Telomere Exploration

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November 3rd, 2023

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Analyses

Purpose:

This is an exploratory analysis of 971 participants, 487 cases with Schizophrenia and 484 controls, with telomere and phenotype data from NGAP Pyschosis. The original sample has 1000 participants, evenly split between cases and controls, but 29 participants were not included in the analysis due to insufficient DNA or failed samples, leaving a total of n=971 participants.

Here, we explore associations between Telomere length and Schizophrenia, age, and sex. We also explore other associations on available phenotypic data.

Methods:

We viewed association between phenotype variables with sufficient prevalence with log telomere length (ltl). We used linear regression to view associations between said variables and ltl at first without adjusting for age, and then adjusting for age. Additionally, we ran an ANOVA test with models with and without interaction terms to determine whether there was any functional difference between these models.

Results

In the unadjusted model, case status (is_case), sex (msex), and educational attainment ($educ_ord$) were associated with ltl (p<0.05).

Of particular note, congruent with previous literature, men on average had shorter telomere lengthboth adjusting and not adjusting for age (T-test, p = 0.0018).

In the adjusted linear regression model, controlling for age and sex, schizophrenia ($is_case=1$) is associated with LTL (p=0.016)

Data

Libraries

```
library(tidyverse)
library(knitr)
library(batchtma)
library(ggplot2)
library(stargazer)
library(tableone)
library(arsenal)
```

```
library(ggpubr)
library(rstatix)
library(broom)
library(ggrepel)
library(knitr)
library(kableExtra)
library(gtsummary)
library(table1)
library(flextable)
library(janitor)
library(table1)
library(ggeasy)
library(rcompanion)
library(nnet)
#devtools::install_github("thomasp85/patchwork")
library(patchwork)
```

Data specifics

- I was sent the data specified in tel_data from Steven Senese seenese@hsph.harvard.edu on 09-26-2023. This data includes the telomere lengths. This is a modified version of the *Koenen Sorted Data*, that included 5 sets of data based on the plates run for telomere analysis. I copied those into one document, tel_data , but specified in column Set which plate it came from.
- I was sent the data specified in manifest from Patrice Soule spsoule@hsph.harvard.edu on 10-16-2023.
- The data specified in *freeze* is from the latest data freeze file located in DropBox NeuroGAP Psychosis called *NeuroGAP-P_Release8_Kenya_as-of-2023-08-29*

Loading and merging data

```
#Adjust to fit your computer

path_analytic <- "/Users/ham593/Dropbox (Harvard University)/NeuroGAP-Psychosis/Telomeres/Data/Analytic
path_raw <- "/Users/ham593/Dropbox (Harvard University)/NeuroGAP-Psychosis/Telomeres/Data/Raw Data"

#Load in data-----
#Analytical Data
setwd(path_analytic)
tel_data <- read.csv('Telomeredata_new.csv')
freeze <- read.csv('Copy of NeuroGAP_DataFreeze8.csv')

#Raw Data
setwd(path_raw)
manifest <- read.csv('Full manifest Broad PDO-31674 Plate Map.csv')

colnames(manifest)[colnames(manifest) == "Collaborator.Participant.ID"] <- "subjid"
colnames(freeze)[colnames(freeze) == "subj_id"] <- "subjid"

#Remove all Reference/QC rows and Insufficient DNA
tel_data <- tel_data[grep1("^SM-", tel_data$Sample.ID),]</pre>
```

```
tel_data <- tel_data[!grepl("Failed", tel_data$Tel.CT.1),]
tel_data <- tel_data[!grepl("Insufficient DNA", tel_data$Tel.CT.1),]

#MERGE
tel_all <- merge(manifest, tel_data, by="Sample.ID")
tel_all <- merge(tel_all, freeze, by="subjid")

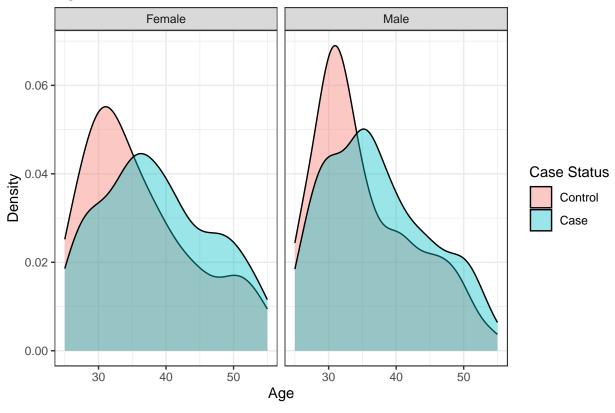
#We found that Age was duplicated, the below will remove on of the Age variables.
#Remove duplicated columns
tel_all <- tel_all %>%
    subset(select=which(!duplicated(names(.))))

setwd("/Users/ham593/Dropbox (Harvard University)/NeuroGAP-Psychosis/Data Working Group (DAWG)/Data/Dat ngap_7 <- read.csv("NeuroGAP-P_Release7_Final.csv")
setwd(path_analytic)</pre>
```

Distribution of age among cases and controls

```
ggplot(tel_all, aes(x=age_at_iview, group = is_case, fill=is_case))+
  facet_wrap(~msex) +
  geom_density(alpha=0.4) +
  xlab("Age") + ylab("Density") +
  ggtitle("Age Distribution of Cases and Controls") +
  ggeasy::easy_center_title() +
  labs(fill='Case Status') +
  #scale_y_continuous(labels = scales::percent_format(scale = 1)) +
  theme_bw() +
  scale_fill_discrete(name = "Case Status", labels = c("Control", "Case"))
```

Age Distribution of Cases and Controls



Analysis

:			Dependent	variable:	
:		·		:1	
: :	(1))	(2)	(3)	(4)
Age	-0.00	02*	0.002	-0.009	-0.002
	(0.00	J1)	(0.009)	(0.018)	(0.065)
I(Age2)			-0.0001		0.0001
			(0.0001)		(0.002)
: : T(A==0 E)				2 222	
I(Age0.5)				0.096	
:				(0.218)	
: I(Age3)					-0.00000
: I(Mgeo)					(0.00001)
:					(0.00
Constant	0.02	22	-0.052	-0.271	0.003
:	(0.03	35)	(0.172)	(0.664)	(0.814)
: : Observations	 97:	 1	971	 971	971
R2	0.00		0.003	0.003	0.003
Adjusted R2	0.00		0.001	0.001	0.0002
•				0.226 (df = 968)	
F Statistic) 1.577 (df = 2; 968)	

Bivariate Analysis (Unadjusted)

Age was not statistically associated with LTL (p=0.0856)

Case status (p=0.034) and sex (p=0.0012) are associated with LTL.

Participants who completed College from educ_ord were also significantly associated with LTL (p=3e-04)... I'm not sure what to make of this.

The following variables were shown to be associated with LTL:

- 1. Binary
- msex, p = 0.0012
- is_case, p= 0.034
- 2. Ordinal
- educ ord

Secondary, p=0.1236 College, p= 3e-04

```
results=NULL
for (ii in 1:length(vars)) {
  # Create a formula for the linear regression
  fmla <- as.formula(paste("ltl ~", vars[ii]))</pre>
  # Fit the linear regression model
  model <- lm(fmla, data = tel_all)</pre>
  fit<-lm(fmla, data= tel_all)</pre>
  res<-coef(summary(fit))</pre>
  ci<-round(confint(fit),3)</pre>
  if(nrow(res)==2){
rownames(res)[2]<-vars[ii]</pre>
rownames(ci)[2]<-vars[ii]</pre>
est<-round(coef(fit)[2:length(coef(fit))],3)</pre>
p<-coef(summary(fit))[2:nrow(coef(summary(fit))), "Pr(>|t|)"]
if(p<=0.05){est<-paste(est, "*", sep="")}</pre>
if(p<=0.001){est<-paste(est, "*", sep="")}</pre>
if(p<=0.0001){est<-paste(est, "*", sep="")}</pre>
ci<-paste(ci[vars[ii],], collapse=", ")</pre>
ci<-paste("(", ci, ")", sep="")
r2<-round(summary(fit)$adj.r.squared,3)
p < -round(p, 4)
p[p<0.0001]<-"<0.001"
results<-rbind(results, c(vars[ii], est, ci, p, r2))
  if(nrow(res)>2){
varLine<-c(vars[ii], rep("", 3), round(summary(fit)$adj.r.squared,3))</pre>
res<-res[grep(vars[ii], rownames(res)),]</pre>
```

```
rownames(res)<-gsub(vars[ii], "", rownames(res))</pre>
est<-round(res[, "Estimate"],3)</pre>
ind<-which(apply(res, 1, function(x) x["Pr(>|t|)"]<0.05))
est[ind] <-paste(est[ind], "*", sep="")</pre>
ind<-which(apply(res, 1, function(x) x["Pr(>|t|)"]<0.001))
est[ind] <-paste(est[ind], "*", sep="")</pre>
ind<-which(apply(res, 1, function(x) x["Pr(>|t|)"]<0.0001))
est[ind] <-paste(est[ind], "*", sep="")</pre>
ci<-ci[grep(vars[ii], rownames(ci)),]</pre>
rownames(ci)<-gsub(vars[ii], "", rownames(ci))</pre>
ci<-apply(ci, 1, function(x) paste(x, collapse=", "))</pre>
ci<-paste("(", ci, ")", sep="")</pre>
p<-res[, "Pr(>|t|)"]
p < -round(p, 4)
p[which(p<=0.0001)]<-"<0.0001"
temp<-cbind(rownames(res), est)</pre>
temp<-cbind(temp, ci)</pre>
temp<-cbind(temp, p)</pre>
temp<-cbind(temp, rep("", nrow(res)))</pre>
temp<-rbind(varLine, temp)</pre>
results<-rbind(results, temp)</pre>
22}
}
rownames(results)<-c(1:nrow(results))</pre>
colnames(results)<-c("Variable", "Beta", "CI", "p", "Adj R2")</pre>
kable(results, booktabs = TRUE, longtable = TRUE,
  caption = "Bivariate associations between phenotypes and LTL") %>%
  kable_styling(latex_options = c("hold_position", "repeat_header"))
```

Table 1: Bivariate associations between phenotypes and LTL

Variable	Beta	CI	p	Adj R2
age_at_iview	-0.002	(-0.003, 0)	0.0856	0.002
$_{ m bmi}$	0.001	(-0.002, 0.005)	0.4421	0
is_case	0.031*	(0.002, 0.059)	0.034	0.004
msex	0.049*	(0.019, 0.079)	0.0012	0.01
$assist_khat$	-0.018	(-0.049, 0.013)	0.249	0
khat_3cat				-0.001
Irregular Users	-0.019	(-0.055, 0.017)	0.3078	
Regular Users	-0.017	(-0.063, 0.029)	0.4743	
assist_alcohol	-0.016	(-0.045, 0.012)	0.2649	0
hbp				-0.001
Low BP	0.102	(-0.212, 0.416)	0.5247	
High BP	0.028	(-0.033, 0.089)	0.3642	
$assist_tobacco$	-0.02	(-0.057, 0.017)	0.2811	0
$tobacco_3cat$				0
Irregular Users	-0.01	(-0.055, 0.036)	0.6767	
Regular Users	-0.037	(-0.092, 0.019)	0.1934	
assist_cannabis	-0.036	(-0.108, 0.035)	0.3223	0

Table 1: Bivariate associations between phenotypes and LTL $\left(continued\right)$

Variable	Beta	CI	p	Adj R2
cannabis_3cat				-0.001
Irregular Users	-0.031	(-0.106, 0.045)	0.4254	
Regular Users	-0.086	(-0.308, 0.137)	0.4492	
$cidi_q9$	-0.061	(-0.156, 0.034)	0.2101	0.001
obese	0.027	(-0.025, 0.078)	0.3042	0
bmi_bin				-0.002
Underweight	0.014	(-0.048, 0.076)	0.6516	
Overweight	0.008	(-0.027, 0.043)	0.6548	
Obese	0.03	(-0.023, 0.082)	0.2651	
hbp				-0.001
Low BP	0.102	(-0.212, 0.416)	0.5247	
High BP	0.028	(-0.033, 0.089)	0.3642	0
bpm_bin				0
Slow Heartrate	-0.023	(-0.124, 0.077)	0.6468	
Fast Heartrate	0.044	(-0.017, 0.105)	0.1612	
$cidi_q15$	0	(0, 0)	0.7996	-0.001
age_group	0.014	(0 040 0 00)	0.4161	0.002
31-40	-0.014	(-0.049, 0.02)	0.4161	
41-50	-0.027	(-0.067, 0.013)	0.1909	
51-60	-0.071*	(-0.136, -0.007)	0.0308	
educ_ord	0.007	(0 000 0 007)	0.1000	0.012
Secondary	-0.027 -0.066**	(-0.062, 0.007) (-0.101, -0.03)	0.1236	
College	-0.000	(-0.101, -0.03)	3e-04	
Set	o o o skaled	(0 100 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		0.099
В	-0.09***	(-0.129, -0.051)	< 0.0001	
C	0.031	(-0.007, 0.07)	0.1125	
D E	0.114***	(0.075, 0.152)	<0.0001 0.3066	
<u> </u>	-0.040	(-0.135, 0.042)	0.5000	

 ${\it \#kable (results, caption="Bivariate associations between phenotypes and LTL.")}$

Rosner Batch Correction

```
# Rosner batch correction
# Step 1: biomarker = +\Sigma[i(batchi)] + (covariate1) + (covariate2)
fit<-lm(ltl~Set, data=tel_all)</pre>
# Step 2: calculate the average for batch: avgb=(\Sigma i)/N
                                                          (Note: N=0)
avgb<-mean(coef(fit)[grep("Set", names(coef(fit)))])</pre>
# Step 3: recalibrate biomarker levelts
                   if batch=1 then adj_biomarker=orig_biomarker-(1-avgb)
#
        if batch=2 then adj_biomarker=orig_biomarker-(2-avgb)
#
        if batch=N then adj_biomarker=orig_biomarker-(N-avgb) (Note: N=0)
#### with package
#install.packages("batchtma")
ros_df <- tel_all %>% select(is_case, subjid, ltl, Set, Age)
ros <- adjust_batch(</pre>
 data = ros_df,
 markers = ltl,
 batch = Set,
 method = simple
```

Multivariate Analysis

After adjusting for age, the following variables were shown to be associated with LTL:

```
    Binary
    is_case, p=0.0177
    msex, p=8e-04
    Ordinal
    educ_ord
    Secondary, p=0.1119
    College p<0.0001</li>
```

```
results=NULL
vars<-vars[!vars%in%"age_at_iview"]</pre>
for(ii in 1:length(vars)){
  fmla<-as.formula(paste("ltl ~ age_at_iview +", vars[ii]))</pre>
  fit<-lm(fmla, data=tel_all)</pre>
  res<-coef(summary(fit))</pre>
  res<-res[-2, ]
  ci<-round(confint(fit),3)</pre>
  ci<-ci[-2,]
  if(nrow(res)==2){
    rownames(res)[2]<-vars[ii]</pre>
    rownames(ci)[2]<-vars[ii]</pre>
    est<-round(res[2, "Estimate"], 3)</pre>
    p<-res[2, "Pr(>|t|)"]
    if(p<=0.05){est<-paste(est, "*", sep="")}</pre>
    if(p<=0.001){est<-paste(est, "*", sep="")}</pre>
    if(p<=0.0001){est<-paste(est, "*", sep="")}</pre>
    ci<-paste(ci[vars[ii],], collapse=", ")</pre>
    ci<-paste("(", ci, ")", sep="")</pre>
    r2<-round(summary(fit)$adj.r.squared,3)
    p < -round(p, 4)
    p[p<0.0001]<-"<0.001"
    results<-rbind(results, c(vars[ii], est, ci, p, r2))</pre>
}
    if(nrow(res)>2){
      varLine<-c(vars[ii], rep("", 3), round(summary(fit)$adj.r.squared,3))</pre>
      res<-res[grep(vars[ii], rownames(res)),]</pre>
      rownames(res)<-gsub(vars[ii], "", rownames(res))</pre>
      est<-round(res[, "Estimate"],3)</pre>
      ind<-which(apply(res, 1, function(x) x["Pr(>|t|)"]<0.05))
      est[ind] <-paste(est[ind], "*", sep="")</pre>
      ind \leftarrow which (apply (res, 1, function(x) x ["Pr(>|t|)"] < 0.001))
      est[ind] <-paste(est[ind], "*", sep="")</pre>
```

```
ind \leftarrow which (apply (res, 1, function(x) x["Pr(>|t|)"] < 0.0001))
      est[ind] <-paste(est[ind], "*", sep="")</pre>
      ci<-ci[grep(vars[ii], rownames(ci)),]</pre>
      rownames(ci)<-gsub(vars[ii], "", rownames(ci))</pre>
      ci<-apply(ci, 1, function(x) paste(x, collapse=", "))</pre>
      ci<-paste("(", ci, ")", sep="")</pre>
      p<-res[, "Pr(>|t|)"]
      p < -round(p, 4)
      p[which(p<=0.0001)]<-"<0.0001"
      temp<-cbind(rownames(res), est)</pre>
      temp<-cbind(temp, ci)</pre>
      temp<-cbind(temp, p)</pre>
      temp<-cbind(temp, rep("", nrow(res)))</pre>
      temp<-rbind(varLine, temp)</pre>
  results<-rbind(results, temp)</pre>
    }
}
rownames(results)<-c(1:nrow(results))</pre>
colnames(results)<-c("Variable", "Beta", "CI", "p", "Adj R2")</pre>
kable(results, booktabs = TRUE, longtable = TRUE,
  caption = "Associations between phenotype and LTL adjusted for age") %>%
  kable_styling(latex_options = c("hold_position", "repeat_header"))
```

Table 2: Associations between phenotype and LTL adjusted for age

Variable	Beta	CI	p	Adj R2
bmi	0.002	(-0.002, 0.005)	0.2905	0.002
is_case	0.035*	(0.006, 0.063)	0.0177	0.007
msex	0.051**	(0.021, 0.081)	8e-04	0.013
$assist_khat$	-0.018	(-0.049, 0.012)	0.2389	0.002
khat_3cat				0.001
Irregular Users	-0.018	(-0.054, 0.018)	0.3254	
Regular Users	-0.019	(-0.065, 0.027)	0.4167	
$assist_alcohol$	-0.016	(-0.045, 0.013)	0.2701	0.002
hbp				0.002
Low BP	0.11	(-0.203, 0.424)	0.4901	
High BP	0.04	(-0.022, 0.102)	0.2049	
$assist_tobacco$	-0.02	(-0.057, 0.016)	0.2794	0.002
$tobacco_3cat$				0.002
Irregular Users	-0.011	(-0.056, 0.034)	0.6437	
Regular Users	-0.035	(-0.09, 0.02)	0.2122	
$assist_cannabis$	-0.042	(-0.114, 0.03)	0.2486	0.002
$cannabis_3cat$				0.002
Irregular Users	-0.037	(-0.112, 0.039)	0.3418	
Regular Users	-0.093	(-0.316, 0.129)	0.4108	
$cidi_q9$	-0.051	(-0.148, 0.045)	0.2972	0.001

Table 2: Associations between phenotype and LTL adjusted for age (continued)

Variable	Beta	CI	p	Adj R2
obese bmi_bin	0.033	(-0.019, 0.085)	0.2153	$0.003 \\ 0.001$
Underweight	0.015	(-0.047, 0.077)	0.6312	
Overweight	0.012	(-0.023, 0.047)	0.5068	
Obese	0.037	(-0.016, 0.09)	0.1723	
hbp				0.002
Low BP	0.11	(-0.203, 0.424)	0.4901	
High BP	0.04	(-0.022, 0.102)	0.2049	
bpm_bin				0.003
Slow Heartrate	-0.02	(-0.12, 0.08)	0.6939	
Fast Heartrate	0.049	(-0.013, 0.11)	0.1203	
$cidi_q15$	0	(0, 0)	0.7982	0.001
age_group				0.002
31-40	-0.031	(-0.083, 0.022)	0.2489	
41-50	-0.067	(-0.171, 0.037)	0.2072	
51-60	-0.128	(-0.279, 0.023)	0.0955	
$educ_ord$, ,		0.016
Secondary	-0.028	(-0.063, 0.007)	0.1119	
College	-0.071***	(-0.107, -0.035)	< 0.0001	
Set				0.103
В	-0.093***	(-0.132, -0.054)	< 0.0001	
\mathbf{C}	0.032	(-0.007, 0.07)	0.1073	
D	0.112***	(0.074, 0.151)	< 0.0001	
E	-0.048	(-0.137, 0.04)	0.2841	

Tables

Case Status

```
label(tel all$bmi bin) <- "BMI"</pre>
label(tel_all$alcohol_3cat) <- "Alcohol Use"</pre>
label(tel_all$khat_3cat) <- "Khat Use"</pre>
label(tel_all$cannabis_3cat) <- "Cannabis Use"</pre>
label(tel_all$tobacco_3cat) <- "Tobacco Use"</pre>
label(tel_all$cidi_q1) <- "Arthritis and Rheumatism"
label(tel_all$cidi_q2) <- "Chronic back or neck problems"
label(tel_all$cidi_q3) <- "Frequent or severe headaches"</pre>
label(tel_all$cidi_q4) <- "Any other chronic pain"</pre>
label(tel_all$cidi_q5) <- "Seasonal allergies"</pre>
label(tel_all$cidi_q6) <- "Stroke"</pre>
label(tel_all$cidi_q7) <- "Heart attack"</pre>
label(tel_all$cidi_q8) <- "Heart disease"</pre>
label(tel_all$cidi_q9) <- "High blood pressure"</pre>
label(tel_all$cidi_q10) <- "Asthma"</pre>
label(tel_all$cidi_q11) <- "Tuberculosis"</pre>
label(tel_all$cidi_q12) <- "Other chronic lung disease"</pre>
label(tel_all$cidi_q13) <- "Diabetes"</pre>
label(tel_all$cidi_q14) <- "Stomach or intestine"</pre>
label(tel_all$cidi_q15) <- "HIV/AIDS"</pre>
label(tel_all$cidi_q16) <- "Epilepsy/Seizure"</pre>
label(tel_all$cidi_q17) <- "Cancer"</pre>
tbl <- table1(~ age_at_iview + msex + educ_ord +
                    bmi_bin + alcohol_3cat + khat_3cat + cannabis_3cat + tobacco_3cat +
                   cidi_q1 + cidi_q2 + cidi_q3 + cidi_q4 + cidi_q5 + cidi_q6 +
                   cidi_q7 + cidi_q8 + cidi_q9 + cidi_q10 + cidi_q11 + cidi_q12 +
                 cidi_q13 + cidi_q14 + cidi_q15 + cidi_q16 + cidi_q17 | is_case_f, data=tel_all, overall=
tbl <- t1flex(tbl)
tbl
```

	Case (N=487)	Control (N=484)
Age		
Mean (SD)	37.8 (7.88)	35.7 (7.68)
Median [Min, Max]	36.0 [25.0, 55.0]	33.5 [25.0, 55.0]
Sex		
Male	317 (65.1%)	315 (65.1%)
Female	170 (34.9%)	169 (34.9%)
Education		
Primary or less	161 (33.1%)	135 (27.9%)

	Case (N=487)	Control (N=484)
Secondary	212 (43.5%)	138 (28.5%)
College	114 (23.4%)	211 (43.6%)
ВМІ		
Underweight	28 (5.7%)	28 (5.8%)
Normal Weight	311 (63.9%)	307 (63.4%)
Obese	48 (9.9%)	33 (6.8%)
Overweight	100 (20.5%)	116 (24.0%)
Alcohol Use		
Never Users	233 (47.8%)	201 (41.5%)
Irregular Users	215 (44.1%)	188 (38.8%)
Regular Users	39 (8.0%)	95 (19.6%)
Khat Use		
Never Users	262 (53.8%)	404 (83.5%)
Irregular Users	144 (29.6%)	52 (10.7%)
Regular Users	81 (16.6%)	28 (5.8%)
Cannabis Use		
Never Users	453 (93.0%)	478 (98.8%)
Irregular Users	30 (6.2%)	6 (1.2%)
Regular Users	4 (0.8%)	0 (0%)
Tobacco Use		
Never Users	307 (63.0%)	484 (100%)
Irregular Users	110 (22.6%)	0 (0%)
Regular Users	70 (14.4%)	0 (0%)
Arthritis and Rheumatism		
0	443 (91.0%)	484 (100%)
1	42 (8.6%)	0 (0%)
777	2 (0.4%)	0 (0%)
Chronic back or neck problems		
0	423 (86.9%)	484 (100%)
1	63 (12.9%)	0 (0%)
777	1 (0.2%)	0 (0%)
Frequent or severe headaches		
0	406 (83.4%)	484 (100%)
1	79 (16.2%)	0 (0%)

	Case (N=487)	Control (N=484)
777	2 (0.4%)	0 (0%)
Any other chronic pain		
0	472 (96.9%)	484 (100%)
1	15 (3.1%)	0 (0%)
Seasonal allergies		
0	461 (94.7%)	484 (100%)
1	26 (5.3%)	0 (0%)
Stroke		
0	484 (99.4%)	484 (100%)
1	1 (0.2%)	0 (0%)
777	2 (0.4%)	0 (0%)
Heart attack		
0	474 (97.3%)	484 (100%)
1	12 (2.5%)	0 (0%)
777	1 (0.2%)	0 (0%)
Heart disease		
0	476 (97.7%)	484 (100%)
1	8 (1.6%)	0 (0%)
777	3 (0.6%)	0 (0%)
High blood pressure		
0	458 (94.0%)	484 (100%)
1	22 (4.5%)	0 (0%)
777	7 (1.4%)	0 (0%)
Asthma		
0	477 (97.9%)	484 (100%)
1	8 (1.6%)	0 (0%)
777	2 (0.4%)	0 (0%)
Tuberculosis		
0	457 (93.8%)	484 (100%)
1	27 (5.5%)	0 (0%)
777	3 (0.6%)	0 (0%)
Other chronic lung disease		
0	475 (97.5%)	484 (100%)
1	9 (1.8%)	0 (0%)

	Case (N=487)	Control (N=484)
777	3 (0.6%)	0 (0%)
Diabetes		
0	465 (95.5%)	484 (100%)
1	9 (1.8%)	0 (0%)
777	13 (2.7%)	0 (0%)
Stomach or intestine		
0	451 (92.6%)	484 (100%)
1	34 (7.0%)	0 (0%)
777	2 (0.4%)	0 (0%)
HIV/AIDS		
0	440 (90.3%)	484 (100%)
1	10 (2.1%)	0 (0%)
777	37 (7.6%)	0 (0%)
Epilepsy/Seizure		
0	480 (98.6%)	484 (100%)
1	4 (0.8%)	0 (0%)
777	3 (0.6%)	0 (0%)
Cancer		
0	476 (97.7%)	484 (100%)
777	11 (2.3%)	0 (0%)

Tables

Set

A	В	С	D	Е
(N=235)	(N=236)	(N=238)	(N=237)	(N=25)
37.2 (7.73)	35.7 (7.56)	37.4 (8.29)	36.5 (7.79)	36.2 (7.22)
36.0 [25.0, 55.0]	34.0 [25.0, 55.0]	36.0 [25.0, 55.0]	35.0 [25.0, 55.0]	36.0 [25.0, 52.0
142 (60.4%)	170 (72.0%)	145 (60.9%)	158 (66.7%)	17 (68.0%)
93 (39.6%)	66 (28.0%)	93 (39.1%)	79 (33.3%)	8 (32.0%)
63 (26.8%)	75 (31.8%)	76 (31.9%)	78 (32.9%)	4 (16.0%)
96 (40.9%)	83 (35.2%)	75 (31.5%)	87 (36.7%)	9 (36.0%)
76 (32.3%)	78 (33.1%)	87 (36.6%)	72 (30.4%)	12~(48.0%)
9 (3.8%)	23 (9.7%)	11 (4.6%)	11 (4.6%)	2 (8.0%)
149 (63.4%)	159 (67.4%)	147 (61.8%)	151 (63.7%)	12 (48.0%)
28 (11.9%)	17 (7.2%)	17 (7.1%)	17 (7.2%)	2 (8.0%)
49~(20.9%)	$37\ (15.7\%)$	$63\ (26.5\%)$	58~(24.5%)	9 (36.0%)
107 (45.5%)	93 (39.4%)	122 (51.3%)	98 (41.4%)	14 (56.0%)
92 (39.1%)	106 (44.9%)	84 (35.3%)	111 (46.8%)	10 (40.0%)
36 (15.3%)	37 (15.7%)	32 (13.4%)	28 (11.8%)	1 (4.0%)
179 (76.2%)	150 (63.6%)	168 (70.6%)	153 (64.6%)	16 (64.0%)
43 (18.3%)	48 (20.3%)	49 (20.6%)	52 (21.9%)	4 (16.0%)
13 (5.5%)	38 (16.1%)	21 (8.8%)	32 (13.5%)	5 (20.0%)
226 (96.2%)	224 (94.9%)	231 (97.1%)	225 (94.9%)	25 (100%)
9 (3.8%)	10 (4.2%)	6 (2.5%)	11 (4.6%)	0 (0%)
0 (0%)	2 (0.8%)	1 (0.4%)	1 (0.4%)	0 (0%)
191 (81.3%)	$183\ (77.5\%)$	206~(86.6%)	191~(80.6%)	20~(80.0%)
33 (14.0%)	29 (12.3%)	14 (5.9%)	33 (13.9%)	1 (4.0%)
11 (4.7%)	24 (10.2%)	18 (7.6%)	13 (5.5%)	4 (16.0%)
226 (96.2%)	229 (97.0%)	224 (94.1%)	223 (94.1%)	25 (100%)
9 (3.8%)	7 (3.0%)	12 (5.0%)	14 (5.9%)	0 (0%)
0 (0%)	0 (0%)	2~(0.8%)	0 (0%)	0 (0%)
218 (92.8%)	223 (94.5%)	222 (93.3%)	219 (92.4%)	25 (100%)
17 (7.2%)	13 (5.5%)	15 (6.3%)	18 (7.6%)	0 (0%)
0 (0%)	0 (0%)	1 (0.4%)	0 (0%)	0 (0%)
218 (92.8%)	218 (92.4%)	216 (90.8%)	214 (90.3%)	24 (96.0%)
17 (7.2%)	17 (7.2%)	21 (8.8%)	23 (9.7%)	1 (4.0%)
0 (0%)	1 (0.4%)	1 (0.4%)	0 (0%)	0 (0%)
232 (98.7%)	233 (98.7%)	233 (97.9%)	233 (98.3%)	25 (100%)
3 (1.3%)	3 (1.3%)	5 (2.1%)	4 (1.7%)	0 (0%)
228 (97.0%)	231 (97.9%)	233 (97.9%)	228 (96.2%)	25 (100%)
7 (3.0%)	5 (2.1%)	5 (2.1%)	9 (3.8%)	0 (0%)
	17			
234 (99.6%)	235 (99.6%)	238 (100%)	236 (99.6%)	25 (100%)
0 (0%)	0 (0%)	0 (0%)	1 (0.4%)	0 (0%)
1 (0.4%)	1 (0.4%)	0 (0%)	0 (0%)	0 (0%)
	(N=235) 37.2 (7.73) 36.0 [25.0, 55.0] 142 (60.4%) 93 (39.6%) 63 (26.8%) 96 (40.9%) 76 (32.3%) 9 (3.8%) 149 (63.4%) 28 (11.9%) 49 (20.9%) 107 (45.5%) 92 (39.1%) 36 (15.3%) 179 (76.2%) 43 (18.3%) 13 (5.5%) 226 (96.2%) 9 (3.8%) 0 (0%) 191 (81.3%) 33 (14.0%) 11 (4.7%) 226 (96.2%) 9 (3.8%) 0 (0%) 218 (92.8%) 17 (7.2%) 0 (0%) 218 (92.8%) 17 (7.2%) 0 (0%) 218 (92.8%) 17 (7.2%) 0 (0%) 232 (98.7%) 3 (1.3%) 228 (97.0%) 7 (3.0%) 234 (99.6%)	(N=235) (N=236) 37.2 (7.73) 35.7 (7.56) 36.0 [25.0, 55.0] 142 (60.4%) 170 (72.0%) 66 (28.0%) 63 (26.8%) 75 (31.8%) 96 (40.9%) 83 (35.2%) 76 (32.3%) 78 (33.1%) 9 (3.8%) 23 (9.7%) 149 (63.4%) 159 (67.4%) 28 (11.9%) 17 (7.2%) 49 (20.9%) 37 (15.7%) 107 (45.5%) 93 (39.4%) 92 (39.1%) 106 (44.9%) 36 (15.3%) 37 (15.7%) 179 (76.2%) 150 (63.6%) 43 (18.3%) 48 (20.3%) 13 (5.5%) 38 (16.1%) 226 (96.2%) 224 (94.9%) 9 (3.8%) 10 (4.2%) 2 (0.8%) 191 (81.3%) 183 (77.5%) 33 (14.0%) 29 (12.3%) 11 (4.7%) 24 (10.2%) 226 (96.2%) 229 (97.0%) 9 (3.8%) 7 (3.0%) 0 (0%) 0 (0%) 218 (92.8%) 223 (94.5%) 17 (7.2%) 13 (5.5%) 0 (0%) 0 (0%) 218 (92.8%) 218 (92.4%) 17 (7.2%) 17 (7.2%) 13 (5.5%) 0 (0%) 0 (0%) 218 (92.8%) 218 (92.4%) 17 (7.2%) 17 (7.2%) 17 (7.2%) 17 (7.2%) 10 (0%) 228 (97.0%) 231 (97.9%) 7 (3.0%) 5 (2.1%) 17 234 (99.6%) 235 (99.6%) 0 (0%) 17 234 (99.6%) 235 (99.6%) 0 (0%)	(N=235) (N=236) (N=238) 37.2 (7.73) 35.7 (7.56) 37.4 (8.29) 36.0 [25.0, 55.0] 34.0 [25.0, 55.0] 36.0 [25.0, 55.0] 142 (60.4%) 170 (72.0%) 145 (60.9%) 93 (39.6%) 66 (28.0%) 93 (39.1%) 63 (26.8%) 75 (31.8%) 76 (31.9%) 96 (40.9%) 83 (35.2%) 75 (31.5%) 76 (32.3%) 78 (33.1%) 87 (36.6%) 9 (3.8%) 23 (9.7%) 11 (4.6%) 149 (63.4%) 159 (67.4%) 147 (61.8%) 28 (11.9%) 17 (7.2%) 17 (7.1%) 49 (20.9%) 37 (15.7%) 63 (26.5%) 107 (45.5%) 93 (39.4%) 122 (51.3%) 92 (39.1%) 106 (44.9%) 84 (35.3%) 36 (15.3%) 37 (15.7%) 63 (26.5%) 179 (76.2%) 150 (63.6%) 168 (70.6%) 43 (18.3%) 48 (20.3%) 49 (20.6%) 13 (5.5%) 38 (16.1%) 21 (8.8%) 226 (96.2%) 224 (94.9%) 231 (97.1%) 9 (3.8%) 10 (4.2%)	$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$

	A	В	С	D	Е
	(N=235)	(N=236)	(N=238)	(N=237)	(N=25)
Asthma	(04)		(04)	(04)	/ 04)
0	235 (100%)	230 (97.5%)	236 (99.2%)	236 (99.6%)	24 (96.0%)
1	0 (0%)	4 (1.7%)	2(0.8%)	1 (0.4%)	1 (4.0%)
777	0 (0%)	2 (0.8%)	0 (0%)	0 (0%)	0 (0%)
Tuberculosis	(04)	((((04)
0	229 (97.4%)	229 (97.0%)	229 (96.2%)	231 (97.5%)	23 (92.0%)
1	6 (2.6%)	5 (2.1%)	9 (3.8%)	5 (2.1%)	2(8.0%)
777 Other chronic lung disease	0 (0%)	2~(0.8%)	0 (0%)	1 (0.4%)	0 (0%)
0	232 (98.7%)	231 (97.9%)	237 (99.6%)	235 (99.2%)	24 (96.0%)
1	2 (0.9%)	4 (1.7%)	1 (0.4%)	2(0.8%)	0 (0%)
777	1 (0.4%)	1 (0.4%)	0 (0%)	0 (0%)	1 (4.0%)
Diabetes 0	229 (97.4%)	229 (97.0%)	233 (97.9%)	233 (98.3%)	25 (100%)
•	,	, , ,	` ′	` ′	, ,
1	2 (0.9%)	4 (1.7%)	2 (0.8%)	1 (0.4%)	0 (0%)
777	4 (1.7%)	3 (1.3%)	3 (1.3%)	3 (1.3%)	0 (0%)
Stomach or intestine	999 (00 707)	224 (04 007)	220 (06 207)	200 (05 407)	04 (06 007)
0	232 (98.7%) 2 (0.9%)	224 (94.9%) 11 (4.7%)	229 (96.2%) 9 (3.8%)	226 (95.4%) 11 (4.6%)	24 (96.0%) 1 (4.0%)
	, ,	, ,	, ,	` /	,
777	1 (0.4%)	1 (0.4%)	0 (0%)	0 (0%)	0 (0%)
HIV/AIDS	010 (00 007)	201 (07 007)	207 (05 407)	000 (04 107)	05 (10007)
0	218 (92.8%)	231 (97.9%) 1 (0.4%)	227 (95.4%) 2 (0.8%)	223 (94.1%)	25 (100%)
1 777	5 (2.1%) 12 (5.1%)	4 (0.4%)	9(3.8%)	2 (0.8%) $12 (5.1%)$	$0 (0\%) \\ 0 (0\%)$
	12 (3.170)	4 (1.770)	9 (3.670)	12 (3.170)	0 (070)
Epilepsy/Seizure	000 (00 107)	202 (00 004)	207 (00 604)	007 (100%)	OF (10007)
0 1	233 (99.1%)	232 (98.3%)	237 (99.6%)	237 (100%)	25 (100%)
777	$ \begin{array}{ccc} 2 & (0.9\%) \\ 0 & (0\%) \end{array} $	$ \begin{array}{c} 1 \ (0.4\%) \\ 3 \ (1.3\%) \end{array} $	$ \begin{array}{ccc} 1 & (0.4\%) \\ 0 & (0\%) \end{array} $	0 (0%) 0 (0%)	0 (0%) 0 (0%)
Cancer	0 (0/0)	o (1.o/0)	0 (0/0)	0 (0/0)	0 (070)
	224 (22.25%)	224 (22.25%)	224 (22.25%)	222 (22.25)	07 (400°C)
0	231 (98.3%)	234 (99.2%)	234 (98.3%)	236 (99.6%)	25 (100%)
777	4 (1.7%)	2~(0.8%)	4~(1.7%)	1 (0.4%)	0 (0%)

tbl

Propensity score analysis

Predicting case status using all covariates

```
family="binomial")
#tbl_regression(qlm_fit, exp=TRUE)
glm_probs = data.frame(probs = predict(glm_fit, type="response"), tel_all$is_case_f)
glm probs$is case <- tel all$is case</pre>
res <- as.data.frame(coef(summary(glm fit)))</pre>
res <- res %>% mutate(ci_ll = round(exp(Estimate - qnorm(0.975)*`Std. Error`),3))
res <- res %>% mutate(ci_ul = round(exp(Estimate + qnorm(0.975)*`Std. Error`),3))
res <- res %>% mutate(ci = paste0("(",ci_ll,",",ci_ul,")"))
res <- res %>% mutate(OR = exp(Estimate)) %>%
  mutate(OR_round = case_when())
    OR < 100 ~ sprintf("%.2f", OR),
    OR>100 ~ "Error"))
res <- res %>% mutate(pval = round(`Pr(>|z|)`,3)) %>%
  mutate(pval2 = case_when(
                     paste0(pval, "***"),
    pval<0.001 ~
    pval<0.01 & pval>0 ~ paste0(pval,"**"),
    pval<0.05 & pval>0 ~ paste0(pval,"*"),
    TRUE ~ as.character(pval)
  ))
res <- res %>% select(OR_round, ci, pval2)
rownames(res) <- c("Intercept", "Age", "Female",</pre>
                   "Education - Secondary", "Education - College",
                   "BMI - Underweight", "BMI- Overweight", "BMI - Obese",
                   "Alcohol - Irregular", "Alcohol - Regular",
                   "Khat - Irregular", "Khat - Regular",
                   "Cannabis - Irregular", "Cannabis - Regular",
                   "Tobacco - Irregular", "Tobacco - Regular",
                   "Arthritis - Yes", "Arthritis - Missing",
                   "Chronic back/neck problems - Yes",
                   "Chronic back/neck problems - Missing",
                   "Headaches - Yes", "Headaches - Missing",
                   "Other chronic pain - Yes",
                   "Allergies - Yes",
                   "Stroke - Yes", "Stroke - Missing",
                   "Heart attack - Yes", "Heart Attack - Missing",
                   "Heart disease - Yes", "Heart Disase - Missing",
                   "High blood pressure - Yes", "High blood pressure - Missing",
                   "Asthma - Yes", "Asthma - Missing",
                   "Tuberculosis - Yes", "Tuberculosis - Missing",
                   "Other chronic lung disease - Yes",
                   "Other chronic lung disease - Missing",
```

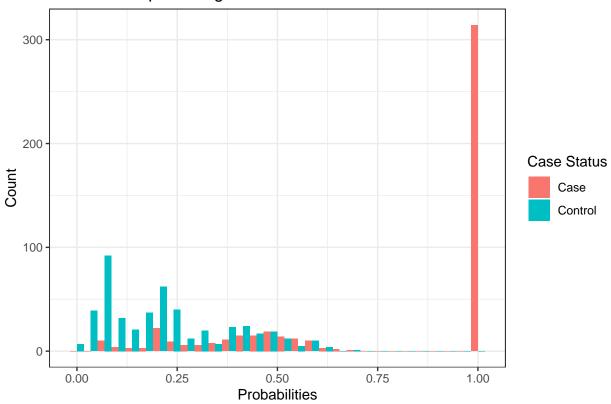
Coefficient	OR	95% CI	p-value
Intercept Age Female Education - Secondary Education - College	0.27	(0.069,1.026)	0.055
	1.02	(0.995,1.045)	0.121
	0.88	(0.567,1.353)	0.551
	1.33	(0.847,2.093)	0.215
	0.40	(0.24,0.654)	0***
BMI - Underweight BMI- Overweight BMI - Obese Alcohol - Irregular Alcohol - Regular	1.43 2.55 1.47 0.31 0.08	(0.572,3.577) (0.858,7.557) (0.555,3.882) (0.199,0.471) (0.034,0.199)	0.444 0.092 0.44 0***
Khat - Irregular	1.65	(0.914,2.977)	0.096
Khat - Regular	0.94	(0.391,2.262)	0.891
Cannabis - Irregular	0.00	(0,Inf)	0.988
Cannabis - Regular	0.00	(0,Inf)	0.999
Tobacco - Irregular	Error	(0,Inf)	0.982
Tobacco - Regular	Error	(0,Inf)	0.985
Arthritis - Yes	Error	(0,Inf)	0.992
Arthritis - Missing	2.00	(0,Inf)	1
Chronic back/neck problems - Yes	Error	(0,Inf)	0.99
Chronic back/neck problems - Missing	0.00	(0,Inf)	0.997
Headaches - Yes	Error	(0,Inf)	0.99
Headaches - Missing	Error	(0,Inf)	0.997
Other chronic pain - Yes	3.47	(0,Inf)	1
Allergies - Yes	Error	(0,Inf)	0.993
Stroke - Yes	0.00	(0,Inf)	0.998
Stroke - Missing Heart attack - Yes Heart Attack - Missing Heart disease - Yes Heart Disase - Missing	0.00	(0,Inf)	0.996
	0.30	(0,Inf)	1
	0.00	(0,Inf)	0.998
	Error	(0,Inf)	0.997
	Error	(0,Inf)	0.999
High blood pressure - Yes High blood pressure - Missing Asthma - Yes Asthma - Missing Tuberculosis - Yes	Error	(0,Inf)	0.994
	0.33	(0,Inf)	1
	Error	(0,Inf)	0.997
	0.00	(0,Inf)	0.998
	Error	(0,Inf)	0.994
Tuberculosis - Missing	2.84	(0,Inf)	1
Other chronic lung disease - Yes	Error	(0,Inf)	0.996
Other chronic lung disease - Missing	4.74	(0,Inf)	1
Diabetes - Yes	Error	(0,Inf)	0.997

Diabetes - Missing	Error	(0,Inf)	0.996
Ulcer - Yes	Error	(0,Inf)	0.993
Ulcer - Missing	0.00	(0,Inf)	0.999
HIV/AIDS - Yes	Error	(0,Inf)	0.998
HIV/AIDS - Missing	Error	(0,Inf)	0.992
Epilepsy/Seizure - Yes	Error	(0,Inf)	0.997
Cancer - Missing	Error	(0,Inf)	0.996

```
ggplot(data=glm_probs, aes(x=probs, fill=tel_all.is_case_f)) +
  geom_histogram(position="dodge") +
  ggtitle("Probabilities predicting Case Status") +
  xlab("Probabilities") +
  ylab("Count") +
  guides(fill = guide_legend(title = "Case Status")) +
  theme_bw()
```

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

Probabilities predicting Case Status



Propensity score analysis for set

```
#With chronic conditions
#Missingness in cidi_q9 and cidi_q13
glm_fit <- multinom(factor(Set) ~ age_at_iview + factor(msex) + educ_ord +</pre>
                bmi_bin + alcohol_3cat + khat_3cat + cannabis_3cat + tobacco_3cat +
                cidi_q1 + cidi_q2 + cidi_q3 + cidi_q4 + cidi_q5 + cidi_q6 +
                cidi_q7 + cidi_q8 + cidi_q9 + cidi_q10 + cidi_q11 + cidi_q12 +
               cidi_q13 + cidi_q14 + cidi_q15 + cidi_q16 + cidi_q17,
               data=tel all)
## # weights: 240 (188 variable)
## initial value 1562.764213
## iter 10 value 1384.195516
## iter 20 value 1331.352221
## iter 30 value 1320.585962
## iter 40 value 1316.832197
## iter 50 value 1314.470208
## iter 60 value 1313.613017
## iter 70 value 1313.487470
## iter 80 value 1313.446385
## iter 90 value 1313.434546
## iter 100 value 1313.429003
## final value 1313.429003
## stopped after 100 iterations
# tidy(glm_fit, conf.int = TRUE) %>%
  kable() %>%
  kable_styling("basic", full_width = FALSE,
                  latex_options = "scale_down")
```

```
## i Multinomial models have a different underlying structure than the models
## gtsummary was designed for. Other gtsummary functions designed to work with
## tbl_regression objects may yield unexpected results.
```

```
## Table printed with `knitr::kable()`, not {gt}. Learn why at
## https://www.danieldsjoberg.com/gtsummary/articles/rmarkdown.html
```

tbl_regression(glm_fit, exp = TRUE)

To suppress this message, include `message = FALSE` in code chunk header.

Outco	ome Characteristic	OR	95% CI	p-value
В	Age	0.97	0.95, 1.00	0.042
	factor(msex)			
	Male	_	_	
	Female	0.82	0.51, 1.32	0.4
	Education			
	Primary or less	_	_	
	Secondary	0.66	0.41, 1.07	0.092
	College	0.75	0.45, 1.23	0.3

Outcome	Characteristic	OR	95% CI	p-value
	BMI			
	Underweight	_	_	
	Normal Weight	0.34	0.15, 0.82	0.015
	Obese	0.26	0.09, 0.77	0.015
	Overweight	0.29	0.11, 0.75	0.010
	Alcohol Use			
	Never Users	_	_	
	Irregular Users	1.23	0.79, 1.90	0.4
	Regular Users	0.96	0.54, 1.72	>0.9
	Khat Use		,	
	Never Users	_	_	
	Irregular Users	1.70	0.91, 3.18	0.093
	Regular Users	3.96	$1.67^{'}, 9.42$	0.002
	Cannabis Use		,	
	Never Users	_	_	
	Irregular Users	0.74	0.25, 2.14	0.6
	Regular Users	131,020,801	31,787,706, 540,034,255	< 0.001
	Tobacco Use	- ,,	- , : , , ,	
	Never Users	_	_	
	Irregular Users	0.62	0.29, 1.33	0.2
	Regular Users	0.93	0.34, 2.59	0.9
	Arthritis and	0.00	0.01, 2.00	0.0
	Rheumatism			
	Asthma	_	_	
	Tuberculosis	0.70	0.18, 2.73	0.6
	777	0.01	0.01, 0.01	< 0.001
	Chronic back or neck	0.01	0.01, 0.01	70.001
	problems			
	0	_	_	
	1	0.65	0.23,1.82	0.4
	777	0.12	0.12, 0.12	< 0.001
	Frequent or severe	0.12	0.12, 0.12	V0.001
	headaches			
	0	_	_	
	1	0.80	0.32, 2.00	0.6
	777	3.42	0.92, 2.00	0.0
	Any other chronic	0.42		
	pain			
	0	_	_	
	1	0.58	0.07, 4.86	0.6
	Seasonal allergies	0.90	0.07, 4.00	0.0
	0	_	_	
	1	0.52	0.12, 2.26	0.4
	Stroke	0.02	0.12, 2.20	0.4
	0		_	
	1	0.01		
	777	0.00	0.00, 0.00	< 0.001
	Heart attack	0.00	0.00, 0.00	<0.001
	neart attack	_		
	1	1.04	0.11 0.06	>0.9
	1 777	0.00	0.11, 9.96	>0.9 <0.001
		0.00	0.00, 0.00	<0.001
	Heart disease			

Outcome	Characteristic	OR	95% CI	p-value
	0	_	_	
	1	7,091,086	$1,901,324,\ 26,446,574$	< 0.001
	777	3.59	0.03, 440	0.6
	High blood pressure		,	
	0	_	_	
	1	1.43	0.29, 7.12	0.7
	- 777	5.52	0.03, 883	0.5
	Asthma	3.3 <u>-</u>	3.03, 333	0.0
	0			
	1	12,854,804,039	3,694,455,664, 44,728,101,211	< 0.001
	777	3,856,318,629,538,427	3,855,644,690,325,668, 3,856,992,686,550,957	< 0.001
	Tuberculosis		3,000,000,000,000,000	
	0	_	_	
	1	1.44	0.36, 5.79	0.6
	- 777	1.19	1.19, 1.19	< 0.001
	Other chronic lung disease	1110	1.10, 1.10	(0.001
	0	_	_	
	1	5.03	0.31,82.7	0.3
	777	0.00	0.00, 0.00	< 0.001
	Diabetes			
	0	_	_	
	1	3.08	0.45, 21.2	0.3
	777	0.73	0.06, 8.20	0.8
	Stomach or intestine		,	
	0	_	_	
	1	7.67	1.36, 43.2	0.021
	- 777	246,750	246,747, 246,753	< 0.001
	HIV/AIDS	210,100	210,111, 210,100	(0.001
	0	_	_	
	1	0.39	0.04, 4.18	0.4
	777	0.20	0.04, 4.18	0.47
		0.20	0.04, 0.98	0.047
	Epilepsy/Seizure			
	0			0.5
	1	0.23	0.00, 11.4	0.5
	777	107,678,680,707,522,736	$107,659,862,530,800,944, \\107,697,502,173,527,712$	< 0.001
	Cancer			
	0			.0.001
~	777	0.00	0.00, 0.00	< 0.001
С	Age	1.00	0.98, 1.03	>0.9
	factor(msex)			
	Male	_	_	
	Female	1.01	0.64, 1.59	> 0.9
	Education			
	Primary or less	_	_	
	Secondary	0.63	0.39, 1.02	0.059
	College	0.96	0.59, 1.57	0.9
	BMI			
	Underweight		_	
	Normal Weight	0.71	0.27, 1.86	0.5
	,,	V., ±	5.2., 2.00	•••

Outcome	Characteristic	OR	95% CI	p-value
	Obese	0.49	0.15, 1.54	0.2
	Overweight	1.10	0.40, 3.05	0.8
	Alcohol Use			
	Never Users	_	_	
	Irregular Users	0.82	0.53, 1.26	0.4
	Regular Users	0.70	0.39, 1.25	0.2
	Khat Use			
	Never Users			
	Irregular Users	2.13	1.16, 3.91	0.015
	Regular Users	2.52	1.00, 6.33	0.049
	Cannabis Use			
	Never Users			0.0
	Irregular Users	0.74	0.22, 2.47	0.6
	Regular Users	88,961,117	$17,166,585,\ 461,016,574$	< 0.001
	Tobacco Use			
	Never Users			-0.001
	Irregular Users	0.22	0.09, 0.52	< 0.001
	Regular Users	0.92	0.33, 2.57	0.9
	Arthritis and Rheumatism			
	Asthma			
	Tuberculosis	1.29	0.41,4.01	0.7
	777	116,592,899	116,592,899, 116,592,900	< 0.001
	Chronic back or neck	110,592,699	110,592,899, 110,592,900	<0.001
	problems			
	0		_	
	1	0.92	0.36, 2.35	0.9
	777	1,088	1,088, 1,088	< 0.001
	Frequent or severe	1,000	1,000, 1,000	(0.001
	headaches			
	0	_	_	
	1	1.50	0.63, 3.54	0.4
	777	392	,	
	Any other chronic			
	pain			
	0		_	
	1	1.48	0.26,8.55	0.7
	Seasonal allergies			
	0	_	_	
	1	0.41	0.10, 1.79	0.2
	Stroke			
	0		_	
	1	0.00	0.00, 0.00	< 0.001
	777	0.00	0.00, 0.00	< 0.001
	Heart attack			
	0	_	_	
	1	0.10	0.01, 2.00	0.13
	777	0.00	0.00,0.00	< 0.001
	Heart disease			
	0	_	_	
	1	15,578,354	4,328,565, 56,065,951	< 0.001
	777	2.28	0.02,323	0.7

Outcome	Characteristic	OR	$95\%~\mathrm{CI}$	p-value
	High blood pressure			
	0	_	_	
	1	0.55	0.10, 2.96	0.5
	777	1.58	0.01, 271	0.9
	Asthma			
	0	_	_	
	1	7,077,832,784	$1,837,633,900,\ 27,260,988,671$	< 0.001
	777	0.08	0.08, 0.08	< 0.001
	Tuberculosis		,	
	0	_	_	
	1	1.82	0.53,6.27	0.3
	777	0.01	0.000, 0.21	0.0
	Other chronic lung	0.01		
	disease			
	0			
	1	0.98	0.04,26.2	>0.9
	777	0.00	0.00, 0.00	< 0.001
	Diabetes			
	0	_	_	
	1	1.22	0.14,10.5	0.9
	777	1.07	0.16, 7.14	> 0.9
	Stomach or intestine			
	0	_	_	
	1	6.79	$1.21,\ 38.1$	0.029
	777	0.00	0.00, 0.00	< 0.001
	HIV/AIDS			
	0	_	_	
	1	0.40	0.05, 3.29	0.4
	777	0.96	0.34, 2.70	>0.9
	Epilepsy/Seizure	0.00	3.3 =, =3	,
	0		_	
	1	1.17	0.05, 29.0	>0.9
	777	0.03	0.03, 0.03	< 0.001
	Cancer	0.05	0.05, 0.05	<0.001
	0	1.00	0.07 7.10	0.7
D	777	1.32	0.25, 7.12	0.7
D	Age	0.99	0.96, 1.01	0.3
	factor(msex)			
	Male	_	_	
	Female	0.87	0.55, 1.39	0.6
	Education			
	Primary or less	_	_	
	Secondary	0.75	0.47, 1.20	0.2
	College	0.71	0.43, 1.17	0.2
	BMI			
	Underweight	_	_	
	Normal Weight	0.77	0.29, 2.01	0.6
	Obese	0.52	0.16, 1.62	0.3
	Overweight	1.02	0.37, 2.80	>0.9
	Alcohol Use	1.02	0.01, 2.00	/ 0.0
	Never Users			
	Irregular Users	1.22	0.07.0.04	0.0
	irregiliar Users	1.33	0.87, 2.04	0.2

Outcome	Characteristic	OR	95% CI	p-value
	Regular Users	0.72	0.39, 1.31	0.3
	Khat Use			
	Never Users	_	_	
	Irregular Users	1.91	1.04, 3.50	0.036
	Regular Users	4.47	1.88, 10.6	< 0.001
	Cannabis Use			
	Never Users		_	
	Irregular Users	0.98	0.34, 2.80	> 0.9
	Regular Users	159,156,860	$31,195,433,\ 812,006,883$	< 0.001
	Tobacco Use			
	Never Users	_	_	
	Irregular Users	0.48	$0.23,\ 0.99$	0.048
	Regular Users	0.35	0.12, 1.05	0.060
	Arthritis and			
	Rheumatism			
	Asthma			
	Tuberculosis	1.62	0.55, 4.82	0.4
	777	0.01	0.01,0.01	< 0.001
	Chronic back or neck			
	problems			
	0	_		o =
	1	0.82	0.34, 2.00	0.7
	777	0.25	0.25, 0.25	< 0.001
	Frequent or severe			
	headaches			
	0		0.50, 0.70	0.7
	1	1.20	0.52, 2.78	0.7
	777	0.12		
	Any other chronic			
	pain			
	0 1		0.08, 4.21	0.6
	Seasonal allergies	0.59	0.08, 4.21	0.0
	0			
	1	0.84	0.24, 2.96	0.8
	Stroke	0.04	0.24, 2.30	0.0
	0	<u></u>	_	
	1	20,075,119,220,975	20,075,119,220,952,	< 0.001
	1	20,010,113,220,310	20,075,119,220,998	\0.001
	777	0.03	0.03, 0.03	< 0.001
	Heart attack		0.00, 0.00	
	0			
	1	0.67	0.07, 6.19	0.7
	- 777	0.00	0.00, 0.00	< 0.001
	Heart disease		0.00, 0.00	
	0			
	1	11,227,360	3,288,907, 38,326,906	< 0.001
	777	0.00	0.00, 0.00	< 0.001
	High blood pressure		,	
	0	_	_	
	1	1.13	0.24,5.37	0.9
	- 777	4.08	0.03, 606	0.6
	•		, •••	

Outcome	Characteristic	OR	95% CI	p-value
	Asthma			
	0	_	_	
	1	$2,\!214,\!169,\!781$	$467,895,478,\ 10,477,869,639$	< 0.001
	777	0.00	0.00, 0.00	< 0.001
	Tuberculosis			
	0	_	_	
	1	1.13	0.29, 4.33	0.9
	777	$9,\!104,\!847,\!967,\!347$	$9,104,847,967,162,\ 9,104,847,967,532$	< 0.001
	Other chronic lung			
	disease			
	0	_	_	
	1	1.21	0.04, 33.7	> 0.9
	777	0.00	0.00, 0.00	< 0.001
	Diabetes			
	0	_	_	
	1	0.49	0.04, 6.45	0.6
	777	1.17	0.19, 7.17	0.9
	Stomach or intestine			
	0	_	_	
	1	7.24	1.33, 39.3	0.022
	777	0.00	0.00, 0.00	< 0.001
	HIV/AIDS			
	0	_	_	
	1	0.34	0.05, 2.49	0.3
	777	1.11	0.43, 2.89	0.8
	Epilepsy/Seizure			
	0	_	_	
	1	0.00	0.00, 0.00	< 0.001
	777	0.00	0.00, 0.00	< 0.001
	Cancer			
	0	_	<u> </u>	
	777	0.35	0.03, 3.72	0.4
E	Age	0.97	0.91, 1.04	0.4
	factor(msex)			
	Male	_	_	
	Female	0.87	0.30, 2.58	0.8
	Education			
	Primary or less	_	<u> </u>	
	Secondary	1.69	0.42, 6.83	0.5
	College	2.81	0.70, 11.2	0.14
	BMI			
	Underweight	_	_	
	Normal Weight	0.36	0.06, 2.24	0.3
	Obese	0.43	0.04, 4.53	0.5
	Overweight	1.29	0.19, 8.58	0.8
	Alcohol Use			
	Never Users	_	_	
	Irregular Users	0.95	0.36, 2.49	> 0.9
	Regular Users	0.16	0.02, 1.44	0.10
	Khat Use			
	Never Users	_	_	
	Irregular Users	1.72	0.36, 8.08	0.5
	~		*	

Outcome	Characteristic	OR	95% CI	p-value
	Regular Users	6.63	1.16, 37.8	0.033
	Cannabis Use			
	Never Users	_	_	
	Irregular Users	0.00	0.00, 0.00	< 0.001
	Regular Users	0.00	0.00, 0.00	< 0.001
	Tobacco Use			
	Never Users	_	_	
	Irregular Users	0.34	0.03, 3.62	0.4
	Regular Users	1.91	0.27, 13.7	0.5
	Arthritis and			
	Rheumatism			
	Asthma	_	_	
	Tuberculosis	0.00	0.00,0.00	< 0.001
	777	0.23	0.23,0.23	< 0.001
	Chronic back or neck			
	problems			
	0	_		
	1	0.00	0.00, 0.00	< 0.001
	777	1.14	1.14, 1.14	< 0.001
	Frequent or severe			
	headaches			
	0			0.0
	1	0.71	0.05, 10.2	0.8
	777	1.35	1.35, 1.35	< 0.001
	Any other chronic			
	pain			
	0			-0.001
	1	0.00	0.00, 0.00	< 0.001
	Seasonal allergies			
	0			-0.001
	1 Stroke	0.00	0.00, 0.00	< 0.001
	0 1	1 10	1 10 1 10	<0.001
	1 777	$1.18 \\ 1.37$	1.18, 1.18	<0.001 <0.001
	Heart attack	1.57	1.37, 1.37	<0.001
	0 1	0.00	0.00, 0.00	< 0.001
	1 777	1.08	1.08, 1.08	< 0.001
	Heart disease	1.00	1.08, 1.08	<0.001
	0	_	_	
	1	0.00	0.00, 0.00	< 0.001
	777	0.00	0.00, 0.00 $0.02, 0.02$	< 0.001
	High blood pressure	0.02	0.02, 0.02	\0.001
	0	_	_	
	1	12.5	0.92, 171	0.058
	777	2.11	2.11, 2.11	< 0.001
	Asthma	2.11	2.11, 2.11	\0.001
	0	_		
	1	10,847	10,847, 10,847	< 0.001
	777	1.18	1.18, 1.18	< 0.001
	Tuberculosis	1.10	1.10, 1.10	\0.001

Outcome Characteristic	\mathbf{OR}	95% CI	p-value
0	_	_	
1	9.16	1.17, 71.8	0.035
777	1.26	1.26, 1.26	< 0.001
Other chronic lung			
disease			
0	_	_	
1	0.00	0.00, 0.00	< 0.001
777	6,650,678	$6,650,673,\ 6,650,682$	< 0.001
Diabetes			
0	_	_	
1	0.00	0.00, 0.00	< 0.001
777	0.00	0.00, 0.00	< 0.001
Stomach or intestine			
0	_	_	
1	13.8	0.88, 217	0.062
777	0.00	0.00, 0.00	< 0.001
HIV/AIDS			
0	_	_	
1	0.00	0.00, 0.00	< 0.001
777	0.00	0.00, 0.00	< 0.001
Epilepsy/Seizure			
0	_	_	
1	0.00	0.00, 0.00	< 0.001
777	1.39	1.39, 1.39	< 0.001
Cancer			
0	_	_	
777	0.00	0.00, 0.00	< 0.001

```
glm_probs = data.frame(probs = predict(glm_fit, type="probs"), tel_all$Set,
                       tel_all$is_case_f)
predA <- ggplot(data=glm_probs, aes(x=probs.A, fill=tel_all.Set)) + geom_histogram(position="dodge") +</pre>
facet_wrap(~tel_all.is_case_f) + ggtitle("Probabilities predicting Set A") +
  xlab("Probabilities") +
  ylab("Count") +
  guides(fill = guide_legend(title = "Set")) +
  theme bw()
predB <- ggplot(data=glm_probs, aes(x=probs.B, fill=tel_all.Set)) + geom_histogram(position="dodge") +</pre>
facet_wrap(~tel_all.is_case_f) + ggtitle("Probabilities predicting Set B") +
  xlab("Probabilities") +
  ylab("Count") +
  guides(fill = guide_legend(title = "Set")) +
  theme_bw()
predC <- ggplot(data=glm_probs, aes(x=probs.C, fill=tel_all.Set)) + geom_histogram(position="dodge") +</pre>
facet_wrap(~tel_all.is_case_f) + ggtitle("Probabilities predicting Set C") +
  xlab("Probabilities") +
  ylab("Count") +
  guides(fill = guide_legend(title = "Set")) +
  theme_bw()
```

```
predD <- ggplot(data=glm_probs, aes(x=probs.D, fill=tel_all.Set)) + geom_histogram(position="dodge") +
facet_wrap(~tel_all.is_case_f) + ggtitle("Probabilities predicting Set D") +
    xlab("Probabilities") +
    ylab("Count") +
    guides(fill = guide_legend(title = "Set")) +
    theme_bw()

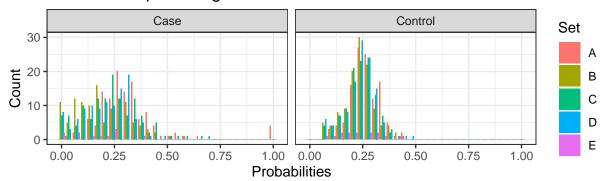
predE <- ggplot(data=glm_probs, aes(x=probs.E, fill=tel_all.Set)) + geom_histogram(position="dodge") +
    facet_wrap(~tel_all.is_case_f) + ggtitle("Probabilities predicting Set E") +
    xlab("Probabilities") +
    ylab("Count") +
    guides(fill = guide_legend(title = "Set")) +
    theme_bw()

predF <- ggplot(data = NULL) +
    geom_blank()</pre>
```

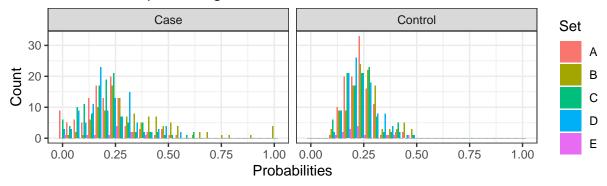
predA / predB

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

Probabilities predicting Set A



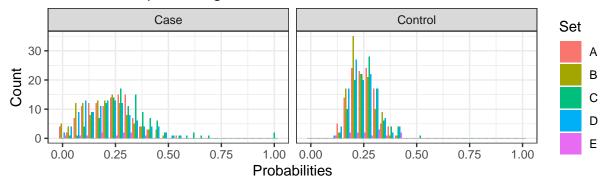
Probabilities predicting Set B



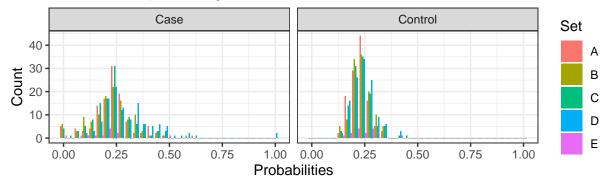
predC / predD

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

Probabilities predicting Set C



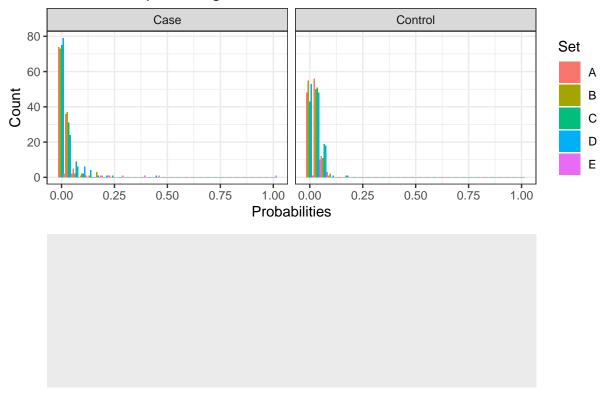
Probabilities predicting Set D



predE / predF

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

Probabilities predicting Set E



Comparing freeze 7 vs freeze 8

```
ngap_7$bmi_bin<-0
ngap_7[which(ngap_7$bmi<=18.5), "bmi_bin"]<-"Underweight"</pre>
ngap_7[which(ngap_7$bmi>18.5 & ngap_7$bmi<=24.9),
            "bmi bin"] <-"Normal Weight"
ngap_7[which(ngap_7$bmi>24.9 & ngap_7$bmi<30),
            "bmi bin"] <- "Overweight"
ngap_7[which(ngap_7$bmi>=30),
            "bmi bin"] <- "Obese"
ngap_7[which(is.na(ngap_7$bmi)), "bmi_bin"] <- NA</pre>
ngap_7$bmi_bin<-factor(ngap_7$bmi_bin, levels=c("Normal Weight",</pre>
                                             "Underweight",
                                             "Overweight",
                                             "Obese"
                                           ))
ngap_7$bmi_bin <- relevel(ngap_7$bmi_bin, ref = "Normal Weight")</pre>
tab<-sumTab(ngap_7$bmi_bin)</pre>
names(tab)[which(is.na(names(tab)))]<-"Missing"</pre>
ngap_7_new <- ngap_7 %>% select(subj_id, bmi_bin, hbp, cidi_q8, cidi_q9, cidi_q13) %>%
    rename(
    subjid = subj_id,
    ngap_7_bmi_bin = bmi_bin,
    ngap_7hbp = hbp,
    ngap_7_cidi_q8 = cidi_q8,
    ngap_7_cidi_q9 = cidi_q9,
    ngap_7_cidi_q13 = cidi_q13
  )
ngap_7_merge <- merge(ngap_7_new, tel_all, by="subjid")</pre>
ngap_7_merge <- ngap_7_merge %>% select( subjid,
    ngap_7_bmi_bin, bmi_bin,
    ngap_7_hbp, hbp,
    ngap_7_cidi_q8, cidi_q8,
    ngap_7_cidi_q9, cidi_q9,
    ngap_7_cidi_q13, cidi_q13)
```

Comparing cardiometabolic variables

There are no differences in cardiometabolic conditions between Freeze 7 and Freeze 8.

```
#BMI
tab <- table(ngap_7_merge$ngap_7_bmi_bin, ngap_7_merge$bmi_bin)
kable(tab)</pre>
```

	Underweight	Normal Weight	Obese	Overweight
Normal Weight	0	618	0	0
Underweight	56	0	0	0
Overweight	0	0	0	216
Obese	0	0	81	0

#High Blood Pressure - Calculated

tab <- table(ngap_7_merge\$ngap_7_hbp, ngap_7_merge\$ngap_7_hbp)
kable(tab)</pre>

	Normal BP	Low BP	High BP
Normal BP	913	0	0
Low BP	0	2	0
High BP	0	0	56

#Heart Disease

tab <- table(ngap_7_merge\$ngap_7_cidi_q8, ngap_7_merge\$cidi_q8)
kable(tab)</pre>

	0	1	777
0	960	0	0
1	0	8	0
777	0	0	3

#High Blood Pressure - CIDI

tab <- table(ngap_7_merge\$ngap_7_cidi_q9, ngap_7_merge\$cidi_q9)
kable(tab)</pre>

	0	1	777
0	942	0	0
1	0	22	0
777	0	0	7

#Diabetes or high blood sugar

tab <- table(ngap_7_merge\$ngap_7_cidi_q13, ngap_7_merge\$cidi_q13)
kable(tab)</pre>

	0	1	777
0	949	0	0
1	0	9	0
777	0	0	13

Analysis with clean/unclean cases and controls

Distribution of ltl/age with clean cases , unclean cases, controls Update powerpoint with controls as reference group

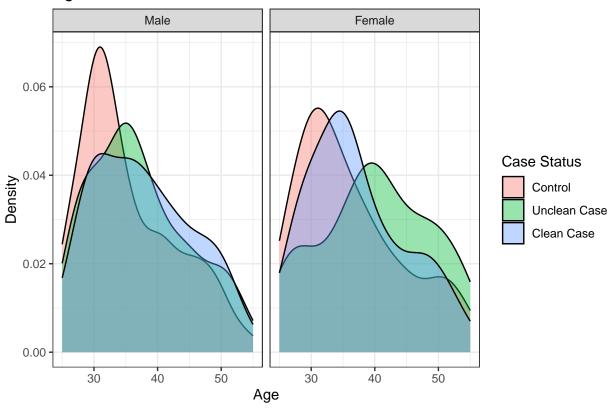
```
tel_all <- tel_all %>% mutate(is_case_clean = case_when(
                       is_case == 1 &
                       (tobacco_3cat %in% c("Irregular Users",
                                            "Regular Users") |
                       cidi_q1 == 1 |
                       cidi_q2 == 1 |
                       cidi_q3 == 1 |
                       cidi_q4 == 1 |
                       cidi_q5 == 1 |
                       cidi_q6 == 1 |
                       cidi_q7 == 1 |
                       cidi q8 == 1
                       cidi_q9 == 1 |
                       cidi_q10 == 1 |
                       cidi_q11 == 1 |
                       cidi_q12 == 1
                       cidi_q13 == 1
                       cidi_q14 == 1
                       cidi_q15 == 1
                       cidi_q16 == 1 |
                       cidi_q17 == 1) ~ "Unclean Case",
                       is_case == 0 ~ "Control",
                       TRUE
                                       ~ "Clean Case"
                         )) %>% mutate(is_case_clean = factor(is_case_clean,
                         levels = c("Control", "Unclean Case", "Clean Case")))
```

Age distribution of clean/unclean cases and controls.

The unclean cases among Females screw older than the rest of the population

```
ggplot(tel_all, aes(x=age_at_iview, group = is_case_clean, fill=is_case_clean))+
  facet_wrap(~msex) +
  geom_density(alpha=0.4) +
  xlab("Age") + ylab("Density") +
  ggtitle("Age Distribution of Clean/Unclean Cases and Controls") +
  ggeasy::easy_center_title() +
  labs(fill='Case Status') +
  #scale_y_continuous(labels = scales::percent_format(scale = 1)) +
  theme_bw()
```

Age Distribution of Clean/Unclean Cases and Controls



Rosner batch correction with clean/unclean cases Models

	Control	Unclean Case	Clean Case	Control	Unclean Case
	(N=315)	(N=214)	(N=103)	(N=169)	(N=85)
Age	(11-919)	(11-214)	(11-109)	(11-109)	(11-09)
	25 4 (7 41)	27.1 (7.70)	27.7 (7.66)	26.2 (0.15)	40.0 (9.60)
Mean (SD)	35.4 (7.41)	37.1 (7.70)	37.7 (7.66)	36.2 (8.15)	40.0 (8.60)
Median [Min, Max]	33.0 [25.0, 55.0]	36.0 [25.0, 55.0]	36.0 [25.0, 55.0]	34.0 [25.0, 55.0]	40.0 [25.0, 55
Education	00 (00 004)	01 (00 FOV)	9F (94 00%)	KO (00 004)	04 (40 007)
Primary or less	83 (26.3%)	61 (28.5%)	35 (34.0%)	52 (30.8%)	34 (40.0%)
Secondary	75 (23.8%)	94 (43.9%)	39 (37.9%)	63 (37.3%)	38 (44.7%)
College	157 (49.8%)	59 (27.6%)	29 (28.2%)	54 (32.0%)	13 (15.3%)
BMI					
Underweight	19 (6.0%)	15 (7.0%)	6 (5.8%)	9 (5.3%)	5 (5.9%)
Normal Weight	229 (72.7%)	148 (69.2%)	72 (69.9%)	78 (46.2%)	47 (55.3%)
Obese	5 (1.6%)	10 (4.7%)	6 (5.8%)	28 (16.6%)	17 (20.0%)
Overweight	62 (19.7%)	41 (19.2%)	19 (18.4%)	54 (32.0%)	16 (18.8%)
Alcohol Use	, , ,	, , ,	, , ,	, , ,	
Never Users	118 (37.5%)	49 (22.9%)	74 (71.8%)	83 (49.1%)	46 (54.1%)
Irregular Users	116 (36.8%)	136 (63.6%)	23 (22.3%)	72 (42.6%)	36 (42.4%)
Regular Users	81 (25.7%)	29 (13.6%)	6 (5.8%)	14 (8.3%)	3 (3.5%)
Khat Use	(=3.1,70)	20 (20.070)	3 (3.370)	-1 (0.070)	3 (3.370)
Never Users	240 (76.2%)	42 (19.6%)	73 (70.9%)	164 (97.0%)	68 (80.0%)
Irregular Users	47 (14.9%)	102 (47.7%)	22 (21.4%)	5 (3.0%)	16 (18.8%)
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Regular Users	28 (8.9%)	70 (32.7%)	8 (7.8%)	0 (0%)	1 (1.2%)
Cannabis Use	211 (00 =04)	101 (01 001)	100 (10004)	107 (00 00)	04 (00 00)
Never Users	311 (98.7%)	181 (84.6%)	103 (100%)	167 (98.8%)	84 (98.8%)
Irregular Users	4 (1.3%)	29 (13.6%)	0 (0%)	2 (1.2%)	1 (1.2%)
Regular Users	0 (0%)	4 (1.9%)	0 (0%)	0 (0%)	0 (0%)
Tobacco Use					
Never Users	315 (100%)	43 (20.1%)	103 (100%)	169 (100%)	76 (89.4%)
Irregular Users	0 (0%)	102 (47.7%)	0 (0%)	0 (0%)	8 (9.4%)
Regular Users	0 (0%)	69 (32.2%)	0 (0%)	0 (0%)	1 (1.2%)
Arthritis and Rheumatism	, ,	` '	, ,	,	, ,
0	315 (100%)	193 (90.2%)	103 (100%)	169 (100%)	62 (72.9%)
<u> </u>	0 (0%)	20 (9.3%)	0 (0%)	0 (0%)	22 (25.9%)
777	0 (0%)	1 (0.5%)	0 (0%)	0 (0%)	1 (1.2%)
Chronic back or neck problems	0 (070)	1 (0.070)	0 (070)	0 (0/0)	1 (1.2/0)
0	315 (100%)	175 (81.8%)	103 (100%)	169 (100%)	60 (70.6%)
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1	0 (0%)	38 (17.8%)	0 (0%)	0 (0%)	25 (29.4%)
777	0 (0%)	1 (0.5%)	0 (0%)	0 (0%)	0 (0%)
Frequent or severe headaches	04 = (1000)	4 = 0 (00 100)	100 (100%)	100 (1000)	10 (5:100)
0	315 (100%)	172 (80.4%)	103 (100%)	169 (100%)	46 (54.1%)
1	0 (0%)	40 (18.7%)	0 (0%)	0 (0%)	39 (45.9%)
777	0 (0%)	2 (0.9%)	0 (0%)	0 (0%)	0 (0%)
Any other chronic pain					
0	315 (100%)	206 (96.3%)	103 (100%)	169 (100%)	78 (91.8%)
1	0 (0%)	8 (3.7%)	0 (0%)	0 (0%)	7 (8.2%)
Seasonal allergies	\ ' "/	,	\ · */	\ ' */	
0	315 (100%)	200 (93.5%)	103 (100%)	169 (100%)	73 (85.9%)
<u>.</u> 1	0 (0%)	14 (6.5%)	0 (0%)	0 (0%)	12 (14.1%)
Stroke	0 (0/0)	11 (0.070)	0 (0/0)	0 (0/0)	12 (17.1/0)
	315 (100%)	211 (98.6%)	103 (100%)	169 (100%)	85 (100%)
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[0 (0%)	1 (0.5%)	0 (0%)	0 (0%)	0 (0%)
777	0 (0%)	2 (0.9%)	0 (0%)	0 (0%)	0 (0%)
Heart attack	, -	, -	, -	, -	
)	315 (100%)	208 (97.2%)	103 (100%)	169 (100%)	78 (91.8%)
1	0 (0%)	36 (2.8%)	0 (0%)	0 (0%)	6 (7.1%)
	0 (004)	0 (0%)	0 (0%)	0 (004)	1 (1 007)
777	0 (0%)	0 (0%)	0 (0%)	0 (0%)	1 (1.2%)
777 Heart disease	0 (0%)	0 (0%)	0 (0%)	0 (0%)	1 (1.2%)