Kaitlin's EDA

$Kaitlin\ Maciejewski$ 12/1/2018

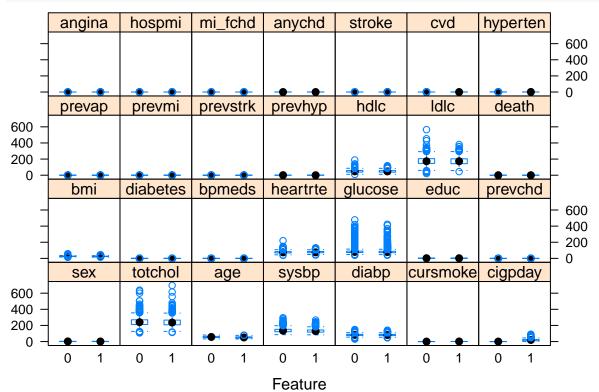
Exploratory

```
library(dplyr)
data <- read.csv("../final_data/frmgham2.csv") %>% janitor::clean_names()
attach(data)
library(psych)
knitr::kable(describe(data)[,c(2,3,4,5,8,9,10,13)], digits = 3)
```

	n	mean	sd	median	min	max	range	se
randid	11627	5004740.917	2900877.440	5006008.00	2448.00	9999312.0	9996864.00	26902.680
sex	11627	1.568	0.495	2.00	1.00	2.0	1.00	0.005
totchol	11218	241.162	45.368	238.00	107.00	696.0	589.00	0.428
age	11627	54.793	9.564	54.00	32.00	81.0	49.00	0.089
sysbp	11627	136.324	22.799	132.00	83.50	295.0	211.50	0.211
diabp	11627	83.038	11.660	82.00	30.00	150.0	120.00	0.108
$\operatorname{cursmoke}$	11627	0.433	0.495	0.00	0.00	1.0	1.00	0.005
cigpday	11548	8.250	12.187	0.00	0.00	90.0	90.00	0.113
bmi	11575	25.877	4.103	25.48	14.43	56.8	42.37	0.038
diabetes	11627	0.046	0.209	0.00	0.00	1.0	1.00	0.002
bpmeds	11034	0.086	0.280	0.00	0.00	1.0	1.00	0.003
heartrte	11621	76.782	12.463	75.00	37.00	220.0	183.00	0.116
glucose	10187	84.125	24.994	80.00	39.00	478.0	439.00	0.248
educ	11332	1.990	1.027	2.00	1.00	4.0	3.00	0.010
prevchd	11627	0.072	0.259	0.00	0.00	1.0	1.00	0.002
prevap	11627	0.054	0.226	0.00	0.00	1.0	1.00	0.002
prevmi	11627	0.032	0.176	0.00	0.00	1.0	1.00	0.002
prevstrk	11627	0.013	0.114	0.00	0.00	1.0	1.00	0.001
prevhyp	11627	0.460	0.498	0.00	0.00	1.0	1.00	0.005
$_{ m time}$	11627	1957.019	1758.777	2156.00	0.00	4854.0	4854.00	16.311
period	11627	1.899	0.807	2.00	1.00	3.0	2.00	0.007
hdlc	3027	49.365	15.627	48.00	10.00	189.0	179.00	0.284
ldlc	3026	176.467	46.863	173.00	20.00	565.0	545.00	0.852
death	11627	0.303	0.460	0.00	0.00	1.0	1.00	0.004
angina	11627	0.164	0.370	0.00	0.00	1.0	1.00	0.003
hospmi	11627	0.099	0.299	0.00	0.00	1.0	1.00	0.003
mi_fchd	11627	0.154	0.361	0.00	0.00	1.0	1.00	0.003
anychd	11627	0.272	0.445	0.00	0.00	1.0	1.00	0.004
stroke	11627	0.091	0.288	0.00	0.00	1.0	1.00	0.003
cvd	11627	0.249	0.433	0.00	0.00	1.0	1.00	0.004
hyperten	11627	0.743	0.437	1.00	0.00	1.0	1.00	0.004
$_{\rm timeap}$	11627	7241.557	2477.780	8766.00	0.00	8766.0	8766.00	22.979
$_{ m timemi}$	11627	7593.847	2136.730	8766.00	0.00	8766.0	8766.00	19.816
$_{\rm timemifc}$	11627	7543.037	2192.120	8766.00	0.00	8766.0	8766.00	20.330
timechd	11627	7008.154	2641.345	8766.00	0.00	8766.0	8766.00	24.496

	\mathbf{n}	mean	sd	median	\min	max	range	se
timestrk	11627	7660.880	2011.077	8766.00	0.00	8766.0	8766.00	18.651
timecvd	11627	7166.083	2541.668	8766.00	0.00	8766.0	8766.00	23.571
timedth	11627	7854.103	1788.370	8766.00	26.00	8766.0	8740.00	16.585
timehyp	11627	3598.956	3464.165	2429.00	0.00	8766.0	8766.00	32.127

```
library(caret)
data_2 <- select(data, -c(randid, time, period, timeap:timehyp))
featurePlot(data_2,as.factor(cursmoke),"box")</pre>
```



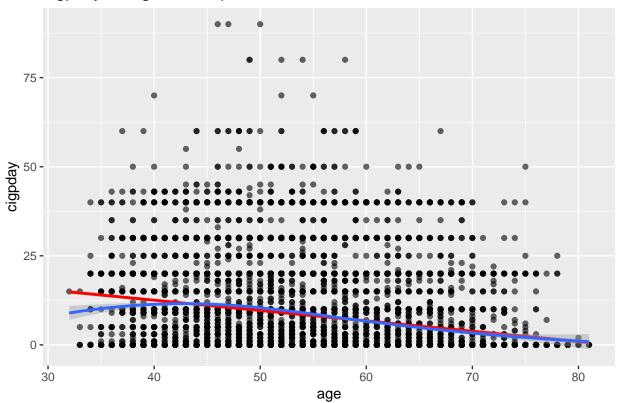
```
# library(GGally)
# ggpairs(data_2)

# spaghetti
# lorellogram
#

library(ggplot2)

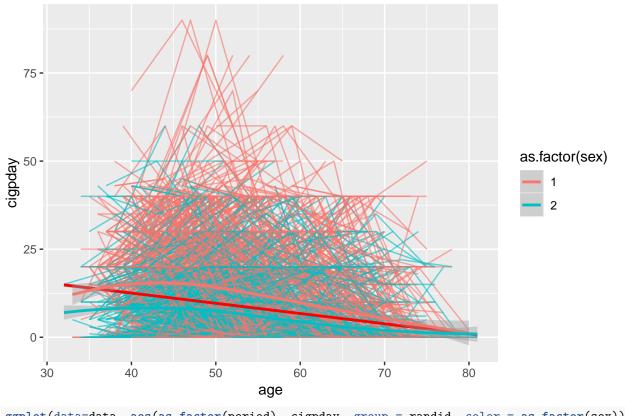
ggplot(data= data, aes(age, cigpday)) +
    geom_point(alpha = .6) +
    geom_smooth(method = 'lm', col = 'red') +
    geom_smooth(method = 'loess') +
    ggtitle("cigpday vs Age Scatterplot")
```

cigpday vs Age Scatterplot



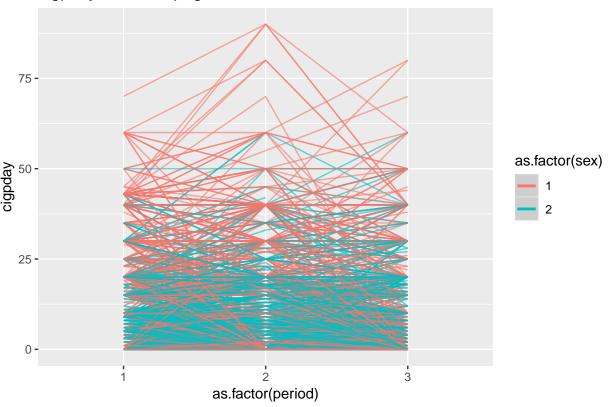
```
ggplot(data=data, aes(age, cigpday, group = randid, color = as.factor(sex))) +
  geom_path(alpha = .6) +
  geom_smooth(aes(group = NULL), method = 'lm', col = 'red') +
  geom_smooth(aes(group = NULL), method = 'loess') +
  ggtitle("cigpday vs Age Spaghetti Plot")
```

cigpday vs Age Spaghetti Plot



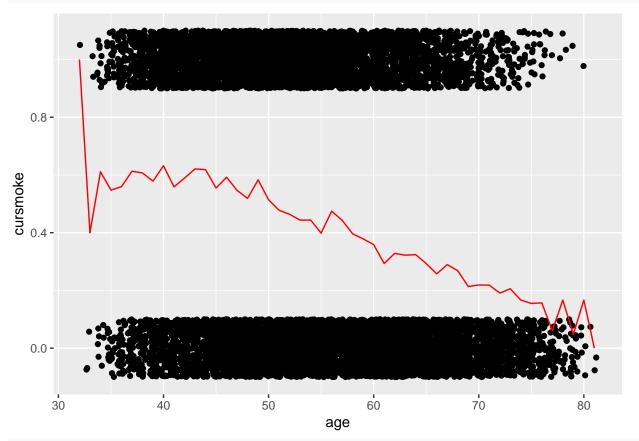
```
ggplot(data=data, aes(as.factor(period), cigpday, group = randid, color = as.factor(sex))) +
  geom_path(alpha = .6) +
  geom_smooth(aes(group = NULL), method = 'lm', col = 'red') +
  geom_smooth(aes(group = NULL), method = 'loess') +
  ggtitle("cigpday vs time Spaghetti Plot")
```

cigpday vs time Spaghetti Plot



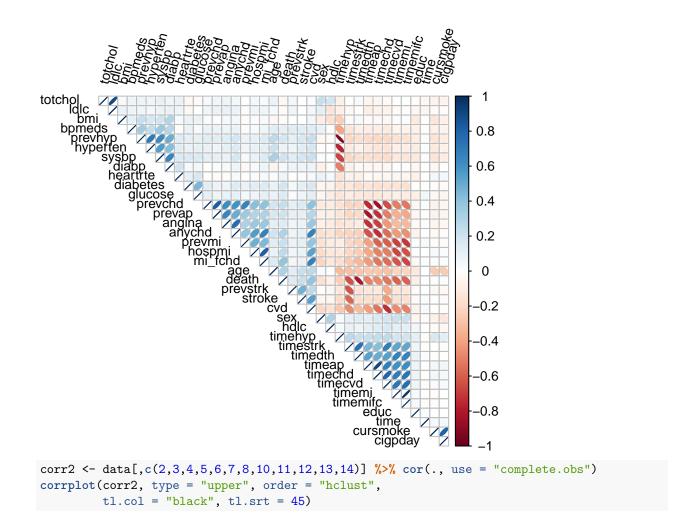
```
### Select samples from quartiles -- doesnt work
# data.residual <- data %>% na.omit() %>%
# group_by(age) %>%
# mutate(mean.cigpday = mean(cigpday)) %>%
# ungroup() %>%
# mutate(residuals = cigpday - mean.cigpday) %>%
# group_by(randid) %>%
# mutate(median.residual = median(residuals)) %>%
# ungroup()
# data.stats <- c(min(data.residual$median.residual),</pre>
# quantile(data.residual$median.residual,
\# c(.25, .5, .75)), \max(\text{data.residual}\$\text{median.residual}))
# data.id.select <- data.residual %>%
# filter(median.residual %in% data.stats)
# data.residual.plot <- ggplot() +</pre>
# geom_line(data = data.id.select,
\# aes(x = age, y = cigpday, group = randid)) +
# ggtitle("cigpday by time, Selected from Residuals") +
# geom_smooth(data = data,
\# aes(x = age, y = cigpday))
###
```

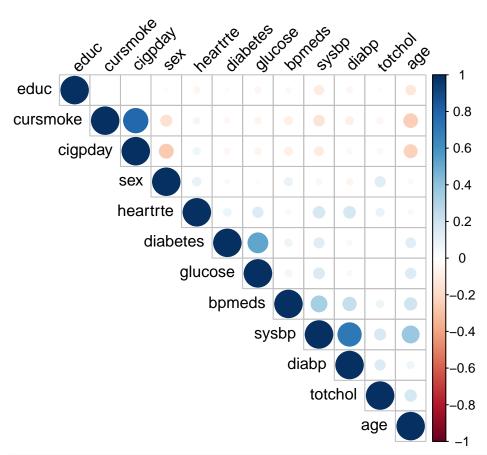
```
ggplot(data, aes(y = cursmoke, x = age)) + geom_jitter(height = 0.1) +
stat_summary(fun.y = 'mean', geom="line", col = 'red')
```



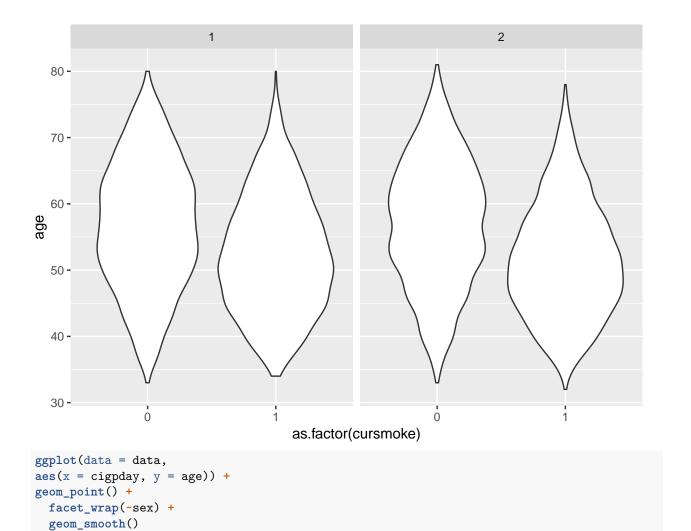
select(data, -c(randid, time,timeap:timehyp)) %>% GGally::ggcorr(.)

```
hyper
                                                                cvd
                                                            stroke
                                                      anychd
mi_fchd
                                                    hospmi
                                                  angina
                                                death
                                hdlk
period
prevhyp
prevstrk
prevmi
evap
                                               Idlc
                                                                              1.0
                                                                              0.5
                            prevap
prevo<mark>hd</mark>
                                                                              0.0
                           educ
                                                                              -0.5
                       glucose
                     heartrte
                                                                              -1.0
                  bpmeds
                diabetes
                bmi
        cigpday
cursm<mark>ok</mark>e
        diabp
     sysbp
    age
totchol
sex
corr <- data[,c(-1,-21)] %>% cor(., use = "complete.obs")
library(corrplot)
corrplot(corr, type = "upper", order = "hclust",
          tl.col = "black", tl.srt = 75, tl.offset = 1, tl.cex = .8, method = "ellipse")
```

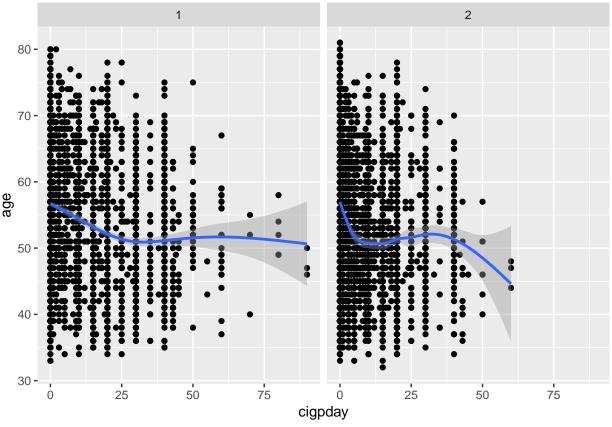




```
ggplot(data = data,
aes(x = as.factor(cursmoke), y = age)) +
geom_violin() +
facet_wrap(~sex)
```



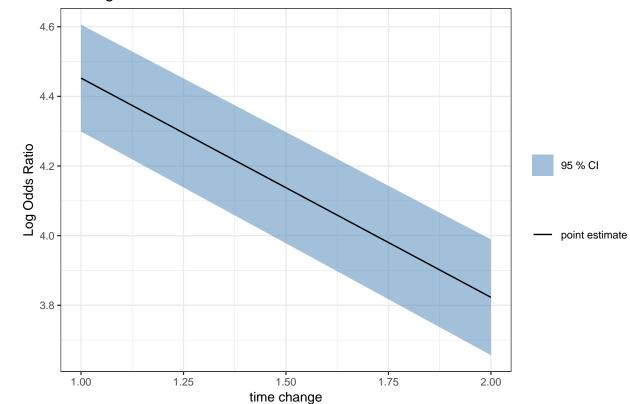
```
10
```



```
resp.long_2 <- data
theme_set(theme_bw(base_size = 10))
##lorelogram
lorelogram <- function(id, time, y, title){</pre>
  data <- data_frame(id,time,y)</pre>
  #function that innumerates all combinations of a given time
  #and all futures times for an individual and then
  #returns a dataframe of time differences and results at both time points
  calc_time_deltas <- function(ind_data){</pre>
   results <- expand.grid(ind_data$time,ind_data$time) %>%
      rename(time1 = Var1, time2 = Var2) %>% #rename columns
      right_join(ind_data[,c("time", "y")], by = c("time1" = "time")) %>%
      rename(y1 = y) %>% #add results for time1 from the full data
      right_join(ind_data[,c("time", "y")], by = c("time2" = "time")) %>%
      rename(y2 = y) %>% #add results for time2 from the full data.
      mutate(time_diff = time1 - time2, id = ind_data$id[1]) %>% #find difference in times
      filter(time_diff > 0) #cleanup output.
 }
  Z <- data %>%
    group_by(id) %>% #qrab an individuals data in isolation
    do(calc_time_deltas(.)) #run the immumeration function
  #Predict past outcomes from future by utilizing time differences
  outcome_model <- glm(y1 ~ y2:factor(time_diff), data=Z, family=binomial)</pre>
```

```
#grab the parameter estimates (ignoring intercept)
  LOR_estimates <- data_frame(</pre>
    time_diff = sort(unique(Z$time_diff)),
    point est = summary(outcome model)$coef[-1,1],
    std_err = summary(outcome_model)$coef[-1,2] ) %>%
    mutate(lower_bound = point_est - 1.96*std_err,
           upper_bound = point_est + 1.96*std_err)
  #plot it
  ggplot(LOR_estimates, aes(x = time_diff)) +
    geom_ribbon(aes(ymin = lower_bound, ymax = upper_bound, fill = "95 % CI"),
                alpha = 0.5) +
    geom_line(aes(y = point_est, color = "point estimate")) +
    scale_colour_manual("", values="black")+
    scale_fill_manual("",values="steelblue") +
    labs(x = "time change", y = "Log Odds Ratio", title = title)
}
lorelogram(id = resp.long_2$randid,
             time = resp.long_2$period,
             y = resp.long_2$cursmoke,
             title = 'Lorelogram')
```

Lorelogram



Models GLMER

(1) Is there a relationship between age and smoking status? Does this relationship differ by sex?

```
# possible confounders of smoking status...
# education
# disease status

library(lme4)

smoke_stat_age <- glmer(cursmoke~age+educ + (1|randid), family=binomial, na.action = "na.omit")

knitr::kable(summary(smoke_stat_age)$coefficients,digits = 3)</pre>
```

	Estimate	Std. Error	z value	$\Pr(> z)$
(Intercept)	9.854	0.516	19.101	0.000
age	-0.196	0.009	-22.278	0.000
educ	-0.182	0.102	-1.787	0.074

```
smoke_stat_sex <- glmer(cursmoke~age+educ + as.factor(sex) + (1|randid), family=binomial, na.action = "sknitr::kable(summary(smoke_stat_sex)$coefficients,digits = 3)</pre>
```

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	12.402	0.623	19.918	0.000
age	-0.206	0.009	-22.583	0.000
educ	-0.211	0.110	-1.914	0.056
as.factor(sex)2	-3.715	0.332	-11.191	0.000

(2) Is there a relationship between the number of cigarettes smoked per day and age? Does this relationship differ by sex?

```
cigpday_age <- glmer(cigpday~age+educ +(1|randid), family=poisson, na.action = "na.omit")
knitr::kable(summary(cigpday_age)$coefficients, digits = 3)</pre>
```

	Estimate	Std. Error	z value	$\Pr(> z)$
(Intercept)	0.239 -0.018	0.144 0.001	1.662 -25.560	0.096 0.000
age educ	0.021	0.058	0.362	0.717

```
ncig_gee <- glmer(cigpday~age+as.factor(sex) + educ+ (1|randid), family=poisson, na.action = "na.omit")
knitr::kable(summary(ncig_gee)$coefficients,digits = 3)</pre>
```

	Estimate	Std. Error	z value	$\Pr(> z)$
(Intercept)	1.358	0.148	9.163	0.000
age	-0.018	0.001	-25.622	0.000
as.factor(sex)2	-1.994	0.118	-16.845	0.000
educ	0.023	0.056	0.409	0.683

(1) The relationship between current smoking status and systolic blood pressure.

```
smoke_sys<-glmer(cursmoke~sysbp + (1|randid), family=binomial, na.action = "na.omit")
knitr::kable(summary(smoke_sys)$coefficients,digits = 3)</pre>
```

	Estimate	Std. Error	z value	$\Pr(> z)$
(Intercept)	3.778	0.364	10.39	0
sysbp	-0.035	0.003	-13.20	0

```
smoke_sys<-glmer(cursmoke~sysbp + as.factor(sex) + (1|randid), family=binomial, na.action = "na.omit")
knitr::kable(summary(smoke_sys)$coefficients,digits = 3)</pre>
```

	Estimate	Std. Error	z value	$\Pr(> z)$
(Intercept)	5.328	0.414	12.860	0
sysbp	-0.035	0.003	-12.767	0
as.factor(sex)2	-2.927	0.264	-11.071	0

(2) The relationship between current smoking status and diastolic blood pressure.

```
smoke_dias<-glmer(cursmoke~diabp + (1|randid), family=binomial, na.action = "na.omit")
knitr::kable(summary(smoke_dias)$coefficients,digits = 3)</pre>
```

	Estimate	Std. Error	z value	$\Pr(> z)$
(Intercept)	0.731	0.388	1.886	0.059
diabp	-0.020	0.005	-4.431	0.000

```
smoke_dias<-glmer(cursmoke~diabp + as.factor(sex) + (1|randid), family=binomial, na.action = "na.omit"
knitr::kable(summary(smoke_dias)$coefficients,digits = 3)</pre>
```

	Estimate	Std. Error	z value	$\Pr(> z)$
(Intercept)	2.410	0.423	5.701	0
diabp	-0.023	0.005	-4.825	0
as.factor(sex)2	-2.920	0.260	-11.244	0

(3) The relationship between current smoking status and serum total cholesterol.

```
smoke_chol<-glmer(cursmoke~totchol + (1|randid), family=binomial, na.action = "na.omit")
knitr::kable(summary(smoke_chol)$coefficients,digits = 3)</pre>
```

	Estimate	Std. Error	z value	$\Pr(> z)$
(Intercept)	-0.531	0.317	-1.673	0.094
totchol	-0.002	0.001	-1.305	0.192

```
smoke_chol<-glmer(cursmoke~totchol + as.factor(sex) + (1|randid), family=binomial, na.action = "na.omi
knitr::kable(summary(smoke_chol)$coefficients,digits = 3)</pre>
```

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	0.523	0.343	1.525	0.127
totchol	0.000	0.001	0.038	0.970
as.factor(sex)2	-2.991	0.277	-10.798	0.000

```
###
my.data.complete <- data %>%
  dplyr::select(-c(hdlc,ldlc)) %>%
  na.omit()
model.saturated <- geepack::geeglm(formula = cursmoke ~ as.factor(sex) +</pre>
    age + age * as.factor(sex) + sysbp + diabp + sysbp * diabp +
   bpmeds + as.factor(educ) + totchol + bmi + glucose + diabetes +
   heartrte + prevap + prevchd + prevmi + prevstrk + prevhyp,
   family = binomial, data = my.data.complete, id = randid,
    corstr = ("unstructured"))
model <- aov(model.saturated)</pre>
car::vif(model)
                           GVIF Df GVIF^(1/(2*Df))
## as.factor(sex)
                      35.783769 1
                                          5.981954
## age
                      2.661847 1
                                          1.631517
## sysbp
                      35.429075 1
                                          5.952233
## diabp
                      25.680738 1
                                          5.067617
## bpmeds
                       1.194381 1
                                          1.092878
## as.factor(educ)
                       1.108310 3
                                          1.017287
## totchol
                       1.107321 1
                                          1.052293
## bmi
                       1.183872 1
                                          1.088059
## glucose
                       1.414456 1
                                          1.189309
## diabetes
                       1.392703 1
                                          1.180128
## heartrte
                       1.081176 1
                                          1.039796
## prevap
                       4.787315 1
                                          2.187993
## prevchd
                       7.474492 1
                                          2.733952
                       2.402776 1
## prevmi
                                          1.550089
## prevstrk
                       1.027812 1
                                          1.013811
```

Models GEE

sysbp:diabp

as.factor(sex):age 37.891150 1

prevhyp

(1) Is there a relationship between age and smoking status? Does this relationship differ by sex?
smoke_stat_age <- gee::gee(cursmoke-age, id =randid, family=binomial, corstr = "unstructured", na.action
(Intercept) age</pre>

1.439414

6.155579

9.589336

2.071912 1

91.955361 1

2.80332932 -0.05649341

```
knitr::kable(summary(smoke_stat_age)$coefficients[,c(1,4,5)],digits = 3)
```

	Estimate	Robust S.E.	Robust z
${(Intercept)}$ age	2.449	0.116	21.200
	-0.049	0.002	-24.196

smoke_stat_sex <- gee::gee(cursmoke~age+as.factor(sex), id =randid, family=binomial, corstr = "unstruct")</pre>

(Intercept) age as.factor(sex)2 ## 3.16756901 -0.05686335 -0.61580832

knitr::kable(summary(smoke_stat_sex)\$coefficients[,c(1,4,5)],digits = 3)

	Estimate	Robust S.E.	Robust z
(Intercept)	2.844	0.122	23.363
age	-0.050	0.002	-24.292
as.factor(sex)2	-0.636	0.057	-11.092

knitr::kable(round(2*pnorm(abs(coef(summary(smoke_stat_sex))[,5]), lower.tail = FALSE), 3))

	Х
(Intercept)	0
age	0
as.factor(sex)2	0

(2) Is there a relationship between the number of cigarettes smoked per day and age? Does this relationship differ by sex?

differ by sex?
cigpday_age <- gee::gee(cigpday~age, id=randid, family=poisson, corstr = "unstructured", na.action = "n</pre>

(Intercept) age ## 4.05032331 -0.03649173

knitr::kable(summary(cigpday_age)\$coefficients[,c(1,4,5)],digits = 3)

	Estimate	Robust S.E.	Robust z
(Intercept) age	3.451	0.066	52.064
	-0.025	0.001	-19.607

ncig_gee <- gee::gee(cigpday~age+as.factor(sex), id=randid, family=poisson, corstr = "unstructured", na</pre>

```
## (Intercept) age as.factor(sex)2
## 4.3362004 -0.0355547 -0.7015079
```

summary(ncig_gee)

##

GEE: GENERALIZED LINEAR MODELS FOR DEPENDENT DATA

gee S-function, version 4.13 modified 98/01/27 (1998)

```
##
## Model:
##
  Link:
                               Logarithm
   Variance to Mean Relation: Poisson
##
   Correlation Structure:
                               Unstructured
##
## Call:
   gee::gee(formula = cigpday ~ age + as.factor(sex), id = randid,
       na.action = "na.omit", family = poisson, corstr = "unstructured")
##
##
##
  Summary of Residuals:
##
          Min
                      1Q
                             Median
                                             ЗQ
                                                       Max
  -19.386146 -7.058707
                          -4.393591
##
                                      7.136533 77.315884
##
##
## Coefficients:
##
                      Estimate Naive S.E.
                                            Naive z Robust S.E.
                                                                  Robust z
## (Intercept)
                    3.78802175 0.07973074 47.51018 0.065953995 57.43430
                   -0.02495343 0.00147914 -16.87023 0.001258888 -19.82180
## as.factor(sex)2 -0.73580910 0.03810161 -19.31176 0.038909532 -18.91077
##
## Estimated Scale Parameter: 16.06327
## Number of Iterations: 3
##
## Working Correlation
             [,1]
                       [,2]
## [1,] 1.0000000 0.7397549 0.4860163
## [2,] 0.7397549 1.0000000 0.6480732
## [3,] 0.4860163 0.6480732 1.0000000
knitr::kable(summary(ncig_gee)$coefficients[,c(1,4,5)],digits = 3)
```

	Estimate	Robust S.E.	Robust z
$\overline{\text{(Intercept)}}$	3.788	0.066	57.434
age as. $factor(sex)2$	-0.025 -0.736	$0.001 \\ 0.039$	-19.822 -18.911

knitr::kable(round(2*pnorm(abs(coef(summary(ncig_gee))[,5]), lower.tail = FALSE),3))

(1) The relationship between current smoking status and systolic blood pressure.

	Estimate	Robust S.E.	Robust z
(Intercept) sysbp	0.902	0.112	8.025
	-0.009	0.001	-10.704

(2) The relationship between current smoking status and diastolic blood pressure.

	Estimate	Robust S.E.	Robust z
(Intercept)	0.201	0.123	1.628
diabp	-0.006	0.001	-3.934

(3) The relationship between current smoking status and serum total cholesterol.

```
smoke_chol<-gee::gee(cursmoke~totchol, id =randid, family=binomial, corstr = "unstructured", na.action
## (Intercept) totchol
## 0.180474997 -0.001869103
knitr::kable(summary(smoke_chol)$coefficients[,c(1,4,5)],digits = 3)</pre>
```

	Estimate	Robust S.E.	Robust z
(Intercept)	-0.034	0.107	-0.313
totchol	-0.001	0.000	-2.304

What to Turn In: For this assignment you will turn in a single pdf document with (a) your summary of findings and (b) an Appendix with figures and (c) an Appendix with all R code (in the form of a knited pdf).

OBJECTIVE:

An objective or description of the goals of the analysis

We are interested to describe the smoking habits of the participants in the Framingham Heart study as they age and the impact of smoking on certain health outcomes. The Framingham heart study asks participants about their smoking habits at each visit. In particular, participants are asked if they are currently smoking at this visit (0 = Not a current smoker, 1 = Current smoker), which we will refer to as current smoking status. In addition, participants also report the number of cigarettes they are smoking per day. A more complete description of each of variables in the Framingham Heart study can be found in the Framingham Heart Study Longitudinal Data Documentation.

We are interested to answer the following questions:

- (1) Is there a relationship between age and smoking status? Does this relationship differ by sex?
- (2) Is there a relationship between the number of cigarettes smoked per day and age? Does this relationship differ by sex?

While answering these questions, please account for any confounders that you have evidence may impact the relationship between age and sex with smoking.

Next you are interested in the relationship between certain health outcomes and smoking status. In particular you are interested in :

- (1) The relationship between current smoking status and systolic blood pressure.
- (2) The relationship between current smoking status and diastolic blood pressure.
- (3) The relationship between current smoking status and serum total cholesterol.

Again, while answering these questions, please account for any confounders that you have evidence may impact these relationships.

STUDY DESIGN:

A brief description of the study design and the data

METHODS:

A methods section describing your statistical analysis (please justify all modeling choices that were made with evidence).

RESULTS:

A results section that includes a) descriptive statistics for the data b) a summary of your key findings including supporting numerical summaries (i.e. confidence intervals, pvalues, etc.) c) interpretations of your key findings (i.e. interpretations of coefficients).

CONCLUSION:

A conclusion specifically answering the objective of the analysis.

APPENDIX: