Data Warehousing Dimension Changes

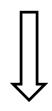
Esteban Zimányi ezimanyi@ulb.ac.be Slides by Toon Calders

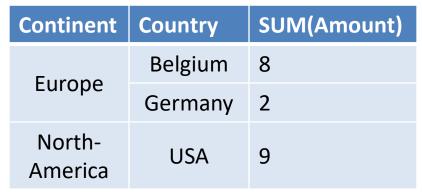


- Properties of measures
 - Additive along a dimension
 - Aggregable
- Properties of aggregation operators
 - Distributive
 - Algebraic
 - Holistic

Continent	Country	City	Amount
Europe	Dolgium	Brussels	5
	Belgium	Antwerp	3
	Germany	Berlin	2
North- America	USA	Chicago	1
		Tampa	8

Properties of measures: Distributivity





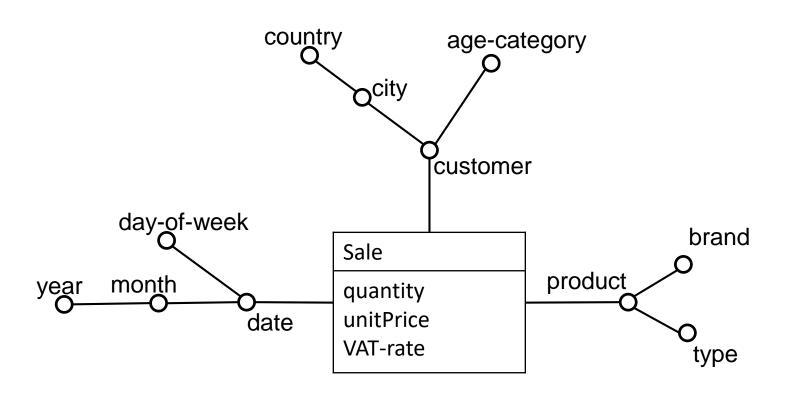


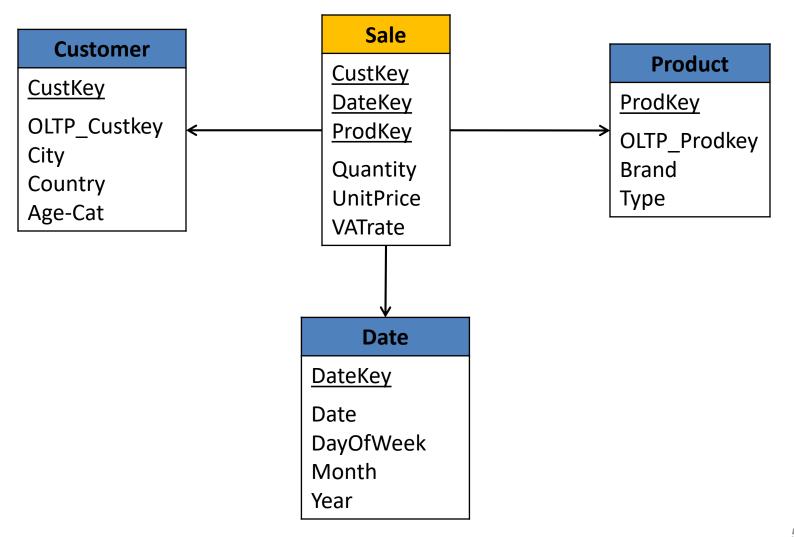
Continent	Sum(Amount)		
Europe	10		
North-America	9		



Sum(Amount)

19





Star schema

- Dimension tables are not normalized
 - Use surrogate key
- Dimensions such as Date are materialized
- Key for the fact table consists of the foreign keys to the dimension tables

Snowflake schema

has (partially) normalized dimensions

- Ways to deal with the different conceptual modeling constructions
 - Non-standard hierarchies
 - Multiple arcs
 - Cross-dimensional attributes

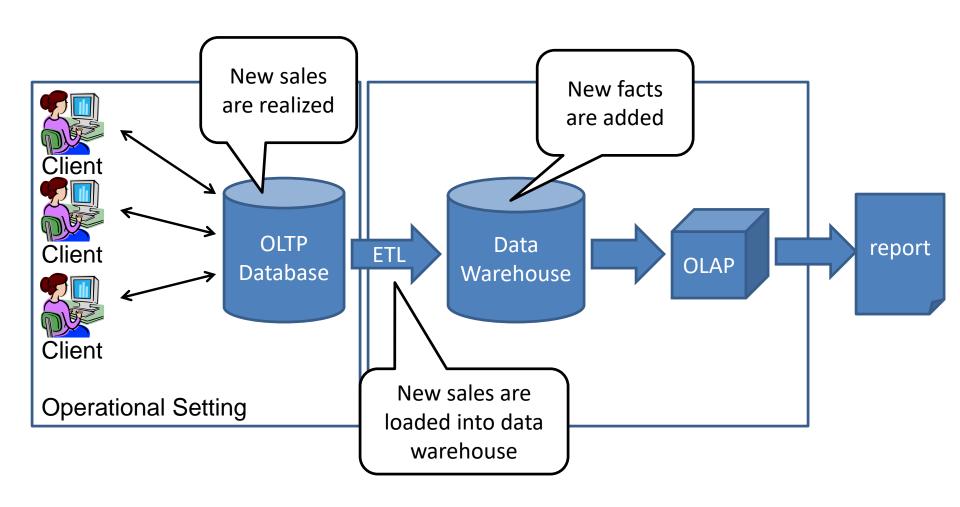
Outline

- Dealing with changing dimensions
 - Slowly Changing Dimensions
 - Type 1, 2, and 3
 - Rapidly changing dimensions
 - Type 4: Mini dimension
- Specific dimension types
 - Junk dimension
 - Outriggers
 - Degenerate dimension
 - Time and Data Quality dimensions

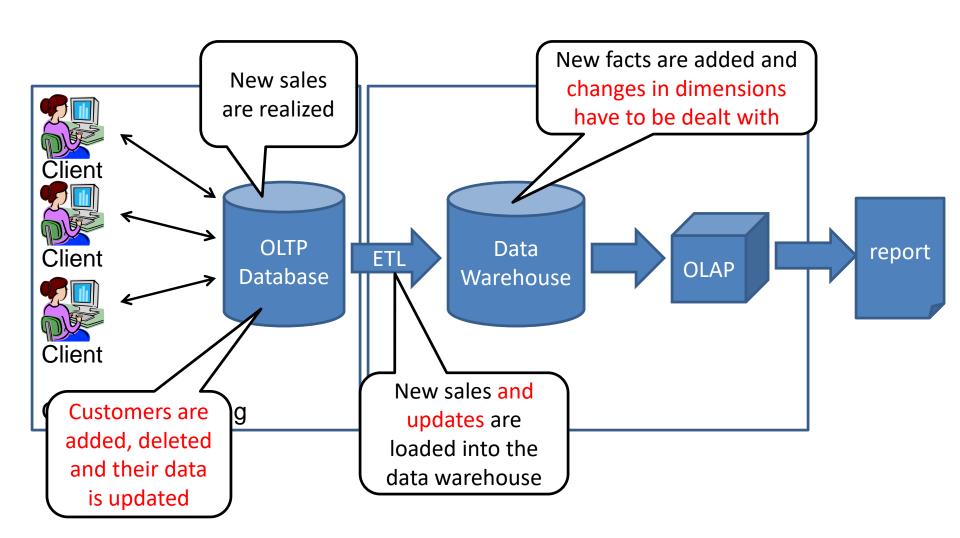
Changes in Dimensions

- Dimensions are not stable over time
 - New rows can be inserted
 - Existing rows can be updated
- We will see techniques for handling changing dimensions
 - Slowly changing dimensions
 - Rapidly changing dimensions

Idealized Picture



More Realistic Picture



What is the problem?

Customer

Oustorner

CID	Name	Address
001	John	Dallas
002	Mary	Dallas
003	Pete	New York

Sales

CID	Product	Price
001	Gun	5\$
002	Beef	20\$
003	Lava lamp	150\$

Customer

2010

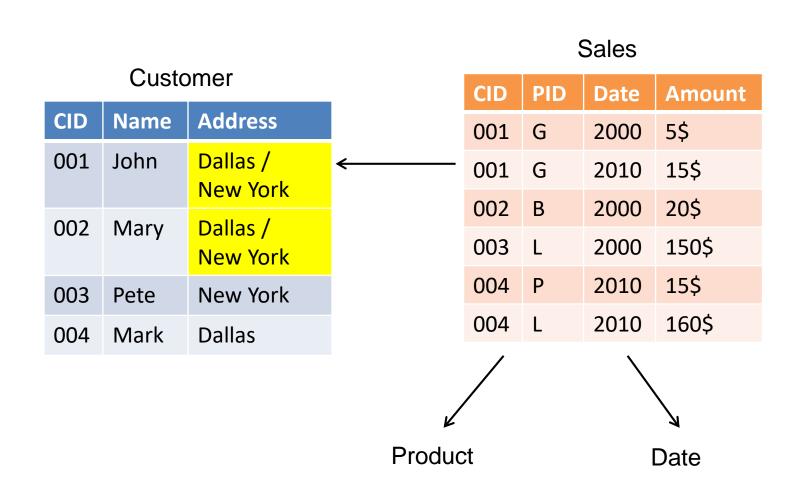
2000

CID	Name	Address
001	John	New York
002	Mary	New York
004	Mark	Dallas

Sales

CID	Product	Price
001	Gun	15\$
004	Pork	15\$
004	Lava lamp	160\$

What is the problem?



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Different Types of Handling

"Standardized" types of handling changes

- Type 1: No special handling
- Type 2: Versioning dimension values
 - 2A. Special facts
 - 2B. Time stamping
- Type 3: Capturing the previous and the current value

Type 1 - Updating

- Type 1 is updating the value
 - Suitable in case of mistakes
 - For dimensions with static attributes; e.g., last name; date of birth

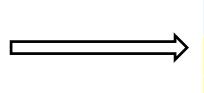
CID	Name	Address
001	John	NY
002	Mary	New York
003	Pete	NY
003	rete	IVI

Correct inconsistency in the city names of John and Pete Mark is a new client

Type 2 - Versioning

- Whenever there is a change, create a new version of the affected row
 - Need for surrogate key!

SID	CID	Name	Address
1	001	John	Dallas
2	002	Mary	Dallas
3	003	Pete	New York



SID	CID	Name	Address
1	001	John	Dallas
2	002	Mary	Dallas
3	003	Pete	New York
4	001	John	New York
5	002	Mary	New York
6	004	Mark	Dallas

John and Mary move to New York Mark is a new client

Type 2 – Valid Time

 It may be useful to know at what point a certain version was valid

- Different ways to store the valid time of a version
 - 2A: use time dimension and special facts
 - 2B: time stamping of rows

Type 2A

Use special facts and time dimension of facts

	Customer					S	ales	
SID	CID	Name	Address		SID	PID	Date	Amount
1	001	John	Dallas		1	G	D1	5\$
2	002	Mary	Dallas		2	G	D1	15\$
3	003	Pete	New York		2	В	D2	20\$
4	001	John	New York		3	L	D3	150\$
5	002	Mary	New York		4	Р	D4	15\$
6	004	Mark	Dallas		6	L	D4	160\$
					5	-	D5	-
Consider to the								
	Special fact to store date of							
	change							

Type 2B (More popular one)

Keep valid time as explicit attributes

Customer

SID	CID	Name	Address	Start	End	Valid
1	001	John	Dallas	D1	D2	
2	002	Mary	Dallas	D1	D3	
3	003	Pete	New York	D2	-	X
4	001	John	New York	D2	-	X
5	002	Mary	New York	D3	-	X
6	004	Mark	Dallas	D5	-	X

Type 3 – Limited Versions

- Foresee limited number of changes
 - Add attribute for every change
- Advantage:
 - If the change itself is useful information
 - How does calling volume change if people move?
 - No need to chain primary events
- Various disadvantages:
 - When did the change happen?
 - What if there are more changes?

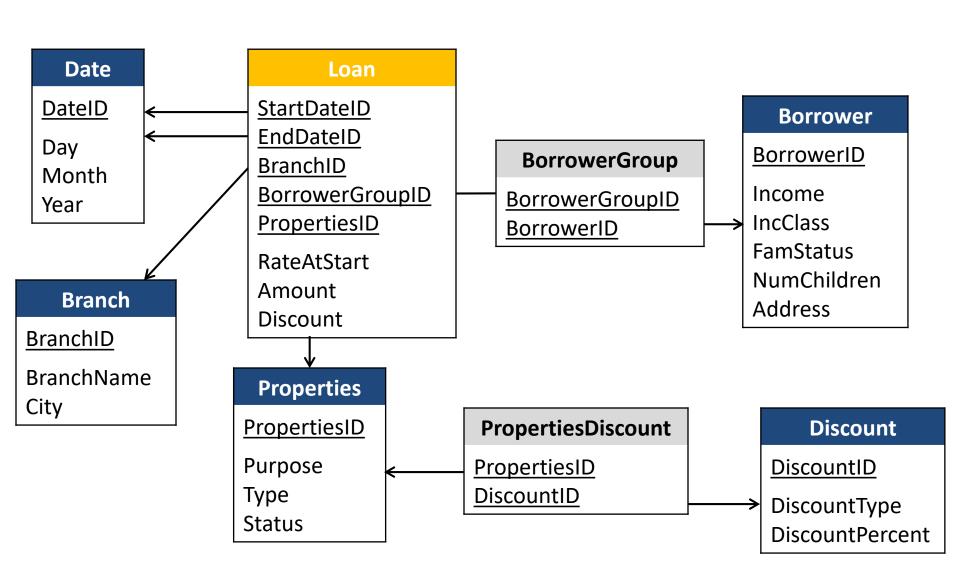
Type 3 – Limited Versions

Customer

SID	CID	Name	OldAddress	NewAddress
1	001	John	Dallas	New York
2	002	Mary	Dallas	New York
3	003	Pete	New York	New York
6	004	Mark	Dallas	Dallas

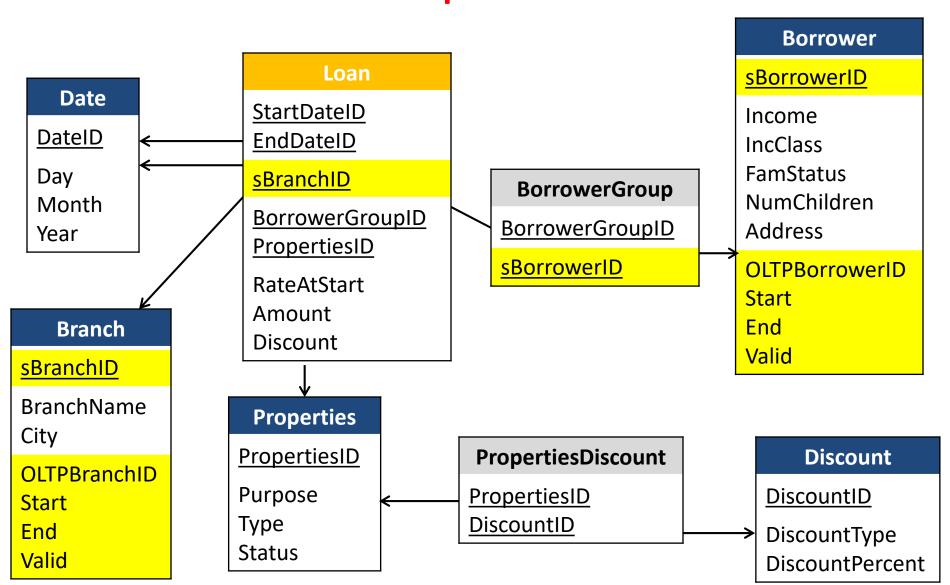
ONLY TYPE 1 CHANGES

Example: SCD



Example: SCD

TYPE 2 CHANGES for brand and borrower



Example: SCD

Borrower

sBID	OLTP_BID	Address	•••	Valid
S1	001	Brussels		X
S2	002	Brussels		X

Borrower_Group

BGroup	sBID
G1	S1
G1	S2

BGroup	Date		Amount
G1	D1	•••	1000

Example: SCD

002 moves

Borrower

sBID	OLTP_BID	Address	•••	Valid
S1	001	Brussels		X
S2	002	Brussels	•••	
S3	002	Antwerp		X

Borrower_Group

001 and 002 get a new loan

Loan

BGroup	sBID
G1	S1
G1	S2
G2	S1
G2	S3

BGroup	Date		Amount
G1	D1	•••	1000
G2	D2		7000

SCD and Bridge Table

bridgeAccountCustomer dimCustomer dimAccount OLTPkey **Premium** sCID VIP **Adress sAID sCID sAID** Type Key Savings SC1 Brussels ← SA1 SC1 \rightarrow SA1 001 C1 Υ SA₂ 002 Debit SC₂ **C2** SC2 N Antwerp SA1 SC3 **C3** SC3 Ν Brussels SA2

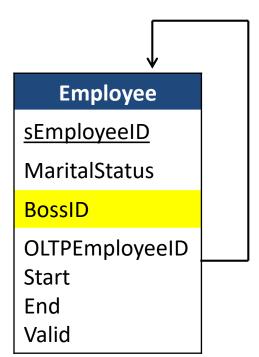
	SAID	•••	Amount	
Changes possible to:		•••	1000	
changes possible to:	SA2		2000	

- Account
- Customer
- Relation between Account and Customer

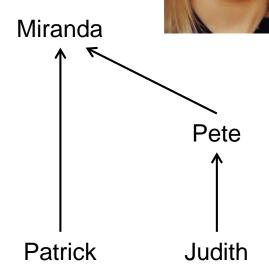
SCD and Bridge Table

- Solution: add valid start and end time to bridge table; Customer and Account versioned
- New tuples are added to the bridge table whenever:
 - Customer changes
 - Account changes
 - Relation between accounts and customers changes

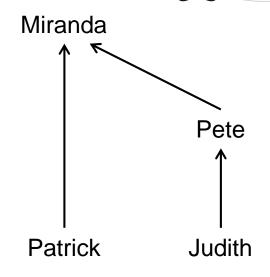
- Be very careful when to apply type-2 changes; consider the following Employee dimension with an unbalanced hierarchy expressing who is whose boss.
 - What happens if the CEO of the company gets married?



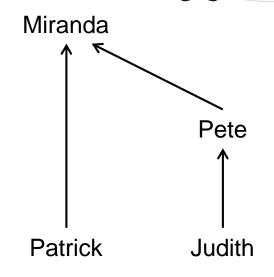
sID	OLTPkey	BossID	MStat	Start	End	Val
1	miranda	none	single	D1	-	X
2	patrick	1	married	D1	-	Χ
3	pete	1	single	D1	-	X
4	judith	3	married	D1	-	X



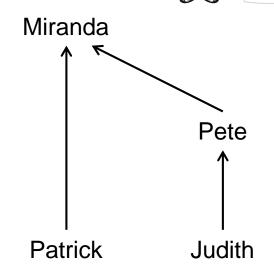
sID	OLTPkey	BossID	MStat	Start	End	Val
1	miranda	none	single	D1	-	X
2	patrick	1	married	D1	-	X
3	pete	1	single	D1	-	Χ
4	judith	3	married	D1	-	Χ



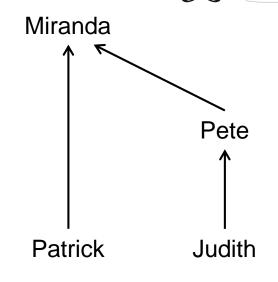
sID	OLTPkey	BossID	MStat	Start	End	Val
1	miranda	none	single	D1	D2	
5	miranda	none	married	D2	-	X
2	patrick	1	married	D1	-	X
3	pete	1	single	D1	-	X
4	judith	3	married	D1	-	Χ



sID	OLTPkey	BossID	MStat	Start	End	Val
1	miranda	none	single	D1	D2	
5	miranda	none	married	D2	-	Χ
2	patrick	1	married	D1	D2	
6	patrick	5	married	D2	-	X
3	pete	1	single	D1	D2	
7	pete	5	single	D2	-	X
4	judith	3	married	D1	-	Χ



sID	OLTPkey	BossID	MStat	Start	End	Val
1	miranda	none	single	D1	D2	
5	miranda	none	married	D2	-	Χ
2	patrick	1	married	D1	D2	
6	patrick	5	married	D2	-	Χ
3	pete	1	single	D1	D2	
7	pete	5	single	D2	-	Χ
4	judith	3	married	D1	D2	
8	judith	7	married	D2	-	X



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Rapidly Changing Dimensions

- Some attributes may change frequently
 - Many dimensional
 attributes in a changing
 dimension of Type 2 results
 in many repeated values
 when there is a change
 - Some attributes never change, but are duplicated nevertheless

Customer <u>SID</u> CID **Fname** Lname Dob Gender **MaritalStatus** NumChildren CreditScoreGroup BuyingStatusGroup IncomeGroup EducationGroup

Example: Changes

Customer
<u>SID</u>
CID
Fname
Lname
Dob
Gender
Address
City
Country
MaritalStatus
NumChildren
CreditScoreGroup
BuyingStatusGroup

Customer
SID
CID
Fname
Lname
Dob
Gender
Address
City
Country
MaritalStatus
NumChildren
CreditScoreGroup
BuyingStatusGroup
IncomeGroup
EducationGroup

Customer
SID
CID
Fname
Lname
Dob
Gender
Address
City
Country
MaritalStatus
NumChildren
CreditScoreGroup
BuyingStatusGroup
IncomeGroup
EducationGroup

Customer SID CID **Fname** Lname Dob Gender Address City Country MaritalStatus NumChildren CreditScoreGroup BuyingStatusGroup IncomeGroup EducationGroup

D1

EducationGroup

IncomeGroup

D2

D3

D4

Solution Type 4: Split Dimension

Customer

<u>SID</u>

CID

Fname

Lname

Dob

Gender

MaritalStatus

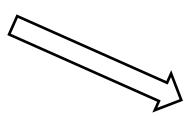
NumChildren

CreditScoreGroup

BuyingStatusGroup IncomeGroup

EducationGroup





Customer

SID

CID

Fname

Lname

Dob

Gender

Demographics

DemID

MaritalStatus

NumChildren

CreditScoreGroup

BuyingStatusGroup

IncomeGroup

EducationGroup

Add as a FK in the fact table;

not in customer dimension!

Mini-Dimension

- Suppose frequently changing attributes have small domains
 - We could force this situation by discretizing some attributes with many values
- We can fully populate the dimension with all possible combinations of values
 - No type-II changes to the mini-dimension!
 - Demography is not updated; a fact is about a customer with a new demography

Mini-Dimension

- What if we need to keep the changes in the dimensions?
 - Make an empty fact. Caveat: what with COUNT?
 - Can be considered a new kind of fact that needs to be stored → add a separate measureless fact table. Include additional identifying attribute, f.i. date/time as one of the dimensions (why?)
- Often surrogate key of the most recent minidimension value is added to dimension table

Solution Type 4: Split Dimension

Customer

<u>SID</u>

CID

Fname

Lname

Dob

Gender

MaritalStatus

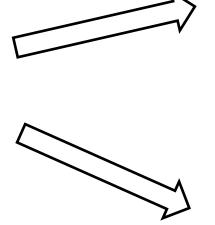
NumChildren

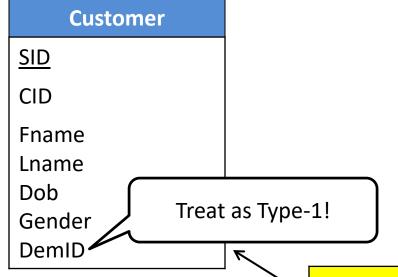
CreditScoreGroup

BuyingStatusGroup

IncomeGroup

EducationGroup





Demographics

<u>DemID</u>

MaritalStatus

NumChildren

CreditScoreGroup
BuyingStatusGroup

IncomeGroup EducationGroup

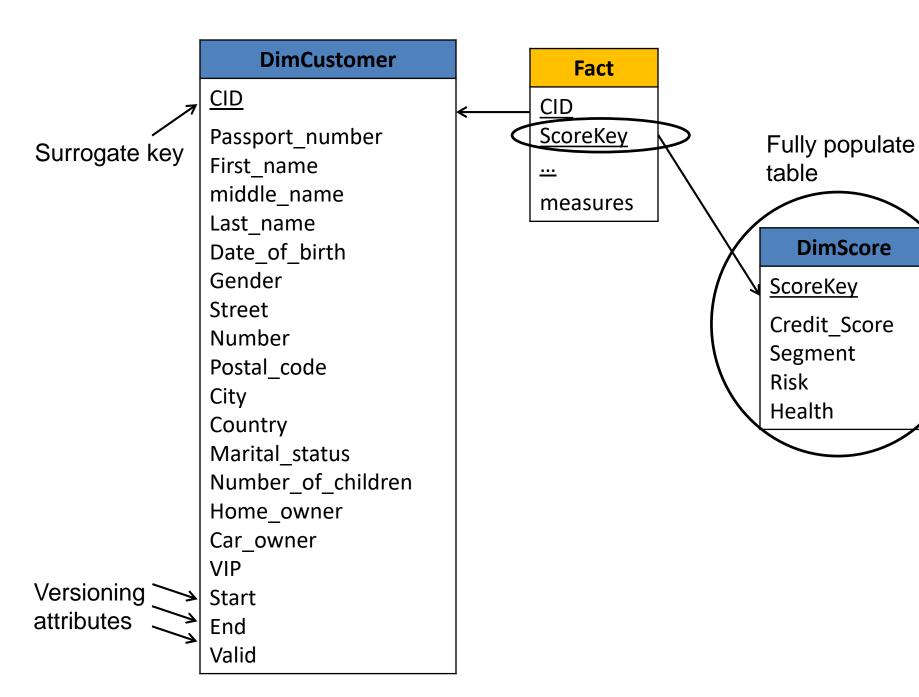
fNewDem

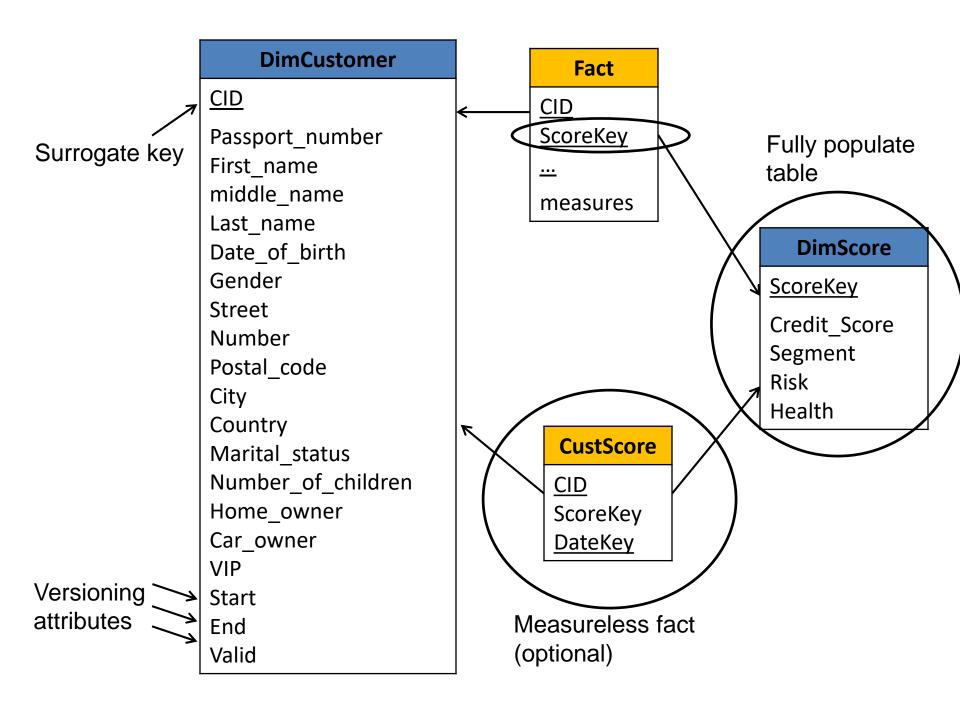
SID DemID

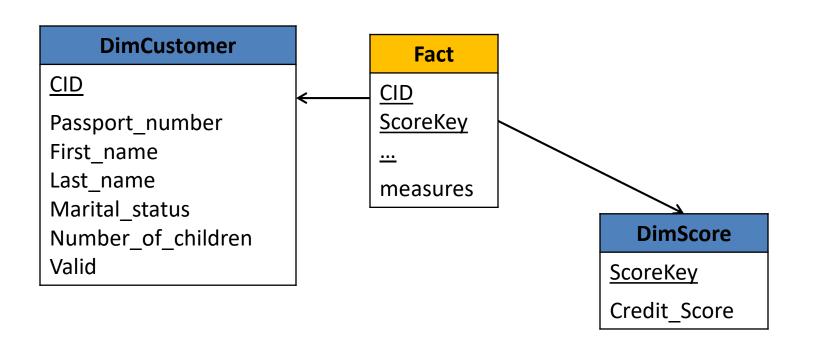
<u>DateID</u>

Date needed!

Model a dimension Customer that stores the following information: passport number first name, middle name, last name No changes date of birth and gender street, number, postal code, city, country marital status (single, married, divorced, widowed) number of children Some changes whether or not he or she is a home owner whether or not he or she owns a car whether or not he or she is a VIP client credit score (A+, A, B, C, or D) market segment (a number from 1 to 8) risk categorization as a car driver Frequent changes (number 1 to 10; starts at 5, decreases 1 per year, increases if person has an accident) health categorization (low risk, medium, or high risk)







CID	passport	fname	sname	Marital stat	nChildren	Valid
	•••	•••	•••	•••		

CID	ScoreKey	 measure
	•••	 •••

ScoreKey	credit
SK1	A+
•••	•••
SK8	DF

 A new customer with passport number 1234 and name Jan Janssens is added to the database. He is single, has no children and has a perfect credit score A+.

CID	passport	fname	sname	Marital stat	nChildren	Valid
•••	•••		•••	•••	•••	***
C01	1234	Jan	Janssens	single	0	Υ

CID	ScoreKey	•••	measure

ScoreKey	credit
SK1	A+
•••	•••
SK8	DF

Customer 1234 makes a sale; measure is 5

CID	passport	fname	sname	Marital stat	nChildren	Valid
•••	•••	•••		•••		***
C01	1234	Jan	Janssens	single	0	Υ

CID	ScoreKey	 measure
	•••	 •••
C01	SK1	 5

ScoreKey	credit
SK1	A+
SK8	DF

The customer with passport number 1234 gets married.

CID	passport	fname	sname	Marital stat	nChildren	Valid
•••	•••	•••			•••	•••
C01	1234	Jan	Janssens	single	0	N
C02	1234	Jan	Janssens	married	0	Υ

CID	ScoreKey	•••	measure
			•••
C01	SK1		5

ScoreKey	credit
SK1	A+
•••	•••
SK8	DF

 The name of customer 1234 is corrected to Jan Jansens (one s removed from last name).

CID	passport	fname	sname	Marital stat	nChildren	Valid
		•••			•••	•••
C01	1234	Jan	Jansens	single	0	N
C02	1234	Jan	Jansens	married	0	Υ

CID	ScoreKey	•••	measure
C01	SK1		5

ScoreKey	credit
SK1	A+
SK8	DF

• Customer 1234 becomes the father of twins. His credit score drops to B. He makes a sale.

CID	passport	fname	sname	Marital stat	nChildren	Valid
•••	•••	•••		•••	•••	•••
C01	1234	Jan	Jansens	single	0	N
C02	1234	Jan	Jansens	married	0	N
C03	1234	Jan	Jansens	married	2	Υ

CID	ScoreKey	•••	measure

C01	SK1		5
C03	SK3		3

ScoreKey	credit
SK1	A+
SK8	DF

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Outrigger

- Dimension referred to by another dimension
 - Not exactly the same as a snow-flake

Example:

Date that an employee joined the company can be stored by referring to the date dimension

Degenerate Dimension

Dimension without any content

CustID	ProdID	DateID	Transaction	quantity	Price
001	P003	D101	1	2	5.10
001	P001	D101	1	3	3.24
002	P002	D101	2	6	7.99
003	P003	D102	3	1	2.13
003	P005	D102	3	2	8.15

- Transaction is needed to be able to aggregate data at transaction level
- Dimension table is not necessary

Junk Dimension

- Many small dimensions combined into one
 - "Junk drawer"
 - Typically flags; promotion; how-displayed, ...
- Combine into one dimension; fully populate

JunkID	Packed	Shipped	Delivered	Returned	Refunded
001	N	N	N	N	N
002	N	N	N	N	Υ
003	N	N	N	Υ	N
004	N	N	N	Υ	Υ
032	Υ	Υ	Υ	Υ	Υ

Date/Time Dimension

- Date-time dimensions can become extremely large
 - Enormous number of possible combinations
 - Either get from data (expensive) or generate all possibilities (infeasible)
- Therefore: usually split into Date dimension at granularity day and a Time-of-day dimension
 - Limited number of dates
 - Only 1440 minutes in a day

Data Quality Dimension

- Is sometimes added to comment on the quality of a fact
 - Normal value
 - Out-of-bounds value
 - Unlikely value
 - Verified value
 - Unverified value

— ...

Summary

- Different techniques to store changes in dimensions
 - Type 1: update
 - Type 2: keep versions
 - Type 3: limited changes
 - Type 4: dimension splitting
- Special types of dimensions
 - Outrigger, degenerate, junk
 - Date, data quality

Exercise

 Devise a relational schema that can accommodate changes to the following attributes:

Band name

Band members

Instrument

