

Data Warehouse

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| Data Warehouse, 2025

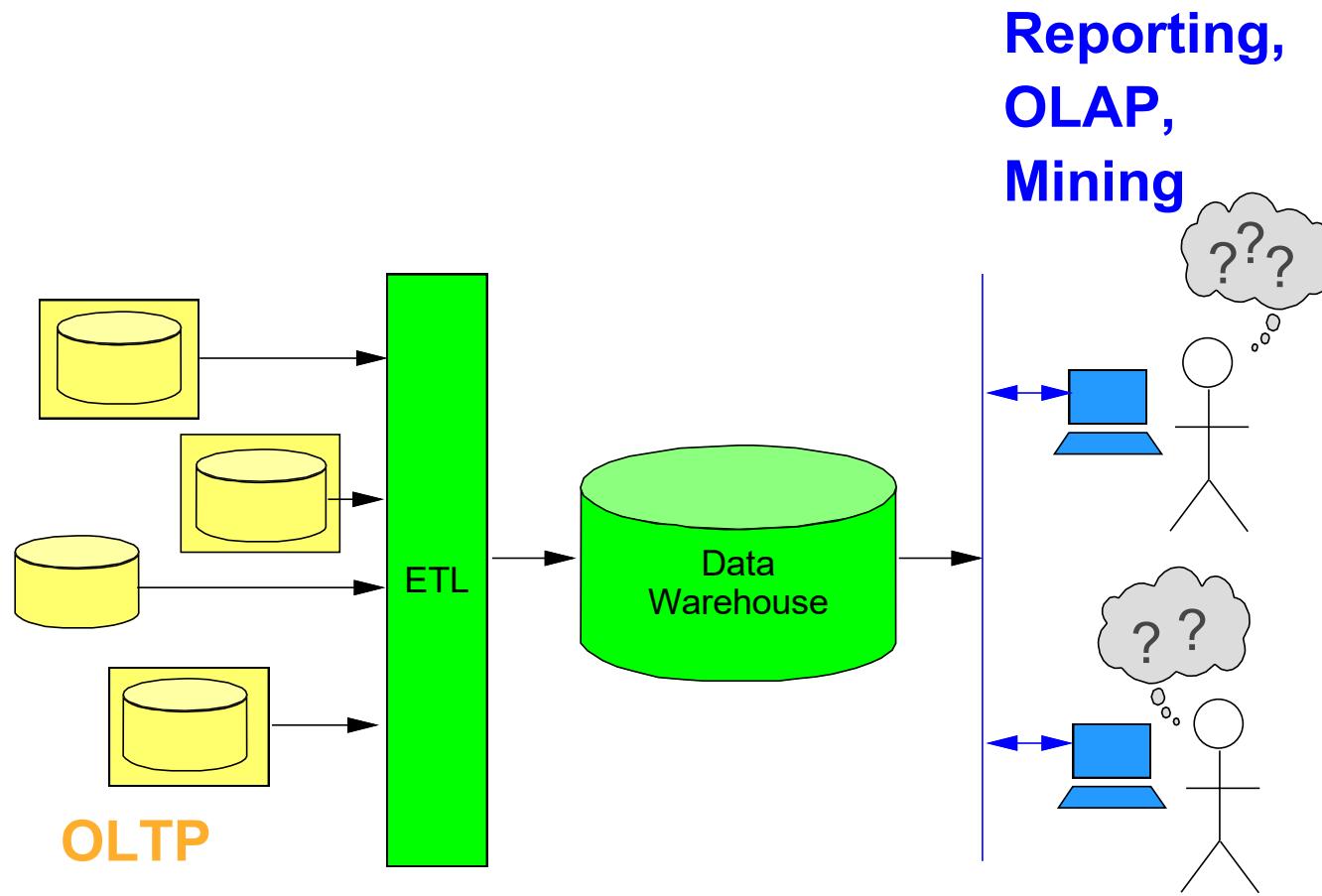
Extract –Transform - Load

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TIMETOACT
GROUP

Basic Data Warehouse architecture

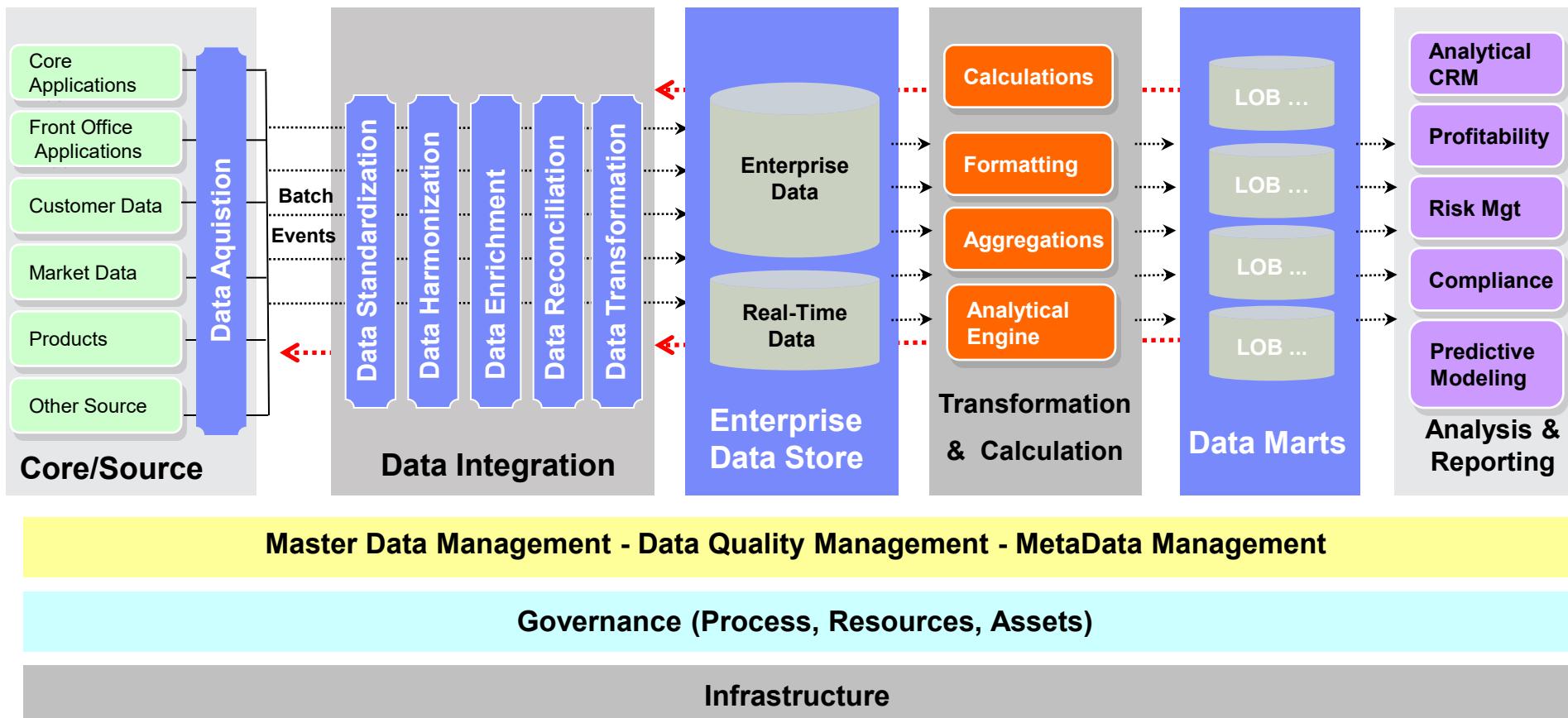


Tasks of the ETL Process

- Monitor changes in source systems
- Extract
 - ▶ capture and copy data from source systems (e.g. operational systems)
 - ▶ many different types of sources
 - Relational, hierarchical DBMSs
 - Flat files
 - Legacy systems
 - ERP systems
 - Other internal/external sources
- Transform
 - ▶ Filter data
 - ▶ Check and cleanse data
 - ▶ Transform data to CDW data model
- Load
 - ▶ fast load into staging area/CDW tables

Data Integration Architecture

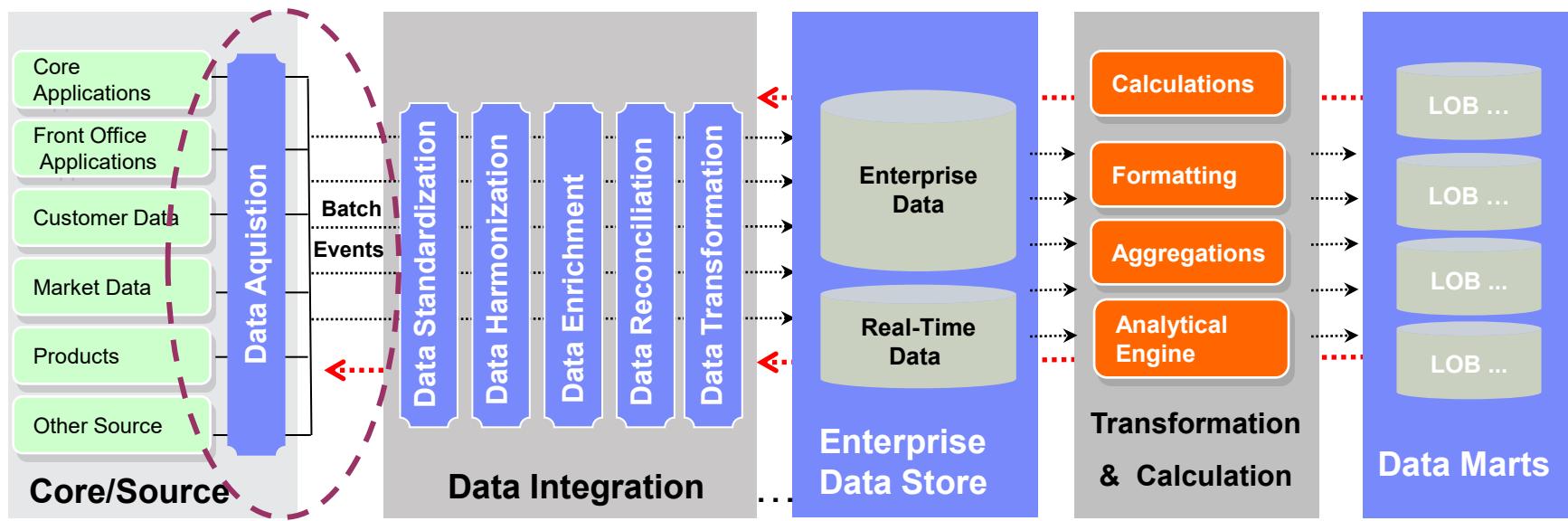
Data-Driven Reference Architecture



Prerequisite of ETL - Understanding The Data

- Profile Existing Data Sources, Extracted Data
 - ▶ Analyze data structure, content, and quality
 - ▶ Find data relationships across systems
 - ▶ Uncover data issues that can affect subsequent transformation steps
 - missing values
 - Duplicates
 - inconsistencies

Data Integration Process



- **Data Acquisition**

- ▶ Support data acquisition in the source systems with minimal impact into transaction systems
- ▶ Capture of deltas for batch and real-time processing
- ▶ Depends on data latency requirements

Monitoring

- Extracts from source systems
 - ▶ Intial extract for setting up the data warehouse
 - ▶ Periodical extracts for adding new/changed information to the data warehouse
 - Question: How to determine what is new or what has changed in the source systems?
- Task of „monitoring“

Monitor types

- The monitor determines all changes itself
 - ▶ the result of monitoring is a delta-file containing all changes
- The monitor determines only if something has changed (not what)
 - ▶ The changes have to be determined by the extraction-component

Aspects of monitoring

- Discovery of all changes vs. determining the net effect at extract/load time only
 - ▶ Example: an attribute value can be changed in two ways:
 - by one update operation
 - by one delete and one insert operation
 - ▶ The net effect of both is the same
 - ▶ However, history information is lost if the net effect is recorded only.
- Notification vs. Polling
 - ▶ Data sources notify monitor about changes (e.g. through triggers) or the monitor determines the changes itself
 - ▶ Frequency of polling
 - High → high load on source systems
 - Low → history information may be lost

Aspects of monitoring

- Internal vs. External monitoring
 - ▶ Internal:
the source systems offers sufficient hooks for determining the changes
 - E.g. Notification about changes
 - ▶ External:
the changes have to determined from outside the source systems
 - E.g.; by comparinig consequutive extracts

Monitoring techniques

- Depend on characteristics of the data sources
 - ▶ based on modern relational DBMS
 - ▶ ...
 - ▶ legacy systems with proprietary data storage
- Types of techniques
 - ▶ Based on DBMS
 - Active mechanisms
 - Replication techniques
 - Protocol-based discovery
 - ▶ Controlled by application
 - Timestamp-based discovery
 - Snapshot-based discovery

Monitoring based on DBMS

- Active monitoring mechanisms
 - ▶ Based on (database) triggers
 - Example:
 - If new record is inserted in sales transaction table
then insert transaction id and timestamp in change table
 - ▶ Advantage:
 - Triggers do not change operational applications
 - ▶ Disadvantage:
 - Performance impact on operation systems if triggers are used extensively

Monitoring based on DBMS

- Replication techniques
 - ▶ Snapshot
 - Read-only, local copies of one or multiple tables of the source systems
 - Incremental update through snapshot-log
 - Snapshot log
 - contains information of changes wrt to last snapshot operation
 - implemented through internal triggers
 - The monitor can use snapshot log to determine what has changed
 - ▶ Data replication
 - Target tables (like snapshot tables), not necessarily on local system
 - Creates delta-tables with the all (committed) changes, not only net effect

Monitoring based on DBMS

- Protocol-based discovery
 - ▶ Usage of database logs to determine changes
 - DBMSs write transaction logs in order to be able to undo partially executed transactions
 - This information can be used to determine all changes
 - ▶ Requires exact knowledge of the protocol mechanisms of the DBMS
 - ▶ Protocol files can be transferred to other systems to avoid additional load on source systems

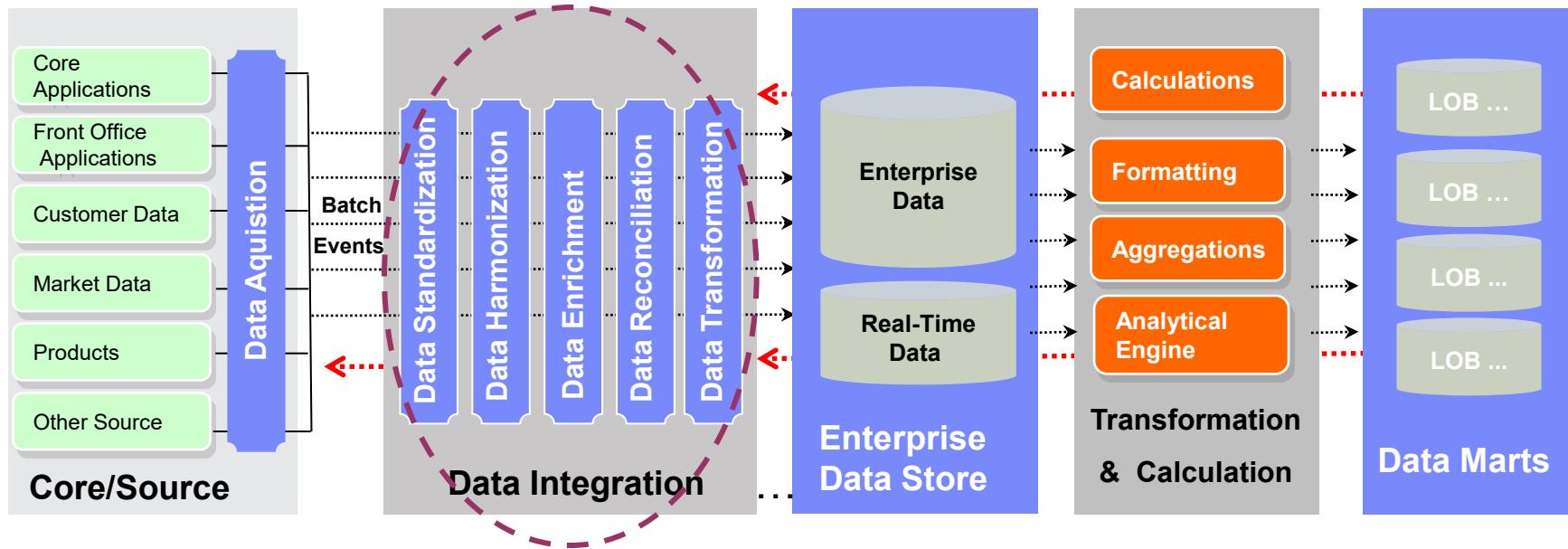
Monitoring controlled by the applications

- Approaches
 - ▶ Operational applications log themselves the changes in the operational data
 - Can require changes in the source code of the applications
 - Very difficult for legacy applications
 - Can have an impact on the performance
 - ▶ Independant monitoring applications that discover the changes in the operational applications

Application based monitoring techniques

- Timestamp based discovery
 - ▶ Every data item is associated with timestamp information about its validity period
 - ▶ Changed data can be determined from this timestamp information
 - ▶ Operational applications have to keep a limited change history
- Data comparison
 - ▶ Comparison of snapshots of the operational data at different points in time
 - ▶ Can be very complex
 - ▶ Sometimes the only possibility, for instance for legacy applications

Data Integration Process



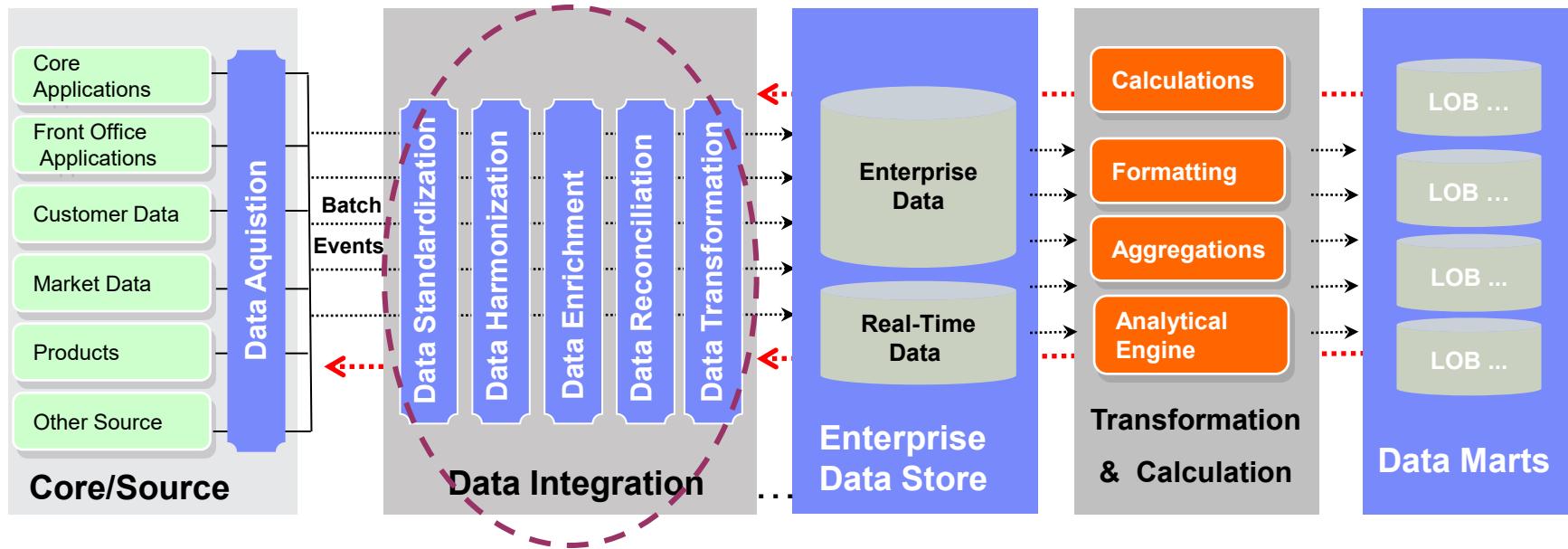
■ Data Integration

- ▶ Integrate source data into the consolidated Enterprise Data Warehouse
- ▶ Implement complex integration processes
- ▶ Building the „Single Version of Truth“

Extract

- Extraction intervals
 - ▶ Periodically – in regular intervals
 - Every day, week, etc.
 - Depends on the requirements on timeliness of the data warehouse data
 - ▶ Triggered by a specific request
 - Addition of a new product
 - Query which involves more recent data
 - ▶ Triggered by specific events
 - Number of changes in operational data exceeds threshold
 - ▶ Instantly
 - Every change is directly propagated into the data warehouse
 - „real time data warehouse“

Data Integration Process



■ Data Standardization

- ▶ Match and reconcile related data elements
- ▶ Remove duplicate, redundant data
- ▶ Reengineer data to match single corporate standard
- ▶ Quality processing to ensure data integrity

Transform - Integration of data

- Different keys for same entities in source data

AD_SYS	...	Kunde_Text	LOADTIME
AD-FX8257		Müller	31DEC1999:23:03:08
AD-FH2454		Meier	31DEC1999:23:03:08
AD-FX7059		Schulz	31DEC1999:23:03:08
AD-FT2567		Schmitz	31DEC1999:23:03:08
...

AC_SYS	Kunde_Text	Kunde_Status
3857_ACC	Müller	A
3525_ACC	Meier	A
3635_ACC	Schulz	A
3566_ACC	Schmitz	B
...

CC_SYS	Kunde_Grp	Kunde_Text	LOADTIME
59235395	Handel	Müller	31DEC1999:23:03:08
08485356	Industrie	Meier	31DEC1999:23:03:08
08555698	Industrie	Schulz	31DEC1999:23:03:08
85385386	Handel	Schmitz	31DEC1999:23:03:08
...

Kunde_ID	Kunde_Text	...	AD_SYS	CC_SYS	AC_SYS	...	LOADTIME
0001	Müller		AD-FX8257	59235395	3857_ACC		31DEC1999:23:03:08
0002	Meier		AD-FH2454	08485356	3525_ACC		31DEC1999:23:03:08
0003	Schulz		AD-FX7059	08555698	3635_ACC		31DEC1999:23:03:08
0004	Schmitz		AD-FT2567	85385386	3566_ACC		31DEC1999:23:03:08
...

Legende: AD – Außendienstsystem, CC – Call-Center-Anwendung, AC – Abrechnungs-/Accounting-System

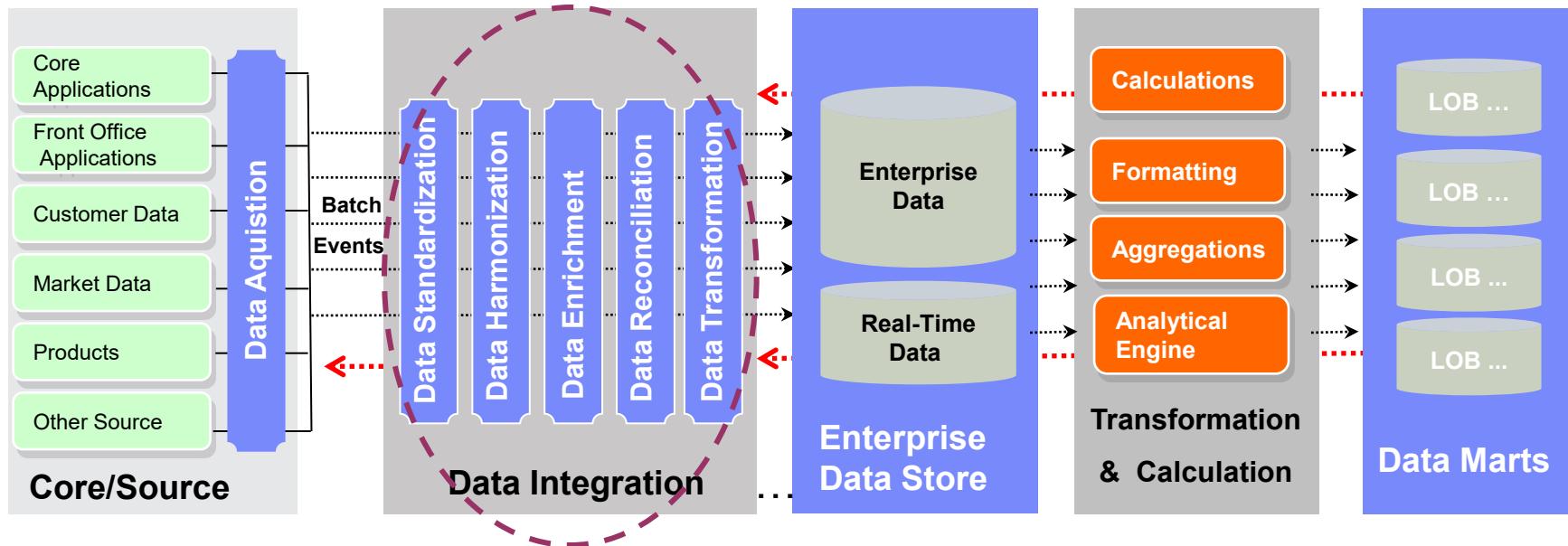
Quelle: Finger, R. (2002), Historisierungskonzepte, Vortrag im Rahmen der Seminarreihe „Data Warehouses und Data Marts – Effizienter Einsatz für das Controlling“, Frankfurt am Main 2002.

© Kemper, Mehanna, Unger: Business Intelligence, Vieweg 2004, ISBN 3-528-05802-1

Transform - Integration of data

- Different keys for same entities in source data
 - ▶ Solution
 - Creation of a unique surrogate key
 - Storing the association of surrogate keys to the keys in the operational system in a mapping table
 - ▶ Determining if different keys point to the same entity is a difficult problem
 - Different ways of how a name can be written
 - Peter Maier
 - Maier, Peter
 - P. Maier

Data Integration Process



- **Data Harmonization**
 - ▶ Standardization of data so that they can be matched with other data and information regardless of the format.
 - ▶ Consolidate, Integrate, Cleanse, Normalize & Harmonize

Transform - Integration of data

- Unification of data
 - ▶ Unification of data types
 - Character string → date
„20.01.2006“ → 20.01.2006
 - Character string → number
„12345“ → 12345
 - ▶ Unification of encodings
 - For instance for gender F and M
 - Lookup-tables contain the mapping from old to new encodings
 - ▶ Unification of character strings
 - Unique way of naming and writing certain concepts
 - Küchengerät → Kuechengeraet
 - customer → client
 - Names:
 - „last name“, „first name“ like „Maier, Peter“

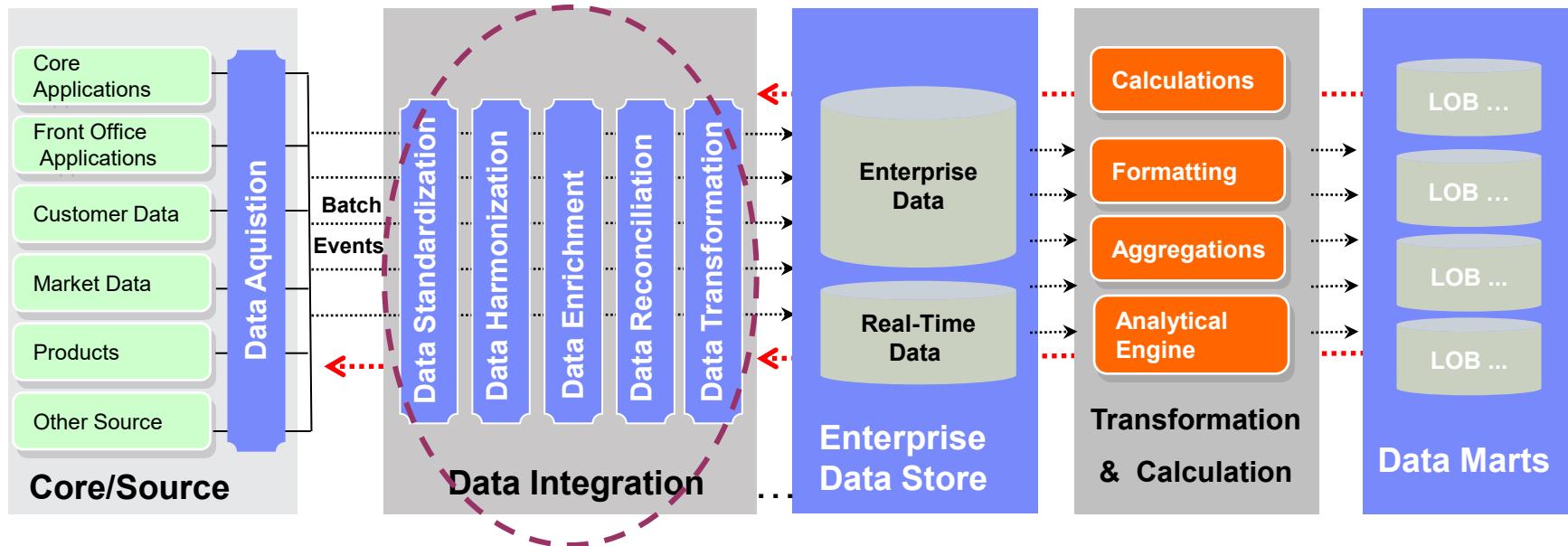
Transform - Integration of data

- Unification of data (cont.)
 - ▶ Unification of dates and timestamps
 - Rules for representing incomplete date information
 - If only month and year are known
 - Dates and timestamps wrt one specific timezone
 - Important for multi-national organizations
 - ▶ Combination of different attributes to one attributes
 - day, month, year → date
 - ▶ Split of one attribute into two or more
 - Name → first name, last name
 - Product name - „Cola, 0.33 l“ →
Product short name - „Cola“, size in liters - 0.33

Transform - Integration of data

- Unification of data (cont.)
 - ▶ Computation of derived values
 - Profit = sales price – purchase price
 - ▶ Aggregations
 - Revenue of the month/week computed from revenues of the day

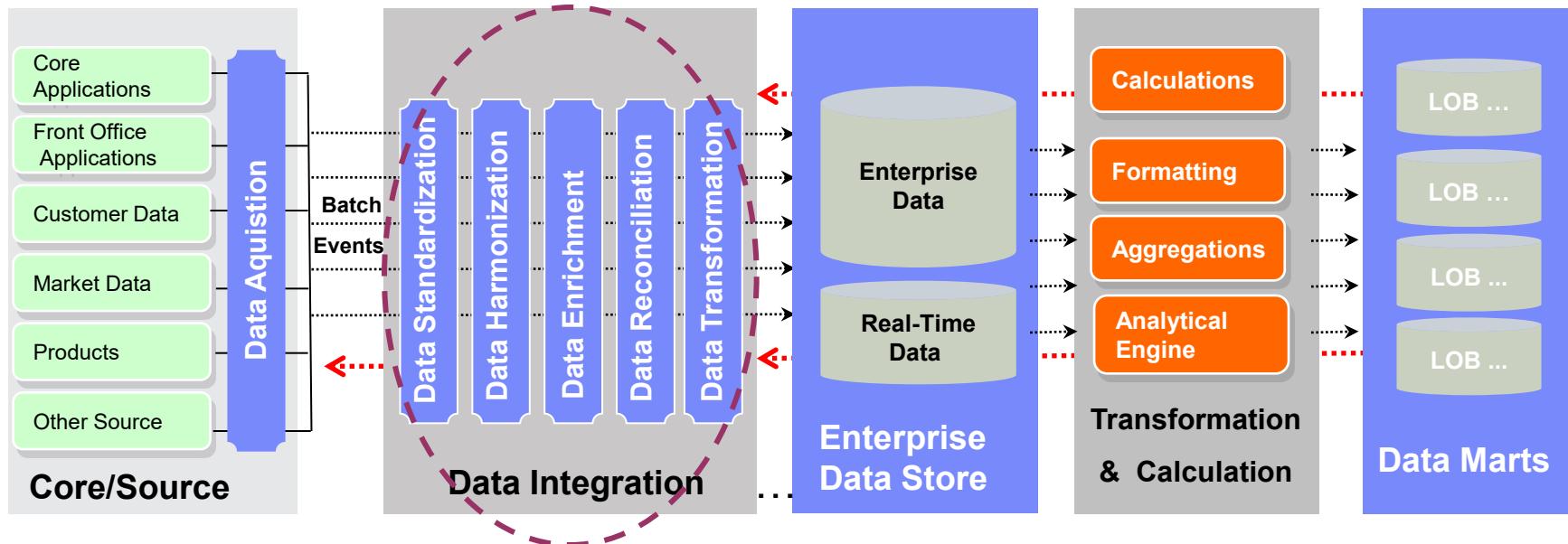
Data Integration Process



■ Data Enrichment

- ▶ Supplementing internal data with data from external sources
- ▶ Personal data such as date-of-birth and gender codes
- ▶ Geographical data
- ▶ Postal Data, Demographic information, Economic data , etc

Data Integration Process



■ Data Reconciliation

- ▶ Process to integrate and reconcile a view of data of the organization
 - Comparing Information from Multiple Sources
 - Correct one or both sources
 - Create a single source

Transform – Data cleansing

- Types of data quality problems
 - ▶ Wrong values of attributes
 - Phone number as address
 - Outliers for numeric attributes
 - Unlikely high price for a product
 - Can be discovered by statistical methods
 - ▶ Inconsistencies
 - Order date after payment date
 - Can be discovered by statistical and data mining methods
 - ▶ Multiple representations for one entity
 - Keys
 - Encodings

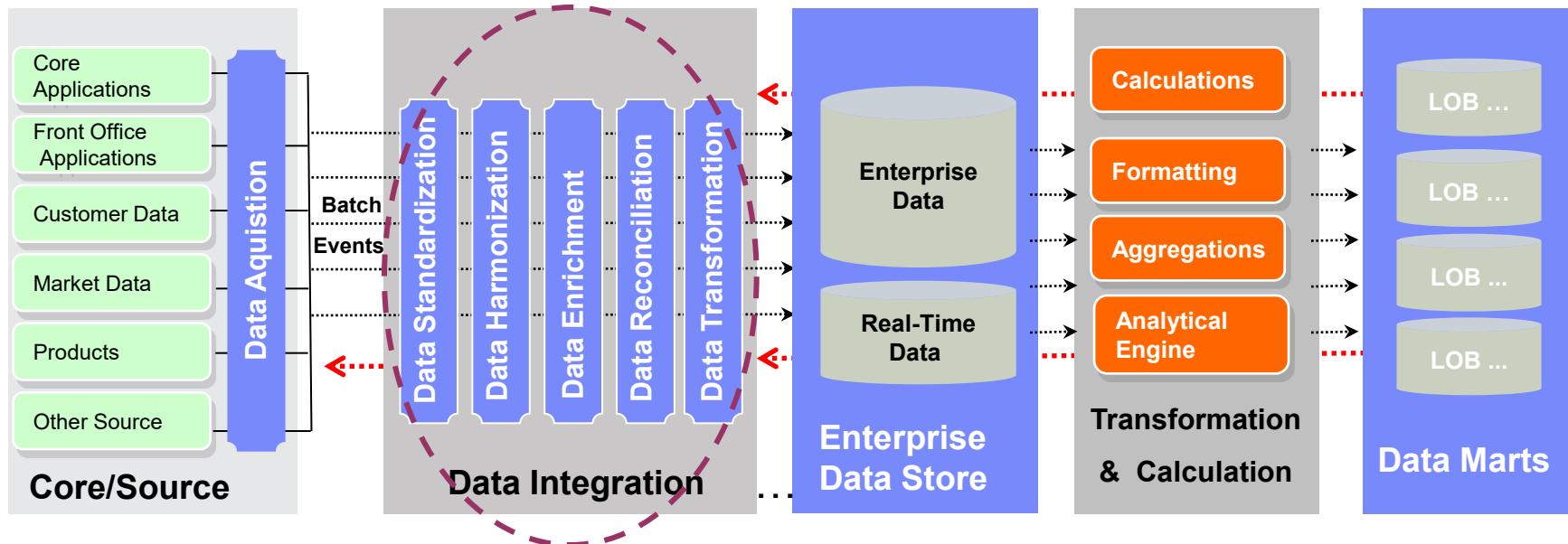
Transform – Data cleansing

- Types of data quality problems (cont.)
 - ▶ Missing values
 - In applications missing values are encoded by specific values like 01.01.1900 for unknown date
 - In databases NULL represents an unknown value
 - Missing values can represent
 - an unknown value
 - like date of birth of a customer
 - a value that does not exist
 - like „engine type“ for bicycle in a vehicles table
 - it is possible that both cases are possible
 - like e-mail address of a customer

Transform – Data cleansing

- Correcting the data
 - ▶ Automatically
 - E.g., address of a customer if a correct reference table exists
 - ▶ Manually
 - ▶ At the source systems
 - Common master data management across all operational applications

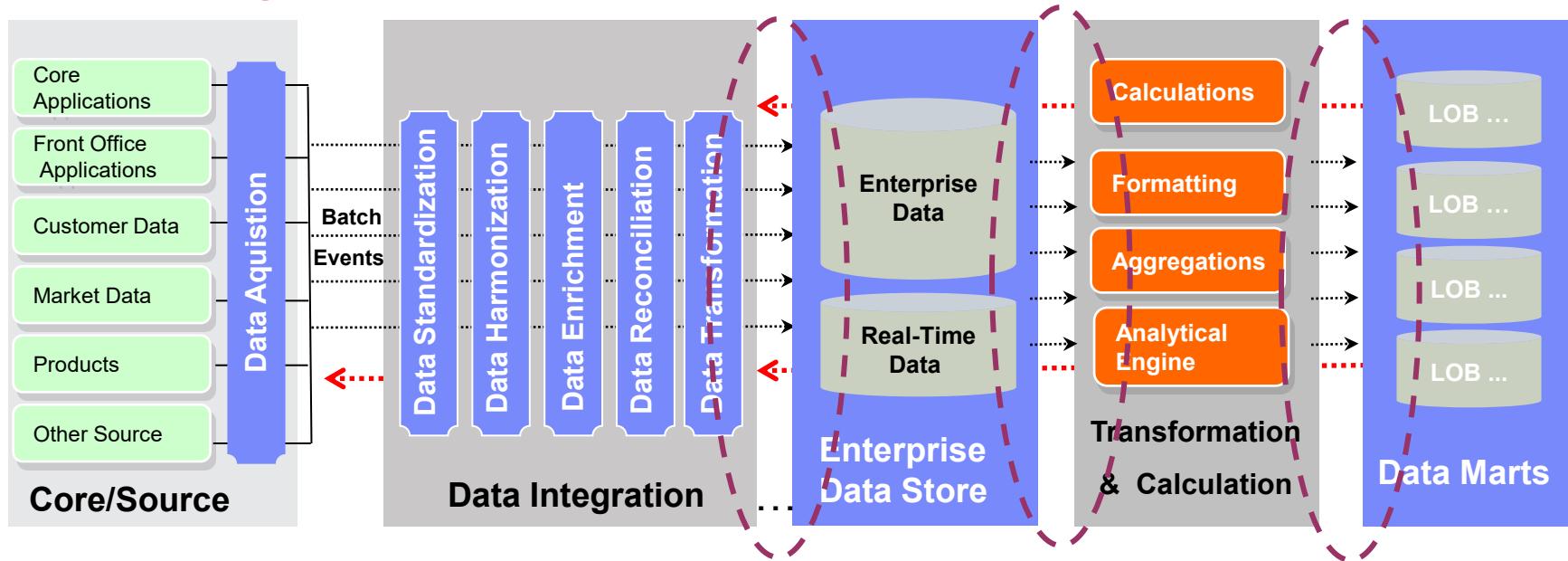
Data Integration Process



■ Data Transformation

- ▶ Apply transformations to source data to adequate to corporate standards (based on transformation rules), so that data can be leveraged for data analysis

Data Integration Process

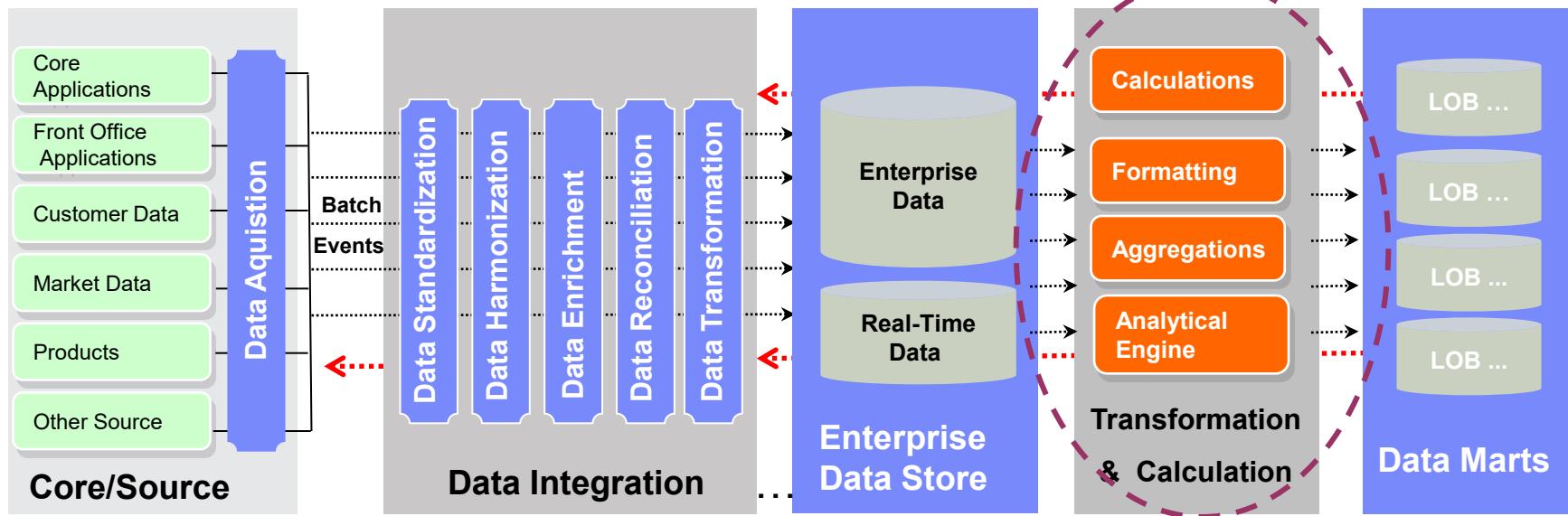


- **Data Publish and Delivery**
 - ▶ Guarantee delivery of the data with integrity
 - ▶ Populates targets structures in the EDW and DM
 - ▶ Publish information for close loop analysis (integration with operational systems)

Load

- Efficient load operations are important
 - ▶ „bulk load“
- Online load
 - ▶ Data warehouse is still accessible
 - ▶ For incremental updates
- Offline load
 - ▶ Data warehouse is offline
 - ▶ For updates that require the recomputation of a cube

Data Integration Process



- **Transformations and Calculations**

- ▶ Calculations: To meet specific business requirements for LOB (data marts)
- ▶ Formatting: Format data for easy understanding and analysis by end users
- ▶ Aggregations: Summarize data for analysis
- ▶ Analytical Engines: Data Mining and others analytical engines required for advanced analysis

Excercise

- For one of the following companies
 - ▶ Retail Bank
 - ▶ Telecommunication company
 - ▶ Online book store (like Amazon.com)
 - ▶ Discount furniture store (like IKEA)
 - ▶ Supermarket
- describe 5 potential data quality problems.
- Which impact might these problems have on its business?

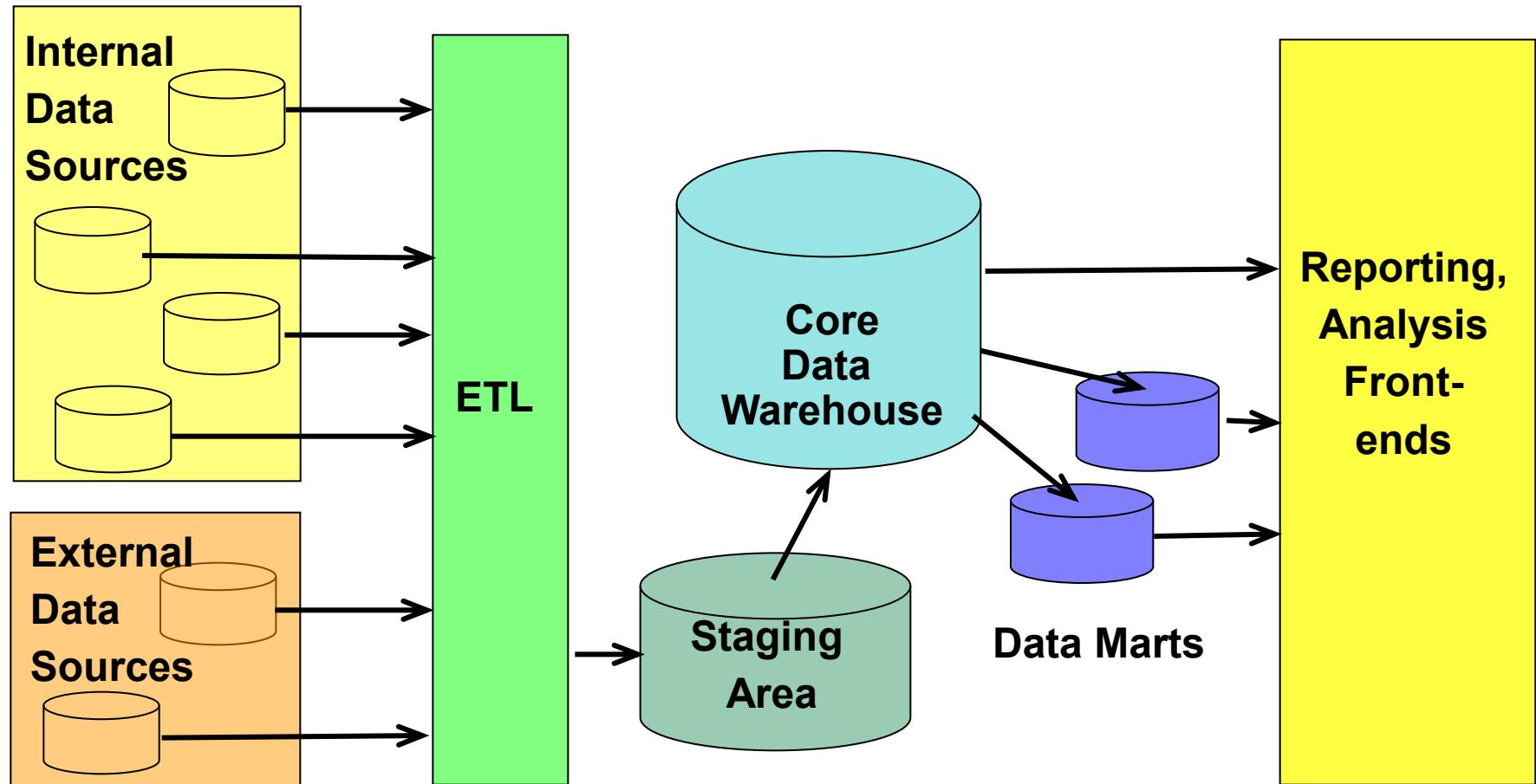
Metadata

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Standard Data Warehouse Architecture



Metadata

What Is Metadata, anyway?



Types of metadata

- Business Metadata
 - ▶ Definition of business vocabulary and relationships
 - ▶ Definition of the value range
 - ▶ Linkage to physical representation

GL Account Number

The ten digit account number for general ledger. Sometimes referred to as the account ID. This value is of the form L-FIIIIIVVV.



Business



Technical

Database = DB2

Schema = NAACCT

Table = DLYTRANS

Column = ACCT_NO

data type = char(11)

Types of metadata (cont.)

- Technical metadata
 - ▶ Report metadata
 - Report definitions
 - Data sources
 - Column definitions
 - ▶ Logical and physical metadata of data model
 - Table structure
 - Definition of columns
 - Relationships between tables and columns
 - ▶ ETL metadata
 - Job design
 - Input-/output tables
 - Computations
- Operational metadata (of ETL jobs)
 - Start time and duration
 - Return code

Why to use a common metadata repository

- Components of a data warehouse system are interconnected
 - ▶ BI report designer has to know
 - the table definitions
 - the meaning of the column values
 - ▶ ETL job designer has to know
 - the table definitions
 - the exact definition of the measures
 - ▶ ...
- Common metadata repository ensures consistency across all components
- Enables cross component metadata analysis
 - ▶ Data Lineage
 - ▶ Impact Analysis

The Areas of Metadata

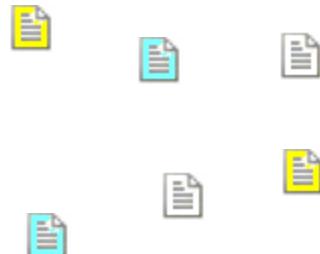
Business Glossary/Metadata



ETL Operational Metadata (Job Run Information)



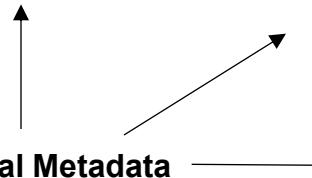
BI Reports



Physical Schemas



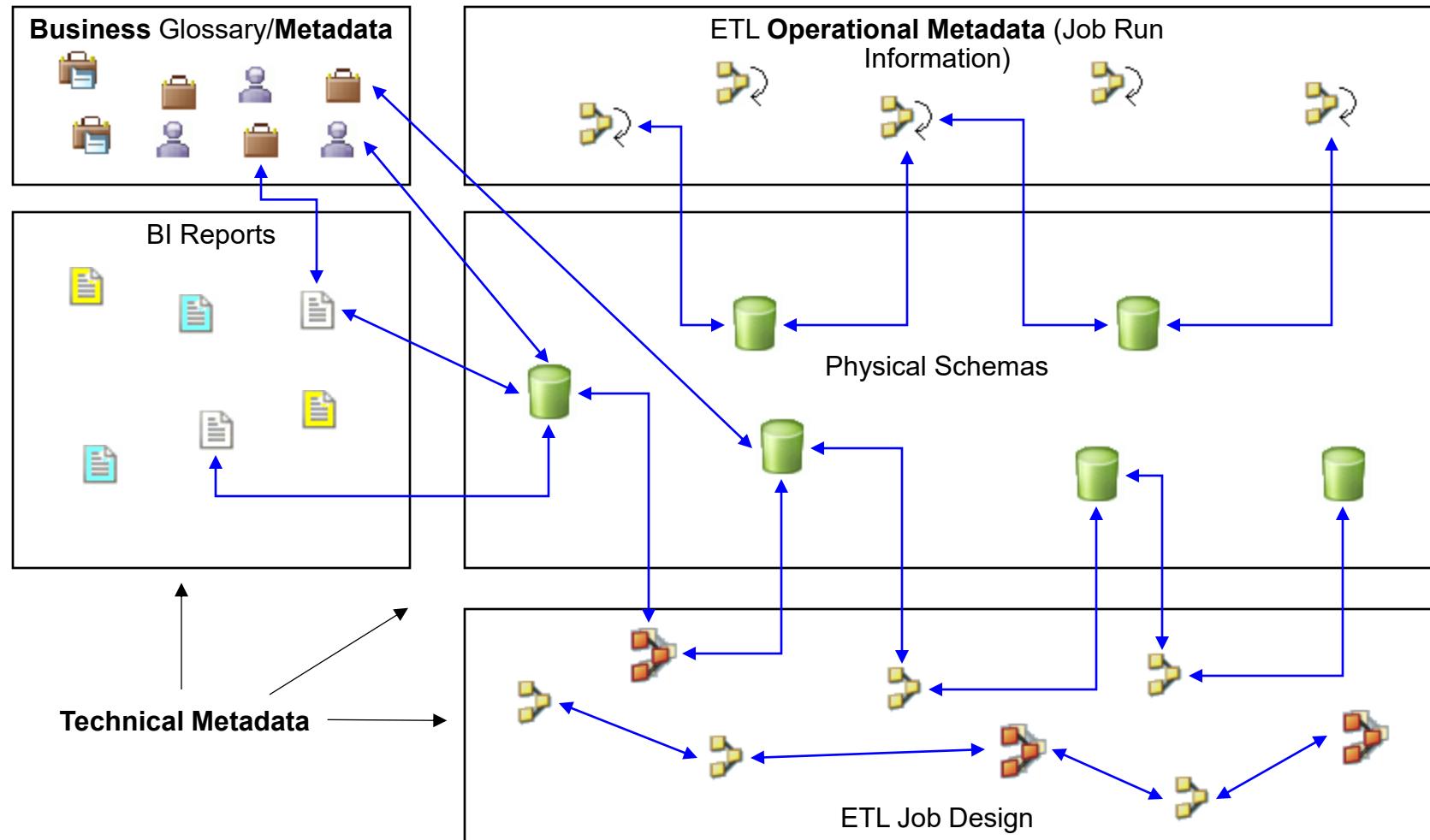
Technical Metadata



ETL Job Design

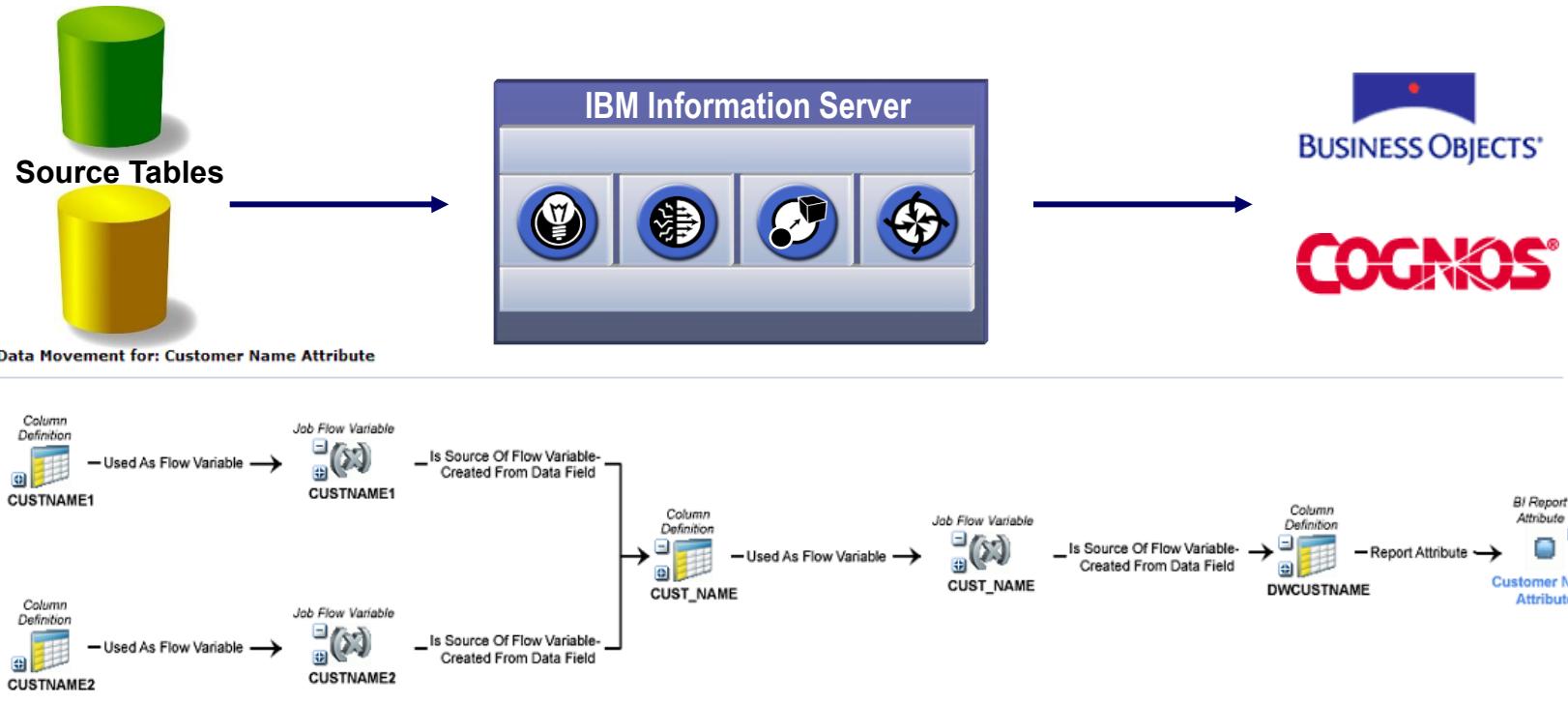


The Areas of Metadata Connected



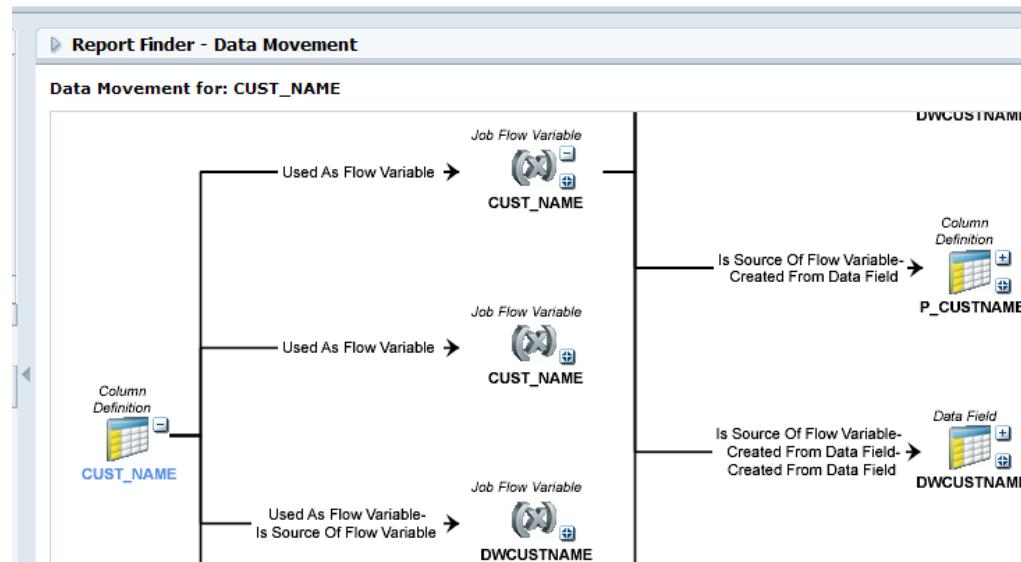
Where does a field of data in this report come from?

- “Data lineage”
 - ▶ Import & Browse Full BI Report Metadata
 - ▶ Navigate through report attributes
 - ▶ Visually navigate through data lineage across tools
 - ▶ Combines operational & design viewpoint



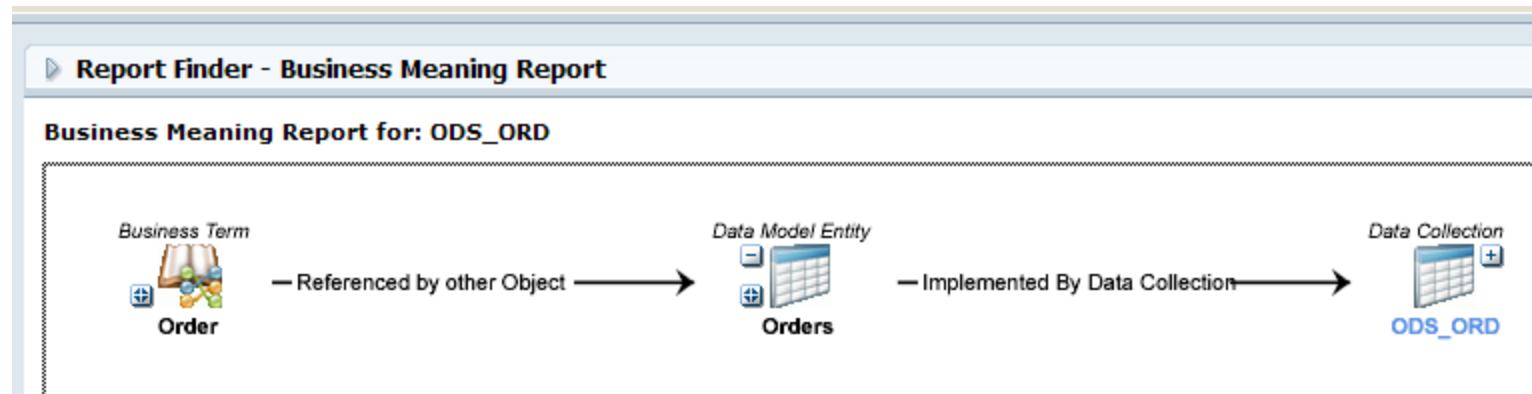
What happens if I change this column?

- “Impact Analysis”
 - ▶ Show complete change impact in graphical or list form
 - ▶ Includes impact on reports in BI tools
 - ▶ Visually navigate through impacted objects across tools
 - ▶ Allows impact analysis on any object type



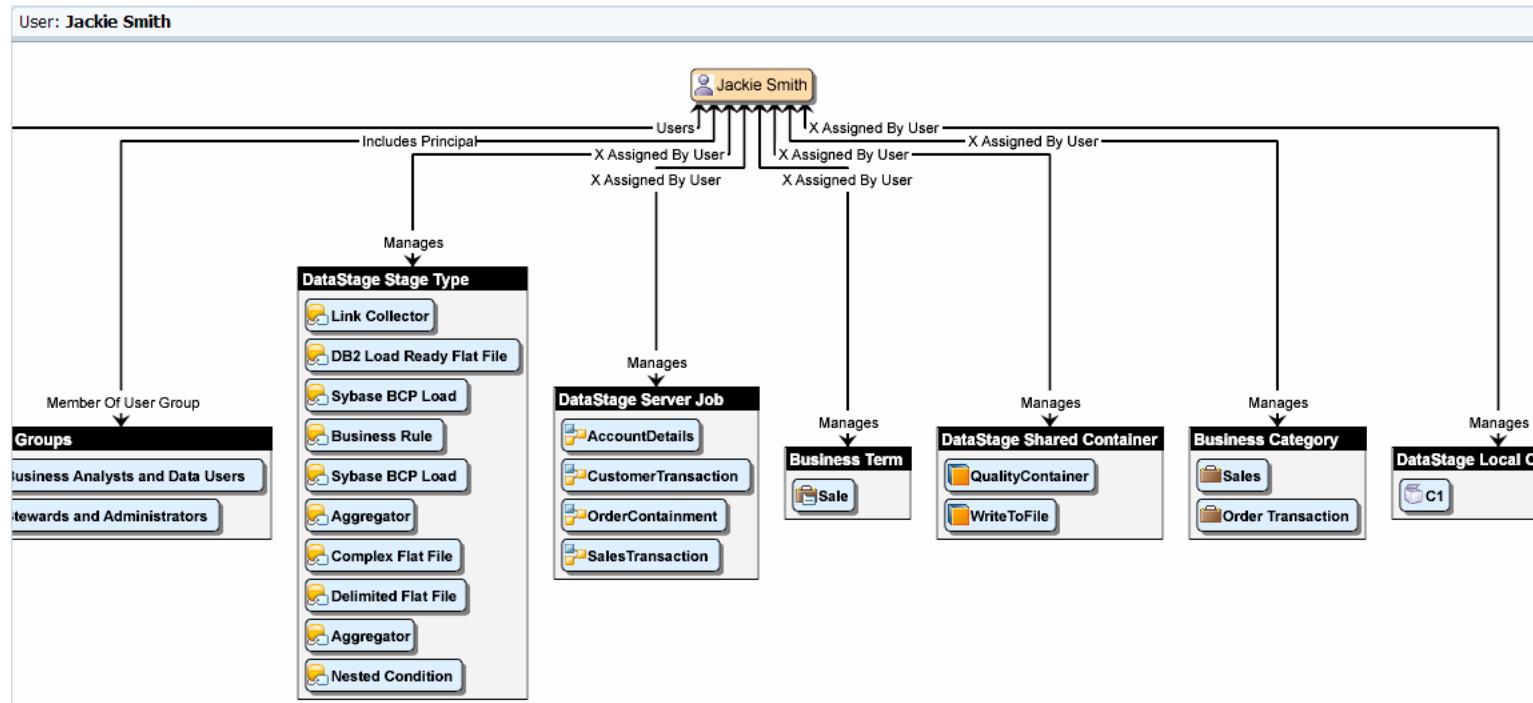
What does this field mean?

- Show relationships between business terms, data model entities, and technical and report fields
- Provides cross-tool mapping of business terms
- Allows field meaning to be understood
- Allows business term relationships to be understood



What objects does this user own?

- Shows objects that user manages
- Shows stewardship relationships on business terms
- Shows user group associations

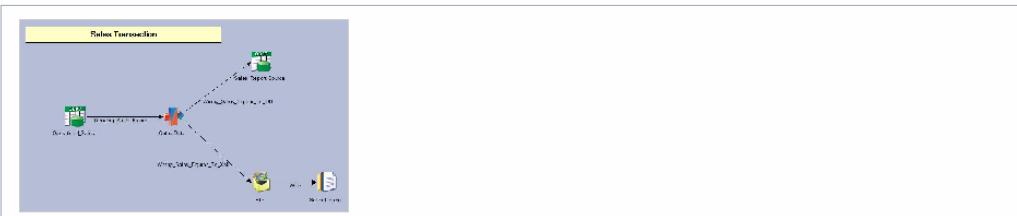


What happened on the last job run?

- Navigation through complete job details
- Navigation of complete operational metadata

DataStage Server Job: SalesTransaction

Image



Sales_TX.JPG

DataStage Job

Job	SalesTransaction
Type	SERVER
Project	LAURIE-DEMO1:TAB
Folder	\Jobs\ACME_Sales
Description	Reconcile all current transactional information into the operational data store, comprising activity of the past 12 months.
Data Steward	Jackie Smith
Contains DataStage Stages	<ul style="list-style-type: none"> File Operational_Sales OutputData Sales_Report Sales_Report_Source
Contains DataStage Links	<ul style="list-style-type: none"> Reading_Sales_Figures Write Writing_Sales_Figures_To_DB Writing_Sales_Figures_To_XML
Contains DataStage Local Containers	None

DataStage Job Usage

Operational Job Locator Information	SoftwareExecutable(Job) = "SalesTransaction";
Job Runs	<ul style="list-style-type: none"> SalesTransaction 2007-01-24 16:21:12 SalesTransaction 2007-01-24 18:26:44 SalesTransaction 2007-01-25 12:06:21
Previous DataStage Job	OrderContainment

Exercises

- List 3 reasons why common metadata is important in the context of warehousing.
- Define criteria for the evaluation of an ETL tool
- How does a relational DBMS (like Oracle, DB2, MS SQL Server) meet these requirements?