

Statistics and Data Analysis

Unit 01 – Lecture 01: Data Types, Sources, and Cleaning Basics

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<https://github.com/tali7c/Statistics-and-Data-Analysis>

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Learning Outcomes

- Identify common data types and formats used in analytics

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- List common data sources and acquisition methods
- Detect typical data quality issues (missing values, duplicates, outliers)
- Apply basic cleaning steps in Python and save a cleaned dataset

Dataset, Observation, Variable

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Goal: convert raw data into a form suitable for analysis and modeling.

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- **Observation:** one record (e.g., one student)
- **Variable/Feature:** one attribute (e.g., attendance, CGPA)

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- **Date/Time:** join date, timestamp
- **Text:** feedback, comments

Exercise 1: Classify Variable Types

For each variable, write the type
(numeric/categorical/binary/datetime/text):

- 1 Age
- 2 Program/Branch (CSE, ECE, ...)
- 3 Attendance (%)
- 4 Join date
- 5 Feedback comment

Solution 1

- Age: numeric (integer)
- Program/Branch: categorical (nominal)
- Attendance (%): numeric (real)
- Join date: datetime
- Feedback: text (unstructured)

Data Formats

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Examples: JSON, XML
- **Unstructured:** free-form content
Examples: text documents, images, audio

Structured Example (Table)

student_id	program	attendance_pct	cgpa
1001	CSE	92	8.2
1002	CSE	85	7.5
1003	ECE	105	8.9

Note: 105% attendance is an example of an out-of-range value.

Semi-structured Example (JSON)

```
{  
  "student_id": 1001,  
  "program": "CSE",  
  "attendance_pct": 92,  
  "courses": ["Math", "DSA", "Stats"]  
}
```

Keys may vary from record to record (flexible schema).

Unstructured Example (Text/Log)

2026-02-08 10:02:11 INFO login user=1007 device=android city=Del

Useful information exists, but it requires parsing and feature extraction.

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- Databases (student records, attendance systems)
- Web and app logs (clickstream)
- Sensors/IoT (temperature, GPS)
- Public datasets (government portals, research repositories)

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- Database query: SQL → extract tables
- API calls: JSON responses → parse and store
- Manual entry: small datasets (careful with errors)

Exercise 2: Choose a Source

For each case, suggest a likely source (survey/database/log/API):

- 1 Daily attendance of students
- 2 Online learning platform clicks
- 3 Student feedback comments
- 4 Weather readings every minute

Solution 2

- Attendance: database export (or CSV from attendance system)
- Clicks: logs (web/app logs)
- Feedback: survey + text field (unstructured text)
- Weather readings: sensors/IoT or API

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- Models and statistics assume data is meaningful and consistent
- “Garbage in, garbage out” → wrong conclusions
- Cleaning improves: accuracy, fairness, and reproducibility

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- Missing values (blank, NaN, NULL)
- Duplicates (same record repeated)
- Inconsistent categories (cse, CSE, CSE)
- Out-of-range values (attendance 105%, CGPA 12)
- Wrong data type (“nine” instead of 9.0)

Handling Missing Values (Basic Options)

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- **Domain rule:** fill with a meaningful default (carefully)
- **Flag:** create an indicator feature “was_missing”

Exercise 3: Missingness Decision

In a dataset of 20 students, the column cgpa has 2 missing values.

- What is the missingness percentage?
- Suggest one reasonable action for this column.

Solution 3

- Missingness = $2/20 \times 100\% = 10\%$
- Action: impute using **median** CGPA (robust) and optionally add a flag

Outliers (Basic Idea)

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An outlier is a value that is unusually far from typical values.

- Outliers can be **errors** (wrong entry) or **real extremes**
- They can strongly affect mean, variance, and some models
- Use rules like IQR fences as a **screening** step

IQR Rule (Fences)

Lower fence = $Q_1 - 1.5 \times \text{IQR}$, Upper fence = $Q_3 + 1.5 \times \text{IQR}$

$$\text{IQR} = Q_3 - Q_1$$

Values outside fences are *possible* outliers.

Exercise 4: IQR Outlier Check

Attendance (%): 70, 75, 80, 85, 90, 95, 150

Task: Compute Q_1 , Q_3 , IQR, fences, and decide if 150 is an outlier.

Solution 4

Sorted data: 70, 75, 80, 85, 90, 95, 150 ($n=7$). Median = 85.

Lower half: 70, 75, 80 $\Rightarrow Q_1 = 75$

Upper half: 90, 95, 150 $\Rightarrow Q_3 = 95$

$IQR = 95 - 75 = 20$

Fences: $75 - 30 = 45$ and $95 + 30 = 125$

Conclusion: $150 > 125 \Rightarrow$ outlier (by IQR rule).

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- Standardize categories (trim whitespace, normalize case)
- Check ranges and impossible values
- Save a cleaned version (do not overwrite raw file)

Mini Demo (Python)

Run from the lecture folder:

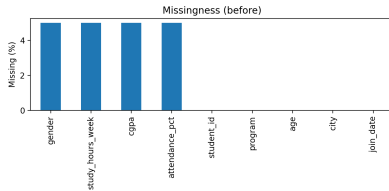
```
python demo/cleaning_demo.py
```

Outputs:

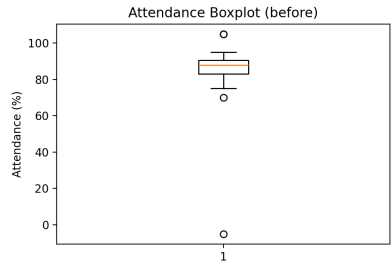
- data/students_clean.csv
- plots in images/ (missingness and outlier visual)

Demo Output (Example)

Missingness



Attendance Outliers



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Summary

- Data types and formats determine how we store and process data
- Different sources require different acquisition and validation steps
- Cleaning deals with missing values, duplicates, inconsistencies, and outliers
- Always save a cleaned dataset and document the rules you applied

Exit question: In one sentence, why can “attendance 105%” be dangerous in analysis?