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INFO90002 Database Systems & Information Modelling

Lecture 19
Data Warehousing

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By the end of this class you should be able to:

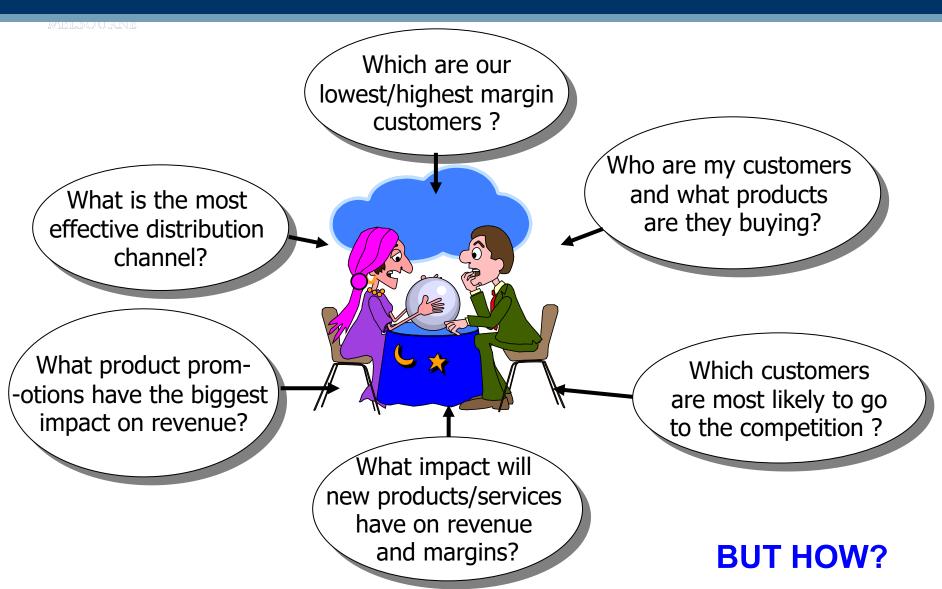
- Articulate the differences between transactional (operational) and informational (dimensional) databases
- Explain the characteristics of a DW
- Understand and explain the overall architecture of a DW
- Design Star Schemas



Part 1: Introduction to Data Warehousing



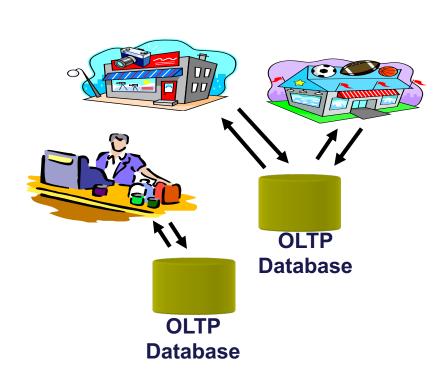
Motivations: A manager wants to know....





Relational Databases for Operational Processing

- MIELDUUKNI
- Used to run day to day business operations
- Automation of routine business processes
 - Accounting
 - Inventory
 - Purchasing
 - Sales
- Created huge efficiencies





MELBOURNE OLTP Databases

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- OLTP = "OnLine Transaction Processing"
- Transaction processing supports daily (routine, repetitive) operations
 - Mundane but crucial
 - Become even more important with the growth of the internet
- Definition:
 - Collection of read/write operations
 - Processed as one unit
 - Reliably and efficiently processed
 - No data loss due to interference and failures (operating system, program, disk, ...)



MELBOURNE OLTP Data Characteristics

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- Characteristics of data:
 - Transaction oriented
 - DML
 - Inserts Updates Deletes
 - May be inconsistent and incomplete
 - Data may not be in its final form
 - Volatile continually changing
 - Data maybe subject to change
 - Current
 - Data related to the operation of the business TODAY!



Databases are great, BUT for business...

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- Too many of them
 - Everybody wanted one, or two, or more
 - Production, Marketing, Sales, Accounting ...
- Everybody got what was best for them
 - IBM, Oracle, Access, Microsoft
- Eventually this re-created the problem databases were meant to solve
 - Duplicated data
 - Inaccessible data
 - Inconsistent data

But data is useful for analysis and decision making



MELBOURNE What can we do about it?

- Need an integrated way of getting the ENTIRE organisational data
- Its really an Informational Database, rather than a Transactional Database
 - A single database that allows all of the organisations data to be stored in a form that can be used to support organisational decision processes

- A centralised repository for decision making
 - Populated from operational databases and external data sources
 - Integrated and transformed data
 - Optimised for reporting
- Single Point of Truth about the data (SPOT)



Warehouse: An Informational Database

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Data Warehouse:

- A single repository of organisational data
- Integrates data from multiple sources
 - Extracts data from source systems, transforms, loads into the warehouse
- Makes data available to managers/users
- Supports analysis and decision-making
- Involve a large data store (often several Terabytes, Petabytes of data)



Difference between Transactional & Informational Systems

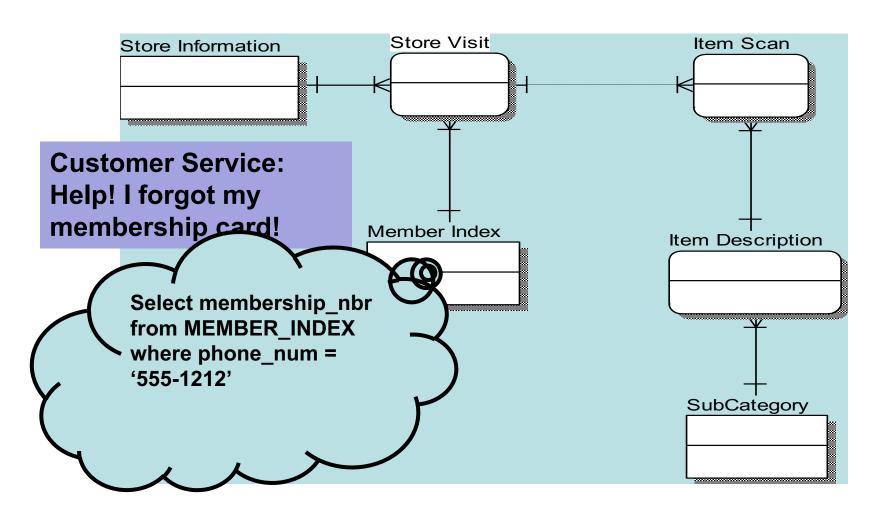
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Characteristic	Transactional	Informational
Primary Purpose	Run the day to day business	Support decision making
Type of Data	Current data – representing the state of the business	Historical data – snapshots and predictions
Primary Users	Customers, clerks and other employees	Managers, analysts
Scope of Usage	Narrow, planned, fixed interfaces	Broad, ad hoc, complex interfaces
Design Goal	Performance and availability	Flexible use and data accessibility
Volume	Many constant updates and queries on a few tables or rows	Periodic batch updates, complex querying on multiple or all rows



Transactional (Operational) Questions

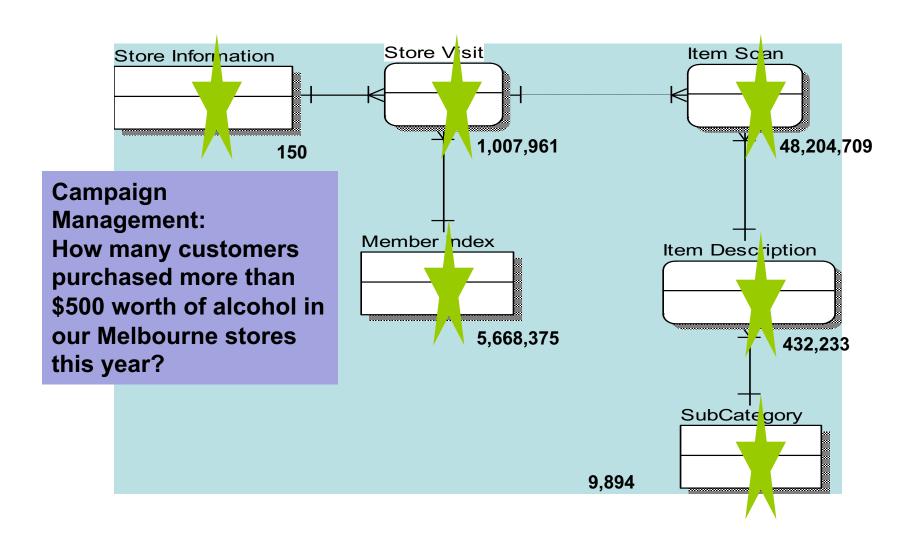
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MELBOURNE Analytical Questions

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MELBOURNE DW Supports Analytical Queries

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- One is interested in numerical aggregations
 - How many?
 - What is the average?
 - What is the total cost?
- One is interested in understanding dimensions
 - Sales by state by customer type
 - Sales by product by store by quarter

DW will help answer these questions



MELBOURNE Characteristics of a DW

- Subject oriented
 - Data warehouses are organised around particular subjects (sales, customers, products)
- Validated, Integrated data
 - Data from different systems converted to a common format: allows comparison and consolidation of data from different sources
 - Data from various sources validated before storing it in a data warehouse



MELBOURNE Characteristics of a DW

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Time variant

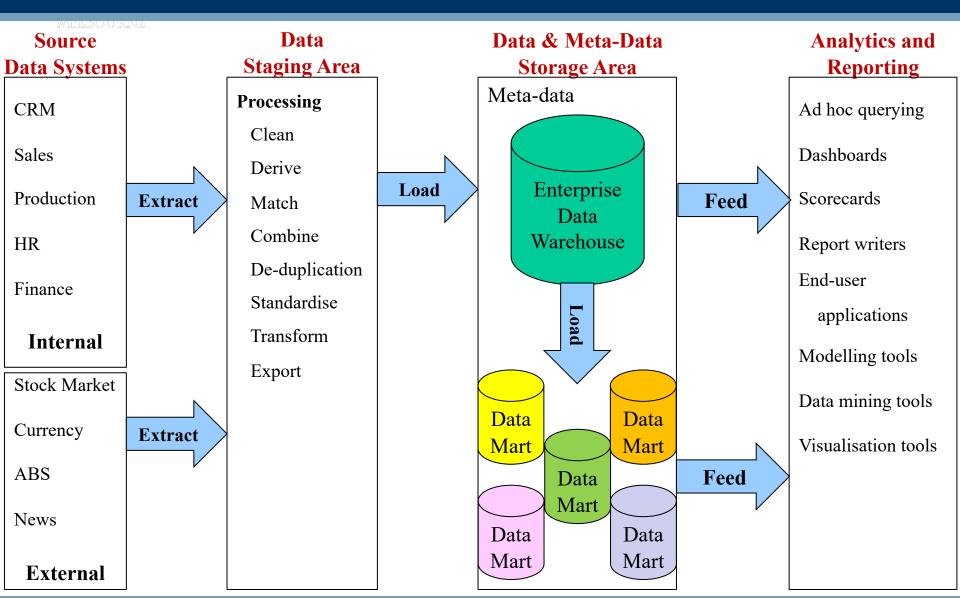
- Historical data
- Trend analysis crucial for decision support: requires historical data
- Data consists of a series of "snapshots" which are time stamped

Non-volatile

 Users have read access only – all updating done automatically by ETL process and periodically by a DBA

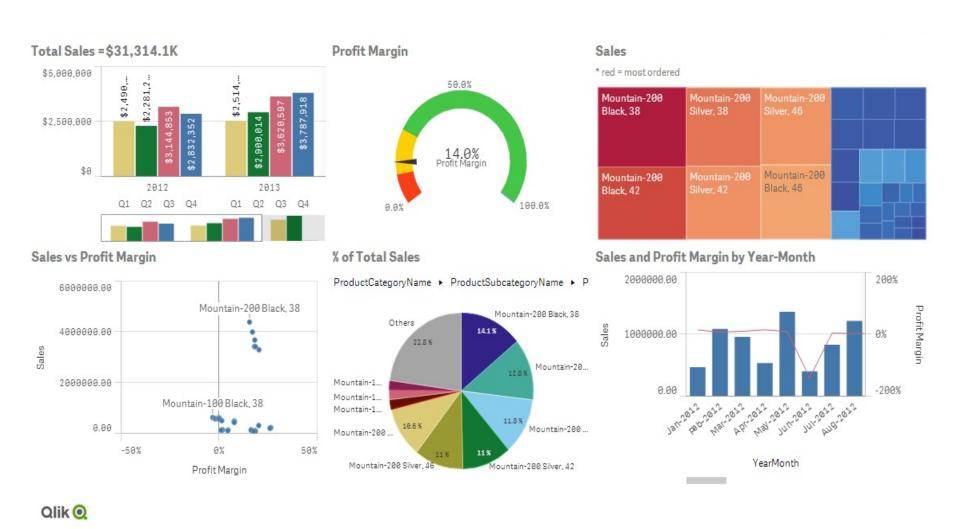


MELBOURNE A DW Architecture





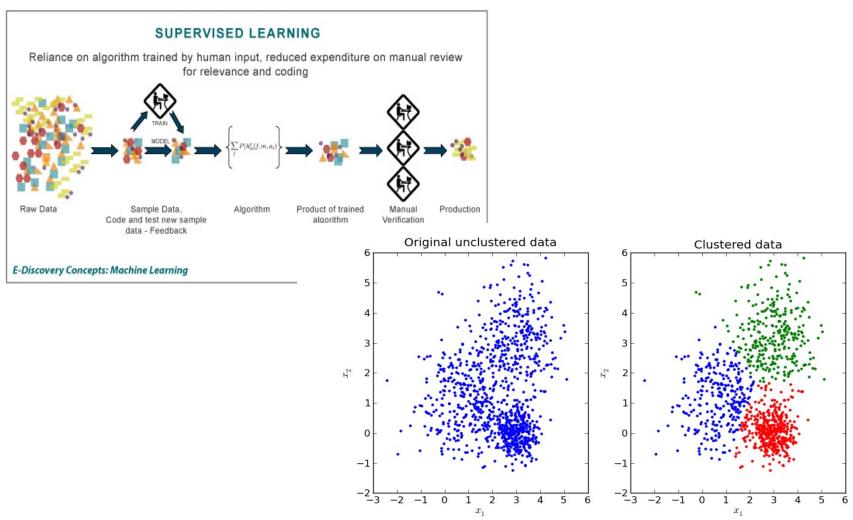
Business Intelligence Dashboard





MELBOURNE DW Supports Advanced Analytics

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http://us.hudson.com/legal/blog/postid/513/predictive-analytics-artificial-intelligence-science-fiction-e-discovery-truth http://pypr.sourceforge.net/kmeans.html



Part 2- Dimensional Modelling

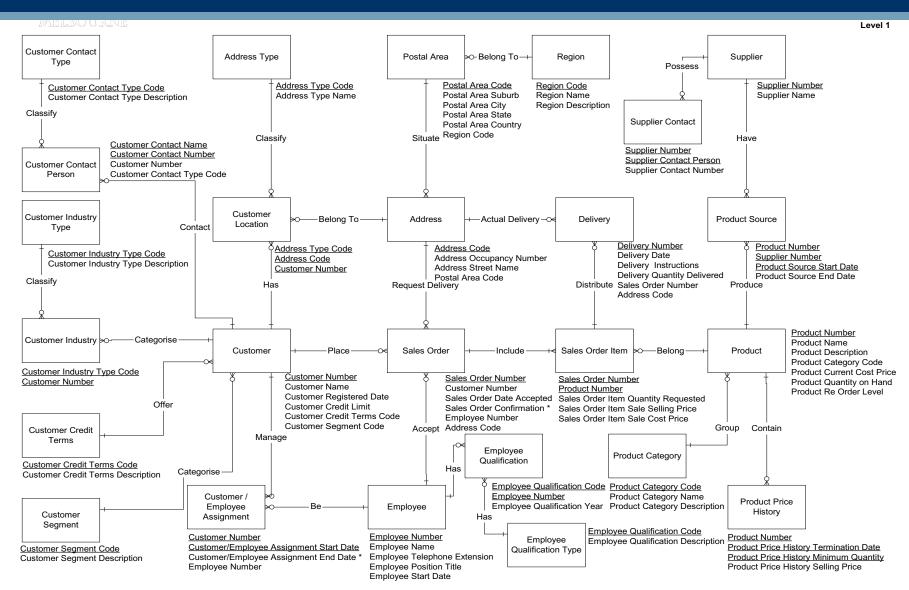


MELBOURNE Business Analyst World

- How much revenue did the product G generate in the last three months, broken down by month for the south eastern sales region, by individual stores, broken down by promotions, compared to estimates and to the previous version of the product
 - Analysis starts usually with a single indication of something strange, then goes deep into the data, left to a new dimension, right to another, up to the summary, back down and left and right again, until the problem is identified...
 - Dimensional Analysis: To support business analysts view
 - Revenue per product per customer per location?
 Fact Dimension Dimension Dimension



Example ER model





Introduction to Dimensional Modelling

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- Popularised by Ralph Kimball in the 1990s
- Based on the multi-dimensional model of data and designed for retrieval-only databases
- Very simple, intuitive, and easily-understood structure
- Also known as star schema design



MELBOURNE Dimensional Modelling

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- A dimensional model consists of:
 - Fact table
 - Several dimensional tables
 - (Sometimes) hierarchies in the dimensions
- Essentially a simple and restricted type of ER model

MELBOURNE Fact Table

- A fact table contains the actual business measures (additive, aggregates), called *facts*
- The fact table also contains foreign keys pointing to dimensions





MELBOURNE Fact Table - example

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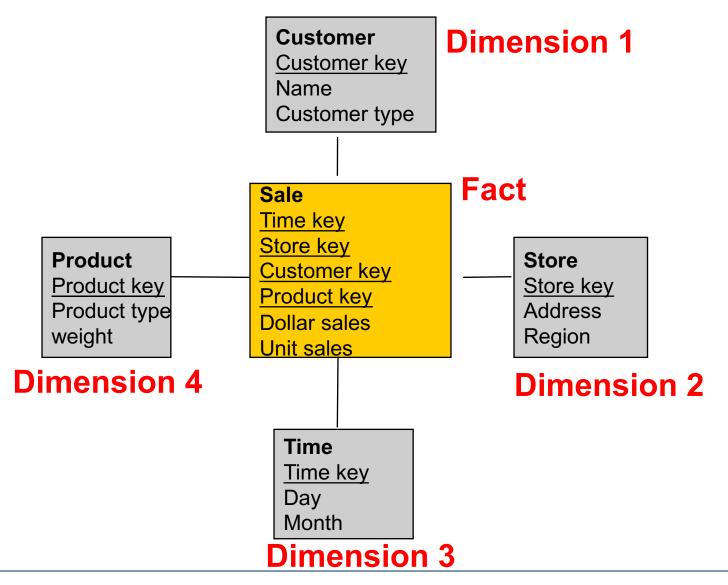
- Actual data might look like this
- Granularity, or level of detail, is a key issue
 - Finest level of detail for a fact table, determined by the finest level of each dimension

Time-id	Store-id	Cust-id	Prod-id	Dollar	Unit Sales
				sales	
T100	S303	C101	P98	\$120,000	5,000
T101	S303	C256	P98	\$240000	10,000
T102	S387	C101	P10	\$456,000	27,899
T100	S234	C400	P56	\$100,200	5,600



MELBOURNE Star schema – dimensional model

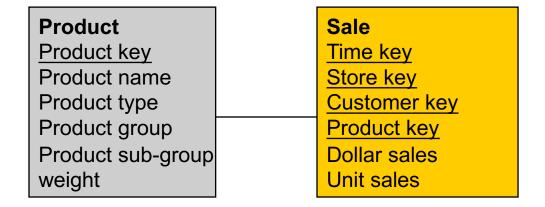
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Dimension Hierarchies

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Product name e.g. Hammer

- Product type e.g. Tool
 - Product group e.g. Hardware



Dimension Table - example

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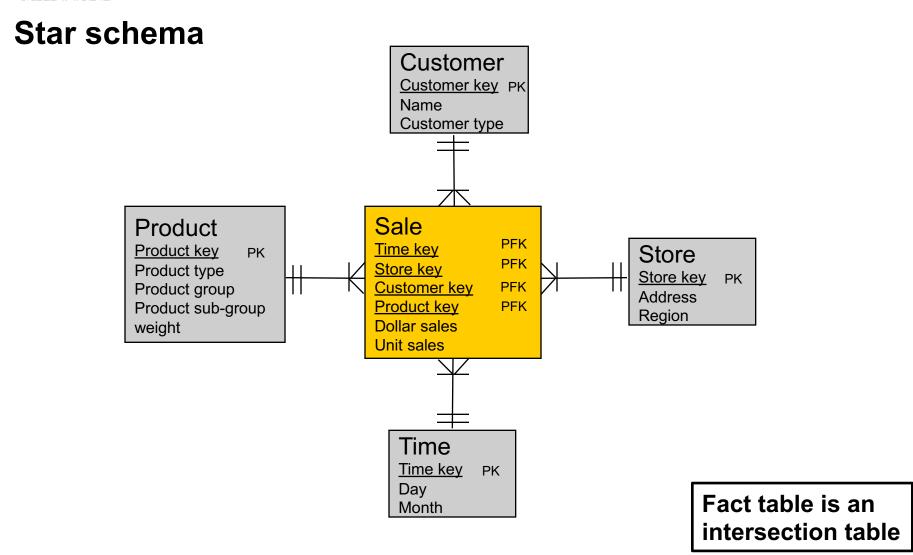
- Actual data might look like this
- Hierarchy evident in data

Prod-id	Prod-Name	Prod-Group	Prod-	Weight
			Subgroup	
P10	Hammer	Hardware	Tool	5kg
P56	10cm Nails	Hardware	Nails	1kg
P98	Plastic Pipe	Plumbing	Pipe	1kg



Dimensional model as an ER model

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MELBOURNE Designing a Dimensional Model

Steps:

- 1. Choose a Business Process
- 2. Choose the measured facts (usually numeric, additive quantities)
- 3. Choose the granularity of the fact table
- 4. Choose the dimensions
- 5. Complete the dimension tables

(Kimball, 1996)



Embedded Hierarchies in Dimensional Tables

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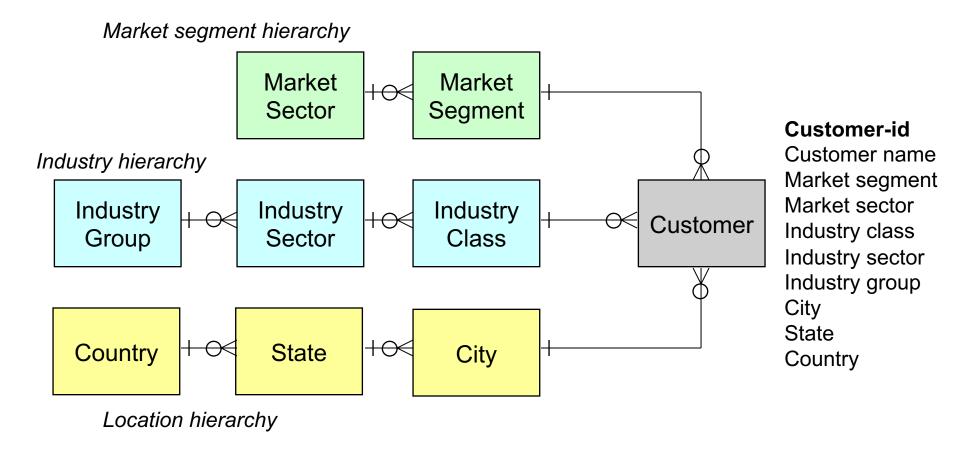
Customer

Customer-id
Customer name
Market segment
Market sector
Industry class
Industry sector
Industry group
City
State
Country



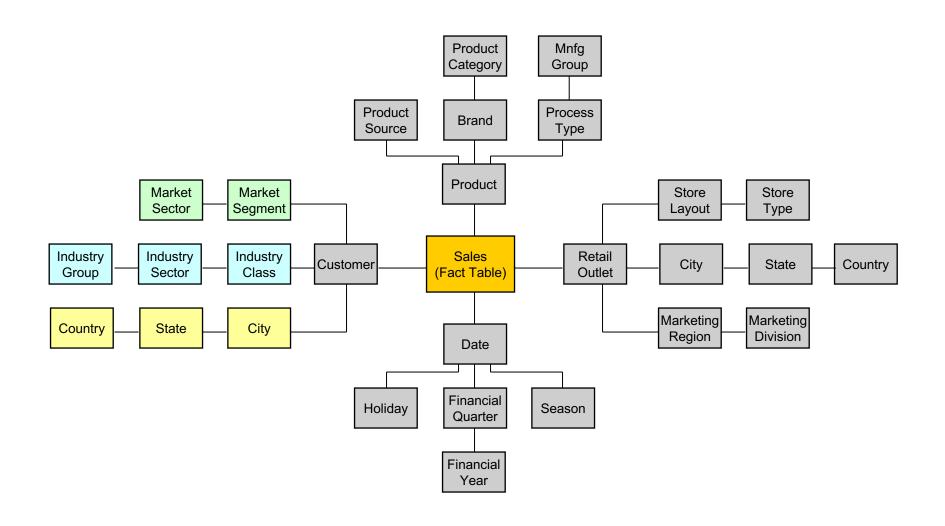
MELBOURNE Embedded Hierarchies in Dimensional Tables

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MELBOURNE Snowflake Schema: hierarchy in dimensions





Design Outcomes: Normalised or Denormalised?

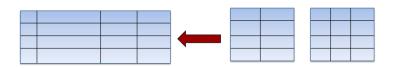
Normalisation

- Eliminates redundancy
- Storage efficiency
- Referential Integrity



Denormalisation

- Fewer tables (fewer joins)
- Fast querying
- Design is tuned for end-user analysis



Homework...

- MELDQUKNI
- We are making a data warehouse for a real estate agency. The company wants to track information about the **selling** of their properties. This warehouse keeps information about the agents (license#, first name, last name, phone #), buyers that come in (buyer id, first name, last name, phone #), and property (property#, property address, price). The information managers want to be able to find is the number of times a property is viewed, sales price. The information needs to be accessible by rental agent, by buyer, by property and for different time (day, week, month, quarter and year).
- Draw a star schema to support the design of this data warehouse.



MELBOURNE What is Examinable?

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- Differences between transactional and informational databases
- Designing a star schema
- Defining facts and dimension tables



MELBOURNE If you want to know more

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More technical details (Not assessed)

https://www.youtube.com/watch?v=w-S0fj0fmqg&list=PLdQddgMBv5zHcEN9RrhADq3CBColhY2hl&index=17

Next Lecture

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Security & Ethics