

***Unit 11***  
***Online Analytical Processing (OLAP)***  
***Basic Concepts***

# ***OLAP vs. OLTP***

- ◆ We have focused until now on ***OLTP***: Online Transaction Processing
- ◆ This dealt with storing data both logically and physically and managing transactions querying and modifying the data
- ◆ We will now focus on providing support for analytical queries, essentially statistical and summary information for decision-makers, that is on ***OLAP***: Online Analytical Processing
- ◆ This may be accomplished by preprocessing, for efficiency purposes, and producing special types of views, which are also not necessarily up to date
  - Not up to date may not be a problem in OLAP
- ◆ Data for OLAP (and more generally for data mining) is frequently stored in a ***Data Warehouse***

## ***Example***

- ◆ Our company has several stores and sells several products
- ◆ The stores are in different locations
- ◆ The locations, identified by (city, state) pairs are grouped into several regions
- ◆ We divide the time of sale into four quarters
- ◆ The quarters are grouped into two half-years

# ***Our Company***

Store	<u>Store#</u>	City	State	Region
	Alpha	New York	NY	NE
	Beta	Albany	NY	NE

Quarter	<u>Quarter#</u>	Half_Year
	1	First
	2	First
	3	Second
	4	Second

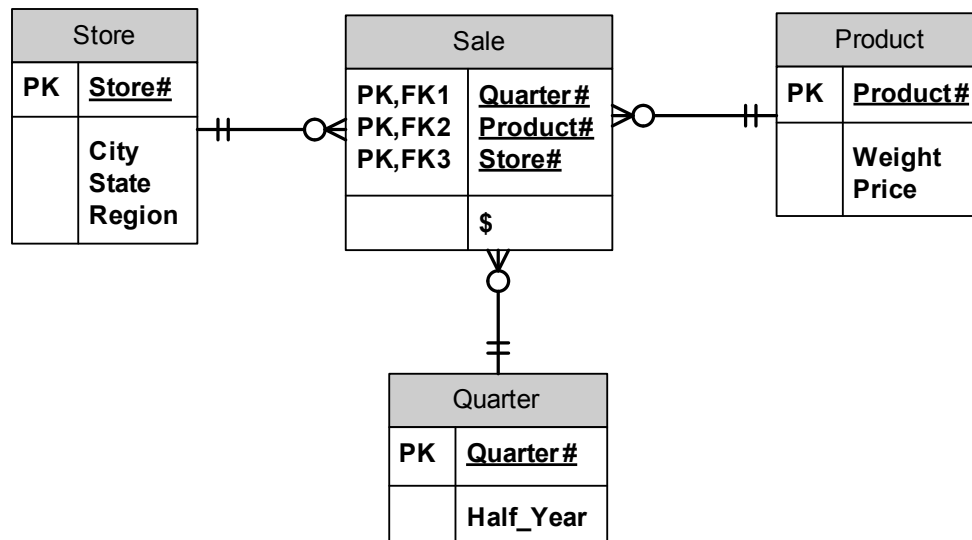
Product	<u>Product#</u>	Weight	Price
	Book	4	100
	Glass	15	200

# Our Sales

Sale	<u>Store#</u>	<u>Product#</u>	<u>Quarter#</u>	\$
	Alpha	Book	1	70,000
	Alpha	Glass	1	90,000
	Beta	Book	1	90,000
	Beta	Glass	1	80,000
	Alpha	Book	2	90,000
	Alpha	Glass	2	90,000
	Beta	Book	2	60,000
	Beta	Glass	2	50,000
	Alpha	Book	3	60,000
	Alpha	Glass	3	80,000
	Beta	Book	3	50,000
	Beta	Glass	3	90,000
	Alpha	Book	4	50,000
	Alpha	Glass	4	50,000
	Beta	Book	4	70,000
	Beta	Glass	4	70,000

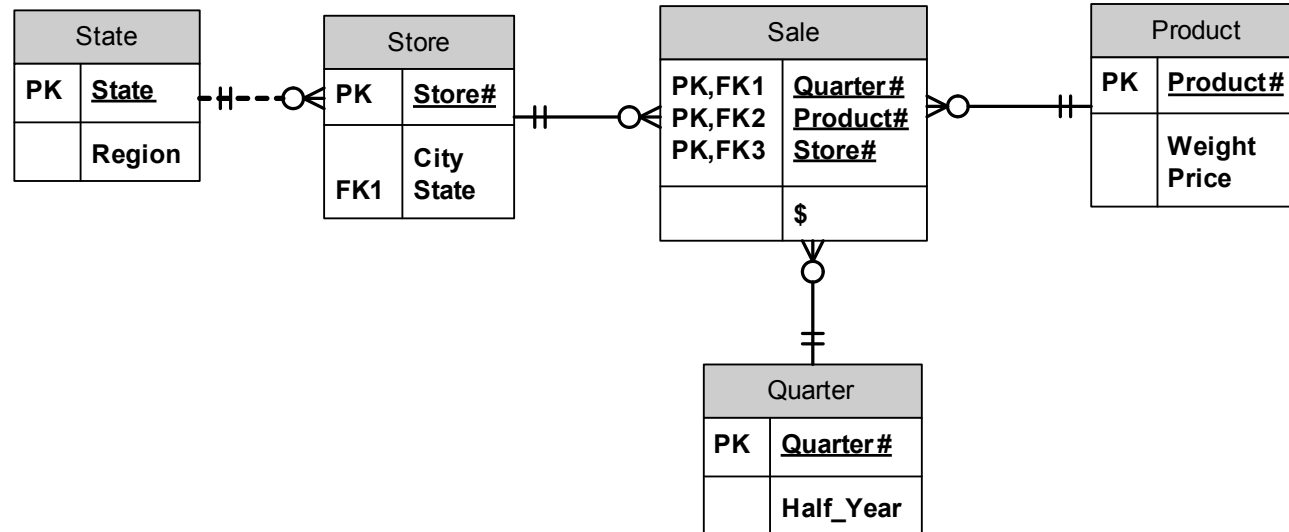
# Star Schema

- ◆ We want to support queries useful for statistical analysis by computing various sums, averages, etc.
- ◆ The structure we have is a **star schema**
- ◆ In the middle we have our **facts table**



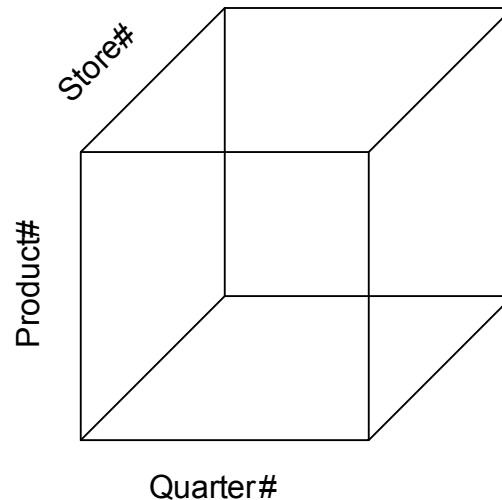
# Snowflake Schema: Normalized Star Schema

- ◆ One could also normalize, as table Store is not normalized, since State → Region
- ◆ Then, one could get, which we will not consider further, a **snowflake schema**



# Cube

- ◆ We could think of each row of fact table as occupying a voxel (volume element) in a **cube**



- ◆ Cubes, in general, can have any number of dimensions; in our example there are three
- ◆ This cube can then be **sliced and diced**

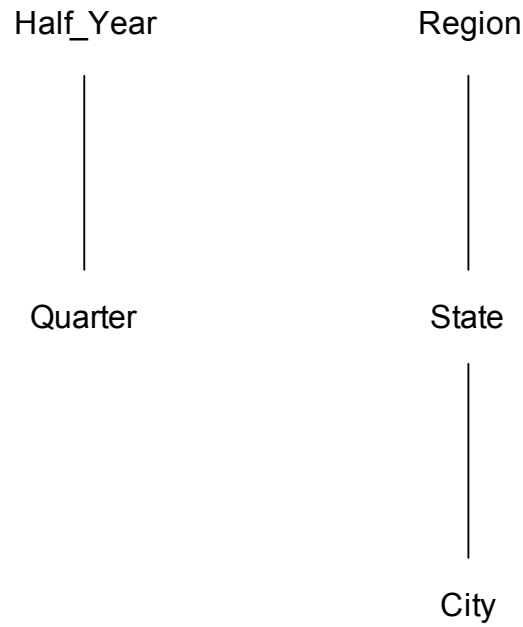


# ***Slice***

- ◆ SELECT Store#, Product#, SUM(\$)  
FROM Sale  
GROUP BY Store#, Product#
- ◆ We can do all kinds of such slices

# ***Dimension Hierarchies***

- ◆ Dimensions could have hierarchies (or more generally even lattices)
- ◆ We have two very simple hierarchies
  - One temporal: quarters are in half years
  - One geographical: cities are in states are in regions



## ***Using Hierarchies***

- ◆ `SELECT Sale.Product#, Quarter.Half_Year, SUM($)  
FROM Sale, Quarter  
WHERE Sale.Quarter# = Quarter.Quarter#  
GROUP BY Half_Year;`
- ◆ Will produce summaries by half years, not quarters

## ***New Operator: CUBE***

- ◆ `SELECT Store#, Product#, SUM($)`  
`FROM Sale`  
`GROUP BY CUBE (Store#, Product#);`
- ◆ Will produce all possible aggregations based on subsets of {Store#, Product#}, best explained by looking at what will come out

	Store#	Product#	\$
	Alpha	Book	270,000
	Alpha	Glass	310,000
	Beta	Book	270,000
	Beta	Glass	290,000
	Alpha	NULL	580,000
	Beta	NULL	560,000
	NULL	Book	540,000
	NULL	Glass	600,000
	NULL	NULL	1,140,000

## ***New Operator: ROLLUP***

- ◆ ROLLUP produces only some of the aggregate operators produced by CUBE, best explained by example
- ◆ `SELECT Store#, Product#, SUM($)  
FROM Sale  
GROUP BY ROLLUP (Store#,Product#);`

	Store#	Product#	\$
	Alpha	Book	270,000
	Alpha	Glass	310,000
	Beta	Book	270,000
	Beta	Glass	290,000
	Alpha	NULL	580,000
	Beta	NULL	560,000
	NULL	NULL	1,140,000

# ***Key Ideas***

- ◆ OLAP vs. OLTP
- ◆ Star schema
- ◆ Snowflake schema
- ◆ Cube
- ◆ Slicing and dicing
- ◆ Dimension hierarchies
- ◆ ROLAP
- ◆ MOLAP