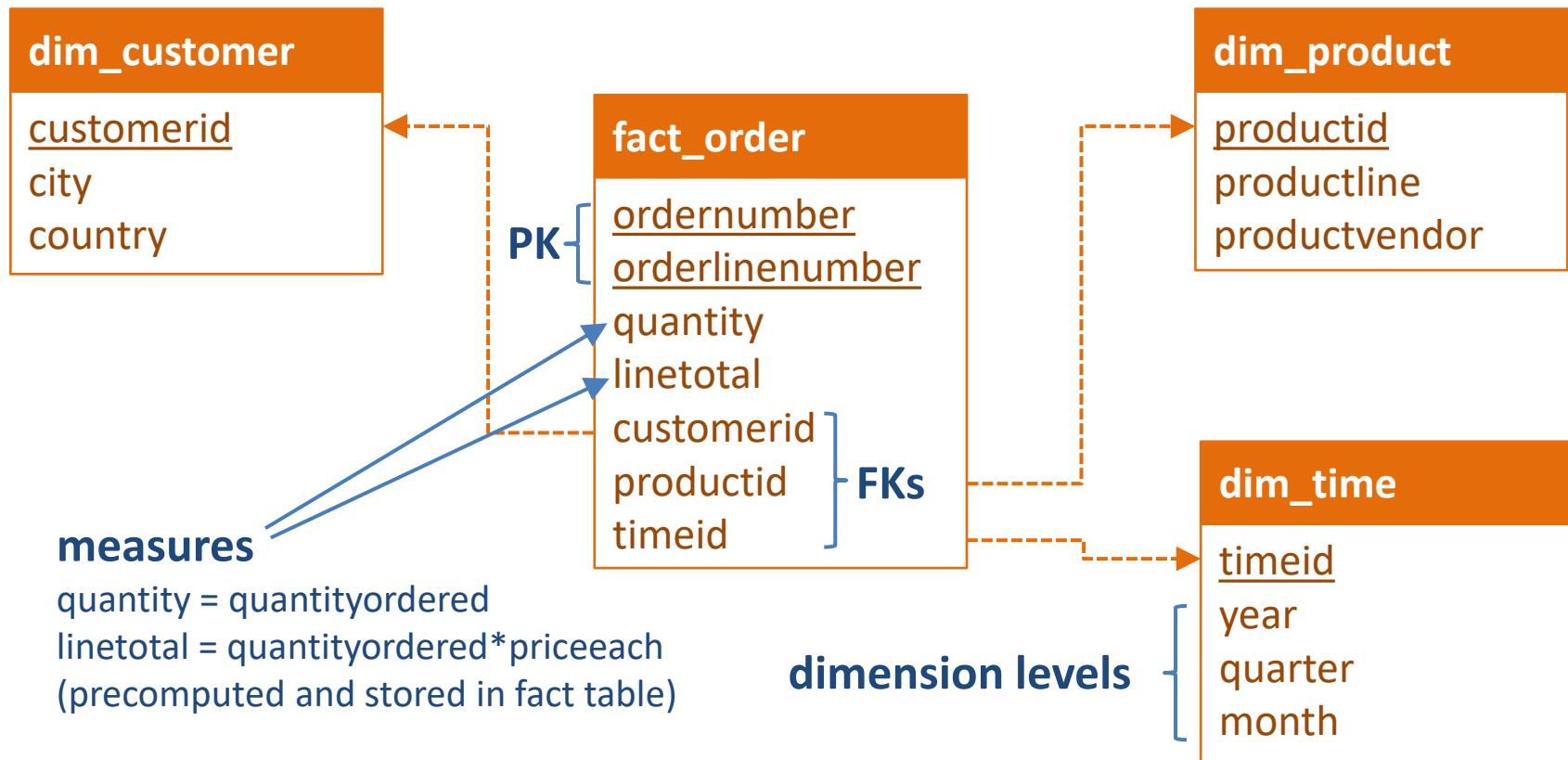


Data Analysis and Integration

OLAP operations

Data warehousing

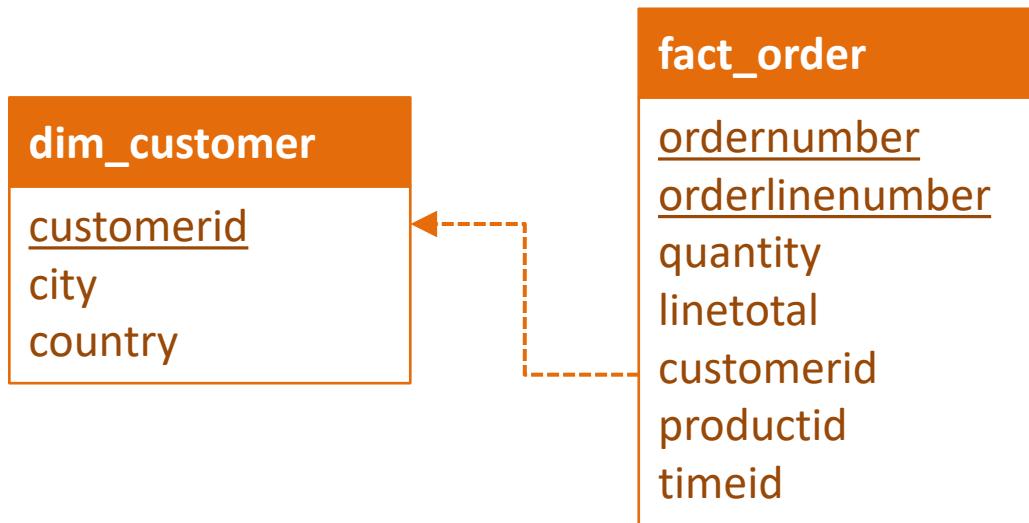
- Star schema



Data warehousing

- Sales by customer country

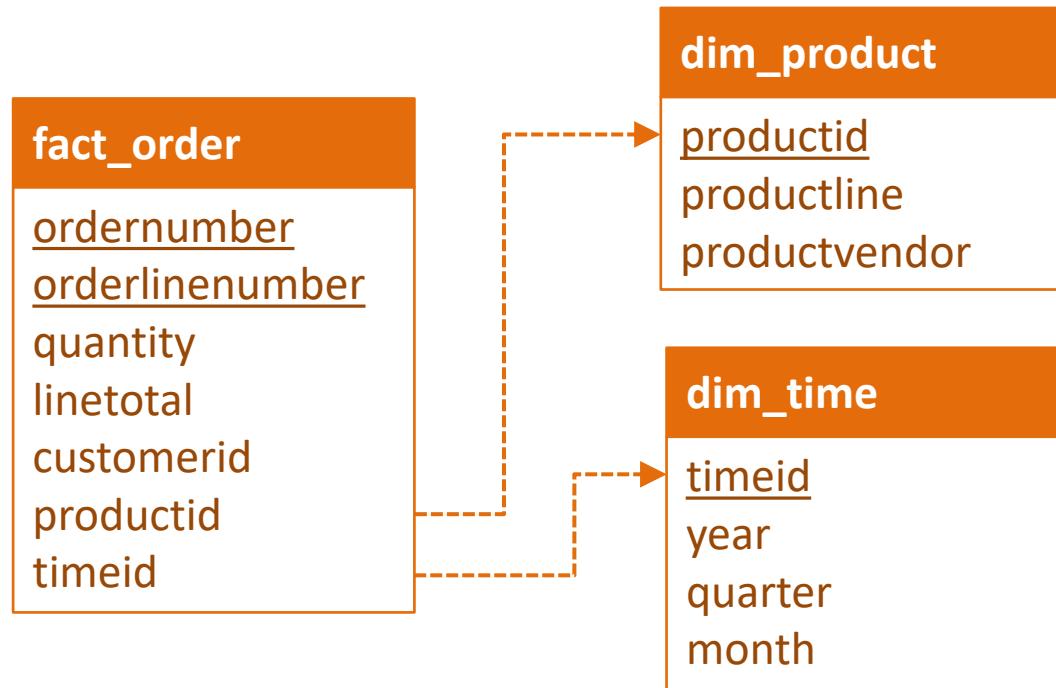
```
select country, sum(linetotal) as sales
from fact_order natural join dim_customer
group by country;
```



Data warehousing

- Sales by product line and year

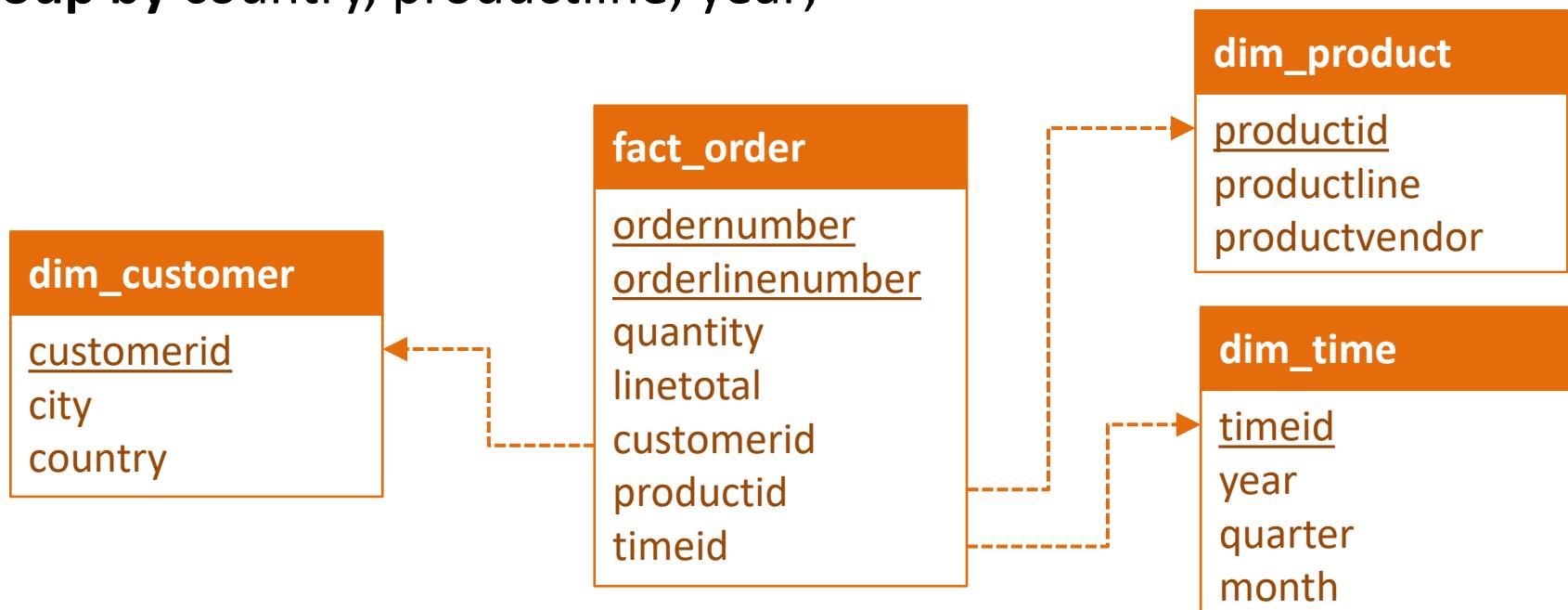
```
select productline, year, sum(linetotal) as sales  
from fact_order natural join dim_product natural join dim_time  
group by productline, year;
```



Data warehousing

- Sales by customer country, product line and year

```
select country, productline, year, sum(linetotal) as sales  
from fact_order natural join dim_customer natural join dim_product  
natural join dim_time  
group by country, productline, year;
```



Dimension levels

- Each **dimension** can have different **levels**
 - customer (city; country)
 - product (productline; productvendor)
 - time (year; quarter; month)

Drill-down

- Going from a higher to a lower level is called **drill-down**
 - e.g. sales by customer country → sales by customer city

```
select country, sum(linetotal) as sales  
from fact_order natural join dim_customer  
group by country;
```



```
select city, sum(linetotal) as sales  
from fact_order natural join dim_customer  
group by city;
```



Drill-down

- Going from a higher to a lower level is called **drill-down**
 - e.g. sales by customer country → sales by customer city

```
select country, sum(linetotal) as sales  
from fact_order natural join dim_customer  
group by country;
```



```
select country, city, sum(linetotal) as sales  
from fact_order natural join dim_customer  
group by country, city;
```



Drill-down

- Going from a higher to a lower level is called **drill-down**
 - e.g. sales by customer country → sales by customer city

country	sales
Australia	630638
Austria	202089
Belgium	108485
Canada	224085
Denmark	245582
Finland	329472
France	1111022
Germany	220354
Hong Kong	48766
Ireland	57788
Italy	403696
Japan	188212
...	...

21 rows in set (0.02 sec)

country	city	sales
Australia	Chatswood	151631
Australia	Glen Waverly	64621
Australia	Melbourne	200845
Australia	North Sydney	154070
Australia	South Brisbane	59471
Austria	Graz	52218
Austria	Salzburg	149871
Belgium	Bruxelles	75037
Belgium	Charleroi	33448
Canada	Montréal	74224
Canada	Tsawassen	74665
Canada	Vancouver	75196
...

81 rows in set (0.03 sec)

Drill-down

- Another example of **drill-down**
 - e.g. sales by year → sales by month

```
select year, sum(linetotal) as sales  
from fact_order natural join dim_time  
group by year;
```



```
select year, month, sum(linetotal) as sales  
from fact_order natural join dim_time  
group by year, month;
```

Drill-down

- Another example of **drill-down**
 - e.g. sales by year → sales by month

year	sales
2003	4312435
2004	4987780
2005	1980850

3 rows in set (0.02 sec)

year	month	sales
2003	1	764883
2003	2	140920
2003	3	174467
2003	4	201557
2003	5	192785
2003	6	170533
2003	7	225638
2003	8	197822
2003	9	263836
2003	10	589773
2003	11	1086757
2003	12	303464
2004	1	316662
2004	2	318663
...

29 rows in set (0.01 sec)

Roll-up

- Going from a lower to a higher level is called **roll-up**
 - e.g. sales by customer city → sales by customer country

```
select country, city, sum(linetotal) as sales  
from fact_order natural join dim_customer  
group by country, city;
```



```
select country, sum(linetotal) as sales  
from fact_order natural join dim_customer  
group by country;
```

Roll-up

- Going from a lower to a higher level is called **roll-up**
 - e.g. sales by customer city → sales by customer country

country	city	sales
Australia	Chatswood	151631
Australia	Glen Waverly	64621
Australia	Melbourne	200845
Australia	North Sydney	154070
Australia	South Brisbane	59471
Austria	Graz	52218
Austria	Salzburg	149871
Belgium	Bruxelles	75037
Belgium	Charleroi	33448
Canada	Montréal	74224
Canada	Tsawassen	74665
Canada	Vancouver	75196
...

81 rows in set (0.03 sec)

country	sales
Australia	630638
Austria	202089
Belgium	108485
Canada	224085
Denmark	245582
Finland	329472
France	1111022
Germany	220354
Hong Kong	48766
Ireland	57788
Italy	403696
Japan	188212
...	...

21 rows in set (0.02 sec)

Roll-up

- Another example of **roll-up**
 - e.g. sales by month → sales by year

```
select year, month, sum(linetotal) as sales  
from fact_order natural join dim_time  
group by year, month;
```



```
select year, sum(linetotal) as sales  
from fact_order natural join dim_time  
group by year;
```

Roll-up

- Another example of **roll-up**
 - e.g. sales by month → sales by year

year	month	sales
2003	1	764883
2003	2	140920
2003	3	174467
2003	4	201557
2003	5	192785
2003	6	170533
2003	7	225638
2003	8	197822
2003	9	263836
2003	10	589773
2003	11	1086757
2003	12	303464
2004	1	316662
2004	2	318663
...

29 rows in set (0.01 sec)

year	sales
2003	4312435
2004	4987780
2005	1980850

3 rows in set (0.02 sec)

Slice

- Selecting a particular value of dimension level is a **slice**
 - e.g. sales by product line in 2003

```
select productline, year, sum(linetotal) as sales  
from fact_order natural join dim_product natural join dim_time  
group by productline, year;
```



```
select productline, year, sum(linetotal) as sales  
from fact_order natural join dim_product natural join dim_time  
group by productline, year  
having year = 2003;
```



Slice

- Selecting a particular value of dimension level is a **slice**
 - e.g. sales by product line in 2003

```
select productline, year, sum(linetotal) as sales  
from fact_order natural join dim_product natural join dim_time  
group by productline, year;
```



```
select productline, year, sum(linetotal) as sales  
from fact_order natural join dim_product natural join dim_time  
where year = 2003  
group by productline, year;
```



Slice

- Selecting a particular value of dimension level is a **slice**
 - e.g. sales by product line in 2003

productline	year	sales
Classic Cars	2003	1513998
Classic Cars	2004	1837904
Classic Cars	2005	738587
Motorcycles	2003	397392
Motorcycles	2004	590632
Motorcycles	2005	286327
Planes	2003	347924
Planes	2004	529129
Planes	2005	200077
Ships	2003	244652
Ships	2004	375498
Ships	2005	128219
Trains	2003	72857
Trains	2004	124885
Trains	2005	36920
...

21 rows in set (0.04 sec)



productline	year	sales
Classic Cars	2003	1513998
Motorcycles	2003	397392
Planes	2003	347924
Ships	2003	244652
Trains	2003	72857
Trucks and Buses	2003	420523
Vintage Cars	2003	1315089

7 rows in set (0.01 sec)

Dice

- Applying multiple slicing conditions is called a **dice**
 - e.g. sales of Motorcycles in 2003 and 2004

```
select productline, year, sum(linetotal) as sales  
from fact_order natural join dim_product natural join dim_time  
group by productline, year;
```



```
select productline, year, sum(linetotal) as sales  
from fact_order natural join dim_product natural join dim_time  
where productline = 'Motorcycles' and year in (2003, 2004)  
group by productline, year;
```

Dice

- Applying multiple slicing conditions is called a **dice**
 - e.g. sales of Motorcycles in 2003 and 2004

productline	year	sales
Classic Cars	2003	1513998
Classic Cars	2004	1837904
Classic Cars	2005	738587
Motorcycles	2003	397392
Motorcycles	2004	590632
Motorcycles	2005	286327
Planes	2003	347924
Planes	2004	529129
Planes	2005	200077
Ships	2003	244652
Ships	2004	375498
Ships	2005	128219
Trains	2003	72857
Trains	2004	124885
Trains	2005	36920
...

21 rows in set (0.04 sec)



productline	year	sales
Motorcycles	2003	397392
Motorcycles	2004	590632

2 rows in set (0.01 sec)

Pivot

- Changing the order of dimensions is called **pivot**
 - e.g. sales by product line, year → sales by year, product line

```
select productline, year, sum(linetotal) as sales  
from fact_order natural join dim_product natural join dim_time  
group by productline, year;
```



```
select year, productline, sum(linetotal) as sales  
from fact_order natural join dim_product natural join dim_time  
group by year, productline;
```

Pivot

- Changing the order of dimensions is called **pivot**
 - e.g. sales by product line, year → sales by year, product line

productline	year	sales
Classic Cars	2003	1513998
Classic Cars	2004	1837904
Classic Cars	2005	738587
Motorcycles	2003	397392
Motorcycles	2004	590632
Motorcycles	2005	286327
Planes	2003	347924
Planes	2004	529129
Planes	2005	200077
Ships	2003	244652
Ships	2004	375498
Ships	2005	128219
Trains	2003	72857
Trains	2004	124885
Trains	2005	36920
...

21 rows in set (0.04 sec)



year	productline	sales
2003	Classic Cars	1513998
2003	Motorcycles	397392
2003	Planes	347924
2003	Ships	244652
2003	Trains	72857
2003	Trucks and Buses	420523
2003	Vintage Cars	1315089
2004	Classic Cars	1837904
2004	Motorcycles	590632
2004	Planes	529129
2004	Ships	375498
2004	Trains	124885
2004	Trucks and Buses	532024
2004	Vintage Cars	997708
2005	Classic Cars	738587
...

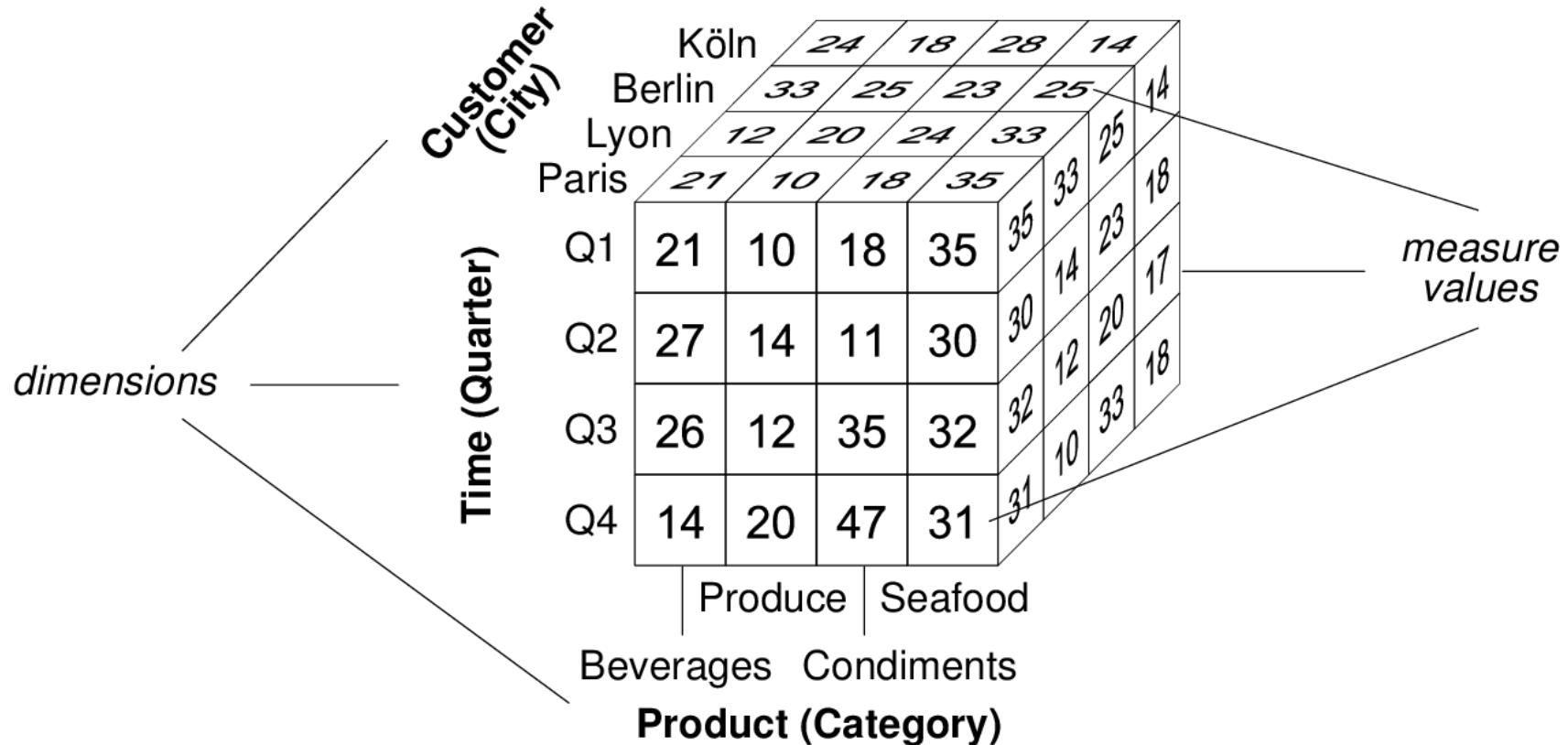
21 rows in set (0.04 sec)

OLAP operations

- Typical analytical operations
 - **drill-down** and **roll-up** between levels
 - **slice** and **dice** to select particular values
 - **pivot** to rotate dimensions
 - etc.

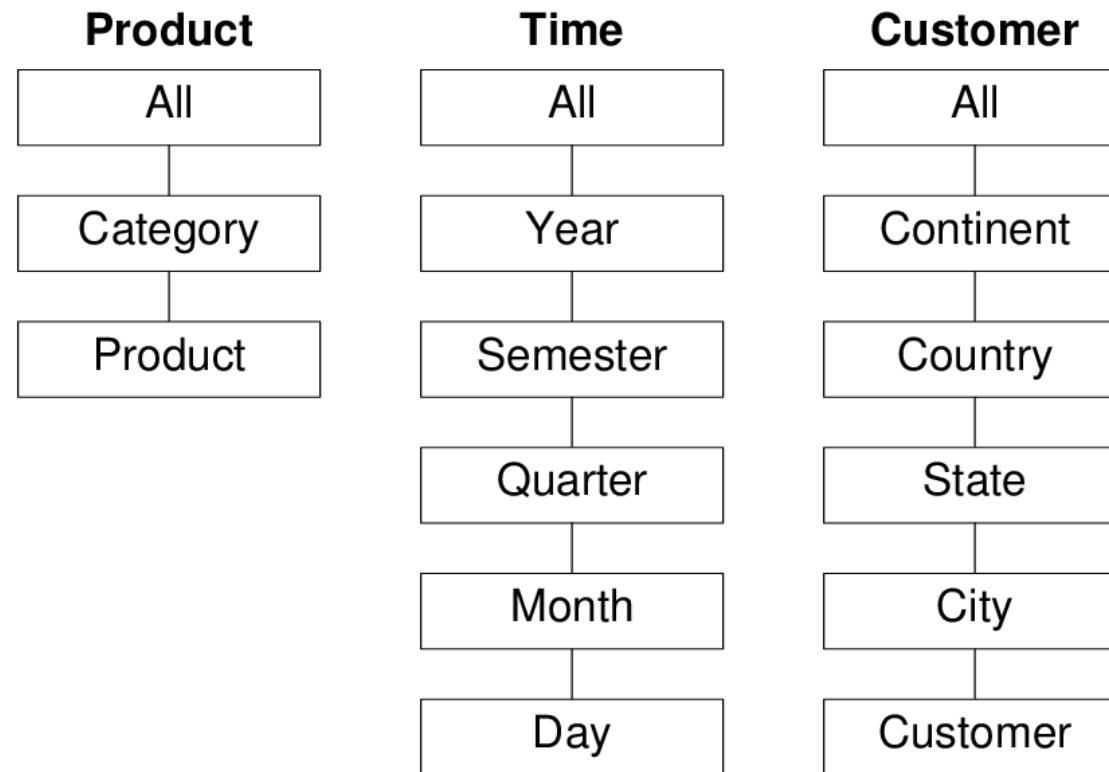
Multidimensional model

- Data can be viewed as a **cube**



Hierarchies

- Each **dimension** has multiple **levels (hierarchy)**



OLAP operations

- Drill-down to month

Customer (City)		Product (Category)				Time (Quarter)			
		Beverages	Condiments	Produce	Seafood	Q1	Q2	Q3	Q4
Köln	24	18	28	14	21	10	18	35	35
Berlin	33	25	23	25	27	14	11	30	30
Lyon	12	20	24	33	26	12	35	32	32
Paris	21	10	18	35	14	20	47	31	31
						10	33	12	18
						14	23	20	11
						12	20	33	18

OLAP operations

- Roll-up to country

Customer (City)	Köln				Berlin			
	Beverages	Condiments	Produce	Seafood	Beverages	Condiments	Produce	Seafood
Time (Quarter)	24	18	28	14	33	25	23	25
Q1	12	20	24	33	35	33	25	14
Q2	21	10	18	35	35	14	23	17
Q3	27	14	11	30	30	12	20	18
Q4	26	12	35	32	32	10	33	18
	14	20	47	31	31			

Customer (Country)	Germany				France			
	Beverages	Condiments	Produce	Seafood	Beverages	Condiments	Produce	Seafood
Time (Quarter)	57	43	51	39	33	30	42	68
Q1	33	30	42	68	68	41		
Q2	39	26	41	44	44	37		
Q3	30	22	46	44	44	51		
Q4	25	29	49	41	41			

OLAP operations

- Slice on city (Paris)

Customer (City)	Köln				Berlin			
	Beverages	Condiments	Produce	Seafood	Beverages	Condiments	Produce	Seafood
Q1	21	10	18	35	35	14	23	17
Q2	27	14	11	30	30	12	20	18
Q3	26	12	35	32	32	10	33	19
Q4	14	20	47	31	31	30	45	28

Time (Quarter)	Product (Category)			
	Beverages	Condiments	Produce	Seafood
Q1	21	10	18	35
Q2	27	14	11	30
Q3	26	12	35	32
Q4	14	20	47	31

OLAP operations

- Dice on city {Paris, Lyon} and quarter {Q1, Q2}

		Customer (City)							
		Köln	18	28	14				
		Berlin	33	25	23	25	14		
		Lyon	12	20	24	33	25		
		Paris	21	10	18	35	33	23	18
Time (Quarter)		Q1	21	10	18	35	35	14	23
		Q2	27	14	11	30	30	12	20
		Q3	26	12	35	32	32	10	33
		Q4	14	20	47	31	31		
					Produce	Seafood			
					Beverages	Condiments			
Product (Category)									

		Customer (City)					
		Lyon	20	24	33	33	
		Paris	12	18	35	35	14
		Q1	21	10	18	35	35
		Q2	27	14	11	30	30
			Produce	Seafood			
			Beverages	Condiments			
Product (Category)							

OLAP operations

- **Pivot (rotate) dimensions**

Customer (City)	Köln				Berlin				Lyon				Paris			
	Produce	Condiments	Beverages	Seafood												
Time (Quarter)	24	18	28	14	33	25	23	25	12	20	24	33	21	10	18	35
Q1	21	10	18	35	35	33	25	14	30	12	20	23	35	30	32	31
Q2	27	14	11	30	30	32	26	12	32	10	33	17	27	20	20	31
Q3	26	12	35	32	32	31	30	10	31	12	28	21	26	27	21	33
Q4	14	20	47	31	31	30	27	18	30	13	47	33	24	23	25	18

Customer (City)	Product (Category)				Customer (City)				Product (Category)				Customer (City)			
	Seafood	Condiments	Produce	Beverages	Seafood	Condiments	Produce	Beverages	Seafood	Condiments	Produce	Beverages	Seafood	Condiments	Produce	Beverages
Time (Quarter)	35	30	32	31	18	11	35	47	10	14	12	20	21	27	20	10
Paris	21	27	26	14	14	17	21	33	13	17	20	33	21	27	20	10
Lyon	12	14	11	13	12	14	11	13	13	17	20	18	13	17	20	18
Berlin	33	28	35	32	32	35	47	33	32	35	47	33	28	35	47	33
Köln	24	23	25	18	18	23	25	18	18	23	25	18	24	23	25	18

Other OLAP operations

- Sort by product category

Customer (City)	Köln				Berlin				Lyon				Paris			
	Produce	Condiments	Beverages	Seafood												
Time (Quarter)	Q1	Q2	Q3	Q4												
Köln	24	18	28	14	33	25	23	25	12	20	24	33	21	10	18	35
Berlin	33	25	23	14	21	18	10	18	27	14	11	30	26	12	18	35
Lyon	12	20	24	33	30	14	10	18	26	11	14	32	14	12	10	33
Paris	28	18	35	18	33	25	23	17	35	30	20	33	31	21	18	35
Q1	21	10	18	35	35	14	23	17	21	18	10	35	35	21	18	35
Q2	27	14	11	30	30	12	20	18	27	11	14	30	30	21	18	30
Q3	26	12	35	32	32	10	33	31	26	35	12	32	32	10	33	31
Q4	14	20	47	31	31	31	31	31	14	47	20	31	31	31	31	31

Customer (City)	Köln				Berlin				Lyon				Paris			
	Condiments	Seafood	Beverages	Produce												
Time (Quarter)	Q1	Q2	Q3	Q4												
Köln	24	28	18	14	33	23	25	25	12	24	20	33	21	18	10	35
Berlin	33	23	25	25	21	18	10	18	27	11	14	30	26	21	18	35
Lyon	12	20	24	33	30	12	20	18	26	11	14	32	32	21	18	33
Paris	28	18	35	18	33	25	23	17	35	30	20	33	31	21	18	35
Q1	21	18	10	35	35	14	23	17	21	18	10	35	35	21	18	35
Q2	27	11	14	30	30	12	20	18	27	11	14	30	30	21	18	30
Q3	26	35	12	32	32	10	33	31	26	35	12	32	32	21	18	33
Q4	14	47	20	31	31	31	31	31	14	47	20	31	31	31	31	31

Other OLAP operations

- Cube for 2012 vs. cube for 2011 (previous year)

Customer (City)	Köln				Berlin				Lyon				Paris			
	Produce	Beverages	Condiments	Seafood												
Time (Quarter)	Q1	Q2	Q3	Q4												
Köln	24	18	28	14	33	25	23	25	12	20	24	33	21	10	18	35
Berlin	14	16	26	16	30	22	21	26	14	18	22	28	19	12	31	28
Lyon	28	21	16	18	35	30	28	26	28	26	20	19	28	16	14	21
Paris	18	17	19	16	33	35	33	31	10	12	20	23	23	21	20	19
Product (Category)	Produce	Beverages	Condiments	Seafood												

Customer (City)	Köln				Berlin				Lyon				Paris			
	Produce	Beverages	Condiments	Seafood												
Time (Quarter)	Q1	Q2	Q3	Q4												
Köln	20	22	24	16	30	22	21	26	14	18	22	28	19	12	31	28
Berlin	16	16	26	16	14	18	22	28	19	12	31	28	28	26	21	19
Lyon	26	21	19	18	14	18	22	28	28	12	31	28	29	14	20	19
Paris	18	19	28	28	19	12	31	28	28	28	28	28	29	12	31	29
Product (Category)	Produce	Beverages	Condiments	Seafood												

Other OLAP operations

- Drill-across 2011 and 2012

Customer (City)	Köln			
	Berlin	14	22	24
Beverages	30	22	21	26
Condiments	19	12	31	28
Product (Category)	28	26	21	16
Time (Quarter)	19	12	31	28
Q1	30	12	10	29
Q2	28	11	31	28
Q3	12	22	45	29
Q4	20	22	24	16
	Produce	Seafood		

Customer (City)	Köln			
	Berlin	12	20	24
Beverages	33	25	23	25
Condiments	12	20	24	33
Product (Category)	21	10	18	35
Time (Quarter)	21	10	18	35
Q1	27	14	11	30
Q2	26	12	35	32
Q3	14	20	47	31
Q4	24	18	28	14
	Produce	Seafood		

Customer (City)	Köln			
	Berlin	14	22	24
Beverages	30	22	21	26
Condiments	12	20	24	33
Product (Category)	19	10	18	35
Time (Quarter)	21	10	18	35
Q1	30	12	22	28
Q2	27	14	11	30
Q3	28	11	31	28
Q4	12	22	45	29
	Produce	Seafood		

Other OLAP operations

- Percentage change between 2011 and 2012

Customer (City)	Köln				
	Berlin	Lyon	Paris		
Time (Quarter)	Q1	Q2	Q3	Q4	
Produce	20 30 14 19	22 22 18 12	24 28 22 31	16 16 26 28	16 18 18 19
Seafood					
Beverages	28 30 28 12	11 12 10 22	31 29 28 45	28 29 12 29	28 16 20 19
Condiments					
Product (Category)					

Customer (City)	Köln				
	Berlin	Lyon	Paris		
Time (Quarter)	Q1	Q2	Q3	Q4	
Produce	24 33 12 21	18 25 20 10	28 23 24 18	14 25 33 35	14 14 18 17
Seafood					
Beverages	21 27 26 14	10 14 12 20	18 11 35 47	35 30 32 31	35 23 33 31
Condiments					
Product (Category)					

Customer (City)	Köln				
	Berlin	Lyon	Paris		
Time (Quarter)	Q1	Q2	Q3	Q4	
Produce	20 10 -14 11	-18 14 11 9	17 10 18 25	-13 -4 4 18	-13 -4 0 0
Seafood					
Beverages	11 -10 -7 17	-17 17 9 -9	-42 10 13 4	25 3 25 7	10 -13 6 -11
Condiments					
Product (Category)					

Data warehousing

- What is a data warehouse?
 - a data warehouse is a relational database that stores historical data in a convenient schema for multidimensional analysis using OLAP operations

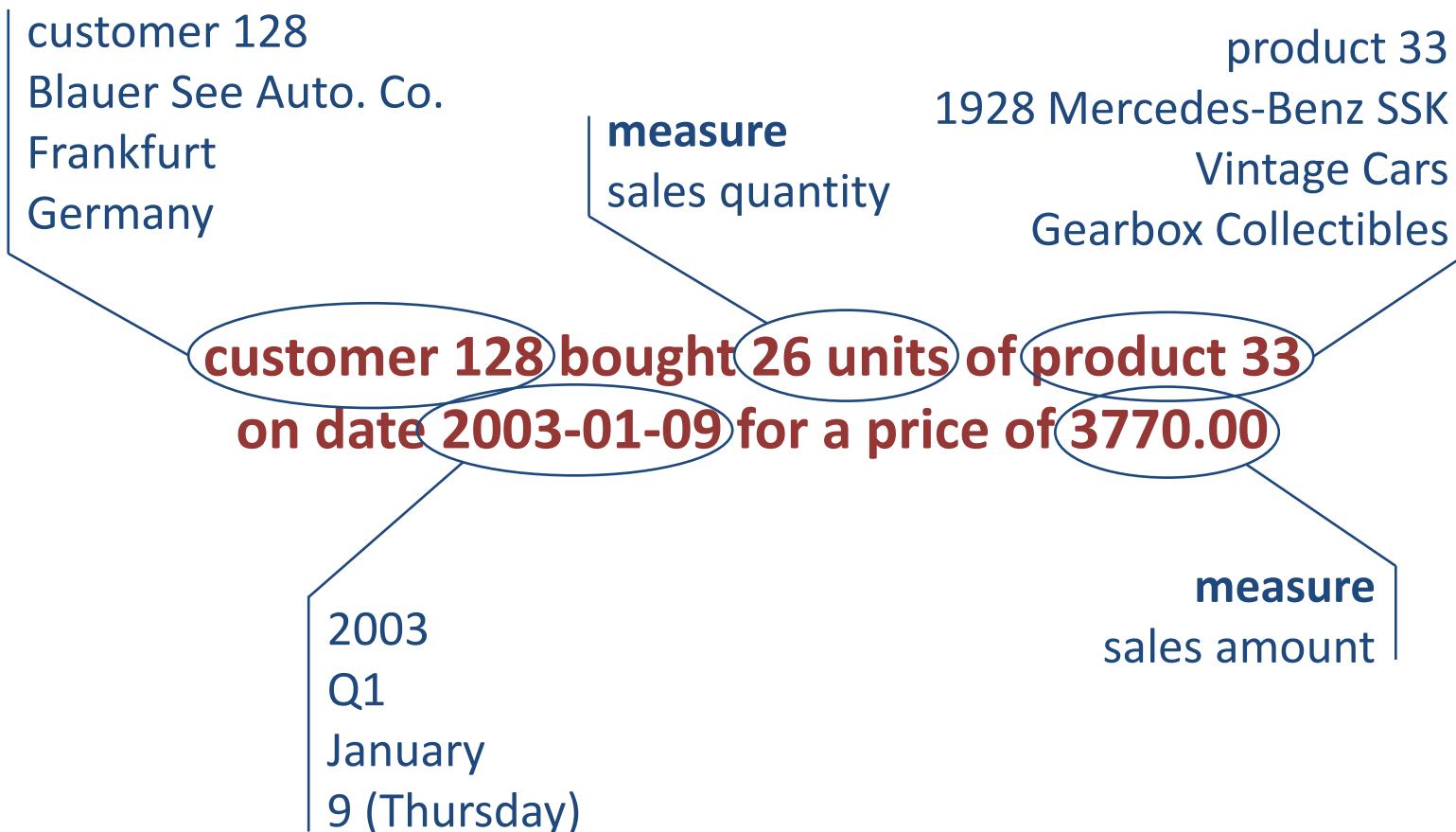
Data warehousing

- A data warehouse stores **facts**

**customer 128 bought 26 units of product 33
on date 2003-01-09 for a price of 3770.00**

Data warehousing

- Facts have associated **measures**



Data warehousing

- **Facts** define the possible analysis dimensions
 - customer c bought product p (2D)
 - customer c bought product p on date d (3D)
 - customer c bought product p on date d in store s (4D)
 - customer c bought product p on date d in store s using payment method m (5D)
 - m may be 'cash', 'credit card', etc.
 - customer c bought product p on date d in store s using payment method m through sales representative e (6D)
 - e is an employee number
 - etc.

Data warehousing

- **Measures** define the quantity being analyzed
 - sales
 - quantity sold
 - sales amount
 - production
 - quantity produced
 - production cost
 - logistics
 - distance traveled
 - transported weight
 - etc.

Data warehousing

- **Facts can be grouped by dimensions**
 - examples
 - by customer (1D)
 - by product (1D)
 - by time (1D)
 - by customer and product (2D)
 - by customer and time (2D)
 - by product and time (2D)
 - by customer, product and time (3D)

Data warehousing

- As **facts** are grouped, their **measures** are aggregated
 - examples
 - sales amount by customer (1D)
 - sales amount by product (1D)
 - sales amount by time (1D)
 - sales amount by customer and product (2D)
 - sales amount by customer and time (2D)
 - sales amount by product and time (2D)
 - sales amount by customer, product and time (3D)

Data warehousing

- **Measures are aggregated at a certain level of detail**
 - examples
 - sales amount by customer country
 - sales amount by customer city
 - sales amount by customer country and product line
 - sales amount by customer city and product vendor
 - sales amount by customer country, product line and year
 - sales amount by customer city, product vendor and month

Data warehousing

- **Levels** are organized into **hierarchies**
 - examples
 - country, state, city
 - year, month, day
- A **dimension** can have one or more **hierarchies**
 - example: time dimension
 - year, month, day (for calendar year, from Jan to Dec)
 - year, semester, period (for school year, from Sep to Aug)