Data Warehousing and Data Mining

— L2: OLAP —

Part I

- Why and What are Data Warehouses?
  - Transaction Processing vs. Analytical Processing
  - Databases v

Data is meaningless nalysis!

Example in a finance department

- Daily transaction tasks
  - E.g., account receivable, account payable, payroll, etc.

Columns:

Description
G/L Account
Branch
cost center
G/L account name
Tax code
Total
…

Example/2

- Weekly…monthly…yearly analytical tasks
  - E.g., Finance reports

Why OLAP Servers?

- Different workload:
  - OLTP (on-line transaction processing)
    - Major task of traditional relational DBMS
    - Day-to-day operations: purchasing, inventory, banking, manufacturing, payroll, registration, accounting, etc.
  - OLAP (on-line
    - Major task of
    - Data analysis and decision making
- Queries hard/infeasible for OLTP, e.g.,
  - Which **week** we have the largest sales?
  - Does the sales of **dairy products** increase over time?
  - Generate a **spread sheet** of total sales by state and by year.
- Difficult to represent these queries by using SQL ← Why?

OLTP vs. OLAP

	OLTP	OLAP
users	clerk, IT professional	knowledge worker
function	day to day operations	decision support
DB design	application-oriented	subject-oriented
data	current, up-to-date det iso	historical, zed, multidimensional d, consolidated
usage	repetitive	
access	read/write index/hash on prim. key	
unit of work	short, simple transaction	complex query
# records accessed	tens	millions
#users	thousands	hundreds
DB size	100MB-GB	100GB-TB
metric	transaction throughput	query throughput, response

Data Analysis Problems

- The same data found in many different systems
  - Example: customer data across different departments
  - The same ifferently
- Heterogeneo
  - Relational DBMS, OnLi ction Processing (OLTP)
  - Unstructured data in files (e.g., MS Excel) and documents (e.g., MS Word)

Data Analysis Problems (Cont'd)

- Data is suited for operational systems
  - Accounting,billing,etc.
  - Do not support analysis across business functions
- Data quality i
  - Missing data, imprecise ferent use of systems
- Data are "volatile"
  - Data deleted in operational systems (6months)
  - Data change over time – no historical information

Solution: Data Warehouse

- Defined in many different ways, but not rigorously.
  - A decision support database that is maintained separately from the organization's operational database
  - Support inform g a solid platform of consolidated, h
- "A data warehouse is a subject-oriented, integrated, time-variant, and nonvolatile collection of data in support of management's decision-making process."—W. H. Inmon
- Data warehousing:
  - The process of constructing and using data warehouses

Data Warehouse—Subject-Oriented

- Organized around major subjects, such as customer, product, sales.
- Focusing on the modeling and analysis of data for decision makers, not on daily operations or transaction processing.
- Provide a simple and concise view around particular subject issues by excluding data that are not useful in the decision support process.

Data Warehouse—Integrated

- Constructed by integrating multiple, heterogeneous data sources
  - relational databases, flat files, on-line transaction records
- Data cleaning and data integration techniques are applied.
  - Ensure consistency in naming conventions, encoding structures, attribute measures, etc. among different data sources
    - E.g., Hotel price: currency, tax, breakfast covered, etc.
  - When data is moved to the warehouse, it is converted.

Data Warehouse—Time Variant

- The time horizon for the data warehouse is significantly longer than that of operational systems.
  - Operational database: current value data.
  - Data warehouse: provides information from a historical perspective (e.g., past 5-10 years)
- Every key structure in the data warehouse
  - Contains an element of time, explicitly or implicitly
  - But the key of operational data may or may not contain "time element".

Data Warehouse—Non-Volatile

1. A physically separate store of data transformed from the operational environment.
2. Operational update of data does not occur in the data warehouse en

   - Does not require transact sing, recovery, and concurrency control s
   - Requires only two operations in data accessing:
     - *initial loading of data* and *access of data*.

Data Warehouse Architecture

- Extract data from operational data sources
  - clean, transform
- Bulk load/refresh
  - warehouse is of
- OLAP-server prov multidimensional view
- Multidimensional-olap (Essbase, oracle express)
- Relational-olap (Redbrick, Informix, Sybase, SQL server)

Data Warehouse Architecture

All subjects, integrated

Advanced analysis

Function-oriented systems

Subject-oriented systems

Why Separate Data Warehouse?

- High performance for both systems
  - DBMS— tuned for OLTP: access methods, indexing, concurrency control, recovery
  - Warehouse—t P queries, multidimensio
- Different functions and different da
  - missing data: Decision support rical data which operational DBs do not typically maintain
  - data consolidation: DS requires consolidation (aggregation, summarization) of data from heterogeneous sources
  - data quality: different sources typically use inconsistent data representations, codes and formats which have to be reconciled

Comparisons

	Databases	Data Warehouses
Purpose	Many purposes; Flexible and general	One purpose: Data analysis
Conceptual Model	ER	ltidimensional
Logical Model	(Normalized) Relational	ormalized) Star schema / cube/cuboids
Physical Model	Relational Tables	ROLAP: Relational tables MOLAP: Multidimensional arrays
Query Language	SQL (hard for analytical queries)	MDX (easier for analytical queries)
Query Processing	B+-tree/hash indexes, Multiple join optimization, Materialized	Bitmap/Join indexes, Star join, Materialized data cube

Comparisons/2

Source：https://www.zhihu.com/question/20623931

Comparisons/2

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- The Multidimensional Model

The Multidimensional Model

- A data warehouse is based on a multidimensional data model which views data in the form of a data cube, which is a multidimensional generalization of 2D spread sheet.
- Key concepts:
  - **Facts**: the s
    - Typically tr er types includes snapshots, etc.
    - Measures: numbers that can d
    - Dimensions: context of the measure
  - Hierarchies:
    - Provide contexts of different granularities (aka. grains)
- Goals for dimensional modeling:
  - Surround facts with as much relevant context (dimensions) as possible ← Why?

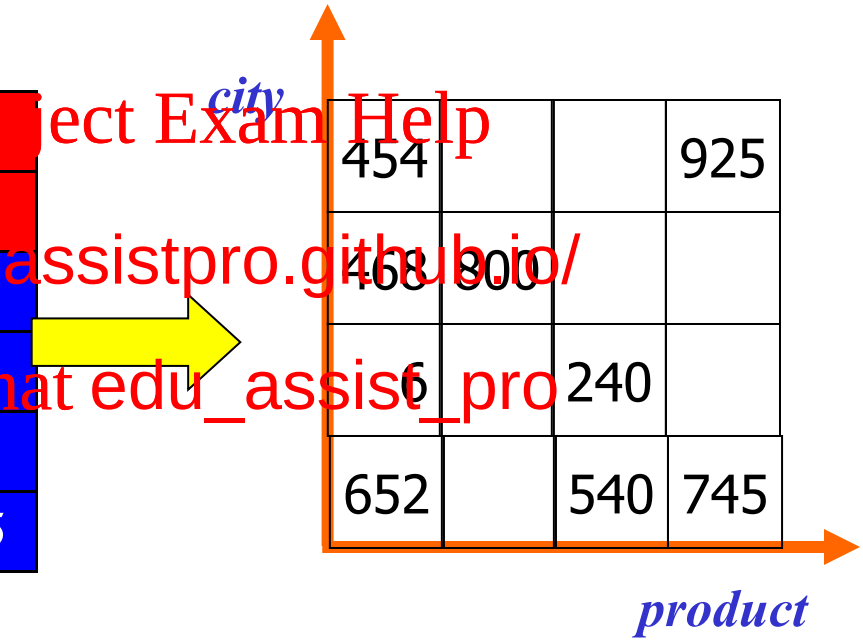
Supermarket Example

- Subject: analyze total sales and profits
- Fact: Each Sales **Transaction**
  - Measure: Dollars_Sold, Amount_Sold, Cost
  - Calculated M
- Dimensions:
  - Store
  - Product
  - Time

Visualizing the Cubes

- A valid instance of the model is a data cube

total Sales		product			
		p1	p2		
city	NY	$454	-		
	LA	$468	$800		-
	SD	$296	-	$240	-
	SF	$652	-	$540	$745



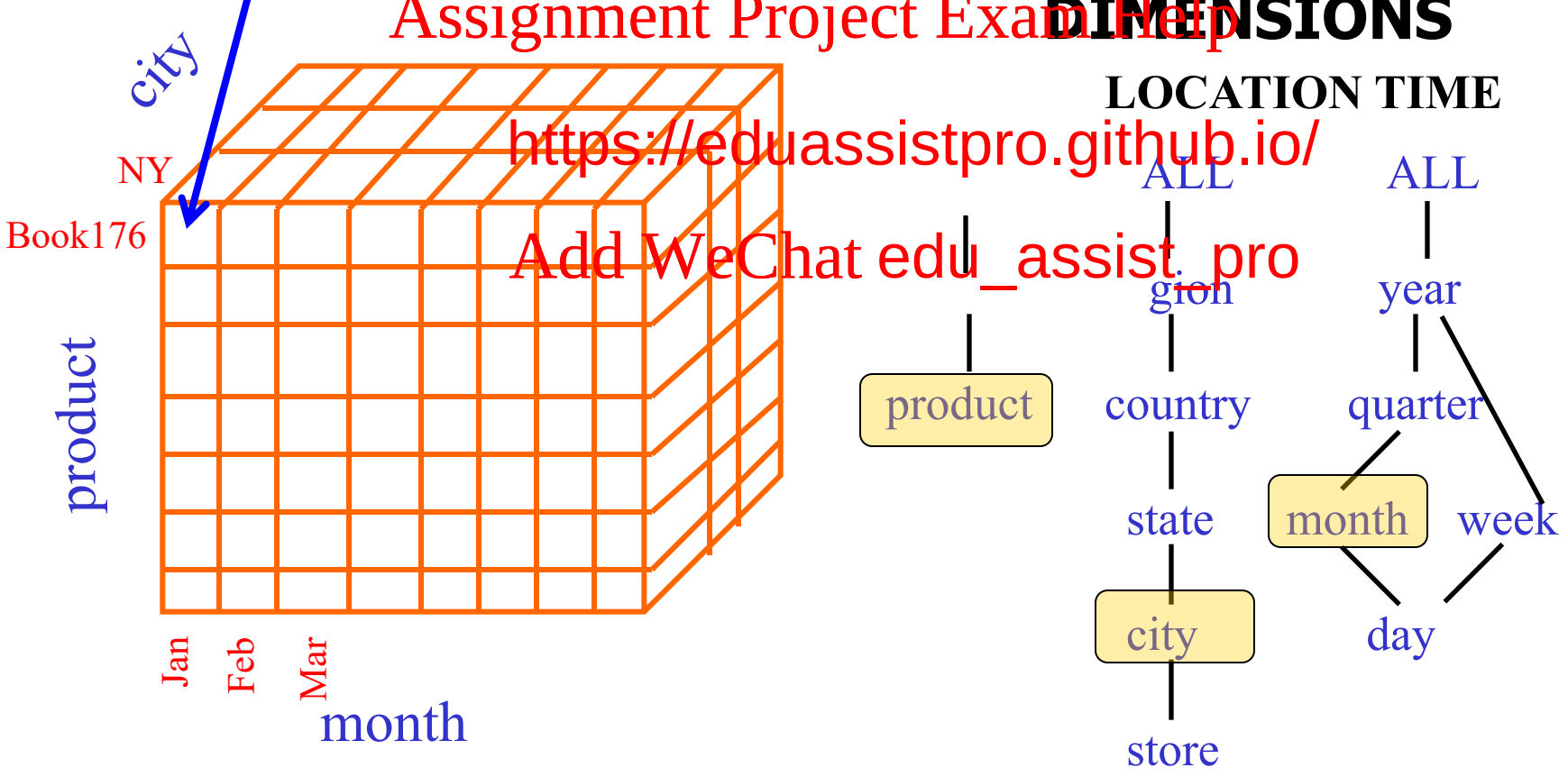
Concepts: cell, fact (=non-empty cell), measure, dimensions

Q: How to generalize it to 3D?

3D Cube and Hierarchies

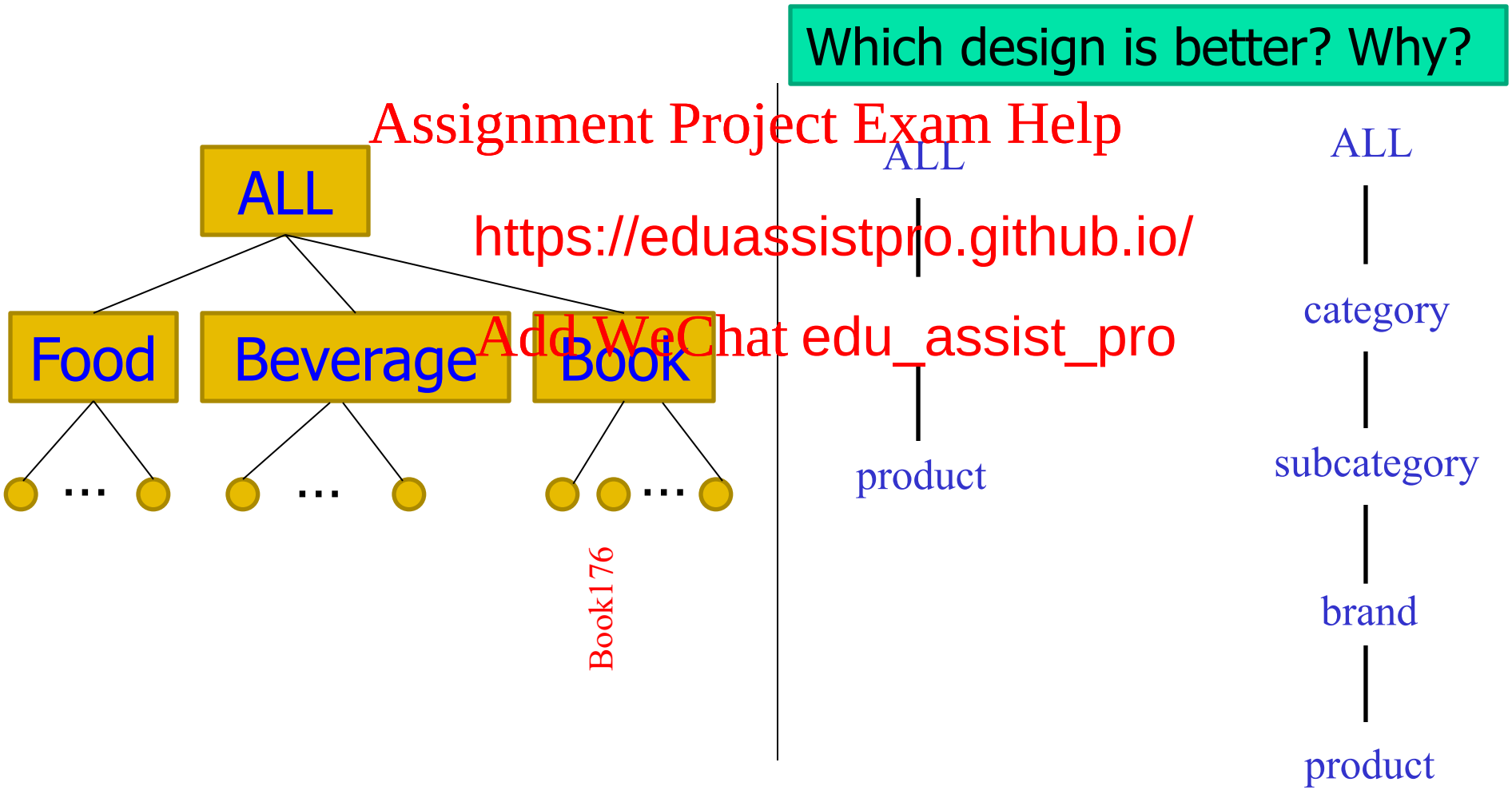
Concepts: hierarchy (a tree of dimension values), level

Sales of book176 in NY in Jan can be found in this cell



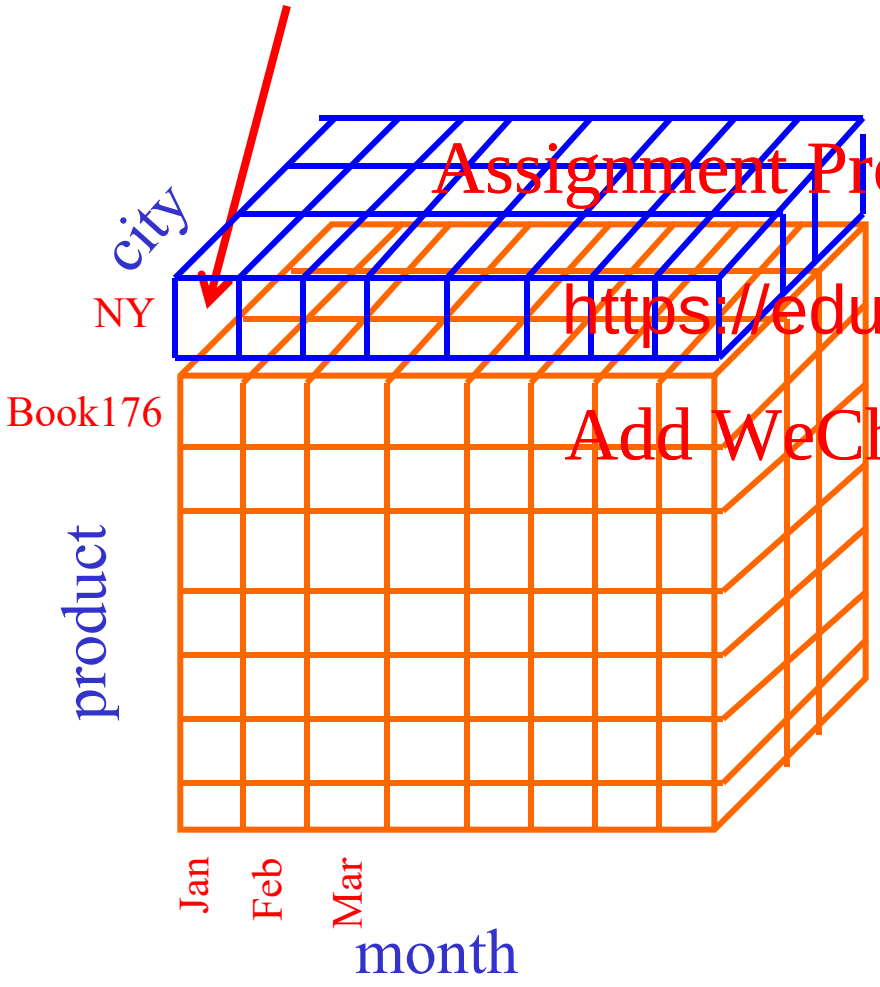
Hierarchies

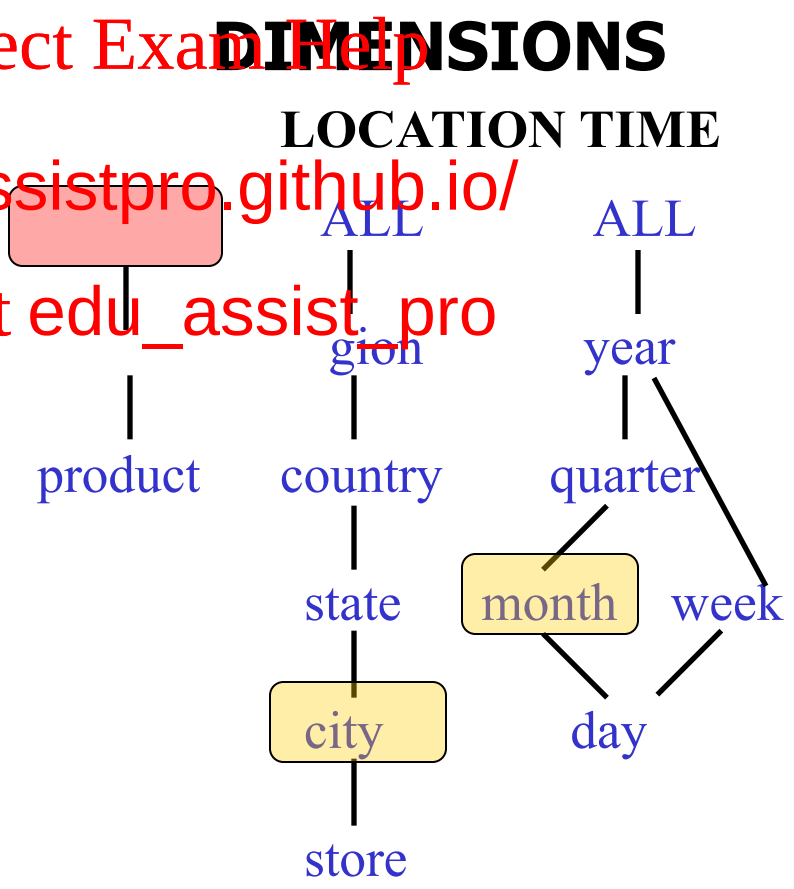
Concepts: hierarchy (a tree of dimension values), level



The (city, moth) Cuboid

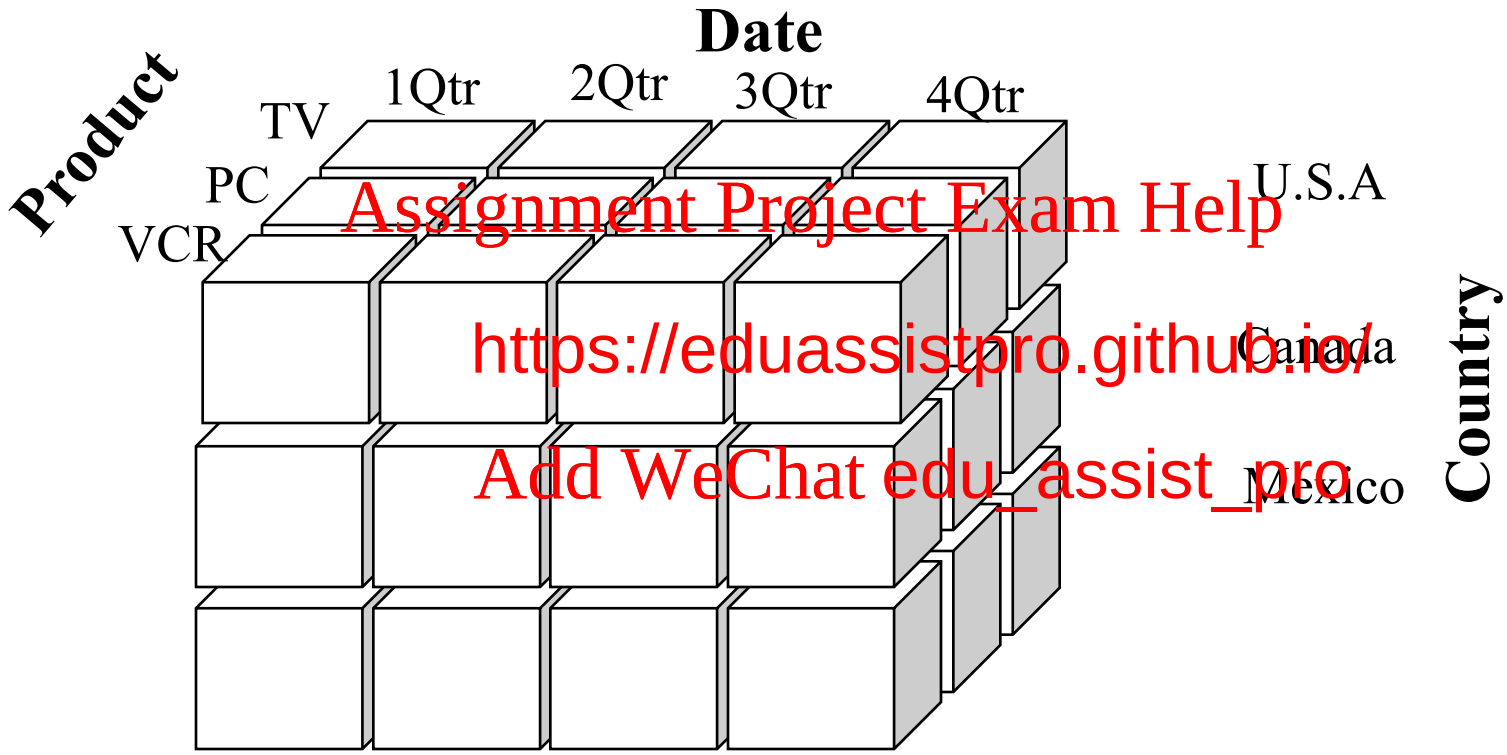
Sales of ALL_PROD in NY in Jan





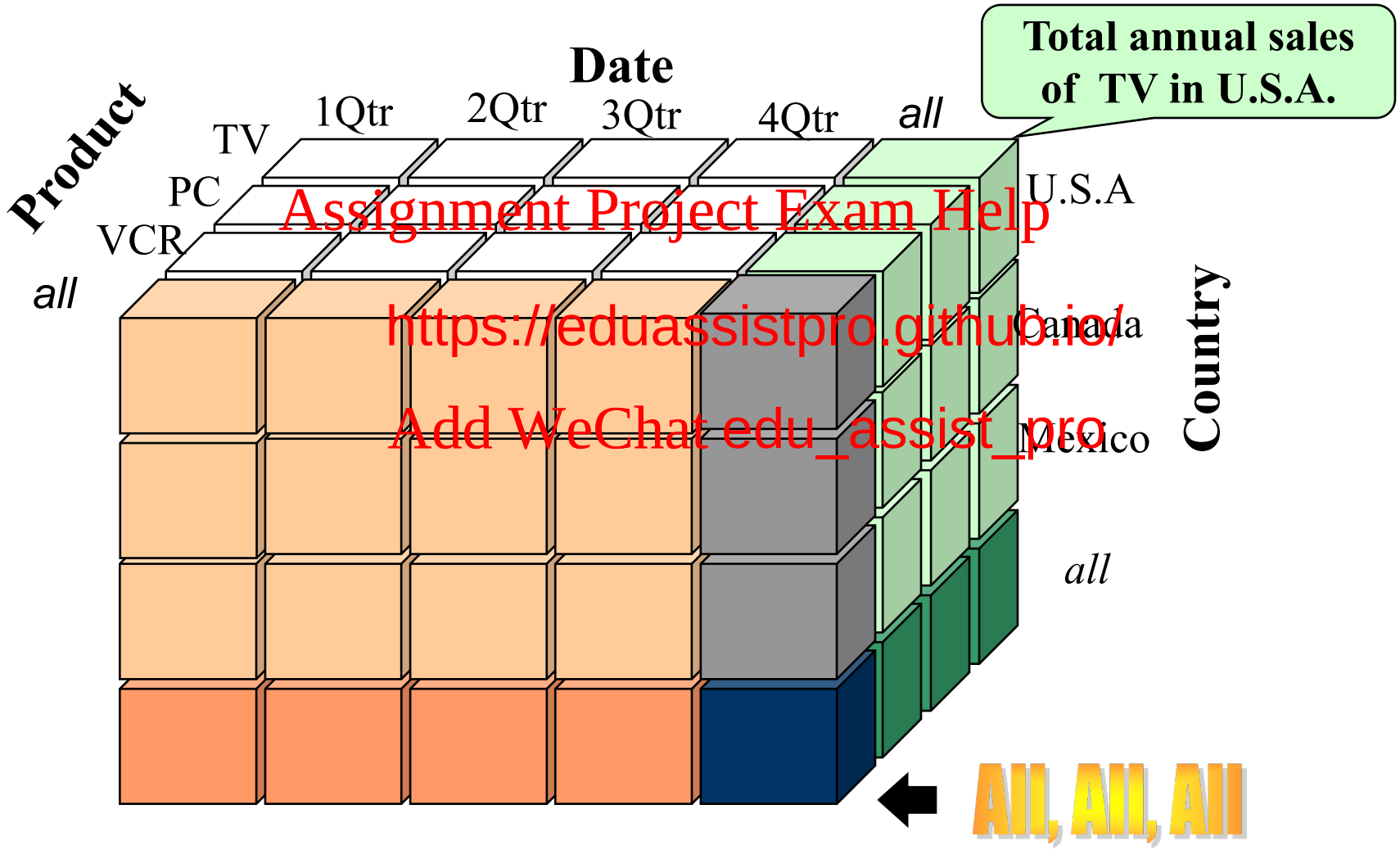
All the Cuboids

Assume: no other non-ALL levels on all dimensions.

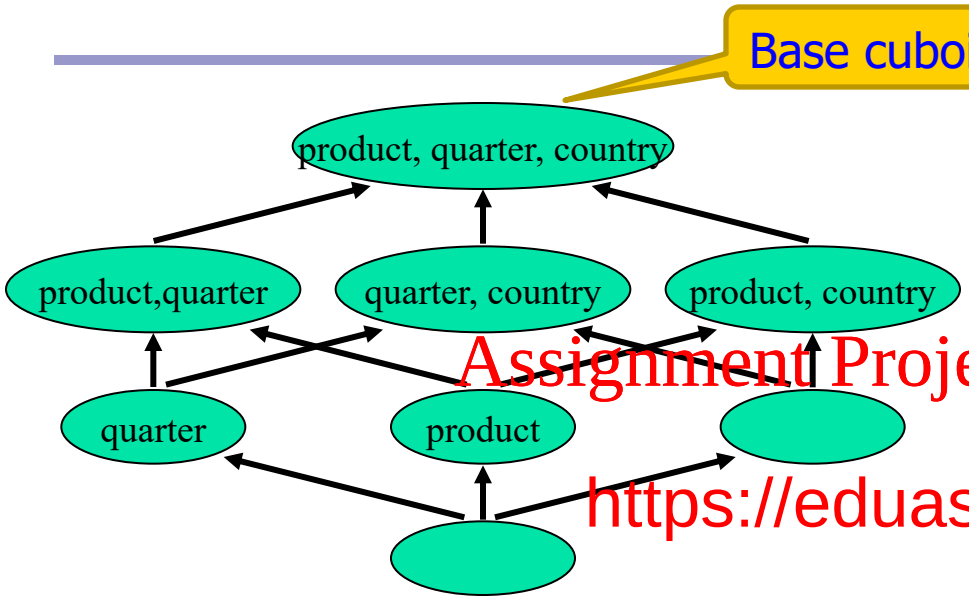


All the Cuboids /2

Assume: no other non-ALL levels on all dimensions.



Lattice of the cuboids



- n-dim cube can be represented as ( ), where D_i is the set of allowed values on the i-th dimen
  - if $D_i = L_i$ (a particular level), then Di = all descendant dimension values of L_i.
  - ALL can be omitted and hence reduces the effective dimensionality
- A complete cube of d-dimensions consists of $\prod_{i=1}^{d}(n_i + 1)$ cuboids, where n_i is the number of levels (excluding ALL) on i-th dimension.
  - They collectively form a lattice.

Properties of Operations

- All operations are closed under the multidimensional model
  - i.e., both input and output of an operation is a cube
- So that they

Q: What's the analogy in the Relational Model?