



Data-Warehouse-, Data-Mining- und OLAP-Technologien

Data Warehouse Design

Bernhard Mitschang Universität Stuttgart

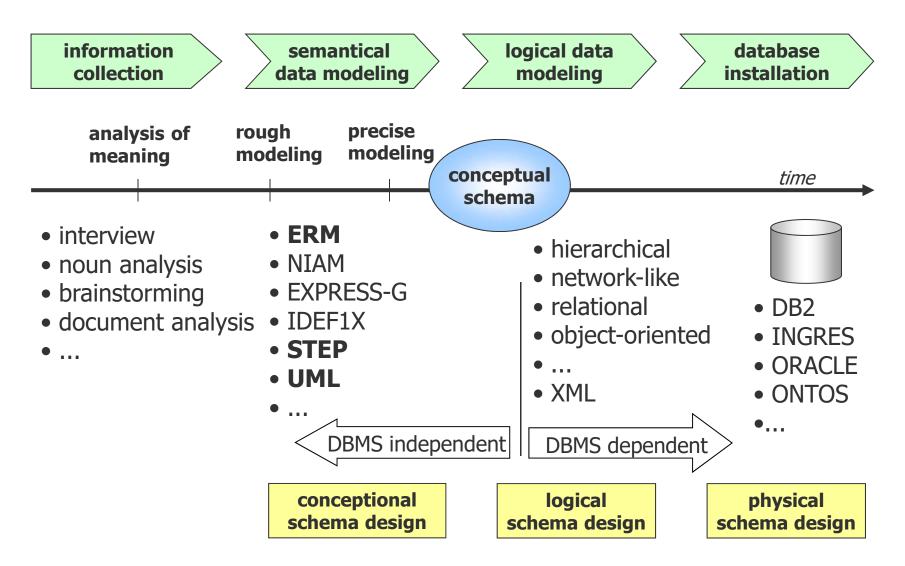
Winter Term 2017/2018

Overview

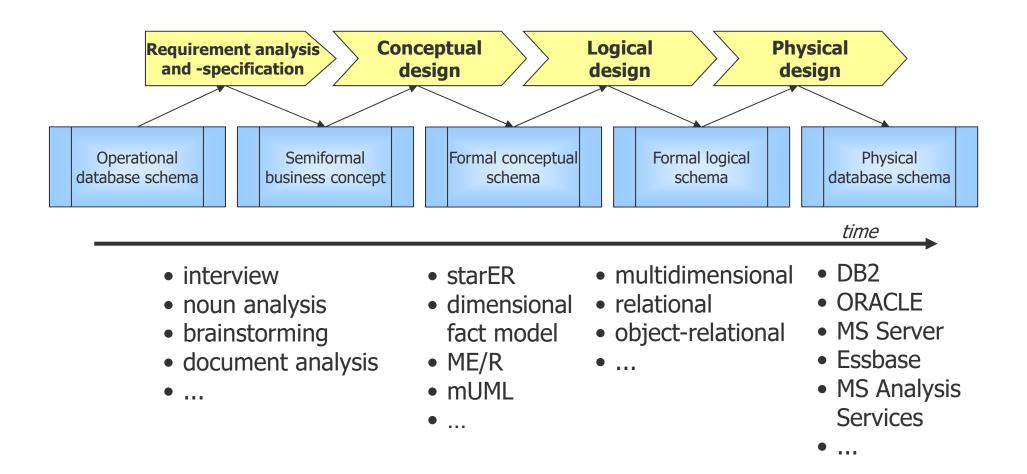
- Data Warehouse Design Process
 - Conceptual Design
 - Logical Design
 - Details of Logical Design
 - Physical Design

Database Design

Process Model



Data Warehouse Design Process



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Overview

- Data Warehouse Design Process
- Conceptual Design
 - Multidimensional Model
 - Dimensional Fact Model
 - starER
 - UML Profile for Multidimensional Modeling
 - Logical Design
 - Details of Logical Design
 - Physical Design

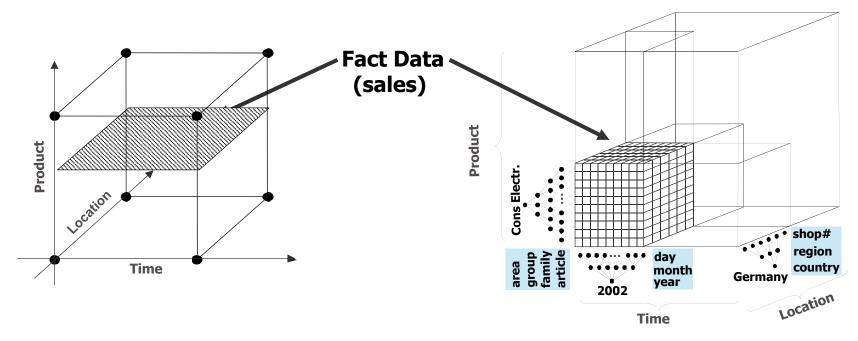
Conceptual Design

Transformation of the semi-formal business requirements specification into a formalized conceptual multidimensional schema

- Most approaches produce a graphical multidimensional schema Dimensional Fact Model, starER, UML-based, ...
- Conceptual Design is based on
 - business requirement specification
 - ER schema of the operational systems
- Process model to conceptual data warehouse design
 - context definition of measures
 - dimensional hierarchy design
 - definition of summarizability constraints

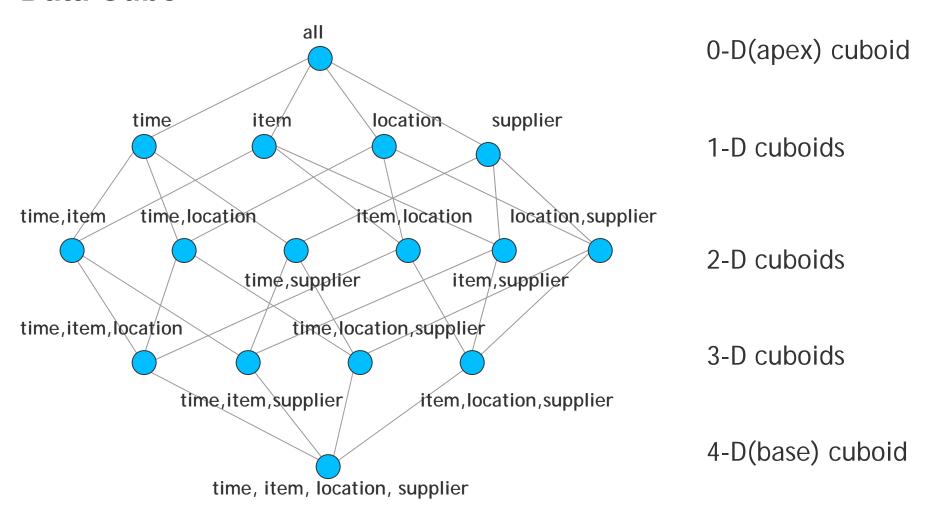
Multidimensional Model

"Cube" Metaphor



- A data warehouse is based on a multidimensional data model which views data in the form of a data cube. A data cube, such as sales, allows data to be modeled and viewed in multiple dimensions.
- In data warehousing literature, an n-D base cube is called a base cuboid. The top most 0-D cuboid, which holds the highest-level of summarization, is called the apex cuboid. The lattice of cuboids forms the data cube.

Data Cube



 Example Data Cube for SALES modeled and viewed in multiple dimensions (item, time, location, and supplier)

Basic Elements of a Conceptual Model

Fact data

- Mostly numeric data that is observed or measured
- Example: turnover/sales, number of pieces, ...

Qualities

- Represent a state, a status or a mode
- Cannot be aggregated (in contrast to fact data)
- Example: shipping mode, status

Attributes

- Describe dimension objects
- Mostly text-based descriptions
- Example: product descriptions, customer profiles, addresses

Dimensions

- Strongly associated attributes
- Typically 5 to 20 attributes
- Showing mostly hierarchical relationships
- Example: product information, customer information, time references

Conformed Dimensions

A conformed dimension is a dimension that means the same thing with every possible fact to which it can be joined

- A conformed dimension is identically the same dimension in each data mart
- Conformed dimensions support
 - a single dimension can be used against multiple facts in the same database space
 - user interfaces and data content are consistent whenever the dimension is used
 - there is a consistent interpretation of attributes and, therefore, rollups across data marts

Dimensional Fact Model (DFM)

- Part of work on a complete and consistent design methodology
- Multidimensional conceptual model including a graphical representation
- Representation of reality consists of a set of fact schemes
- Basic elements of a fact scheme f = (M, A, N, R, O, S) are facts, measures, attributes, dimensions, and hierarchies
- Other features which may be represented on fact schemes are the additivity of fact attributes along dimensions, the optionality of dimension attributes, and the existence of non-dimension attributes
- The multidimensional model may be mapped on the logical level differently depending on the underlying DBMS

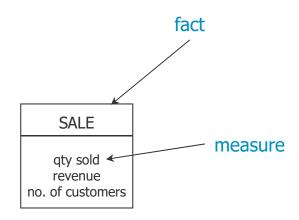
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Basic Definitions

- Let g=(V, E) be a directed, acyclic and weakly connected graph
- g is a quasi-tree with root in $v_0 \in V$ if each other vertex $v_j \in V$ can be reached from v_0 through at least one directed path. path_{0j}(g) \subseteq g denotes a directed path starting in v_0 and ending in v_i .
- Given v_i∈path_{0i}(g),
 - path_{ij}(g)⊆g denotes a directed path starting in v_i and ending in v_j,
 - sub(g,v_i)⊂g denotes the quasi-tree rooted in v_i≠v₀.

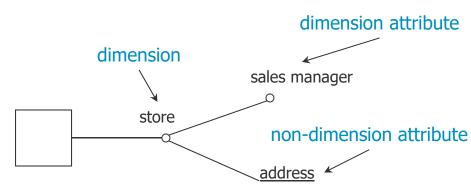
Facts and Measures

- A fact is a focus of interest for the decision-making process. It models an event occurring in the enterprise world (e.g. sales and shipments)
- M is a set of measures. Each measure m_i∈M is defined by a numeric or Boolean expression which involves values acquired from the operational information systems
- Graphical representation
 - A fact is represented by a box which reports the fact name and, typically, one or more measures



Attributes and Dimensions

- Dimensions are discrete attributes which determine the minimum granularity adopted to represent facts
- A is a set of dimension attributes
 - Each dimension attribute a_i∈A is characterized by a discrete domain of values, Dom(a_i)
- N is a set of non-dimension attributes
 - A non-dimension attribute contains additional information about a dimension attribute and is connected by a -to-one relationship
 - Unlike dimension attributes, non-dimension attributes cannot be used for aggregation
- Graphical representation
 - Dimension attributes are represented by circles, non-dimension attributes by lines

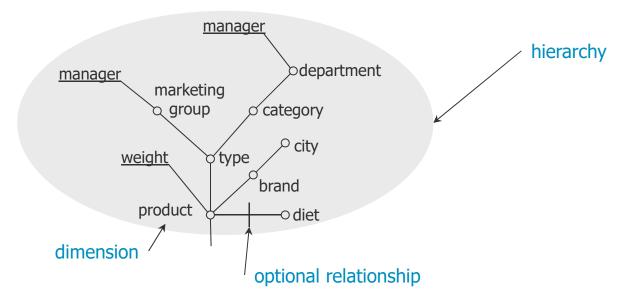


Hierarchies

- Hierarchies are made up of discrete dimension attributes linked by -toone relationships, and determine how facts may be aggregated and selected significantly for the decision-making process
 - R is a set of ordered tuples, each having the form (a_i, a_j) where $a_i \in A \cup \{a_0\}$ and $a_j \in A \cup N$ $(a_i \neq a_j)$, such that the graph $qt(f) = (A \cup N \cup a_0, R)$ is a quasi-tree with root a_0
 - a₀ is a dummy attribute playing the role of the fact f on which the scheme is centered. The couple (a_i, a_j) models a -to-one relationship between attributed a_i and a_j
 - Each element in Dim(f) = {a_i∈A | (a₀, a_i)∈R} is a dimension
 - The hierarchy on dimension d_i∈Dim(f) is the quasi-tree rooted in d_i: sub(qt(f), d_i)
 - O \subset R is a set of optional relationships. The domain of each dimension attribute a_j such that $\exists (a_i, a_j) \in O$ includes a NULL value

Hierarchies

- Graphical representation
 - Subtrees rooted in dimensions are hierarchies. The arc connecting two attributes represents a -to-one relationship between them.
 - Optional relationships are represented by marking with a dash the corresponding arc.
- Example



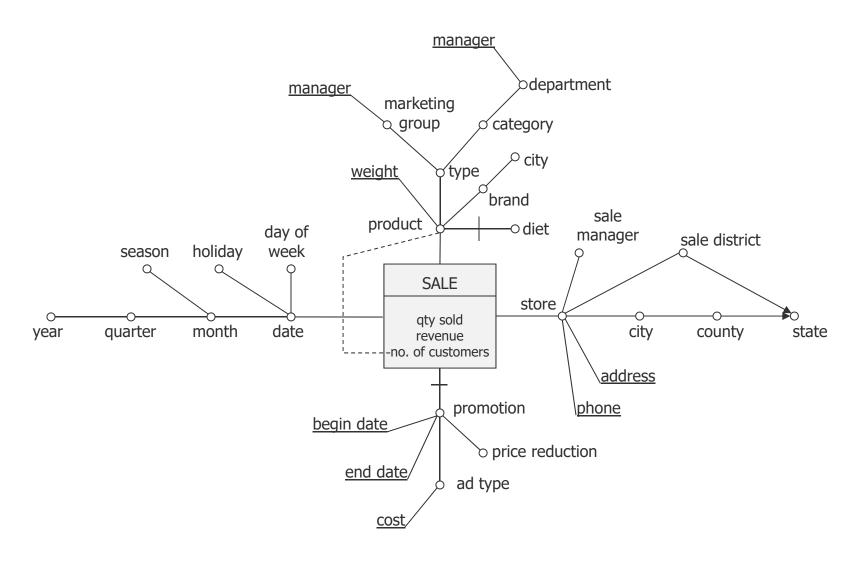
Aggregation

- S is a set of aggregation statements
 - Each aggregation statement consists of a triple (m_j, d_i, Ω) where m_j∈M, d_i∈Dim(f) and Ω∈{'SUM', 'AVG', 'COUNT', 'MIN', 'MAX', 'AND', 'OR', ...} (aggregation/grouping operator)
 - Statement $(m_j, d_i, \Omega) \in S$ declares that measure m_j can be aggregated along dimension d_i by means of the operator Ω
 - If no aggregation statement exists for a given pair (m_j, d_i), then m_j cannot be aggregated at all along d_i

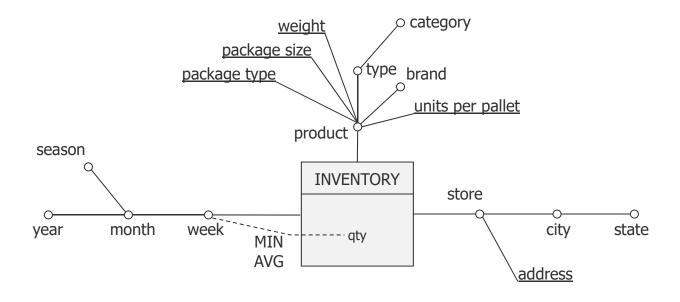
Aggregation

- A measure is additive on a dimension if its values can be aggregated along the corresponding hierarchy by the 'SUM' operator
- Graphical representation
 - If $(m_j, d_i, 'SUM') \notin S$, m_j and d_i are connected by a dashed line labelled with all aggregation operators Ω such that $(m_i, d_i, \Omega) \in S$
 - If (m_i, d_i, 'SUM')∈S
 - If $\neg \exists \ \Omega \neq 'SUM' \mid (m_i, d_i, \Omega) \in S$, m_i and d_i are not graphically connected
 - Otherwise, m_j and d_i are connected by a dashed line labelled with the symbol '+' followed by all the other operators $\Omega \neq$ 'SUM' such that $(m_j, d_i, \Omega) \in S$

SALE Fact Schema



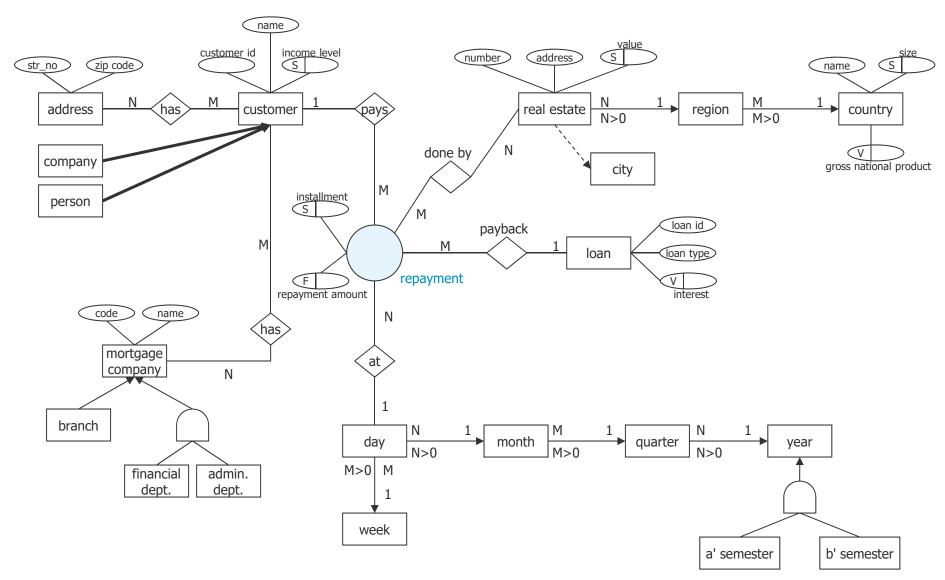
INVENTORY Fact Schema



starER

- Combines the star structure with the semantically rich constructs of the ER model
- Adds special types of relationships to support hierarchies
- starER vs. DFM
 - starER allows many-to-many relationships between dimensions and facts
 - starER allows objects participating in the data warehouse, but not in the form of a dimension
 - Specialized relationships on dimensions are permitted in starER (specialization/generalization)

starER: Mortgage Company Data Warehouse



UML Profile for Multidimensional Modeling

- UML profile: extend UML by
 - stereotypes represented by icons

e.g





- tagged valuese.g., {name = value}
- constraints, e.g.,
 {Quantity Is Not SUM Along Salesperson}

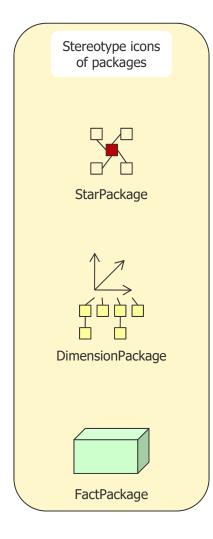
Main characteristics

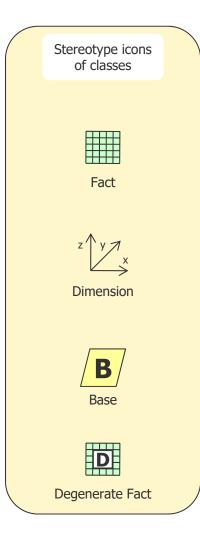
- Accurate
 - major features of multidimensional models supported
- No redundancy
 - concepts and elements are only defined once in the model and imported where needed
- Simplicity
 - minimal subset of UML and minimal extensions
- Understandable
 - define three levels of abstraction
 - use packages as grouping mechanism
 - support conformed dimensions

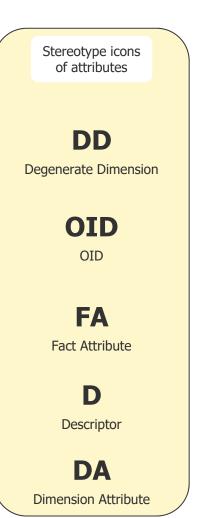
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UML Profile

Overview







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Three Levels of Detail

Level 1 Model definition

- A package represents a star schema
- A dependency between two packages indicates that star schemas share at least one dimension

Level 2 Star schema definition

- A package represents a fact or a dimension
- A dependency between two packages indicates that they share at least one level of the dimension hierarchy

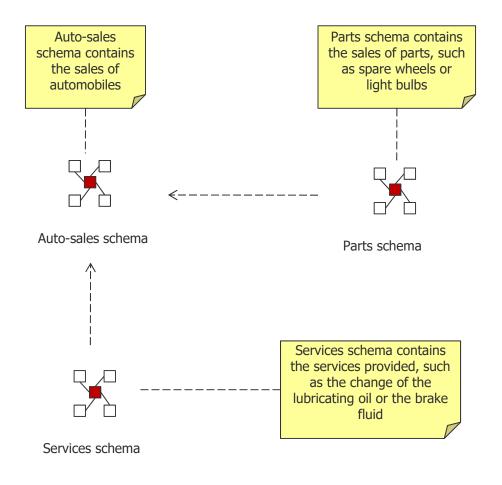
Level 3 Dimension Definition fact definition

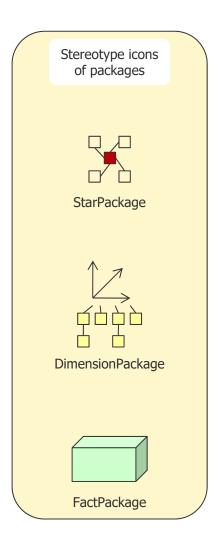
- Set of classes representing the hierarchy levels, or
- set of classes representing the entire star schema

Running Example

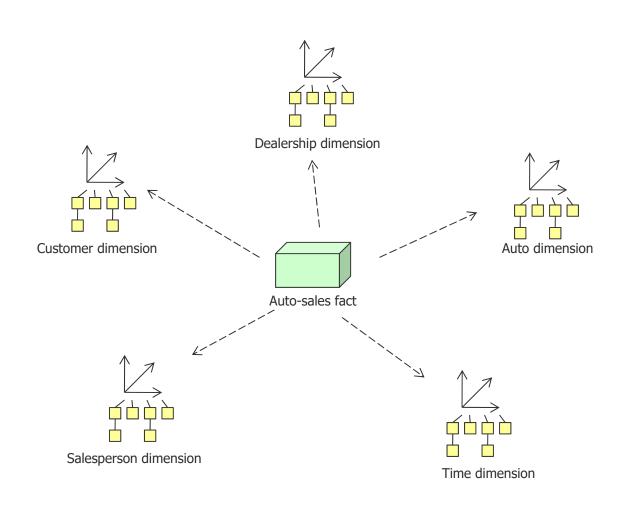
- Company having several dealerships that sell automobiles (cars and vans) across several states
- Data warehouse should support analysis of
 - sales of automobiles
 - sales of parts such as spare wheels or light bulbs
 - service works such as change of lubricating oil or brake oil

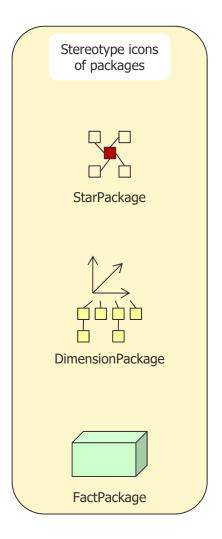
Level 1: Model Definition



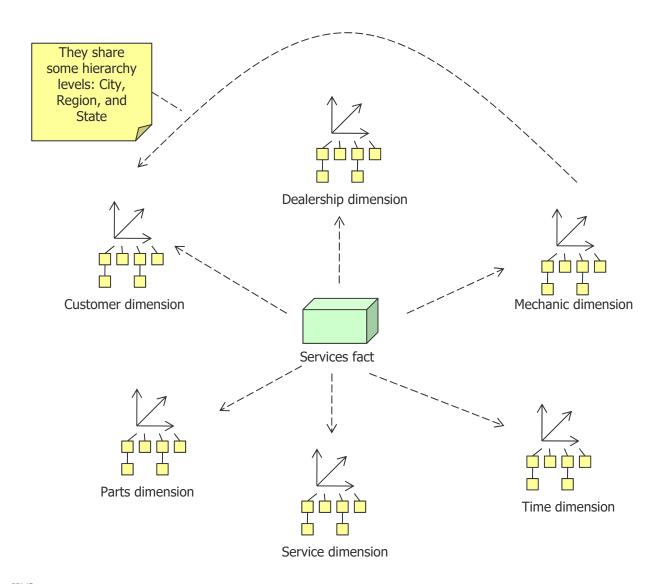


Level 2: Auto-Sales Schema



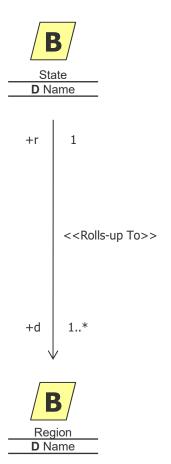


Level 2: Services Schema

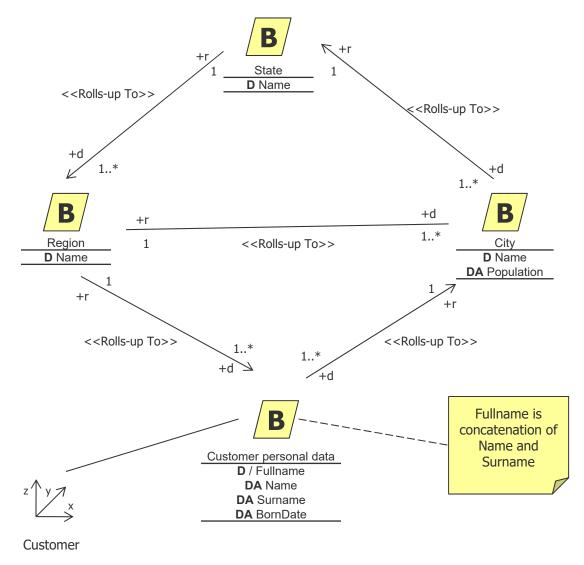


Dimension Hierarchies

- Classification hierarchy levels are specified by base classes
- Associations show hierarchical relationship
 - · drill-down direction is marked by +d
 - roll-up direction is marked by +r
 - default roll-up or drill-down directions are marked by arrows
- Multiplicity may be defined for each association
- Various attribute types
 - DA: dimension attribute
 - **D**: description
 - OID: object id



Level 3: Customer Dimension



Types of Classification Hierarchies

Strict hierarchy

Each object of the lower level (+d) belongs to only one object at the higher level (+r).



Non-strict hierarchy

An object of the lower level (+d) may belong to several objects at the higher level (+r).



Completeness for drill-down

For each object at the higher level there exists an object at the lower level.



Completeness for roll-up

For each object at the lower level there exists an object at the higher level.

Non-strict and incomplete hierarchies may yield to inconsistent totals!

not complete!

Summarizability Issues

Drill-Down



Country	Sales
Germany	50
France	40
Andorra	15

Total = 105

Region	Sales
NRW	20
BW	10
BAY	20
Paris	30
Province	10

• Roll-up Total = 90





+r +d

CategorySalesDairy55Bakery25

Category

Total = 80

Product	Sales
Milk	15
Butter	10
Bread	25
Yogurt	30
Glasses	100

Non-strict



Season	Sales
Winter	560
Spring	500
Summer	100

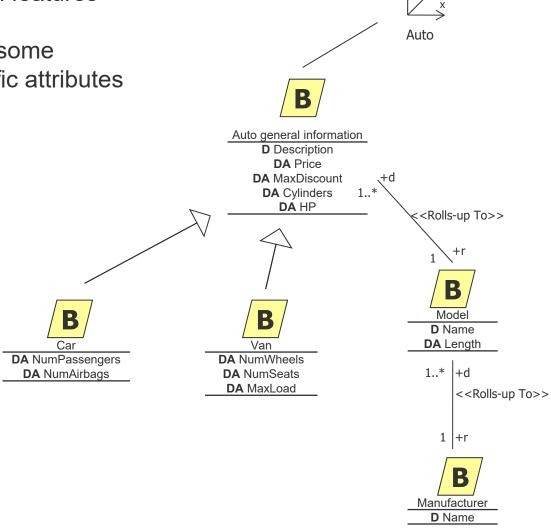
Total = 1160

Month	Sales
Jan	200
Feb	180
Mar	180
Apr	160
May	160
Jun	100

Total = 980

Categorization Hierarchies

 Allows to model additional features for subtypes of a class, e.g., cars and vans have some common and some specific attributes



Level 3: Auto-Sales Fact

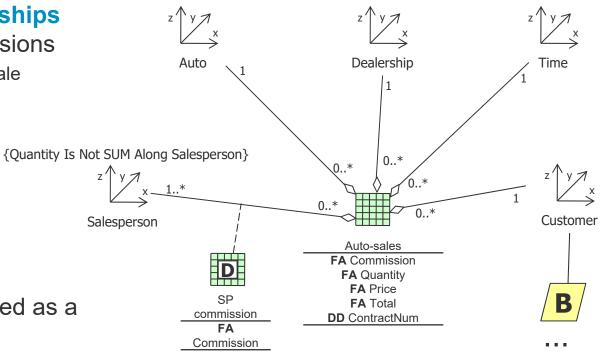
- Many-to-many-Relationships between facts and dimensions
 - Several salespersons for a sale

 Summarizability is defined as a constraint





- for m:n-relationships between facts and dimensions
- specific attribute that is provided for each instance combination in such a relationship



Degenerate Dimension

- Most of the properties are already presented by other elements (facts, dimensions)
- Remaining attributes are necessary to uniquely identify fact instances.

Important Features of a Conceptual Model

- Explicit hierarchies in dimensions
 Multiple hierarchies in a dimension
 Non-strict and non-complete hierarchies
 Conformed dimensions

 Dimensions
- Support correct aggregation (summarizability)
 Degenerated facts
 Degenerated dimensions

 Facts

Facts and Dimensions

- Many-to-many relationships between facts and dimensions
- Symmetric treatment of dimensions and measures

Others

Changes of data over time

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- Reflect uncertainty
- Different levels of granularity

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Sources: [PJ99, LTS06]

Overview

- Data Warehouse Design Process
- Conceptual Design
- → Logical Design
 - TPCH Benchmark Schema
 - Star Schema
 - Snowflake Schema
 - Informix Schema
 - Details of Logical Design
 - Physical Design

Logical Design

Convert the conceptual schema to a logical one with respect to the target logical data model

- Logical Design is based on
 - conceptual diagrams
 - summarizability constraints
 - transformation rules
- Logical data models
 - relational
 - multidimensional

Sample Scenario

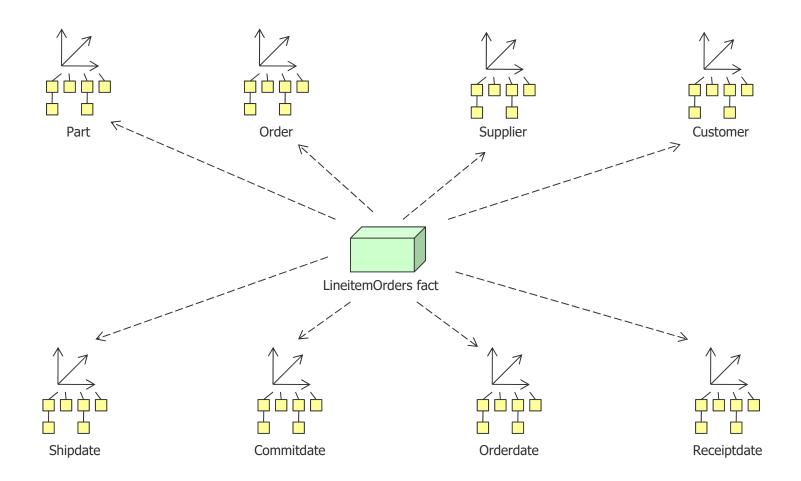
- Company that manages, sells, or distributes a product worldwide (e.g., car rental, food distribution, parts, suppliers, etc.).
- Scenario taken from TPC-H benchmark



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- Each order entry refers to a part, an order, a customer, a supplier and the dates for the ordering, the commitment, the shipment and the receipt
- Order entries are further characterized by a line number, a status, shipment instructions and a return flag

Conceptual Schema: Overview



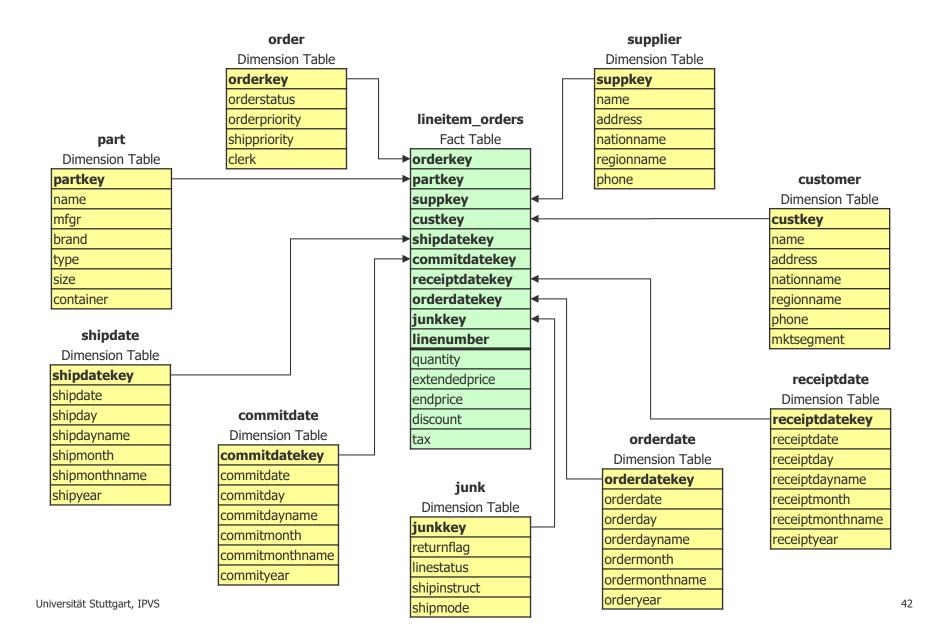
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Logical Schema Types

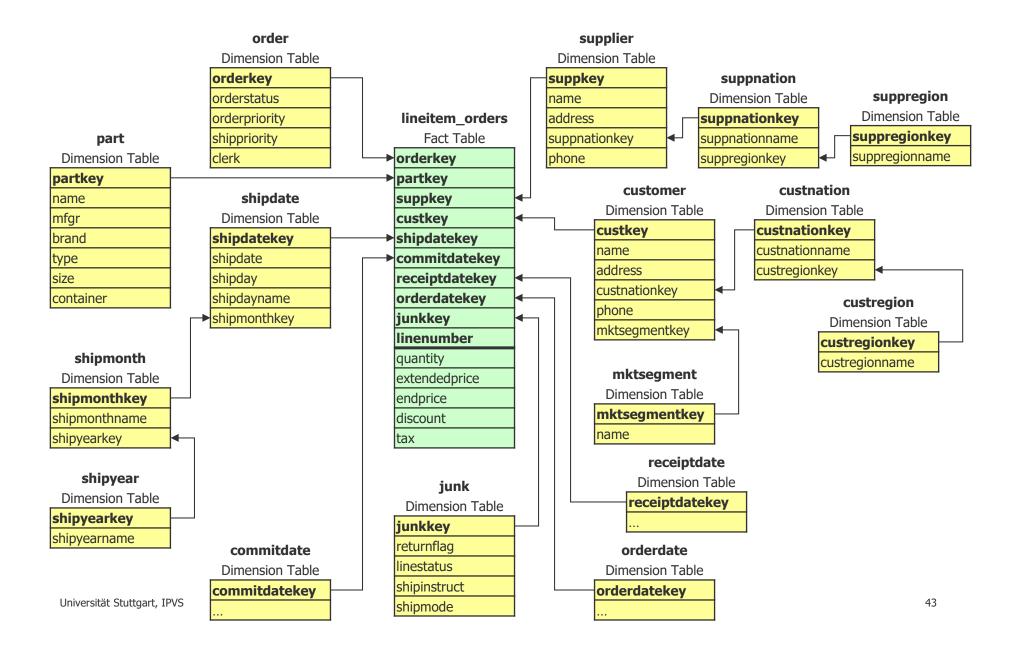
Star Schema

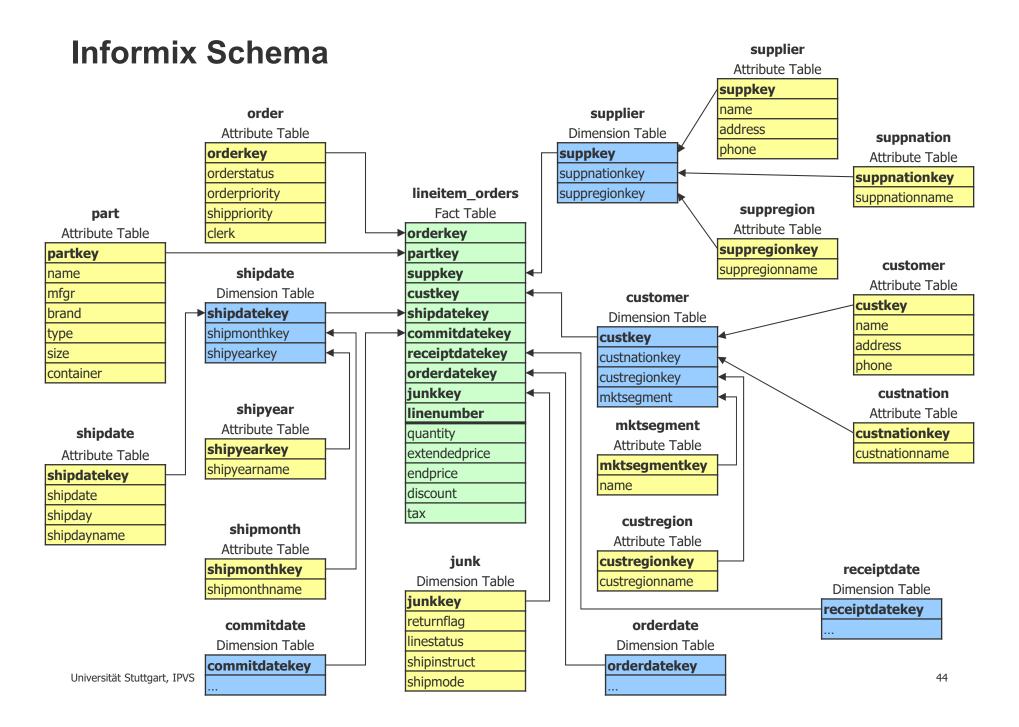
- Each dimension is modeled by a single table
- Redundancy within the dimension tables, e.g. nationname, regionname
- Qualities are combined into one or a few 'junk' dimensions
- Snowflake Schema
 - Dimensions get normalized (3NF)
 - One dimension table per hierarchy level
- Informix Schema
 - Normalized attribute tables
 - Hierarchies are partly represented in dimension tables

Star Schema



Snowflake Schema





Comparing Logical Schema Types

- Is all the information of the conceptual model representable in the logical schema?
 - Explicit hierarchies in dimensions
 - Multiple hierarchies in a dimension
 - Non-strict and non-complete hierarchies
 - Many-to-many relationships between facts and dimensions
 - Symmetric treatment of dimensions and measures
 - Support correct aggregation (summarizability)

- What effort is needed to reflect changes of the conceptual design in the logical schema
 - insert hierarchy level
 - delete hierarchy level
 - insert measure
 - delete measure
 - insert dimension
 - delete dimension
 - modify granularity

→ logical schema?

→ metadata?

Comparison

	Star Schema	Snowflake Schema	Informix Schema
Clearness	one table per dimension	multiple tables per dimension	multiple tables per dimension
Redundancy	in the dimension table	normalization takes care of	normalization takes care of
Data Volume	high(er)	low	low
Hierarchies	not represented	represented in dimension tables	only one hierarchy
Summarizability	-	-	-
Adding a Dimension	one additional table	several additional tables	several additional tables
Adding an Attribute	one attribute appended to dimension table	changes to several tables	changes to several tables
Performance (Queries)	max. one join per dimension	several joins among the dimension tables	several joins among the dimension tables

Overview

- Data Warehouse Design Process
- Conceptual Design
- Logical Design
- Details of Logical Design
 - Extended Dimension Table Design
 - Production keys / surrogate keys
 - · Roles of Dimensions
 - Hierarchies
 - Slowly changing dimensions
 - Time stamping in large dimensions
 - · Large dimensions with frequent changes
 - Many-to-Many Dimensions
 - Time Dimensions
 - Extended Fact Table Design
 - Modeling Events and Coverage (Factless Fact Tables)
 - Multinational Currency Tracking
 - Physical Design

Production Keys / Surrogate Keys

- Production key: Attributes that make up a primary key in the production system (source system)
 - keys may be reused in the production system
 - rules for building production keys may change
 - production keys are kept although the dimension has changed
- Surrogate key: Single attribute (INTEGER) that is used as primary key in the data warehouse
 - results in small keys
 - surrogate keys have no meaning
 - keys may be changed independent of the source system (this is especially important for type two slowly changing dimensions)

Roles of Dimensions

- In some situations a single dimension appears several times in the same fact table
- Example
 A fact table on customer orders
 may include
 - Order date
 - Packaging date
 - Shipping date
 - Delivery date
 - Payment date
 - Return date
 - Order status
 - Customer
 - ...

- Problems
 - using the same dimension for each role may result in wrong query results
 - result may contain many columns with identical names

time in several roles

Single Dimension Tables for all Roles

SELECT ...

FROM Orders AS O,

Time AS T

WHERE O.OrderDate = T.TimeID

AND O.PackagingDate = T.TimeID

AND O.ShippingDate = T.TimeID

AND O.OrderStatus = 'completed'

GROUP BY...

SELECT ...

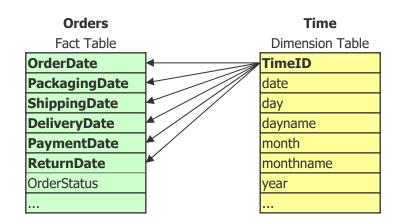
FROM Orders AS O,

Time AS OT, Time AS PT, Time AS ST

WHERE O.OrderDate = OT.TimeID

AND O.PackagingDate = PT.TimeID
AND O.ShippingDate = ST.TimeID
AND O.OrderStatus = 'completed'

GROUP BY...



Single Dimension Table for each Role

SELECT ...

FROM Orders AS O,

OrderTime AS OT, PackTime AS PT, ShipTime AS ST

WHERE O.OrderDate = OT.OrderTimeID

AND O.PackagingDate =

PT.PackTimeID

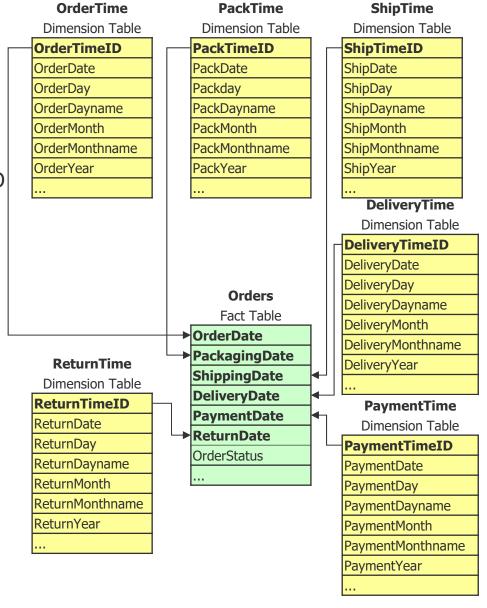
AND O.ShippingDate =

ST.ShipTimeID

AND O.Orderstatus = 'completed'

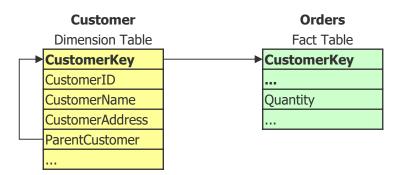
GROUP BY...

Role-playing dimension tables can be provided as physical tables or as views



Hierarchies

- Organization hierarchies
- Parts explosion hierarchies



- Compact way to represent the hierarchy
- Structure results in complex queries

Goals

- Keep the original grain of the customer dimension
- Support aggregation for the entire hierarchy
- Support aggregation for the immediate subsidiaries as well as for the lowest-level subsidiaries
- Support the efficient search for the immediate parent or the top-most parent

Bridge Table for Hierarchies

- Dimension table and fact table are kept
- Bridge table contains
 - All combinations of CustomerKeys and all their SubsidiaryCustomers
 - Path-length of all combinations including paths of length zero
 - A flag identifying the bottom level of the hierarchy

Customer Dimension Table CustomerKey CustomerID CustomerName CustomerAddress

ParentCustomer



CustomerHierarchy

Bridge Table

ParentCustomerKey

SubsidiaryCustomerKey

NumberOfLevels

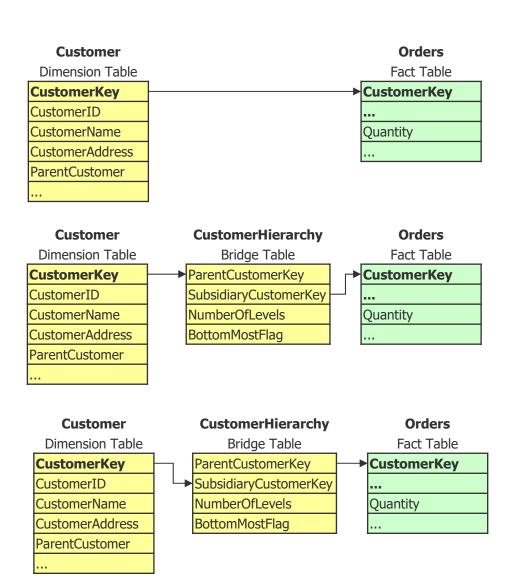
BottomMostFlag

Usage of Bridge Tables

 Aggregation ignoring the customer hierarchy

 Descending the customer hierarchy

 Ascending the customer hierarchy

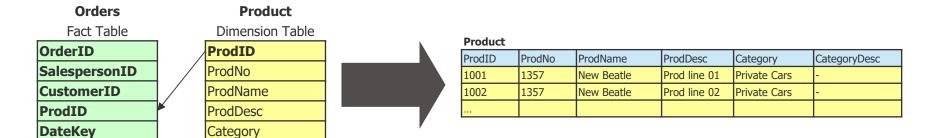


Slowly Changing Dimensions

- Example: The description of a product changes.
- Modeling alternatives

Type One: Substitution of the old description by the new one.

Type Two: Product together with its new description gets a new ID.



Type Three:

CityName

Quantity

TotalPrice

Associated product entry has an additional attribute that holds the old description.

CategoryDesc

UnitPrice

QOH

Product

Slowly Changing Dimensions

	Type One	Type Two	Type Three
Keeps history			
Schema changed			
Volume of data changes			
Effect on queries			
Usage			

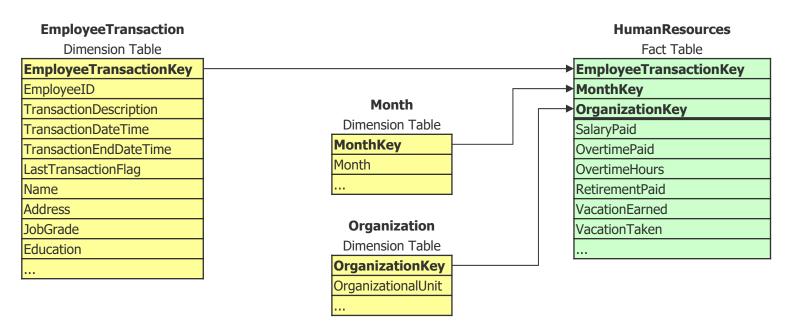
Time Stamping in Large Dimensions

- Example: A human resource department has to store many attributes on many employees and precisely track all the changes on these attributes
- Reports that should be provided
 - R1: Summary status of the entire employee base on a regular (monthly) basis
 - R2: Profile the employee population at any precise instant of time
 - R3: Provide every action taken on a given employee, with the correct transaction sequence and the correct timing of the transaction
- Slowly changing dimensions, type 2
 - R1, R2: Does not provide the valid information for an employee for a certain point in time

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R3: Does not track the transaction time for actions taken on employees

Human Resource Environment



- Employee transaction table contains a complete snapshot of the employee record for each transaction
- Human resource table contains the regular facts stored for each employee each month
- Which tables are needed for reports R1, R2 and R3?

Example

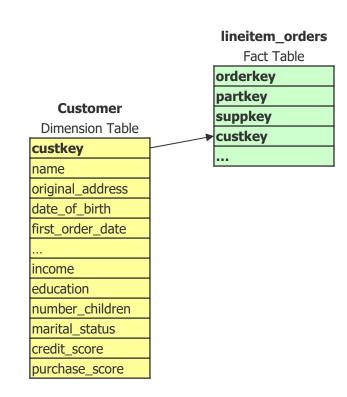
- For an insurance company the customer dimension might hold millions of tuples
- Large amounts of demographic data are kept for each customer
- Customer data change constantly and these changes have to be propagated into the warehouse

Customer

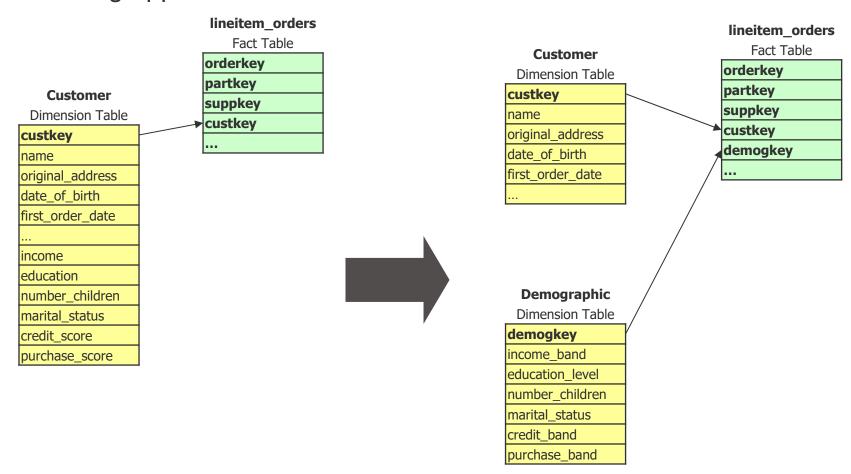
Dimension Table

Difficision rabic		
custkey		
name		
original_address		
date_of_birth		
first_order_date		
income		
education		
number_children		
marital_status		
credit_score		
purchase_score		
-		

- Design goals for this kind of dimensions
 - support rapid browsing, e.g. of low cardinality attributes
 - support efficient browsing of crossconstrained values in the dimension table
 - do not penalize the fact table query for using a large dimension
 - find and suppress duplicate entries in the dimension
 - do not create additional records to handle the changing dimension problem



Modeling approach: Break off dimension into several dimension tables



 One (or several) additional dimension table (Demographics) includes all changing attributes

10000 20000 21000 22000 40000 90000 95000

band 3

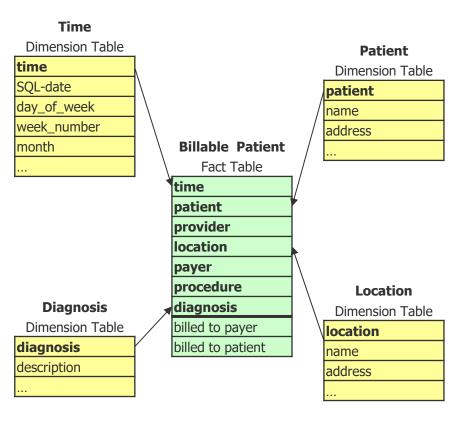
band 4

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band 1 band 2

- Attributes in this new dimension are forced to have a relatively small number of discrete values
 - e.g., income is mapped to income bands
- The additional dimension table
 - is populated with all possible discrete attribute combinations, and
 - comes with its own surrogate key
- The surrogate key of the additional dimension is used in the fact table
- Changes in the demographics dimension are reflected by entries in the fact table

Many-to-Many Dimensions



- Dimensions may have zero, one or many values for a given fact record
- The number of dimension values is not known in advance

Example:

- A fact table record for each line item of a hospital bill or each line item of a treatment performed in a physician's office
- The design has to cover several diagnoses per patient

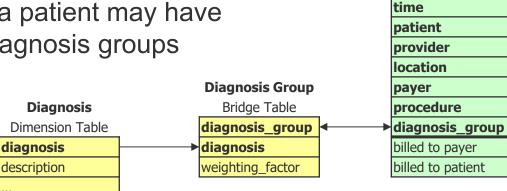
Bridge Tables

- Bridge table between the fact table and the dimension table
- Use a special diagnosis group key in the fact table
- The diagnosis group table contains a set of records for each patient. Each record points to a specific diagnosis and provides a weighting factor
- Over time a patient may have different diagnosis groups

- Questions that can be answered
 - Correctly weighted summary of all charges grouped by diagnosis
 - Impact report that totals the impact each diagnosis has in terms of total amounts associated with that diagnosis

Billable Patient

Fact Table



Time Dimension

- Some transactions need a finescale tracking to the minute or even the second
- Solution 1
 - Time dimension with one record for every minute or second
- Solution 2
 - Time dimension on a daily basis
 - Transaction time is treated as a fact
 - Timestamp may be provided for various time zones

Sales

Fact Table

DateTimeKey
ProductKey
StoreKey
CustomerKey
ClerkKey
RegisterKey
PromotionKey
DollarsSold
UnitsSold

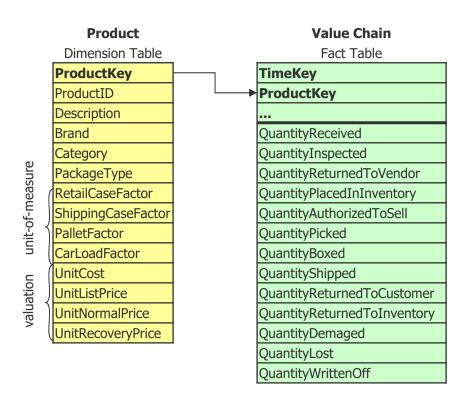
Sales

Fact Table

DateKey		
ProductKey		
StoreKey		
CustomerKey		
ClerkKey		
RegisterKey		
PromotionKey		
TimeOfDay		
GMTTimeOfDay		
DollarsSold		
Donarosora		
UnitsSold		

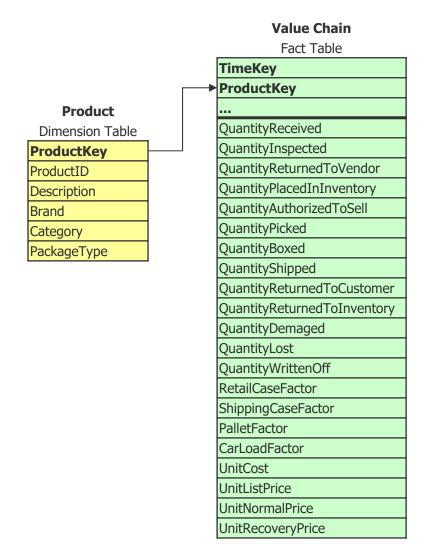
Various Units of Measure

- Several business processes monitor the flow of products or the inventory. For some of these processes numbers should be expressed in different units of measure.
- Example: In the value chain several quantity facts have five different unit-of-measure interpretations and four valuation schemas
- Solution 1
 - The unit-of-measure interpretations as well as the valuation schemas are presented in the dimension table



Various Units of Measure

- Solution 2
 - The unit-of-measure interpretations as well as the valuation schemas are provided in the fact table
- Advantages of this solution
 - Eliminates the possibility of choosing the wrong factors
 - Changes in factors do not lead to changes in the dimension table

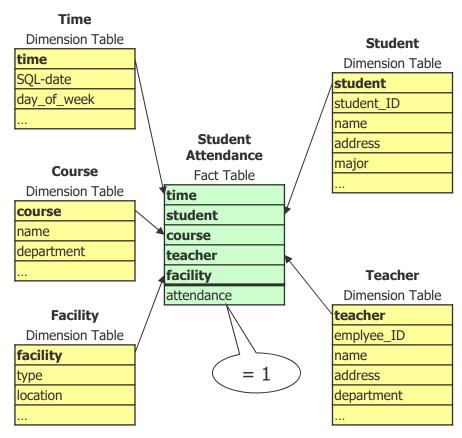


Modeling Events and Coverage

- How to model events?
- Example
 - A fact table should express the fact that students participate in courses and exercises (student's attendance event)
- What are the dimensions?What are the facts?
- Dimensions: Time, Student,
 Course, Teacher, Facility
- No obvious fact:
 Combination of all relevant keys express the fact

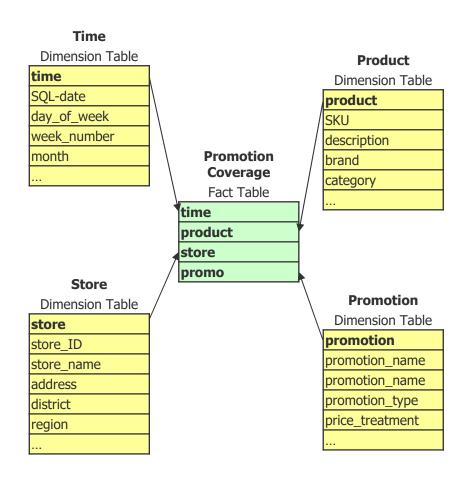
- How to model coverage?
- Example
 - A fact table should record the sales of products in stores on particular days under each promotion condition
- Sales fact table may be sparse.
 How to handle products that did not sell?
- Dimensions: Time, Store, Product, Promotion
- No obvious fact: Combination of all relevant keys express the fact

Factless Fact Tables



- Dummy fact 'attendance' is optional but makes SQL more readable
- Many questions can be answered
 - Which classes were the most heavily attended?
 - Which classes were the most consistently attended?
 - Which teachers taught the most students?
 - Which teachers taught classes in facilities belonging to other departments?
 - Which facilities where the most lightly used?

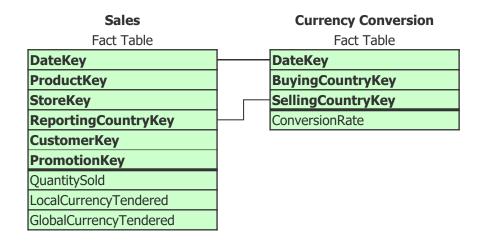
Factless Fact Tables



- Answering the question of which products were on promotion but did not sell requires a two-step application
 - first, consult coverage table
 - second, consult sales table
- Use sales fact table instead and fill in records representing zero sales for all possible products?
 - The coverage factless fact table can be made much smaller

Multinational Currency Tracking

- Transactions have to be expressed in a multitude of currencies
- Solution:
 - Value is reported in the local currency as well as in the global currency for a multinational enterprise (e.g. US dollars)
 - Conversion table provides conversion rates on a daily basis in both directions



Overview

- Data Warehouse Design Process
- Conceptual Design
- Logical Design
- Details of Logical Design
- Physical Design

Physical Design

Physical implementation of the logical schemata with respect to the individual properties of the target database system

- What to consider in physical design
 - indexing
 - partitioning
 - denormalization
 - pre-aggregation
 - ...
- See also 'Database Support for Data Warehousing'

Summary

- Main steps of the Data Warehouse Design Process
 - req. analysis, conceptual, logical and physical design
- Different conceptual models are available. Common elements
 - attributes, dimensions, hierarchies, facts, measures
- Conceptual models usually provide a graphical notation
- Logical design is based on the multidimensional model or the relational model
- Conceptual schema may be mapped to different logical schema types (relational model)
 - star schema, snowflake schema, vendor-specific schemes
- Physical design deals with optimization steps for the target database system

Papers



- [GMR98] M. Golfarelli, D. Maio, S. Rizzi: The Dimensional Fact Model: A Conceptual Model for Data Warehouses. International Journal of Cooperative Information Systems, Vol. 7, No. 2&3, 1998.
- [HLV00] B. Hüsemann, J. Lechtenbörger, G. Vossen: Conceptual Data Warehouse Design. Proc. of the Second International Workshop on Design and Management of Data Warehouses, Stockholm, 2000.
- [LTS06] S. Luján-Mora, J. Trujillo, I.-Y. Song: A UML profile for multidimensional modeling in data warehouses. Data & Knowledge Engineering 59 (2006) 725-769.
- [PJ99] T. B. Pedersen, C. S. Jensen: Multidimensional Data Modeling for Complex Data. Proc. of the 15th International Conference on Data Engineering, Sydney, Austrialia, 1999.
- [TBC99] N. Tryfona, F. Busborg, J. G. Christiansen: starER: A Conceptual Model for Data Warehouse Design. Proc. of the ACM Second International Workshop on Data Warehousing and OLAP, Kansas City, Missouri, USA, 1999.

Appendix: starER

- Combines the star structure with the semantically rich constructs of the ER model
- Adds special types of relationships to support hierarchies
- starER vs. DFM
 - starER allows many-to-many relationships between dimensions and facts
 - starER allows objects participating in the data warehouse, but not in the form of a dimension
 - Specialized relationships on dimensions are permitted in starER (specialization/generalization)

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Appendix: starER: Fact Sets

Fact set

- Represents a set of real-world facts sharing the same characteristics or properties.
- Semantically, a fact set points to the process of generating data over time, i.e., data is generated in terms of facts, each time an event related to the fact takes place.
- Graphical representation

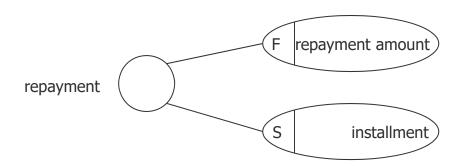


Appendix: starER: Fact Properties

Fact properties

- Properties that characterize a fact
- Usually numerical data that can be summarized
- There are three different types of properties
 - stock (S): records the state of something at a specific point in time
 - flow (F): records the commutative effect over a period of time
 - value-per-unit (V): like 'stock', but the unit of the property is different

Graphical representation



Appendix: starER: Entity sets and relationship sets

Entity sets

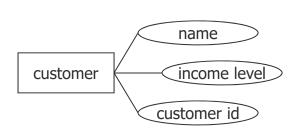
- Represent a set of real-world objects with similar properties
- It has the same meaning as in traditional application modeling

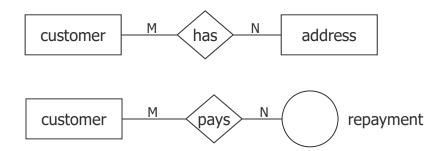
Relationship sets

- Represents a set of associations among entity sets or among entity sets and fact sets
- Its cardinality can be many-to-many, many-to-one or one-to-many

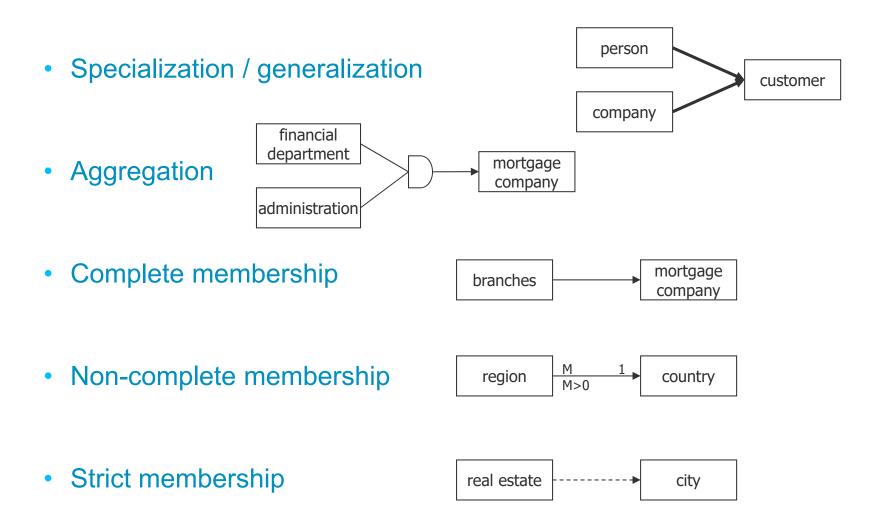
Attributes

- Static properties of entity sets, relationship sets, and fact sets
- Graphical representation





Appendix: starER: Relationships



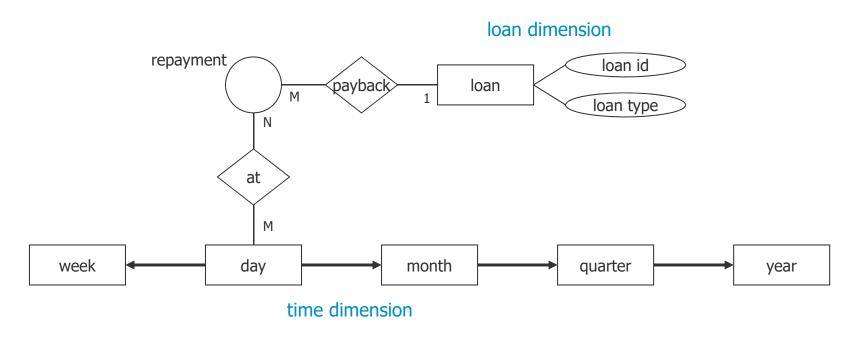
Appendix: starER: Dimensions and Hierarchies

Dimension

Entity sets associated to a fact are the dimensions of that fact

Hierarchies

 Dimensions consist of hierarchies and other relationships among other entity sets



Appendix: Mortgage Company DW

