

# Introduction to Data Engineering

## What is Data Engineering:

- the modern data Ecosystem:
- the core concept is that data is a valuable asset
- The ecosystem's purpose is to turn raw data into "actionable insights" for decision making.
- this is achieved through a collaboration of several specialized roles.

## Key players in data Ecosystem:

- data Engineer: build the highway for data *The Architect*
- data analyst: drives on the highway to support on traffic *The Story teller*
- data scientist: use the highway to predict where new cities will be built *Predictor*
- Business Analyst: decides why we need the highway and where it should go for economic growth *Translator*

## Data Engineer tasks:

- collecting source data
- processing data
- storing data
- making data Available to users securely.

## 2) the Data Ecosystem and Languages for Data professionals:

### Data types & Sources:

- Structured: rigid Schema (DB table, spreadsheet)
- semi-structured: Flexib. with tags, json, email
- unstructured: not defn schema (img, video, social media posts)
- Source: relational / non relational DB APIs, web scraping, Data streams (IoT, GPS)

### Data repositories (storage):

- OLTP (Transactional): For day to day operation (e.g. banking transaction) optimized for fast writes.
- OLAP (Analytical): For complex analysis and reporting optimized for fast reads on large Datasets.
- Data Warehouse: central repository for structured cleaned data from multiple sources.
- Data Mart: subset of a data warehouse
- Data lake: stores vast amounts of raw data in its native format (structured - semi structured)

### Data Process & Pipeline:

- ETL: data transformation before loaded (like data warehouse)
- ELT: data loaded first (data lake) by data engineer

## Essential Languages: SQL, python, R, shell scripting.

### Comparison:

Repository Type	Description & Purpose	Key Characteristics	Example
DB	collecting data for input storage search, modification	relational RDBMS: table format, SQL, ACID compliant non relational (NoSQL) Schemaless, scales easily	RDBMS: OLTP (MySQL, PostgreSQL) NoSQL: Large, diverse, fast changing data
D Warehouse	structured	optimized for OLAP Analysis ready	Business Analy (Business intelligence & reporting) google big query, snowflake
Data (mart)			
D Lake	massive amount of raw data (stru/unstru/semi)	stores everything highly scalable	AWS S3 Azure Data Lake
Big data	handle high Volume, Velocity, Variety.	handles high Volume Velocity, Variety	Hadoop HDFS

### No SQL DB types:

- Key-value store (Redis, Dynamo DB)
- Document based (like JSON) (MongoDB, CouchDB)
- column-base (Cassandra HBase)
- Graph-base uses Nodes (data) and edges (Neo4j, Cosmos DB)

RDBMS (SQL) when have structured data on need ACID compliance.

NO SQL you have high Volume, variety, or need high Scalability and Flexibility (No ACID)

### Data Gathering & Import methods:

- SQL: extract data from relational databases (select, Filter, Sort, Group)
- APIs: how applic program access data from an endpoint (like web service or data market place).
- Web Scraping: process autom to download data from websites (text, cat info, images, ... from unstructured web pages).
- Data streams: Handling continuous real time data flows from sources (IoT, sensor, social media ...).

Data exchange: using formal platforms to securely share data.

- where in part data?!
- structured: from relation DB
- unstr/semi: Geo into NoSQL or Data lakes

### Data Wrangling (Data clean)

- process to make raw data usable it involves Exploration, transformation, validation.
- Key tasks: *parallel - map*
- Join: combine columns from diff tables
- Union: ... rows from diff tables
- Normalization: removing biased data
- Denormalization: combine data from multi-ple tables.
- Cleaning tasks: handle missing values, remove duplicate data, remove irrelev data, fix syntax error, standardize data (make text lower case) convert data types.

### Governance & Compliance

- is assure that Data secure, private, trustworthy
- adher laws & regulation