Kaitlin's EDA

$Kaitlin\ Maciejewski$ 12/1/2018

Exploratory

```
# save data in separate folder "final_data" one directory up
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)
```

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 2.00 1.00 2.00 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 136.324 222.799 132.00 83.50 295.0 211.50 0.211 diabp 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 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.		n	mean	sd	median	min	max	range	se
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 83.038 11.660 82.00 30.00 150.0 120.00 0.108 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 0.013 bmi 11575 25.877 4.103 25.48 14.43 56.8 42.37 0.038 diabetes 11627 0.046 0.299 0.00 0.00 1.0 1.00 0.00 bpmeds 11034 0.086 0.280 0.00 0.00 1.0 1.00 0.00 glucose 10187 84.125 24.994 80.00 39.00 478.0 439.00 0.248	randid	11627	5004740.917	2900877.440	5006008.00	2448.00	9999312.0	9996864.00	26902.680
age 11627 54.793 9.564 54.00 32.00 81.0 49.00 0.089 sysbp 11627 136.324 22.799 132.00 83.50 295.0 211.55 0.211 diabp 11627 83.038 11.660 82.00 30.00 150.0 120.00 0.108 cursmoke 11627 0.433 0.495 0.00 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 11627 0.046 0.229 0.00 0.00 1.0 1.00 0.002 bpmeds 11621 76.782 12.463 75.00 37.00 220.0 183.00 0.116 <td>sex</td> <td>11627</td> <td>1.568</td> <td>0.495</td> <td>2.00</td> <td>1.00</td> <td>2.0</td> <td>1.00</td> <td>0.005</td>	sex	11627	1.568	0.495	2.00	1.00	2.0	1.00	0.005
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 cursmoke 11627 0.433 0.495 0.00 0.00 1.0 1.00 0.005 cigday 11548 8.250 12.187 0.00 0.00 90.0 90.00 0.013 bmi 11575 25.877 4.103 25.48 14.43 56.8 42.37 0.038 diabetes 11627 0.046 0.299 0.00 0.00 1.0 1.00 0.002 bmeds 11034 0.086 0.280 0.00 0.00 1.0 1.00 0.002 bmeds 11627 0.046 0.280 0.00 0.00 1.0 1.00 0.003 glucose 10187 84.125 24.994 80.00 39.00 478.0 439.00 0.248	totchol	11218	241.162	45.368	238.00	107.00	696.0	589.00	0.428
diabr 11627 83.038 11.660 82.00 30.00 150.0 120.00 0.108 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 0.002 bpmeds 11037 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 2.09 1.00 0.00 1.0 1.00 0.002	age	11627	54.793	9.564	54.00	32.00	81.0	49.00	0.089
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 0.003 bemeds 11034 0.086 0.280 0.00 0.00 1.0 1.00 0.003 heartre 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 40.0 3.00 0.010 prevath 11627 0.054 0.226 0.00 0.00 1.0 1.00 0.002 <	sysbp	11627	136.324	22.799	132.00	83.50	295.0	211.50	0.211
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 heartre 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 30.0 0.010 prevap 11627 0.054 0.226 0.00 0.00 1.0 1.00 0.002 prevap 11627 0.032 0.176 0.00 0.00 1.0 1.00 0.002 p	diabp		83.038	11.660	82.00	30.00	150.0	120.00	0.108
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 heartre 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 0.00 prevap 11627 0.054 0.226 0.00 0.00 1.0 1.00 0.00 prevmi 11627 0.032 0.176 0.00 0.00 1.0 1.00 0.00 prevmir	$\operatorname{cursmoke}$	11627	0.433	0.495	0.00	0.00	1.0	1.00	0.005
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 0.00 prevap 11627 0.032 0.176 0.00 0.00 1.0 1.00 0.00 prevstrk 11627 0.013 0.114 0.00 0.00 1.0 1.00 0.00 prevstrk 11627 0.460 0.498 0.00 0.00 1.0 1.00 0.00 time<	cigpday	11548	8.250	12.187	0.00	0.00	90.0	90.00	0.113
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 prevap 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 prevstrk 11627 0.460 0.498 0.00 0.00 1.0 1.00 0.005 tim	bmi	11575	25.877	4.103	25.48	14.43	56.8	42.37	0.038
heartrete 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 time 11627 1.899 0.807 2.00 1.00 3.0 2.00 0.007 hdlc<	diabetes	11627	0.046	0.209		0.00	1.0	1.00	0.002
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 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	bpmeds	11034	0.086	0.280	0.00	0.00	1.0		0.003
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 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	heartrte	11621	76.782	12.463	75.00	37.00		183.00	0.116
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 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	glucose	10187	84.125	24.994	80.00	39.00	478.0	439.00	0.248
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 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 0.003 <	educ					1.00		3.00	0.010
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 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 <	prevchd		0.072			0.00		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 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 <	prevap			0.226		0.00		1.00	0.002
prevhyp 11627 0.460 0.498 0.00 0.00 1.0 1.00 0.005 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_chd 11627 0.154 0.361 0.00 0.00 1.0 1.00 0.003 <th< td=""><td>prevmi</td><td></td><td></td><td></td><td></td><td></td><td></td><td></td><td>0.002</td></th<>	prevmi								0.002
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 <td< td=""><td>prevstrk</td><td></td><td></td><td>0.114</td><td></td><td></td><td>1.0</td><td></td><td>0.001</td></td<>	prevstrk			0.114			1.0		0.001
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 0.249 0.433 0.00 0.00 1.0 1.00 0.004 hyperten	prevhyp							1.00	
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	$_{ m time}$		1957.019	1758.777	2156.00	0.00		4854.00	16.311
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	period	11627	1.899	0.807	2.00	1.00	3.0	2.00	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	hdlc				48.00	10.00			
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 timeap 11627 7241.557 2477.780 8766.00 0.00 8766.0 8766.00 22.979 timemi 11627 7593.847 2136.730 8766.00 0.00 8766.0 8766.00 19.816									
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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 timeap 11627 7241.557 2477.780 8766.00 0.00 8766.0 8766.00 22.979 timemi 11627 7593.847 2136.730 8766.00 0.00 8766.0 8766.00 19.816									
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timeap 11627 7241.557 2477.780 8766.00 0.00 8766.0 8766.00 22.979 timemi 11627 7593.847 2136.730 8766.00 0.00 8766.0 8766.00 19.816	cvd								
timemi 11627 7593.847 2136.730 8766.00 0.00 8766.0 8766.00 19.816	hyperten								
	timeap								
$\begin{array}{cccccccccccccccccccccccccccccccccccc$								8766.00	
	timemifc	11627	7543.037	2192.120	8766.00	0.00	8766.0	8766.00	20.330

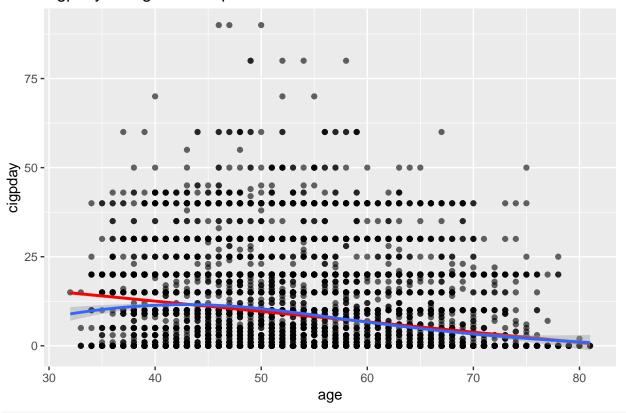
	n	mean	sd	median	min	max	range	se
timechd	11627	7008.154	2641.345	8766.00	0.00	8766.0	8766.00	24.496
timestrk	11627	7660.880	2011.077	8766.00	0.00	8766.0	8766.00	18.651
timecvd	11627	7166.083	2541.668	8766.00	0.00	8766.0	8766.00	23.571
timedth	11627	7854.103	1788.370	8766.00	26.00	8766.0	8740.00	16.585
timehyp	11627	3598.956	3464.165	2429.00	0.00	8766.0	8766.00	32.127

```
# spaghetti
# lorellogram
#

library(ggplot2)

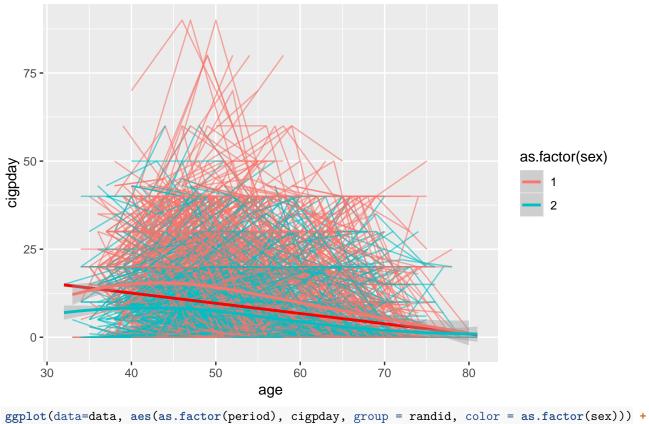
ggplot(data= data, aes(age, cigpday)) +
    geom_point(alpha = .6) +
    geom_smooth(method = 'lm', col = 'red') +
    geom_smooth(method = 'loess') +
    ggtitle("cigpday vs Age Scatterplot")
```

cigpday vs Age Scatterplot



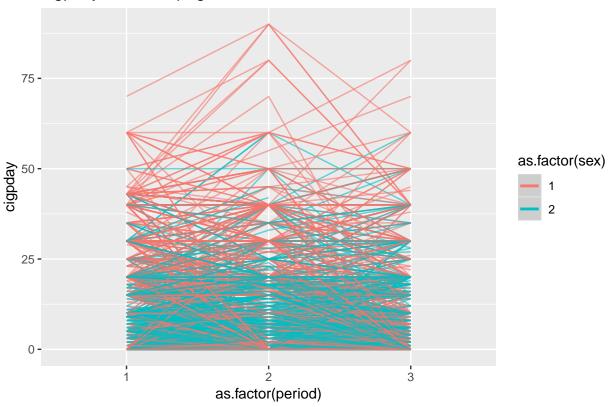
```
ggplot(data=data, aes(age, cigpday, group = randid, color = as.factor(sex))) +
  geom_path(alpha = .6) +
  geom_smooth(aes(group = NULL), method = 'lm', col = 'red') +
  geom_smooth(aes(group = NULL), method = 'loess') +
  ggtitle("cigpday vs Age Spaghetti Plot")
```

cigpday vs Age Spaghetti Plot



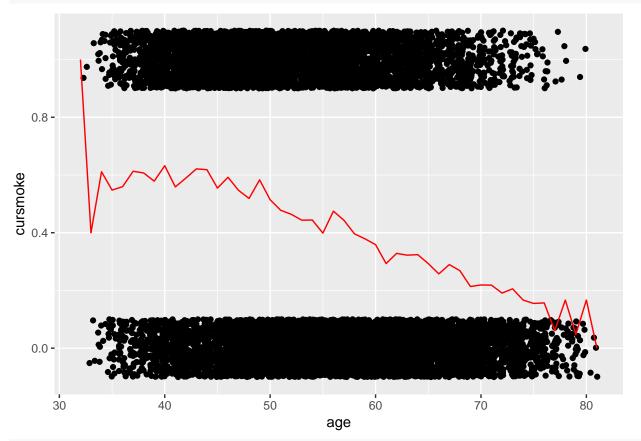
```
ggplot(data=data, aes(as.factor(period), cigpday, group = randid, color = as.factor(sex)))
geom_path(alpha = .6) +
geom_smooth(aes(group = NULL), method = 'lm', col = 'red') +
geom_smooth(aes(group = NULL), method = 'loess') +
ggtitle("cigpday vs time Spaghetti Plot")
```

cigpday vs time Spaghetti Plot

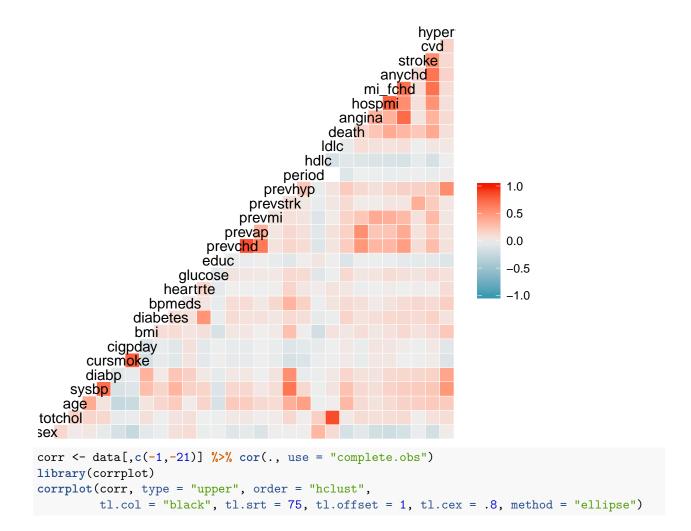


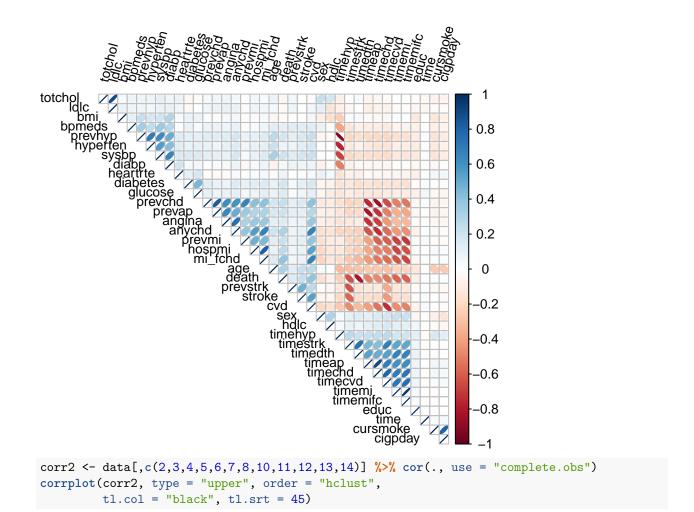
```
###
# data.residual <- data %>% na.omit() %>%
# group_by(age) %>%
# mutate(mean.cigpday = mean(cigpday)) %>%
# ungroup() %>%
# mutate(residuals = cigpday - mean.cigpday) %>%
# group_by(randid) %>%
# mutate(median.residual = median(residuals)) %>%
# ungroup()
# data.stats <- c(min(data.residual$median.residual),</pre>
# quantile(data.residual$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))
###
```

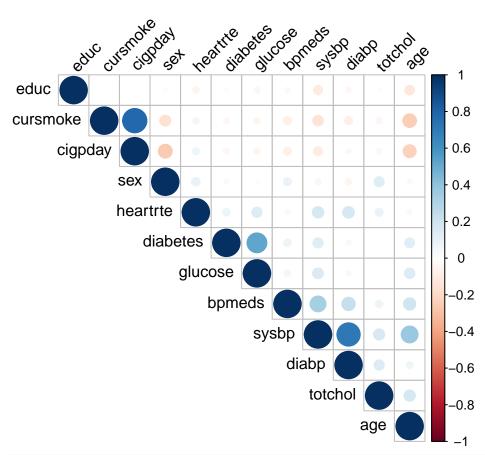
```
ggplot(data, aes(y = cursmoke, x = age)) + geom_jitter(height = 0.1) +
stat_summary(fun.y = 'mean', geom="line", col = 'red')
```



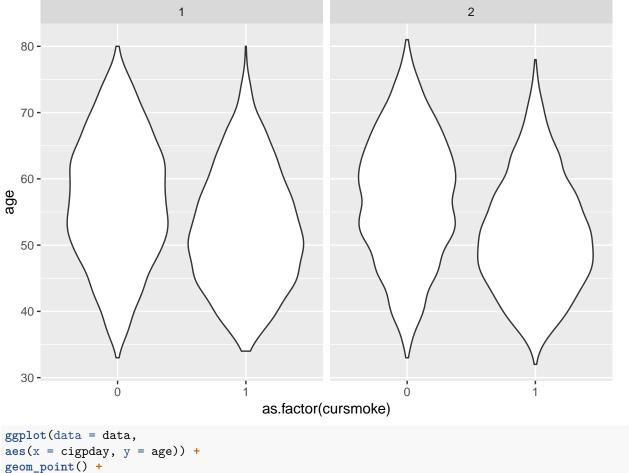
select(data, -c(randid, time,timeap:timehyp)) %>% GGally::ggcorr(.)



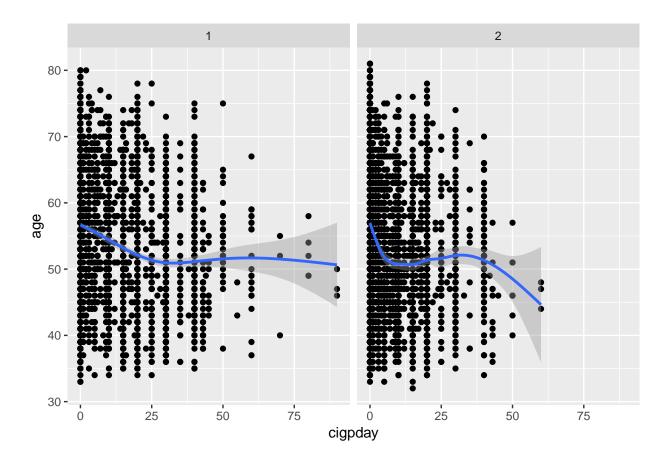




```
ggplot(data = data,
aes(x = as.factor(cursmoke), y = age)) +
geom_violin() +
facet_wrap(~sex)
```



```
aes(x = cigpday, y = age)) +
geom_point() +
  facet_wrap(~sex) +
  geom_smooth()
```



Models

(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 ## 2.80332932 -0.05649341

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

	Estimate	Robust S.E.	Robust z
(Intercept)	2.449	0.116	21.200
age	-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

	Estimate	Robust S.E.	Robust z
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))

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

	Estimate	Robust S.E.	Robust z
(Intercept)	3.451	0.066	52.064
age	-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
##
  gee::gee(formula = cigpday ~ age + as.factor(sex), id = randid,
       na.action = "na.omit", family = poisson, corstr = "unstructured")
##
##
## Summary of Residuals:
                      1Q
                             Median
## -19.386146 -7.058707 -4.393591
                                      7.136533 77.315884
##
##
## Coefficients:
##
                      Estimate Naive S.E.
                                            Naive z Robust S.E. Robust z
```

3.78802175 0.07973074 47.51018 0.065953995 57.43430

(Intercept)

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

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

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

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

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

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

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

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

```
smoke_dias<-gee::gee(cursmoke~diabp, id =randid, family=binomial, corstr = "unstructured", na.action =
## (Intercept) diabp
## 0.83258871 -0.01331811</pre>
```

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

	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.

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

```
smoke_chol<-gee::gee(cursmoke~totchol, id =randid, family=binomial, corstr = "unstructured", na.action
## (Intercept) totchol
## 0.180474997 -0.001869103</pre>
```

	Estimate	Robust S.E.	Robust z
(Intercept) totchol	-0.034	0.107	-0.313
	-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: