Data Warehouse Projects: A short course for IT executives

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Bob's Background

- IT professional, 17 years
- Entrepreneur
- Education
 - BS Business Administration (MIS) from Kansas State University
 - MBA (finance concentration) from University of Kansas
 - Coursework in Mathematics at Washburn University
 - Graduate Certificate Data Science from Rockhurst University
- Addicted to everything data



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Motivations for the Course

- Give IT managers the tools and language necessary to successfully manage a data warehouse project.
- Set proper expectations for a data warehouse project.
- Help IT managers hire the right talent.
- Help IT managers navigate potential land mines.
- Help IT managers understand what they need from the rest of the organization.



This is an Opinionated Course!

- You can't find this information on the internet
- Based on 17 years of building databases
- Material is based on Mass Street's philosophy and practices
- Material does not invalidate any other opinions or methods



Course Outline

Data Warehouse Overview

Data Warehouse ETL

Organizational Considerations



Let's answer the following question

Should I invest time and money learning about traditional databases?



The current reality of Big Data

- The job market isn't as big as the data
- Most companies don't have big data
- There is still a wide gap between the perception and the reality
- Having a lot of data and being able to do anything with it are two separate things



The current reality of Big Data

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What is Medium Data®?

- · I'm just kidding. You can use it.
- Not big data but not small data either
- Far more ubiquitous than Big Data



Getting the Course Material

• The latest course material can be found on GitHub.

•https://github.com/MassStreetAnalytics/data-warehouse-projects



Data Warehouse Overview What is some important terminology?

- Data model
- Dimensional modeling
- Dimension table
- Fact table
- Slowly changing dimension (SCD)
- Star schema
- Snowflake
- Online Analytical Processing (OLAP)
- Master Data Management (MDM)
- Data Contract



Data Warehouse Overview Why do we need a data warehouse?

- Historical analysis in transactional systems isn't efficient.
- Removes data silos.
- Provides a central source of truth.
- Enables 360 analysis.
- Bonus: Helps uncover errors and inefficiencies in existing business processes.

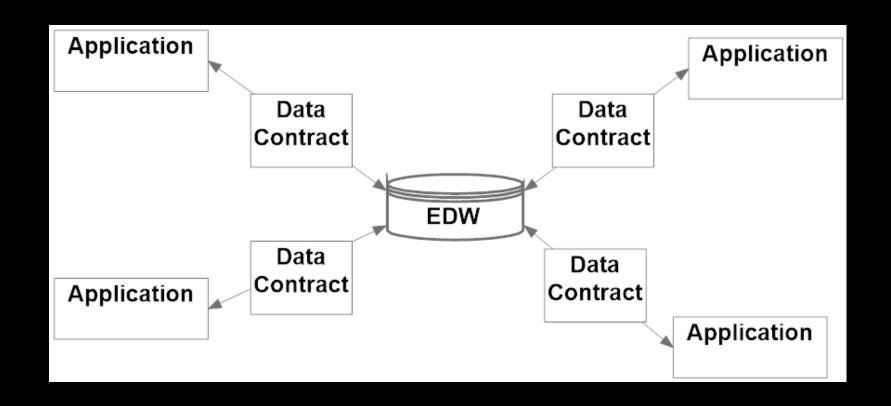


Data Warehouse Overview What is a data warehouse?

- Enterprise Data Warehouse (EDW)
 - Central to any organization's analytic efforts
 - Should be a central location of all historical data
 - Single source of truth
 - There are strict rules around processing and storage



Data Warehouse Overview What is a data warehouse?





Data Warehouse Overview What is NOT a data warehouse?

- Data Mart
 - Subset of data warehouse
 - Usually organized by business process
 - Finance, Marketing, HR, etc.



Data Warehouse Overview What is NOT a data warehouse?

- Data Lake
 - A massive dumping ground for data.
 - There is no complex ETL.
 - Structure and relationships are determined after the load.



- A server
 - Needs to be a dedicated box
 - Specs driven by selected DB
- A database
- Possibly a point-and-click ETL Tool
- A Master Data Management Tool
- Possibly an OLAP technology
- An analysis and visualization tool



- Relational DBs
 - PostgreSQL, MySQL
 - SQL Server
 - Oracle
- NoSQL DBs
 - Druid
- Massively Parallel Processing DBs (MPP)
 - MySQL Cluster CGE
 - SQL Server Parallel Data Warehouse
 - Greenplum



- An ETL Tool
 - Possibly
 - Usually used by junior people
 - More trouble than they are worth
- Popular ETL Tools
 - Informatica
 - SQL Server Integration Services (SSIS)
 - DataStage
- Less Popular Tools
 - Talend
 - Pentaho?



- Master Data Management Tools (MDM)
 - SQL Server Master Data Services
 - It has some shortcomings
 - Some critical functions can't be automated
 - Talend
 - Informatica



- OLAP software
- Can be used to create data marts
- OLAP isn't that popular
- SQL Server Analysis Services
- There are others
 - Pyramid Analytics



- Orchestration
- Called job scheduler in the old world
- Tools
 - SQL Server Agent
 - Control M
 - Apache Airflow



- Analysis and visualization tools
- Traditional point-and-click tools
 - SQL Server Reporting Services
 - Power BI
 - Tableau
- New Tools
 - Jupyter Notebook
 - R Notebooks
 - Apache Zeppelin



- There are two approaches to data warehouse design
 - Kimball
 - Inmon
- We build them as a destination for everything
 - Build a data mart from that

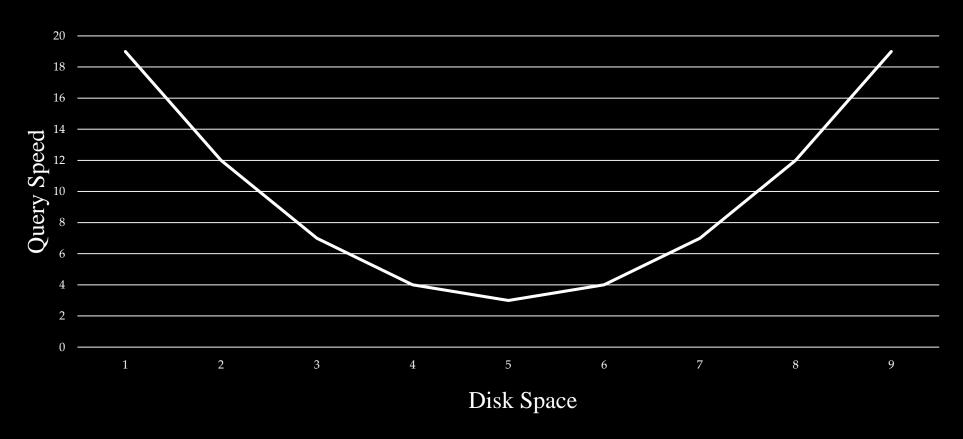


- In general, there are two approaches to designing databases
 - Transactional
 - Dimensional
- Transactional
 - Highly normalized
 - Many tables
 - Many tables leads to many joins
 - Little, if any, data duplication



- Dimensional
 - Denormalized
 - Far fewer tables
 - Fewer tables leads to fewer joins
 - Data duplication is ok





Nobody likes a slow query.



- More than one way to skin the cat
- Convert the transactional database into a dimensional one.
 - Works if all the business rules are captured in the data model
 - Not resource-intensive
- Develop a two dimensional dataset to answer a specific question and model that.
 - This requires significant resources to support
 - Can be a much faster process



- Recommended process flow for new models
 - Source data identified
 - Data model created
 - ETL created
 - Data validation/UAT
 - Deploy
- This should be a 1-3 week process
- If it takes longer, something is wrong



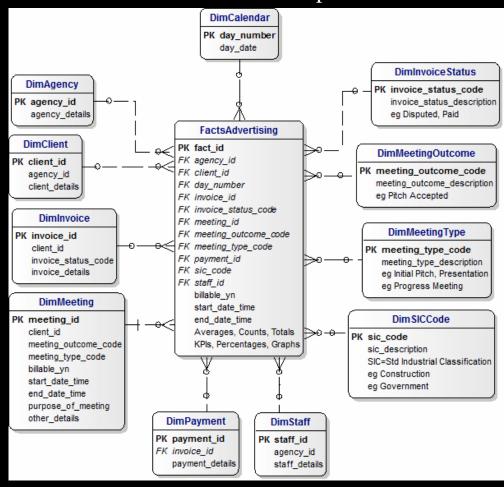
Data Warehouse Overview What is a data model?

- Abstract model that organizes elements of data and standardizes how they relate to one another and to properties of the real world
- Models are built around fact tables
- Two dimensional representation of how data is stored
- Errors in data models will cause serious issues
 - Query performance suffers
 - Numbers won't crunch correctly
 - Errors don't get fixed because of priorities
 - Hacky workarounds worsen the situation
 - Fixing errors is expensive



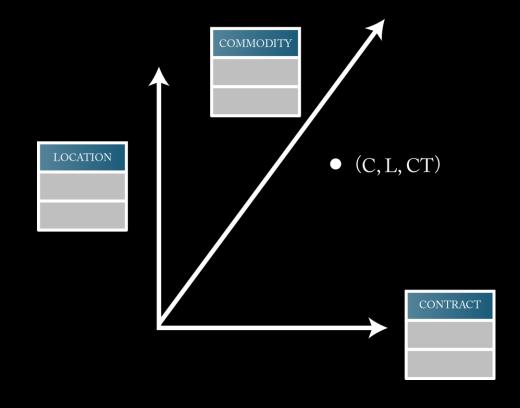
Data Warehouse Overview What is a data model?

Data Model Example





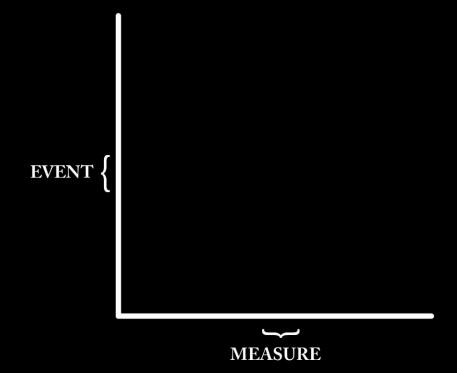
- Using the word dimension isn't arbitrary
- A dimensional model is a two dimensional representation of n dimensional space
- A dimension is like an axis on a chart
- Issuing a query is like issuing coordinates
- Facts (numbers) reside at the coordinates in n dimensional space

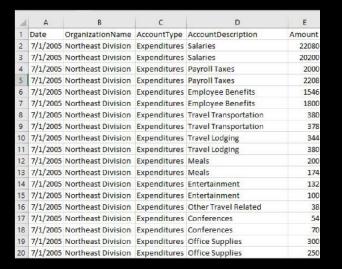




MENTAL MODEL

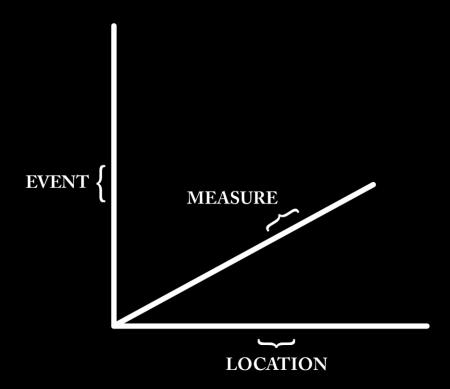
REAL WORLD OBJECT



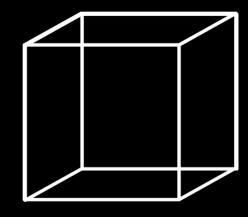




MENTAL MODEL



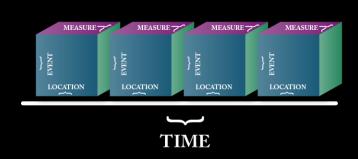
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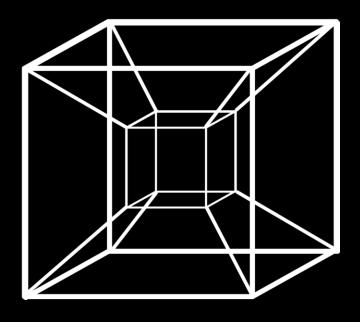




MENTAL MODEL

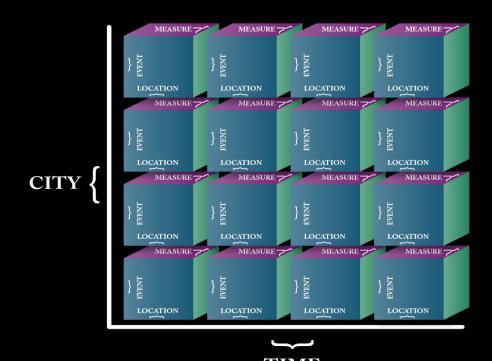
REAL WORLD OBJECT





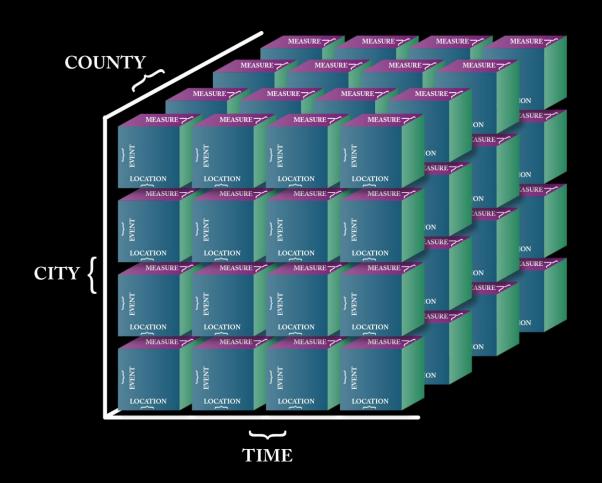


FIVE DIMENSIONS



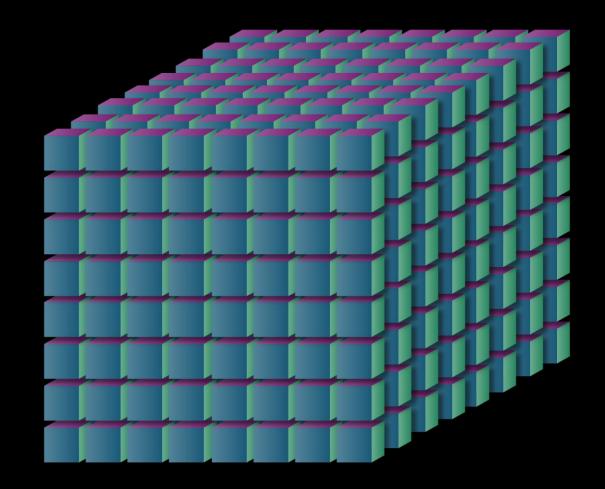


SIX DIMENSIONS



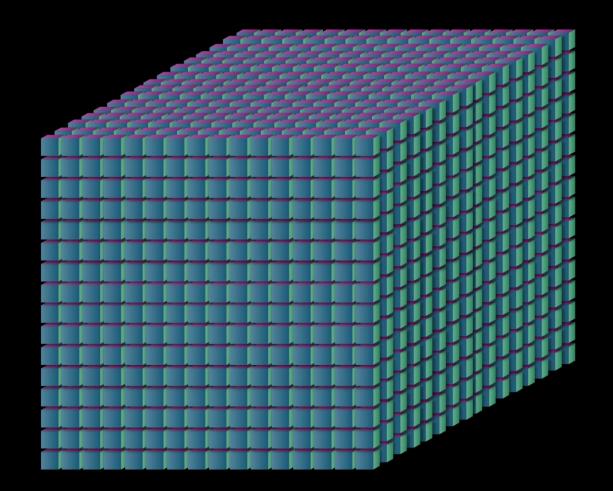


NINE DIMENSIONS



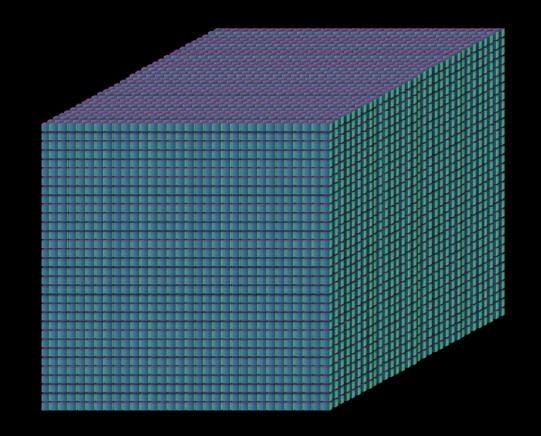


TWELVE DIMENSIONS





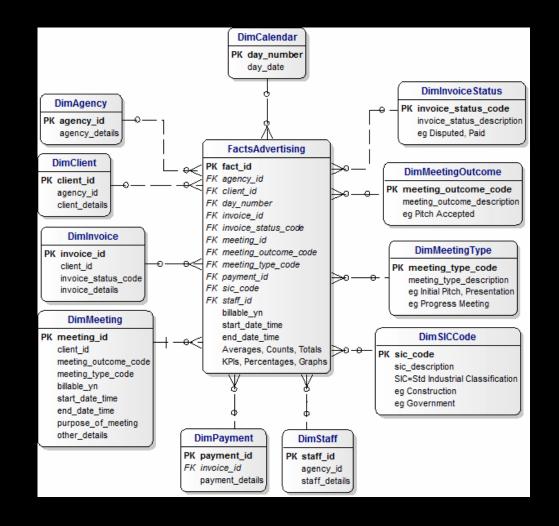
FIFTEEN DIMENSIONS





- Primary keys are meaningless (contrived key)
- Dimensions surround fact tables
- Dimensions contain textual filter data
- Dimensions can contain numbers as long as they aren't being used to add
- There are various kinds of dimensions
- Data in dimensions can get duplicated
- Dimensional data changes slowly, if ever







- Most dimensions should change rarely
- Slowly Changing Dimension Types
 - Type I Make the change. Don't keep history.
 - Type II Add a row. Retire the old row.
 - Type III Put the change in a different column.
 - Type X − Use some combination of the above.



- Types of Dimensions
 - Date dimension Used to make time-based queries
 - Not strictly a type of dimension
 - Junk unique combinations of low cardinality values
 - Degenerate Dimension that goes in fact table
 - Role Play Copy of an existing dimension but renamed
 - Conformed Dimension used to jump across models
 - Outrigger Used to snowflake



- Date dimension
- Absolutely critical piece of any data warehouse
- Allows you to easily slice and dice data based on date parameters
- Can be customized for your particular organization
 - More columns can be added
 - Can be adjusted for your fiscal calendar
- Almost every fact table will connect to the date dimension



- Junk Dimension
 - Unique combination of low cardinality values
 - Usually a table filled with flags or binary values
 - You can use math to predict the number of rows in the table
 - $X_1 * X_2 * X_3 * X_n$
 - Where X = number of unique values



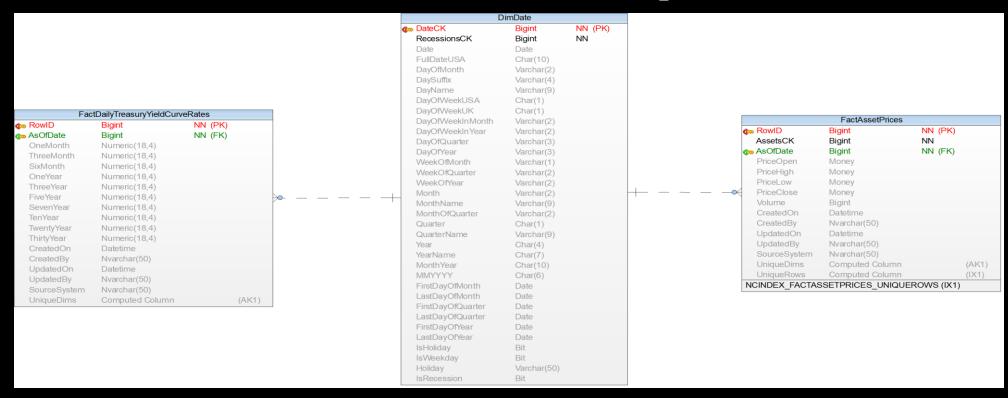
- Degenerate Dimension
 - A dimension that does not belong in its own table
 - Goes in the fact table
 - Invoice numbers are a good example
 - Resist the temptation to create inappropriate degenerate dimensions
 - If it repeats infinitely, it is NOT a degenerate dimension



- Role Play
 - Close copy of an existing dimension
 - Data is the same. Table and columns are named differently
 - Implemented with views
 - Used to make things clearer for users and creates cleaner code for data professionals
 - Most commonly implemented with the date dimension



Conformed Dimension Example





Outrigger Dimension Example

DimReces	sions	
RecessionsCK	Bigint	NN (PFK)
RecessionName	Nvarchar(50)	
RecessionBeginDate	Date	
RecessionEndDate	Date	
Duration	Nvarchar(50)	
TimeSincePreviousRecession	Nvarchar(50)	
PeakUnemployement	Numeric(16,4)	
WhenPeakUnemploymentAchieved	Nvarchar(50)	
MaxGDPDecline	Numeric(16,4)	
DecreaseInBusinessActivity	Numeric(16,4)	
DecreaseInTradeAndIndustrialActivity	Numeric(16,4)	
Characteristics	Nvarchar(max)	
CreatedBy	Nvarchar(50)	
CreatedOn	Datetime	
UpdatedBy	Nvarchar(50)	
UpdatedOn	Datetime	
SourceSystem	Nvarchar(100)	
SourceSystemKey	Nvarchar(100)	
EffectiveFrom	Datetime	
EffectiveTo	Datetime	
IsMostRecentRecord	Bit	
RowHash	Binary(16)	



Data Warehouse Overview What is a fact table?

- Two dimensional representation of an answer to a question
- Each record is a single event
- All records have to be at the same grain
- Records have to be additive (usually)
 - Corrections have to be complete or incremental
 - There are also semi-additive and non-additive records
- Types of fact tables
 - Snapshot
 - Cumulative

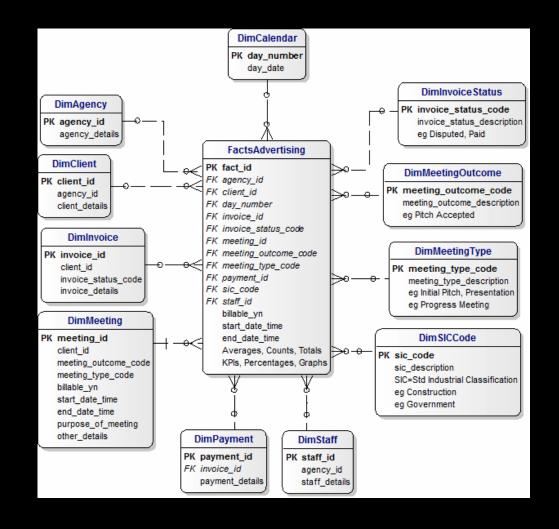


Data Warehouse Overview What is a fact table?

- The primary key of a fact table is the unique combination of dimensions associated with that fact
- That key combination defines the record's signature
- You can add an auto increment key but not for record identification
- Defining unique records is critical



Data Warehouse Overview What is a fact table?





Plan to make MDM part of your data architecture from Day 1!



- Master data is the set of data objects that are at the center of business activities (Customers, Products, Cost Centers, Locations, Assets, Task)
- Central concept necessary for any customer data integration project.
- MDM is a human in the loop process.



- The disciplines, technologies, and solutions that are used to create and maintain consistent and accurate master data for all stakeholders across and beyond the enterprise
- The fundamental purpose of an MDM System is to serve as the authoritative source for master data; an MDM System is a system that provides clean, consistent master data to the enterprise.



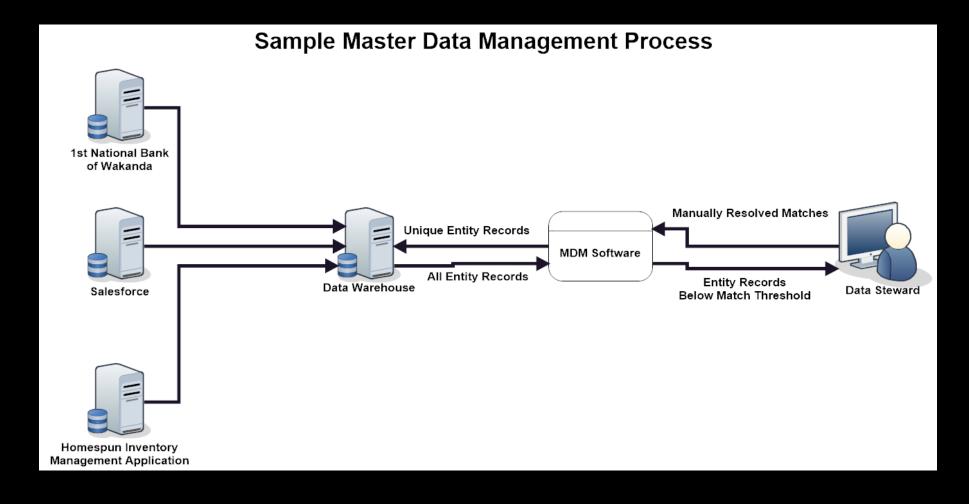
• The problem is actually technical in nature

De-duplicate identical entities from different systems

• De-duplicate identical entities from the same systems

• Manage all those keys so fact records flow through properly







- How to do MDM from Day 1
 - Identify entities in your organization
 - Identify the source systems
 - Identify a data steward
 - Begin the process of selecting MDM software



Data Warehouse Overview Section Conclusion



E = Extract T = Transform L = Load



- The process of moving data from one location to the next
- Usually reserved for talking about loading historical data
- This is different from the process of moving data in and out of a transactional system
- Usually involves moving large amounts of data in a batch process

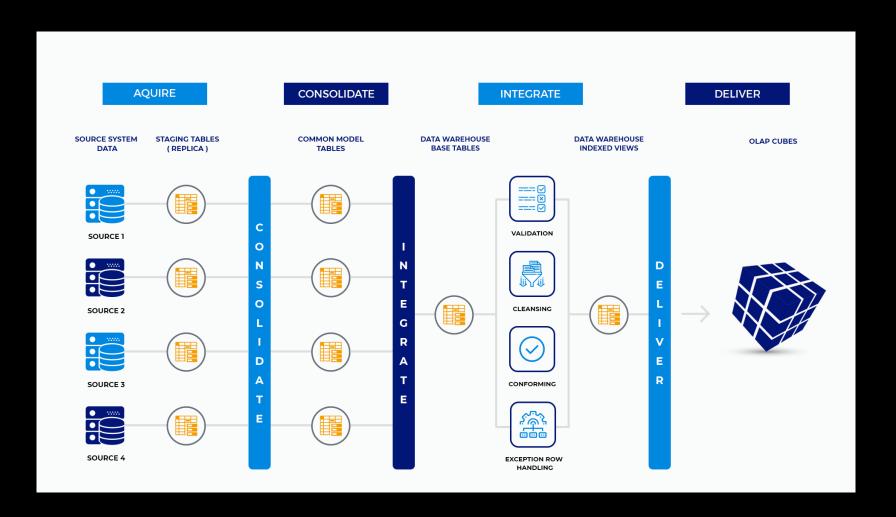


- Also involves combining data sources
- Usually a rigorous process with many business rules
- Being able to write high performing code is important
- Can easily turn into a hot mess
- Might be referred to as ELT

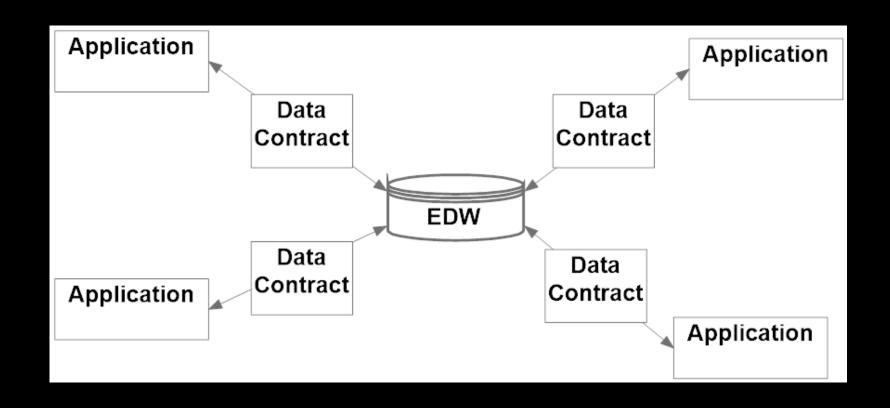


- Two methods for ETL
 - Supply push
 - Demand pull
- Demand pull is the most common and recommended approach
- Supply push is often necessary for third party integration











- An agreement between systems that outlines the exchange of data
- Engineers on both ends of a pipe need to agree on what is flowing through that pipe
- Data contracts have two general implementations
 - As actual software via an interface
 - A written agreement between two departments



- Data contracts as interfaces
 - An API is a type of data contract
 - Not appropriate for data warehouse work
 - Specialized APIs can be written but not recommended



- Data contracts as written agreements
 - This is the recommended approach
 - It's critical for supply push processes that might change frequently
 - Change control needs to be a part of the contract



Data contracts are messy in the real world

A mixed approach may be necessary

Be prepared to adapt and overcome



Data Warehouse ETL What is an ETL Framework?

- A standardized methodology to move data into and out of the data warehouse
- Can be generalized to moving data around the entire organization
- Enables management of all data pipelines
- Helps significantly reduce the cost of ETL maintenance



Data Warehouse ETL What is an ETL Framework?

- Why do we need an ETL Framework?
 - Helps prevent the Rube Goldberg machine
- Maintenance is not a data engineer's highest or best use
- We see maintenance as a waste of cash



Data Warehouse ETL What is the Mass Street ETL Framework?

- An open source ETL Framework for SQL Server
- Technically for any RDBMS
- Six years in the making
- Its use is taught in implementing data warehouses in the real world
- Freely available on the MIT license
- Pull requests are more than welcome



Data Warehouse ETL What is the Mass Street ETL Framework?

- Mass Street ETL Framework components
 - Full documentation on usage
 - Sample model creation template
 - SQL Scripts for creating all necessary objects
 - Sample scripts for creating new objects
 - Sample scripts for data processing
 - Helper scripts to tackle common task



Data Warehouse ETL Section Conclusion



Organizational Considerations Getting Organizational Buy-In

- Organizational buy-in is critical for project success
- This requires leadership from the top down
- You're going to change the status quo
 - People will resist
 - People will be ambivalent
- You need to involve everybody at every level
 - Marketing is a great place to start



Organizational Considerations Getting Organizational Buy-In

- The fastest way to get buy-in is quick wins!
 - The lure of big problems is seductive
- Sometimes quick wins won't do it.
 - Involve business users and get them invested
- Don't hide your tech resources
- Implementation should be a big deal
 - Anything having to do with the warehouse should be a newsworthy corporate event



Organizational Considerations Getting Organizational Buy-In

- The data warehouse is exciting!
- It reduces cost!
- It saves time!
- It allows you to do analysis you could never do before!
- It's a necessary step before any data science efforts!
- It cures cancer! (Not really.)

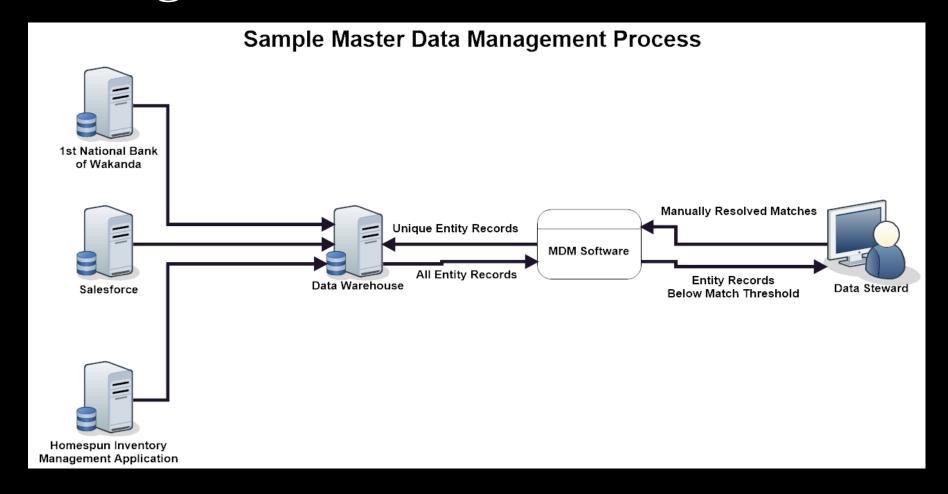


Organizational Considerations Data Governance

- Data governance is a necessity
- This is how you keep junk data out of the warehouse
- Data governance looks at:
 - Who can access data
 - What they are accessing
 - Keeping data clean
 - Keeping data secure
- You can scale the team
 - You need at least one business user defining what does and doesn't go into the warehouse



Organizational Considerations MDM Again





Organizational Considerations MDM Again

- MDM is not done in isolation
- Every master object needs a human business owner
- You'll have to convince a manager to let go of someone's time for a bit
- It needs to be someone who is super familiar with the data
- After an initial clean-up, this should not take up a lot of time



Organizational Considerations Documentation

- Users of the data warehouse
 - Highly technical data professionals
 - Somewhat technical business analysts and report writers
 - Non-technical business analysts
 - Non-technical report consumers (managers, executives, etc.)
- All these people will need documentation



Organizational Considerations <u>Documentation</u>

- Types of documents
 - Source to target mapping
 - For engineers
 - Entity Relationship Diagram
 - For engineers, technical analysts, report writers
 - Data dictionary
 - For everybody



Organizational Considerations How to Hire a Data Engineer

- Options to get this work done
 - Hire a specialized services firm like Mass Street
 - Engage a recruiter
 - You'll get mixed results with this
 - Try to hire them yourself (not recommended)
- There are not a lot of "right" answers in data warehousing
- Candidates should be selected on depth of experience
- I've provided a list of recommended interview questions



Organizational Considerations How to Hire a Data Engineer

- You're looking for an engineer, not a DBA
- Having an understanding of the inner workings of a specific RDBMS is not necessary
- Understanding dimensional modeling techniques and efficient ETL is what is required
- They should have a good understanding of all the material presented in this course



- •Caution! Opinions ahead!
- The questions are ok. The thought processes behind the questions is more important.



Explain the difference between Kimball and Inmon methodologies.



Discuss the different types of slowly changing dimensions.



When you are pulling large amounts of data from a transactional source system, how do you keep from locking users out?



There is a number we look at that is the result of a calculation. Where in the ETL process does the math for that calculation belong?



We are getting started and this is what we are planning. [List off the things you are planning.] What else do you recommend we do?



What are the challenges surrounding managing the history of a Type II SCD?



We have customers stored in three different systems. Because they are entered three different ways, we often cannot tell when a customer is entered twice. We don't want duplicates in the warehouse. How would you solve this problem?



How do you decide where to put indexes on tables?



Describe how you go about transforming a transactional system into a dimensional model.



We get bank reconciliation data in a feed. The data comes in at the transaction level on a daily basis but it also gives us a month-to-date rollup. How would you go about storing this?



Organizational Considerations Section Conclusion



Course Wrap Up

