Bios 6301: Assignment 6

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Due Tuesday, 25 October, 1:00 PM

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

40 points total.

Submit a single knitr file (named homework6.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 homework6.rmd or include author name may result in 5 points taken off.

Question 1

16 points

Obtain a copy of the football-values lecture (https://github.com/couthcommander/football-values). Save the five 2021 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 orderd by value descendingly. Do not round dollar values.

Note that the returned data.frame should have sum(posReq)*nTeams rows.

Define the function as such (10 points):

```
## ---- load package & set wd
# assumed that the current working directory (and the location that this function will be calle
d) is in the 2021 year folder of football-values-main
library(tidyverse)
```

```
## - Attaching packages -
                                                            – tidyverse 1.3.2 —
## √ ggplot2 3.3.6
                      √ purrr
                                0.3.4
## √ tibble 3.1.8

√ dplyr

                                1.0.9
## √ tidyr 1.2.0
                      ✓ stringr 1.4.1
## √ readr
            2.1.2
                      ✓ forcats 0.5.2
## -- Conflicts ----
                                                      – tidyverse conflicts() —
## X dplyr::filter() masks stats::filter()
## X dplyr::lag()
                    masks stats::lag()
```

```
## ---- define the function
# 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)) {
  ## ---- 1. read in CSV files
  year <- 2021
  positions <- c('k','qb','rb','te','wr')</pre>
  files <- paste0(path,'/','proj_',positions,substr(year,3,4),'.csv')</pre>
  names(files) <- positions</pre>
  # read csv from each position
  k <- read.csv(files['k'], header=TRUE, stringsAsFactors=FALSE)</pre>
  qb <- read.csv(files['qb'], stringsAsFactors=FALSE)</pre>
  rb <- read.csv(files['rb'])</pre>
  te <- read.csv(files['te'])</pre>
  wr <- read.csv(files['wr'])</pre>
  # make a column name for integrated df
  cols <- unique(c(names(k), names(qb), names(rb), names(te), names(wr)))</pre>
  k[,'pos'] <- 'k'
  qb[,'pos'] <- 'qb'
  rb[,'pos'] <- 'rb'
  te[,'pos'] <- 'te'
  wr[,'pos'] <- 'wr'
  cols <- c(cols, 'pos')</pre>
  k[,setdiff(cols, names(k))] <- 0
  qb[,setdiff(cols, names(qb))] <- 0</pre>
  rb[,setdiff(cols, names(rb))] <- 0</pre>
  te[,setdiff(cols, names(te))] <- 0</pre>
  wr[,setdiff(cols, names(wr))] <- 0</pre>
  x <- rbind(k[,cols], qb[,cols], rb[,cols], te[,cols], wr[,cols])</pre>
  ## calculate dollar values
  x[,'p fg'] \leftarrow x[,'fg']*points['fg']
  x[,'p_xpt'] <- x[,'xpt']*points['xpt']
  x[,'p_pass_yds'] <- x[,'pass_yds']*points['pass_yds']</pre>
  x[,'p_pass_tds'] <- x[,'pass_tds']*points['pass_tds']</pre>
```

```
x[,'p_pass_ints'] <- x[,'pass_ints']*points['pass_ints']</pre>
x[,'p_rush_yds'] <- x[,'rush_yds']*points['rush_yds']</pre>
x[,'p_rush_tds'] <- x[,'rush_tds']*points['rush_tds']</pre>
x[,'p_fumbles'] <- x[,'fumbles']*points['fumbles']</pre>
x[,'p_rec_yds'] <- x[,'rec_yds']*points['rec_yds']</pre>
x[,'p rec tds'] <- x[,'rec tds']*points['rec tds']</pre>
# calculate the point
x[,'points'] <- rowSums(x[,grep("^p_", names(x))])</pre>
x2 <- x[order(x[,'points'], decreasing=TRUE),]</pre>
rownames(x2) <-NULL
# get an index for each position
k.idx <- which(x2[,'pos']=='k')
qb.idx <- which(x2[,'pos']=='qb')</pre>
rb.idx <- which(x2[,'pos']=='rb')
te.idx <- which(x2[,'pos']=='te')</pre>
wr.idx <- which(x2[,'pos']=='wr')</pre>
# calculate marginal values
n_req.k <- posReq['k']*nTeams</pre>
n_req.qb <- posReq['qb']*nTeams</pre>
n_req.rb <- posReq['rb']*nTeams</pre>
n_req.te <- posReq['te']*nTeams</pre>
n_req.wr <- posReq['wr']*nTeams</pre>
# If the number of required position is zero, make marginal values -Inf, so that it will not b
# included in the final data frame
if (n_req.k==0) {
  x2[k.idx, 'marg'] <- -Inf
} else {
  x2[k.idx, 'marg'] <- x2[k.idx,'points'] - x2[k.idx[n_req.k],'points']</pre>
}
if (n_req.qb==0){
  x2[qb.idx, 'marg'] <- -Inf
} else {
  x2[qb.idx, 'marg'] <- x2[qb.idx,'points'] - x2[qb.idx[n_req.qb],'points']</pre>
}
if (n req.rb==0){
  x2[rb.idx, 'marg'] <- -Inf</pre>
} else {
  x2[rb.idx, 'marg'] <- x2[rb.idx,'points'] - x2[rb.idx[n_req.rb],'points']</pre>
}
if (n_req.te==0){
  x2[te.idx, 'marg'] <- -Inf</pre>
} else {
  x2[te.idx, 'marg'] <- x2[te.idx,'points'] - x2[te.idx[n_req.te],'points']</pre>
}
```

```
if (n_req.wr==0){
    x2[wr.idx, 'marg'] <- -Inf</pre>
  } else {
    x2[wr.idx, 'marg'] <- x2[wr.idx,'points'] - x2[wr.idx[n_req.wr],'points']</pre>
  # create data.frame by removing players with negative marginal values
  x3 \leftarrow x2[x2[,'marg'] >= 0,]
  x3 <- x3[order(x3[,'marg'], decreasing=TRUE),]</pre>
  rownames(x3) <- NULL
  # calculate a dollar value!
  x3[,'value'] \leftarrow (nTeams*cap-nrow(x3)) * (x3[,'marg'] / sum(x3[,'marg'])) + 1
  # genearte a final data frame
  x4 <- x3[,c('PlayerName','pos','points','value')]</pre>
  ## save dollar values as CSV file
  write.csv(x=x4, file=paste0(path,'/',file))
  ## return data.frame with dollar values
  return(x4)
}
```

1. Call x1 <- ffvalues('.')</pre>

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

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

""
sum(sum(x1$value > 20))

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

David Montgomery

""
x1[which(x1[,'pos']=='rb')[15],'PlayerName']

""
## [1] "David Montgomery"
""
""
```

```
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)
46 players

""r
sum(x2$value > 20)
""

## [1] 46

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

""r
sum(x2[1:40,'pos']=='wr')
""
## [1] 8
""
## [1] 8
```

1. Call:

1. How many players are worth more than \$20? (1 point) 43 players

```
sum(x3$value >20)
```

```
## [1] 43
```

1. How many quarterbacks (qb) are in the top 30? (1 point) 13 quarterbacks

```
sum(x3[1:30,'pos']=='qb')
```

```
## [1] 13
```

Question 2

24 points

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

```
# read haart data and Load packages.
haart <- read_csv("haart.csv",show_col_types=FALSE)
haart <- data.frame(haart)
library(lubridate)</pre>
```

```
##
## Attaching package: 'lubridate'
```

```
## The following objects are masked from 'package:base':
##

## date, intersect, setdiff, union
```

```
library(dplyr)
library(tidyverse)
```

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

```
haart$init.date <- mdy(haart$init.date)
haart$last.visit <- mdy(haart$last.visit)
haart$date.death <- mdy(haart$date.death)
head(haart)</pre>
```

			aids ≍dbl>	cd4baseline <dbl></dbl>	logvl <dbl></dbl>	weight <dbl></dbl>	hemoglobin <dbl></dbl>	init.reg <chr></chr>	init.date <date></date>
1	1	25	0	NA	NA	NA	NA	3TC,AZT,EFV	2003-07-01
2	1	49	0	143	NA	58.0608	11	3TC,AZT,EFV	2004-11-23
3	1	42	1	102	NA	48.0816	1	3TC,AZT,EFV	2003-04-30
4	0	33	0	107	NA	46.0000	NA	3TC,AZT,NVP	2006-03-25
5	1	27	0	52	4	NA	NA	3TC,D4T,EFV	2004-09-01
6	0	34	0	157	NA	54.8856	NA	3TC,AZT,NVP	2003-12-02
6 ro	ws I	1-10	of 13 co	lumns					

```
haart_t <- haart %>%
  mutate(ydiff_last.visit=year(last.visit)-year(init.date)) %>%
  mutate(ydiff_date.death=year(date.death)-year(init.date))

print('last.visit ( a count of years from initial date )')
```

```
## [1] "last.visit ( a count of years from initial date )"
```

```
table(haart_t$ydiff_last.visit)
```

```
##
## 0 1 2 3 4 5 6 7 9
## 174 178 194 165 153 112 9 3 1
```

```
print('date.death ( a count of years from initial date )')
```

```
## [1] "date.death ( a count of years from initial date )"
```

```
table(haart_t$ydiff_date.death)
```

```
##
## 0 1 2 3 4
## 72 28 11 3 3
```

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?

92 observations died in year 1.

```
# create indicator variable
haart_t <- haart_t %>% mutate(ind = as.numeric((date.death-init.date)<=365))

# calculate the number of observations died in year 1 (within 365 days)
sum(haart_t$ind, na.rm=TRUE)</pre>
```

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

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.

```
haart[,'followup'] <- NA
haart[,'ddiff last.visit'] <- as.numeric(haart$last.visit - haart$init.date)</pre>
haart[,'ddiff_date.death'] <- as.numeric(haart$date.death - haart$init.date)</pre>
# create a initial followup variable
for (i in 1:length(haart[,'followup'])) {
  if (haart[i, 'death'] == 0) { # if date.death is NA
    # in this case, last.visit should be used (because date.death would be NA)
    # if last.visit is NA, it would be just NA
    haart[i,'followup'] <- haart[i,'ddiff last.visit']</pre>
  } else if (is.na(haart[i,'ddiff_last.visit'])) { # if last.visit is NA
    haart[i,'followup'] <- haart[i,'ddiff date.death']</pre>
    # compare which one is earlier between date.death & last.visit
    if (haart[i,'ddiff_last.visit']>haart[i,'ddiff_date.death']){
      haart[i, 'followup'] <- haart[i, 'ddiff date.death']</pre>
      } else {
        haart[i, 'followup'] <- haart[i, 'ddiff date.death']</pre>
  }
}
# censor followup value
cut.val <- 365
# conduct censoring
# if its value is larger than cut.val make it as cut.val.
# if not, don't change it.
haart <- haart %>% mutate(followup=if else(followup>cut.val, cut.val, followup))
# calculate quantile of followup variable (after censoring)
quantile(haart$followup)
```

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

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? 173 records

```
haart[,'loss'] <- NA
for (i in 1:length(haart[,'loss'])) {
   if (haart[i, 'death'] == 0 & haart[i, 'ddiff_last.visit'] < 365) {
     haart[i, 'loss'] <- 1
   } else
     haart[i, 'loss'] <- 0
}
sum(haart$loss)</pre>
```

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

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? 3TC, AZT, EFV, NVP, D4T are found over 100 times.

```
init.reg <- as.character(haart[,'init.reg'])
init.reg.unique <- unique(unlist(strsplit(haart[,'init.reg'],",")))

# initialize
mat_drugs <- matrix(0, nrow=nrow(haart), ncol=length(init.reg.unique))

# create new columns and fill-in
for (i in 1:nrow(mat_drugs)){
    for (j in 1:length(init.reg.unique))
        mat_drugs[i,j] <- grepl(pattern=init.reg.unique[[j]], x=init.reg[[i]])
}
mat_drugs <- data.frame(mat_drugs)
names(mat_drugs) <- init.reg.unique
haart_merged <- cbind(haart, mat_drugs)

# check which regimen are found over 100 times?
colSums(haart_merged[,init.reg.unique]) > 100
```

```
##
    3TC
          AZT
                EFV
                      NVP
                           D4T
                                 ABC
                                       DDI
                                             IDV
                                                  LPV
                                                        RTV
                                                              SOV
                                                                    FTC
                                                                          TDF
##
   TRUE
         TRUE
               TRUE
                     TRUE
                          TRUE FALSE FALSE FALSE FALSE FALSE FALSE
    DDC
                           FPV
##
          NFV
                T20
                     ATV
## FALSE FALSE FALSE FALSE
```

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.

```
haart2 <- read_csv("haart2.csv", show_col_types=FALSE)
haart2[,setdiff(names(haart), names(haart2))] <- 0
merged_data <- rbind(haart, haart2)
merged_data</pre>
```

 <dbl></dbl>	age <dbl></dbl>	ai <dbl></dbl>	cd4baseline <dbl></dbl>	logvl <dbl></dbl>	weight <dbl></dbl>	•	init.reg <chr></chr>	init.date <date></date>
1	25.00000	0	NA	NA	NA	NA	3TC,AZT,EFV	2003-07-01
1	49.00000	0	143	NA	58.06080	11.000000	3TC,AZT,EFV	2004-11-23
1	42.00000	1	102	NA	48.08160	1.000000	3TC,AZT,EFV	2003-04-30
0	33.00000	0	107	NA	46.00000	NA	3TC,AZT,NVP	2006-03-25
1	27.00000	0	52	4.000000	NA	NA	3TC,D4T,EFV	2004-09-01

 <dbl></dbl>	age <dbl></dbl>	ai <dbl></dbl>	cd4baseline <dbl></dbl>	logvl <dbl></dbl>	weight <dbl></dbl>	•	init.reg <chr></chr>	init.date <date></date>
0	34.00000	0	157	NA	54.88560	NA	3TC,AZT,NVP	2003-12-02
0	39.00000	0	65	NA	55.33920	11.000000	3TC,AZT,NVP	2004-02-06
1	31.00000	0	NA	NA	NA	NA	3TC,AZT,EFV	2001-09-06
1	52.00000	0	NA	NA	NA	NA	3TC,ABC,AZT	2002-08-13
1	23.00000	1	3	5.718295	NA	NA	3TC,DDI,NVP	2005-06-21
I-10 of	f 1,004 row	rs 1-10) of 16 columns	3	Pr	evious 1 2	2 3 4 5	6 101 Next

head(merged_data)

	n :dbl>		aids ×dbl>	cd4baseline <dbl></dbl>	logvl <dbl></dbl>	weight <dbl></dbl>	hemoglobin <dbl></dbl>	init.reg <chr></chr>	init.date <date></date>
1	1	25	0	NA	NA	NA	NA	3TC,AZT,EFV	2003-07-01
2	1	49	0	143	NA	58.0608	11	3TC,AZT,EFV	2004-11-23
3	1	42	1	102	NA	48.0816	1	3TC,AZT,EFV	2003-04-30
4	0	33	0	107	NA	46.0000	NA	3TC,AZT,NVP	2006-03-25
5	1	27	0	52	4	NA	NA	3TC,D4T,EFV	2004-09-01
6	0	34	0	157	NA	54.8856	NA	3TC,AZT,NVP	2003-12-02

tail(merged_data)

		•	ai	cd4baseline	logvi	•	hemoglobin	•	init.dat
	<dbl></dbl>	· <dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<chr></chr>	<date< td=""></date<>
999	0	31.00000	0	102	NA	61.6896	11	3TC,AZT,NVP	2003-05-2
1000	0	40.00000	1	131	NA	46.2672	8	3TC,D4T,NVP	2003-07-0
1001	0	27.00000	0	232	NA	NA	NA	3TC,AZT,NVP	0012-01-0
1002	1	38.72142	0	170	NA	84.0000	NA	3TC,AZT,NVP	<na< td=""></na<>
1003	1	23.00000	NA	154	3.995635	65.5000	14	3TC,DDI,EFV	<na< td=""></na<>
1004	0	31.00000	0	236	NA	45.8136	NA	3TC,D4T,NVP	0012-03-0