



# Calculating Values in a Query



### Math Operators

- addition +
- subtraction -
- multiplication \*
- division /
- modulus %
- Work differently on different data types





#### Calculating Difference

```
In [1]: stmt = select([census.columns.age,
            (census.columns.pop2008-
            census.columns.pop2000).label('pop_change')
   ...: ])
In [2]: stmt = stmt.group_by(census.columns.age)
In [3]: stmt = stmt.order_by(desc('pop_change'))
In [4]: stmt = stmt.limit(5)
In [5]: results = connection.execute(stmt).fetchall()
In [6]: print(results)
Out[6]: [(61, 52672), (85, 51901), (54, 50808), (58, 45575), (60,
44915)]
```



#### Case Statement

- Used to treat data differently based on a condition
- Accepts a list of conditions to match and a column to return if the condition matches
- The list of conditions ends with an else clause to determine what to do when a record doesn't match any prior conditions



#### Case Example

```
In [1]: from sqlalchemy import case
In [2]: stmt = select([
           func.sum(
               case([
                   (census.columns.state == 'New York',
                   census.columns.pop2008)
       ], else_=0))])
In [3]: results = connection.execute(stmt).fetchall()
In [4]: print(results)
Out[4]:[(19465159,)]
```



#### Cast Statement

- Converts data to another type
- Useful for converting
  - integers to floats for division
  - strings to dates and times
- Accepts a column or expression and the target Type



#### Percentage Example

```
In [1]: from sqlalchemy import case, cast, Float
In [2]: stmt = select([
            (func.sum(
               case([
                    (census.columns.state == 'New York',
                    census.columns.pop2008)
       ], else_=0)) /
   ...: cast(func.sum(census.columns.pop2008),
                 Float) * 100).label('ny_percent')])
   • • • •
In [3]: results = connection.execute(stmt).fetchall()
In [4]: print(results)
Out[4]: [(Decimal('6.4267619765'),)]
```





# Let's practice!





# SQL Relationships



#### Relationships

- Allow us to avoid duplicate data
- Make it easy to change things in one place
- Useful to break out information from a table we don't need very often





# Relationships

Census

state	sex	age	pop2000	pop2008
New York	F	O	120355	122194
New York	F	1	118219	119661
New York	F	2	119577	116413

State\_Fact

name	abbreviation	type
New York	NY	state
Washington DC	DC	capitol
Washington	WA	state





#### Automatic Joins

```
In [1]: stmt = select([census.columns.pop2008,
   ...: state_fact.columns.abbreviation])
In [2]: results = connection.execute(stmt).fetchall()
In [3]: print(results)
Out[3]: [(95012, u'IL'),
 (95012, u'NJ'),
 (95012, u'ND'),
 (95012, u'OR'),
 (95012, u'DC'),
 (95012, u'WI'),
```



#### Join

- Accepts a Table and an optional expression that explains how the two tables are related
- The expression is not needed if the relationship is predefined and available via reflection
- Comes immediately after the select() clause and prior to any where (), order\_by or group\_by() clauses



#### Select\_from

- Used to replace the default, derived FROM clause with a join
- Wraps the join() clause





#### Select\_from Example



#### Joining Tables without Predefined Relationship

- Join accepts a Table and an optional expression that explains how the two tables are related
- Will only join on data that match between the two columns
- Avoid joining on columns of different types





#### Select\_from Example

```
In [1]: stmt = select([func.sum(census.columns.pop2000)])
In [2]: stmt = stmt.select_from(
   ...: census.join(state_fact, census.columns.state
   ...: == state_fact.columns.name))
In [3]: stmt = stmt.where(
   ...: state_fact.columns.census_division_name ==
   'East South Central')
In [4]: result = connection.execute(stmt).scalar()
In [5]: print(result)
Out[5]: 16982311
```





# Let's practice!





# Working with Hierarchical Tables



#### Hierarchical Tables

- Contain a relationship with themselves
- Commonly found in:
  - Organizational
  - Geographic
  - Network
  - Graph





### Hierarchical Tables - Example

Employees

id	name	job	manager
1	Johnson	Admin	6
2	Harding	Manager	9
3	Taft	Sales I	2
4	Hoover	Sales I	2



#### Hierarchical Tables - alias()

- Requires a way to view the table via multiple names
- Creates a unique reference that we can use



#### Querying Hierarchical Data

```
In [1]: managers = employees.alias()
In [2]: stmt = select(
           [managers.columns.name.label('manager'),
   employees.columns.name.label('employee')])
In [3]: stmt = stmt.select_from(employees.join()
   ...: managers, managers.columns.id ==
   ...: employees.columns.manager)
In [4]: stmt = stmt.order_by(managers.columns.name)
In [5]: print(connection.execute(stmt).fetchall())
Out[5]: [(u'FILLMORE', u'GRANT'),
 (u'FILLMORE', u'ADAMS'),
 (u'HARDING', u'TAFT'), ...
```



#### Group\_by and Func

- It's important to target group\_by() at the right alias
- Be careful with what you perform functions on
- If you don't find yourself using both the alias and the table name for a query, don't create the alias at all



#### Querying Hierarchical Data

```
In [1]: managers = employees.alias()
In [2]: stmt = select([managers.columns.name,
        func.sum(employees.columns.sal)])
In [3]: stmt = stmt.select_from(employees.join(
   ...: managers, managers.columns.id ==
   ...: employees.columns.manager)
In [4]: stmt = stmt.group_by(managers.columns.name)
In [5]: print(connection.execute(stmt).fetchall())
Out[5]: [(u'FILLMORE', Decimal('96000.00')),
 (u'GARFIELD', Decimal('83500.00')),
 (u'HARDING', Decimal('52000.00')),
 (u'JACKSON', Decimal('197000.00'))]
```





# Let's practice!





# Handling Large ResultSets



#### Dealing with Large ResultSets

- fetchmany() lets us specify how many rows we want to act upon
- We can loop over fetchmany()
- It returns an empty list when there are no more records
- We have to close the ResultProxy afterwards





#### Fetching Many Rows





# Let's practice!