

BIOS6301: Homework 5

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Question 1

24 points

Import the HAART dataset (haart.csv) from the GitHub repository into R, and perform the following manipulations: (4 points each)

```
library(lubridate)
haart <- read.csv("https://raw.githubusercontent.com/fonnesbeck/Bios6301/master/datasets/haart.csv")
head(haart) #preview structure of base dataset
```

```
##   male age aids cd4baseline logvl  weight hemoglobin  init.reg init.date
## 1    1  25   0         NA     NA      NA           NA 3TC,AZT,EFV    7/1/03
## 2    1  49   0        143     NA  58.0608          11 3TC,AZT,EFV   11/23/04
## 3    1  42   1        102     NA  48.0816           1 3TC,AZT,EFV    4/30/03
## 4    0  33   0        107     NA  46.0000          NA 3TC,AZT,NVP    3/25/06
## 5    1  27   0         52      4      NA           NA 3TC,D4T,EFV    9/1/04
## 6    0  34   0        157     NA  54.8856          NA 3TC,AZT,NVP   12/2/03
##   last.visit death date.death
## 1    2/26/07     0      <NA>
## 2    2/22/08     0      <NA>
## 3   11/21/05     1    1/11/06
## 4     5/5/06     1     5/7/06
## 5   11/13/07     0      <NA>
## 6    2/28/08     0      <NA>
```

1. Convert date columns into a usable (for analysis) format. Use the table command to display the counts of the year from init.date.

```
# convert date columns
haart[, "init.date"] <- as.POSIXct(haart[, "init.date"], format = "%m/%d/%y")
haart[, "last.visit"] <- as.POSIXct(haart[, "last.visit"], format = "%m/%d/%y")
haart[, "date.death"] <- as.POSIXct(haart[, "date.death"], format = "%m/%d/%y")

# display counts of the year from init.date
table(year(haart[, "init.date"]))
```

```
##
## 1998 2000 2001 2002 2003 2004 2005 2006 2007
##     1     5    17    60   270   292   207   104    44
```

- 1.2. Create an indicator variable (one which takes the values 0 or 1 only) to represent death within 1 year of the initial visit. How many observations died in year 1?

```

# create indicator variable to represent death within 1 year of the initial
# visit
haart$death.within.year <- (difftime(haart$date.death, haart$last.visit, units = "days") <=
365)
haart$death.within.year[is.na(haart$death.within.year)] <- 0
head(haart) #prove I added the column correctly

```

```

##   male age aids cd4baseline logvl  weight hemoglobin  init.reg
## 1    1  25    0         NA    NA      NA          NA 3TC,AZT,EFV
## 2    1  49    0        143    NA  58.0608         11 3TC,AZT,EFV
## 3    1  42    1        102    NA  48.0816          1 3TC,AZT,EFV
## 4    0  33    0        107    NA  46.0000         NA 3TC,AZT,NVP
## 5    1  27    0         52     4      NA          NA 3TC,D4T,EFV
## 6    0  34    0        157    NA  54.8856         NA 3TC,AZT,NVP
##   init.date last.visit death date.death death.within.year
## 1 2003-07-01 2007-02-26     0      <NA>                0
## 2 2004-11-23 2008-02-22     0      <NA>                0
## 3 2003-04-30 2005-11-21     1 2006-01-11                1
## 4 2006-03-25 2006-05-05     1 2006-05-07                1
## 5 2004-09-01 2007-11-13     0      <NA>                0
## 6 2003-12-02 2008-02-28     0      <NA>                0

```

```

# calculate number of participants who died in year 1
(yr1.deaths <- sum(haart$death.within.year))

```

```
## [1] 104
```

1.3. Use the init.date, last.visit and death.date columns to calculate a followup time (in days), which is the difference between the first and either the last visit or a death event (whichever comes first).

```

#function that takes who dataframe, adds a followup time column and calculates values
#then returns the new and improved dataframe
calculate.followup <- function(dataframe)
{
  for (i in 1:nrow(dataframe))
  {
    if (dataframe$death[i] == 1) #check to see if death came first
    {
      dataframe$followup.time[i] <- difftime(dataframe$date.death[i],dataframe$init.date[i],
units="days")
    }
    else
    {
      dataframe$followup.time[i] <- difftime(dataframe$last.visit[i],dataframe$init.date[i],
units="days")
    }
  }
  return(dataframe)
}

haart <- calculate.followup(haart)
head(haart)

```

```
##   male age aids cd4baseline logvl  weight hemoglobin  init.reg
## 1    1  25   0      NA      NA      NA      NA 3TC,AZT,EFV
## 2    1  49   0     143     NA  58.0608     11 3TC,AZT,EFV
## 3    1  42   1     102     NA  48.0816      1 3TC,AZT,EFV
## 4    0  33   0     107     NA  46.0000     NA 3TC,AZT,NVP
## 5    1  27   0      52      4      NA     NA 3TC,D4T,EFV
## 6    0  34   0     157     NA  54.8856     NA 3TC,AZT,NVP
##   init.date last.visit death date.death death.within.year followup.time
## 1 2003-07-01 2007-02-26     0      <NA>                0    1336.04167
## 2 2004-11-23 2008-02-22     0      <NA>                0    1186.00000
## 3 2003-04-30 2005-11-21     1 2006-01-11                1     987.04167
## 4 2006-03-25 2006-05-05     1 2006-05-07                1      42.95833
## 5 2004-09-01 2007-11-13     0      <NA>                0    1168.04167
## 6 2003-12-02 2008-02-28     0      <NA>                0    1549.00000
```

1.4. If these times are longer than 1 year, censor them (this means if the value is above 365, set followup to 365).

```
#function that takes in a vector of followup.times and
#censors them so that 365 is the maximum, then returns the censored vector
max(haart$followup.time) #see the max of followup.time as calculated
```

```
## [1] 3533.042
```

```
censor.followup <- function(followup.time)
{
  for (i in 1:length(followup.time))
  {
    if (followup.time[i] > 365)
    {
      followup.time[i] <- 365
    }
  }
  return(followup.time)
}
haart$followup.time <- censor.followup(haart$followup.time)
max(haart$followup.time) #see new max of followup.time after censoring
```

```
## [1] 365
```

1.5. Print the quantile for this new variable.

```
quantile(haart$followup.time)
```

```
##    0%   25%   50%   75%  100%
##   0.0 329.5 365.0 365.0 365.0
```

1.6. Create another indicator variable representing loss to followup; this means the observation is not known to be dead but does not have any followup visits after the first year. How many records are lost-to-followup?

```

#function that takes in whole dataframe and appends an indicator
#column for whether or not individuals were lost fo followup
calculate.losstofollowup <- function(dataframe)
{
  for (i in 1:nrow(dataframe))
  {
    if (dataframe$death[i] == 1)#if dead, then they weren't "lost-to-followup"
    {
      dataframe$loss.to.followup[i] <- 0
    }
    else
    {
      #if not dead, haven't had followup after first year, "lost-to-followup"
      if (as.integer(difftime(dataframe$last.visit[i],dataframe$init.date[i],units="days")) <= 365)
      {
        dataframe$loss.to.followup[i] <- 1
      }
      #if not dead, have had a followup after first year, not "lost-to-followup"
      else dataframe$loss.to.followup[i] <- 0
    }
  }
  return(dataframe)
}

#add lost-to-followup column to the haart dataframe
haart <- calculate.losstofollowup(haart)
head(haart) #appreciate shiny new column

```

```

##   male age aids cd4baseline logvl  weight hemoglobin  init.reg
## 1    1  25   0         NA      NA      NA      NA 3TC,AZT,EFV
## 2    1  49   0        143     NA 58.0608      11 3TC,AZT,EFV
## 3    1  42   1        102     NA 48.0816       1 3TC,AZT,EFV
## 4    0  33   0        107     NA 46.0000      NA 3TC,AZT,NVP
## 5    1  27   0         52      4      NA      NA 3TC,D4T,EFV
## 6    0  34   0        157     NA 54.8856      NA 3TC,AZT,NVP
##   init.date last.visit death date.death death.within.year followup.time
## 1 2003-07-01 2007-02-26     0      <NA>                0      365.00000
## 2 2004-11-23 2008-02-22     0      <NA>                0      365.00000
## 3 2003-04-30 2005-11-21     1 2006-01-11                1      365.00000
## 4 2006-03-25 2006-05-05     1 2006-05-07                1      42.95833
## 5 2004-09-01 2007-11-13     0      <NA>                0      365.00000
## 6 2003-12-02 2008-02-28     0      <NA>                0      365.00000
##   loss.to.followup
## 1                  0
## 2                  0
## 3                  0
## 4                  0
## 5                  0
## 6                  0

```

```

#sum of indicator variables gives number of patients "lost-to-followup"
sum(haart$loss.to.followup)

```

```
## [1] 173
```

So, from this we can see that 173 records were lost-to-followup.

1.7. Recall our work in class, which separated the `init.reg` field into a set of indicator variables, one for each unique drug. Create these fields and append them to the database as new columns.

```
create.regimens <- function(dataframe) {
  init.reg <- as.character(dataframe[, "init.reg"])
  dataframe[["init.reg_list"]] <- strsplit(init.reg, ",")
  unique.drugs <- unique(unlist(dataframe$init.reg_list))
  reg.drugs <- matrix(FALSE, nrow = nrow(dataframe), ncol = length(unique.drugs))
  for (i in seq_along(unique.drugs)) {
    reg.drugs[, i] <- sapply(dataframe$init.reg_list, function(x) unique.drugs[i] %in%
      x)
  }
  reg.drugs <- data.frame(reg.drugs)
  names(reg.drugs) <- unique.drugs
  dataframe <- cbind(dataframe, reg.drugs)
  return(dataframe)
}

haart <- create.regimens(haart)
head(haart) #why not
```

```
##   male age aids cd4baseline logvl  weight hemoglobin  init.reg
## 1    1  25   0         NA      NA      NA          NA 3TC,AZT,EFV
## 2    1  49   0        143     NA 58.0608         11 3TC,AZT,EFV
## 3    1  42   1        102     NA 48.0816          1 3TC,AZT,EFV
## 4    0  33   0        107     NA 46.0000         NA 3TC,AZT,NVP
## 5    1  27   0         52      4      NA          NA 3TC,D4T,EFV
## 6    0  34   0        157     NA 54.8856         NA 3TC,AZT,NVP
##   init.date last.visit death date.death death.within.year followup.time
## 1 2003-07-01 2007-02-26     0      <NA>                0      365.00000
## 2 2004-11-23 2008-02-22     0      <NA>                0      365.00000
## 3 2003-04-30 2005-11-21     1 2006-01-11                1      365.00000
## 4 2006-03-25 2006-05-05     1 2006-05-07                1      42.95833
## 5 2004-09-01 2007-11-13     0      <NA>                0      365.00000
## 6 2003-12-02 2008-02-28     0      <NA>                0      365.00000
##   loss.to.followup init.reg_list 3TC  AZT  EFV  NVP  D4T  ABC  DDI
## 1                0 3TC, AZT, EFV TRUE  TRUE  TRUE FALSE FALSE FALSE
## 2                0 3TC, AZT, EFV TRUE  TRUE  TRUE FALSE FALSE FALSE
## 3                0 3TC, AZT, EFV TRUE  TRUE  TRUE FALSE FALSE FALSE
## 4                0 3TC, AZT, NVP TRUE  TRUE FALSE  TRUE FALSE FALSE
## 5                0 3TC, D4T, EFV TRUE FALSE  TRUE FALSE  TRUE FALSE
## 6                0 3TC, AZT, NVP TRUE  TRUE FALSE  TRUE FALSE FALSE
##   IDV  LPV  RTV  SQV  FTC  TDF  DDC  NFV  T20  ATV  FPV
## 1 FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## 2 FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## 3 FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## 4 FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## 5 FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## 6 FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
```

1.8. Which drug regimen are found over 100 times?

```
regimen <- matrix(nrow = nrow(haart), ncol = 1)
for (j in 17:34) {
  for (i in 1:nrow(haart)) {
    if (haart[i, j] == TRUE) {
      if (j == 17) {
        regimen[i] <- colnames(haart)[j]
      } else {
        regimen[i] <- paste(regimen[i], colnames(haart)[j])
      }
    }
  }
}
haart <- cbind(haart, regimen)
haart[, 17:34] <- NULL #get rid of the indicators before preview
head(haart) #check what that actually did
```

```
##   male age aids cd4baseline logvl  weight hemoglobin  init.reg
## 1    1  25   0         NA     NA      NA          NA 3TC,AZT,EFV
## 2    1  49   0        143     NA  58.0608         11 3TC,AZT,EFV
## 3    1  42   1        102     NA  48.0816          1 3TC,AZT,EFV
## 4    0  33   0        107     NA  46.0000         NA 3TC,AZT,NVP
## 5    1  27   0         52     4      NA          NA 3TC,D4T,EFV
## 6    0  34   0        157     NA  54.8856         NA 3TC,AZT,NVP
##   init.date last.visit death date.death death.within.year followup.time
## 1 2003-07-01 2007-02-26     0      <NA>                0      365.00000
## 2 2004-11-23 2008-02-22     0      <NA>                0      365.00000
## 3 2003-04-30 2005-11-21     1 2006-01-11                1      365.00000
## 4 2006-03-25 2006-05-05     1 2006-05-07                1      42.95833
## 5 2004-09-01 2007-11-13     0      <NA>                0      365.00000
## 6 2003-12-02 2008-02-28     0      <NA>                0      365.00000
##   loss.to.followup init.reg_list  regimen
## 1                0 3TC, AZT, EFV 3TC AZT EFV
## 2                0 3TC, AZT, EFV 3TC AZT EFV
## 3                0 3TC, AZT, EFV 3TC AZT EFV
## 4                0 3TC, AZT, NVP 3TC AZT NVP
## 5                0 3TC, D4T, EFV 3TC EFV D4T
## 6                0 3TC, AZT, NVP 3TC AZT NVP
```

```
sort(table(regimen))
```

```
## regimen
##      3TC ABC IDV RTV      3TC ABC RTV      3TC AZT ABC LPV RTV
##              1              1              1
##      3TC AZT ABC RTV SQV      3TC AZT DDI      3TC AZT EFV NFV
##              1              1              1
##      3TC AZT RTV FPV      3TC EFV TDF      3TC LPV RTV TDF
##              1              1              1
##      3TC RTV TDF FPV      NA ABC DDI LPV RTV      NA ABC DDI RTV ATV
##              1              1              1
##      NA D4T ABC LPV RTV      NA D4T ABC RTV SQV      NA D4T RTV SQV
##              1              1              1
```

## NA DDI LPV RTV SQV T20	NA EFV D4T DDC	NA EFV DDI FTC
## 1	1	1
## NA NVP FTC TDF	NA RTV FTC TDF ATV	3TC D4T LPV RTV
## 1	1	2
## 3TC NVP ABC	NA AZT EFV DDI	NA EFV D4T ABC
## 2	2	2
## NA LPV RTV FTC TDF	NA NVP D4T DDI	NA NVP LPV RTV
## 2	2	2
## 3TC D4T NFV	NA EFV FTC TDF	3TC ABC RTV SQV
## 3	3	4
## 3TC AZT NFV	3TC DDI LPV RTV	NA EFV D4T DDI
## 4	4	4
## 3TC D4T IDV RTV	3TC NVP DDI	3TC AZT IDV RTV
## 6	6	8
## 3TC D4T RTV SQV	3TC EFV ABC	3TC AZT IDV
## 8	11	12
## 3TC AZT RTV SQV	3TC EFV DDI	3TC AZT LPV RTV
## 13	15	16
## 3TC AZT ABC	3TC EFV D4T	3TC NVP D4T
## 29	54	61
## 3TC AZT NVP	3TC AZT EFV	
## 284	421	

From this, we can see that the only regimens that were prescribed more than 100 times were “3TC AZT NVP” and “3TC AZT EFV”.

Turning this into a data frame is as simple as a call to `data.frame`, using `all_drugs` as a set of column labels:

1.9. The dataset `haart2.csv` contains a few additional observations for the same study. Import these and append them to your master dataset (if you were smart about how you coded the previous steps, cleaning the additional observations should be easy!). Show the first five records and the last five records of the complete (and clean) data set.

```
haart <- data.frame(read.csv("https://raw.githubusercontent.com/fonnesbeck/Bios6301/master/datasets/haart.csv"))
haart2 <- data.frame(read.csv("https://raw.githubusercontent.com/fonnesbeck/Bios6301/master/datasets/haart2.csv"))
haart.merged <- merge(haart, haart2, all = TRUE)

# convert date columns
haart.merged[, "init.date"] <- as.POSIXct(haart.merged[, "init.date"], format = "%m/%d/%y")
haart.merged[, "last.visit"] <- as.POSIXct(haart.merged[, "last.visit"], format = "%m/%d/%y")
haart.merged[, "date.death"] <- as.POSIXct(haart.merged[, "date.death"], format = "%m/%d/%y")

# create indicator variable to represent death within 1 year of the initial
# visit
haart.merged$death.within.year <- (difftime(haart.merged$date.death, haart.merged$last.visit,
      units = "days") <= 365)
haart.merged$death.within.year[is.na(haart.merged$date.death)] <- 0

# create followup time field
haart.merged <- calculate.followup(haart.merged)

# censor followup time
haart.merged$followup.time <- censor.followup(haart.merged$followup.time)

# add indicator for loss-to-followup
```

```

haart.merged <- calculate.losstofollowup(haart.merged)

# add indicators for regimen
haart.merged <- create.regimens(haart.merged)

regimen <- matrix(nrow = nrow(haart.merged), ncol = 1)
for (j in 17:34) {
  for (i in 1:nrow(haart.merged)) {
    if (haart.merged[i, j] == TRUE) {
      if (j == 17) {
        regimen[i] <- colnames(haart.merged)[j]
      } else {
        regimen[i] <- paste(regimen[i], colnames(haart.merged)[j])
      }
    }
  }
}
haart.merged <- cbind(haart.merged, regimen)

haart.merged[1:5, ] #first 5

```

```

##   male age aids cd4baseline   logvl weight hemoglobin   init.reg
## 1    0  18    0         89 5.184231    NA          NA 3TC,AZT,EFV
## 2    0  18    0        280      NA 52.164          11 3TC,AZT,EFV
## 3    0  18    0        431 5.342423  58.000          NA 3TC,AZT,NVP
## 4    0  19    0         51 5.618615  48.600          NA 3TC,AZT,NVP
## 5    0  19    0        180 4.121330    NA          NA 3TC,AZT,NVP
##   init.date last.visit death date.death death.within.year followup.time
## 1 2003-11-03 2006-04-12    0      <NA>                0          365
## 2 2004-02-19 2008-03-14    0      <NA>                0          365
## 3 2007-03-13 2007-03-13    0      <NA>                0           0
## 4 2005-12-07 2007-04-17    0      <NA>                0          365
## 5 2006-09-08 2006-10-15    0      <NA>                0           37
##   loss.to.followup init.reg_list  3TC  AZT   EFV   NVP   NFV   ABC   LPV
## 1                0 3TC, AZT, EFV TRUE TRUE  TRUE FALSE FALSE FALSE FALSE
## 2                0 3TC, AZT, EFV TRUE TRUE  TRUE FALSE FALSE FALSE FALSE
## 3                1 3TC, AZT, NVP TRUE TRUE  TRUE FALSE FALSE FALSE FALSE
## 4                0 3TC, AZT, NVP TRUE TRUE  TRUE FALSE FALSE FALSE FALSE
## 5                1 3TC, AZT, NVP TRUE TRUE  TRUE FALSE FALSE FALSE FALSE
##   RTV   D4T   DDI   IDV   SQV   T20   FPV   TDF   ATV   FTC   DDC
## 1 FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## 2 FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## 3 FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## 4 FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## 5 FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
##       regimen
## 1 3TC AZT EFV
## 2 3TC AZT EFV
## 3 3TC AZT NVP
## 4 3TC AZT NVP
## 5 3TC AZT NVP

```



```
haart.merged[1000:1004, ] #last 5
```

```
##      male age aids cd4baseline   logvl   weight hemoglobin      init.reg
## 1000    1  66   0      298 4.09496      NA      NA      3TC,AZT,EFV
## 1001    1  67   0       95      NA 66.6792      16      3TC,AZT,EFV
## 1002    1  69   0      NA      NA      NA      NA 3TC,AZT,RTV,SQV
## 1003    1  80   0      267      NA 53.0712      NA      3TC,AZT,NVP
## 1004    1  89   0       9      NA 43.5456      10      3TC,ABC,AZT
##      init.date last.visit death date.death death.within.year
## 1000 2006-06-08 2007-02-12    0      <NA>                0
## 1001 2004-02-13 2008-02-21    0      <NA>                0
## 1002 2006-04-01 2007-09-13    0      <NA>                0
## 1003 2004-11-08 2006-11-20    1 2006-11-26                1
## 1004 2004-12-15 2006-04-11    0      <NA>                0
##      followup.time loss.to.followup      init.reg_list 3TC AZT EFV
## 1000      249.0417                1      3TC, AZT, EFV TRUE TRUE TRUE
## 1001      365.0000                0      3TC, AZT, EFV TRUE TRUE TRUE
## 1002      365.0000                0 3TC, AZT, RTV, SQV TRUE TRUE FALSE
## 1003      365.0000                0      3TC, AZT, NVP TRUE TRUE FALSE
## 1004      365.0000                0      3TC, ABC, AZT TRUE TRUE FALSE
##      NVP NFV ABC LPV RTV D4T DDI IDV SQV T20 FPV
## 1000 FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## 1001 FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## 1002 FALSE FALSE FALSE FALSE TRUE FALSE FALSE FALSE TRUE FALSE FALSE
## 1003 TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## 1004 FALSE FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
##      TDF ATV FTC DDC      regimen
## 1000 FALSE FALSE FALSE FALSE      3TC AZT EFV
## 1001 FALSE FALSE FALSE FALSE      3TC AZT EFV
## 1002 FALSE FALSE FALSE FALSE 3TC AZT RTV SQV
## 1003 FALSE FALSE FALSE FALSE      3TC AZT NVP
## 1004 FALSE FALSE FALSE FALSE      3TC AZT ABC
```

Question 2

14 points

Use the following code to generate data for patients with repeated measures of A1C (a test for levels of blood glucose).

```
genData <- function(n) {
  if(exists(".Random.seed", envir = .GlobalEnv)) {
    save.seed <- get(".Random.seed", envir = .GlobalEnv)
    on.exit(assign(".Random.seed", save.seed, envir = .GlobalEnv))
  } else {
    on.exit(rm(".Random.seed", envir = .GlobalEnv))
  }
  set.seed(n)
  subj <- ceiling(n / 10)
  id <- sample(subj, n, replace=TRUE)
  times <- as.integer(difftime(as.POSIXct("2005-01-01"), as.POSIXct("2000-01-01"), units='secs'))
  dt <- as.POSIXct(sample(times, n), origin='2000-01-01')
  mu <- runif(subj, 4, 10)
```

```

a1c <- unsplit(mapply(rnorm, tabulate(id), mu, SIMPLIFY=FALSE), id)
data.frame(id, dt, a1c)
}
x <- genData(500)

```

Perform the following manipulations: (2 points each)

2.1. Order the data set by id and dt.

```

x <- x[order(x$id,x$dt),]

```

2.2. For each id, determine if there is more than a one year gap in between observations. Add a new row at the one year mark, with the a1c value set to missing. A two year gap would require two new rows, and so forth.

```

#take subset of observations with a parameter for the id
subset.id <- function(id.num)
{
  return(subset(x,x$id==id.num))
}

#find gaps between observation i and the observation below it that are greater than 365 days
#this is where we will need to insert rows
find.gaps <- function(id)
{
  id.vector <- subset.id(id)
  gaps <- matrix(nrow=nrow(id.vector),ncol=1)
  for (i in 1:nrow(id.vector)-1)
  {
    gaps[i] <- as.integer(difftime(id.vector[i+1,2],id.vector[i,2],units="days"))
  }
  return(gaps)
}

#apply the find.gaps function to all ids 1:50
gaps <- NULL
for (id in 1:50)
{
  gaps <- rbind(gaps,find.gaps(id))
}
x <- cbind(x,gaps)

#find rows where the gap (after the observation) > 365
(gaps.positions <- which(abs(x$gaps)>365))

```

```

## [1] 36 47 49 55 56 69 71 81 87 110 117 119 126 128 135 136 143
## [18] 151 152 159 190 193 196 209 210 215 216 258 267 279 286 294 310 319
## [35] 326 327 336 341 356 360 370 380 381 388 416 420 424 433 436 444 454
## [52] 471 474 481

```

```

#insert rows for gaps of 1 year
for (i in 1:length(gaps.positions))

```

```

{
  row.below <- gaps.positions[i]
  save.above <- x[1:row.below,] #save all of the rows above
  save.below <- x[(row.below+1):nrow(x),] #save all of the rows below
  x[row.below+1,1] <- x[row.below,1]
  x[row.below+1,2] <- x[row.below,2] + days(365)
  x[row.below+1,3] <- NA
  x[row.below+1,4] <- NA
  x <- rbind(save.above,x[row.below+1,],save.below)
  gaps.positions <- gaps.positions + 1 #increment gap indices to account for new inserted row
}

#apply the function again to allow for two year gaps
gaps2 <- NULL
for (id in 1:50)
{
  gaps2 <- rbind(gaps2,find.gaps(id))
}
x <- cbind(x,gaps2)

(gaps2.positions <- which(abs(x$gaps2)>365)) #find gaps that were greater than 1 year

```

```
## [1] 169 179
```

```

for (i in 1:length(gaps2.positions))
{
  row.below <- gaps2.positions[i]
  save.above <- x[1:row.below,] #save all of the rows above
  save.below <- x[(row.below+1):nrow(x),] #save all of the rows below
  x[row.below+1,1] <- x[row.below,1]
  x[row.below+1,2] <- x[row.below,2] + days(365)
  x[row.below+1,3] <- NA
  x[row.below+1,4] <- NA
  x <- rbind(save.above,x[row.below+1,],save.below)
  gaps2.positions <- gaps2.positions + 1 #increment gap indices to account for new inserted row
}

```

2.3. Create a new column visit. For each id, add the visit number. This should be 1 to n where n is the number of observations for an individual. This should include the observations created with missing alc values.

```

#function that will create a visit field for the parameterized id
count.visits <- function(id)
{
  id.vector <- subset.id(id)
  visit.no <- matrix(nrow=nrow(id.vector),ncol=1)
  for (i in 1:nrow(visit.no))
  {
    visit.no[i] <- i
  }
  return(visit.no)
}

```

```

#run the count.visits function on all ids to create a complete column for visits
visit <- NULL
for (id in 1:50)
{
  visit <- rbind(visit,count.visits(id))
}

#add the visit column to the original x dataframe
x <- cbind(x,visit)

```

2.4. For each id, replace missing values with the mean a1c value for that individual.

```

#function that will find NA in the a1c field and replace them with that id's mean a1c
replace.na <- function(id)
{
  id.vector <- subset.id(id)
  a1c.mean <- mean(id.vector$a1c, na.rm=TRUE)
  new.a1c <- matrix(nrow=nrow(id.vector),ncol=1)
  for (i in 1:nrow(id.vector))
  {
    if (is.na(id.vector[i,3])==TRUE)
    {
      new.a1c[i] <- a1c.mean
    }
    else
    {
      new.a1c[i] <- id.vector[i,3]
    }
  }
  return(new.a1c)
}

#run the replace.na function for every id from 1:50
a1c.replaced <- NULL
for (id in 1:50)
{
  a1c.replaced <- rbind(a1c.replaced,replace.na(id))
}

x <- cbind(x,a1c.replaced)

```

2.5. Print mean a1c for each id.

```

id.means <- function(id)
{
  return(mean(subset.id(id)$a1c.replaced))
}

id <- seq(1:50)

indiv.means <- lapply(id,id.means)

```

```
(avg.a1c.by.id <- cbind(id,indiv.means))
```

```
##      id indiv.means
## [1,] 1  4.063372
## [2,] 2  7.544643
## [3,] 3  6.75764
## [4,] 4  3.892127
## [5,] 5  9.512311
## [6,] 6  7.555965
## [7,] 7  9.161686
## [8,] 8  7.189064
## [9,] 9  9.283873
## [10,] 10 7.975217
## [11,] 11 6.917562
## [12,] 12 7.034021
## [13,] 13 9.145282
## [14,] 14 6.623756
## [15,] 15 8.012406
## [16,] 16 4.222158
## [17,] 17 3.996034
## [18,] 18 9.164873
## [19,] 19 5.50721
## [20,] 20 3.726675
## [21,] 21 8.140939
## [22,] 22 5.637501
## [23,] 23 7.366889
## [24,] 24 7.439316
## [25,] 25 6.877135
## [26,] 26 6.556759
## [27,] 27 4.926457
## [28,] 28 7.433917
## [29,] 29 4.508086
## [30,] 30 6.045577
## [31,] 31 7.116586
## [32,] 32 6.568791
## [33,] 33 6.494069
## [34,] 34 6.768615
## [35,] 35 8.4767
## [36,] 36 9.60441
## [37,] 37 9.606253
## [38,] 38 5.355979
## [39,] 39 6.917013
## [40,] 40 9.530136
## [41,] 41 9.802424
## [42,] 42 3.89177
## [43,] 43 6.095849
## [44,] 44 9.09167
## [45,] 45 6.737204
## [46,] 46 9.621763
## [47,] 47 9.231489
## [48,] 48 6.4046
## [49,] 49 6.096076
## [50,] 50 8.962319
```

2.6. Print total number of visits for each id.

```
total.visits <- function(id)
{
  return(nrow(subset.id(id)))
}
id <- seq(1:50)
total.visits <- lapply(id,total.visits)
(total.visits.by.id <- cbind(id,total.visits))
```

```
##      id total.visits
## [1,] 1  11
## [2,] 2  20
## [3,] 3  14
## [4,] 4  12
## [5,] 5  14
## [6,] 6  10
## [7,] 7   9
## [8,] 8  12
## [9,] 9  11
## [10,] 10 12
## [11,] 11 10
## [12,] 12 10
## [13,] 13  8
## [14,] 14 12
## [15,] 15  8
## [16,] 16  9
## [17,] 17 12
## [18,] 18 10
## [19,] 19 10
## [20,] 20  9
## [21,] 21 10
## [22,] 22  8
## [23,] 23  8
## [24,] 24 15
## [25,] 25 12
## [26,] 26 14
## [27,] 27 11
## [28,] 28 14
## [29,] 29 10
## [30,] 30  7
## [31,] 31 11
## [32,] 32  5
## [33,] 33  8
## [34,] 34 12
## [35,] 35 11
## [36,] 36  9
## [37,] 37 17
## [38,] 38 15
## [39,] 39  8
## [40,] 40  7
## [41,] 41 17
## [42,] 42 14
## [43,] 43 11
```

```
## [44,] 44 11
## [45,] 45 14
## [46,] 46 9
## [47,] 47 12
## [48,] 48 11
## [49,] 49 12
## [50,] 50 10
```

2.7. Print the observations for id = 15.

```
subset.id(15)
```

```
##      id      dt      a1c gaps gaps2 visit a1c.replaced
## 11    15 2000-04-30 00:34:50 7.527105 262 262      1      7.527105
## 406   15 2001-01-17 21:11:02 5.898371  97  97      2      5.898371
## 306   15 2001-04-25 06:23:05 8.566593 772 365      3      8.566593
## 484   15 2002-04-25 06:23:05      NA  NA  407      4      8.012406
## 4841  15 2003-04-25 06:23:05      NA  NA  365      5      8.012406
## 48411 15 2003-06-06 14:06:00 9.133769 441 365      6      9.133769
## 263   15 2004-06-05 14:06:00      NA  NA   76      7      8.012406
## 2631  15 2004-08-20 17:47:11 8.936190  NA  NA   8      8.936190
```

Question 3

10 points

Import the addr.txt file from the GitHub repository. This file contains a listing of names and addresses (thanks google). Parse each line to create a data.frame with the following columns: lastname, firstname, streetno, streetname, city, state, zip. Keep middle initials or abbreviated names in the firstname column. Print out the entire data.frame.

```
addr <- read.delim("https://raw.githubusercontent.com/fonnesbeck/Bios6301/master/datasets/addr.txt",
  stringsAsFactors = FALSE, head = FALSE)

find.spaces <- function(textRow) {
  spaces <- c(unlist(gregexpr(" {2,}", textRow)), nchar(textRow))
  return(spaces) #return a vector of values where spaces of length 2+ are located in the textRow
}

trim <- function(x) gsub("^\\s+|\\s+$", "", x)
# credit:
# http://stackoverflow.com/questions/2261079/how-to-trim-leading-and-trailing-whitespace-in-r

findNumbers <- function(textRow) {
  numberPos <- NULL
  for (i in 1:nchar(textRow)) {
    if ((substr(textRow, i, i) %in% seq(0, 9)) == TRUE) {
      numberPos <- c(numberPos, i)
    }
  }
  return(numberPos)
}
```

```

lastname <- vector()
firstname <- vector()
streetno <- vector()
streetname <- vector()
city <- vector()
state <- vector()
zip <- vector()

fixText <- function(text) {
  for (i in 1:nrow(text)) {
    row.spaces <- find.spaces(text[i, ])
    lastname[i] <- substr(text[i, ], 1, (row.spaces[1] - 1))
    firstname[i] <- trim(substr(text[i, ], (row.spaces[1] + 1), (row.spaces[2] -
      1)))
    streetno.pos <- findNumbers(substr(text[i, ], (row.spaces[2] + 1), (row.spaces[3] -
      1))) #find indices of the numbers in the street address
# some addresses contain numerical street names, so streetno.pos can be
# longer than just the first 3-4 streetno values
    if (length(streetno.pos) > 4) {
      streetno[i] <- substr(substr(text[i, ], (row.spaces[2] + 1), (row.spaces[3] -
        1)), min(streetno.pos), (min(streetno.pos) + 3))
      streetname[i] <- trim(substr(substr(text[i, ], (row.spaces[2] +
        1), (row.spaces[3] - 1)), (min(streetno.pos) + 4), row.spaces[3] -
        1))
    } else {
      streetno[i] <- substr(substr(text[i, ], (row.spaces[2] + 1), (row.spaces[3] -
        1)), min(streetno.pos), max(streetno.pos))
      streetname[i] <- trim(substr(substr(text[i, ], (row.spaces[2] +
        1), (row.spaces[3] - 1)), (max(streetno.pos) + 1), row.spaces[3] -
        1))
    }

    city[i] <- trim(substr(text[i, ], row.spaces[3], row.spaces[4]))
    state[i] <- trim(substr(text[i, ], row.spaces[4], row.spaces[5]))
    zip[i] <- trim(substr(text[i, ], row.spaces[5], row.spaces[6]))
  }
  zip <- sub("0", "00", zip) #replace mistaken 0s in zipcodes with 0s
  return(cbind(lastname, firstname, streetno, streetname, city, state, zip))
}

(addr <- data.frame(fixText(addr)))

```

	lastname	firstname	streetno	streetname	city	state
## 1	Bania	Thomas M.	725	Commonwealth Ave.	Boston	MA
## 2	Barnaby	David	373	W. Geneva St.	Wms. Bay	WI
## 3	Bausch	Judy	373	W. Geneva St.	Wms. Bay	WI
## 4	Bolatto	Alberto	725	Commonwealth Ave.	Boston	MA
## 5	Carlstrom	John	933	E. 56th St.	Chicago	IL
## 6	Chamberlin	Richard A.	111	Nowelo St.	Hilo	HI
## 7	Chuss	Dave	2145	Sheridan Rd	Evanston	IL
## 8	Davis	E. J.	933	E. 56th St.	Chicago	IL
## 9	Depoy	Darren	174	W. 18th Ave.	Columbus	OH
## 10	Griffin	Greg	5000	Forbes Ave.	Pittsburgh	PA

## 11	Halvorsen	Nils	933	E. 56th St.	Chicago	IL
## 12	Harper	Al	373	W. Geneva St.	Wms. Bay	WI
## 13	Huang	Maohai	725	W. Commonwealth Ave.	Boston	MA
## 14	Ingalls	James G.	725	W. Commonwealth Ave.	Boston	MA
## 15	Jackson	James M.	725	W. Commonwealth Ave.	Boston	MA
## 16	Knudsen	Scott	373	W. Geneva St.	Wms. Bay	WI
## 17	Kovac	John	5640	S. Ellis Ave.	Chicago	IL
## 18	Landsberg	Randy	5640	S. Ellis Ave.	Chicago	IL
## 19	Lo	Kwok-Yung	1002	W. Green St.	Urbana	IL
## 20	Loewenstein	Robert F.	373	W. Geneva St.	Wms. Bay	WI
## 21	Lynch	John	4201	Wilson Blvd	Arlington	VA
## 22	Martini	Paul	174	W. 18th Ave.	Columbus	OH
## 23	Meyer	Stephan	933	E. 56th St.	Chicago	IL
## 24	Mrozek	Fred	373	W. Geneva St.	Wms. Bay	WI
## 25	Newcomb	Matt	5000	Forbes Ave.	Pittsburgh	PA
## 26	Novak	Giles	2145	Sheridan Rd	Evanston	IL
## 27	Odalen	Nancy	373	W. Geneva St.	Wms. Bay	WI
## 28	Pernic	Dave	373	W. Geneva St.	Wms. Bay	WI
## 29	Pernic	Bob	373	W. Geneva St.	Wms. Bay	WI
## 30	Peterson	Jeffrey	5000	Forbes Ave.	Pittsburgh	PA
## 31	Pryke	Clem	933	E. 56th St.	Chicago	IL
## 32	Rebull	Luisa	5640	S. Ellis Ave.	Chicago	IL
## 33	Renbarger	Thomas	2145	Sheridan Rd	Evanston	IL
## 34	Rottman	Joe	8730	W. Mountain View Ln	Littleton	CO
## 35	Schartman	Ethan	933	E. 56th St.	Chicago	IL
## 36	Spotz	Bob	373	W. Geneva St.	Wms. Bay	WI
## 37	Thoma	Mark	373	W. Geneva St.	Wms. Bay	WI
## 38	Walker	Chris	933	N. Cherry St.	Tucson	AZ
## 39	Wehrer	Cheryl	5000	Forbes Ave.	Pittsburgh	PA
## 40	Wirth	Jesse	373	W. Geneva St.	Wms. Bay	WI
## 41	Wright	Greg	791	Holmdel-Keyport Rd.	Holmdel	NY
## 42	Zingale	Michael	5640	S. Ellis Ave.	Chicago	IL
##	zip					
## 1	02215					
## 2	53191					
## 3	53191					
## 4	02215					
## 5	60637					
## 6	96720					
## 7	60208-3112					
## 8	60637					
## 9	43210					
## 10	15213					
## 11	60637					
## 12	53191					
## 13	02215					
## 14	02215					
## 15	02215					
## 16	53191					
## 17	60637					
## 18	60637					
## 19	61801					
## 20	53191					
## 21	22230					

```
## 22      43210
## 23      60637
## 24      53191
## 25      15213
## 26 60208-3112
## 27      53191
## 28      53191
## 29      53191
## 30      15213
## 31      60637
## 32      60637
## 33 60208-3112
## 34      80125
## 35      60637
## 36      53191
## 37      53191
## 38      85721
## 39      15213
## 40      53191
## 41 07733-1988
## 42      60637
```

Question 4

2 points

The first argument to most functions that fit linear models are formulas. The following example defines the response variable death and allows the model to incorporate all other variables as terms. . is used to mean all columns not otherwise in the formula.

```
url <- "https://github.com/fonnesbeck/Bios6301/raw/master/datasets/haart.csv"
haart_df <- read.csv(url)[,c('death','weight','hemoglobin','cd4baseline')]
coef(summary(glm(death ~ ., data=haart_df, family=binomial(logit))))
```

```
##              Estimate Std. Error  z value    Pr(>|z|)
## (Intercept)  3.576411744 1.226870535  2.915069 0.0035561039
## weight      -0.046210552 0.022556001 -2.048703 0.0404911395
## hemoglobin   -0.350642786 0.105064078 -3.337418 0.0008456055
## cd4baseline  0.002092582 0.001811959  1.154872 0.2481427160
```

Now imagine running the above several times, but with a different response and data set each time. Here's a function:

```
myfun <- function(dat, response) {
  form <- as.formula(response ~ .)
  coef(summary(glm(form, data=dat, family=binomial(logit))))
}
```

Unfortunately, it doesn't work. tryCatch is "catching" the error so that this file can be knit to PDF.

```
tryCatch(myfun(haart_df, death), error = function(e) e)
```

```
## <simpleError in eval(expr, envir, enclos): object 'death' not found>
```

What do you think is going on? Consider using debug to trace the problem.

The tryCatch error message reads: <simpleError in eval(expr, envir, enclos): object 'death' not found>, so I wondered if the problem could be in the way that the "response" variable was included in the parameters. To test my theory, I ran the line again with a minor correction:

```
tryCatch(myfun(haart_df, haart_df$death), error = function(e) e)
```

```
## Warning: glm.fit: algorithm did not converge
```

```
##              Estimate Std. Error      z value Pr(>|z|)
## (Intercept) -2.656607e+01 115935.1724 -2.291459e-04 0.9998172
## death       5.313214e+01  69028.4183  7.697140e-04 0.9993859
## weight     -4.499694e-15   1939.0571 -2.320558e-18 1.0000000
## hemoglobin  5.124642e-14   9774.8190  5.242697e-18 1.0000000
## cd4baseline 1.830771e-16    184.0846  9.945271e-19 1.0000000
```

Hooray! Now the function runs completely. Although, the algorithm does not converge, but that is more an artifact of the data than the code written above.

Bonus

5 bonus points

Create a working function.

```
myfun <- function(dat, response) {
  attach(dat, warn.conflicts = FALSE)
  response.name <- deparse(substitute(response))
  df.name <- deparse(substitute(dat))
  reponse.df <- paste(df.name, response.name, sep="$")
  formula <- paste(reponse.df, " ~ .", sep="")
  #by attaching the dataframe names we can call response without an identifier
  #form <- as.formula(response ~ .)
  print(coef(summary(glm(formula, data=dat, family=binomial(logit)))))
  detach(dat)
}
```

```
tryCatch(myfun(haart_df, death), error = function(e) e)
```

```
##              Estimate Std. Error      z value      Pr(>|z|)
## (Intercept)  3.576411744 1.226870535  2.915069 0.0035561039
## weight      -0.046210552 0.022556001 -2.048703 0.0404911395
## hemoglobin  -0.350642786 0.105064078 -3.337418 0.0008456055
## cd4baseline  0.002092582 0.001811959  1.154872 0.2481427160
```