Data Transformation

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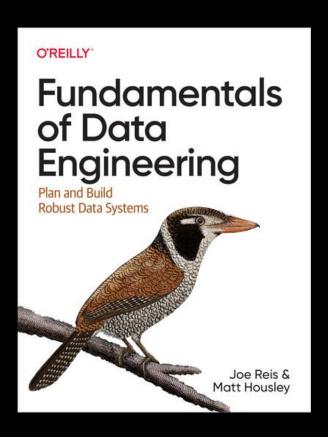
Data Engineering 2024 Fall

Agenda

- Data querying
- Data transformation
- Data wrangling
- Quiz session

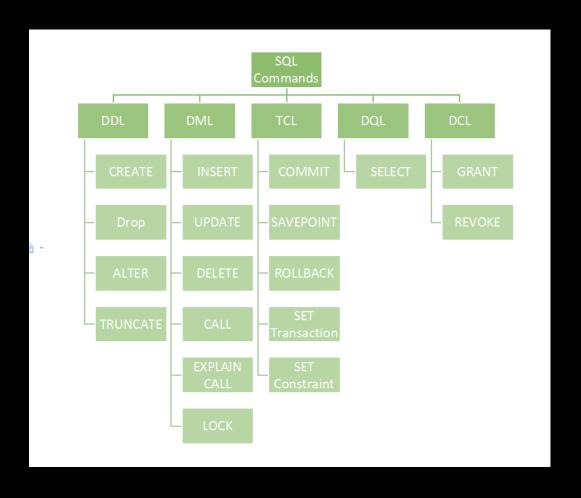
Reading

Chapter VIII



- What is a query?
- Query -> Retrieve and act on data
- CRUD
 - Create-Read-Update-Delete
- Most basic query in SQL (Read)
 - "SELECT [columns] FROM [table];"

- Categorization
 - DDL Data Definition Language (create/modify/delete db objects)
 - DML Data Manipulation Language (CRUD for actual data)
 - DCL Data Control Language (access rights)
 - TCL Transaction Control Language (commits, rollbacks)
 - DQL Data Query Language ("SELECT", debated)



Life of a query



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Data transformation

- Query vs transformation
 - "SQL query"
 - Query → read (SELECT)
 - SELECT customerld, SUM(salesInEur) FROM ... ?
- Generally speaking, transformation:
 - Manipulates, enhances, and saves data
 - For downstream use
 - Increases the value of data

Data transformation

- Transformation (vs query) persists the results
 - Ephemerally
 - Permanently
- Transformation (vs query) generally more complex
 - Query single CTEs, scripts, etc
 - Transformation "bundle" of queries

Data transformation

- Persistence types
 - Table
 - View
 - Materialized view
 - Federated query (Data virtualization)

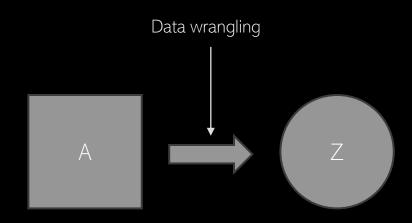
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What is data wrangling?

- Data cleaning
- Data cleansing
- Data munging
- Data preprocessing
- Data preparation
- Data mapping
- Data transformation

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We need data in format Z
We have data in format A

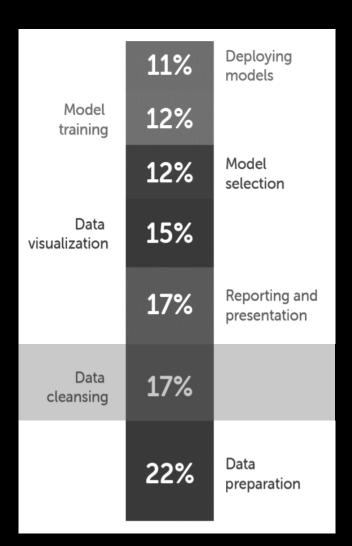
What is data wrangling?

- We want to change
 - Data format
 - Data type
- We want to fix
 - Missing values
 - Duplicates
- We want to
 - Augment
 - Group
 - Aggregate
 - Filter



Why is data wrangling necessary?

• The ML view

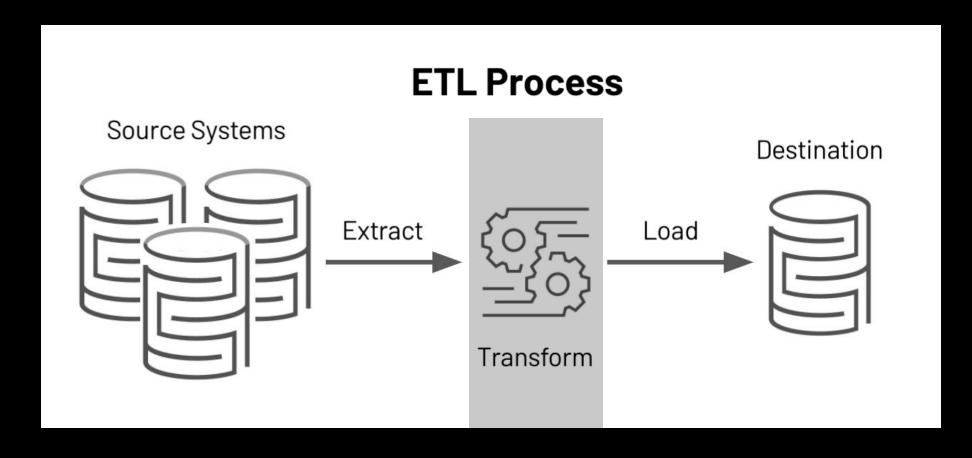


How do data scientists spend their time?

Source: 2021 State of Data Science Report Anaconda $N=2\,030$

Why is data wrangling necessary?

• The BI view



- Validity of values
 - Phone numbers
 - "+372 51234567"
 - "00372 51234567"
 - "512 345 67"
 - "51234567"
 - 51234567
 - Checksum
 - SSN
 - Invoice reference number
 - Credit card

- Consistent values
 - Does zip code + city make sense?
 - Can there be a sales order worth more than 1 million euros?
 - If customer has conflicting emails in different systems, which system is correct?

- Duplicates
 - Do we accept duplicates?
 - Is it possible to set validity of data (updated timestamp)?
 - Keep only one row (try to make the choice idempotent)
 - Can we trust the source system to have unique values?

- Business rules
 - Does a premium customer have the associated premium services?
 - Is this product allowed to have this discount?

- Conforming values
 - If customer has conflicting emails in different systems, which system is correct?
 - Is there a unique code across systems for defining a ... (product, customer, location, ...)
 - Eg product name in ERP vs sales system vs website?

- Missing data
 - What is NULL?
 - Is NULL acceptable?
 - Aggregation over NULL is (usually) correct
 - -1 (or similar) to use for referential integrity
 - In data science/ML:
 - Delete data
 - Impute data
 - Easiest/fastest: median

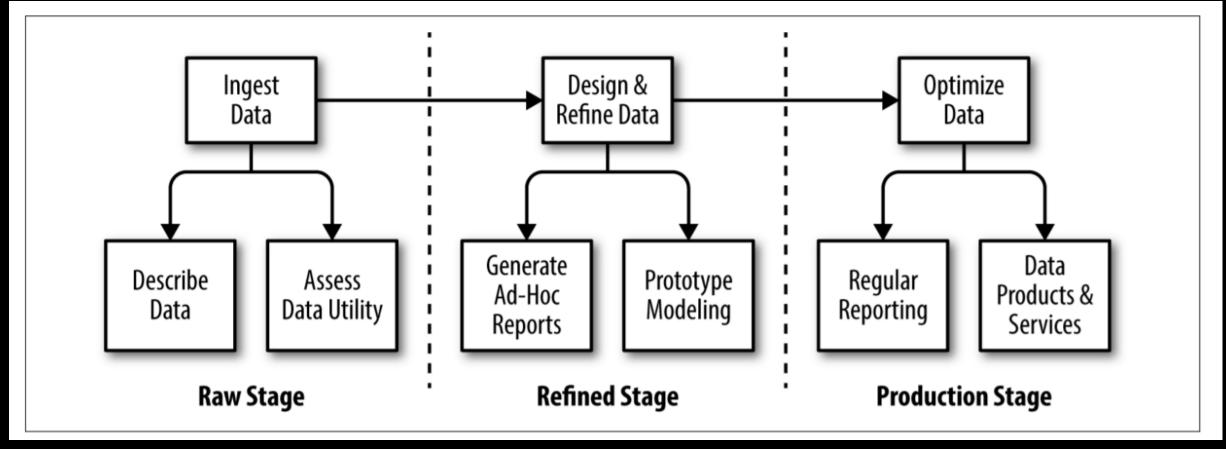
- Wrong data type
 - Schema definition
 - Load in csv without schema everything is a string
 - Inferring schema (eg Spark) can end up with wrong type
 - Unit
 - String vs integer vs decimal type
 - A hundred pieces
 - 100 pieces
 - 100.55 pieces
 - Timestamps
 - UTC vs local
 - UNIX timestamp
 - Avro files date/timestamp is integer. Eg how many days since 1 January 1970 (ISO calendar)

- Wrong data structure
 - Structured data
 - CSV
 - Excel
 - Conventional database
 - Nested data (semistructured data)
 - JSON
 - Parquet
 - Struct, array (in Spark, BigQuery, etc)
 - Unstructured data
 - Text
 - Images
 - Audio/Video

- Aggregations
 - Grouping
 - Correct grouping columns (level of slice/dice)
 - How will the data be used (visualization, reporting, ML)

Levels of data wrangling

• Various definitions. Commonly 3 stages are defined



Source: Trifacta

Levels of data wrangling

Building reliable, performant data pipelines with 🛕 DELTA LAKE Improve Data Quality SXP **Raw Data** Gold **Bronze** Silver **Batch** Filtered, Cleaned, Business-Level Raw Integration Augmented Aggregates **Streaming** "Landing zone" for raw data, no Define structure, enforce schema, Deliver continuously updated, clean schema needed evolve schema as needed data to downstream users and apps

Source: Databricks

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Further reading

- Chapter VIII
 - Joins in distributed systems
 - Update patterns
- https://www.databricks.com/glossary/what-is-data-transformation
- https://www.getdbt.com/blog/what-exactly-is-dbt

