بنام خدا

مبانى فناورى اطلاعات

Business Intelligence هوش تجاری

بصیری دانشکده برق و کامپیوتر دانشگاه صنعتی اصفهان

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مراجع

- Han, Jiawei, Micheline Kamber, and Data Mining. "Concepts and techniques." *Morgan Kaufmann* 340 (2006): 94104-3205.
- Kimball, Raiph. The data warehouse toolkit. John Wiley & Sons, 2006.
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Outline

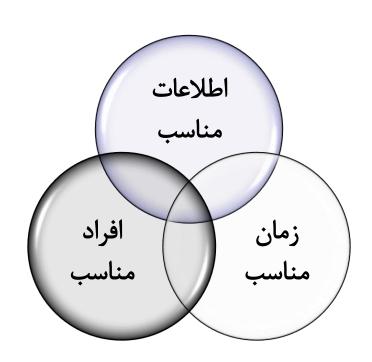
- Business Intelligence
 - Definition
 - □ Important
 - ☐ History
 - □ Architecture
- Data Warehouse
 - Definition
 - □ Characteristics
 - □ OLAP
 - Conceptual Modeling
 - □ Cube

هوش تجاری چیست؟

- 1989 Howard Dresner (Gartner Group) ■
- مجموعه ای از مفاهیم و متدها جهت بهبود تصمیم سازی تجاری با بکارگیری سیستم های حقیقت–محور
 - ابزار ETL
 - ابزار OLAP
 - ابزار گزارشگیری
- جمعآوری، نگهداری، تجزیه و تحلیل و انتشار داده الطلاعات به منظور تسهیل در تصمیم گیریهای مدیریتی.



هوش تجاری (Business Intelligence)

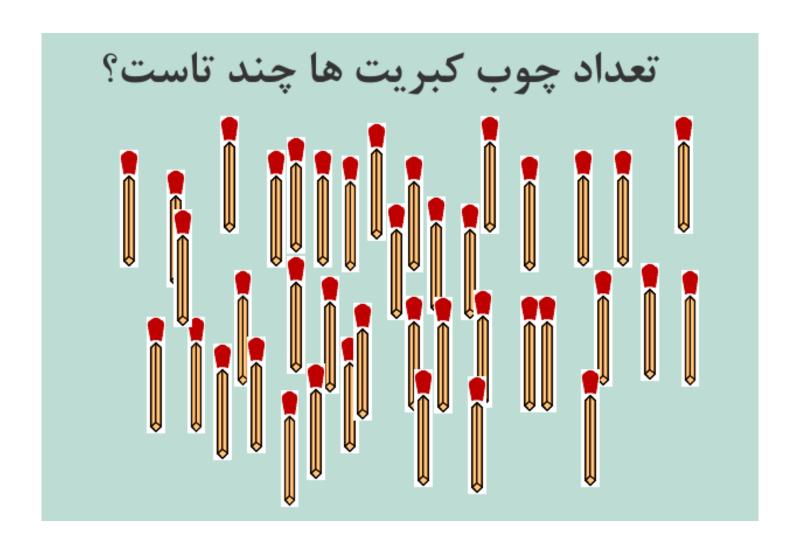


هوش تجاری یعنی دراختیار قراردادن اطلاعات مناسب به افراد مناسب در زمان مناسب برای اخذ تصمیم مناسب

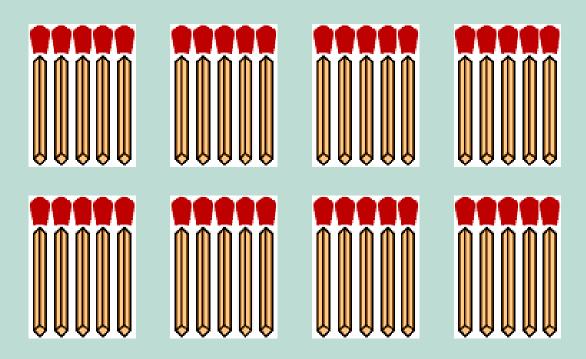
چرا به انبار داده و هوش تجاری نیاز داریم؟



- نیاز به یک چارچوب تصمیم گیری برای تسریع در اتخاذ تصمیمات موثر
 - دسترسی سریع به تمام اطلاعات موجود
 - امکان ساخت هر گونه گزارش به صورت دینامیک
- امکان تحلیل اطلاعات به صورت Historical(مبتنی بر زمان) از کل به جزء(Drill Down) با استفاده از تجمیع ها(Aggregation)
 - امکان انجام تحلیلهای آماری و مبتنی بر داده کاوی
 - ایجاد مرجع واحد آمار و گزارشات
 - امکان برنامه نویسی و تولید نرم افزارهای خاص منظوره



تعداد چوب كبريت ها اكنون چند تاست؟



ابزارهای هوش تجاری

- BizzScore Suite
- Board Management IntelligenceToolkit
- Business Objects Enterprise
- IBM Cognos
- JasperSoft
- Microsoft BI tools
- Microstrategy

- Oracle
- WebFocus
- Tableau Software
- Style Intelligence
- SAS
- SAP
- QlikView
- Pentaho BI Suite
- Actuate
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History of Business Intelligence

Previous Names

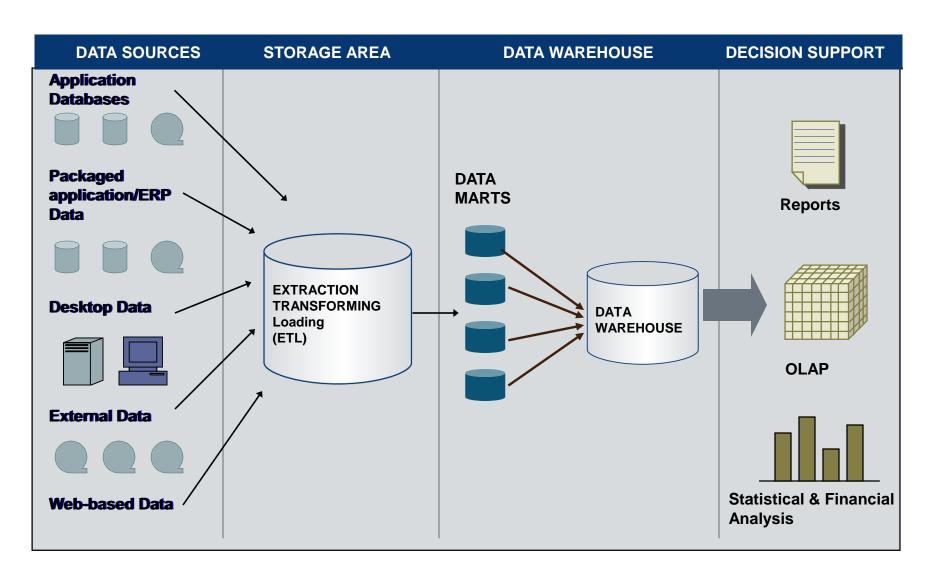
- Decision Support Systems
- Executive InformationSystems
- Online Analytic Processing

Related Concepts

- Competitive Intelligence
- Market Intelligence
- Customer Intelligence
- Strategic Intelligence
- Technical Intelligence
- Knowledge Management
- Data Miming

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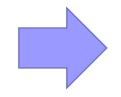
BI Architecture





انبار داده یک منبع واحد از دادههاست که دادههای منابع مختلف اطلاعاتی سازمان در آن جمع آوری، دستهبندی، خلاصهسازی و ذخیره شده تا تصمیم گیری را در سازمان تسهیل نماید.

برای دسترسی آسان کاربران به حجم زیادی از دادهها طراحی شده است، و دسترسی به دیتا عموما به وسیله ابزارهای تحلیلی ویژه و اپلیکیشنها پشتیبانی میشود.



OLAPابزارهای دسترسی به انبارداده را تسهیل نموده است.

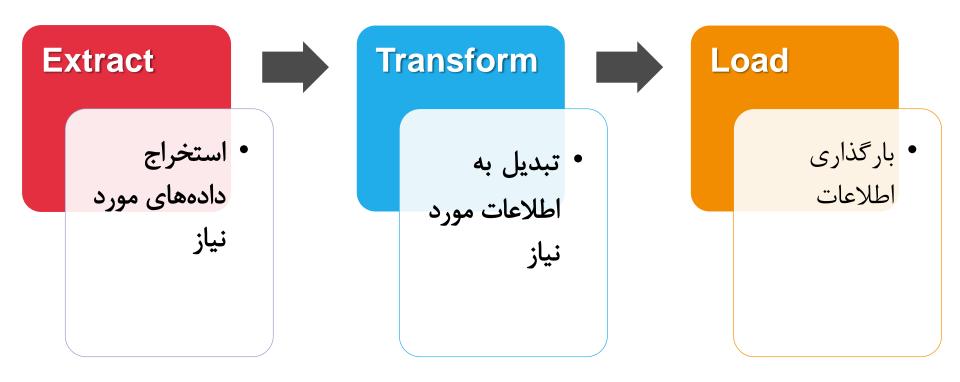
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Extraction, Transformation, and Loading (ETL)

- The process of data consolidation is often called Extraction, Transformation, and Loading (ETL)
 - □ The ETL process extracts data from the various source systems
 - Data is then transformed to make it consistent and improve data quality
 - The consolidated, consistent, and cleaned data is then loaded into a data repository
- Developing the ETL process often consumes 80% of the development time

Extraction, Transformation, and Loading (ETL)

- استخراج داده ها از منابع داده ای مختلف
- تحلیل و تبدیل دادهها مطابق با نیازمندیهای کسب و کار
 - بارگذاری اطلاعات





(ETL) Tools

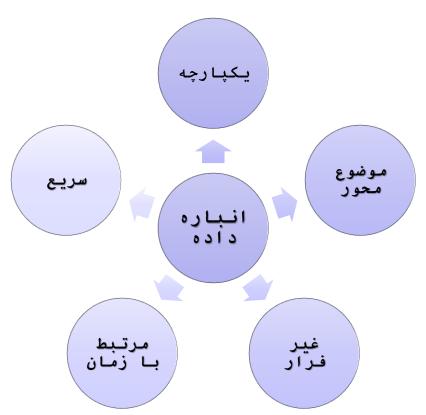
- Some ETL Tools
 - □ Oracle Data Integrator (ODI)
 - □ Informatica
 - □ IBM Ascential
 - □ SSIS (Microsoft SQL SERVER)



Data Warehouse Characteristics

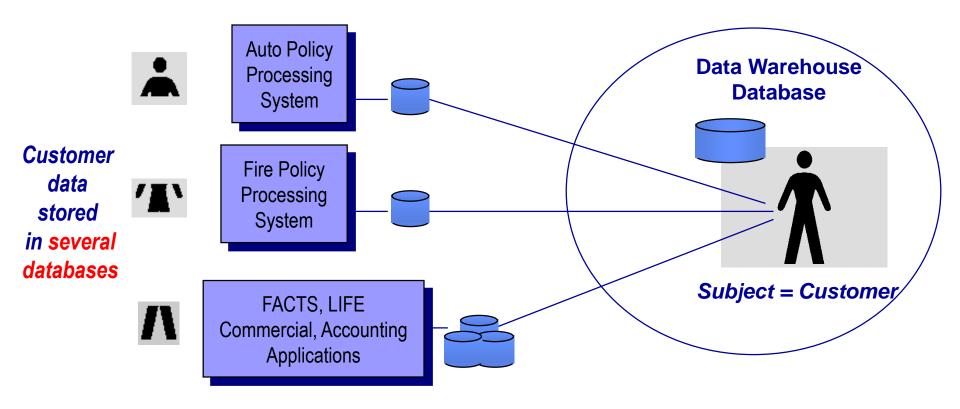
Key Characteristics of a Data Warehouse

- Integrated
- Time-variant
- Non-volatile
- Subject-Oriented
- Fast



Integrated

 Data is stored once in a single integrated location (e.g. insurance company)

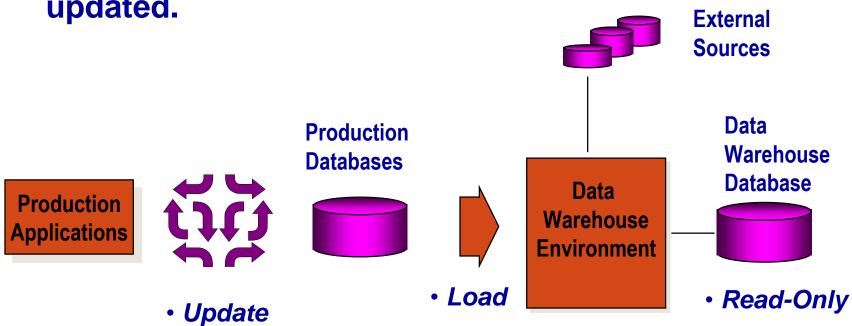


Data Warehouse—Time Variant

- The time horizon for the data warehouse is significantly longer than that of operational systems
 - Operational database: current value data
 - Data warehouse data: provide 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"

Non-Volatile

Existing data in the warehouse is not overwritten or updated.



- Insert
- Delete



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 vs. Operational DBMS

- 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 analytical processing)
 - Major task of data warehouse system
 - Data analysis and decision making
- Distinct features (OLTP vs. OLAP):
 - User and system orientation: customer vs. market
 - □ Data contents: current, detailed vs. historical, consolidated
 - □ Database design: ER + application vs. star + subject
 - View: current, local vs. evolutionary, integrated
 - Access patterns: update vs. read-only but complex queries

OLAP

- √ On-Line Analytical Processing) OLAP) با استفاده از تجمیع مقیاسها و شکستن آنها بر روی ابعاد مختلف کسبوکار، یک شمای متناسب با نیازهای گوناگون بخشهای مختلف در اختیار بانک قرار می دهد.
 - ✓ Cube قالب نگهداری داده ها در مدل چند بعدی است.
- √با استفاده از Cube داده ها به آسانی و به سرعت می تواند در اختیار کاربران غیر فنی قرار گیرد.
- √ در Cube اطلاعات به صورت تجمیع شده محاسبه و نگهداری می شوند و در زمان واکشی داده از Cube محاسبات انجام نمی گیرد.

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OLTP vs. OLAP

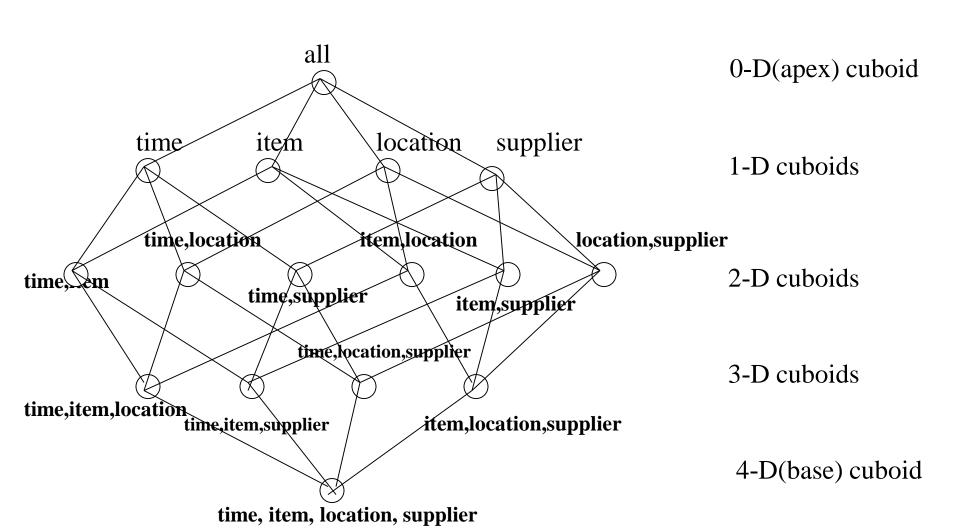
	OLTP	OLAP
users	clerk, IT professional	knowledge worker
function	day to day operations	decision support
DB design	application-oriented	subject-oriented
data	current, up-to-date detailed, flat relational isolated	historical, summarized, multidimensional integrated, consolidated
usage	repetitive	ad-hoc
access	read/write index/hash on prim. key	lots of scans
unit of work	short, simple transaction	complex query
# records accessed	tens	millions
#users	thousands	hundreds
DB size	GB to high-order GB	>=TB
metric	transaction throughput	query throughput, response



From Tables and Spreadsheets to Data Cubes

- A data warehouse is based on a multidimensional data model which views data in the form of a data cube
- A data cube, such as sales, allows data to be modeled and viewed in multiple dimensions
 - □ Dimension tables, such as item (item_name, brand, type), or time(day, week, month, quarter, year)
 - Fact table contains measures (such as dollars_sold) and keys to each of the related dimension tables
- In data warehousing literature, an n-D base cube is called a base cuboid. The top most 0-D cuboid, which holds the highest-level of summarization, is called the apex cuboid. The lattice of cuboids forms a data cube.

Cube: A Lattice of Cuboids





Conceptual Modeling of Data Warehouses

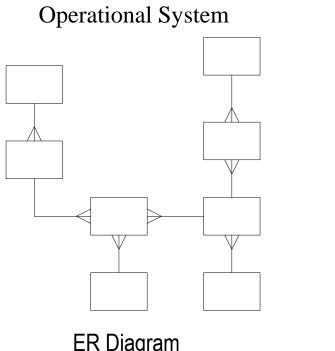
- Modeling data warehouses: dimensions & measures
 - ☐ <u>Star schema</u>: A fact table in the middle connected to a set of dimension tables
 - □ <u>Snowflake schema</u>: A refinement of star schema where some dimensional hierarchy is <u>normalized</u> into a set of smaller dimension tables, forming a shape similar to snowflake
 - ☐ <u>Fact constellations</u>: Multiple fact tables share dimension tables, viewed as a collection of stars, therefore called <u>galaxy schema</u> or fact constellation

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Dimensional Modeling

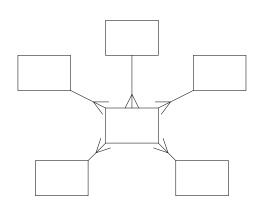
- Dimensional modeling = data warehouse modeling technique
- 2 types of tables: facts and dimensions.
- A fact table contains one or more measures (usually numerical) of a subject that is being modeled for analysis.
- Dimension tables contain various descriptive attributes (usually textual) that are related to the subject depicted by the fact table.
- The intent of the dimensional model is to represent relevant questions whose answers enable appropriate decision making in a specific business area

مدل ذخيره داده



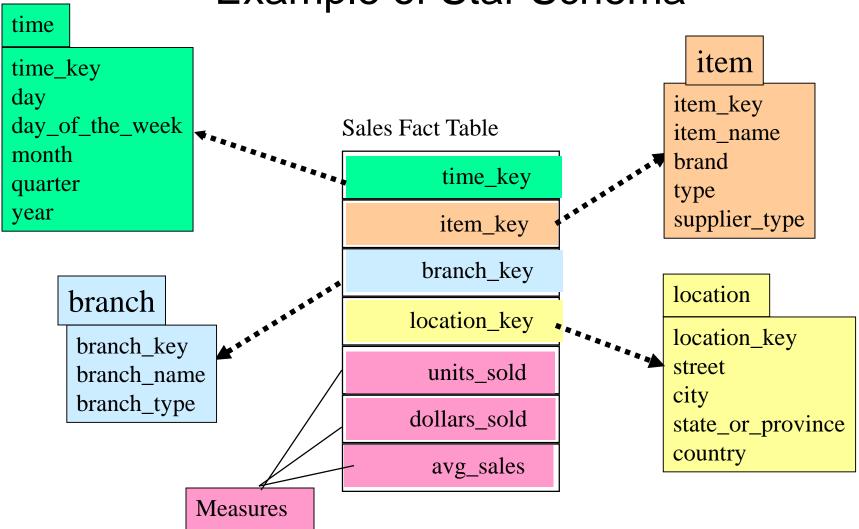
ER Diagram

Data Warehouse

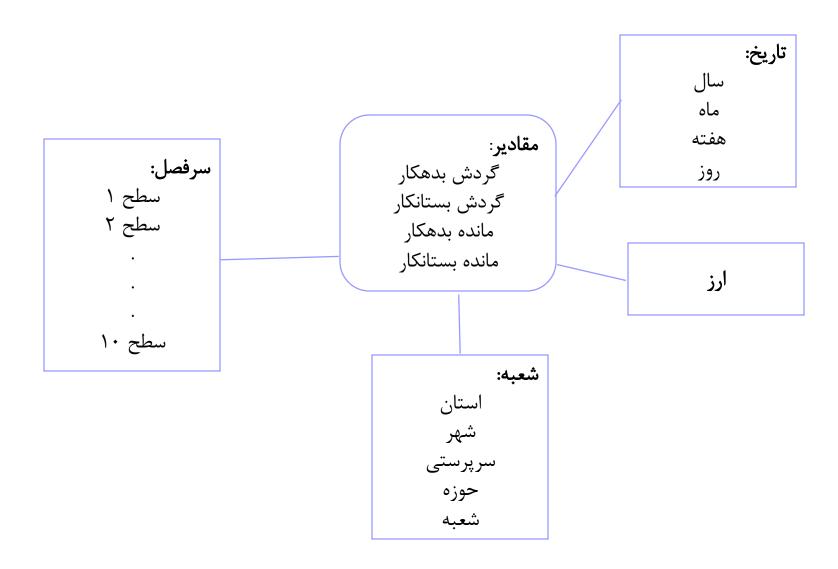


Star Schema

Example of Star Schema



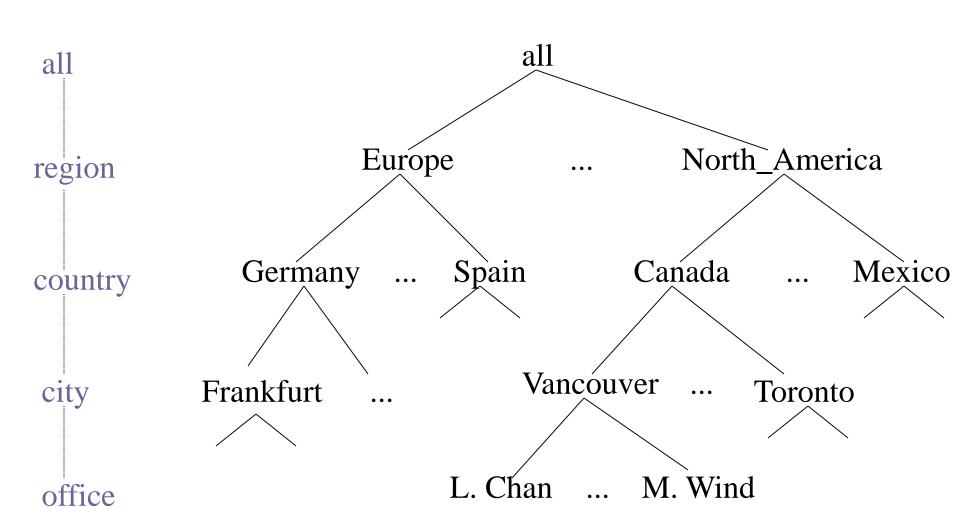
مثال مدل STAR



Measures of Data Cube: Three Categories

- Distributive: if the result derived by applying the function to n aggregate values is the same as that derived by applying the function on all the data without partitioning
 - E.g., count(), sum(), min(), max()
- Algebraic: if it can be computed by an algebraic function with M arguments (where M is a bounded integer), each of which is obtained by applying a distributive aggregate function
 - E.g., avg(), min_N(), standard_deviation()
- Holistic: if there is no constant bound on the storage size needed to describe a subaggregate.
 - E.g., median(), mode(), rank()

A Concept Hierarchy: Dimension (location)

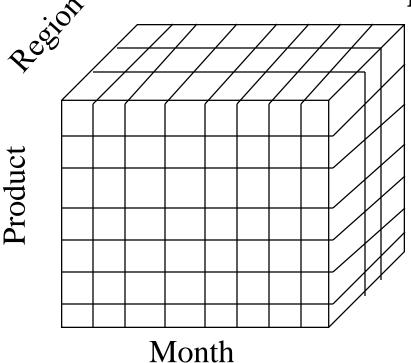


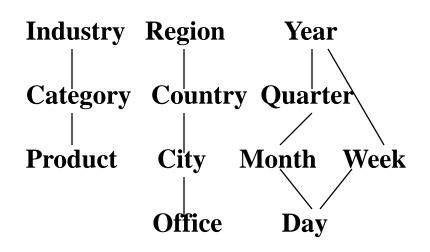
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Multidimensional Data

Sales volume as a function of product, month, and region

Dimensions: Product, Location, Time Hierarchical summarization paths





Typical OLAP Operations

- Roll up (drill-up): summarize data
 - by climbing up hierarchy or by dimension reduction
- Drill down (roll down): reverse of roll-up
 - □ from higher level summary to lower level summary or detailed data, or introducing new dimensions
- Slice and dice: project and select
- Pivot (rotate):
 - reorient the cube, visualization, 3D to series of 2D planes

