

Data Warehousing

Read chapter 13 of Riguzzi et al Sistemi Informativi

Slides derived from those by Hector Garcia-Molina

What is a Warehouse?

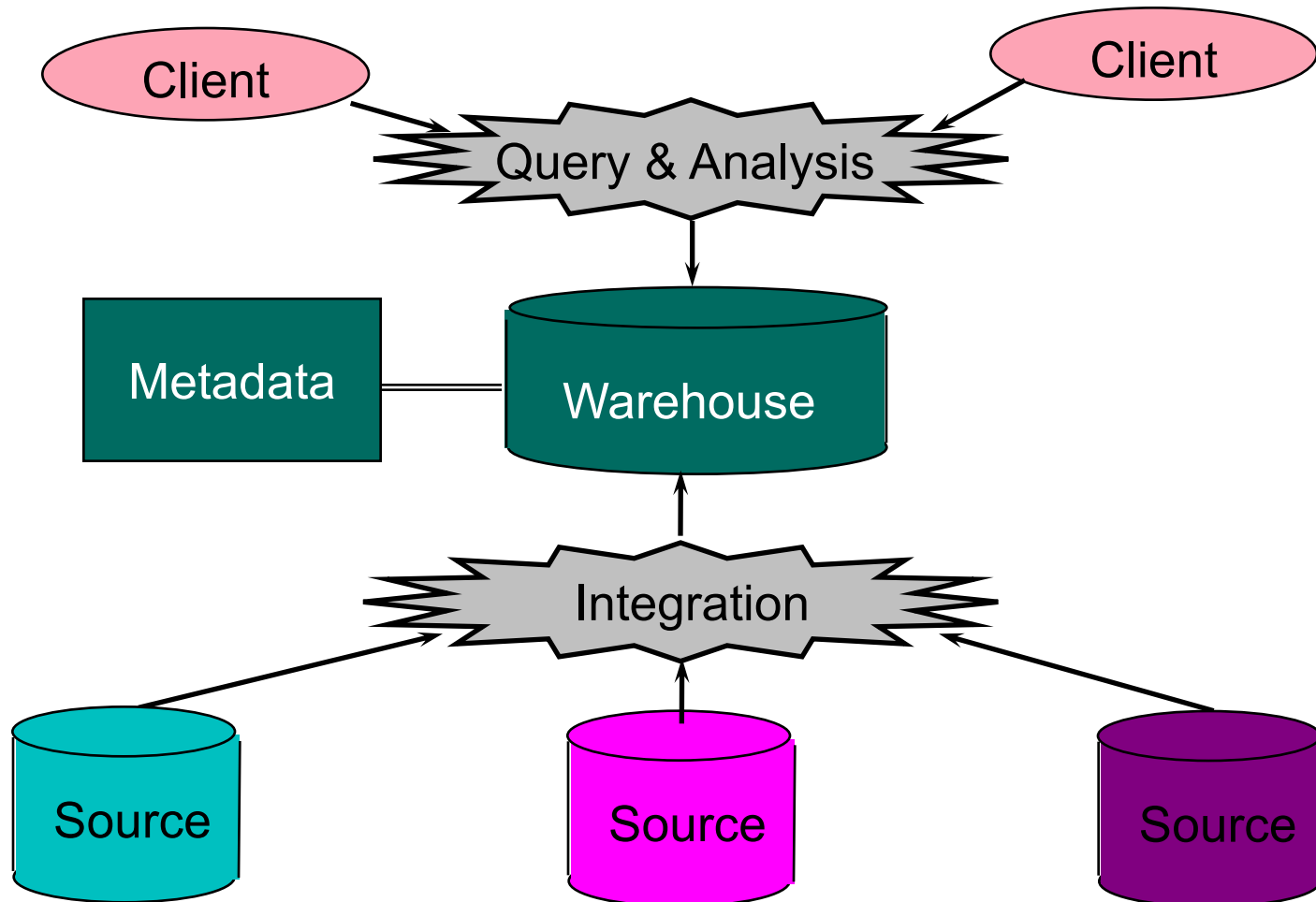
- Collection of diverse data
 - subject oriented
 - aimed at executive, decision maker
 - often a copy of operational data
 - with value-added data (e.g., summaries, history)
 - integrated
 - time-varying
 - non-volatile



What is a Warehouse?

- Collection of tools
 - gathering data
 - cleansing, integrating, ...
 - querying, reporting, analysis
 - data mining
 - monitoring, administering warehouse

Warehouse Architecture



Motivating Examples

- Forecasting
- Comparing performance of units
- Monitoring, detecting fraud
- Visualization

OLTP vs. OLAP

- OLTP: On Line Transaction Processing
 - Describes processing at operational sites
- OLAP: On Line Analytical Processing
 - Describes processing at warehouse

OLTP vs. OLAP

OLTP

- Mostly updates
- Many small transactions
- Mb-Gb of data
- Raw data
- Clerical users
- Up-to-date data
- Consistency, recoverability critical

OLAP

- Mostly reads
- Queries long, complex
- Gb-Tb of data
- Summarized, consolidated data
- Decision-makers, analysts as users

Data Marts

- Smaller warehouses
- Spans a part of an organization
 - e.g., marketing (customers, products, sales)
- Do not require enterprise-wide consensus
 - but long term integration problems?

Warehouse Models & Operators

- Data Models
 - relations
 - stars & snowflakes
 - cubes
- Operators
 - slice & dice
 - roll-up, drill down
 - pivoting
 - other

Star

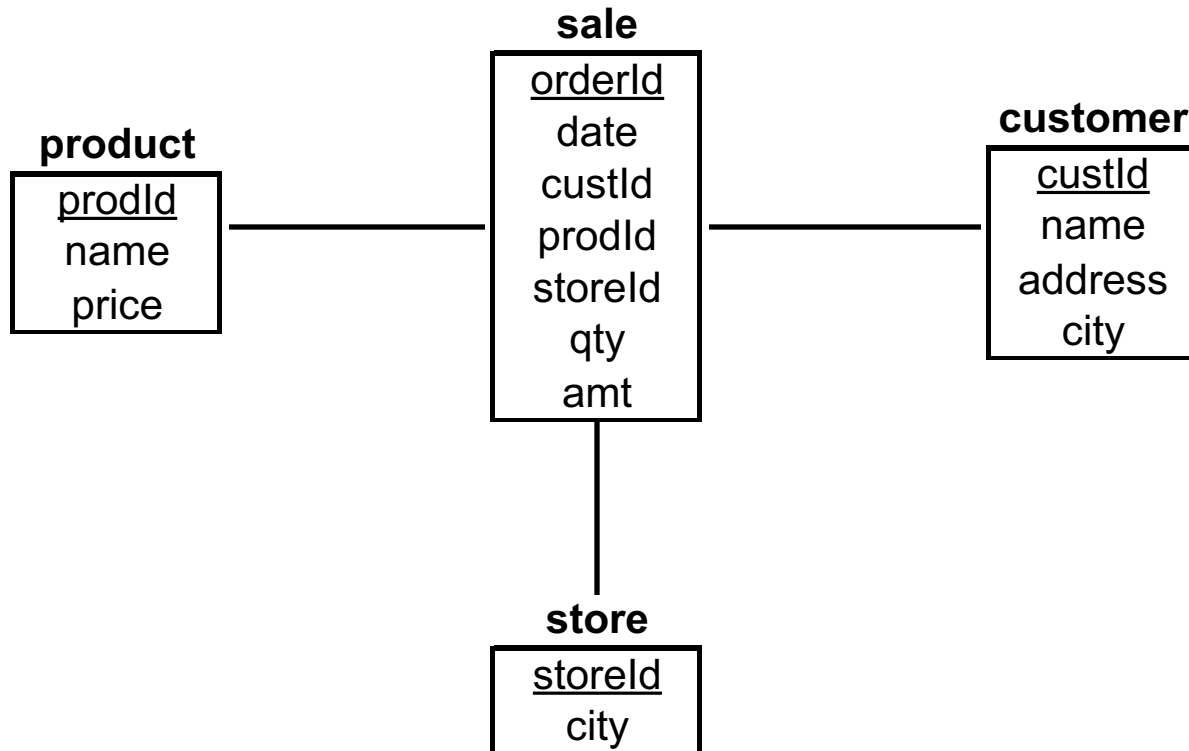
product	<u>prodId</u>	name	price
	p1	bolt	10
	p2	nut	5

store	<u>storeId</u>	city
	c1	nyc
	c2	sfo
	c3	la

sale	<u>oderId</u>	date	custId	<u>prodId</u>	<u>storeId</u>	qty	amt
	o100	1/7/97	53	p1	c1	1	12
	o102	2/7/97	53	p2	c1	2	11
	105	3/8/97	111	p1	c3	5	50

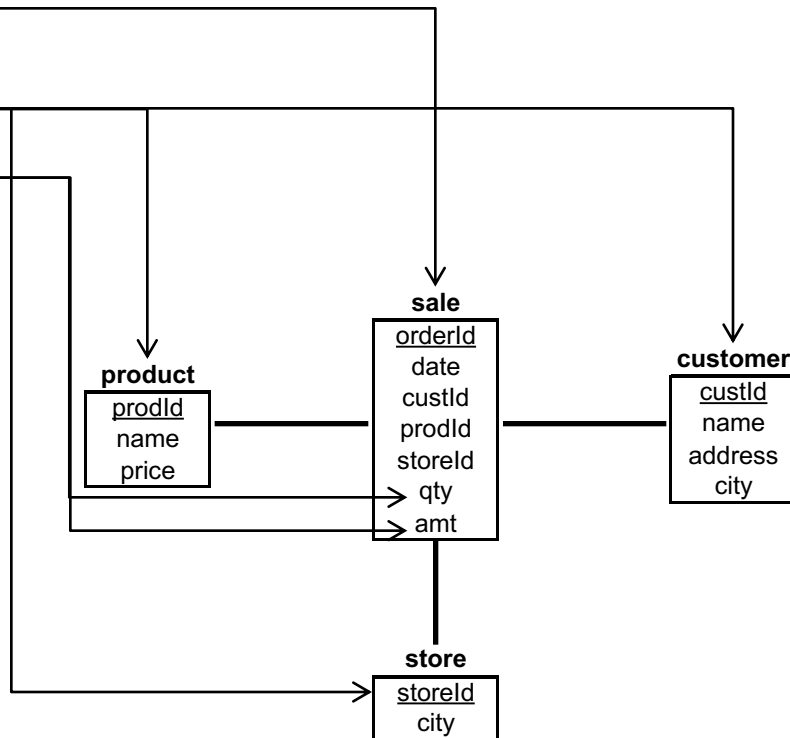
customer	<u>custId</u>	name	address	city
	53	joe	10 main	sfo
	81	fred	12 main	sfo
	111	sally	80 willow	la

Star Schema

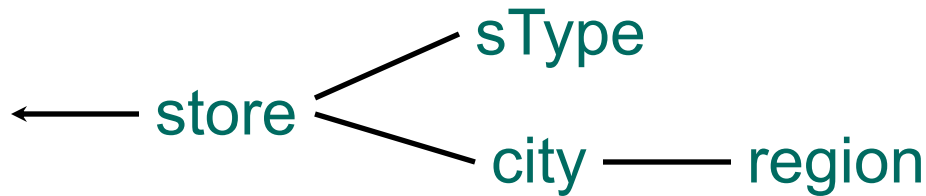


Terms

- Fact table
- Dimension tables
- Measures



Dimension Hierarchies



store	<u>storeld</u>	cityld	tld	mgr
	s5	sfo	t1	joe
	s7	sfo	t2	fred
	s9	la	t1	nancy

sType	<u>tld</u>	size	location
	t1	small	downtown
	t2	large	suburbs

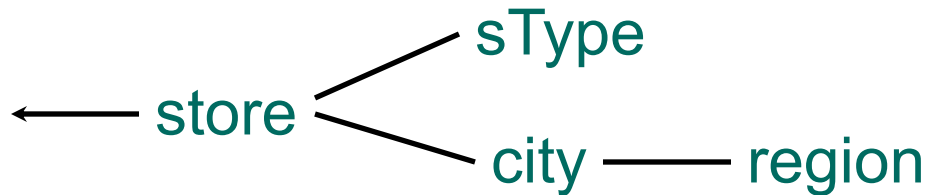
city	<u>cityld</u>	pop	regld
	sfo	1M	north
	la	5M	south

→ snowflake schema

region	<u>regld</u>	name
	north	cold region
	south	warm region

Snowflake Schema

Sometimes not normalized: not in third normal form



store	<u>storeld</u>	cityld	tld	mgr
	s5	sfo	t1	joe
	s7	sfo	t2	fred
	s9	la	t1	nancy

sType	<u>tld</u>	size	location
	t1	small	downtown
	t2	large	suburbs

city	<u>cityld</u>	pop	regld	name
	sfo	1M	north	cold region
	la	5M	south	warm region

Cube

Fact table view:

sale	prodId	storeId	amt
	p1	c1	12
	p2	c1	11
	p1	c3	50
	p2	c2	8



Multi-dimensional cube:

	c1	c2	c3
p1	12		50
p2	11	8	

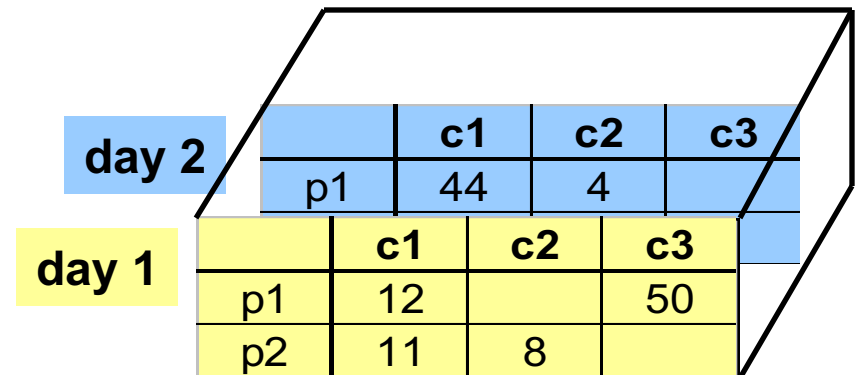
dimensions = 2

3-D Cube

Fact table view:

sale	prodId	storeId	date	amt
	p1	c1	1	12
	p2	c1	1	11
	p1	c3	1	50
	p2	c2	1	8
	p1	c1	2	44
	p1	c2	2	4

Multi-dimensional cube:

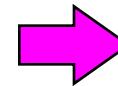


dimensions = 3

Aggregates

- Add up amounts for day 1
- In SQL: `SELECT sum(amt) FROM SALE WHERE date = 1`

sale	prodlid	storeid	date	amt
	p1	c1	1	12
	p2	c1	1	11
	p1	c3	1	50
	p2	c2	1	8
	p1	c1	2	44
	p1	c2	2	4



81

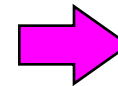
—— rollup —→

←—— drill-down

Aggregates

- Add up amounts for days 1 and 2
- In SQL: `SELECT sum(amt) FROM SALE WHERE date >= 1 AND date <=2`

sale	prodlid	storeid	date	amt
	p1	c1	1	12
	p2	c1	1	11
	p1	c3	2	40
	p2	c2	2	8
	p1	c1	3	44
	p1	c2	3	4



71

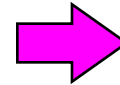
—— rollup ——→

←—— drill-down

Aggregates

- Add up amounts by day
- In SQL: `SELECT date, sum(amt) FROM SALE GROUP BY date`

sale	prodlid	storeld	date	amt
	p1	c1	1	12
	p2	c1	1	11
	p1	c3	1	50
	p2	c2	1	8
	p1	c1	2	44
	p1	c2	2	4



ans	date	sum
	1	81
	2	48

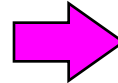
— rollup —→

← drill-down —

Another Example

- Add up amounts by day, product
- In SQL: `SELECT prodId, date, sum(amt)`
`FROM SALE GROUP BY date, prodId`

sale	prodId	storeId	date	amt
	p1	c1	1	12
	p2	c1	1	11
	p1	c3	1	50
	p2	c2	1	8
	p1	c1	2	44
	p1	c2	2	4



sale	prodId	date	amt
	p1	1	62
	p2	1	19
	p1	2	48

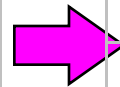
— rollup —→

← drill-down —

Another Example

- Add up amounts by month
- In SQL: `SELECT month, prodId, storeId, sum(amt)`
`FROM SALE JOIN DATE GROUP BY month, prodId, storeId`

sale	prodId	storeId	date	amt
	p1	c1	1	12
	p1	c1	1	11
	p2	c2	1	50
	p2	c2	1	8
	p1	c3	2	44
	p1	c3	2	4



sale	prodId	storeId	month	amt
	p1	c1	sep	23
	p2	c2	sep	58
	p1	c3	oct	48

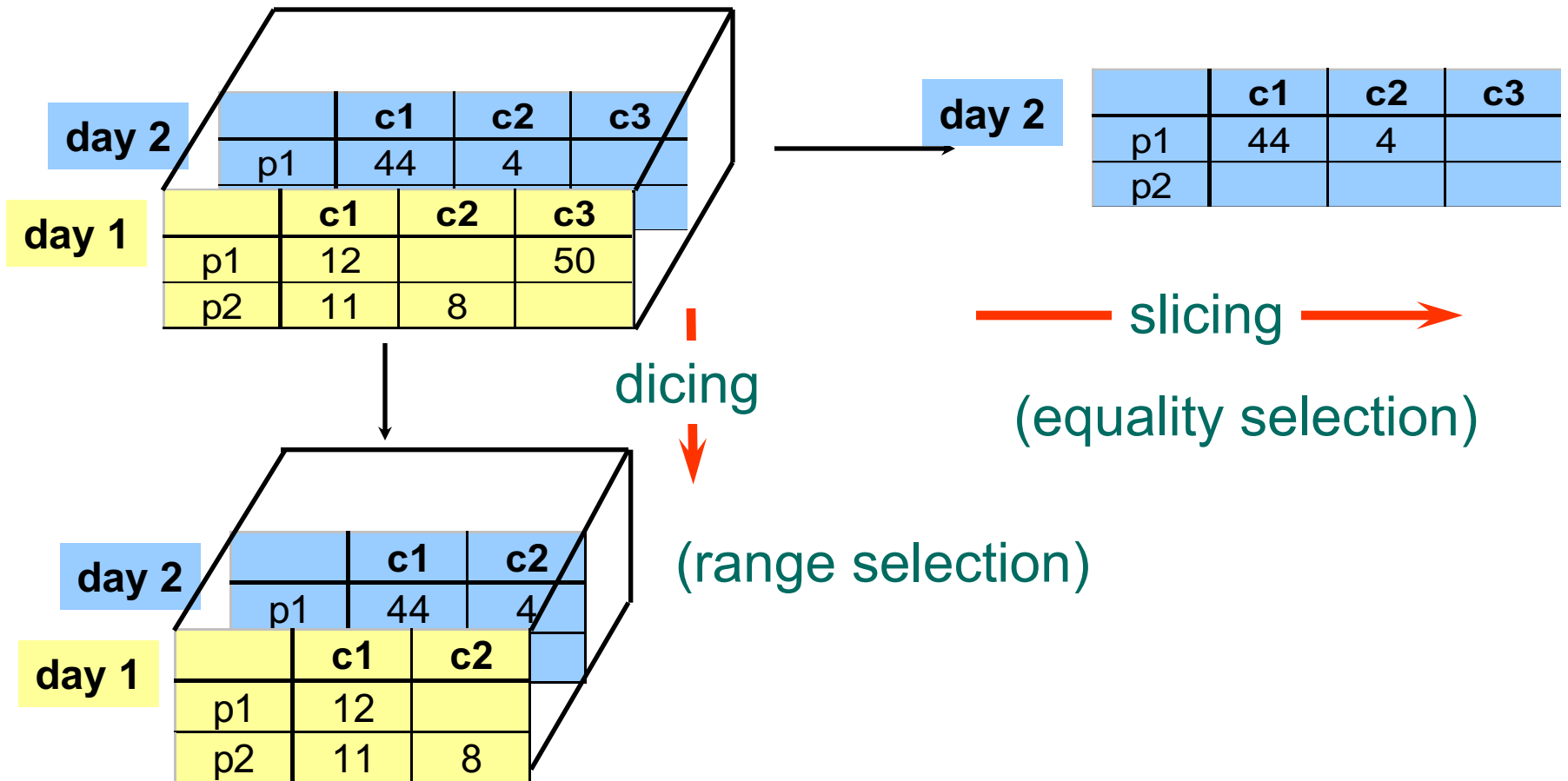
—— rollup ———→

←—— drill-down ——

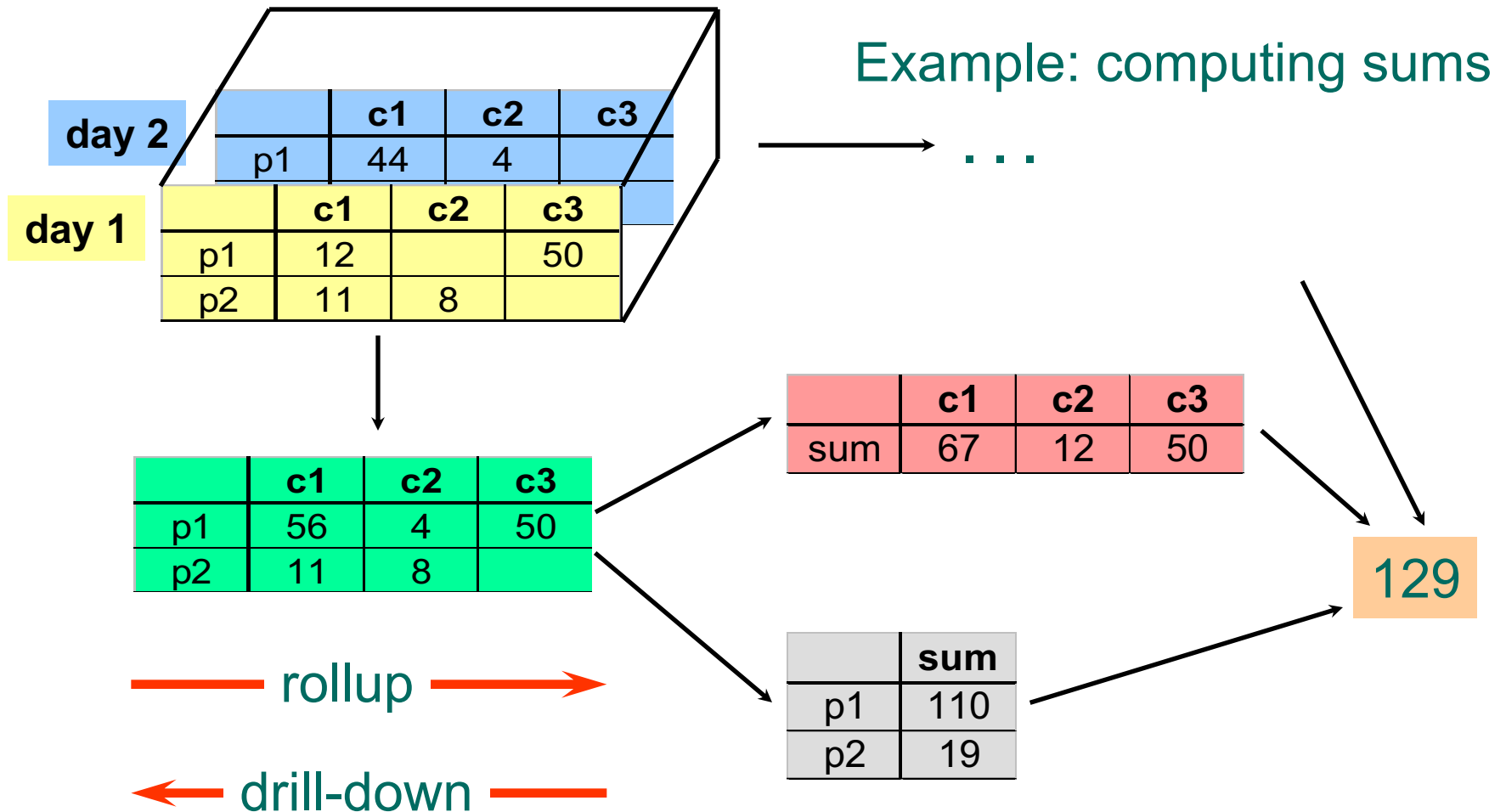
Aggregates

- Operators: sum, count, max, min, median, ave
- “Having” clause
- Using dimension hierarchy
 - average by region (within store)
 - maximum by month (within date)

Operations on the Cube



Cube Aggregation



Aggregation Using Hierarchies

day 2		c1	c2	c3
	p1	44	4	
day 1		c1	c2	c3
	p1	12		50
	p2	11	8	

rollup



	region A	region B
p1	56	54
p2	11	8

drill-down



customer
|
region
|
country

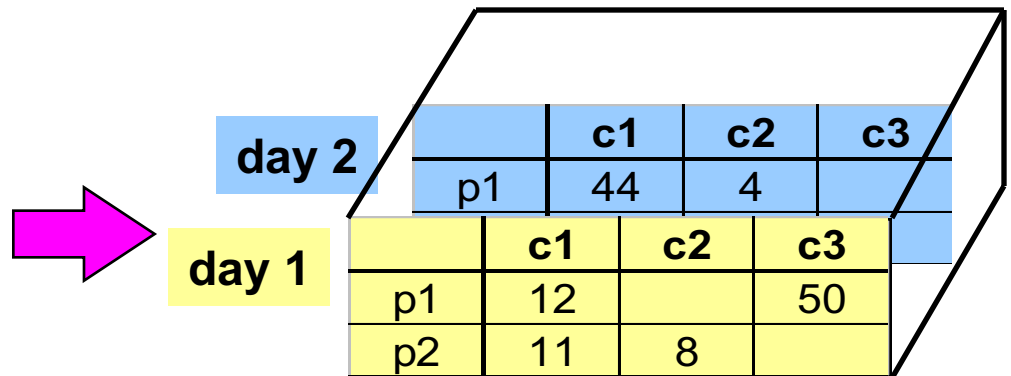
(customer c1 in Region A;
customers c2, c3 in Region B)

Pivoting

Fact table view:

sale	prodId	storeId	date	amt
	p1	c1	1	12
	p2	c1	1	11
	p1	c3	1	50
	p2	c2	1	8
	p1	c1	2	44
	p1	c2	2	4

Multi-dimensional cube:



Query & Analysis Tools

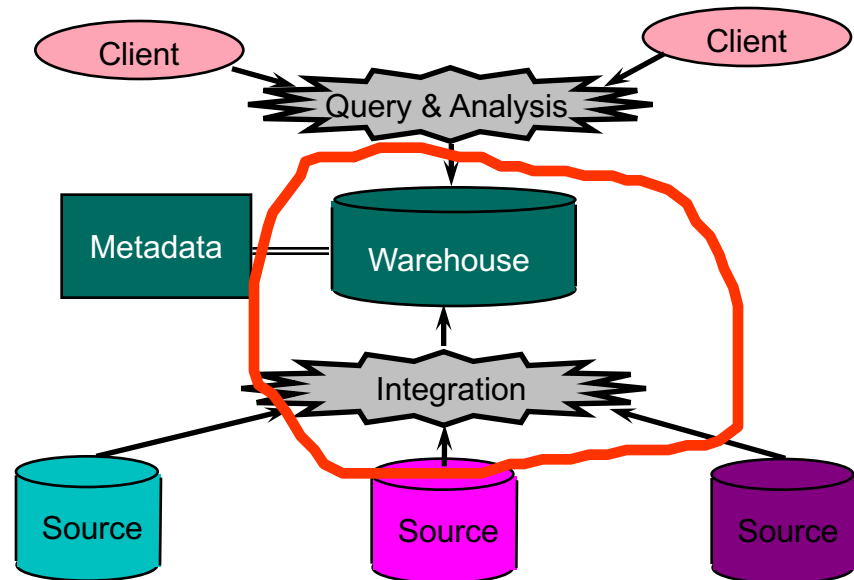
- Query Building
- Report Writers (comparisons, growth, graphs,...)
- Spreadsheet Systems
- Web Interfaces
- Data Mining

Implementing a Warehouse

- *Integrating*: Loading, cleansing,...
- *Processing*: Query processing, indexing, ...
- *Managing*: Metadata, tools

Integration

- Data Loading
- Data Cleaning



Loading

- Source Types: relational, flat files, IMS, VSAM, WWW, news-wire, ...
- Incremental vs. Refresh

customer	<u>id</u>	name	address	city
	53	joe	10 main	sfo
	81	fred	12 main	sfo
	111	sally	80 willow	la



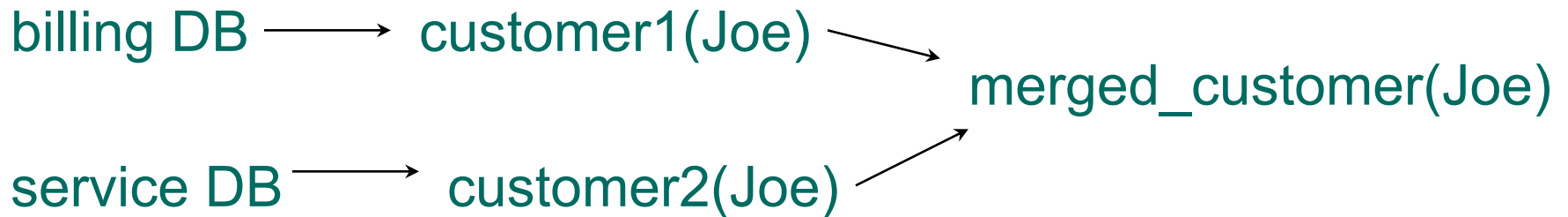
Loading Techniques

- Periodic snapshots
 - Off-line vs. on-line
 - Frequency
 - At night, 1x a week/month
 - Parallel/Partitioned load
- Polling (queries to source)
- Database triggers
- Log shipping
- Data shipping (replication service)
- Transaction shipping
- Application level monitoring

➔ Advantages & Disadvantages!!

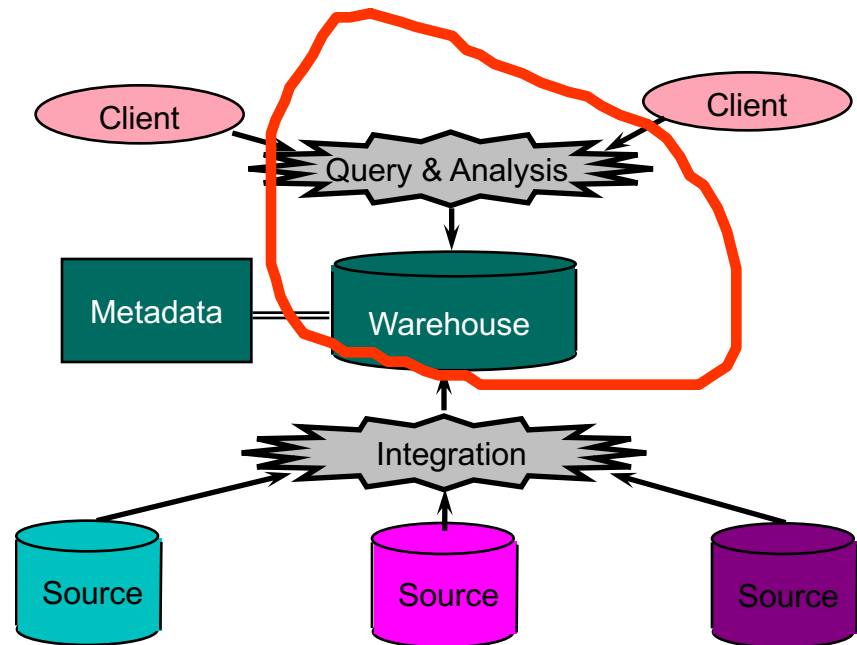
Data Cleaning

- Migration (e.g., yen $\Rightarrow$ dollars)
- Scrubbing: use domain-specific knowledge (e.g., social security numbers)
- Fusion (e.g., customer merging)



Processing

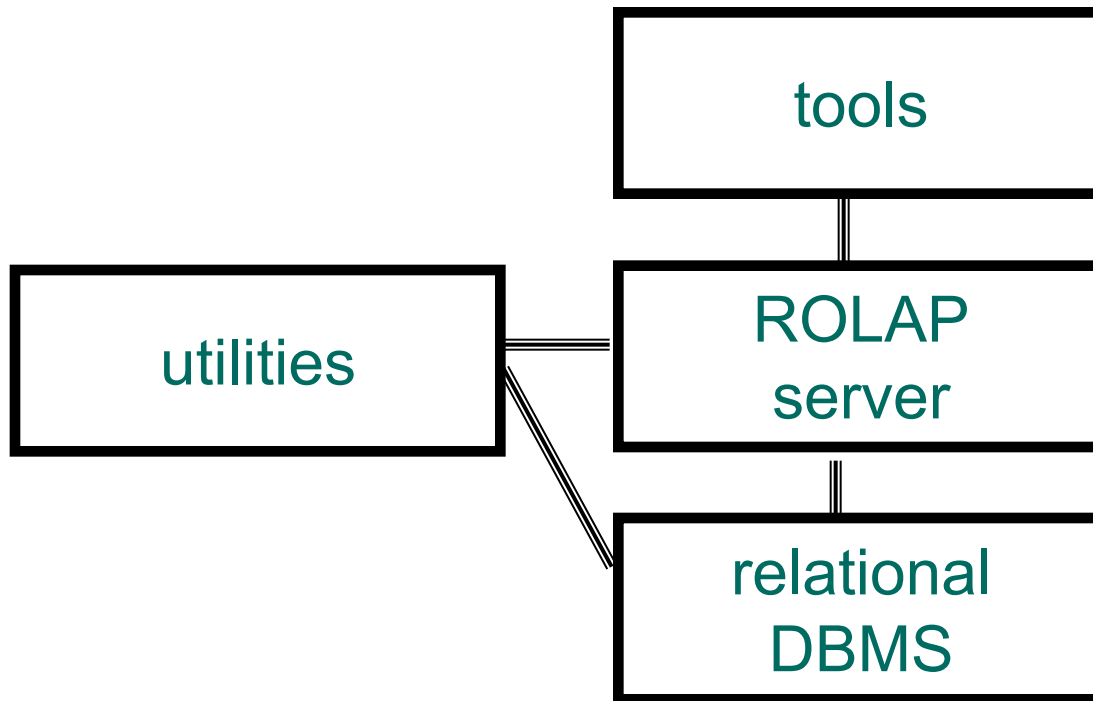
- ROLAP: Relational On-Line Analytical Processing
- MOLAP: Multi-Dimensional On-Line Analytical Processing
- Index Structures
- What to Materialize?
- Algorithms



ROLAP Server

- Relational OLAP Server

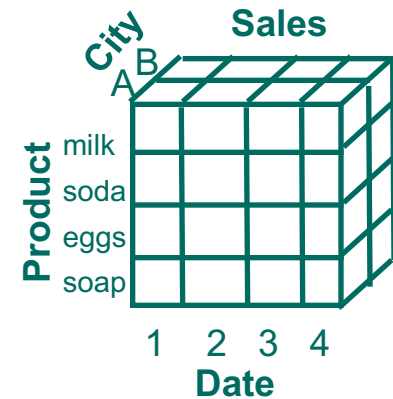
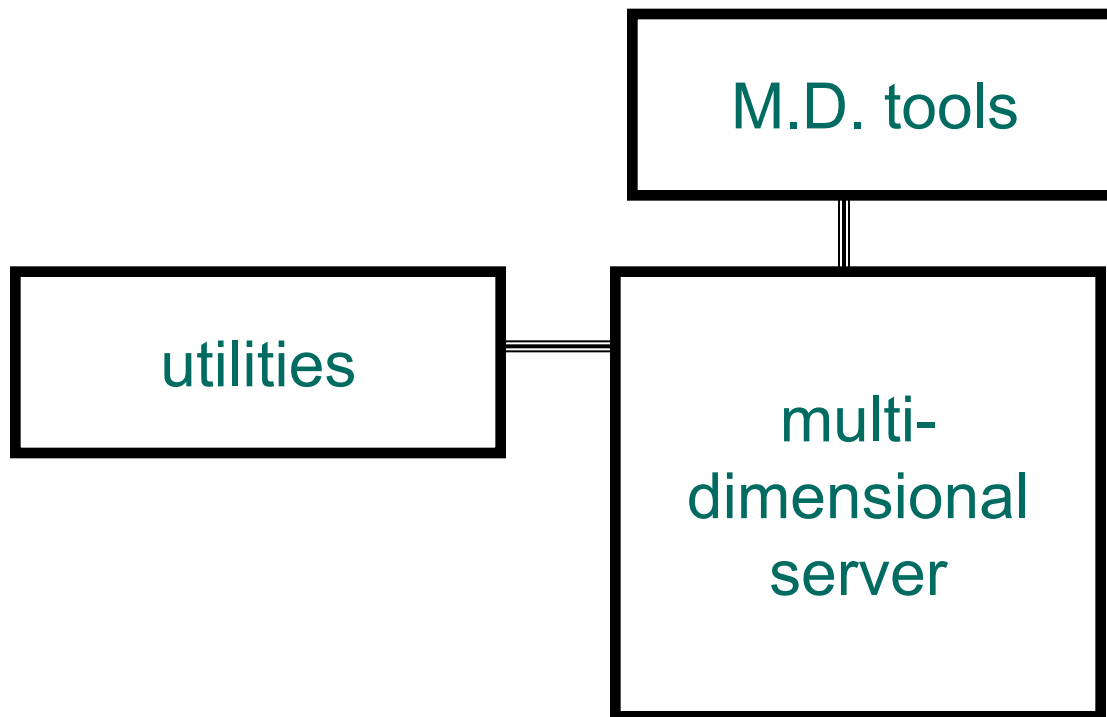
sale	prodId	date	sum
	p1	1	62
	p2	1	19
	p1	2	48



.....Special indices, tuning;
Schema is “denormalized”

MOLAP Server

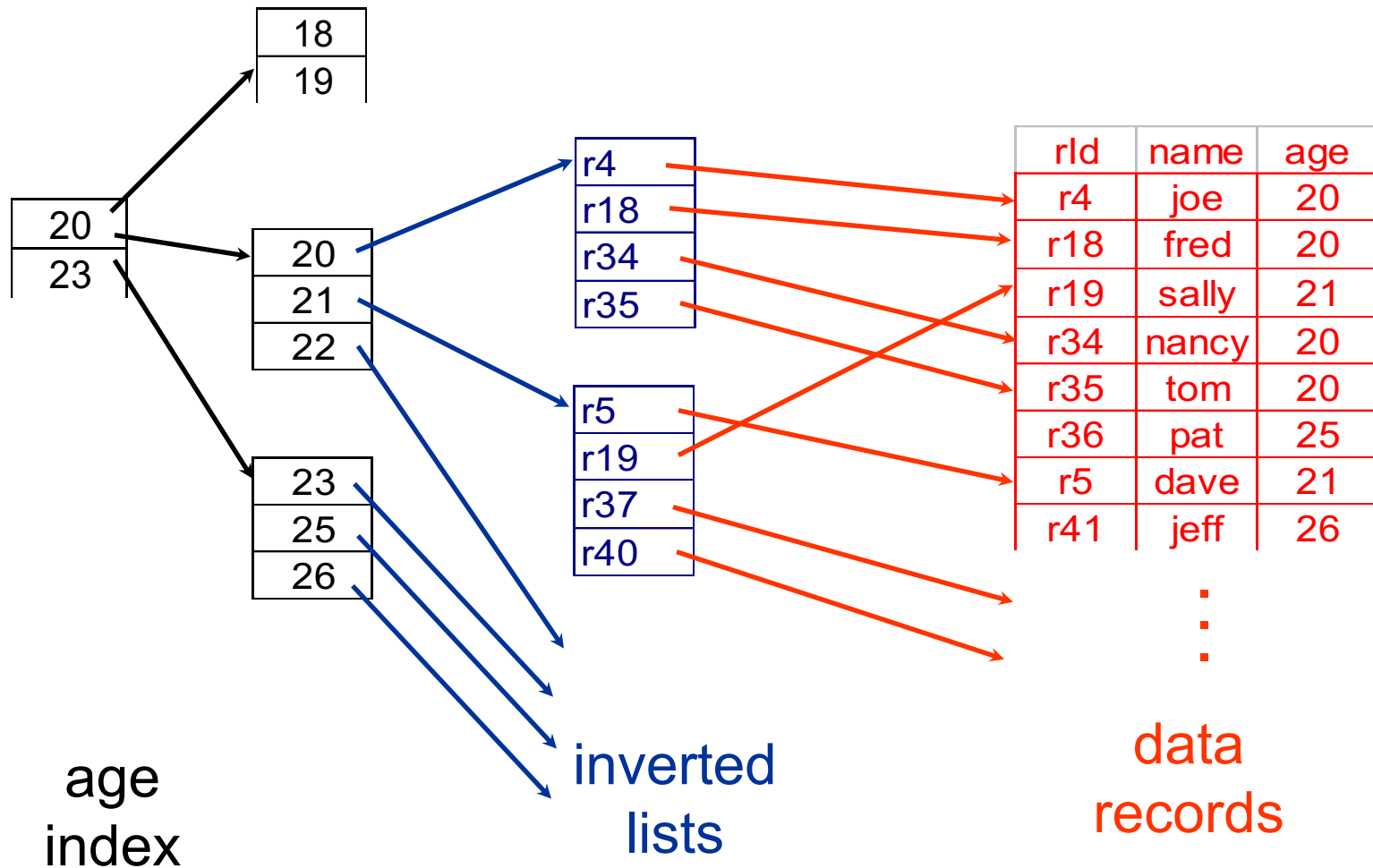
- Multi-Dimensional OLAP Server



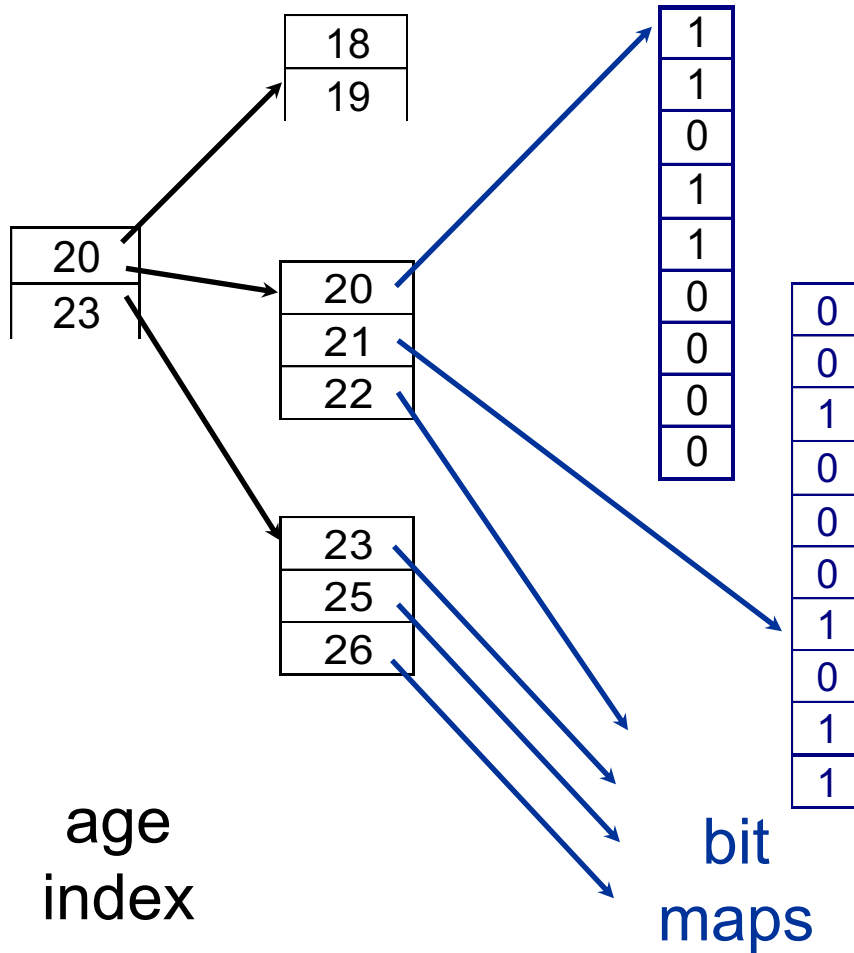
Index Structures

- Traditional Access Methods
 - B-trees, hash tables, grids, ...
- Popular in Warehouses
 - inverted lists
 - bit map indexes

Inverted Lists



Bit Maps



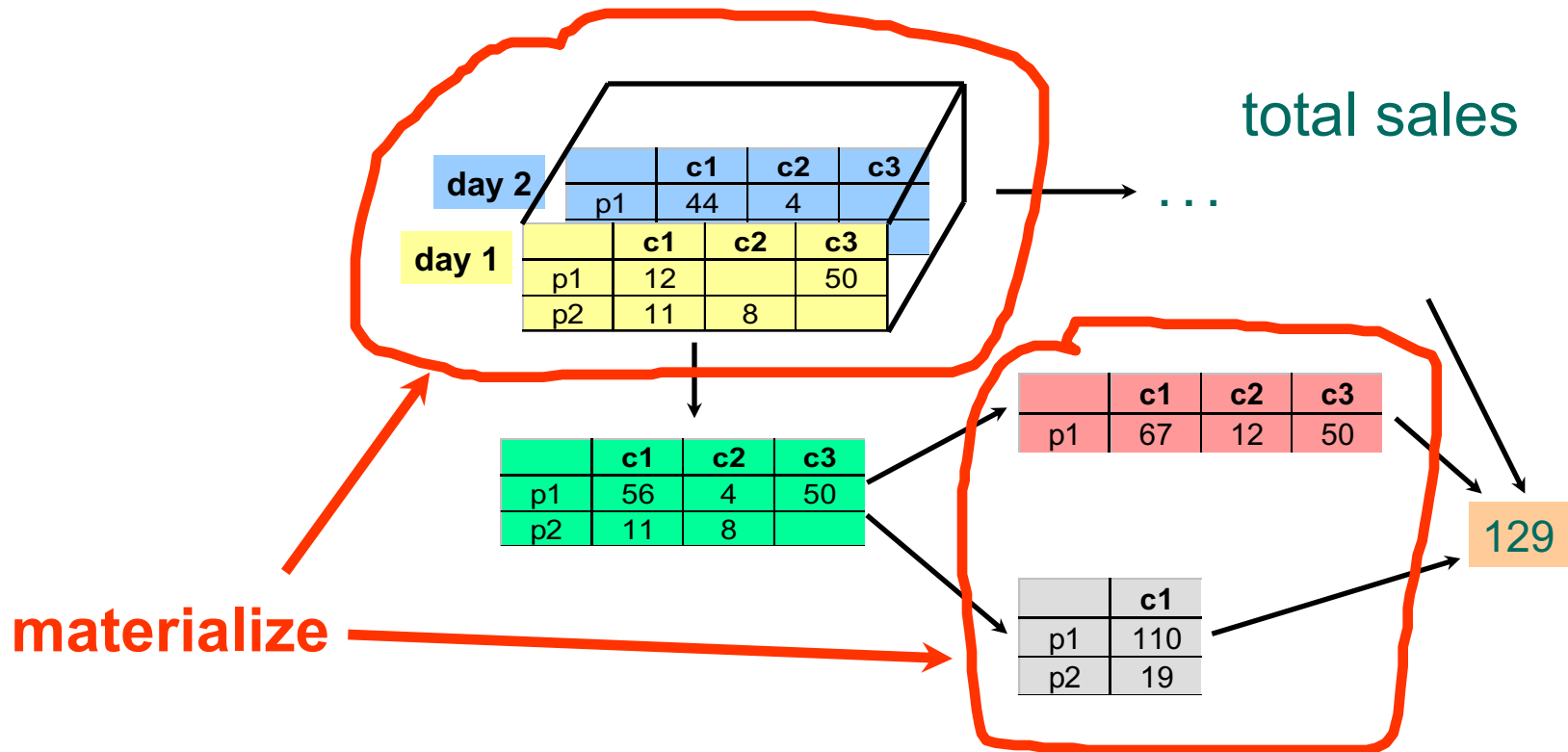
id	name	age
1	joe	20
2	fred	20
3	sally	21
4	nancy	20
5	tom	20
6	pat	25
7	dave	21
8	jeff	26

⋮

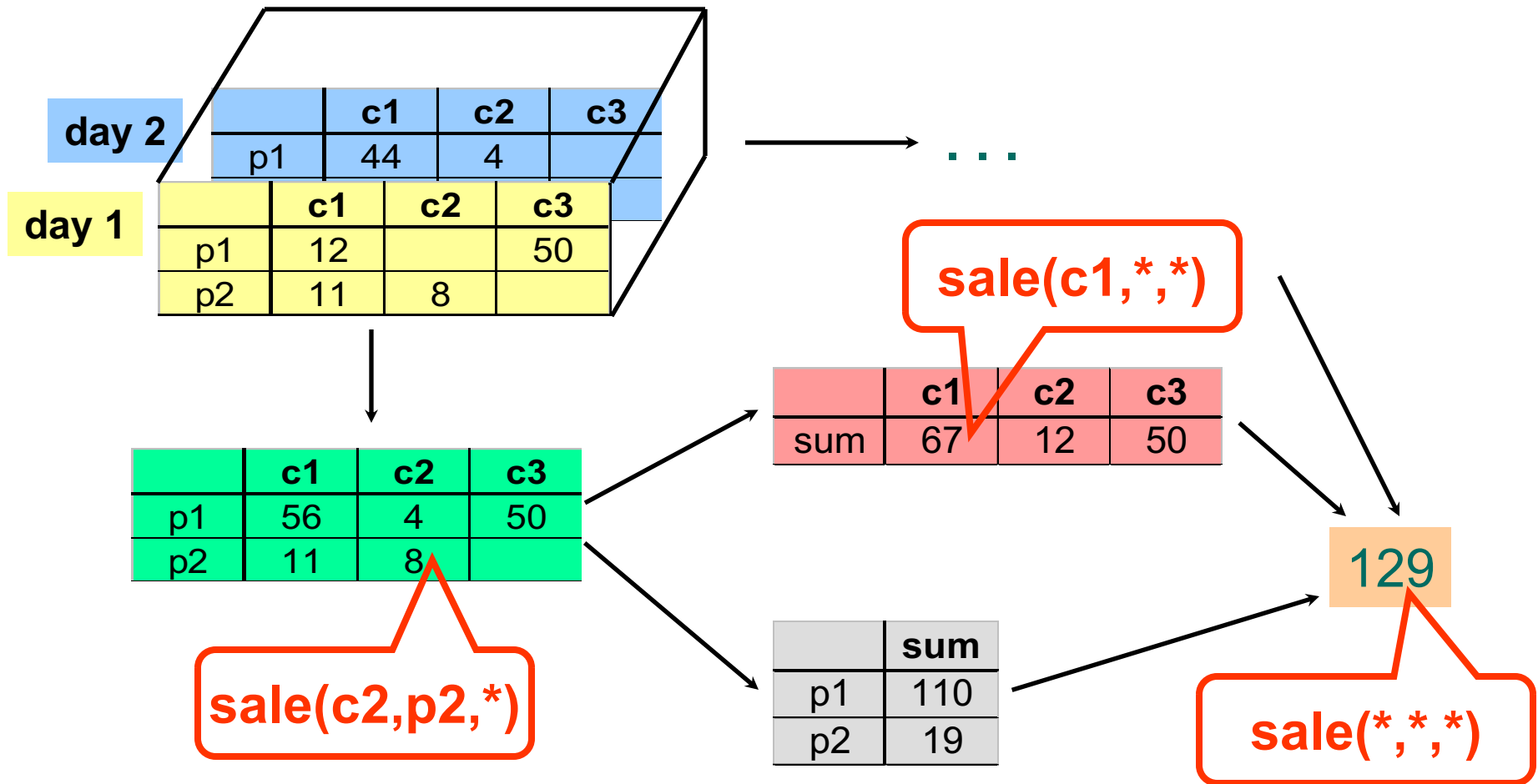
data
records

What to Materialize?

- Store in warehouse results useful for common queries
- Example:



Intermediate Results



Extended Cube

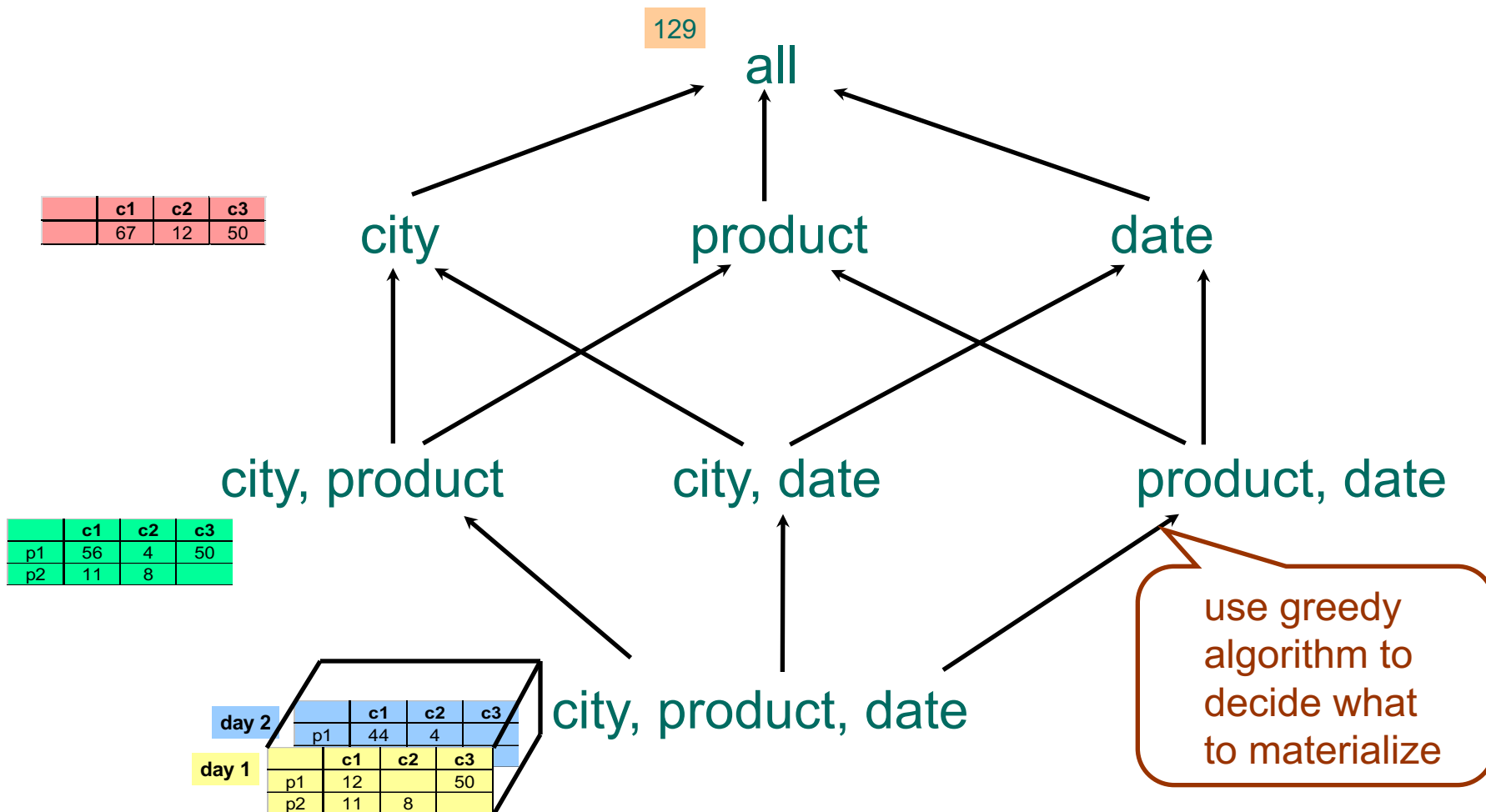
The diagram illustrates an extended cube with three stacked tables. The tables are color-coded: yellow for day 1, blue for day 2, and pink for the third day. A red callout box points to the value 19 in the third table, labeled **sale(*,p2,*)**.

		c1	c2	c3	*
	p1	56	4	50	110
	p2	11	8		19
					129
day 2					
	p1	44	4		48
day 1					
	p1	12		50	62
	p2	11	8		19
	*	23	8	50	81

Materialization Factors

- Type/frequency of queries
- Query response time
- Storage cost
- Update cost

Cube Aggregates Lattice

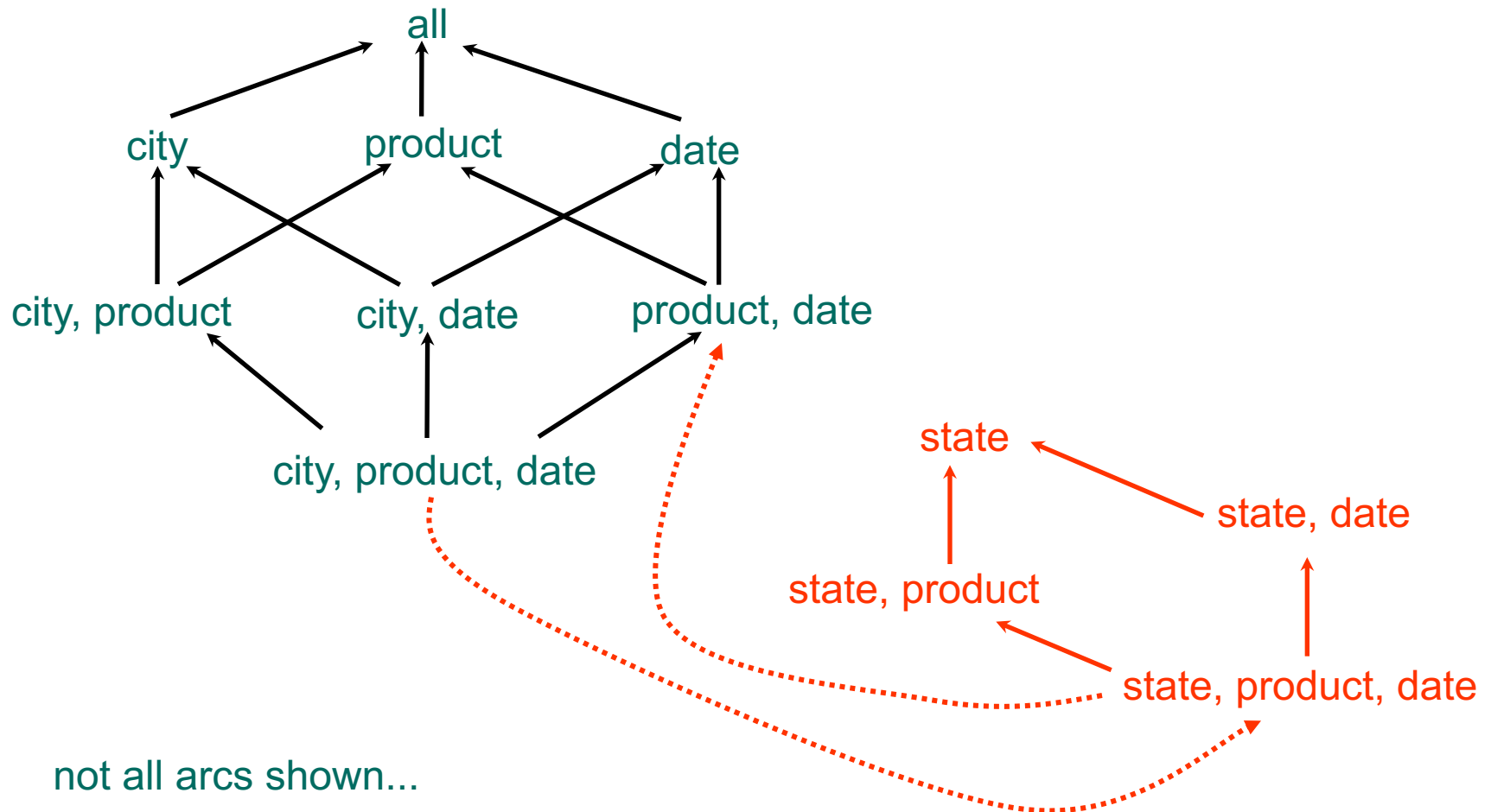


Dimension Hierarchies

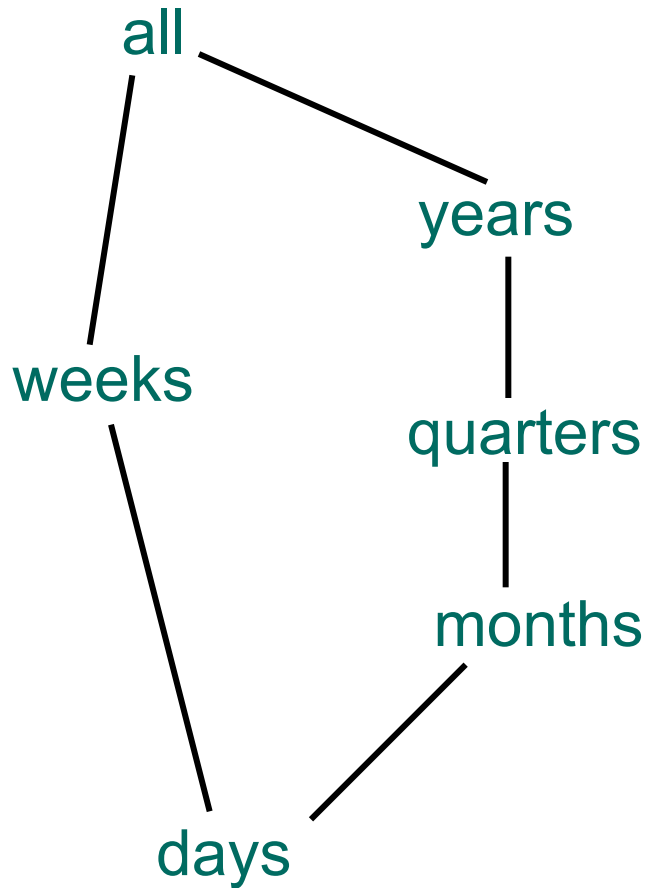
all
|
state
|
city

cities	city	state
	c1	CA
	c2	NY

Dimension Hierarchies



Interesting Hierarchy

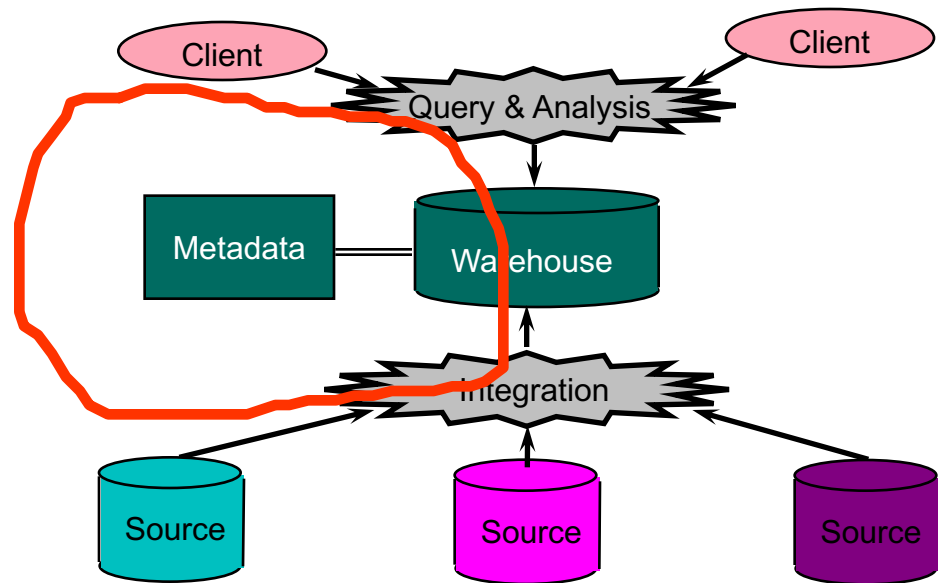


time	day	week	month	quarter	year
	1	1	1	1	2000
	2	1	1	1	2000
	3	1	1	1	2000
	4	1	1	1	2000
	5	1	1	1	2000
	6	1	1	1	2000
	7	1	1	1	2000
	8	2	1	1	2000

conceptual
dimension table

Managing

- Metadata



Metadata

- Administrative
 - definition of sources, tools,
 - schemas, dimension hierarchies,
 - rules for extraction, cleaning,
 - refresh, purging policies
 - user profiles, access control

Current State of Industry

- Extraction and integration done off-line
 - Usually in large, time-consuming, batches
- Everything copied at warehouse
 - Not selective about what is stored
 - Query benefit vs storage & update cost