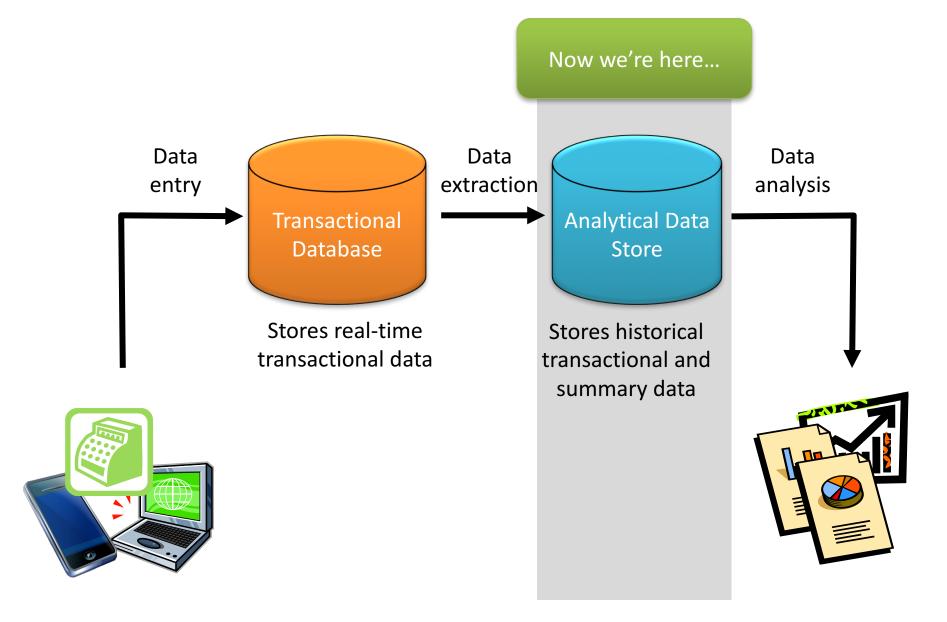


# MIS2502: Data Analytics Dimensional Data Modeling

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## Where we are...





## What do we know so far?

Why are relational databases good for storing transaction data?

Why are they bad for analytical processing?

What's the solution?

## Some terminology

## Data Warehouse

- Takes many forms
- Really is just a repository for historical data

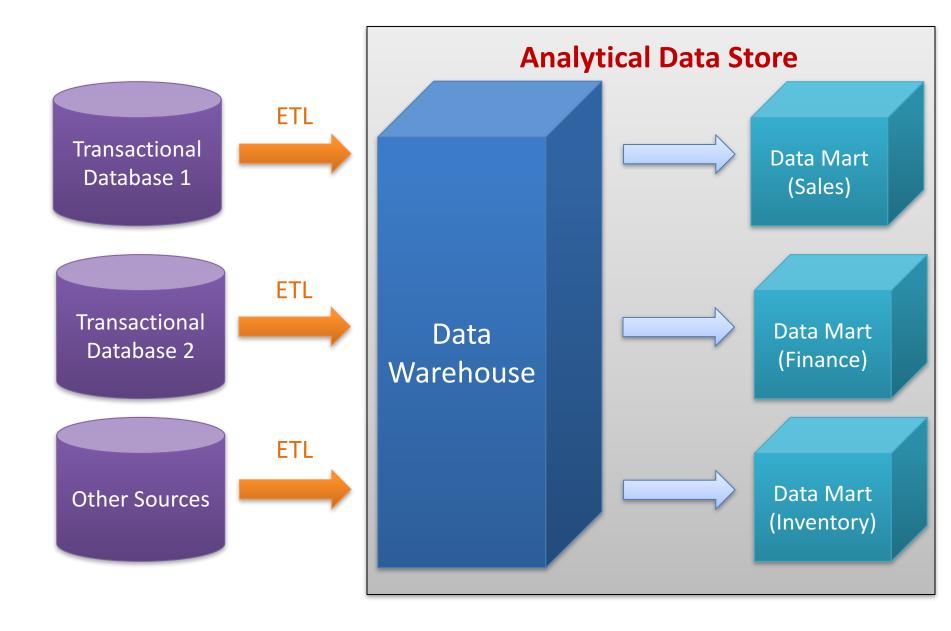
#### Data Mart

- **Subset** of the Data Warehouse
- Designed for specific analysis

#### Data Cube

- Organization of data as a "multidimensional matrix"
- Implementation of a Data Mart

## The Actual Process



## How they all relate

The data in the transactional database...

...is put into a data warehouse...

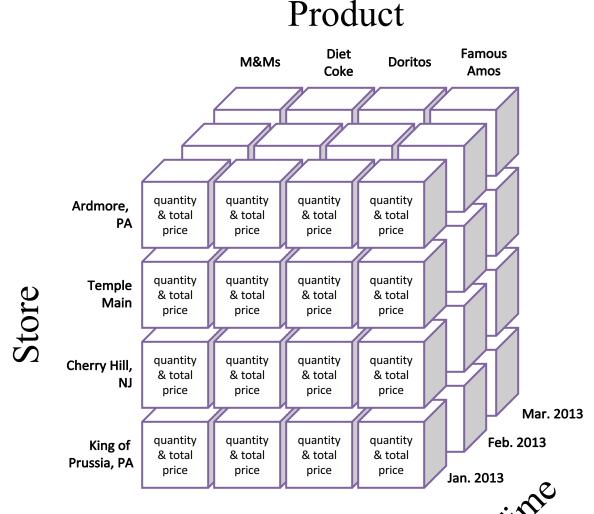
...which feeds the data mart... ...and is analyzed as a cube.

We'll start here.

#### The Data Cube

 Core component of Online Analytical Processing (OLAP) and Multidimensional Data Analysis

 Made up of "facts" and "dimensions"



Quantity sold and total price are measured facts.

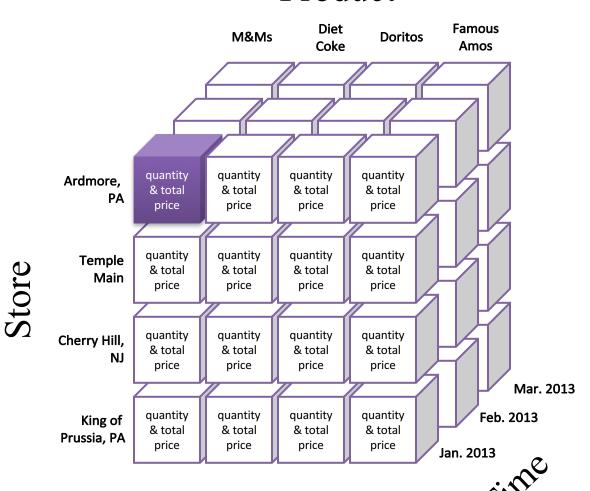
Why isn't product price a measured fact?

## The Data Cube

The highlighted element represents all the M&Ms sold in Ardmore, PA in January, 2013

A single summary record representing a business event (monthly sales).

#### **Product**

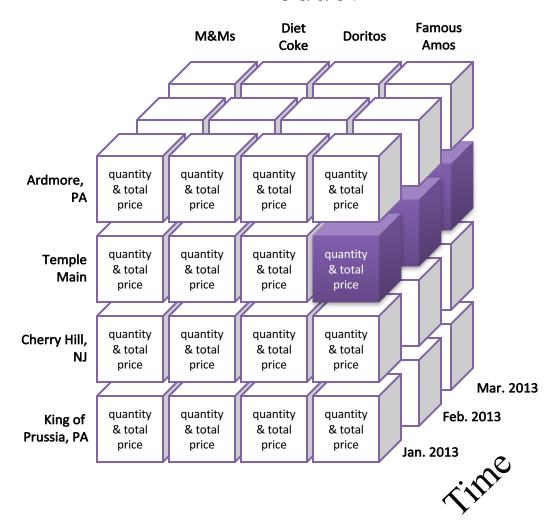


## Slicing the Data

The highlighted elements represent Famous Amos cookies sold on Temple's Main campus from January to March, 2013

This is called "SLICING the data."

#### **Product**



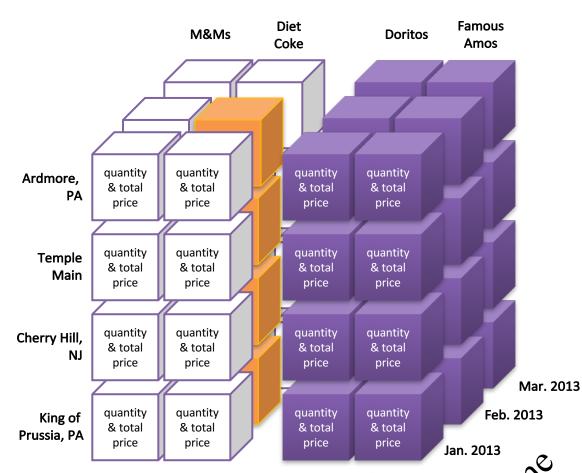
## Dicing the Data

#### **Product**

What do the orange highlighted elements represent?

What do the purple highlighted elements represent?

Store



This is called "DICING the data"



## Modeling a data cube: The Star Schema

Transactional databases aren't built around dimensions

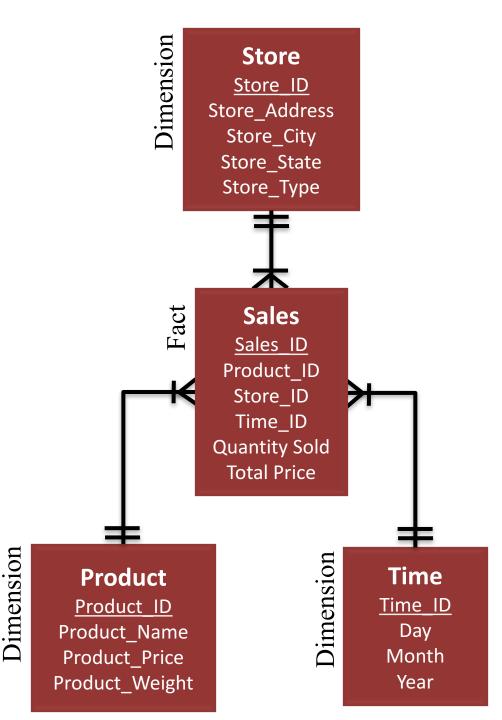
- They don't map well to cubes
- They aren't set up for summarization

#### So we build a star schema

- Built around "dimensions" and "facts"
- Simplified relational model

#### The star schema facilitates

- Aggregating individual transactions
- Creation of cubes



#### Fact Table

- Contain facts (numeric measurements) associated with a specific business process
- Contain foreign keys that refer to dimension tables

Sales

Sales ID

Product\_ID

Store\_ID

Time\_ID

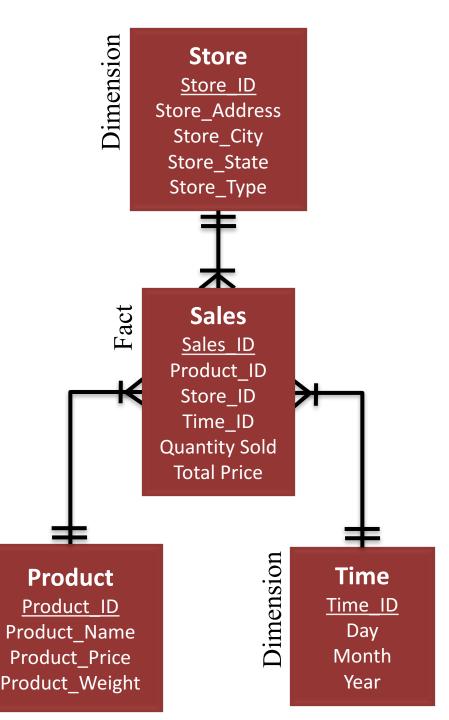
Quantity Sold

Total Price

#### **Dimension Tables**

Contain text and descriptive information

 Provide the "who, what, where, when, why, and how" context surrounding a business process event

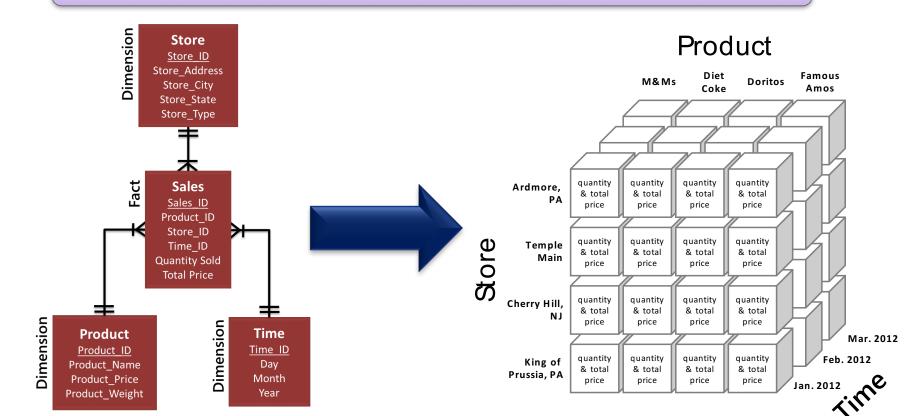


#### From Star Schema to Data Cube

A Cube typically uses a Star Schema as its source

and stores precomputed summarized (aggregated) data

Much more efficient, but can't be changed (non-volatile)



## Designing the Star Schema

Kimball's Four Step Process for Data Cube Design (Kimball et al., 2008)

Choose the business process

2. Identify the fact

3. Decide on the level of granularity

4. Identify the dimensions

## Choose the business process

- What your data cube is "about"
- Determined by the questions you want to answer about your organization

Question	<b>Business Process</b>
Who is my best customer?	Sales
What are my highest selling products?	Sales
Which teachers have the best student performance?	Standardized testing
Which supplier is offering us the best deals?	Purchasing

Note that a "business process" is not always about business.

## Identify the fact

The fact table contains data associated with the business process event

#### Keys

- Primary key for each event
- Foreign keys for the associated dimensions

 Example: Sales has Sales\_ID as primary key, and Product\_ID, Store\_ID, and Time\_ID as foreign keys

#### Measured, numeric data

- Facts: Quantifiable information for each business event
- Describes a particular combination of dimensional data
- Example: Sales has quantity\_sold and total\_price.

Try it for the "student performance" example.

## Decide on the level of granularity

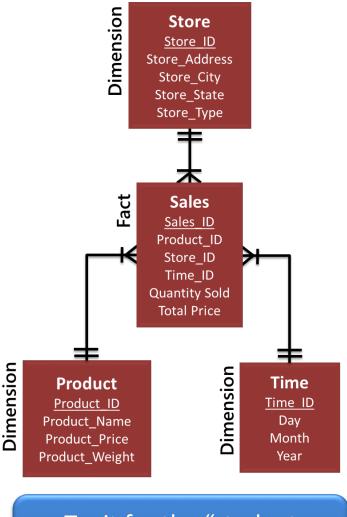
- Level of detail for each event (row in the table)
- Will determine the data in the dimensions
- Example: Who is my best customer?
  - The "event" is a sales transaction
  - Choices for time: yearly, quarterly, monthly, daily
  - Choices for store: store, city, state



How would you select the right granularity?

## Identify the dimensions

- Description of the context of the business process
- The key elements of the process needed to answer the question ("fact")
  - who, what, where, when, why, and how
- Example: Sales transaction
  - A "sale" is the fact
  - Occurs for a particular product, store, and time
  - Could this data mart tell you
    - What is the best selling product?
    - Who is the best customer?



Try it for the "student performance" example.

## Advantages of data cube

Speed

 Fast response to give you the information you have previously designed in the cube

**Analysis** 

 The data multi-dimensional data structure allows the data to be analyzed in the most logical way.

### Data cube caveats

- The cube is "non volatile," so you're locked in
  - Measured facts
  - Dimensions
  - Granularity

- So choose wisely!
  - For example: You can't track daily sales if "date" is monthly
  - So why not include every single sale and do no aggregation?

## Pivot tables in Excel

- PivotTable is a data summarization tool in Excel
  - the easiest way to learn multidimensional data and generate simple reports

 Data cubes can act as the data source for Pivot Table in Excel

## Summary

Data warehouse vs. data mart vs. data cube

Data Cube

Star schema

Kimball's four step process for data mart design

Pivot tables in Excel