

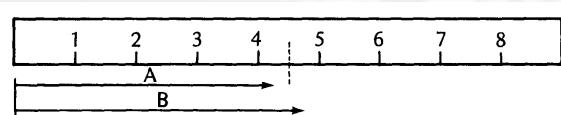
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### Errors of observation

- **Error:** the difference between the observed value of any physical quantity and the 'actual' value.
- **Human error:** due to observational reasons.
- **Systematic error:** due to the instruments and it determines the accuracy of the reading.
- **Random error:** due to environmental reasons or nature of the measurement
- Note difference between **accuracy** and **precision**.

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### Estimation of Errors



- The maximum possible error is 0.5 cm
- Therefore  $A = 4.0 \pm 0.5$  and  $B = 5.0 \pm 0.5$
- Absolute error is 0.5 cm, but relative (percentage) error changes.
- Relative error of A is  $0.5/4.0 \times 100\% = 12.5\%$
- If C is 50, then relative error of C is  $0.5/50 \times 100\% = 1\%$ .

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### Propagation of Errors

$$\text{If } x = a + b, \text{ then } \Delta x = \Delta a + \Delta b$$

$$\text{If } x = a - b, \text{ then } \Delta x = \Delta a + \Delta b$$

$$\text{If } x = a \times b, \text{ then } \frac{\Delta x}{x} = \frac{\Delta a}{a} + \frac{\Delta b}{b}$$

$$\text{If } x = \frac{a}{b}, \text{ then } \frac{\Delta x}{x} = \frac{\Delta a}{a} + \frac{\Delta b}{b}$$

$$\text{If } x = ka, \text{ where } k \text{ is a constant, then } \Delta x = k\Delta a$$

In general, if  $u = u(x, y, z, \dots)$ ,

$$\text{then } \delta u = \frac{\partial u}{\partial x} \delta x + \frac{\partial u}{\partial y} \delta y + \frac{\partial u}{\partial z} \delta z + \Lambda$$

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## Examples

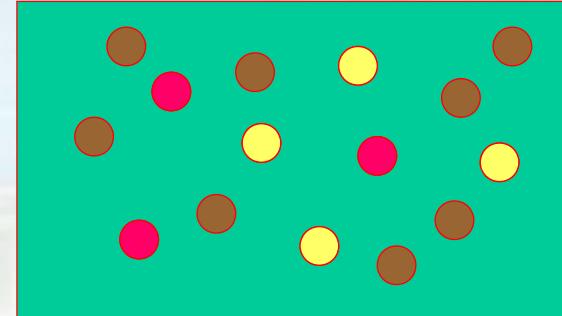
Given  $a = 10 \pm 2$ ,  $b = 7 \pm 3$ .

- $d = a + b = 17 \pm (2+3)$ .
- $e = a - b = 3 \pm (2+3)$ .
- $f = a \times b = 70 \pm \Delta f$   
where  $\Delta f = (2/10 + 3/7) \times 70 = 44$ .
- $g = a/b = 1.43 \pm \Delta g$   
where  $\Delta g = (2/10 + 3/7) \times 1.43 = 0.90$ .

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## Statistical Nature of Random Error

- No “accurate” value
- Mean value with uncertainty



e.g. radioactive decay

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## Terminologies

- Expectation value
- Frequency distribution function
- Mean value

– For discrete x values, the mean

$$\bar{x} = \frac{\sum_{i=1}^n x_i}{n}$$

– For continuous distribution,

$$\bar{x} = \frac{\int_{x_1}^{x_2} xf(x)dx}{\int_{x_1}^{x_2} f(x)dx}$$

– The mean value is only a sample mean, it may be quite different from the true mean value.

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## Terminologies

- Deviation =  $x_i - \bar{x}$
- Sample variance  $s^2$

$$s^2 = \frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n-1} \quad \text{or} \quad s^2 = \frac{\int_{x_1}^{x_2} (x - \bar{x})^2 f(x)dx}{\int_{x_1}^{x_2} f(x)dx}$$

- Standard deviation =  $(variance)^{1/2} = s$
- Normalization – a distribution function is said to be normalized if the total area under the distribution curve is equal to 1. i.e.

$$\int_{-\infty}^{\infty} f(x)dx = 1$$

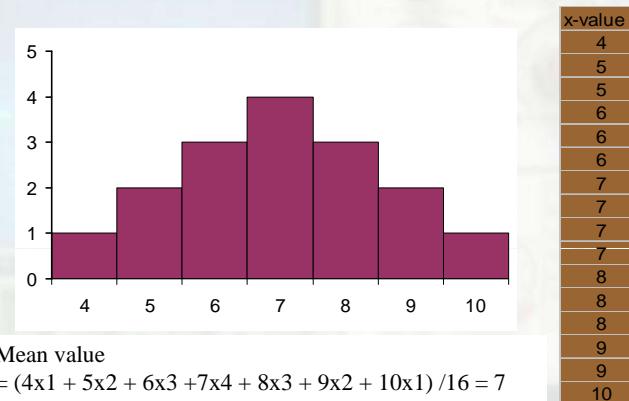
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## Statistical Models

- Binomial distribution
- Poisson distribution
- Gaussian or Normal distribution
  - standard deviation  $\sigma = (\text{mean})^{1/2}$
  - distribution is symmetric about mean

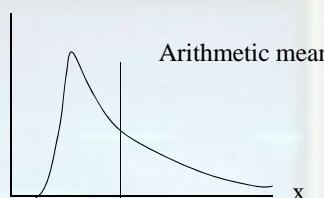
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## Example - mean

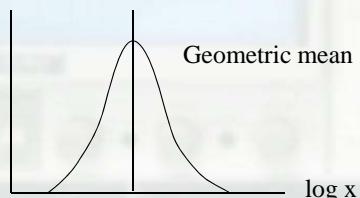


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## Arithmetic & Geometric Mean



e.g. the size distribution of atmospheric aerosols is Guassian when the size is plotted in a log scale.



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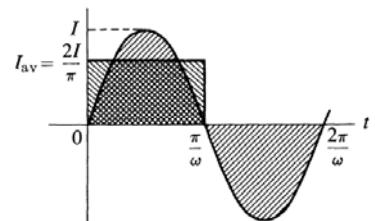
## Mean and Root-mean-squared Value

- R.M.S. equals the square root of the mean of the squared value.
- E.g. for five measurement results: 9.81, 9.80, 10.0, 9.60, 9.71
  - Mean  $= (9.81 + 9.80 + 10.0 + 9.60 + 9.71)/5 = 9.784$
  - R.M.S. =

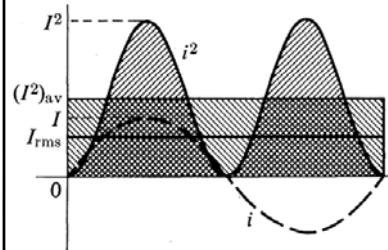
$$\sqrt{\frac{9.81^2 + 9.80^2 + 10.0^2 + 9.60^2 + 9.71^2}{5}} = 9.785$$

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## R.M.S. of AC



$$I_{av} = \frac{\omega}{\pi} \int_0^{\pi/\omega} I \sin \omega t dt = \frac{2I}{\pi}$$



$$i^2 = I^2 \sin^2 \omega t = \frac{1}{2} I^2 - \frac{1}{2} I^2 \cos 2\omega t$$

$$(I^2)_{av} = \frac{I^2}{2}$$

$$I_{rms} = \sqrt{(I^2)_{av}} = \frac{I}{\sqrt{2}}$$

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## Example - standard deviation

What is the standard deviation of the previous distribution?

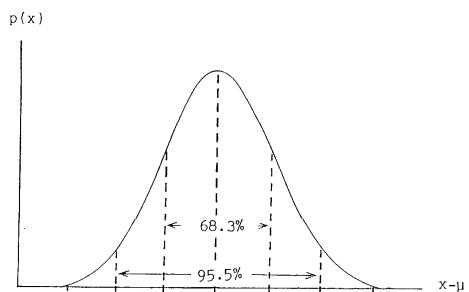
$$1. \quad \sigma = \sqrt{\frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n-1}} = 1.6$$

$$2. \quad \sigma = \sqrt{\bar{x}} = \sqrt{7} = 2.6$$

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## Confidence Levels

The confidence of observing a value within the range  $\mu-r$  to  $\mu+r$  is equal to the probability of its occurrence given by  $\int_{\mu-r}^{\mu+r} P(x)dx$ .



Range r	0.675 $\sigma$	$\sigma$	2 $\sigma$	3 $\sigma$
Confidence	50 %	68.3 %	95.4 %	99.7 %

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## Example - relative error

$$X = 100$$

$$X = 10000$$

$$\sigma = 10$$

$$\sigma = 100$$

$$\sigma/X = 0.1 = 10 \%$$

$$\sigma/X = 0.01 = 1 \%$$

$$X = 100 \pm 10$$

$$X = 10000 \pm 100$$

at 68.3 % C.L.

at 68.3 % C.L.

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## Useful statistical equations

$$\sigma_x = \sqrt{\bar{x}}$$

$$\frac{\sigma_x}{\bar{x}} = \frac{1}{\sqrt{\bar{x}}}$$

$$\sigma_R = \frac{\sigma_x}{t} = \sqrt{\frac{R}{t}}$$

$$\frac{\sigma_R}{R} = \frac{1}{\sqrt{Rt}}$$

$$\sigma_{\bar{x}} = \sqrt{\frac{\bar{x}}{n}}$$

$$\frac{\sigma_{\bar{x}}}{\bar{x}} = \frac{1}{\sqrt{n\bar{x}}}$$

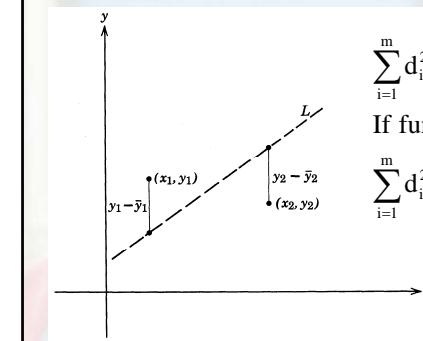
If  $u = u(x, y, z, \dots)$  and if the errors are individually small and symmetric about zero, then

$$\sigma_u^2 = \left( \frac{\partial u}{\partial x} \right)^2 \sigma_x^2 + \left( \frac{\partial u}{\partial y} \right)^2 \sigma_y^2 + \left( \frac{\partial u}{\partial z} \right)^2 \sigma_z^2 + \Lambda$$

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## Least Squares Approximations

- A curve fitting technique
- Minimize the sum of the squares of the deviations



$$\sum_{i=1}^m d_i^2 = \sum_{i=1}^m (y_i - \bar{y}_i)^2$$

If function is linear, then  $\bar{y} = a_1 x + a_0$

$$\sum_{i=1}^m d_i^2 = \sum_{i=1}^m (y_i - a_1 x_i - a_0)^2$$

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## Least Square Fit

To minimize S with respect to a & b

$$\frac{\partial S}{\partial a_0} = \sum_{i=1}^m 2(y_i - a_1 x_i - a_0)(-1) = 0$$

$$\frac{\partial S}{\partial a_1} = \sum_{i=1}^m 2(y_i - a_1 x_i - a_0)(-x_i) = 0$$

Rearranging terms we get

$$ma_0 + (\sum x_i)a_1 = \sum y_i \quad \text{and} \quad (\sum x_i)a_0 + (\sum x_i^2)a_1 = \sum x_i y_i$$

Solving the two linear equations, we have

$$a_0 = \frac{\sum y_i \sum x_i^2 - \sum x_i \sum x_i y_i}{m \sum x_i^2 - (\sum x_i)^2}$$

$$a_1 = \frac{m \sum x_i y_i - \sum x_i \sum y_i}{m \sum x_i^2 - (\sum x_i)^2}$$

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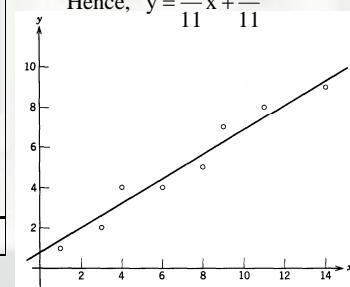
## Example

x	y	$x^2$	xy
1	1	1	1
3	2	9	6
4	3	16	16
6	3	36	24
8	4	64	40
9	7	81	63
11	8	121	88
14	9	196	126
sums:		56	40
		524	364

$$a_0 = \frac{40 \times 524 - 56 \times 364}{8 \times 524 - (56)^2} = \frac{6}{11}$$

$$a_1 = \frac{8 \times 364 - 56 \times 40}{8 \times 524 - (56)^2} = \frac{7}{11}$$

$$\text{Hence, } y = \frac{7}{11}x + \frac{6}{11}$$



## Other curves

Polynomial:  $y = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0$

Hyperbola:  $y = \frac{1}{a_0 + a_1 x}$

Let  $z = \frac{1}{y}$ , then  $z = a_0 + a_1 x$

Exponential:  $y = a(b^x) \Rightarrow \log y = \log a + x \log b$

Geometric:  $y = ax^b \Rightarrow \log y = \log a + b \log x$

Trigonometric:  $y = a_0 + a_1 \cos \omega x$

or more generally  $y = a_0 + \sum_{k=1}^n a_k \cos(k\omega x) + \sum_{k=1}^n b_k \sin(k\omega x)$