

Bios 6301: Assignment 4

Josh DeClercq

Grade 48/50

Due Tuesday, 01 November, 1:00 PM

$5^{n=\text{day}}$ points taken off for each day late.

50 points total.

Submit a single knitr file (named `homework4.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 `homework4.rmd` or include author name may result in 5 points taken off.

Question 1

15 points

A problem with the Newton-Raphson algorithm is that it needs the derivative f' . If the derivative is hard to compute or does not exist, then we can use the *secant method*, which only requires that the function f is continuous.

Like the Newton-Raphson method, the **secant method** is based on a linear approximation to the function f . Suppose that f has a root at a . For this method we assume that we have *two* current guesses, x_0 and x_1 , for the value of a . We will think of x_0 as an older guess and we want to replace the pair x_0, x_1 by the pair x_1, x_2 , where x_2 is a new guess.

To find a good new guess x_2 we first draw the straight line from $(x_0, f(x_0))$ to $(x_1, f(x_1))$, which is called a secant of the curve $y = f(x)$. Like the tangent, the secant is a linear approximation of the behavior of $y = f(x)$, in the region of the points x_0 and x_1 . As the new guess we will use the x -coordinate x_2 of the point at which the secant crosses the x -axis.

The general form of the recurrence equation for the secant method is:

$$x_{i+1} = x_i - f(x_i) \frac{x_i - x_{i-1}}{f(x_i) - f(x_{i-1})}$$

Notice that we no longer need to know f' but in return we have to provide *two* initial points, x_0 and x_1 .

Write a function that implements the secant algorithm. Validate your program by finding the root of the function $f(x) = \cos(x) - x$. Compare its performance with the Newton-Raphson method – which is faster, and by how much? For this example $f'(x) = -\sin(x) - 1$.

```
secant <- function (guess1, guess2, f, tol = 10e-7, maxiter = 10000)
{
  i <- 1
  while (abs(f(guess1)) > tol && i < maxiter){
    guess3 <- guess2 - f(guess2)*((guess2 - guess1)/(f(guess2)-f(guess1)))
    guess1 <- guess2
    guess2 <- guess3
    i <- i + 1
  }
}
```

```

    if(i == maxiter) {
      print('failed to converge')
      return (NULL)
    }else{
      print(sprintf('converges at %s', i))
    }
    guess1
  }
}

newton <- function(guess,f, fp, tol = 10e-7, maxiter = 10000){
  i <- 1
  while (abs(f(guess)) > tol && i < maxiter){
    guess <- guess - f(guess)/fp(guess)
    i <- i + 1
  }
  if(i == maxiter) {
    print('failed to converge')
    return (NULL)
  }else{
    print(sprintf('converges at %s', i))
  }
  guess
}

```

```

f <- function(x) cos(x) - x
fp <- function(x) -sin(x) - 1
secant(-10,10, f)

```

```
## [1] "converges at 9"
```

```
## [1] 0.7390851
```

```
newton(10,f,fp)
```

```
## [1] "converges at 49"
```

```
## [1] 0.7390852
```

```

# time1 <- system.time(replicate(10000,newton(10,f,fp)))
# time2 <- system.time(replicate(10000,secant(-10,10, f)))

```

```

library(tictoc)
tic()
invisible(capture.output(replicate(10000,newton(10,f,fp))))
newton_time <-toc()

```

```
## 3.149 sec elapsed
```

```
newton_time <- newton_time$toc - newton_time$tic
```

```

tic()
invisible(capture.output(replicate(10000,secant(-10,10, f))))
secant_time <-toc()

```

```
## 1.659 sec elapsed
```

```
secant_time <- secant_time$toc - secant_time$tic
```

```
result <- newton_time - secant_time
ratio <- round(newton_time/secant_time,2)
```

The secant method is 1.49 seconds faster than the newton method, when both are simulated 10,000 times. The secant method is 1.9 times faster than the newton method.

Question 2

18 points

The game of craps is played as follows. First, you roll two six-sided dice; let x be the sum of the dice on the first roll. If $x = 7$ or 11 you win, otherwise you keep rolling until either you get x again, in which case you also win, or until you get a 7 or 11 , in which case you lose.

Write a program to simulate a game of craps. You can use the following snippet of code to simulate the roll of two (fair) dice:

```
x <- sum(ceiling(6*runif(2)))
```

1. The instructor should be able to easily import and run your program (function), and obtain output that clearly shows how the game progressed. Set the RNG seed with `set.seed(100)` and show the output of three games. (lucky 13 points)

```
set.seed(100)
craps <- function(){
  win <- 2
  i <- 1
  x <- sum(ceiling(6*runif(2)))
  rolls <- x
  if (x == 7 | x == 11){
    i <- 0
    win <- 1
  }else{point <- x}

  while(i != 0){
    x <- sum(ceiling(6*runif(2)))
    rolls[i+1] <- x
    if(x == 7 | x == 11){
      i <- 0
      win <- 0
    }else if(x == point){
      i <- 0
      win <- 1
    }else {i <- i+1}
  }

  if (win == 1) {call <- "You Win!"}
  } else {call <- "You lose!"}

  result <- list("rolls" = rolls, "outcome" = call)
  return(result)
}
craps()

## $rolls
```

```
## [1] 4 5 6 8 6 10 5 10 5 8 9 9 5 11
##
## $outcome
## [1] "You lose!"
```

```
craps()
```

```
## $rolls
## [1] 6 9 9 11
##
## $outcome
## [1] "You lose!"
```

```
craps()
```

```
## $rolls
## [1] 6 7
##
## $outcome
## [1] "You lose!"
```

1. Find a seed that will win ten straight games. Consider adding an argument to your function that disables output. Show the output of the ten games. (5 points)

```
TEN <- FALSE
i <- 1
while (TEN == FALSE){
  set.seed(i)
  if (sum(invisible(replicate(10, craps())$outcome == "You Win!"))) == 10)
  {
    TEN <- TRUE
    print(i)
  }else{i <- i+1}
}
```

```
## [1] 880
```

```
set.seed(i) #The first seed that works is 880.
x <- list("game" = replicate(10, craps())$rolls)
x
```

```
## $game
## $game[[1]]
## [1] 7
##
## $game[[2]]
## [1] 8 9 3 10 6 8
##
## $game[[3]]
## [1] 10 10
##
## $game[[4]]
## [1] 9 9
##
## $game[[5]]
## [1] 11
##
## $game[[6]]
```

```
## [1] 8 8
##
## $game[[7]]
## [1] 5 5
##
## $game[[8]]
## [1] 7
##
## $game[[9]]
## [1] 9 9
##
## $game[[10]]
## [1] 7
```

Question 3

12 points

Obtain a copy of the football-values lecture. Save the five 2016 CSV files in your working directory.

Modify the code to create a function. This function will create dollar values given information (as arguments) about a league setup. It will return a data.frame and write this data.frame to a CSV file. The final data.frame should contain the columns 'PlayerName', 'pos', 'points', 'value' and be ordered by value descendingly. Do not round dollar values.

Note that the returned data.frame should have $\text{sum}(\text{posReq}) * \text{nTeams}$ rows.

Define the function as such (6 points):

```
# path: directory path to input files
# file: name of the output file; it should be written to path
# nTeams: number of teams in league
# cap: money available to each team
# posReq: number of starters for each position
# points: point allocation for each category
ffvalues <- function(path, file='outfile.csv', nTeams=12, cap=200,
                     posReq=c(qb=1, rb=2, wr=3, te=1, k=1),
                     points=c(fg=4, xpt=1, pass_yds=1/25, pass_tds=4, pass_ints=-2,
                               rush_yds=1/10, rush_tds=6, fumbles=-2, rec_yds=1/20,
                               rec_tds=6)) {

  #read in csv files
  positions <- c('k','qb','rb','te','wr')
  year <- 2016
  csvfile <- paste('proj_', positions, substr(year, 3, 4), '.csv', sep='')
  files <- file.path(year, csvfile)
  names(files) <- positions
  k <- read.csv(files['k'], header=TRUE, stringsAsFactors=FALSE)
  qb <- read.csv(files['qb'], header=TRUE, stringsAsFactors=FALSE)
  rb <- read.csv(files['rb'], header=TRUE, stringsAsFactors=FALSE)
  te <- read.csv(files['te'], header=TRUE, stringsAsFactors=FALSE)
  wr <- read.csv(files['wr'], header=TRUE, stringsAsFactors=FALSE)

  # generate unique list of column names
  cols <- unique(c(names(k), names(qb), names(rb), names(te), names(wr)))

  # create a new column in each data.frame
```

```

# values are recycled
k[, 'pos'] <- 'k'
qb[, 'pos'] <- 'qb'
rb[, 'pos'] <- 'rb'
te[, 'pos'] <- 'te'
wr[, 'pos'] <- 'wr'

# append 'pos' to unique column list
cols <- c(cols, 'pos')

# create common columns in each data.frame
# initialize values to zero
k[, setdiff(cols, names(k))] <- 0
qb[, setdiff(cols, names(qb))] <- 0
rb[, setdiff(cols, names(rb))] <- 0
te[, setdiff(cols, names(te))] <- 0
wr[, setdiff(cols, names(wr))] <- 0

# combine data.frames by row, using consistent column order
x <- rbind(k[, cols], qb[, cols], rb[, cols], te[, cols], wr[, cols])

## calculate dollar values
x[, 'p_fg'] <- x[, 'fg'] * points['fg']
x[, 'p_xpt'] <- x[, 'xpt'] * points['xpt']
x[, 'p_pass_yds'] <- x[, 'pass_yds'] * points['pass_yds']
x[, 'p_pass_tds'] <- x[, 'pass_tds'] * points['pass_tds']
x[, 'p_pass_ints'] <- x[, 'pass_ints'] * points['pass_ints']
x[, 'p_rush_yds'] <- x[, 'rush_yds'] * points['rush_yds']
x[, 'p_rush_tds'] <- x[, 'rush_tds'] * points['rush_tds']
x[, 'p_fumbles'] <- x[, 'fumbles'] * points['fumbles']
x[, 'p_rec_yds'] <- x[, 'rec_yds'] * points['rec_yds']
x[, 'p_rec_tds'] <- x[, 'rec_tds'] * points['rec_tds']

# sum selected column values for every row
x[, 'points'] <- rowSums(x[, grep("^p_", names(x))])

# create new data.frame ordered by points descendingly
x2 <- x[order(x[, 'points'], decreasing=TRUE),]

# determine the row indices for each position
k.ix <- which(x2[, 'pos'] == 'k')
qb.ix <- which(x2[, 'pos'] == 'qb')
rb.ix <- which(x2[, 'pos'] == 'rb')
te.ix <- which(x2[, 'pos'] == 'te')
wr.ix <- which(x2[, 'pos'] == 'wr')

# calculate marginal points by subtracting "baseline" player's points
x2[k.ix, 'marg'] <- x2[k.ix, 'points'] - x2[k.ix[nTeams], 'points']
x2[qb.ix, 'marg'] <- x2[qb.ix, 'points'] - x2[qb.ix[nTeams], 'points']
x2[rb.ix, 'marg'] <- x2[rb.ix, 'points'] - x2[rb.ix[2*nTeams], 'points']
x2[te.ix, 'marg'] <- x2[te.ix, 'points'] - x2[te.ix[nTeams], 'points']
x2[wr.ix, 'marg'] <- x2[wr.ix, 'points'] - x2[wr.ix[3*nTeams], 'points']

```

```

# create a new data.frame subset by non-negative marginal points
x3 <- x2[x2[, 'marg'] >= 0,]

# re-order by marginal points
x3 <- x3[order(x3[, 'marg'], decreasing=TRUE),]

# reset the row names
rownames(x3) <- NULL

# calculation for player value
x3[, 'value'] <- x3[, 'marg']*(nTeams*cap-nrow(x3))/sum(x3[, 'marg']) + 1

# create a data.frame with more interesting columns
x4 <- x3[,c('PlayerName', 'pos', 'points', 'value')]
## save dollar values as CSV file
write.csv(x4, file)
## return data.frame with dollar values
return(x4)
}

```

1. Call `x1 <- ffvalues('.',.)`

```
x1 <- ffvalues('.',.)
```

1. How many players are worth more than \$20? (1 point)

```
nrow(x1[x1[, 'value'] > 20,])
```

```
## [1] 46
```

1. Who is 15th most valuable running back (rb)? (1 point)

```
x1$PlayerName[x1$pos == 'rb'][15]
```

```
## [1] "Carlos Hyde"
```

1. Call `x2 <- ffvalues(getwd(), '16team.csv', nTeams=16, cap=150)`

```
x2 <- ffvalues(getwd(), '16team.csv', nTeams=16, cap=150)
```

1. How many players are worth more than \$20? (1 point)

```
nrow(x2[x2[, 'value'] > 20,])
```

```
## [1] 49
```

1. How many wide receivers (wr) are in the top 40? (1 point)

```
nrow(na.exclude(x2[1:40,][x2$pos == 'wr',]))
```

```
## [1] 18
```

1. Call:

```

x3 <- ffvalues('.', 'qbheavy.csv', posReq=c(qb=2, rb=2, wr=3, te=1, k=0),
           points=c(fg=0, xpt=0, pass_yds=1/25, pass_tds=6, pass_ints=-2,
                    rush_yds=1/10, rush_tds=6, fumbles=-2, rec_yds=1/20, rec_tds=6))

```

1. How many players are worth more than \$20? (1 point)

```
nrow(x3[x3[, 'value'] > 20,])
```

```
## [1] 46
```

JC Grading -1

Should be 51. Look at where comment is: # calculate marginal points by subtracting “baseline” player’s points. Here you hard code the number required at each position rather than using the posReq function input.

```
x3[x3$value>20,]
```

##	PlayerName	pos	points	value
## 1	Aaron Rodgers	qb	395.586	88.48827
## 2	Adrian Peterson	rb	201.420	77.60832
## 3	Cam Newton	qb	384.832	77.36949
## 4	Antonio Brown	wr	146.740	75.24582
## 5	Todd Gurley	rb	198.005	74.07749
## 6	Russell Wilson	qb	374.956	67.15849
## 7	Odell Beckham Jr.	wr	138.855	67.09335
## 8	Jamaal Charles	rb	190.290	66.10079
## 9	Julio Jones	wr	132.535	60.55898
## 10	David Johnson	rb	183.780	59.36997
## 11	Andrew Luck	qb	366.980	58.91194
## 12	Lamar Miller	rb	181.915	57.44171
## 13	Rob Gronkowski	te	114.950	57.14704
## 14	DeAndre Hopkins	wr	128.560	56.44915
## 15	Ezekiel Elliott	rb	172.825	48.04337
## 16	Dez Bryant	wr	120.155	47.75905
## 17	Drew Brees	qb	356.022	47.58224
## 18	Le'Veon Bell	rb	170.180	45.30865
## 19	A.J. Green	wr	116.775	44.26439
## 20	Jordy Nelson	wr	114.615	42.03113
## 21	Eddie Lacy	rb	164.770	39.71515
## 22	LeSean McCoy	rb	164.065	38.98623
## 23	Devonta Freeman	rb	162.295	37.15619
## 24	Doug Martin	rb	161.830	36.67542
## 25	Allen Robinson	wr	109.280	36.51516
## 26	Alshon Jeffery	wr	108.915	36.13778
## 27	Mark Ingram	rb	160.260	35.05216
## 28	Thomas Rawls	rb	158.400	33.12907
## 29	Brandon Marshall	wr	105.195	32.29160
## 30	Sammy Watkins	wr	104.380	31.44895
## 31	Ben Roethlisberger	qb	340.088	31.10776
## 32	Mike Evans	wr	103.930	30.98369
## 33	T.Y. Hilton	wr	103.875	30.92682
## 34	Brandin Cooks	wr	103.055	30.07901
## 35	Randall Cobb	wr	100.625	27.56658
## 36	Greg Olsen	te	86.065	27.28225
## 37	Jordan Reed	te	84.965	26.14494
## 38	Demaryius Thomas	wr	98.515	25.38501
## 39	Carson Palmer	qb	334.408	25.23509
## 40	Keenan Allen	wr	98.330	25.19373
## 41	C.J. Anderson	rb	150.680	25.14721
## 42	Amari Cooper	wr	97.015	23.83413
## 43	Doug Baldwin	wr	95.070	21.82315
## 44	Kelvin Benjamin	wr	94.980	21.73010
## 45	Eric Decker	wr	94.740	21.48196


```
## 46      Jeremy Maclin  wr  94.075 20.79440
1.  How many quarterbacks (qb) are in the top 30? (1 point)
nrow(na.exclude(x3[1:30,][x3$pos == 'qb',]))
```

```
## [1] 5
```

JC Grading -1

Should be 10. See comment above.

Question 4

5 points

This code makes a list of all functions in the base package:

```
objs <- mget(ls("package:base"), inherits = TRUE)
funs <- Filter(is.function, objs)
```

Using this list, write code to answer these questions.

1. Which function has the most arguments? (3 points)

```
arg_length <- sapply(funs, function(x) length(formals(x)))
max_arg <- which(arg_length == max(arg_length))
names(funs[max_arg])
```

```
## [1] "scan"
```

1. How many functions have no arguments? (2 points)

```
length(which(arg_length == 0))
```

```
## [1] 225
```

Hint: find a function that returns the arguments for a given function.