

Data





What is Data?

raw information in the form of numbers, text, images, or symbols











Data Formats

Structured (spreadsheets, databases)
Unstructured (emails, videos, social media posts)
Semi-structured (JSON, XML)

Data Processing Cycle

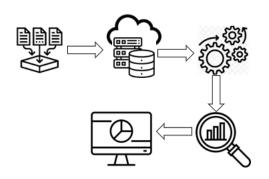
Collection (Sensors, surveys, transactions)

Storage (Databases, cloud, servers)

Processing (Sorting, filtering, analyzing)

Analysis (Trends, patterns, insights)

Visualization (Graphs, charts, dashboards)





Data Types & Examples

Quantitative (Numbers) \rightarrow Sales figures, temperature Qualitative (Descriptions) \rightarrow Customer reviews, comments Big Data (Massive sets) \rightarrow Social media trends, IoT sensor data



Importance of Data

Better Decisions (Business strategies, healthcare, AI)

Efficiency (Automation, predictive models)

Innovation (Machine learning, scientific research)









Data Challenges

Data Privacy & Security (Hacks, leaks, GDPR)

Data Overload (Too much data, hard to analyze)

Bias & Accuracy (Incorrect or misleading data)

Data Analysis

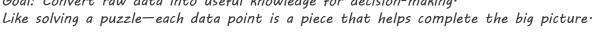


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What is Data Analysis?

Process of cleaning, transforming, and interpreting data To find meaningful patterns, trends, and insights

Goal: Convert raw data into useful knowledge for decision-making.





Why is Data Analysis Important?

Better Decision-Making - Data-driven insights lead to smarter choices
Problem-Solving - Identifies inefficiencies, risks, and opportunities
Predicting Trends - Helps businesses prepare for future changes
Competitive Advantage - Effective data analysis outperform others

Types of Data Analysis

Descriptive Analysis - "What happened?" (e·g·, sales reports, trend charts)

Diagnostic Analysis - "Why did it happen?" (e·g·, correlation, root cause analysis)

Predictive Analysis - "What might happen?" (e·g·, forecasting, machine learning)

Prescriptive Analysis - "What should we do?" (e.g., decision-making models)





Common Data Analysis Techniques

Statistical Analysis - Mean, median, variance, hypothesis testing

Data Visualization - Charts, graphs, heatmaps for easy understanding

Correlation & Regression - Finding relationships between variables

Machine Learning Models - Al-driven pattern recognition

Text Analysis - Extracting insights from words and language

Challenges in Data Analysis

Dirty Data - Incomplete, inconsistent, or incorrect data

Data Overload - Too much data without clear focus

Bias & Misinterpretation - Drawing incorrect conclusions

Lack of Skills & Tools - Not everyone is trained in data analysis









Data Mesh

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What is Data Mesh?

a decentralized approach to data architecture Moves away from centralized data lakes to domain-driven, selfserve data ownership

Instead of one giant warehouse, each team has its own organized data store

Why Data Mesh? (Benefits)

Scalability - No central team bottleneck

Faster Insights - Teams access the data they need without delays

Ownership & Quality - Teams take responsibility for reliable, high-quality data

Flexibility - Works with data lakes, warehouses, and real-time processing

Core Principles of Data Mesh

Domain-Oriented Ownership - Teams own & manage their data as a product

Data as a Product - Treat data like a service with defined consumers & quality standards

Self-Serve Infra - Empower teams to store, process, & share data independently

Federated Governance - Enforce global security, privacy, and standards

How Data Mesh Works

Each business unit (Finance, Marketing, HR, etc.) manages its own data Data is discoverable, shareable, and reusable across teams A common platform ensures security & interoperability without central bottlenecks

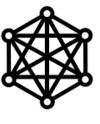


Challenges of Data Mesh

Cultural Shift - Teams must take ownership of data

Standardization Needed - Common governance rules must be enforced

Tech Complexity - Requires the right tools for seamless self-service





AI

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What is AI?

simulation of human intelligence in machines

Learning (Adapts from data)

Reasoning (Makes decisions)

Self-correction (Improves over time)







Types of AI

 $Narrow \ Al \ (Weak \ Al) \rightarrow Specialized in one task (Siri, Google Translate)$

General AI (Strong AI) \rightarrow Thinks like a human (still theoretical)

Super $AI \rightarrow More$ intelligent than humans (future concept)

Al Subfields

Machine Learning (ML) - Learns from data (Netflix recommendations)

Deep Learning (DL) - Al mimicking the human brain (self-driving cars)

Natural Language Processing (NLP) - Understands human language (Chatbots)

Computer Vision - Recognizes images (Face recognition)

How Al Works



Data Collection - Al learns from massive datasets

Training - Models adjust through experience

Decision Making - Al analyzes patterns

Output & Improvement - Al refines predictions over time

Al in Everyday Life

Voice Assistants (Alexa, Google Assistant)
Recommendation Systems (Netflix, YouTube)

Healthcare (Disease diagnosis, robotic surgery)

Autonomous Vehicles (Self-driving cars)

Finance & Security (Fraud detection, stock predictions)













Al Challenges & Ethics

Bias in Al - Unfair outcomes due to biased data

Privacy Issues - Al tracking and surveillance concerns

Job Automation - Al replacing jobs

Ethical Al - Ensuring Al benefits society









What is Explainable AI (XAI)?

Al models often behave like black boxes—the 'why' remains missing XAI aims to make decisions understandable & interpretable







Why Does Explainability Matter?

Trust - to trust Al decisions

Fairness - to prevent bias & discrimination in Al models

Regulations - to abide by Laws (i.e. GDPR)

Debugging - to improve Al performance

Safety -in healthcare, finance, autonomous systems

How AI Becomes Explainable?

Feature Importance -data points influencing the decision?

Decision Trees - breaking down decision path

Local vs. Global Explanations

Local: Why was this decision made?

Global: How does the model behave in general?

SHAP & LIME - Techniques for interpreting black-box Al

Model Transparency - Using simpler, more interpretable models





Trade-offs: Accuracy vs. Explainability

Deep Learning Models (Black Box)

- Highly accurate but hard to interpret
- Used in image recognition, NLP, etc

Simple Models (Transparent but Less Powerful)

- Decision trees, linear regression are more interpretable
- Used when explanations are critical (e·g· healthcare, finance)

Challenges & Future of XAI

Trade-off: More explainability can reduce performance

Human Interpretation: Even simple explanations can be misunderstood

Bias Detection: XAI helps, but bias elimination is tough

Future: Al that explains itself in human-like language







GenAI



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What is Generative AI (GenAI)?

A type of Al that can create new content—text, images, music, code, and more—rather than just analyzing data

Like an Al artist, writer, or musician that generates original work based on patterns it has learned.

How Generative AI Works?



Training on Data: Al learns from vast datasets (text, images, code, etc.)

Pattern Recognition: Identifies relationships, structures, and styles

Content Generation: Uses learned patterns to create new content

Refinement & Feedback: Adjusts output based on user input or corrections

Popular Generative AI Models

GPT (Text) - Writes articles, chat responses, and summaries DALL·E (Images) - Creates artwork from text descriptions Codex (Code) - Writes and completes programming code Jukebox (Music) - Generates songs and instrumental music





Challenges & Risks of GenAl

Misinformation - AI can generate fake news & deepfakes

Bias & Ethics - AI can reflect biases in its training data

Creativity Debate - Is AI-generated content real creativity?

Data Privacy - AI models are trained on vast amounts of public data



The Future of Generative Al

More human-like Al assistants

Personalized Al-generated content for individuals

Al that co-creates with humans in art, music, and writing

Ethical guidelines for responsible Al use



Agentic Al



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What is Agentic AI?

Al systems that act autonomously, making decisions, setting goals, and taking actions without constant human intervention Like a self-driving car that plans its route, adapts to traffic, and makes real-time decisions all by itself







Autonomous Decision-Making - sets its own tasks and goals
Planning & Reasoning - doesn't just respond; it strategizes
Adaptability & Learning - improves based on feedback
Memory & Context Awareness - remembers past interactions
Action Execution - takes real-world actions, not just predictions

How Agentic Al Works?

Perception: observes the environment (data, sensors, user input)

Decision-Making: determines the best action based on goals

Action Execution: performs tasks autonomously

Feedback Loop: learns from successes and failures





Traditional vs Agentic Al

Aspect	Traditional AI	Agentic AI
Task Execution	Predefined responses	Self-directed decision-making
Adaptability	Limited, follows rules	Learns and adapts
Autonomy	Requires human input	Acts independently
Memory	Short-term	Long-term memory & context





Challenges & Risks of Agentic Al

Loss of Control - Al taking actions beyond human oversight

Ethical Concerns - Who is responsible for Al decisions?

Unintended Consequences - Al optimizing for unintended goals

Safety & Security - Preventing rogue Al behaviour



Ethics





What is AI Ethics?

Study of moral principles that guide the development and use of Al ensuring Al is fair, safe, and accountable while respecting human rights.

Al is like a powerful car—without ethical "rules of the road," it can cause harm.



Why Does AI Ethics Matter?

Trust - People must trust Al to use it safely Bias & Fairness - Prevent discrimination in Al decisions Privacy - Protect personal data from misuse Accountability - Who is responsible when Al makes mistakes? Safety & Security - Al should not cause harm or be misused



Examples of Ethical AI Challenges

Hiring Bias - Al in job screening favouring certain groups unfairly Deepfakes - Al-generated fake videos spreading misinformation Facial Recognition - Privacy concerns in surveillance and law enforcement Al in Warfare - Autonomous weapons making life-and-death decisions

Solutions for Ethical Al

Fair Al Training - Diverse, unbiased training datasets Explainable AI (XAI) - Making AI decisions understandable Regulations & Guidelines - Laws ensuring ethical Al use (e.g. GDPR, Al Act) Human Oversight - Al should assist, not replace, human decision-making Al for Good - Using Al in healthcare, climate change, and education

The Future of Al Ethics

Stronger AI regulations worldwide More transparency in Al systems Al designed for social good and fairness Better Al-human collaboration with ethical safeguards



