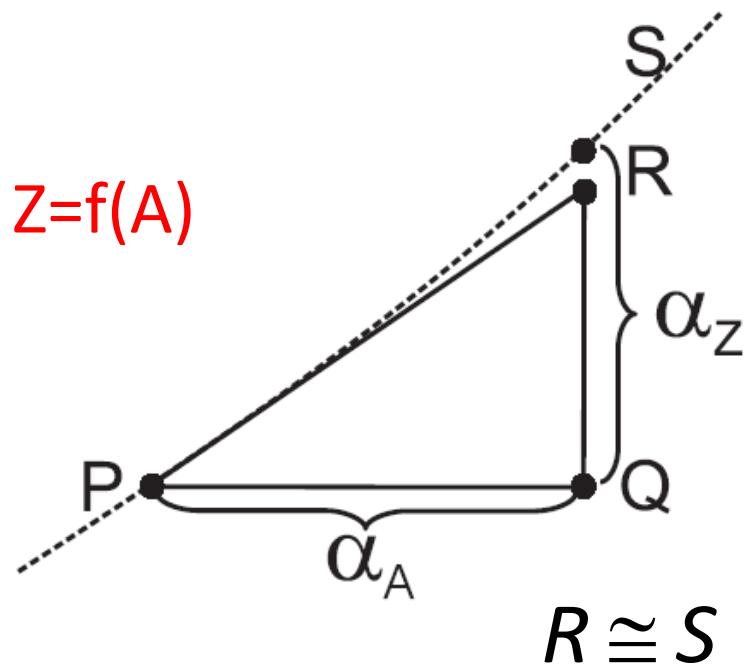
$Z=f(A)$



$$\alpha_Z = |f(\bar{A} + \alpha_A) - f(\bar{A})|.$$

$$\begin{aligned}\bar{Z} \pm \alpha_Z &= f(\bar{A} + \alpha_A), \\ \bar{Z} &= f(\bar{A}), \\ \bar{Z} \mp \alpha_Z &= f(\bar{A} - \alpha_A).\end{aligned}$$

Measurements and their Uncertainties, p 37



$$f(\bar{A}) + \frac{df(A)}{dA} \alpha_A = f(\bar{A} + \alpha_A),$$

$$\alpha_Z = \left| \frac{dZ}{dA} \right| \alpha_A.$$

Measurements and their Uncertainties, p 38

Level 1  
Physics  
Dealing with experimental uncertainty

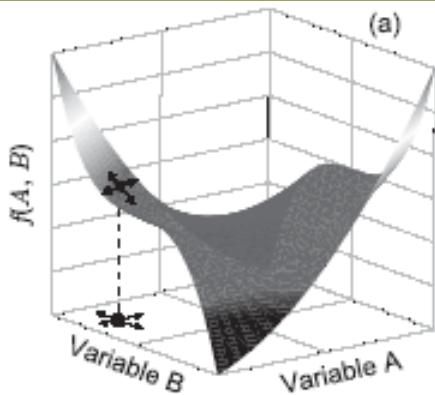
15th October 2019

- Calculus approach

| Function, $Z(A)$ | $\frac{dZ}{dA}$       | Uncertainty   |
|------------------|-----------------------|---|
| $\frac{1}{A}$    | $-\frac{1}{A^2}$      | $\alpha_z = \frac{\alpha_A}{A^2} = Z^2 \alpha_A$ OR $\left  \frac{\alpha_Z}{Z} \right  = \left  \frac{\alpha_A}{A} \right $ |
| $\exp A$         | $\exp A$              | $\alpha_z = \exp A \alpha_A = Z \alpha_A$   |
| $\ln A$          | $\frac{1}{A}$         | $\alpha_z = \frac{\alpha_A}{A}$   |
| $\log A$         | $\frac{1}{\ln(10) A}$ | $\alpha_z = \frac{\alpha_A}{\ln(10) A}$   |
| $A^n$            | $n A^{n-1}$           | $\alpha_z =  n A^{n-1}  \alpha_A$ OR $\left  \frac{\alpha_Z}{Z} \right  = \left  n \frac{\alpha_A}{A} \right $              |
| $10^A$           | $10^A \ln(10)$        | $\alpha_z = 10^A \ln(10) \alpha_A$  |
| $\sin A$         | $\cos A$              | $\alpha_z =  \cos A  \alpha_A$  |
| $\cos A$         | $-\sin A$             | $\alpha_z =  \sin A  \alpha_A$  |
| $\tan A$         | $1 + \tan^2 A$        | $\alpha_z = (1 + Z^2) \alpha_A$   |

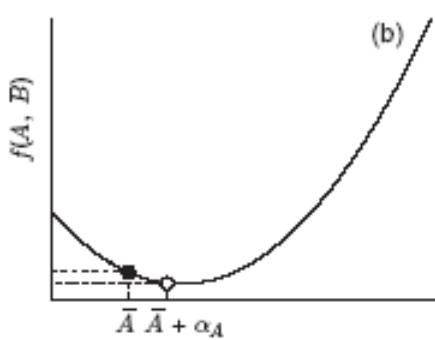
Measurements and their Uncertainties, inside front cover

# The functional approach for multi-variable functions

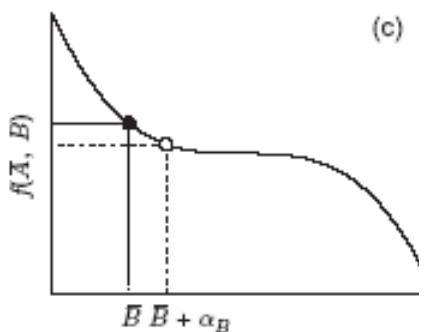


$$\alpha_Z^A = f(\bar{A} + \alpha_A, \bar{B}) - f(\bar{A}, \bar{B}), \quad \alpha_Z^B = f(\bar{A}, \bar{B} + \alpha_B) - f(\bar{A}, \bar{B})$$

$$(\alpha_Z)^2 = (\alpha_Z^A)^2 + (\alpha_Z^B)^2 + (\alpha_Z^C)^2 + \dots$$



$$\begin{aligned} (\alpha_Z)^2 = & [f(\bar{A} + \alpha_A, \bar{B}, \bar{C}, \dots) - f(\bar{A}, \bar{B}, \bar{C}, \dots)]^2 \\ & + [f(\bar{A}, \bar{B} + \alpha_B, \bar{C}, \dots) - f(\bar{A}, \bar{B}, \bar{C}, \dots)]^2 \\ & + [f(\bar{A}, \bar{B}, \bar{C} + \alpha_C, \dots) - f(\bar{A}, \bar{B}, \bar{C}, \dots)]^2 \\ & + \dots \end{aligned}$$



A calculus approximation to this result is

$$(\alpha_Z)^2 = \left(\frac{\partial Z}{\partial A}\right)^2 (\alpha_A)^2 + \left(\frac{\partial Z}{\partial B}\right)^2 (\alpha_B)^2 + \left(\frac{\partial Z}{\partial C}\right)^2 (\alpha_C)^2 + \dots$$

It is frequently possible to bypass the need to use the rigorous expression if one performs a quick back of the envelope calculation.

For example, if  $Z = A \times B \times C \times D$ , and  $A$  is known to 5%, and  $B$ ,  $C$  and  $D$  to 1%, what is the percentage error in  $Z$ ?

The experienced practitioner will not have to use the appropriate formula from Table 4.2 as the addition of the percentage errors in quadrature will yield, to one significant figure, 5%.

**RULE OF THUMB:** Perform a quick calculation to identify any dominant errors. Consider whether a more rigorous calculation is useful.

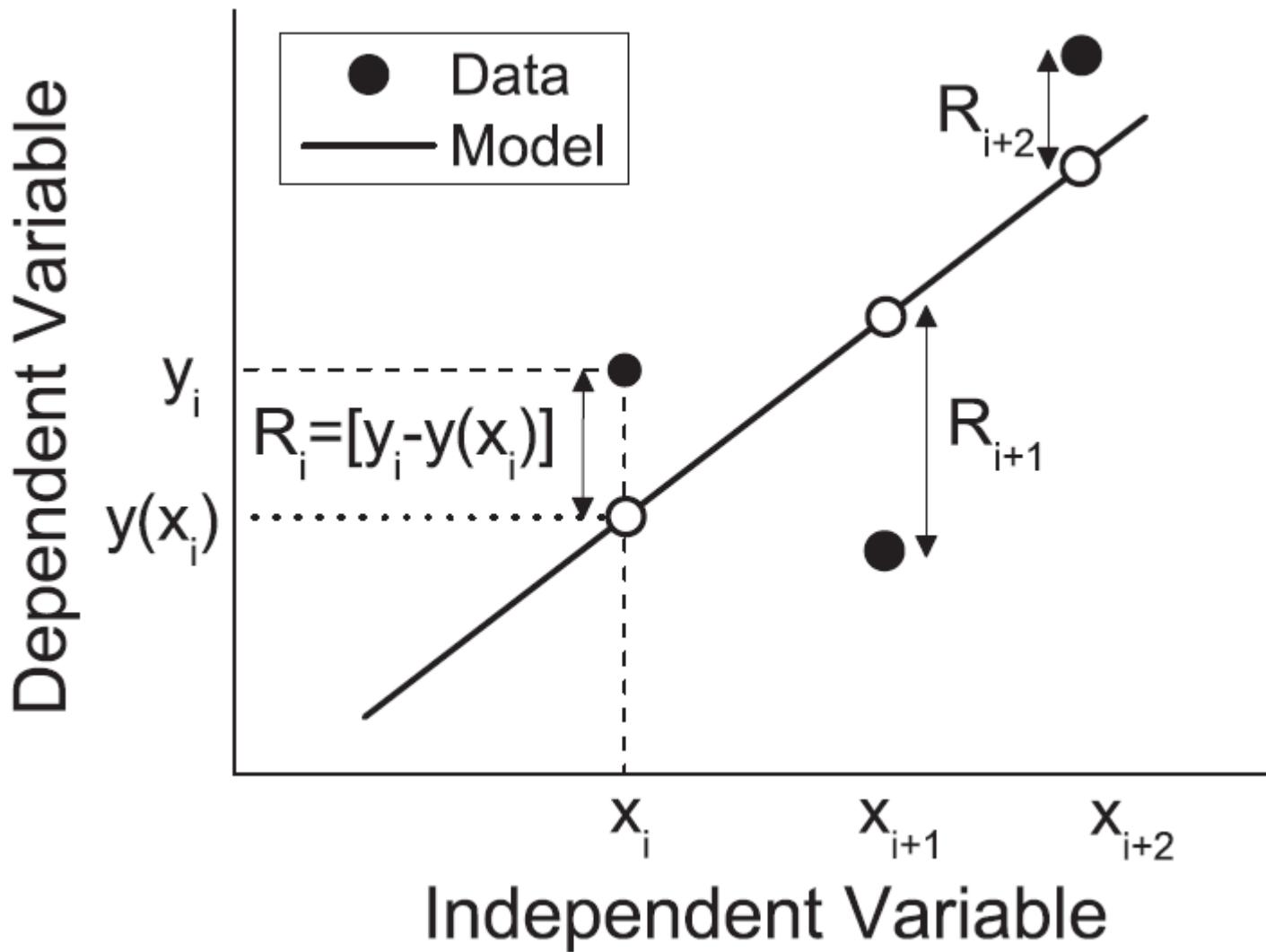
$$Z = \frac{(A - B)}{(A + B)}$$

One might be tempted to use the look-up tables and make the substitutions  
 $X(A, B) = (A - B)$  and  $Y(A, B) = (A + B)$  such that  $Z = X/Y$

This would be **WRONG** as the errors in  $X$  and  $Y$  are correlated

Only substitute single-variable functions when using the look-up tables.  
Recall that the look-up tables are only valid for independent variables or single-variable functions.

# The importance of residuals



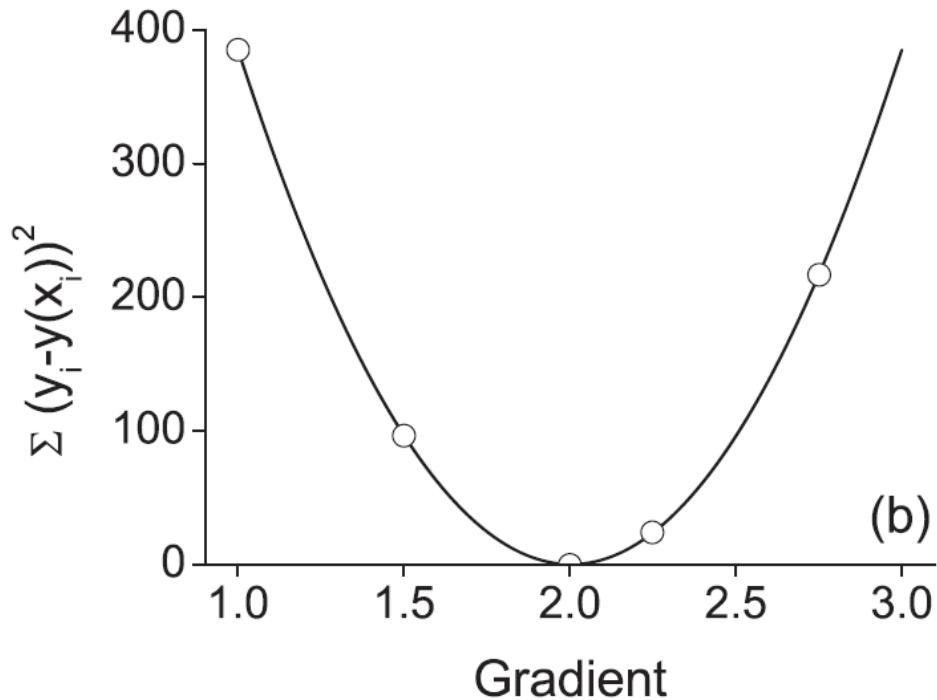
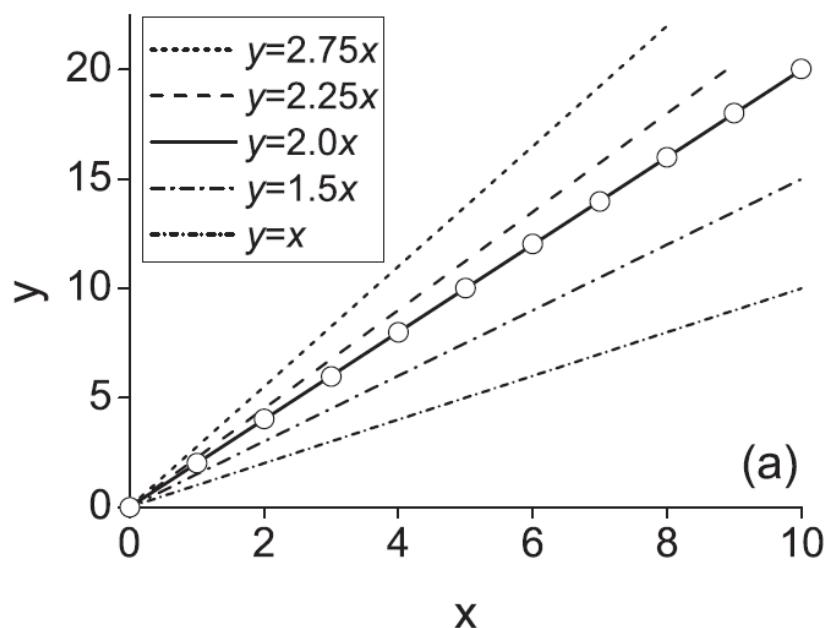
Measurements and their Uncertainties, Pg. 60

15th October 2019

# The goodness-of-fit parameter

$$\chi^2 = \sum_i \frac{(y_i - y(x_i))^2}{\alpha_i^2}$$

Note that  $\chi^2$  is dimensionless



$$\chi^2 = \sum_i \frac{(O_i - E_i)^2}{E_i}$$

$i$  is the number of counts (integer)

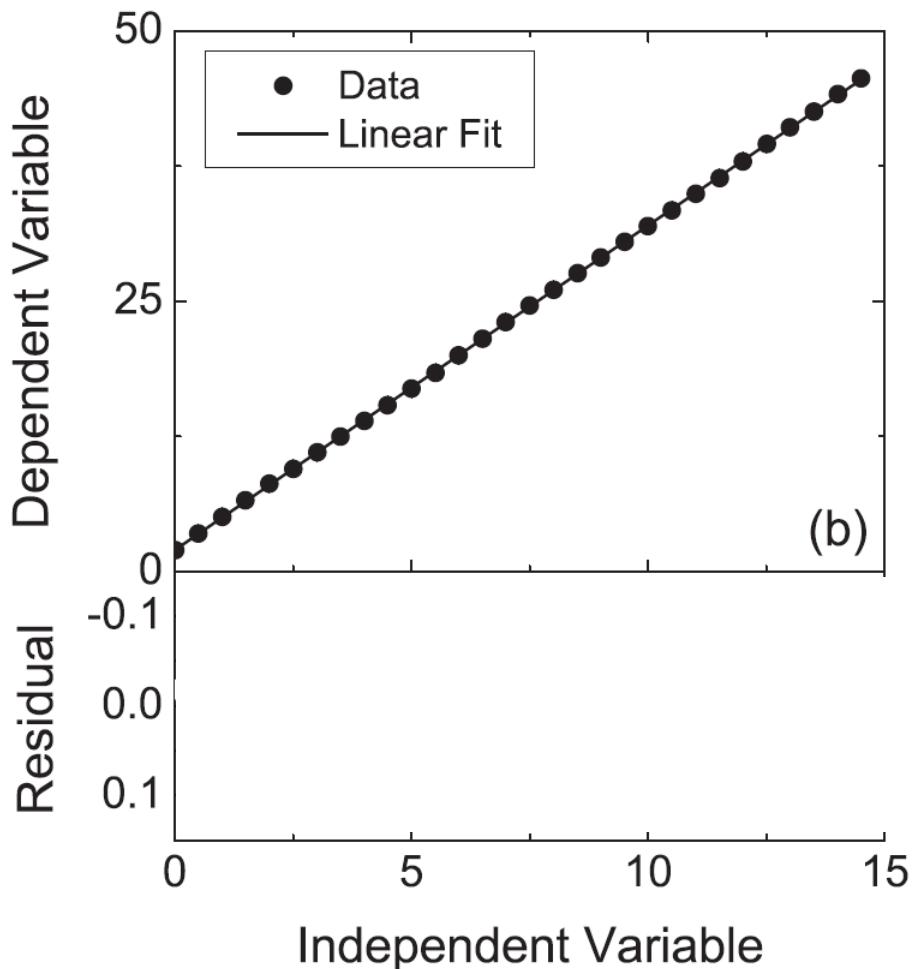
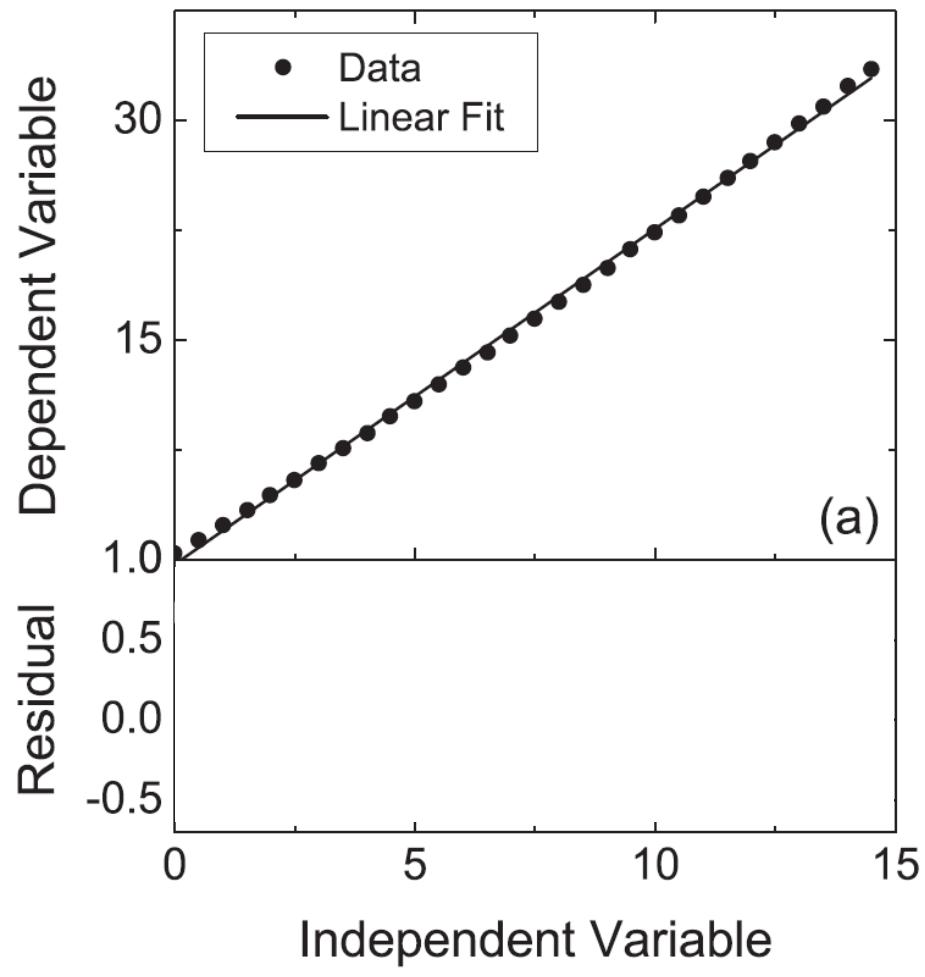
$O_i$  is the observed number of occurrences (integer)

$E_i$  is the expected number of occurrences

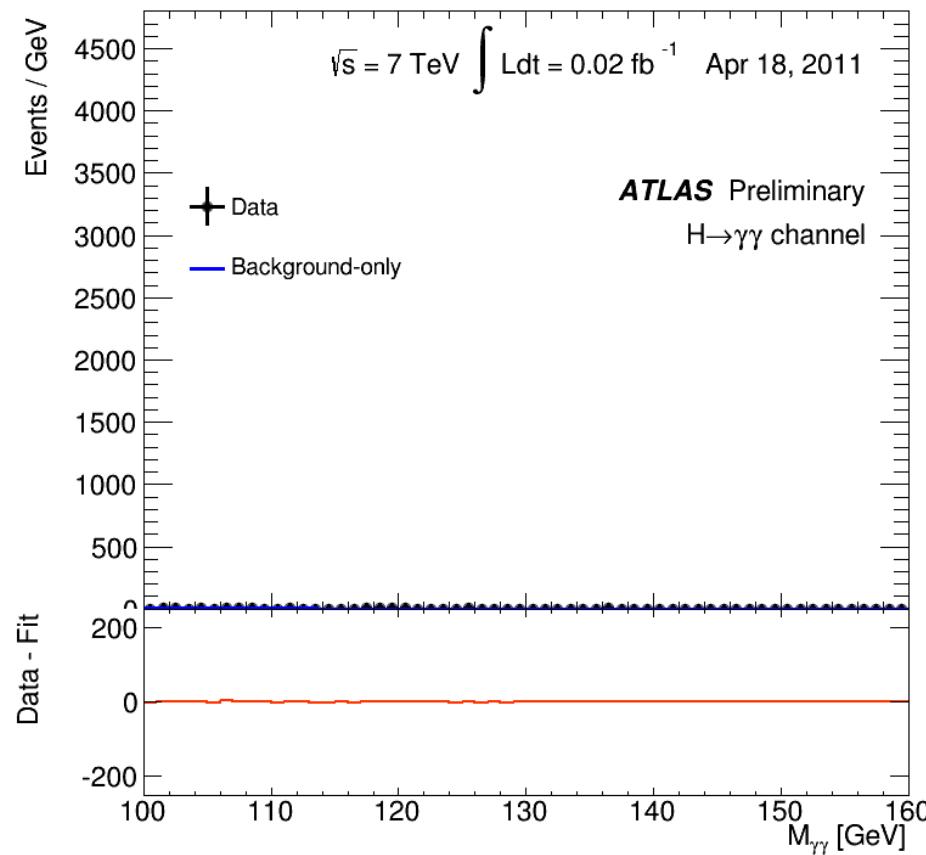
$O_i$  is the observed number of occurrences (integer)

Note that  $\chi^2$  is (again) dimensionless

# After fitting a straight-line graph – plot residuals



Plotting the residuals makes certain features (the existence of a Higgs boson) more apparent



**Observation of a New Particle in the Search for the Standard Model Higgs Boson with the ATLAS Detector at the LHC**  
Phys.Lett. B716 1-29 (2012)

# Minimisation: The Method of Least Squares

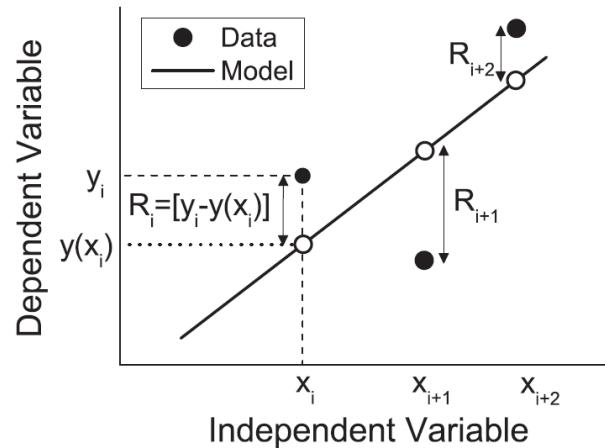
$$S = \sum_i (y_i - y(x_i))^2 = \sum_i R_i^2$$

$$S = \sum_i (y_i - mx_i - c)^2$$

$S$  is a minimum when  $\frac{\partial S}{\partial m} = \frac{\partial S}{\partial c} = 0$

$$\frac{\partial S}{\partial m} = -2 \sum_i (x_i [y_i - mx_i - c]) = 0$$

$$\frac{\partial S}{\partial c} = -2 \sum_i (y_i - mx_i - c) = 0.$$



Measurements and their Uncertainties, p 58

Level 1  
Physics  
Dealing with experimental uncertainty

15th October 2019

# Minimisation: The Method of Least Squares

Therefore the required values of  $m$  and  $c$  are obtained from the two simultaneous equations

$$\begin{aligned} m \sum_i x_i^2 + c \sum_i x_i &= \sum_i x_i y_i \\ m \sum_i x_i + cN &= \sum_i y_i \end{aligned}$$

Note that the last equations demonstrates that the best-fit straight line must go through the point  $(\bar{x}, \bar{y})$ , where  $\bar{x} = \frac{1}{N} \sum_i x_i$ ,  $\bar{y} = \frac{1}{N} \sum_i y_i$ .

Measurements and their Uncertainties, p 58

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In this case by ignoring the  $y$ -error bars we obtain a simple  
**ANALYTIC SOLUTIONS**

for the      **GRADIENT, INTERCEPT** and their  
**UNCERTAINTIES**

$$c = \frac{\sum_i x_i^2 \sum_i y_i - \sum_i x_i \sum_i x_i y_i}{\Delta},$$

$$\Delta = N \sum x_i^2 - \left( \sum x_i \right)^2$$
$$\alpha_{\text{CU}} = \sqrt{\frac{1}{N-2} \sum_i (y_i - mx_i - c)^2}$$

$$m = \frac{N \sum_i x_i y_i - \sum_i x_i \sum_i y_i}{\Delta},$$

$$\alpha_c = \alpha_{\text{CU}} \sqrt{\frac{\sum_i x_i^2}{\Delta}} \quad \alpha_m = \alpha_{\text{CU}} \sqrt{\frac{N}{\Delta}}$$

Measurements and their Uncertainties, p 58

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We use a weighted fit using the  
**Normalised residual**  $R_i$

$$R_i = \frac{y_i - y(x_i)}{\alpha_i}$$

The sum of the squares of the normalised residuals is called  $\chi^2$

This is now a **weighted fit** with the weight  $w_i$  being the inverse of the square of the error bar – points with small errors are more important!

Measurements and their Uncertainties, Pg. 67

$$\chi^2 = \sum_i \frac{(y_i - y(x_i))^2}{\alpha_i^2} = \sum_i R_i^2$$

$$\chi^2 = \sum_i w_i (y_i - mx_i - c)^2$$

$\chi^2$  is a minimum when  $\frac{\partial \chi^2}{\partial m} = \frac{\partial \chi^2}{\partial c} = 0$

Measurements and their Uncertainties, Pg. 67

Therefore the required values of  $m$  and  $c$  are obtained from the two simultaneous equations

$$m \sum_i w_i x_i^2 + c \sum_i w_i x_i = \sum_i w_i x_i y_i$$
$$m \sum_i w_i x_i + c \sum_i w_i = \sum_i w_i y_i$$

Note that the last equations demonstrates that the best-fit straight line must go through the point  $(\tilde{x}, \tilde{y})$ , where  $\tilde{x} = \frac{\sum_i w_i x_i}{\sum_i w_i}$ ,  $\tilde{y} = \frac{\sum_i w_i y_i}{\sum_i w_i}$ .

Measurements and their Uncertainties, Pg. 67

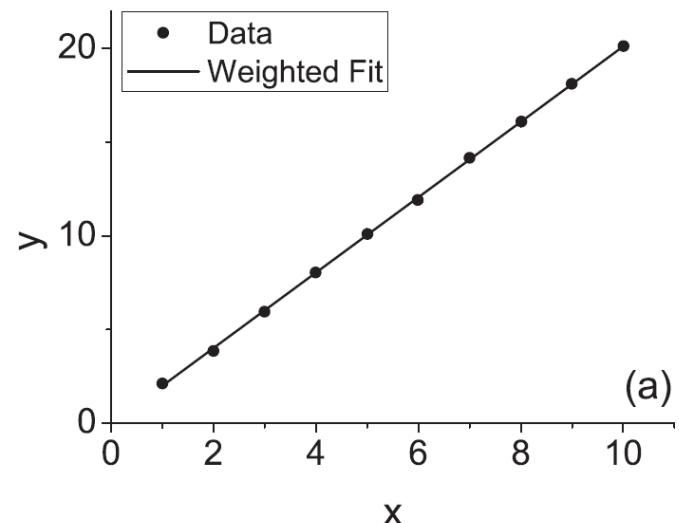
$$c = \frac{\sum_i w_i x_i^2 \sum_i w_i y_i - \sum_i w_i x_i \sum_i w_i x_i y_i}{\Delta'}, \quad \alpha_c = \sqrt{\frac{\sum_i w_i x_i^2}{\Delta'}}$$

$$m = \frac{\sum_i w_i \sum_i w_i x_i y_i - \sum_i w_i x_i \sum_i w_i y_i}{\Delta'}, \quad \alpha_m = \sqrt{\frac{\sum_i w_i}{\Delta'}}$$

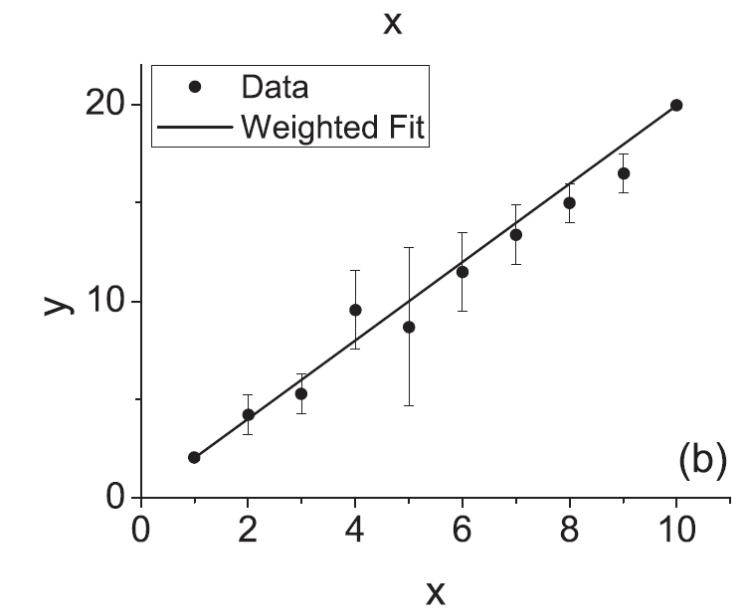
$$\Delta' = \sum_i w_i \sum_i w_i x_i^2 - \left( \sum_i w_i x_i \right)^2 \quad \tilde{w}_i = \alpha_i^{-2}$$

Measurements and their Uncertainties, Pg. 69

# Strategies for a straight-line fit



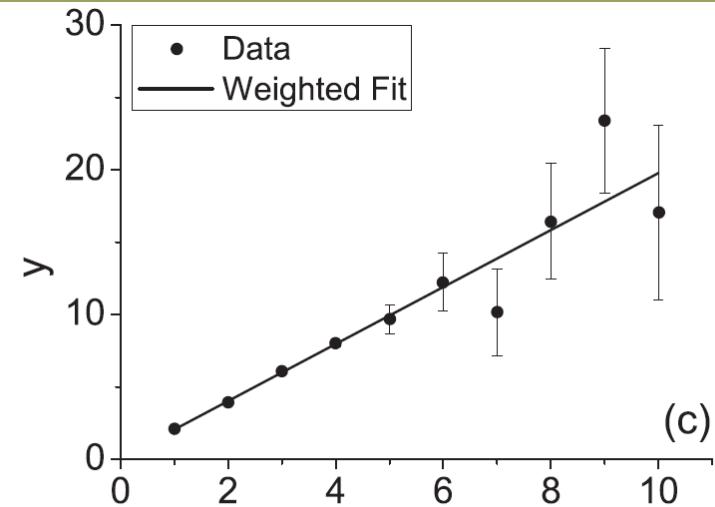
$$m = 2.01 \pm 0.01$$
$$c = -0.02 \pm 0.07$$



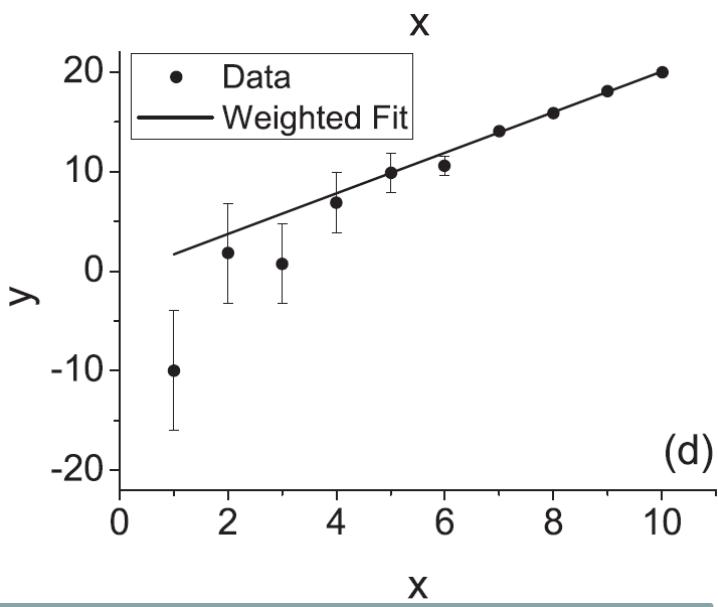
$$m = 1.99 \pm 0.01$$
$$c = 0.04 \pm 0.08$$

Measurements and their Uncertainties, Pg. 71

# Strategies for a straight-line fit



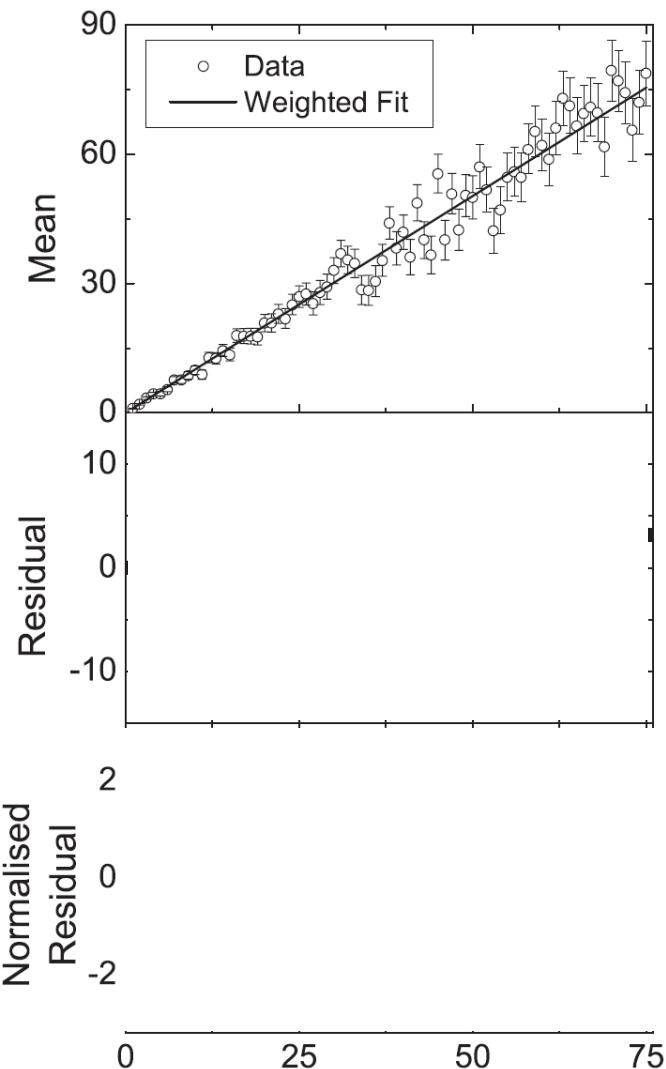
$$m = 1.97 \pm 0.03$$
$$c = 0.09 \pm 0.06$$



$$m = 2.04 \pm 0.04$$
$$c = -0.3 \pm 0.4$$

Measurements and their Uncertainties, Pg. 71

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## normalised residual,

$$R_i = \frac{y_i - y(x_i)}{\alpha_i},$$

65%  
are scattered within  $\pm 1$   
96% are within  $\pm 2$ .

“...error analysis is a participation, rather than a spectator, sport.”

Please attempt homework before 5pm next Monday  
(21<sup>st</sup> October).

For MiSCaDA students via  
<https://notebooks.dmaitre.phyip3.dur.ac.uk/miscada-da/hub/login>

For PhD students either the notebook server OR email me a document