

HW 1 Solutions

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```
library(sqldf)

## Loading required package: gsubfn
## Loading required package: proto
## Loading required package: RSQLite
library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.2 --
## v ggplot2 3.4.0      v purrr  1.0.0
## v tibble  3.1.8      v dplyr  1.0.10
## v tidyr   1.2.1      v stringr 1.5.0
## v readr   2.1.3      v forcats 0.5.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()

order_details <- read.csv("order_details.csv")
orders <- read.csv("orders.csv")
territories <- read.csv("territories.csv")
regions <- read.csv("regions.csv")
employee_territories <- read.csv("employee_territories.csv")
employees <- read.csv("employees.csv")
customers <- read.csv("customers.csv")
shippers <- read.csv("shippers.csv")
suppliers <- read.csv("suppliers.csv")
products <- read.csv("products.csv")
categories <- read.csv("categories.csv")

#1 Perform a sort of orders by employeeID, then by shipVia, and then by freight, # for those orders by
shipped to France. Order_France <- sqldf("SELECT * FROM orders WHERE shipCountry = 'France'
ORDER BY shipVia, freight")

#1 Perform a sort of orders by employeeID, then by shipVia, and then by freight,
# for those orders by shipped to France.
Order_France <- sqldf("SELECT * FROM orders WHERE shipCountry = 'France' ORDER BY shipVia, freight")
glimpse(Order_France)

## Rows: 77
## Columns: 14
## $ orderID      <int> 10371, 10631, 10738, 10683, 10274, 10826, 10559, 10331,~
## $ customerID   <chr> "LAMAI", "LAMAI", "SPEC", "DUMON", "VINET", "BLONP", "~
## $ employeeID   <int> 1, 8, 2, 2, 6, 6, 6, 9, 8, 3, 5, 4, 5, 8, 1, 7, 3, 6, 1~
```

```
## $ orderDate      <chr> "1996-12-03 00:00:00.000", "1997-08-14 00:00:00.000", "~
## $ requiredDate   <chr> "1996-12-31 00:00:00.000", "1997-09-11 00:00:00.000", "~
## $ shippedDate     <chr> "1996-12-24 00:00:00.000", "1997-08-15 00:00:00.000", "~
## $ shipVia        <int> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1~
## $ freight        <dbl> 0.45, 0.87, 2.91, 4.40, 6.01, 7.09, 8.05, 10.19, 11.26,~
## $ shipName       <chr> "La maison d'Asie", "La maison d'Asie", "Spécialités du~
## $ shipAddress     <chr> "1 rue Alsace-Lorraine", "1 rue Alsace-Lorraine", "25 r~
## $ shipCity       <chr> "Toulouse", "Toulouse", "Paris", "Nantes", "Reims", "St~
## $ shipRegion     <chr> "NULL", "NULL", "NULL", "NULL", "NULL", "NULL", "NULL",~
## $ shipPostalCode <chr> "31000", "31000", "75016", "44000", "51100", "67000", "~
## $ shipCountry    <chr> "France", "France", "France", "France", "France", "Fran~
```

#2 Which shipVia has the largest average cost?

```
order_cost <-sqldf("SELECT ShipVia, AVG(freight) FROM orders GROUP BY ShipVia")
glimpse(order_cost)
```

```
## Rows: 3
## Columns: 2
## $ shipVia      <int> 1, 2, 3
## $ `AVG(freight)` <dbl> 65.00133, 86.64064, 80.44122
```

*# shipVia 2 has the largest cost. This corresponds to shipper United Package
(since the field had a different name I will accept either).*

#3 Which product category has the highest average UnitPrice? The Lowest?

```
average_cost <-sqldf("SELECT CategoryID, AVG(UnitPrice) FROM products GROUP BY CategoryID")
category_costs <- sqldf("Select average_cost.*, categories.categoryName FROM average_cost INNER JOIN ca
category_costs
```

```
##   CategoryID AVG(UnitPrice)  categoryName
## 1          1      37.97917    Beverages
## 2          2      23.06250    Condiments
## 3          3      25.16000    Confections
## 4          4      28.73000 Dairy Products
## 5          5      20.25000 Grains/Cereals
## 6          6      54.00667  Meat/Poultry
## 7          7      32.37000    Produce
## 8          8      20.68250    Seafood
```

Highest category cost is Meat/Poultry, Lowest is Grain/Cereals

#4 Which products are supplied by a company in the United States?

```
USproducts <- sqldf("SELECT products.ProductName FROM products INNER JOIN suppliers WHERE products.Supp
USproducts
```

```
##               ProductName
## 1  Chef Anton's Cajun Seasoning
## 2      Chef Anton's Gumbo Mix
## 3 Louisiana Fiery Hot Pepper Sauce
## 4      Louisiana Hot Spiced Okra
## 5  Grandma's Boysenberry Spread
## 6  Northwoods Cranberry Sauce
## 7  Uncle Bob's Organic Dried Pears
## 8      Laughing Lumberjack Lager
## 9                Sasquatch Ale
## 10               Steeleye Stout
## 11             Boston Crab Meat
```

```
## 12 Jack's New England Clam Chowder
```

```
#5 Which shipper is shipping the largest number of units of product?
```

```
# Answer in terms of units; you do not need to consider quantityPerUnit here.
```

```
full_orders <- sqldf("SELECT orders.*, order_details.productID, order_details.unitPrice, order_details.quantity FROM orders INNER JOIN order_details ON orders.orderID = order_details.orderID")
total_shipQ <- sqldf("SELECT SUM(full_orders.quantity), shippers.companyName FROM full_orders INNER JOIN shippers ON full_orders.shipperID = shippers.shipperID")
total_shipQ
```

```
##      SUM(full_orders.quantity)      companyName
## 1              15919      Speedy Express
## 2              19945      United Package
## 3              15453 Federal Shipping
```

```
# United Package, which is shipping 19,945 units
```

```
#6 Which employee is tied to the most sales revenue?
```

```
# Give the name, not the code, along with the total revenue for the employee.
```

```
order_revenue <- sqldf("SELECT *, unitPrice*quantity*(1-discount) as revenue from order_details")
employee_revenue <- sqldf("select orders.employeeID, SUM(order_revenue.revenue) FROM orders INNER JOIN order_revenue ON orders.orderID = order_revenue.orderID")
revenue_by_name <- sqldf("SELECT employee_revenue.*, employees.lastName, employees.firstName FROM employee_revenue INNER JOIN employees ON employee_revenue.employeeID = employees.employeeID")
revenue_by_name
```

```
##      employeeID SUM(order_revenue.revenue)  lastName  firstName
## 1              1          192107.60    Davolio    Nancy
## 2              2          166537.76      Fuller    Andrew
## 3              3          202812.84 Leverling    Janet
## 4              4          232890.85    Peacock    Margaret
## 5              5           68792.28 Buchanan    Steven
## 6              6           73913.13     Suyama    Michael
## 7              7          124568.24       King    Robert
## 8              8          126862.28 Callahan    Laura
## 9              9           77308.07 Dodsworth     Anne
```

```
# Margeret Peacock has the most revenue
```

```
#7 Find the total revenue for each product category.
```

```
product_category <- sqldf("SELECT categories.categoryName, products.ProductID FROM categories INNER JOIN products ON categories.categoryID = products.categoryID")
Cat_revenue <- sqldf("SELECT product_category.categoryName, SUM(order_revenue.revenue) FROM product_category INNER JOIN order_revenue ON product_category.ProductID = order_revenue.ProductID")
Cat_revenue
```

```
##      categoryName SUM(order_revenue.revenue)
## 1      Beverages          267868.18
## 2    Condiments          106047.09
## 3    Confections          167357.22
## 4 Dairy Products          234507.29
## 5 Grains/Cereals           95744.59
## 6  Meat/Poultry          163022.36
## 7      Produce           99984.58
## 8      Seafood          131261.74
```

```
#8 Consider the amount of revenue for each customer. If there were no discounts applied, which customer would see the largest increase in cost?
```

```
order_disc <- sqldf("SELECT unitPrice*quantity*discount as revenue_disc, orders.customerID FROM order_details INNER JOIN orders ON order_details.orderID = orders.orderID")
total_disc <- sqldf("SELECT customers.companyName, SUM(order_disc.revenue_disc) FROM customers INNER JOIN order_disc ON customers.customerID = order_disc.customerID")
# Answer is Save-a-lot Markets
```

```
#9 Which order(s) has the most number of items (and how many)? Give the orderID for this one.
```

```
item_count = sqldf("SELECT orderID, COUNT(productID) FROM order_details GROUP BY orderID ORDER BY count DESC")
```

```
glimpse(item_count)
```

```
## Rows: 830
## Columns: 2
## $ orderID      <int> 11077, 10979, 10847, 10657, 11064, 11031, 11021, 10~
## $ `COUNT(productID)` <int> 25, 6, 6, 6, 5, 5, 5, 5, 5, 5, 5, 5, 5, 5, 5,~
# OrderID 11077 has 25 items
```

10 Create a new field called “InventoryOrderRatio” which is, for each product, the UnitsInStock (the inventory) for the product (across all customers) divided by the quantity ordered for that product.

A high value represents sufficient product in stock, while a low number represents products that are in

danger of running out. What 3 products are most in danger of running out?

```
inventory_data <- sqldf("SELECT products.ProductName, products.UnitsInStock, SUM(order_details.quantity)
                        FROM products
                        INNER JOIN order_details
                        ON products.productID = order_details.productID
                        GROUP BY order_details.productID")

inventory_data$quant <- as.numeric(inventory_data$total_q)

inventory_ratio <- sqldf("SELECT ProductName, UnitsInStock/quant AS InventoryOrderRatio
                        FROM inventory_data
                        WHERE UnitsInStock > 0 AND total_q > 0
                        ORDER BY UnitsInStock/quant")

inventory_ratio
```

```
##           ProductName InventoryOrderRatio
## 1      Sir Rodney's Scones      0.002952756
## 2    Scottish Longbreads      0.007509387
## 3          Rogede sild      0.009842520
## 4    Camembert Pierrot      0.012048193
## 5        Longlife Tofu      0.013468013
## 6        Tarte au sucre      0.015697138
## 7              Chang      0.016083254
## 8 Northwoods Cranberry Sauce      0.016129032
## 9      Nord-Ost Matjeshering      0.016339869
## 10   Gnocchi di nonna Alice      0.016627078
## 11 Louisiana Hot Spiced Okra      0.016736402
## 12   Mozzarella di Giovanni      0.017369727
## 13   Guaraná Fantástica      0.017777778
## 14      Outback Lager      0.018359853
```

## 15	Maxilaku	0.019230769
## 16	Uncle Bob's Organic Dried Pears	0.019659240
## 17	Gumbär Gummibärchen	0.019920319
## 18	Manjimup Dried Apples	0.022573363
## 19	Steeleye Stout	0.022650057
## 20	Flotemysost	0.024597919
## 21	Pavlova	0.025043178
## 22	Konbu	0.026936027
## 23	Côte de Blaye	0.027287319
## 24	Tourtière	0.027814570
## 25	Ipoh Coffee	0.029310345
## 26	Wimmers gute Semmelknödel	0.029729730
## 27	Mascarpone Fabioli	0.030303030
## 28	Queso Cabrales	0.031161473
## 29	Teatime Chocolate Biscuits	0.034578147
## 30	Gudbrandsdalsost	0.036414566
## 31	Singaporean Hokkien Fried Mee	0.037302726
## 32	Aniseed Syrup	0.039634146
## 33	Original Frankfurter grüne Soße	0.040455120
## 34	Rössle Sauerkraut	0.040625000
## 35	Ikura	0.041778976
## 36	Gula Malacca	0.044925125
## 37	Chai	0.047101449
## 38	Raclette Courdavault	0.052807487
## 39	Vegie-spread	0.053932584
## 40	Lakkalikööri	0.058103976
## 41	Zaanse koeken	0.074226804
## 42	Filo Mix	0.076000000
## 43	Carnarvon Tigers	0.077922078
## 44	Ravioli Angelo	0.082949309
## 45	Tofu	0.086633663
## 46	Jack's New England Clam Chowder	0.086646279
## 47	Chartreuse verte	0.087011349
## 48	Gravad lax	0.088000000
## 49	Louisiana Fiery Hot Pepper Sauce	0.102013423
## 50	Tunnbröd	0.105172414
## 51	Rhönbräu Klosterbier	0.108225108
## 52	Chocolade	0.108695652
## 53	Boston Crab Meat	0.111514053
## 54	Escargots de Bourgogne	0.116104869
## 55	Chef Anton's Cajun Seasoning	0.116997792
## 56	Pâté chinois	0.127353267
## 57	Sir Rodney's Marmalade	0.127795527
## 58	Schoggi Schokolade	0.134246575
## 59	Inlagd Sill	0.139130435
## 60	Geitost	0.148344371
## 61	Spegesild	0.173357664
## 62	Sirop d'érable	0.187396352
## 63	Sasquatch Ale	0.219367589
## 64	NuNuCa Nuß-Nougat-Creme	0.238993711
## 65	Queso Manchego La Pastora	0.250000000
## 66	Valkoinen suklaa	0.276595745
## 67	Laughing Lumberjack Lager	0.282608696
## 68	Gustaf's Knäckebröd	0.298850575

```
## 69          Mishi Kobe Niku          0.305263158
## 70          Genen Shouyu             0.319672131
## 71          Röd Kaviar               0.344709898
## 72  Grandma's Boysenberry Spread    0.398671096
```

*# 3 smallest inventory_ratios are for Sir Rodney's Scones, Scottish Longbreads, and Rogede sild
ok if ones where inventory are 0 are included--these will be a different list.*

*# 11 A recommender engine looks at which pairs of products tend to be bought by the same # customer, so
(Hint: this will require a creative join with the orders data)*

```
full_orders <- sqldf("SELECT orders.customerID, order_details.productID
                      FROM orders
                      INNER JOIN order_details
                      ON orders.orderID = order_details.orderID")

fo2 <- sqldf("SELECT customerID, productID AS prod1 FROM full_orders")
order_combos <- sqldf("SELECT fo2.customerID, fo2.prod1, full_orders.productID AS prod2
                      FROM fo2
                      INNER JOIN full_orders
                      ON fo2.customerID = full_orders.customerID
                      WHERE fo2.prod1 > full_orders.productID")

combo_count <- sqldf("SELECT prod1, prod2, COUNT(customerID) FROM order_combos
                     GROUP BY prod1, prod2
                     ORDER BY COUNT(customerID) DESC")
glimpse(combo_count)
```

```
## Rows: 2,893
## Columns: 3
## $ prod1      <int> 56, 60, 31, 75, 41, 41, 71, 75, 62, 31, 31, 59, 59~
## $ prod2      <int> 31, 2, 2, 2, 2, 31, 31, 41, 56, 17, 24, 2, 24, 31,~
## $ `COUNT(customerID)` <int> 62, 61, 59, 58, 57, 57, 57, 57, 56, 55, 54, 54, 54~
```

*# Most common order pair is product ID 31 and 56, corresponding to Gorgonzola Telino and Gnocchi di non
Gnocchi with a Gorgonzola sauce is amazing.*