Module13

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Exercise 1 - Table for gender and year combinations

```
#a. Load the data.
Survey <- read.csv('http://www.lock5stat.com/datasets/StudentSurvey.csv', na.strings=c('','''))
#b. Cleanup and order by the Year variable. We need the year and gender columns.
Survey = Survey %>%
  select(Gender, Year) %>%
  drop_na()
Survey$Year <- factor(Survey$Year, levels = c('FirstYear', 'Sophomore', 'Junior', 'Senior'))</pre>
classCounts <- Survey[order(Survey$Year), ]</pre>
classCounts = classCounts %>%
  count(Year, Gender)
classCounts
          Year Gender n
                  F 43
## 1 FirstYear
## 2 FirstYear
                    M 51
## 3 Sophomore
                   F 96
## 4 Sophomore
                   M 99
                    F 18
## 5
        Junior
## 6
        Junior
                    M 17
## 7
       Senior
                    F 10
## 8
       Senior
                    M 26
classCounts = classCounts %>%
  pivot_wider(names_from=Year, values_from=n) %>%
  mutate(Gender = if_else(Gender=='F', 'Female', 'Male'))
head(classCounts)
## # A tibble: 2 x 5
    Gender FirstYear Sophomore Junior Senior
     <chr> <int> <int> <int> <int>
## 1 Female
                  43
                            96
                                           10
                                    18
```

26

2 Male

51

99

17

Exercise 2 - US Govt Expenditures

```
#a. Download and load the data, skip the first two garbage rows.
expDf = read_excel("hist03z2.xls", skip=2)
#b. Rename function or subfunction to Department
expDf = rename(expDf, `Department` = `Function and Subfunction`)
#c. Remove rows Total, Subtotal, (On-budget), (Off-budget). Remove rows at the bottom.
#First, remove bottom rows.
expDf = expDf %>% filter(row_number() <= n()-4)</pre>
#Then filter out terms we do not want in our rows.
expDf = filter(expDf, !str_detect(Department, "(Total|Subtotal|(On-budget)|(Off-budget))"))
#d.Create a new column for ID_number and parse the Department column for it.
#Parse Department and extract ids using str_extract
expDf = expDf %>% mutate(
  Id_number = str_extract(Department,"([0-9][0-9][0-9])")
#e. Create Function and Subfunction columns.
#First, copy Department to Subfunction
expDf = expDf %>% mutate(
 Subfunction = Department,
 Function = ifelse(is.na(`2015`), Department, NA)
#Fill in NAs
expDf = expDf %>%
 fill(Function)
#f. Remove function rows that did not have data for the 2015 column
expDf = expDf %>%
 na.omit(cols=seq_along(`2015`))
#q. Reshape the data into four columns: Function, Subfunction, Year, and Amount.
reshapedDf = expDf %>%
  pivot_longer(
    `1962`:`2015`,
   names to = 'Year',
   values_to = 'Amount') %>%
    select(Subfunction:Amount) %>%
   mutate(Function = str_remove_all(Function, '\\d|:'))
head(reshapedDf)
## # A tibble: 6 x 4
##
    Subfunction
                                        Function
                                                                        Year Amount
     <chr>>
##
                                        <chr>>
                                                                        <chr> <chr>
## 1 053 Atomic energy defense activit~ " Department of Defense-Milit~ 1962
                                                                              2074
## 2 053 Atomic energy defense activit~ " Department of Defense-Milit~ 1963
## 3 053 Atomic energy defense activit~ " Department of Defense-Milit~ 1964
                                                                              1902
## 4 053 Atomic energy defense activit~ " Department of Defense-Milit~ 1965 1620
## 5 053 Atomic energy defense activit~ " Department of Defense-Milit~ 1966 1466
## 6 053 Atomic energy defense activit~ " Department of Defense-Milit~ 1967 1277
```

```
#h, i. Remove rows with ..... values and non-numeric values.
reshapedDf = filter(reshapedDf, !str_detect(Amount, "....")) %>%
  filter(!str detect(Year, "(TQ)"))
reshapedDf$Amount = as.numeric(as.character(reshapedDf$Amount))
reshapedDf$Year = as.numeric(as.character(reshapedDf$Year))
head(reshapedDf)
## # A tibble: 6 x 4
    Subfunction
##
                                        Function
                                                                       Year Amount
##
     <chr>
                                        <chr>>
                                                                       <dbl> <dbl>
## 1 053 Atomic energy defense activit~ " Department of Defense-Milit~
                                                                       1962
                                                                              2074
## 2 053 Atomic energy defense activit~ " Department of Defense-Milit~ 1963
## 3 053 Atomic energy defense activit~ " Department of Defense-Milit~ 1964
                                                                               1902
## 4 053 Atomic energy defense activit~ " Department of Defense-Milit~ 1965
                                                                               1620
## 5 053 Atomic energy defense activit~ " Department of Defense-Milit~ 1966
                                                                               1466
## 6 053 Atomic energy defense activit~ " Department of Defense-Milit~ 1967
                                                                              1277
#j. Line graph to compare spending for National Defense, Health, Medicare,
# Income Security, and Social Security for each of the years 2001 through 2015
#First, grab only measurements from 2001 to 2015.
finalDf = filter(reshapedDf, Year >= 2001 & Year <= 2015 ) %>%
 filter(str_detect(Function, "Department of Defense-Military|Health|Medicare|Income Security|Social Se
  group_by(Function, Year)
  #Some sort of sum function that I don't know how to use, then the graphing.
```

Exercise 3 - Working with two simple data sets.

```
#Given data
A <- tribble(
  ~Name, ~Car,
  'Alice', 'Ford F150',
  'Bob', 'Tesla Model III',
  'Charlie', 'VW Bug')
B <- tribble(
  ~First.Name, ~Pet,
  'Bob', 'Cat',
  'Charlie', 'Dog',
  'Alice', 'Rabbit')
#a. Join together with both cbind and join commands.
bindedData = cbind(A, B)
head(bindedData)
##
        Name
                         Car First.Name
                                            Pet.
## 1
       Alice
                   Ford F150
                                     Rob
                                            Cat
         Bob Tesla Model III
                                 Charlie
                                            Dog
## 3 Charlie
                      VW Bug
                                   Alice Rabbit
#left join
joinedData = left_join(A,B, by=c("Name"="First.Name"))
joinedData
```

```
## # A tibble: 3 x 3
##
    Name
             Car
                             Pet.
     <chr>
##
             <chr>
                             <chr>>
## 1 Alice Ford F150
                             Rabbit
## 2 Bob
             Tesla Model III Cat
## 3 Charlie VW Bug
#b. Add row for Alice's guinea pig, add to table B!! Not the join!
guineaRow = tibble(First.Name="Alice", Pet="Guinea Pig")
B = rbind(B, guineaRow)
## # A tibble: 4 x 2
   First Name Pet
     <chr>
              <chr>
## 1 Bob
                Cat
## 2 Charlie
## 3 Alice
                Rabbit
## 4 Alice
                Guinea Pig
#c. Combine the two datasets together with cbind and dplyr join.
#In order to bind the two dataframes, we need to add a row to the B dataset so we can directly column b
placeholderRow = tibble(Name=NA, Car=NA)
biggerA = rbind(A, placeholderRow)
biggerA
## # A tibble: 4 x 2
##
   Name
             Car
     <chr>
             <chr>
## 1 Alice
             Ford F150
## 2 Bob
             Tesla Model III
## 3 Charlie VW Bug
## 4 <NA>
             <NA>
bindedData2 = cbind(biggerA,B)
bindedData2 #This is messier than the join operation.
##
                         Car First.Name
                                                Pet
        Name
## 1
       Alice
                   Ford F150
                                    Bob
                                                Cat
         Bob Tesla Model III
                                Charlie
                                               Dog
                      VW Bug
## 3 Charlie
                                  Alice
                                             Rabbit
## 4
        <NA>
                        <NA>
                                  Alice Guinea Pig
#Accomplish this with a simple join instead.
joinedData2 = right_join(A,B, by=c("Name"="First.Name"))
joinedData2 #Much cleaner! No NAs necessary!
## # A tibble: 4 x 3
    Name
             Car
                             Pet
##
     <chr>
             <chr>
                             <chr>
## 1 Alice
           Ford F150
                             Rabbit
## 2 Alice
           Ford F150
                             Guinea Pig
             Tesla Model III Cat
## 3 Bob
## 4 Charlie VW Bug
```

Exercise 4 - Table joins for Customers, Retailers, Cards, and Transactions.

```
Customers <- tribble(
  ~PersonID, ~Name, ~Street, ~City, ~State,
  1, 'Derek Sonderegger', '231 River Run', 'Flagstaff', 'AZ',
  2, 'Aubrey Sonderegger', '231 River Run', 'Flagstaff', 'AZ',
  3, 'Robert Buscaglia', '754 Forest Heights', 'Flagstaff', 'AZ',
  4, 'Roy St Laurent', '845 Elk View', 'Flagstaff', 'AZ')
Retailers <- tribble(</pre>
  ~RetailID, ~Name, ~Street, ~City, ~State,
  1, 'Kickstand Kafe', '719 N Humphreys St', 'Flagstaff', 'AZ',
  2, 'MartAnnes', '112 E Route 66', 'Flagstaff', 'AZ',
  3, 'REI', '323 S Windsor Ln', 'Flagstaff', 'AZ')
Cards <- tribble(</pre>
  ~CardID, ~PersonID, ~Issue_DateTime, ~Exp_DateTime,
  '9876768717278723', 1, '2019-9-20 0:00:00', '2022-9-20 0:00:00',
  '5628927579821287', 2, '2019-9-20 0:00:00', '2022-9-20 0:00:00',
  '7295825498122734', 3, '2019-9-28 0:00:00', '2022-9-28 0:00:00',
  '8723768965231926', 4, '2019-9-30 0:00:00', '2022-9-30 0:00:00')
Transactions <- tribble(</pre>
  ~CardID, ~RetailID, ~DateTime, ~Amount,
  '9876768717278723', 1, '2019-10-1 8:31:23',
  '7295825498122734', 2, '2019-10-1 12:45:45', 25.67,
  '9876768717278723', 1, '2019-10-2 8:26:31',
  '9876768717278723', 1, '2019-10-2 8:30:09',
                                                9.23,
  '5628927579821287', 3, '2019-10-5 18:58:57', 68.54,
  '7295825498122734', 2, '2019-10-5 12:39:26', 31.84,
  '8723768965231926', 2, '2019-10-10 19:02:20', 42.83)
Cards <- Cards %>%
  mutate( Issue_DateTime = lubridate::ymd_hms(Issue_DateTime),
          Exp_DateTime = lubridate::ymd_hms(Exp_DateTime) )
Transactions <- Transactions %>%
  mutate( DateTime = lubridate::ymd_hms(DateTime))
#a. Create a table that gives the credit card statement for Derek. Gives all transactions, amounts, and
head(Customers)
## # A tibble: 4 x 5
## PersonID Name
                                 Street
                                                    City
                                                              State
       <dbl> <chr>
                                 <chr>
                                                    <chr>>
                                                              <chr>
           1 Derek Sonderegger 231 River Run
## 1
                                                    Flagstaff AZ
           2 Aubrey Sonderegger 231 River Run
                                                    Flagstaff AZ
## 3
           3 Robert Buscaglia 754 Forest Heights Flagstaff AZ
           4 Roy St Laurent
                                 845 Elk View
                                                    Flagstaff AZ
#Filter the customers to Derek
custDerek = Customers %>%
  filter(Name == "Derek Sonderegger") %>%
 select(Name)
```

```
#Filter the cards to Derek
cardDerek = Cards %>%
  filter(PersonID == 1) %>%
 select(PersonID, CardID)
#Grab the card ID
derekCardID = cardDerek$CardID[1]
#Filter the transactions based on CardID
transacDerek = Transactions %>%
  filter(CardID == derekCardID)
#Filter the retailers
retailersDerek = Retailers %>%
 filter(RetailID == 1) %>%
  select(RetailID, Name, Street)
#Finally, join these filtered dataframes.
derekStatement = left_join(transacDerek, retailersDerek) %>%
 select(!RetailID)
## Joining, by = "RetailID"
derekStatement
## # A tibble: 3 x 5
##
   CardID
                                        Amount Name
                     DateTime
                                                                Street
    <chr>
                     <dttm>
                                          <dbl> <chr>
                                                                <chr>>
## 1 9876768717278723 2019-10-01 08:31:23 5.68 Kickstand Kafe 719 N Humphreys St
## 2 9876768717278723 2019-10-02 08:26:31 \, 5.68 Kickstand Kafe 719 N Humphreys St
## 3 9876768717278723 2019-10-02 08:30:09 9.23 Kickstand Kafe 719 N Humphreys St
#b. Aubrey lost her credit card on Oct 15 2019
#Close Aubrey's credit card at 4:28:21PM and issue a new credit card in the Cards table.
#Filter the Customers dataframe to get needed info
custAubrey = Customers %>%
 filter(Name == "Aubrey Sonderegger") %>%
  select(Name, PersonID)
#Grab Aubrey's ID
aubreyID = custAubrey$PersonID[1]
aubreyCard = Cards %>%
 filter(PersonID == aubreyID)
#head(aubreyCard)
#Make a new card entry for Aubrey
newCard <- tibble( CardID='6969927542021287', PersonID=2, Issue_DateTime=ymd_hms('2019-10-15 4:28:22'),
#Add the new card
Cards = Cards %>%
 mutate(Exp_DateTime = if_else(PersonID == aubreyID, ymd_hms('2019-10-15 4:28:21'), Exp_DateTime )) %
 full_join(.,newCard)
## Joining, by = c("CardID", "PersonID", "Issue_DateTime", "Exp_DateTime")
```

```
#c. Generate new transaction for new card at KickStand Kafe.
card = newCard$CardID
retailId = 1
datetime = ymd hms('2019-10-16 14:30:21')
amount = 4.98
newPurchase = tibble(CardID = card, RetailID=retailId, DateTime=datetime, Amount = amount)
#d. If the card is currently valid, this should return exactly 1 row.
Valid Cards <- Cards %>%
 filter(CardID == card, Issue_DateTime <= datetime, datetime <= Exp_DateTime)</pre>
# If the transaction is valid, insert the transaction into the table
if( nrow(Valid_Cards) == 1){
  Transactions = rbind(Transactions, newPurchase)
}else{
 print('Card Denied')
#e. Create statement for Aubrey's cards.
#Filter the customers to Aubrey
custAubrey = Customers %>%
  filter(Name == "Aubrey Sonderegger") %>%
  select(Name)
#Filter the cards to Aubrey
cardsAubrey = Cards %>%
 filter(PersonID == 2) %>%
  select(PersonID, CardID)
#Grab the card ID
aubreyCardIDs = cardsAubrey$CardID[1:2]
#Filter the transactions based on CardID
transacAubrey = Transactions %>%
  filter(CardID == aubreyCardIDs[1:2])
#Finally, join the transactionAubrey and retailers dataframes.
aubreyStatement = left_join(transacAubrey, Retailers) %>%
 select(CardID, DateTime, Amount, Name)
## Joining, by = "RetailID"
aubreyStatement
## # A tibble: 2 x 4
##
    CardID
                      DateTime
                                          Amount Name
     <chr>
                                           <dbl> <chr>
##
                      <dttm>
## 1 5628927579821287 2019-10-05 18:58:57 68.5 REI
## 2 6969927542021287 2019-10-16 14:30:21
                                           4.98 Kickstand Kafe
#Exercise 5 - nycflights13
data('flights', package='nycflights13')
write.csv(flights, 'flights.csv')
data('airports', package='nycflights13')
```

```
write.csv(airports, 'airports.csv')
data('airlines', package='nycflights13')
write.csv(airlines, 'airlines.csv')
head(flights)
## # A tibble: 6 x 19
                  day dep_time sched_dep_time dep_delay arr_time sched_arr_time
      year month
                                       <int>
##
     <int> <int> <int>
                        <int>
                                                   <dbl>
                                                            <int>
## 1 2013
              1
                     1
                            517
                                           515
                                                       2
                                                              830
                                                                             819
## 2 2013
              1
                     1
                           533
                                           529
                                                       4
                                                              850
                                                                             830
## 3 2013
                            542
                                           540
                                                       2
                                                              923
                                                                             850
              1
                     1
## 4 2013
                                           545
                            544
                                                      -1
                                                             1004
                                                                            1022
              1
                     1
## 5 2013
               1
                     1
                            554
                                           600
                                                      -6
                                                              812
                                                                             837
## 6 2013
                            554
                                           558
                                                      -4
                                                              740
              1
                                                                             728
                     1
## # ... with 11 more variables: arr_delay <dbl>, carrier <chr>, flight <int>,
      tailnum <chr>, origin <chr>, dest <chr>, air_time <dbl>, distance <dbl>,
     hour <dbl>, minute <dbl>, time_hour <dttm>
head(airports)
## # A tibble: 6 x 8
##
   faa
          name
                                         lat
                                               lon
                                                     alt
                                                            tz dst
                                                                     tzone
##
     <chr> <chr>
                                       <dbl> <dbl> <dbl> <chr> <chr>
## 1 04G
          Lansdowne Airport
                                        41.1 -80.6 1044
                                                            -5 A
                                                                     America/New_Y~
## 2 06A
          Moton Field Municipal Airp~ 32.5 -85.7
                                                     264
                                                            -6 A
                                                                     America/Chica~
## 3 06C
          Schaumburg Regional
                                       42.0 -88.1
                                                            -6 A
                                                                     America/Chica~
                                                     801
## 4 06N
          Randall Airport
                                       41.4 -74.4
                                                     523
                                                            -5 A
                                                                     America/New Y~
## 5 09J
          Jekyll Island Airport
                                       31.1 -81.4
                                                     11
                                                            -5 A
                                                                     America/New_Y~
## 6 OA9
          Elizabethton Municipal Air~ 36.4 -82.2 1593
                                                            -5 A
                                                                     America/New Y~
head(airlines)
## # A tibble: 6 x 2
##
   carrier name
##
     <chr> <chr>
## 1 9E
            Endeavor Air Inc.
## 2 AA
            American Airlines Inc.
## 3 AS
            Alaska Airlines Inc.
## 4 B6
            JetBlue Airways
## 5 DL
            Delta Air Lines Inc.
## 6 EV
            ExpressJet Airlines Inc.
#Filter to Virign America and the date/time, and then join with selected columns.
virginAmerica = flights %>%
 filter(str_detect(carrier, "VX")) %>%
 filter(month == 2 & day == 14) %>% #Filter to the date we want!
  select(carrier, dest, dep_time, air_time) %>%
  left_join(., airports, by=c("dest"="faa"))
head(virginAmerica)
## # A tibble: 6 x 11
     carrier dest dep_time air_time name
                                                lat
                                                      lon
                                                            alt
                                                                   tz dst
                                                                            tzone
                               <dbl> <chr>
                                              <dbl> <dbl> <dbl> <chr> <chr>
     <chr>
            <chr>
                      <int>
## 1 VX
            LAX
                        706
                                 347 Los Ang~ 33.9 -118.
                                                            126
                                                                   -8 A
                                                                            Americ~
```

## 2 VX	SF0	732	344 San Fra~	37.6 -122.	13	-8 A	Americ~
## 3 VX	LAX	909	341 Los Ang~	33.9 -118.	126	-8 A	Americ~
## 4 VX	LAS	934	307 Mc Carr~	36.1 -115.	2141	-8 A	Americ~
## 5 VX	SF0	1029	351 San Fra~	37.6 -122.	13	-8 A	Americ~
## 6 VX	T.AX	1317	349 Los Ang~	33.9 -118.	126	-8 A	Americ~