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21st Century Data Modeling for Effective Analytics



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EWSolutions' Background

EWSolutions is a Chicago-headquartered strategic partner and full life-cycle systems integrator providing both award winning strategic consulting and full-service implementation services. This combination affords our client partners a full range of services for any size enterprise information management, metadata management, data governance and data warehouse/business intelligence initiative. Our notable client partner projects have been featured in the Chicago Tribune, Federal Computer Weekly, Journal of the American Medical Informatics Association (JAMIA), Crain's Chicago Business, FBI, The Doings and won the 2004 Intelligent Enterprise's RealWare award, 2007 Excellence in Information Integrity Award nomination, DM Review's 2005 World Class Solutions award, 2016 CIO Review 20 Most Promising Enterprise Architecture providers and 2018 Innovation





2007 Excellence in Information Integrity Award Nomination



Best Business Intelligence Application Information Integration Client: Department of Defense









Excellence and Thought Leadership





2018: Most Innovative in **Data Management Solutions**





For more information on our Strategic Consulting Services, Implementation Services, or World-Class Training, email us at info@EWSolutions.com

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Anne Marie Smith, Ph.D.

- Internationally recognized expert and speaker in the fields of enterprise information management, data architecture and data modeling, metadata management and data warehousing
- One of the top industry experts in data governance and information management strategy and planning
- Over 20 years experience in delivering solutions in enterprise information management for numerous companies / clients across varied industries
- Published almost 100 articles that have been featured in some of the industry's most prestigious magazines and journals
- Has taught at several institutions of higher learning including LaSalle University
- Earned a Ph.D. in Management Information Systems and holds several certifications
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Agenda

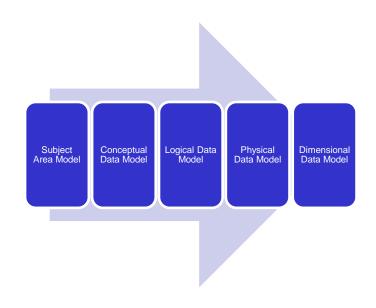
- Data Modeling and Data Architecture
- □ Data Modeling for Analytics Best Practices
- Data Modeling at Scale
- ☐ Future of Data Modeling Data Modeling for 21st Century



Data Modeling and Data Architecture

Data Modeling Definition

- Data modeling: process used to define and analyze data requirements that support the business processes for information systems
 - Process involves professional data modelers working closely with business stakeholders and potential users of the information system
- Using proper data modeling techniques and methodologies allow an organization to model data in a standard, consistent, predictable manner to manage data as a resource
 - Model should reflect concepts of the organization and how they relate to one another and to the business processes of the application (operational or analytical)
 - Data models should be part of the organization's business documentation
 - Data literate organizations develop and maintain data models for all applications and at all levels



Data Architecture Definition

- Data architecture describes the structure of an organization's logical and physical data assets and data management resources
 - ➤ Includes models, policies, rules, and standards that govern the collection, storage, arrangement, integration, and use of data in organization
- ☐ The goal of data architecture is to translate business needs into data and system requirements and to manage data and its flow through the enterprise and all applications
- Data literate organizations use data architecture to manage data and process flow throughout organization

Data Architecture vs. Data Modeling

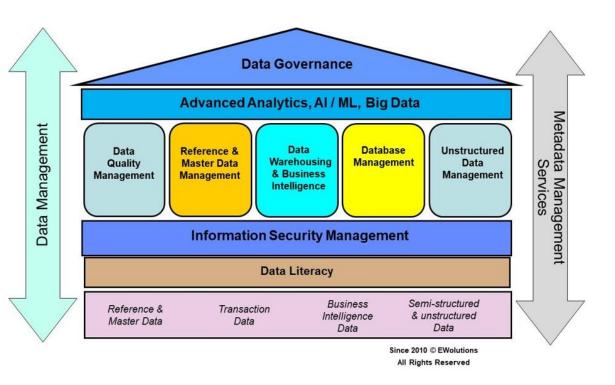
Data Architecture: defines the blueprint for managing data assets to establish strategic data requirements and designs to meet those requirements using models and other artifacts

Data Modeling: process of discovering, analyzing, representing, and communicating data requirements in a precise form called the data model

Data architecture and data modeling are essential components of effective 21st century systems, data management capabilities, and user empowerment

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Data Management Framework



- •Data Governance: Planning, supervision and control over data management and use.
- •Data Architecture Management: Defining the blueprint for managing data assets.
- •Data Modeling and Design: Analysis, design, implementation, testing, deployment, maintenance.
- •Data Storage and Operations: Providing support from data acquisition to purging.
- •Data / Information Security Management: Insuring privacy, confidentiality and appropriate access.
- •Data Quality Management: Defining, monitoring and improving data quality.
- •Reference and Master Data Management: Managing golden versions and replicas of data.
- •Data Warehousing and Business Intelligence Management: Enabling reporting and analysis.
- •Unstructured Data Management: Managing data found outside of databases.
- •Metadata Management: Integrating, controlling and providing Metadata.
- •Data Integration: Enabling transfer of data across applications and systems

Data Modeling Goals, Benefits and Phases

Goals of Data Modeling

- Ensures that all data objects required by the business process are accurately represented
 - Omission (or repetition) of data will lead to creation of faulty reports and produce incorrect results for operations and decisions
- A data model helps identify and organize data needed at the conceptual, logical and physical levels
 - Developing the series of data models can identify missing and redundant data early in the development process (conceptual and logical levels)
 - Data model structure helps to define the relational tables, primary and foreign keys and stored procedures to support a robust database

Benefits and Value of Data Modeling

☐ Higher quality data

- Data model diagrams display requirements and business rules; can anticipate large-scale data corruption opportunities in advance
- Data models define rules that monitor data quality, which reduces the chance of errors
- Data model development includes definition of concepts and attributes, establishing terms and meaning throughout organization

☐ Increased internal communication about data and data processes

- Enables business to define how data is generated and moved throughout applications based on business needs, establishes organization-supported business definitions for terms (attributes) and concepts
- Supports data governance processes through identification of relevant data stewards and clear data definitions

Reduced development and maintenance costs

Data modeling identifies errors and inconsistencies early, so they are far easier and cheaper to correct

☐ Improved performance

An organized database operates more efficiently; building a database from a well-designed data model stops endless searching across databases; enables faster results

Data Modeling Phases

- ☐ Five types of data models:
 - > Subject Area
 - Conceptual
 - Logical
 - > Physical
 - Dimensional for analytics, data marts, business intelligence
- Once the organization has developed a robust subject area model for the entire organization, all data modeling efforts should begin with a conceptual data model based on the subject area model. The conceptual data model is expanded to a logical data model, then into the physical data model
 - Starting at the physical data model level prevents discovery of business requirements and hinders ability to scale the database appropriately

Subject Area Model

- Each subject area is a high-level classification that represents a group of concepts pertaining to a major topic of interest to an organization
- Subject areas can represent generic business concepts (customer, product, employee and finance), as well as industry specific areas
- SAM is not intended to represent each subject area as a "silo", it is a view of the business
- ☐ There can be some discussion about where to place a concept, since each concept belongs in one and only one subject area
 - Resolve conflicts or identify concepts that can relate to multiple subject areas through conceptual model's relationships

Subject Area Model Example

Finance

- Regulation
- Reporting
- **Financial** Accounting
- Accounts Payable
- Accounts Receivable
- Asset
- Treasury

Product

- **Finished Good**
- Manufacturing Service
- Manufacturing Equipment
- **Product Taxonomy**
- Research & Development
- Design
- **Testing**
- Bill of Materials
- **Build Plan**

Location

- Site
- Land
- Building
- Office
- Factory
- Port
- Airfield
- Warehouse
- Geographical area

Customer

- Individual
- Company
- **Pricing**
- Sale
- Sales

Legal

- Policy
- Contract
- Compliance and
- **Ethics**
- Dispute
- Data Security and Privacy
- Government Affairs

Material

- Raw Material
- Part
- Work in Process (WiP)
- Inventory

Supplier

- Supplier Type
- Supplier Organization
- Commitment
- Performance
- Pricing
- Specification

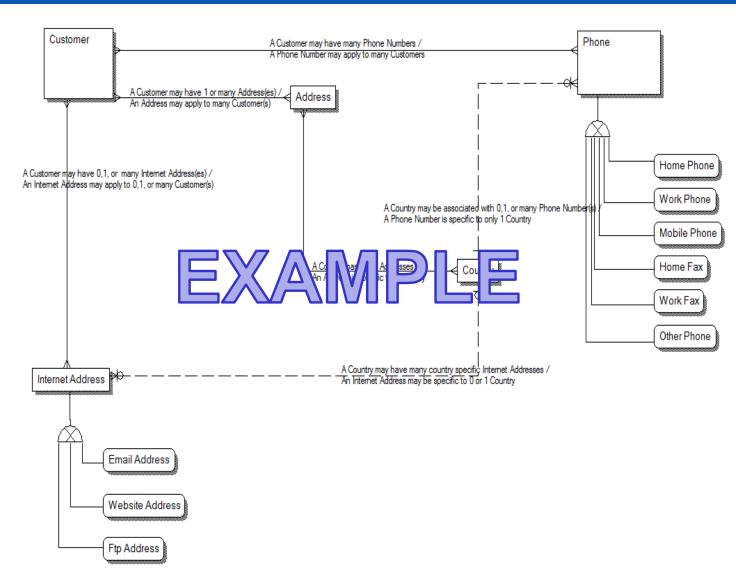
Party

- **Employee**
- Partner
- **Tenant**
- Third Party
- Government Agency

Conceptual Data Model

- Identifies business concepts to support the organization's activities and requirements
 - Represents data concepts and their relationships, defined for organizational use
 - > Based on the organization's subject area model
- Conceptual data models are designed and developed for a business audience
 - ➤ Business requirements should drive all database and application design, starting at conceptual level helps avoid redundant applications or conflicting databases
- □ Developed independently of hardware specifications like data storage capacity, location, or software specifications like DBMS-specific content
 - no technology requirements

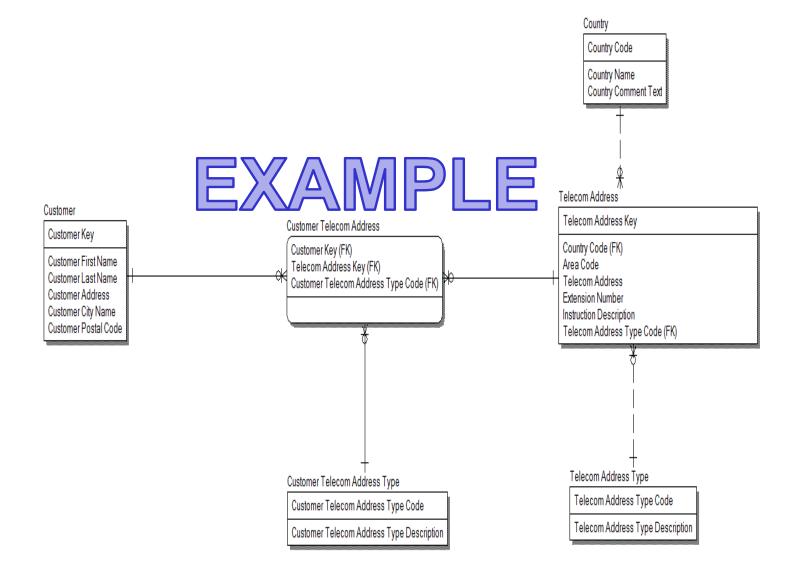
Conceptual Data Model Example



Logical Data Model

- Describes data requirements in more detail than the conceptual level data model
 - Designed based on the conceptual data model
 - ➤ Identifies **business requirements**, including attribute definitions to support business requirements
 - Definitions should be documented and accessible across organization
- Designed and developed independently from the DBMS includes business terms, business relationships
- Each entity has associated attributes, and each attribute has one location in the logical data model
 - Data attributes include datatypes and usual length, derived from business requirements

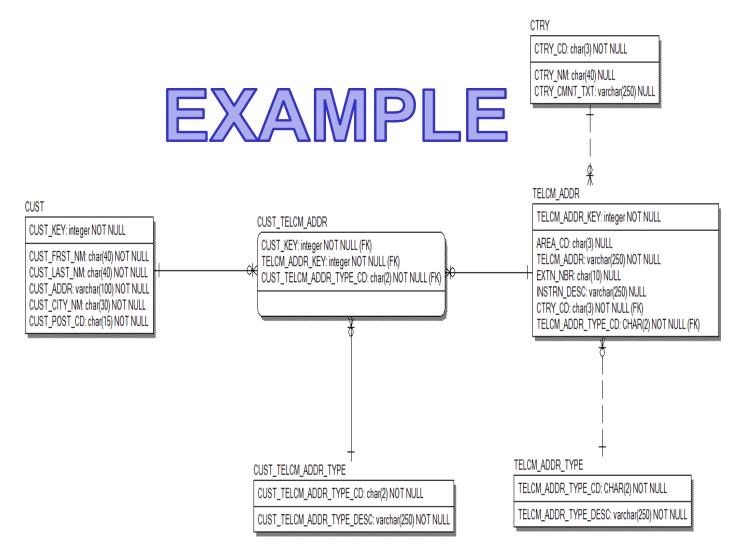
Logical Data Model Example



Physical Data Model

- Describes data to be used in an application with a specific database management system (DBMS)
 - Location, data storage or technology to be used in the project are included in physical data model content
- Columns (attributes) should have exact datatypes, lengths assigned, and default values
 - Each table (entity) may contain columns that are repeated for performance or application reasons
 - > Table and column names may be abbreviated based on modeling standards
- Primary and foreign keys (drawn from logical model), views, indexes, access profiles, and authorizations, etc. are defined at physical level

Physical Data Model Example

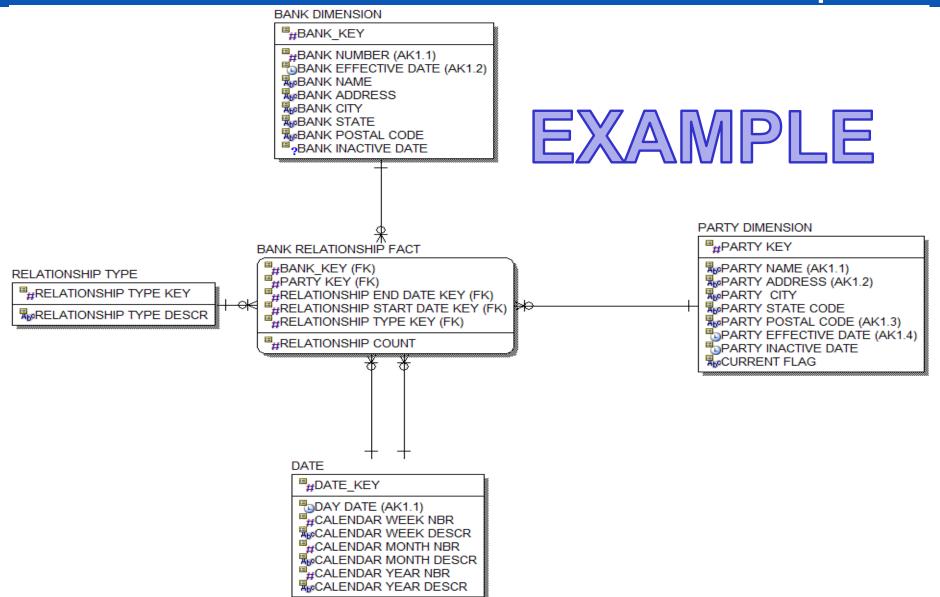


Dimensional Data Model

- Describes data in a format optimized for data storage in a data warehouse or data mart, used for reporting and analytics
- Optimizes the database for faster retrieval of data, using "fact" and "dimension" tables
 - > Facts are the measurements/metrics or facts from a business process. Example: quarterly sales number
 - ▶ Dimensions provide context for a business process event; who, what, where of a fact. Example: for the fact quarterly sales number, dimensions include customer names, customer location, product name
- Structure used by BI query, reporting and analytical tools
 - Dimensional models have specific advantages for reporting and analysis, arranging data so it is easier to retrieve information and generate reports

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Dimensional Model Example



Some Data Modeling Best Practices for Analytics

Data Modeling and Data Analysis

- Data modeling is different from data analysis. Well-designed and implemented data models optimize data to support data analysis (operational or analytical)
- □ Developing data models in all stages (conceptual, logical, physical, dimensional) enables improved understanding of data and supports effective data analysis
 - Developing a dimensional data model based on the logical data model can reduce time for development and improve usability of data
- Data definition, part of best practices in data modeling, is an essential component of effective analytics
 - Tables and attributes without definitions are counterintuitive
 - Data catalog, business glossary, data dictionary can support all forms of analytics

Human Factors in Data Modeling



Business domain subject experts – provide requirements to data modeler / data architect, validates data model content, may include data analysts and data scientists Data modeler / data architect – develops conceptual and logical data models from business requirements, supports physical and dimensional database creation

Database developer, database administrator, programmer – builds physical and dimensional data structures from data models and related content

Data Models for Analytics

- ☐ Requirements for queries and reports will change, models must reflect changing needs with relevant data and analytical processes
- Optimizing performance is a primary goal of analytically-focused data models; techniques include:
 - De-normalizing data from logical data model to physical version can support improved access for queries and reports; denormalized database may not resemble normalized logical data model
 - Creating views based on specific and repeated business needs
 - Using pre-calculated results in appropriate tables, appending detail data to master record table, redundant data storage, etc.
 - > Analytics and reporting are not data mining. Data mining uses data structures unique to a specific set of questions or a set of relationships under exploration

Data Modeling at Scale

Data Modeling at Scale

- □ Data modeling decisions affect performance and usability of data
 - Initially
 - Over time
- ☐ Effective data architecture and data modeling are key aspects of performance improvements, operations effectiveness, and user access to data for analytics
 - > Comprehensive, pragmatic data model facilitates BI tool usage for business value
- Scalability of any application is based, in part, on the data architecture and the content of each data model
 - Cannot "tune" around a poorly designed data architecture and inaccurate data models
 - Business requirements must be documented and understood for effective data models regardless of application/database scale
 - > Accurate and validated data definitions are essential for scalable data models and applications

Design for Scalability with Data Models

- Initial investment in proper enterprise data architecture and accurate data models is essential to scaling any application (operational, analytical)
 - Application/database scale may start small, can grow properly only with clear requirements and accurate data models
- Overly complex architecture and "clever" data models often do not scale well, and negatively affect performance and data usage
 - Clear, well-documented, and business-focused data models offer best opportunities to support scaling application and/or database as needed
- Design for practicality while following best practices and industry standards for effective enterprise and application data architecture with accurate data models

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Proven Strategies for Scalability with Data Models

Define	Define and document data architecture and data modeling standards for the enterprise
Communicate	Communicate standards to organization – clearly and consistently – expect them to be followed
Train	Train business analysts and application developers to work with data models to reduce communication barrier with SMEs and data architects / data modelers
Incorporate	Incorporate data modeling practices and techniques into all business subject areas to maintain enterprise view while supporting subject area's requirements
Capture and store	Capture and store all relevant metadata centrally and consistently
Use	Use the data to test the data model's accuracy – validate each model before proceeding to next stage
Insist on	Insist on conceptual and logical data modeling (before physical data model design) for all business areas throughout the lifecycle to support actual business requirements
Follow	Follow release management process consistently to manage and document data model changes

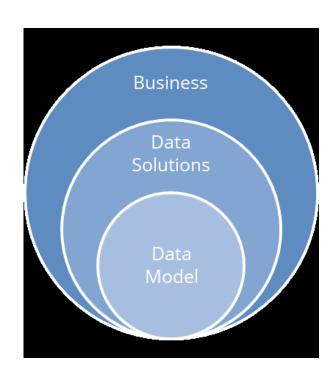
EMSolutions Challenges with Scaling Traditional Data Models

- Data variety, especially using big data and analytics, can offer challenges to data modeling practices
 - Data models traditionally focused on only structured, relational data
 - > Big data and analytics requirements can (and should be) part of a data model
- Organizations think that time-to-delivery value is a concern for developing the data model forms
 - Data modeling does not have to be a lengthy and tedious effort skill and best practices
 - Most organizations focus on development and struggle to see value in staged data modeling
 - Conceptual
 - Logical
 - Physical
 - Dimensional
- Many traditional data model processes are cumbersome and can be streamlined to reduce challenges
 - Building conceptual and logical data models before physical can identify data requirements from business perspective to support effective database and application access
 - > Documenting data definitions is essential and can be done during data model development
- Perception of data modeling value and lack of data modeling skills are major challenges in 21st century organizations

Future of Data Modeling

Fundamental Pillar of 21st Century Businesses

- ☐ High-quality data foundation is important for success in digital technologies and business competitive advantage
 - ➤ High data quality based on underlying structure that connects an organization's data assets from business requirements
 - Business case for all development should include all stages of data models
- Starting with an effective data model set is essential for enabling high-quality data that can transform an organization
 - Advanced analytics
 - Self-service business intelligence and reporting capabilities
 - Artificial intelligence and machine learning
 - Other digital business applications



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Data Modeling Success Tips

- ☐ Identify business goals for data with each initiative do not focus only on processes
 - Develop a standard approach to enterprise data architecture that includes all stages of data models and related documentation
 - > Develop good business use cases for data modeling to demonstrate effectiveness and value of the technique
 - Focus on business needs and data assets from project inception
- Select the right data modeling approach, without sacrificing the essential model stages (conceptual, logical, physical, dimensional)
- ☐ Start with the organization's subject area model to identify the relevant subjects as focus
 - Base approach on the business use case and underlying technology
 - > Design each model to the appropriate level of detail for the business requirements
- □ Data model must fully address the requirements, but not over-engineer them
- Seamless and open communication across all teams and stakeholders are essential for successful data modeling
 - Ensure all relevant data elements are incorporated and have the same meaning and interpretation across the organization, with supporting documentation
- Remember a data model is a living artifact that requires updating and maintenance
 - Creating a formal process for updating all models is critical, including updating model documentation

21st Century Data Modeling Skills



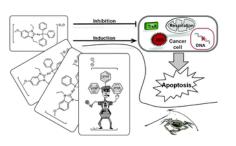
Conceptual and logical thinking and design



Critical thinking and analysis



Effective communications across business and IT



Abstract thinking and detailed analysis capabilities



Documentation skills



Data representation and definition



Data modeling tool competency



Adaptability and teamwork

Future Data Modeling Trends and Techniques



Agile Data Modeling: "just-in-time data modeling" using "best practices with a minimally sufficient design" and "the right data model for specific situations" for a mix of unstructured data, relational data, master data, and dimensional data



Graph Databases: Graph databases provide easy, quick, visuals of business cases, and they need data models



Business Intelligence and Reporting: Data modelers will continue to diagram and document relational databases, focused on specific business and technical use cases



Cloud-based Computing: Data modeling for cloud applications focuses on scalability, speed, and reduced cost. Cloud computing will encourage "larger applications that circulate more data through their hardware", and require well-documented data models and effective metadata management



Wider Focus: data models must address and document multi-level concerns: business level, solution level, and implementation challenges

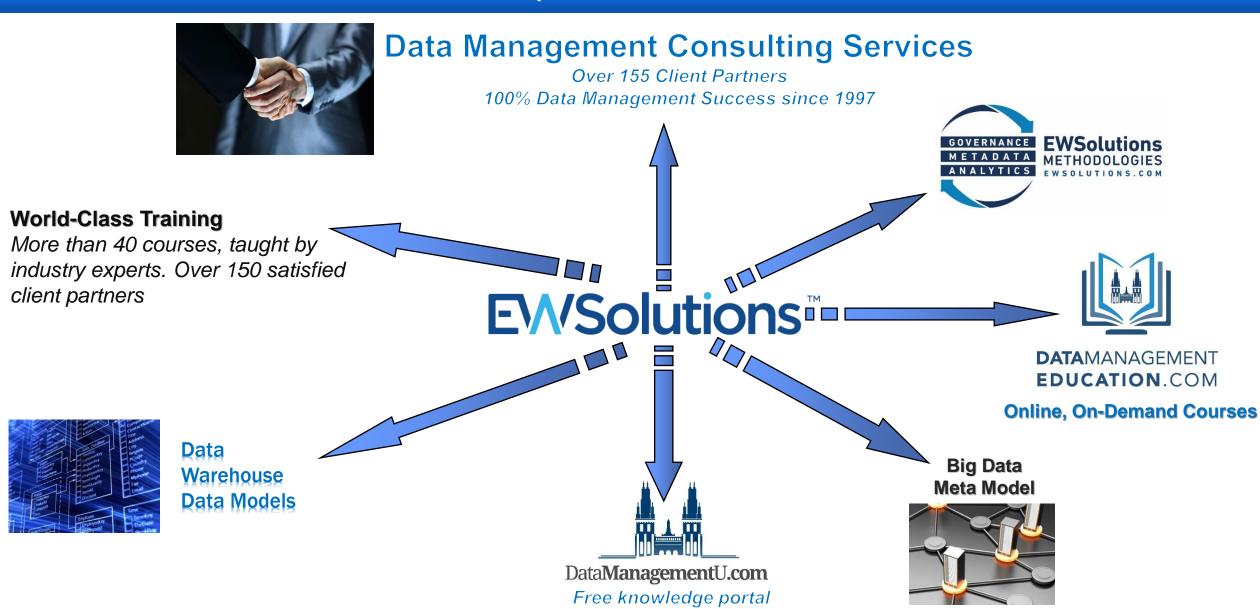
In the Final Analysis

In Conclusion

- □ Data modeling is an essential component for any application / system operational, analytical, etc.
 - > Data modeling is a component of an organization's enterprise data architecture
- Adopting data modeling best practices can offer a competitive advantage for organizations that value data as an organizational asset
 - > Data modeling techniques should be practiced consistently throughout organization
- There are many aspects to effective data modeling, including data management, metadata management, data architecture, etc.
- ☐ Building and sustaining an effective data modeling competency requires commitment across the organization, from business and IT
- ☐ Continuous learning is a crucial component of data modeling competency

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Questions & Answers



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