

# Miniazon Architecture

High-Performance Data Structures & Algorithms

---

An engineering deep-dive into hybrid storage systems.



# Executive Summary: The Hybrid Engine

## The Problem

Standard SQL databases are optimized for storage, not high-frequency interactive speed. Searching via **LIKE %query%** requires scanning every row ( $O(N)$ ), which causes latency at scale.

## The Solution

Miniazon implements a **Hybrid Architecture**:

- **Performance Layer:** Custom Data Structures in RAM for sub-millisecond operations.
- **Storage Layer:** SQLite for persistent data reliability.



# 1. The Search Engine: Inverted Index


## Why not SQL?

Scanning a database row-by-row is slow. Instead, we use an **Inverted Index**.

Think of the index at the back of a textbook. Instead of reading every page to find a word, you look at the word to find the page numbers.

**Concept:** Hash Table + Sets

Maps a single keyword to a set of multiple Product IDs.

 Abstract geometric diagram of an inverted index data structure



# Implementation: search\_index.py

```
class SearchIndex:
    def __init__(self):
        # Hash table mapping words to product sets
        self.index = {}

    def search(self, query):
        # ... logic to split query ...
        # Intersect sets (AND logic)
        results &= self.index[word]
```

## Complexity Analysis

- **Add Product:**  $O(k)$   
Depends only on word count ( $k$ ), not database size.
- **Search:**  $O(m)$   
Depends only on query length ( $m$ ).

*"I avoided the Slow Scan ( $O(N)$ ). I went straight to the answer."*



## 2. The Shopping Cart: Nested Hash Table

600 × 400

### The "Holy Grail" of Speed

If we used a simple list, updating a cart would require looping through items to find the right one. That is inefficient.

We use a **Nested Hash Table** (Dictionary of Dictionaries):

- **Outer Key:** User ID
- **Inner Key:** Product ID
- **Value:** Quantity



# Implementation: shopping\_cart.py

## Constant Time Access

This structure allows us to access any specific item for any user instantly, without looping.

**Complexity:**  $O(1)$

No matter if I have 10 users or 10 million users, adding an item takes exactly 1 step.

```
self.carts = {  
    "erospai_1337": { # Outer Key (User)  
        1: 5          # Inner Key -> Value  
    }  
}  
  
def add_item(self, user_id, prod_id, qty):  
    # Direct access, no loops  
    self.carts[user_id][prod_id] = qty
```



# 3. Recommendations: Product Graph

## Modeling Relationships

How do we know that a "Laptop" is related to a "Mouse"?  
Storing this in a grid (Matrix) is wasteful because most products aren't related.

We use an **Adjacency List**. Each product holds a list of its "friends" (neighbors).

**Type:** Undirected Graph.



600 × 400



# Implementation: recommendations.py

```
self.graph = {  
    1: [2, 3, 4] # Laptop -> [Mouse, Keyboard, Monitor]  
}  
  
def add_relationship(self, a, b):  
    # Bidirectional logic  
    self.graph[a].append(b)  
    self.graph[b].append(a)
```

## Efficient Traversal

**Complexity:**  $O(1)$  (Effective)

Writing a name in an address book takes 1 step. Finding a product's friends is instantaneous because we just retrieve the list directly from the dictionary key.



# System Architecture: RAM vs Disk



600 × 400

## Search & Cart Flow

1. **User Action:** User types "Mouse" or adds item.
2. **RAM Processing:** Python uses the Inverted Index or Hash Map to find IDs or update Quantities. (Fast)
3. **SQL Storage:** Only accessed when we need rich details (Image, Price) for the final display. (Slow)



# Recommendation Workflow



## View Product

User visits "Laptop" page.



## Graph Query

Python checks neighbor list in RAM.



## ID Retrieval

Returns IDs: [Mouse, Monitor].



## Render

Frontend displays "Also bought".



# Algorithmic Efficiency

Comparison of Miniazon's engineered approach vs. a naive database approach.

Feature	DSA Implemented	Complexity	Naive Approach (SQL)
Search	Inverted Index	$O(k)$	$O(N)$ (Full Table Scan)
Cart	Nested Hash Map	$O(1)$	$O(N)$ (List Iteration)
Recs	Adjacency List	$O(1)$	$O(N^2)$ (Complex Joins)



# Engineering, Not Just Coding.

Miniazon proves that intelligent data structure selection solves performance problems that raw database power cannot.

 [kwisdomk/Miniazon](#)