

Social Inclusion Analysis

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```
dat <- read_csv("~/urbanplanning/preliminary_data.csv")

## Rows: 169989 Columns: 26

## -- Column specification -----
## Delimiter: ","
## chr (9): education, gender, ethnicity, marriage, diabete_hypertension, rece...
## dbl (17): participant, like.current.city, natives.like.me, natives.lookdown....

##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.

dat$migration.type <- NA
dat$migration.type[dat$migration.self == 1] <- "self"
dat$migration.type[dat$migration.fellows == 1] <- "fellows"
dat$migration.type[dat$migration.relatives == 1] <- "relatives"

col_names <- c("ethnicity", "marriage", "diabete_hypertension", "group", "recent_disease", "insuranced", "gender")
dat[, col_names] <- lapply(dat[, col_names], factor)

dat$money.left <- dat$income - dat$expenche
dat$participant <- as.character(dat$participant)

dat$health <- as.numeric(dat$health)*1000

head(dat)

## # A tibble: 6 x 28
##   participant like.current.city natives.like.me natives.lookdown~ native.customs.~
##   <chr>          <dbl>          <dbl>          <dbl>          <dbl>
## 1 0              1             -1              1             -1
## 2 1              2              1              1             -1
## 3 2              2              1              1              1
## 4 3              1              1             -1             -1
## 5 4              1              1             -2              1
## 6 5              2              1             -2              2
## # ... with 23 more variables: i.am.native <dbl>, income <dbl>, education <chr>,
## #   gender <fct>, ethnicity <fct>, marriage <fct>, worked_before5.1 <fct>,
## #   diabete_hypertension <fct>, recent_disease <fct>, group <fct>,
## #   participated_in_group_activity <dbl>, insuranced <fct>,
## #   migration.self <dbl>, migration.relatives <dbl>, migration.fellows <dbl>,
## #   health <dbl>, migration_scale <chr>, expence <dbl>, job <chr>,
## #   willing.to.movein <dbl>, willing.to.stay <dbl>, migration.type <chr>, ...
```

```

dat$natives_inclusion <- dat$natives.like.me-dat$natives.lookdown.me
dat$city_inclusion <- dat$like.current.city-dat$native.customs.better+dat$i.am.native
dat$tendency.livehere <- dat$willing.to.movein + dat$willing.to.stay
dat$migration.companionship <- - dat$migration.self + dat$migration.relative + dat$migration.fellows
# dat$loneliness.level <- (dat$loneliness.level-min(dat$loneliness.level, na.rm=TRUE))/(max(dat$loneliness.level, na.rm=TRUE))

dat

## # A tibble: 169,989 x 32
##   participant like.current.city natives.like.me natives.lookdown.me
##   <chr>          <dbl>          <dbl>          <dbl>
## 1 0              1            -1             1
## 2 1              2             1             1
## 3 2              2             1             1
## 4 3              1             1            -1
## 5 4              1             1            -2
## 6 5              2             1            -2
## 7 6              1             1            -2
## 8 7              1             1            -2
## 9 8              2             1            -1
##10 9              2             2            -2
## # ... with 169,979 more rows, and 28 more variables:
## #   native.customs.better <dbl>, i.am.native <dbl>, income <dbl>,
## #   education <chr>, gender <fct>, ethnicity <fct>, marriage <fct>,
## #   worked_before5.1 <fct>, diabete_hypertension <fct>, recent_disease <fct>,
## #   group <fct>, participated_in_group_activity <dbl>, insured <fct>,
## #   migration.self <dbl>, migration.relative <dbl>, migration.fellows <dbl>,
## #   health <dbl>, migration_scale <chr>, expence <dbl>, job <chr>, ...

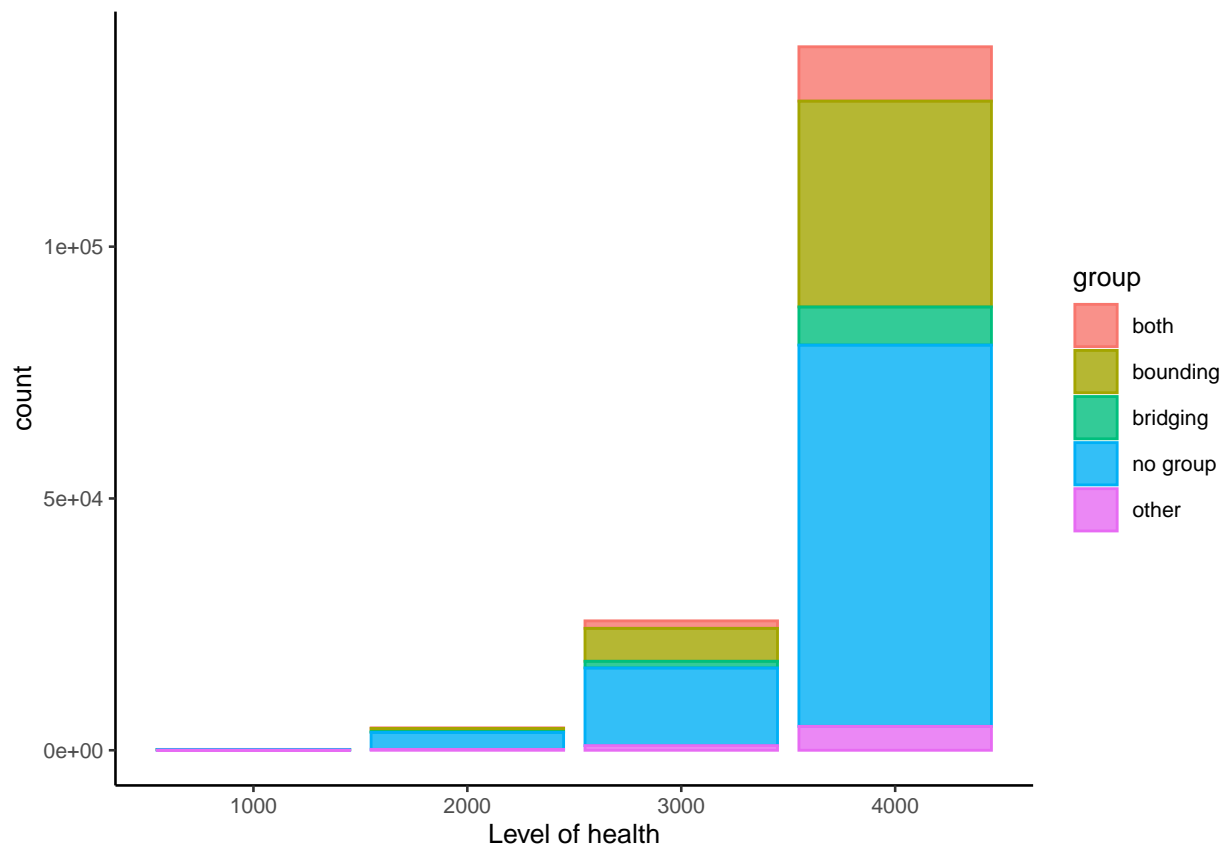
## regroup education
dat$education.group <- NA
dat$education.group[dat$education == 0] <- "0"
dat$education.group[(dat$education == 1) || (dat$education == 2)] <- "1, 2"
dat$education.group[dat$education == 3] <- "3"
dat$education.group[(dat$education == 4) || (dat$education == 5)] <- "4, 5"
dat$education.group[dat$education == 6] <- "6"

## regroup ethnicity
dat$ethnicity.group <- "other"
dat$ethnicity.group[dat$ethnicity == 1] <- "han"

cbPalette <- c("#e61212", "#ffb300", "#22ff00", "#0015ff", "#00bfff")

p1 <- ggplot(dat, aes(health, colour = group, fill=group)) +
  geom_bar(alpha=0.8) +
  labs(
    x = "Level of health",
    colour = "group") +
  theme_classic(base_size = 10)
p1

```



```
m0 <- lmer(health ~ group + like.current.city + natives.like.me + natives.lookdown.me + native.customs.l
data=dat)
```

```
## boundary (singular) fit: see ?isSingular
```

```
m1 <- lmer(health ~ group + natives_inclusion + city_inclusion + tendency.livehere + (1|participated_in
+ (1|migration.type) + (1|money.left) + (1|job) + (1|migration_scale)+ (1| education) + (1
data=dat)
```

```
## boundary (singular) fit: see ?isSingular
```

```
screenreg(c(m0,m1))
```

```
##
## =====
##                                     Model 1      Model 2
## -----
## (Intercept)                      3642.59 ***    3665.86 ***
##                                     (81.47)      (80.09)
## groupbounding                     -3.21          -2.40
##                                     (4.37)      (4.38)
## groupbridging                    -21.20 ***    -21.05 ***
##                                     (5.82)      (5.83)
## groupno group                    -26.42 ***    -25.89 ***
##                                     (4.33)      (4.33)
## groupother                       -29.23 ***    -30.01 ***
##                                     (7.03)      (7.05)
## like.current.city                 18.60 ***
##                                     (1.80)
```

```

## natives.like.me          3.04
##                          (1.57)
## natives.lookdown.me     -14.81 ***
##                          (1.09)
## native.customs.better   -1.76 *
##                          (0.85)
## i.am.native             -2.06
##                          (1.06)
## willing.to.movein       2.80 *
##                          (1.32)
## willing.to.stay         25.97 ***
##                          (2.45)
## natives_inclusion                11.45 ***
##                                (0.80)
## city_inclusion                   2.51 ***
##                                (0.60)
## tendency.livehere              9.13 ***
##                                (1.05)
## -----
## AIC                2070219.01      2064382.89
## BIC                2070455.37      2064579.79
## Log Likelihood     -1035085.51     -1032171.44
## Num. obs.          139831          139418
## Num. groups: money.left      1088          1082
## Num. groups: education        7              7
## Num. groups: marriage         6              6
## Num. groups: migration.companionship      4
## Num. groups: participated_in_group_activity 4
## Num. groups: recent_disease      3              3
## Num. groups: migration_scale      3              3
## Num. groups: job                3              3
## Num. groups: insured            2              2
## Num. groups: ethnicity.group      2              2
## Num. groups: gender              2              2
## Var: money.left (Intercept)    3916.30      4169.99
## Var: education (Intercept)    7513.81      7466.88
## Var: marriage (Intercept)     7063.23      7169.16
## Var: migration.companionship (Intercept)  856.61
## Var: participated_in_group_activity (Intercept) 171.61      168.06
## Var: recent_disease (Intercept) 11138.29     11060.88
## Var: migration_scale (Intercept)  282.66      274.87
## Var: job (Intercept)          605.20      618.57
## Var: insured (Intercept)       19.67      29.93
## Var: ethnicity.group (Intercept)  38.97      36.05
## Var: gender (Intercept)        0.00      0.00
## Var: Residual                157014.70     157312.69
## Num. groups: migration.type      3
## Var: migration.type (Intercept)      69.51
## =====
## *** p < 0.001; ** p < 0.01; * p < 0.05
write_csv(dat,"processed_data.csv")

```