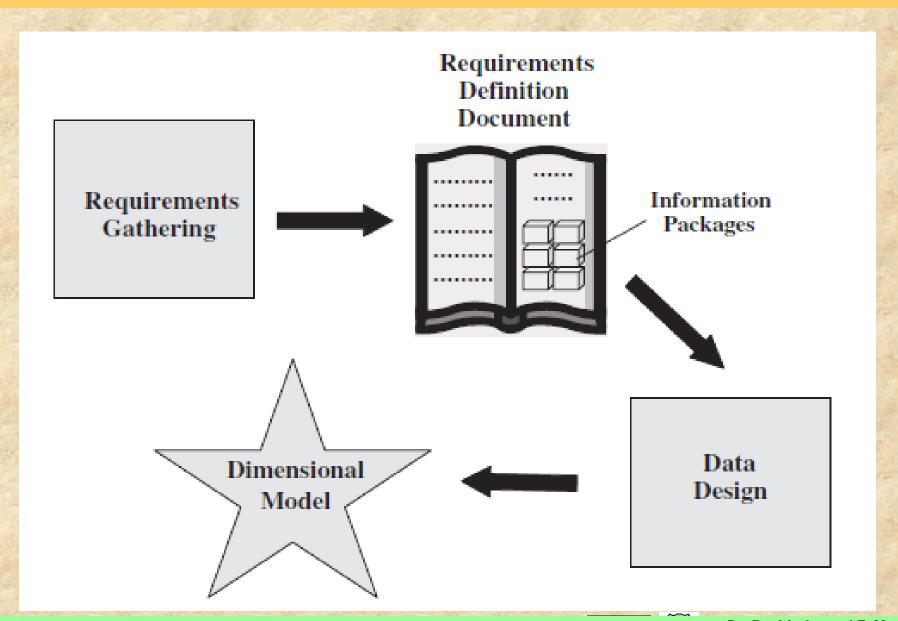


Data Warehouse Design Process

- Top-down, bottom-up approaches or a combination of both
 - Top-down: Starts with overall design and planning (mature)
 - Bottom-up: Starts with experiments and prototypes (rapid)
- From software engineering point of view
 - Waterfall: structured and systematic analysis at each step before proceeding to the next
 - Spiral: rapid generation of increasingly functional systems, short turn around time, quick turn around
- Typical data warehouse design process
 - Choose a business process to model, e.g., orders, invoices, etc.
 - Choose the <u>grain</u> (atomic level of data) of the business process
 - Choose the dimensions that will apply to each fact table record
 - Choose the measure that will populate each fact table record

FROM REQUIREMENTS TO DATA DESIGN



Dimensional Modeling Basics

- logical design technique to structure the business dimensions and the metrics
- based on a multidimensional data model which views data in the form of a data cube
- Terminology:
 - Dimension tables: subjects
 - Fact table : units/metrics

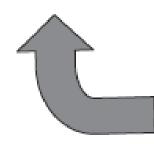
Example

Dimensions

Automaker Sales

Fact Table

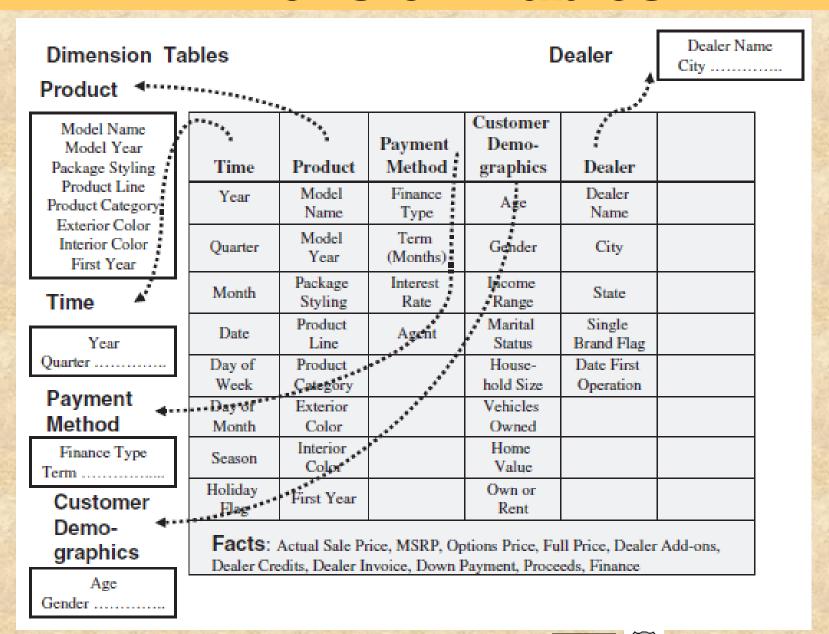
Actual Sale Price
MSRP
Options Price
Full Price
Dealer Add-ons
Dealer Credits
Dealer Invoice
Down Payment
Proceeds Finance



Time	Product	Payment Method	Customer Demo- graphics	Dealer	
Year	Model Name	Finance Type	Age	Dealer Name	
Quarter	Model Year	Term (Months)	Gender	City	
Month	Package Styling	Interest Rate	Income Range	State	
Date	Product Line	Agent	Marital Status	Single Brand Flag	
Day of Week	Product Category		House- hold Size	Date First Operation	
Day of Month	Exterior Color		Vehicles Owned		
Season	Interior Color		Home Value		
Holiday Flag	First Year		Own or Rent		

Facts: Actual Sale Price, MSRP, Options Price, Full Price, Dealer Add-ons, Dealer Credits, Dealer Invoice, Down Payment, Proceeds, Finance

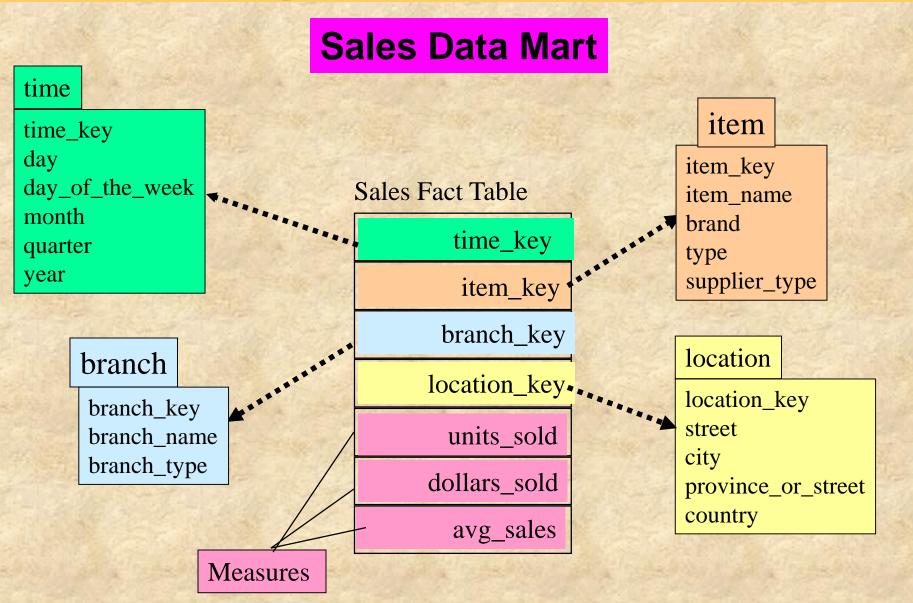
Dimension Tables



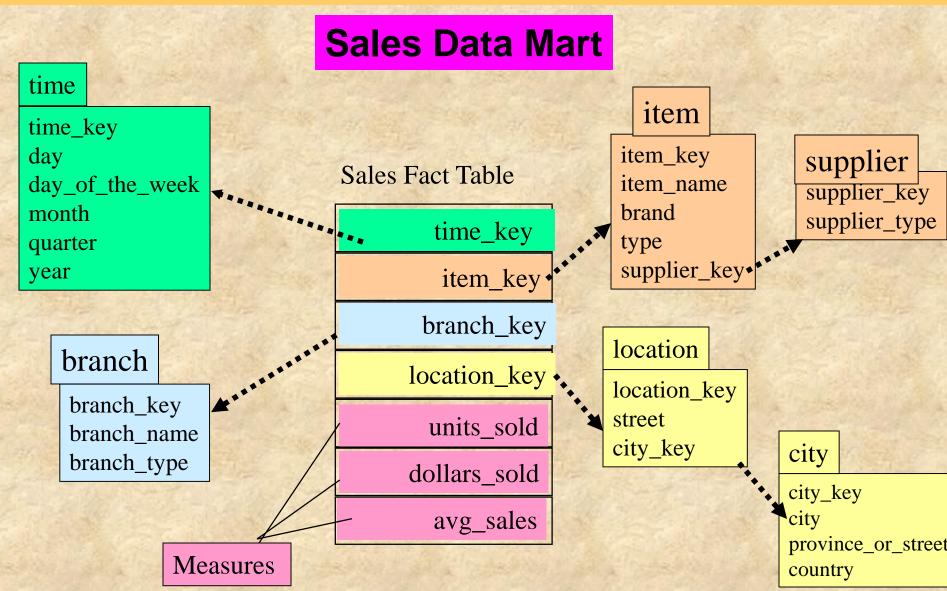
Conceptual Modeling of Data Warehouses

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

Example of Star Schema

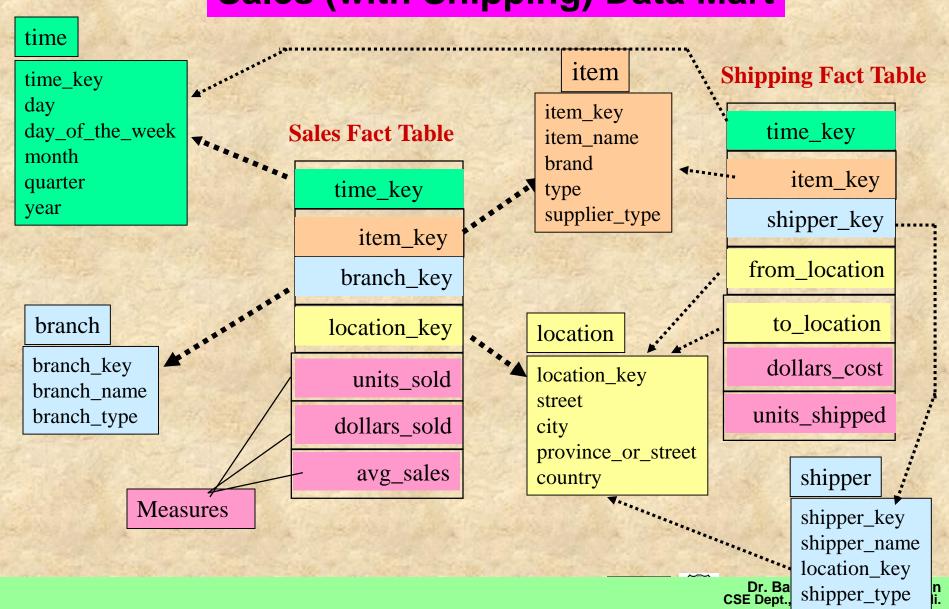


Example of Snowflake Schema



Example of Fact Constellation

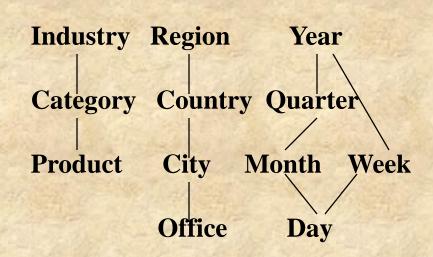
Sales (with Shipping) Data Mart



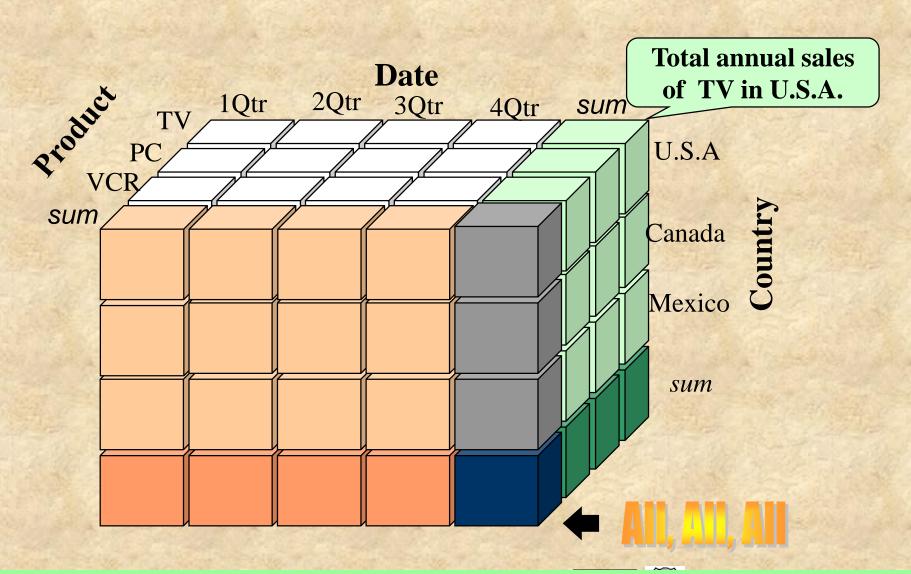
Multidimensional Data: Cube

 Sales volume as a function of product, month, and region **Dimensions: Product, Location, Time**

Hierarchical summarization paths Product Month



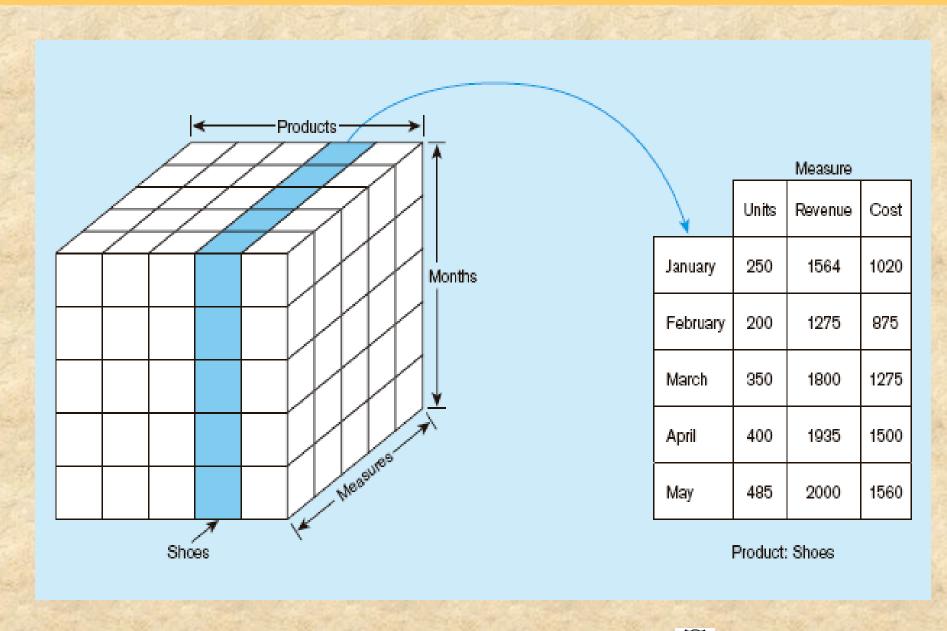
A Sample Data Cube



Typical Cube Operations

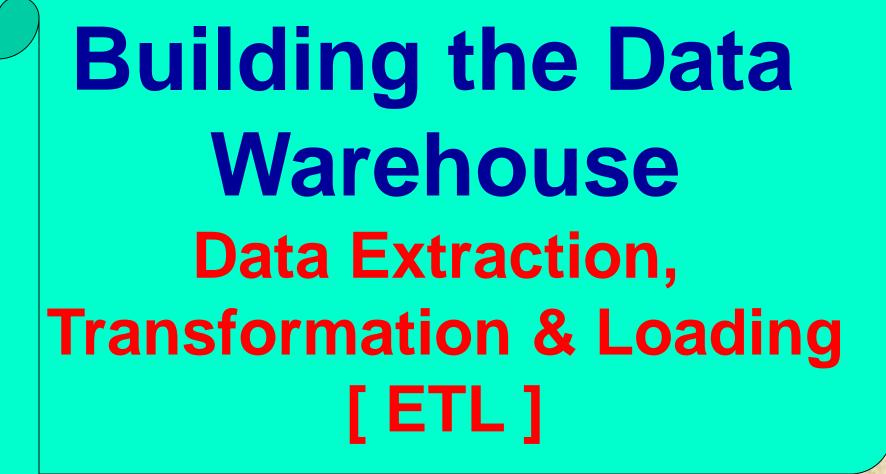
- Roll up (drill-up): summarize data
 - by climbing up hierarchy or by dimension reduction
- Drill down (roll down): reverse of roll-up
 - from higher level summary to lower level summary or detailed data, or introducing new dimensions
- Slice and dice:
 - project and select
- Pivot (rotate):
 - reorient the cube, visualization, 3D to series of 2D planes.
- Other operations
 - drill across: involving (across) more than one fact table
 - drill through: through the bottom level of the cube to its back-end relational tables (using SQL)

Example: Slicing a data cube



Exercise 2.1

Design a STAR schema for a retail company to track the sales units and sales dollars with three dimension tables.



Data Warehouse Back-End Tools and Utilities

Data extraction:

get data from multiple, heterogeneous, and external sources

Data cleaning:

- detect errors in the data and rectify them when possible

Data transformation:

convert data from legacy or host format to warehouse format

Load:

 sort, summarize, consolidate, compute views, check integrity, and build indicies and partitions

· Refresh:

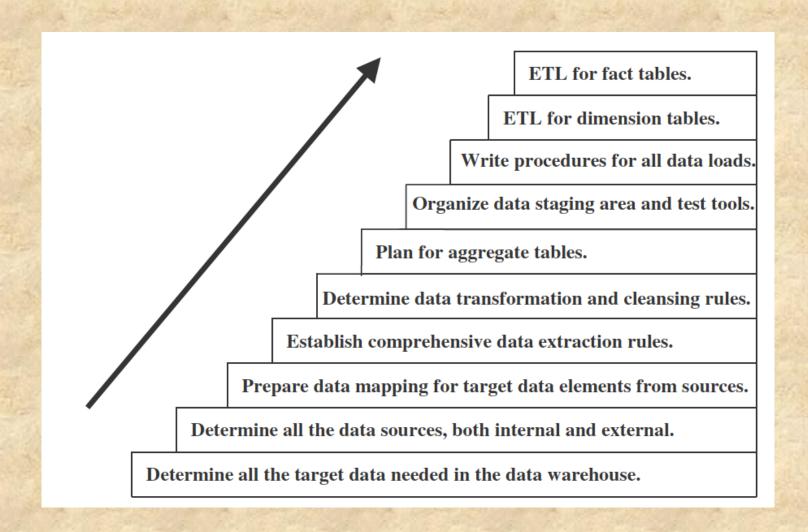
propagate the updates from the data sources to the warehouse

Extract, Transform, Load (ETL)

- Extract only relevant data from the internal source systems or external systems, instead of dumping all data ("data junkhouse")
- The ETL completion can take up to 50-70% of your total effort while developing a data warehouse.
- These ETL efforts depends on various factors



Major steps in ETL



Data Extraction

- Data can be extracted using third party tools or in-house programs or scripts
- Data extraction issues:
- 1. Identify sources
- 2. Method of extraction for each source (manual, automated)
- 3. When and how much frequently data will be extracted for each source
- 4. Time window
- 5. Sequencing of extraction processes



How data is stored in operational systems

- Current value: Values continue to changes as daily transactions are performed. We need to monitor these changes to maintain history for decision making process, e.g., bank balance, customer address, etc.
- Periodic status: sometimes the history of changes is maintained in the source system



Example

VALUES OF ATTRIBUTES AS STORED IN EXAMPLES OF ATTRIBUTES OPERATIONAL SYSTEMS AT DIFFERENT DATES

Storing Current Value

Attribute: Customer's State of Residence

6/1/2000 Value: OH

9/15/2000 Changed to CA

1/22/2001 Changed to NY

3/1/2001 Changed to NJ 6/1/2000

9/15/2000

1/22/2001

3/1/2001

OH

6/1/2000 RE

CA

NY

NJ

Storing Periodic Status

Attribute: Status of Property consigned

to an auction house for sale. 6/1/2000

6/1/2000 Value: RE (property receipted)

9/15/2000 Changed to ES (value estimated)

1/22/2001 Changed to AS (assigned to auction)

3/1/2001 Changed to SL (property sold)

9/15/2000

6/1/2000 RE 9/15/2000 ES

6/1/2000 RE 9/15/2000 ES 1/22/2001 AS

1/22/2001

6/1/2000 RE 9/15/2000 ES 1/22/2001 AS 3/1/2001 SL

3/1/2001



Data Extraction Method

Static data extraction:

- 1. Extract the data at a certain time point.
- It will include all transient data and periodic data along with its time/date status at the extraction time point
- 3. Used for initial data loading
- Data of revisions
- 1. Data is loaded in increments thus preserving history of both changing and periodic data

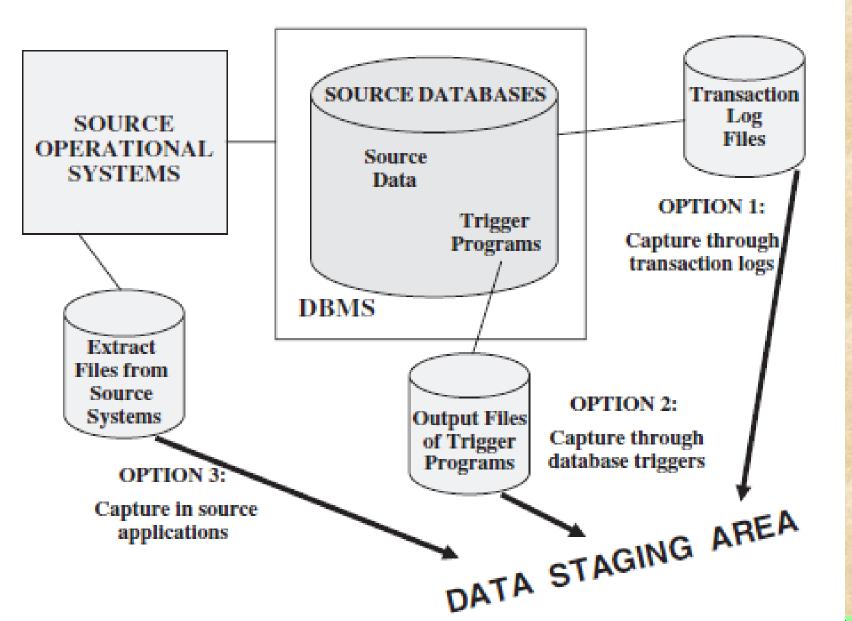


Incremental data extraction

- Immediate data extraction: involves data extraction in real time.
- Possible options:
- 1. Capture through transactions logs
- 2. Make triggers/Stored procedures
- 3. Capture via source application
- 4. Capture on the basis of time and date stamps
- 5. Capture by comparing files



Options for Immediate Extraction



Data Transformation

- Transformation means to integrate or consolidate data from various sources
- Major tasks:
- 1. Format conversions (change in data type, length)
- 2. Decoding of fields (1,0 → male, female)
- Calculated and derived values (units sold, price, cost→ profit)
- 4. Splitting of single fields (House No 11, ABC Road, Sangli, Maharashtra State, INDIA)
- 5. Merging of information (information from different sources regarding any entity, attribute)
- 6. Character set conversion



Data Transformation (Cont.)

- 8. Conversion of unit of measures
- 9. Date/time conversion
- 10. Key restructuring
- 11. De-duplication
- 12. Entity identification
- 13. Multiple source problem



Data Loading

- Determine when (time) and how (as a whole or in chunks) to load data
- Four modes to load data
- Load: removes old data if available otherwise load data
- 2. Append: The old data is not removed, the new data is appended with the old data
- 3. Destructive Merge: If primary key of the new record matched with the primary key of old record then update old record
- 4. Constructive Merge: If primary key of the new record matched with the primary key of old record then do not update old record just add the new record and mark it as superseding record

Refresh / Update

After the initial load, the data warehouse is kept up-to-date by

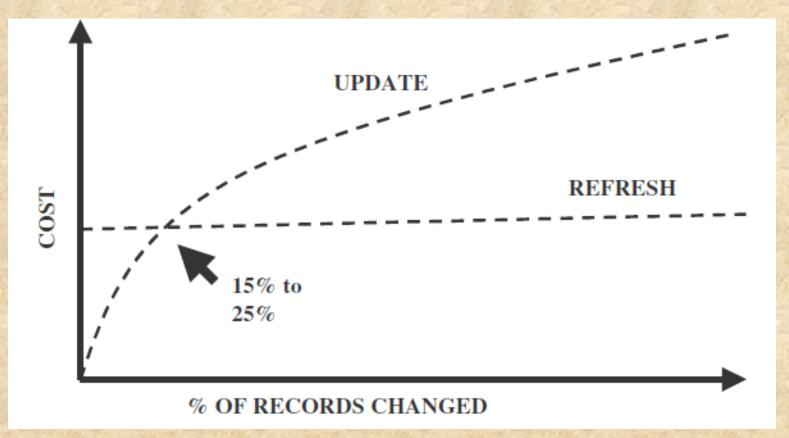
REFRESH - complete reload at specified intervals

UPDATE - application of incremental changes

Data Loading (Cont.)

Data Refresh Vs. Data Update

Full refresh reloads whole data after deleting old data and data updates are used to update the changing attributes



Data Loading (Cont.)

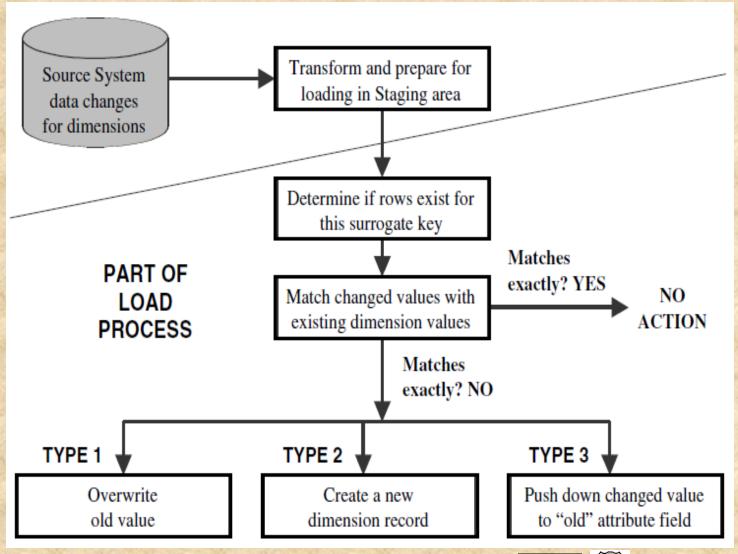
Loading for dimensional tables:

You need to define a mapping between source system key and system generated key in data warehouse, otherwise you will not be able to load/update data correctly



Data Loading (Cont.)

Updates to dimension table



Loading Fact Table

- Concatenation of the keys of dimensional table
- Load dimension records first
- Create concatenated key for the fact table record from the keys of the corresponding dimension record
- History load:
 - Loads historical data useful and interesting
- Incremental load:
 - Load as frequently as possible
- Use partitioned files/indexes, parallel processing

ETL Summary

DATA EXTRACTION

Extraction from heterogeneous source systems and outside sources.

DATA INTEGRATION

Combining all related data from various sources based on source-to-target mapping.

DATA SUMMARIZATION

Creating aggregate datasets based on predefined procedures.

METADATA UPDATES

Maintain and use metadata for Extraction, Transformation, and Load functions.

DATA TRANSFORMATION

Conversion and restructuring according to transformation rules.

DATA CLEANSING

Scrubbing and enriching according to cleansing rules.

INITIAL DATA LOADING

Apply initial data in large volumes to the warehouse.

ONGOING LOADING

Apply ongoing incremental loads and periodic refreshes to the warehouse.