

# BASICS OF ENGINEERING MEASUREMENTS

### (AGE 2340)

# Dr. Nasser Mohamed Shelil

Assistant Professor, Mechanical Engineering Dept., College of Applied Engineering, King Saud University

B.Sc. & M.Sc. , Suez Canal University; PhD, Cardiff University/UK



**Calibration & Uncertainty Analysis** 

Applied Mechanical Engineering Program

# **Basic Terminology of Measurement**

### Measurement

The International Vocabulary of Basic and General Terms in Metrology, using International Organization for Standardization (ISO) norms, has defined measurement as "a set of operations having the object of determining the value of a quantity". In other words, a measurement is the evaluation of a quantity made after comparing it to a quantity of the same type which we use as a "unit".



# **Basic Terminology of Measurement**

### Metrology

the science and "grammar" of measurement is defined as "the field of knowledge concerned with measurement". Standardized measurement units mean that scientific and economic figures can be understood, reproduced, and converted with a high degree of certitude.



Applied Mechanical Engineering Program Basics of Engineering Measurements *Chapter 2* Uncertainty

# **Basic Terminology of Measurement**

### Instrumentation

refers to a group of permanent systems which help us measure objects. In this sense, instruments and systems of measurement constitute the "tools" of measurement and metrology.



Applied Mechanical Engineering Program

# **Basic Terminology of Measurement**

### Load Effects

- ✓ measurement operations may require connection or without contact.
- This linking of an instrument to an object or site of investigation means that a transfer of energy and/or information termed "a load effect" takes place.
- ✓ An example of this is shown by the insertion of a measuring probe into a cup of tea which takes some heat from the tea, leading to a difference between the "true" value and the value to be measured.



Applied Mechanical Engineering Program Basics of Engineering Measurements



6

Chapter 2

UNCERTAINT

# Calibration

- The relationship between the value of the input to the measurement system and the system's indicated output value is established during calibration of the measurement system.
- The known value used for the calibration is called the standard.
- The quantity to be measured being the **measurand**, which we call *m*, the sensor must convert *m into an electrical variable called s. The* expression s = F(m) is established by calibration. By using a standard or unit of measurement, we discover for these values of *m* (*m1*, *m2*... *mi*) electrical signals sent by the sensor (s1, s2... si) and we trace the curve s(m), called the sensor calibration curve.



# **Accuracy & Precision**

- Accuracy of a system can be estimated during calibration. If the input value of calibration is known exactly, then it can called the *true value*. The accuracy of a measurement system refers to its ability to indicate a true value exactly.
- Accuracy : It is the ability of instrument to tell the truth
- Accuracy is related to *absolute error*, ε:
   ε = true value indicated value
   from which the percent accuracy is found by :

$$A = \left(1 - \frac{|\varepsilon|}{\text{true value}}\right) x 100$$



Applied Mechanical Engineering Program

UNCERTAINTY

# **Accuracy & Precision**

- **Precision:** or repeatability of a measuring system refers to the ability of the system to indicate a particular value upon repeated but independent applications of a specific value input. Precision of a measurement describes the units used to measure something.
- **Precision** : It is the ability of the instrument to give the same output for the same input under the same conditions







# **Term Used in Instrument Rating**

- **Resolution:** The smallest increment of change in the measured value that can be determined from the instrument's readout scale. The resolution is often on the same order as the precision; sometimes it is smaller.
- **Sensitivity:** The change of an instrument's output per unit change in the measured quantity. Typically, an instrument with higher sensitivity will have also finer resolution, better precision, and higher accuracy.
- **Range:** The proper procedure for calibration is to apply known inputs ranging from the minimum to the maximum values for which the measurement system is to be used. These limits the operating range of the system.

Applied Mechanical Engineering Program

Chapter

# **Error Classifications**

### > 1. Systematic, Fixed or Bias Errors:

- Insidious in nature, exist unnoticed unless deliberately searched.
- Repeated readings to be in error by the same amount.
- Not susceptible to statistical analysis.
  - Calibration errors
  - Certain consistently recurring human error
  - Technique error
  - Uncorrected loading error
  - Limitations of system resolution

Applied Mechanical Engineering Program

# **Error Classifications**

### 2. Precision or Random Errors:

- Distinguished by their lack of consistency. Usually (not always) follow a certain statistical distribution.
- In many instances very difficult to distinguish from bias errors.
  - Error stemming from environmental variations
  - Certain type of human error
  - Error resulting from variations in definition.

Applied Mechanical Engineering Program

Chapter 2
UNCERTAINTY

14

Chapter 2

UNCERTAINT

# **Error Classifications**

### > 3. Illegitimate Errors

**Illegitimate Errors** are simply mistakes on the part of experimenter.

- Can be eliminated through the exercise of care and repetition of the measurement.
  - Blunders and mistakes
  - Computational errors
  - Chaotic errors.



Applied Mechanical Engineering Program

*Chapter 2* 15 **UNCERTAINTY** 

### Ideal Distinction: bias versus random errors

**Bias error** is a systematic inaccuracy caused by a mechanism that we can (ideally) control. We might be able to adjust the way measurement are taken in an attempt to reduce bias errors. We can try to correct bias errors by including adjustments in our data analysis *after* the measurements are taken.

Random error is a non-repeatable inaccuracy caused by an unknown or an uncontrollable influence. Random errors introduce scatter in the measured values, and propagate through the data analysis to produce scatter in values computed from the measurements. Ideally random errors establish the limits on the precision of a measurement, not on the accuracy of a measurement.

Applied Mechanical Engineering Program

Chapter 2 16 UNCERTAINTY

# **Bias & Precision Errors**

### Bias (Systematic) Error

- ✓ is the difference between the average value in a series of repeated calibration measurements and the true value.
- Systematic error causes an offset between the mean value of the data set and its true value
- ✓ systematic error = average value true value

### • Precision Error (Random error)

- ✓ is a measure of the random variation found during repeated measurements.
- ✓ random error = reading average of readings
- Random error causes a random variation in measured values found during repeated measurements of a variable

Chapter 2

**UNCERTAIN1** 

17

□ Both random and systematic errors affect a system's accuracy.

Applied Mechanical Engineering Program









# Uncertainty

- The uncertainty is a numerical estimate of the possible range of the error in a measurement.
- In any measurement, the error is not known exactly since the true value is rarely known exactly.
- that the error is within certain bounds, a plus or minus range of the indicated reading







# <section-header> Notation 2 denotes the addition of a set of values 4 is the variable usually used to represent the individual data values 6 represents the number of data values in a sample 7 represents the number of data values in a population

# Definitions

**Mean (Average):** the number obtained by adding the values and dividing the total by the number of values.

**Median:** the middle value when the original data values are arranged in order of increasing (or decreasing) magnitude.

Variance: It is the expectation of the squared deviation of a random variable from its mean

**Standard Deviation:** a measure of variation of the scores about the mean (average deviation from the mean)

Applied Mechanical Engineering Program Basics of Engineering Measurements

### **Sample and Population Mean**



where  $\sum X$  is sum of all data values

 $N \ \mbox{is number}$  of data items in population

 ${f n}$  is number of data items in sample

Applied Mechanical Engineering Program	Chapter 27			
<b>BASICS OF ENGINEERING MEASUREMENTS</b>	STATISTICS			

Sample and Population Variance  

$$s^{2} = \frac{\sum (x - \overline{x})^{2}}{n - 1} \text{ Sample} \\ \sigma^{2} = \frac{\sum (x - \mu)^{2}}{N} \text{ Population} \\ \text{Variance}$$

### **Sample and Population Standard Deviations**

$$s_x = \sqrt{\frac{\sum_{i=1}^{n} (x_i - \bar{x})^2}{n-1}}$$

- $\eta =$  The number of data points
- $ar{x}=$  The mean of the  $x_i$
- $x_i =$  Each of the values of the data

Applied Mechanical Engineering Program

\_\_\_\_\_

$$\sigma = \sqrt{rac{1}{N}\sum_{i=1}^N (x_i-\mu)^2},$$

Chapter .. Statistics

# **Normal Distribution**

The normal distribution is a continuous probability distribution that has a bell-shaped probability density function, known as the Gaussian function, or informally, the bell curve.





Basics of Engineering Measurements





# **Calculation of bias Uncertainty**

Manufacturers' Specifications

- If you can't do better, you may take it from the manufacturer's specs.
- Accuracy %FS, %reading, offset, or some combination (e.g., 0.1% reading + 0.15 counts)
  - Unless you can identify otherwise, assume that these are at a 95% confidence interval

Independent Calibration

May be deduced from the calibration process

Applied Mechanical Engineering Program Basics of Engineering Measurement

# **Calculation of precision Uncertainty**

- Use Statistics to Estimate Random Uncertainty
- □ Mean: the sum of measurement values divided by the number of measurements.

$$\bar{\mathbf{x}} = \frac{1}{N} \sum_{i=1}^{N} \mathbf{x}_{i}$$

- **Deviation:** the difference between a single result and the mean of many results.  $d_i = x_i \overline{x}$
- Standard Deviation: the smaller standard deviation is the more precise data
  - Large sample size

$$\sigma = \left[\frac{1}{n}\sum(x_i - \bar{x})^2\right]^2$$

E.

Small sample size (n<30)</li>
 Slightly larger value

$$\sigma_{s} = \left\lfloor \frac{1}{n-1} \sum (x_{i} - \bar{x})^{2} \right\rfloor^{2}$$

Applied Mechanical Engineering Program

Chapter 2 Uncertainty

34

Chapter 2

UNCERTAINT

# **Calculation of precision Uncertainty**

The precision uncertainty has to be calculated by estimating mean of the sample reading and standard deviation of the sample.

Let  $x_m$  be the mean and  $S_x$  be the standard deviation of the sample for which n repetitions were made.

Then the precision uncertainty will be given by :  $U_p = \pm t \frac{\alpha}{2} \vartheta \frac{s_x}{\sqrt{n}}$ 

Each of the individual measurement variables  $(X_1, X_2, ..., X_K)$  is subject to Several precision errors.

The bias limits for each of these elemental sources are combined in some manner to obtain the overall bias limit  $(U_{P1}, U_{P2}, ..., U_{PK})$  for each variable.

Applied Mechanical Engineering Program	Chapter 2		
BASICS OF ENGINEERING MEASUREMENTS	UNCERTAINTY		

# **Calculation of precision Uncertainty**

How to calculate:  $U_p = \pm t \frac{\alpha}{2} \vartheta \frac{s_x}{\sqrt{n}}$ 

By using T - distribution:

n = number of samples v = n - 1 = degree of freedom $\alpha = 1 - c , c: is the confidence$ 

$$S_x = \sigma_s = \left[\frac{1}{n-1}\sum(x_i - \bar{x})^2\right]^2$$

$$\overline{X} \pm t_{\alpha/2, n-1} \frac{s}{\sqrt{n}}$$

Applied Mechanical Engineering Program

UNCERTAINT

*Chapter 2* 



### **Student t-distribution (small sample sizes)**

- The t-distribution was formulated by W.S. Gosset, a scientist in the Guinness brewery in Ireland, who published his formulation in 1908 under the pen name (pseudonym) "Student."
- □ The t-distribution looks very much like the Gaussian distribution, bell shaped, symmetric and centered about the mean. The primary difference is that it has stronger tails, indicating a lower probability of being within an interval. The variability depends on the sample size, n.
- □ With a confidence interval of c%

$$\overline{x} - t_{\alpha/2,\nu} \frac{\sigma_s}{\sqrt{n}} < X < \overline{x} + t_{\alpha/2,\nu} \frac{\sigma_s}{\sqrt{n}}$$

Where α=1-c and v=n-1 (Degrees of Freedom)

Don't apply blindly - you may have better information about the population than you think.

Applied Mechanical Engineering Program

$x_1 = \frac{x_1 - x'}{\sigma}$	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09	
0.0	0.0000	0.0040	0.0080	0.0120	0.0160	0.0199	0.0239	0.0279	0.0319	0.0359	
0.1	0.0398	0.0438	0.0478	0.0517	0.0557	0.0596	0.0636	0.0675	0.0714	0.0753	
).2	0.0793	0.0832	0.0871	0.0910	0.0948	0.0987	0.1026	0.1064	0.1103	0.1141	
).3	0.1179	0.1217	0.1255	0.1293	0.1331	0.1368	0.1406	0.1443	0.1480	0.1517	
).4	0.1554	0.1591	0.1628	0.1664	0.1700	0.1736	0.1772	0.1809	0.1844	0.1879	
).5	0.1915	0.1950	0.1985	0.2019	0.2054	0.2088	0.2123	0.2157	0.2190	0.2224	
0.6	0.2257	0.2291	0.2324	0.2357	0.2389	0.2422	0.2454	0.2486	0.2517	0.2549	
0.7	0.2580	0.2611	0.2642	0.2673	0.2794	0.2734	0.2764	0.2794	0.2823	0.2852	570 G
0.8	0.2881	0.2910	0.2939	0.2967	0.2995	0.3023	0.3051	0.3078	0.3106	0.3133	p(x)
).9	0.3159	0.3186	0.3212	0.3238	0.3264	0.3289	0.3315	0.3340	0.3365	0.3389	
0.1	0.3413	0.3438	0.3461	0.3485	0.3508	0.3531	0.3554	0.3577	0.3599	0.3621	
.1	0.3643	0.3665	0.3686	0.3708	0.3729	0.3749	0.3770	0.3790	0.3810	0.3830	
.2	0.3849	0.3869	0.3888	0.3907	0.3925	0.3944	0.3962	0.3980	0.3997	0.4015	1
.3	0.4032	0.4049	0.4066	0.4082	0.4099	0.4115	0.4131	0.4147	0.4162	0.4177	1
.4	0.4192	0.4207	0.4292	0.4236	0.4251	0.4265	0.4279	0.4292	0.4306	0.4319	/
.5	0.4332	0.4345	0.4357	0.4370	0.4382	0.4394	0.4406	0.4418	0.4429	0.4441	
.6	0.4452	0.4463	0.4474	0.4484	0.4495	0.4505	0.4515	0.4525	0.4535	0.4545	
.7	0.4554	0.4564	0.4573	0.4582	0.4591	0.4599	0.4608	0.4616	0.4625	0.4633	
1.8	0.4641	0.4649	0.4656	0.4664	0.4671	0.4678	0.4686	0.4693	0.4699	0.4706	
1.9	0.4713	0.4719	0.4726	0.4732	0.4738	0.4744	0.4750	0.4758	0.4761	0.4767	Sector States
2.0	0.4772	0.4778	0.4803	0.4788	0.4793	0.4799	0.4803	0.4808	0.4812	0.4817	
2.1	0.4821	0.4826	0.4830	0.4834	0.4838	0.4842	0.4846	0.4850	0.4854	0.4857	
2.2	0.4861	0.4864	0.4868	0.4871	0.4875	0.4878	0.4881	0.4884	0.4887	0.4890	1.7.2
2.3	0.4893	0.4896	0.4898	0.4901	0.4904	0.4906	0.4909	0.4911	0.4913	0.4916	
2.4	0.4918	0.4920	0.4922	0.4925	0.4927	0.4929	0.4931	0.4932	0.4934	0.4936	
2.5	0.4938	0.4940	0.4941	0.4943	0.4945	0.4946	0.4948	0.4949	0.4951	0.4952	
2.6	0.4953	0.4955	0.4956	0.4957	0.4959	0.4960	0.4961	0.4962	0.4963	0.4964	
2.7	0.4965	0.4966	0.4967	0.4968	0.4969	0.4970	0.4971	0.4972	0.4973	0.4974	
2.8	0.4974	0.4975	0.4976	0.4977	0.4977	0.4978	0.4979	0.4979	0.4980	0.4981	
2.9	0.4981	0.4982	0.4982	0.4983	0.4984	0.4984	0.4985	0.4985	0.4986	0.4986	
3.0	0.49865	0.4987	0.4987	0.4988	0.4988	0.4988	0.4989	0.4989	0.4989	0.4990	



Basics of Engineering Measurements

UNCERTAINTY

39

JN

### Use Z-distribution

If the population standard deviation is known or the sample is greater than 30.

$$\overline{X} \pm z \frac{\sigma}{\sqrt{n}}$$

### Use *t*-distribution

If the population standard deviation is unknown and the sample is less than 30 and normally distributed.

$$\overline{X} \pm z \frac{\sigma}{\sqrt{n}} \qquad \overline{X} \pm t \frac{s}{\sqrt{n}}$$

$$\int t = \frac{\overline{x} - \mu_0}{s/\sqrt{n}}$$
One-sided
$$\begin{cases}
H_a: \mu > \mu_0 \Box > P(T \ge t) \\
H_a: \mu < \mu_0 \Box > P(T \le t)
\end{cases}$$
Applied Mechanic
$$\int t = \frac{Two-sided}{two-tailed} \qquad H_a: \mu \neq \mu_0 \Box > 2P(T \ge |t|)$$

$$\int t = \frac{\sigma}{|t|}$$



		Reading Number	Volts, mv
Stu	ident t-distribution	1	5.30
Stu	ident t-distribution	2	5.73
		3	6.77
	Example: t-distribution	4	5.26
	Example: t-distribution	5	4.33
	Sample data	6	5.45
Sec.	<b>n</b> = 21	7	6.09
	<b>Degrees of Freedom</b> $-n_1 - 20$	8	5.64
and the		9	5.81
	Desire 95% Confidence Interval	10	5.75
	$\alpha = 1 - c = 0.05$	11	5.42
	$\alpha/2 = 0.025$	12	5.31
	-0.025	13	5.80
100		14	3.70 A 91
	Student t-distribution chart	16	6.02
	t=2.086	13	6.25
		18	4.99
Sec. 5		19	5.61
1072		20	5.81
		21	5.60
	Applied Mechanical Engineering Program	Mean	5.60
		Standard dev.	0.51 43
	BASICS OF ENGINEERING MEASUREMENTS	Variance	0.26



BASICS OF ENGINEERING MEASUREMENTS

# **Propagation of Uncertainty**

- Uncertainty is based on a careful specification of the uncertainties in the various primary experimental measurements.
- If result R = R(x<sub>1</sub>,x<sub>2</sub>, ...,x<sub>n</sub>) is a given function of independent variables x<sub>1</sub>,x<sub>2</sub>, ...,x<sub>n</sub>; and w<sub>1</sub>,w<sub>2</sub>, ...,w<sub>n</sub> are the associated uncertainties, then uncertainty of the result w<sub>R</sub> is given by:

$$w_{R} = \sqrt{\left(\frac{\partial R}{\partial x_{1}} w_{1}\right)^{2} + \left(\frac{\partial R}{\partial x_{2}} w_{2}\right)^{2} + \dots + \left(\frac{\partial R}{\partial x_{n}} w_{n}\right)^{2}}$$

Applied Mechanical Engineering Program

# **Propagation of Uncertainty**

The bias uncertainty  $U_B$  is given by manufacturer. Usually a formula for the calculation is given in the catalogue.

Each of the individual measurement variables  $(X_1, X_2,...,X_K)$  is subject to Several bias and precision errors.

The bias error :

 $U_{B} = (U_{B1}, U_{B2}, ..., U_{Bk})$ 

The precision error :

$$U_{P} = (U_{P1}, U_{P2}, ..., U_{Pk})$$

Applied Mechanical Engineering Program

UNCERTAINTY

46

Chapter 2

Chapter 2

UNCERTAINT

# **Calculation of bias Uncertainty**

1- For measurement variable X this is given by:

 $U_{B} = (U_{B1}^{2} + U_{B2}^{2} + ... + U_{Bk}^{2})^{1/2}$ 

2- The next step in the procedure is to apply uncertainty analysis to determine how the bias limits  $U_B = (U_{B1}, U_{B2}, ..., U_{Bk})$  for the individual variables propagate through the data reduction equation to form the bias limit  $U_B$  for the experimental result. The data reduction equation is taken to be of the form :

 $R=R(X_1, X_2, ..., X_K)$ 

The bias Uncertainty is :

$$\left(\frac{\mathbf{U}_{\mathbf{B}}}{\mathbf{R}}\right)^{2} = \left(\frac{1}{\mathbf{R}}\frac{\partial \mathbf{R}}{\partial \mathbf{X}_{1}}\mathbf{U}_{\mathsf{B1}}\right)^{2} + \left(\frac{1}{\mathbf{R}}\frac{\partial \mathbf{R}}{\partial \mathbf{X}_{2}}\mathbf{U}_{\mathsf{B2}}\right)^{2} + \ldots + \left(\frac{1}{\mathbf{R}}\frac{\partial \mathbf{R}}{\partial \mathbf{X}_{\mathsf{K}}}\mathbf{U}_{\mathsf{Bk}}\right)^{2}$$

Applied Mechanical Engineering Program

# **Calculation of precision Uncertainty**

1- For measurement variable X, Then, the overall precision limit UP is given by the

$$U_{P} = (U_{P1}^{2} + U_{P2}^{2} + ... + U_{Pk}^{2})^{1/2}$$

2- The next step in the procedure is to apply uncertainty analysis to determine how the bias limits  $UP = (U_{P1}, U_{P2}, ..., U_{Pk})$  for the individual variables propagate through the data reduction equation. The data reduction equation is taken to be of the form :

$$R = R(X_1, X_2, ..., X_K)$$

the 95% precision limit for the experimental result UP is found from the uncertainty analysis expression

$$\left(\frac{U_{\text{P}}}{R}\right)^{2} = \left(\frac{1}{R}\frac{\partial R}{\partial X_{1}}U_{\text{P1}}\right)^{2} + \left(\frac{1}{R}\frac{\partial R}{\partial X_{2}}U_{\text{P2}}\right)^{2} + \ldots + \left(\frac{1}{R}\frac{\partial R}{\partial X_{K}}U_{\text{Pk}}\right)^{2}$$

Applied Mechanical Engineering Program

*Chapter 2* UNCERTAINTY

48

Chapter 2

UNCERTAINT





Basics of Engineering Measurements

UNCERTAINTY