

# ThesisAnalysis

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This Rmd file will be used as the file for my Thesis Analysis on causal mediation.

## Libraries

```
library(tidyverse)
library(car)
library(foreign)
library(Hmisc)
library(knitr)
library(dplyr)
library(survival)
library(finalfit)
library(tableone)
library(lme4)
library(e1071)
library(mediation)
```

## Reading of Raw Survey Data

```
ibiccs_readin <- read.csv("Database_recoded_2012-2014_weights_Walkscore_RTA.csv")
ibiccs_readin$TransitScore <- as.numeric(ibiccs_readin$TransitScore)
ibiccs_readin$BikeScore <- as.numeric(ibiccs_readin$BikeScore)
```

## Recode BMI

```
## BMI
ibiccs_clean <- ibiccs_readin %>%
  mutate(bmi_category = case_when(
    bmi < 18.5 ~ "underweight",
    bmi >=30 & bmi <999 ~ "obese",
    bmi >=25 & bmi <30 ~ "overweight",
    bmi >=18.5 & bmi <25 ~ "normal weight",
    bmi == 9999 ~ "NA",
    TRUE ~ "other"
  ))

table(ibiccs_clean$bmi, ibiccs_clean$bmi_category)

##
##          normal weight obese other overweight underweight
## 17.5           0         0     0             0             2
```

##	17.51	0	0	0	0	2
##	17.54	0	0	0	0	4
##	17.56	0	0	0	0	3
##	17.57	0	0	0	0	5
##	17.58	0	0	0	0	6
##	17.59	0	0	0	0	1
##	17.63	0	0	0	0	4
##	17.64	0	0	0	0	2
##	17.65	0	0	0	0	1
##	17.68	0	0	0	0	3
##	17.7	0	0	0	0	2
##	17.71	0	0	0	0	12
##	17.72	0	0	0	0	8
##	17.74	0	0	0	0	4
##	17.75	0	0	0	0	22
##	17.76	0	0	0	0	4
##	17.77	0	0	0	0	1
##	17.79	0	0	0	0	2
##	17.81	0	0	0	0	2
##	17.85	0	0	0	0	7
##	17.87	0	0	0	0	2
##	17.89	0	0	0	0	2
##	17.91	0	0	0	0	1
##	17.92	0	0	0	0	4
##	17.94	0	0	0	0	18
##	17.95	0	0	0	0	6
##	17.96	0	0	0	0	3
##	17.97	0	0	0	0	3
##	17.98	0	0	0	0	1
##	18	0	0	0	0	3
##	18.01	0	0	0	0	15
##	18.02	0	0	0	0	25
##	18.07	0	0	0	0	5
##	18.08	0	0	0	0	9
##	18.11	0	0	0	0	3
##	18.13	0	0	0	0	8
##	18.14	0	0	0	0	5
##	18.16	0	0	0	0	3
##	18.17	0	0	0	0	2
##	18.18	0	0	0	0	4
##	18.19	0	0	0	0	5
##	18.21	0	0	0	0	1
##	18.22	0	0	0	0	1
##	18.24	0	0	0	0	1
##	18.25	0	0	0	0	21
##	18.26	0	0	0	0	2
##	18.27	0	0	0	0	2
##	18.29	0	0	0	0	31
##	18.3	0	0	0	0	33
##	18.31	0	0	0	0	7
##	18.33	0	0	0	0	3
##	18.34	0	0	0	0	7
##	18.35	0	0	0	0	2
##	18.36	0	0	0	0	3

##	18.37	0	0	0	0	11
##	18.38	0	0	0	0	1
##	18.4	0	0	0	0	2
##	18.42	0	0	0	0	3
##	18.44	0	0	0	0	1
##	18.45	0	0	0	0	1
##	18.46	0	0	0	0	23
##	18.47	0	0	0	0	10
##	18.48	0	0	0	0	3
##	18.51	2	0	0	0	0
##	18.52	5	0	0	0	0
##	18.54	9	0	0	0	0
##	18.55	22	0	0	0	0
##	18.56	30	0	0	0	0
##	18.58	1	0	0	0	0
##	18.59	1	0	0	0	0
##	18.6	27	0	0	0	0
##	18.61	1	0	0	0	0
##	18.62	4	0	0	0	0
##	18.64	8	0	0	0	0
##	18.65	15	0	0	0	0
##	18.66	12	0	0	0	0
##	18.68	2	0	0	0	0
##	18.69	1	0	0	0	0
##	18.7	2	0	0	0	0
##	18.71	4	0	0	0	0
##	18.72	2	0	0	0	0
##	18.73	4	0	0	0	0
##	18.74	4	0	0	0	0
##	18.75	13	0	0	0	0
##	18.78	14	0	0	0	0
##	18.79	24	0	0	0	0
##	18.8	7	0	0	0	0
##	18.81	1	0	0	0	0
##	18.82	1	0	0	0	0
##	18.83	10	0	0	0	0
##	18.84	5	0	0	0	0
##	18.85	1	0	0	0	0
##	18.87	4	0	0	0	0
##	18.88	55	0	0	0	0
##	18.89	15	0	0	0	0
##	18.9	2	0	0	0	0
##	18.91	1	0	0	0	0
##	18.92	2	0	0	0	0
##	18.93	1	0	0	0	0
##	18.94	7	0	0	0	0
##	18.95	11	0	0	0	0
##	18.96	1	0	0	0	0
##	18.97	5	0	0	0	0
##	18.98	1	0	0	0	0
##	18.99	24	0	0	0	0
##	19.01	36	0	0	0	0
##	19.02	3	0	0	0	0
##	19.03	1	0	0	0	0

##	19.05	17	0	0	0	0
##	19.06	1	0	0	0	0
##	19.08	5	0	0	0	0
##	19.09	1	0	0	0	0
##	19.1	1	0	0	0	0
##	19.11	9	0	0	0	0
##	19.13	10	0	0	0	0
##	19.14	51	0	0	0	0
##	19.16	8	0	0	0	0
##	19.18	1	0	0	0	0
##	19.19	4	0	0	0	0
##	19.2	40	0	0	0	0
##	19.21	1	0	0	0	0
##	19.22	15	0	0	0	0
##	19.23	4	0	0	0	0
##	19.25	2	0	0	0	0
##	19.26	10	0	0	0	0
##	19.27	7	0	0	0	0
##	19.28	1	0	0	0	0
##	19.3	6	0	0	0	0
##	19.31	11	0	0	0	0
##	19.33	3	0	0	0	0
##	19.37	74	0	0	0	0
##	19.38	6	0	0	0	0
##	19.39	5	0	0	0	0
##	19.4	5	0	0	0	0
##	19.41	1	0	0	0	0
##	19.42	5	0	0	0	0
##	19.44	3	0	0	0	0
##	19.46	12	0	0	0	0
##	19.47	13	0	0	0	0
##	19.48	3	0	0	0	0
##	19.49	53	0	0	0	0
##	19.5	1	0	0	0	0
##	19.53	59	0	0	0	0
##	19.55	1	0	0	0	0
##	19.57	22	0	0	0	0
##	19.58	67	0	0	0	0
##	19.59	2	0	0	0	0
##	19.6	2	0	0	0	0
##	19.61	3	0	0	0	0
##	19.63	1	0	0	0	0
##	19.64	18	0	0	0	0
##	19.65	6	0	0	0	0
##	19.66	5	0	0	0	0
##	19.67	12	0	0	0	0
##	19.68	2	0	0	0	0
##	19.69	8	0	0	0	0
##	19.71	1	0	0	0	0
##	19.73	6	0	0	0	0
##	19.74	49	0	0	0	0
##	19.75	18	0	0	0	0
##	19.77	59	0	0	0	0
##	19.78	1	0	0	0	0

##	19.79	17	0	0	0	0
##	19.8	13	0	0	0	0
##	19.81	4	0	0	0	0
##	19.82	1	0	0	0	0
##	19.83	2	0	0	0	0
##	19.84	37	0	0	0	0
##	19.85	9	0	0	0	0
##	19.86	1	0	0	0	0
##	19.88	1	0	0	0	0
##	19.89	6	0	0	0	0
##	19.9	7	0	0	0	0
##	19.91	6	0	0	0	0
##	19.92	8	0	0	0	0
##	19.93	1	0	0	0	0
##	19.94	34	0	0	0	0
##	19.95	2	0	0	0	0
##	19.96	2	0	0	0	0
##	19.97	62	0	0	0	0
##	19.99	8	0	0	0	0
##	20	7	0	0	0	0
##	20.01	11	0	0	0	0
##	20.02	6	0	0	0	0
##	20.03	6	0	0	0	0
##	20.05	15	0	0	0	0
##	20.06	5	0	0	0	0
##	20.07	13	0	0	0	0
##	20.08	17	0	0	0	0
##	20.09	37	0	0	0	0
##	20.11	2	0	0	0	0
##	20.12	72	0	0	0	0
##	20.14	3	0	0	0	0
##	20.16	5	0	0	0	0
##	20.17	1	0	0	0	0
##	20.18	84	0	0	0	0
##	20.19	10	0	0	0	0
##	20.2	20	0	0	0	0
##	20.22	32	0	0	0	0
##	20.23	3	0	0	0	0
##	20.24	2	0	0	0	0
##	20.25	18	0	0	0	0
##	20.27	1	0	0	0	0
##	20.28	2	0	0	0	0
##	20.29	3	0	0	0	0
##	20.3	19	0	0	0	0
##	20.31	5	0	0	0	0
##	20.32	1	0	0	0	0
##	20.33	1	0	0	0	0
##	20.34	38	0	0	0	0
##	20.35	1	0	0	0	0
##	20.36	93	0	0	0	0
##	20.37	82	0	0	0	0
##	20.38	9	0	0	0	0
##	20.39	1	0	0	0	0
##	20.4	4	0	0	0	0

##	20.41	13	0	0	0	0
##	20.42	1	0	0	0	0
##	20.43	5	0	0	0	0
##	20.45	4	0	0	0	0
##	20.47	13	0	0	0	0
##	20.48	19	0	0	0	0
##	20.5	9	0	0	0	0
##	20.51	21	0	0	0	0
##	20.52	5	0	0	0	0
##	20.53	49	0	0	0	0
##	20.54	13	0	0	0	0
##	20.55	15	0	0	0	0
##	20.56	2	0	0	0	0
##	20.57	3	0	0	0	0
##	20.58	2	0	0	0	0
##	20.6	119	0	0	0	0
##	20.61	3	0	0	0	0
##	20.62	5	0	0	0	0
##	20.63	6	0	0	0	0
##	20.64	3	0	0	0	0
##	20.66	20	0	0	0	0
##	20.67	59	0	0	0	0
##	20.68	7	0	0	0	0
##	20.69	4	0	0	0	0
##	20.7	8	0	0	0	0
##	20.71	1	0	0	0	0
##	20.72	7	0	0	0	0
##	20.73	12	0	0	0	0
##	20.74	2	0	0	0	0
##	20.75	5	0	0	0	0
##	20.76	8	0	0	0	0
##	20.77	3	0	0	0	0
##	20.78	44	0	0	0	0
##	20.8	96	0	0	0	0
##	20.81	34	0	0	0	0
##	20.82	8	0	0	0	0
##	20.83	24	0	0	0	0
##	20.85	18	0	0	0	0
##	20.87	1	0	0	0	0
##	20.89	1	0	0	0	0
##	20.9	24	0	0	0	0
##	20.91	4	0	0	0	0
##	20.92	21	0	0	0	0
##	20.93	4	0	0	0	0
##	20.94	19	0	0	0	0
##	20.95	2	0	0	0	0
##	20.96	4	0	0	0	0
##	20.97	16	0	0	0	0
##	20.98	111	0	0	0	0
##	20.99	9	0	0	0	0
##	21.01	1	0	0	0	0
##	21.02	20	0	0	0	0
##	21.03	69	0	0	0	0
##	21.06	1	0	0	0	0

##	21.08	11	0	0	0	0
##	21.09	8	0	0	0	0
##	21.1	2	0	0	0	0
##	21.11	31	0	0	0	0
##	21.12	3	0	0	0	0
##	21.13	12	0	0	0	0
##	21.14	89	0	0	0	0
##	21.16	16	0	0	0	0
##	21.18	8	0	0	0	0
##	21.19	4	0	0	0	0
##	21.2	2	0	0	0	0
##	21.21	10	0	0	0	0
##	21.22	16	0	0	0	0
##	21.23	2	0	0	0	0
##	21.24	5	0	0	0	0
##	21.25	5	0	0	0	0
##	21.26	106	0	0	0	0
##	21.27	1	0	0	0	0
##	21.28	16	0	0	0	0
##	21.29	84	0	0	0	0
##	21.3	25	0	0	0	0
##	21.31	16	0	0	0	0
##	21.34	9	0	0	0	0
##	21.35	5	0	0	0	0
##	21.37	2	0	0	0	0
##	21.38	1	0	0	0	0
##	21.4	13	0	0	0	0
##	21.41	61	0	0	0	0
##	21.42	1	0	0	0	0
##	21.43	7	0	0	0	0
##	21.44	2	0	0	0	0
##	21.45	7	0	0	0	0
##	21.46	111	0	0	0	0
##	21.47	22	0	0	0	0
##	21.48	50	0	0	0	0
##	21.5	2	0	0	0	0
##	21.51	3	0	0	0	0
##	21.52	69	0	0	0	0
##	21.53	4	0	0	0	0
##	21.54	6	0	0	0	0
##	21.55	2	0	0	0	0
##	21.56	9	0	0	0	0
##	21.57	4	0	0	0	0
##	21.58	22	0	0	0	0
##	21.59	3	0	0	0	0
##	21.6	6	0	0	0	0
##	21.61	30	0	0	0	0
##	21.62	16	0	0	0	0
##	21.63	136	0	0	0	0
##	21.64	2	0	0	0	0
##	21.66	3	0	0	0	0
##	21.67	2	0	0	0	0
##	21.68	1	0	0	0	0
##	21.7	48	0	0	0	0

##	21.71	8	0	0	0	0
##	21.72	7	0	0	0	0
##	21.73	32	0	0	0	0
##	21.74	7	0	0	0	0
##	21.75	1	0	0	0	0
##	21.76	5	0	0	0	0
##	21.77	27	0	0	0	0
##	21.79	114	0	0	0	0
##	21.8	27	0	0	0	0
##	21.81	13	0	0	0	0
##	21.83	17	0	0	0	0
##	21.86	10	0	0	0	0
##	21.87	11	0	0	0	0
##	21.88	1	0	0	0	0
##	21.89	4	0	0	0	0
##	21.9	5	0	0	0	0
##	21.91	8	0	0	0	0
##	21.92	5	0	0	0	0
##	21.93	109	0	0	0	0
##	21.94	6	0	0	0	0
##	21.95	98	0	0	0	0
##	21.96	1	0	0	0	0
##	21.97	65	0	0	0	0
##	21.99	2	0	0	0	0
##	22	3	0	0	0	0
##	22.01	2	0	0	0	0
##	22.03	2	0	0	0	0
##	22.04	12	0	0	0	0
##	22.05	85	0	0	0	0
##	22.06	1	0	0	0	0
##	22.07	2	0	0	0	0
##	22.08	3	0	0	0	0
##	22.1	7	0	0	0	0
##	22.11	20	0	0	0	0
##	22.12	1	0	0	0	0
##	22.13	5	0	0	0	0
##	22.14	92	0	0	0	0
##	22.15	71	0	0	0	0
##	22.16	2	0	0	0	0
##	22.18	3	0	0	0	0
##	22.2	8	0	0	0	0
##	22.21	3	0	0	0	0
##	22.22	14	0	0	0	0
##	22.24	65	0	0	0	0
##	22.25	1	0	0	0	0
##	22.26	7	0	0	0	0
##	22.27	10	0	0	0	0
##	22.28	1	0	0	0	0
##	22.3	22	0	0	0	0
##	22.31	141	0	0	0	0
##	22.32	51	0	0	0	0
##	22.34	3	0	0	0	0
##	22.35	9	0	0	0	0
##	22.37	2	0	0	0	0



##	22.38	55	0	0	0	0
##	22.39	1	0	0	0	0
##	22.4	8	0	0	0	0
##	22.41	1	0	0	0	0
##	22.42	5	0	0	0	0
##	22.43	25	0	0	0	0
##	22.44	5	0	0	0	0
##	22.45	8	0	0	0	0
##	22.46	141	0	0	0	0
##	22.47	20	0	0	0	0
##	22.48	6	0	0	0	0
##	22.49	4	0	0	0	0
##	22.5	45	0	0	0	0
##	22.51	2	0	0	0	0
##	22.52	5	0	0	0	0
##	22.53	6	0	0	0	0
##	22.55	5	0	0	0	0
##	22.58	1	0	0	0	0
##	22.59	13	0	0	0	0
##	22.6	116	0	0	0	0
##	22.62	2	0	0	0	0
##	22.63	10	0	0	0	0
##	22.65	5	0	0	0	0
##	22.66	21	0	0	0	0
##	22.67	66	0	0	0	0
##	22.68	12	0	0	0	0
##	22.7	1	0	0	0	0
##	22.71	80	0	0	0	0
##	22.72	3	0	0	0	0
##	22.73	6	0	0	0	0
##	22.74	3	0	0	0	0
##	22.75	2	0	0	0	0
##	22.76	3	0	0	0	0
##	22.77	5	0	0	0	0
##	22.78	4	0	0	0	0
##	22.79	12	0	0	0	0
##	22.8	13	0	0	0	0
##	22.81	99	0	0	0	0
##	22.82	2	0	0	0	0
##	22.83	14	0	0	0	0
##	22.84	3	0	0	0	0
##	22.85	12	0	0	0	0
##	22.86	75	0	0	0	0
##	22.87	6	0	0	0	0
##	22.88	3	0	0	0	0
##	22.89	66	0	0	0	0
##	22.91	4	0	0	0	0
##	22.92	16	0	0	0	0
##	22.95	1	0	0	0	0
##	22.96	112	0	0	0	0
##	22.97	1	0	0	0	0
##	22.98	5	0	0	0	0
##	22.99	5	0	0	0	0
##	23	9	0	0	0	0

##	23.01	46	0	0	0	0
##	23.02	9	0	0	0	0
##	23.03	100	0	0	0	0
##	23.04	2	0	0	0	0
##	23.05	27	0	0	0	0
##	23.06	48	0	0	0	0
##	23.08	9	0	0	0	0
##	23.09	29	0	0	0	0
##	23.1	7	0	0	0	0
##	23.11	42	0	0	0	0
##	23.12	12	0	0	0	0
##	23.13	14	0	0	0	0
##	23.14	1	0	0	0	0
##	23.15	12	0	0	0	0
##	23.17	93	0	0	0	0
##	23.18	20	0	0	0	0
##	23.19	1	0	0	0	0
##	23.21	5	0	0	0	0
##	23.22	4	0	0	0	0
##	23.23	17	0	0	0	0
##	23.24	16	0	0	0	0
##	23.25	1	0	0	0	0
##	23.26	7	0	0	0	0
##	23.27	1	0	0	0	0
##	23.29	4	0	0	0	0
##	23.3	128	0	0	0	0
##	23.31	3	0	0	0	0
##	23.32	5	0	0	0	0
##	23.33	21	0	0	0	0
##	23.34	17	0	0	0	0
##	23.36	2	0	0	0	0
##	23.37	5	0	0	0	0
##	23.38	9	0	0	0	0
##	23.39	14	0	0	0	0
##	23.4	91	0	0	0	0
##	23.41	23	0	0	0	0
##	23.42	9	0	0	0	0
##	23.43	15	0	0	0	0
##	23.44	54	0	0	0	0
##	23.45	2	0	0	0	0
##	23.46	13	0	0	0	0
##	23.48	14	0	0	0	0
##	23.49	109	0	0	0	0
##	23.5	1	0	0	0	0
##	23.51	3	0	0	0	0
##	23.52	13	0	0	0	0
##	23.53	7	0	0	0	0
##	23.54	5	0	0	0	0
##	23.55	2	0	0	0	0
##	23.56	18	0	0	0	0
##	23.57	84	0	0	0	0
##	23.58	1	0	0	0	0
##	23.59	3	0	0	0	0
##	23.6	6	0	0	0	0

##	23.62	33	0	0	0	0
##	23.63	105	0	0	0	0
##	23.65	3	0	0	0	0
##	23.66	1	0	0	0	0
##	23.67	88	0	0	0	0
##	23.68	3	0	0	0	0
##	23.69	15	0	0	0	0
##	23.7	3	0	0	0	0
##	23.71	64	0	0	0	0
##	23.72	14	0	0	0	0
##	23.73	84	0	0	0	0
##	23.74	11	0	0	0	0
##	23.75	72	0	0	0	0
##	23.77	1	0	0	0	0
##	23.78	102	0	0	0	0
##	23.79	1	0	0	0	0
##	23.8	11	0	0	0	0
##	23.81	17	0	0	0	0
##	23.82	5	0	0	0	0
##	23.83	5	0	0	0	0
##	23.84	3	0	0	0	0
##	23.85	6	0	0	0	0
##	23.86	8	0	0	0	0
##	23.87	9	0	0	0	0
##	23.88	8	0	0	0	0
##	23.89	17	0	0	0	0
##	23.9	1	0	0	0	0
##	23.91	78	0	0	0	0
##	23.92	12	0	0	0	0
##	23.94	5	0	0	0	0
##	23.95	3	0	0	0	0
##	23.96	17	0	0	0	0
##	23.98	1	0	0	0	0
##	23.99	9	0	0	0	0
##	24	7	0	0	0	0
##	24.01	12	0	0	0	0
##	24.02	20	0	0	0	0
##	24.03	120	0	0	0	0
##	24.04	3	0	0	0	0
##	24.05	5	0	0	0	0
##	24.06	1	0	0	0	0
##	24.07	7	0	0	0	0
##	24.09	10	0	0	0	0
##	24.1	1	0	0	0	0
##	24.11	8	0	0	0	0
##	24.12	3	0	0	0	0
##	24.13	77	0	0	0	0
##	24.14	29	0	0	0	0
##	24.15	1	0	0	0	0
##	24.16	4	0	0	0	0
##	24.17	4	0	0	0	0
##	24.18	4	0	0	0	0
##	24.19	15	0	0	0	0
##	24.2	5	0	0	0	0

##	24.21	136	0	0	0	0
##	24.22	19	0	0	0	0
##	24.23	1	0	0	0	0
##	24.24	11	0	0	0	0
##	24.25	5	0	0	0	0
##	24.27	10	0	0	0	0
##	24.28	99	0	0	0	0
##	24.3	2	0	0	0	0
##	24.31	4	0	0	0	0
##	24.33	109	0	0	0	0
##	24.34	18	0	0	0	0
##	24.36	1	0	0	0	0
##	24.37	87	0	0	0	0
##	24.38	3	0	0	0	0
##	24.39	132	0	0	0	0
##	24.41	222	0	0	0	0
##	24.42	1	0	0	0	0
##	24.43	3	0	0	0	0
##	24.44	2	0	0	0	0
##	24.45	14	0	0	0	0
##	24.46	5	0	0	0	0
##	24.47	1	0	0	0	0
##	24.48	3	0	0	0	0
##	24.49	7	0	0	0	0
##	24.5	2	0	0	0	0
##	24.51	15	0	0	0	0
##	24.52	4	0	0	0	0
##	24.53	6	0	0	0	0
##	24.54	5	0	0	0	0
##	24.55	10	0	0	0	0
##	24.56	37	0	0	0	0
##	24.57	1	0	0	0	0
##	24.58	3	0	0	0	0
##	24.59	10	0	0	0	0
##	24.6	4	0	0	0	0
##	24.61	5	0	0	0	0
##	24.62	6	0	0	0	0
##	24.63	31	0	0	0	0
##	24.64	1	0	0	0	0
##	24.65	7	0	0	0	0
##	24.66	3	0	0	0	0
##	24.67	8	0	0	0	0
##	24.68	15	0	0	0	0
##	24.69	94	0	0	0	0
##	24.72	7	0	0	0	0
##	24.73	1	0	0	0	0
##	24.74	1	0	0	0	0
##	24.75	13	0	0	0	0
##	24.77	3	0	0	0	0
##	24.78	8	0	0	0	0
##	24.8	90	0	0	0	0
##	24.81	19	0	0	0	0
##	24.82	7	0	0	0	0
##	24.83	18	0	0	0	0

##	24.85	1	0	0	0	0
##	24.86	12	0	0	0	0
##	24.87	2	0	0	0	0
##	24.89	86	0	0	0	0
##	24.9	9	0	0	0	0
##	24.91	8	0	0	0	0
##	24.92	1	0	0	0	0
##	24.93	2	0	0	0	0
##	24.94	6	0	0	0	0
##	24.95	11	0	0	0	0
##	24.96	120	0	0	0	0
##	24.97	14	0	0	0	0
##	24.98	4	0	0	0	0
##	24.99	3	0	0	0	0
##	25	0	0	0	15	0
##	25.01	0	0	0	2	0
##	25.02	0	0	0	65	0
##	25.03	0	0	0	1	0
##	25.04	0	0	0	14	0
##	25.06	0	0	0	130	0
##	25.07	0	0	0	36	0
##	25.08	0	0	0	2	0
##	25.09	0	0	0	157	0
##	25.1	0	0	0	145	0
##	25.11	0	0	0	69	0
##	25.13	0	0	0	3	0
##	25.14	0	0	0	7	0
##	25.15	0	0	0	10	0
##	25.16	0	0	0	2	0
##	25.18	0	0	0	8	0
##	25.19	0	0	0	3	0
##	25.2	0	0	0	3	0
##	25.22	0	0	0	1	0
##	25.23	0	0	0	11	0
##	25.24	0	0	0	31	0
##	25.25	0	0	0	17	0
##	25.26	0	0	0	1	0
##	25.27	0	0	0	2	0
##	25.28	0	0	0	1	0
##	25.29	0	0	0	9	0
##	25.31	0	0	0	5	0
##	25.32	0	0	0	5	0
##	25.33	0	0	0	11	0
##	25.34	0	0	0	4	0
##	25.35	0	0	0	3	0
##	25.36	0	0	0	4	0
##	25.37	0	0	0	11	0
##	25.38	0	0	0	9	0
##	25.39	0	0	0	48	0
##	25.4	0	0	0	22	0
##	25.41	0	0	0	4	0
##	25.42	0	0	0	6	0
##	25.43	0	0	0	2	0
##	25.44	0	0	0	1	0

##	25.46	0	0	0	11	0
##	25.47	0	0	0	3	0
##	25.48	0	0	0	2	0
##	25.5	0	0	0	20	0
##	25.51	0	0	0	36	0
##	25.52	0	0	0	5	0
##	25.53	0	0	0	4	0
##	25.54	0	0	0	30	0
##	25.55	0	0	0	10	0
##	25.56	0	0	0	9	0
##	25.58	0	0	0	6	0
##	25.59	0	0	0	2	0
##	25.61	0	0	0	77	0
##	25.62	0	0	0	12	0
##	25.63	0	0	0	8	0
##	25.65	0	0	0	1	0
##	25.66	0	0	0	9	0
##	25.68	0	0	0	42	0
##	25.69	0	0	0	70	0
##	25.7	0	0	0	11	0
##	25.71	0	0	0	8	0
##	25.72	0	0	0	1	0
##	25.73	0	0	0	21	0
##	25.75	0	0	0	119	0
##	25.76	0	0	0	2	0
##	25.77	0	0	0	63	0
##	25.78	0	0	0	4	0
##	25.79	0	0	0	51	0
##	25.8	0	0	0	63	0
##	25.82	0	0	0	90	0
##	25.83	0	0	0	98	0
##	25.84	0	0	0	125	0
##	25.85	0	0	0	93	0
##	25.86	0	0	0	4	0
##	25.88	0	0	0	2	0
##	25.89	0	0	0	5	0
##	25.9	0	0	0	3	0
##	25.91	0	0	0	1	0
##	25.92	0	0	0	5	0
##	25.93	0	0	0	6	0
##	25.94	0	0	0	9	0
##	25.95	0	0	0	8	0
##	25.96	0	0	0	4	0
##	25.97	0	0	0	8	0
##	25.98	0	0	0	2	0
##	25.99	0	0	0	10	0
##	26	0	0	0	7	0
##	26.01	0	0	0	1	0
##	26.04	0	0	0	10	0
##	26.05	0	0	0	4	0
##	26.07	0	0	0	10	0
##	26.08	0	0	0	5	0
##	26.09	0	0	0	11	0
##	26.11	0	0	0	6	0

##	26.12	0	0	0	10	0
##	26.13	0	0	0	4	0
##	26.14	0	0	0	5	0
##	26.15	0	0	0	31	0
##	26.16	0	0	0	7	0
##	26.17	0	0	0	9	0
##	26.18	0	0	0	2	0
##	26.19	0	0	0	8	0
##	26.21	0	0	0	1	0
##	26.22	0	0	0	20	0
##	26.23	0	0	0	4	0
##	26.24	0	0	0	4	0
##	26.25	0	0	0	14	0
##	26.26	0	0	0	20	0
##	26.29	0	0	0	17	0
##	26.3	0	0	0	9	0
##	26.31	0	0	0	19	0
##	26.32	0	0	0	13	0
##	26.33	0	0	0	1	0
##	26.34	0	0	0	5	0
##	26.35	0	0	0	1	0
##	26.36	0	0	0	6	0
##	26.37	0	0	0	21	0
##	26.39	0	0	0	46	0
##	26.4	0	0	0	2	0
##	26.41	0	0	0	1	0
##	26.42	0	0	0	1	0
##	26.43	0	0	0	15	0
##	26.45	0	0	0	88	0
##	26.46	0	0	0	11	0
##	26.47	0	0	0	12	0
##	26.48	0	0	0	1	0
##	26.49	0	0	0	1	0
##	26.5	0	0	0	51	0
##	26.51	0	0	0	1	0
##	26.52	0	0	0	44	0
##	26.54	0	0	0	76	0
##	26.56	0	0	0	4	0
##	26.57	0	0	0	76	0
##	26.58	0	0	0	98	0
##	26.59	0	0	0	2	0
##	26.6	0	0	0	1	0
##	26.61	0	0	0	128	0
##	26.62	0	0	0	2	0
##	26.63	0	0	0	231	0
##	26.64	0	0	0	3	0
##	26.65	0	0	0	2	0
##	26.66	0	0	0	6	0
##	26.68	0	0	0	9	0
##	26.69	0	0	0	7	0
##	26.7	0	0	0	11	0
##	26.71	0	0	0	2	0
##	26.72	0	0	0	3	0
##	26.73	0	0	0	3	0

##	26.75	0	0	0	2	0
##	26.76	0	0	0	4	0
##	26.78	0	0	0	19	0
##	26.79	0	0	0	5	0
##	26.81	0	0	0	1	0
##	26.82	0	0	0	1	0
##	26.83	0	0	0	8	0
##	26.84	0	0	0	1	0
##	26.85	0	0	0	6	0
##	26.86	0	0	0	1	0
##	26.87	0	0	0	10	0
##	26.88	0	0	0	7	0
##	26.89	0	0	0	9	0
##	26.9	0	0	0	2	0
##	26.91	0	0	0	8	0
##	26.92	0	0	0	2	0
##	26.93	0	0	0	13	0
##	26.94	0	0	0	6	0
##	26.95	0	0	0	11	0
##	26.96	0	0	0	32	0
##	26.97	0	0	0	5	0
##	26.98	0	0	0	2	0
##	26.99	0	0	0	3	0
##	27.01	0	0	0	2	0
##	27.02	0	0	0	6	0
##	27.04	0	0	0	2	0
##	27.05	0	0	0	17	0
##	27.06	0	0	0	15	0
##	27.07	0	0	0	14	0
##	27.09	0	0	0	1	0
##	27.1	0	0	0	10	0
##	27.12	0	0	0	131	0
##	27.13	0	0	0	3	0
##	27.14	0	0	0	3	0
##	27.15	0	0	0	3	0
##	27.16	0	0	0	4	0
##	27.17	0	0	0	9	0
##	27.18	0	0	0	1	0
##	27.2	0	0	0	31	0
##	27.21	0	0	0	3	0
##	27.22	0	0	0	8	0
##	27.25	0	0	0	8	0
##	27.26	0	0	0	77	0
##	27.27	0	0	0	8	0
##	27.28	0	0	0	6	0
##	27.29	0	0	0	6	0
##	27.31	0	0	0	3	0
##	27.32	0	0	0	45	0
##	27.34	0	0	0	36	0
##	27.35	0	0	0	3	0
##	27.36	0	0	0	2	0
##	27.37	0	0	0	80	0
##	27.39	0	0	0	6	0
##	27.4	0	0	0	14	0



##	27.41	0	0	0	46	0
##	27.44	0	0	0	133	0
##	27.46	0	0	0	144	0
##	27.47	0	0	0	2	0
##	27.48	0	0	0	9	0
##	27.49	0	0	0	1	0
##	27.5	0	0	0	7	0
##	27.51	0	0	0	4	0
##	27.52	0	0	0	1	0
##	27.53	0	0	0	2	0
##	27.54	0	0	0	1	0
##	27.55	0	0	0	3	0
##	27.56	0	0	0	3	0
##	27.57	0	0	0	5	0
##	27.59	0	0	0	1	0
##	27.6	0	0	0	22	0
##	27.61	0	0	0	4	0
##	27.62	0	0	0	14	0
##	27.63	0	0	0	6	0
##	27.64	0	0	0	4	0
##	27.67	0	0	0	12	0
##	27.68	0	0	0	3	0
##	27.69	0	0	0	5	0
##	27.7	0	0	0	1	0
##	27.71	0	0	0	19	0
##	27.72	0	0	0	5	0
##	27.73	0	0	0	6	0
##	27.74	0	0	0	1	0
##	27.75	0	0	0	3	0
##	27.76	0	0	0	21	0
##	27.78	0	0	0	6	0
##	27.79	0	0	0	6	0
##	27.8	0	0	0	34	0
##	27.81	0	0	0	7	0
##	27.82	0	0	0	6	0
##	27.84	0	0	0	7	0
##	27.86	0	0	0	1	0
##	27.87	0	0	0	4	0
##	27.88	0	0	0	6	0
##	27.89	0	0	0	62	0
##	27.91	0	0	0	4	0
##	27.92	0	0	0	2	0
##	27.93	0	0	0	3	0
##	27.94	0	0	0	1	0
##	27.96	0	0	0	11	0
##	27.97	0	0	0	2	0
##	27.98	0	0	0	41	0
##	27.99	0	0	0	5	0
##	28	0	0	0	9	0
##	28.01	0	0	0	3	0
##	28.03	0	0	0	2	0
##	28.04	0	0	0	2	0
##	28.06	0	0	0	68	0
##	28.07	0	0	0	3	0

##	28.08	0	0	0	5	0
##	28.09	0	0	0	2	0
##	28.1	0	0	0	1	0
##	28.11	0	0	0	1	0
##	28.12	0	0	0	16	0
##	28.13	0	0	0	47	0
##	28.15	0	0	0	4	0
##	28.16	0	0	0	1	0
##	28.17	0	0	0	11	0
##	28.19	0	0	0	88	0
##	28.21	0	0	0	11	0
##	28.22	0	0	0	1	0
##	28.23	0	0	0	3	0
##	28.24	0	0	0	1	0
##	28.25	0	0	0	87	0
##	28.27	0	0	0	1	0
##	28.28	0	0	0	7	0
##	28.29	0	0	0	51	0
##	28.3	0	0	0	1	0
##	28.31	0	0	0	4	0
##	28.32	0	0	0	51	0
##	28.34	0	0	0	78	0
##	28.35	0	0	0	33	0
##	28.36	0	0	0	1	0
##	28.37	0	0	0	15	0
##	28.38	0	0	0	3	0
##	28.4	0	0	0	3	0
##	28.41	0	0	0	10	0
##	28.43	0	0	0	2	0
##	28.44	0	0	0	1	0
##	28.45	0	0	0	1	0
##	28.46	0	0	0	1	0
##	28.47	0	0	0	6	0
##	28.48	0	0	0	46	0
##	28.49	0	0	0	3	0
##	28.5	0	0	0	6	0
##	28.52	0	0	0	2	0
##	28.53	0	0	0	5	0
##	28.55	0	0	0	5	0
##	28.57	0	0	0	7	0
##	28.59	0	0	0	30	0
##	28.6	0	0	0	6	0
##	28.62	0	0	0	8	0
##	28.63	0	0	0	4	0
##	28.65	0	0	0	5	0
##	28.66	0	0	0	5	0
##	28.67	0	0	0	4	0
##	28.68	0	0	0	1	0
##	28.7	0	0	0	81	0
##	28.71	0	0	0	3	0
##	28.72	0	0	0	3	0
##	28.73	0	0	0	13	0
##	28.74	0	0	0	6	0
##	28.75	0	0	0	16	0

##	28.76	0	0	0	2	0
##	28.78	0	0	0	2	0
##	28.79	0	0	0	3	0
##	28.8	0	0	0	32	0
##	28.82	0	0	0	1	0
##	28.84	0	0	0	10	0
##	28.85	0	0	0	1	0
##	28.87	0	0	0	4	0
##	28.88	0	0	0	1	0
##	28.89	0	0	0	78	0
##	28.9	0	0	0	8	0
##	28.91	0	0	0	1	0
##	28.93	0	0	0	1	0
##	28.94	0	0	0	1	0
##	28.95	0	0	0	2	0
##	28.97	0	0	0	28	0
##	28.98	0	0	0	5	0
##	29	0	0	0	2	0
##	29.01	0	0	0	7	0
##	29.02	0	0	0	2	0
##	29.03	0	0	0	19	0
##	29.04	0	0	0	2	0
##	29.05	0	0	0	88	0
##	29.07	0	0	0	1	0
##	29.08	0	0	0	1	0
##	29.09	0	0	0	1	0
##	29.1	0	0	0	8	0
##	29.11	0	0	0	1	0
##	29.12	0	0	0	36	0
##	29.13	0	0	0	3	0
##	29.14	0	0	0	1	0
##	29.15	0	0	0	2	0
##	29.16	0	0	0	18	0
##	29.18	0	0	0	65	0
##	29.19	0	0	0	4	0
##	29.2	0	0	0	1	0
##	29.21	0	0	0	6	0
##	29.22	0	0	0	2	0
##	29.23	0	0	0	28	0
##	29.24	0	0	0	5	0
##	29.26	0	0	0	40	0
##	29.27	0	0	0	1	0
##	29.29	0	0	0	92	0
##	29.3	0	0	0	3	0
##	29.32	0	0	0	1	0
##	29.35	0	0	0	4	0
##	29.36	0	0	0	1	0
##	29.37	0	0	0	8	0
##	29.38	0	0	0	8	0
##	29.39	0	0	0	1	0
##	29.4	0	0	0	2	0
##	29.41	0	0	0	26	0
##	29.42	0	0	0	1	0
##	29.43	0	0	0	3	0

##	29.44	0	0	0	4	0
##	29.45	0	0	0	2	0
##	29.49	0	0	0	1	0
##	29.5	0	0	0	3	0
##	29.52	0	0	0	4	0
##	29.53	0	0	0	73	0
##	29.54	0	0	0	3	0
##	29.55	0	0	0	1	0
##	29.57	0	0	0	3	0
##	29.58	0	0	0	2	0
##	29.6	0	0	0	4	0
##	29.62	0	0	0	7	0
##	29.63	0	0	0	5	0
##	29.64	0	0	0	1	0
##	29.65	0	0	0	26	0
##	29.67	0	0	0	2	0
##	29.68	0	0	0	15	0
##	29.69	0	0	0	3	0
##	29.7	0	0	0	5	0
##	29.71	0	0	0	2	0
##	29.74	0	0	0	1	0
##	29.75	0	0	0	1	0
##	29.76	0	0	0	48	0
##	29.79	0	0	0	2	0
##	29.8	0	0	0	4	0
##	29.81	0	0	0	1	0
##	29.82	0	0	0	3	0
##	29.83	0	0	0	2	0
##	29.84	0	0	0	57	0
##	29.85	0	0	0	3	0
##	29.86	0	0	0	25	0
##	29.87	0	0	0	3	0
##	29.88	0	0	0	1	0
##	29.9	0	0	0	5	0
##	29.91	0	0	0	1	0
##	29.92	0	0	0	1	0
##	29.94	0	0	0	2	0
##	29.95	0	0	0	56	0
##	29.98	0	0	0	2	0
##	29.99	0	0	0	19	0
##	30	0	13	0	0	0
##	30.02	0	4	0	0	0
##	30.04	0	50	0	0	0
##	30.05	0	1	0	0	0
##	30.07	0	2	0	0	0
##	30.08	0	4	0	0	0
##	30.11	0	43	0	0	0
##	30.12	0	4	0	0	0
##	30.13	0	54	0	0	0
##	30.14	0	2	0	0	0
##	30.17	0	9	0	0	0
##	30.18	0	29	0	0	0
##	30.21	0	2	0	0	0
##	30.23	0	29	0	0	0

##	30.24	0	4	0	0	0
##	30.25	0	4	0	0	0
##	30.27	0	24	0	0	0
##	30.28	0	4	0	0	0
##	30.29	0	3	0	0	0
##	30.3	0	7	0	0	0
##	30.31	0	2	0	0	0
##	30.34	0	21	0	0	0
##	30.36	0	6	0	0	0
##	30.37	0	3	0	0	0
##	30.38	0	4	0	0	0
##	30.4	0	1	0	0	0
##	30.41	0	74	0	0	0
##	30.42	0	13	0	0	0
##	30.43	0	7	0	0	0
##	30.45	0	2	0	0	0
##	30.46	0	1	0	0	0
##	30.47	0	5	0	0	0
##	30.5	0	2	0	0	0
##	30.51	0	3	0	0	0
##	30.52	0	39	0	0	0
##	30.54	0	17	0	0	0
##	30.55	0	4	0	0	0
##	30.56	0	3	0	0	0
##	30.59	0	2	0	0	0
##	30.61	0	1	0	0	0
##	30.62	0	12	0	0	0
##	30.65	0	4	0	0	0
##	30.66	0	2	0	0	0
##	30.67	0	38	0	0	0
##	30.68	0	43	0	0	0
##	30.69	0	1	0	0	0
##	30.7	0	1	0	0	0
##	30.71	0	4	0	0	0
##	30.72	0	3	0	0	0
##	30.73	0	9	0	0	0
##	30.74	0	1	0	0	0
##	30.75	0	4	0	0	0
##	30.78	0	2	0	0	0
##	30.79	0	26	0	0	0
##	30.8	0	3	0	0	0
##	30.81	0	9	0	0	0
##	30.82	0	2	0	0	0
##	30.83	0	4	0	0	0
##	30.85	0	23	0	0	0
##	30.86	0	16	0	0	0
##	30.87	0	1	0	0	0
##	30.88	0	1	0	0	0
##	30.9	0	54	0	0	0
##	30.91	0	3	0	0	0
##	30.95	0	4	0	0	0
##	30.96	0	2	0	0	0
##	30.99	0	9	0	0	0
##	31	0	35	0	0	0

##	31.01	0	21	0	0	0
##	31.02	0	2	0	0	0
##	31.04	0	3	0	0	0
##	31.05	0	3	0	0	0
##	31.07	0	2	0	0	0
##	31.08	0	1	0	0	0
##	31.09	0	30	0	0	0
##	31.12	0	1	0	0	0
##	31.14	0	4	0	0	0
##	31.15	0	1	0	0	0
##	31.16	0	1	0	0	0
##	31.17	0	11	0	0	0
##	31.18	0	12	0	0	0
##	31.19	0	23	0	0	0
##	31.2	0	1	0	0	0
##	31.21	0	1	0	0	0
##	31.22	0	1	0	0	0
##	31.24	0	5	0	0	0
##	31.25	0	35	0	0	0
##	31.28	0	3	0	0	0
##	31.29	0	1	0	0	0
##	31.31	0	5	0	0	0
##	31.32	0	58	0	0	0
##	31.33	0	1	0	0	0
##	31.35	0	6	0	0	0
##	31.36	0	1	0	0	0
##	31.38	0	15	0	0	0
##	31.4	0	2	0	0	0
##	31.41	0	2	0	0	0
##	31.42	0	1	0	0	0
##	31.45	0	2	0	0	0
##	31.46	0	13	0	0	0
##	31.47	0	15	0	0	0
##	31.48	0	2	0	0	0
##	31.5	0	1	0	0	0
##	31.51	0	2	0	0	0
##	31.52	0	2	0	0	0
##	31.53	0	4	0	0	0
##	31.55	0	1	0	0	0
##	31.57	0	39	0	0	0
##	31.58	0	5	0	0	0
##	31.59	0	1	0	0	0
##	31.6	0	2	0	0	0
##	31.62	0	37	0	0	0
##	31.63	0	5	0	0	0
##	31.64	0	4	0	0	0
##	31.65	0	3	0	0	0
##	31.66	0	20	0	0	0
##	31.67	0	1	0	0	0
##	31.69	0	1	0	0	0
##	31.71	0	3	0	0	0
##	31.72	0	1	0	0	0
##	31.74	0	7	0	0	0
##	31.75	0	32	0	0	0

##	31.78	0	2	0	0	0
##	31.79	0	2	0	0	0
##	31.8	0	5	0	0	0
##	31.82	0	1	0	0	0
##	31.83	0	1	0	0	0
##	31.84	0	1	0	0	0
##	31.85	0	1	0	0	0
##	31.86	0	1	0	0	0
##	31.87	0	18	0	0	0
##	31.89	0	39	0	0	0
##	31.91	0	1	0	0	0
##	31.92	0	2	0	0	0
##	31.93	0	24	0	0	0
##	31.95	0	5	0	0	0
##	31.96	0	4	0	0	0
##	31.99	0	2	0	0	0
##	32	0	3	0	0	0
##	32.01	0	16	0	0	0
##	32.02	0	3	0	0	0
##	32.06	0	1	0	0	0
##	32.08	0	21	0	0	0
##	32.1	0	17	0	0	0
##	32.11	0	10	0	0	0
##	32.12	0	15	0	0	0
##	32.14	0	2	0	0	0
##	32.17	0	1	0	0	0
##	32.22	0	2	0	0	0
##	32.23	0	1	0	0	0
##	32.24	0	1	0	0	0
##	32.26	0	2	0	0	0
##	32.27	0	2	0	0	0
##	32.28	0	89	0	0	0
##	32.29	0	5	0	0	0
##	32.32	0	12	0	0	0
##	32.36	0	2	0	0	0
##	32.38	0	1	0	0	0
##	32.41	0	2	0	0	0
##	32.42	0	1	0	0	0
##	32.43	0	1	0	0	0
##	32.44	0	4	0	0	0
##	32.45	0	8	0	0	0
##	32.49	0	33	0	0	0
##	32.5	0	7	0	0	0
##	32.54	0	1	0	0	0
##	32.55	0	17	0	0	0
##	32.56	0	7	0	0	0
##	32.58	0	1	0	0	0
##	32.6	0	5	0	0	0
##	32.61	0	36	0	0	0
##	32.62	0	3	0	0	0
##	32.64	0	1	0	0	0
##	32.65	0	3	0	0	0
##	32.66	0	1	0	0	0
##	32.69	0	17	0	0	0

##	32.71	0	1	0	0	0
##	32.73	0	2	0	0	0
##	32.74	0	7	0	0	0
##	32.77	0	5	0	0	0
##	32.78	0	12	0	0	0
##	32.8	0	1	0	0	0
##	32.81	0	1	0	0	0
##	32.84	0	2	0	0	0
##	32.85	0	1	0	0	0
##	32.86	0	1	0	0	0
##	32.87	0	3	0	0	0
##	32.88	0	1	0	0	0
##	32.89	0	24	0	0	0
##	32.9	0	1	0	0	0
##	32.91	0	2	0	0	0
##	32.92	0	24	0	0	0
##	32.93	0	2	0	0	0
##	32.95	0	7	0	0	0
##	32.96	0	2	0	0	0
##	32.98	0	21	0	0	0
##	32.99	0	1	0	0	0
##	33	0	24	0	0	0
##	33.02	0	1	0	0	0
##	33.03	0	1	0	0	0
##	33.05	0	2	0	0	0
##	33.07	0	5	0	0	0
##	33.08	0	1	0	0	0
##	33.09	0	11	0	0	0
##	33.11	0	2	0	0	0
##	33.12	0	9	0	0	0
##	33.13	0	2	0	0	0
##	33.14	0	1	0	0	0
##	33.19	0	4	0	0	0
##	33.2	0	14	0	0	0
##	33.21	0	1	0	0	0
##	33.23	0	21	0	0	0
##	33.25	0	2	0	0	0
##	33.28	0	49	0	0	0
##	33.29	0	4	0	0	0
##	33.3	0	1	0	0	0
##	33.31	0	1	0	0	0
##	33.32	0	2	0	0	0
##	33.33	0	3	0	0	0
##	33.36	0	1	0	0	0
##	33.38	0	8	0	0	0
##	33.43	0	1	0	0	0
##	33.44	0	1	0	0	0
##	33.45	0	25	0	0	0
##	33.46	0	1	0	0	0
##	33.47	0	26	0	0	0
##	33.48	0	4	0	0	0
##	33.49	0	1	0	0	0
##	33.52	0	5	0	0	0
##	33.57	0	1	0	0	0



##	33.58	0	1	0	0	0
##	33.59	0	3	0	0	0
##	33.61	0	2	0	0	0
##	33.63	0	2	0	0	0
##	33.64	0	12	0	0	0
##	33.66	0	17	0	0	0
##	33.67	0	11	0	0	0
##	33.68	0	1	0	0	0
##	33.72	0	8	0	0	0
##	33.73	0	1	0	0	0
##	33.75	0	7	0	0	0
##	33.76	0	1	0	0	0
##	33.77	0	1	0	0	0
##	33.79	0	1	0	0	0
##	33.8	0	4	0	0	0
##	33.81	0	2	0	0	0
##	33.82	0	1	0	0	0
##	33.84	0	7	0	0	0
##	33.86	0	1	0	0	0
##	33.87	0	1	0	0	0
##	33.89	0	15	0	0	0
##	33.91	0	29	0	0	0
##	33.93	0	1	0	0	0
##	33.96	0	13	0	0	0
##	33.97	0	2	0	0	0
##	33.99	0	4	0	0	0
##	34.01	0	12	0	0	0
##	34.02	0	7	0	0	0
##	34.03	0	1	0	0	0
##	34.04	0	1	0	0	0
##	34.06	0	5	0	0	0
##	34.08	0	4	0	0	0
##	34.11	0	4	0	0	0
##	34.14	0	2	0	0	0
##	34.15	0	1	0	0	0
##	34.16	0	2	0	0	0
##	34.17	0	6	0	0	0
##	34.18	0	12	0	0	0
##	34.21	0	11	0	0	0
##	34.22	0	4	0	0	0
##	34.27	0	2	0	0	0
##	34.28	0	1	0	0	0
##	34.3	0	5	0	0	0
##	34.31	0	1	0	0	0
##	34.33	0	38	0	0	0
##	34.34	0	3	0	0	0
##	34.36	0	2	0	0	0
##	34.37	0	3	0	0	0
##	34.38	0	1	0	0	0
##	34.39	0	3	0	0	0
##	34.4	0	1	0	0	0
##	34.41	0	2	0	0	0
##	34.44	0	24	0	0	0
##	34.45	0	4	0	0	0

##	34.46	0	23	0	0	0
##	34.48	0	1	0	0	0
##	34.5	0	1	0	0	0
##	34.52	0	1	0	0	0
##	34.53	0	3	0	0	0
##	34.54	0	3	0	0	0
##	34.56	0	1	0	0	0
##	34.57	0	5	0	0	0
##	34.58	0	5	0	0	0
##	34.6	0	3	0	0	0
##	34.61	0	3	0	0	0
##	34.67	0	6	0	0	0
##	34.68	0	1	0	0	0
##	34.69	0	1	0	0	0
##	34.7	0	20	0	0	0
##	34.72	0	4	0	0	0
##	34.74	0	1	0	0	0
##	34.75	0	11	0	0	0
##	34.77	0	1	0	0	0
##	34.78	0	1	0	0	0
##	34.8	0	1	0	0	0
##	34.81	0	2	0	0	0
##	34.85	0	2	0	0	0
##	34.87	0	18	0	0	0
##	34.88	0	1	0	0	0
##	34.9	0	5	0	0	0
##	34.93	0	2	0	0	0
##	34.95	0	14	0	0	0
##	34.96	0	2	0	0	0
##	34.97	0	19	0	0	0
##	34.98	0	1	0	0	0
##	34.99	0	2	0	0	0
##	35	0	2	0	0	0
##	35.02	0	1	0	0	0
##	35.05	0	2	0	0	0
##	35.08	0	1	0	0	0
##	35.11	0	3	0	0	0
##	35.12	0	1	0	0	0
##	35.13	0	1	0	0	0
##	35.15	0	17	0	0	0
##	35.16	0	1	0	0	0
##	35.19	0	4	0	0	0
##	35.24	0	12	0	0	0
##	35.26	0	12	0	0	0
##	35.28	0	4	0	0	0
##	35.29	0	1	0	0	0
##	35.3	0	1	0	0	0
##	35.31	0	5	0	0	0
##	35.33	0	3	0	0	0
##	35.35	0	1	0	0	0
##	35.36	0	3	0	0	0
##	35.4	0	1	0	0	0
##	35.43	0	22	0	0	0
##	35.44	0	16	0	0	0

##	35.48	0	2	0	0	0
##	35.49	0	1	0	0	0
##	35.5	0	1	0	0	0
##	35.51	0	19	0	0	0
##	35.52	0	1	0	0	0
##	35.54	0	2	0	0	0
##	35.55	0	1	0	0	0
##	35.56	0	6	0	0	0
##	35.57	0	1	0	0	0
##	35.58	0	7	0	0	0
##	35.59	0	1	0	0	0
##	35.61	0	2	0	0	0
##	35.62	0	10	0	0	0
##	35.67	0	8	0	0	0
##	35.71	0	3	0	0	0
##	35.73	0	9	0	0	0
##	35.74	0	8	0	0	0
##	35.75	0	1	0	0	0
##	35.78	0	4	0	0	0
##	35.8	0	1	0	0	0
##	35.81	0	1	0	0	0
##	35.82	0	3	0	0	0
##	35.83	0	3	0	0	0
##	35.85	0	1	0	0	0
##	35.87	0	20	0	0	0
##	35.88	0	3	0	0	0
##	35.89	0	1	0	0	0
##	35.9	0	5	0	0	0
##	35.93	0	2	0	0	0
##	35.94	0	11	0	0	0
##	35.95	0	18	0	0	0
##	35.98	0	2	0	0	0
##	35.99	0	1	0	0	0
##	36	0	1	0	0	0
##	36.01	0	1	0	0	0
##	36.02	0	14	0	0	0
##	36.03	0	2	0	0	0
##	36.04	0	3	0	0	0
##	36.05	0	6	0	0	0
##	36.11	0	1	0	0	0
##	36.13	0	1	0	0	0
##	36.15	0	1	0	0	0
##	36.18	0	6	0	0	0
##	36.19	0	2	0	0	0
##	36.2	0	1	0	0	0
##	36.21	0	3	0	0	0
##	36.22	0	1	0	0	0
##	36.25	0	3	0	0	0
##	36.26	0	7	0	0	0
##	36.28	0	7	0	0	0
##	36.29	0	3	0	0	0
##	36.31	0	4	0	0	0
##	36.32	0	6	0	0	0
##	36.33	0	1	0	0	0

##	36.35	0	1	0	0	0
##	36.36	0	4	0	0	0
##	36.38	0	1	0	0	0
##	36.39	0	3	0	0	0
##	36.48	0	1	0	0	0
##	36.49	0	17	0	0	0
##	36.52	0	1	0	0	0
##	36.56	0	2	0	0	0
##	36.58	0	23	0	0	0
##	36.59	0	4	0	0	0
##	36.6	0	2	0	0	0
##	36.61	0	13	0	0	0
##	36.62	0	6	0	0	0
##	36.65	0	1	0	0	0
##	36.68	0	1	0	0	0
##	36.69	0	9	0	0	0
##	36.73	0	1	0	0	0
##	36.78	0	2	0	0	0
##	36.8	0	2	0	0	0
##	36.81	0	4	0	0	0
##	36.82	0	1	0	0	0
##	36.84	0	1	0	0	0
##	36.87	0	7	0	0	0
##	36.88	0	1	0	0	0
##	36.9	0	8	0	0	0
##	36.91	0	2	0	0	0
##	36.92	0	16	0	0	0
##	36.93	0	1	0	0	0
##	36.94	0	5	0	0	0
##	36.95	0	1	0	0	0
##	36.96	0	8	0	0	0
##	37	0	1	0	0	0
##	37.01	0	2	0	0	0
##	37.03	0	1	0	0	0
##	37.08	0	1	0	0	0
##	37.09	0	5	0	0	0
##	37.1	0	10	0	0	0
##	37.11	0	7	0	0	0
##	37.12	0	13	0	0	0
##	37.18	0	1	0	0	0
##	37.2	0	4	0	0	0
##	37.22	0	2	0	0	0
##	37.23	0	3	0	0	0
##	37.25	0	9	0	0	0
##	37.28	0	1	0	0	0
##	37.3	0	6	0	0	0
##	37.31	0	9	0	0	0
##	37.36	0	6	0	0	0
##	37.37	0	6	0	0	0
##	37.38	0	2	0	0	0
##	37.41	0	3	0	0	0
##	37.42	0	2	0	0	0
##	37.44	0	7	0	0	0
##	37.45	0	1	0	0	0

##	37.46	0	1	0	0	0
##	37.49	0	10	0	0	0
##	37.5	0	10	0	0	0
##	37.54	0	1	0	0	0
##	37.57	0	1	0	0	0
##	37.59	0	11	0	0	0
##	37.6	0	2	0	0	0
##	37.61	0	4	0	0	0
##	37.64	0	1	0	0	0
##	37.65	0	1	0	0	0
##	37.66	0	12	0	0	0
##	37.68	0	2	0	0	0
##	37.71	0	1	0	0	0
##	37.73	0	2	0	0	0
##	37.75	0	1	0	0	0
##	37.76	0	19	0	0	0
##	37.77	0	1	0	0	0
##	37.79	0	11	0	0	0
##	37.8	0	2	0	0	0
##	37.83	0	1	0	0	0
##	37.84	0	13	0	0	0
##	37.86	0	1	0	0	0
##	37.87	0	1	0	0	0
##	37.88	0	1	0	0	0
##	37.89	0	6	0	0	0
##	37.91	0	3	0	0	0
##	37.93	0	5	0	0	0
##	37.95	0	3	0	0	0
##	37.96	0	8	0	0	0
##	37.97	0	8	0	0	0
##	37.98	0	14	0	0	0
##	38.01	0	19	0	0	0
##	38.02	0	1	0	0	0
##	38.04	0	2	0	0	0
##	38.06	0	4	0	0	0
##	38.07	0	1	0	0	0
##	38.08	0	2	0	0	0
##	38.09	0	7	0	0	0
##	38.1	0	1	0	0	0
##	38.11	0	1	0	0	0
##	38.14	0	1	0	0	0
##	38.17	0	1	0	0	0
##	38.21	0	3	0	0	0
##	38.22	0	4	0	0	0
##	38.24	0	1	0	0	0
##	38.25	0	1	0	0	0
##	38.26	0	1	0	0	0
##	38.27	0	17	0	0	0
##	38.28	0	4	0	0	0
##	38.31	0	3	0	0	0
##	38.34	0	1	0	0	0
##	38.35	0	3	0	0	0
##	38.37	0	3	0	0	0
##	38.39	0	4	0	0	0

##	38.41	0	4	0	0	0
##	38.42	0	1	0	0	0
##	38.43	0	6	0	0	0
##	38.44	0	2	0	0	0
##	38.45	0	1	0	0	0
##	38.52	0	9	0	0	0
##	38.53	0	1	0	0	0
##	38.58	0	1	0	0	0
##	38.59	0	1	0	0	0
##	38.61	0	1	0	0	0
##	38.62	0	4	0	0	0
##	38.65	0	4	0	0	0
##	38.66	0	1	0	0	0
##	38.67	0	1	0	0	0
##	38.69	0	2	0	0	0
##	38.73	0	2	0	0	0
##	38.74	0	18	0	0	0
##	38.75	0	1	0	0	0
##	38.77	0	2	0	0	0
##	38.79	0	2	0	0	0
##	38.82	0	1	0	0	0
##	38.91	0	1	0	0	0
##	38.92	0	1	0	0	0
##	38.95	0	1	0	0	0
##	38.97	0	14	0	0	0
##	39	0	1	0	0	0
##	39.05	0	5	0	0	0
##	39.06	0	7	0	0	0
##	39.08	0	1	0	0	0
##	39.11	0	1	0	0	0
##	39.13	0	5	0	0	0
##	39.14	0	2	0	0	0
##	39.16	0	16	0	0	0
##	39.2	0	1	0	0	0
##	39.22	0	1	0	0	0
##	39.25	0	1	0	0	0
##	39.27	0	1	0	0	0
##	39.28	0	1	0	0	0
##	39.32	0	5	0	0	0
##	39.33	0	4	0	0	0
##	39.36	0	1	0	0	0
##	39.38	0	1	0	0	0
##	39.45	0	1	0	0	0
##	39.46	0	2	0	0	0
##	39.48	0	9	0	0	0
##	39.49	0	1	0	0	0
##	39.53	0	8	0	0	0
##	39.54	0	1	0	0	0
##	39.58	0	4	0	0	0
##	39.6	0	2	0	0	0
##	39.67	0	1	0	0	0
##	39.68	0	5	0	0	0
##	39.75	0	5	0	0	0
##	39.77	0	2	0	0	0

```
## 39.8      0      4      0      0      0
## 39.84     0      1      0      0      0
## 39.86     0      1      0      0      0
## 39.87     0      8      0      0      0
## 39.89     0      1      0      0      0
## 39.94     0     16      0      0      0
## 40.03     0      1      0      0      0
## 40.09     0      3      0      0      0
## 40.14     0      2      0      0      0
## 40.17     0      2      0      0      0
## 40.18     0      6      0      0      0
## 40.19     0      1      0      0      0
## 40.21     0      1      0      0      0
## 40.24     0      4      0      0      0
## 40.25     0      1      0      0      0
## 40.27     0      1      0      0      0
## 40.29     0      1      0      0      0
## 40.31     0      1      0      0      0
## 40.34     0      4      0      0      0
## 40.35     0     19      0      0      0
## 40.37     0      1      0      0      0
## 40.39     0      1      0      0      0
## 40.41     0      2      0      0      0
## 40.42     0      1      0      0      0
## 40.44     0      1      0      0      0
## 40.45     0      3      0      0      0
## 40.49     0      1      0      0      0
```

```
summary(ibiccs_clean$bmi)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.     NA's
##      17.5   22.1   24.9   25.6   28.2   40.5   2741
```

## Recode Language

```
table(ibiccs_clean$lang)
```

```
##
##  Anglais Espagnol Français
##    22236      44    1621
```

```
## Language
```

```
ibiccs_clean <- ibiccs_clean %>%
  mutate(language = case_when(
    lang == "Anglais" ~ "English",
    lang == "Espagnol" ~ "Fren/Span",
    lang == "Français" ~ "Fren/Span"
  ))
table(ibiccs_clean$lang, ibiccs_clean$language)
```

```
##
##           English Fren/Span
##  Anglais    22236      0
##  Espagnol      0     44
```

```
## Français 0 1621
```

## Recode Gender

```
table(ibiccs_clean$q54)
```

```
##
## Femme Homme
## 14042 9859
```

```
#Gender
```

```
ibiccs_clean <- ibiccs_clean %>%
  mutate(gender = case_when(
    q54 == "Femme" ~ "Female",
    q54 == "Homme" ~ "Male"
  ))
table(ibiccs_clean$q54, ibiccs_clean$gender)
```

```
##
## Female Male
## Femme 14042 0
## Homme 0 9859
```

## Recode Self-Rated Health

```
table(ibiccs_clean$q2)
```

```
##
## Bon Excellent
## 6725 4803
## Mauvais Moyen
## 489 2104
## Ne sais pas/Refuse de répondre Très bon
## 39 9741
```

```
ibiccs_clean <- ibiccs_clean %>%
  mutate(health = case_when(
    q2 == "Excellent" ~ "Excellent",
    q2 == "Très bon" ~ "Very Good",
    q2 == "Bon" ~ "Good",
    q2 == "Moyen" ~ "Poor/Fair",
    q2 == "Mauvais" ~ "Poor/Fair"
  ))
table(ibiccs_clean$q2, ibiccs_clean$health)
```

```
##
## Excellent Good Poor/Fair Very Good
## Bon 0 6725 0 0
## Excellent 4803 0 0 0
## Mauvais 0 0 489 0
## Moyen 0 0 2104 0
## Ne sais pas/Refuse de répondre 0 0 0 0
```



```
##      Très bon                0      0      0      9741
```

## Recode Transportation

```
table(ibiccs_clean$q13)
```

```
##
##      Autre (précisez:)
##      35
##      Marche
##      4484
##      Ne s'applique pas
##      61
##      Ne sais pas
##      36
##      Scooter
##      21
##      Taxi
##      205
##      Transport en commun
##      7886
## Véhicule motorisé (loué, emprunté, covoiturage)
##      9768
##      Vélo en libre-service
##      243
##      Vélo personnel
##      934
##      Voiture personnelle
##      228
```

```
ibiccs_clean <- ibiccs_clean %>%
  mutate(common_transportation = case_when(
    q13 == "Marche" ~ "Walking",
    q13 == "Véhicule motorisé (loué, emprunté, covoiturage)" ~ "Car",
    q13 == "Voiture personnelle" ~ "Car",
    q13 == "Scooter" ~ "Other",
    q13 == "Taxi" ~ "Other",
    q13 == "Autre (précisez:)" ~ "Other",
    q13 == "Transport en commun" ~ "Public Transportation",
    q13 == "Vélo en libre-service" ~ "Bicycle",
    q13 == "Vélo personnel" ~ "Bicycle"
  ))
table(ibiccs_clean$q13, ibiccs_clean$common_transportation)
```

```
##
##      Bicycle  Car  Other
##      Autre (précisez:)      0      0      35
##      Marche                0      0      0
##      Ne s'applique pas      0      0      0
##      Ne sais pas           0      0      0
##      Scooter               0      0      21
##      Taxi                  0      0     205
##      Transport en commun    0      0      0
```

```
## Véhicule motorisé (loué, emprunté, covoiturage)      0 9768      0
## Vélo en libre-service                             243      0      0
## Vélo personnel                                    934      0      0
## Voiture personnelle                                0 228      0
##
##                                     Public Transportation
## Autre (précisez:)                                  0
## Marche                                              0
## Ne s'applique pas                                  0
## Ne sais pas                                         0
## Scooter                                             0
## Taxi                                                0
## Transport en commun                               7886
## Véhicule motorisé (loué, emprunté, covoiturage)      0
## Vélo en libre-service                              0
## Vélo personnel                                     0
## Voiture personnelle                                0
##
##                                     Walking
## Autre (précisez:)                                  0
## Marche                                             4484
## Ne s'applique pas                                  0
## Ne sais pas                                         0
## Scooter                                             0
## Taxi                                                0
## Transport en commun                               0
## Véhicule motorisé (loué, emprunté, covoiturage)      0
## Vélo en libre-service                              0
## Vélo personnel                                     0
## Voiture personnelle                                0
```

## Recode Physically Active (Y/N)

```
table(ibiccs_clean$q14)
```

```
##
## Ne sais pas/Ne s'applique pas      Non
##                                276      4124
##                                Oui      Refuse de répondre
##                                19471      30
```

```
ibiccs_clean <- ibiccs_clean %>%
  mutate(physically_active = case_when(
    q14 == "Oui" ~ "Yes",
    q14 == "Non" ~ "No"
  ))
table(ibiccs_clean$q14, ibiccs_clean$physically_active)
```

```
##
##                                No      Yes
## Ne sais pas/Ne s'applique pas      0      0
## Non                                4124      0
## Oui                                0 19471
```

```
## Refuse de répondre 0 0
```

## Recode Type of Physical Activity

```
# Code into Sedentary, Walking, Moderate, Vigorous Activity
table(ibiccs_clean$q18)
```

```
##
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15
## 266 447 49 97 243 508 976 52 204 78 291 10 53 332 662
## 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30
## 51 31 561 101 181 8 254 102 29 24 13 17 649 5 30
## 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45
## 206 6 55 254 22 229 920 24 17 44 4 9 11 46 1
## 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60
## 12 8 15 16 5 139 58 23 266 3 5 355 225 55 80
## 61 62 63 64 65 66 67 68 69 70 71 96 97 98 99
## 145 30 75 3243 8 1203 20 862 34 52 17 544 3167 594 40
```

```
ibiccs_clean <- ibiccs_clean %>%
  mutate(pa_level = case_when(
    q15 == 1 ~ "Low",
    q15 == 2 ~ "Moderate",
    q15 == 3 ~ "Low",
    q15 == 4 ~ "Moderate",
    q15 == 5 ~ "Vigorous",
    q15 == 6 ~ "Moderate",
    q15 == 7 ~ "Moderate",
    q15 == 8 ~ "Low",
    q15 == 9 ~ "Low",
    q15 == 10 ~ "Vigorous",
    q15 == 11 ~ "Vigorous",
    q15 == 12 ~ "Moderate",
    q15 == 13 ~ "Low",
    q15 == 14 ~ "Vigorous",
    q15 == 15 ~ "Vigorous",
    q15 == 16 ~ "Sedentary",
    q15 == 17 ~ "Low",
    q15 == 18 ~ "Moderate",
    q15 == 19 ~ "Low",
    q15 == 20 ~ "Moderate",
    q15 == 21 ~ "Moderate",
    q15 == 22 ~ "Vigorous",
    q15 == 23 ~ "Vigorous",
    q15 == 24 ~ "Low",
    q15 == 25 ~ "Low",
    q15 == 26 ~ "Low",
    q15 == 27 ~ "Moderate",
    q15 == 28 ~ "Vigorous",
    q15 == 29 ~ "Vigorous",
    q15 == 30 ~ "Vigorous",
    q15 == 31 ~ "Low",
    q15 == 32 ~ "Moderate",
```

```

q15 == 33 ~ "Moderate",
q15 == 34 ~ "Moderate",
q15 == 35 ~ "Moderate",
q15 == 36 ~ "Moderate",
q15 == 37 ~ "Vigorous",
q15 == 38 ~ "Vigorous",
q15 == 39 ~ "Vigorous",
q15 == 40 ~ "Vigorous",
q15 == 41 ~ "Vigorous",
q15 == 42 ~ "Moderate",
q15 == 43 ~ "Moderate",
q15 == 44 ~ "Vigorous",
q15 == 45 ~ "Low",
q15 == 46 ~ "Moderate",
q15 == 47 ~ "Moderate",
q15 == 48 ~ "Vigorous",
q15 == 49 ~ "Moderate",
q15 == 50 ~ "Moderate",
q15 == 51 ~ "Vigorous",
q15 == 52 ~ "Vigorous",
q15 == 53 ~ "Vigorous",
q15 == 54 ~ "Moderate",
q15 == 55 ~ "Low",
q15 == 56 ~ "Vigorous",
q15 == 57 ~ "Moderate",
q15 == 58 ~ "Vigorous",
q15 == 59 ~ "Moderate",
q15 == 60 ~ "Vigorous",
q15 == 61 ~ "Vigorous",
q15 == 62 ~ "Vigorous",
q15 == 63 ~ "Vigorous",
q15 == 64 ~ "Moderate",
q15 == 65 ~ "Moderate",
q15 == 66 ~ "Vigorous",
q15 == 67 ~ "Vigorous",
q15 == 68 ~ "Low",
q15 == 69 ~ "Moderate",
q15 == 96 ~ "NA",
q15 == 97 ~ "Sedentary",
q15 == 98 ~ "NA",
q15 == 99 ~ "NA",
TRUE ~ "NA"
))
table(ibiccs_clean$q15, ibiccs_clean$pa_level)

```

```

##
##      Low Moderate    NA Sedentary Vigorous
##  1    157         0    0         0         0
##  2      0        629    0         0         0
##  3     47         0    0         0         0
##  4      0         81    0         0         0
##  5      0          0    0         0        288
##  6      0        436    0         0         0
##  7      0       1044    0         0         0

```

##	8	26	0	0	0	0
##	9	131	0	0	0	0
##	10	0	0	0	0	74
##	11	0	0	0	0	179
##	12	0	12	0	0	0
##	13	33	0	0	0	0
##	14	0	0	0	0	316
##	15	0	0	0	0	828
##	16	0	0	0	30	0
##	17	27	0	0	0	0
##	18	0	357	0	0	0
##	19	122	0	0	0	0
##	20	0	174	0	0	0
##	21	0	9	0	0	0
##	22	0	0	0	0	156
##	23	0	0	0	0	182
##	24	29	0	0	0	0
##	25	27	0	0	0	0
##	26	8	0	0	0	0
##	27	0	14	0	0	0
##	28	0	0	0	0	852
##	29	0	0	0	0	2
##	30	0	0	0	0	23
##	31	92	0	0	0	0
##	32	0	3	0	0	0
##	33	0	30	0	0	0
##	34	0	188	0	0	0
##	35	0	15	0	0	0
##	36	0	93	0	0	0
##	37	0	0	0	0	1555
##	38	0	0	0	0	25
##	39	0	0	0	0	8
##	40	0	0	0	0	34
##	41	0	0	0	0	15
##	42	0	18	0	0	0
##	43	0	13	0	0	0
##	44	0	0	0	0	56
##	46	0	1	0	0	0
##	47	0	3	0	0	0
##	48	0	0	0	0	11
##	49	0	11	0	0	0
##	50	0	4	0	0	0
##	51	0	0	0	0	167
##	52	0	0	0	0	56
##	53	0	0	0	0	25
##	54	0	142	0	0	0
##	55	8	0	0	0	0
##	56	0	0	0	0	10
##	57	0	300	0	0	0
##	58	0	0	0	0	199
##	59	0	28	0	0	0
##	60	0	0	0	0	60
##	61	0	0	0	0	179
##	62	0	0	0	0	28

```
##    63    0      0    0      0      98
##    64    0    7090    0      0      0
##    65    0      2    0      0      0
##    66    0      0    0      0    1165
##    67    0      0    0      0     13
##    68  840      0    0      0      0
##    69    0     74    0      0      0
##    70    0      0   65      0      0
##    71    0      0   21      0      0
##    96    0      0  362      0      0
##    98    0      0   54      0      0
##    99    0      0   17      0      0
```

## Recode Days Per Week Spent Travelling via Car

```
table(ibiccs_clean$q21)

##
##    0    1    2    3    4    5    6    7   98   99
## 2785 2527 2631 2070 1648 3136 2019 4925 2037 123

ibiccs_clean <- ibiccs_clean %>%
  mutate(day_per_week_motor_vehicle = case_when(
    q21 == 0 ~ 0,
    q21 == 1 ~ 1,
    q21 == 2 ~ 2,
    q21 == 3 ~ 3,
    q21 == 4 ~ 4,
    q21 == 5 ~ 5,
    q21 == 6 ~ 6,
    q21 == 7 ~ 7,
    q21 == 98 ~ NA_real_,
    q21 == 99 ~ NA_real_
  ))
table(ibiccs_clean$day_per_week_motor_vehicle)

##
##    0    1    2    3    4    5    6    7
## 2785 2527 2631 2070 1648 3136 2019 4925
```

## Recode Days Per Week Spent Travelling via Public Transport

```
table(ibiccs_clean$q23)

##
##    0    1    2    3    4    5    6    7    8    9
## 6169 2993 2167 1598 1339 3617 1482 1583 2726 227

ibiccs_clean <- ibiccs_clean %>%
  mutate(day_per_week_public_transit = case_when(
    q23 == 0 ~ 0,
```

```

q23 == 1 ~ 1,
q23 == 2 ~ 2,
q23 == 3 ~ 3,
q23 == 4 ~ 4,
q23 == 5 ~ 5,
q23 == 6 ~ 6,
q23 == 7 ~ 7,
q23 == 8 ~ NA_real_,
q23 == 9 ~ NA_real_
))

table(ibiccs_clean$q23, ibiccs_clean$day_per_week_public_transit)

```

```

##
##      0      1      2      3      4      5      6      7
## 0 6169      0      0      0      0      0      0      0
## 1      0 2993      0      0      0      0      0      0
## 2      0      0 2167      0      0      0      0      0
## 3      0      0      0 1598      0      0      0      0
## 4      0      0      0      0 1339      0      0      0
## 5      0      0      0      0      0 3617      0      0
## 6      0      0      0      0      0      0 1482      0
## 7      0      0      0      0      0      0      0 1583
## 8      0      0      0      0      0      0      0      0
## 9      0      0      0      0      0      0      0      0

```

## Recode Days Per Week Spent Travelling via Walking

```

table(ibiccs_clean$q25)

##
##      0      1      2      3      4      5      6      7      8      9
## 2115 1841 2361 2245 1831 3994 1711 5436 2149 218

ibiccs_clean <- ibiccs_clean %>%
  mutate(day_per_week_walking = case_when(
    q25 == 0 ~ 0,
    q25 == 1 ~ 1,
    q25 == 2 ~ 2,
    q25 == 3 ~ 3,
    q25 == 4 ~ 4,
    q25 == 5 ~ 5,
    q25 == 6 ~ 6,
    q25 == 7 ~ 7,
    q25 == 8 ~ NA_real_,
    q25 == 9 ~ NA_real_
  ))

table(ibiccs_clean$q25, ibiccs_clean$day_per_week_walking)

##
##      0      1      2      3      4      5      6      7
## 0 2115      0      0      0      0      0      0      0

```

```
## 1 0 1841 0 0 0 0 0 0
## 2 0 0 2361 0 0 0 0 0
## 3 0 0 0 2245 0 0 0 0
## 4 0 0 0 0 1831 0 0 0
## 5 0 0 0 0 0 3994 0 0
## 6 0 0 0 0 0 0 1711 0
## 7 0 0 0 0 0 0 0 5436
## 8 0 0 0 0 0 0 0 0
## 9 0 0 0 0 0 0 0 0
```

## Recode Days Per Week Spent Travelling via Bike

```
table(ibiccs_clean$q27)
```

```
##
## 0 1 2 3 4 5 6 7 98 99
## 12010 1793 1185 741 462 632 215 294 6395 174
```

```
ibiccs_clean <- ibiccs_clean %>%
  mutate(day_per_week_bike = case_when(
    q27 == 0 ~ 0,
    q27 == 1 ~ 1,
    q27 == 2 ~ 2,
    q27 == 3 ~ 3,
    q27 == 4 ~ 4,
    q27 == 5 ~ 5,
    q27 == 6 ~ 6,
    q27 == 7 ~ 7,
    q27 == 98 ~ NA_real_,
    q27 == 99 ~ NA_real_
  ))
```

```
table(ibiccs_clean$q27, ibiccs_clean$day_per_week_bike)
```

```
##
## 0 1 2 3 4 5 6 7
## 0 12010 0 0 0 0 0 0
## 1 0 1793 0 0 0 0 0
## 2 0 0 1185 0 0 0 0
## 3 0 0 0 741 0 0 0
## 4 0 0 0 0 462 0 0
## 5 0 0 0 0 0 632 0
## 6 0 0 0 0 0 0 215
## 7 0 0 0 0 0 0 0 294
## 98 0 0 0 0 0 0 0
## 99 0 0 0 0 0 0 0
```

## Recode Age

```
summary(ibiccs_clean$q42)
```



```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      18.0    31.0    41.0    42.6    53.0    94.0
```

*#CCHS/STATCAN or continuous*

## Recode Marital Status

```
table(ibiccs_clean$q44)
```

```
##
##          Célibataire          Divorcé (e)
##          9452          1854
##          En couple Marié(e)/Conjoint de fait
##          79          11287
##          Refuse de répondre          Séparé (e)
##          291          460
##          Veuf (ve)
##          478
```

```
ibiccs_clean <- ibiccs_clean %>%
  mutate(marital_status = case_when(
    q44 == "Célibataire" ~ "Single",
    q44 == "Divorcé(e)" ~ "Divorced/Separated/Widowed",
    q44 == "En couple" ~ "Relationship/Married/Common-Law",
    q44 == "Marié(e)/Conjoint de fait" ~ "Relationship/Married/Common-Law",
    q44 == "Séparé(e)" ~ "Divorced/Separated/Widowed",
    q44 == "Veuf(ve)" ~ "Divorced/Separated/Widowed"
  ))
table(ibiccs_clean$q44, ibiccs_clean$marital_status)
```

```
##
##          Relationship/Married/Common-Law Single
##  Célibataire          0  9452
##  Divorcé (e)          0    0
##  En couple          79    0
##  Marié(e)/Conjoint de fait 11287  0
##  Refuse de répondre          0    0
##  Séparé (e)          0    0
##  Veuf (ve)          0    0
```

## Recode Number of Children in Home

```
table(ibiccs_clean$q45)
```

```
##
##      1      2      3      4      5      6      7      8      9
## 2915 1863  471  119   40   18    9 18191  275
```

```
ibiccs_clean <- ibiccs_clean %>%
  mutate(children_household = case_when(
    q45 == 1 ~ 1,
    q45 == 2 ~ 2,
```

```

q45 == 3 ~ 3,
q45 == 4 ~ 3,
q45 == 5 ~ 3,
q45 == 6 ~ 3,
q45 == 7 ~ 3,
q45 == 8 ~ 0,
q45 == 9 ~ NA_real_
))
table(ibiccs_clean$q45, ibiccs_clean$children_household)

```

```

##
##      0      1      2      3
##  1      0 2915      0      0
##  2      0      0 1863      0
##  3      0      0      0 471
##  4      0      0      0 119
##  5      0      0      0  40
##  6      0      0      0  18
##  7      0      0      0   9
##  8 18191      0      0      0
##  9      0      0      0      0

```

## Recode Months in Current Home?

```
summary(ibiccs_clean$q46b)
```

```

##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.   NA's
##         0         3         5         5         6         81  15358

```

## Recode Ethnicity

```
table(ibiccs_clean$q47)
```

```

##
## Amérindien des États-Unis / Autochtone d'Amérique
##                                     79
##           Arabe (Moyen-Orient, Afrique du Nord)
##                                     135
##           Asiatique / insulaire du Pacifique
##                                     2661
##                                     Autre
##                                     129
##           Blanc(che) / Caucasien
##                                     17154
##           Hispanique / Latino / Espagnol(e)
##                                     933
##           Indien / Pakistanais
##                                     330
##           Je préfère ne pas répondre
##                                     639

```

```
##                                Jewish
##                                16
##                                Mixed / Mixed race / Bi-racial
##                                171
##                                Noir(e) / Africain(e) / Afro-Américain(e)
##                                1654

ibiccs_clean <- ibiccs_clean %>%
  mutate(ethnicity = case_when(
    q47 == "Amérindien des États-Unis / Autochtone d'Amérique" ~ "Native American/Indigenous",
    q47 == "Arabe (Moyen-Orient, Afrique du Nord)" ~ "Other",
    q47 == "Asiatique / insulaire du Pacifique" ~ "Asian",
    q47 == "Autre" ~ "Other",
    q47 == "Blanc(che) / Caucasien" ~ "Caucasian",
    q47 == "Hispanique / Latino / Espagnol(e)" ~ "Hispanic",
    q47 == "Indien / Pakistanais" ~ "Other",
    q47 == "Jewish" ~ "Other",
    q47 == "Mixed / Mixed race / Bi-racial" ~ "Other",
    q47 == "Noir(e) / Africain(e) / Afro-Américain(e)" ~ "African American/Canadian"
  ))
table(ibiccs_clean$q47, ibiccs_clean$ethnicity)

##
##                                African American/Canadian
##  Amérindien des États-Unis / Autochtone d'Amérique      0
##  Arabe (Moyen-Orient, Afrique du Nord)                  0
##  Asiatique / insulaire du Pacifique                      0
##  Autre                                                    0
##  Blanc(che) / Caucasien                                  0
##  Hispanique / Latino / Espagnol(e)                      0
##  Indien / Pakistanais                                    0
##  Je préfère ne pas répondre                              0
##  Jewish                                                  0
##  Mixed / Mixed race / Bi-racial                          0
##  Noir(e) / Africain(e) / Afro-Américain(e)              1654
##
##                                Asian Caucasian
##  Amérindien des États-Unis / Autochtone d'Amérique      0      0
##  Arabe (Moyen-Orient, Afrique du Nord)                  0      0
##  Asiatique / insulaire du Pacifique                     2661     0
##  Autre                                                    0      0
##  Blanc(che) / Caucasien                                  0    17154
##  Hispanique / Latino / Espagnol(e)                      0      0
##  Indien / Pakistanais                                    0      0
##  Je préfère ne pas répondre                              0      0
##  Jewish                                                  0      0
##  Mixed / Mixed race / Bi-racial                          0      0
##  Noir(e) / Africain(e) / Afro-Américain(e)              0      0
##
##                                Hispanic
##  Amérindien des États-Unis / Autochtone d'Amérique      0
##  Arabe (Moyen-Orient, Afrique du Nord)                  0
##  Asiatique / insulaire du Pacifique                      0
##  Autre                                                    0
##  Blanc(che) / Caucasien                                  0
```

```
## Hispanique / Latino / Espagnol(e) 933
## Indien / Pakistanais 0
## Je préfère ne pas répondre 0
## Jewish 0
## Mixed / Mixed race / Bi-racial 0
## Noir(e) / Africain(e) / Afro-Américain(e) 0
##
## Native American/Indigenous
## Amérindien des États-Unis / Autochtone d'Amérique 79
## Arabe (Moyen-Orient, Afrique du Nord) 0
## Asiatique / insulaire du Pacifique 0
## Autre 0
## Blanc(che) / Caucasien 0
## Hispanique / Latino / Espagnol(e) 0
## Indien / Pakistanais 0
## Je préfère ne pas répondre 0
## Jewish 0
## Mixed / Mixed race / Bi-racial 0
## Noir(e) / Africain(e) / Afro-Américain(e) 0
##
## Other
## Amérindien des États-Unis / Autochtone d'Amérique 0
## Arabe (Moyen-Orient, Afrique du Nord) 135
## Asiatique / insulaire du Pacifique 0
## Autre 129
## Blanc(che) / Caucasien 0
## Hispanique / Latino / Espagnol(e) 0
## Indien / Pakistanais 330
## Je préfère ne pas répondre 0
## Jewish 16
## Mixed / Mixed race / Bi-racial 171
## Noir(e) / Africain(e) / Afro-Américain(e) 0
```

## Recode Country Born

```
table(ibiccs_clean$q48)
```

```
##
##      1      2      3      4      5      6      7      8      9     10     11     12
## 7171 13175 246    11    11    20    29    49    20    28    14    17
##    13    14    15    16    17    18    19    20    21    22    23    24
##    15    12   109   153   104   231   166   39     7    22    77    43
##    25    26    27    28    29    30    31    32    33    34    35    36
##    73    28    21    57    27    18    93    63    44    48    66    33
##    37    38    39    40    41    42    43    44    45    46    47    48
##    33    23    25    24   159    51    27    43     8    15     4    40
##    49    50    51    52    53    54    55    56    57    58    59    60
##     5    26    14    11    20    11    13    34     3    18    13    13
##    61    62    63    64    65    66    67    68    69    70    71    72
##    12     9    33    24    13    15    20     9     6    13    10     3
##    73    74    75    76    77    78    79    80    81    96    99
##    16     8     4     5    19    12     6     9    13   196   473
```

```

ibiccs_clean <- ibiccs_clean %>%
  mutate(country_born = case_when(
    q48 == 1 ~ "Canada",
    q48 == 2 ~ "United States",
    q48 >= 3 ~ "Other",
    q48 <= 81 ~ "Other",
    q48 == 96 ~ "Other"
  ))
table(ibiccs_clean$q48, ibiccs_clean$country_born)

```

```

##
##      Canada Other United States
##  1      7171      0              0
##  2         0      0          13175
##  3         0    246              0
##  4         0     11              0
##  5         0     11              0
##  6         0     20              0
##  7         0     29              0
##  8         0     49              0
##  9         0     20              0
## 10         0     28              0
## 11         0     14              0
## 12         0     17              0
## 13         0     15              0
## 14         0     12              0
## 15         0    109              0
## 16         0    153              0
## 17         0    104              0
## 18         0    231              0
## 19         0    166              0
## 20         0     39              0
## 21         0      7              0
## 22         0     22              0
## 23         0     77              0
## 24         0     43              0
## 25         0     73              0
## 26         0     28              0
## 27         0     21              0
## 28         0     57              0
## 29         0     27              0
## 30         0     18              0
## 31         0     93              0
## 32         0     63              0
## 33         0     44              0
## 34         0     48              0
## 35         0     66              0
## 36         0     33              0
## 37         0     33              0
## 38         0     23              0
## 39         0     25              0
## 40         0     24              0
## 41         0    159              0
## 42         0     51              0

```

##	43	0	27	0
##	44	0	43	0
##	45	0	8	0
##	46	0	15	0
##	47	0	4	0
##	48	0	40	0
##	49	0	5	0
##	50	0	26	0
##	51	0	14	0
##	52	0	11	0
##	53	0	20	0
##	54	0	11	0
##	55	0	13	0
##	56	0	34	0
##	57	0	3	0
##	58	0	18	0
##	59	0	13	0
##	60	0	13	0
##	61	0	12	0
##	62	0	9	0
##	63	0	33	0
##	64	0	24	0
##	65	0	13	0
##	66	0	15	0
##	67	0	20	0
##	68	0	9	0
##	69	0	6	0
##	70	0	13	0
##	71	0	10	0
##	72	0	3	0
##	73	0	16	0
##	74	0	8	0
##	75	0	4	0
##	76	0	5	0
##	77	0	19	0
##	78	0	12	0
##	79	0	6	0
##	80	0	9	0
##	81	0	13	0
##	96	0	196	0
##	99	0	473	0

## Recode Motor Vehicle Access

```
table(ibiccs_clean$q50)
```

##		
##	Ne sais pas/Ne s'applique pas	Non
##	230	6295
##	Oui	
##	17376	

```
ibiccs_clean <- ibiccs_clean %>%
  mutate(motor_vehicle_access = case_when(
    q50 == "Non" ~ "No",
    q50 == "Oui" ~ "Yes"
  ))
table(ibiccs_clean$q50, ibiccs_clean$motor_vehicle_access)
```

```
##
##
##      No      Yes
## Ne sais pas/Ne s'applique pas    0      0
## Non                             6295     0
## Oui                              0 17376
```

## Recode Education Level

```
table(ibiccs_clean$q51)
```

```
##
##      Aucun grade, certificat ou diplôme
##                                     32
##      Autre (précisez):
##                                     81
##      Baccalauréat
##                                     8638
##      Cégep
##                                     3197
## Certificat d'école de métier, certificat ou diplôme
##                                     2559
##      Diplôme d'études secondaire ou l'équivalent
##                                     2471
##      Diplôme universitaire supérieur au baccalauréat
##                                     6655
##      École primaire
##                                     88
##      Refuse de répondre
##                                     180
```

```
ibiccs_clean <- ibiccs_clean %>%
  mutate(education = case_when(
    q51 == "Aucun grade, certificat ou diplôme" ~ "High School/Lower",
    q51 == "Baccalauréat" ~ "Baccalaureate",
    q51 == "Cégep" ~ "Cégep",
    q51 == "Certificat d'école de métier, certificat ou diplôme" ~ "Certificate/Diploma",
    q51 == "Diplôme d'études secondaire ou l'équivalent" ~ "High School/Lower",
    q51 == "Diplôme universitaire supérieur au baccalauréat" ~ "Graduate School",
    q51 == "École primaire" ~ "High School/Lower"
  ))
table(ibiccs_clean$q51, ibiccs_clean$education)
```

```
##
##
##      Baccalaureate Cégep
##      Aucun grade, certificat ou diplôme    0      0
##      Autre (précisez):                    0      0
```

##	Baccalauréat	8638	0
##	Cégep	0	3197
##	Certificat d'école de métier, certificat ou diplôme	0	0
##	Diplôme d'études secondaire ou l'équivalent	0	0
##	Diplôme universitaire supérieur au baccalauréat	0	0
##	École primaire	0	0
##	Refuse de répondre	0	0
##			
##		Certificate/Diploma	
##	Aucun grade, certificat ou diplôme		0
##	Autre (précisez):		0
##	Baccalauréat		0
##	Cégep		0
##	Certificat d'école de métier, certificat ou diplôme		2559
##	Diplôme d'études secondaire ou l'équivalent		0
##	Diplôme universitaire supérieur au baccalauréat		0
##	École primaire		0
##	Refuse de répondre		0
##			
##		Graduate School	
##	Aucun grade, certificat ou diplôme		0
##	Autre (précisez):		0
##	Baccalauréat		0
##	Cégep		0
##	Certificat d'école de métier, certificat ou diplôme		0
##	Diplôme d'études secondaire ou l'équivalent		0
##	Diplôme universitaire supérieur au baccalauréat		6655
##	École primaire		0
##	Refuse de répondre		0
##			
##		High School/Lower	
##	Aucun grade, certificat ou diplôme		32
##	Autre (précisez):		0
##	Baccalauréat		0
##	Cégep		0
##	Certificat d'école de métier, certificat ou diplôme		0
##	Diplôme d'études secondaire ou l'équivalent		2471
##	Diplôme universitaire supérieur au baccalauréat		0
##	École primaire		88
##	Refuse de répondre		0

## Recode Occupation Status

```
table(ibiccs_clean$Q52_occupational_status_category)
```

##		
##	Disability	Full time or Self-employed
##	453	15373
##	Homemaker or parental leave	Part time
##	983	1899
##	Retired	Student
##	2434	1586



```
##      Unemployed seeking work
##                               977
```

```
ibiccs_clean <- ibiccs_clean %>%
  mutate(occupation_status = case_when(
    Q52_occupational_status_category == "Disability" ~ "Unemployed",
    Q52_occupational_status_category == "Full time or Self-employed" ~ "Employed",
    Q52_occupational_status_category == "Homemaker or parental leave" ~ "Unemployed",
    Q52_occupational_status_category == "Part time" ~ "Employed",
    Q52_occupational_status_category == "Retired" ~ "Unemployed",
    Q52_occupational_status_category == "Student" ~ "Student",
    Q52_occupational_status_category == "Unemployed seeking work" ~ "Unemployed"
  ))
table(ibiccs_clean$Q52_occupational_status_category, ibiccs_clean$occupation_status)
```

```
##
##      Employed Student Unemployed
## Disability              0         0        453
## Full time or Self-employed 15373      0         0
## Homemaker or parental leave    0         0        983
## Part time                  1899      0         0
## Retired                    0         0       2434
## Student                    0       1586         0
## Unemployed seeking work      0         0        977
```

## Recode Household Income

```
table(ibiccs_clean$q53)
```

```
##
##      Entre 10000 $ and 19999 $ par année
##                               1084
##      Entre 100000 $ and 149999 $ par année
##                               3490
##      Entre 150000 $ and 199999 $ par année
##                               1296
##      Entre 20000 $ and 34999 $ par année
##                               2061
##      Entre 35000 $ and 49999 $ par année
##                               2700
##      Entre 50000 $ and 74999 $ par année
##                               4396
##      Entre 75000 $ and 99999 $ par année
##                               3425
##      Moins de 10000 $ par année
##                               1366
##      Plus de 200000 $ par année
##                               1104
##      Refuse de répondre
##                               2979
```

```
ibiccs_clean <- ibiccs_clean %>%
  mutate(household_income = case_when(
```

```

q53 == "Moins de 10000 $ par année" ~ "$0-$49999",
q53 == "Entre 10000 $ and 19999 $ par année" ~ "$0-$49999",
q53 == "Entre 20000 $ and 34999 $ par année" ~ "$0-$49999",
q53 == "Entre 35000 $ and 49999 $ par année" ~ "$0-$49999",
q53 == "Entre 50000 $ and 74999 $ par année" ~ "$50000-$99999",
q53 == "Entre 75000 $ and 99999 $ par année" ~ "$50000-$99999",
q53 == "Entre 100000 $ and 149999 $ par année" ~ "$100000-$149999",
q53 == "Entre 150000 $ and 199999 $ par année" ~ "$150000+",
q53 == "Plus de 200000 $ par année" ~ "$150000+",
q53 == "Refuse de répondre" ~ "Missing"
))
table(ibiccs_clean$q53, ibiccs_clean$household_income)

```

```

##
##                               $0-$49999 $100000-$149999 $150000+
##  Entre 10000 $ and 19999 $ par année      1084           0           0
##  Entre 100000 $ and 149999 $ par année       0          3490           0
##  Entre 150000 $ and 199999 $ par année       0           0          1296
##  Entre 20000 $ and 34999 $ par année      2061           0           0
##  Entre 35000 $ and 49999 $ par année      2700           0           0
##  Entre 50000 $ and 74999 $ par année        0           0           0
##  Entre 75000 $ and 99999 $ par année        0           0           0
##  Moins de 10000 $ par année      1366           0           0
##  Plus de 200000 $ par année        0           0          1104
##  Refuse de répondre              0           0           0
##
##                               $50000-$99999 Missing
##  Entre 10000 $ and 19999 $ par année        0           0
##  Entre 100000 $ and 149999 $ par année       0           0
##  Entre 150000 $ and 199999 $ par année       0           0
##  Entre 20000 $ and 34999 $ par année        0           0
##  Entre 35000 $ and 49999 $ par année        0           0
##  Entre 50000 $ and 74999 $ par année      4396           0
##  Entre 75000 $ and 99999 $ par année      3425           0
##  Moins de 10000 $ par année        0           0
##  Plus de 200000 $ par année        0           0
##  Refuse de répondre              0          2979

```

## Filtering Out Cities

```

Boston <- filter(ibiccs_clean, ville == "Boston")
Chicago <- filter(ibiccs_clean, ville == "Chicago")
Detroit <- filter(ibiccs_clean, ville == "Détroit")
NewYork <- filter(ibiccs_clean, ville == "New-York")
Philadelphia <- filter(ibiccs_clean, ville == "Philadelphie")
Montreal <- filter(ibiccs_clean, ville == "Montréal")
Toronto <- filter(ibiccs_clean, ville == "Toronto")
Vancouver <- filter(ibiccs_clean, ville == "Vancouver")

```

## Filtering Out Variables in Clean Data & Filtering Cities

```
vars_clean <- c('language', 'ville', 'gender', 'health', 'common_transportation', 'physically_active',

ibiccs <- dplyr::select(ibiccs_clean, vars_clean)
Boston <- dplyr::select(Boston, vars_clean)
Chicago <- dplyr::select(Chicago, vars_clean)
Detroit <- dplyr::select(Detroit, vars_clean)
NewYork <- dplyr::select(NewYork, vars_clean)
Philadelphia <- dplyr::select(Philadelphia, vars_clean)
Montreal <- dplyr::select(Montreal, vars_clean)
Toronto <- dplyr::select(Toronto, vars_clean)
Vancouver <- dplyr::select(Vancouver, vars_clean)
city <- rbind(Philadelphia, NewYork, Detroit, Chicago, Boston, Montreal, Toronto, Vancouver)
```

#Tables for Each City and Combined

```
vars <- c('language', 'ville', 'gender', 'health', 'common_transportation', 'physically_active', 'pa_level')
CreateTableOne(vars = vars, strata = "ville", data = ibiccs)
```

```
##                               Stratified by ville
##                               Boston           Chicago
##  n                               1977           4085
##  language = Fren/Span (%)           3 ( 0.2)        20 ( 0.5)
##  ville (%)
##    Boston           1977 (100.0)           0 ( 0.0)
##    Chicago           0 ( 0.0)        4085 (100.0)
##    Détroit           0 ( 0.0)           0 ( 0.0)
##    Montréal           0 ( 0.0)           0 ( 0.0)
##    New-York           0 ( 0.0)           0 ( 0.0)
##    Philadelphie           0 ( 0.0)           0 ( 0.0)
##    Toronto           0 ( 0.0)           0 ( 0.0)
##    Vancouver           0 ( 0.0)           0 ( 0.0)
##  gender = Male (%)           694 ( 35.1)        1628 ( 39.9)
##  health (%)
##    Excellent           467 ( 23.7)           832 ( 20.4)
##    Good           521 ( 26.4)        1105 ( 27.1)
##    Poor/Fair           166 (  8.4)           412 ( 10.1)
##    Very Good           818 ( 41.5)        1727 ( 42.4)
##  common_transportation (%)
##    Bicycle           104 (  5.3)           170 (  4.2)
##    Car           685 ( 34.7)        1803 ( 44.3)
##    Other           16 (  0.8)           48 (  1.2)
##    Public Transportation           669 ( 33.9)        1473 ( 36.2)
##    Walking           500 ( 25.3)           576 ( 14.2)
##  physically_active = Yes (%)           1703 ( 87.0)        3418 ( 84.5)
##  pa_level (%)
##    Low           132 (  6.7)           285 (  7.0)
##    Moderate           837 ( 42.3)        1712 ( 41.9)
##    NA           317 ( 16.0)           743 ( 18.2)
##    Sedentary           2 (  0.1)           4 (  0.1)
##    Vigorous           689 ( 34.9)        1341 ( 32.8)
##  day_per_week_motor_vehicle (mean (SD))           3.68 (2.47)           4.08 (2.35)
##  day_per_week_public_transit (mean (SD))           2.95 (2.37)           2.77 (2.36)
```

##	day_per_week_walking (mean (SD))	4.62 (2.25)	4.10 (2.30)
##	day_per_week_bike (mean (SD))	0.87 (1.74)	0.84 (1.58)
##	q42 (mean (SD))	37.21 (13.86)	39.59 (12.73)
##	marital_status = Single (%)	1025 ( 57.8)	1844 ( 50.5)
##	children_household (mean (SD))	0.25 (0.62)	0.37 (0.77)
##	ethnicity (%)		
##	African American/Canadian	110 ( 5.7)	449 ( 11.2)
##	Asian	218 ( 11.3)	313 ( 7.8)
##	Caucasian	1480 ( 76.6)	2863 ( 71.5)
##	Hispanic	67 ( 3.5)	300 ( 7.5)
##	Native American/Indigenous	7 ( 0.4)	8 ( 0.2)
##	Other	49 ( 2.5)	72 ( 1.8)
##	country_born (%)		
##	Canada	21 ( 1.1)	31 ( 0.8)
##	Other	220 ( 11.1)	342 ( 8.4)
##	United States	1736 ( 87.8)	3712 ( 90.9)
##	motor_vehicle_access = Yes (%)	1442 ( 73.5)	3266 ( 80.5)
##	education (%)		
##	Baccalaureate	757 ( 38.6)	1624 ( 40.1)
##	Cégep	231 ( 11.8)	495 ( 12.2)
##	Certificate/Diploma	65 ( 3.3)	293 ( 7.2)
##	Graduate School	774 ( 39.5)	1320 ( 32.6)
##	High School/Lower	132 ( 6.7)	317 ( 7.8)
##	occupation_status (%)		
##	Employed	1442 ( 73.5)	3160 ( 78.0)
##	Student	272 ( 13.9)	271 ( 6.7)
##	Unemployed	249 ( 12.7)	622 ( 15.3)
##	household_income (%)		
##	\$0-\$49999	599 ( 30.3)	1248 ( 30.6)
##	\$100000-\$149999	304 ( 15.4)	617 ( 15.1)
##	\$150000+	232 ( 11.7)	438 ( 10.7)
##	\$50000-\$99999	617 ( 31.2)	1403 ( 34.3)
##	Missing	225 ( 11.4)	379 ( 9.3)
##	bmi (mean (SD))	25.06 (4.59)	25.89 (4.86)
##	bmi_category (%)		
##	normal weight	982 ( 49.7)	1745 ( 42.7)
##	obese	248 ( 12.5)	698 ( 17.1)
##	other	192 ( 9.7)	438 ( 10.7)
##	overweight	511 ( 25.8)	1139 ( 27.9)
##	underweight	44 ( 2.2)	65 ( 1.6)
##	WalkScore (mean (SD))	86.19 (17.12)	80.36 (17.45)
##	WalkScoreLabel (%)		
##	Car-Dependent	38 ( 1.9)	51 ( 1.2)
##	Somewhat Walkable	96 ( 4.9)	274 ( 6.7)
##	Very Walkable	85 ( 4.3)	510 ( 12.5)
##	Walker's Paradise	598 ( 30.2)	1680 ( 41.1)
##	TransitScore (mean (SD))	1160 ( 58.7)	1570 ( 38.4)
##	TransitScoreLabel (%)		
##	Excellent Transit	1694 ( 85.7)	3713 ( 90.9)
##	Good Transit	201 ( 10.2)	277 ( 6.8)
##	Minimal Transit	64 ( 3.2)	82 ( 2.0)
##	Rider's Paradise	0 ( 0.0)	0 ( 0.0)
##		16 ( 0.8)	13 ( 0.3)

##	Some Transit	2 ( 0.1)	0 ( 0.0)
##	BikeScore (mean (SD))	106.50 (32.02)	87.29 (29.96)
##	BikeScoreLabel (%)		
##		598 ( 30.2)	558 ( 13.7)
##	Bikeable	475 ( 24.0)	1113 ( 27.2)
##	Biker's Paradise	323 ( 16.3)	0 ( 0.0)
##	Somewhat Bikeable	26 ( 1.3)	264 ( 6.5)
##	Very Bikeable	555 ( 28.1)	2150 ( 52.6)
##	DiningandDrinkingScore (mean (SD))	86.12 (17.23)	82.41 (18.59)
##	GroceryScore (mean (SD))	88.94 (21.26)	78.71 (28.21)
##		Stratified by ville	
##		Détroit	Montréal
##	n	3077	2678
##	language = Fren/Span (%)	3 ( 0.1)	1617 ( 60.4)
##	ville (%)		
##	Boston	0 ( 0.0)	0 ( 0.0)
##	Chicago	0 ( 0.0)	0 ( 0.0)
##	Détroit	3077 (100.0)	0 ( 0.0)
##	Montréal	0 ( 0.0)	2678 (100.0)
##	New-York	0 ( 0.0)	0 ( 0.0)
##	Philadelphie	0 ( 0.0)	0 ( 0.0)
##	Toronto	0 ( 0.0)	0 ( 0.0)
##	Vancouver	0 ( 0.0)	0 ( 0.0)
##	gender = Male (%)	1134 ( 36.9)	1159 ( 43.3)
##	health (%)		
##	Excellent	566 ( 18.4)	454 ( 17.0)
##	Good	928 ( 30.2)	791 ( 29.6)
##	Poor/Fair	390 ( 12.7)	383 ( 14.3)
##	Very Good	1192 ( 38.8)	1047 ( 39.1)
##	common_transportation (%)		
##	Bicycle	75 ( 2.5)	207 ( 7.8)
##	Car	2718 ( 89.2)	914 ( 34.3)
##	Other	11 ( 0.4)	29 ( 1.1)
##	Public Transportation	105 ( 3.4)	1019 ( 38.3)
##	Walking	139 ( 4.6)	492 ( 18.5)
##	physically_active = Yes (%)	2355 ( 77.6)	2107 ( 80.0)
##	pa_level (%)		
##	Low	242 ( 7.9)	160 ( 6.0)
##	Moderate	1246 ( 40.5)	1335 ( 49.9)
##	NA	776 ( 25.2)	647 ( 24.2)
##	Sedentary	11 ( 0.4)	3 ( 0.1)
##	Vigorous	802 ( 26.1)	533 ( 19.9)
##	day_per_week_motor_vehicle (mean (SD))	5.75 (1.74)	3.24 (2.47)
##	day_per_week_public_transit (mean (SD))	0.58 (1.55)	2.63 (2.39)
##	day_per_week_walking (mean (SD))	2.32 (2.37)	3.96 (2.32)
##	day_per_week_bike (mean (SD))	0.65 (1.43)	0.89 (1.67)
##	q42 (mean (SD))	41.11 (12.74)	47.02 (14.17)
##	marital_status = Single (%)	988 ( 36.6)	906 ( 40.5)
##	children_household (mean (SD))	0.60 (0.92)	0.31 (0.68)
##	ethnicity (%)		
##	African American/Canadian	361 ( 11.9)	52 ( 2.0)
##	Asian	123 ( 4.1)	104 ( 4.0)
##	Caucasian	2422 ( 80.0)	2301 ( 88.1)
##	Hispanic	44 ( 1.5)	50 ( 1.9)

##	Native American/Indigenous	15 ( 0.5)	8 ( 0.3)
##	Other	61 ( 2.0)	97 ( 3.7)
##	country_born (%)		
##	Canada	38 ( 1.2)	2144 ( 80.1)
##	Other	172 ( 5.6)	500 ( 18.7)
##	United States	2867 ( 93.2)	34 ( 1.3)
##	motor_vehicle_access = Yes (%)	2936 ( 96.0)	1821 ( 68.5)
##	education (%)		
##	Baccalaureate	951 ( 31.2)	788 ( 29.8)
##	Cégep	506 ( 16.6)	528 ( 19.9)
##	Certificate/Diploma	444 ( 14.6)	330 ( 12.5)
##	Graduate School	736 ( 24.1)	581 ( 21.9)
##	High School/Lower	411 ( 13.5)	421 ( 15.9)
##	occupation_status (%)		
##	Employed	2152 ( 70.4)	1750 ( 65.8)
##	Student	243 ( 7.9)	141 ( 5.3)
##	Unemployed	663 ( 21.7)	769 ( 28.9)
##	household_income (%)		
##	\$0-\$49999	1078 ( 35.0)	1017 ( 38.0)
##	\$100000-\$149999	443 ( 14.4)	271 ( 10.1)
##	\$150000+	193 ( 6.3)	148 ( 5.5)
##	\$50000-\$99999	997 ( 32.4)	888 ( 33.2)
##	Missing	366 ( 11.9)	354 ( 13.2)
##	bmi (mean (SD))	26.75 (5.11)	25.88 (4.65)
##	bmi_category (%)		
##	normal weight	1073 ( 34.9)	1122 ( 41.9)
##	obese	661 ( 21.5)	454 ( 17.0)
##	other	439 ( 14.3)	321 ( 12.0)
##	overweight	864 ( 28.1)	746 ( 27.9)
##	underweight	40 ( 1.3)	35 ( 1.3)
##	WalkScore (mean (SD))	36.05 (24.44)	81.88 (16.05)
##	WalkScoreLabel (%)		
##		27 ( 0.9)	28 ( 1.0)
##	Car-Dependent	2094 ( 68.1)	135 ( 5.0)
##	Somewhat Walkable	733 ( 23.8)	303 ( 11.3)
##	Very Walkable	175 ( 5.7)	1117 ( 41.7)
##	Walker's Paradise	48 ( 1.6)	1095 ( 40.9)
##	TransitScore (mean (SD))	79.31 (7.37)	13.38 (18.57)
##	TransitScoreLabel (%)		
##		3077 (100.0)	45 ( 1.7)
##	Excellent Transit	0 ( 0.0)	71 ( 2.7)
##	Good Transit	0 ( 0.0)	485 ( 18.1)
##	Minimal Transit	0 ( 0.0)	43 ( 1.6)
##	Rider's Paradise	0 ( 0.0)	1911 ( 71.4)
##	Some Transit	0 ( 0.0)	123 ( 4.6)
##	BikeScore (mean (SD))	133.74 (18.39)	86.99 (43.07)
##	BikeScoreLabel (%)		
##		2898 ( 94.2)	624 ( 23.3)
##	Bikeable	90 ( 2.9)	719 ( 26.8)
##	Biker's Paradise	40 ( 1.3)	496 ( 18.5)
##	Somewhat Bikeable	1 ( 0.0)	83 ( 3.1)
##	Very Bikeable	48 ( 1.6)	756 ( 28.2)
##	DiningandDrinkingScore (mean (SD))	40.56 (26.92)	82.34 (17.58)
##	GroceryScore (mean (SD))	38.81 (34.49)	89.47 (17.42)

	Stratified by ville	
	New-York	Philadelphie
n	3824	1478
language = Fren/Span (%)	16 ( 0.4)	3 ( 0.2)
ville (%)		
Boston	0 ( 0.0)	0 ( 0.0)
Chicago	0 ( 0.0)	0 ( 0.0)
Détroit	0 ( 0.0)	0 ( 0.0)
Montréal	0 ( 0.0)	0 ( 0.0)
New-York	3824 (100.0)	0 ( 0.0)
Philadelphie	0 ( 0.0)	1478 (100.0)
Toronto	0 ( 0.0)	0 ( 0.0)
Vancouver	0 ( 0.0)	0 ( 0.0)
gender = Male (%)	1593 ( 41.7)	536 ( 36.3)
health (%)		
Excellent	909 ( 23.8)	297 ( 20.1)
Good	986 ( 25.8)	404 ( 27.4)
Poor/Fair	295 ( 7.7)	144 ( 9.8)
Very Good	1628 ( 42.6)	631 ( 42.8)
common_transportation (%)		
Bicycle	184 ( 4.8)	87 ( 5.9)
Car	360 ( 9.4)	606 ( 41.1)
Other	83 ( 2.2)	17 ( 1.2)
Public Transportation	2207 ( 57.9)	321 ( 21.8)
Walking	978 ( 25.7)	443 ( 30.1)
physically_active = Yes (%)	3088 ( 82.0)	1193 ( 81.5)
pa_level (%)		
Low	236 ( 6.2)	99 ( 6.7)
Moderate	1578 ( 41.3)	627 ( 42.4)
NA	809 ( 21.2)	312 ( 21.1)
Sedentary	2 ( 0.1)	0 ( 0.0)
Vigorous	1199 ( 31.4)	440 ( 29.8)
day_per_week_motor_vehicle (mean (SD))	2.26 (2.21)	3.85 (2.41)
day_per_week_public_transit (mean (SD))	4.17 (2.16)	2.32 (2.22)
day_per_week_walking (mean (SD))	5.15 (2.00)	4.32 (2.31)
day_per_week_bike (mean (SD))	0.89 (1.63)	0.98 (1.76)
q42 (mean (SD))	39.74 (13.55)	39.19 (13.51)
marital_status = Single (%)	1879 ( 54.7)	729 ( 55.6)
children_household (mean (SD))	0.33 (0.70)	0.38 (0.77)
ethnicity (%)		
African American/Canadian	336 ( 9.0)	231 ( 15.9)
Asian	483 ( 13.0)	128 ( 8.8)
Caucasian	2493 ( 66.9)	999 ( 68.6)
Hispanic	317 ( 8.5)	60 ( 4.1)
Native American/Indigenous	9 ( 0.2)	0 ( 0.0)
Other	90 ( 2.4)	38 ( 2.6)
country_born (%)		
Canada	50 ( 1.3)	7 ( 0.5)
Other	497 ( 13.0)	94 ( 6.4)
United States	3277 ( 85.7)	1377 ( 93.2)
motor_vehicle_access = Yes (%)	1671 ( 44.2)	1109 ( 75.9)
education (%)		
Baccalaureate	1589 ( 41.8)	509 ( 34.9)
Cégep	466 ( 12.3)	206 ( 14.1)

##	Certificate/Diploma	168 ( 4.4)	86 ( 5.9)
##	Graduate School	1284 ( 33.8)	447 ( 30.6)
##	High School/Lower	294 ( 7.7)	212 ( 14.5)
##	occupation_status (%)		
##	Employed	2943 ( 77.7)	1049 ( 71.3)
##	Student	270 ( 7.1)	172 ( 11.7)
##	Unemployed	574 ( 15.2)	251 ( 17.1)
##	household_income (%)		
##	\$0-\$49999	1002 ( 26.2)	562 ( 38.0)
##	\$100000-\$149999	631 ( 16.5)	182 ( 12.3)
##	\$150000+	598 ( 15.6)	124 ( 8.4)
##	\$50000-\$99999	1156 ( 30.2)	483 ( 32.7)
##	Missing	437 ( 11.4)	127 ( 8.6)
##	bmi (mean (SD))	25.05 (4.52)	26.02 (4.97)
##	bmi_category (%)		
##	normal weight	1841 ( 48.1)	618 ( 41.8)
##	obese	490 ( 12.8)	273 ( 18.5)
##	other	450 ( 11.8)	164 ( 11.1)
##	overweight	971 ( 25.4)	404 ( 27.3)
##	underweight	72 ( 1.9)	19 ( 1.3)
##	WalkScore (mean (SD))	95.88 (11.04)	86.27 (17.72)
##	WalkScoreLabel (%)		
##		65 ( 1.7)	22 ( 1.5)
##	Car-Dependent	61 ( 1.6)	83 ( 5.6)
##	Somewhat Walkable	33 ( 0.9)	96 ( 6.5)
##	Very Walkable	169 ( 4.4)	244 ( 16.5)
##	Walker's Paradise	3496 ( 91.4)	1033 ( 69.9)
##	TransitScore (mean (SD))	71.19 (24.19)	73.31 (21.71)
##	TransitScoreLabel (%)		
##		3394 ( 88.8)	1371 ( 92.8)
##	Excellent Transit	34 ( 0.9)	0 ( 0.0)
##	Good Transit	7 ( 0.2)	2 ( 0.1)
##	Minimal Transit	0 ( 0.0)	0 ( 0.0)
##	Rider's Paradise	389 ( 10.2)	105 ( 7.1)
##	Some Transit	0 ( 0.0)	0 ( 0.0)
##	BikeScore (mean (SD))	90.15 (24.92)	106.32 (30.52)
##	BikeScoreLabel (%)		
##		178 ( 4.7)	148 ( 10.0)
##	Bikeable	1137 ( 29.7)	236 ( 16.0)
##	Biker's Paradise	145 ( 3.8)	594 ( 40.2)
##	Somewhat Bikeable	71 ( 1.9)	38 ( 2.6)
##	Very Bikeable	2293 ( 60.0)	462 ( 31.3)
##	DiningandDrinkingScore (mean (SD))	96.05 (11.20)	87.70 (16.44)
##	GroceryScore (mean (SD))	97.91 (11.21)	90.43 (22.07)
##		Stratified by ville	
##		Toronto	Vancouver
##	n	4264	2518
##	language = Fren/Span (%)	3 ( 0.1)	0 ( 0.0)
##	ville (%)		
##	Boston	0 ( 0.0)	0 ( 0.0)
##	Chicago	0 ( 0.0)	0 ( 0.0)
##	Détroit	0 ( 0.0)	0 ( 0.0)
##	Montréal	0 ( 0.0)	0 ( 0.0)
##	New-York	0 ( 0.0)	0 ( 0.0)



##	Philadelphia	0 ( 0.0)	0 ( 0.0)
##	Toronto	4264 (100.0)	0 ( 0.0)
##	Vancouver	0 ( 0.0)	2518 (100.0)
##	gender = Male (%)	1984 ( 46.5)	1131 ( 44.9)
##	health (%)		
##	Excellent	811 ( 19.1)	467 ( 18.6)
##	Good	1220 ( 28.7)	770 ( 30.6)
##	Poor/Fair	464 ( 10.9)	339 ( 13.5)
##	Very Good	1761 ( 41.4)	937 ( 37.3)
##	common_transportation (%)		
##	Bicycle	231 ( 5.4)	119 ( 4.7)
##	Car	1665 ( 39.2)	1245 ( 49.5)
##	Other	42 ( 1.0)	15 ( 0.6)
##	Public Transportation	1459 ( 34.3)	633 ( 25.2)
##	Walking	853 ( 20.1)	503 ( 20.0)
##	physically_active = Yes (%)	3493 ( 83.0)	2114 ( 85.0)
##	pa_level (%)		
##	Low	256 ( 6.0)	137 ( 5.4)
##	Moderate	2131 ( 50.0)	1305 ( 51.8)
##	NA	871 ( 20.4)	474 ( 18.8)
##	Sedentary	3 ( 0.1)	5 ( 0.2)
##	Vigorous	1003 ( 23.5)	597 ( 23.7)
##	day_per_week_motor_vehicle (mean (SD))	3.66 (2.41)	4.05 (2.32)
##	day_per_week_public_transit (mean (SD))	2.68 (2.27)	2.22 (2.23)
##	day_per_week_walking (mean (SD))	4.19 (2.25)	4.22 (2.26)
##	day_per_week_bike (mean (SD))	0.95 (1.73)	0.78 (1.53)
##	q42 (mean (SD))	46.98 (14.52)	47.49 (14.51)
##	marital_status = Single (%)	1345 ( 37.6)	736 ( 34.4)
##	children_household (mean (SD))	0.31 (0.68)	0.36 (0.72)
##	ethnicity (%)		
##	African American/Canadian	96 ( 2.3)	19 ( 0.8)
##	Asian	572 ( 14.0)	720 ( 29.9)
##	Caucasian	3065 ( 74.8)	1531 ( 63.6)
##	Hispanic	71 ( 1.7)	24 ( 1.0)
##	Native American/Indigenous	14 ( 0.3)	18 ( 0.7)
##	Other	279 ( 6.8)	95 ( 3.9)
##	country_born (%)		
##	Canada	3058 ( 71.7)	1822 ( 72.4)
##	Other	1100 ( 25.8)	630 ( 25.0)
##	United States	106 ( 2.5)	66 ( 2.6)
##	motor_vehicle_access = Yes (%)	3057 ( 72.7)	2074 ( 83.1)
##	education (%)		
##	Baccalaureate	1543 ( 36.6)	877 ( 35.7)
##	Cégep	451 ( 10.7)	314 ( 12.8)
##	Certificate/Diploma	746 ( 17.7)	427 ( 17.4)
##	Graduate School	995 ( 23.6)	518 ( 21.1)
##	High School/Lower	481 ( 11.4)	323 ( 13.1)
##	occupation_status (%)		
##	Employed	3051 ( 72.2)	1725 ( 69.4)
##	Student	134 ( 3.2)	83 ( 3.3)
##	Unemployed	1040 ( 24.6)	679 ( 27.3)
##	household_income (%)		
##	\$0-\$49999	1069 ( 25.1)	636 ( 25.3)
##	\$100000-\$149999	673 ( 15.8)	369 ( 14.7)

##	\$150000+	463 ( 10.9)	204 ( 8.1)
##	\$50000-\$99999	1399 ( 32.8)	878 ( 34.9)
##	Missing	660 ( 15.5)	431 ( 17.1)
##	bmi (mean (SD))	25.60 (4.46)	24.87 (4.27)
##	bmi_category (%)		
##	normal weight	1843 ( 43.2)	1213 ( 48.2)
##	obese	586 ( 13.7)	262 ( 10.4)
##	other	498 ( 11.7)	239 ( 9.5)
##	overweight	1276 ( 29.9)	747 ( 29.7)
##	underweight	61 ( 1.4)	57 ( 2.3)
##	WalkScore (mean (SD))	79.56 (18.77)	78.79 (19.52)
##	WalkScoreLabel (%)		
##		26 ( 0.6)	27 ( 1.1)
##	Car-Dependent	351 ( 8.2)	243 ( 9.7)
##	Somewhat Walkable	700 ( 16.4)	403 ( 16.0)
##	Very Walkable	1467 ( 34.4)	869 ( 34.5)
##	Walker's Paradise	1720 ( 40.3)	976 ( 38.8)
##	TransitScore (mean (SD))	48.42 (25.00)	45.29 (19.95)
##	TransitScoreLabel (%)		
##		33 ( 0.8)	88 ( 3.5)
##	Excellent Transit	1547 ( 36.3)	1035 ( 41.1)
##	Good Transit	939 ( 22.0)	865 ( 34.4)
##	Minimal Transit	34 ( 0.8)	5 ( 0.2)
##	Rider's Paradise	1671 ( 39.2)	371 ( 14.7)
##	Some Transit	40 ( 0.9)	154 ( 6.1)
##	BikeScore (mean (SD))	70.90 (30.62)	92.61 (36.44)
##	BikeScoreLabel (%)		
##		120 ( 2.8)	512 ( 20.3)
##	Bikeable	2214 ( 51.9)	745 ( 29.6)
##	Biker's Paradise	431 ( 10.1)	422 ( 16.8)
##	Somewhat Bikeable	500 ( 11.7)	179 ( 7.1)
##	Very Bikeable	999 ( 23.4)	660 ( 26.2)
##	DiningandDrinkingScore (mean (SD))	81.64 (19.10)	81.17 (19.59)
##	GroceryScore (mean (SD))	82.99 (23.65)	84.45 (22.22)
##		Stratified by ville	
##		p	test
##	n		
##	language = Fren/Span (%)	<0.001	
##	ville (%)	<0.001	
##	Boston		
##	Chicago		
##	Détroit		
##	Montréal		
##	New-York		
##	Philadelphie		
##	Toronto		
##	Vancouver		
##	gender = Male (%)	<0.001	
##	health (%)	<0.001	
##	Excellent		
##	Good		
##	Poor/Fair		
##	Very Good		
##	common_transportation (%)	<0.001	

```

##      Bicycle
##      Car
##      Other
##      Public Transportation
##      Walking
##      physically_active = Yes (%)          <0.001
##      pa_level (%)                        <0.001
##      Low
##      Moderate
##      NA
##      Sedentary
##      Vigorous
##      day_per_week_motor_vehicle (mean (SD)) <0.001
##      day_per_week_public_transit (mean (SD)) <0.001
##      day_per_week_walking (mean (SD))      <0.001
##      day_per_week_bike (mean (SD))         <0.001
##      q42 (mean (SD))                      <0.001
##      marital_status = Single (%)           <0.001
##      children_household (mean (SD))        <0.001
##      ethnicity (%)                        <0.001
##      African American/Canadian
##      Asian
##      Caucasian
##      Hispanic
##      Native American/Indigenous
##      Other
##      country_born (%)                     <0.001
##      Canada
##      Other
##      United States
##      motor_vehicle_access = Yes (%)        <0.001
##      education (%)                       <0.001
##      Baccalaureate
##      Cégep
##      Certificate/Diploma
##      Graduate School
##      High School/Lower
##      occupation_status (%)                <0.001
##      Employed
##      Student
##      Unemployed
##      household_income (%)                 <0.001
##      $0-$49999
##      $100000-$149999
##      $150000+
##      $50000-$99999
##      Missing
##      bmi (mean (SD))                     <0.001
##      bmi_category (%)                    <0.001
##      normal weight
##      obese
##      other
##      overweight
##      underweight

```

```
## WalkScore (mean (SD)) <0.001
## WalkScoreLabel (%) <0.001
##
## Car-Dependent
## Somewhat Walkable
## Very Walkable
## Walker's Paradise
## TransitScore (mean (SD)) <0.001
## TransitScoreLabel (%) <0.001
##
## Excellent Transit
## Good Transit
## Minimal Transit
## Rider's Paradise
## Some Transit
## BikeScore (mean (SD)) <0.001
## BikeScoreLabel (%) <0.001
##
## Bikeable
## Biker's Paradise
## Somewhat Bikeable
## Very Bikeable
## DiningandDrinkingScore (mean (SD)) <0.001
## GroceryScore (mean (SD)) <0.001
```

## Table of Cities Combined

```
vars <- c('language', 'ville', 'gender', 'health', 'common_transportation', 'physically_active', 'pa_le
CreateTableOne(vars = vars, data = ibiccs)
```

```
##
## Overall
## n 23901
## language = Fren/Span (%) 1665 ( 7.0)
## ville (%)
## Boston 1977 ( 8.3)
## Chicago 4085 (17.1)
## Détroit 3077 (12.9)
## Montréal 2678 (11.2)
## New-York 3824 (16.0)
## Philadelphie 1478 ( 6.2)
## Toronto 4264 (17.8)
## Vancouver 2518 (10.5)
## gender = Male (%) 9859 (41.2)
## health (%)
## Excellent 4803 (20.1)
## Good 6725 (28.2)
## Poor/Fair 2593 (10.9)
## Very Good 9741 (40.8)
## common_transportation (%)
## Bicycle 1177 ( 4.9)
## Car 9996 (42.0)
```

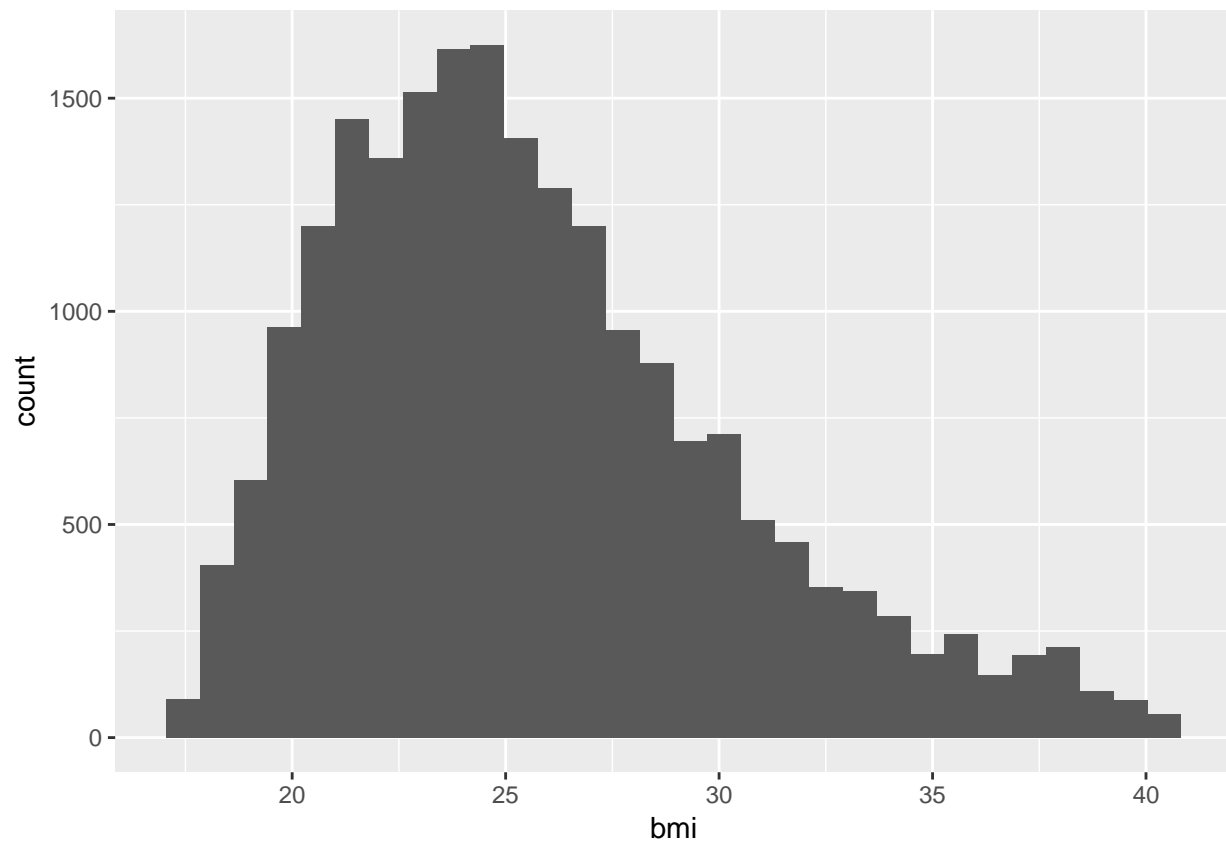
##	Other	261 ( 1.1)
##	Public Transportation	7886 (33.1)
##	Walking	4484 (18.8)
##	physically_active = Yes (%)	19471 (82.5)
##	pa_level (%)	
##	Low	1547 ( 6.5)
##	Moderate	10771 (45.1)
##	NA	4949 (20.7)
##	Sedentary	30 ( 0.1)
##	Vigorous	6604 (27.6)
##	day_per_week_motor_vehicle (mean (SD))	3.81 (2.49)
##	day_per_week_public_transit (mean (SD))	2.65 (2.41)
##	day_per_week_walking (mean (SD))	4.13 (2.38)
##	day_per_week_bike (mean (SD))	0.85 (1.63)
##	q42 (mean (SD))	42.57 (14.20)
##	marital_status = Single (%)	9452 (45.4)
##	children_household (mean (SD))	0.36 (0.75)
##	ethnicity (%)	
##	African American/Canadian	1654 ( 7.1)
##	Asian	2661 (11.4)
##	Caucasian	17154 (73.7)
##	Hispanic	933 ( 4.0)
##	Native American/Indigenous	79 ( 0.3)
##	Other	781 ( 3.4)
##	country_born (%)	
##	Canada	7171 (30.0)
##	Other	3555 (14.9)
##	United States	13175 (55.1)
##	motor_vehicle_access = Yes (%)	17376 (73.4)
##	education (%)	
##	Baccalaureate	8638 (36.5)
##	Cégep	3197 (13.5)
##	Certificate/Diploma	2559 (10.8)
##	Graduate School	6655 (28.2)
##	High School/Lower	2591 (11.0)
##	occupation_status (%)	
##	Employed	17272 (72.9)
##	Student	1586 ( 6.7)
##	Unemployed	4847 (20.4)
##	household_income (%)	
##	\$0-\$49999	7211 (30.2)
##	\$100000-\$149999	3490 (14.6)
##	\$150000+	2400 (10.0)
##	\$50000-\$99999	7821 (32.7)
##	Missing	2979 (12.5)
##	bmi (mean (SD))	25.64 (4.70)
##	bmi_category (%)	
##	normal weight	10437 (43.7)
##	obese	3672 (15.4)
##	other	2741 (11.5)
##	overweight	6658 (27.9)
##	underweight	393 ( 1.6)
##	WalkScore (mean (SD))	77.81 (24.80)
##	WalkScoreLabel (%)	

```
##                284 ( 1.2)
##      Car-Dependent      3337 (14.0)
##      Somewhat Walkable  2863 (12.0)
##      Very Walkable      6319 (26.4)
##      Walker's Paradise  11098 (46.4)
##      TransitScore (mean (SD))  60.30 (28.30)
##      TransitScoreLabel (%)
##                13415 (56.1)
##      Excellent Transit    3165 (13.2)
##      Good Transit         2444 (10.2)
##      Minimal Transit      82 ( 0.3)
##      Rider's Paradise     4476 (18.7)
##      Some Transit         319 ( 1.3)
##      BikeScore (mean (SD))  94.10 (35.92)
##      BikeScoreLabel (%)
##                5636 (23.6)
##      Bikeable            6729 (28.2)
##      Biker's Paradise     2451 (10.3)
##      Somewhat Bikeable    1162 ( 4.9)
##      Very Bikeable        7923 (33.1)
##      DiningandDrinkingScore (mean (SD))  79.53 (24.59)
##      GroceryScore (mean (SD))  80.76 (29.33)
```

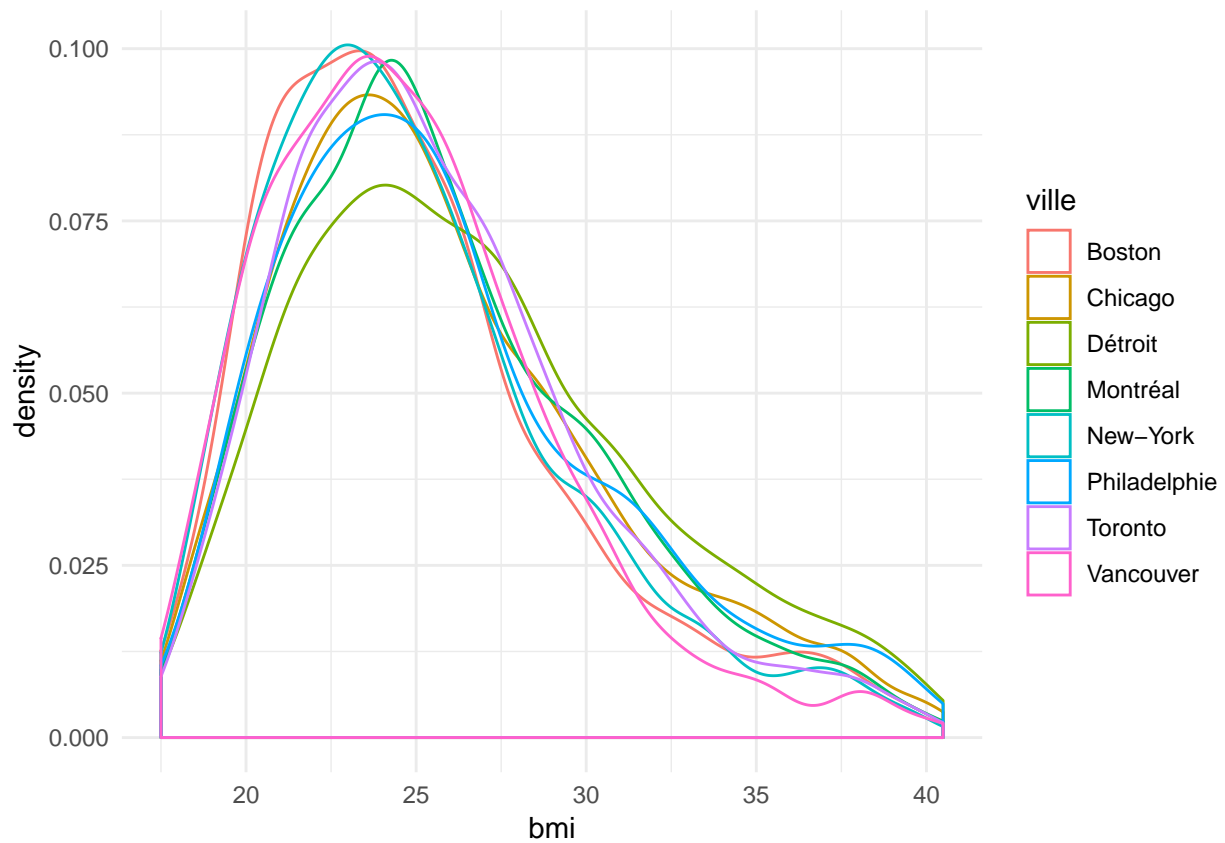
## Histogram of BMI

```
hist_bmi <- ggplot(city, aes(bmi)) +
  geom_histogram()
plot(hist_bmi)
```

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



```
hist_bmi_city <- ggplot(city, aes(bmi)) +  
  geom_density(aes(group = ville, colour = ville)) +  
  theme_minimal()  
plot(hist_bmi_city)
```

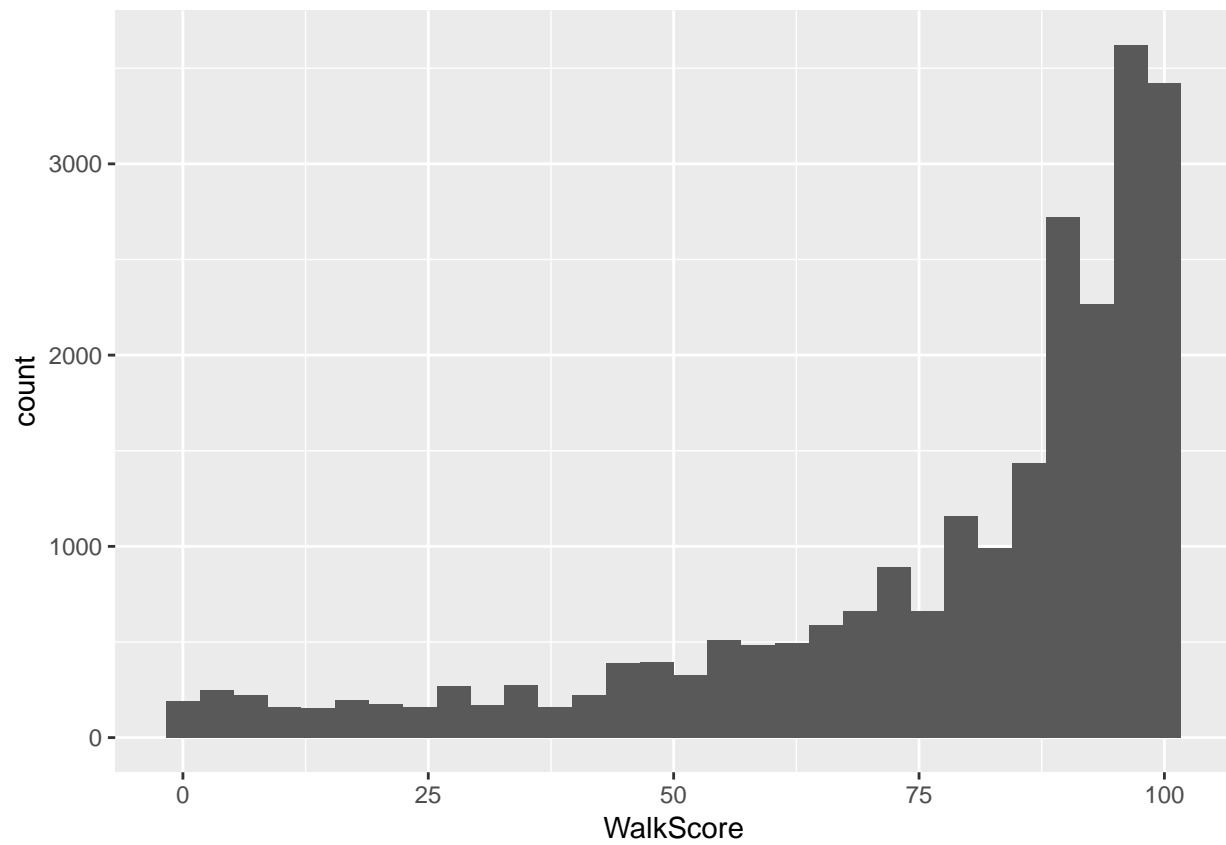


## Histogram of WalkScore

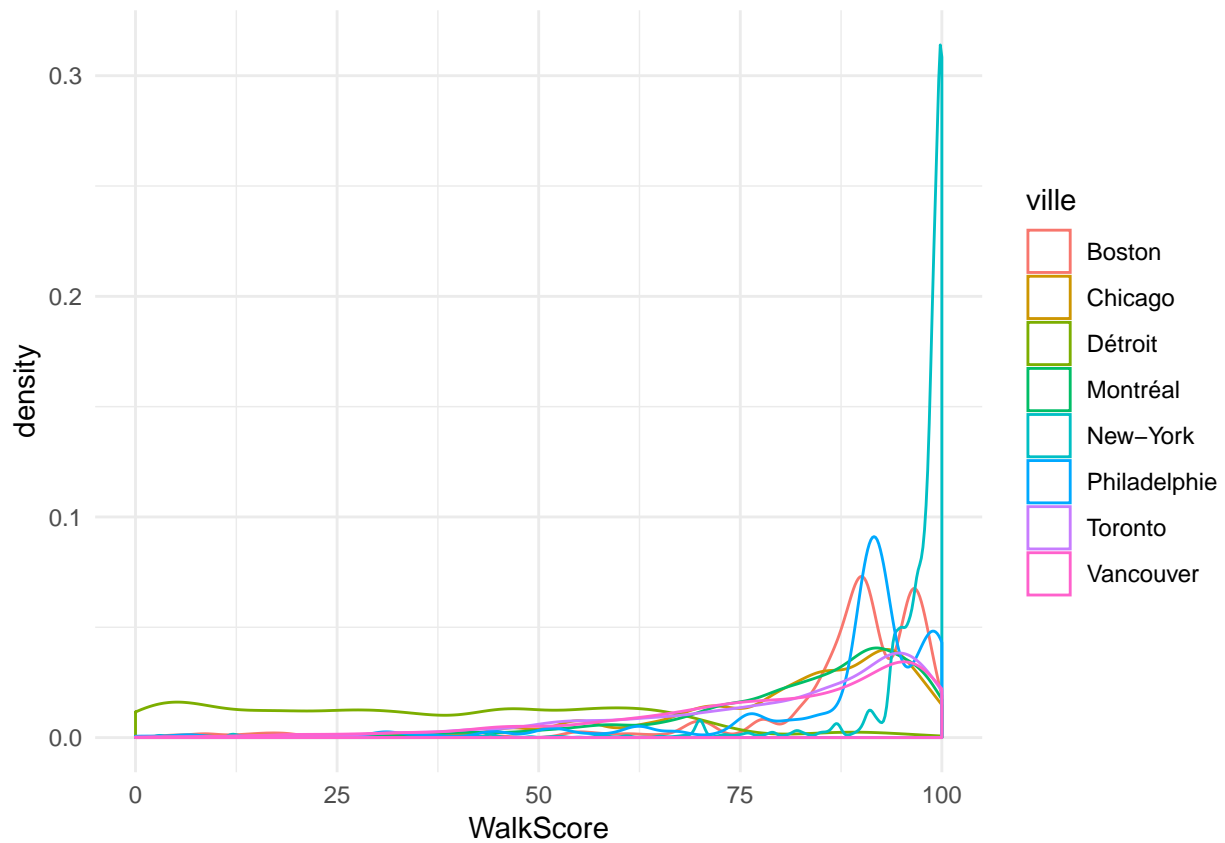
```
hist_walkscore <- ggplot(city, aes(WalkScore)) +  
  geom_histogram()  
plot(hist_walkscore)
```

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



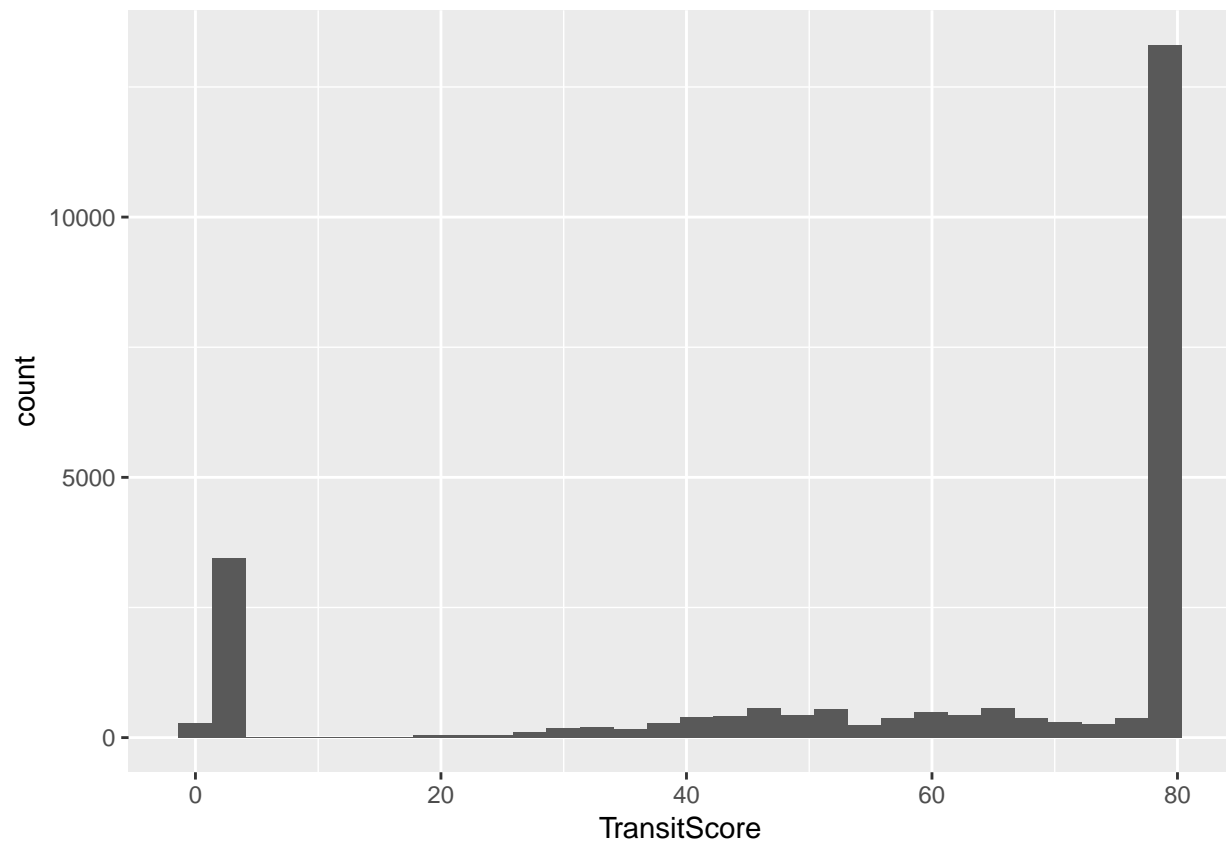


```
hist_walkscore_city <- ggplot(city, aes(WalkScore)) +  
  geom_density(aes(group = ville, colour = ville)) +  
  theme_minimal()  
plot(hist_walkscore_city)
```

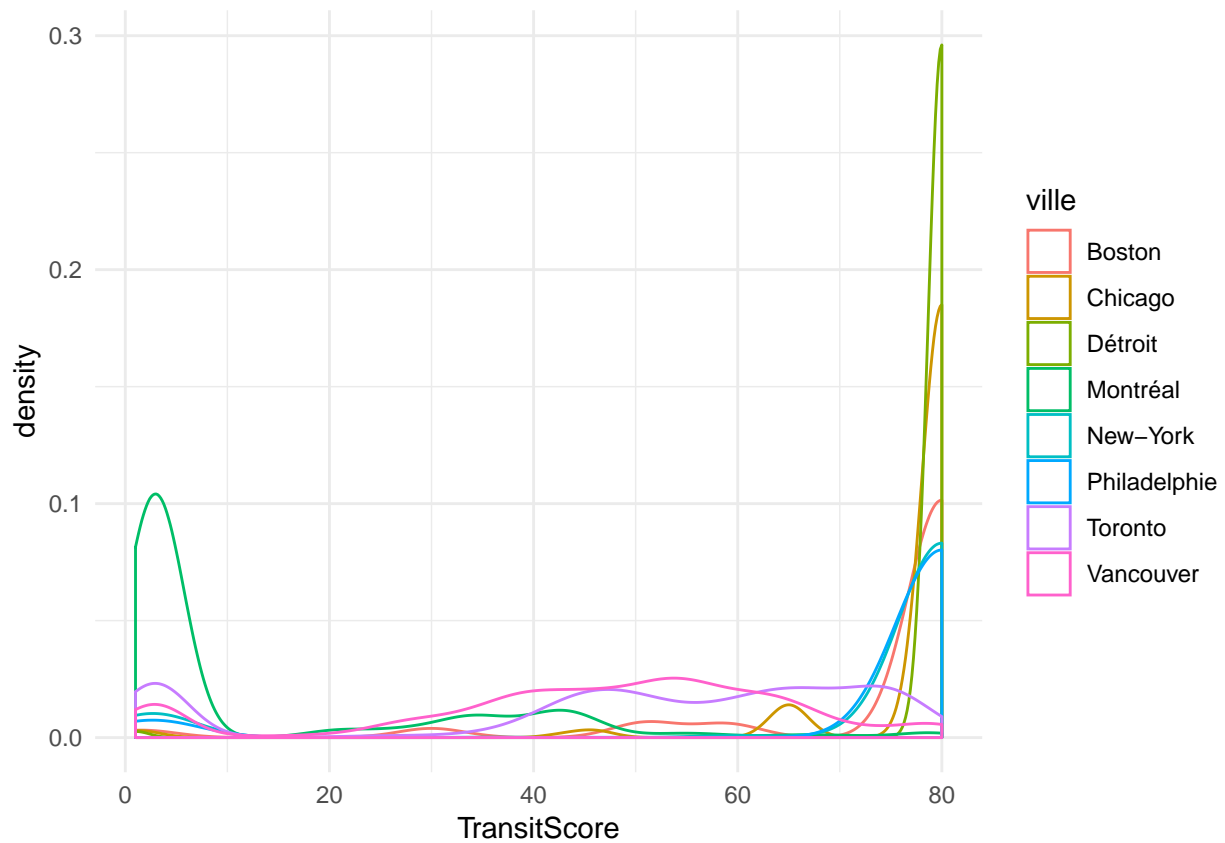


## Histogram of TransitScore

```
hist_trasnsitscore <- ggplot(city, aes(TransitScore)) +  
  geom_histogram()  
plot(hist_trasnsitscore)  
  
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



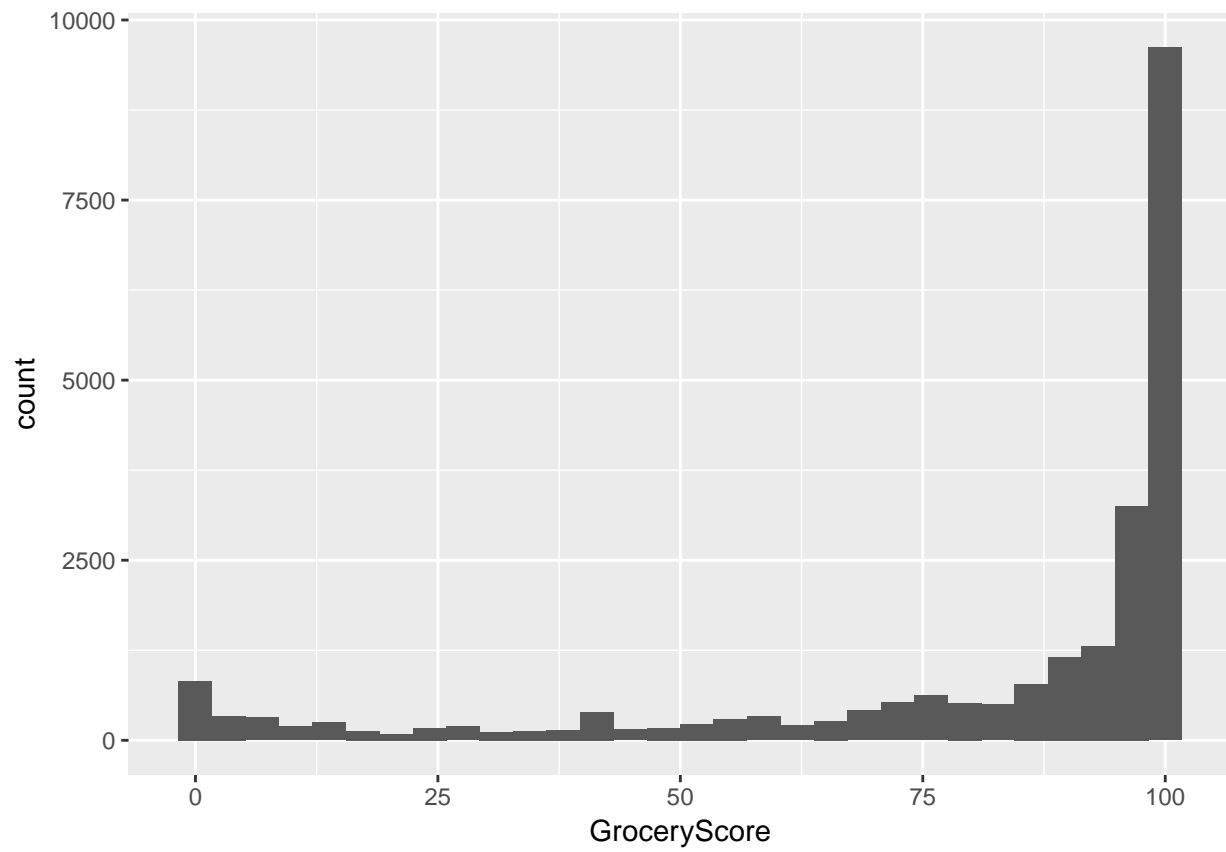
```
hist_trasnsitscore_city <- ggplot(city, aes(TransitScore)) +  
  geom_density(aes(group = ville, colour = ville)) +  
  theme_minimal()  
plot(hist_trasnsitscore_city)
```



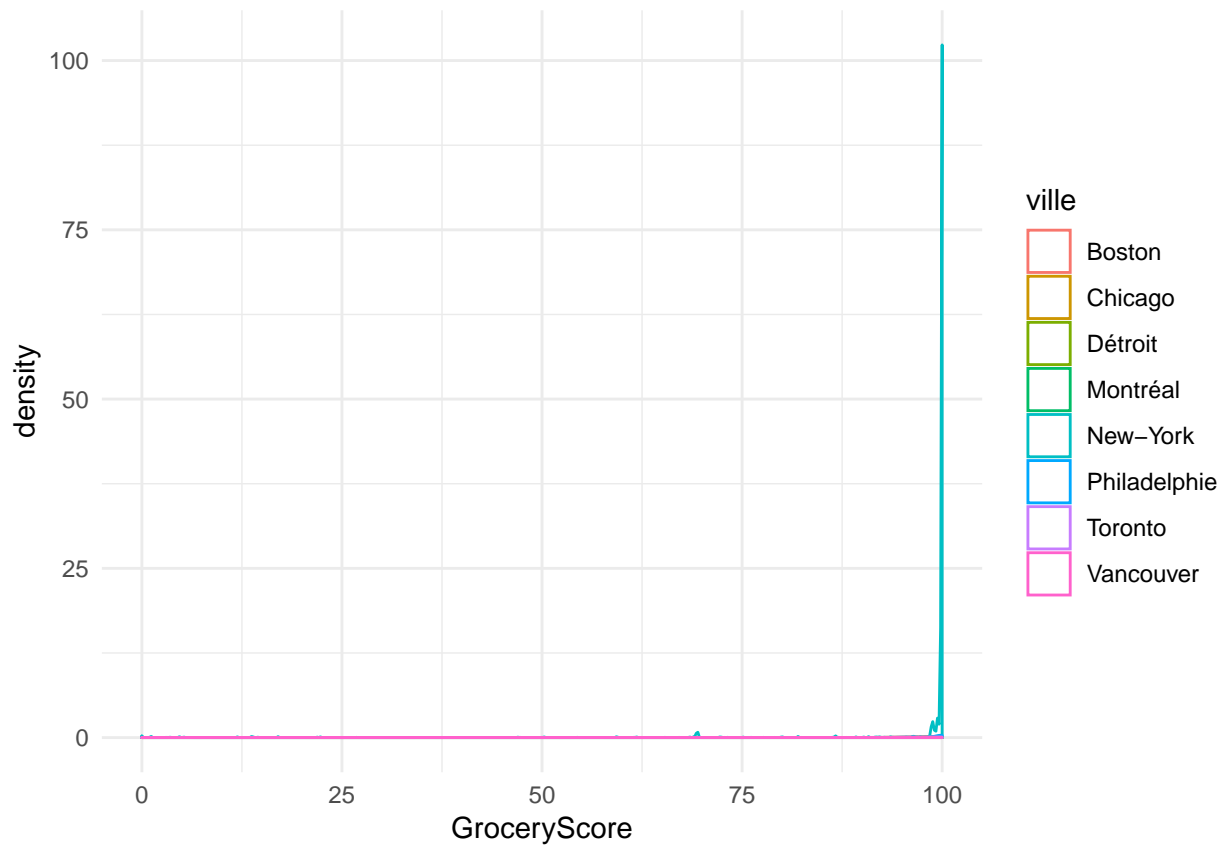
## Histogram of GroceryScore

```
hist_groceryscore <- ggplot(city, aes(GroceryScore)) +
  geom_histogram()
plot(hist_groceryscore)

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



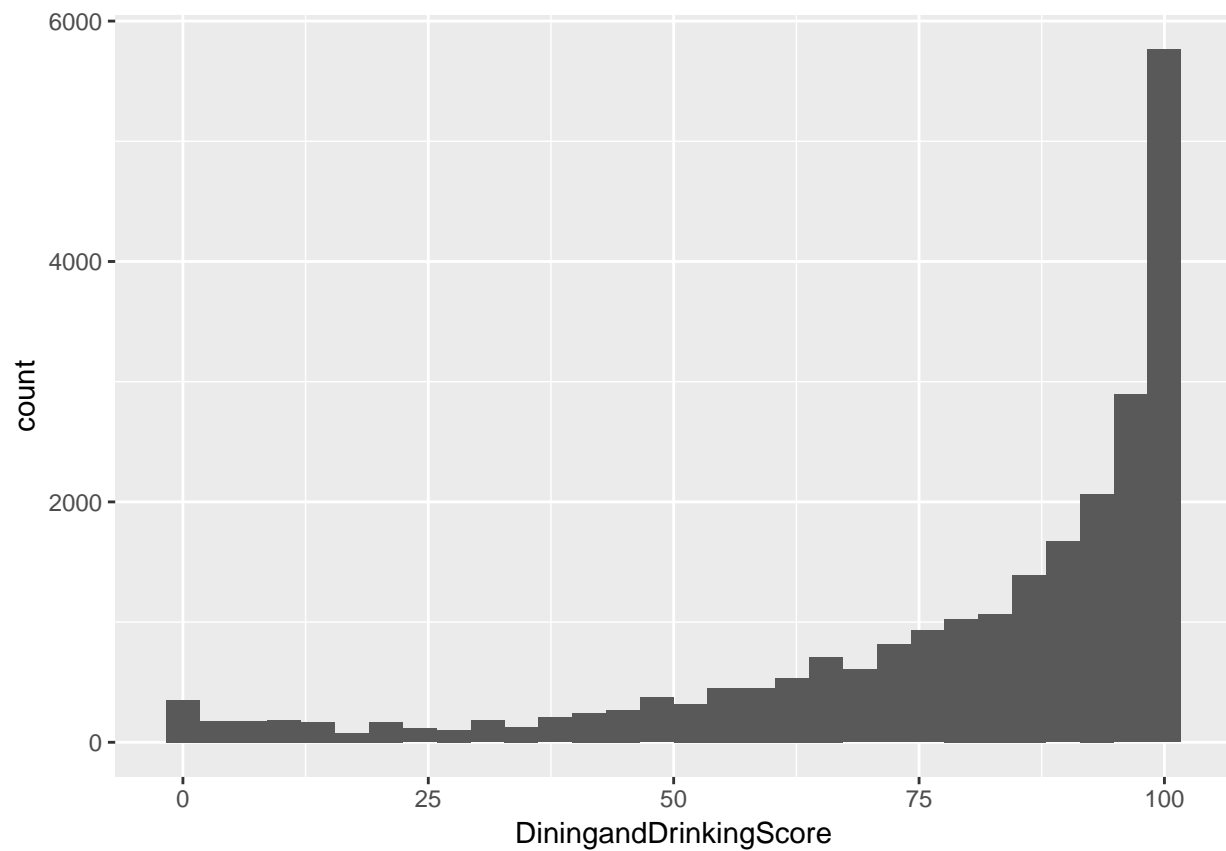
```
hist_groceryscore_city <- ggplot(city, aes(GroceryScore)) +  
  geom_density(aes(group = ville, colour = ville)) +  
  theme_minimal()  
plot(hist_groceryscore_city)
```



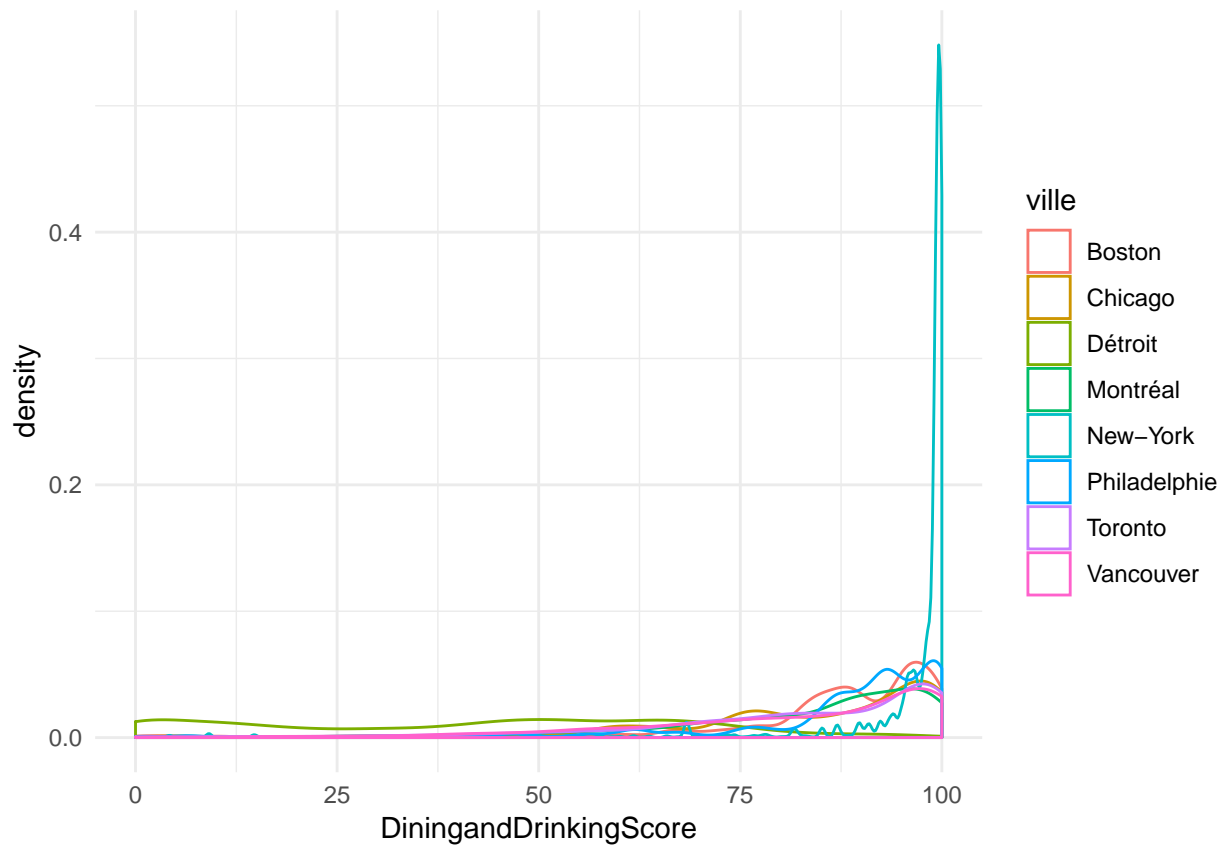
## Histogram of DiningAndDrinkingScore

```
hist_diningdrinkingscore <- ggplot(city, aes(DiningandDrinkingScore)) +  
  geom_histogram()  
plot(hist_diningdrinkingscore)
```

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



```
hist_diningdrinkingscore_city <- ggplot(city, aes(DiningandDrinkingScore)) +  
  geom_density(aes(group = ville, colour = ville)) +  
  theme_minimal()  
plot(hist_diningdrinkingscore_city)
```



## Linear Regression

```
explanatory <- c('language', 'ville', 'gender', 'health', 'physically_active', 'q42', 'marital_status',
dependent <- "bmi"
ibiccs %>%
  finalfit.lm(dependent, explanatory, na.rm = TRUE)
```

```
##           Dependent: bmi                               Mean (sd)
## 26           language                                English 25.6 (4.7)
## 27                                     Fren/Span 26.0 (4.8)
## 39           ville                                    Boston 25.1 (4.6)
## 40                                     Chicago 25.9 (4.9)
## 41                                     Détroit 26.7 (5.1)
## 42                                     Montréal 25.9 (4.7)
## 43                                     New-York 25.0 (4.5)
## 44           Philadelphie 26.0 (5.0)
## 45                                     Toronto 25.6 (4.5)
## 46           Vancouver 24.9 (4.3)
## 19           gender                                   Female 25.0 (4.9)
## 20                                     Male 26.4 (4.3)
## 22           health                                   Excellent 24.0 (3.7)
## 23                                     Good 26.8 (5.0)
## 24           Poor/Fair 28.4 (5.6)
## 25           Very Good 25.1 (4.2)
## 35           physically_active                        No 26.7 (5.3)
```



## 36		Yes	25.4 (4.5)
## 37	q42	[18,94]	25.6 (4.7)
## 28	marital_status	Relationship/Married/Common-Law	25.8 (4.6)
## 29		Single	25.2 (4.8)
## 11	country_born	Canada	25.7 (4.6)
## 12		Other	24.8 (4.3)
## 13		United States	25.9 (4.9)
## 30	motor_vehicle_access	No	25.2 (4.8)
## 31		Yes	25.8 (4.7)
## 32	occupation_status	Employed	25.6 (4.6)
## 33		Student	24.2 (4.8)
## 34		Unemployed	26.4 (4.9)
## 2	bmi_category	normal weight	22.2 (1.7)
## 3		obese	33.6 (2.8)
## 4		overweight	27.1 (1.4)
## 5		underweight	18.1 (0.3)
## 47	WalkScore	[0,100]	25.6 (4.7)
## 38	TransitScore	[1,80]	25.6 (4.7)
## 1	BikeScore	[1,138]	25.6 (4.7)
## 18	DiningandDrinkingScore	[0,100]	25.6 (4.7)
## 21	GroceryScore	[0,100]	25.6 (4.7)
## 15	day_per_week_motor_vehicle	[0,7]	25.6 (4.7)
## 16	day_per_week_public_transit	[0,7]	25.6 (4.7)
## 17	day_per_week_walking	[0,7]	25.6 (4.7)
## 14	day_per_week_bike	[0,7]	25.6 (4.6)
## 7	children_household	0	25.5 (4.7)
## 8		1	25.9 (4.8)
## 9		2	26.1 (4.7)
## 10		3	26.9 (5.2)
## 6	<NA>	<NA>	<NA>
##	Coefficient (univariable)	Coefficient (multivariable)	
## 26	-	-	
## 27	0.43 (0.18 to 0.69, p=0.001)	-0.11 (-0.31 to 0.08, p=0.252)	
## 39	-	-	
## 40	0.83 (0.57 to 1.10, p<0.001)	0.07 (-0.06 to 0.21, p=0.273)	
## 41	1.69 (1.41 to 1.97, p<0.001)	0.06 (-0.12 to 0.23, p=0.529)	
## 42	0.82 (0.53 to 1.11, p<0.001)	0.03 (-0.23 to 0.29, p=0.843)	
## 43	-0.01 (-0.28 to 0.26, p=0.937)	-0.11 (-0.25 to 0.03, p=0.129)	
## 44	0.96 (0.63 to 1.29, p<0.001)	-0.04 (-0.20 to 0.13, p=0.678)	
## 45	0.54 (0.28 to 0.81, p<0.001)	-0.01 (-0.21 to 0.19, p=0.901)	
## 46	-0.18 (-0.47 to 0.11, p=0.213)	-0.22 (-0.43 to -0.01, p=0.037)	
## 19	-	-	
## 20	1.39 (1.26 to 1.52, p<0.001)	0.26 (0.19 to 0.33, p<0.001)	
## 22	-	-	
## 23	2.78 (2.60 to 2.95, p<0.001)	0.39 (0.29 to 0.49, p<0.001)	
## 24	4.38 (4.15 to 4.62, p<0.001)	0.64 (0.49 to 0.78, p<0.001)	
## 25	1.08 (0.91 to 1.24, p<0.001)	0.14 (0.06 to 0.23, p=0.001)	
## 35	-	-	
## 36	-1.27 (-1.44 to -1.10, p<0.001)	-0.03 (-0.13 to 0.06, p=0.520)	
## 37	0.06 (0.05 to 0.06, p<0.001)	0.01 (0.01 to 0.01, p<0.001)	
## 28	-	-	
## 29	-0.57 (-0.71 to -0.44, p<0.001)	0.02 (-0.05 to 0.10, p=0.577)	
## 11	-	-	
## 12	-0.91 (-1.11 to -0.71, p<0.001)	-0.17 (-0.29 to -0.05, p=0.006)	

```
## 13    0.18 (0.04 to 0.33, p=0.012)    0.02 (-0.15 to 0.18, p=0.846)
## 30                                     -                               -
## 31    0.63 (0.48 to 0.77, p<0.001)   -0.02 (-0.12 to 0.07, p=0.607)
## 32                                     -                               -
## 33   -1.32 (-1.57 to -1.06, p<0.001)    0.01 (-0.12 to 0.14, p=0.909)
## 34    0.90 (0.74 to 1.06, p<0.001)   -0.06 (-0.16 to 0.03, p=0.205)
## 2                                     -                               -
## 3   11.44 (11.37 to 11.51, p<0.001) 11.12 (11.03 to 11.22, p<0.001)
## 4     4.89 (4.83 to 4.95, p<0.001)    4.68 (4.60 to 4.76, p<0.001)
## 5   -4.11 (-4.30 to -3.92, p<0.001) -3.99 (-4.23 to -3.74, p<0.001)
## 47  -0.02 (-0.02 to -0.02, p<0.001)    0.00 (-0.00 to 0.01, p=0.458)
## 38    0.01 (0.00 to 0.01, p<0.001)    0.00 (-0.00 to 0.00, p=0.585)
## 1     0.00 (0.00 to 0.00, p=0.035)   -0.00 (-0.00 to 0.00, p=0.100)
## 18  -0.02 (-0.02 to -0.02, p<0.001)   -0.01 (-0.01 to 0.00, p=0.050)
## 21  -0.01 (-0.01 to -0.01, p<0.001)    0.00 (-0.00 to 0.00, p=0.264)
## 15    0.21 (0.19 to 0.24, p<0.001)    0.01 (-0.00 to 0.03, p=0.138)
## 16  -0.18 (-0.21 to -0.15, p<0.001)    0.01 (-0.01 to 0.02, p=0.479)
## 17  -0.25 (-0.27 to -0.22, p<0.001) -0.02 (-0.03 to -0.00, p=0.047)
## 14  -0.15 (-0.19 to -0.10, p<0.001)   -0.01 (-0.03 to 0.01, p=0.365)
## 7                                     -                               -
## 8                                     -                               -
## 9                                     -                               -
## 10                                    -                               -
## 6     0.36 (0.27 to 0.44, p<0.001)    0.09 (0.04 to 0.13, p<0.001)
```

```
# Regressions for variables that won't run in explanatory
educ_reg <- lm(bmi ~ factor(education), data = ibiccs)
summary(educ_reg)
```

```
##
## Call:
## lm(formula = bmi ~ factor(education), data = ibiccs)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -9.275 -3.438 -0.801  2.539 15.409
##
## Coefficients:
##                                Estimate Std. Error t value Pr(>|t|)
## (Intercept)                   25.2285     0.0525   480.19  <2e-16
## factor(education)Cégep         1.1505     0.1024    11.23  <2e-16
## factor(education)Certificate/Diploma 1.3631     0.1126    12.10  <2e-16
## factor(education)Graduate School -0.1476     0.0794    -1.86   0.063
## factor(education)High School/Lower 1.6166     0.1153    14.03  <2e-16
##
## (Intercept)                  ***
## factor(education)Cégep      ***
## factor(education)Certificate/Diploma ***
## factor(education)Graduate School .
## factor(education)High School/Lower ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.65 on 21002 degrees of freedom
## (2894 observations deleted due to missingness)
```

```
## Multiple R-squared:  0.021, Adjusted R-squared:  0.0208
## F-statistic: 112 on 4 and 21002 DF,  p-value: <2e-16
```

```
confint(educ_reg)
```

```
##                2.5 %    97.5 %
## (Intercept)      25.1255 25.331477
## factor(education)Cégep      0.9497  1.351290
## factor(education)Certificate/Diploma  1.1424  1.583923
## factor(education)Graduate School    -0.3032  0.007946
## factor(education)High School/Lower   1.3907  1.842495
```

```
tran_reg <- lm(bmi ~ factor(common_transportation), data = ibiccs)
summary(tran_reg)
```

```
##
## Call:
## lm(formula = bmi ~ factor(common_transportation), data = ibiccs)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -8.876 -3.374 -0.764  2.619 15.631
##
## Coefficients:
##                Estimate Std. Error
## (Intercept)      24.556      0.147
## factor(common_transportation)Car      1.860      0.155
## factor(common_transportation)Other     1.723      0.374
## factor(common_transportation)Public Transportation    0.768      0.158
## factor(common_transportation)Walking     0.163      0.164
##
##                t value Pr(>|t|)
## (Intercept)      166.62  < 2e-16 ***
## factor(common_transportation)Car      11.97  < 2e-16 ***
## factor(common_transportation)Other     4.61 0.0000041 ***
## factor(common_transportation)Public Transportation    4.87 0.0000011 ***
## factor(common_transportation)Walking     0.99      0.32
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.65 on 21113 degrees of freedom
## (2783 observations deleted due to missingness)
## Multiple R-squared:  0.023, Adjusted R-squared:  0.0228
## F-statistic: 124 on 4 and 21113 DF,  p-value: <2e-16
```

```
confint(tran_reg)
```

```
##                2.5 %    97.5 %
## (Intercept)      24.2673 24.8451
## factor(common_transportation)Car      1.5556  2.1650
## factor(common_transportation)Other     0.9900  2.4565
## factor(common_transportation)Public Transportation    0.4589  1.0764
## factor(common_transportation)Walking    -0.1590  0.4854
```

```
income_reg <- lm(bmi ~ factor(household_income), data = ibiccs)
summary(income_reg)
```

```
##
```

```
## Call:
## lm(formula = bmi ~ factor(household_income), data = ibiccs)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -8.541 -3.490 -0.801  2.610 15.506
##
## Coefficients:
##              Estimate Std. Error t value
## (Intercept)      26.0414    0.0597  435.99
## factor(household_income)$100000-$149999  -0.4132    0.1016   -4.07
## factor(household_income)$150000+        -0.6987    0.1151   -6.07
## factor(household_income)$50000-$99999    -0.4018    0.0815   -4.93
## factor(household_income)Missing          -1.1772    0.1143  -10.30
##              Pr(>|t|)
## (Intercept)      < 2e-16 ***
## factor(household_income)$100000-$149999  4.7e-05 ***
## factor(household_income)$150000+        1.3e-09 ***
## factor(household_income)$50000-$99999    8.4e-07 ***
## factor(household_income)Missing          < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.69 on 21155 degrees of freedom
## (2741 observations deleted due to missingness)
## Multiple R-squared:  0.00553,    Adjusted R-squared:  0.00534
## F-statistic: 29.4 on 4 and 21155 DF,  p-value: <2e-16
```

```
confint(income_reg)
```

```
##              2.5 %  97.5 %
## (Intercept)      25.9244 26.1585
## factor(household_income)$100000-$149999 -0.6123 -0.2141
## factor(household_income)$150000+        -0.9244 -0.4730
## factor(household_income)$50000-$99999    -0.5616 -0.2420
## factor(household_income)Missing          -1.4012 -0.9532
```

```
ethn_reg <- lm(bmi ~ factor(ethnicity), data = ibiccs)
summary(ethn_reg)
```

```
##
## Call:
## lm(formula = bmi ~ factor(ethnicity), data = ibiccs)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -10.008  -3.398  -0.758   2.542  16.259
##
## Coefficients:
##              Estimate Std. Error t value
## (Intercept)      27.748    0.127  218.21
## factor(ethnicity)Asian      -4.067    0.158  -25.76
## factor(ethnicity)Caucasian  -1.970    0.132  -14.87
## factor(ethnicity)Hispanic   -1.539    0.212   -7.26
## factor(ethnicity)Native American/Indigenous -2.131    0.606   -3.52
```

```
## factor(ethnicity)Other          -2.458      0.217  -11.32
##                                Pr(>|t|)
## (Intercept)                    < 2e-16 ***
## factor(ethnicity)Asian          < 2e-16 ***
## factor(ethnicity)Caucasian      < 2e-16 ***
## factor(ethnicity)Hispanic       4.1e-13 ***
## factor(ethnicity)Native American/Indigenous 0.00044 ***
## factor(ethnicity)Other          < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.63 on 20730 degrees of freedom
## (3165 observations deleted due to missingness)
## Multiple R-squared:  0.0345, Adjusted R-squared:  0.0343
## F-statistic: 148 on 5 and 20730 DF, p-value: <2e-16
```

```
confint(ethn_reg)
```

```
##                                2.5 % 97.5 %
## (Intercept)                  27.498 27.9968
## factor(ethnicity)Asian       -4.376 -3.7575
## factor(ethnicity)Caucasian   -2.230 -1.7103
## factor(ethnicity)Hispanic    -1.955 -1.1232
## factor(ethnicity)Native American/Indigenous -3.319 -0.9437
## factor(ethnicity)Other       -2.883 -2.0325
```

```
pa_lvl_reg <- lm(bmi ~ factor(pa_level), data = ibiccs)
summary(pa_lvl_reg)
```

```
##
## Call:
## lm(formula = bmi ~ factor(pa_level), data = ibiccs)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
##  -9.03  -3.42  -0.75   2.54  15.61
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      25.030      0.127  197.34 < 2e-16 ***
## factor(pa_level)Moderate    0.780      0.135    5.76 8.5e-09 ***
## factor(pa_level)NA         1.562      0.147   10.63 < 2e-16 ***
## factor(pa_level)Sedentary   2.593      0.923    2.81  0.005 **
## factor(pa_level)Vigorous   -0.153      0.140   -1.09  0.274
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.66 on 21155 degrees of freedom
## (2741 observations deleted due to missingness)
## Multiple R-squared:  0.0172, Adjusted R-squared:  0.017
## F-statistic: 92.4 on 4 and 21155 DF, p-value: <2e-16
```

```
confint(pa_lvl_reg)
```

```
##                                2.5 % 97.5 %
## (Intercept)                  24.7818 25.2790
```

```
## factor(pa_level)Moderate    0.5147  1.0454
## factor(pa_level)NA          1.2743  1.8504
## factor(pa_level)Sedentary   0.7835  4.4034
## factor(pa_level)Vigorous    -0.4279  0.1214
```

Interpretation: What covariates are we keeping? 'language', 'ville', 'gender', 'health', 'physically\_active', 'q42', 'marital\_status', 'country\_born', 'motor\_vehicle\_access', 'occupation\_status', 'bmi\_category', 'WalkScore', 'DiningandDrinkingScore', 'GroceryScore', 'day\_per\_week\_motor\_vehicle', 'day\_per\_week\_public\_transit', 'day\_per\_week\_walking', 'day\_per\_week\_bike', 'children\_household', 'education', 'common\_transportation', 'household\_income', 'ethnicity', 'pa\_level'

## Linear Regression

```
lm1 <- lm(bmi ~ WalkScore + DiningandDrinkingScore + GroceryScore + factor(ville), data = ibiccs)
summary(lm1)
```

```
##
## Call:
## lm(formula = bmi ~ WalkScore + DiningandDrinkingScore + GroceryScore +
##     factor(ville), data = ibiccs)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -9.983  -3.455  -0.787   2.590  15.602
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    26.29791    0.19130   137.47 < 2e-16 ***
## WalkScore       -0.01675    0.00756    -2.22  0.02662 *
## DiningandDrinkingScore -0.01317    0.00580    -2.27  0.02325 *
## GroceryScore     0.01474    0.00265     5.57  2.6e-08 ***
## factor(ville)Chicago    0.88916    0.13675     6.50  8.1e-11 ***
## factor(ville)Détroit    1.00145    0.17314     5.78  7.4e-09 ***
## factor(ville)Montréal   0.71927    0.14821     4.85  1.2e-06 ***
## factor(ville)New-York    0.16378    0.13874     1.18  0.23780
## factor(ville)Philadelphie 1.00341    0.17099     5.87  4.5e-09 ***
## factor(ville)Toronto     0.48620    0.13599     3.58  0.00035 ***
## factor(ville)Vancouver  -0.26448    0.15016    -1.76  0.07820 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.66 on 20911 degrees of freedom
## (2979 observations deleted due to missingness)
## Multiple R-squared:  0.0201, Adjusted R-squared:  0.0197
## F-statistic:  43 on 10 and 20911 DF,  p-value: <2e-16
```

```
confint(lm1)
```

```
##              2.5 %      97.5 %
## (Intercept)  25.922939 26.672884
## WalkScore    -0.031561 -0.001943
## DiningandDrinkingScore -0.024546 -0.001795
## GroceryScore  0.009547  0.019924
```

```

## factor(ville)Chicago      0.621122  1.157201
## factor(ville)D  troit     0.662073  1.340821
## factor(ville)Montr  al    0.428758  1.009773
## factor(ville)New-York     -0.108149  0.435718
## factor(ville)Philadelphie 0.668248  1.338564
## factor(ville)Toronto      0.219654  0.752745
## factor(ville)Vancouver    -0.558809  0.029848

lm2 <- lm(bmi ~ WalkScore + DiningandDrinkingScore + GroceryScore + factor(ville) + education + occupat.
summary(lm2)

##
## Call:
## lm(formula = bmi ~ WalkScore + DiningandDrinkingScore + GroceryScore +
##     factor(ville) + education + occupation_status + household_income,
##     data = ibiccs)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -10.778  -3.325  -0.713   2.529  16.596
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    25.99174    0.20607   126.13 < 2e-16 ***
## WalkScore      -0.00983    0.00750    -1.31  0.190
## DiningandDrinkingScore -0.01174    0.00575    -2.04  0.041 *
## GroceryScore     0.01157    0.00263     4.40 1.1e-05 ***
## factor(ville)Chicago  0.66853    0.13557     4.93 8.2e-07 ***
## factor(ville)D  troit  0.80563    0.17164     4.69 2.7e-06 ***
## factor(ville)Montr  al 0.15903    0.14864     1.07  0.285
## factor(ville)New-York -0.06918    0.13766    -0.50  0.615
## factor(ville)Philadelphie 0.75884    0.16953     4.48 7.6e-06 ***
## factor(ville)Toronto  0.04029    0.13657     0.30  0.768
## factor(ville)Vancouver -0.72432    0.15113    -4.79 1.7e-06 ***
## educationC  gep      1.19719    0.10425    11.48 < 2e-16 ***
## educationCertificate/Diploma 1.21253    0.11504    10.54 < 2e-16 ***
## educationGraduate School -0.15236    0.07950    -1.92  0.055 .
## educationHigh School/Lower 1.35470    0.11961    11.33 < 2e-16 ***
## occupation_statusStudent -1.69600    0.13499   -12.56 < 2e-16 ***
## occupation_statusUnemployed 0.62974    0.08410     7.49 7.3e-14 ***
## household_income$100000-$149999 -0.02180    0.10440    -0.21  0.835
## household_income$150000+ -0.09668    0.11962    -0.81  0.419
## household_income$50000-$99999 -0.18349    0.08341    -2.20  0.028 *
## household_incomeMissing -0.89151    0.11609    -7.68 1.7e-14 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.58 on 20675 degrees of freedom
## (3205 observations deleted due to missingness)
## Multiple R-squared:  0.0511, Adjusted R-squared:  0.0502
## F-statistic: 55.7 on 20 and 20675 DF, p-value: <2e-16

confint(lm2)

##              2.5 %      97.5 %

```

```
## (Intercept)                25.587840 26.3956476
## WalkScore                  -0.024533  0.0048823
## DiningandDrinkingScore     -0.023016 -0.0004666
## GroceryScore                0.006414  0.0167289
## factor(ville)Chicago        0.402795  0.9342657
## factor(ville)D            0.469209  1.1420588
## factor(ville)Montr       -0.132327  0.4503827
## factor(ville)New-York       -0.339011  0.2006416
## factor(ville)Philadelphie   0.426545  1.0911267
## factor(ville)Toronto        -0.227395  0.3079706
## factor(ville)Vancouver      -1.020539 -0.4280943
## educationC                0.992850  1.4015378
## educationCertificate/Diploma 0.987049  1.4380132
## educationGraduate School   -0.308187  0.0034599
## educationHigh School/Lower  1.120256  1.5891351
## occupation_statusStudent    -1.960597 -1.4314060
## occupation_statusUnemployed  0.464902  0.7945871
## household_income$100000-$149999 -0.226420  0.1828271
## household_income$150000+    -0.331150  0.1377965
## household_income$50000-$99999 -0.346984 -0.0199879
## household_incomeMissing     -1.119041 -0.6639695

lm3 <- lm(bmi ~ WalkScore + DiningandDrinkingScore + GroceryScore + factor(ville) + education + occupation_status + household_income + language + gender + marital_status + ethnicity, data = ibiccs)
summary(lm3)
```

```
##
## Call:
## lm(formula = bmi ~ WalkScore + DiningandDrinkingScore + GroceryScore +
##     factor(ville) + education + occupation_status + household_income +
##     language + gender + marital_status + ethnicity, data = ibiccs)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -11.126   -3.141   -0.821    2.360   17.148
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    27.30407     0.26097  104.63 < 2e-16
## WalkScore      -0.01643     0.00793   -2.07  0.03834
## DiningandDrinkingScore -0.00320     0.00609   -0.53  0.59949
## GroceryScore     0.01110     0.00275    4.03  5.5e-05
## factor(ville)Chicago  0.49302     0.14094    3.50  0.00047
## factor(ville)D         0.62159     0.18022    3.45  0.00056
## factor(ville)Montr       0.01985     0.19880    0.10  0.92045
## factor(ville)New-York  -0.14329     0.14257   -1.01  0.31490
## factor(ville)Philadelphie 0.57673     0.17592    3.28  0.00105
## factor(ville)Toronto    0.07630     0.14441    0.53  0.59723
## factor(ville)Vancouver  -0.50180     0.16154   -3.11  0.00190
## educationC             0.95751     0.11192    8.56 < 2e-16
## educationCertificate/Diploma 0.99254     0.12359    8.03  1.0e-15
## educationGraduate School -0.17618     0.08251   -2.14  0.03275
## educationHigh School/Lower 0.98075     0.12989    7.55  4.6e-14
## occupation_statusStudent -1.23730     0.13869   -8.92 < 2e-16
## occupation_statusUnemployed 0.56095     0.09220    6.08  1.2e-09
## household_income$100000-$149999 -0.22411     0.11350   -1.97  0.04834
```



```

## household_income$150000+ -0.33973 0.12982 -2.62 0.00888
## household_income$50000-$99999 -0.23013 0.09011 -2.55 0.01067
## household_incomeMissing -0.74197 0.12889 -5.76 8.7e-09
## languageFren/Span 0.03894 0.20376 0.19 0.84846
## genderMale 1.45267 0.06859 21.18 < 2e-16
## marital_statusSingle -0.34171 0.07605 -4.49 7.0e-06
## ethnicityAsian -3.23678 0.17403 -18.60 < 2e-16
## ethnicityCaucasian -1.67354 0.14714 -11.37 < 2e-16
## ethnicityHispanic -1.22842 0.22453 -5.47 4.5e-08
## ethnicityNative American/Indigenous -1.73431 0.64279 -2.70 0.00698
## ethnicityOther -2.02624 0.23340 -8.68 < 2e-16
##
## (Intercept) ***
## WalkScore *
## DiningandDrinkingScore
## GroceryScore ***
## factor(ville)Chicago ***
## factor(ville)Détroit ***
## factor(ville)Montréal
## factor(ville)New-York
## factor(ville)Philadelphie **
## factor(ville)Toronto
## factor(ville)Vancouver **
## educationCégep ***
## educationCertificate/Diploma ***
## educationGraduate School *
## educationHigh School/Lower ***
## occupation_statusStudent ***
## occupation_statusUnemployed ***
## household_income$100000-$149999 *
## household_income$150000+ **
## household_income$50000-$99999 *
## household_incomeMissing ***
## languageFren/Span
## genderMale ***
## marital_statusSingle ***
## ethnicityAsian ***
## ethnicityCaucasian ***
## ethnicityHispanic ***
## ethnicityNative American/Indigenous **
## ethnicityOther ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.47 on 17820 degrees of freedom
## (6052 observations deleted due to missingness)
## Multiple R-squared: 0.0927, Adjusted R-squared: 0.0913
## F-statistic: 65 on 28 and 17820 DF, p-value: <2e-16
confint(lm3)

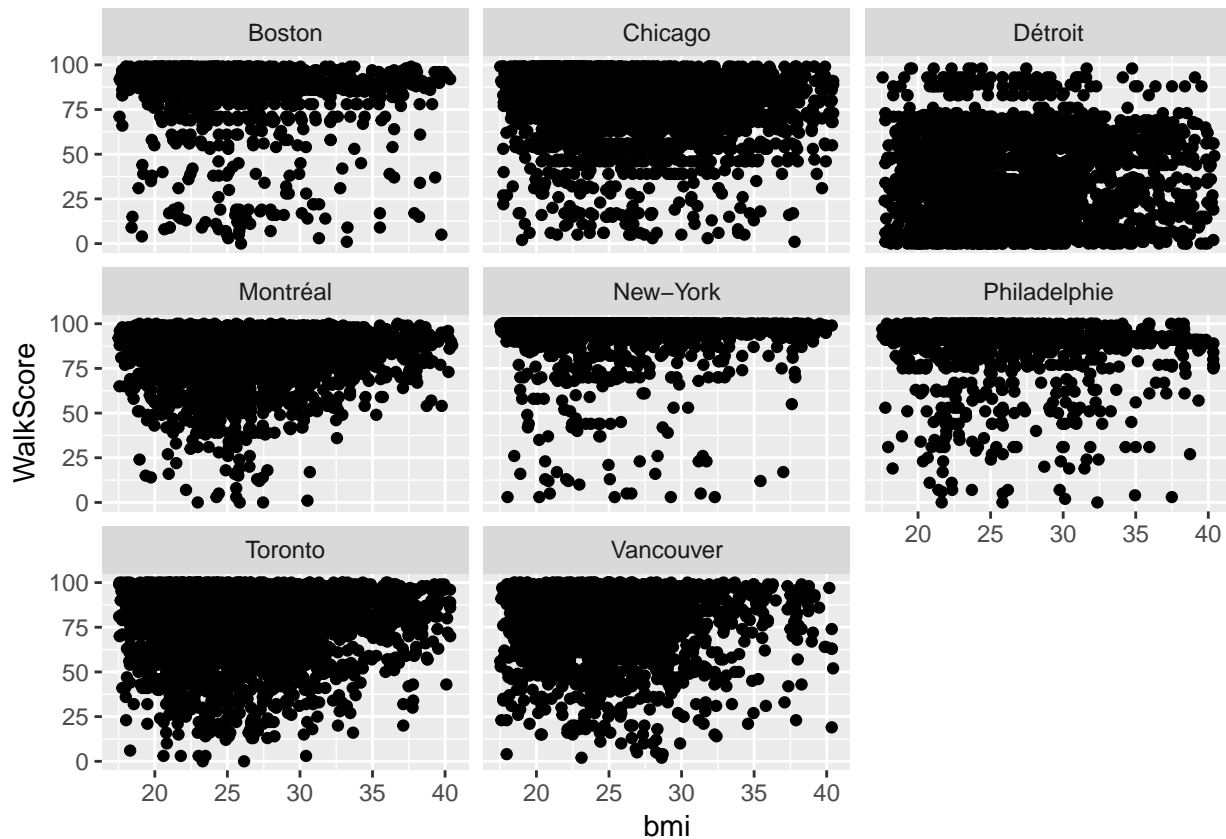
## 2.5 % 97.5 %
## (Intercept) 26.792544 27.8155918
## WalkScore -0.031970 -0.0008823
## DiningandDrinkingScore -0.015142 0.0087429

```

## GroceryScore	0.005707	0.0164928
## factor(ville)Chicago	0.216752	0.7692812
## factor(ville)Détroit	0.268343	0.9748362
## factor(ville)Montréal	-0.369811	0.4095176
## factor(ville)New-York	-0.422745	0.1361683
## factor(ville)Philadelphie	0.231905	0.9215589
## factor(ville)Toronto	-0.206745	0.3593518
## factor(ville)Vancouver	-0.818441	-0.1851673
## educationCégep	0.738128	1.1768826
## educationCertificate/Diploma	0.750296	1.2347745
## educationGraduate School	-0.337902	-0.0144561
## educationHigh School/Lower	0.726141	1.2353511
## occupation_statusStudent	-1.509155	-0.9654525
## occupation_statusUnemployed	0.380220	0.7416748
## household_income\$100000-\$149999	-0.446582	-0.0016357
## household_income\$150000+	-0.594179	-0.0852755
## household_income\$50000-\$99999	-0.406762	-0.0534958
## household_incomeMissing	-0.994620	-0.4893275
## languageFren/Span	-0.360446	0.4383183
## genderMale	1.318228	1.5871178
## marital_statusSingle	-0.490772	-0.1926575
## ethnicityAsian	-3.577904	-2.8956639
## ethnicityCaucasian	-1.961941	-1.3851394
## ethnicityHispanic	-1.668516	-0.7883207
## ethnicityNative American/Indigenous	-2.994243	-0.4743765
## ethnicityOther	-2.483733	-1.5687506

## Scatter Plots for BMI & Variables

```
#BMI & WalkScore
ggplot(ibiccs, aes(x = bmi, y = WalkScore)) +
  geom_point() +
  facet_wrap(~ ville)
```



## Complete cases

```
ibiccs_cc <- ibiccs %>%
  drop_na("bmi", "GroceryScore", "WalkScore", "DiningandDrinkingScore")
```

We remove missing data for all variables. We go from 23901 observations to 20922 observations.

## Causal Mediation

### GroceryScore

```
med.fit <- glm(GroceryScore ~ WalkScore, family = gaussian(link = "identity"), data = ibiccs_cc)
summary(med.fit)
```

```
##
## Call:
## glm(formula = GroceryScore ~ WalkScore, family = gaussian(link = "identity"),
##      data = ibiccs_cc)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -66.59   -3.83   -0.08    5.60   63.41
```

```
##
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.96590    0.30735   -3.14  0.0017 **
## WalkScore    1.04793    0.00375  279.45  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 180.5)
##
## Null deviance: 17875112  on 20921  degrees of freedom
## Residual deviance:  3776883  on 20920  degrees of freedom
## AIC: 168088
##
## Number of Fisher Scoring iterations: 2
out.fit <- glm(bmi ~ WalkScore * GroceryScore, family = gaussian(link = "identity"), data = ibiccs_cc)
summary(out.fit)

##
## Call:
## glm(formula = bmi ~ WalkScore * GroceryScore, family = gaussian(link = "identity"),
##      data = ibiccs_cc)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -9.844  -3.458  -0.797   2.567  15.301
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    26.5684245  0.1547173  171.72 < 2e-16 ***
## WalkScore      -0.0202338  0.0040430   -5.00 5.6e-07 ***
## GroceryScore     0.0290941  0.0033992    8.56 < 2e-16 ***
## WalkScore:GroceryScore -0.0002451  0.0000442   -5.55 2.9e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 21.83)
##
## Null deviance: 463192  on 20921  degrees of freedom
## Residual deviance: 456602  on 20918  degrees of freedom
## AIC: 123887
##
## Number of Fisher Scoring iterations: 2
med.out <- mediate(med.fit, out.fit, treat = "WalkScore", mediator = "GroceryScore", sims = 100)
summary(med.out)

##
## Causal Mediation Analysis
##
## Quasi-Bayesian Confidence Intervals
##
##              Estimate 95% CI Lower 95% CI Upper p-value
## ACME (control)    0.030738    0.024893    0.04 <2e-16 ***
```

```
## ACME (treated)          0.030478      0.024677      0.04 <2e-16 ***
## ADE (control)          -0.019982     -0.028575     -0.01 <2e-16 ***
## ADE (treated)          -0.020242     -0.028805     -0.01 <2e-16 ***
## Total Effect           0.010496      0.000724      0.02  0.04 *
## Prop. Mediated (control) 2.710278      1.599247     16.62  0.04 *
## Prop. Mediated (treated) 2.686551      1.583944     16.50  0.04 *
## ACME (average)         0.030608      0.024785      0.04 <2e-16 ***
## ADE (average)          -0.020112     -0.028690     -0.01 <2e-16 ***
## Prop. Mediated (average) 2.698415      1.591596     16.56  0.04 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Sample Size Used: 20922
##
##
## Simulations: 100
```

## Step 2: Add all covariates *Focus*

```
med.fit <- glm(GroceryScore ~ WalkScore + ville + gender + health + physically_active + q42 + marital_s
summary(med.fit)
```

```
##
## Call:
## glm(formula = GroceryScore ~ WalkScore + ville + gender + health +
##      physically_active + q42 + marital_status + day_per_week_motor_vehicle +
##      day_per_week_public_transit + day_per_week_walking + day_per_week_bike +
##      children_household + common_transportation + ethnicity +
##      pa_level, family = gaussian(link = "identity"), data = ibiccs_cc)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -62.42   -4.19   -0.03    5.77   64.91
##
## Coefficients:
##
##              Estimate Std. Error t value
## (Intercept)    -2.03606     1.60125   -1.27
## WalkScore         1.10960     0.00714  155.42
## villeChicago     -5.26472     0.48525  -10.85
## villeDétroit      4.20804     0.63107    6.67
## villeMontréal     5.19837     0.54304    9.57
## villeNew-York    -1.47891     0.50618   -2.92
## villePhiladelphie 0.66852     0.61399    1.09
## villeToronto      1.35832     0.50924    2.67
## villeVancouver    2.93633     0.56379    5.21
## genderMale        0.05285     0.24984    0.21
## healthGood        1.03684     0.35356    2.93
## healthPoor/Fair   1.58766     0.51120    3.11
## healthVery Good   0.35585     0.31514    1.13
## physically_activeYes -1.53172     0.88539   -1.73
## q42              -0.00381     0.01050   -0.36
## marital_statusSingle 0.30554     0.27847    1.10
```

## day_per_week_motor_vehicle	0.19731	0.06642	2.97
## day_per_week_public_transit	0.01767	0.07817	0.23
## day_per_week_walking	-0.16560	0.06017	-2.75
## day_per_week_bike	0.13071	0.09290	1.41
## children_household	0.60600	0.17497	3.46
## common_transportationCar	-0.20703	0.67860	-0.31
## common_transportationOther	-0.11057	1.49468	-0.07
## common_transportationPublic Transportation	-0.56948	0.66703	-0.85
## common_transportationWalking	-0.89135	0.65753	-1.36
## ethnicityAsian	-3.65495	0.64422	-5.67
## ethnicityCaucasian	-4.03778	0.54736	-7.38
## ethnicityHispanic	-2.00917	0.83280	-2.41
## ethnicityNative American/Indigenous	-0.67417	2.25998	-0.30
## ethnicityOther	-4.25760	0.87012	-4.89
## pa_levelModerate	0.21310	0.49775	0.43
## pa_levelNA	-0.74982	0.94585	-0.79
## pa_levelSedentary	1.65715	3.55151	0.47
## pa_levelVigorous	-0.27657	0.50931	-0.54
##	Pr(> t )		
## (Intercept)	0.20356		
## WalkScore	< 2e-16	***	
## villeChicago	< 2e-16	***	
## villeDétroit	2.7e-11	***	
## villeMontréal	< 2e-16	***	
## villeNew-York	0.00349	**	
## villePhiladelphie	0.27626		
## villeToronto	0.00766	**	
## villeVancouver	1.9e-07	***	
## genderMale	0.83247		
## healthGood	0.00337	**	
## healthPoor/Fair	0.00190	**	
## healthVery Good	0.25884		
## physically_activeYes	0.08366	.	
## q42	0.71712		
## marital_statusSingle	0.27258		
## day_per_week_motor_vehicle	0.00298	**	
## day_per_week_public_transit	0.82119		
## day_per_week_walking	0.00593	**	
## day_per_week_bike	0.15945		
## children_household	0.00054	***	
## common_transportationCar	0.76031		
## common_transportationOther	0.94103		
## common_transportationPublic Transportation	0.39325		
## common_transportationWalking	0.17525		
## ethnicityAsian	1.4e-08	***	
## ethnicityCaucasian	1.7e-13	***	
## ethnicityHispanic	0.01586	*	
## ethnicityNative American/Indigenous	0.76547		
## ethnicityOther	1.0e-06	***	
## pa_levelModerate	0.66856		
## pa_levelNA	0.42794		
## pa_levelSedentary	0.64079		
## pa_levelVigorous	0.58712		
## ---			

```

## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 172.8)
##
## Null deviance: 10515687 on 12231 degrees of freedom
## Residual deviance: 2108402 on 12198 degrees of freedom
## (8690 observations deleted due to missingness)
## AIC: 97773
##
## Number of Fisher Scoring iterations: 2
out.fit <- glm(bmi ~ WalkScore * GroceryScore + ville + gender + health + physically_active + q42 + marital_status + day_per_week_motor_vehicle + day_per_week_public_transit + day_per_week_walking + day_per_week_bike + children_household + common_transportation + ethnicity + pa_level, family = gaussian(link = "identity"), data = ibiccs_cc)
summary(out.fit)

##
## Call:
## glm(formula = bmi ~ WalkScore * GroceryScore + ville + gender + health + physically_active + q42 + marital_status + day_per_week_motor_vehicle + day_per_week_public_transit + day_per_week_walking + day_per_week_bike + children_household + common_transportation + ethnicity + pa_level, family = gaussian(link = "identity"), data = ibiccs_cc)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -11.972   -2.855   -0.608    2.141   17.593
##
## Coefficients:
##              Estimate Std. Error t value
## (Intercept)      22.2641750   0.5246506   42.44
## WalkScore        -0.0095479   0.0051525   -1.85
## GroceryScore      0.0171813   0.0042736    4.02
## villeChicago      0.3211243   0.1562272    2.06
## villeDétroit      0.2195619   0.2012180    1.09
## villeMontréal    -0.4138707   0.1737665   -2.38
## villeNew-York    -0.0277660   0.1623542   -0.17
## villePhiladelphie 0.4169788   0.1954018    2.13
## villeToronto     -0.2999863   0.1625453   -1.85
## villeVancouver   -0.8963295   0.1799889   -4.98
## genderMale        1.3503195   0.0794868   16.99
## healthGood        2.5916722   0.1125020   23.04
## healthPoor/Fair   4.0801239   0.1626854   25.08
## healthVery Good   0.9746545   0.1002516    9.72
## physically_activeYes -0.1342122   0.2816550   -0.48
## q42               0.0517281   0.0033411   15.48
## marital_statusSingle 0.2120547   0.0885805    2.39
## day_per_week_motor_vehicle 0.0598396   0.0211683    2.83
## day_per_week_public_transit -0.0226448   0.0248632   -0.91
## day_per_week_walking -0.0664917   0.0191535   -3.47
## day_per_week_bike -0.0289107   0.0295544   -0.98
## children_household 0.3285687   0.0556898    5.90
## common_transportationCar 0.5233395   0.2159322    2.42
## common_transportationOther 0.3525731   0.4754329    0.74
## common_transportationPublic Transportation 0.5139321   0.2121792    2.42
## common_transportationWalking 0.1158420   0.2092267    0.55
## ethnicityAsian    -3.0633145   0.2053521  -14.92

```

```

## ethnicityCaucasian          -1.4440697  0.1747277  -8.26
## ethnicityHispanic           -0.7734630  0.2650995  -2.92
## ethnicityNative American/Indigenous -1.4654113  0.7190937  -2.04
## ethnicityOther              -1.6945561  0.2771437  -6.11
## pa_levelModerate            0.3856558  0.1583209   2.44
## pa_levelNA                  0.4683318  0.3008611   1.56
## pa_levelSedentary           1.3108716  1.1297077   1.16
## pa_levelVigorous            0.0363756  0.1620093   0.22
## WalkScore:GroceryScore      -0.0000906  0.0000558  -1.62
##                               Pr(>|t|)
## (Intercept)                  < 2e-16 ***
## WalkScore                     0.06390 .
## GroceryScore                  5.8e-05 ***
## villeChicago                  0.03985 *
## villeDétroit                  0.27522
## villeMontréal                 0.01725 *
## villeNew-York                 0.86421
## villePhiladelphie             0.03287 *
## villeToronto                  0.06498 .
## villeVancouver                6.4e-07 ***
## genderMale                    < 2e-16 ***
## healthGood                    < 2e-16 ***
## healthPoor/Fair               < 2e-16 ***
## healthVery Good               < 2e-16 ***
## physically_activeYes          0.63372
## q42                           < 2e-16 ***
## marital_statusSingle          0.01668 *
## day_per_week_motor_vehicle    0.00471 **
## day_per_week_public_transit   0.36243
## day_per_week_walking          0.00052 ***
## day_per_week_bike             0.32798
## children_household            3.7e-09 ***
## common_transportationCar       0.01538 *
## common_transportationOther     0.45835
## common_transportationPublic Transportation 0.01544 *
## common_transportationWalking  0.57982
## ethnicityAsian                < 2e-16 ***
## ethnicityCaucasian            < 2e-16 ***
## ethnicityHispanic              0.00353 **
## ethnicityNative American/Indigenous 0.04159 *
## ethnicityOther                1.0e-09 ***
## pa_levelModerate              0.01487 *
## pa_levelNA                    0.11958
## pa_levelSedentary             0.24592
## pa_levelVigorous              0.82235
## WalkScore:GroceryScore        0.10479
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 17.49)
##
## Null deviance: 260861 on 12231 degrees of freedom
## Residual deviance: 213274 on 12196 degrees of freedom
## (8690 observations deleted due to missingness)

```



```
## AIC: 69752
##
## Number of Fisher Scoring iterations: 2
med.out <- mediate(med.fit, out.fit, treat = "WalkScore", mediator = "GroceryScore", sims = 100)
summary(med.out)

##
## Causal Mediation Analysis
##
## Quasi-Bayesian Confidence Intervals
##
##           Estimate 95% CI Lower 95% CI Upper p-value
## ACME (control)      0.01906      0.00916      0.03 <2e-16 ***
## ACME (treated)      0.01896      0.00915      0.03 <2e-16 ***
## ADE (control)      -0.00874     -0.02220      0.00  0.10 .
## ADE (treated)      -0.00884     -0.02214      0.00  0.10 .
## Total Effect        0.01022     -0.00755      0.02  0.18
## Prop. Mediated (control) 1.64249    -2.59397      7.29  0.18
## Prop. Mediated (treated) 1.63167    -2.59824      7.27  0.18
## ACME (average)      0.01901      0.00916      0.03 <2e-16 ***
## ADE (average)      -0.00879     -0.02217      0.00  0.10 .
## Prop. Mediated (average) 1.63708    -2.59611      7.28  0.18
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Sample Size Used: 12232
##
##
## Simulations: 100
```

## DiningandDrinkingScore

```
med.fit <- glm(DiningandDrinkingScore ~ WalkScore, family = gaussian(link = "identity"), data = ibiccs_
summary(med.fit)

##
## Call:
## glm(formula = DiningandDrinkingScore ~ WalkScore, family = gaussian(link = "identity"),
##      data = ibiccs_cc)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -46.43   -2.30   -0.15    2.20   51.14
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  4.69628    0.13849   33.9    <2e-16 ***
## WalkScore    0.96181    0.00169  569.2    <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 36.66)
##
```

```

##      Null deviance: 12643236  on 20921  degrees of freedom
## Residual deviance:   766854  on 20920  degrees of freedom
## AIC: 134731
##
## Number of Fisher Scoring iterations: 2
out.fit <- glm(bmi ~ WalkScore * DiningandDrinkingScore, family = gaussian(link = "identity"), data = i
summary(out.fit)

##
## Call:
## glm(formula = bmi ~ WalkScore * DiningandDrinkingScore, family = gaussian(link = "identity"),
##      data = ibiccs_cc)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -9.052  -3.437  -0.773   2.539  15.392
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      25.9238288   0.1673918  154.87 < 2e-16
## WalkScore         0.0407126   0.0064922    6.27 3.7e-10
## DiningandDrinkingScore -0.0001460  0.0056775   -0.03  0.98
## WalkScore:DiningandDrinkingScore -0.0005065  0.0000475  -10.66 < 2e-16
##
## (Intercept)          ***
## WalkScore             ***
## DiningandDrinkingScore
## WalkScore:DiningandDrinkingScore ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 21.77)
##
##      Null deviance: 463192  on 20921  degrees of freedom
## Residual deviance: 455398  on 20918  degrees of freedom
## AIC: 123832
##
## Number of Fisher Scoring iterations: 2
med.out <- mediate(med.fit, out.fit, treat = "WalkScore", mediator = "DiningandDrinkingScore", sims = 1000)
summary(med.out)

##
## Causal Mediation Analysis
##
## Quasi-Bayesian Confidence Intervals
##
##              Estimate 95% CI Lower 95% CI Upper p-value
## ACME (control)      -0.0000167  -0.0099706      0.01   0.94
## ACME (treated)      -0.0005029  -0.0104558      0.01   0.84
## ADE (control)        0.0382083   0.0254589      0.05 <2e-16 ***
## ADE (treated)        0.0377222   0.0250133      0.05 <2e-16 ***
## Total Effect         0.0377055   0.0269357      0.05 <2e-16 ***
## Prop. Mediated (control) -0.0062384  -0.3129620      0.23   0.94

```

```
## Prop. Mediated (treated) -0.0190664    -0.3273799          0.22    0.84
## ACME (average)          -0.0002598    -0.0102132          0.01    0.92
## ADE (average)           0.0379652     0.0252361          0.05 <2e-16 ***
## Prop. Mediated (average) -0.0126524    -0.3201709          0.22    0.92
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Sample Size Used: 20922
##
##
## Simulations: 100
```

## Step 2: Add all covariates *Focus*

```
med.fit <- glm(DiningandDrinkingScore ~ WalkScore + ville + gender + health + physically_active + q42 +
summary(med.fit)
```

```
##
## Call:
## glm(formula = DiningandDrinkingScore ~ WalkScore + ville + gender +
##     health + physically_active + q42 + marital_status + day_per_week_motor_vehicle +
##     day_per_week_public_transit + day_per_week_walking + day_per_week_bike +
##     children_household + common_transportation + ethnicity +
##     pa_level, family = gaussian(link = "identity"), data = ibiccs_cc)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -43.69   -2.20   -0.30    1.94   49.51
##
## Coefficients:
##
##              Estimate Std. Error t value
## (Intercept)   -0.03762    0.71122  -0.05
## WalkScore       0.99074    0.00317 312.44
## villeChicago    2.38884    0.21553  11.08
## villeDétroit    4.43658    0.28030  15.83
## villeMontréal   0.41079    0.24120   1.70
## villeNew-York   0.19733    0.22483   0.88
## villePhiladelphie 1.49760    0.27272   5.49
## villeToronto    1.97616    0.22619   8.74
## villeVancouver  2.12289    0.25042   8.48
## genderMale     -0.06918    0.11097  -0.62
## healthGood     -0.26165    0.15704  -1.67
## healthPoor/Fair  0.14426    0.22706   0.64
## healthVery Good -0.22549    0.13997  -1.61
## physically_activeYes 0.02698    0.39326   0.07
## q42            -0.02362    0.00467  -5.06
## marital_statusSingle -0.19618    0.12369  -1.59
## day_per_week_motor_vehicle -0.03979    0.02950  -1.35
## day_per_week_public_transit -0.01697    0.03472  -0.49
## day_per_week_walking -0.00712    0.02673  -0.27
## day_per_week_bike -0.00800    0.04126  -0.19
## children_household -0.30141    0.07772  -3.88
```

```

## common_transportationCar -0.00204 0.30141 -0.01
## common_transportationOther -0.16017 0.66389 -0.24
## common_transportationPublic Transportation 0.34185 0.29627 1.15
## common_transportationWalking 0.02380 0.29206 0.08
## ethnicityAsian 2.74359 0.28614 9.59
## ethnicityCaucasian 2.36625 0.24312 9.73
## ethnicityHispanic 1.96747 0.36990 5.32
## ethnicityNative American/Indigenous 3.37670 1.00381 3.36
## ethnicityOther 2.76221 0.38648 7.15
## pa_levelModerate -0.09570 0.22108 -0.43
## pa_levelNA -0.02652 0.42012 -0.06
## pa_levelSedentary -0.55193 1.57747 -0.35
## pa_levelVigorous -0.00288 0.22622 -0.01
## Pr(>|t|)
## (Intercept) 0.95782
## WalkScore < 2e-16 ***
## villeChicago < 2e-16 ***
## villeDétroit < 2e-16 ***
## villeMontréal 0.08857 .
## villeNew-York 0.38013
## villePhiladelphie 4.1e-08 ***
## villeToronto < 2e-16 ***
## villeVancouver < 2e-16 ***
## genderMale 0.53303
## healthGood 0.09572 .
## healthPoor/Fair 0.52523
## healthVery Good 0.10722
## physically_activeYes 0.94531
## q42 4.2e-07 ***
## marital_statusSingle 0.11274
## day_per_week_motor_vehicle 0.17749
## day_per_week_public_transit 0.62504
## day_per_week_walking 0.78987
## day_per_week_bike 0.84624
## children_household 0.00011 ***
## common_transportationCar 0.99461
## common_transportationOther 0.80936
## common_transportationPublic Transportation 0.24859
## common_transportationWalking 0.93505
## ethnicityAsian < 2e-16 ***
## ethnicityCaucasian < 2e-16 ***
## ethnicityHispanic 1.1e-07 ***
## ethnicityNative American/Indigenous 0.00077 ***
## ethnicityOther 9.4e-13 ***
## pa_levelModerate 0.66510
## pa_levelNA 0.94967
## pa_levelSedentary 0.72643
## pa_levelVigorous 0.98985
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 34.1)
##
## Null deviance: 7447212 on 12231 degrees of freedom

```

```

## Residual deviance: 415956 on 12198 degrees of freedom
## (8690 observations deleted due to missingness)
## AIC: 77919
##
## Number of Fisher Scoring iterations: 2
out.fit <- glm(bmi ~ WalkScore * DiningandDrinkingScore + ville + gender + health + physically_active +
summary(out.fit)

##
## Call:
## glm(formula = bmi ~ WalkScore * DiningandDrinkingScore + ville +
##      gender + health + physically_active + q42 + marital_status +
##      day_per_week_motor_vehicle + day_per_week_public_transit +
##      day_per_week_walking + day_per_week_bike + children_household +
##      common_transportation + ethnicity + pa_level, family = gaussian(link = "identity"),
##      data = ibiccs_cc)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -11.849   -2.858   -0.615    2.134   17.638
##
## Coefficients:
##
##              Estimate Std. Error t value
## (Intercept)      21.8651080   0.5359572  40.80
## WalkScore          0.0229981   0.0083463   2.76
## DiningandDrinkingScore
## villeChicago      0.0007473   0.0069495   0.11
## villeDétroit      0.2888759   0.1552512   1.86
## villeMontréal     -0.3728092   0.1732538  -2.15
## villeNew-York      0.0082895   0.1629227   0.05
## villePhiladelphie  0.4537656   0.1957420   2.32
## villeToronto     -0.2763245   0.1627472  -1.70
## villeVancouver   -0.8465175   0.1800289  -4.70
## genderMale         1.3527104   0.0795035  17.01
## healthGood         2.5954334   0.1125264  23.07
## healthPoor/Fair    4.0865021   0.1627307  25.11
## healthVery Good    0.9714643   0.1003007   9.69
## physically_activeYes
## q42                0.0512740   0.0033462  15.32
## marital_statusSingle
## day_per_week_motor_vehicle
## day_per_week_public_transit
## day_per_week_walking
## day_per_week_bike
## children_household
## common_transportationCar
## common_transportationOther
## common_transportationPublic Transportation
## common_transportationWalking
## ethnicityAsian    -3.0315261   0.2066587 -14.67
## ethnicityCaucasian
## ethnicityHispanic
## ethnicityNative American/Indigenous
## ethnicityOther    -1.6746712   0.2779898  -6.02

```

```

## pa_levelModerate          0.3862290  0.1583640   2.44
## pa_levelNA                0.4654172  0.3009417   1.55
## pa_levelSedentary         1.3225545  1.1299697   1.17
## pa_levelVigorous          0.0392605  0.1620632   0.24
## WalkScore:DiningandDrinkingScore -0.0002147  0.0000618  -3.47
##                           Pr(>|t|)
## (Intercept)               < 2e-16 ***
## WalkScore                 0.00587 **
## DiningandDrinkingScore    0.91436
## villeChicago              0.06281 .
## villeDétroit              0.07102 .
## villeMontréal             0.03143 *
## villeNew-York             0.95942
## villePhiladelphie         0.02046 *
## villeToronto              0.08956 .
## villeVancouver            2.6e-06 ***
## genderMale                 < 2e-16 ***
## healthGood                 < 2e-16 ***
## healthPoor/Fair           < 2e-16 ***
## healthVery Good           < 2e-16 ***
## physically_activeYes       0.60031
## q42                        < 2e-16 ***
## marital_statusSingle       0.01672 *
## day_per_week_motor_vehicle 0.00573 **
## day_per_week_public_transit 0.36790
## day_per_week_walking       0.00056 ***
## day_per_week_bike          0.32515
## children_household         3.9e-09 ***
## common_transportationCar    0.01930 *
## common_transportationOther  0.44520
## common_transportationPublic Transportation 0.01650 *
## common_transportationWalking 0.56497
## ethnicityAsian             < 2e-16 ***
## ethnicityCaucasian         8.9e-16 ***
## ethnicityHispanic          0.00536 **
## ethnicityNative American/Indigenous 0.05701 .
## ethnicityOther             1.7e-09 ***
## pa_levelModerate          0.01475 *
## pa_levelNA                 0.12200
## pa_levelSedentary          0.24185
## pa_levelVigorous           0.80859
## WalkScore:DiningandDrinkingScore 0.00051 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 17.5)
##
## Null deviance: 260861 on 12231 degrees of freedom
## Residual deviance: 213389 on 12196 degrees of freedom
## (8690 observations deleted due to missingness)
## AIC: 69759
##
## Number of Fisher Scoring iterations: 2

```

```
med.out <- mediate(med.fit, out.fit, treat = "WalkScore", mediator = "DiningandDrinkingScore", sims = 1000)
summary(med.out)
```

```
##
## Causal Mediation Analysis
##
## Quasi-Bayesian Confidence Intervals
##
##           Estimate 95% CI Lower 95% CI Upper p-value
## ACME (control)      0.000612   -0.011790      0.01    0.94
## ACME (treated)      0.000402   -0.011958      0.01    0.96
## ADE (control)       0.022205    0.004622      0.04 <2e-16 ***
## ADE (treated)       0.021996    0.004474      0.04 <2e-16 ***
## Total Effect        0.022607    0.006406      0.04    0.02 *
## Prop. Mediated (control) 0.037019  -0.967764      0.65    0.92
## Prop. Mediated (treated) 0.027902  -0.979982      0.64    0.94
## ACME (average)      0.000507   -0.011874      0.01    0.94
## ADE (average)       0.022100    0.004548      0.04 <2e-16 ***
## Prop. Mediated (average) 0.032460  -0.973873      0.64    0.92
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Sample Size Used: 12232
##
##
## Simulations: 100
```

### List of next steps

- Add in the city level model with glmer and interpretation
- Income stratified model
  - Rerun the model for categories of income
  - Create a new dataframe with only one level of income

## Area-Level Analysis

```
med.fit <- lm(smorale ~ free + #area-covariates, data = school)
out.fit <- lmer(late ~ free + smorale + #area-covariates + #covariates + (1 | SCH_ID), data = student)
med.out <- mediate(med.fit, out.fit, treat = "free", mediator = "smorale", + control.value = 3, treat.value =
4, sims = 100) summary(med.out) ““
```