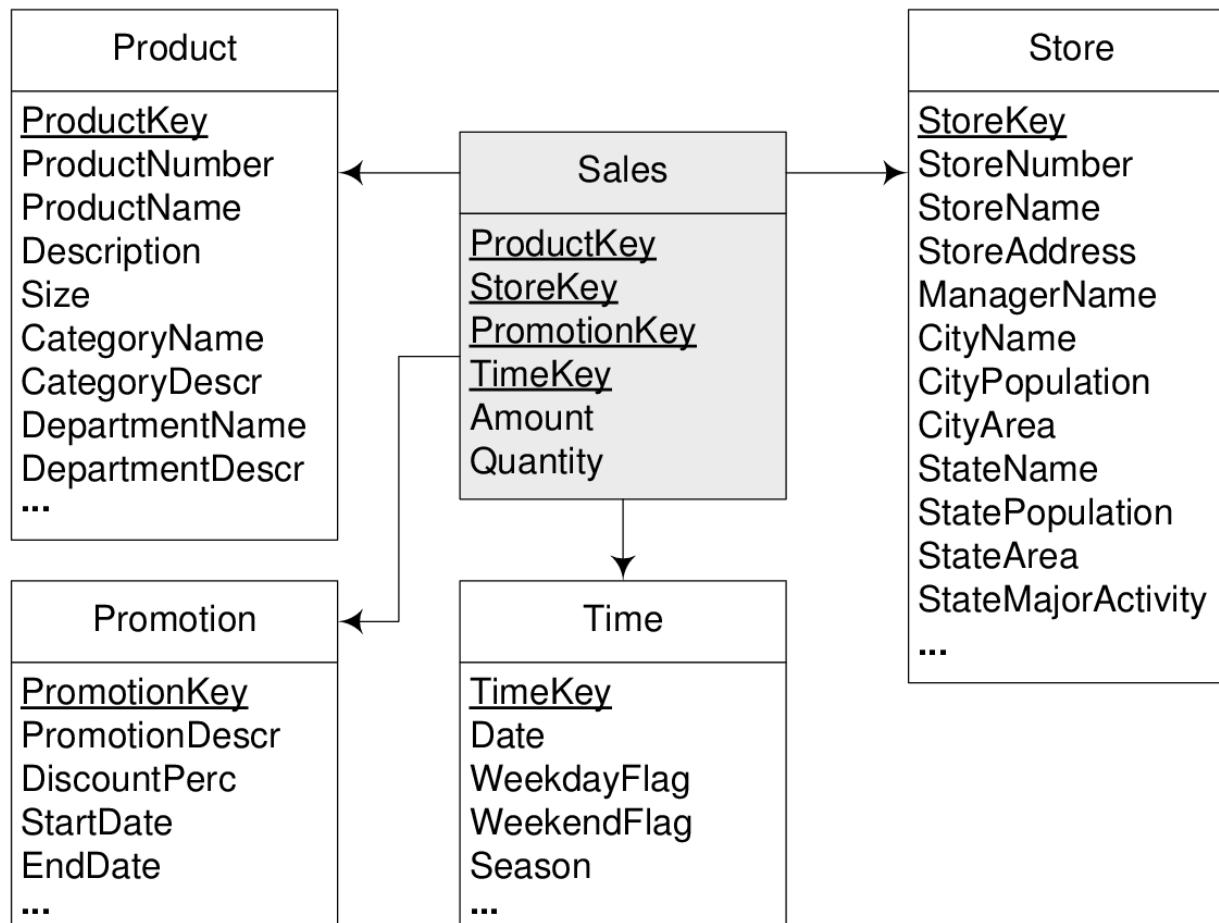


Data Analysis and Integration

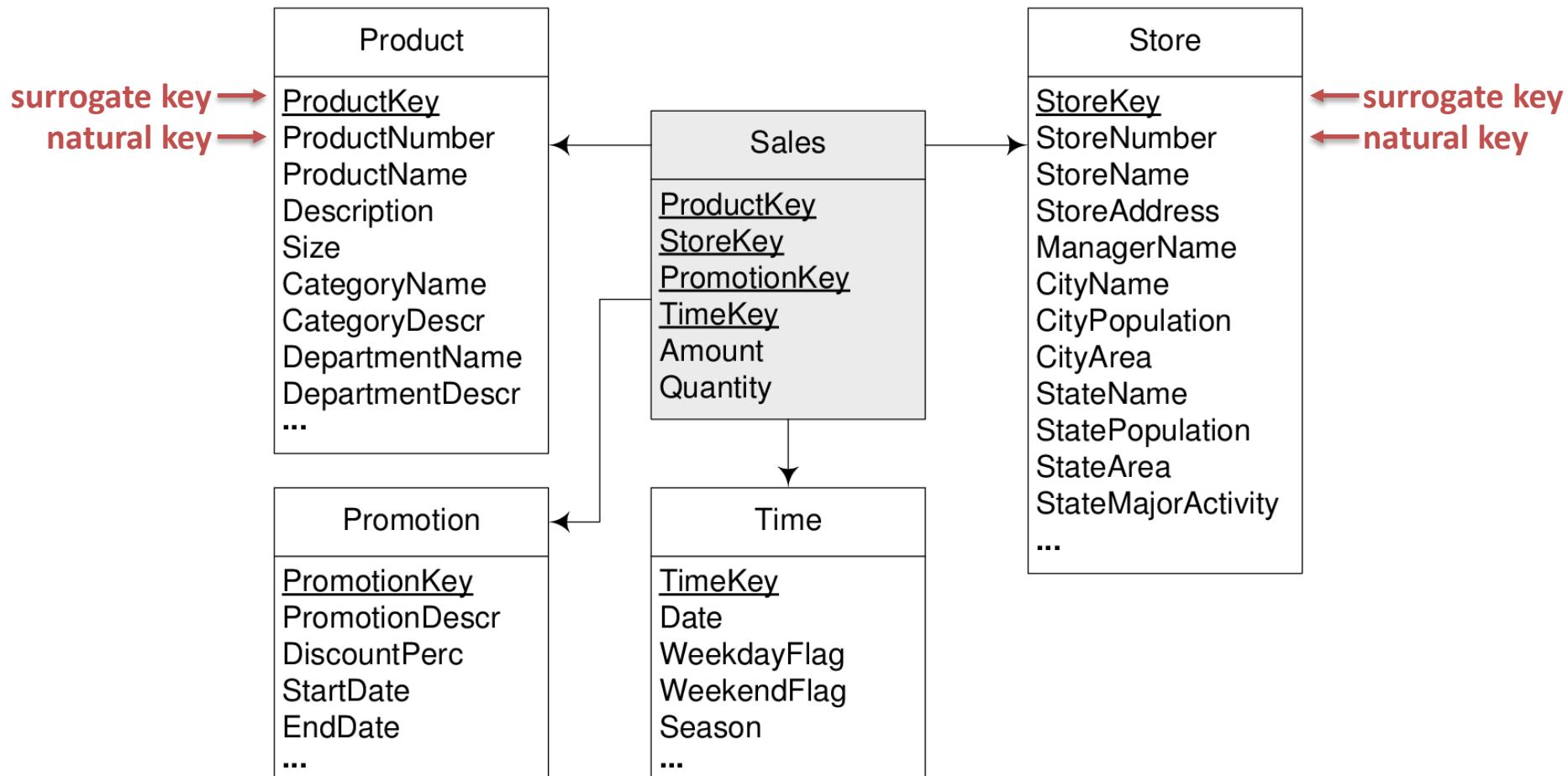
Data warehouse design

Star schema



Surrogate keys

- Each **dimension** has its own key



Surrogate keys

- A data warehouse has its own primary keys
 - these are called **surrogate keys** or **technical keys**
 - **ProductNumber** identifies products in the original database
 - **ProductKey** identifies products in the data warehouse
 - **surrogate keys** replace the original primary keys (**natural keys**)
 - provide independence from keys in the original data sources
 - solve inconsistencies between keys from multiple sources
 - represent keys as **integers** to improve efficiency
 - avoid less efficient data types, such as strings

Types of DW schema

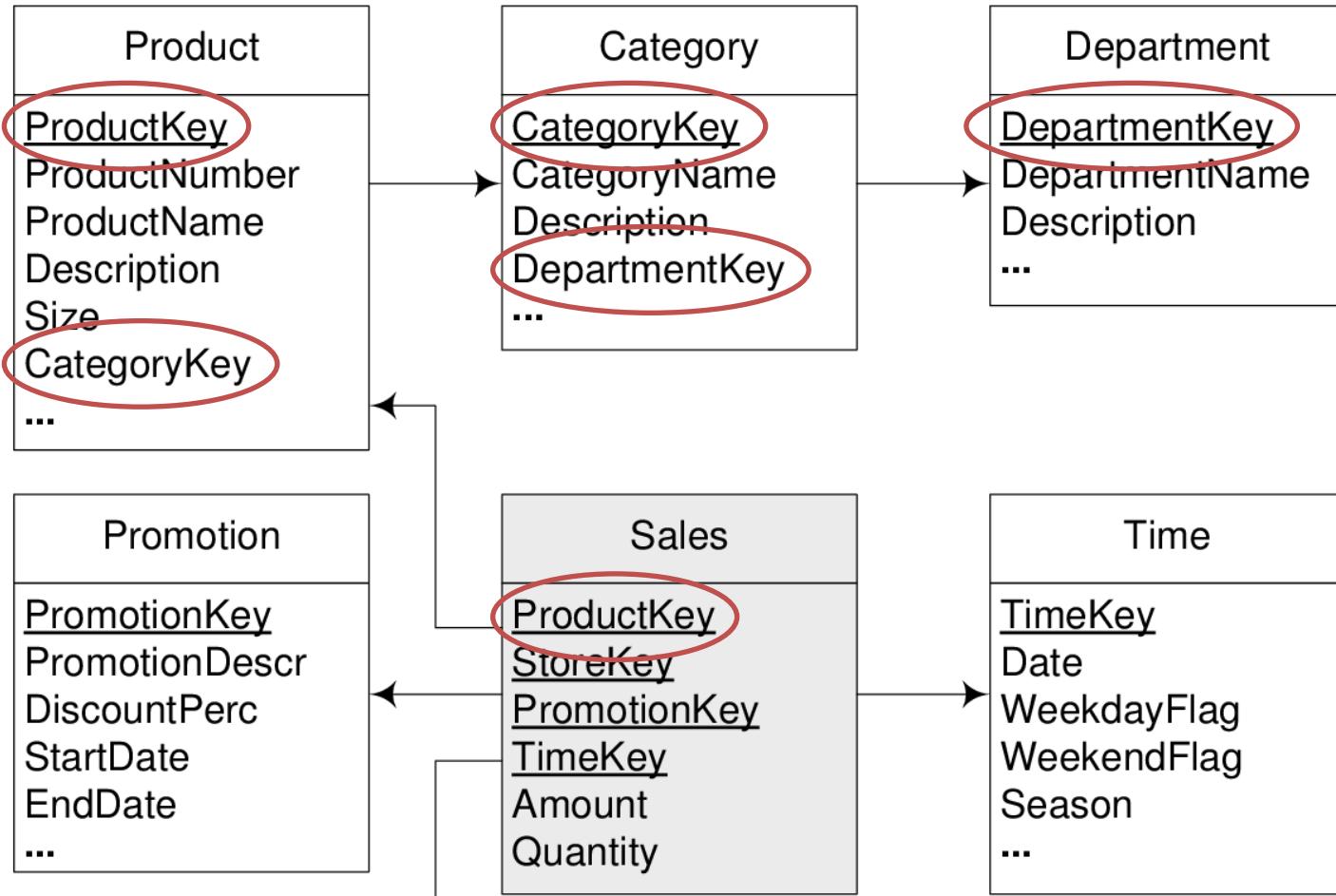
- Star schema
 - has a single table for each dimension, even in the presence of hierarchies (denormalized dimension tables)
- Snowflake schema
 - has normalized tables for all dimensions and their hierarchies
- Starflake schema
 - is a mix between star and snowflake schemas (both normalized and denormalized dimensions)
- Constellation schema
 - has multiple fact tables, possibly with shared dimension tables, and is viewed as a collection of stars

Snowflake schema

- When dimensions store redundant data (are denormalized)
 - **CategoryName**, **CategoryDescr** are the same for multiple products
 - replace by **CategoryKey** and move them to another table
 - **DepartmentName**, **DepartmentDescr** are the same for multiple categories
 - replace by **DepartmentKey** and move them to another table

Product
<u>ProductKey</u>
ProductNumber
ProductName
Description
Size
CategoryName
CategoryDescr
DepartmentName
DepartmentDescr
...

Snowflake schema

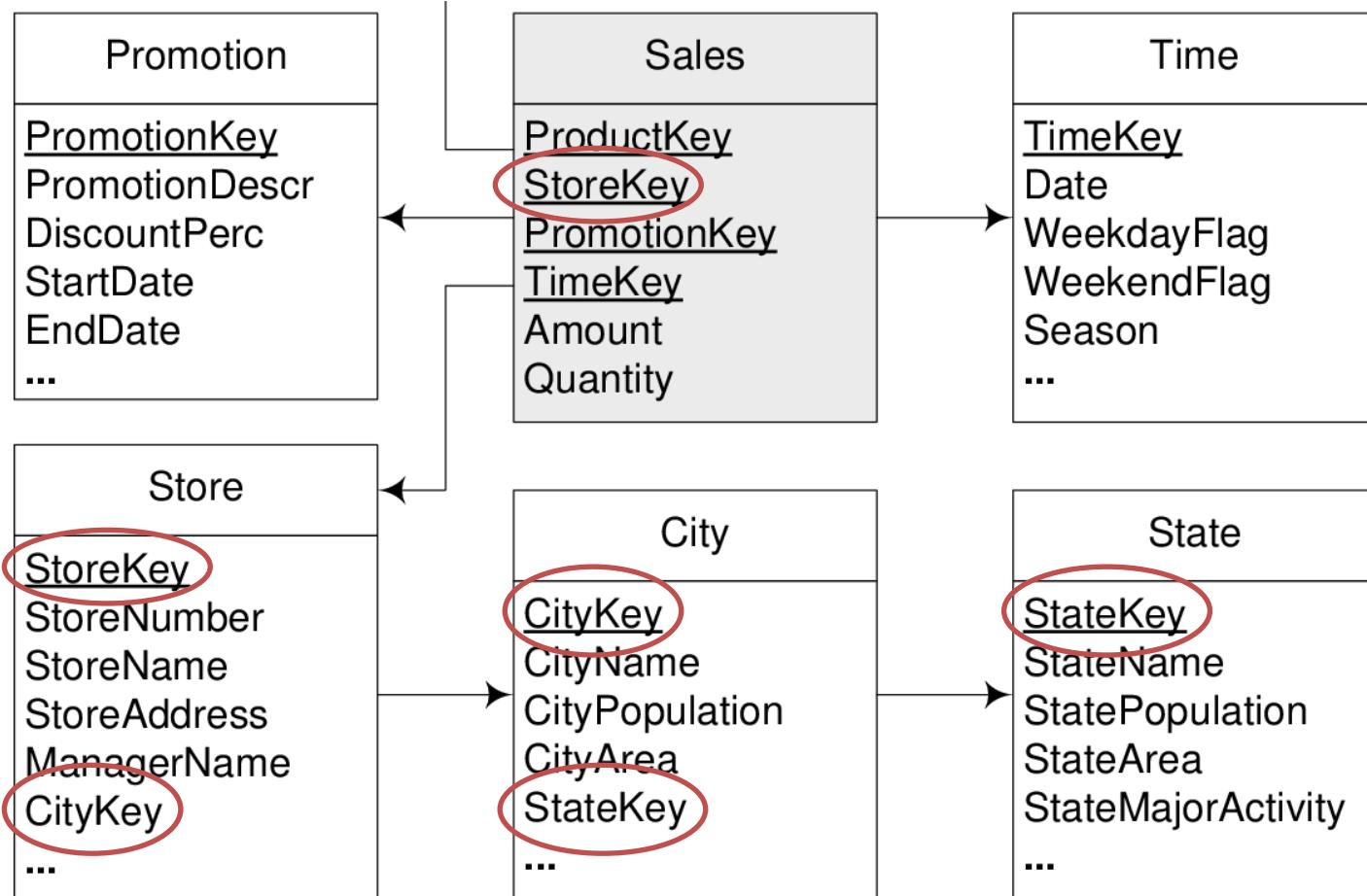


Snowflake schema

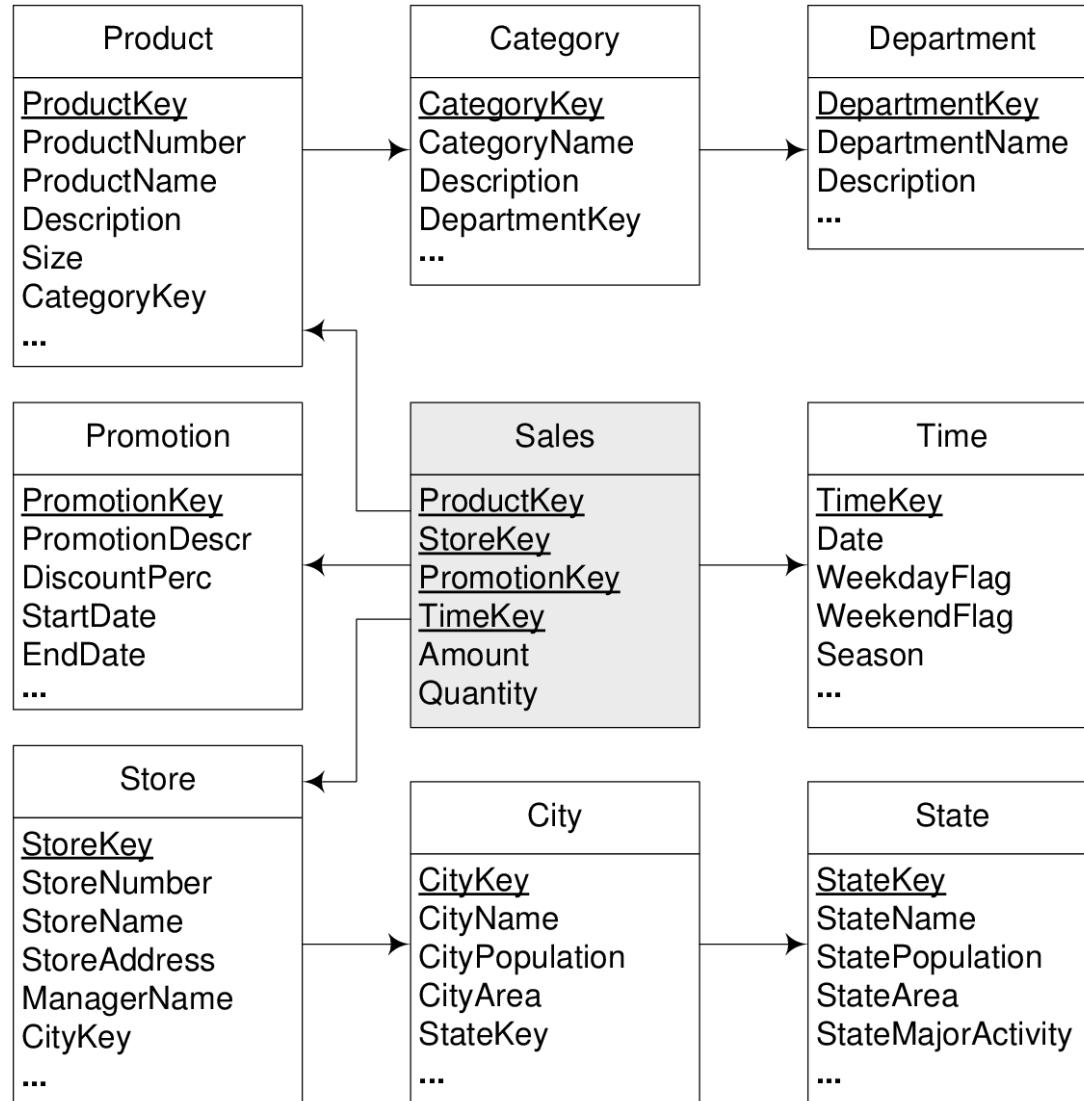
- Another example
 - **CityName**, **CityPopulation**, **CityArea** are the same for multiple stores
 - replace by **CityKey** and move them to another table
 - **StateName**, **StatePopulation**, **StateArea**, etc. are the same for multiple products
 - replace by **StateKey** and move them to another table

Store
<u>StoreKey</u>
StoreNumber
StoreName
StoreAddress
ManagerName
CityName
CityPopulation
CityArea
StateName
StatePopulation
StateArea
StateMajorActivity
...

Snowflake schema

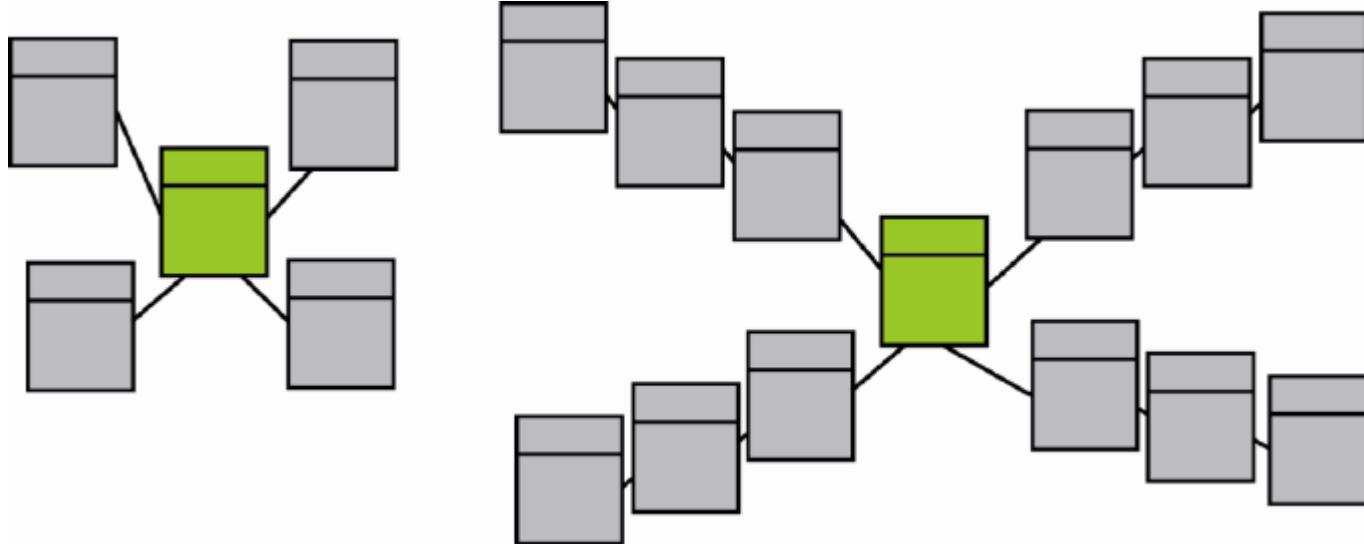


Snowflake schema



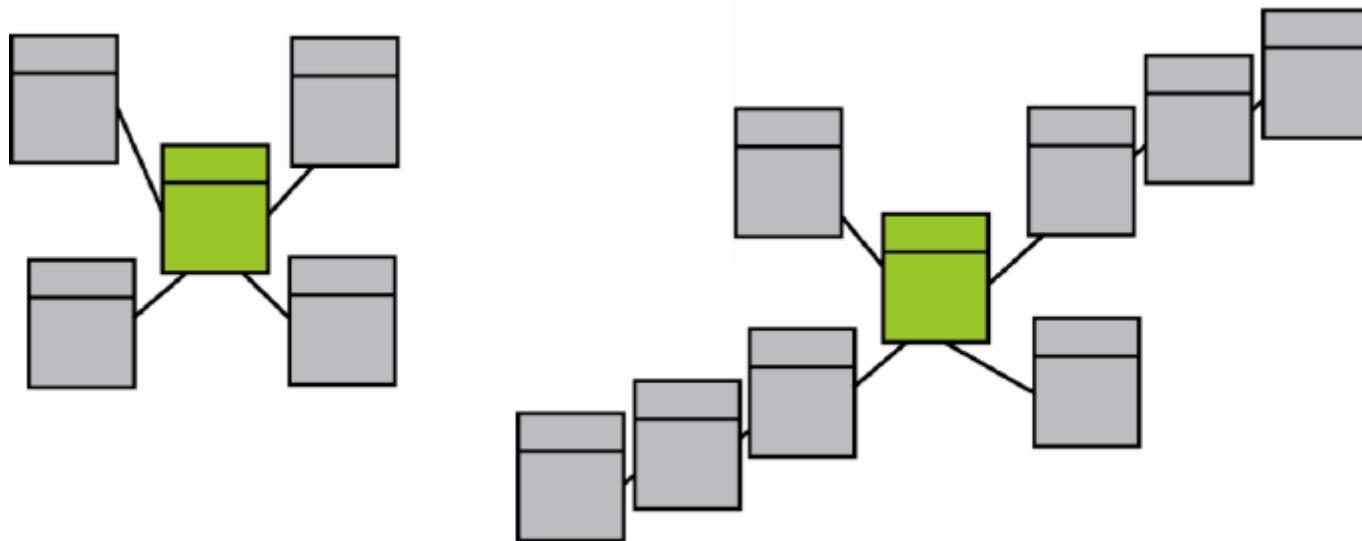
Snowflake schema

- Star vs. snowflake schema
 - all dimensions are normalized



Starflake schema

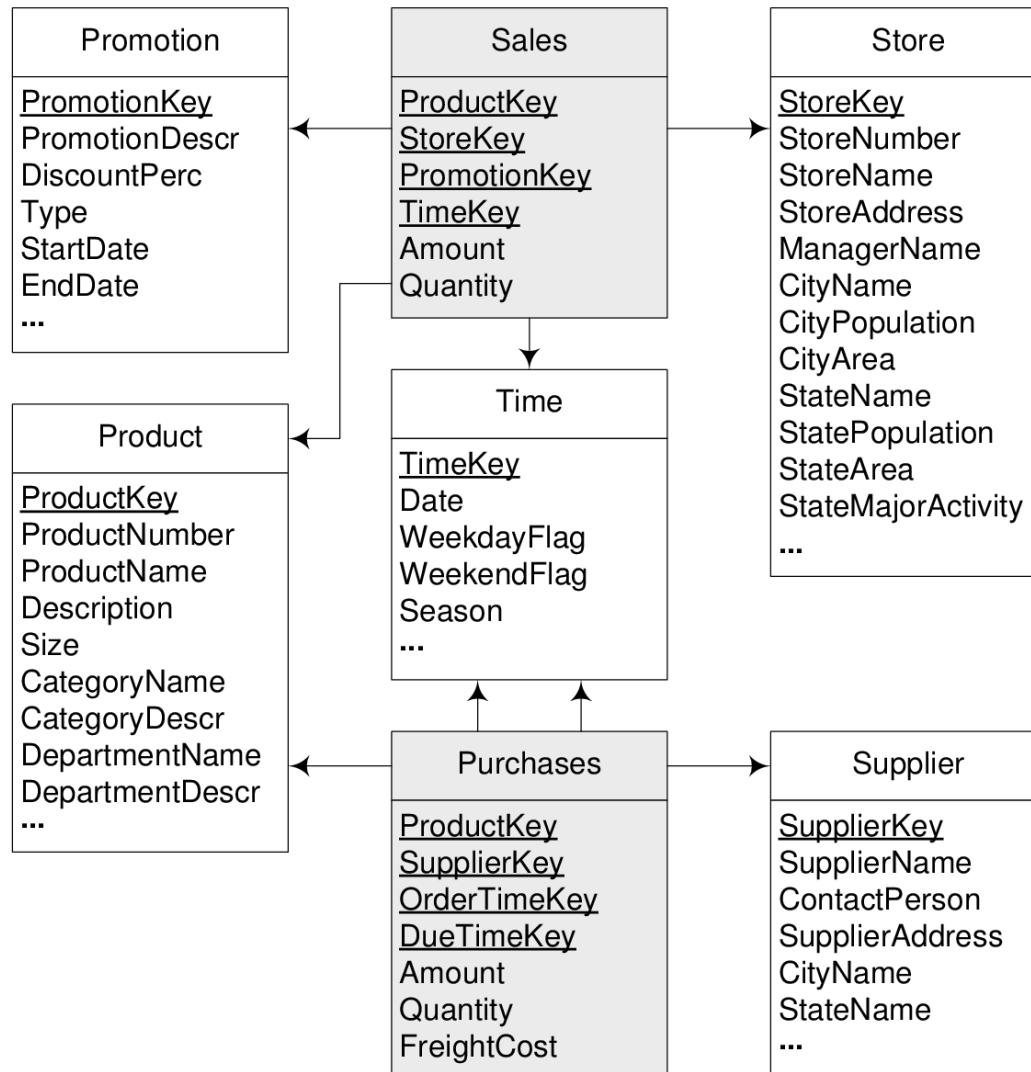
- Star vs. starflake schema
 - some dimensions are normalized



Constellation schema

- Multiple fact tables
 - e.g. sales facts, purchase facts
- Some dimension tables may be shared
 - e.g. product, time

Constellation schema

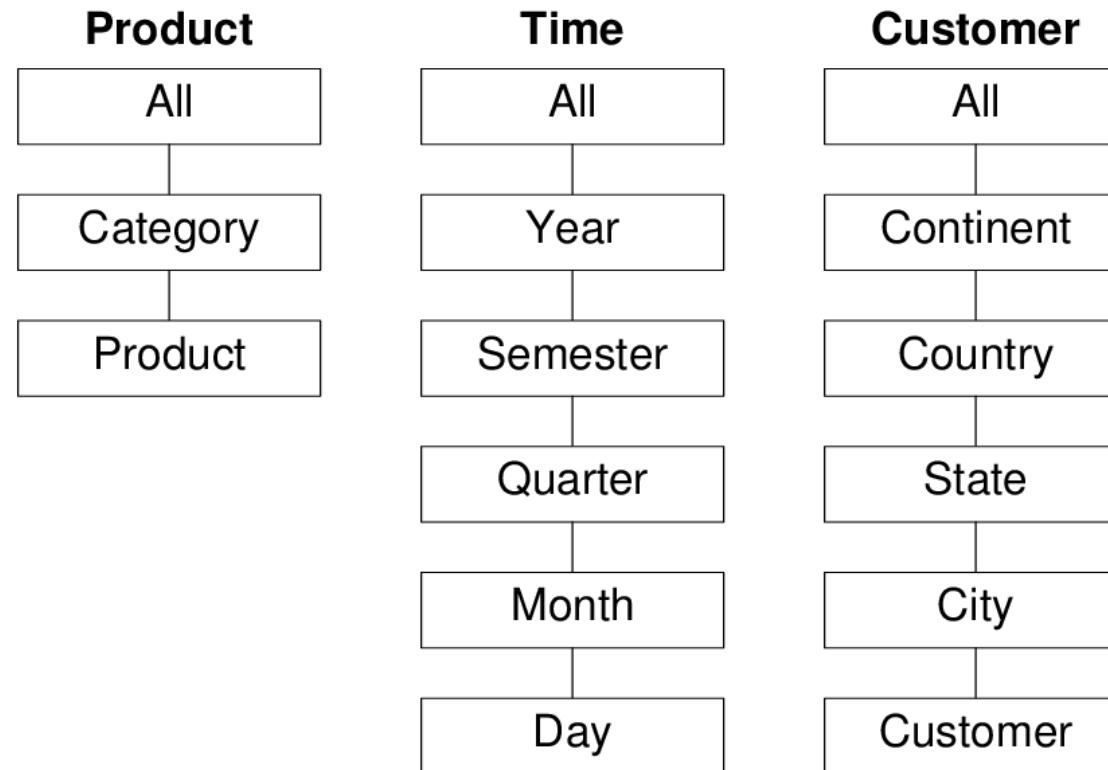


Hierarchies

- Dimension hierarchies are essential to enable analysis at different levels of detail
 - define a hierarchical structure of levels relating lower-level members to higher-level ones
- In real-world applications, users must deal with complex hierarchies of various kinds
 - however, current DW and OLAP systems support only a limited set of hierarchies
 - here, we are going to study additional types of hierarchies

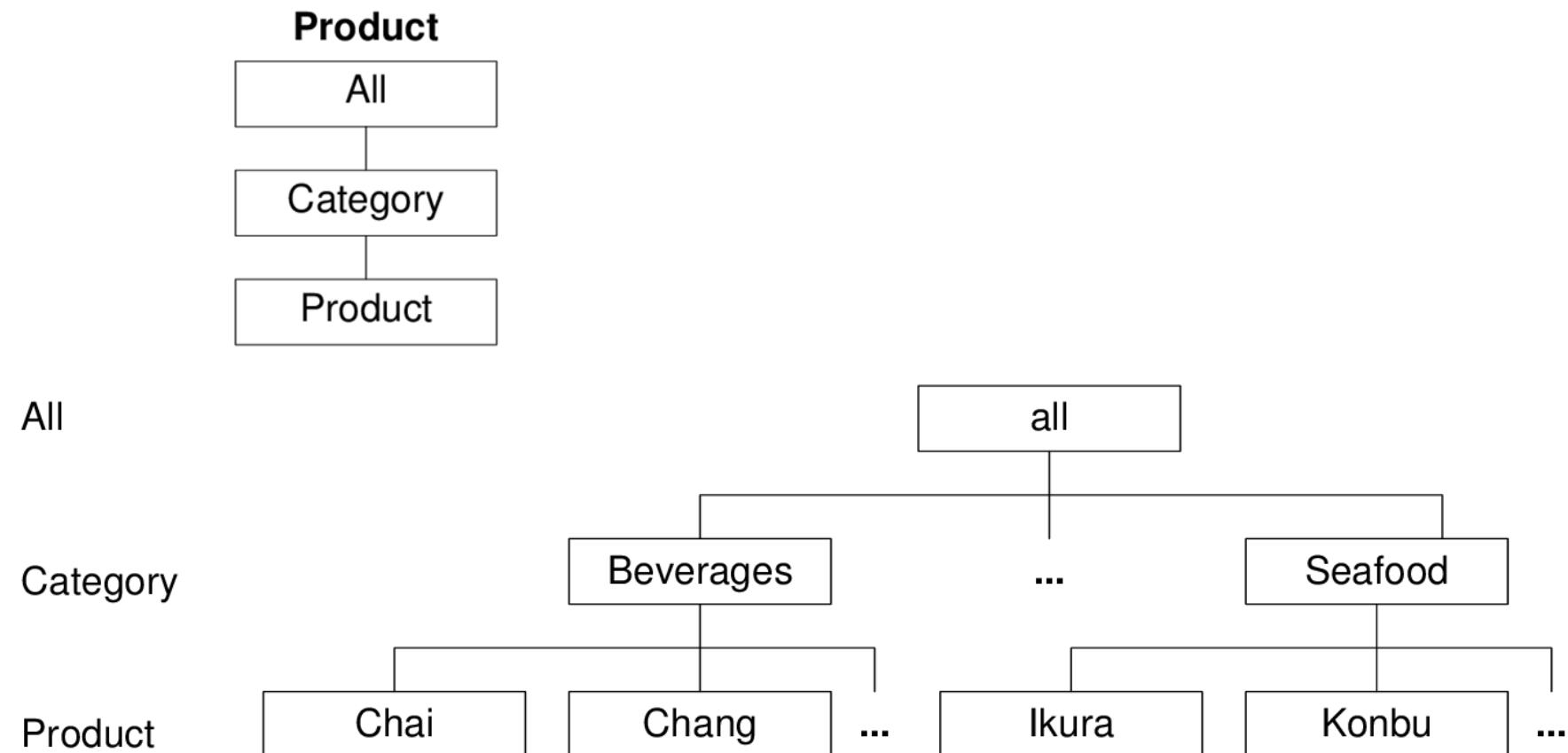
Hierarchies

- Product, time and customer dimensions



Hierarchy members

- Example of a hierarchy and its members

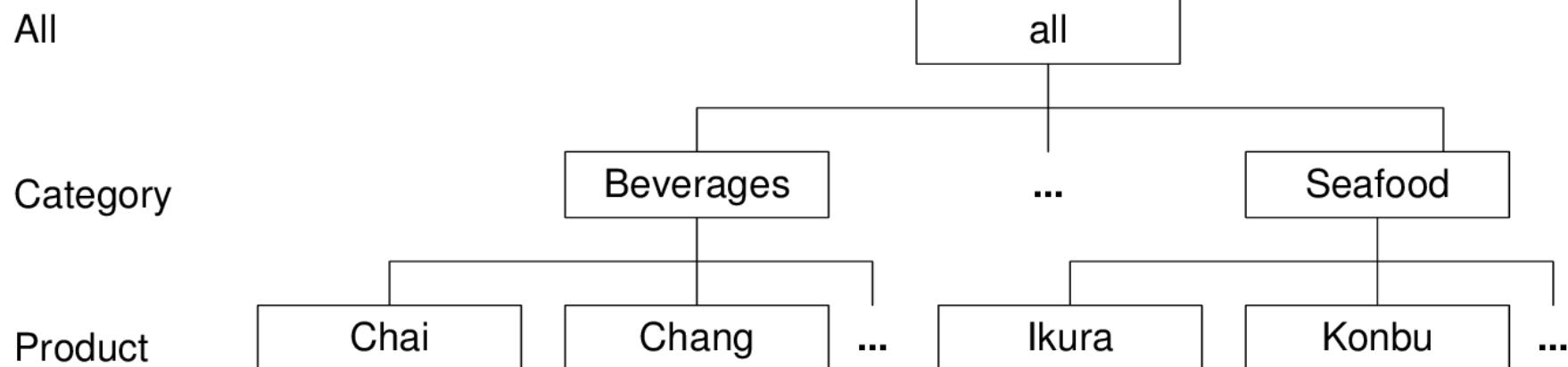


Types of hierarchy

- Balanced hierarchy
- Unbalanced hierarchy
- Recursive hierarchy
- Generalized hierarchy
- Ragged hierarchy
- Alternative hierarchy
- Parallel hierarchies
- Non-strict hierarchy

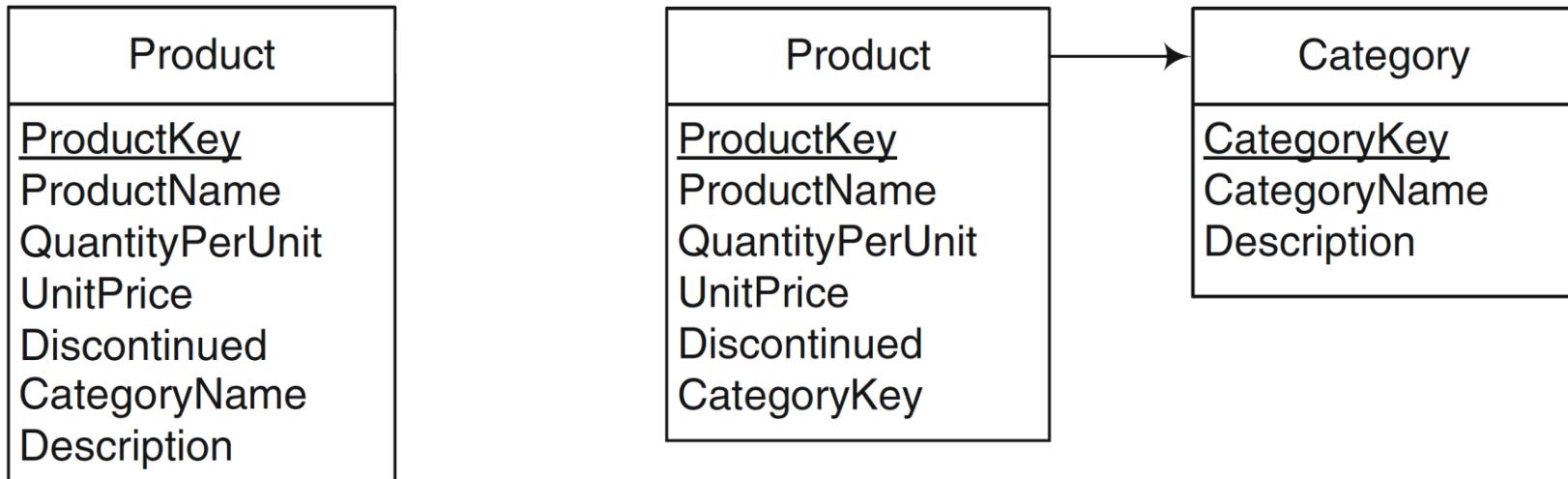
Types of hierarchy

- Balanced hierarchy
 - all levels are mandatory
 - all branches have the same length
 - a child member belongs to only one parent



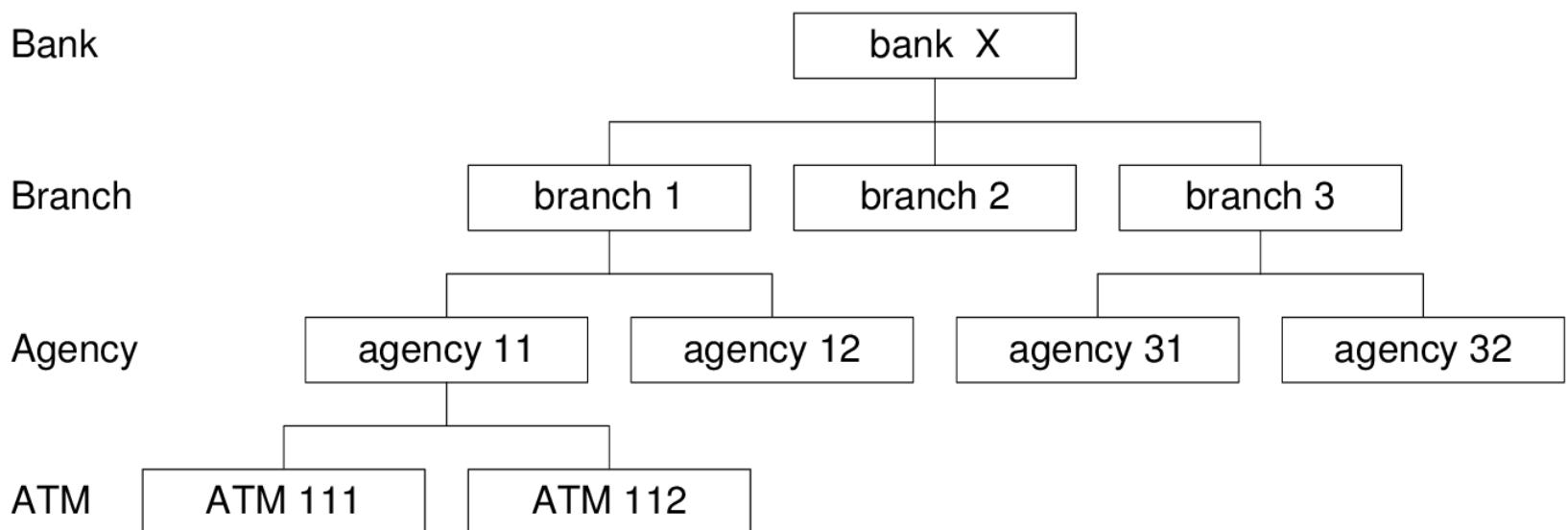
Types of hierarchy

- Balanced hierarchy
 - flat table (as in star schema) or snowflake structure



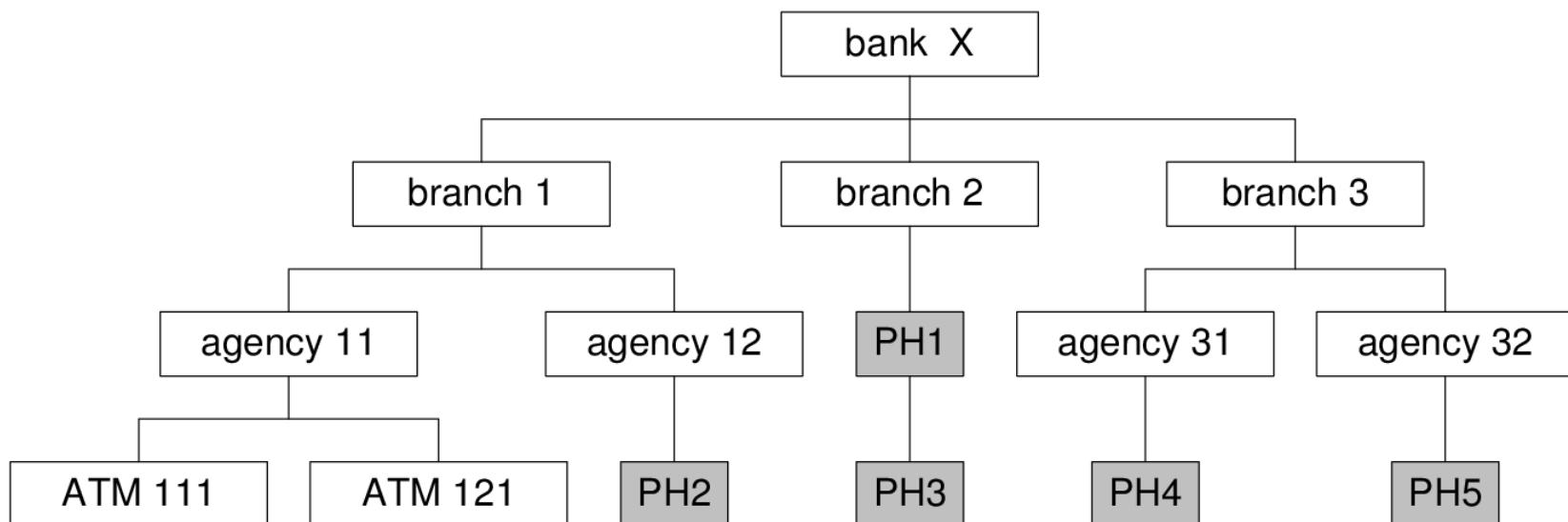
Types of hierarchy

- Unbalanced hierarchy
 - some levels are optional
 - branches may have different lengths
 - a child member belongs to only one parent



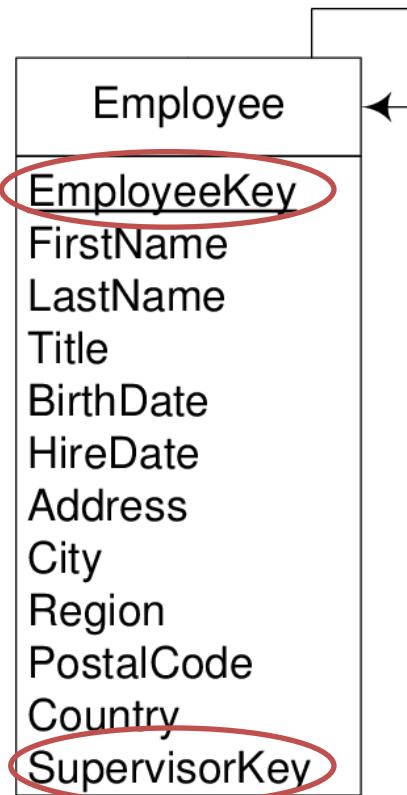
Types of hierarchy

- Unbalanced hierarchy
 - transform into balanced hierarchy by using placeholders
 - then use flat table or snowflake structure



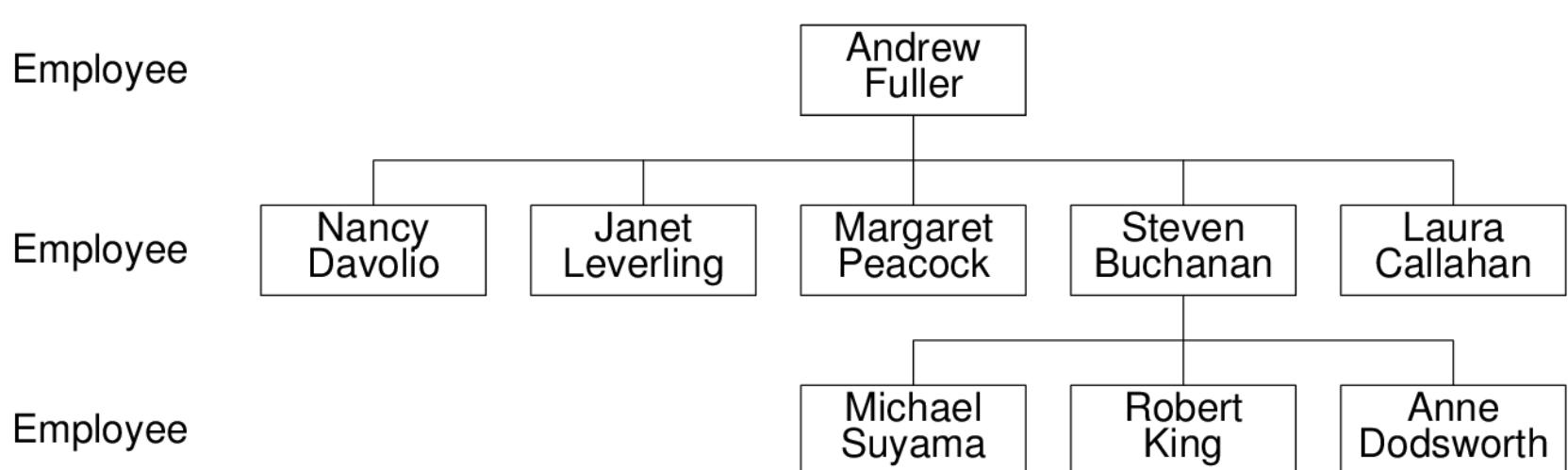
Types of hierarchy

- Recursive hierarchy
 - a scenario that we have seen before



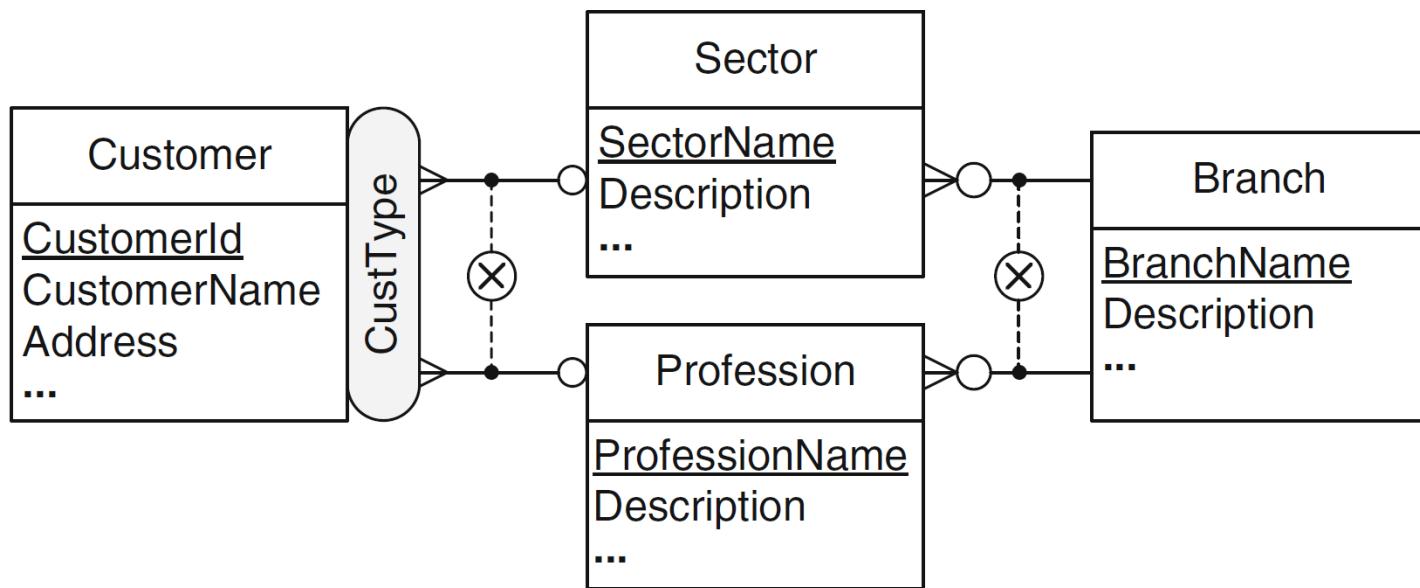
Types of hierarchy

- Recursive hierarchy
 - all levels are of the same type
 - can easily become an unbalanced hierarchy
 - requires recursive queries to traverse the hierarchy



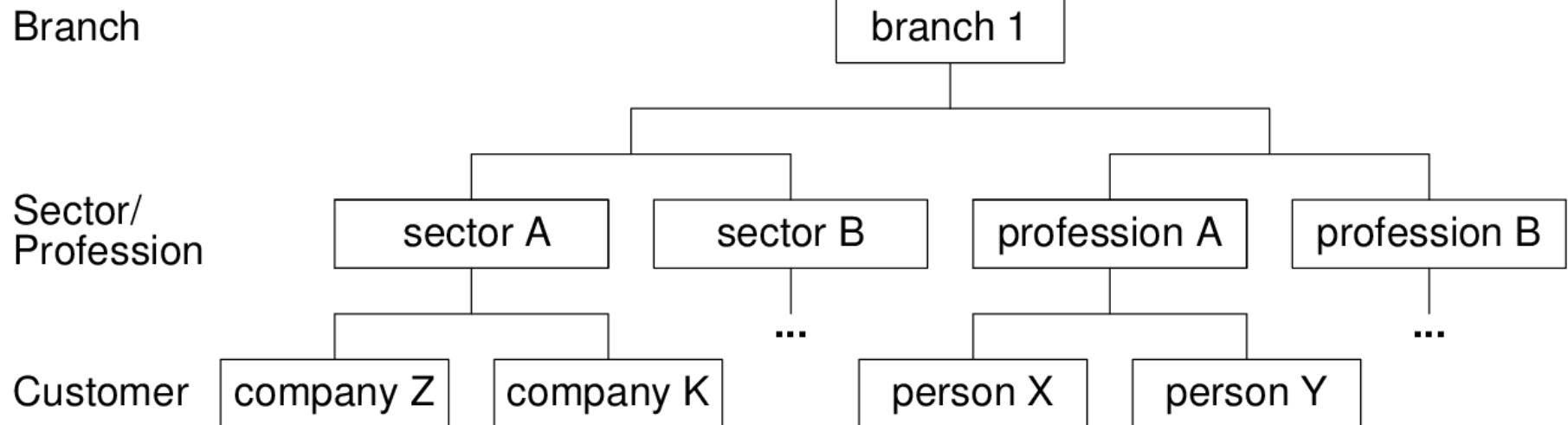
Types of hierarchy

- Generalized hierarchy
 - the same level may have different types
 - e.g. customers of a bank may be companies (with an industry **sector**) or individual persons (with a **profession**)



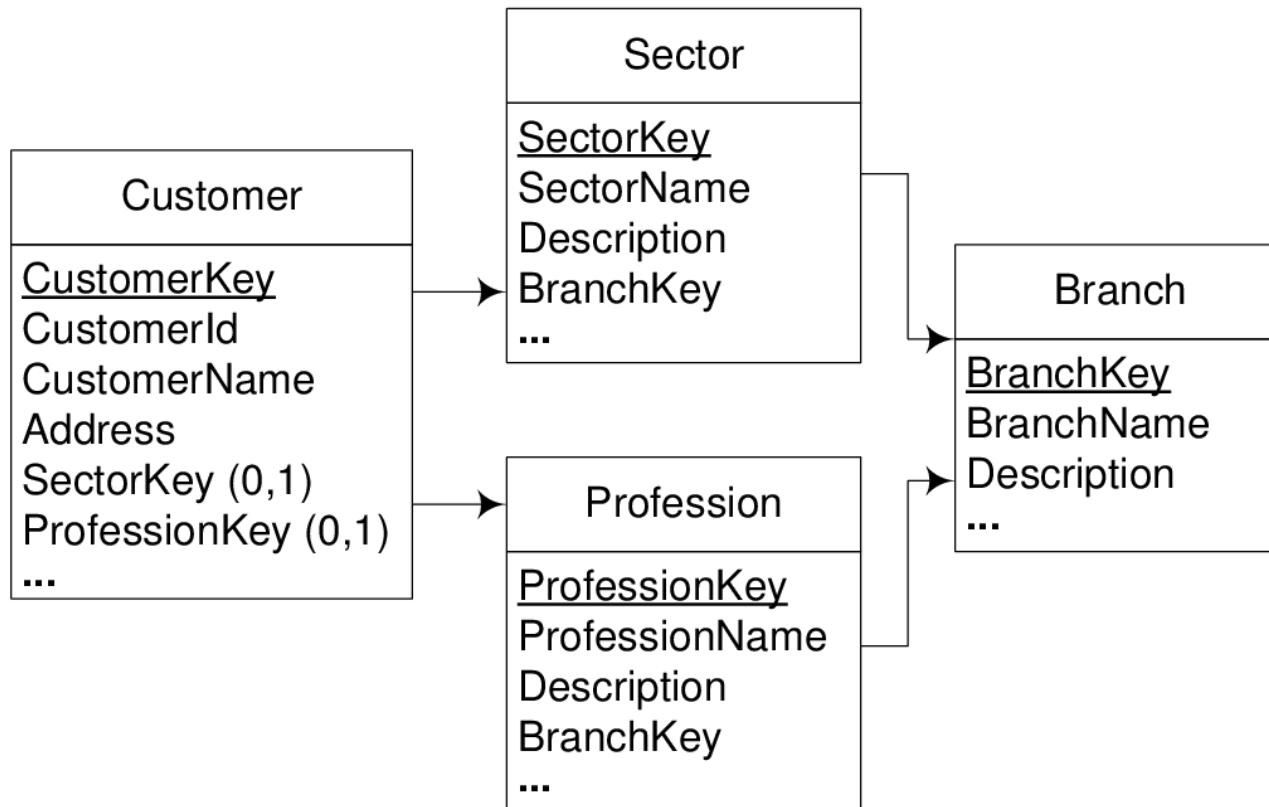
Types of hierarchy

- Generalized hierarchy
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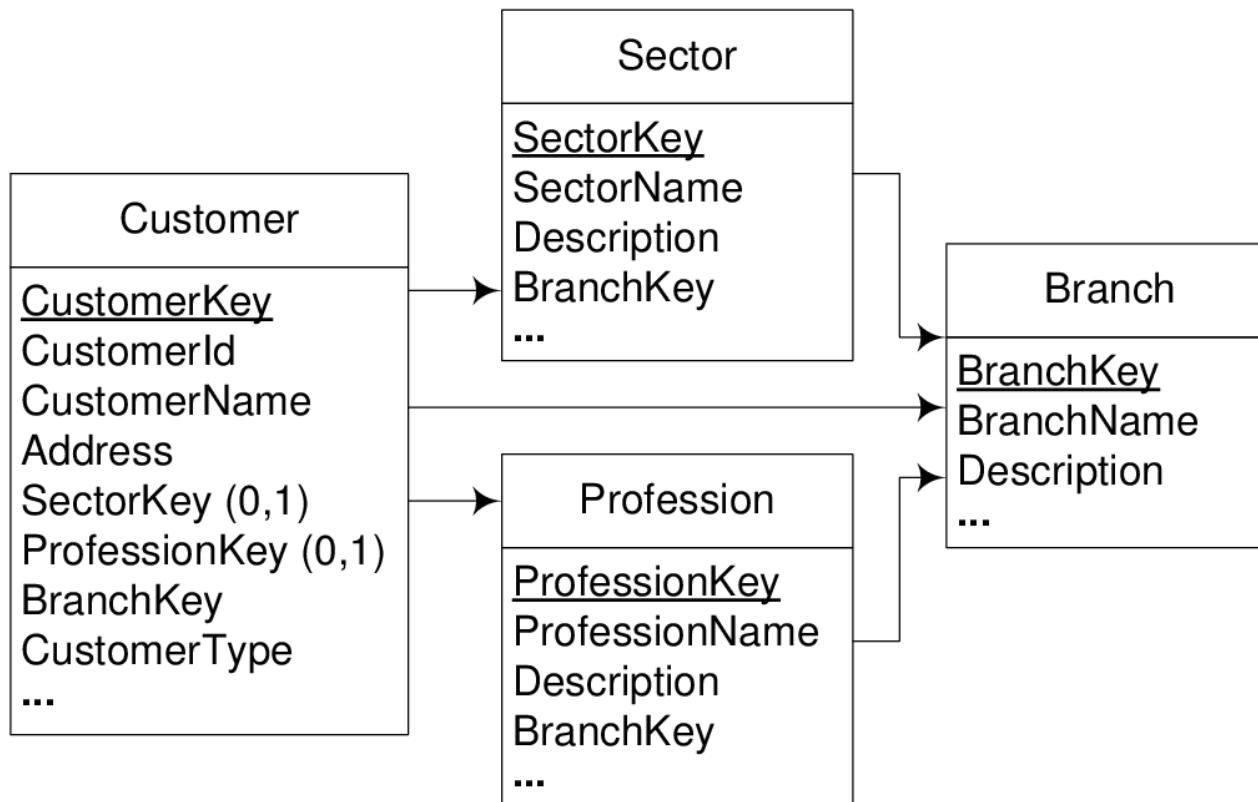
Types of hierarchy

- Generalized hierarchy
 - flat table with NULLs or snowflake structure (preferred)
 - different aggregation paths for different types of customer



Types of hierarchy

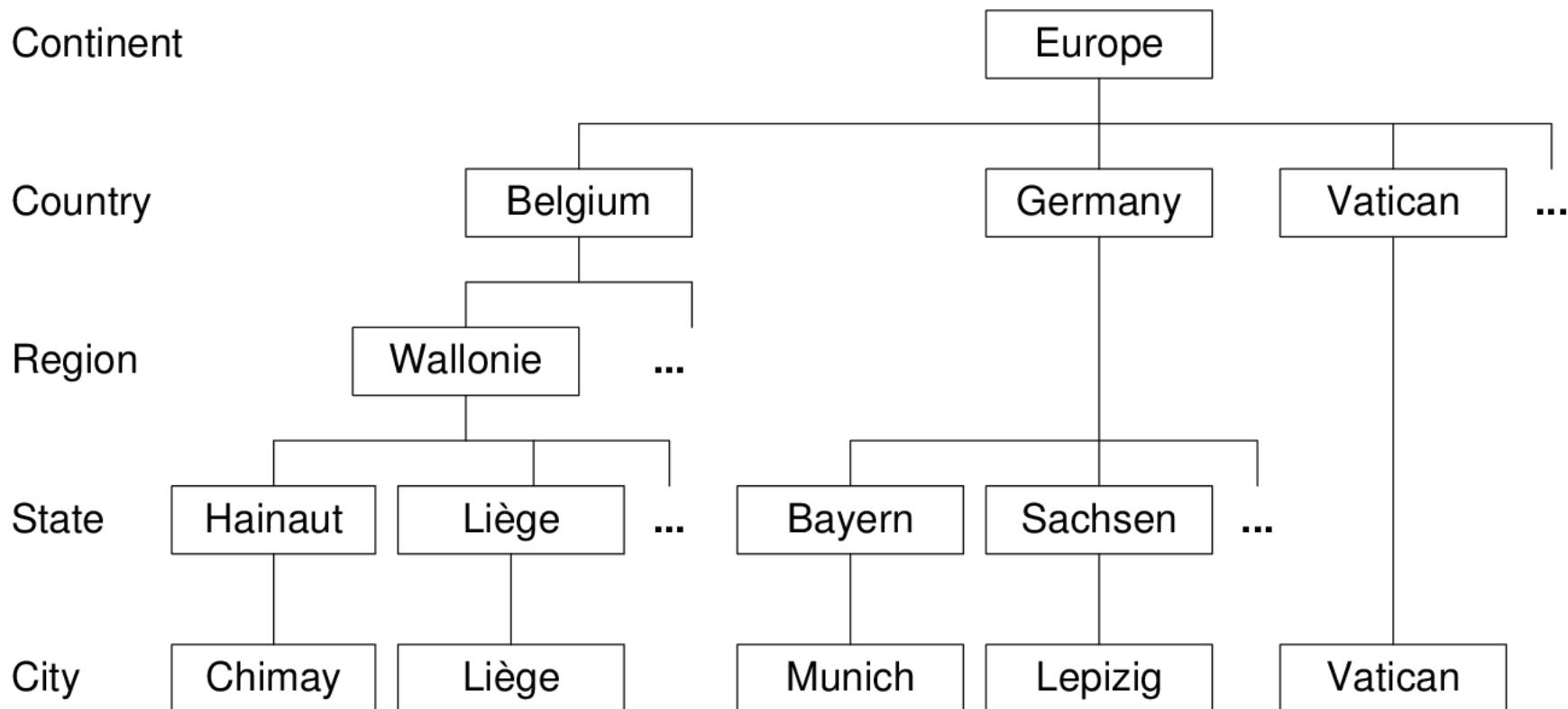
- Generalized hierarchy
 - possibility of using extra foreign key to roll-up directly to upper level



Types of hierarchy

- Ragged hierarchy
 - one or more levels can be skipped

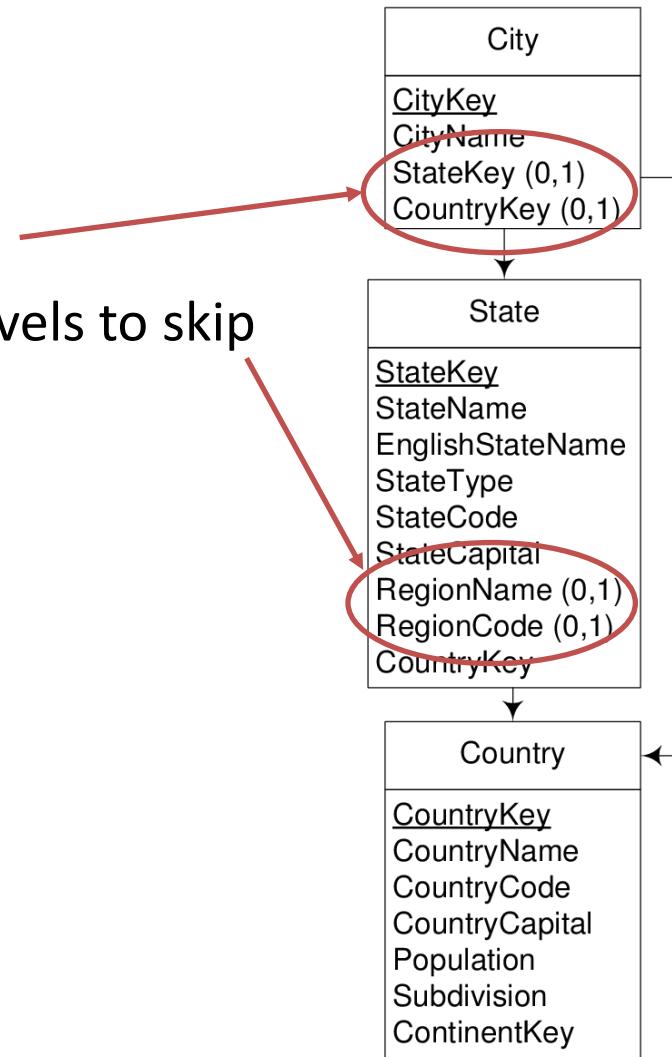
Continent



Types of hierarchy

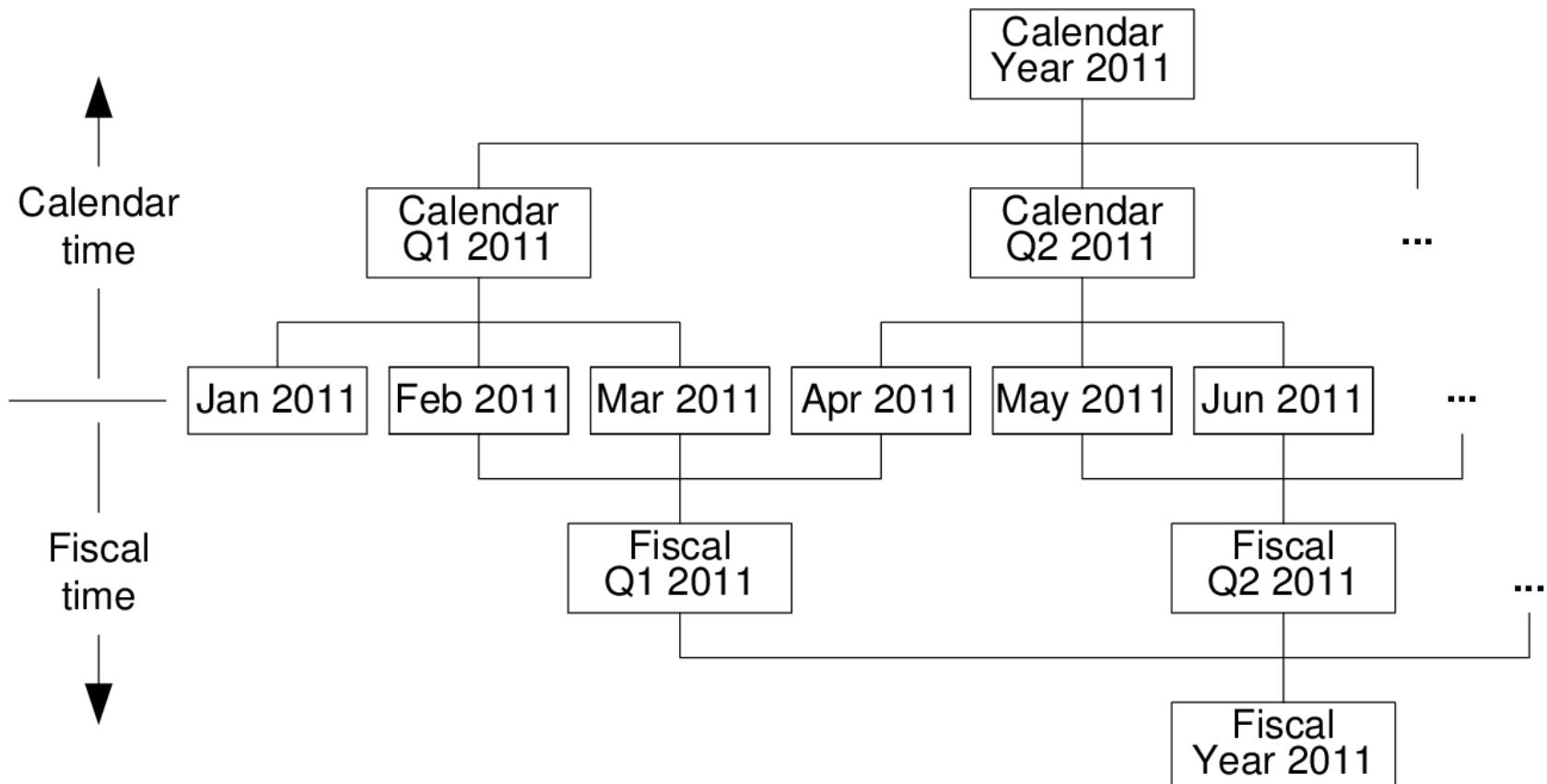
- Ragged hierarchy
 - several implementations

- add extra foreign keys to skip levels
- or use optional attributes for the levels to skip
- or use placeholders



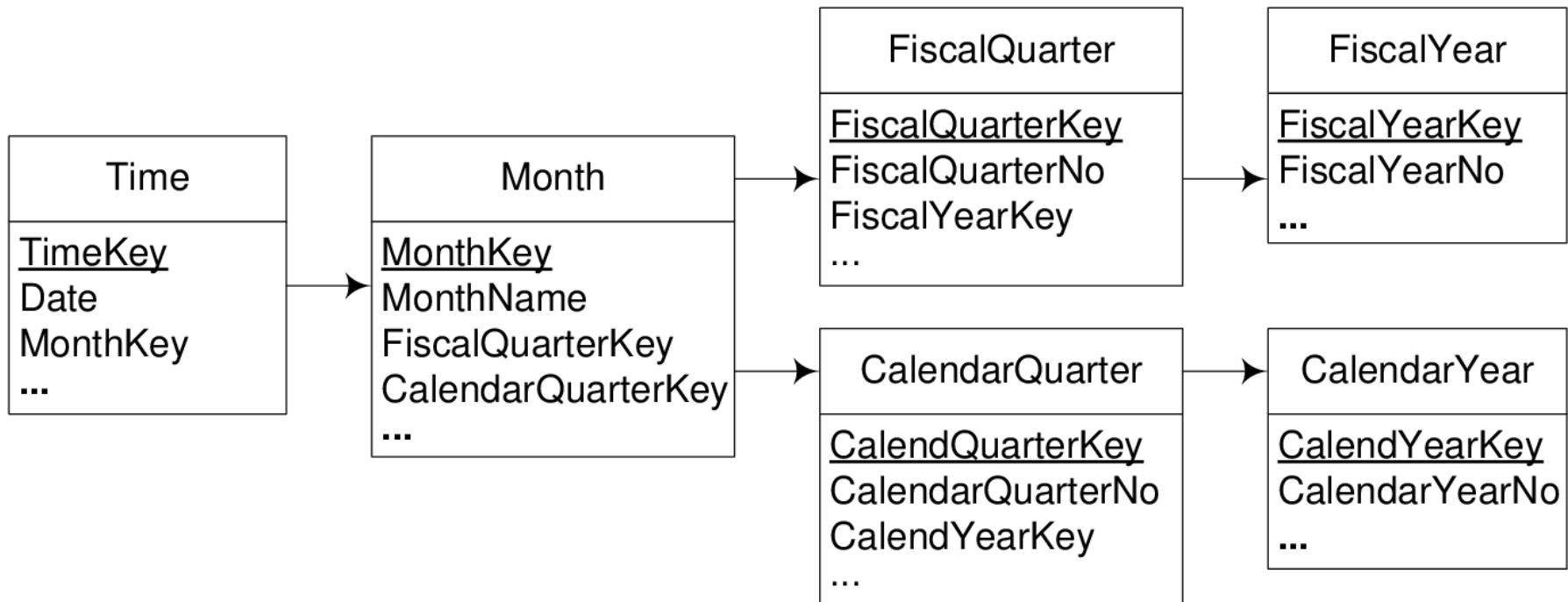
Types of hierarchy

- Alternative hierarchy
 - the same level has alternative aggregation paths



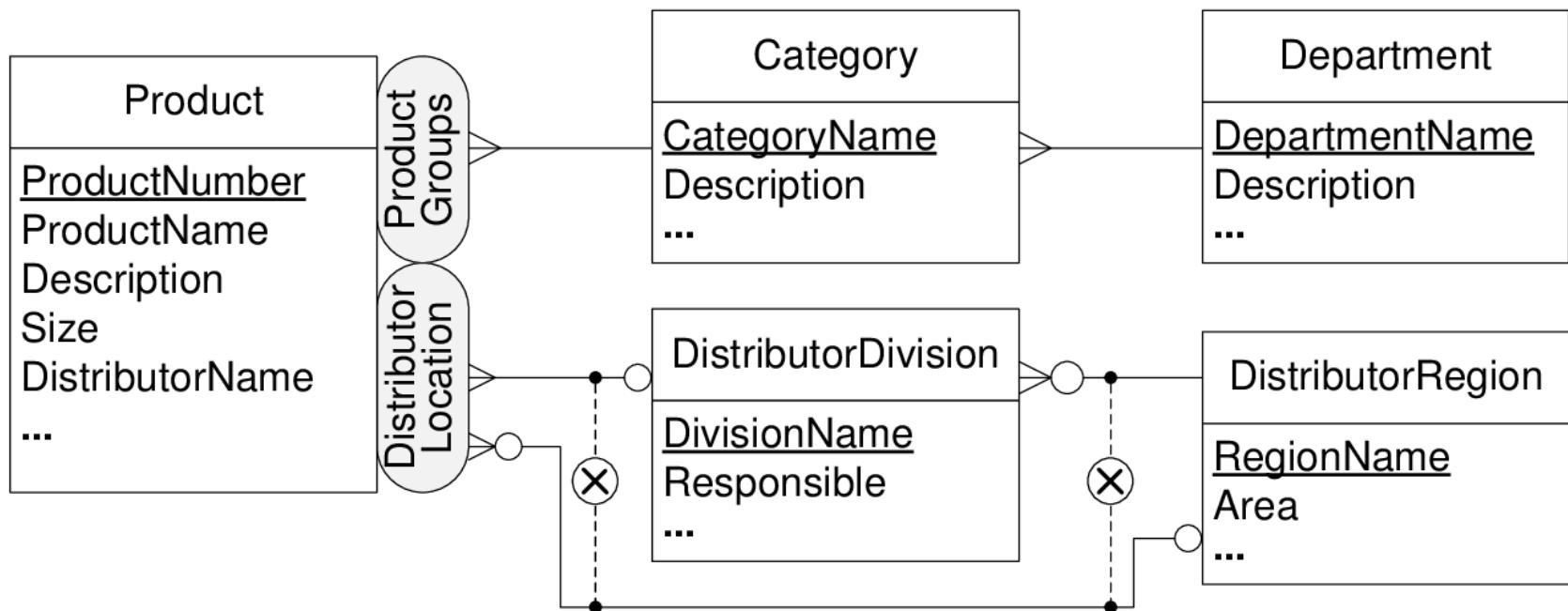
Types of hierarchy

- Alternative hierarchy
 - use snowflake structure



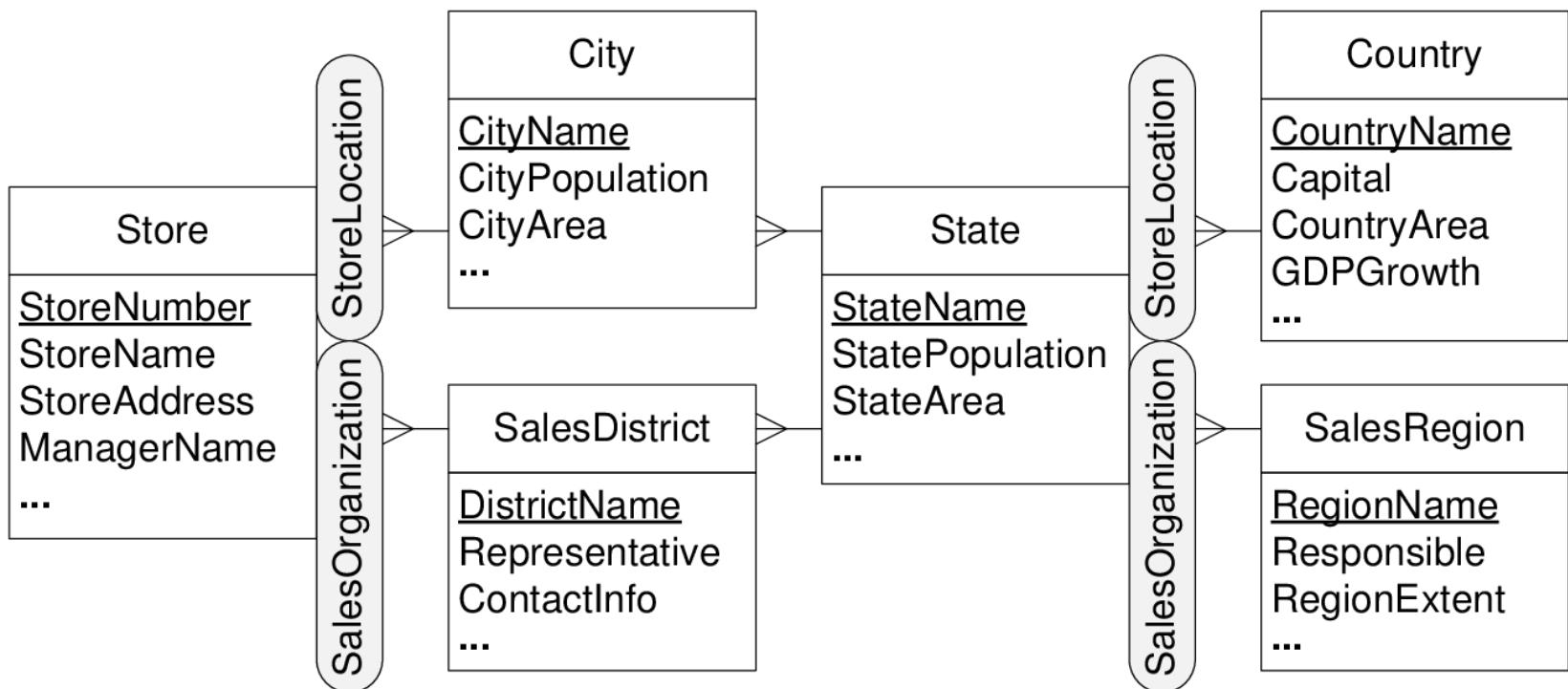
Types of hierarchy

- Parallel hierarchies
 - the same dimension has several hierarchies
 - example of two parallel **independent** hierarchies



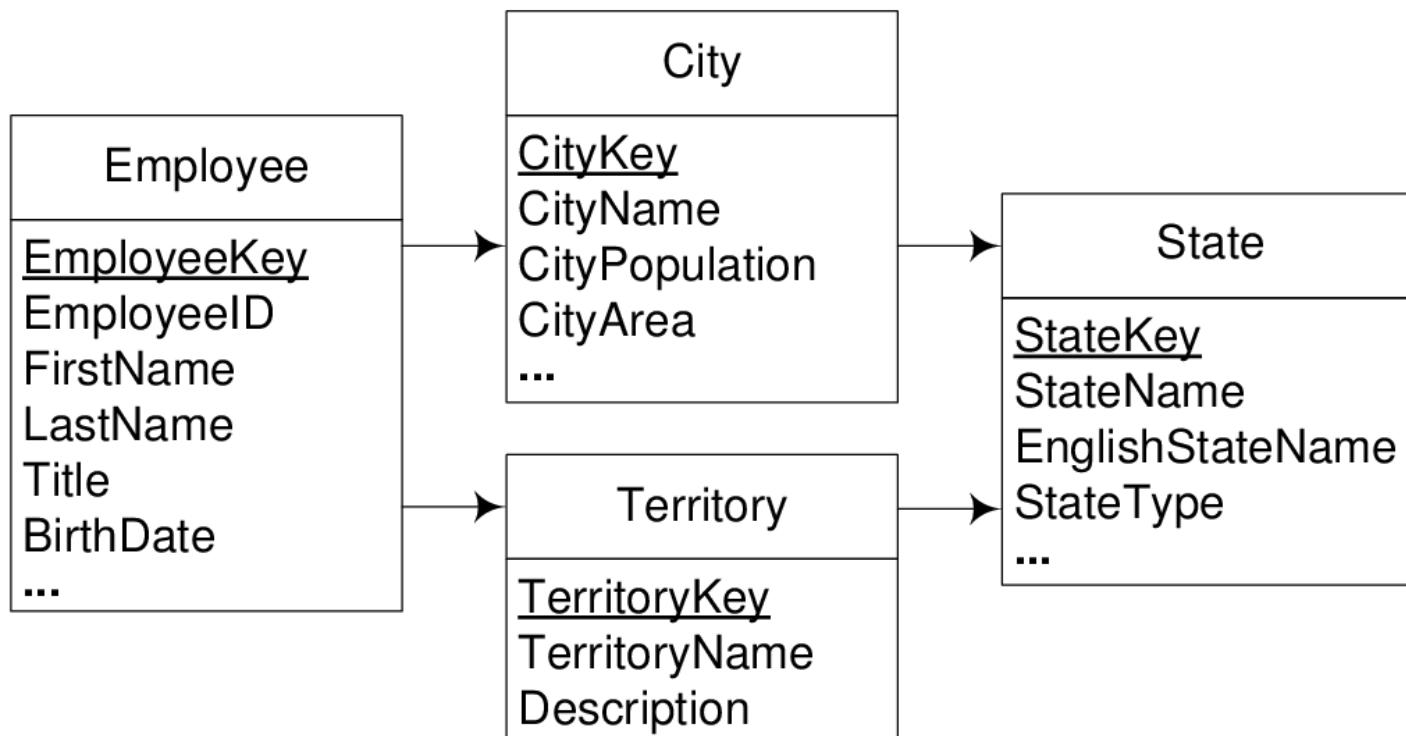
Types of hierarchy

- Parallel hierarchies
 - the same dimension has several hierarchies
 - example of two parallel **dependent** hierarchies



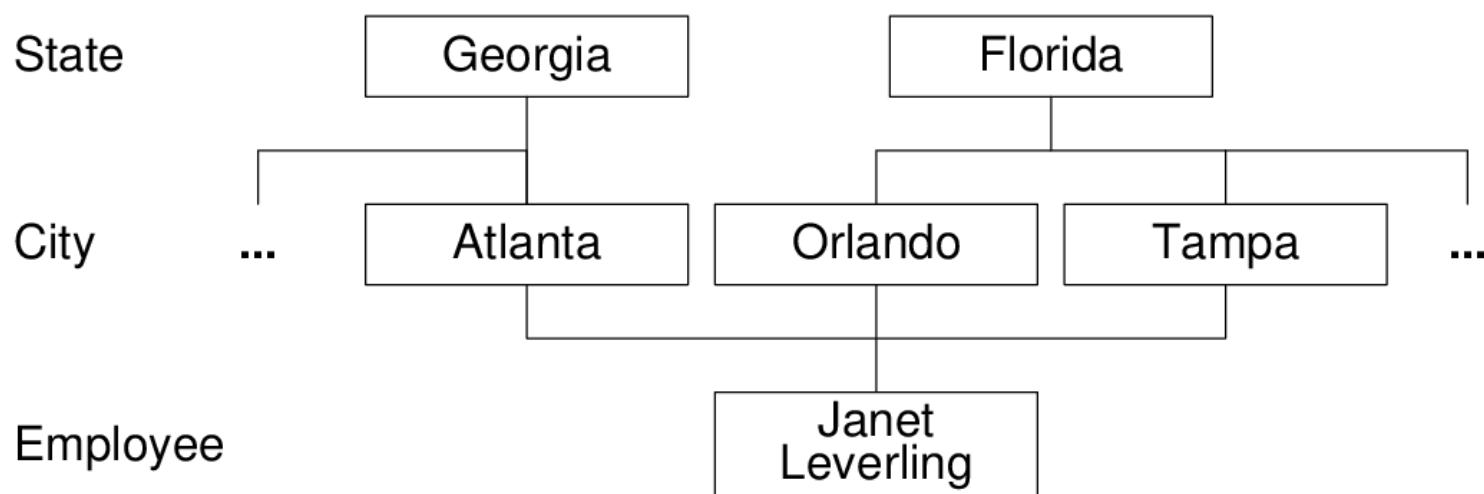
Types of hierarchy

- Parallel hierarchies
 - use snowflake structure



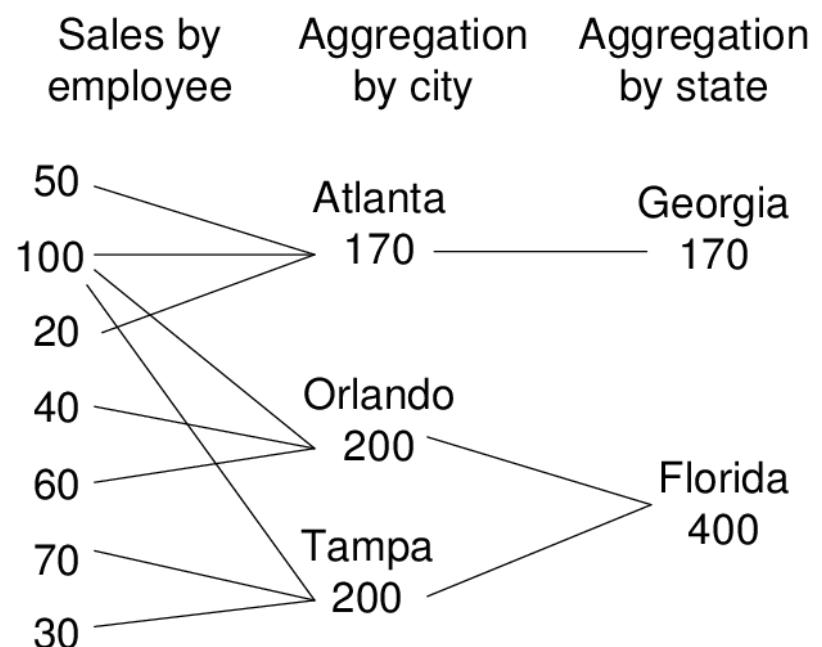
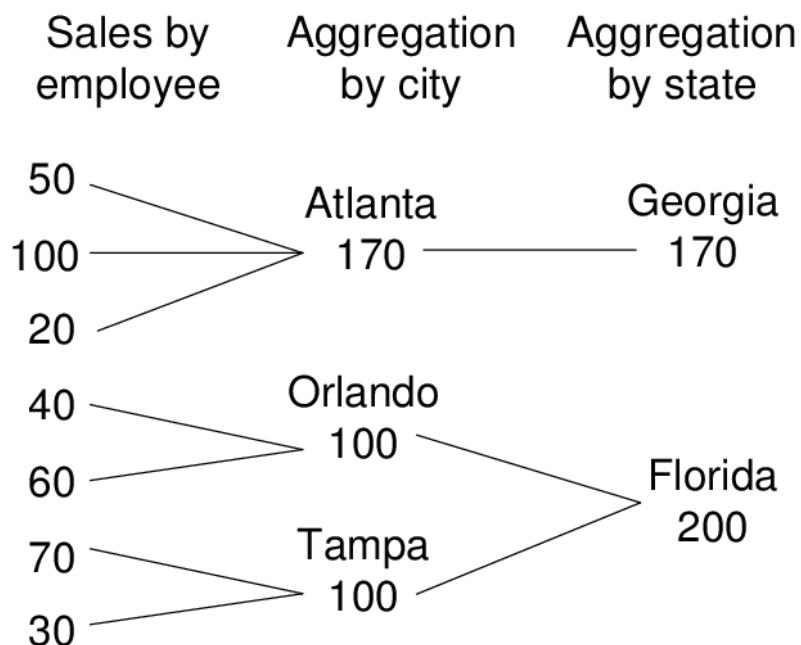
Types of hierarchy

- Non-strict hierarchy
 - a member may have several parents
 - e.g. an employee that works in multiple cities
 - e.g. a week that belongs to two months



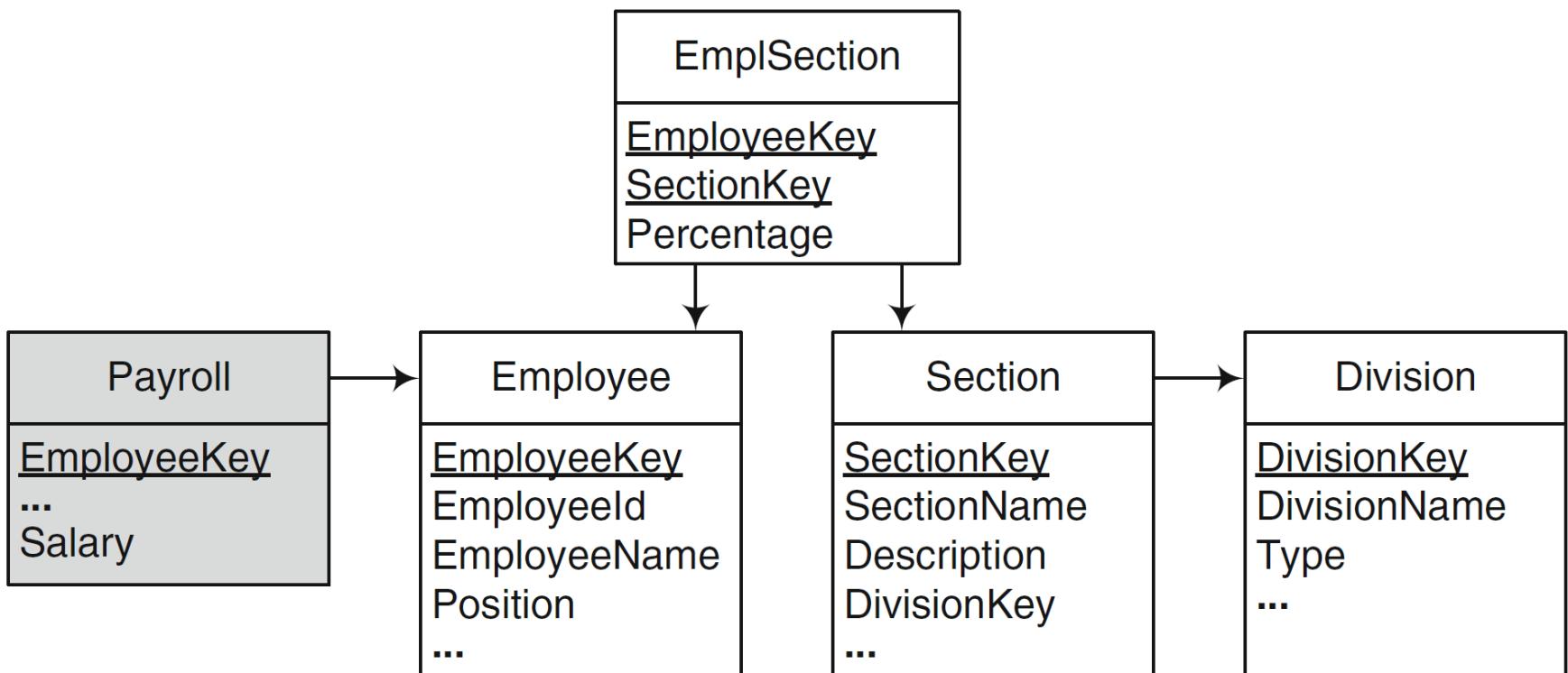
Types of hierarchy

- Non-strict hierarchy
 - must be careful to avoid **double counting** on roll-up



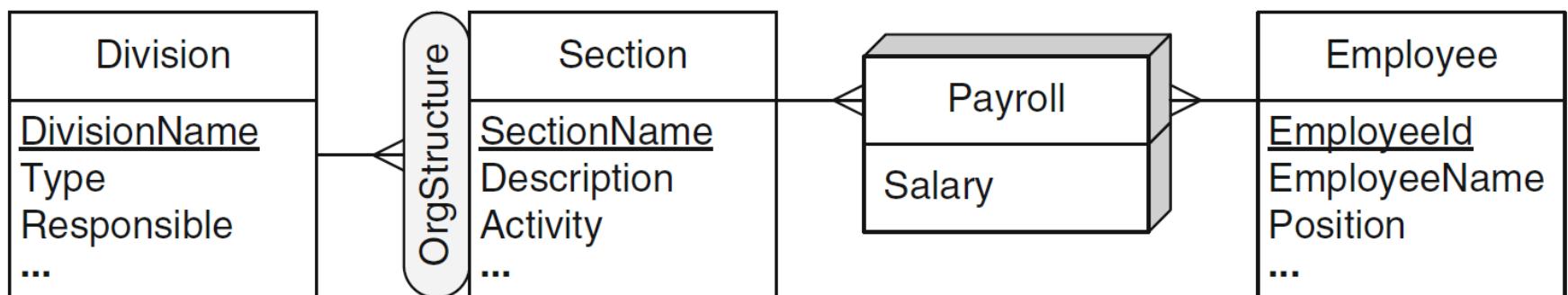
Types of hierarchy

- Non-strict hierarchy
 - use **bridge table** with percentage (%)
 - distributing attribute



Types of hierarchy

- Non-strict hierarchy
 - another solution is to re-design the DW schema using two separate dimensions



Measures

- Each measure is associated to an **aggregation function** that combines several values into a single one
 - the aggregation takes place whenever we change to a different level in a dimension hierarchy
- When defining a measure we must decide the associated aggregation function
 - **sum** is the most typical, but it may not always apply
 - some aggregation functions may not apply to a measure, or to a measure on a certain dimension

Measures

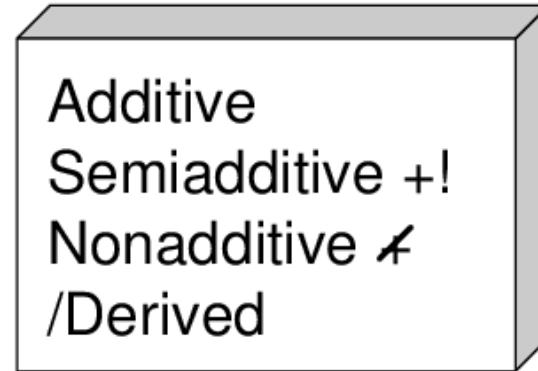
- Additive measures
 - can be aggregated along all dimensions using addition (sum)
 - e.g. sales amount along customer, product and time
- Semi-additive measures
 - can be added along some, but not all dimensions
 - e.g. inventory level cannot be summed along time
- Non-additive measures
 - cannot be added along any dimension
 - e.g. unit price, exchange rate

Measures

- What to do about semi- or non-additive measures
 - use other forms of aggregation
 - average (e.g. average inventory level over time)
 - minimum (e.g. minimum exchange rate over space or time)
 - maximum (e.g. maximum unit price over space or time)

Measures

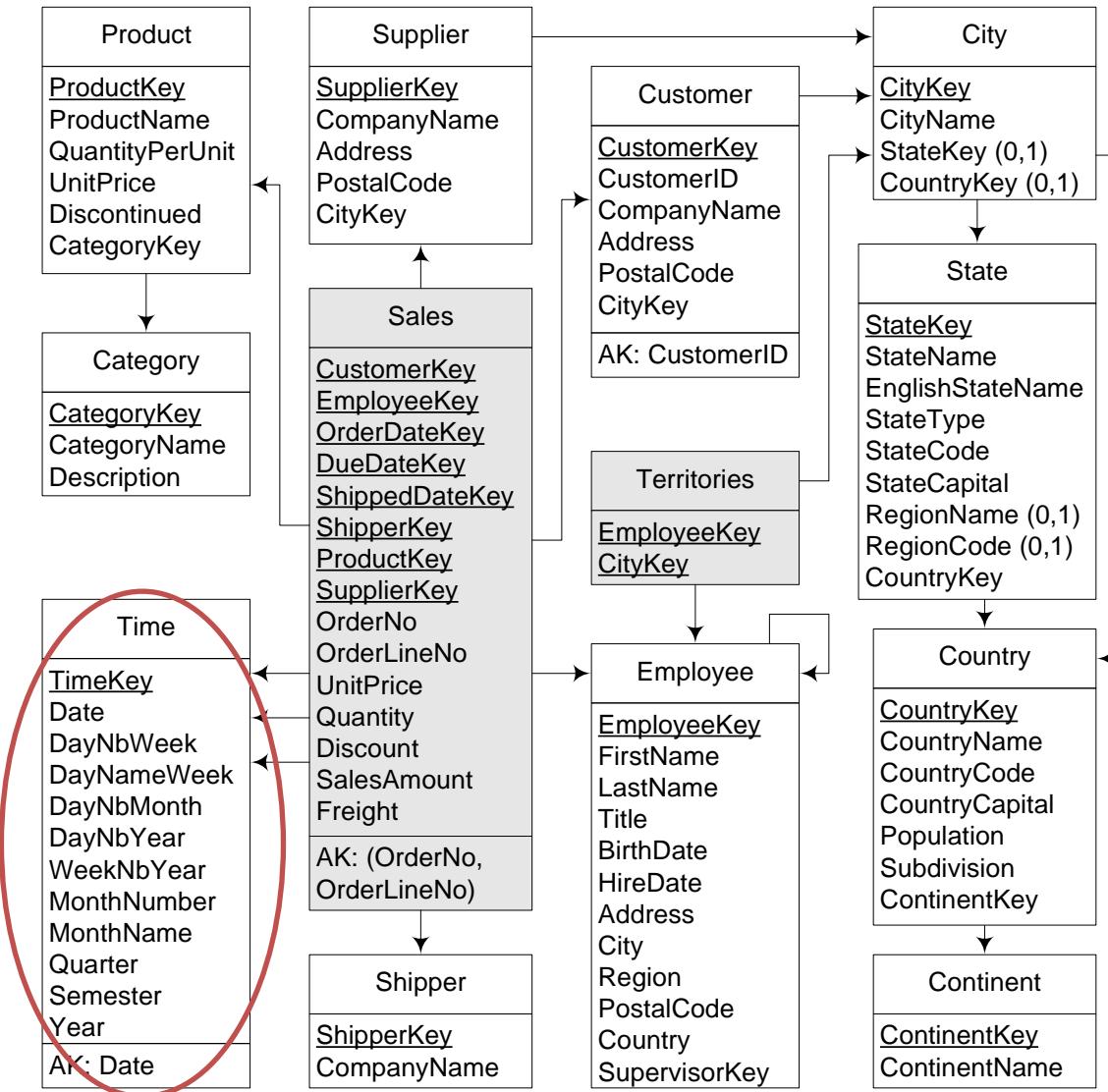
- Derived measures
 - can be calculated from other measures or attributes
 - e.g. two measures: **sales amount** and **tax amount**
 - then **net amount** can be derived as a third measure
net amount = sales amount – tax amount



Time dimension

- A data warehouse is a historical database so the time dimension is present in almost all DWs
 - in star/snowflake schema, time is included both as a foreign key in the fact table and as a time dimension containing the associated hierarchy levels
- In transactional databases, time information is stored in attributes of a DATE data type
 - e.g. weekend is computed on-the-fly using appropriate functions
- In a data warehouse, time information is stored explicitly in multiple attributes in the time dimension
 - easier to compute queries, e.g. total sales during weekends

Time dimension



Time dimension

- Granularity of time dimension depends on use
 - if interested in monthly data only, define the time dimension with granularity of month
 - time dimension with a month granularity spanning 5 years will have $5*12 = 60$ tuples
 - if the granularity is second, then the dimension time will have:
 $5*12*30*24*3600 = 155\,520\,000$ tuples
- Time dimension may have more than one hierarchy
 - e.g. fiscal and calendar year
- Time dimension can often be populated automatically