

Module 1

What is Research?

Research is basically a **systematic way of finding answers** to questions we don't know yet. Think of it like this: whenever you are curious about something—say, “*Why do plants grow faster in sunlight?*”—you don't just guess. You carefully **observe, collect data, and test** to find the real truth.

So, research = a **planned and scientific method** of discovering hidden facts or truths.

Example:

If someone says “*Eating almonds improves memory*”, research is the process where scientists test this claim by studying a group of people who eat almonds daily and comparing their memory skills with others who don't.

Objectives of Research

The purpose of research is not just to study something randomly—it has clear goals. Here are the main objectives in simple words:

1. **To Explore (Exploratory Research)**
 - This is when you study something **new or less known** to gain basic understanding.
 - Example: Imagine scientists suddenly discover a new species of fish in the ocean. They don't know much yet, so their first research is just to **explore and learn more about it**.
 2. **To Describe (Descriptive Research)**
 - This is about **clearly describing details** about people, situations, or groups.
 - Example: A survey about “*How many college students prefer online classes over offline?*” is descriptive research, because it tells us the **characteristics and preferences** of that group.
 3. **To Diagnose (Diagnostic Research)**
 - Here, research looks at **how often something happens** or whether it is connected with something else.
 - Example: Doctors may study “*How often do people with high sugar intake develop diabetes?*” This shows the **frequency** and **relationship** between sugar intake and diabetes.
 4. **To Test (Hypothesis-Testing Research)**
 - This is when researchers already have an **assumption (hypothesis)** and they want to check if it's true or false.
 - Example: A scientist believes “*Drinking green tea reduces stress.*” Now they test this by conducting experiments. If results match the assumption, the hypothesis is correct; otherwise, it's rejected.
-

 In short:

- Research = a **systematic search for truth**.
- Its objectives can be to **explore something new, describe things clearly, diagnose frequency/relationships, or test a belief/hypothesis**.

Motivation in Research

So, the big question is: **Why do people do research at all?**

Research is not just about sitting in a lab with test tubes or crunching data. Behind every researcher, there's always a **reason**—something that **pushes or pulls them** to search for answers.

Here are the main motivations (reasons):

1. Getting a Research Degree

Some people do research because they want a **Ph.D. or Master's degree**.

- Example: A student who wants to become a professor knows that a research degree will open doors to **better jobs, scholarships, or promotions**.

So here, the motivation is **career and academic benefits**.

2. Solving Practical Problems (Challenges)

Sometimes, research starts because people face a **real-world problem** that doesn't have a solution yet.

- Example: Farmers struggling with pests → scientists research to create **eco-friendly pesticides**.
- Example: During COVID-19, researchers rushed to find vaccines.

So here, the motivation is **to tackle unsolved problems**.

3. Intellectual Joy (Creative Work)

Some people love the **excitement of discovery**. For them, solving a mystery or inventing something new feels like fun.

- Example: Think of an inventor who spends nights building gadgets in their garage—not for money, but because they enjoy **creative problem-solving**.

So here, research is like a **game for the mind**.

4. 🤝 Service to Society

Many researchers want to **help people and improve society**.

- Example: A doctor researching cancer treatments is motivated by the dream of **saving lives**.
- Example: Environmental researchers work on **reducing pollution** so future generations live healthier lives.

So here, the motivation is **social responsibility**.

5. 🙌 Respect and Recognition

Research can also bring **status and respect**.

- Example: When scientists discover something big, like Einstein with relativity, they are remembered and respected worldwide.
 - For some, the motivation is to **earn name, fame, and recognition** in their field.
-

⌚ Other Motivations (Extra)

The list above is not complete—there are many more reasons:

- **Government orders or funding** (like defense research).
 - **Job requirement** (some jobs demand research projects).
 - **Curiosity** (the simple human “I want to know why this happens”).
 - **Understanding relationships** (like “Does exercise really improve memory?”).
 - **Awareness and social change** (studying poverty, education, gender issues, etc.).
-

👉 In short:

People do research for **personal growth** (degree, respect), **intellectual excitement** (curiosity, creativity), **practical needs** (solving problems, jobs), and **social service** (helping others, creating impact).

⚡ Imagine research like being a **detective**:

- Some detectives work for money (degree/job).

- Some for **solving mysteries** (problems).
- Some for the **thrill** of cracking a case (intellectual joy).
- Some to **protect society** (service).
- And some because they want **fame as the best detective** (respect).



Types of Research – Explained Simply

Research is like a big **toolbox**—depending on what problem you want to solve, you pick a different tool. That's why there are many “types” of research. Let's look at them one by one:

1. Descriptive vs. Analytical

- **Descriptive Research** → This type simply **describes what is happening right now**. The researcher just observes and reports facts without changing anything.
 - Example: If I survey “*How many students in my college prefer online classes?*” I'm only **describing the present situation**. I don't change anything, just record facts.
- **Analytical Research** → Here, the researcher **uses existing information or data** and tries to **analyze or interpret it**.
 - Example: Using crime data from the past 10 years to **find patterns and reasons why crime rates increased or decreased**.

👉 Descriptive = “What is happening?”

👉 Analytical = “Why is it happening?”

2. Applied vs. Fundamental

- **Applied Research (Action Research)** → This type is done to **solve an immediate, practical problem**.
 - Example: A company researching “*Why are customers leaving our app?*” and finding solutions to fix it.
 - Example: Developing a new vaccine for a spreading disease.
- **Fundamental Research (Pure/Basic Research)** → This type is not about solving a problem directly but about **gaining knowledge for the sake of knowledge**. It builds theories that may help later.
 - Example: Studying how memory works in the human brain.
 - Example: Pure mathematics research without immediate use, but useful later.

👉 Applied = “Solve today's problem.”

👉 Fundamental = “Build knowledge for the future.”

3. Quantitative vs. Qualitative

- **Quantitative Research** → Based on **numbers and measurements**. It deals with “how much,” “how many,” or “how often.”
 - Example: A survey that shows *70% of people prefer online shopping over offline.*
 - Example: Measuring the increase in crop yield after using new fertilizer.
- **Qualitative Research** → Focuses on **qualities, feelings, motives, and reasons** instead of numbers. It answers “why” and “how.”
 - Example: Interviewing people to understand **why they prefer online shopping** (convenience, discounts, variety).
 - Example: Psychologists studying why children behave in a certain way in school.

👉 Quantitative = Numbers 

👉 Qualitative = Feelings, Motives 

4. 💡 Conceptual vs. Empirical

- **Conceptual Research** → Based on **ideas, theories, and abstract thinking**. Often done by philosophers or thinkers.
 - Example: A philosopher developing a **new theory of justice**.
 - Example: Reinterpreting old concepts like “freedom” or “democracy.”
- **Empirical Research** → Based on **observation and experiments**. The researcher collects real-world data and tests a hypothesis.
 - Example: A scientist testing *“Does this new medicine reduce fever faster than the old one?”*
 - Example: Conducting experiments to check if “listening to music improves memory.”

👉 Conceptual = Thinking and theorizing 

👉 Empirical = Testing and experimenting 

5. ⏳ Other Types of Research

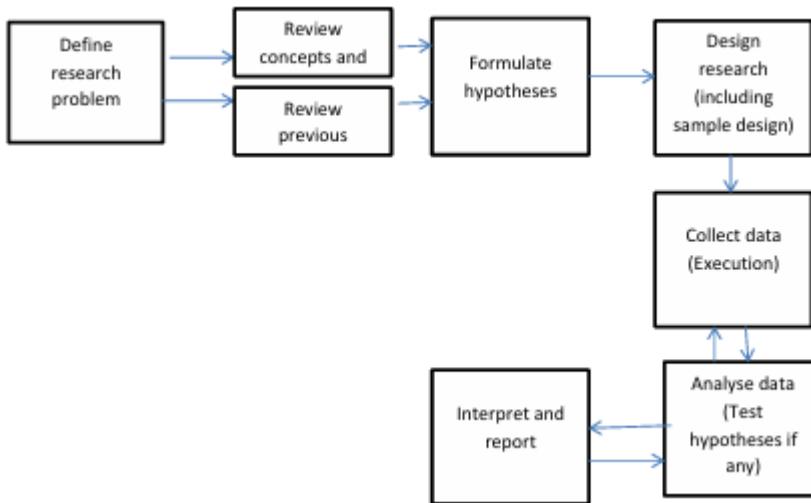
Apart from the main categories, there are special types too:

- **One-time vs. Longitudinal Research** →
 - One-time = research done in a single period (e.g., one survey in 2025).
 - Longitudinal = research done over a long period (e.g., studying child growth from age 2 to 18).
- **Field vs. Laboratory vs. Simulation** →
 - Field = in the real environment (e.g., studying farmers in their fields).
 - Lab = in controlled labs (e.g., testing medicine on mice).
 - Simulation = creating artificial conditions (e.g., flight simulators for pilot training).

- **Clinical/Diagnostic Research** → Like a doctor diagnosing illness, this type tries to find the **causes of problems** in detail, often with small samples.
 - **Exploratory vs. Formalized Research** →
 - Exploratory = done when little is known, to generate hypotheses (like exploring new disease symptoms).
 - Formalized = structured, with clear hypotheses to test (like clinical drug trials).
 - **Historical Research** → Using past records, documents, or artifacts to study events.
 - Example: Studying India's freedom struggle using letters and newspapers.
 - **Conclusion-oriented vs. Decision-oriented Research** →
 - Conclusion-oriented = researcher is free to explore and form conclusions.
 - Decision-oriented = done to help decision-makers (e.g., Operations Research in business to decide how to cut costs).
-

🎯 Simple Summary

- **Descriptive vs. Analytical** → Describe vs. Analyze
- **Applied vs. Fundamental** → Solve problems now vs. Build knowledge
- **Quantitative vs. Qualitative** → Numbers vs. Feelings
- **Conceptual vs. Empirical** → Ideas vs. Experiments
- **Other Types** → Time-based (one-time/longitudinal), Environment (field/lab), Purpose (exploratory, historical, decision-oriented)



Formulating the Research Problem (Simple Explanation)

When someone starts research, the very first step is to **decide what exactly they want to study**. This decision is called *formulating the research problem*. Without a clear problem, research will have no direction—like starting a journey without knowing the destination.

Now, research problems can be of **two types**:

1. **Problems related to states of nature** → These are about describing or understanding "what is".
 - Example: "What is the unemployment rate in rural India?"
 - Here, we are not testing a relationship, just describing a situation.
 2. **Problems related to relationships between variables** → These are about finding connections.
 - Example: "Does better education reduce unemployment?"
 - Here, we are checking the relationship between two things: education (one variable) and unemployment (another variable).
-

Steps in Formulating the Research Problem

1. **Choose a broad area of interest**
 - At first, a researcher only has a rough idea. Example: "I want to study about health issues."
 2. **Narrow it down**
 - That broad idea has to be made specific. Example: From "health issues" → "impact of junk food on teenagers' health."
 3. **Check feasibility**
 - Before finalizing, the researcher must ask: Do I have the data? Do I have time and resources? Example: If you want to study "impact of junk food on teenagers in the whole world," that's too big. But "impact of junk food on teenagers in my city" is feasible.
 4. **Understand the problem thoroughly**
 - Discuss with experts, guides, or colleagues. Example: A student talks with their professor about how junk food impacts obesity, and the professor guides them to focus on one measurable aspect (like BMI or weight).
 5. **Review literature**
 - Read books, research papers, or reports related to the topic. There are two kinds:
 - **Conceptual literature:** Theories and concepts (like "junk food has high sugar which may cause obesity").
 - **Empirical literature:** Previous studies (like "Research in 2020 showed teenagers eating fast food 3 times a week gained 5 kg in 6 months").
 - This helps the researcher know what is already studied and what is still left unexplored.
 6. **Rephrase the problem in clear, specific terms**
 - Instead of a vague problem like "junk food is bad," reframe it into something measurable:
 - Example: "Does eating junk food more than 3 times a week increase obesity among teenagers aged 13–18 in Bangalore?"
-

Why is this step so important?

Because if the research problem is **not clear**, everything else (data collection, analysis, report writing) will go wrong. A properly defined problem ensures:

- We know exactly **what data to collect**.
- We can separate **relevant information from irrelevant information**.
- We can choose the **right techniques** for analysis.
- We can prepare a **meaningful report** at the end.

Think of it like a doctor diagnosing a patient. If the doctor wrongly assumes the problem is "cold" when it's actually "pneumonia," the treatment will fail. Similarly, if a research problem is wrongly framed, the whole research will fail.

Example Story

Suppose you are a college student interested in social media. You start broadly:

- "I want to study about social media." (Too vague)

Then you narrow it:

- "I want to study how social media affects students." (Still broad)

After discussion and reading, you refine it:

- "I want to study how daily Instagram usage affects the academic performance of undergraduate students in my college."

This is now a **specific, measurable research problem** that can be studied with surveys, data, and analysis.

What does "literature survey" mean?

When researchers say **literature**, they don't mean novels or poems  . Here, **literature** means **all the studies, research papers, books, articles, government reports, and other documents related to the research topic**.

So, a **literature survey** means:

👉 Reading and studying what other researchers have already found about the problem you want to work on.

◆ Why is it needed?

Think of it like this:

Suppose you want to **build a new mobile app for food delivery**. Before you start coding, you would check:

- What apps like Swiggy, Zomato, Uber Eats already do
- What their strengths and weaknesses are
- What customers complain about
- What new ideas are missing

Only after knowing all this, you can create something **new and useful** instead of just repeating what others have already done.

Similarly, in **research**, a literature survey helps you:

1. **Understand the background** of your topic.
 2. **Avoid repeating old work** (so you don't "reinvent the wheel").
 3. **Find gaps** – things that have not been studied properly yet.
 4. **Get ideas** for how to design your own research.
-

◆ How is it done?

The process usually goes like this:

1. **Write a summary of your research problem**
 - Example: "I want to study how social media affects students' mental health."
 2. **Search for existing studies**
 - Start with **abstracting and indexing journals** (these are like search engines or databases for research, e.g., Scopus, Web of Science).
 - Look at **academic journals, conference papers, books, government reports, or even unpublished bibliographies**.
 3. **Follow the references**
 - One paper will usually mention other studies → you can track those too (like following links on Wikipedia).
 4. **Study earlier research similar to yours**
 - Example: If you are researching social media & students, you might find studies on **Instagram & anxiety, Facebook & academic performance**, etc.
 5. **Use libraries and online sources**
 - Universities have good libraries and access to online research databases.
-

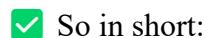
◆ Example (simple case)

Let's say your topic is "**The effect of online learning on school students' performance.**"

- You check journals and find a study on **college students**.
- You read a government report about **digital education policies**.
- You find a conference paper comparing **online vs. classroom learning**.
- From these, you realize:
 - 👉 "Oh, many studies are on college students, but very few focus on **school kids (ages 10–15)**. That's the gap I can study."

◆ Key point

The literature survey is like **doing your homework before starting a big project**. It gives you a strong foundation, helps you avoid mistakes, and shows the academic community that you know what has already been done.



A literature survey means **reviewing all the research already done on your problem**, using libraries, journals, reports, and books, so that you can build on past knowledge and clearly define your own research contribution.

💡 What is a Working Hypothesis?

A **working hypothesis** is like a **smart guess or assumption** that a researcher makes before starting their detailed research.

- It's not the final truth, but a **temporary idea** that can be tested.
- Think of it like a **map** for your research journey—it guides where to go, what to test, and what data to collect.

👉 Example: Suppose you want to research why students in your college score low in mathematics.

- A working hypothesis could be: "*Students score low in mathematics because they spend more time on social media than on practice.*"
This is not proven yet—it's just an assumption you will test with data.

🌟 Why is it Important?

1. **Focuses the research** – Without a hypothesis, you may collect too much useless data. With a hypothesis, you know *what exactly to check*.
2. **Saves time** – It stops you from going off-track.
3. **Guides data collection & analysis** – The kind of hypothesis you make tells you what data you need and which analysis method to use.
 - Example: If your hypothesis is about "time spent on social media," you need survey or usage data.

🛠️ How do researchers develop a Working Hypothesis?

The text gave **4 main approaches**. Let's simplify them with real-world examples:

1. **Discussions with experts/colleagues**
 - Talk to teachers, professors, or specialists about the problem.
 - Example: You discuss with a math teacher who says “*students often skip basics, that's why they fail.*” That gives you an idea for a hypothesis.
 2. **Examine data/records**
 - Look at past reports, attendance records, test scores, or performance patterns.
 - Example: You find in records that students with <50% attendance in tutorials fail more. Hypothesis: “*Attendance is directly linked with performance.*”
 3. **Review similar studies**
 - Check what other researchers have already found in similar cases.
 - Example: Another study from a different college shows that *students who solve practice problems daily score higher*. You can build on that idea.
 4. **Exploratory personal investigation**
 - Do small interviews or surveys on your own.
 - Example: You ask 20 students why they struggle with math. 12 of them say “*We don't practice regularly.*” That can shape your hypothesis.
-

◆ **Key Points to Remember:**

- A hypothesis must be **clear, specific, and testable**.
 - It should **relate only to your research problem**, not to everything in general.
 - Sometimes, especially in **exploratory research** (when you're just trying to explore and don't have enough info), you may not need a hypothesis.
-

◆ **Final Example:**

Research Topic: “Impact of Online Classes on Student Learning”

- Possible Working Hypothesis: “*Students who attend live online classes perform better than those who only watch recorded lectures.*”
Now, you'll test this hypothesis by collecting data (exam scores, attendance logs, surveys, etc.).

◆ **What is a Research Design?**

Research design is like a **blueprint or plan** for your research.

Imagine you are building a house. Before construction, you need a map of where rooms, doors, and windows will go. Similarly, before doing research, you need a design (plan) that shows:

- How you'll collect data
- From where you'll collect it
- How much time and money you'll need
- Which methods you'll use to analyze the data

So, **research design** = a structured plan to solve your research problem in the best way possible.

◆ Why is Research Design Important?

The main goal is to make research **efficient** → less time, less cost, less effort but more **useful information**.

Think of it like cooking:

- If you plan your recipe (ingredients, quantity, cooking time) in advance, you save time and get tasty food.
- If you don't plan, you may waste ingredients and time, and the food may not even taste good.

Similarly, a proper research design ensures your research gives **valid and reliable results**.

◆ Types of Research Purpose

Depending on **what you want from your research**, design changes. There are 4 main purposes:

1. **Exploration** → To explore something new, when not much is known.
Example: A researcher wants to know *why young people spend so much time on Instagram*.
→ Flexible design is used (open interviews, surveys, group talks).
 2. **Description** → To describe a situation accurately.
Example: A survey to find out *how many people in a city use electric bikes*.
→ Design must minimize **bias** and maximize **reliability** of data.
 3. **Diagnosis** → To find causes of a problem.
Example: A company sees a sudden drop in sales. Research is done to *diagnose why customers are leaving*.
 4. **Experimentation** → To test something under controlled conditions.
Example: A doctor testing a new medicine on two groups: one gets the medicine, the other doesn't.
-

◆ Types of Research Designs

There are two big categories:

1. **Experimental Designs** (you test cause-and-effect)
 - **Informal designs** (less strict control):

- Before-and-after without control → measure before & after, but no comparison group.
 - After-only with control → only measure results after experiment, compare with a control group.
 - Before-and-after with control → measure both groups before and after experiment.
 - **Formal designs** (very structured & scientific):
 - Completely randomized design
 - Randomized block design
 - Latin square design
 - Factorial designs (study effects of multiple factors at once)
2. **Non-Experimental Designs** (no manipulation, just observe and analyze existing data).
-

◆ **Things to Consider While Preparing Research Design**

When making a research design, researcher must think about:

1. **How will I get the information?** (surveys, interviews, observations, etc.)
 2. **Do I or my team have the skills to collect and analyze data?**
 3. **Why am I choosing this method? Is it logical?**
 4. **How much time do I have?**
 5. **How much money is available?**
-

✓ **Simple Example to Understand Everything Together:**

Suppose you want to research: “*Does online learning improve student performance compared to classroom learning?*”

- **Research purpose** → Experimentation (test effect)
- **Research design** → Before-and-after with control group
- **How to collect data** → Pre-test and post-test scores
- **Skills needed** → Ability to conduct tests, analyze results statistically
- **Time & cost** → 2 months, limited budget, so maybe test on a small group of students

This planned structure is your **research design**.

What is a Sample Design?

When we do research, we usually can't study **every single person or item** in the whole population (that would be a **census**).

Instead, we pick a smaller group (a **sample**) to represent the population.

👉 Example:

If you want to study the eating habits of **all students in your university** (say 10,000

students), it's almost impossible to ask every student.

So you might pick **500 students** from different departments → That's your **sample**.

But *how* you choose those 500 students is what we call the **Sample Design**.

◆ Why not Census (study everyone)?

- It takes too much **time** 
- It costs too much **money** 
- It uses too much **energy/effort** 
- Sometimes it's not even possible (like testing blood of every citizen – we only take samples).

So, researchers use **sampling**.

◆ Main Types of Sampling Designs

1. Deliberate (Purposive) Sampling

Here, the researcher **chooses intentionally** who will be in the sample.

- **Convenience Sampling** → Choose whoever is easiest to reach.
Example: If you want to study fuel buyers, you stand at 5 petrol stations and ask people there.
- **Judgment Sampling** → Researcher uses personal judgment.
Example: A teacher picks a group of students she believes represent the whole class.

⚠ Risk → Can be **biased**, since it depends on convenience or judgment.

2. Simple Random Sampling

Everyone in the population has an **equal chance** of being selected.

- Like a **lottery system**.
- Example: Writing names of 10,000 students on slips, then randomly drawing 500 names.

This method is **fair** but can be difficult with very large populations.

3. Systematic Sampling

Pick every **nth** item from a list.

- Example: If you have 10,000 students listed alphabetically, and you want 500 → select every **20th student**.

👉 To avoid bias, you randomly choose a **starting point**.

4. Stratified Sampling

Used when the population is **not the same (not homogeneous)**.

- Population is divided into **groups (strata)** like male/female, departments, age groups, etc.
- Then you take **samples from each group**.

Example: If you study health of university students → divide by year (freshman, sophomore, etc.) and then randomly select students from each year.

5. Quota Sampling

Similar to stratified sampling, but instead of random selection, researchers are told to fill a **quota**.

- Example: Interviewers are told to take 50 men and 50 women. They choose them however they like (not random).

⚠ Quota sampling = faster and cheaper, but **less scientific**.

6. Cluster Sampling

Instead of choosing individuals, you choose **groups (clusters)**.

- Example: A supermarket wants to survey 500 customers out of 15,000. Instead of picking 500 randomly, they divide customers into **100 clusters (150 each)** and then select 3 clusters randomly.

Cluster sampling = easier but may be **less accurate**.

7. Area Sampling

Similar to cluster, but used for **geographical areas**.

- Example: If you want to survey farmers across India → divide India into regions → randomly choose some regions → survey all farmers there.
-

8. Multi-Stage Sampling

A **step-by-step cluster sampling**.

- Example: To study families in India:
1st stage → Select states
2nd stage → Select districts from those states
3rd stage → Select towns/villages
4th stage → Select families

This is used in **big national surveys** (like Census Bureau or National Family Health Surveys).

9. Sequential Sampling

Here, the **sample size is not fixed** from the start.

- You collect some data, analyze it, and then decide if you need **more samples**.
 - Example: In product quality testing, you check 20 items. If many defects are found, you check more until results are reliable.
-



Summary (Super Simple):

- **Census** → Study everyone (almost impossible).
 - **Sample Design** → Plan for how you pick people/items.
 - **Random sampling** → Equal chance for all.
 - **Systematic sampling** → Every nth person.
 - **Stratified sampling** → Divide into groups, then sample.
 - **Quota sampling** → Like stratified but not random, just fill numbers.
 - **Cluster/Area sampling** → Pick groups, not individuals.
 - **Multi-stage sampling** → Sampling step by step.
 - **Sequential sampling** → Keep adding samples until results are clear.
-



👉 Think of it like **choosing food at a buffet**:

- Census → You eat everything 😊

- Random → Close eyes, pick dishes randomly
- Systematic → Taste every 5th dish
- Stratified → Take from every category (starters, main, desserts)
- Quota → Told to take 3 starters, 2 mains, 2 desserts (any you like)
- Cluster → Choose one table's food only
- Multi-stage → First choose a country, then city, then restaurant, then dishes
- Sequential → Keep eating until you're sure you've tried enough 😊

Why Collect Data?

When solving a real-world research problem, the information already available (secondary data like reports, books, or existing surveys) is often **not enough**.

So, researchers need to collect **new and specific information** → called **primary data**.

Ways of Collecting Primary Data

1. By Observation

- **What it means:** The researcher observes events, behaviors, or situations directly without asking people questions.
 - **Example (Real-life):** A traffic researcher stands at a junction and counts how many cars pass during peak hours.
 - **Advantages:**
 - Data is **fresh and real-time**.
 - No influence from respondents' biases.
 - **Disadvantages:**
 - Expensive (needs time and manpower).
 - Provides **limited information** (e.g., you see how many cars, but not *why* they are there).
 - **Best for:** Small-scale, real-time studies (like monitoring customer behavior in a shop).
-

2. Through Personal Interview

- **What it means:** Researcher asks pre-decided (structured) questions face-to-face.
 - **Example (Real-life):** A marketing researcher interviews shoppers in a mall about their brand preferences.
 - **Advantages:**
 - Rich and detailed data.
 - Researcher can clarify doubts instantly.
 - **Disadvantages:**
 - Time-consuming.
 - Costly if large samples are required.
 - **Best for:** Studies where opinions, attitudes, and detailed explanations are needed.
-

3. Through Telephone Interview

- **What it means:** Interviewing respondents over phone calls.
 - **Example (Real-life):** A telecom company calls users to get feedback on their network service.
 - **Advantages:**
 - Faster and cheaper than face-to-face.
 - Can cover a large geographic area.
 - **Disadvantages:**
 - Limited to short questions.
 - Some people may refuse to answer unknown calls.
 - **Best for:** Quick surveys, industrial research, or when time is limited.
-

4. By Mailing Questionnaires

- **What it means:** Researcher sends printed or digital forms (questionnaires) to respondents, who fill them and return.
 - **Example (Real-life):** A company emails a survey form to employees asking about job satisfaction.
 - **Advantages:**
 - Cost-effective for large samples.
 - Respondents can answer at their convenience.
 - **Disadvantages:**
 - Low response rate (people may ignore it).
 - No control over whether the respondent filled it honestly.
 - **Special Step: Pilot Study** – A small trial run of the questionnaire is done first to test for weaknesses.
 - **Best for:** Large-scale business or economic surveys.
-

5. Through Schedules

- **What it means:** Enumerators (trained field workers) visit respondents with a set of questions (schedules). They ask and record answers on the spot.
 - **Example (Real-life):** Government census workers visiting homes to collect family details.
 - **Advantages:**
 - Can be used even when respondents are illiterate.
 - More reliable (since enumerators ensure form is filled).
 - **Disadvantages:**
 - Quality depends on enumerator's skill.
 - Expensive because training and supervision are required.
 - **Best for:** Large-scale surveys where accuracy is critical (e.g., population census).
-

Factors to Consider Before Choosing a Method

- **Nature of investigation** (behavioral study, economic survey, etc.).
 - **Objectives** (why data is being collected).
 - **Scope** (small/local vs. large/national).
 - **Resources** (money, manpower, technology).
 - **Time available.**
 - **Desired accuracy level.**
-

Expert View

Dr. A.L. Bowley said:

 “In collection of statistical data, commonsense is the chief requisite and experience the chief teacher.”

This means a researcher must be practical, not just theoretical, and should rely on experience to avoid mistakes.

In summary:

Collecting data is the most **crucial step** in research. The method chosen depends on the **type of study, resources, and accuracy required**. Each method has pros and cons, and often researchers use a **combination** (e.g., questionnaires + interviews) to get better results.

What does “Execution of the Project” mean?

It simply means: **doing the actual research work in a proper way** (collecting data, asking people questions, training helpers, etc.) after you’ve planned everything.

If this step is not done properly, the research will fail—even if your plan was perfect.

Simple Story Example

Imagine you are doing a research project on “**How many students in your college like online classes vs offline classes?**”

1. **Systematic Work** – You need to collect answers from students in an organized way. If you are lazy or random, the results will be wrong.
2. **Using Questionnaires** – If you make a printed form with questions like:
 - Do you prefer online or offline?
 - Why?You can then **code answers** (for example: Online = 1, Offline = 2) so that the computer can easily count results.
3. **Using Interviewers** – If you hire some volunteers to ask these questions, you must **train them** properly. For example:
 - They should ask the question in the same way to every student.

- They should not add their personal opinion.
You can even give them a small guidebook/manual that explains exactly how to do the job.
4. **Checking Their Work** – You (the researcher) should sometimes **check in the field** to see if your volunteers are asking correctly or just filling fake answers.
 5. **Unexpected Problems** – Some students may refuse to answer. Instead of ignoring them, you can:
 - Make a list of such students.
 - Choose a small group of them again.
 - Try harder to convince them with the help of experts/teachers.
 6. **Statistical Control** – This means you must always ensure that the way you collect answers follows the rules of accuracy and fairness. For example:
 - Don't only ask your close friends (biased sample).
 - Make sure the data is complete and not missing many students.
-

👉 In short: **Execution = doing the actual survey or data collection properly, on time, with trained helpers, and handling problems smartly so that your data is trustworthy**

8. Analysis of Data

Once a researcher has collected raw data (like survey answers, measurements, test results), the next step is to **make sense of it**. Raw data is usually messy and cannot directly give conclusions. So, the researcher organizes it step by step:

- **Classification & Coding** → Divide the data into meaningful categories and give them symbols/numbers (e.g., male = 1, female = 2).
- **Editing** → Clean the data to remove errors or inconsistencies.
- **Tabulation** → Put the data into tables so patterns can be seen more clearly.
Nowadays, this is mostly done using computers which can handle large volumes of data and multiple variables at once.
- **Statistical Analysis** → Apply formulas like averages, percentages, correlation, regression, etc. to see patterns.

For example: if two factories report different average weekly wages, the researcher will check whether this difference is **real** or just due to **random chance**. Statistical tests (like test of significance) help in deciding this. Similarly, for agriculture research, analysis of variance (ANOVA) can show if different seed types really give different yields.

So, **data analysis = turning raw data into meaningful results with the help of statistics**.

9. Hypothesis Testing

After data analysis, the researcher is ready to **test the hypothesis** (the assumption or prediction made earlier). This step answers:

👉 “Do my results support my hypothesis or not?”

For this, various **statistical tests** are used:

- **Chi-square test** → checks relationships between categories (e.g., gender vs. product preference).
- **t-test** → checks differences between two groups (e.g., average marks of two classes).
- **F-test/ANOVA** → checks differences among three or more groups (e.g., crop yields of 3 seed varieties).

The outcome will be either:

- **Accept hypothesis** → data supports it.
- **Reject hypothesis** → data goes against it.

If no hypothesis was made at the beginning, then the researcher may form one now based on patterns found, and future researchers can test it further.

10. Generalisation and Interpretation

Finally, if a hypothesis is repeatedly tested and confirmed, the researcher can make **generalizations** (broad rules or theories that apply to many cases). For example: “Better training improves employee productivity.”

If the researcher didn’t start with a hypothesis, they can still **interpret results** by connecting findings to existing theories. This stage is very important because it gives meaning to the numbers and results.

Also, interpretation often leads to **new questions**, which may inspire future research. For example: if a study finds that one fertilizer improves crop yield, the next question might be: “*Does it also improve nutritional quality?*”

KHALCHA READ KRNE

1. What is Generalisation?

- Generalisation means: taking the results of your research and making a **broad statement or rule** that can apply to many similar situations.
- It’s like saying: “Based on my study, this thing usually happens in most cases.”

👉 Example:
If you test in 5 companies and find that employees who got **better training performed better**, you can generalise:
“Better training improves employee productivity.”
This statement is now a **general rule**, not just for the 5 companies you studied but also for many other companies.

2. What is Interpretation?

- Interpretation means: **explaining the meaning** of your findings.
- Even if you didn't have a clear hypothesis at the start, you can still **connect your results to theories or real-world situations**.

👉 Example:

Suppose you studied crops and found that a new fertilizer gives **20% more yield**.

- Your interpretation could be: "This fertilizer improves soil nutrient balance, which helps crops grow faster."
 - You are **giving meaning** to the numbers.
-

3. Why is this step important?

Because numbers or data by themselves don't make sense until you **explain what they mean**.

- Generalisation helps build **theories** (rules for future).
 - Interpretation helps give **meaning and context** to the findings.
-

4. New Questions (Future Research)

When you interpret, you often discover **new questions**.

👉 Example:

- You found that fertilizer improves crop yield.
 - Now you may ask: "Does this fertilizer also improve the **nutritional quality** of the crops?"
This becomes a new research topic.
-

✓ So, in short:

- **Data Analysis** = Organize & statistically process raw data.
- **Hypothesis Testing** = Check whether your assumption is supported by the data.
- **Generalisation & Interpretation** = Turn results into theories or explanations, often leading to new research.

Step 8: Preparation of the Report or Thesis

This is the final stage of the research process where the researcher has to present everything done in a well-structured written document. Writing the report is not just about putting words

together; it must be done carefully, systematically, and with clarity so that others can easily understand and evaluate the research.

1. Layout of the Report

- The report usually has **three main parts**:
 - **Preliminary pages** → Includes title, date, acknowledgements, foreword, table of contents, and lists of tables/figures.
 - **Main text** → This is the core part and should include:
 - **Introduction** (objectives, scope, methodology, limitations)
 - **Summary of Findings** (main results and recommendations in simple non-technical language)
 - **Main Report** (detailed results, presented logically in sections)
 - **Conclusion** (clear and precise summary of results).
 - **End matter** → Appendices (technical data), bibliography (sources used), and index (for easy navigation).

2. Writing Style

- The report should be **concise, objective, and written in simple language**.
- Avoid vague expressions like “*it seems*” or “*there may be*”.

3. Use of Visuals

- **Charts, graphs, and illustrations** should only be used if they make the information easier to understand and more impactful.

4. Accuracy and Transparency

- Mention **confidence limits** (statistical reliability of results).
- Clearly state any **constraints or challenges** faced during research.

Module 2

Meaning of Research Design

- A **research design** is basically a **blueprint or plan** for your research.
- Just like an architect makes a plan before building a house, a researcher makes a design before starting the study.
- It answers all the “**W**” **questions**:
 1. **What** is the study about? (topic/problem)
 2. **Why** is the study being done? (purpose/objective)
 3. **Where** will it take place? (location/field of study)
 4. **What type of data** is needed? (qualitative, quantitative, or both)
 5. **Where** will you get the data from? (sources)
 6. **What time period** is covered? (duration)
 7. **Who will be studied?** (sample design)
 8. **How** will the data be collected? (methods like survey, interview, observation, etc.)
 9. **How** will data be analysed? (statistical tools, software, etc.)
 10. **In what format** will the report be presented? (style of report writing)

 In short: **Research Design = Roadmap from problem → solution.**

Features of Research Design

1. It is a **plan** that specifies what information is needed and from where.
 2. It gives a **strategy** for collecting and analysing data.
 3. It considers **time and cost limits** (since research must be practical).
 4. It must include:
 - Clear research problem.
 - Methods for gathering information.
 - Population/sample to be studied.
 - Ways to process and analyse data.
-

Need for Research Design (Why is it important?)

- Without a proper design, research can become **confusing, wasteful, and incomplete**.
 - A good research design ensures:
 - Smooth flow of research operations.
 - Maximum useful information.
 - Minimum effort, time, and money wasted.
-

Example (Story Style)

Imagine you want to study: “**Does online learning improve student performance compared to classroom learning?**”

Your **research design** would answer:

- **What is the study about?** → Impact of online learning on performance.
- **Why study it?** → To know which method is more effective.
- **Where?** → Selected schools in your city.
- **What data?** → Exam scores, student feedback.
- **From where?** → School records + student surveys.
- **Time period?** → Academic year 2024–25.
- **Sample?** → 200 students (100 online, 100 classroom).
- **Techniques?** → Structured questionnaires + exam result analysis.
- **How analysed?** → Statistical tests (t-test, ANOVA).
- **Report style?** → Academic research report with tables/graphs.

This way, your **design works like a roadmap** to carry out the study successfully.

“**Selecting the Research Problem**” in very simple and clear terms, so you understand not just the theory but also the logic behind it.

Meaning

Selecting a research problem = **choosing the topic or issue** you want to study.

It looks simple, but it's one of the **most difficult and important steps** because a wrong choice can waste time, money, and effort.

👉 Think of it like choosing a seed to grow into a tree. If the seed is weak or unsuitable, the plant won't grow well. Similarly, the research problem must be carefully chosen so that the study becomes meaningful.

Guidelines for Selecting a Research Problem

1. Avoid overdone subjects

- If a topic is already researched too much, it will be hard to add anything new.
- Example: "Effect of exercise on health" has thousands of studies. Unless you bring a *new angle* (like effect of exercise on night-shift workers), it won't be useful.

2. Avoid highly controversial topics (for beginners/average researchers)

- Political, religious, or socially sensitive topics may create unnecessary difficulties.
- Example: Researching "impact of religious beliefs on government policies" may invite conflicts.

3. Avoid too narrow or too vague problems

- Too **narrow** → Not enough material to study.
- Too **vague** → No clear direction.
- Example:
 - Narrow: "Impact of tea on memory of 12-year-old students in one school."
 - Vague: "Impact of food on people's life."
 - Balanced: "Impact of caffeine drinks on concentration of high school students."

4. Choose a familiar and feasible subject

- Pick something you have background knowledge about and where research material is accessible.
- Example: If you are from a Computer Science background, studying "AI in healthcare" makes more sense than "Genetics in agriculture."

5. Consider practical factors

- **Importance** → Is the study useful in real life?
- **Qualification and training** → Do you have enough knowledge/skills to study it?
- **Cost** → Can you afford the expenses?
- **Time** → Can it be completed within the given time?
- **Cooperation** → Will participants or institutions agree to provide data?

Self-check Questions Before Final Selection

A researcher must ask himself:

- Am I well-equipped (skills/knowledge) to study this?
 - Can I afford the cost of data collection?
 - Do I have enough time to finish it?
 - Will people/organizations cooperate with me?
-

Preliminary Study (Feasibility Study)

- Before finalizing the problem, do a **small initial study** to check if it's doable.
 - Example: If you plan to study “impact of stress on IT employees,” first check whether companies will allow you to collect data from their employees.
-

Example (Story)

Suppose you want to research: “**Impact of social media on student performance.**”

- If you choose “impact of Instagram on one student,” → too narrow.
- If you choose “impact of social media on the world,” → too vague.
- Balanced problem → “Impact of daily Instagram use on academic performance of college students in Bangalore.”

This is **feasible, specific, and research-worthy**.

Techniques Involved in Defining a Research Problem

Defining the research problem = **setting a clear, precise, and workable statement** of what you are going to study.

If this step is done badly, the whole research becomes confusing or useless. That's why researchers follow a **systematic technique** instead of rushing.

Steps Involved

1. Statement of the Problem in a General Way

- First, state the problem broadly.
 - Example: “Many college students spend long hours on social media, which may affect their academic performance.”
- 2. Understanding the Nature of the Problem**
- Study it deeply to know what exactly you are trying to solve—cause, effect, variables involved.
 - Example: Is the problem about “time spent” or “type of content” that affects performance?
- 3. Survey of Available Literature**
- Review books, journals, articles, and earlier research.
 - This helps avoid duplication and sharpens your focus.
 - Example: Previous studies may show that social media impacts concentration, so you can refine your angle.
- 4. Developing Ideas through Discussions**
- Discuss with peers, guides, or experts.
 - Fresh perspectives often lead to clearer, sharper problem statements.
- 5. Rephrasing the Problem into a Working Proposition**
- Finally, convert the broad idea into a **clear, testable, and researchable problem statement.**
 - Example: “To study the effect of daily Instagram use (in hours) on the academic performance (CGPA) of undergraduate students in Bangalore colleges.”
-



Other Key Considerations

- **Define technical terms** → If you use words with special meaning, explain them clearly.
(Example: “Academic performance” = exam grades, attendance, or overall CGPA?)
 - **State assumptions/postulates** → Mention any basic assumptions you are making.
(Example: Assuming students will honestly report their social media usage.)
 - **Show value of the investigation** → Why is this study worth doing?
(Example: Results may help colleges create awareness about digital distractions.)
 - **Check suitability of time and data** → Is the time period realistic and is data available?
 - **Define scope/limits** → Be clear about what is included and excluded.
(Example: Only Instagram, not Facebook or Twitter; only undergraduate students, not school kids.)
-



Exam-Ready Answer (Short Form)

Techniques in Defining a Research Problem

1. State the problem in a general way.
2. Understand the nature of the problem.
3. Survey the available literature.

4. Develop ideas through discussions.
5. Rephrase the problem into a working proposition.

Other considerations:

- Define technical terms clearly.
- State assumptions and postulates.
- Mention value/importance of study.
- Consider time period and data sources.
- Clearly define scope and limitations.

Measurement & Scaling Techniques

When researchers study people, objects, or events, they often need a way to **measure** or **classify** information. Measurement scales are like "rules" that tell us how numbers or labels are assigned to data.

There are **4 main types of measurement scales**:

1 Nominal Scale (Just Labels – No Order)

- **Meaning:** Numbers or words are used only as **labels** (not for calculation).
- **No order, no arithmetic meaning.**
- **Example:**
 - Gender: Male = 1, Female = 2
 - Jersey numbers in football: Player 7, Player 10, Player 99
 - Blood type: A, B, AB, O

👉 You **cannot add, subtract, or rank** them. It only **classifies** data into categories.

2 Ordinal Scale (Order, but No Equal Gaps)

- **Meaning:** Items can be **ranked or ordered**, but the difference between them is **not equal**.
- **Example:**
 - Movie ratings: 1st, 2nd, 3rd place
 - Survey responses: Satisfied, Neutral, Dissatisfied
 - Education level: High School < Graduate < Postgraduate

👉 We know the **order**, but not the **exact difference**. (Example: The difference between rank 1 and 2 may not be the same as between rank 2 and 3).

3 Interval Scale (Order + Equal Intervals, but No True Zero)

- **Meaning:** Equal differences between values, but **zero is arbitrary** (not absolute).
- **Example:**
 - Temperature in Celsius: 20°C, 30°C, 40°C (difference of 10° is meaningful, but 0°C does not mean "no temperature").
 - IQ scores: A person with IQ 140 is smarter than IQ 120, but not necessarily "twice as smart".

👉 You can do **addition & subtraction**, but not ratios (e.g., 40°C is NOT "twice as hot" as 20°C).

4 Ratio Scale (Order + Equal Intervals + True Zero)

- **Meaning:** Has **all properties** – order, equal intervals, and a **true zero**.
- **Example:**
 - Weight: 60 kg, 30 kg (zero means no weight; 60 is twice 30).
 - Height: 180 cm, 90 cm
 - Income: ₹50,000, ₹1,00,000

👉 Allows **all mathematical operations** – addition, subtraction, multiplication, division.

Scale Construction Techniques (Super Simple)

1. Arbitrary Approach (By Researcher's Choice)

- **Meaning:** Researcher makes the scale on their own.
 - **Example:** Teacher writes 5 questions on "Happiness" without testing.
 - **Key Idea:** Fast, but not very reliable.
-

2. Consensus Approach (By Experts' Agreement)

- **Meaning:** A group of experts/judges check whether questions are good.
 - **Example:** Doctor + Psychologist approve questions about "Stress."
 - **Key Idea:** More reliable because experts confirm.
-

3. Item Analysis Approach (By Testing Items)

- **Meaning:** Many questions are given to people, then bad ones are removed.

- **Example:** Out of 20 exam anxiety questions, only 8 differentiate well between “high anxiety” and “low anxiety” students → keep only those 8.
 - **Key Idea:** Keep only strong questions.
-

4. Cumulative Scales (Step-by-Step Intensity)

- **Meaning:** Questions arranged from easy → strong.
 - **Example:**
 - Q1: “I talk to classmates.”
 - Q2: “I visit classmates at home.”
 - Q3: “I want to live with classmates.”If someone agrees with Q3 (strong), they also agree with Q2 and Q1.
 - **Key Idea:** Agreement builds step by step.
-

5. Factor Scales (By Statistics/Grouping)

- **Meaning:** Use maths (factor analysis) to group similar questions.
- **Example:** 30 questions on Job Satisfaction → grouped into 3 factors:
 - Salary & Perks
 - Work Environment
 - Promotions
- **Key Idea:** Finds hidden groups/dimensions.

Data Collection: Concept, Types & Methods

◆ Concept

- Data collection = the process of **gathering information** for research.
 - Purpose = to answer questions, test hypotheses, or understand a problem.
 - Example: If you want to study “**How social media affects students’ sleep**”, you must collect data about students’ social media use and their sleeping hours.
-

◆ Types of Data

1. Primary Data (First-hand, collected directly)

- Collected by the researcher **for the first time**.
- More reliable, but costly and time-consuming.
- Example: You personally interview 100 students about social media use.

2. Secondary Data (Already available, collected earlier)

- Collected and analyzed by **someone else**, but reused by you.
 - Cheaper, quicker, but may be outdated or less reliable.
 - Example: You take student sleep data from a government health report.
-

◆ **Methods of Primary Data Collection**

1. Observation Method

- Researcher watches and records behavior/events.
- Example: A shopkeeper observes how many people stop at a display shelf.

2. Interview Method

- Researcher asks questions face-to-face, by phone, or online.
- Example: Interviewing farmers about fertilizer use.

3. Questionnaires

- A set of written questions given to respondents to fill.
- Example: Google Form sent to students asking about study habits.

4. Schedules

- Similar to questionnaires, but filled by the **researcher** instead of respondents.
- Example: Census officers visit homes and fill the form themselves.

5. Other Methods

- **Warranty cards** → Collect feedback after product sale.
 - **Distributor audits** → Check sales from distributors.
 - **Pantry audits** → Examine household consumption by checking pantry.
 - **Consumer panels** → Selected group gives opinions regularly.
 - **Mechanical devices** → Devices measure behavior (e.g., eye-tracking in ads).
 - **Projective techniques** → Ask indirect questions to reveal hidden feelings.
 - **Depth interviews** → Long interviews to get deep insights.
 - **Content analysis** → Study documents/media to find trends.
-

◆ **Methods of Secondary Data Collection**

Sources of Published Data:

- Government reports (Census, RBI reports)
- International bodies (UN, WHO)
- Journals, books, magazines, newspapers

- Trade associations, banks, stock exchange reports
- Research reports from universities/scholars
- Public records, statistics, historical documents

Sources of Unpublished Data:

- Diaries, letters, autobiographies
 - Unpublished research work
 - Records of trade associations, labor bureaus, private companies
-

◆ Selection of Appropriate Method Depends On:

1. **Nature, scope, and objective of the study**
 - Example: For studying personal emotions → interviews work best.
 2. **Availability of funds**
 - Limited budget? → Use secondary data.
 3. **Time factor**
 - Less time? → Use secondary data or quick surveys.
 4. **Precision required**
 - High accuracy? → Use primary data with careful methods.
-

✓ In short:

- **Primary Data** = Freshly collected → more accurate, but costly.
- **Secondary Data** = Already available → cheaper, but may not perfectly fit your study.
- **Method chosen** depends on **objective, money, time, and accuracy needs**.

Processing and Analysis of Data

1. Processing Data (Cleaning + Organizing)

Once data is collected, it's usually messy. Processing makes it **ready for analysis**.

- **Editing** → Checking data for errors.
 - Example: If someone writes age = 250 in a survey, we correct/remove it.
- **Coding** → Giving numbers/symbols to responses.
 - Example: Gender → Male = 1, Female = 2.
- **Classification** → Grouping data into categories.
 - Example: Income → Low, Medium, High.
- **Tabulation** → Putting data in tables.
 - Example: A table showing “Number of students using social media vs hours of sleep.”

👉 This step makes data **organized** and **ready for analysis**.

2. Analysis of Data

Analysis means applying **statistical or logical techniques** to find patterns, test hypotheses, and draw conclusions.

Here are the **main types of analysis** (don't worry, I'll simplify):

(i) Multiple Regression Analysis

- Used when you want to **predict one thing** (dependent variable) from **many factors** (independent variables).
 - Example: Predicting **student grades** (dependent) using **study hours, attendance, and sleep time** (independent variables).
-

(ii) Multiple Discriminant Analysis

- Used when outcome is not a number, but a **category/group**.
 - Example: Predicting whether a customer will “**Buy**” or “**Not Buy**” a product based on income, age, and education.
-

(iii) Multivariate Analysis of Variance (MANOVA)

- Extension of **ANOVA** (which compares means of groups).
 - MANOVA compares **multiple dependent variables across groups**.
 - Example: A researcher tests if **diet type (vegetarian, keto, regular)** affects **both weight loss and blood pressure** together.
-

(iv) Canonical Analysis

- Used when you have **multiple dependent variables** and **multiple independent variables**.
 - It looks at how groups of variables are related.
 - Example: A company studies how **training hours + motivation** (independent) affect **employee performance + job satisfaction** (dependent).
-

◆ Experimental Design

✿ Meaning

Experimental design = **the plan of how an experiment is carried out** to test a hypothesis **scientifically**.

It answers:

- Which variables to study?
- How to control unwanted factors?
- How to collect/analyze data fairly?

◆ Why Important?

- Prevents **bias** (researcher's influence).
 - Reduces **random errors**.
 - Increases **accuracy** and **validity** of results.
 - Makes research **repeatable** by others.
-

◆ Example of Experimental Design

Suppose you want to test:

👉 “*Does a new fertilizer increase crop yield?*”

Steps:

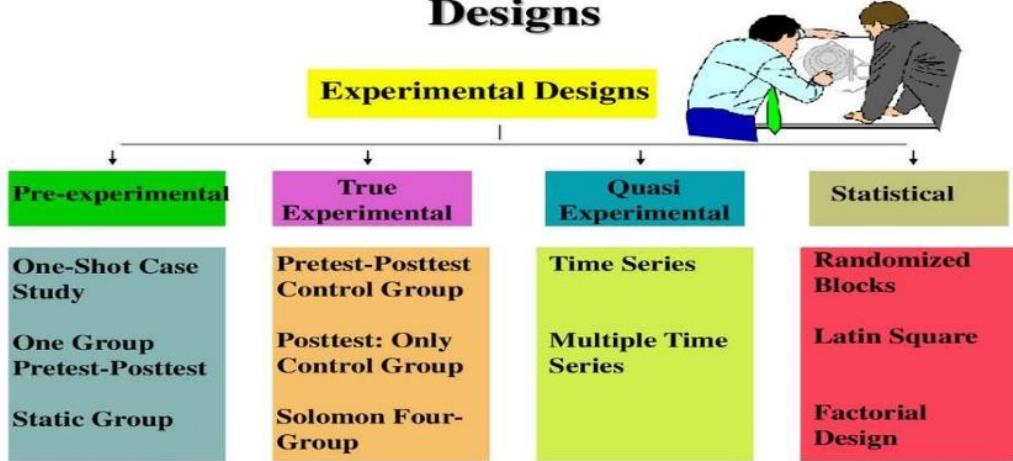
1. **Form research question** → Will new fertilizer improve yield?
 2. **Select variables** →
 - Independent = Fertilizer type (new vs old).
 - Dependent = Crop yield (kg per acre).
 3. **Control conditions** → Keep water, sunlight, soil same for both groups.
 4. **Experiment** → Apply new fertilizer to Group A, old fertilizer to Group B.
 5. **Collect data** → Record crop yield after harvest.
 6. **Analyze** → Compare yields using statistical tests.
 7. **Conclusion** → If Group A > Group B significantly, then fertilizer works.
-

✓ In short:

- **Processing** = Clean, organize, and prepare data.
- **Analysis** = Apply statistical methods to test relationships.
- **Experimental design** = The blueprint of how to test hypotheses scientifically with controlled experiments.

Design of Experiment

Figure 7.1 A Classification of Experimental Designs



TIME ASEL TAR READ KAR

Classification of Experimental Designs

Experimental designs are broadly grouped into **three categories**:

1. **Pre-Experimental Designs** (weak, less reliable)
2. **True Experimental Designs** (strong, high reliability)
3. **Quasi-Experimental Designs** (middle ground, practical but less control)

1. Pre-Experimental Designs (very basic, no strong control group)

(a) One-Shot Case Study

- Only **one group** is studied after a treatment.
- No pretest, no control group.
- Example: A teacher tries a new teaching method in one class and later checks scores. But we can't know if improvement was due to the method or something else.

(b) One-Group Pretest–Posttest Design

- One group is tested **before and after** treatment.
- Example: Measure students' test scores → give them a new learning app → measure scores again.
- Problem: Still no comparison with another group.

(c) Static Group Design

- Two groups: one gets treatment, one doesn't.
 - Only **posttest** is taken, no pretest.
 - Example: Two classes → one taught with a new method, another with old → compare test scores.
-

2. True Experimental Designs (most scientific, strong control)

(a) Pretest–Posttest Control Group Design

- Two groups: Experimental + Control.
- Both pretested, then only the experimental group gets treatment, then both posttested.
- Example: Two groups of patients → both tested for blood pressure → one group given new drug, one placebo → test both again.

(b) Posttest-Only Control Group Design

- Two groups: Experimental + Control.
- No pretest, only posttest after treatment.
- Example: One group given new fertilizer, one not → only final yield is measured.

(c) Solomon Four-Group Design

- Combination of pretest–posttest + posttest-only.
 - Four groups:
 - Group 1: Pretest → Treatment → Posttest
 - Group 2: Pretest → No Treatment → Posttest
 - Group 3: No Pretest → Treatment → Posttest
 - Group 4: No Pretest → No Treatment → Posttest
 - Best design (removes bias from pretesting).
-

3. Quasi-Experimental Designs (used when random assignment is not possible)

(a) Time Series Design

- One group is tested **repeatedly before and after** treatment.
- Example: Measure air pollution for 3 months → introduce new traffic rules → measure again for 3 months.

(b) Multiple Time Series Design

- Similar to time series, but with an additional **control group**.
 - Example: Compare pollution in **two cities** → one with new traffic rules, one without.
-

4. Statistical Experimental Designs

(a) Randomized Block Design

- Subjects are grouped into **blocks** (similar categories) to reduce variation.
- Example: Students grouped by intelligence level (low, medium, high) → within each group, some get new teaching method, some old.

(b) Latin Square Design

- Used when **two factors** affect the outcome and both need to be controlled.
- Example: Crop yield affected by fertilizer type and irrigation method → arrange experiment in a Latin Square grid.

(c) Factorial Design

- Studies the effect of **two or more independent variables simultaneously**.
 - Example: Test effect of fertilizer type (organic/chemical) and water level (low/high) on crop yield.
-



In Short

- **Pre-Experimental** → Weak, no proper control (One-shot, One-group, Static).
 - **True Experimental** → Strong, with control + randomization (Pretest-posttest, Posttest-only, Solomon four-group).
 - **Quasi-Experimental** → Practical, no full control (Time series, Multiple time series).
 - **Statistical Designs** → Advanced, use math for precision (Block, Latin Square, Factorial).
-

👉 Think of it like **testing a new medicine**:

- Pre-experimental → Give drug to few people, see effect (not reliable).
- True experimental → Random groups, one with drug, one with placebo (scientific).
- Quasi-experimental → Study real-world groups where randomization isn't possible (like two cities).
- Statistical → Use math to control extra factors (age, diet, gender).

Module 3

Quantitative Research (Meaning)

- It is the process of collecting and analyzing **numerical data (numbers, measurements, statistics)**.
- Purpose: To find **patterns, averages, predictions, cause-and-effect relationships**, and then apply results to a larger population.

👉 Example: A company wants to know if training improves employee productivity.

- Collect **data**: number of tasks done before training vs after training.
 - Analyze: average productivity increased → conclude training works.
-

◆ Quantitative vs Qualitative Research

Aspect	Quantitative Research	Qualitative Research
Data type	Numbers, measurable values	Words, images, videos, observations
Goal	Find patterns, test hypotheses, generalize results	Understand experiences, opinions, meanings
Example	Survey of 1000 people → % of people who like a product	Interview 10 people → why they like/dislike product

◆ Where is Quantitative Research Used?

- **Natural sciences**: Biology, Chemistry (measurements, lab experiments)
 - **Social sciences**: Psychology, Economics, Sociology, Marketing (surveys, experiments, statistics)
-

◆ Types of Quantitative Research

1. Descriptive Research

- Aim: To summarize data and describe what is happening.
- Example: Average age of employees in a company = 29 years.
- No cause-effect, just facts.

2. Correlational Research

- Aim: To check if two variables are related.
- Example: Is there a relationship between **hours studied** and **exam marks**?
- If more study hours = higher marks → positive correlation.
- But ~~X~~ correlation ≠ causation (studying more might be linked, but not the only reason for higher marks).

3. Experimental Research

- Aim: To test **cause-and-effect relationship**.
 - Researcher manipulates one variable (independent variable) and observes its effect on another (dependent variable).
 - Example: Give one group of students extra tutoring (treatment) and another group no tutoring (control). Compare exam scores → proves tutoring **causes** higher marks.
-

◆ Hypotheses Testing in Quantitative Research

- Hypothesis = prediction about relationship between variables.
 - Example: “Better training increases productivity.”
 - Correlational and experimental research use **statistics** to test whether the prediction is true.
-

◆ Operational Definitions (Very Important)

- Abstract concepts must be turned into measurable numbers.
- Example:
 - “Mood” → not directly measurable.
 - But we can measure it as:
 - Self-rating scale from 1–10
 - Number of smiles in a day
 - Energy level in activity

This makes the abstract concept **quantifiable**.

Quantitative Techniques



What is Sampling?

Sampling is simply the **selection of a small part (sample)** from a **big group (population)**, and then using that small part to make conclusions about the whole group.

👉 Example:

- Suppose a company has **10,000 customers**.
- Instead of asking **all 10,000** about their satisfaction (which will take time and money), the researcher may ask **500 customers** (sample).
- If chosen properly, these 500 customers will represent the opinions of all 10,000.
- **Population** = all 10,000 customers.
- **Sample** = 500 customers selected.
- **Sample design** = the method used to select the 500 (random, stratified, etc.).
- **Sample survey** = the actual survey done on those 500.

✓ The sample must be **representative**, meaning it should truly reflect the population, not be biased.

If the sample is biased (e.g., only rich customers asked), the conclusion will be **wrong**.

◆ Why Do We Need Sampling?

1. Saves Time and Money

- Surveying everyone is costly and slow.
- Sampling is cheaper and faster.

👉 Example: Instead of testing all 10,000 bulbs in a factory, test only 200 bulbs.

2. More Accurate Results

- A smaller group can be studied carefully with **trained investigators**.
- In census (studying everyone), mistakes may happen due to size.

3. Infinite Population

- Some populations are extremely large or infinite.
👉 Example: You cannot test **all grains of rice** in a sack, so you just test a cupful.

4. When Testing Destroys the Item

- Some studies involve destruction.
👉 Example: To check the quality of a matchstick, you need to burn it. You cannot burn the entire stock, so you test only a sample.

5. Helps in Estimating Errors

- Sampling allows researchers to calculate **sampling error** (difference between sample result and true population value).
👉 Example: If the sample average height = 165 cm, the real population height may be 166 cm. The difference = sampling error.
-

◆ Short Story Example (to fix in memory)

Imagine you are a **chef** cooking a big pot of soup 🍲 .

- You don't taste the **entire soup** (population).
 - You take **one spoonful** (sample).
 - From that spoonful, you judge whether it has enough salt or not.
 - If your spoonful is taken **properly** (mixed well, not just from the top), then your judgment will be correct.
 - That's **sampling**.
-

✓ In summary:

- **Sampling = studying a small part to understand the whole.**
- It is necessary because it saves time, money, effort, and sometimes is the only possible way.
- But the sample must be chosen carefully to be **representative**.

What is a Hypothesis?

A **hypothesis** is like an **educated guess** or **prediction** that a researcher makes before testing. It's not just a random assumption—it is a **formal statement** that can be **tested scientifically**.

👉 Example:

- “Students who receive counselling will score higher on creativity tests than those who do not.”

Here, we are predicting a **cause-effect relationship** (counselling → creativity). This is testable, because we can collect data on two groups and compare.

👉 Another example:

- “Car A performs as well as Car B.”
We can test mileage, speed, fuel consumption, etc., to see if it’s true.

✓ In short: A **hypothesis tells us what we are looking for** and gives us a **direction** in research.

◆ Why do we Test Hypotheses?

- To check if our assumptions are correct or not.
- To make decisions based on data, not guesses.
- To confirm relationships between variables (like smoking causes health issues).

👉 Think of it like a **courtroom trial** 🕴️ :

- A lawyer says: “The accused is guilty.” (Hypothesis).
 - Evidence (data) is collected.
 - The judge tests this claim and decides whether to accept or reject it.
-

◆ Steps in Hypothesis Testing

1. **State the hypothesis**
 - Null Hypothesis (H_0): There is no effect or difference.
👉 Example: “Counselling has no effect on creativity.”
 - Alternative Hypothesis (H_1): There is an effect or difference.
👉 Example: “Counselling increases creativity.”
 2. **Choose a significance level (α)**
 - Usually 5% (0.05).
 - Means: we are okay with 5 chances in 100 of being wrong.
 3. **Collect data**
 - Conduct surveys, experiments, or observations.
 4. **Apply a statistical test**
 - Depending on the data type and problem.
 5. **Make a decision**
 - If test result $< \alpha \rightarrow$ Reject H_0 (alternative is true).
 - If test result $\geq \alpha \rightarrow$ Fail to reject H_0 (stick with null).
-

◆ Various Tests for Hypothesis Testing (including multivariate analysis)

1. t-test (Student's t-test)

- Used when comparing the **means of two groups**.
👉 Example: Compare exam scores of students with and without counselling.

2. Chi-square test (χ^2 test)

- Used for **categorical data** (like Yes/No, Male/Female).
👉 Example: Testing whether gender is related to preference for product A or B.

3. ANOVA (Analysis of Variance)

- Used to compare the **means of more than 2 groups**.
👉 Example: Comparing creativity scores of students receiving 0, 1, or 2 counselling sessions.

4. Regression Analysis

- Checks **cause-effect relationship** between variables.
👉 Example: Predicting exam score based on study hours.
-

◆ Multivariate Analysis in Hypothesis Testing

When we deal with **more than two variables at a time**, we use **multivariate analysis**.

Examples:

1. **Multiple Regression**
 - One dependent variable + many independent variables.
👉 Example: Predicting students' exam score (dependent) using **study hours + sleep + motivation**.
 2. **Multivariate ANOVA (MANOVA)**
 - Used when there are multiple dependent variables.
👉 Example: Testing if counselling affects both **creativity and confidence** at the same time.
 3. **Discriminant Analysis**
 - Classifies objects into groups based on multiple predictors.
👉 Example: Predicting whether a customer is **loyal or not loyal** based on age, income, and shopping habits.
-

◆ Story Example to Remember

Imagine you are a **doctor** .

- You believe that **a new medicine reduces fever faster than the old medicine**.
- That belief = **hypothesis**.
- You give medicine to two groups of patients.
- Collect their recovery times (data).
- Run a **t-test** to compare average recovery times.
- If new medicine works significantly better → you reject the null hypothesis and accept your claim.

This is exactly how researchers test hypotheses in real life.

Characteristics of a Good Hypothesis

A hypothesis is not just a guess — it needs to have some qualities to be **useful in research**. Let's go one by one:

1. Clarity and Precision

 The hypothesis must be **clear and precise**, not confusing.

- If it's vague, the results based on it won't be reliable.

 Example:

 Bad: "Students do better if they get help." (Too vague — what kind of help?)

 Good: "Students who attend weekly counselling sessions score higher in creativity tests than those who do not."

2. Testability

 A hypothesis should be something that we can **actually test with data/experiments**.

- If you cannot check it in real life, it's not useful.

 Example:

- "Eating breakfast improves memory performance in children." (Can be tested by experiments).

 "Aliens influence human creativity." (Cannot be tested → unscientific).

3. Shows Relationship Between Variables

👉 A good hypothesis should show how **two or more things are related**.

- Independent Variable → Dependent Variable

✓ Example:

- “More hours of study (independent variable) increases exam performance (dependent variable).”
-

4. Limited in Scope and Specific

👉 A hypothesis should not be too broad. Narrow, specific hypotheses are easier to test.

✓ Example:

✗ Bad: “Exercise improves health.” (Too broad).

✓ Good: “30 minutes of walking daily reduces blood pressure in adults aged 30–40.”

5. Simplicity

👉 It should be written in **simple and understandable language**.

- Everyone should easily understand it, even if the research is complex.

✓ Example:

- “Listening to soft music while studying improves concentration.” (Simple and clear).
-

6. Consistency with Known Facts

👉 A hypothesis should not **contradict established scientific facts**.

- It should agree with what is already known.

✓ Example:

- “Excessive sugar intake increases the risk of diabetes.” (Matches known medical facts).

✗ “Drinking more sugar reduces diabetes.” (Contradicts science).

7. Can be Tested in a Reasonable Time

👉 The hypothesis should be **practical** → testable within a reasonable time frame.

- If testing takes a lifetime, it's useless for research.

✓ Example:

- “A new fertilizer increases crop yield in 3 months.” (Practical).
 - ✗ “A new fertilizer will increase soil fertility after 500 years.” (Impractical).
-

8. Explains the Problem

👉 A hypothesis must help explain the problem that led to the research.

- It should provide a possible explanation that fits the situation.

✓ Example:

- Problem: Students in school are performing poorly in maths.
 - Hypothesis: “Lack of practice tests reduces maths performance.”
 - This hypothesis actually **explains the problem** and can be tested.
-



Quick Story to Remember

Imagine you are a **detective** 🔎.

- You find a crime scene (research problem).
- You make a **guess (hypothesis)** about the suspect.
But for your guess to be useful, it must:
 - Be **clear** (“The suspect is John”),
 - Be **testable** (you can check John’s alibi),
 - Show a **relationship** (John was at the place → crime happened),
 - Be **specific** (not just “someone did it”),
 - Be **simple** so others understand,
 - Match **known facts** (John has a criminal history),
 - Be testable **quickly** (not after 50 years),
 - And actually **explain the crime**.

That’s exactly how a good hypothesis works in research.

- ✓ In summary, a **good hypothesis** is:

Clear, Testable, Relational, Specific, Simple, Consistent with facts, Practical (time-bound), and Explains the problem.

Multivariate Analysis – Explained Simply

1. What it means

- **Univariate analysis** = You study **one variable** at a time.
Example: Checking only people's **height** in a study.
- **Bivariate analysis** = You study **two variables** and their relationship.
Example: Checking if **height and weight** are related.
- **Multivariate analysis** = You study **many variables at the same time** and how they are related.
Example: Looking at how **height, weight, diet, and exercise** together affect health.

👉 In short: **Multivariate = many variables together.**

2. Why do we need it?

If we analyze each variable separately (univariate), we might miss the **connections** between them.

- Example: Imagine you're studying **students' performance**.
 - If you only look at "study hours," you might say: *More hours → better performance.*
 - But if you include **sleep, nutrition, motivation, and teaching quality** together, you get a more **realistic and correct picture**.

This is why **multivariate analysis** is powerful – it shows the **full story**, not just pieces.

3. Mathematics behind it (easy version)

- It uses **linear algebra** (matrices, vectors).
- Instead of just "mean" and "variance" (used in single-variable statistics), it uses:
 - **Mean vector** → average of many variables.
 - **Covariance matrix** → shows how variables move together.

👉 Example: If **height and weight** increase together, their covariance is high.

4. Growth of Multivariate Techniques

- In the last **50 years**, statisticians developed these tools because single-variable studies gave **misleading results**.
 - With computers becoming powerful, multivariate analysis became easier and very common.
-

5. Where is it used?

Multivariate analysis is used in many fields where variables are connected:

- **Economics** → Predicting inflation using many factors (money supply, demand, employment, etc.).
 - **Medicine** → Studying how diet, exercise, stress, and genetics together affect heart disease.
 - **Psychology** → Analyzing personality traits, behavior, and social background together.
 - **Agriculture** → Studying how rainfall, fertilizer, and soil quality affect crop yield.
 - **Business/Marketing** → Understanding how price, packaging, advertising, and brand image affect sales.
-

6. Example to Imagine

Think of baking a cake 🎂.

- If you only check **sugar**, you don't know if the cake will taste good.
- If you only check **flour**, still incomplete.
- But if you consider **sugar + flour + butter + eggs + baking time**, you get the correct outcome.

That's exactly what **multivariate analysis** does – it looks at **all ingredients (variables)** together instead of separately.

In summary:

- **Multivariate analysis** = analyzing many variables at the same time.
- It avoids mistakes from looking at variables separately.
- Uses math tools like mean vector & covariance matrix.
- Widely used in economics, medicine, psychology, agriculture, etc.
- Computers made it easy and powerful in recent decades.

Use of Standard Statistical Software

1. What is it?

- Statistical software = computer programs that help us **collect, organize, analyze, interpret, and present data.**
- It makes complex statistics easier to use and understand.
- Useful for **business managers, researchers, analysts, and decision-makers.**

👉 Think of it like a **calculator + brain + artist**:

- **Calculator** → does all the math quickly.
 - **Brain** → applies statistical logic.
 - **Artist** → shows results in graphs, charts, and dashboards.
-

2. Who uses it?

- **Business** → trends & forecasts.
 - **Academia & Research** → sociology, economics, political science.
 - **Healthcare & Biology** → patient data, drug research.
 - **Engineering & Manufacturing** → quality testing, reliability.
 - **Finance** → risk analysis, predictions.
-

3. Key Features

1. **Data Importation**
 - Import data from **databases, spreadsheets, text files, PDFs, RTF, HTML**.
 - Example: Pull sales data from Excel or customer data from SQL.
2. **Preparing Data**
 - Clean and organize before analysis.
 - Techniques:
 - Data sampling
 - Variable transformations
 - Coding & discretization
 - Creating contingency tables
 - Example: Converting age into categories (Child, Teen, Adult).
3. **Modeling Data**
 - Apply statistical tests and models:
 - Regression (predicting sales from price/ads)
 - ANOVA / ANCOVA (compare group means)
 - MANOVA (compare groups on multiple outcomes)
 - Logistic regression (yes/no outcomes like loan approval).
4. **Dashboards & Visualization**
 - Create **charts, graphs, maps, animations**.
 - Drag-and-drop dashboards → customize reports.
 - Example: CEO views a dashboard showing profit trends by region.
5. **Analysis & Reporting**
 - Perform **regression, cluster analysis, multivariate analysis, Bayesian analysis**.
 - Generate **easy-to-read business reports**.

- Example: Report on customer segments for targeted marketing.
 - 6. **Multi-platform Support**
 - Works on **Windows, Linux, UNIX, MacOS**.
 - Supports multiple programming languages (C#, Java, Python, C++).
-

4. Benefits

- **Efficiency** → Organizes and cleans data faster.
 - **Informed Decisions** → Managers base decisions on facts, not guesses.
 - **Predictions** → Forecast future trends (e.g., sales, demand, disease spread).
 - **Visualization** → Easy to explain results with graphs/charts.
-

5. Popular Statistical Software

- **Business & Research:** IBM SPSS, Stata, SAS, Minitab, XLSTAT.
 - **Academia & Science:** MATLAB, Mathematica, EViews, JMP.
 - **Biology & Healthcare:** MedCalc, NCSS, Qlucore.
 - **Engineering & Forecasting:** Forecast Pro, Statgraphics, SigmaPlot.
 - **Open/Programming-based:** R, Python (with libraries like Pandas, NumPy, SciPy, StatsModels).
-

Quick Example to Imagine

Suppose a company wants to **predict sales**.

- They import last year's sales, advertising spend, and seasonal data into SPSS.
- Clean the data (remove errors).
- Run **regression analysis** → find that *ads + discounts increase sales by 20%*.
- Create a **dashboard** → managers see graphs of sales trend.
- Report helps them **plan the budget for next year**.

 Without software = takes weeks.

 With software = done in minutes.

In summary:

- Standard statistical software is essential for handling big, complex data.
- It helps in **data preparation, modeling, visualization, and reporting**.
- It improves **efficiency, predictions, and decision-making**.
- Examples: SPSS, SAS, Stata, R, Python, MATLAB, Minitab.

Data Processing

1. What is Data Processing?

Imagine you are making **fruit juice** .

- You first **collect fruits** (raw data).
- Then you **wash, cut, and blend** them (processing).
- Finally, you **pour juice into a glass** so people can drink it (usable information).

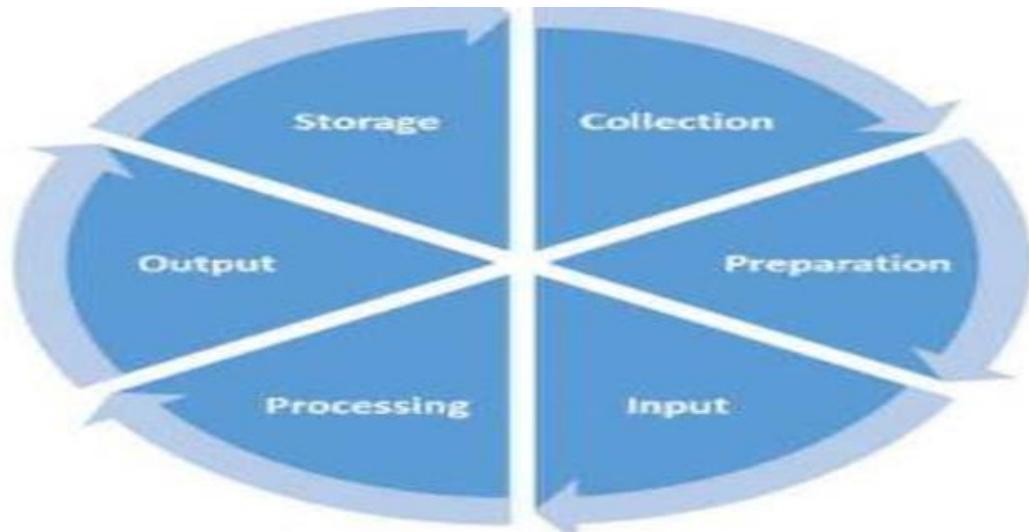
That's exactly what companies do with **raw data**:

- Raw data by itself (like customer age, sales numbers, survey responses) is not useful.
- It must be into a form that employees can actually use → like **reports, charts, or dashboards**.

 So, **Data Processing** = turning raw data into meaningful information.

2. Why is it Important?

- **Better business decisions** → Managers can see trends (e.g., sales drop in winter, rise in summer).
- **Competitive edge** → A company that understands its data can beat rivals.
- **Easy understanding** → Data shown as graphs or charts can be understood by anyone, not just data scientists.
- **Future planning** → Helps predict demand, customer behavior, risks, etc.



3. Data Processing Cycle

This is the **step-by-step process**. Think of it as a **loop**, because the output of one cycle can be reused as input for another.

Steps:

1. **Collection**
 - Gather raw data from sources → surveys, sensors, transactions, online forms, databases.
 - Example: A retail store collects purchase data from billing machines.
 2. **Preparation (Cleaning & Sorting)**
 - Remove errors, duplicates, or incomplete data.
 - Example: If “age = -5” is in data, it must be corrected or removed.
 3. **Input**
 - Feed cleaned data into the system/software for processing.
 - Example: Enter sales data into Excel/SPSS/database.
 4. **Processing**
 - Apply formulas, algorithms, or statistical methods.
 - Example: Calculate *average sales per month* or *growth rate*.
 5. **Storage**
 - Save processed data for future use.
 - Example: Store it in cloud databases or data warehouses.
 6. **Output (Information)**
 - Present results in user-friendly formats → reports, graphs, dashboards.
 - Example: A pie chart showing 40% sales from online vs 60% from offline.
 7. **Feedback / Reuse (Cycle Repeats)**
 - The output can become input for another cycle to get deeper insights.
 - Example: Use last year’s sales analysis to improve this year’s forecast.
-

4. Simple Example

A **hospital** collects data on patients:

- Raw data: Names, ages, symptoms, diagnosis, treatment cost.
- Data processing:
 1. Collect patient records.
 2. Clean errors (remove duplicates, fix typos).
 3. Enter into hospital database.
 4. Process → Group patients by disease type, calculate average cost per patient.
 5. Store in digital records.
 6. Output → A report for doctors showing “*Most common diseases this month*” or “*Average cost of treatment*”.

This helps doctors and managers make better decisions.

In summary:

- **Data Processing** = turning raw data into useful information.
- It follows a **cycle**: Collection → Preparation → Input → Processing → Storage → Output → Feedback.
- It is essential for **decision-making, planning, and competitive advantage**.

Preliminary Data Analysis and Interpretation

1. What is it?

Think of it like **tasting food while cooking** .

- You don't serve it yet (final analysis).
- First, you **check the taste, smell, salt, and spice levels** → this helps you know if you need to adjust.

Similarly, in research:

- **Preliminary Data Analysis (PDA)** = first check of raw data before doing full statistical analysis.
 - It helps to **detect problems early** and understand **basic patterns**.
-

2. Objectives of Preliminary Data Analysis

 The main goals are:

1. **Edit the data** – Clean the dataset (remove mistakes, fill missing values).
 2. **Describe the data** – Look at features like mean, median, mode, range, etc.
 3. **Summarize results** – Create simple summaries (tables, charts) for a quick understanding.
-

3. Steps / Things to Consider

1. **Look at the dataset**
 - Check number of rows & columns.
 - See if there are missing or incorrect values.
 - Example: Age = 200 years → clearly wrong!
2. **Check reliability**
 - Make sure your measures are consistent.
 - Example: If a survey has repeated questions, answers should not conflict.
3. **Evaluate effectiveness of manipulations**

- In experiments, check if your treatment or manipulation worked.
 - Example: If you gave training, did test scores improve at least a little?
4. **Examine distributions**
 - Look at variable distributions (normal, skewed, etc.).
 - Example: If most people scored between 40–60, but some got 5 or 100 → the distribution may be skewed.
 5. **Identify outliers**
 - Detect extreme values that can distort results.
 - Example: One student scoring 100% in a class where average is 50%.
-

4. Use in Different Research Types

- **Quantitative Research** → Checking numbers, averages, trends.
 - **Quasi-experimental Research** → Checking if groups differ before applying treatments.
 - **Non-experimental Research** → Understanding survey responses, patterns, etc.
-

5. Example

A company conducts a **survey on employee satisfaction** (scale 1–10):

1. **Edit data** → Remove duplicate responses, fix blank entries.
 2. **Describe data** → Average score = 6.2, median = 6.
 3. **Summarize results** → Most employees rate 5–7 (moderate satisfaction).
 4. **Check distributions** → A few people rated “1” → possible unhappy employees.
 5. **Interpretation** → Company may need to improve working conditions for some employees.
-

✓ In short:

- **Preliminary Data Analysis** = first look at data before formal tests.
- Helps in **cleaning, summarizing, detecting errors/outliers, and refining research**.
- Ensures final results are **valid and trustworthy**.

Univariate and Bivariate Analysis

1. Univariate Analysis (Single Variable)

👉 “Uni” = **one**.

- Here, you study **only one variable** at a time.
- Purpose = **describe, summarize, and find patterns** in that variable.

Example:

- A teacher wants to know the **average marks** of students in Math.
- Variable = *Marks in Math*.
- Methods: Mean, Median, Mode, Frequency table, Histogram, Box plot.
- Result: “Most students scored between 50–70 marks.”

👉 So, univariate = **describing one variable** (its distribution, spread, and center).

2. Bivariate Analysis (Two Variables)

👉 “Bi” = **two**.

- Here, you study **two variables together**.
- Purpose = **see if they are related** (correlation, association, cause-effect).

Example:

- Same teacher now checks if **study hours (X)** affect **Math marks (Y)**.
- Variables = *Study hours* and *Marks*.
- Methods: Correlation, Regression, Cross-tabulation, Scatterplot.
- Result: “Students who studied more hours scored higher marks.”

👉 So, bivariate = **relationship between two variables**.

3. Comparison Table

Feature	Univariate Analysis	Bivariate Analysis
Meaning	Study of one variable	Study of two variables
Purpose	Describe, summarize	Compare, find relationship
Example	Average age of students	Relation between age & height
Tools	Mean, Median, Histogram, Boxplot	Correlation, Regression, Scatterplot
Output	Describes patterns	Explains associations

4. Real-Life Analogy

- **Univariate** = Looking at one flower  and describing its color, size, and petals.
 - **Bivariate** = Comparing two flowers   → “Does more sunlight make the flower bigger?”
-

In short:

- **Univariate** = single variable, descriptive analysis.
- **Bivariate** = two variables, relationship analysis.
- Researchers often use both together: first describe variables (univariate), then explore relationships (bivariate).

Hypothesis Testing

Definition:

Hypothesis testing is a **statistical method** used to check whether the assumption (hypothesis) you made about a population parameter is **true or false** using sample data.

It's like a **court trial**: you assume the person is "not guilty" (null hypothesis), then test if evidence (data) proves otherwise.

Steps in Hypothesis Testing

1. State the Hypotheses

- **Null Hypothesis (H_0)**: Assumes **no effect / no difference / no relationship**.
Example: "*The average marks of Group A = Group B.*"
- **Alternative Hypothesis (H_1)**: Assumes **effect / difference / relationship exists**.
Example: "*The average marks of Group A \neq Group B.*"

 Think: H_0 = **boring / status quo**, H_1 = **what we want to test**.

2. Collect Data

- Gather data using surveys, experiments, or observations.
- Ensure sample is representative of population.
Example: Take **50 students' marks** from Group A and Group B.

3. Perform a Statistical Test

- Choose a test depending on data type & research design:
 - **t-test** → Compare means of 2 groups.
 - **Chi-square test** → Compare categories (yes/no, male/female).
 - **ANOVA** → Compare means of 3+ groups.
 - **Regression / Correlation** → Test relationships between variables.

4. Decide Whether to Reject H_0

- Choose **significance level ($\alpha = 0.05$)** → means 5% risk of being wrong.
- Compare **p-value** (probability from test) with α :
 - If $p \leq 0.05$ → Reject H_0 (there is evidence for H_1).
 - If $p > 0.05$ → Do not reject H_0 (no evidence for H_1).

Example:

t-test result shows $p = 0.02$ → Since $0.02 < 0.05$ → Reject H_0 → Group A and B have different average marks.

5. Present the Findings

- Report results with clarity:

Example:

“A t-test revealed a significant difference between Group A and Group B ($t = 2.45, p = 0.02$). Hence, we reject the null hypothesis and conclude that the two groups differ in their average marks.”

Summary

- Hypothesis testing = a **scientific decision-making tool**.
 - Steps = (1) State H_0 & H_1 → (2) Collect Data → (3) Statistical Test → (4) Accept/Reject H_0 → (5) Report Results.
-

✓ Real-Life Analogy:

- H_0 : “Medicine has no effect.”
- H_1 : “Medicine reduces fever.”
- After testing → If $p < 0.05$ → We **reject H_0** and conclude the medicine works.

PYQ:

A) What is Research? Explain Types of Research with Examples

Definition:

Research is a **systematic and scientific investigation** undertaken to increase knowledge

about a topic, solve problems, or discover new facts.

It involves collecting, analyzing, and interpreting information to answer specific questions.

Types of Research:

1. **Basic (Fundamental) Research:**
 - Objective: To gain knowledge for **the sake of knowledge**.
 - Example: Studying the effects of a new enzyme on human cells.
 2. **Applied Research:**
 - Objective: To **solve practical problems**.
 - Example: Developing a new vaccine for a disease.
 3. **Descriptive Research:**
 - Objective: To **describe characteristics** of a population or phenomenon.
 - Example: Surveying students' study habits.
 4. **Analytical Research:**
 - Objective: To **analyze existing information** to explain phenomena.
 - Example: Studying causes of inflation by analyzing economic data.
 5. **Exploratory Research:**
 - Objective: To **explore an area** where little information exists.
 - Example: Investigating social media trends among teenagers.
 6. **Experimental Research:**
 - Objective: To **test cause-and-effect relationships**.
 - Example: Testing the effect of fertilizer type on crop yield.
-

A) What is Research Design? Explain Need and Features

Definition:

Research design is a **blueprint or plan for conducting research**. It specifies **what, where, when, how, and by whom** the research is carried out.

Need for Research Design:

1. Ensures smooth research operations.
2. Saves time, effort, and money.
3. Guides proper collection and analysis of data.
4. Helps in making accurate conclusions.

Features of Research Design:

1. Specifies **sources and types of data**.
2. Provides **strategy for data collection and analysis**.
3. Considers **time and cost budgets**.
4. Contains:
 - Clear statement of research problem.
 - Procedures/techniques for gathering information.
 - Population to be studied.
 - Methods for processing and analyzing data.

A) Explain Research Process with Block Diagram

Research Process Steps:

1. Identify the problem
2. Review literature
3. Formulate hypothesis
4. Prepare research design
5. Collect data
6. Process and analyze data
7. Interpret results
8. Report findings

Block Diagram:

Problem Identification → Literature Review → Hypothesis Formulation → Research Design → Data Collection → Data Processing & Analysis → Interpretation → Report Writing

A) Define Research Problem and Technique Involved in Defining Problem

Research Problem:

A research problem is a **specific issue, difficulty, or gap in knowledge** that a researcher wants to investigate.

Techniques to Define a Problem:

1. State problem in **general terms**.
2. Understand **nature of the problem**.
3. **Survey existing literature**.
4. Discuss ideas with experts.
5. Rephrase problem into a **working proposition**.

Other Considerations:

- Define technical terms clearly.
 - State basic assumptions.
 - Mention scope, time, and sources of data.
-

B) Define Sampling, Fundamentals and Need

Sampling:

Sampling is the process of selecting a **part of a population** to make inferences about the **whole population**.

Fundamentals of Sampling:

1. Sample should be **representative**.
2. Sampling method should be **unbiased**.
3. Sample size should be **adequate**.

Need for Sampling:

1. Saves **time and money**.
 2. Allows **more accurate measurement**.
 3. Necessary when **population is infinite**.
 4. Useful when **testing destroys the subject**.
 5. Helps **estimate errors** and population characteristics.
-

A) Use of Standard Statistical Software (in brief)

Purpose:

Statistical software helps in **collecting, analyzing, interpreting, and presenting data** for decision-making.

Functions:

1. **Data Import:** From spreadsheets, databases, PDFs.
2. **Data Preparation:** Cleaning, coding, transformations.
3. **Data Modeling:** Regression, ANOVA, MANOVA, Logistic regression.
4. **Visualization:** Graphs, charts, dashboards.
5. **Analysis & Reporting:** Regression, cluster analysis, hypothesis testing.
6. **Multi-platform Support:** Works on Windows, Linux, Mac; supports Python, R, etc.

Examples: SPSS, SAS, R, Stata, Minitab, MATLAB.

A) What is Quantitative Technique? Explain with Diagram

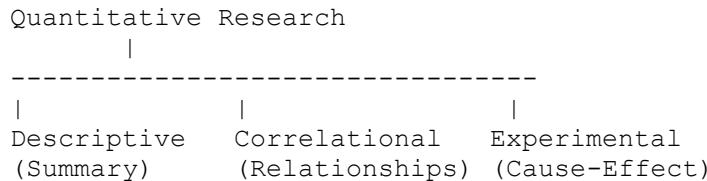
Definition:

Quantitative research uses **numerical data** to find patterns, make predictions, and test hypotheses.

Purpose:

- Summarize variables (descriptive).
- Study relationships (correlational).
- Test cause-effect (experimental).

Diagram:



Example:

- Descriptive: Average salary of employees in a company.
 - Correlational: Relation between hours of study and exam score.
 - Experimental: Effect of fertilizer type on crop yield.
-

B) Processing and Analysis of Data

Data Processing:

- Converting **raw data** into meaningful information.
- Steps: **Editing → Coding → Classification → Tabulation → Analysis → Interpretation**

Analysis of Data:

- **Univariate Analysis:** Single variable (mean, median, mode, frequency distribution).
- **Bivariate Analysis:** Two variables (correlation, regression).
- **Multivariate Analysis:** Multiple variables (MANOVA, Canonical Analysis).

Example:

- Univariate: Average height of students.
- Bivariate: Relationship between height and weight.
- Multivariate: Predicting weight using height, age, and exercise hours.

Steps in Research Process:

1. **Selecting a Research Topic:**
 - Choose something **interesting, feasible, and new**.
 - Example: Studying the effect of mobile apps on student learning efficiency.
2. **Defining the Research Problem:**
 - Narrow down the topic into a **specific question**.
 - Example: “Do mobile learning apps improve students’ test scores?”
3. **Literature Review:**

- Check **what has already been studied** in books, journals, and articles.
 - Helps in avoiding duplication.
4. **Formulating Hypothesis:**
- A **prediction** that can be tested.
 - Example: “Students using app X will score higher than those not using it.”
5. **Research Design:**
- Decide **how** to collect and analyze data.
 - Example: Survey or experiment with two student groups.
6. **Data Collection:**
- Collect information using **questionnaires, interviews, or experiments**.
7. **Data Processing & Analysis:**
- Clean, code, and **analyze data** using graphs, charts, or statistical software.
8. **Interpretation:**
- Draw meaning from data.
 - Example: Group using the app scored 20% higher → app is effective.
9. **Reporting Findings:**
- Present results in **reports, tables, charts** with clear conclusions and recommendations.
-

2. Significance of Peer-Reviewed Journals

- Articles are **checked by experts** for accuracy and quality.
- Ensures your research is **trustworthy**.
- Helps researchers **stay updated** and **avoid mistakes**.
- Example: Publishing in *Nature* or *Science* means your research is recognized globally.

Finding Reputable Journals:

- Indexed in **Scopus or Web of Science**.
 - High **impact factor** (widely cited).
 - Published by **recognized academic publishers**.
-

3. Data Collection

Primary Data:

- **Original data** collected by researcher.
- Example: Asking 100 students about study hours using a survey.

Secondary Data:

- Data already **collected by someone else**.
- Example: Using census data or school exam records.

Significance: Primary data = **fresh and specific**; Secondary = **easy and cheap**.

4. Intellectual Property Rights (IPR)

- Legal rights protecting **creations of mind**.
- Encourages innovation and protects ownership.

Types of IPR with Example:

1. **Patent:** Protects inventions → Example: New battery design.
2. **Copyright:** Protects books/music → Example: Harry Potter book.
3. **Trademark:** Protects brand/logo → Example: Coca-Cola logo.
4. **Industrial Design:** Protects design → Example: Car shape.
5. **Geographical Indications:** Linked to place → Example: Darjeeling Tea.

Factors in Technical Report Writing:

- Clear and simple language
 - Proper structure: Introduction → Methods → Results → Conclusion
 - Visuals: Graphs, charts
 - Accurate citations
-

5. Copyright

- Protects **literary, artistic, or musical work**.
- **Classes:**
 1. Books, articles
 2. Paintings, sculptures
 3. Songs, music
 4. Films
 5. Software

6. Geographical Indications (GI)

- **Definition:** A sign for products from a specific place with unique quality.
 - **Example:** Darjeeling Tea (India), Champagne (France)
 - **Rights for GI Holders:**
 - Exclusive use of GI
 - Prevent imitation
 - Legal protection
-

7. Difference Between Univariate, Bivariate, and Multivariate Data

Type	No. of Variables	Use	Example
Univariate	1	Describe a single variable	Average marks of students
Bivariate	2	Study relationship between 2 variables	Height vs weight
Multivariate	3 or more	Study relationships among many variables	Predict weight using height, age, diet

8. Statistical Significance

- Shows if results are **real or due to chance**.
- Example: $p\text{-value} < 0.05 \rightarrow$ difference is significant.
- Importance:** Validates research and ensures results are reliable.

9. Literature Review

- Helps **identify gaps** and avoid repeated studies.
- Helps in **defining the research problem**.
- Example: Reading journals to see what studies exist on mobile learning apps.

10. Process of Filing Patent Applications (Flow Chart)

Idea → Patent Search → Draft Application → Submit to Patent Office → Examination → Publication → Grant → Renewal/Licensing

- Helps protect **new inventions** legally.
- Example: Inventing a new eco-friendly water purifier.

11. Hypothesis

- Definition:** Testable prediction linking variables.
- Example:** “Students who use mobile apps learn faster than those who don’t.”

Significance:

- Gives **direction to research**

- Helps in **planning experiments**
- Guides **data collection & analysis**
- Supports **drawing valid conclusions**