

# **SNS COLLEGE OF TECHNOLOGY**

**An Autonomous Institution  
Coimbatore-35**



## **DEPARTMENT OF ARTIFICIAL INTELLIGENCE & DATA SCIENCE**

### **23ADT202 – FUNDAMENTALS OF DATA SCIENCE AND ANALYTICS**

**II YEAR IV SEM**

#### **UNIT I – INTRODUCTION TO DATA SCIENCE**

**FACETS OF DATA**

# FACETS OF DATA

In data science and big data, you encounter **many different types of data**, and **each type often requires different tools, storage methods, processing techniques, and analysis strategies..**



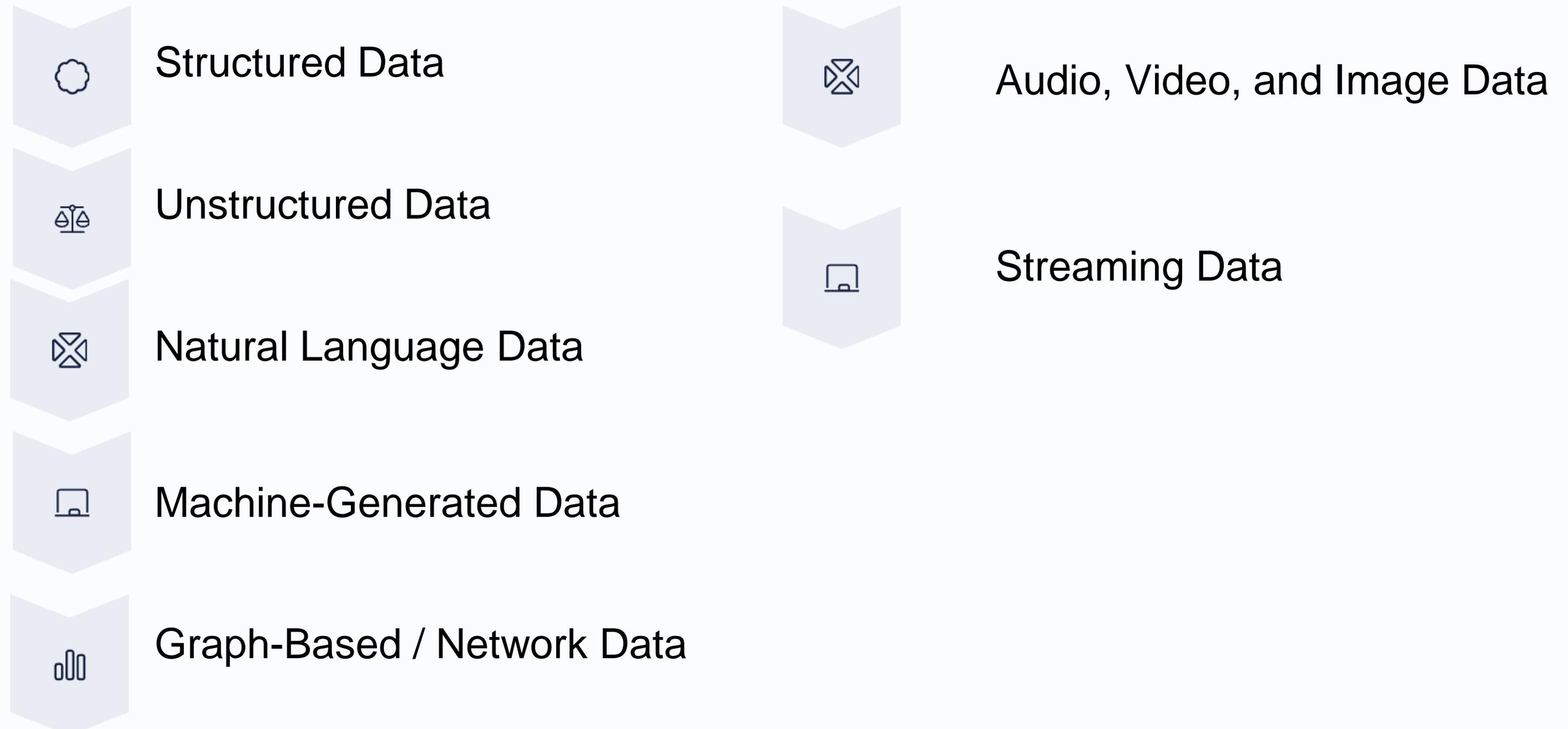
# 1. Empathy

- Data scientists face difficulty handling diverse data formats. Organizations lack clarity on which tools suit each data type.
- Decision-makers need combined insights from multiple data sources.
- Traditional databases fail to process unstructured and streaming data. There is confusion in selecting analytics methods for different data facets.

## Problem Statement: Facets of Data in Data Science

- Modern organisations collect data in multiple forms such as structured tables, unstructured text, images, audio, machine-generated logs, network graphs, and real-time streams.
- Each facet of data differs in format, velocity, volume, and analytical requirements.
- However, many organisations struggle to integrate and analyse these diverse data types using a unified approach.
- This leads to fragmented insights, inefficient data utilisation, and poor decision-making.
- There is a need for a systematic framework to understand, manage, and analyse different facets of data effectively within the data science lifecycle.

# FACETS of Data in Data Science



## 1. Structured Data

- Data that *fits a predefined model* and has a predictable format (e.g., rows and columns in tables).
- Easy to store, query, and analyze using SQL and relational databases.
- Examples: spreadsheets, transaction records.

## 2. Unstructured Data

- Data that *does not conform to a fixed schema* and is not easily stored in traditional databases.
- Often text-heavy or context-specific, such as emails, documents, logs.
- Requires specialized tools and techniques for meaningful analysis.

## 2. Define

Defined Problem:

How can data science effectively handle and analyse different facets of data—such as structured, unstructured, machine-generated, graph-based, multimedia, and streaming data—to produce meaningful and actionable insights?

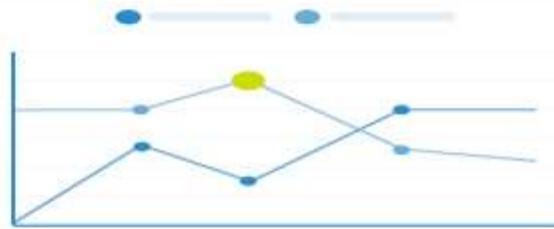
# Standard **Data Visualization** Examples

## Bar Chart



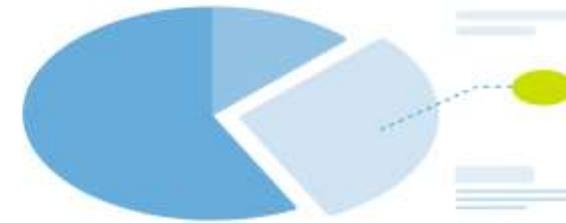
**BEST FOR:** Comparing discrete categories or showing simple distributions.

## Line Chart



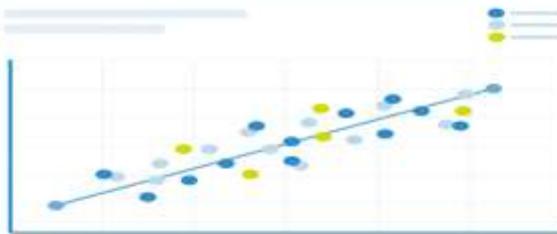
**BEST FOR:** Showing trends over time or continuous data.

## Pie Chart



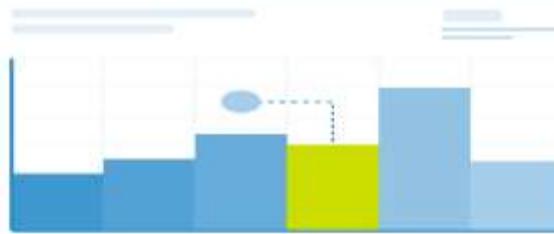
**BEST FOR:** Illustrating the composition or proportion of parts within a whole.

## Scatter Plot



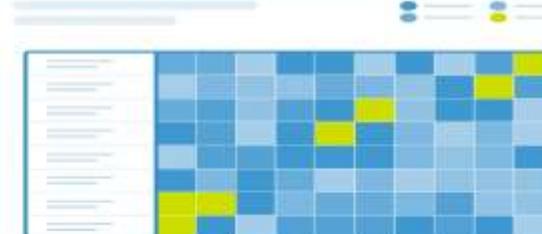
**BEST FOR:** Presenting the relationship between two variables.

## Histogram



**BEST FOR:** Showing the distribution of continuous data.

## Heat Map



**BEST FOR:** Visualizing density and distribution in large datasets.



## UNSTRUCTURED DATA EXAMPLES



Text documents



Emails



Images



Audio files



Video files



Log files



Sensor data



Social media posts

### 3. Natural Language Data

- A *special type of unstructured data* that consists of human language (text or speech).
- Analysis often uses Natural Language Processing (NLP) techniques
- for tasks like sentiment analysis, entity recognition, summarization, etc.

### 4. Machine-Generated Data

- Automatically produced by systems, sensors, or machines.
- Includes logs, telemetry from devices, clickstreams, and server logs.
- Often high-volume and streaming in nature.

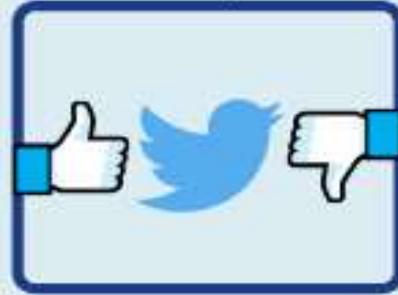
### 3. Ideate

- Classify data based on structure, source, and behaviour. Apply suitable storage systems (SQL, NoSQL, graph databases).
- Use NLP for text data and computer vision for image/video data. Apply graph analytics for relationship-based data. Use stream processing frameworks for real-time data analysis.

Information Retrieval



Sentiment Analysis



Information Extraction



# Natural Language Processing (NLP)

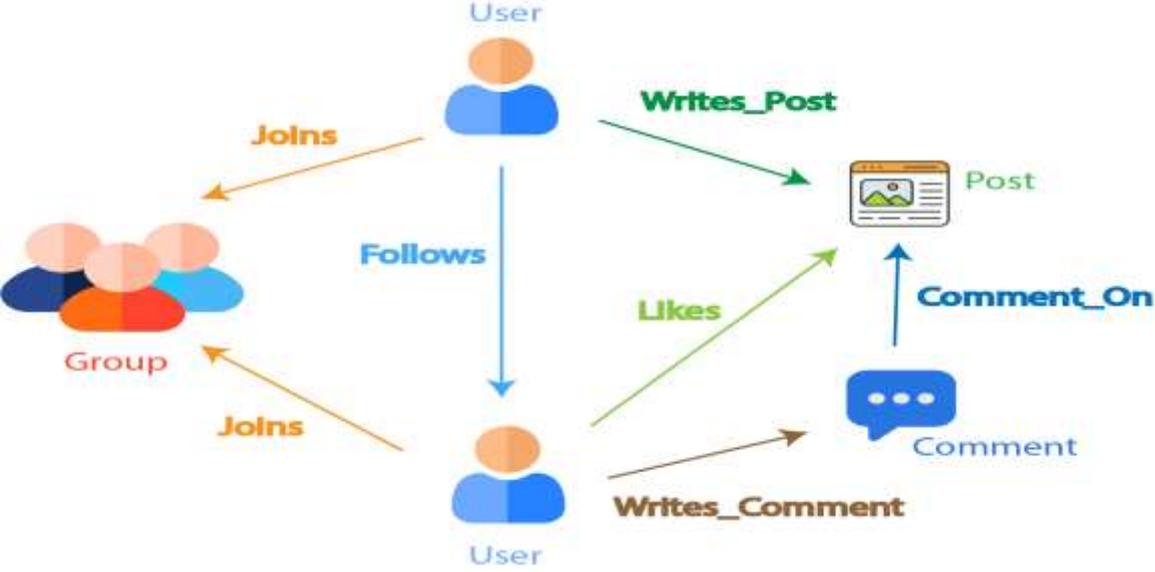
Machine Translation



Question Answering

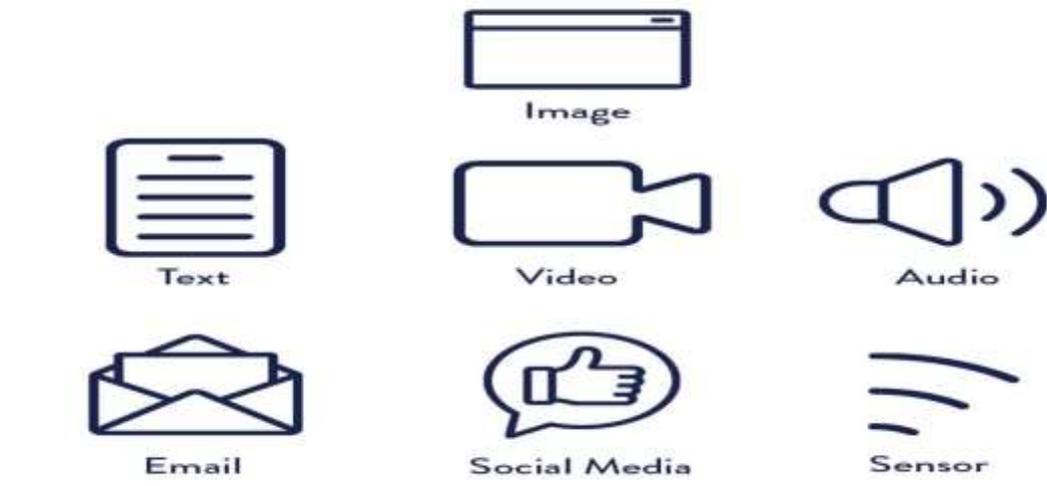






## 5. Graph-Based / Network Data

- Data that represents *relationships* between entities (nodes and edges).
- Useful in social network analysis, recommendation systems, and network optimisation.

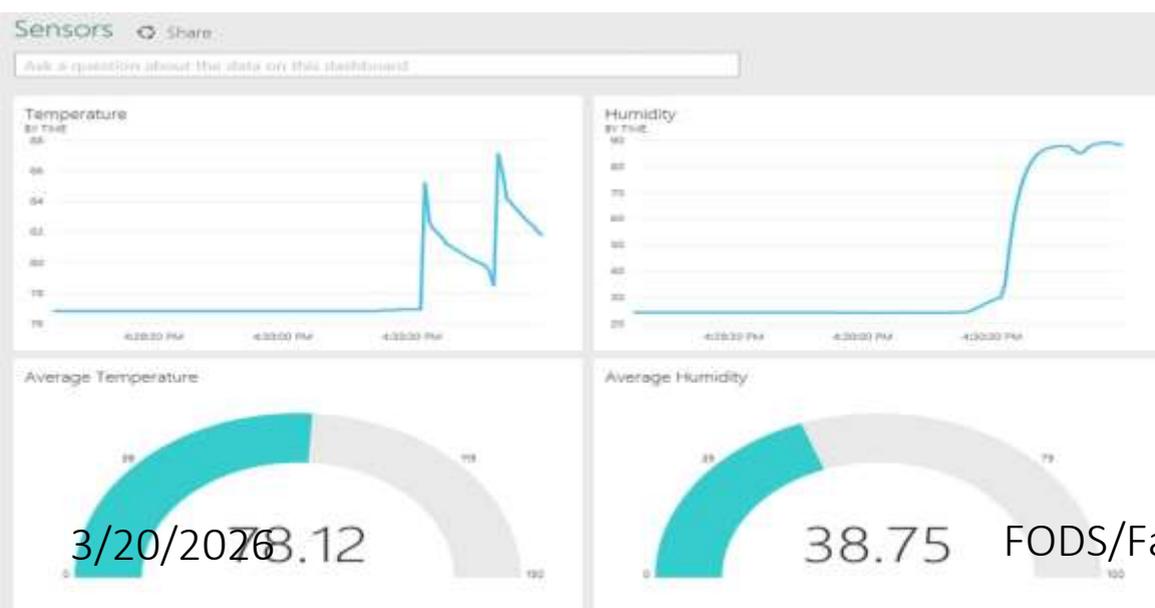


## 6. Audio, Video, and Image Data

- *Multimedia data*, which is unstructured and often requires computer vision, pattern recognition, or signal processing techniques for interpretation.

## 7. Streaming Data

- Data that *flows continuously in real time*, rather than being stored first and processed later.
- Examples include IoT sensor streams, live user interactions, and financial tick data.



## 4. Prototype

- Design a multi-facet data pipeline. Implement structured data analysis using relational databases.
- Build NLP models for text and language data. Develop dashboards for machine-generated and streaming data. Integrate graph visualisations for network-based insights.

## Why These Facets Matter in Data Science

The **diversity of data formats and sources** directly impacts how data scientists approach analysis:

- ✓ Determines *data storage and management systems* (e.g., SQL vs NoSQL)
- ✓ Affects *data preprocessing and transformation strategies*
- ✓ Guides the selection of *analysis and machine learning techniques*
- ✓ Influences *tools and libraries* used (e.g., NLP vs image processing frameworks)

## 5. Testing

- Validate accuracy and performance for each data type.
- Test system scalability with increasing data volume. Evaluate real-time response for streaming data. Compare insights generated from individual vs integrated data facets.
- Collect user feedback and refine analytical models.

### Expected Outcomes / Benefits

- Clear understanding of different facets of data. Improved data integration and analytics efficiency. Better tool and technique selection.
- Enhanced decision-making from diverse data sources. Scalable and flexible data science solutions