DA5020 - Week 10 SQLite and comparing dplyr to SQL Developed By Milad Tatari

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This week you are responsible for chapters 10, 11, 12 in the "Data Collection, Integration and Analysis" textbook. Review each chapter separately and work through all examples in the text BEFORE starting the assignment. You will use the schema you developed in homework 6 to store data in SQLite.

This week's assignment you use the relational schema you designed in week 6 and store data into the SQLite relational database system. Load the Unemployment and Educational data files into R studio. One file contains yearly unemployment rates from 1970 to 2015, for counties in the US. The other file contains aggregated data percentages on the highest level of education achieved for each census member. The levels of education are: "less than a high school diploma", "high school diploma awarded", "attended some college", "college graduate and beyond". The census tracks the information at the county level and uses a fips number to represent a specific county within a U.S. state. The fips number is a 5 digit number where the first two digits of the fips number represents a U.S. state, while the last three digits represent a specific county within that state.

Questions

1. Revisit the census schema you created for homework 6. After installing SQLite, implement the tables for your database design in SQLite and load the data into the correct tables using either SQL INSERT statements or CSV loads. Make sure the database design is normalized (at least 3NF) and has minimal redundancy. Make sure your SQLite tables have primary keys as well as foreign keys for relationships. (20 points)

```
a <- read.csv("FipsEducationsDA5020v2.csv")</pre>
b <- read.csv("FipsUnemploymentDA5020.csv")</pre>
#install.packages("stringr")
library(stringr)
#install.packages("tidyr")
library(tidyr)
library(dplyr)
#every measurement for a year and fips is reapeted 4 times which is not good, so we use spread function
a.new <- a%>%
  spread(key=percent_measure, value=percent)
#Seperating the state and counties
a.sep <- a.new %>%
  separate(county state, into = c("state", "county"))
## Warning: Expected 2 pieces. Additional pieces discarded in 15721 rows
## [6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24,
## 25, ...].
d <- select(a.sep, fips,county,state)</pre>
#making the FIPS data frame to make the table
fipsdf <- as_data_frame(d)%>%
```

```
group_by(fips,county,state) %>%
  summarize()
## Warning: `as_data_frame()` is deprecated, use `as_tibble()` (but mind the new semantics).
## This warning is displayed once per session.
# Renaming the a.sep colmns to amke it easier to work with.
colnames(a.sep)<- c("fips", "year","state","county","rural","description","PLUS4College","SOMEcollege",</pre>
#install.packages("RSQLite")
library("RSQLite")
\# open a connection to SQLite and create the EDUEMPDB database
db<-dbConnect(SQLite(),dbname="EDUEMPDBmiladTA.sqlite")</pre>
summary(db)
##
             Length
                               Class
                                                  Mode
                  1 SQLiteConnection
##
                                                    S4
dbListTables(db)
## character(0)
# In SQLite foreign key constraints are disabled by default, so they must be enabled for each database
dbSendQuery(conn = db, "pragma foreign_keys=on;")
## <SQLiteResult>
##
    SQL pragma foreign_keys=on;
    ROWS Fetched: 0 [complete]
##
          Changed: 0
# Use the unique function to remove redundancy
FIPS.DF<-unique(cbind.data.frame(as.integer(a.sep$fips),as.character(a.sep$county),as.character(a.sep$s
colnames(FIPS.DF)<- c("fipsID", "County", "State")</pre>
# Create the FIPS table, specifying fipsID as the PRIMARY KEY
# Since we are specifying a primary ID, there is no need for autoincremented rowid that is automaticall
dbSendQuery(conn = db, "CREATE TABLE FIPS (
            fipsID INTEGER PRIMARY KEY,
            County TEXT,
            State TEXT)
            WITHOUT ROWID")
## Warning: Closing open result set, pending rows
## <SQLiteResult>
     SQL CREATE TABLE FIPS (
##
               fipsID INTEGER PRIMARY KEY,
##
               County TEXT,
##
```

```
State TEXT)
##
               WITHOUT ROWID
##
     ROWS Fetched: 0 [complete]
##
          Changed: 0
##
dbWriteTable(conn = db, name = "FIPS", value = FIPS.DF, row.names=FALSE, append = TRUE)
## Warning: Closing open result set, pending rows
dbListTables(db)
## [1] "FIPS"
#dbReadTable(db, "FIPS")
# Making the rural table
s1 <- select(a.sep, rural, description)</pre>
rural.DF <- as_data_frame(s1)%>%
  group_by(rural,description) %>%
  summarize()
rural.DF <- unique(cbind.data.frame(as.integer(rural.DF$rural),as.character(rural.DF$description)))
colnames(rural.DF)<- c("ruraliD","description")</pre>
library(dplyr)
#rural.DF<-rural.DF[1:9,]</pre>
dbSendQuery(conn = db, "CREATE TABLE RURAL(
            ruraliD INTEGER PRIMARY KEY,
            description TEXT)
            WITHOUT ROWID")
## <SQLiteResult>
     SQL CREATE TABLE RURAL(
##
##
               ruraliD INTEGER PRIMARY KEY,
##
               description TEXT)
##
               WITHOUT ROWID
##
    ROWS Fetched: 0 [complete]
##
          Changed: 0
# insert the RURAL data frame into the Student table in the EDUEMPDB.sqlite database make sure you set
dbWriteTable(conn = db, name = "RURAL", value = rural.DF, row.names=FALSE, append = TRUE)
## Warning: Closing open result set, pending rows
dbListTables(db)
## [1] "FIPS" "RURAL"
dbReadTable(db, "RURAL")
```

```
##
      ruraliD
## 1
            1
## 2
            2
## 3
            3
## 4
            4
## 5
            5
## 6
            6
## 7
            7
## 8
            8
## 9
            9
## 10
           10
##
                                                                               description
## 1
                                 Counties in metro areas of 1 million population or more
## 2
                              Counties in metro areas of 250,000 to 1 million population
## 3
                                Counties in metro areas of fewer than 250,000 population
## 4
                            Urban population of 20,000 or more, adjacent to a metro area
## 5
                        Urban population of 20,000 or more, not adjacent to a metro area
## 6
                           Urban population of 2,500 to 19,999, adjacent to a metro area
## 7
                       Urban population of 2,500 to 19,999, not adjacent to a metro area
## 8
          Completely rural or less than 2,500 urban population, adjacent to a metro area
## 9
      Completely rural or less than 2,500 urban population, not adjacent to a metro area
## 10
#Create the education table, specifying fips ID and ruralID as foreign keys.
# In this table there is no column that can be used as a primary ID, so we will have to use and autoinc
Education.DF<-unique(cbind.data.frame(as.integer(a.sep$fips),as.character(a.sep$year),as.integer(a.sep$
colnames(Education.DF) <- c("fipsID", "YEAR", "rural", "PLUS4College", "SOMEcollege", "DIPLOMA", "LESSDiploma
f <- Education.DF%>%
  select(rural)%>%
  group_by(rural)%>%
  summarise()
dbSendQuery(conn = db, "CREATE TABLE Education(
            fipsID INTEGER,
            YEAR INTEGER,
            rural INTEGER,
            PLUS4College REAL,
            SOMEcollege REAL,
            DIPLOMA REAL,
            LESSDiploma REAL,
            FOREIGN KEY(fipsID) REFERENCES FIPS(fipsID)
            FOREIGN KEY(rural) REFERENCES RURAL(ruraliD))")
## <SQLiteResult>
##
     SQL CREATE TABLE Education(
##
               fipsID INTEGER,
##
               YEAR INTEGER,
               rural INTEGER,
##
##
               PLUS4College REAL,
##
               SOMEcollege REAL,
##
               DIPLOMA REAL,
               LESSDiploma REAL,
##
```

```
##
               FOREIGN KEY(fipsID) REFERENCES FIPS(fipsID)
              FOREIGN KEY(rural) REFERENCES RURAL(ruraliD))
##
     ROWS Fetched: 0 [complete]
##
          Changed: 0
##
dbWriteTable(conn = db, name = "Education", value = Education.DF,row.names = FALSE,append = TRUE)
## Warning: Closing open result set, pending rows
head(dbReadTable(db, "Education"))
##
     fipsID YEAR rural PLUS4College SOMEcollege DIPLOMA LESSDiploma
      1000 1970
                               7.8
                                            7.5
## 1
                   10
                                                   25.9
                                                               58.7
      1000 1980
                               12.2
                                           12.5
                                                   31.8
                                                               43.5
## 2
                   10
      1000 1990
                                                   29.4
                                                               33.1
## 3
                 10
                               15.7
                                           21.7
      1000 2000
## 4
                 10
                               19.0
                                           25.9
                                                   30.4
                                                               24.7
                 10
## 5
      1000 2015
                               23.5
                                           29.7
                                                   31.0
                                                               15.7
## 6 1001 1970
                               6.4
                                                   31.1
                                                               54.8
                   2
                                           7.7
EMPLOY.DF<-unique(cbind.data.frame(as.integer(b$fips),as.integer(b$year),as.double(b$percent_unemployed
colnames(EMPLOY.DF)<- c("fips","year","unemployedRate")</pre>
dbSendQuery(conn = db, "CREATE TABLE EMPLOYMENT (
            fips INTEGER,
            year INTEGER,
            unemployedRate REAL,
            FOREIGN KEY (fips) REFERENCES FIPS(fipsID))")
## <SQLiteResult>
    SQL CREATE TABLE EMPLOYMENT (
##
##
              fips INTEGER,
##
              year INTEGER,
##
              unemployedRate REAL,
##
              FOREIGN KEY (fips) REFERENCES FIPS(fipsID))
##
    ROWS Fetched: 0 [complete]
          Changed: 0
dbWriteTable(conn = db, name = "EMPLOYMENT", value = EMPLOY.DF,row.names = FALSE,append = TRUE)
## Warning: Closing open result set, pending rows
head(dbReadTable(db, "EMPLOYMENT"))
     fips year unemployedRate
## 1 1000 2007
## 2 1000 2008
                          5.7
## 3 1000 2009
                         11.0
## 4 1000 2010
                         10.5
## 5 1000 2011
                         9.6
## 6 1000 2012
                          8.0
```

```
#We have created 4 tables as follows which have the minimum redundancy and are 3NF:
#FIPS: fipsID is the PrimaryKEY
#RURAL: ruraliD is the primary key. "NULL" is conisdered as 10 (integer)
#EDUCATION: fipsID and rural are the foreign keys
#EMPLOYMENT: fipsID is the foreign key
```

- 2. Write SQL expressions to answer the following queries: (40 points)
- 2.0 In the year 1970, what is the population percent that did not earn a high school diploma for the Nantucket county in Massachusetts? What about the year 2015?

• 2.1 What is the average population percentage that did not earn a high school diploma for the counties in Alabama for the year 2015?

```
dbGetQuery(db, "SELECT AVG(LESSDiploma) FROM Education LEFT JOIN FIPS ON Education.fipsID = FIPS.fipsI
```

```
## AVG(LESSDiploma)
## 1 19.75882
```

#returns 5.2

• 2.2 What is the average percentage of college graduates for the counties in the state of Massachusetts for the year 2015?

```
dbGetQuery(db, "SELECT AVG(PLUS4College) FROM Education LEFT JOIN FIPS ON Education.fipsID = FIPS.fips
```

```
## AVG(PLUS4College)
## 1 38.52667
```

```
#It is 38.52%
```

• 2.3 Determine the average percentage of the population that did not earn a high school diploma for the counties in Alabama for each year within the dataset. The result should return the calendar year and the average percentage drop out rate for that year. dbGetQuery(db, paste("SELECT YEAR, AVG(LESSDiploma)", "FROM Education LEFT JOIN FIPS ON Education.fips

```
## YEAR AVG(LESSDiploma)
## 1 1970 65.15882
## 2 1980 50.62059
## 3 1990 40.10000
## 4 2000 30.26471
## 5 2015 19.75882
```

• 2.4 What is the most common rural_urban code for the U.S. counties?

dbGetQuery(db, paste("SELECT rural, COUNT(rural)", "FROM Education", "GROUP BY rural", "ORDER BY COUNT(ru

```
rural COUNT(rural)
##
## 1
           6
                      2961
## 2
           7
                      2165
## 3
           1
                      2153
## 4
           9
                      2091
           2
## 5
                      1890
           3
## 6
                      1779
## 7
           8
                      1097
## 8
           4
                      1070
## 9
           5
                       460
## 10
                        255
          10
```

```
\#the\ most\ common\ rural\ code\ is\ 6
```

• 2.5 Which counties have not been coded with a rural urban code? Return a result that contains two fields: County, State for the counties that has not been assigned a rural urban code. Do not return duplicate values in the result. Order the result alphabetically by state.

#rural code number 10 has not been assigned any rural ID
dbGetQuery(db, paste("SELECT county, state", "FROM Education LEFT JOIN FIPS ON Education.fipsID = FIPS

```
##
              County State
## 1
              Alaska
                         ΑK
## 2
             Alabama
                         AL
## 3
           Arkansas
                         AR
## 4
             Arizona
                         AZ
## 5
         California
                         CA
## 6
           Colorado
                         CO
## 7
        Connecticut
                         CT
## 8
           District
                         DC
## 9
           Delaware
                         DE
                         FL
## 10
             Florida
## 11
             Georgia
                         GA
## 12
              Hawaii
                         HI
## 13
                Iowa
                         ΙA
## 14
               Idaho
                         ID
## 15
           Illinois
                         IL
             Indiana
## 16
                         IN
```

```
## 17
              Kansas
                         KS
## 18
            Kentucky
                         ΚY
## 19
           Louisiana
                         LA
## 20 Massachusetts
                         MA
## 21
            Maryland
                         MD
## 22
                         ME
               Maine
## 23
            Michigan
                         ΜI
## 24
           Minnesota
                         MN
## 25
            Missouri
                         MO
## 26
                         MS
         Mississippi
## 27
             Montana
                         MT
                         NC
## 28
               North
## 29
                         NE
            Nebraska
## 30
                         NH
                 New
## 31
              Nevada
                         NV
## 32
                 Ohio
                         OH
## 33
                         OK
            Oklahoma
## 34
              Oregon
                         OR
       Pennsylvania
## 35
                         PA
## 36
               Rhode
                         RΙ
## 37
               South
                         SC
## 38
           Tennessee
                         TN
## 39
                         \mathsf{TX}
               Texas
## 40
                Utah
                         UT
## 41
                         VA
            Virginia
## 42
             Vermont
                         VT
## 43
          Washington
                         WA
## 44
           Wisconsin
                          WΙ
                          WV
## 45
                West
## 46
                          WY
             Wyoming
```

• 2.6 What is the minimal percentage of college graduates for the counties in the state of Mississippi for the year 2010?

```
#year 2010 does not exist, I consider 2015
dbGetQuery(db, paste("SELECT county,PLUS4College FROM Education LEFT JOIN FIPS ON Education.fipsID = F
```

```
##
             County PLUS4College
## 1
                              7.2
         Issaquena
## 2
             Greene
                              8.2
## 3
                              8.3
              Perry
## 4
      Tallahatchie
                              8.5
## 5
                             10.6
             Benton
## 6
         Chickasaw
                             10.7
## 7
          Walthall
                             11.0
## 8
           Calhoun
                             11.2
## 9
                             11.2
              Scott
## 10
                             11.2
        Tishomingo
## 11
             Kemper
                             11.5
## 12
             Leake
                             11.8
## 13
           Noxubee
                             11.9
## 14
          Prentiss
                             11.9
## 15
             George
                             12.0
```

##	16	Amite	12.1
##	17	Humphreys	12.1
##	18	Lawrence	12.1
##	19	Holmes	12.3
##	20	Yalobusha	12.4
##	21	Simpson	12.7
##	22	Quitman	12.9
##	23	Tippah	12.9
##	24	Marion	13.0
##	25	Pontotoc	13.0
##	26	Smith	13.0
##	27	Itawamba	13.2
##	28	Yazoo	13.3
##	29	Jasper	13.4
##	30	Marshall	13.4
##	31	Stone	13.4
##	32	Choctaw	13.5
##	33	Neshoba	13.5
##	34	Carroll	13.6
##	35	Wayne	13.6
##	36	Union	13.8
##	37	Clarke	13.9
##	38	Pearl	13.9
##	39	Copiah	14.1
##	40	Jefferson	14.2
##	41	Sunflower	14.4
##	42	Wilkinson	14.5
##	43	Monroe	14.6
##	44	Attala	14.9
##	45	Panola	14.9
##	46	Lincoln	15.2
##	47	Covington	15.3
##	48	Claiborne	15.6
##	49	Grenada	15.6
##	50	Montgomery	15.8
##	51	Newton	15.8
##	52	Alcorn	16.2
##	53	Pike	16.3
##	54	Jefferson	16.5
##	55	Winston	16.5
##	56	Tate	16.7
##	57	Webster	17.2
##	58	Tunica	17.3
##	59	Leflore	17.5
##	60	Coahoma	17.6
##	61	Adams	17.8
##	62	Clay	17.8
##	63	Franklin	17.9
##	64	Jones	18.4
##	65	Lauderdale	18.7
##	66	Washington	18.8
##	67	Sharkey	20.0
##	68	Jackson	20.1
##	69	Mississippi	20.7

```
## 70
           Hancock
                             21.0
## 71
           Bolivar
                             21.1
## 72
                             21.3
          Harrison
## 73
                             21.8
               Lee
## 74
           Lowndes
                             21.8
## 75
            DeSoto
                             22.4
## 76
            Warren
                             24.4
                             26.7
## 77
           Forrest
## 78
             Hinds
                             27.7
## 79
            Rankin
                             29.0
## 80
             Lamar
                             35.9
                             38.3
## 81
         Lafayette
## 82
         Oktibbeha
                             43.0
## 83
           Madison
                             46.0
```

#Minimum percent is 7.2 for "Issaquena" county

• 2.7 Which state contains the most number of counties that have not been provided a rural urban code?

dbGetQuery(db, paste("SELECT state, COUNT(county)", "FROM Education LEFT JOIN FIPS ON Education.fipsID

```
##
      State COUNT(county)
## 1
                           5
          ΑK
## 2
                           5
          AL
## 3
          AR
                           5
## 4
          AZ
                           5
## 5
          CA
                           5
## 6
          CO
                           5
## 7
                           5
          CT
## 8
          DC
                           5
## 9
          DE
                           5
## 10
          FL
                           5
## 11
          GA
                           5
## 12
                           5
          HI
## 13
          ΙA
                           5
## 14
                           5
          ID
## 15
          IL
                           5
## 16
          IN
                           5
## 17
                           5
          KS
## 18
          KY
                           5
## 19
          LA
                           5
## 20
                           5
          MA
## 21
          MD
                           5
## 22
                           5
          ME
## 23
                           5
          ΜI
## 24
          MN
                           5
## 25
                           5
          MO
## 26
          MS
                           5
## 27
          MT
                           5
## 28
          NC
                           5
## 29
          ND
                           5
## 30
          NE
                           5
## 31
          NH
                           5
```

```
## 32
          NJ
                            5
## 33
          NM
                            5
## 34
          NV
                            5
## 35
          NY
                            5
## 36
          OH
                            5
## 37
          OK
                            5
## 38
          OR
                            5
## 39
          PA
                            5
## 40
          RΙ
                            5
                            5
## 41
          SC
## 42
          SD
                            5
## 43
          TN
                            5
## 44
          TX
                            5
## 45
                            5
          UT
## 46
                            5
          VA
## 47
          VT
                            5
## 48
                            5
          WA
## 49
          WI
                            5
## 50
          WV
                            5
## 51
          WY
                            5
```

#In all states, there are 5 counties that have not been assigned a rural ID

• 2.8 In the year 2015, which fip counties, U.S. states contain a higher percentage of unemployed citizens than the percentage of college graduates? List the county name and the state name. Order the result alphabetically by state.

```
##
          County State PLUS4College unemployedRate
## 1
        Conecuh
                     AL
                                  8.2
                                                  9.2
## 2
          Greene
                     AL
                                 10.9
                                                 11.0
## 3
          Wilcox
                     AL
                                 12.5
                                                 14.7
## 4
         Bethel
                     AK
                                 11.6
                                                 14.4
## 5
       Kusilvak
                    AK
                                  5.0
                                                 23.2
## 6
      Northwest
                                                 15.5
                     AK
                                 10.6
## 7
          Yukon
                     AK
                                 11.2
                                                 18.0
## 8
          Apache
                     ΑZ
                                 10.8
                                                 13.4
## 9
            Yuma
                                 14.4
                                                 21.8
                     ΑZ
## 10
          Colusa
                     CA
                                 14.6
                                                 15.3
## 11
       Imperial
                     CA
                                 14.1
                                                 24.0
## 12
          Hendry
                    FL
                                  9.8
                                                 10.3
## 13
                                  7.8
                                                 11.3
            Clay
                     GA
## 14
          Macon
                     GA
                                  7.9
                                                  8.9
## 15
        Webster
                     GA
                                  7.6
                                                  8.9
## 16
        Wheeler
                                  5.6
                                                 10.7
                     GA
## 17 Alexander
                                  8.0
                                                  8.6
                     IL
```

```
## 18
            Clay
                     ΚY
                                   9.6
                                                    9.7
## 19
                     KY
                                                   10.0
        Elliott
                                   7.5
## 20
             Lee
                     ΚY
                                   7.9
                                                    8.5
## 21
          Leslie
                                   8.6
                                                   10.8
                     ΚY
## 22
       McCreary
                     ΚY
                                   7.0
                                                    8.3
## 23
       Magoffin
                     ΚY
                                   8.5
                                                   14.7
## 24
          Martin
                     ΚY
                                   6.5
                                                    9.6
## 25
            East
                     LA
                                   8.8
                                                   13.9
## 26
            West
                     LA
                                 11.1
                                                   13.3
## 27 Humphreys
                     MS
                                 12.1
                                                   12.9
## 28
      Issaquena
                     MS
                                  7.2
                                                   16.9
                     NM
                                 12.1
                                                   17.6
## 29
            Luna
## 30
         Tyrrell
                     NC
                                   8.0
                                                    9.4
## 31
          Monroe
                     OH
                                   9.9
                                                   10.0
## 32
       Marlboro
                     SC
                                   8.5
                                                   10.1
## 33
          Oglala
                     SD
                                 11.4
                                                   11.6
                                                    7.6
## 34
          Morgan
                     TN
                                   6.4
## 35
           Scott
                     TN
                                   9.0
                                                    9.6
## 36
                                   8.1
                                                    8.2
           Duval
                     TX
## 37
          Loving
                     TX
                                   1.9
                                                    5.1
## 38
          Newton
                     ΤX
                                   6.4
                                                    7.5
## 39
           Starr
                                   9.1
                                                   13.6
                     TX
         Willacy
## 40
                     ΤX
                                   8.3
                                                   13.1
## 41
          Zavala
                     TX
                                   9.0
                                                   11.1
## 42
       Buchanan
                     VA
                                   9.9
                                                   10.8
## 43
           Boone
                     WV
                                   8.8
                                                    9.6
## 44
        Calhoun
                     WV
                                 10.4
                                                   12.5
                     WV
## 45
            Clay
                                   9.8
                                                   11.2
## 46
                     WV
        Lincoln
                                   9.5
                                                    9.7
## 47
                     WV
                                   8.1
                                                   11.4
           Logan
## 48
       McDowell
                     WV
                                  5.1
                                                   13.0
## 49
           Mingo
                     WV
                                 10.2
                                                   13.1
## 50
           Roane
                     WV
                                 11.3
                                                   11.5
## 51
                                   7.9
                                                    9.7
                     WV
         Wyoming
```

#MAX happens in Falls county, VA state in 2015

• 2.9 Return the county, U.S. state and year that contains the highest percentage of college graduates in this dataset?

```
dbGetQuery(db, paste("SELECT county, state, YEAR, MAX(PLUS4College)", "FROM Education LEFT JOIN FIPS ON
```

```
## County State YEAR MAX(PLUS4College)
## 1 Falls VA 2015 78.8
```

3. Compare your SQL SELECT statements to your dplyr statements written to answer the same questions. Do you have a preference between the two methods? State your reasons for your preference. (10 points)

```
#RSQLite is a database management system, But if you use Exploratory and/or modern R, most likely you a # Based on my experience the total number of codes are more or less same. It is a little bit harder to
```

3.0 In the year 1970, what is the population percent that did not earn a high school diploma for the Nantucket county in Massachusetts? What about the year 2015?

```
# Percent not attaining a high school diploma in MA and Nantucket county in 1970 and 2015
#Filter works on the rows
#select works on the columns (variables)
#qroup_by gathers all the same parameters in column and make them ready for other analysis by summarize
a <- read.csv("FipsEducationsDA5020v2.csv")</pre>
b <- read.csv("FipsUnemploymentDA5020.csv")</pre>
#install.packages("stringr")
library(stringr)
#install.packages("tidyr")
library(tidyr)
library(dplyr)
#every measurement for a year and fips is reapeted 4 times which is not good, so we use spread function
  spread(key=percent_measure, value=percent)
#Seperating the state and counties
a.sep <- a.new %>%
  separate(county_state, into = c("state", "county"))
## Warning: Expected 2 pieces. Additional pieces discarded in 15721 rows
## [6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24,
## 25, ...].
filter(a.sep, state=="MA",county=="Nantucket",year=="1970") %>%
  select(`percent_less than_hs_diploma`) %>%
head() ##33.7%
     percent_less than_hs_diploma
## 1
                             33.7
filter(a.sep, state=="MA",county=="Nantucket",year=="2015") %>%
  select(`percent_less than_hs_diploma`) %>%
head() #5.2%
##
    percent_less than_hs_diploma
## 1
                              5.2
```

• 3.1 What is the average population percentage that did not earn a high school diploma for the counties in Alabama for the year 2015?

```
s<- filter (a.sep, state=="AL",year== "2015") %>%
  select(`percent_less than_hs_diploma`)
head(mean(s$`percent_less than_hs_diploma`))
```

```
## [1] 19.75882
```

• 3.2 What is the average percentage of college graduates for the counties in the state of Massachusetts for the year 2015?

```
x<- filter (a.sep, state=="MA",year== "2015") %>%
    select(percent_four_plus_years_college)
head(mean(x*percent_four_plus_years_college))
```

```
## [1] 38.52667
```

• 3.3 Determine the average percentage of the population that did not earn a high school diploma for the counties in Alabama for each year within the dataset. The result should return the calendar year and the average percentage drop out rate for that year.

```
filter (a.sep, state=="AL") %>%
  select(year, `percent_less than_hs_diploma`) %>%
  group_by(year) %>%
  summarise(avg.not.hs.diploma=mean(`percent_less than_hs_diploma`)) %>%
  head()
```

```
## # A tibble: 5 x 2
##
      year avg.not.hs.diploma
##
     <int>
                         <dbl>
## 1
     1970
                          65.2
## 2 1980
                          50.6
## 3
     1990
                          40.1
## 4
      2000
                          30.3
## 5
     2015
                          19.8
```

• 3.4 What is the most common rural urban code for the U.S. counties?

```
a.sep %>%
count(rural_urban_cont_code) %>%
arrange(desc(n))
```

```
## # A tibble: 10 x 2
##
      rural urban cont code
##
      <fct>
                              <int>
##
    1 6
                                2961
##
    2 7
                               2165
##
    3 1
                               2153
   4 9
##
                               2091
##
    5 2
                                1890
##
    6 3
                                1779
##
    7 8
                                1097
##
    8 4
                                1070
    9 5
##
                                 460
## 10 NULL
                                 255
```

• 3.5 Which counties have not been coded with a rural urban code? Return a result that contains two fields: County, State for the counties that has not been assigned a rural urban code. Do not return duplicate values in the result. Order the result alphabetically by state.

```
#whenever the name of county is exactly the name of state, rural urban code is NULL. for 5 years it has
q <- a.sep %>%
filter (rural_urban_cont_code=="NULL")%>%
select(state,county,rural_urban_cont_code) %>%
group_by(state,county,rural_urban_cont_code) %>%
summarise()
q <- q[order(q$state),] #making in alphabetical order</pre>
```

• 3.6 What is the minimal percentage of college graduates for the counties in the state of Mississippi for the year 2010?

```
#There is no data for year 2010, I calculate it for 2015
a.sep %>%
filter (state=="MS", year== "2015") %>%
  select(county,percent_four_plus_years_college) %>%
  arrange(desc(percent_four_plus_years_college)) %>%
 tail()
##
            county percent_four_plus_years_college
## 78
         Chickasaw
                                               10.6
## 79
            Benton
## 80 Tallahatchie
                                                8.5
## 81
             Perry
                                                8.3
## 82
            Greene
                                                8.2
## 83
                                                7.2
         Issaquena
a.sep %>%
filter (state=="MS", year== "2015") %>%
  select(county,percent_four_plus_years_college) %>%
  summarise(min(percent_four_plus_years_college))
     min(percent_four_plus_years_college)
## 1
                                       7.2
#the mimimum percentage belongs to Issaquena which is 7.2 %
```

• 2.7 Which state contains the most number of counties that have not been provided a rural urban code?

```
v <- b %>%
  filter(year=="2015")
mean(v$percent_unemployed) #average is 5.528102

## [1] 5.528102

d <- select(a.sep, fips,county,state)
fips <- as_data_frame(d)%>%
  group_by(fips,county,state) %>%
  summarize()
z <- inner_join(v,fips, by="fips")
desc.2015 <- z%>%
```

```
filter(percent_unemployed>5.528102) %>%
  arrange(desc(percent_unemployed)) %>%
  select(state,county,percent_unemployed)
```

• 3.8 In the year 2015, which fip counties, U.S. states contain a higher percentage of unemployed citizens than the percentage of college graduates? List the county name and the state name. Order the result alphabetically by state.

```
n <- filter(a.sep,year=="2015")
m <- filter(b,year=="2015") %>%
    select(fips,percent_unemployed)
l<- merge(n,m,by="fips")
k <- 1 %>%
    filter(percent_unemployed>percent_four_plus_years_college) %>%
    select(state,county,percent_unemployed,percent_four_plus_years_college)
k <- k[order(k$state),] #making in alphabetical order</pre>
```

• 3.9 Return the county, U.S. state and year that contains the highest percentage of college graduates in this dataset?

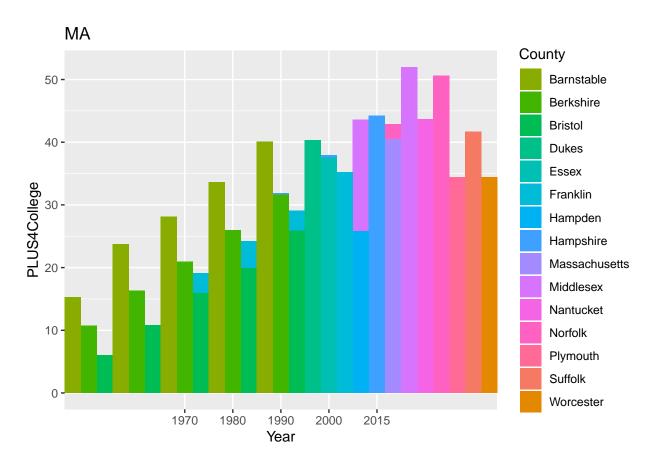
```
a.sep %>%
  select(county, year, state, percent_four_plus_years_college) %>%
  arrange(desc(percent_four_plus_years_college)) %>%
 head()
##
         county year state percent_four_plus_years_college
## 1
                                                        78.8
          Falls 2015
                        VA
## 2 Arlington 2015
                                                        72.9
                        VΑ
            Los 2015
                                                        64.2
## 3
                        NM
## 4
          Falls 2000
                        VA
                                                        63.7
## 5 Alexandria 2015
                                                        61.4
                        VΑ
## 6
         Howard 2015
                        MD
                                                        60.6
# The highest percentage goes to county "Falls" and state "VA" in 2015
```

4. Write a R function named get_state_county_education_data_dplyr(edf, state), it accepts a data frame containing education data and a state's abbreviation for arguments and produces a chart that shows the change in education across time for each county in that state. Use dplyr to extract the data. Write a few R statements that call the function with different state values. (5 points)

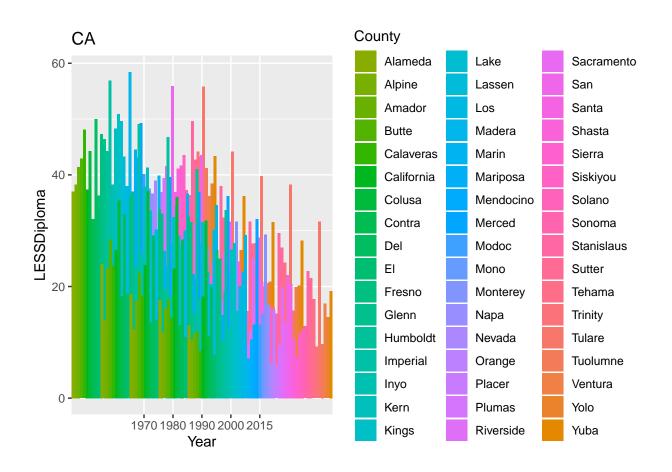
```
library("ggplot2")
get_state_county_education_data_dplyr <- function(EDF, STATEID, EL){
   stateEDUData <- filter(EDF, State==STATEID)

ggplot(stateEDUData, aes(fill=County, y=stateEDUData[,EL], x=YEAR)) +
   geom_bar(width=5, position="dodge", stat="identity", show.legend = T) +
   scale_fill_hue(h = c(100, 400)) +
    xlab('Year') + ylab(EL) + ggtitle(STATEID)
}
EDUCATION <- left_join(Education.DF,FIPS.DF)
#Joining 2 dataframes
get_state_county_education_data_dplyr(EDUCATION, "MA", "PLUS4College")</pre>
```

Warning: position_dodge requires non-overlapping x intervals

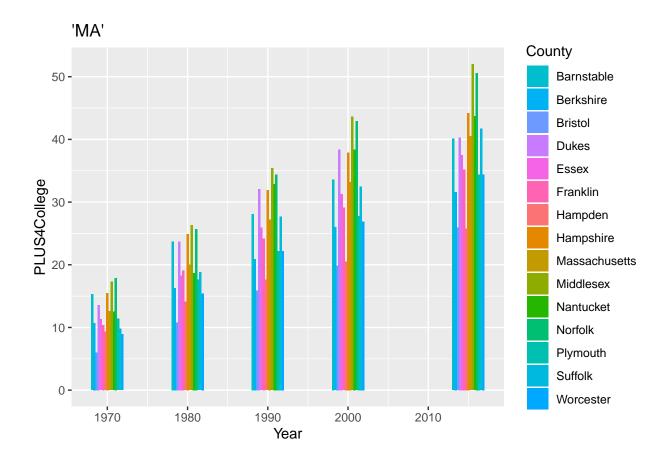


#The number of people with the 4 year college degree is increasing from 1970 to 2015 in almost all counget_state_county_education_data_dplyr(EDUCATION, "CA", "LESSDiploma")

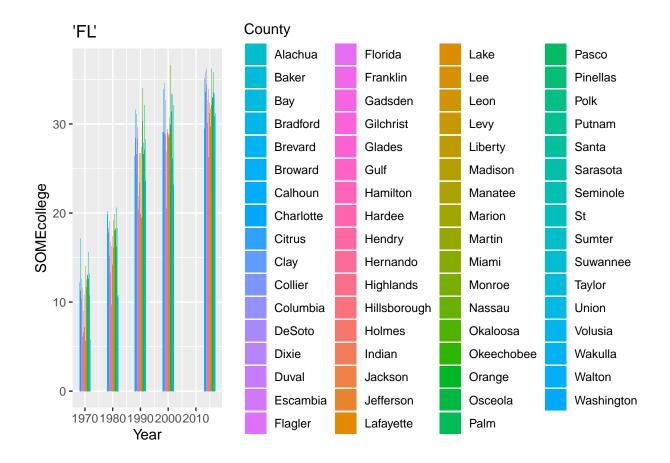


 $\textit{\# we see that the number of students with less diploma degree is decreasing from 1970 to 2015 in \textit{Califoration of the control of the con$

5. Write a R function named get_state_county_education_data_sql(edSQL, state), it accepts a SQL database connection containing education data and a state's abbreviation for arguments and produces a chart that shows the change in education across time for each county in that state. Use SQL SELECT to extract the data from the database. Write a few R statements that call the function with different state values. (10 points)



get_state_county_education_data_sql(db, "'FL'", 'SOMEcollege')



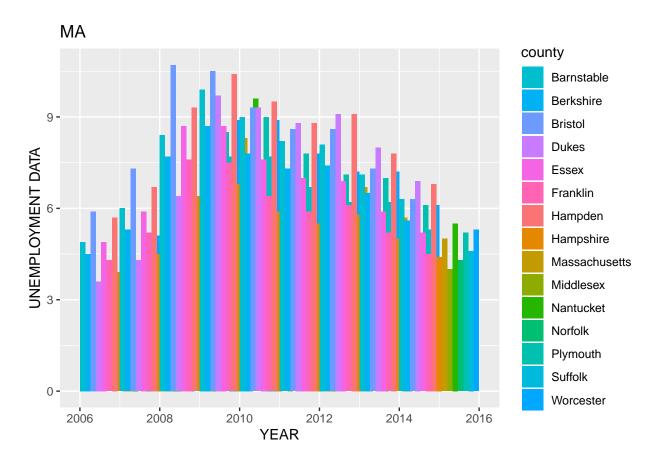
6. Write a R function named get_state_county_unemployment_data_dplyr(udf, state), it accepts a data frame containing unemployment data and state's abbreviation and produces a chart that shows the change in unemployment across time for each county in that state. Use dplyr to extract the data. Write a few R statements that call the function with different state values. (5 points)

```
UNEM.ST <- left_join(EMPLOY.DF, fips, by='fips')

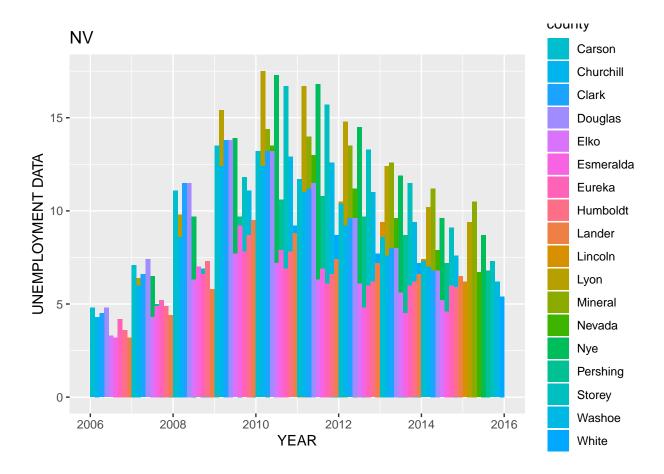
get_state_county_unemployment_data_dplyr <- function(udf, STATEID){
    stateEMPData <- filter(udf, state==STATEID)

    ggplot(stateEMPData, aes(fill=county, y=unemployedRate, x=year)) +
    geom_bar(width=2, position="dodge", stat="identity", show.legend = T) +
    scale_fill_hue(h = c(200, 600)) +
    xlab("YEAR") + ylab("UNEMPLOYMENT DATA") + ggtitle(STATEID)
}

get_state_county_unemployment_data_dplyr(UNEM.ST, "MA")</pre>
```

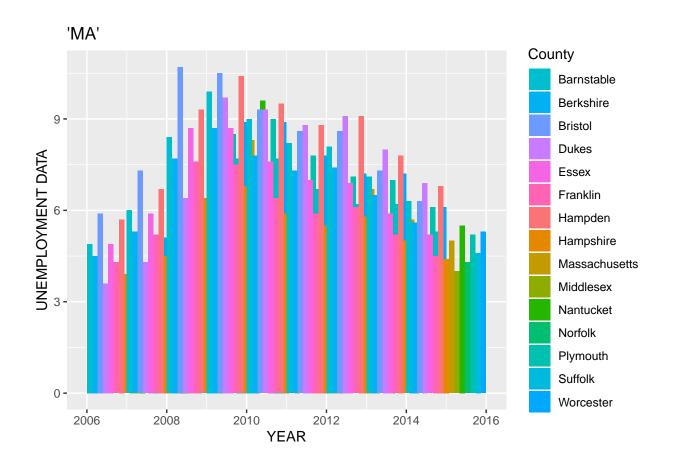


#unemployment rate is first increasing and then decreasing as the recession is passed
get_state_county_unemployment_data_dplyr(UNEM.ST, "NV")



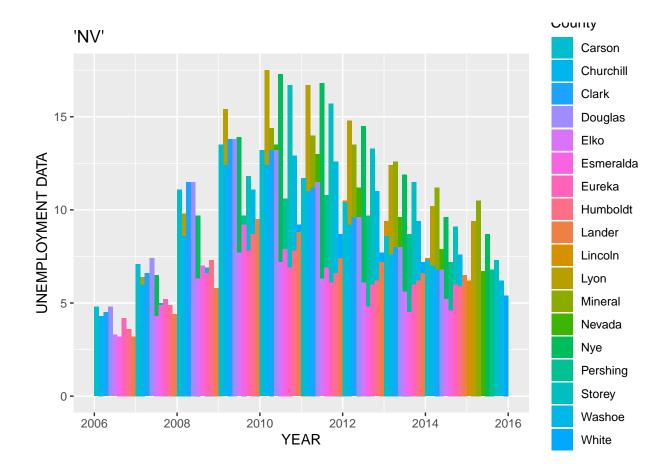
 $\#In\ nevada$, the pattern is the same

7. Write a R function named get_state_county_unemployment_data_sql(udfSQL, state), it accepts a SQL database oject containing unemployment data and state's abbreviation and produces a chart that shows the change in education across time for each county in that state. Use SQL SELECT to extract the data. Write a few R statements that call the function with different state values. (10 points)



get_state_county_unemployment_data_sql(db, "'NV'") #Plot the data in Nevada

Warning: position_dodge requires non-overlapping x intervals



Submission

You need to submit an .Rmd extension file as well as the generated pdf file. Be sure to state all the assumptions and give explanations as comments in the .Rmd file wherever needed to help us assess your submission. Please name the submission file LAST_FirstInitial_1.Rmd for example for John Smith's 1st assignment, the file should be named Smith_J_1.Rmd.