

# Bios 6301: Assignment 5

*Elizabeth Sigworth*

*Tuesday 15 November*

*Due Tuesday, 15 November, 1:00 PM*

$5^{n=\text{day}}$  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)
```

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?

```
for (i in 1:nrow(haart)) {
  haart$one.year[i] <- ifelse(abs(unclass(difftime(haart$init.date[i],
                                                    haart$date.death[i], units='days'))[1]) > 365, 0, 1)
}
sum(haart$one.year, na.rm = TRUE)
```

```
## [1] 92
```

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)
  }
  else {
    difference <- unclass(difftime(haart$date.death[i], haart$init.date[i], 'days'))[1]
    haart$follow.up[i] <- min(365,difference)
  }
}
quantile(haart$follow.up)
```

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

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$date.death[i]) && unclass(difftime(haart$last.visit[i],
                                                    haart$init.date[i], 'days'))[1] < 365){
    haart$lost[i] <- 1
  }
  else {
    haart$lost[i] <- 0
  }
}
sum(haart$lost, na.rm=TRUE)
```

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

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)))
```

```
## [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)))

## [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))
for(i in seq_along(all_drugs)) {
  reg_drugs[,i] <- sapply(haart$init.reg_list, function(x) all_drugs[i] %in% x)
}
reg_drugs <- data.frame(reg_drugs)
names(reg_drugs) <- all_drugs
haart_merged <- cbind(haart, reg_drugs)
for (i in 17:34){
  for (j in 1:nrow(haart_merged)){
    if(haart_merged[j,i]==TRUE){
      haart_merged[j,i] <- colnames(haart_merged)[i]
    }
    else {
      haart_merged[j,i] <- NA
    }
  }
}
haart_merged$regimen <- NA
for(i in 1:nrow(haart_merged)){
  drugs <- vector()
  for(j in 17:34){
    if(is.na(haart_merged[i,j])) == FALSE){
      drugs <- c(drugs,haart_merged[i,j])
    }
  }
  haart_merged$regimen[i] <- paste(drugs, collapse = "-")
}
table(haart_merged$regimen)[which(table(haart_merged$regimen) > 100)]

##
## 3TC-AZT-EFV 3TC-AZT-NVP
##          421          284

```

The 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]
      haart2$follow.up[i] <- min(365,difference)
    }
    else {
      difference <- unclass(difftime(haart2$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'])
haart2[['init.reg_list']] <- strsplit(init.reg, ",")

reg_drugs <- matrix(FALSE, nrow=nrow(haart2), ncol=length(all_drugs))
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)
names(reg_drugs) <- all_drugs
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]
    }
    else {
      haart2_merged[j,i] <- NA
    }
  }
}

haart2_merged$regimen <- NA
for(i in 1:nrow(haart2_merged)){
  drugs <- vector()
  for(j in 17:34){
    if(is.na(haart2_merged[i,j]) == FALSE){
      drugs <- c(drugs,haart2_merged[i,j])
    }
  }
}

```

```

  haart2_merged$regimen[i] <- paste(drugs, collapse = "-")
}

```

```

haart_final <- rbind(haart_merged,haart2_merged)
haart_final[c(1:5,1000:1004),]

```

##	male	age	aids	cd4baseline	logv1	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	1	102	NA	48.0816	1					
## 4	0	33.00000	0	107	NA	46.0000	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	NA					
## 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					
## 3	3TC,AZT,EFV	2003-04-30	2005-11-21	1	2006-01-11	0	365					
## 4	3TC,AZT,NVP	2006-03-25	2006-05-05	1	2006-05-07	1	43					
## 5	3TC,D4T,EFV	2004-09-01	2007-11-13	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	0	<NA>	NA	35					
## 1002	3TC,AZT,NVP	2002-09-26	2004-03-29	0	<NA>	NA	365					
## 1003	3TC,DDI,EFV	2007-01-31	2007-04-16	0	<NA>	NA	75					
## 1004	3TC,D4T,NVP	2003-12-03	2007-10-11	0	<NA>	NA	365					
##	lost	init.reg_list	3TC	AZT	EFV	NVP	D4T	ABC	DDI	IDV	LPV	RTV
## 1	0	3TC, AZT, EFV	3TC	AZT	EFV	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>
## 2	0	3TC, AZT, EFV	3TC	AZT	EFV	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>
## 3	0	3TC, AZT, EFV	3TC	AZT	EFV	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>
## 4	0	3TC, AZT, NVP	3TC	AZT	<NA>	NVP	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>
## 5	0	3TC, D4T, EFV	3TC	<NA>	EFV	<NA>	D4T	<NA>	<NA>	<NA>	<NA>	<NA>
## 1000	0	3TC, D4T, NVP	3TC	<NA>	<NA>	NVP	D4T	<NA>	<NA>	<NA>	<NA>	<NA>
## 1001	1	3TC, AZT, NVP	3TC	AZT	<NA>	NVP	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>
## 1002	0	3TC, AZT, NVP	3TC	AZT	<NA>	NVP	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>
## 1003	1	3TC, DDI, EFV	3TC	<NA>	EFV	<NA>	<NA>	<NA>	DDI	<NA>	<NA>	<NA>
## 1004	0	3TC, D4T, NVP	3TC	<NA>	<NA>	NVP	D4T	<NA>	<NA>	<NA>	<NA>	<NA>
##	SQV	FTC	TDF	DDC	NFV	T20	ATV	FPV	regimen			
## 1	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	3TC-AZT-EFV			
## 2	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	3TC-AZT-EFV			
## 3	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	3TC-AZT-EFV			
## 4	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	3TC-AZT-NVP			
## 5	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	3TC-EFV-D4T			
## 1000	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	3TC-NVP-D4T			
## 1001	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	3TC-AZT-NVP			
## 1002	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	3TC-AZT-NVP			
## 1003	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	3TC-EFV-DDI			
## 1004	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>	3TC-NVP-D4T			

## 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)

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 a1c 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)
      }
    }
  }
  return(insert.at)
}
```

```
#Write a function that fills gaps
add.gap <- function(df,insertion){
```

```

df <- rbind(df[1:(insertion-1),],data.frame(id=df$id[insertion-1],
                                             dt=df$dt[insertion-1]+years(1),a1c=NA),
            df[insertion:nrow(df),])
return(df)
}

p <- x
insert.at <- check.dates(p$id,p$dt)
lines <- insert.at+seq(from=0,by=1,length.out=length(insert.at))
for (i in 1:length(lines)){
  p <- add.gap(p,lines[i])
}

#Check again to fix 2-year gaps
insert.at <- check.dates(p$id,p$dt)
lines <- insert.at+seq(from=0,by=1,length.out=length(insert.at))
for (i in 1:length(lines)){
  p <- add.gap(p,lines[i])
}

#Check for any 3-year gaps
(insert.at <- check.dates(p$id,p$dt))

```

```
## 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 `a1c` values.

```

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

```

4. For each `id`, replace missing values with the mean `a1c` 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
    }
  }
}

```

5. Print mean `a1c` for each `id`.

```

for (i in 1:length(unique(x$id))){
  rows <- which(x$id==i)
  meana1c <- mean(x$a1c[rows[1]:tail(rows,n=1)])
  print(c(as.integer(i),meana1c))
}

```

```

## [1] 1.000000 4.063372
## [1] 2.000000 7.544643
## [1] 3.000000 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)
```

```
##
##  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25
## 11 20 14 12 14 10  9 12 11 12 10 10  8 12  8  9 12 10 10  9 10  8  8 15 12
## 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49 50
## 14 11 14 10  7 11  5  8 12 11  9 17 15  8  7 17 14 11 11 14  9 12 11 12 10
```

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    1
## 406  15 2001-01-17 21:11:02 5.898371    2
## 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    8
```

### 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.

```
url3 <- "https://raw.githubusercontent.com/fonnesbeck/Bios6301/master/datasets/addr.txt"
addr <- read.delim(url3, header=FALSE, stringsAsFactors=FALSE)
parsed.data <- data.frame(lastname=rep(NA,nrow(addr)),firstname=rep(NA,nrow(addr)),
                          streetno=rep(NA,nrow(addr)),streetname=rep(NA,nrow(addr)),
                          city=rep(NA,nrow(addr)),state=rep(NA,nrow(addr)),zip=rep(NA,nrow(addr)))

#Write function to trim leading and trailing whitespace to use after splitting strings into sections
trim <- function(x) gsub("^\\s+|\\s+$", "", x)

#Loop through addr file to parse into fields
for(i in 1:nrow(addr)){
  #Identify whitespace of 2 spaces or more
  cutpoints <- c(1,unlist(gregexpr(" {2,}", addr[i,])),nchar(addr[i,]))
  fields <- vector()
  #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])
  fields[j] <- trim(fields[j])
}
#Assign parts that don't need more splitting to the appropriate columns
parsed.data[i,"lastname"] <- fields[1]
parsed.data[i,"firstname"] <- fields[2]
parsed.data[i,"city"] <- fields[4]
parsed.data[i,"state"] <- fields[5]
parsed.data[i,"zip"] <- fields[6]
#Split street address into number and street name
name.cut <- unlist(gregexpr("[:alpha:]", fields[3]))
number <- substring(fields[3],1,name.cut[1]-1)
parsed.data[i,"streetno"] <- trim(number)
street <- substring(fields[3],name.cut[1],nchar(fields[3]))
parsed.data[i,"streetname"] <- trim(street)
}
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	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(response ~ ., 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 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
```

```
##              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
```

### 5 bonus points

Create a working function.

*#We will use `deparse(substitute(x))` to convert the inputs into strings, create the formula manually, and*

```
myfun.2 <- function(dat, response) {  
  c <- deparse(substitute(response))  
  d <- deparse(substitute(dat))  
  e <- paste(d,c,sep="$")  
  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 ~ ."  
##               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
```