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**B. Pharmacy — 8th Semester**  
**BIOSTATISTICS AND RESEARCH**  
**METHODOLOGY**  
**UNIT 1 — Comprehensive Notes**

Subject Code	Semester	Unit	Source
BP801T	8th Semester	Unit 1 of 5	Noteskarts.com

**UNIT 1 SYLLABUS**  
**COVERAGE**

1.1	<b>Introduction to Statistics &amp; Biostatistics</b>	<i>Definition, Scope, Importance</i>
1.2	<b>Frequency Distribution</b>	<i>Types, Tabulation, Histogram, Ogive</i>
1.3	<b>Measures of Central Tendency</b>	<i>Mean, Median, Mode with Pharma Examples</i>
1.4	<b>Measures of Dispersion</b>	<i>Range, Variance, Standard Deviation</i>
1.5	<b>Correlation</b>	<i>Karl Pearson's, Multiple Correlation – Pharma Problems</i>

# Introduction to Statistics & Biostatistics

## ◆ What is Statistics?

Statistics is the science of collecting, organizing, summarizing, analyzing, and interpreting numerical data to draw meaningful conclusions and make informed decisions.

Statistics can be divided into two main branches:

Descriptive Statistics	Inferential Statistics
Summarizes and describes the collected data	Draws conclusions about a population from a sample
<i>Mean, Median, Mode, SD, Range</i>	<i>t-test, ANOVA, Chi-square, Regression</i>

## ◆ What is Biostatistics?

Biostatistics (Biological Statistics) is the application of statistical methods to biological, health, pharmaceutical, and medical sciences. It helps in designing experiments, analyzing drug data, and validating research findings.

**Applications of Biostatistics in Pharmacy:**

- Designing and analyzing clinical trials for new drugs
- Evaluating drug efficacy and safety profiles
- Quality control in pharmaceutical manufacturing
- Epidemiological studies and disease surveillance
- Stability testing of pharmaceutical formulations
- Bioavailability and bioequivalence studies
- Pharmacokinetic and pharmacodynamic modeling

## ◆ Importance of Biostatistics in Pharmacy

S.N.	Area	Role of Biostatistics
1	<b>Drug Development</b>	Determines sample size, analyzes Phase I-IV trial data
2	<b>Quality Control</b>	Monitors batch variability, acceptance sampling
3	<b>Pharmacokinetics</b>	Analyzes drug concentration-time data
4	<b>Epidemiology</b>	Studies disease prevalence, drug side-effects
5	<b>Regulatory Affairs</b>	CDSCO/FDA submission requires statistical analysis

## ◆ Basic Statistical Terms

Term	Definition
<b>Population</b>	Complete set of individuals/items from which data is collected (e.g., all diabetic patients in India)
<b>Sample</b>	A subset of population selected for study (e.g., 200 diabetic patients from Delhi)
<b>Variable</b>	Any characteristic that can vary between individuals (e.g., blood glucose level, drug dose)
<b>Data</b>	Observed values of variables (e.g., 120 mg/dL, 250 mg tablet)
<b>Parameter</b>	Numerical summary of a population ( $\mu$ = population mean, $\sigma$ = population SD)
<b>Statistic</b>	Numerical summary of a sample ( $\bar{x}$ = sample mean, $s$ = sample SD)

## Frequency Distribution

A Frequency Distribution is an organized tabular arrangement showing how frequently each value or group of values occurs in a dataset. It converts raw data into a meaningful form.

## ◆ Terms in Frequency Distribution

Term	Meaning
<b>Class Interval</b>	A range of values grouped together (e.g., 10–20, 20–30)
<b>Class Width</b>	Difference between upper and lower limits of a class (e.g., 20–10 = 10)
<b>Frequency (f)</b>	Number of times a value falls within a class interval
<b>Relative Frequency</b>	Frequency of a class divided by total frequency ( $f/N$ )
<b>Cumulative Frequency</b>	Running total of frequencies from the beginning
<b>Class Mark / Midpoint</b>	Middle value of a class interval = $(\text{Lower} + \text{Upper}) / 2$

## ◆ Pharmaceutical Example — Frequency Distribution Table

**Example: Dissolution (%) values of 30 tablets of Paracetamol 500 mg:**

*Raw data:* 76, 82, 89, 74, 91, 85, 78, 83, 87, 90, 79, 84, 88, 76, 93, 81, 86, 77, 92, 84, 80, 85, 89, 75, 91, 88, 83, 78, 87, 86

Class Interval (%)	Frequency (f)	Class Mark (xi)	Rel. Freq.	Cumul. Freq.
70–75	2	72.5	0.067	2
75–80	6	77.5	0.200	8
80–85	8	82.5	0.267	16
85–90	9	87.5	0.300	25
90–95	5	92.5	0.167	30
<b>Total</b>	<b>30</b>	–	<b>1.000</b>	–

▮ Interpretation: Maximum tablets (9 out of 30) showed dissolution between 85–90%, indicating most tablets meet pharmacopoeial dissolution standards (NLT 80%).

## ◆ Types of Frequency Distribution

Type	Description & Example
<b>Simple / Ungrouped</b>	Frequency of each individual value. Used for discrete data. Ex: Number of tablets passing/failing
<b>Grouped Frequency</b>	Values grouped into class intervals. Used for continuous data. Ex: Dissolution %, Particle size
<b>Relative Frequency</b>	Each class frequency expressed as proportion of total. Used for comparison
<b>Cumulative Frequency</b>	Running sum of frequencies. Basis for Ogive curve
<b>Normal Distribution</b>	Bell-shaped, symmetric frequency distribution around mean. Common in pharmacokinetics

## Measures of Central Tendency

Measures of Central Tendency are statistical values that represent the 'centre' or 'typical value' of a dataset. The three main measures are Mean, Median, and Mode.

### ◆ Mean (Arithmetic Mean)

The Mean is the sum of all observations divided by the total number of observations. It is the most commonly used measure of central tendency.

<b>MEAN</b>	$\bar{x} = \Sigma x / n$	x=values, n=count
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<b>MEAN (Grouped)</b>	$\bar{x} = \Sigma (f \cdot x_i) / \Sigma f$	f=freq, x <sub>i</sub> =midpoint
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**Properties of Mean:**

- Uses all data values — most sensitive to extreme values (outliers)
- Sum of deviations from mean is always zero:  $\Sigma(x - \bar{x}) = 0$
- Algebraically tractable — can be used in further calculations
- Affected by skewed distribution and extreme values

**💡 Pharmaceutical Example — Mean**

Problem: A pharmacist measured the weight of 8 tablets (mg): 498, 502, 497, 503, 499, 501, 500, 500 Mean =  $(498+502+497+503+499+501+500+500) / 8 = 4000 / 8 = 500$  mg  
 Interpretation: Average tablet weight = 500 mg, which meets USP ±5% limit (475–525 mg). All tablets within spec.

**◆ Median**

The Median is the middle value when data is arranged in ascending or descending order. It divides the data into two equal halves.

Odd Number (n) of Values	Even Number (n) of Values
<b>Median = value at position <math>(n+1)/2</math></b>	<b>Median = Average of <math>n/2</math> and <math>(n/2+1)</math> values</b>
<i>n=7: Median = 4th value</i>	<i>n=8: Median = avg of 4th &amp; 5th values</i>

**Formula for Grouped Data:**

<b>MEDIAN (Grouped)</b>	$M = L + [(N/2 - cf) / f] \times h$	L=lower limit, cf=cumul freq before, f=class freq, h=width
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**💡 Pharmaceutical Example — Median**

Problem: Disintegration times (seconds) of 7 tablets: 45, 52, 38, 60, 47, 55, 42 Step 1: Arrange in order: 38, 42, 45, 47, 52, 55, 60 Step 2: n = 7 (odd), Median =  $(7+1)/2 = 4$ th value = 47 seconds Interpretation: 50% tablets disintegrate in ≤47 seconds. Not affected by extreme values.

## ◆ Mode

Mode is the value that appears most frequently in a dataset. A dataset can have No mode, One mode (Unimodal), Two modes (Bimodal), or Multiple modes (Multimodal).

**MODE (Grouped)**

$$Mo = L + \frac{(f1 - f0) / (2f1 - f0 - f2)}{h} \times h$$

f1=modal class freq,  
f0=preceding freq,  
f2=following freq

### 💡 Pharmaceutical Example — Mode

Problem: Side effects reported by patients: Nausea(15), Headache(22), Dizziness(8), Headache(22), Vomiting(10) Mode = Headache (appears most frequently = 22 cases)  
Pharmaceutical Use: Identifies the most common adverse drug reaction (ADR) — useful in pharmacovigilance reporting.

## ◆ Comparison of Mean, Median & Mode

Property	Mean	Median	Mode
<b>Definition</b>	Arithmetic average	Middle value	Most frequent
<b>Uses all values</b>	Yes	No	No
<b>Affected by outliers</b>	Highly	Less	Not affected
<b>Can be calculated for</b>	Continuous data	Continuous data	All data types
<b>Pharma application</b>	Avg tablet weight	Median shelf life	Most common ADR
<b>Mathematical use</b>	Further calculations	Non-parametric tests	Categorical data

📌 Empirical Relationship:  $Mode = 3 \times Median - 2 \times Mean$  (holds for moderately skewed distributions)

## Measures of Dispersion

Measures of Dispersion quantify the spread or variability in a dataset. They show how much the individual values deviate from the central value (mean). High dispersion = less uniformity; Low dispersion = more uniformity.

## ◆ What is Dispersion?

Two datasets can have the same mean but very different variability. Example:

Dataset	Values	Mean
Batch A Tablet Weights (mg)	498, 500, 499, 501, 502	500 mg
Batch B Tablet Weights (mg)	480, 510, 490, 520, 500	500 mg

Both batches have the same mean (500 mg), but Batch B has much greater variability — this is pharmacologically unacceptable. Measures of dispersion help detect such differences.

## ◆ Range

Range is the simplest measure of dispersion. It is the difference between the largest and smallest values.

<b>RANGE</b>	<b><math>R = X_{max} - X_{min}</math></b>	Max minus Min value
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### 🔑 Pharmaceutical Example — Range

Problem: Hardness (kg/cm<sup>2</sup>) of tablets: 5.2, 6.8, 5.9, 7.2, 6.1, 5.5, 7.0 Range = 7.2 – 5.2 = 2.0 kg/cm<sup>2</sup> Interpretation: The hardness varies by 2 kg/cm<sup>2</sup>. USP limit: 4–8 kg/cm<sup>2</sup>. All in range but variability should be minimized in production.

### Limitations of Range:

- Considers only two extreme values — ignores all other data
- Highly affected by outliers
- Not suitable for open-ended frequency distributions
- Not useful when comparing datasets with different sizes

## ◆ Variance

Variance is the average of squared deviations from the mean. It gives a measure of how spread the data is.

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

Used when all data of population is known

Used for sample data (Bessel's correction:  $n-1$ )

### ◆ 1.4.4 Standard Deviation (SD)

Standard Deviation (SD) is the square root of variance. It is expressed in the same units as the original data, making it the most widely used measure of dispersion in pharmaceutical sciences.

Population SD ( $\sigma$ )	Sample SD ( $s$ )
$\sigma = \sqrt{[\Sigma(x - \mu)^2 / N]}$	$s = \sqrt{[\Sigma(x - \bar{x})^2 / (n-1)]}$

#### Step-by-Step Formula for SD Calculation:

- Step 1: Calculate the mean ( $\bar{x}$ )
- Step 2: Find deviation of each value from mean:  $(x - \bar{x})$
- Step 3: Square each deviation:  $(x - \bar{x})^2$
- Step 4: Sum all squared deviations:  $\Sigma(x - \bar{x})^2$
- Step 5: Divide by  $(n-1)$  for sample:  $s^2 = \Sigma(x - \bar{x})^2 / (n-1)$
- Step 6: Take square root:  $s = \sqrt{s^2}$

#### 💡 Pharmaceutical Problem — Standard Deviation

Problem: Blood glucose levels (mg/dL) of 6 patients after Metformin therapy: 120, 135, 118, 142, 128, 133  
 Step 1: Mean =  $(120+135+118+142+128+133)/6 = 776/6 = 129.33$  mg/dL  
 Step 2: Deviations:  $-9.33, 5.67, -11.33, 12.67, -1.33, 3.67$   
 Step 3: Squared deviations:  $87.05, 32.15, 128.37, 160.53, 1.77, 13.47$   
 Step 4: Sum =  $423.34$   
 Step 5:  $s^2 = 423.34 / (6-1) = 84.67$   
 Step 6:  $s = \sqrt{84.67} = 9.20$  mg/dL  
 Result: SD =  $9.20$  mg/dL | Interpretation: Average blood glucose =  $129.33 \pm 9.20$  mg/dL

### ◆ Coefficient of Variation (CV)

CV expresses SD as a percentage of the mean. It is used to compare variability between datasets with different units or means.

CV	$CV = (s / \bar{x}) \times 100 \%$	Lower CV = more uniform batch
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✦ Pharmaceutical Standard: CV < 2% for tablet weight uniformity (USP). CV < 6% for content uniformity. If CV > 5%, batch-to-batch variability is high and must be investigated.

## ◆ Measures of Dispersion

Measure	Formula	Advantage	Disadvantage
Range	$X_{\max} - X_{\min}$	Simple, easy	Ignores other values
Variance	$\Sigma(x-\bar{x})^2/(n-1)$	Uses all values	Unit = square of data
Std. Dev.	$\sqrt{\text{Variance}}$	Same unit as data	Affected by outliers
CV	$(\text{SD}/\text{Mean}) \times 100$	Unitless comparison	Meaningless if $\text{mean} \approx 0$

## Correlation

Correlation is a statistical measure that describes the degree and direction of the linear relationship between two or more variables. It helps understand how one variable changes with respect to another.

## ◆ Types of Correlation

Type	Direction	Example in Pharmacy
Positive Correlation	Both variables move in same direction	<i>Drug dose</i> $\uparrow \rightarrow$ <i>Drug effect</i> $\uparrow$
Negative Correlation	Variables move in opposite direction	<i>Drug concentration</i> $\uparrow \rightarrow$ <i>Survival rate</i> $\downarrow$ (in toxicity)
Zero / No Correlation	No relationship between variables	<i>Patient height vs drug efficacy</i>
Perfect Positive	$r = +1.0$ , exact linear relationship	<i>Ideal standard calibration curve</i>
Perfect Negative	$r = -1.0$ , exact inverse relationship	<i>Drug clearance vs drug half-life</i>

## ◆ Karl Pearson's Coefficient of Correlation (r)

Karl Pearson's coefficient of correlation (r) is the most widely used method to measure the linear correlation between two continuous variables.

Pearson's r

$$r = \frac{\sum(x-\bar{x})(y-\bar{y})}{\sqrt{[\sum(x-\bar{x})^2 \cdot \sum(y-\bar{y})^2]}}$$

Population formula

r (simplified)

$$r = \frac{[n\sum xy - \sum x \sum y]}{\sqrt{\{[n\sum x^2 - (\sum x)^2][n\sum y^2 - (\sum y)^2]\}}}$$

Working formula

### ► Interpretation of r Value:

Value of r	Strength	Pharmaceutical Interpretation
+1.0	Perfect Positive	<i>Ideal calibration curve linearity</i>
+0.7 to +0.99	Strong Positive	<i>High dose-response correlation</i>
+0.3 to +0.69	Moderate Positive	<i>Drug conc. vs partial effect</i>
0 to +0.29 / 0 to -0.29	Weak / No correlation	<i>Unrelated variables</i>
-0.3 to -0.69	Moderate Negative	<i>Clearance vs blood level</i>
-0.7 to -0.99	Strong Negative	<i>Toxicity vs dose (therapeutic window)</i>
-1.0	Perfect Negative	<i>Inverse relationship</i>

### ► Properties of r:

- r is dimensionless (no units)
- $-1 \leq r \leq +1$  always
- r is independent of change of origin and scale
- If  $r = 0$ , variables are uncorrelated (but may have non-linear relationship)
- Pearson's r is sensitive to outliers — use Spearman's rank correlation for non-normal data

### 🔑 Pharmaceutical Problem — Karl Pearson's r

A study investigated correlation between drug concentration (x, µg/mL) and response (%inhibition, y): x: 2, 4, 6, 8, 10 y: 18, 32, 48, 65, 75 Calculations: n=5,  $\sum x=30$ ,  $\sum y=238$ ,  $\sum x^2=220$ ,  $\sum y^2=13282$ ,  $\sum xy=1660$   $r = \frac{[5 \times 1660 - 30 \times 238]}{\sqrt{\{[5 \times 220 - 30^2][5 \times 13282 - 238^2]\}}$   $r = \frac{[8300 - 7140]}{\sqrt{\{[1100 - 900][66410 - 56644]\}}$   $r = \frac{1160}{\sqrt{200 \times 9766}}$   $r = \frac{1160}{\sqrt{1953200}}$   $r = \frac{1160}{1397.57}$   $r = +0.830$  Result:  $r = +0.83 \rightarrow$  Strong positive correlation between drug concentration and inhibitory response.

## ◆ Multiple Correlation

Multiple Correlation measures the relationship between one dependent variable (Y) and two or more independent variables (X1, X2, X3...). The Multiple Correlation Coefficient is denoted as R (capital R).

Multiple R ( $R^2$ )

$$R^2_{y.12} = (r^2_{y1} + r^2_{y2} - 2r_{y1} \cdot r_{y2} \cdot r_{12}) / (1 - r^2_{12})$$

R = multiple correlation coefficient

### Where:

- $R^2_{y.12}$  = Coefficient of multiple determination
- $r_{y1}$  = correlation of Y with X1
- $r_{y2}$  = correlation of Y with X2
- $r_{12}$  = correlation between X1 and X2

### ► Pharmaceutical Example — Multiple Correlation:

Variable	Represents	Role
Y	Drug Dissolution (%)	Dependent variable
X1	Binder concentration (%)	Independent variable 1
X2	Compression Force (kN)	Independent variable 2

Given:  $r_{y1} = 0.85$ ,  $r_{y2} = 0.78$ ,  $r_{12} = 0.60$

$$R^2_{y.12} = (0.85^2 + 0.78^2 - 2 \times 0.85 \times 0.78 \times 0.60) / (1 - 0.60^2)$$

$$R^2_{y.12} = (0.7225 + 0.6084 - 0.7956) / (1 - 0.36)$$

$$R^2_{y.12} = 0.5353 / 0.64 = 0.8364$$

$$R_{y.12} = \sqrt{0.8364} = 0.915$$

Interpretation:  $R = 0.915$  — Strong multiple correlation. Binder concentration and compression force together explain 83.6% of the variation in drug dissolution ( $R^2 = 0.836$ ).

## ◆ Correlation vs Causation

⚠ IMPORTANT: Correlation does NOT imply causation. A high correlation between two variables does not mean one causes the other. Always interpret correlation in the context of pharmacological logic and experimental design.

Correlation

Causation

Statistical relationship between variables

*Measured by  $r$  or  $R$  coefficient*

Example: High  $r$  between drug dose and response

One variable directly influences another

*Established by controlled experiments*

Example: Increasing dose CAUSES increased response (pharmacodynamics)

## ◆ Expected Exam Questions

Q	Question	Marks
1	Define Statistics and Biostatistics. What is its importance in pharmacy?	2–5 Marks
2	What is a frequency distribution? Draw and explain frequency distribution table for given pharmaceutical data.	5 Marks
3	Explain Mean, Median, Mode with pharmaceutical examples.	5–10 Marks
4	What are measures of dispersion? Explain range, variance, and SD with examples.	5–10 Marks
5	Define standard deviation. Calculate SD for given tablet weight data.	5 Marks
6	Explain Karl Pearson's coefficient of correlation. Calculate $r$ for a dose-response dataset.	5–10 Marks
7	Write a note on: (a) Multiple Correlation (b) Correlation vs Causation.	5 Marks
8	Differentiate between Descriptive and Inferential Statistics.	2 Marks
9	The dissolution % of 5 tablets are: 78, 82, 79, 85, 91. Find Mean, Median, SD.	5 Marks
10	What is coefficient of variation? How is it used in pharmaceutical QC?	3 Marks