

R Notebook

Code ▼

Hide

```
library(ggplot2)
library(tidyr)
library(knitr)
library(xtable)
```

Attaching package: 'xtable'

The following object is masked from 'package:arm':

```
display
```

Hide

```
library(dplyr)
library(AER)
```

```
Loading required package: car
Loading required package: carData
Registered S3 methods overwritten by 'car':
  method                      from
  influence.merMod             lme4
  cooks.distance.influence.merMod lme4
  dfbeta.influence.merMod      lme4
  dfbetas.influence.merMod     lme4

Attaching package: 'car'

The following object is masked from 'package:purrr':

  some

The following object is masked from 'package:dplyr':

  recode

The following object is masked from 'package:arm':

  logit

Loading required package: lmtest
Loading required package: zoo

Attaching package: 'zoo'

The following objects are masked from 'package:base':

  as.Date, as.Date.numeric

Loading required package: sandwich
Loading required package: survival
```

[Hide](#)

```
library(plm)
```

```
Attaching package: 'plm'

The following objects are masked from 'package:dplyr':

  between, lag, lead
```

[Hide](#)

```
library(multiwayvcov)
library(magrittr)
```

```
Attaching package: 'magrittr'
```

```
The following object is masked from 'package:purrr':
```

```
  set_names
```

```
The following object is masked from 'package:tidyr':
```

```
  extract
```

[Hide](#)

```
library(miceadds)
```

```
Loading required package: mice
```

```
Attaching package: 'mice'
```

```
The following object is masked from 'package:stats':
```

```
  filter
```

```
The following objects are masked from 'package:base':
```

```
  cbind, rbind
```

```
* miceadds 3.10-28 (2020-07-29 21:56:24)
```

[Hide](#)

```
library(car)  
library(lmtest)  
library(sandwich)  
library(psych)
```

```
Attaching package: 'psych'
```

```
The following object is masked from 'package:car':
```

```
  logit
```

```
The following objects are masked from 'package:ggplot2':
```

```
  %+%, alpha
```

```
The following objects are masked from 'package:arm':
```

```
  logit, rescale, sim
```

Hide

```
library(stargazer)
```

Please cite as:

Hlavac, Marek (2018). stargazer: Well-Formatted Regression and Summary Statistics Tables.
R package version 5.2.2. <https://CRAN.R-project.org/package=stargazer>

Hide

```
library(haven)  
library(Amelia)
```

```
Loading required package: Rcpp  
##  
## Amelia II: Multiple Imputation  
## (Version 1.7.6, built: 2019-11-24)  
## Copyright (C) 2005-2020 James Honaker, Gary King and Matthew Blackwell  
## Refer to http://gking.harvard.edu/amelia/ for more information  
##
```

Hide

```
library(mice)  
library(lattice)  
library(VIM)
```

```
Loading required package: colorspace  
VIM is ready to use.
```

Suggestions and bug-reports can be submitted at: <https://github.com/statistikat/VIM/issues>

Attaching package: 'VIM'

The following object is masked from 'package:datasets':

```
sleep
```

Hide

```
library(mitools)
library(miceadds)
library(ggplot2)
library(cobalt)
library(gridExtra)
library(tidyverse)
library(gtable)
library(grid)
```

Hide

```
#!/diagnostics suppress=<comma-separated list of variables>
```

Hide

```
dat <- read_dta("/Users/swimmingcircle/Documents/CS112_Decision/social_network/0422analysis.dta")
```

Hide

```
dat
```

id	address	region	village	takeup_survey	...	ag...	e...	rice_inc	ricearea_2010					
<dbl>	<chr>	<dbl>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>					
1111030	fusheng21	1	fusheng	1	76	5	1	40	5.00					
1111031	fusheng21	1	fusheng	0	54	4	0	100	10.00					
1111032	fusheng21	1	fusheng	1	67	2	2	60	10.00					
1111034	fusheng21	1	fusheng	1	74	2	1	30	0.74					
1111035	fusheng21	1	fusheng	1	60	6	1	50	11.00					
1111036	fusheng21	1	fusheng	0	56	6	1	90	13.00					
1111038	fusheng21	1	fusheng	1	61	13	1	30	4.00					
1111040	fusheng21	1	fusheng	1	40	4	2	100	30.00					
1111041	fusheng21	1	fusheng	0	63	3	1	70	2.00					
1111042	fusheng21	1	fusheng	1	57	9	2	NA	1.00					
1-10 of 4,902 rows 1-10 of 59 columns					Previous	1	2	3	4	5	6	...	100	Next

Descriptive statistics

Hide

```
panelA <- dat %>%
```

Warning messages:

```
1: Unknown or uninitialised column: `session`.
2: Unknown or uninitialised column: `session`.
3: Unknown or uninitialised column: `session`.
4: Unknown or uninitialised column: `session`.
```

Hide

```
select(male, age, agpop, educ, ricearea_2010, rice_inc, disaster_yes, disaster_loss, risk_averse, disaster_prob, understanding) %>%
  summarise_all(funs(n = sum(!is.na(.)), mean = mean(., na.rm = TRUE), sd = sd(., na.rm = TRUE))) %>%
  matrix(nrow = 11, ncol = 3) %>%
  set_rownames(c("male", "age", "agpop", "educ", "ricearea_2010", "rice_inc", "disaster_yes", "disaster_loss", "risk_averse", "disaster_prob", "understanding")) %>%
  set_colnames(c('n', 'mean', 'sd'))
```

`funs()` is deprecated as of dplyr 0.8.0.

Please use a list of either functions or lambdas:

```
# Simple named list:
list(mean = mean, median = median)

# Auto named with `tibble::lst()` :
tibble::lst(mean, median)

# Using lambdas
list(~ mean(., trim = .2), ~ median(., na.rm = TRUE))
```

This warning is displayed once every 8 hours.

Call `lifecycle::last_warnings()` to see where this warning was generated.

Hide

```
panelB <- dat %>%
  select(network_obs, network_rate_preintensive, network_twoside, network_second) %>%
  summarise_all(funs(n = sum(!is.na(.)), mean = mean(., na.rm = TRUE), sd = sd(., na.rm = TRUE))) %>%
  matrix(nrow = 4, ncol = 3) %>%
  set_rownames(c("network_obs", "network_rate_preintensive", "network_twoside", "network_second")) %>%
  set_colnames(c('n', 'mean', 'sd'))

panelC <- dat %>%
  select(indegree, path_out_ind, eigenvector) %>%
  summarise_all(funs(n = sum(!is.na(.)), mean = mean(., na.rm = TRUE), sd = sd(., na.rm = TRUE))) %>%
  matrix(nrow = 3, ncol = 3) %>%
  set_colnames(c('n', 'mean', 'sd')) %>%
  set_rownames(c("indegree", "path_out_ind", "eigenvector"))
```

Hide

```
panelA
```

	n	mean	sd
male	4894	0.9139763	0.2804277
age	4893	51.49438	12.03191
agpop	4893	4.915185	2.133418
educ	4841	1.191902	0.8533748
ricearea_2010	4883	13.62975	19.50735
rice_inc	4789	74.11861	27.67502
disaster_yes	4453	0.6328318	0.482087
disaster_loss	2518	27.50675	18.19906
risk_averse	4902	0.1888209	0.3133002
disaster_prob	4902	33.63282	16.61935
understanding	4723	0.4572518	0.3030129

[Hide](#)

```
print('')
```

```
[1] ""
```

[Hide](#)

```
panelB
```

	n	mean	sd
network_obs	4588	4.918265	0.433588
network_rate_preintensive	4588	0.1649738	0.1896223
network_twoside	4588	0.0416921	0.098993
network_second	4521	0.1677539	0.1172213

[Hide](#)

```
print('')
```

```
[1] ""
```

[Hide](#)

```
panelC
```

	n	mean	sd
indegree	4588	3.244182	1.912167
path_out_ind	4588	2.672834	0.8662739
eigenvector	4588	0.1437882	0.08318533

[Hide](#)

```
takeup_summary <- dat %>%
  select(takeup_survey) %>%
  summarise_all(funs(n = sum(!is.na(.)), mean = mean(., na.rm = TRUE), sd = sd(., na.rm
= TRUE))) %>%
  as.data.frame(row.names = c('n', 'mean', 'sd')) %>%
  mutate(session = "total")

takeup_summary
```

n	mean	sd	session
<int>	<dbl>	<dbl>	<chr>
4902	0.4455324	0.4970751	total

1 row

Hide

```
# matrix(takeup_summary[1,], nrow = 1, ncol = 3)
```

Hide

```
dat$session <- 0
dat$session[dat$delay == 0 & dat$intensive == 0] <- 11
dat$session[dat$delay == 0 & dat$intensive == 1] <- 12
dat$session[dat$delay == 1 & dat$intensive == 0] <- 21
dat$session[dat$delay == 1 & dat$intensive == 1] <- 22
dat$session <- as.factor(dat$session)
as.data.frame(table(dat$session))
```

Var1	Freq
<fctr>	<int>
11	1079
12	1096
21	1374
22	1353

4 rows

Hide

dat

id	address	region	village	takeup_survey	...	ag...	e...	rice_inc	ricearea_2010
<dbl>	<chr>	<dbl>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
1111030	fusheng21	1	fusheng	1	76	5	1	40	5.00
1111031	fusheng21	1	fusheng	0	54	4	0	100	10.00

id	address	region	village	takeup_survey	...	ag...	e...	rice_inc	ricearea_2010					
<dbl>	<chr>	<dbl>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>					
1111032	fusheng21	1	fusheng	1	67	2	2	60	10.00					
1111034	fusheng21	1	fusheng	1	74	2	1	30	0.74					
1111035	fusheng21	1	fusheng	1	60	6	1	50	11.00					
1111036	fusheng21	1	fusheng	0	56	6	1	90	13.00					
1111038	fusheng21	1	fusheng	1	61	13	1	30	4.00					
1111040	fusheng21	1	fusheng	1	40	4	2	100	30.00					
1111041	fusheng21	1	fusheng	0	63	3	1	70	2.00					
1111042	fusheng21	1	fusheng	1	57	9	2	NA	1.00					
1-10 of 4,902 rows 1-10 of 60 columns				Previous		1	2	3	4	5	6	...	100	Next

Hide

```
takeup_by_session <- dat %>%
  select(takeup_survey, info_none, session) %>%
  filter(info_none == 1) %>%
  group_by(session) %>%
  summarise(n = sum(!is.na(takeup_survey)), mean = mean(takeup_survey, na.rm = TRUE), sd
= sd(takeup_survey, na.rm = TRUE)) %>%
  data.frame()
```

`summarise()` ungrouping output (override with `.groups` argument)

Hide

```
panelD <- rbind(takeup_by_session, takeup_summary)
panelD
```

session	n	mean	sd
<fctr>	<int>	<dbl>	<dbl>
11	1079	0.3521779	0.4778706
12	1096	0.5036496	0.5002149
21	657	0.4429224	0.4971099
22	660	0.4651515	0.4991624
total	4902	0.4455324	0.4970751
5 rows			

Intensive Session Effect

Table 2 replication column 1

Hide

```
# Section 2: Intensive Session Effect -----
-
# With extension on bootstrap cluster standard errors
#The dataset is now called dat - SORRY!
#####Table 2 in the paper #####
###Table2, col 1 - intensive session effect
round1dat <- dat[dat$delay == 0, ]

lm_t2c1 <- lm(takeup_survey ~ intensive + male + age + agpop + ricearea_2010 + literacy
+
               village, data = round1dat)

lm_t2c1
```

Call:

```
lm(formula = takeup_survey ~ intensive + male + age + agpop +
    ricearea_2010 + literacy + village, data = round1dat)
```

Coefficients:

(Intercept)	intensive	male	age	agpop
0.3634212	0.1408242	0.0407053	0.0017669	-0.0029449
ricearea_2010	literacy	villagebeilian	villagebeixing	villagecaijia
0.0007886	0.0832050	-0.4123784	-0.1940895	-0.2424232
villagechaxi	villagedaqiao	villagedaxi	villagedayu	villagedazhou
-0.1673066	0.2331480	-0.2599335	-0.2345645	-0.2557939
villagedongan	villagedukou	villagefusheng	villagefuzhou	villagegangtou
-0.2849145	-0.1285052	0.0229908	0.0009335	-0.1539832
villagegangxia	villageguojia	villagehefeng	villagehelin	villagehongxing
0.1156739	-0.5002053	-0.4074481	-0.0584232	-0.2921655
villagehuangshan	villagejingang	villagejinggang	villagelianhe	villagelianqian
-0.1324631	-0.2456389	-0.2850252	-0.2555737	-0.1181014
villagelianxing	villagelongqing	villagelusikou	villagemazhou	villageminzhu
0.0228554	0.0943500	-0.1907178	0.0683045	-0.3818980
villageshigang	villagewanshang	villagexiabao	villagexianghu	villagexiaofang
-0.2715477	0.0914096	0.2450654	0.0920943	-0.2613101
villagexiashatou	villagexiecheng	villagexihe	villagexihu	villagexilian
-0.1738035	-0.1826537	-0.2841952	0.0587897	-0.4263741
villagexinguang	villagexingzeng	villagexinlian	villageyanjiang	villageyazhou
-0.1489744	-0.0932542	-0.3919875	-0.1611961	-0.4803191
villageyongfeng	villagezhangxi	villagezixi		
-0.0252228	-0.3062709	-0.2889457		

#Extension 1: Use matching to check the intensive session effect

Hide

```
#import matching library
library(Matching)
```

```
##  
## Matching (Version 4.9-7, Build Date: 2020-02-05)  
## See http://sekhon.berkeley.edu/matching for additional documentation.  
## Please cite software as:  
## Jasjeet S. Sekhon. 2011. ``Multivariate and Propensity Score Matching  
## Software with Automated Balance Optimization: The Matching package for R.''  
## Journal of Statistical Software, 42(7): 1-52.  
##
```

Hide

```
library(MatchIt)
```

```
Attaching package: 'MatchIt'
```

```
The following object is masked _by_ '.GlobalEnv':
```

```
lalonge
```

Hide

```
#Subset the variable we will be match on  
roundl1dat_NA <- roundl1dat %>% select(intensive, male,age, agpop, ricearea_2010, literac  
y, village, takeup_survey)  
  
#Find the columns that has NA  
colnames(roundl1dat_NA)[colSums(is.na(roundl1dat_NA)) > 0]
```

```
character(0)
```

Hide

```
#Drop NA  
roundl1dat<- roundl1dat_NA %>% drop_na()
```

Hide

```
#set the non-numerical variables as factors  
roundl1dat$village <- as.factor(roundl1dat$village)  
  
X <- cbind(roundl1dat$male, roundl1dat$age, roundl1dat$agpop, roundl1dat$ricearea_2010, roun  
dl1dat$literacy,roundl1dat$village)  
  
X<- as.data.frame(X) %>% rename(make=V1, age=V2, agpop = V3, ricearea_2010=V4, literacy=  
V5, village=V6)
```

Hide

```
#Genetic Matching
```

```
invisible(capture.output(genout1 <- GenMatch(Tr = roundldat$intensive, X=X , M=1, estimation = 'ATT', max.generations = 10)))
```

```
Loading required namespace: rgenoud
```

[Hide](#)

```
mout1 <- Match(Y = roundldat$takeup_survey, Tr= roundldat$intensive, X = X, M =1, estimation = 'ATT', Weight.matrix = genout1, exact = c(1,0,0,0,1,1), caliper = c(1e16, 1e16, 0.5, 0.3,1e16, 1e16 ))
```

```
mbgen1 <- MatchBalance(intensive~ male + age + agpop + ricearea_2010 + literacy + village, data = roundldat, match.out=mout1, nboots=500)
```

***** (V1) male *****

	Before Matching	After Matching
mean treatment.....	0.92208	0.96327
mean control.....	0.90746	0.96327
std mean diff.....	5.451	0
mean raw eQQ diff.....	0.014164	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.007309	0
med eCDF diff.....	0.007309	0
max eCDF diff.....	0.014618	0
var ratio (Tr/Co).....	0.85559	1
T-test p-value.....	0.22663	1

***** (V2) age *****

	Before Matching	After Matching
mean treatment.....	50.974	50.443
mean control.....	51.478	50.284
std mean diff.....	-4.147	1.3604
mean raw eQQ diff.....	0.60434	0.91589
med raw eQQ diff.....	1	1
max raw eQQ diff.....	6	14
mean eCDF diff.....	0.0087386	0.015088
med eCDF diff.....	0.0054993	0.01285
max eCDF diff.....	0.037233	0.03972
var ratio (Tr/Co).....	1.0444	1.195
T-test p-value.....	0.33269	0.62004
KS Bootstrap p-value..	0.314	0.396
KS Naive p-value.....	0.44942	0.50923
KS Statistic.....	0.037233	0.03972

***** (V3) agpop *****

	Before Matching	After Matching
mean treatment.....	4.8451	4.6043
mean control.....	4.9046	4.5924
std mean diff.....	-2.8457	0.74298
mean raw eQQ diff.....	0.090652	0.074766
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	1
mean eCDF diff.....	0.0050011	0.0067969
med eCDF diff.....	0.0022692	0.0035047
max eCDF diff.....	0.035327	0.023364

var ratio (Tr/Co).....	0.96673	1.0308
T-test p-value.....	0.51441	0.51515
KS Bootstrap p-value..	0.172	0.62
KS Naive p-value.....	0.51754	0.9736
KS Statistic.....	0.035327	0.023364

***** (V4) ricearea_2010 *****

	Before Matching	After Matching
mean treatment.....	14.853	11.391
mean control.....	13.719	11.32
std mean diff.....	4.395	0.8151
mean raw eQQ diff.....	1.1145	0.4128
med raw eQQ diff.....	0	0
max raw eQQ diff.....	445	6
mean eCDF diff.....	0.0053186	0.012663
med eCDF diff.....	0.0042309	0.010514
max eCDF diff.....	0.017511	0.036215
var ratio (Tr/Co).....	2.9867	1.0749
T-test p-value.....	0.21287	0.4743
KS Bootstrap p-value..	0.97	0.474
KS Naive p-value.....	0.99668	0.62848
KS Statistic.....	0.017511	0.036215

***** (V5) literacy *****

	Before Matching	After Matching
mean treatment.....	0.78479	0.85545
mean control.....	0.78093	0.85545
std mean diff.....	0.93911	0
mean raw eQQ diff.....	0.0037771	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0019306	0
med eCDF diff.....	0.0019306	0
max eCDF diff.....	0.0038612	0
var ratio (Tr/Co).....	0.98722	1
T-test p-value.....	0.82874	1

***** (V6) villagebeilian *****

	Before Matching	After Matching
mean treatment.....	0.0083488	0.0082938
mean control.....	0.01322	0.0082938
std mean diff.....	-5.3511	0
mean raw eQQ diff.....	0.0056657	0

med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0024356	0
med eCDF diff.....	0.0024356	0
max eCDF diff.....	0.0048712	0
var ratio (Tr/Co).....	0.63463	1
T-test p-value.....	0.27639	1

***** (V7) villagebeixing *****

	Before Matching	After Matching
mean treatment.....	0.018553	0.014218
mean control.....	0.016997	0.014218
std mean diff.....	1.1524	0
mean raw eQQ diff.....	0.00094429	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.00077785	0
med eCDF diff.....	0.00077785	0
max eCDF diff.....	0.0015557	0
var ratio (Tr/Co).....	1.0898	1
T-test p-value.....	0.7856	1

***** (V8) villagecaijia *****

	Before Matching	After Matching
mean treatment.....	0.012987	0.010664
mean control.....	0.014164	0.010664
std mean diff.....	-1.0394	0
mean raw eQQ diff.....	0.0018886	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.00058865	0
med eCDF diff.....	0.00058865	0
max eCDF diff.....	0.0011773	0
var ratio (Tr/Co).....	0.91796	1
T-test p-value.....	0.81424	1

***** (V9) villagechaxi *****

	Before Matching	After Matching
mean treatment.....	0.0064935	0.0035545
mean control.....	0.0084986	0.0035545
std mean diff.....	-2.4952	0
mean raw eQQ diff.....	0.0028329	0

med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0010025	0
med eCDF diff.....	0.0010025	0
max eCDF diff.....	0.0020051	0
var ratio (Tr/Co).....	0.7656	1
T-test p-value.....	0.5915	1

***** (V10) villagedaqiao *****

	Before Matching	After Matching
mean treatment.....	0.0055659	0.0035545
mean control.....	0.00661	0.0035545
std mean diff.....	-1.4028	0
mean raw eQQ diff.....	0.0018886	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.00052207	0
med eCDF diff.....	0.00052207	0
max eCDF diff.....	0.0010441	0
var ratio (Tr/Co).....	0.84291	1
T-test p-value.....	0.7566	1

***** (V11) villagedaxi *****

	Before Matching	After Matching
mean treatment.....	0.032468	0.030806
mean control.....	0.032106	0.030806
std mean diff.....	0.20402	0
mean raw eQQ diff.....	0	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	0	0
mean eCDF diff.....	0.00018089	0
med eCDF diff.....	0.00018089	0
max eCDF diff.....	0.00036177	0
var ratio (Tr/Co).....	1.0109	1
T-test p-value.....	0.96229	1

***** (V12) villagedayu *****

	Before Matching	After Matching
mean treatment.....	0.03525	0.035545
mean control.....	0.037771	0.035545
std mean diff.....	-1.3664	0
mean raw eQQ diff.....	0.0028329	0

med	raw eQQ diff.....	0	0
max	raw eQQ diff.....	1	0
mean	eCDF diff.....	0.0012605	0
med	eCDF diff.....	0.0012605	0
max	eCDF diff.....	0.002521	0
var ratio	(Tr/Co).....	0.93569	1
T-test	p-value.....	0.75622	1

***** (V13) villagedazhou *****

	Before Matching	After Matching
mean treatment.....	0.026902	0.029621
mean control.....	0.023607	0.029621
std mean diff.....	2.0353	0
mean raw eQQ diff.....	0.0028329	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0016472	0
med eCDF diff.....	0.0016472	0
max eCDF diff.....	0.0032945	0
var ratio (Tr/Co).....	1.1357	1
T-test p-value.....	0.62754	1

***** (V14) villagedongan *****

	Before Matching	After Matching
mean treatment.....	0.014842	0.007109
mean control.....	0.012276	0.007109
std mean diff.....	2.1215	0
mean raw eQQ diff.....	0.0018886	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0012833	0
med eCDF diff.....	0.0012833	0
max eCDF diff.....	0.0025666	0
var ratio (Tr/Co).....	1.2059	1
T-test p-value.....	0.60805	1

***** (V15) villagedukou *****

	Before Matching	After Matching
mean treatment.....	0.050093	0.054502
mean control.....	0.049103	0.054502
std mean diff.....	0.45356	0
mean raw eQQ diff.....	0.00094429	0

med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.00049492	0
med eCDF diff.....	0.00049492	0
max eCDF diff.....	0.00098984	0
var ratio (Tr/Co).....	1.0191	1
T-test p-value.....	0.91612	1

***** (V16) villagefusheng *****

	Before Matching	After Matching
mean treatment.....	0.038033	0.040284
mean control.....	0.035883	0.040284
std mean diff.....	1.1238	0
mean raw eQQ diff.....	0.0018886	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0010752	0
med eCDF diff.....	0.0010752	0
max eCDF diff.....	0.0021505	0
var ratio (Tr/Co).....	1.0575	1
T-test p-value.....	0.79229	1

***** (V17) villagefuzhou *****

	Before Matching	After Matching
mean treatment.....	0.033395	0.033175
mean control.....	0.03305	0.033175
std mean diff.....	0.19201	0
mean raw eQQ diff.....	0	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	0	0
mean eCDF diff.....	0.00017256	0
med eCDF diff.....	0.00017256	0
max eCDF diff.....	0.00034513	0
var ratio (Tr/Co).....	1.0101	1
T-test p-value.....	0.96452	1

***** (V18) villagegangtou *****

	Before Matching	After Matching
mean treatment.....	0.030612	0.030806
mean control.....	0.032106	0.030806
std mean diff.....	-0.86659	0
mean raw eQQ diff.....	0.0018886	0

med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.00074676	0
med eCDF diff.....	0.00074676	0
max eCDF diff.....	0.0014935	0
var ratio (Tr/Co).....	0.95494	1
T-test p-value.....	0.8431	1

***** (V19) villagegangxia *****

	Before Matching	After Matching
mean treatment.....	0.0092764	0.010664
mean control.....	0.011331	0.010664
std mean diff.....	-2.1426	0
mean raw eQQ diff.....	0.0028329	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0010275	0
med eCDF diff.....	0.0010275	0
max eCDF diff.....	0.002055	0
var ratio (Tr/Co).....	0.82033	1
T-test p-value.....	0.63845	1

***** (V20) villageguojia *****

	Before Matching	After Matching
mean treatment.....	0.010204	0.0082938
mean control.....	0.012276	0.0082938
std mean diff.....	-2.0604	0
mean raw eQQ diff.....	0.0028329	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0010358	0
med eCDF diff.....	0.0010358	0
max eCDF diff.....	0.0020717	0
var ratio (Tr/Co).....	0.83297	1
T-test p-value.....	0.65	1

***** (V21) villagehefeng *****

	Before Matching	After Matching
mean treatment.....	0.040816	0.043839
mean control.....	0.043437	0.043839
std mean diff.....	-1.324	0
mean raw eQQ diff.....	0.0028329	0

med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0013104	0
med eCDF diff.....	0.0013104	0
max eCDF diff.....	0.0026209	0
var ratio (Tr/Co).....	0.94222	1
T-test p-value.....	0.76315	1

***** (V22) villagehelin *****

	Before Matching	After Matching
mean treatment.....	0.018553	0.015403
mean control.....	0.014164	0.015403
std mean diff.....	3.2507	0
mean raw eQQ diff.....	0.0037771	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0021943	0
med eCDF diff.....	0.0021943	0
max eCDF diff.....	0.0043886	0
var ratio (Tr/Co).....	1.304	1
T-test p-value.....	0.42389	1

***** (V23) villagehongxing *****

	Before Matching	After Matching
mean treatment.....	0.037106	0.040284
mean control.....	0.037771	0.040284
std mean diff.....	-0.35204	0
mean raw eQQ diff.....	0.00094429	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.00033287	0
med eCDF diff.....	0.00033287	0
max eCDF diff.....	0.00066573	0
var ratio (Tr/Co).....	0.98304	1
T-test p-value.....	0.93544	1

***** (V24) villagehuangshan *****

	Before Matching	After Matching
mean treatment.....	0.038033	0.046209
mean control.....	0.036827	0.046209
std mean diff.....	0.63031	0
mean raw eQQ diff.....	0.00094429	0

med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0006031	0
med eCDF diff.....	0.0006031	0
max eCDF diff.....	0.0012062	0
var ratio (Tr/Co).....	1.0314	1
T-test p-value.....	0.88329	1

***** (V25) villagejingang *****

	Before Matching	After Matching
mean treatment.....	0.030612	0.035545
mean control.....	0.028329	0.035545
std mean diff.....	1.325	0
mean raw eQQ diff.....	0.0018886	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0011418	0
med eCDF diff.....	0.0011418	0
max eCDF diff.....	0.0022836	0
var ratio (Tr/Co).....	1.0781	1
T-test p-value.....	0.75507	1

***** (V26) villagejinggang *****

	Before Matching	After Matching
mean treatment.....	0.014842	0.011848
mean control.....	0.016997	0.011848
std mean diff.....	-1.7812	0
mean raw eQQ diff.....	0.0028329	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0010774	0
med eCDF diff.....	0.0010774	0
max eCDF diff.....	0.0021549	0
var ratio (Tr/Co).....	0.87512	1
T-test p-value.....	0.69094	1

***** (V27) villagelianhe *****

	Before Matching	After Matching
mean treatment.....	0.012987	0.010664
mean control.....	0.014164	0.010664
std mean diff.....	-1.0394	0
mean raw eQQ diff.....	0.0018886	0

med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.00058865	0
med eCDF diff.....	0.00058865	0
max eCDF diff.....	0.0011773	0
var ratio (Tr/Co).....	0.91796	1
T-test p-value.....	0.81424	1

***** (V28) villagelianqian *****

	Before Matching	After Matching
mean treatment.....	0.019481	0.016588
mean control.....	0.022663	0.016588
std mean diff.....	-2.3016	0
mean raw eQQ diff.....	0.0037771	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0015912	0
med eCDF diff.....	0.0015912	0
max eCDF diff.....	0.0031824	0
var ratio (Tr/Co).....	0.86236	1
T-test p-value.....	0.60888	1

***** (V29) villagelianxing *****

	Before Matching	After Matching
mean treatment.....	0.01577	0.016588
mean control.....	0.017941	0.016588
std mean diff.....	-1.7422	0
mean raw eQQ diff.....	0.0028329	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0010858	0
med eCDF diff.....	0.0010858	0
max eCDF diff.....	0.0021715	0
var ratio (Tr/Co).....	0.8809	1
T-test p-value.....	0.69687	1

***** (V30) villagelongqing *****

	Before Matching	After Matching
mean treatment.....	0.012987	0.011848
mean control.....	0.016053	0.011848
std mean diff.....	-2.7067	0
mean raw eQQ diff.....	0.0037771	0

med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0015329	0
med eCDF diff.....	0.0015329	0
max eCDF diff.....	0.0030659	0
var ratio (Tr/Co).....	0.81152	1
T-test p-value.....	0.55399	1

***** (V31) villagelusikou *****

	Before Matching	After Matching
mean treatment.....	0.01577	0.018957
mean control.....	0.012276	0.018957
std mean diff.....	2.8034	0
mean raw eQQ diff.....	0.0028329	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0017471	0
med eCDF diff.....	0.0017471	0
max eCDF diff.....	0.0034942	0
var ratio (Tr/Co).....	1.2801	1
T-test p-value.....	0.49218	1

***** (V32) villagemazhou *****

	Before Matching	After Matching
mean treatment.....	0.016698	0.0094787
mean control.....	0.016997	0.0094787
std mean diff.....	-0.23369	0
mean raw eQQ diff.....	0.00094429	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.00014979	0
med eCDF diff.....	0.00014979	0
max eCDF diff.....	0.00029958	0
var ratio (Tr/Co).....	0.98266	1
T-test p-value.....	0.95712	1

***** (V33) villageminzhu *****

	Before Matching	After Matching
mean treatment.....	0.03525	0.0391
mean control.....	0.03305	0.0391
std mean diff.....	1.1927	0
mean raw eQQ diff.....	0.0018886	0

med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0011002	0
med eCDF diff.....	0.0011002	0
max eCDF diff.....	0.0022004	0
var ratio (Tr/Co).....	1.0641	1
T-test p-value.....	0.77955	1

***** (V34) villageshigang *****

	Before Matching	After Matching
mean treatment.....	0.022263	0.024882
mean control.....	0.024551	0.024882
std mean diff.....	-1.5501	0
mean raw eQQ diff.....	0.0028329	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.001144	0
med eCDF diff.....	0.001144	0
max eCDF diff.....	0.002288	0
var ratio (Tr/Co).....	0.90892	1
T-test p-value.....	0.72672	1

***** (V35) villagewanshang *****

	Before Matching	After Matching
mean treatment.....	0.0083488	0.007109
mean control.....	0.0056657	0.007109
std mean diff.....	2.9474	0
mean raw eQQ diff.....	0.0018886	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0013415	0
med eCDF diff.....	0.0013415	0
max eCDF diff.....	0.0026831	0
var ratio (Tr/Co).....	1.4696	1
T-test p-value.....	0.45708	1

***** (V36) villagexiabao *****

	Before Matching	After Matching
mean treatment.....	0.010204	0.0059242
mean control.....	0.00661	0.0059242
std mean diff.....	3.5746	0
mean raw eQQ diff.....	0.0028329	0

med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.001797	0
med eCDF diff.....	0.001797	0
max eCDF diff.....	0.0035941	0
var ratio (Tr/Co).....	1.5381	1
T-test p-value.....	0.3627	1

***** (V37) villagexianghu *****

	Before Matching	After Matching
mean treatment.....	0.010204	0.010664
mean control.....	0.011331	0.010664
std mean diff.....	-1.1213	0
mean raw eQQ diff.....	0.0018886	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.00056368	0
med eCDF diff.....	0.00056368	0
max eCDF diff.....	0.0011274	0
var ratio (Tr/Co).....	0.90152	1
T-test p-value.....	0.80084	1

***** (V38) villagexiaofang *****

	Before Matching	After Matching
mean treatment.....	0.010204	0.0059242
mean control.....	0.0075543	0.0059242
std mean diff.....	2.6354	0
mean raw eQQ diff.....	0.0018886	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0013249	0
med eCDF diff.....	0.0013249	0
max eCDF diff.....	0.0026498	0
var ratio (Tr/Co).....	1.3471	1
T-test p-value.....	0.5138	1

***** (V39) villagexiashatou *****

	Before Matching	After Matching
mean treatment.....	0.011132	0.0011848
mean control.....	0.00661	0.0011848
std mean diff.....	4.3078	0
mean raw eQQ diff.....	0.0037771	0

med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0022609	0
med eCDF diff.....	0.0022609	0
max eCDF diff.....	0.0045217	0
var ratio (Tr/Co).....	1.6764	1
T-test p-value.....	0.26471	1

***** (V40) villagexiecheng *****

	Before Matching	After Matching
mean treatment.....	0.03154	0.03436
mean control.....	0.024551	0.03436
std mean diff.....	3.9967	0
mean raw eQQ diff.....	0.00661	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0034942	0
med eCDF diff.....	0.0034942	0
max eCDF diff.....	0.0069884	0
var ratio (Tr/Co).....	1.2754	1
T-test p-value.....	0.32789	1

***** (V41) villagexihe *****

	Before Matching	After Matching
mean treatment.....	0.017625	0.020142
mean control.....	0.020774	0.020142
std mean diff.....	-2.3921	0
mean raw eQQ diff.....	0.0037771	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0015745	0
med eCDF diff.....	0.0015745	0
max eCDF diff.....	0.0031491	0
var ratio (Tr/Co).....	0.85113	1
T-test p-value.....	0.59617	1

***** (V42) villagexihu *****

	Before Matching	After Matching
mean treatment.....	0.012059	0.013033
mean control.....	0.014164	0.013033
std mean diff.....	-1.9276	0
mean raw eQQ diff.....	0.0028329	0

med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0010525	0
med eCDF diff.....	0.0010525	0
max eCDF diff.....	0.0021049	0
var ratio (Tr/Co).....	0.8532	1
T-test p-value.....	0.66916	1

***** (V43) villagexilian *****

	Before Matching	After Matching
mean treatment.....	0.0092764	0.0082938
mean control.....	0.011331	0.0082938
std mean diff.....	-2.1426	0
mean raw eQQ diff.....	0.0028329	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0010275	0
med eCDF diff.....	0.0010275	0
max eCDF diff.....	0.002055	0
var ratio (Tr/Co).....	0.82033	1
T-test p-value.....	0.63845	1

***** (V44) villagexinguang *****

	Before Matching	After Matching
mean treatment.....	0.025046	0.031991
mean control.....	0.033994	0.031991
std mean diff.....	-5.7234	0
mean raw eQQ diff.....	0.0094429	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.004474	0
med eCDF diff.....	0.004474	0
max eCDF diff.....	0.008948	0
var ratio (Tr/Co).....	0.74359	1
T-test p-value.....	0.22225	1

***** (V45) villagexingzeng *****

	Before Matching	After Matching
mean treatment.....	0.016698	0.016588
mean control.....	0.017941	0.016588
std mean diff.....	-0.97029	0
mean raw eQQ diff.....	0.0018886	0

med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.00062193	0
med eCDF diff.....	0.00062193	0
max eCDF diff.....	0.0012439	0
var ratio (Tr/Co).....	0.93183	1
T-test p-value.....	0.82571	1

***** (V46) villagexinlian *****

	Before Matching	After Matching
mean treatment.....	0.038033	0.043839
mean control.....	0.034939	0.043839
std mean diff.....	1.6172	0
mean raw eQQ diff.....	0.0028329	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0015474	0
med eCDF diff.....	0.0015474	0
max eCDF diff.....	0.0030948	0
var ratio (Tr/Co).....	1.0851	1
T-test p-value.....	0.70294	1

***** (V47) villageyanjiang *****

	Before Matching	After Matching
mean treatment.....	0.0064935	0.0047393
mean control.....	0.0056657	0.0047393
std mean diff.....	1.0301	0
mean raw eQQ diff.....	0	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	0	0
mean eCDF diff.....	0.00041389	0
med eCDF diff.....	0.00041389	0
max eCDF diff.....	0.00082778	0
var ratio (Tr/Co).....	1.1451	1
T-test p-value.....	0.80564	1

***** (V48) villageyazhou *****

	Before Matching	After Matching
mean treatment.....	0.019481	0.020142
mean control.....	0.021719	0.020142
std mean diff.....	-1.6186	0
mean raw eQQ diff.....	0.0028329	0

med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.001119	0
med eCDF diff.....	0.001119	0
max eCDF diff.....	0.0022381	0
var ratio (Tr/Co).....	0.89899	1
T-test p-value.....	0.71594	1

***** (V49) villageyongfeng *****

	Before Matching	After Matching
mean treatment.....	0.027829	0.026066
mean control.....	0.020774	0.026066
std mean diff.....	4.2872	0
mean raw eQQ diff.....	0.00661	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0035275	0
med eCDF diff.....	0.0035275	0
max eCDF diff.....	0.007055	0
var ratio (Tr/Co).....	1.3299	1
T-test p-value.....	0.28954	1

***** (V50) villagezhangxi *****

	Before Matching	After Matching
mean treatment.....	0.058442	0.07109
mean control.....	0.053824	0.07109
std mean diff.....	1.9674	0
mean raw eQQ diff.....	0.0037771	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0023086	0
med eCDF diff.....	0.0023086	0
max eCDF diff.....	0.0046172	0
var ratio (Tr/Co).....	1.0805	1
T-test p-value.....	0.64304	1

***** (V51) villagezixi *****

	Before Matching	After Matching
mean treatment.....	0.018553	0.014218
mean control.....	0.022663	0.014218
std mean diff.....	-3.0444	0
mean raw eQQ diff.....	0.0047214	0

```

med   raw eQQ diff.....      0      0
max   raw eQQ diff.....      1      0

mean  eCDF diff.....    0.002055    0
med   eCDF diff.....    0.002055    0
max   eCDF diff.....    0.00411    0

var ratio (Tr/Co).....    0.82207    1
T-test p-value.....    0.50413    1

```

Before Matching Minimum p.value: 0.172

Variable Name(s): agpop Number(s): 3

After Matching Minimum p.value: 0.396

Variable Name(s): age Number(s): 2

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```

#mbgen2 <- MatchBalance(intensive~ male + age + agpop + ricearea_2010 + literacy + villa
ge, data = round1dat, match.out=mout1, nboots=500)

```

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```
summary(mout1)
```

```

Estimate... 0.12204
AI SE..... 0.022709
T-stat..... 5.3739
p.val..... 7.7051e-08

Original number of observations..... 2137
Original number of treated obs..... 1078
Matched number of observations..... 844
Matched number of observations (unweighted). 856

Number of obs dropped by 'exact' or 'caliper' 234

```

#Social network effect ## Table 2 replication column 2-6

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```

#Round 1 participants with information
dat_delay_noinfo <- dat[dat$delay == 1 & dat$info_none == 1, ]
lm_t2c2 <- lm(takeup_survey ~ network_rate_preintensive + male + age + agpop + ricearea_
2010 + literacy +
               intensive + risk_averse + disaster_prob + friend1 + friend2 + friend3 +
               friend4 + friend5 + village,data = dat_delay_noinfo)

lm_t2c2

```

Call:

```
lm(formula = takeup_survey ~ network_rate_preintensive + male +  
  age + agpop + ricearea_2010 + literacy + intensive + risk_averse +  
  disaster_prob + friend1 + friend2 + friend3 + friend4 + friend5 +  
  village, data = dat_delay_noinfo)
```

Coefficients:

(Intercept)	network_rate_preintensive	male
-0.309686	0.291185	0.015673
age	agpop	ricearea_2010
0.004502	-0.010157	0.004314
literacy	intensive	risk_averse
0.087433	0.029836	0.109170
disaster_prob	friend1	friend2
0.001940	NA	-0.320185
friend3	friend4	friend5
-0.199158	-0.138805	NA
villagebeixing	villagecaijia	villagedaqiao
0.163130	0.304711	0.518810
villagedaxi	villagedayu	villagedazhou
0.332533	0.366318	0.249093
villagedongan	villagedukou	villagefusheng
0.332227	0.254961	0.447132
villagefuzhou	villagegangtou	villagegangxia
0.410264	0.162942	0.587894
villageguojia	villagehefeng	villagehelin
0.439132	0.156094	0.402205
villagehongxing	villagehuangshan	villagejingang
0.122871	0.426051	0.207088
villagejinggang	villagelianqian	villagelianxing
0.271775	0.361079	0.575290
villagelongqing	villagelusikou	villagemazhou
0.331263	0.462447	0.422276
villageminzhu	villageshigang	villagewanshang
0.135621	0.233987	0.262976
villagexiabao	villagexianghu	villagexiaofang
0.532218	0.429912	0.417308
villagexiashatou	villagexiecheng	villagexihe
0.088401	0.102240	0.285431
villagexihu	villagexilian	villagexinguang
0.334351	0.324259	0.177317
villagexingzeng	villagexinlian	villageyanjiang
0.287571	0.177150	0.884911
villageyazhou	villageyongfeng	villagezhangxi
0.351340	0.539276	0.168329
villagezixi		
0.198938		

Hide

```
###check the effects of household covariates and exclude other covariates except intensive session
lm_t2c3 <- lm(takeup_survey ~ network_rate_preintensive + network_rate_presimple + intensive
              + friend1 + friend2 + friend3 + friend4 + friend5 + village,
              data = dat_delay_noinfo)

lm_t2c3
```

Call:

```
lm(formula = takeup_survey ~ network_rate_preintensive + network_rate_presimple +
    intensive + friend1 + friend2 + friend3 + friend4 + friend5 +
    village, data = dat_delay_noinfo)
```

Coefficients:

(Intercept)	network_rate_preintensive	network_rate_presimple
0.13193	0.27809	-0.10795
intensive	friend1	friend2
0.02562	NA	-0.45760
friend3	friend4	friend5
-0.19241	-0.12522	NA
villagebeixing	villagecaijia	villagedaqiao
0.18229	0.22555	0.44728
villagedaxi	villagedayu	villagedazhou
0.27910	0.37532	0.18164
villagedongan	villagedukou	villagefusheng
0.43820	0.20799	0.40045
villagefuzhou	villagegangtou	villagegangxia
0.39530	0.16291	0.55931
villageguojia	villagehefeng	villagehelin
0.45743	0.14484	0.40749
villagehongxing	villagehuangshan	villagejingang
0.14528	0.39272	0.19683
villagejinggang	villagelianqian	villagelianxing
0.24562	0.33922	0.57993
villagelongqing	villagelusikou	villagemazhou
0.27162	0.47283	0.36705
villageminzhu	villageshigang	villagewanshang
0.14525	0.16044	0.20277
villagexiabao	villagexianghu	villagexiaofang
0.51706	0.42758	0.36684
villagexiashatou	villagexiecheng	villagexihe
0.11203	0.07079	0.25564
villagexihu	villagexilian	villagexinguang
0.34227	0.29979	0.16259
villagexingzeng	villagexinlian	villageyanjiang
0.28246	0.15698	0.86202
villageyazhou	villageyongfeng	villagezhangxi
0.31007	0.51916	0.14673
villagezixi		
0.18147		

[Hide](#)

```
###Table 2, col 4
```

Warning messages:

```
1: Unknown or uninitialised column: `pre_takeup_maj`.
2: Unknown or uninitialised column: `pre_takeup_maj`.
3: Unknown or uninitialised column: `pre_takeup_maj`.
4: Unknown or uninitialised column: `pre_takeup_maj`.
```

[Hide](#)

```
lm_t2c4 <- lm(takeup_survey ~ network_rate_preintensive + intensive + network_rate_prein
tensive:intensive + male + age + agpop + ricearea_2010
          + literacy + risk_averse + disaster_prob + friend1 + friend2 + friend3 + f
riend4 + friend5 + village
          , data = dat_delay_noinfo)
```

[Hide](#)

```
#Column 5
lm_t2c5 <- lm(takeup_survey ~ network_onlyone + network_onlytwo + network_twomore + inte
nsive
          + network_onlyone:intensive + network_onlytwo:intensive + network_twomore:
intensive +
          + male + age + agpop + ricearea_2010 + literacy + risk_averse + disaster
_prob
          + friend1 + friend2 + friend3 + friend4 + friend5 + village
          , data = dat_delay_noinfo)
```

[Hide](#)

```
#For people who is in the first round or people who is in second round samples who have
  friends and receive info

dat$nofriend = 1
dat$nofriend[dat$delay == 1 & dat$info_none == 1 & dat$network_yes == 1] = 0 #set no fri
end = 0 (which means they have friends) if they are in the second round, receive informa
tion, and have social network

dat_t2c6 = dat[ (dat$delay == 0 | (dat$delay == 1 & dat$info_none == 1 & dat$nofriend =
= 1)), ]
# dat_t2c6$inter = dat_t2c6$intensive * dat_t2c6$delay
lm_t2c6 <- lm(takeup_survey ~ intensive + delay + intensive*delay
          + male + age + agpop + ricearea_2010 + literacy + risk_averse + disaster_p
rob
          + village
          , data = dat_t2c6)
```

[Hide](#)

```
stargazer(lm_t2c1, lm_t2c2, lm_t2c3, lm_t2c4, lm_t2c5, lm_t2c6,  
          omit = c("village", "friend1", "friend2", "friend3", "friend4", "friend5"), st  
yle = "aer", no.space = T, df = FALSE)
```

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of result is not a multiple of vector length (arg 2)

```

% Table created by stargazer v.5.2.2 by Marek Hlavac, Harvard University. E-mail: hlavac
at fas.harvard.edu
% Date and time: Fri, Dec 18, 2020 - 18:50:29
\begin{table}[!htbp] \centering
  \caption{}
  \label{}
\begin{tabular}{@{\extracolsep{5pt}}lcccccc}
\\[-1.8ex]\hline
\hline \\[-1.8ex]
\\[-1.8ex] & \multicolumn{6}{c}{takeup\_survey} \\
\\[-1.8ex] & (1) & (2) & (3) & (4) & (5) & (6) \\
\hline \\[-1.8ex]
network\_rate\_preintensive & & 0.291$^{***}$ & 0.278$^{***}$ & 0.444$^{***}$ & & \\
& & (0.078) & (0.080) & (0.106) & & \\
network\_rate\_presimple & & & -$0.108 & & & \\
& & & (0.090) & & & \\
network\_onlyone & & & & 0.097$^{**}$ & & \\
& & & & (0.043) & & \\
network\_onlytwo & & & & 0.177$^{*}$ & & \\
& & & & (0.104) & & \\
network\_twomore & & & & 0.137 & & \\
& & & & (0.090) & & \\
intensive & 0.141$^{***}$ & 0.030 & 0.026 & 0.081$^{**}$ & 0.094$^{**}$ & 0.140$^{***}$ \\
$ \\
& (0.020) & (0.027) & (0.027) & (0.036) & (0.040) & (0.021) \\
delay & & & & 0.032 \\
& & & & (0.031) \\
risk\_averse & & 0.109$^{**}$ & 0.105$^{**}$ & 0.104$^{**}$ & 0.072$^{**}$ \\
& & (0.048) & (0.048) & (0.048) & (0.028) \\
disaster\_prob & & 0.002$^{**}$ & 0.002$^{**}$ & 0.002$^{**}$ & 0.000 \\
& & (0.001) & (0.001) & (0.001) & (0.001) \\
male & 0.041 & 0.016 & 0.019 & 0.029 & 0.048 \\
& (0.041) & (0.057) & (0.057) & (0.057) & (0.036) \\
age & 0.002$^{*}$ & 0.005$^{***}$ & 0.005$^{***}$ & 0.005$^{***}$ & 0.002$^{***}$ \\
\\
& (0.001) & (0.001) & (0.001) & (0.001) & (0.001) \\
agpop & -$0.003 & -$0.010 & -$0.010 & -$0.008 & -$0.004 \\
& (0.005) & (0.006) & (0.006) & (0.006) & (0.004) \\
ricearea\_2010 & 0.001 & 0.004$^{***}$ & 0.004$^{***}$ & 0.004$^{***}$ & 0.001$^{*}$ \\
*$ \\
& (0.001) & (0.001) & (0.001) & (0.001) & (0.000) \\
literacy & 0.083$^{***}$ & 0.087$^{**}$ & 0.087$^{**}$ & 0.090$^{**}$ & 0.063$^{**}$ \\
$ \\
& (0.028) & (0.036) & (0.036) & (0.036) & (0.025) \\
network\_rate\_preintensive:intensive & & & & -$0.329$^{**}$ & & \\
& & & & (0.155) & & \\
network\_onlyone:intensive & & & & -$0.087 & & \\
& & & & (0.061) & & \\
network\_onlytwo:intensive & & & & -$0.091 & & \\
& & & & (0.154) & & \\
network\_twomore:intensive & & & & -$0.141 & & \\
& & & & (0.140) & &

```

```
intensive:delay & & & & & $-0.053 \\
& & & & (0.043) \\
Constant & 0.363$^{**}$ & $-0.310$^{*}$ & 0.132 & $-0.342$^{**}$ & $-0.380$^{**}$ & 
0.333$^{**}$ \\
& (0.158) & (0.169) & (0.142) & (0.169) & (0.170) & (0.157) \\
Observations & 2,137 & 1,255 & 1,274 & 1,255 & 1,255 & 2,756 \\
R$^{\{2\}}$ & 0.125 & 0.119 & 0.091 & 0.123 & 0.129 & 0.107 \\
Adjusted R$^{\{2\}}$ & 0.103 & 0.079 & 0.055 & 0.082 & 0.085 & 0.089 \\
Residual Std. Error & 0.469 & 0.476 & 0.482 & 0.475 & 0.474 & 0.473 \\
F Statistic & 5.732$^{***}$ & 2.952$^{***}$ & 2.508$^{***}$ & 2.989$^{***}$ & 2.941$^{***}$ & 
*$ & 5.786$^{***}$ \\
\\hline \\[-1.8ex]
\\textit{Notes:} & \\multicolumn{6}{l}{$^{***}$Significant at the 1 percent level.} \\
& \\multicolumn{6}{l}{$^{**}$Significant at the 5 percent level.} \\
& \\multicolumn{6}{l}{$^{*}$Significant at the 10 percent level.} \\
\\end{tabular}
\\end{table}
```

```
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(arg 2)number of rows of result is not a multiple of vector length (arg 2)number of rows
of result is not a multiple of vector length (arg 2)
```

#Extension 2: Use Genetic Matching to measure Social network effect – Compare with Table 2 column 2

Distribution Balance: check the distribution balance plot before matching

Hide

```
male <- bal.plot(dat_gen3$dense_network ~ dat_gen3$male, treat = dat_gen3$dense_network)
+ ggtitle("Distributional Balance for Male") + xlab('Male')
```

No 'var.name' was provided. Displaying balance for dat gen3\$male.

Hide

```
age <- bal.plot(dat_gen3$dense_network ~ dat_gen3$age, treat = dat_gen3$dense_network) +
  ggtitle("Distributional Balance for Age") + xlab('Age')
```

No 'var.name' was provided. Dispalying balance for dat gen3\$age.

Hide

```
agpop <- bal.plot(dat_gen3$dense_network ~ dat_gen3$agpop, treat = dat_gen3$dense_network) + ggtitle("Distributional Balance for Age Population") + xlab('Age Population')
```

No 'var.name' was provided. Dispalying balance for dat_gen3\$agpop.

Hide

```
ricearea_2010 <- bal.plot(dat_gen3$dense_network ~ dat_gen3$ricearea_2010, treat = dat_gen3$dense_network) + ggtitle("Distributional Balance for Rice Area") + xlab('Rice Area')
```

No 'var.name' was provided. Dispalying balance for dat_gen3\$ricearea_2010.

Hide

```
literacy <- bal.plot(dat_gen3$dense_network ~ dat_gen3$literacy, treat = dat_gen3$dense_network) + ggtitle("Distributional Balance for Literacy") + xlab('Literacy')
```

No 'var.name' was provided. Dispalying balance for dat_gen3\$literacy.

Hide

```
village <- bal.plot(dat_gen3$dense_network ~ dat_gen3$village, treat = dat_gen3$dense_network) + ggtitle("Distributional Balance for Village") + xlab('Village')
```

No 'var.name' was provided. Dispalying balance for dat_gen3\$village.

Hide

```
intensive <- bal.plot(dat_gen3$dense_network ~ dat_gen3$intensive, treat = dat_gen3$dense_network) + ggtitle("Distributional Balance for Intensive") + xlab('Intensive')
```

No 'var.name' was provided. Dispalying balance for dat_gen3\$intensive.

Hide

```
risk_averse <- bal.plot(dat_gen3$dense_network ~ dat_gen3$risk_averse, treat = dat_gen3$risk_averse) + ggtitle("Distributional Balance for Risk Averse") + xlab('Risk Averse')
```

No 'var.name' was provided. Dispalying balance for dat_gen3\$risk_averse.

Hide

```
disaster_prob <- bal.plot(dat_gen3$dense_network ~ dat_gen3$disaster_prob, treat = dat_gen3$dense_network) + ggtitle("Distributional Balance for Disaster Probability") + xlab('Disaster Probability')
```

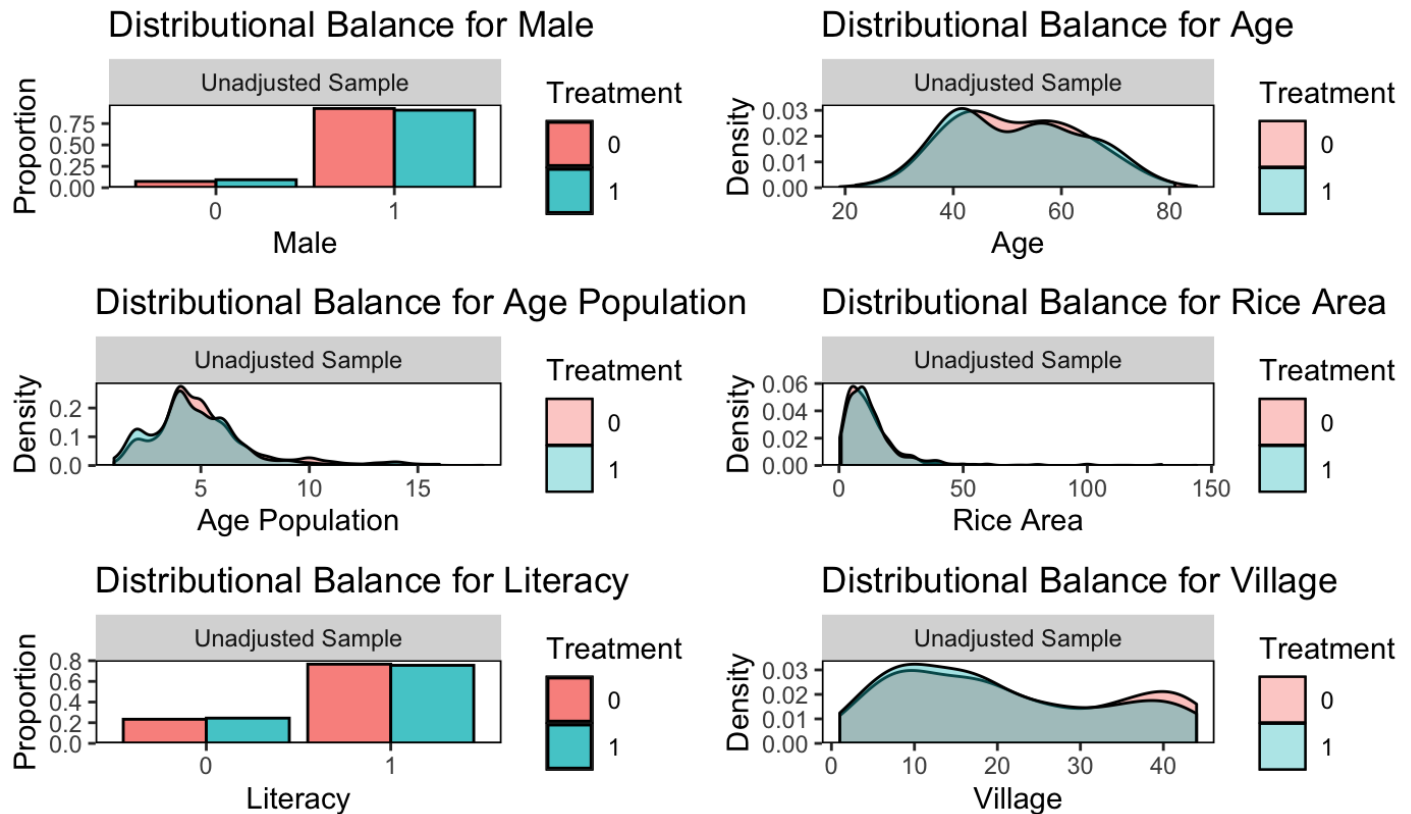
No 'var.name' was provided. Dispalying balance for dat_gen3\$disaster_prob.

Hide

```
#Distribution Balance for household characteristics
```

```
grid.arrange(male, age, agpop, ricearea_2010, literacy, village, nrow =3, top=textGrob(
"Distribution balance for household characteristics", gp=gpar(fontsize=15,font=15)))
```

Distribution balance for household characteristics

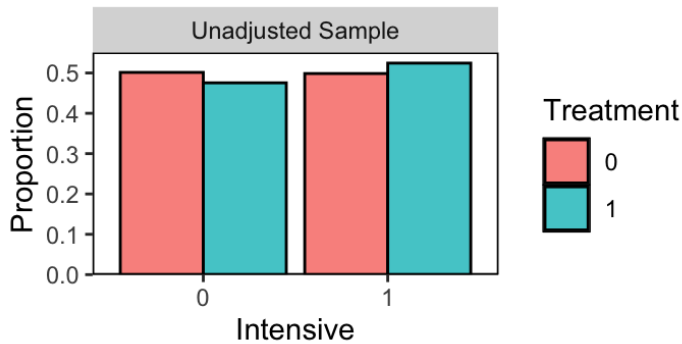


Hide

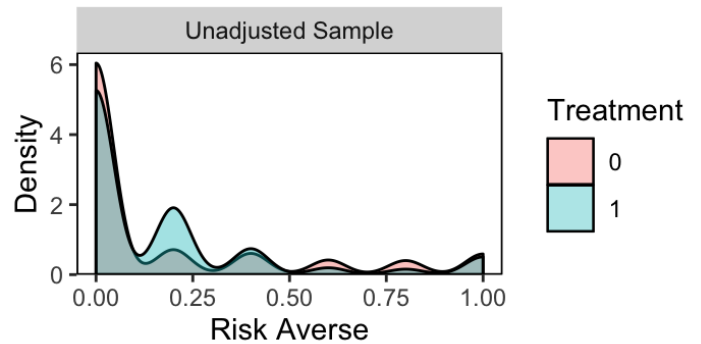
```
grid.arrange(intensive, risk_averse, disaster_prob, nrow =2, top=textGrob("Distribution
balance plot of non-househole variables", gp=gpar(fontsize=15,font=15)))
```

Distribution balance plot of non-household variables

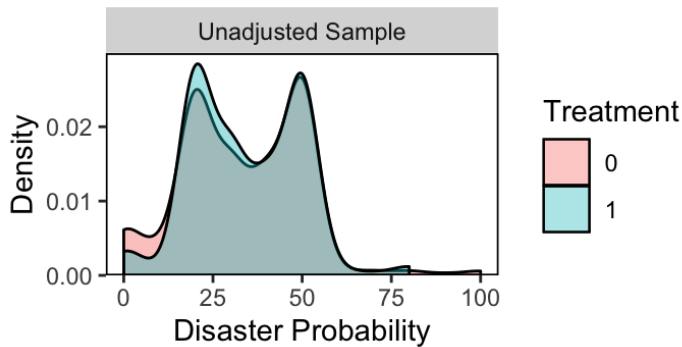
Distributional Balance for Intensive



Distributional Balance for Risk Averse



Distributional Balance for Disaster Probability



Network effect 1-1

- treatment: people have dense network
- sample: 2nd round with no take-up information given
- matching variables: household characteristics

Hide

```
#Define treatment: create a binary variable dense_network as our treatment, if network_r  
ate_preintensive > 0.4  
dat$dense_network <- 0  
dat$dense_network[dat$network_rate_preintensive >=0.4] <-1
```

Hide

```
#Data preparation for genetic matching
```

```
#create dataframe for included variables
```

```
dat_gen <- cbind(dat_delay_noinfo$male, dat_delay_noinfo$age, dat_delay_noinfo$agpop, dat_delay_noinfo$ricearea_2010, dat_delay_noinfo$literacy, factor(dat_delay_noinfo$village), dat_delay_noinfo$intensive, dat_delay_noinfo$dense_network, dat_delay_noinfo$takeup_survey)
```

```
dat_gen<- as.data.frame(dat_gen) %>% rename(male=V1, age=V2, agpop = V3, , ricearea_2010=V4, literacy=V5, village=V6, intensive = V7, dense_network = V8, takeup_survey = V9)
```

```
#drop NAs so we can do genetic matching
```

```
dat_gen<- dat_gen %>% drop_na()
```

```
#Dataframe for matched variable
```

```
X2 <- subset(dat_gen, select = -c(dense_network,takeup_survey))
```

[Hide](#)

```
#Genetic Matching
```

```
invisible(capture.output(genout2 <- GenMatch(Tr = dat_gen$dense_network, X=X2 , M=1, estimand = 'ATT', max.generations = 10)))
```

[Hide](#)

```
mout2 <- Match(Y = dat_gen$takeup_survey, Tr= dat_gen$dense_network, X=X2, M =1, estimand = 'ATT', Weight.matrix = genout2)
```

```
# mout2 <- Match(Y = roundldat$takeup_survey, Tr= roundldat$intensive, X = X, M =2, estimand = 'ATT', Weight.matrix = genout1, exact = c(1,0,0,0,1,1), caliper = c(1e16, 1e16, 1e16, 0.3,1e16, 1e16 ))
```

```
mbgen2 <- MatchBalance(dense_network~ male + age + agpop + ricearea_2010 + literacy + village, data = dat_gen, match.out=mout2, nboots=500)
```


***** (V1) male *****

	Before Matching	After Matching
mean treatment.....	0.90667	0.90667
mean control.....	0.92624	0.90667
std mean diff.....	-6.7126	0
mean raw eQQ diff.....	0.017778	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0097852	0
med eCDF diff.....	0.0097852	0
max eCDF diff.....	0.01957	0
var ratio (Tr/Co).....	1.2429	1
T-test p-value.....	0.35246	1

***** (V2) age *****

	Before Matching	After Matching
mean treatment.....	51.591	51.591
mean control.....	51.794	51.689
std mean diff.....	-1.6387	-0.79111
mean raw eQQ diff.....	0.72444	0.80531
med raw eQQ diff.....	1	1
max raw eQQ diff.....	4	3
mean eCDF diff.....	0.011375	0.01432
med eCDF diff.....	0.0053035	0.013274
max eCDF diff.....	0.054865	0.044248
var ratio (Tr/Co).....	1.1009	1.1513
T-test p-value.....	0.82193	0.68637
KS Bootstrap p-value..	0.472	0.956
KS Naive p-value.....	0.63032	0.97982
KS Statistic.....	0.054865	0.044248

***** (V3) agpop *****

	Before Matching	After Matching
mean treatment.....	4.8489	4.8489
mean control.....	5.0233	4.8578
std mean diff.....	-7.1092	-0.36223
mean raw eQQ diff.....	0.30667	0.11504
med raw eQQ diff.....	0	0
max raw eQQ diff.....	3	2
mean eCDF diff.....	0.018446	0.0071903
med eCDF diff.....	0.012512	0.0044248
max eCDF diff.....	0.061438	0.022124

var ratio (Tr/Co).....	1.2471	1.1378
T-test p-value.....	0.32466	0.7997
KS Bootstrap p-value..	0.182	0.984
KS Naive p-value.....	0.4841	1
KS Statistic.....	0.061438	0.022124

***** (V4) ricearea_2010 *****

	Before Matching	After Matching
mean treatment.....	13.09	13.09
mean control.....	12.863	12.924
std mean diff.....	1.752	1.2827
mean raw eQQ diff.....	0.90249	1.0027
med raw eQQ diff.....	0.4	0.35
max raw eQQ diff.....	20	20
mean eCDF diff.....	0.016307	0.017257
med eCDF diff.....	0.01014	0.013274
max eCDF diff.....	0.058898	0.061947
var ratio (Tr/Co).....	0.94481	1.0256
T-test p-value.....	0.81226	0.53624
KS Bootstrap p-value..	0.378	0.604
KS Naive p-value.....	0.53906	0.77872
KS Statistic.....	0.058898	0.061947

***** (V5) literacy *****

	Before Matching	After Matching
mean treatment.....	0.75556	0.75556
mean control.....	0.76657	0.75556
std mean diff.....	-2.558	0
mean raw eQQ diff.....	0.0088889	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.0055089	0
med eCDF diff.....	0.0055089	0
max eCDF diff.....	0.011018	0
var ratio (Tr/Co).....	1.0358	1
T-test p-value.....	0.72667	1

***** (V6) village *****

	Before Matching	After Matching
mean treatment.....	20.28	20.28
mean control.....	21.72	20.022
std mean diff.....	-11.386	2.0383
mean raw eQQ diff.....	1.4444	0.61504

```

med   raw eQQ diff.....      1          0
max   raw eQQ diff.....      5          3

mean eCDF diff.....      0.032979      0.014328
med   eCDF diff.....      0.037628      0.011062
max   eCDF diff.....      0.061998      0.044248

var ratio (Tr/Co).....      0.9221      1.0325
T-test p-value.....      0.1242      0.48299
KS Bootstrap p-value..      0.276      0.936
KS Naive p-value.....      0.47233      0.97982
KS Statistic.....      0.061998      0.044248

```

Before Matching Minimum p.value: 0.1242
Variable Name(s): village Number(s): 6

After Matching Minimum p.value: 0.48299
Variable Name(s): village Number(s): 6

Hide

```
#mbgen2 <- MatchBalance(intensive~ male + age + agpop + ricearea_2010 + literacy + villa
ge, data = round1dat, match.out=mout1, nboots=500)
```

Hide

```
summary(mout2)
```

```

Estimate...  0.11111
AI SE.....  0.049772
T-stat.....  2.2324
p.val.....  0.025587

Original number of observations..... 1296
Original number of treated obs..... 225
Matched number of observations..... 225
Matched number of observations (unweighted). 226

```

Network effect 1-2

- treatment: people have dense network
- sample: 2nd round with no take-up information given
- matching variables: household characteristics + risk_averse + disaster_prob(exact the same as column 2)

Hide

```
#Data preperation for genetic matching
```

```
#create a dataframe for
```

```
dat_gen3 <- cbind(dat_delay_noinfo$male, dat_delay_noinfo$age, dat_delay_noinfo$agpop, d
at_delay_noinfo$ricearea_2010, dat_delay_noinfo$literacy, factor(dat_delay_noinfo$villag
e), dat_delay_noinfo$intensive, dat_delay_noinfo$risk_averse, dat_delay_noinfo$disaster_
prob ,dat_delay_noinfo$dense_network, dat_delay_noinfo$takeup_survey)
```

```
dat_gen3<- as.data.frame(dat_gen3) %>% rename(male=V1, age=V2, agpop = V3, , ricearea_20
10=V4, literacy=V5, village=V6, intensive = V7, risk_averse= V8,disaster_prob = V9 ,dens
e_network = V10, takeup_survey = V11)
```

```
#drop NAs so we can do genetic matching
```

```
dat_gen3<- dat_gen3 %>% drop_na()
```

```
#Dataframe for matched variable
```

```
X4 <- subset(dat_gen3, select = -c(dense_network,takeup_survey))
```

```
X4
```

	ag...	ricearea_2010	literacy	village	intensive	risk_averse	disaster_prob						
	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>						
1	1	54	4	10.00	0	10	0	0.0	20						
2	1	67	2	10.00	1	10	1	0.0	30						
3	1	61	13	4.00	1	10	1	0.0	50						
4	1	45	4	15.00	1	10	0	0.2	50						
5	1	70	7	3.80	1	10	0	0.0	20						
6	1	76	4	5.00	1	10	1	0.2	50						
7	1	59	7	4.00	0	10	1	0.0	20						
8	1	48	6	8.00	1	10	0	0.4	50						
9	1	72	12	4.55	0	10	1	0.0	50						
10	1	48	7	4.50	1	10	0	0.0	20						
1-10 of 1,296 rows						Previous	1	2	3	4	5	6	...	100	Next

Hide

```
table(dat_gen3$dense_network)
```

```
0      1
1071 225
```

[Hide](#)

```
#Genetic Matching
invisible(capture.output(genout4 <- GenMatch(Tr = dat_gen3$dense_network, X=X4 , M=1, estimand = 'ATT', max.generations = 10)))
```

[Hide](#)

```
mout4 <- Match(Y = dat_gen3$takeup_survey, Tr= dat_gen3$dense_network, X=X4, M =1, estimand = 'ATT', Weight.matrix = genout4)

mbgen4 <- MatchBalance(dense_network~ male + age + agpop + ricearea_2010 + literacy + village+risk_averse+disaster_prob, data = dat_gen3, match.out=mout4, nboots=500)
```

***** (V1) male *****

	Before Matching	After Matching
mean treatment.....	0.90667	0.90667
mean control.....	0.92624	0.91111
std mean diff.....	-6.7126	-1.5244
mean raw eQQ diff.....	0.017778	0.0044444
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	1
mean eCDF diff.....	0.0097852	0.0022222
med eCDF diff.....	0.0097852	0.0022222
max eCDF diff.....	0.01957	0.0044444
var ratio (Tr/Co).....	1.2429	1.0449
T-test p-value.....	0.35246	0.31731

***** (V2) age *****

	Before Matching	After Matching
mean treatment.....	51.591	51.591
mean control.....	51.794	51.702
std mean diff.....	-1.6387	-0.89898
mean raw eQQ diff.....	0.72444	1.6133
med raw eQQ diff.....	1	2
max raw eQQ diff.....	4	7
mean eCDF diff.....	0.011375	0.030675
med eCDF diff.....	0.0053035	0.026667
max eCDF diff.....	0.054865	0.071111
var ratio (Tr/Co).....	1.1009	1.3425
T-test p-value.....	0.82193	0.89129
KS Bootstrap p-value..	0.486	0.518
KS Naive p-value.....	0.63032	0.62002
KS Statistic.....	0.054865	0.071111

***** (V3) agpop *****

	Before Matching	After Matching
mean treatment.....	4.8489	4.8489
mean control.....	5.0233	4.8267
std mean diff.....	-7.1092	0.90558
mean raw eQQ diff.....	0.30667	0.13778
med raw eQQ diff.....	0	0
max raw eQQ diff.....	3	1
mean eCDF diff.....	0.018446	0.0086111
med eCDF diff.....	0.012512	0.0066667
max eCDF diff.....	0.061438	0.022222

var ratio (Tr/Co).....	1.2471	1.1686
T-test p-value.....	0.32466	0.64973
KS Bootstrap p-value..	0.192	0.99
KS Naive p-value.....	0.4841	1
KS Statistic.....	0.061438	0.022222

***** (V4) ricearea_2010 *****

	Before Matching	After Matching
mean treatment.....	13.09	13.09
mean control.....	12.863	12.729
std mean diff.....	1.752	2.7849
mean raw eQQ diff.....	0.90249	0.87684
med raw eQQ diff.....	0.4	0.6
max raw eQQ diff.....	20	20
mean eCDF diff.....	0.016307	0.018272
med eCDF diff.....	0.01014	0.017778
max eCDF diff.....	0.058898	0.053333
var ratio (Tr/Co).....	0.94481	1.2006
T-test p-value.....	0.81226	0.2068
KS Bootstrap p-value..	0.392	0.77
KS Naive p-value.....	0.53906	0.90621
KS Statistic.....	0.058898	0.053333

***** (V5) literacy *****

	Before Matching	After Matching
mean treatment.....	0.75556	0.75556
mean control.....	0.76657	0.76
std mean diff.....	-2.558	-1.0319
mean raw eQQ diff.....	0.0088889	0.0044444
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	1
mean eCDF diff.....	0.0055089	0.0022222
med eCDF diff.....	0.0055089	0.0022222
max eCDF diff.....	0.011018	0.0044444
var ratio (Tr/Co).....	1.0358	1.0126
T-test p-value.....	0.72667	0.31731

***** (V6) village *****

	Before Matching	After Matching
mean treatment.....	20.28	20.28
mean control.....	21.72	20.084
std mean diff.....	-11.386	1.5463
mean raw eQQ diff.....	1.4444	0.52444

med raw eQQ diff.....	1	0
max raw eQQ diff.....	5	3
mean eCDF diff.....	0.032979	0.011919
med eCDF diff.....	0.037628	0.0088889
max eCDF diff.....	0.061998	0.04
var ratio (Tr/Co).....	0.9221	1.0237
T-test p-value.....	0.1242	0.56366
KS Bootstrap p-value..	0.328	0.95
KS Naive p-value.....	0.47233	0.99376
KS Statistic.....	0.061998	0.04

***** (V7) risk_averse *****

	Before Matching	After Matching
mean treatment.....	0.16267	0.16267
mean control.....	0.17628	0.15911
std mean diff.....	-5.0316	1.3138
mean raw eQQ diff.....	0.049778	0.0088889
med raw eQQ diff.....	0	0
max raw eQQ diff.....	0.4	0.2
mean eCDF diff.....	0.050052	0.022222
med eCDF diff.....	0.04747	0.0044444
max eCDF diff.....	0.10413	0.11111
var ratio (Tr/Co).....	0.75532	1.0038
T-test p-value.....	0.50483	0.44991
KS Bootstrap p-value..	0.002	0.036
KS Naive p-value.....	0.035476	0.12432
KS Statistic.....	0.10413	0.11111

***** (V8) disaster_prob *****

	Before Matching	After Matching
mean treatment.....	34.147	34.147
mean control.....	33.243	34.342
std mean diff.....	5.9119	-1.2797
mean raw eQQ diff.....	1.5911	0.77333
med raw eQQ diff.....	0	0
max raw eQQ diff.....	20	10
mean eCDF diff.....	0.018797	0.010159
med eCDF diff.....	0.0079365	0.0066667
max eCDF diff.....	0.062932	0.026667
var ratio (Tr/Co).....	0.77756	1.016
T-test p-value.....	0.4319	0.45969
KS Bootstrap p-value..	0.17	0.906
KS Naive p-value.....	0.45305	1
KS Statistic.....	0.062932	0.026667


```
Before Matching Minimum p.value: 0.002
Variable Name(s): risk_averse Number(s): 7

After Matching Minimum p.value: 0.036
Variable Name(s): risk_averse Number(s): 7
```

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```
summary(mout4)
```

```
Estimate... 0.044444
AI SE..... 0.051029
T-stat..... 0.87096
p.val..... 0.38378

Original number of observations..... 1296
Original number of treated obs..... 225
Matched number of observations..... 225
Matched number of observations (unweighted). 225
```

#Social network effect 2: we only examine if they have their network coming or not

Network effect 1-2

- treatment: as long as people have network(network_yes)
- sample: 2nd round with no take-up information given
- matching variables: household characteristics + risk_averse + disaster_prob(exact the same as column 2)

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```
#Data preperation for genetic matching
```

```
dat_gen2 <- cbind(dat_delay_noinfo$male, dat_delay_noinfo$age, dat_delay_noinfo$agpop, dat_delay_noinfo$ricearea_2010, dat_delay_noinfo$literacy, factor(dat_delay_noinfo$village), dat_delay_noinfo$intensive, dat_delay_noinfo$risk_averse, dat_delay_noinfo$disaster_prob, dat_delay_noinfo$network_yes, dat_delay_noinfo$takeup_survey)
```

```
dat_gen2<- as.data.frame(dat_gen2) %>% rename(male=V1, age=V2, agpop = V3, , ricearea_2010=V4, literacy=V5, village=V6, intensive = V7, risk_averse = V8, disaster_prob= V9, network_yes = V10, takeup_survey = V11)
```

```
#drop NAs so we can do genetic matching
```

```
dat_gen2<- dat_gen2 %>% drop_na()
```

```
#Dataframe for matched variable
```

```
X3 <- subset(dat_gen2, select = -c(network_yes,takeup_survey))
```

X3

	ag...	ricearea_2010	literacy	village	intensive	risk_averse	disaster_prob						
	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>						
1	1	54	4	10.00	0	10	0	0.0	20						
2	1	67	2	10.00	1	10	1	0.0	30						
3	1	61	13	4.00	1	10	1	0.0	50						
4	1	45	4	15.00	1	10	0	0.2	50						
5	1	70	7	3.80	1	10	0	0.0	20						
6	1	76	4	5.00	1	10	1	0.2	50						
7	1	59	7	4.00	0	10	1	0.0	20						
8	1	48	6	8.00	1	10	0	0.4	50						
9	1	72	12	4.55	0	10	1	0.0	50						
10	1	48	7	4.50	1	10	0	0.0	20						
1-10 of 1,255 rows						Previous	1	2	3	4	5	6	...	100	Next

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NA

NA

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```
#Genetic Matching
```

```
invisible(capture.output(genout3 <- GenMatch(Tr = dat_gen2$network_yes, X=X3 , M=1, estimand = 'ATT', max.generations = 10)))
```

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```
mout3 <- Match(Y = dat_gen2$takeup_survey, Tr= dat_gen2$network_yes, X=X3, M =2, estimand = 'ATT', Weight.matrix = genout3, caliper = c(1e16, 1e16, 1e16, 0.3,1e16, 1e16, 1e16, 1e16, 1e16), exact = c(0,0,0,0,1,1,0,1,0))
```

```
# mout2 <- Match(Y = round1dat$takeup_survey, Tr= round1dat$intensive, X = X, M =2, estimand = 'ATT', Weight.matrix = genout1, exact = c(1,0,0,0,1,1), ))#
```

```
mbgen3 <- MatchBalance(network_yes~ male + age + agpop + ricearea_2010 + literacy + village + risk_averse + disaster_prob, data = dat_gen2, match.out=mout3, nboots=500)
```

***** (V1) male *****

	Before Matching	After Matching
mean treatment.....	0.92614	0.93929
mean control.....	0.91696	0.95536
std mean diff.....	3.5112	-6.7179
mean raw eQQ diff.....	0.0086505	0.016071
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	1
mean eCDF diff.....	0.0045949	0.0080357
med eCDF diff.....	0.0045949	0.0080357
max eCDF diff.....	0.0091897	0.016071
var ratio (Tr/Co).....	0.89803	1.3371
T-test p-value.....	0.5474	0.29542

***** (V2) age *****

	Before Matching	After Matching
mean treatment.....	51.198	52.079
mean control.....	52.107	52.198
std mean diff.....	-7.8043	-1.0466
mean raw eQQ diff.....	0.97751	0.87321
med raw eQQ diff.....	1	1
max raw eQQ diff.....	4	6
mean eCDF diff.....	0.015231	0.015831
med eCDF diff.....	0.011074	0.013393
max eCDF diff.....	0.041338	0.055357
var ratio (Tr/Co).....	0.98607	1.1487
T-test p-value.....	0.17005	0.84813
KS Bootstrap p-value..	0.476	0.252
KS Naive p-value.....	0.66098	0.35746
KS Statistic.....	0.041338	0.055357

***** (V3) agpop *****

	Before Matching	After Matching
mean treatment.....	5.0502	5.05
mean control.....	4.9204	4.9482
std mean diff.....	5.5303	4.3419
mean raw eQQ diff.....	0.15052	0.15893
med raw eQQ diff.....	0	0
max raw eQQ diff.....	3	3
mean eCDF diff.....	0.0089757	0.009933
med eCDF diff.....	0.0076564	0.011607
max eCDF diff.....	0.022246	0.025

var ratio (Tr/Co).....	1.321	1.3465
T-test p-value.....	0.29502	0.51504
KS Bootstrap p-value..	0.828	0.792
KS Naive p-value.....	0.99785	0.9948
KS Statistic.....	0.022246	0.025

***** (V4) ricearea_2010 *****

	Before Matching	After Matching
mean treatment.....	13.371	9.3604
mean control.....	12.435	9.2888
std mean diff.....	6.7298	1.3227
mean raw eQQ diff.....	1.1971	0.43704
med raw eQQ diff.....	1	0.2
max raw eQQ diff.....	44	3
mean eCDF diff.....	0.026637	0.021942
med eCDF diff.....	0.023367	0.024107
max eCDF diff.....	0.079408	0.055357
var ratio (Tr/Co).....	1.1642	1.0961
T-test p-value.....	0.21658	0.63694
KS Bootstrap p-value..	0.018	0.248
KS Naive p-value.....	0.039201	0.35746
KS Statistic.....	0.079408	0.055357

***** (V5) literacy *****

	Before Matching	After Matching
mean treatment.....	0.7548	0.86071
mean control.....	0.77855	0.86071
std mean diff.....	-5.5156	0
mean raw eQQ diff.....	0.024221	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	1	0
mean eCDF diff.....	0.011873	0
med eCDF diff.....	0.011873	0
max eCDF diff.....	0.023746	0
var ratio (Tr/Co).....	1.0732	1
T-test p-value.....	0.32121	1

***** (V6) village *****

	Before Matching	After Matching
mean treatment.....	20.954	19.864
mean control.....	21.59	19.864
std mean diff.....	-4.8467	0
mean raw eQQ diff.....	0.69377	0

med raw eQQ diff.....	1	0
max raw eQQ diff.....	3	0
mean eCDF diff.....	0.015278	0
med eCDF diff.....	0.010222	0
max eCDF diff.....	0.049613	0
var ratio (Tr/Co).....	1.03	1
T-test p-value.....	0.38851	1
KS Bootstrap p-value..	0.3	1
KS Naive p-value.....	0.42662	1
KS Statistic.....	0.049613	0

***** (V7) risk_averse *****

	Before Matching	After Matching
mean treatment.....	0.15185	0.022143
mean control.....	0.16055	0.022143
std mean diff.....	-3.1041	0
mean raw eQQ diff.....	0.02526	0
med raw eQQ diff.....	0	0
max raw eQQ diff.....	0.2	0
mean eCDF diff.....	0.023296	0
med eCDF diff.....	0.026435	0
max eCDF diff.....	0.040035	0
var ratio (Tr/Co).....	0.85628	1
T-test p-value.....	0.59977	1
KS Bootstrap p-value..	0.198	1
KS Naive p-value.....	0.69967	1
KS Statistic.....	0.040035	0

***** (V8) disaster_prob *****

	Before Matching	After Matching
mean treatment.....	33.996	32.011
mean control.....	33.202	33.018
std mean diff.....	4.6204	-5.9797
mean raw eQQ diff.....	1.0061	1.5786
med raw eQQ diff.....	0	0
max raw eQQ diff.....	20	20
mean eCDF diff.....	0.008897	0.019643
med eCDF diff.....	0.0034359	0.017857
max eCDF diff.....	0.034942	0.041071
var ratio (Tr/Co).....	1.0116	1.1987
T-test p-value.....	0.41329	0.44923
KS Bootstrap p-value..	0.386	0.334
KS Naive p-value.....	0.84101	0.73233
KS Statistic.....	0.034942	0.041071

Before Matching Minimum p.value: 0.018
Variable Name(s): ricearea_2010 Number(s): 4

After Matching Minimum p.value: 0.248
Variable Name(s): ricearea_2010 Number(s): 4

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summary(mout3)

```
Estimate... 0.057143
AI SE..... 0.019358
T-stat..... 2.9519
p.val..... 0.0031583

Original number of observations..... 1255
Original number of treated obs..... 677
Matched number of observations..... 280
Matched number of observations (unweighted). 560

Number of obs dropped by 'exact' or 'caliper' 397
```