

# Error Analysis (PH1102)

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# Purpose of experimental error analysis

- In teaching lab course (PH1102)

- In research

## ➤ Experimental errors/ uncertainties

- Errors are inevitable: seek the best estimate

- $x_{\text{true}}$ : true value of the physical quantity we measure

- $x_{\text{best}}$ : observed value

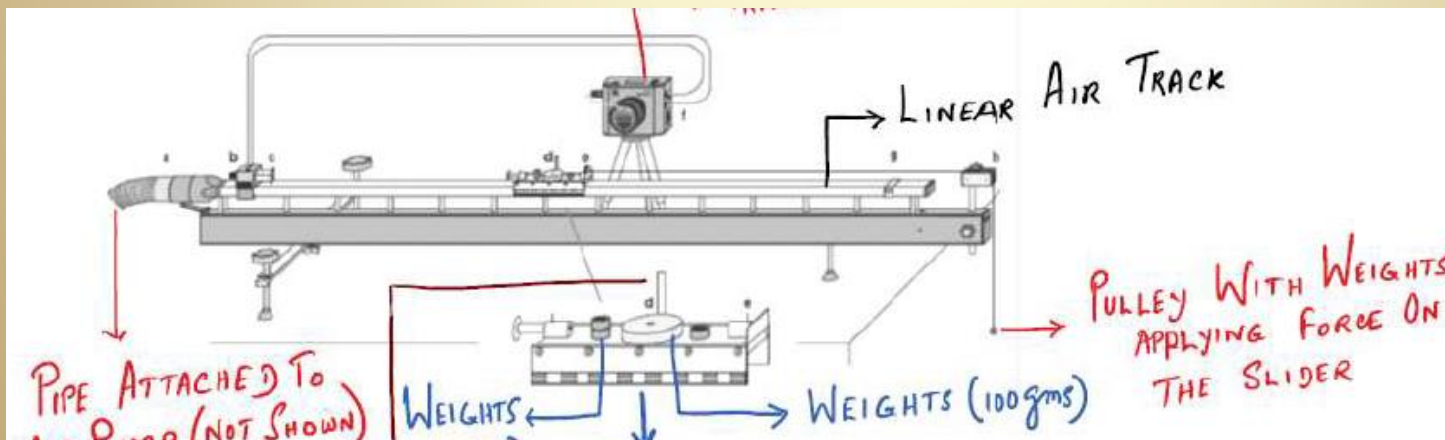
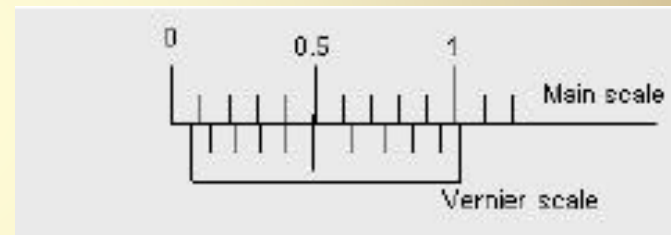
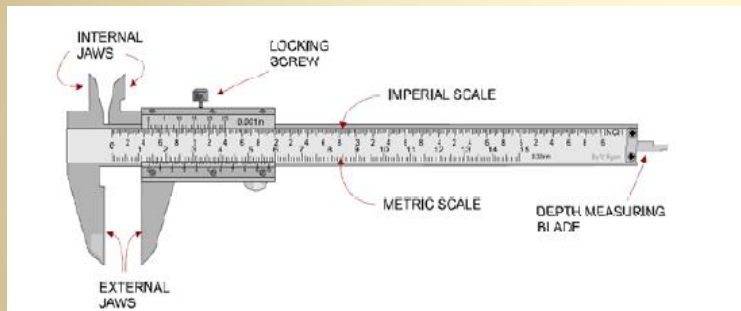
- Error:  $x_{\text{best}} - x_{\text{true}}$

# Sources of experimental errors

1. Real mistakes: wrong measurements

2. Systematic errors

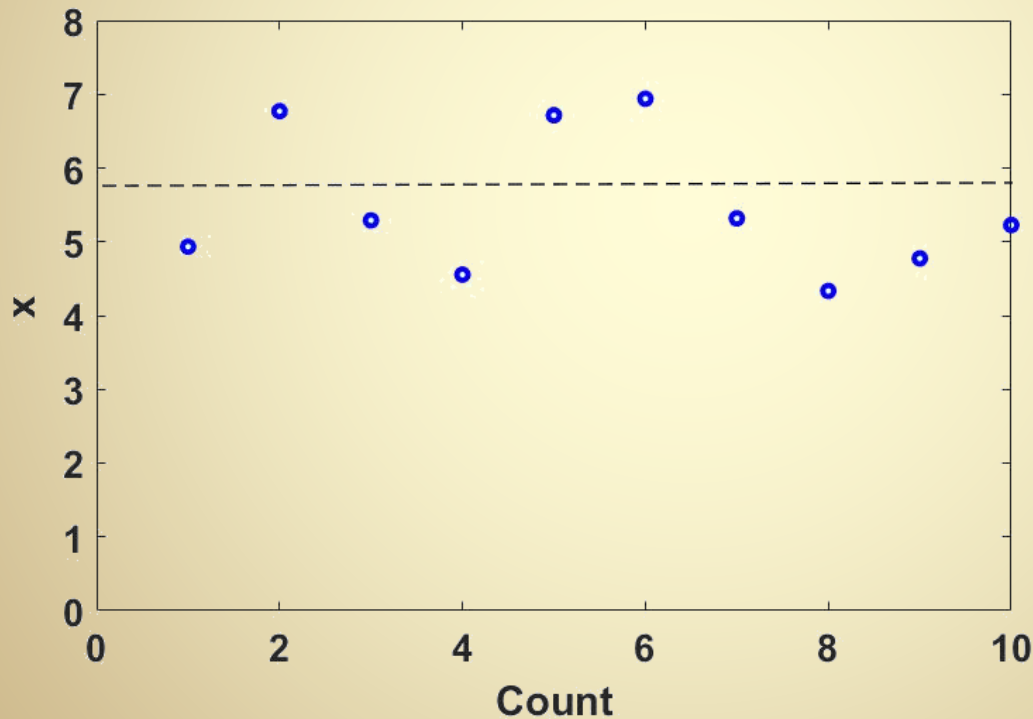
zero error in vernier calipers



Air Track

# Sources of experimental errors

## 3. Random errors:



# Estimation of errors

- **Propagation method:**

Estimation through propagation of primary errors

- **Statistical method:**

Estimation through statistical analysis of data

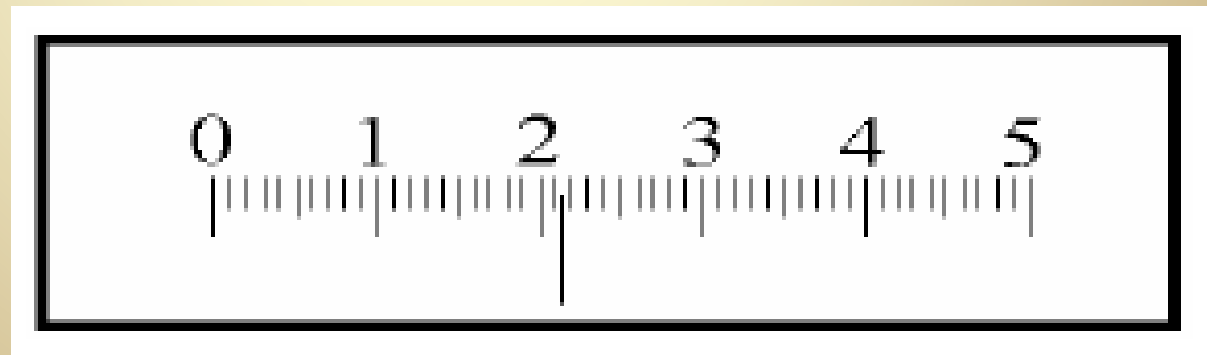
# Estimation of errors

- Estimating a primary error:

- Best estimate and least count

- Estimation of error:  $\delta x = x_{\text{best}} - x_{\text{true}}$

- Effective least count:





# Reporting of results

$$x_{measured} = x_{best} \pm \delta x$$

- upper limit:  $x_{best} + \delta x$

- lower limit:  $x_{best} - \delta x$

# Propagation of estimated errors

$$f = f(x, y); \quad f = x + y$$

We measure:  $x = x_{best} \pm \delta x$

$$y = y_{best} \pm \delta y$$

$$f = x_{best} + y_{best}$$

- error/ uncertainty in  $f$ :  $\delta f = \delta x + \delta y$

- while reporting:  $f \pm \delta f$   
 $= (x_{best} + y_{best}) \pm (\delta x + \delta y)$



# Propagation of estimated errors

$$f = f(x, y); \quad f = x - y$$

We measure:  $x = x_{best} \pm \delta x$

$$y = y_{best} \pm \delta y$$

$$f = x_{best} - y_{best}$$

- error/ uncertainty in  $f$ :  $\delta f = \delta x + \delta y$

- while reporting:  $f \pm \delta f$

$$= (x_{best} - y_{best}) \pm (\delta x + \delta y)$$

$$\delta f = \sqrt{(\delta x)^2 + (\delta y)^2}; \quad (\delta x + \delta y) \geq \sqrt{(\delta x)^2 + (\delta y)^2}$$

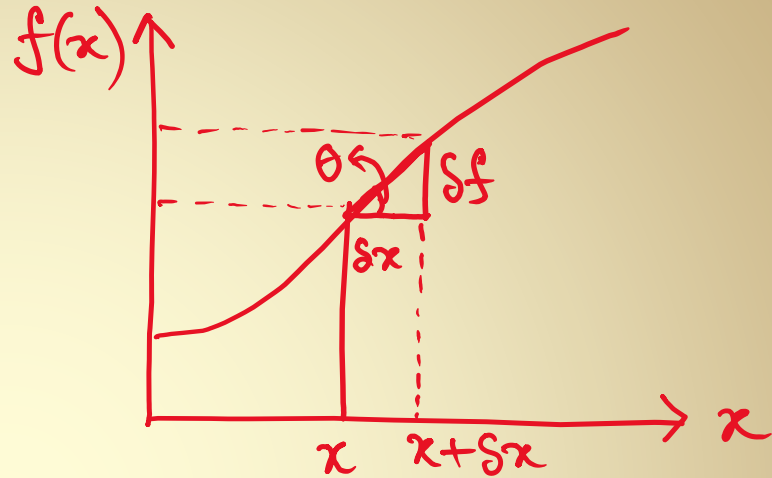
# Propagation of estimated errors

$$f = f(x);$$

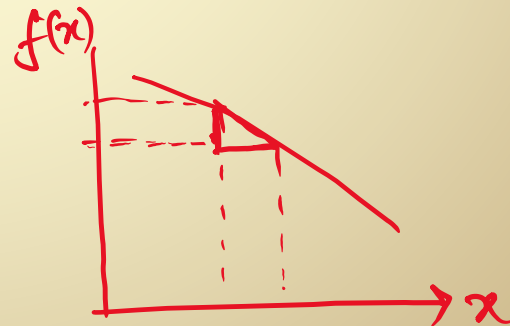
$$\frac{\delta f}{\delta x} = \tan \theta = \frac{df}{dx}$$

$$\delta f = \frac{df}{dx} \delta x$$

$$\delta f = \left| \frac{df}{dx} \right| \delta x$$



$$\lim_{\delta x \rightarrow 0} \frac{f(x + \delta x) - f(x)}{\delta x} = \frac{df}{dx}$$



# Propagation of estimated errors

$$f = f(x, y, z);$$

$$df = \left| \frac{\partial f}{\partial x} \right| \delta x + \left| \frac{\partial f}{\partial y} \right| \delta y + \left| \frac{\partial f}{\partial z} \right| \delta z$$

**Example (1):**  $f = x + y - z;$

$$\delta f = \delta x + \delta y + \delta z$$

# Propagation of estimated errors

Example (2):  $f(x) = x^3$

$$\delta f = \left| \frac{\partial f}{\partial x} \right| \delta x = |3x^2| \delta x$$

Fractional error/ uncertainty:

$$\frac{\delta f}{|f|} = \left| \frac{3x^2}{x^3} \right| \delta x = \frac{3}{|x|} \delta x$$

$$\frac{\delta f}{|f|} = 3 \frac{\delta x}{|x|}$$

# Propagation of estimated errors

**Example (3):**  $f(a, b, c) = \frac{a}{b} - c^{3/2}$

$$\delta f = \left| \frac{\partial f}{\partial a} \right| \delta a + \left| \frac{\partial f}{\partial b} \right| \delta b + \left| \frac{\partial f}{\partial c} \right| \delta c$$

$$= \left| \frac{1}{b} \right| \delta a + \left| \frac{a}{b^2} \right| \delta b + \left| \frac{3}{2} c^{1/2} \right| \delta c$$

# Propagation of estimated errors

**Example (4):**  $f(a, b, c) = \frac{ab}{c} - a^2$

$$\Delta f = \frac{\partial f}{\partial a} \Delta a + \frac{\partial f}{\partial b} \Delta b + \frac{\partial f}{\partial c} \Delta c$$

$$= \left(\frac{b}{c}\right) \Delta a - (2a) \Delta a + \left(\frac{a}{c}\right) \Delta b + \left(\frac{ab}{-c^2}\right) \Delta c$$

$$\Delta f = \left| \frac{\partial f}{\partial a} \right| \Delta a + \left| \frac{\partial f}{\partial b} \right| \Delta b + \left| \frac{\partial f}{\partial c} \right| \Delta c$$

$$\Delta f = \left| \frac{b}{c} - 2a \right| \Delta a + \left| \frac{a}{c} \right| \Delta b + \left| \frac{ab}{c^2} \right| \Delta c$$



# Propagation of estimated errors

Example (5):  $f(a, b, c) = \frac{ab}{c}$

$$\begin{aligned} \Delta f &= \left| \frac{\partial f}{\partial a} \right| \Delta a + \left| \frac{\partial f}{\partial b} \right| \Delta b + \left| \frac{\partial f}{\partial c} \right| \Delta c \\ &= \left| \frac{b}{c} \right| \Delta a + \left| \frac{a}{c} \right| \Delta b + \left| \frac{ab}{c^2} \right| \Delta c \end{aligned}$$

Now,  $\frac{\Delta f}{|f|} = \left| \frac{1}{a} \right| \Delta a + \left| \frac{1}{b} \right| \Delta b + \left| \frac{1}{c} \right| \Delta c$

# Propagation of estimated errors

Example (6):  $f(a, b, c) = \frac{ab}{c}$

**Prescription: take a log and differentiate**

$$\log f = \log a + \log b - \log c$$

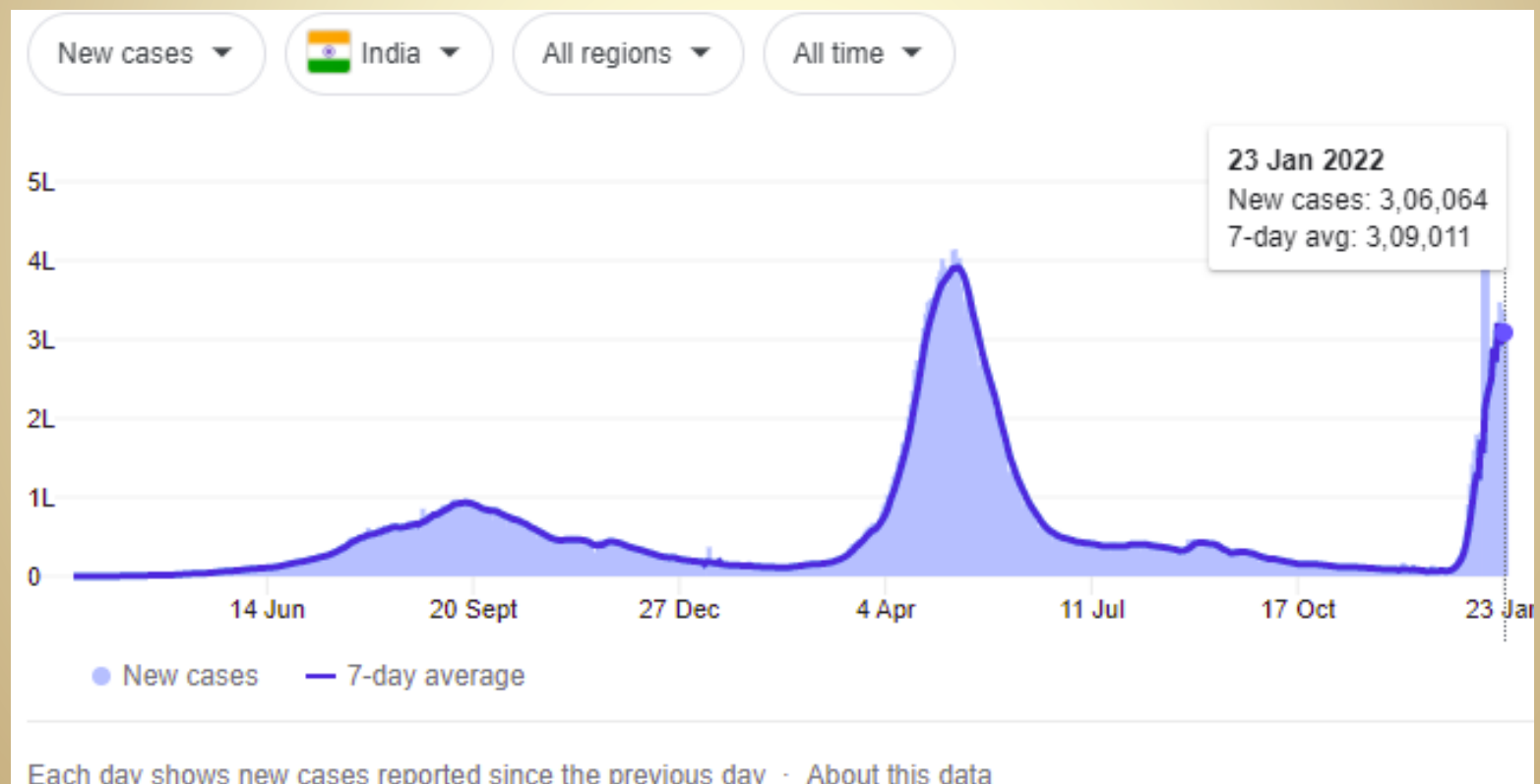
differentiate :  $\frac{df}{f} = \frac{da}{a} + \frac{db}{b} - \frac{dc}{c}$

Error form :

$$\frac{\delta f}{|f|} = \frac{\delta a}{|a|} + \frac{\delta b}{|b|} + \frac{\delta c}{c}$$

# Random processes: Statistical analysis of errors

**COVID-19: Statistics data: Number of new cases**  
(14 March 2020- 23 Jan 2022)

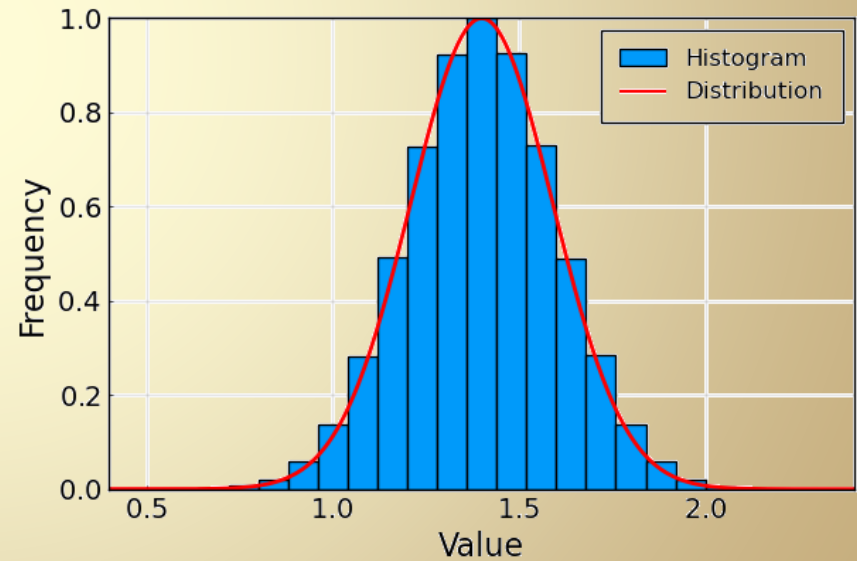
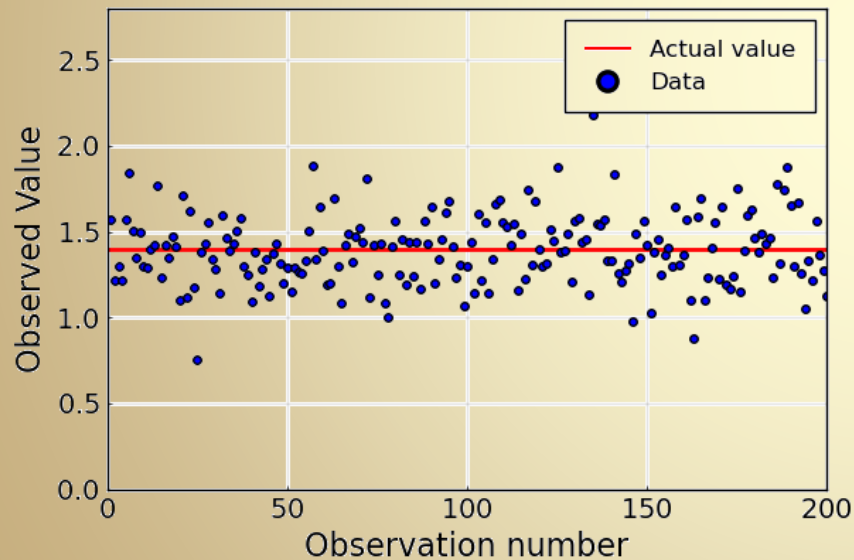


(Source: Google: Johns Hopkins University:  
<https://systems.jhu.edu/research/public-health/ncov/>)

# Statistical analysis of errors

Measurement of the quantity  $x$  made  $N$  times:

$$x_1, x_2, x_3, \dots, x_{N-1}, x_N$$



# Statistical analysis of errors

**Measurement of the quantity  $x$  made  $N$  times:**

$$x_1, x_2, x_3, \dots, x_{N-1}, x_N$$

**Mean:**  $\bar{x} = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{i=1}^N x_i$

**Deviation for single measurement :**  $d_i = x_i - \bar{x}$

**Average deviation:**  $d = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{i=1}^N (x_i - \bar{x})$

$$= \lim_{N \rightarrow \infty} \frac{1}{N} \sum x_i - \lim_{N \rightarrow \infty} \frac{1}{N} \sum \bar{x}$$

$$= \bar{x} - \bar{x} = 0$$

# Statistical analysis of errors

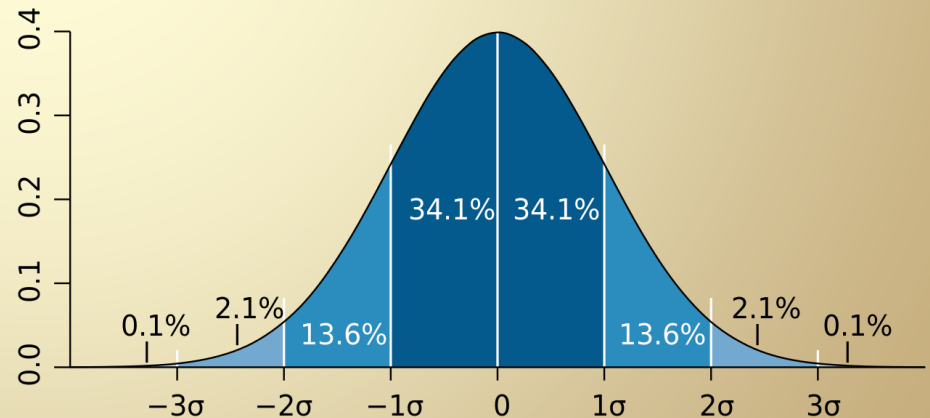
**Variance: mean of deviation squared:**

$$\begin{aligned}\sigma^2 &= \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{i=1}^N (x_i - \bar{x})^2 \\ &= \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{i=1}^N x_i^2 - (\bar{x})^2 = \overline{x_i^2} - (\bar{x})^2 = \langle x^2 \rangle - \langle x \rangle^2\end{aligned}$$

**Standard deviation (s.d.): square root of the variance:**

$$\sigma = \sqrt{\langle x^2 \rangle - \langle x \rangle^2}$$

• while reporting:  
(mean  $\pm$  s. d.)



**Normal Distribution (Source: Wikipedia)**



# Method of least square curve fitting

**N data points:**  $(x_i, y_i)$

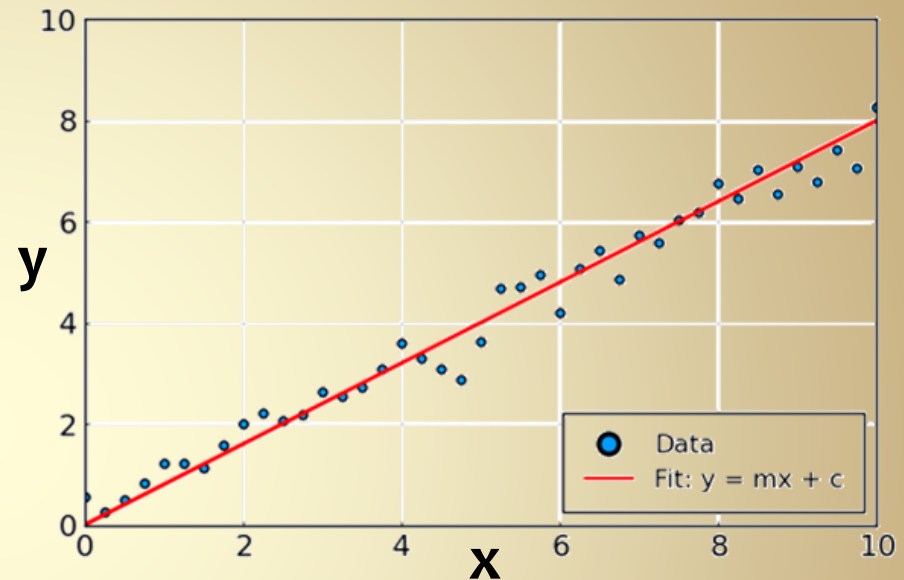
$$y = mx + c$$

**Deviation between observed values and calculated values:**

$$\delta y_i = y_i - (mx_i + c)$$

**Minimizing the mean square deviation:**

$$S = \frac{1}{n} \sum_{i=1}^n (y_i - mx_i - c)^2$$



# Reporting the error in experiments

- **Accuracy:** how close to true/given value
- **Precision:** how well the value is determined (regardless of true value)
- **Example:**  $X_{\text{true}} = 30$  (unit)
  - $X_{\text{measured}} = 29 \pm 6$  : accurate but imprecise (random/statistical error)
  - $X_{\text{measured}} = 22 \pm 1$  : precise but inaccurate (might have systematic error)

# Reporting the error in experiments

- **while reporting:**  $f \pm \delta f$  (with appropriate unit)

**Significant digits:** example: (a)  $3.45 \pm 0.22$ ;

(b)  $3.4587 \pm 0.22$ ; **X**

(c)  $10.82 \pm 0.01857$ ; **X**

- Check precision and then round it to significant digit
- **Compare with the given/true value; percentage error**
- **Discussion on sources of errors**



**Questions??**