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4.1.1 Selection

- Takes place at the beginning of data transformation (usually forms part of the extraction function itself).
- You select either whole records or parts of several records from the source systems.
- In some cases, the composition of the source structure may not be amenable for selection of the necessary parts during data extraction.
 - It is prudent to extract the whole record and then do the selection as part of the transformation function.

4.1.2 Conversion

- This is an **all-inclusive task**.
- It includes a large variety of rudimentary conversions of single fields for two primary reasons:
 - 1. To standardize among the data extractions from disparate source systems,
 - 2. To make the fields usable and understandable to the users.

4.1.3 Summarization

- Sometimes, it is not feasible to keep data at the lowest level of detail in your DW.
- May be none of DW users need data at the lowest granularity for analysis or querying.
 - For example, for a grocery chain, sales data at the lowest level of detail for every transaction at the checkout may not be needed. Storing sales by product by store by day in the DW may be quite adequate.
 - Here, data transformation function includes
 summarization of daily sales by product and by store.

4.1.4 Enrichment

 This task is the rearrangement and simplification of individual fields to make them more useful for DW.

 You may use one or more fields from the same input record to create a better view of the data for the DW.

4.1.5 Format Revisions

 Revisions include changes to the data types and lengths of individual fields.

– Product package types may be indicated by codes and names in which the fields are numeric and text data types.

– Again, the lengths of the package types may vary among the different source systems. It is wise to standardize and change the data type to text to provide values meaningful to the users.

4.1.6 Decoding of Fields

- When you have the same data items described by a plethora of field values.
 - Gender \rightarrow M, F or 1,2
 - Department \rightarrow IS, CS
- You need to decode all such cryptic codes and change these into values that make sense to the users..

4.1.7 Calculated and Derived Values

Depending on DW users and their needs.

- Calculated and derived column produce new data to enrich query and data analysis.
 Quarter, Month, Week Day ← Date
 - Amount in Dinar ← Amount in Dollar/1200IQ

You should use the tools that derive the new data or column.

4.1.8 Splitting of Single Fields

- Earlier legacy systems stored names and addresses of customers and employees in large text fields.
 - Address (Basrah-Ashar-Street10).
 - Ali Sami Ali.
- You need to store individual components of names and addresses in separate fields in your DW for two reasons:
 - Improve the operating performance by indexing on individual components.
 - Your users may need to perform analysis by using individual components such as city, state, and zip code.

4.1.9 Merging of Information

- This is not the opposite of splitting of single fields.
- Merging Info. does not literally mean the merging of several fields to create a single field of data.
 - For example, information about a product may come from different data sources. The product code and description may come from one data source. The relevant package types may be found in another data source. The cost data may be from yet another source. In this case, merging of information denotes the combination of the product code, description, package types, and cost into a single entity.

4.1.10 Character set conversion

- Refers to conversion of character sets to an agreed standard character set for textual data in the DW.
 - If you have mainframe legacy systems as source systems, the source data from these systems will be in EBCDIC characters.
 - If PC-based architecture is the choice for your DW, then you must convert the mainframe EBCDIC format to the ASCII format.

4.1.11 Conv. of Units of Measurements

- Many companies today have global branches.
- Measurements in many European countries are in metric units.
- If your company has overseas operations, you may have to convert the metrics so that the numbers are all in **one standard unit of measurement**.

4.1.12 Date/Time Conversion

 This type relates to representation of date and time in standard formats.

 For example, the American and the British date formats may be standardized to an international format.

-The date of October 11, 2008 is written as 10/11/2008 in the U.S. format and as 11/10/2008 in the British format. This date may be standardized to be written as 11 OCT 2008.

4.1.13 Key Restructuring

- If the product is moved to another DW, the DW part of the product key will have to be changed.
 - Solution is **surrogate key**.
- When choosing keys for your DW database tables, avoid such keys with built-in meanings.
- The key restructuring is transform such keys into generic keys generated by the system itself.

4.1.14 Deduplication

- In many companies, the customer files have several records for the same customer.
 - Mostly, the duplicates are the result of creating additional records by mistake.
- In your DW, you want to keep a single record for one customer and link all the duplicates in the source systems to this single record.
 - This process is called **deduplication** of the customer file.
 - Employee files and, sometimes, product master files have this kind of duplication problem.

4.2 Data Integration and Consolidation



4.2 Data Integration and Consolidation

- A challenge is the pulling together of all the source data from many disparate, dissimilar source systems.
- Data Integration problems occur when many data sources from different platforms used to implement DW.
- Sources do not conform the same set of business rules (naming conventions and varied standards for data representation)
- Two problems affect the data integration (Entity Identification and Multiple Sources).

4.2.1 Entity Identification Problem

 You have to design complex algorithms to match records from all the source files and form groups of matching records.

No matching algorithm can completely determine the groups.

 If the matching criteria are too tight, then some records will escape the groups.

4.2.2 Multiple Sources Problem

• Less common and less complex than the entity identification problem.

 This problem results from a single data element having more than one source.

 A straightforward solution is to assign a higher priority to one of the two sources and pick up the other sources.

4.3 Transformation for Dimension Attributs

- Three ways to handle the three types of slowly changing dimensions.
 - Type 1 changes are corrections of errors. These changes are applied to the DW without any need to preserve history.
 - Type 2 changes preserve the history in the DW.
 - Type 3 changes are tentative changes where your users need the ability to analyze the metrics in both ways with the changes and without the changes.

4.3 Transformation for Dimension Attributs



5. Data Loading

- The whole process of moving data into the DW repository is referred to in several ways (applying the data, loading the data, and refreshing the data).
- During the loads, the **DW** has to be **offline**.
- You need to find a window of time when the loads may be scheduled without affecting your DW users.
- Data may be applied in the following four different modes (Load, Append, Destructive merge, Constructive merge).

5.1 Load Modes



5.1.1 Load

 If the target table to be loaded already exists and data exists in the table, the load process wipes out the existing data and applies the data from the incoming file.

 If the table is already empty before loading, the load process simply applies the data from the incoming file.

5.1.2 Append

- If data already exists in the table, the append process unconditionally adds the incoming data, preserving the existing data in the target table.
- When an incoming record is a duplicate of an already existing record ,you may define how to handle an incoming duplicate.
- The incoming record may be allowed to be added as a duplicate or may be rejected during the append process.

5.1.3 Destructive Merge

- Merge In this mode, you apply the incoming data to the target data.
- If the primary key of an incoming record matches with the key of an existing record, update the matching target record.
- If the incoming record is a new record without a match with any existing record, add the incoming record to the target table.

5.1.4 Constructive Merge

- This mode is slightly different from the destructive merge.
- If the primary key of an incoming record matches with the key of an existing record, leave the existing record, add the incoming record, and mark the added record as superseding the old record.

5.2 Load Stages

- Load stages refer to what you intend to do with data once it's loaded to DW.
- Load stages include:
 - Initial Load
 - Incremental Load
 - Full Refresh

5.2.1 Initial Load

- Bring the data from staging area into DW.
- Takes time and memory intensive due to size of data.
- Must divide the load process, schedule the loads and use a specific mode.

5.2.2 Incremental Load

- Detect and load the ongoing changes from the source systems.
- Changes to the source systems are always tied to specific times, irrespective of whether or not they are based on explicit time stamps in the source systems (need method).
- If the time stamp is also part of the primary key or if the time stamp is included in the comparison between the incoming and the existing records, then constructive merge may be used to preserve the periodic nature of the changes.

5.2.3 Full Refresh

- Involves periodically rewriting the entire DW.
 - Sometimes, you may also do partial refreshes to rewrite only specific tables.
- Partial refreshes are rare because every dimension table is intricately tied to the fact table.
- In the case of full refreshes, data exists in the target tables before incoming data is applied.
- The existing data must be erased before applying the incoming data.
 - Just as in the case of the initial load, the load and append modes are applicable to full refresh.

5.3 Update VS Refresh

- Update application of incremental changes in the data sources.
- Refresh complete reload at specified intervals.
- Refresh is a much simpler option than update.
- To use the update option, you have to devise the proper strategy to extract the changes from each data source.
- Then you have to determine the best strategy to apply the changes to the DW.

5.3 Update VS Refresh

- The refresh option simply involves the periodic replacement of complete DW tables.
- But refresh jobs can take a long time to run.
- If you have to run refresh jobs every day, you may have to keep the DW down for unacceptably long times.
- The case worsens if your database has large tables.

5.4 Dimension and Fact Table Load

- The key of the fact table is the concatenation of the keys of the dimension tables.
 - Therefore, for this reason, dimension records are loaded first.
- You have to create the concatenated key for the fact table record from the keys of the corresponding dimension records.
- Perform fact table surrogate key look-up.

End of ETL-2