Explaining the 5 types of database and how to choose between them

Tibs (they / he)

18th July 2025, EuroPython 2025

Slides available at https://github.com/Aiven-Labs/the-5-types-of-database



I think there are 5 database shapes

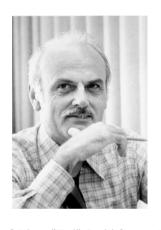
- Relational
- Columnar
- Document
- Key Value
- Graph

1. Relational

A table called **books**

id	title	author
1	This Book	Tibs
2	That Book	Tibs
3	John's Book	John Smith

Edgar F. Codd and relational theory



1970 "A Relational Model of Data for Large Shared Data Banks"

- Just simple enough
- Just abstract enough
- Represent just about anything

Picture of Edgar "Ted" Codd from wikipedia

- *Relation* ≡ table
- Took until the mid-1980s to "win"

Relational tables

books

id	title	author_id
1	This Book	273
2	That Book	273
3	John's Book	301

authors

id	name	
273	Tibs	
301	John Smith	
308	John Smith	

Concept: SQL

"a domain-specific language used to manage data, especially in a relational database management system" – en.wikipedia.org/wiki/SQL

- Originates in the 1970s
- Originally called "SEQUEL" (Structured English Query Language)
- Standardised in the 1980s
- Latest version 2023

How to create those tables

```
CREATE TABLE authors (
 id INT NOT NULL PRIMARY KEY,
 name TEXT NOT NULL,
CREATE TABLE books (
 id INT NOT NULL PRIMARY KEY,
  title TEXT NOT NULL,
 author id INT REFERENCES authors(id)
```

Finding my books...

```
SELECT books.title FROM books
JOIN authors ON authors.id=books.author_id
WHERE authors.name="Tibs";
```

gives the results

This book
That book

Concept: Transactions

- If data is split between multiple tables
 - ▶ then we'll need to change multiple tables "at the same time"
- Transactions let us do this

- 1. START a transaction
- 2. Do all the edits
- 3. Either COMMIT or ROLLBACK

Transaction example

```
START TRANSACTION;
UPDATE authors SET name = "Eric Smith" WHERE id = 301;
UPDATE books SET name = "Eric's book" WHERE id = 3;
COMMIT;
```

gives

books		
id	title	author_id
1	This Book	273
2	That Book	273
3	Eric's Book	301

authors		
id	name	
273	Tibs	
301	Eric Smith	
308	John Smith	

Characteristics of relational databases

- Tables and rows and columns
- Schema design up front
- Transactions (pretty much always)
 - OLTP (online transaction processing)

Relational example 1: PostgreSQL®



"PostgreSQL is a powerful, open source object-relational database system with over 35 years of active development that has earned it a strong reputation for reliability, feature robustness, and performance."

www.postgresql.org/about

More on PostgreSQL

- Rich datatypes
- Stored functions
- Extensibility
- Excellent documentation
- Always a good place to start

Relational example 2: SQLite



"SQLite is a C-language library that implements a small, fast, self-contained, high-reliability, full-featured, SQL database engine. SQLite is the most used database engine in the world." – www.sqlite.org

"Small. Fast. Reliable. Choose any three." - www.sqlite.org

More on SQLite

- A library
- Built into the Python standard library
 - ► It's everywhere
- Single user
- Slightly odd in some ways (schema is optional)
- Use it instead of JSON/YAML/TOML for local storage!

When to use a relational database

- Almost always a good place to start
- If your data fits
 - ► It probably does...
- · Whatever you need to do, some RDB can probably do it
 - and likely fast enough

... but please still stay for the rest of this talk!

2. Columnar

book sales

dt
20250101 12:01
20250101 12:14
20250101 19:27
20250101 20:14
20250101 20:14

id	
1	
1	
1	
3	
2	

This Book This Book This Book Eric's Book That Book	title
This Book Eric's Book	This Book
Eric's Book	This Book
	This Book
That Book	Eric's Book
	That Book

price
5.20
4.50
5.20
4.00
5.20

quantity 1 2 1 1 1 1	
1 2 1 1 1	quantity
2 1 1 1	1
1 1 1	2
1 1	1
1	1
	1

customer_id
1005
923
85
1002
1002

Characteristics of columnar databases

- Essentially an optimisation of the relational idea
- Store data as columns, not rows
- · We know the column datatype, so we can compress column data
 - Giving more efficient data storage
 - Good for data that doesn't change a lot

Compressed columns

book sales

dt	
20250101	12:01
20250101	12:14
20250101	19:27
20250101	20:14

id		title
1		This Book
	-	
3		Eric's Book
7		That Book

5.20
4.50
5.20
4.00
5.20

quantity		
		1
		2
		1

customer_	id
10	05
9	23
	85
10	002

What's fast, what's slow

Fast:

- Adding new rows
- Adding new columns
- Querying a few columns out of many

Slow:

20 / 58

Changing or deleting rows

What data does that suit?

- Log data
- Sensor data
- Time series data in general
- Data stored for historical purposes

Concept: OLAP - Online Analytical Processing.

At the highest level, you can just read these words backward:

- **Processing**: some source data is processed ...
- Analytical: to produce some analytical reports and insights ...
- Online: in real-time.

clickhouse.com/docs/concepts/olap

In contrast to OLTP (Online Transaction Processing)

Columnar example: ClickHouse®

|||| ClickHouse

"ClickHouse® is an open-source column-oriented database management system that allows generating analytical data reports in real-time." – github.com/ClickHouse/ClickHouse

"ClickHouse is the fastest and most resource efficient real-time data warehouse and open-source database." – clickhouse.com

More about ClickHouse

- Queries are still SQL
 - With some extras and useful utility functions
- Records don't have to have a unique primary key
 - Although having one can help
- "Full fledged" transactions aren't supported
 - ▶ Do we really need them for OLTP?

Create book sales table

```
CREATE TABLE book sales (
  dt DateTime,
  id BIGINT,
  title String,
  price Decimal(8,2),
  quantity Int,
  customer id BIGINT,
) ENGINE = MergeTree()
PARTITION BY toYYMM(dt)
ORDER BY (title, dt)
```

Find the 10 top sellers

```
id, title, sum(quantity)
AS
        total_quantity
FROM book_sales
GROUP BY id
ORDER BY total_quantity DESC
LIMIT 10
```

When to use a columnar database

- When you want to query on columns not rows
- When you have lots of columns
- When you have a lot of data
 - Which you don't want to alter

3. Document

```
"title": "This Book",
"author": "Tibs",
"isbn": null,
"publisher": "self-published",
"tags": ["nonFiction, humour"]
"summary": "It's just very good",
"chapterContent": [<chapter 1>, <chapter 2>, ...]
```

Document database concepts

- Documents are essentially JSON
- An *index* is a collection of documents
- When you search
 - you get back all data that matched
 - with a relevance score for how well it matched

Characteristics of document databases

- Relatively unstructured data
- But want indexing
- And rich querying
- OLTP Store and query rather than update
- No transactions

Document example: OpenSearch®



OpenSearch is an open-source, enterprise-grade search and observability suite that brings order to unstructured data at scale

opensearch.org

More about OpenSearch

- Technology origins in document processing, indexing and searching for large bodies of text
- Backed by Apache Lucene
- Queries are written in JSON
- Schema design up front is optional
 - but sometimes advised
- Data visualisation tools built in

Queries: Query DSL

```
query body = {
  "query": {
    "bool": {
       must": {"match": {"author": "Tibs"}},
       must not": {"match": {"title": "That Book"}},
resp = client.search(index=INDEX NAME, body=query body)
```

Queries: Lucene syntax

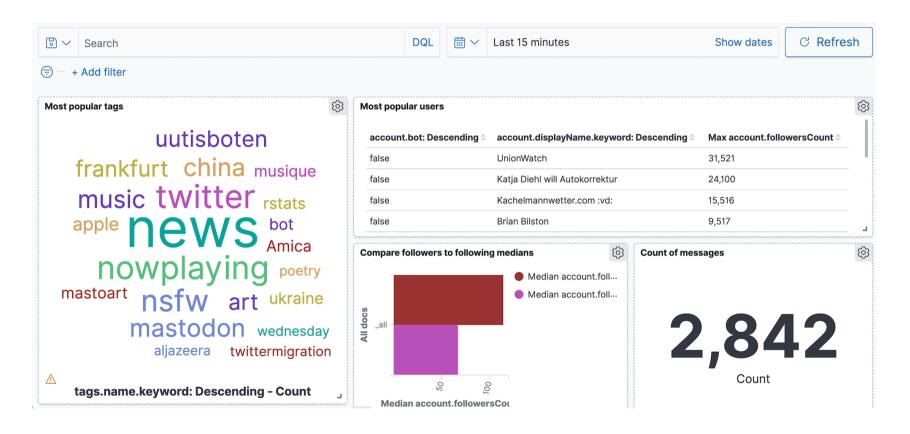
```
client = OpenSearch(SERVICE_URI, use_ssl=True)

client.search({
   index: 'recipes',
   q: 'author:Tibs AND title: (-That Book)'
})
```

Queries: SQL

untested code!

A dashboard about mastodon messages



When to use a document database

- Fast, scalable full text search
- Storage of indexable JSON documents
- OpenSearch: sophisticated analytics visualisation

4. Key Value

A picture of a dictionary \bigcirc

Characteristics of key value databases

- Fast
- Simple
- Sophisticated value data types
- Think like a Python dictionary!

Key Value example: ValkeyTM



"Valkey is an open source (BSD) high-performance key/value datastore that supports a variety of workloads such as caching, message queues, and can act as a primary database." – https://valkey.io

Datatypes

Key: a binary sequence

Value:

- Strings
- Lists
- Sets and Sorted sets
- Hashes
- Streams

- Geospatial indexes
- Bitmaps
- Bitfields
- Hyperloglog
- Bloom filter
- ...plus extensions

Queries

It's actually rather lovely...

SET current:greeting "Hello" EX 60

LSET booklist 0 "This Book"

HGET "book: This Book" author

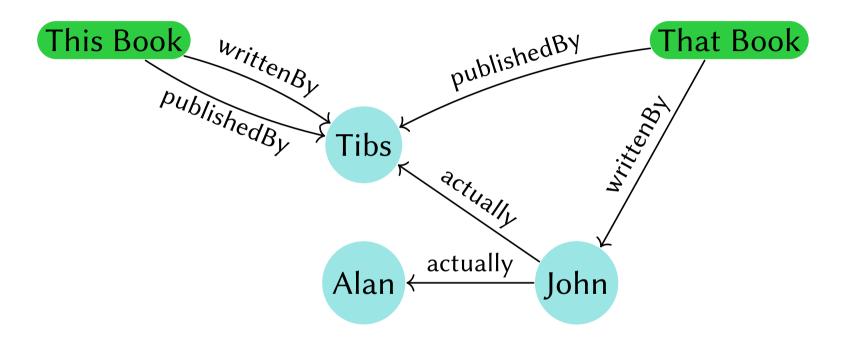
More about Valkey

- In-memory, but persistent to disk
- Use cases include:
 - Data storage and retrieval
 - Caching, leveraging the value expiry support
 - ► Pub/Sub messaging (SUBSCRIBE, UNSUBSCRIBE, PUBLISH)
 - Streams (append-only log) for message queues (XADD, XREAD)

When to use a key value database

- When your data fits the "key" -> "value" idea
- Caching (for instance, URL -> page results)
- Valkey:
 - when you want your data to expire
 - pub/sub messaging
 - message queues
 - for its datatypes

5. Graph



not an XY data graph 🙂

Characteristics of graph databases

Nodes, relationships and properties

- or objects, references and attributes
- or nodes, edges and values

Schemas might be implicit, gradual or designed up-front

Nodes

Nodes have

- a type
- properties
- are linked by relationships

Relationships

Relationships

- are between nodes
- are 1:1 or 1:many or many:1
- depending on design (I have opinions):
 - may have properties

Graph example: Neo4J®



"the world's most-loved graph database" - neo4j.com

"The programmer works with a flexible network structure of nodes and relationships rather than static tables—yet enjoys all the benefits of enterprise-quality database." – github.com/neo4j/neo4j

More about Neo4J nodes

Nodes

- have *labels*
- have key:value properties
- are indexed

More about Neo4J relationships

Relationships

- have a name
- must have a type, a start node and an end node
- must have a direction
- can have properties

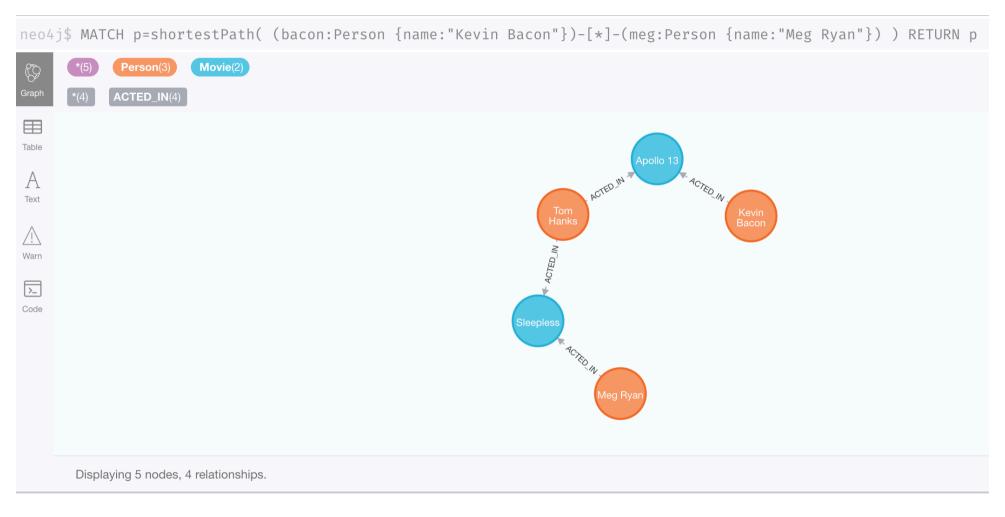
Queries: Neo4J has Cypher

```
CREATE (p:Book
    {name:'This Book'})-[r:IS_WRITTEN_BY]->
         (p:Person {name:'Tibs'}
)
```

From Neo4J's own examples:

```
MATCH p=shortestPath(
   (bacon:Person {name:"Kevin Bacon"})-[*]-
        (meg:Person {name:"Meg Ryan"})
) RETURN p
```

Graph



When to use a graph database

- You have a knowledge graph shaped puzzle
- Neo4J: You want to build structures as you learn them
- Neo4J: You want to leverage existing techniques & solutions

Things just about all the shapes give you

- Transactions (not really OpenSearch)
- JSON support
- Vector embeddings (Valkey with a module; SQLite has an extension)
- Extensibility

What we've looked at

Five different kinds (shapes) of database

Relational	PostgreSQL®	Use for just about anything
	SQLite	Use in your programs, use locally
Columnar	ClickHouse®	Use for analytics, historical data
Document	OpenSearch®	Use for text corpuses, semi-structured data, indexing
Key Value	Valkey™	Use for caching, pub/sub, simple queues
Graph	Neo4J®	Use for graph/network data

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