# Bios 6301: Assignment 5

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Due Tuesday, 15 November, 1:00 PM  $5^{n=day}$  points taken off for each day late.

50 points total.

Submit a single knitr file (named homework5.rmd), along with a valid PDF output file. Inside the file, clearly indicate which parts of your responses go with which problems (you may use the original homework document as a template). Add your name as author to the file's metadata section. Raw R code/output or word processor files are not acceptable.

Failure to name file homework5.rmd or include author name may result in 5 points taken off.

## Question 1

#### 24 points

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

```
url1 <- "https://github.com/fonnesbeck/Bios6301/raw/master/datasets/haart.csv"
haart <- read.csv(url1,stringsAsFactors=FALSE)</pre>
```

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

```
haart$init.date <- as.Date(haart$init.date, format="%m/%d/%y")
haart$last.visit <- as.Date(haart$last.visit, format="%m/%d/%y")
haart$date.death <- as.Date(haart$date.death, format="%m/%d/%y")
table(format(haart$init.date,'%Y'))

##
```

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

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?

In this data, 92 observations died in year 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). If these times are longer than 1 year, censor them (this means if the value is above 365, set followup to 365). Print the quantile for this new variable.

```
for (i in 1:nrow(haart)){
  if(is.na(haart$date.death[i]) == TRUE ) {
    difference <- unclass(difftime(haart$last.visit[i], haart$init.date[i], 'days'))[1]
    haart$follow.up[i] <- min(365,difference)</pre>
  }
  else if(is.na(haart$date.death[i]) == FALSE && is.na(haart$last.visit[i])==TRUE){
    difference <- unclass(difftime(haart$date.death[i], haart$init.date[i], 'days'))[1]
    haart$follow.up[i] <- min(365,difference)</pre>
  }
  else {
    first <- min(haart$last.visit[i],haart$date.death[i])</pre>
    difference <- unclass(difftime(first, haart$init.date[i], 'days'))[1]</pre>
    haart$follow.up[i] <- min(365, difference)
  }
}
quantile(haart$follow.up)
```

```
## 0% 25% 50% 75% 100%
## 0.00 320.75 365.00 365.00 365.00
```

4. 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?

```
for (i in 1:nrow(haart)){
  if(is.na(haart$date.death[i]) && haart$follow.up[i] < 365){
    haart$lost[i] <- 1
  }
  else {
    haart$lost[i] <- 0
  }
}
lost.to <- sum(haart$lost, na.rm=TRUE)</pre>
```

There were 173 records lost-to-followup.

5. 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. Which drug regimen are found over 100 times?

```
init.reg <- as.character(haart[,'init.reg'])
haart[['init.reg_list']] <- strsplit(init.reg, ",")
(all_drugs <- unique(unlist(haart$init.reg_list)))</pre>
```

```
## [1] "3TC" "AZT" "EFV" "NVP" "D4T" "ABC" "DDI" "IDV" "LPV" "RTV" "SQV" ## [12] "FTC" "TDF" "DDC" "NFV" "T20" "ATV" "FPV"
```

```
(unique_drugs <- unique(unlist(haart$init.reg_list)))</pre>
## [1] "3TC" "AZT" "EFV" "NVP" "D4T" "ABC" "DDI" "IDV" "LPV" "RTV" "SQV"
## [12] "FTC" "TDF" "DDC" "NFV" "T20" "ATV" "FPV"
reg_drugs <- matrix(FALSE, nrow=nrow(haart), ncol=length(all_drugs))</pre>
for(i in seq_along(all_drugs)) {
  reg_drugs[,i] <- sapply(haart$init.reg_list, function(x) all_drugs[i] %in% x)
reg_drugs <- data.frame(reg_drugs)</pre>
names(reg_drugs) <- all_drugs</pre>
haart_merged <- cbind(haart, reg_drugs)</pre>
for (i in 17:34){
  for (j in 1:nrow(haart_merged)){
    if(haart_merged[j,i]==TRUE){
      haart_merged[j,i] <- colnames(haart_merged)[i]</pre>
    }
    else {
      haart_merged[j,i] <- NA
    }
 }
}
haart_merged$regimen <- NA
for(i in 1:nrow(haart_merged)){
  drugs <- vector()</pre>
  for(j in 17:34){
    if(is.na(haart_merged[i,j]) == FALSE){
      drugs <- c(drugs,haart_merged[i,j])</pre>
    }
  }
 haart_merged$regimen[i] <- paste(drugs, collapse = "-")</pre>
table(haart merged$regimen)[which(table(haart merged$regimen) > 100)]
##
## 3TC-AZT-EFV 3TC-AZT-NVP
                        284
##
           421
apply(X = haart_merged[,17:34],2,table)
## 3TC AZT EFV NVP D4T ABC DDI IDV LPV RTV SQV FTC TDF DDC NFV T20 ATV FPV
## 973 794 516 358 146 56 38 27
                                    31 79
                                             29
                                                    8
                                                      10
                                                            1
                                                                8
                                                                     1
                                                                         2
```

The individual drugs 3TC, AZT, EFV, NVP, and D4T are each found over 100 times, while the combined drug regimens 3TC-AZT-NVP and 3TC-AZT-EFV are both found over 100 times.

6. 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.

```
url2 <- "https://raw.githubusercontent.com/fonnesbeck/Bios6301/master/datasets/haart2.csv"
haart2 <- read.csv(url2, stringsAsFactors=FALSE)
haart2$init.date <- as.Date(haart2$init.date, format="%m/%d/%y")
haart2$last.visit <- as.Date(haart2$last.visit, format="%m/%d/%y")
haart2$date.death <- as.Date(haart2$date.death, format="%m/%d/%y")
for (i in 1:nrow(haart2)) {
 haart2$one.year[i] <- ifelse(abs(unclass(difftime(haart2$init.date[i], haart2$date.death[i],
                                                      units='days'))[1]) > 365, 0, 1)
}
for (i in 1:nrow(haart2)){
  if(is.na(haart2$date.death[i]) == TRUE) {
    difference <- unclass(difftime(haart2$last.visit[i], haart2$init.date[i], 'days'))[1]</pre>
    haart2$follow.up[i] <- min(365,difference)</pre>
  else {
    difference <- unclass(difftime(haart2$date.death[i], haart2$init.date[i], 'days'))[1]
    haart2$follow.up[i] <- min(365,difference)
}
for (i in 1:nrow(haart2)){
  if(is.na(haart2$date.death[i]) && unclass(difftime(haart2$last.visit[i],
                                                       haart2$init.date[i], 'days'))[1] < 365){
    haart2$lost[i] <- 1
  }
 else {
    haart2$lost[i] <- 0
  }
}
init.reg <- as.character(haart2[,'init.reg'])</pre>
haart2[['init.reg_list']] <- strsplit(init.reg, ",")</pre>
reg_drugs <- matrix(FALSE, nrow=nrow(haart2), ncol=length(all_drugs))</pre>
for(i in seq_along(all_drugs)) {
  reg_drugs[,i] <- sapply(haart2$init.reg_list, function(x) all_drugs[i] %in% x)
reg_drugs <- data.frame(reg_drugs)</pre>
names(reg_drugs) <- all_drugs</pre>
haart2_merged <- cbind(haart2, reg_drugs)
for (i in 17:34){
  for (j in 1:nrow(haart2_merged)){
    if(haart2 merged[j,i] == TRUE){
      haart2_merged[j,i] <- colnames(haart2_merged)[i]</pre>
    }
    else {
      haart2_merged[j,i] <- NA
 }
haart2_merged$regimen <- NA
for(i in 1:nrow(haart2_merged)){
 drugs <- vector()</pre>
```

```
for(j in 17:34){
    if(is.na(haart2_merged[i,j]) == FALSE){
      drugs <- c(drugs,haart2_merged[i,j])</pre>
  }
  haart2_merged$regimen[i] <- paste(drugs, collapse = "-")</pre>
haart_final <- rbind(haart_merged,haart2_merged)</pre>
haart_final[c(1:5,1000:1004),]
##
        male
                   age aids cd4baseline
                                             logvl weight hemoglobin
## 1
            1 25.00000
                           0
                                       NA
                                                NA
                                                         NA
                                                                     NA
## 2
            1 49.00000
                           0
                                      143
                                                NA 58.0608
                                                                     11
## 3
            1 42.00000
                                      102
                                                NA 48.0816
                           1
                                                                      1
## 4
                                      107
                                                NA 46.0000
           0 33.00000
                           0
                                                                     NA
## 5
            1 27.00000
                           0
                                       52 4.000000
                                                         NA
                                                                     NA
## 1000
           0 40.00000
                           1
                                     131
                                                NA 46.2672
                                                                      8
## 1001
           0 27.00000
                           0
                                      232
                                                NA
                                                         NA
                                                                     NA
## 1002
           1 38.72142
                           0
                                      170
                                                NA 84.0000
                                                                     NΑ
## 1003
           1 23.00000
                          NA
                                      154 3.995635 65.5000
                                                                     14
## 1004
           0 31.00000
                           0
                                      236
                                                NA 45.8136
                                                                     NA
##
            init.reg init.date last.visit death date.death one.year follow.up
## 1
        3TC, AZT, EFV 2003-07-01 2007-02-26
                                                 0
                                                          <NA>
                                                                      NA
                                                                                365
## 2
        3TC, AZT, EFV 2004-11-23 2008-02-22
                                                  0
                                                          <NA>
                                                                      NA
                                                                                365
                                                 1 2006-01-11
## 3
                                                                                365
        3TC, AZT, EFV 2003-04-30 2005-11-21
                                                                       0
## 4
        3TC, AZT, NVP 2006-03-25 2006-05-05
                                                  1 2006-05-07
                                                                                 41
                                                                       1
        3TC, D4T, EFV 2004-09-01 2007-11-13
## 5
                                                 0
                                                          < NA >
                                                                      NA
                                                                                365
## 1000 3TC,D4T,NVP 2003-07-03 2008-02-29
                                                 0
                                                          <NA>
                                                                      NA
                                                                                365
## 1001 3TC, AZT, NVP 2003-12-01 2004-01-05
                                                                      NA
                                                                                 35
                                                 0
                                                          <NA>
## 1002 3TC, AZT, NVP 2002-09-26 2004-03-29
                                                 0
                                                                      NA
                                                                                365
                                                           <NA>
## 1003 3TC,DDI,EFV 2007-01-31 2007-04-16
                                                                                 75
                                                  0
                                                           <NA>
                                                                      NA
## 1004 3TC,D4T,NVP 2003-12-03 2007-10-11
                                                  0
                                                           <NA>
                                                                      NA
                                                                                365
```

#### 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) {</pre>
    if(exists(".Random.seed", envir = .GlobalEnv)) {
        save.seed <- get(".Random.seed", envir= .GlobalEnv)</pre>
        on.exit(assign(".Random.seed", save.seed, envir = .GlobalEnv))
    } else {
        on.exit(rm(".Random.seed", envir = .GlobalEnv))
    }
    set.seed(n)
    subj <- ceiling(n / 10)</pre>
    id <- sample(subj, n, replace=TRUE)</pre>
    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')</pre>
    mu <- runif(subj, 4, 10)</pre>
    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)

1. Order the data set by id and dt.

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

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 alc value set to missing. A two year gap would require two new rows, and so forth.

```
#Write a function that finds gaps
check.dates <- function(identity,date){
  insert.at <- vector()
  rows.fin <- vector()
  for (i in unique(identity)){
    rows <- which(identity==i)[1:length(which(identity==i))-1]
    for (j in rows){
      rows.fin <- c(rows.fin, j)
      if(unclass(difftime(date[j+1], date[j], "days"))[1] > 366){
        insert.at <- c(insert.at,j+1)
      }
    }
}</pre>
```

```
return(insert.at)
}
#Write a function that fills gaps
add.gap <- function(df,insertion){</pre>
    df <- rbind(df[1:(insertion-1),],data.frame(id=df$id[insertion-1],</pre>
                                                  dt=df$dt[insertion-1]+years(1),a1c=NA),
                 df[insertion:nrow(df),])
  return(df)
}
p <- x
insert.at <- check.dates(p$id,p$dt)</pre>
lines <- insert.at+seq(from=0,by=1,length.out=length(insert.at))</pre>
for (i in 1:length(lines)){
  p <- add.gap(p,lines[i])</pre>
#Check again to fix 2-year gaps
insert.at <- check.dates(p$id,p$dt)</pre>
lines <- insert.at+seq(from=0,by=1,length.out=length(insert.at))</pre>
for (i in 1:length(lines)){
  p <- add.gap(p,lines[i])</pre>
#Check for any 3-year gaps
(insert.at <- check.dates(p$id,p$dt))</pre>
## logical(0)
x <- p
```

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.

```
for (i in 1:length(unique(x$id))){
  visits <- seq(1:table(x$id)[[i]])
  x$visit[x$id==i] <- visits
}</pre>
```

4. For each id, replace missing values with the mean alc value for that individual.

```
for (i in 1:length(unique(x$id))){
  rows <- which(x$id==i)
  meana1c <- mean(x$a1c[rows[1]:tail(rows,n=1)],na.rm = TRUE)
  for (j in rows){
    if(is.na(x$a1c[j])){
        x$a1c[j] <- meana1c
    }
  }
}</pre>
```

5. Print mean alc for each id.

```
for (i in 1:length(unique(x$id))){
  rows <- which(x$id==i)</pre>
  meana1c <- mean(x$a1c[rows[1]:tail(rows,n=1)])</pre>
  print(c(as.integer(i),meana1c))
}
## [1] 1.000000 4.063372
## [1] 2.000000 7.544643
## [1] 3.00000 6.75764
## [1] 4.000000 3.892127
## [1] 5.000000 9.512311
## [1] 6.000000 7.555965
## [1] 7.000000 9.161686
## [1] 8.000000 7.189064
## [1] 9.000000 9.283873
## [1] 10.000000 7.975217
## [1] 11.000000 6.917562
## [1] 12.000000
                  7.034021
## [1] 13.000000 9.145282
## [1] 14.000000 6.623756
## [1] 15.000000 8.012406
## [1] 16.000000
                  4.222158
## [1] 17.000000
                  3.996034
## [1] 18.000000
                  9.164873
## [1] 19.00000 5.50721
## [1] 20.000000
                  3.726675
## [1] 21.000000
                  8.140939
## [1] 22.000000
                  5.637501
## [1] 23.000000
                  7.366889
## [1] 24.000000
                  7.439316
## [1] 25.000000
                  6.877135
## [1] 26.000000
                  6.556759
## [1] 27.000000
                  4.926457
## [1] 28.000000
                  7.433917
## [1] 29.000000 4.508086
## [1] 30.000000
                  6.045577
## [1] 31.000000
                  7.116586
## [1] 32.000000
                  6.568791
## [1] 33.000000
                  6.494069
## [1] 34.000000
                  6.768615
## [1] 35.0000 8.4767
## [1] 36.00000 9.60441
## [1] 37.000000 9.606253
## [1] 38.000000 5.355979
## [1] 39.000000
                  6.917013
## [1] 40.000000 9.530136
## [1] 41.000000 9.802424
## [1] 42.00000 3.89177
## [1] 43.000000 6.095849
## [1] 44.00000 9.09167
## [1] 45.000000 6.737204
```

## [1] 46.000000 9.621763

```
## [1] 47.000000 9.231489
## [1] 48.0000 6.4046
## [1] 49.000000 6.096076
## [1] 50.000000 8.962319
```

6. Print total number of visits for each id.

```
table(x$id)
```

7. Print the observations for id = 15.

```
x[which(x$id==15),]
```

```
##
       id
                         dt
                                a1c visit
## 11
       15 2000-04-30 00:34:50 7.527105
## 306 15 2001-04-25 06:23:05 8.566593
                                       3
## 1117 15 2002-04-25 06:23:05 8.012406
                                       4
## 1154 15 2003-04-25 06:23:05 8.012406
                                       5
## 484 15 2003-06-06 14:06:00 9.133769
                                       6
## 1118 15 2004-06-06 14:06:00 8.012406
                                       7
## 263 15 2004-08-20 17:47:11 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.

```
\#Loop\ through\ each\ line\ and\ cut\ into\ parts,\ then\ trim\ trailing\ and\ leading\ whitespace
  for (j in 1:(length(cutpoints)-1)){
    fields[j] <- substring(addr[i,],cutpoints[j],cutpoints[j+1])</pre>
    fields[j] <- trim(fields[j])</pre>
  #Assign parts that don't need more splitting to the appropriate columns
  parsed.data[i,"lastname"] <- fields[1]</pre>
  parsed.data[i,"firstname"] <- fields[2]</pre>
  parsed.data[i,"city"] <- fields[4]</pre>
  parsed.data[i,"state"] <- fields[5]</pre>
  parsed.data[i,"zip"] <- fields[6]</pre>
  #Split street address into number and street name
  name.cut <- unlist(gregexpr("[[:alpha:]]", fields[3]))</pre>
  number <- substring(fields[3],1,name.cut[1]-1)</pre>
  parsed.data[i,"streetno"] <- trim(number)</pre>
  street <- substring(fields[3],name.cut[1],nchar(fields[3]))</pre>
  parsed.data[i,"streetname"] <- trim(street)</pre>
print(parsed.data)
```

##		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	${\tt Loewenstein}$	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.	${\tt Pittsburgh}$	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.	${\tt Pittsburgh}$	PΑ
##	31	Pryke	Clem	933		E. 56th St.	Chicago	IL
##	32	Rebull	Luisa	5640		S. Ellis Ave.	Chicago	IL

##		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				
##		53191				
##		15213				
##		60637				
##		60637				
		60208-3112				
##		80125				
##		60637				
##		53191				
##		53191				
##		85721				
##		15213				
##		53191				
		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))))</pre>
```

```
## 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(response ~ ., data=dat, family=binomial(logit))))
}</pre>
```

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 function is failing because it cannot find the object 'death' in the function call. This occurs in the third line when we find the coefficients of the summary of the glm of death ~ . using haart\_df. This is happening because the object "death" is defined inside of the data set haart\_df, but we are trying to call the variable from outside of the data set itself. However, trying to call haart\_df\$death results in the model not properly using death as the outcome variable.

```
myfun(haart_df, haart_df$death)
```

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

```
## (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
```

## 5 bonus points

Create a working function.

```
#We will use departs (substitute(x)) to convert the inputs into strings, create the formula manually, an
myfun.2 <- function(dat, response) {</pre>
 c <- deparse(substitute(response))</pre>
 d <- deparse(substitute(dat))</pre>
 e <- paste(d,c,sep="$")</pre>
 f <- paste(e, " ~ .",sep="")
 print(f)
  print(coef(summary(glm(f, data=dat, family=binomial(logit)))))
myfun.2(haart_df,death)
## [1] "haart_df$death ~ ."
                                                        Pr(>|z|)
##
                   Estimate Std. Error z value
## (Intercept) 3.576411744 1.226870535 2.915069 0.0035561039
              -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