

# STAT 331 Final Project

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## 1 Summary

## 2 Descriptive Statistics

First, take a look at summary statistics of the `fhsd` dataset.

Table 1: Summary Statistics

chdrisk	sex	totchol	age	sysbp	diabp	cursmoke	cigday	bmi	diabetes	hpmeds	hearttte	glucose	prevmi	prevstrk	prevhdp	hdlc	ldlc
Min: 0.0050	Female:1305	Min: 112.0	Min: 44.00	Min: 86.0	Min: 30.00	No:1504	Min: 0.00	Min: 14.43	No:2142	No:1973	Min: 44.00	Min: 46.00	No:2189	No:2260	No: 957	Min: 10.00	Min: 20.0
1st Qu: 0.1320	Male:1001	1st Qu: 207.0	1st Qu: 53.00	1st Qu: 122.5	1st Qu: 73.00	Yes: 802	1st Qu: 0.00	1st Qu: 23.22	Yes: 164	Yes: 353	1st Qu: 70.00	1st Qu: 75.00	Yes: 117	Yes: 46	Yes:1349	1st Qu: 38.00	1st Qu: 152.0
Median: 0.2240	NA	Median: 235.5	Median: 60.00	Median: 136.0	Median: 80.00	NA	Median: 0.00	Median: 25.40	NA	NA	Median: 76.00	Median: 83.00	NA	NA	NA	Median: 47.00	Median: 180.0
Mean: 0.2655	NA	Mean: 237.8	Mean: 60.23	Mean: 139.2	Mean: 81.07	NA	Mean: 6.84	Mean: 25.78	NA	NA	Mean: 77.61	Mean: 89.07	NA	NA	NA	Mean: 48.89	Mean: 183.1
3rd Qu: 0.3448	NA	3rd Qu: 265.0	3rd Qu: 67.00	3rd Qu: 153.0	3rd Qu: 88.00	NA	3rd Qu: 10.00	3rd Qu: 27.91	NA	NA	3rd Qu: 85.00	3rd Qu: 95.00	NA	NA	NA	3rd Qu: 57.00	3rd Qu: 210.0
Max: 0.9770	NA	Max: 625.0	Max: 81.00	Max: 246.0	Max: 130.00	NA	Max: 80.00	Max: 46.52	NA	NA	Max: 150.00	Max: 478.00	NA	NA	NA	Max: 189.00	Max: 565.0

Then take a look at `chdrisk` grouped by `sex` as well as `chdrisk` grouped by `cursmoke`.

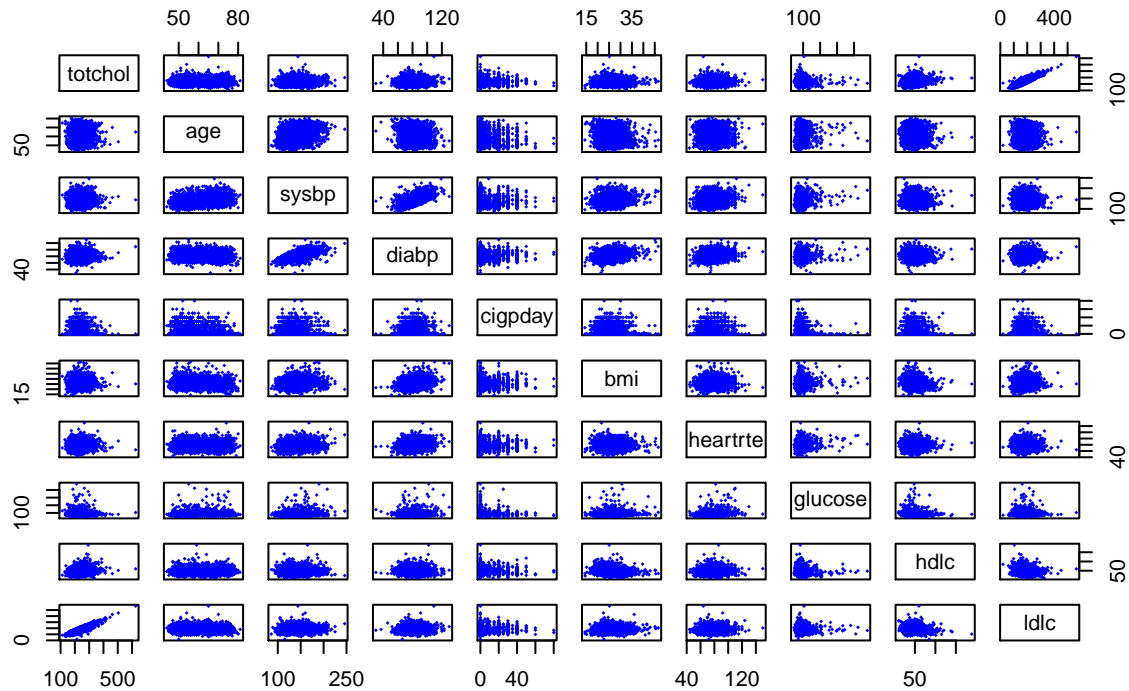
```
## fhsd$sex: Female
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      0.005   0.104   0.179   0.215   0.285   0.949
## -----
## fhsd$sex: Male
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      0.0210  0.1860  0.2860  0.3314  0.4060  0.9770

## fhsd$cursmoke: No
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      0.0050  0.1390  0.2350  0.2754  0.3580  0.9770
## -----
## fhsd$cursmoke: Yes
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      0.0080  0.1220  0.1995  0.2471  0.3140  0.9710
```

[ADD SOME COMMENTS HERE REGARDING SUMMARY]

Now take a look at pair plots of all numeric explanatory variates i.e. variates excluding response variate `chdrisk` and logical variates such as `cursmoke`.

## Pair Plots of Continuous Variates



From the pair plots, we can observe a strong correlation between low density lipoprotein cholesterol and serum total cholesterol. This correlation could be explained by the fact that there could be a relationship between the amount [TO BE CONTINUED]

Now take a look at the VIFs of these variates.

##	sexMale	totchol	age	sysbp	diabp	cursmokeYes
##	1.225191	10.634882	1.489926	2.918660	2.406411	2.978609
##	cigpday	bmi	diabetesYes	bpmedsYes	hearttrte	glucose
##	2.973594	1.181865	1.286401	1.214744	1.105902	1.308923
##	prevmiYes	prevstrkYes	prevhypYes	hdlc	ldlc	
##	1.067134	1.045746	1.823014	2.287571	10.367649	

[ADD COMMENTS]

## 3 Candidate Models

### 3.1 Automated Model Selection

```
library(gtools)

load_calcs = TRUE
# model with only intercept
M0 <- lm(I(logit(chdrisk)) ~ 1, data = fhds)
Mmax <- lm(I(logit(chdrisk)) ~ (. )^2, data = fhds)
# starting model for stepwise selection
Mstart <- lm(I(logit(chdrisk)) ~ ., data = fhds)

# find model coefficients which are NA
beta.max <- coef(Mmax)
names(beta.max)[is.na(beta.max)]

## [1] "cursmokeYes:cigpday" "bpmedsYes:prevhypYes"
# find the problem with the NA coeffs
kable(table(fhds[c("cursmoke", "cigpday")]), "latex")
```

	0	1	2	3	4	5	6	7	8	9	10	12	14	15	16	17	18	19	20	23	25	26	27
No	1504	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
Yes	0	16	18	34	11	18	24	9	18	5	76	3	3	50	6	1	8	1	279	1	14	1	1

```
kable(table(fhds[c("bpmeds", "prevhyp")]), "latex")
```

	No	Yes
No	957	1016
Yes	0	333

```
# remove the coeffs with the problem and add quadratic terms for the continuous variables
Mmax <- lm(I(logit(chdrisk)) ~ (. )^2 - cursmoke:cigpday - bpmeds:prevhyp +
  I(totchol ^ 2) + I(sysbp ^ 2) + I(diabp ^ 2)
  + I(bmi ^ 2) + I(glucose ^ 2)
  + I(hdlc ^ 2) + I(ldlc ^ 2), data = fhds)

anyNA(coef(Mmax)) # check if there are any remaining NAs
```

```
## [1] FALSE
```

```
if(!load_calcs){
  #forward model selection
  system.time({
    Mfwd <- step(object = M0,
      scope = list(lower = M0, upper = Mmax),
      direction = "forward", trace = FALSE)
  })

  #backward model selection
  system.time({
    Mback <- step(object = Mmax,
      scope = list(lower = M0, upper = Mmax),
      direction = "backward", trace = FALSE)
  })
}
```

```

#stepwise model selection
system.time({
  Mstep <- step(object = Mstart,
                scope = list(lower = M0, upper = Mmax),
                direction = "both", trace = FALSE)
})

}

# the caching/loading block
if(!load_calcs) {
  saveRDS(list(Mfwd = Mfwd, Mback = Mback, Mstep = Mstep), file = "models_automated.rds")
} else {
  # just load the calculations
  tmp <- readRDS("models_automated.rds")
  Mfwd <- tmp$Mfwd
  Mback <- tmp$Mback
  Mstep <- tmp$Mstep
  rm(tmp) # optionally remove tmp from workspace
}

# Stepwise model selection
Mstep$call

## lm(formula = I(logit(chdrisk)) ~ sex + totchol + age + sysbp +
##   diabp + cursmoke + cigpday + bmi + diabetes + bpmeds + hearttrte +
##   glucose + prevmi + prevstrk + prevhyp + hdlc + ldlc + I(hdlc^2) +
##   I(bmi^2) + I(diabp^2) + I(sysbp^2) + sysbp:prevmi + totchol:prevhyp +
##   diabetes:prevmi + prevhyp:ldlc + sysbp:prevhyp + totchol:hearttrte +
##   sysbp:diabetes + diabp:bmi + diabp:hdlc + prevmi:hdlc + prevmi:prevhyp +
##   sex:glucose + age:ldlc + age:hearttrte + cigpday:hdlc + bmi:ldlc +
##   totchol:hdlc + totchol:prevmi + sysbp:hearttrte + sysbp:bpmeds +
##   cursmoke:hdlc + prevmi:prevstrk + diabetes:hdlc + sex:sysbp +
##   cigpday:glucose + hearttrte:glucose + diabp:glucose + cursmoke:ldlc +
##   age:cigpday + age:hdlc + hdlc:ldlc + age:prevhyp + diabp:prevhyp +
##   diabp:cursmoke + diabp:cigpday + bmi:bpmeds + bpmeds:glucose +
##   age:prevmi + sex:ldlc + cigpday:hearttrte + cigpday:prevmi +
##   glucose:prevmi + hearttrte:prevmi + bpmeds:prevstrk, data = fhds)

# Forward model selection
Mfwd$call

## lm(formula = I(logit(chdrisk)) ~ prevmi + sysbp + sex + age +
##   ldlc + prevhyp + diabetes + hdlc + I(hdlc^2) + cigpday +
##   I(bmi^2) + bmi + totchol + I(glucose^2) + I(sysbp^2) + bpmeds +
##   hearttrte + cursmoke + prevstrk + prevmi:sysbp + sysbp:age +
##   prevhyp:hdlc + prevmi:diabetes + sysbp:prevhyp + prevhyp:totchol +
##   sysbp:diabetes + prevmi:hdlc + prevmi:prevhyp + age:ldlc +
##   age:cigpday + hdlc:cigpday + prevhyp:bmi + ldlc:bmi + prevmi:totchol +
##   ldlc:prevhyp + sysbp:bpmeds + sysbp:hdlc + hdlc:totchol +
##   totchol:hearttrte + age:hearttrte + diabetes:hdlc + sysbp:hearttrte +
##   bmi:bpmeds + sysbp:sex + ldlc:hdlc + prevmi:bmi + age:bmi +

```

```
##      prevmi:age + sysbp:cursmoke + hdlc:cursmoke + ldlc:cursmoke +
##      prevmi:cigpday + sex:diabetes + prevmi:prevstrk, data = fhds)

# Backward model selection
Mback$call

## lm(formula = I(logit(chdrisk)) ~ sex + totchol + age + sysbp +
##      diabp + cursmoke + cigpday + bmi + diabetes + bpmeds + hearttrte +
##      glucose + prevmi + prevstrk + prevhyp + hdlc + ldlc + I(totchol^2) +
##      I(sysbp^2) + I(diabp^2) + I(bmi^2) + I(hdlc^2) + I(ldlc^2) +
##      sex:totchol + sex:sysbp + sex:glucose + sex:prevstrk + sex:prevhyp +
##      totchol:age + totchol:bpmeds + totchol:hearttrte + totchol:prevmi +
##      totchol:prevstrk + totchol:prevhyp + totchol:hdlc + totchol:ldlc +
##      age:cursmoke + age:bmi + age:hearttrte + age:prevmi + age:prevhyp +
##      age:hdlc + sysbp:diabetes + sysbp:bpmeds + sysbp:hearttrte +
##      sysbp:prevmi + sysbp:prevhyp + diabp:cursmoke + diabp:cigpday +
##      diabp:bmi + diabp:glucose + diabp:prevhyp + diabp:hdlc +
##      cursmoke:bmi + cursmoke:hdlc + cursmoke:ldlc + cigpday:bmi +
##      cigpday:hearttrte + cigpday:glucose + cigpday:prevmi + cigpday:hdlc +
##      bmi:prevmi + bmi:prevhyp + bmi:ldlc + diabetes:prevmi + diabetes:hdlc +
##      bpmeds:glucose + bpmeds:prevstrk + bpmeds:ldlc + hearttrte:glucose +
##      hearttrte:prevmi + glucose:prevmi + prevmi:prevhyp + prevmi:hdlc +
##      prevhyp:ldlc, data = fhds)

beta.fwd = coef(Mfwd)
beta.back = coef(Mback)
beta.step = coef(Mstep)

identical(names(beta.fwd)[names(beta.fwd) %in% names(beta.back)], names(beta.fwd))

## [1] FALSE

identical(names(beta.fwd)[names(beta.fwd) %in% names(beta.step)], names(beta.fwd))

## [1] FALSE

identical(names(beta.back)[names(beta.back) %in% names(beta.step)], names(beta.back))

## [1] FALSE
```

### 3.2 Manual Model Selection

## 4 Model Diagnostics

## 5 Model Selection

## 6 Discussion