

Comparing outcomes between groups

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Seiro Ito

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There are a few key variables.

- **receivedCredit**: The actual treatment status. This is time-invariant. It is T if a subject is receives a loan in our observation period.
- **assignment**: The original treatment assignment. This is time-invariant. This differs from **receivedCredit** because in our design everyone is deemed to get treated at the end (but there are subjects who opted out of a loan but remained in a group).
- **disbursed**: If a subject received a loan. This is time-variant.
- **elapsed**: The number of days since receiving a loan at rd 3 interview date. This is time-invariant as it is computed only at rd 3. This defines the eventual treatment dose and should be our main covariate.

Following results are obtained.

Figure 1 This plots the mean of “3 meals per day” in each round. Left panel is control vs. treated in **receivedCredit** (actual assignment). Right panel is control vs. treated in **assignment** (original assignment). The question is changed in rd 2 onwards so direct comparison across rd 1 and 2,3 are not tenable. However, one sees a rising “3 meals per day”, but almost paralel trend between rd 2 and 3 (which are comparable).

Table 1 This is a first-difference (linear probability) estimation result of “3 meals per day” on **disbursed**, arm, their interactions, **assignment**, **elapsed** days, using the post treatment data of rds 2, 3. It shows positive impacts of receiving a loan (**disbursed** under the covariate name “credit”) aafter controlling for arm. **control/treated** are relative to “lost to flood” or “rejections,” so it is not surprising to have better food intake.

Figure 2 Livestock is the main stated usage of loans.

Figure 3 Work hours seem to get longer.

- Figure 4 New loans increased in rd 2, whose recall period corresponds to the timing disbursement.
- Figure 5 Asset holding by receivedCredit= T/F and rds = 1,2,3. Asset holding is computed with rd 1 asset holding, asset addition in rd 2, 3, while assuming an annual rate of 5% depreciation. receivedCredit not randomised allocation as loan receipt must be agreed by subjects. It shows that the loan receivers have higher mean asset in rd 3, but not in rd 1 or 2, where the latter is good. Red dotted lines are medians, blue dotted lines are means.
- Figure 6 Asset holding by disbursed=T/F and rds=1,2,3. This is also an endogenous switch. The basic picture is the same as Figure 5.
- Figure 7 Asset holding by elapsed days grouped into “early receivers” and “late receivers” according to median elapsed day. This is (roughly) a randomised switch. It shows increasing asset levels, but no mean or median difference between early and late receivers.
- Figure 8 Asset holding by elapsed days and arms. Not much to see here.
- Figure 9 Difference in group-average asset holding between the treated and the control in assignment by difference in elapsed and arms. So it is group differences among original treatment assignment (treated - control) within the same cluster. It controls for cluster FE, and dose and outcome differences are taken between randomly assigned treatment status. This should be one of our main comparisons. (This is not DID: If I plot the first-difference version of the plots between rds, it will be double difference estimates.) When I draw loess curves, I see no trend over various elapsed day (“treatment dose”) differences. It hints a zero gradient of dose levels.
- Figure 10 Same identification idea as Figure 9 on livestock values. Again, we do not see markedly strong impacts, but we see some differences in terms of dispersion. cow arm has smaller variations around the loess curves in rds 2 and 3 relative to the large grace arm. traditional arm has a similar pattern as large grace.
- Figure 11 Livestock holding by elapsed days. Substantial heterogeneity in rd 3 but no statistical significant changes.
- Figure 13 Total asset holding by elapsed days. Substantial heterogeneity in rd 3 but no statistical significant changes.
- Table 3 DID estimation of total asset holding by elapsed days. Zero impact.

I Read

List folder names and read files.

```
setwd(pathsource.mar)
foldername <- list.dirs(path = ".", recursive = T, full.names = T)
foldername <- foldername[!grepl("add|ori|^\\.\\.\\.\\$|1\\$", foldername)]
fn <- unique(list.files(path = foldername, pattern = ".prn$",
  recursive = T, full.names = T))
X = lapply(fn, fread, integer64 = "double")
```

II Treatment through time

Treatment assignment file. There are 220 cases of attrition who are group rejections (140) and lost to flood (80).

```
setwd(pathsave)
```

```
tr <- fread("treatment_assignment.prn")
tr[, disburseDate := as.POSIXct(disburseDate, format = "%Y-%m-%d")]
tr[, purchaseDate := as.POSIXct(purchaseDate, format = "%Y-%m-%d")]
tr[, arm := factor(arm)]
tr[, memstatus := factor(memstatus)]
tr[, assignment := factor(assignment)]
tr0 <- tr[, .(gid, hhid, memstatus, assignment, arm, receivedCredit, elapsed, daysFromSta
table0(tr0[, .(memstatus, assignment))]
```

memstatus	assignment		
	control	treated	
group rejection	140	0	0
individual rejection	0	69	90
lost to flood	80	0	0
new group	0	230	230
old	0	620	599
replacement	0	69	90

```
table0(tr0[, .(arm, assignment)])
```

arm	assignment		
	control	treated	
before intervention	46	0	0
cow	0	254	256
large	40	233	239
large grace	0	239	243
lost to flood	56	0	0
traditional	78	262	271

```
table0(tr0[, .(memstatus, arm)])
```

memstatus	arm					
	before intervention	cow	large	large grace	lost to flood	
group rejection	22	0	40	0	0	
individual rejection	0	72	12	22	0	
lost to flood	24	0	0	0	56	
new group	0	60	100	100	0	
old	0	306	348	338	0	
replacement	0	72	12	22	0	

memstatus	arm	
	traditional	
group rejection	78	
individual rejection	53	
lost to flood	0	
new group	200	
old	227	
replacement	53	

```
tr1 <- tr0[, -grep("Cre", colnames(tr0)), with = F]
```

tr0 and tr1 are based on the information at rd 3.

```
setwd(pathsave)
indate <- fread("interview_dates_long.prn", integer64 = "double")
indate[, intDate := as.POSIXct(intDate, format = "%Y-%m-%d")]
indate[, daysSince2014 :=
  asn(intDate - as.POSIXct("2014-01-01", format = "%Y-%m-%d"))]
indate <- reshape(indate, direction = "wide", idvar = "hhid",
  timevar = "rd",
  v.names = c("disbursed", "purchased", "intDate", "daysSince2014"))
```

Merge interview dates with treatment assignment info tr1.

```
setkey(indate , hhid); setkey(tr0 , hhid); setkey(tr1 , hhid)
tr0 <- indate[tr0]
tr1 <- indate[tr1]
```

```
tr11 <- reshape(tr1 , direction = "long", idvar = c("gid", "hhid",
            "assignment", "arm", "memstatus"), varying = grepout("\\.\\d", colnames(tr0)))
setnames(tr11 , "time", "rd"); setkey(tr11 , hhid , rd)
table(tr11[ ,.(rd , disbursed , assignment)], useNA = "ifany")
```

```
, , assignment =
      disbursed
rd  FALSE TRUE <NA>
1    220    0    0
2    173    0    47
3    168    0    52

, , assignment = control
      disbursed
rd  FALSE TRUE <NA>
1    988    0    0
2    390  409  189
3    148  798   42

, , assignment = treated
      disbursed
rd  FALSE TRUE <NA>
1   1009    0    0
2    127  770  112
3    103  871   35
```

```
table(tr11[ ,.(rd , disbursed , memstatus)], useNA = "ifany")
```

```
, , memstatus = group rejection
      disbursed
rd  FALSE TRUE <NA>
1    140    0    0
2    118    0   22
3    114    0   26

, , memstatus = individual rejection
      disbursed
rd  FALSE TRUE <NA>
1    159    0    0
2    124    0   35
3    130    0   29

, , memstatus = lost to flood
      disbursed
rd  FALSE TRUE <NA>
1     80    0    0
2     55    0   25
3     54    0   26
```

```
, , memstatus = new group

      disbursed
rd  FALSE TRUE <NA>
  1    460    0    0
  2     90   332   38
  3     37   416    7

, , memstatus = old

      disbursed
rd  FALSE TRUE <NA>
  1   1219    0    0
  2    265   756  198
  3     64  1117   38

, , memstatus = replacement

      disbursed
rd  FALSE TRUE <NA>
  1    159    0    0
  2     38    91   30
  3     20   136    3
```

III Food consumption (23B in rd 1, 3B in rds 2, 3)

```
setwd(pathsource.mar)
```

```
grepout("sect.*\\_3b|23b-m|23b.prn", fn)
```

```
[1] "../1/combined/s23b.prn" "../2/section_3b.prn" "../3/section_3b.prn"
```

```
sec3b = copy(X[ grep("sect.*\\_3b|23b-m|23b.prn", fn)])
```

```
setnames(sec3b[[2]], "id", "hhid")
```

```
setnames(sec3b[[3]], "id", "hhid")
```

There is pure duplication in rd 1 files. Drop them.

```
lapply(sec3b[1],
```

```
function(x) x[hhid %in% x[duplicated(x[, .(hhid, mid)]), hhid], ])
```

```
[1]]
      hhid mid s23b_1 s23b_2 s23b_31_fish s23b_32_meat s23b_33_egg s23b_5
1: 9808148207 1      3      2              1          NA           1      2
2: 9808148207 1      3      3              1          NA           2      2
3: 9808148207 2     NA     NA              NA          NA           NA     NA
4: 9808148207 2     NA     NA              NA          NA           NA     NA
5: 9808148207 3     NA     NA              NA          NA           NA     NA
6: 9808148207 3     NA     NA              NA          NA           NA     NA
7: 9808148207 4     NA     NA              NA          NA           NA     NA
8: 9808148207 5     NA     NA              NA          NA           NA     NA
9: 9808148207 6     NA     NA              NA          NA           NA     NA
10: 9808148207 7     NA     NA              NA          NA           NA     NA
11: 9808148220 1      3      2              1          NA           1      2
12: 9808148220 1      3      3              2          NA           2      2
13: 9808148220 2     NA     NA              NA          NA           NA     NA
14: 9808148220 2     NA     NA              NA          NA           NA     NA
15: 9808148220 3     NA     NA              NA          NA           NA     NA
```

16:	9808148220	3	NA	NA	NA	NA	NA	NA
17:	9808148220	4	NA	NA	NA	NA	NA	NA
18:	9808148220	5	NA	NA	NA	NA	NA	NA
	s23b_6	s23b_71_fish	s23b_72_meat	s23b_73_egg	u_id			
1:	1	99	NA	NA	NA			
2:	2	2	NA	NA	9808148480			
3:	NA	NA	NA	NA	NA			
4:	NA	NA	NA	NA	9808148480			
5:	NA	NA	NA	NA	NA			
6:	NA	NA	NA	NA	9808148480			
7:	NA	NA	NA	NA	9808148480			
8:	NA	NA	NA	NA	9808148480			
9:	NA	NA	NA	NA	9808148480			
10:	NA	NA	NA	NA	9808148480			
11:	1	99	NA	NA	NA			
12:	2	1	NA	NA	9808148480			
13:	NA	NA	NA	NA	NA			
14:	NA	NA	NA	NA	9808148480			
15:	NA	NA	NA	NA	NA			
16:	NA	NA	NA	NA	9808148480			
17:	NA	NA	NA	NA	9808148480			
18:	NA	NA	NA	NA	9808148480			

```
sec3b[[1]] <- lapply(sec3b[[1]],
  function(x) x[!duplicated(x[, .(hhid, mid)]), ])
sec3b[[2]] <- sec3b[[2]][!duplicated(hhid), ]
as1(lapply(sec3b[[1]],
  function(x) any(duplicated(x[, .(hhid, mid)]))))
```

```
[1] FALSE
```

There is only one HH that reports food intake of a non-head member. I will drop this non-head member.

```
table0(sec3b[[1]][!is.na(s23b_1), mid])
```

1	4
2216	1

```
sec3b[[1]][hhid %in% hhid[!is.na(s23b_1) & mid != 1], ]
```

	hhid	mid	s23b_1	s23b_2	s23b_31_fish	s23b_32_meat	s23b_33_egg	s23b_5
1:	99081412516	1	3	2	1	NA	NA	2
2:	99081412516	2	NA	NA	NA	NA	NA	NA
3:	99081412516	3	NA	NA	NA	NA	NA	NA
4:	99081412516	4	3	2	1	NA	1	2
	s23b_6	s23b_71_fish	s23b_72_meat	s23b_73_egg	u_id			
1:	1	99	NA	NA	NA			
2:	NA	NA	NA	NA	NA			
3:	NA	NA	NA	NA	NA			
4:	2	1	NA	NA	NA			

```
# drop all non-head members
sec3b[[1]] <- sec3b[[1]][mid == 1, ]
```

Drop hhid = NA.

```
sec3b <- lapply(sec3b, function(x) x[!is.na(hhid), ])
```

Merge treatment assignment info.

```
invisible(lapply(sec3b, setkey, hhid)); setkey(tr0, hhid)
sec3b <- lapply(sec3b, merge, tr0, by = "hhid", all.x = T)
```

Some gids are missing in sec3b. Check if the merge is done correctly. Check if this is due to hhid = 980... cases. Strip leading 980/990 and see if the matched observations have variables originally from tr0.

```
nahhids <- lapply(sec3b, function(x) asn(unique(x[is.na(gid), hhid])))
nahhids <- lapply(nahhids, gsub, pattern = "^980|^990", replacement = "")
nahhids <- lapply(nahhids, asn)
table0(sec3b[[1]][hhid %in% nahhids[[1]], assignment])
```

```
control treated
      7      3
```

```
table0(sec3b[[2]][hhid %in% nahhids[[2]], assignment])
```

```
<NA>
      2
```

```
table0(sec3b[[3]][hhid %in% nahhids[[3]], assignment])
```

```
<NA>
      4
```

Rd 1 seems to be merged OK. Rd 2, 3 show that there are duplicated hhid so drop all entries with duplication.

```
sec3b[[2]] <- sec3b[[2]][!(hhid %in% nahhids[[2]]), ]
sec3b[[3]] <- sec3b[[3]][!(hhid %in% nahhids[[3]]), ]
```

There still remains unmatched observations as seen in NAs in assignment (found in Sec 3B files but not in identification files.) We drop these observations.

```
lapply(sec3b, function(x) table0(x[is.na(gid), assignment]))
```

```
[[1]]
<NA>
      35

[[2]]
<NA>
      4

[[3]]
named integer(0)
```

```
sec3b <- lapply(sec3b, function(x) x[!is.na(gid), ])
asn(lapply(sec3b, dim))
```

```
[1] 2183    38 2072    59 2089    59
```

Three meals. In rd 1, we ask for all the members about the number of times they eat meals, during munga and off-munga seasons. On average, there is only 1 out of 1 HH members repoding to the question, which are all HH head members. In rds 2 and 3, we ask a blanket question if all the members eat three times a day for the whole year. So rd 1 question is more likely to be responded as “3 times” than in rd 2, 3 questions, *cetris paribus*. So observing more “3 times” responses in the latter rds indicate that there may be improvements in household food intake.

Combine rd 1 original and additional into a single file, then put into a list with rds 2, 3.

```
meal3.0 ← c(asn(table0(grepl("3", sec3b[[1]][, s23b_1]) & grepl("3", sec3b[[1]][, s23b_2]),
  grepl("3", sec3b[[1]][, s23b_5]) & grepl("3", sec3b[[1]][, s23b_6])),
  asn(lapply(sec3b[2:3], function(x) table0(grepl("y", x[, s8bq1])))))
# leave monga out
meal3 ← c(asn(table0(grepl("3", sec3b[[1]][, s23b_1]) & grepl("3", sec3b[[1]][, s23b_2])),
  asn(lapply(sec3b[2:3], function(x) table0(grepl("y", x[, s8bq1])))))
meal3 ← matrix(meal3, byrow = T, ncol = 2)
dimnames(meal3) ← list(paste0("rd", 1:3), c("FALSE", "TRUE"))
meal3
```

	FALSE	TRUE
rd1	1926	257
rd2	1218	854
rd3	984	1105

```
iiD1 ← sec3b[[1]][, receivedCredit]
iiD2 ← sec3b[[2]][, receivedCredit]
iiD3 ← sec3b[[3]][, receivedCredit]
iiI1 ← grepl("treated", sec3b[[1]][, assignment])
iiI2 ← grepl("treated", sec3b[[2]][, assignment])
iiI3 ← grepl("treated", sec3b[[3]][, assignment])
meal3D1 ←
  c(asn(table0(grepl("3", sec3b[[1]][iiD1, s23b_1]) & grepl("3", sec3b[[1]][iiD1, s23b_2]),
    asn(table0(grepl("y", sec3b[[2]][iiD2, s8bq1]))),
    asn(table0(grepl("y", sec3b[[3]][iiD3, s8bq1]))))
meal3D0 ←
  c(asn(table0(grepl("3", sec3b[[1]][!iiD1, s23b_1]) & grepl("3", sec3b[[1]][!iiD1, s23b_2]),
    asn(table0(grepl("y", sec3b[[2]][!iiD2, s8bq1]))),
    asn(table0(grepl("y", sec3b[[3]][!iiD3, s8bq1]))))
meal3I1 ←
  c(asn(table0(grepl("3", sec3b[[1]][iiI1, s23b_1]) & grepl("3", sec3b[[1]][iiI1, s23b_2]),
    asn(table0(grepl("y", sec3b[[2]][iiI2, s8bq1]))),
    asn(table0(grepl("y", sec3b[[3]][iiI3, s8bq1]))))
meal3I0 ←
  c(asn(table0(grepl("3", sec3b[[1]][!iiI1, s23b_1]) & grepl("3", sec3b[[1]][!iiI1, s23b_2]),
    asn(table0(grepl("y", sec3b[[2]][!iiI2, s8bq1]))),
    asn(table0(grepl("y", sec3b[[3]][!iiI3, s8bq1]))))
meal3D1 ← matrix(meal3D1, byrow = T, ncol = 2)
meal3D0 ← matrix(meal3D0, byrow = T, ncol = 2)
meal3I1 ← matrix(meal3I1, byrow = T, ncol = 2)
meal3I0 ← matrix(meal3I0, byrow = T, ncol = 2)
dimnames(meal3D1) ← dimnames(meal3D0) ←
dimnames(meal3I1) ← dimnames(meal3I0) ←
  list(paste0("rd", 1:3), c("FALSE", "TRUE"))
meal3DI ← data.table(rbind(reshape(c("D=1", "D=0", "I=1", "I=0"), 2),
  cbind(meal3D1, meal3D0, meal3I1, meal3I0)))
meal3DI
```

	FALSE	TRUE	FALSE	TRUE	FALSE	TRUE	FALSE	TRUE
1:	D=1	D=1	D=0	D=0	I=1	I=1	I=0	I=0
2:	1526	206	400	51	869	118	1057	139
3:	1017	685	201	169	573	394	645	460
4:	793	924	191	181	466	508	518	597

```
#z.0 ← parse(text = "s8bq")
#sec3b[[1]][, zval := eval(z.0)]
```

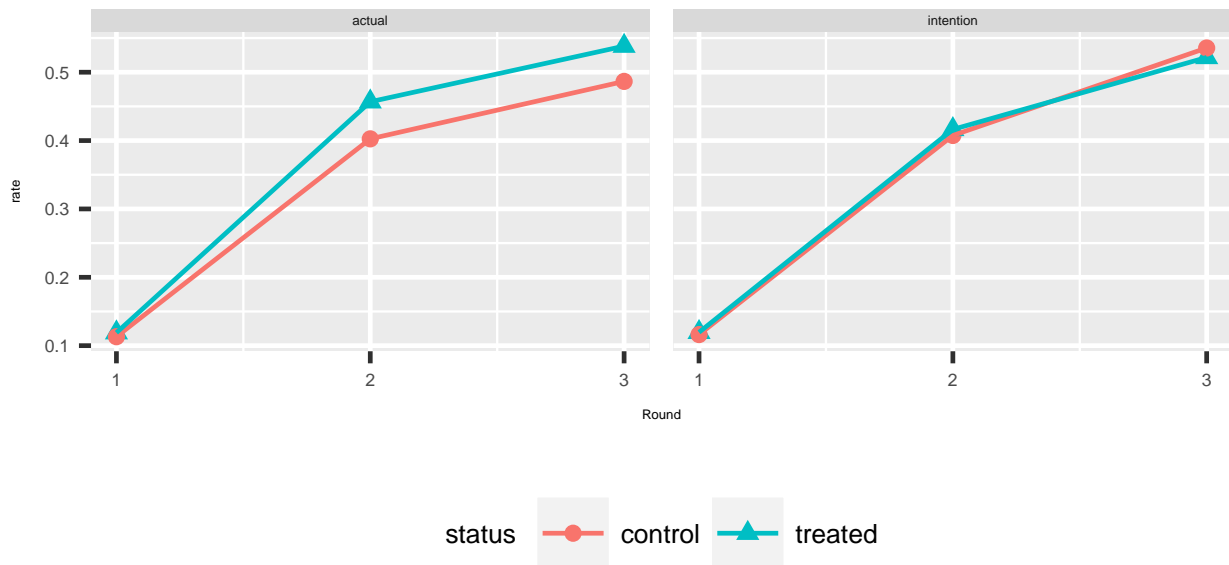



Figure 1 3 meals per day

Given the questions are different, it is not surprising that we have different proportion of subjects with three meals per day. Despite this limitation, we have an increasing food consumption security which is promising.

Form data for regression.

```
s3b <- merge(sec3b[[1]][, .(gid, hhid, s23b_1, s23b_2, arm, assignment,
  disbursed.1, purchased.1, receivedCredit,
  daysFromStart, daysSince2014.1, intDate.1)],
  sec3b[[2]][, .(gid, hhid, s8bq1, arm, assignment,
  disbursed.2, purchased.2, receivedCredit, daysSince2014.2, intDate.2)],
  by = c("gid", "hhid", "arm", "assignment"), all = T, suffixes = c(".1", ".2"))
s3b <- merge(s3b, sec3b[[3]][, .(gid, hhid, s8bq1, arm, assignment,
  disbursed.3, purchased.3, receivedCredit, daysSince2014.3, intDate.3)],
  by = c("gid", "hhid", "arm", "assignment"), all = T, suffixes = c(".2", ".3"))
setnames(s3b, "receivedCredit", "receivedCredit.3")
dim(s3b); dim(s3b <- s3b[!is.na(hhid) | !is.na(gid), ])
```

```
[1] 2215 24
```

```
[1] 2215 24
```

```
s3b[, c("disbursed.1", "purchased.1", "receivedCredit.1") := F]
if (nrow(s3b[is.na(assignment), ]) > 0)
  s3b[is.na(assignment), assignment := "drop out"]
dim(s3b <- s3b[!duplicated(s3b), ])
```

```
[1] 2215 24
```

3 meals per day in regular times for rd 1. For rd 2, 3, yes to the question.

```
s3b[, c("m3.1", "m3.2", "m3.3") :=
  list(grepl("3", s3b[, s23b_1]) & grepl("3", s3b[, s23b_2]),
  grepl("y", s3b[, s8bq1.2]),
  grepl("y", s3b[, s8bq1.3]))]
```

```
s3b[is.na(s23b_1) | is.na(s23b_2), m3.1 := NA]
s3b[is.na(s3b[, s8bq1.2]), m3.2 := NA]
s3b[is.na(s3b[, s8bq1.3]), m3.3 := NA]
```

Rescale days by 100. Note that assignment has empty observations who either group rejected or lost to flood. They form the reference group for assignment (control, treated).

```
s3b[, daysFromStart := daysFromStart/100]
dim(s3b <- s3b[, !grepl("^s\\d", colnames(s3b)), with = F])
```

```
[1] 2215 23
```

```
#s3bl <- reshape(s3b, direction = "long",
#   idvar = c("gid", "hhid", "assignment", "arm"),
#   varying = grepout("\\.\\d", colnames(s3b)))
dim(s3b.comp <- s3b[!is.na(m3.2) & !is.na(m3.3) &
!is.na(receivedCredit.2) & !is.na(receivedCredit.3), ])
```

```
[1] 2043 23
```

```
s3bl <- reshape(s3b.comp, direction = "long",
  idvar = c("gid", "hhid", "assignment", "arm"),
  varying = grepout("\\.\\d", colnames(s3b.comp)))
m3data <- s3bl[time > 1, ]
setkey(m3data, hhid, time)
table(table(m3data[, hhid]))
```

```
2
2043
```

```
m3data[, m3 := m3+0]
m3data[, arm := factor(arm, levels = c("traditional", "large", "large grace", "cow", "lost"))]
m3data[, assigncredit := grepl("tre", assignment) * receivedCredit]
```

m3data: Rd 2-3 data on three meals per day.

```
dm3 <- data.table(m3data[seq(1, nrow(m3data), 2),
  .(gid, arm, assignment, receivedCredit, assigncredit, daysFromStart)],
  m3data[seq(2, nrow(m3data), 2), .(disbursed, purchased, daysSince2014, m3)]-
  m3data[seq(1, nrow(m3data), 2), .(disbursed, purchased, daysSince2014, m3)])
l1 <- glm(m3 ~ arm, data = dm3)
l2 <- glm(m3 ~ assignment, data = dm3)
l3 <- glm(m3 ~ arm*disbursed, data = dm3)
l4 <- glm(m3 ~ assignment + disbursed + assigncredit, data = dm3)
l5 <- glm(m3 ~ arm + daysFromStart, data = dm3)
l6 <- glm(m3 ~ arm*disbursed + daysFromStart, data = dm3)
#p1 <- glm(m3 ~ arm, family=binomial(link="probit"), data = m3data)
linprob <- list(l1, l2, l3, l4, l5, l6)
linest <- lapply(linprob, clx.regobj, Cluster = "gid")
linest <- lapply(linest, function(x) x[, -3])
linest <- tabs2latex(linest)
R2 <- round(asn(lapply(linprob,
  function(x) 1-crossprod(summary(x)$deviance.res)/summary(x)$null.dev)), 3)
en <- asn(lapply(linprob, function(x) length(x$y)))
rn <- rownames(linest)
rn <- gsub("arm|assignment|^se.*", "", rn)
```

TABLE 1: FD ESTIMATES OF THREE MEALS PER DAY, ROUND 2, 3

rn	(1)	(2)	(3)	(4)	(5)	(6)
(Intercept)	0.118* (0.062)	-0.060 (0.111)	0.081 (0.066)	-0.060 (0.111)	0.141** (0.055)	0.051 (0.055)
large	-0.014 (0.099)		0.067 (0.096)		-0.008 (0.097)	0.124 (0.086)
large grace	0.093 (0.090)		0.162 (0.098)		0.087 (0.084)	0.197** (0.091)
cow	-0.079 (0.101)		-0.051 (0.101)		-0.069 (0.098)	-0.009 (0.098)
lost to flood	-0.043 (0.133)		-0.005 (0.134)			
control		0.211* (0.117)		0.236** (0.119)		
treated		0.176 (0.117)		0.178 (0.141)		
credit			0.147** (0.064)	-0.062 (0.062)		0.142** (0.065)
large * credit			-0.367** (0.126)			-0.423** (0.125)
large grace * credit			-0.230** (0.090)			-0.265** (0.091)
cow * credit			-0.108 (0.134)			-0.122 (0.136)
treated * credit				-0.006 (0.086)		
elapsed days * 100					-0.001 (0.006)	0.006 (0.009)
R^2	0.01	0.009	0.022	0.011	0.009	0.029
n	2043	2043	1838	1838	1657	1527

- Notes: 1. First-difference estimates of having three meals per day using rd 2 and 3 information. Standard errors are clustered at the group level.
2. large, large grace, cow, lost to flood, control, treated, credit are all time invariant and are interacted with a trend term. Regressions (1) - (4) include subjects who group-rejected or lost to flood as a reference group. Regressions (5) - (6) drop subjects who group-rejected or lost to flood and use the subjects who were initially assigned to the control as a reference group.
3. *, **, *** indicate significance levels at 10%, 5%, 1%, respectively.

```

rn ← gsub("assignedcredit", "treated * credit", rn)
rn ← gsub("disbursed", "credit", rn)
rn ← gsub(":", " * ", rn)
rn ← gsub("daysFromStart", "elapsed days * 100", rn)
ltab ← rbind(as.matrix(cbind(rn, linest)), c("$R^{2}$", R2),
             c("$n$", en))
write.tablev(latextab(ltab, delimiterline = NULL, alternatelinecolor2 = "gray90",
                    hleft = c("\\footnotesize", rep("\\scriptsize\\hfil$", ncol(ltab)-1)),
                    hcenter = c(2.2, rep(1.25, ncol(ltab)-1)),
                    hright = c("\\hfill", rep("$", ncol(ltab)-1)),
                    adjustlineskip = "-.4ex"),
            paste0(pathsave, "3meals.tex"), colnamestrue = F)

```

We see no impacts of intervention when comparing two periods after the disbursement.

IV Credit use

```

cruse.files ← grepout("21", fn)
cruse.files ← cruse.files[!grepl("com", cruse.files)]

```

File names of rd 3 files are named for the page ordering. For example, ./2/section.21a.prn are the first 2 questions of Section 20, which is named as 21 as it is an unnumbered page that comes right after Section 20. ./2/section.22a.prn is Section 18.

```
setwd(pathsource.mar)
fread(cruse.files[1], integer64 = "double")
```

	id	s11q1	s11q2
1:	7010102	Yes	Yes
2:	7010105	Yes	Yes
3:	7010106	Yes	Yes
4:	7010107	Yes	Yes
5:	7010108	Yes	Yes

2079:	99081912415	Yes	Yes
2080:	99081912417	Yes	Yes
2081:	99081912418	Yes	Yes
2082:	99081912419	Yes	Yes
2083:	99081912420	Yes	Yes

```
fn21 ← grepout("2/section_23.prn|3/section_21", fn)
```

In rd 2, Section 21 is stored under ./2/section_23.prn, in rd 3, ./3/section_21_use_of_credit_1.prn, ./3/section_21_use_of_credit_2.prn.

```
setwd(pathsource.mar)
foldername ← list.dirs(path = ".", recursive = T, full.names = T)
foldername ← foldername[!grepl("add|ori|^\\.\\.\\.\\$|1\\$", foldername)]
fn ← unique(list.files(path = foldername, pattern = ".prn$",
  recursive = T, full.names = T))
X = lapply(fn, fread, integer64 = "double")
```

```
Cr = copy(X[fn %in% fn21])
Cr ← lapply(Cr, function(x) if (any(grepl("^id$", colnames(x))))
  setnames(x, "id", "hhid") else x)
invisible(lapply(Cr, setkey, hhid))
Cr2 ← Cr[[1]]
setnames(Cr2, colnames(Cr2)[-1],
  paste0("V", 1:(ncol(Cr2)-1), "_", colnames(Cr2)[-1]))
setnames(Cr2, colnames(Cr2),
  gsub("seca3_", "", colnames(Cr2)))
setnames(Cr2, colnames(Cr2),
  gsub("-\\d-|_q\\d-|_--", "_", colnames(Cr2)))
setnames(Cr2, colnames(Cr2),
  gsub("-\\d-", "_", colnames(Cr2)))
Cr3 ← Cr[[2]][Cr[[3]]]
setnames(Cr3, colnames(Cr3)[-1],
  paste0("V", 1:(ncol(Cr3)-1), "_", colnames(Cr3)[-1]))
setnames(Cr3, colnames(Cr3),
  gsub("_q.*?_([a-z])", "_\\1", colnames(Cr3)))
setnames(Cr3, colnames(Cr3),
  gsub("_a-", "_", colnames(Cr3)))
Cr3 ← Cr3[!is.na(hhid), ]
setkey(Cr3, hhid, V5_from_when_you_started_1,
  V6_from_when_you_started_2)
#setnames(Cr3, colnames(Cr3)[-1],
#  paste0("V", putzeroontop(1:(ncol(Cr3)-1), totaldigits = 2)))
Cr3[, iga := 1:N, by = hhid]
Cr3[, igas := .N, by = hhid]
setkey(Cr3, hhid, iga)
```

Merge rd 2 and 3.

```

setnames(Cr2, grepout("from.oth", colnames(Cr2)), "loanFromOther")
setnames(Cr3, grepout("from.oth", colnames(Cr3)), "loanFromOther")
setnames(Cr3, grepout("deta.*j$", colnames(Cr3)), "igaContent")
setnames(Cr3, grepout("deta.*i$", colnames(Cr3)), "specify")
setnames(Cr3, grepout("am.*ed", colnames(Cr3)), "investValue")
setnames(Cr3, grepout("start.*_1", colnames(Cr3)), "startY")
setnames(Cr3, grepout("start.*_2", colnames(Cr3)), "startM")
setnames(Cr3, grepout("du", colnames(Cr3)), "investDuration")
setnames(Cr3, grepout("8.*othe.*g", colnames(Cr3)), "investSame")
setnames(Cr3, grepout("9.how.*", colnames(Cr3)), "investSameNum")
setnames(Cr3, grepout("0.*any.*g", colnames(Cr3)), "investSameExper")
setnames(Cr3, grepout("1.how.*", colnames(Cr3)), "investSameExperNum")
setnames(Cr3, grepout("2.*still", colnames(Cr3)), "investSameStill")
Cr2 <- Cr2[!is.na(hhid), ]
Cr3 <- Cr3[!is.na(hhid), ]
lapply(list(Cr2, Cr3), colnames)

```

```

[[1]]
[1] "hhid"
[2] "loanFromOther"
[3] "V2_way_of_using_guk_credit"
[4] "V3_use1"
[5] "V4_use2"
[6] "V5_use3"
[7] "V6_other_specifyc"
[8] "V7_credit_usage1"
[9] "V8_plan_to_repay"
[10] "V9_plan1"
[11] "V10_plan2"
[12] "V11_plan3"
[13] "V12_other_specifyd"
[14] "V13_years_income_expectation"
[15] "V14_expected_income"
[16] "V15_how_to_spend_extra_income_if_any"
[17] "V16_how_to_spend_extra_income_if_anz"
[18] "V17_ow_to_spend_extra_income_if_anya"
[19] "V18_how_to_spend_specify"
[20] "V19_hh_word_hours_increase"
[21] "V20_other_members_work_hours_increas"
[22] "V21_other_members_work_hours_increat"
[23] "V22_other_members_work_hours_decreas"
[24] "V23_other_members_work_hours_decreat"
[25] "V24_other_members_work_hours_same_mi"
[26] "V25_other_members_work_hours_same_mj"

[[2]]
[1] "hhid"                "loanFromOther"      "igaContent"
[4] "specify"              "investValue"        "startY"
[7] "startM"              "investDuration"     "investSame"
[10] "investSameNum"       "investSameExper"    "investSameExperNum"
[13] "investSameStill"     "iga"                "igas"

```

```

setwd(pathsave)
write.tablev(Cr2, "credit_use_rd_2.prn")
write.tablev(Cr3, "credit_use_rd_2.prn")

```

Intended use of credit, mostly livestock (cows). It is interesting to note that the majority of our subjects choose livestock for an investment.

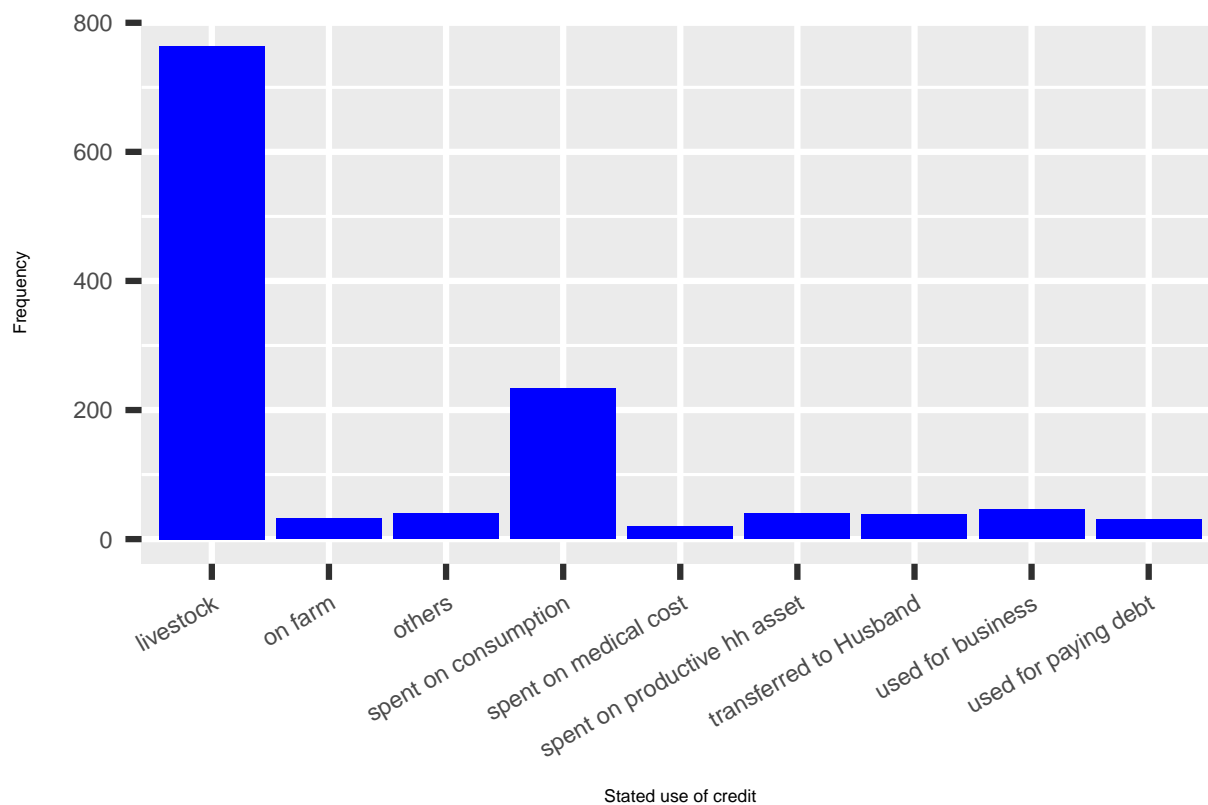


Figure 2 Stated use of credit in rd 2

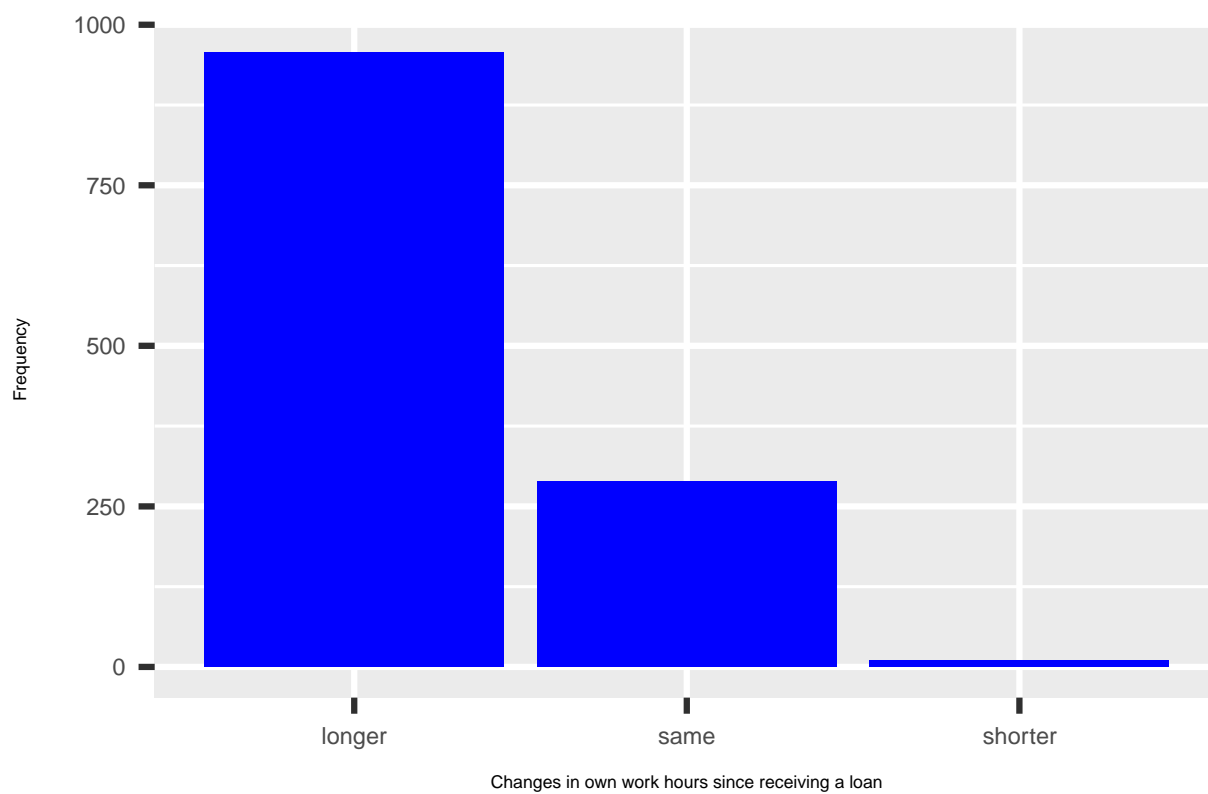


Figure 3 Work hours

Work hours.

V New loans

Loans in rds 1, 2, and 3.

```
(fn20c ← grepout("2/s.*20c|3/s.*19c", fn))
```

```
[1] "./2/section_20c.prn" "./3/section_19c.prn"
```

```
(fn19.1 ← grepout("d/s.*19", fn))
```

```
[1] "./1/combined/s19.prn"
```

```
bol = copy(X[[which(fn %in% fn19.1)]])
bol ← bol[!is.na(s19_1_1), ]
bol ← bol[!duplicated(hhid), ]
lendersin1 ← c("other", "relative", "moneylender")
setnames(bol, paste0("s19_", repseq(1:3, 5), "_", rep(1:5, 3)),
          paste0(c("ask", "askAmount", "cashAmount", "interest", "usage"), "."),
          repseq(lendersin1, 5))
bol1 ← reshape(bol, direction = "long", idvar = "hhid",
               varying = grepout("\\.", colnames(bol)))
setnames(bol1, "time", "lender")
setkey(bol1, hhid, lender)
bol1[, totalSum := sum(cashAmount, na.rm = T), by = hhid]
```

```
Warning in `[.data.table`(bol1, , `:=`(totalSum, sum(cashAmount, na.rm = T)), : Invalid .i
```

```
bol1[grepl("oth", lender), lender := "other NGO/MFI"]
bol1[grepl("rel", lender), lender := "friends , relatives"]
bol1[grepl("mo", lender), lender := "money lenders"]
Bo1 ← cbind(rd = 1, bol1)
```

In rd 1, there are only 14 subjects who have borrowed from other NGO/MFI in the last 12 months. Most of the loans are taken from friends, relatives and money lenders, for about 9%, 13% of subjects, respectively.

```
bor1 ← by(bol1[, cashAmount], bol1[, lender], destat)
bor1des ← data.frame(rbindlist(lapply(bor1, function(x) data.table(t(matrix(x))))))
dimnames(bor1des) ← list(names(bor1), colnames(bor1[[1]]))
bor1des
```

	min	25\\%	median	75\\%	max	mean	std	0s	NAs	n
friends , relatives	100	500	1000	2000	35000	1923.1	2800.3	0	1932	2218
money lenders	100	1500	2000	4000	30000	3344.6	3517.8	0	2041	2218
other NGO/MFI	500	1850	2500	6750	50000	7914.3	13634.9	0	2204	2218

```
Bo = copy(X[fn %in% fn20c])
Bo ← lapply(Bo, function(x) if (any(grepl("^id$", colnames(x))))
            setnames(x, "id", "hhid") else x)
invisible(lapply(Bo, setkey, hhid))
Bo ← rbindlist(list(data.table(rd = 2, Bo[[1]]), data.table(rd = 3, Bo[[2]])))
Bo ← Bo[!is.na(hhid), ]
Bo[, bo := 1:N, by = c("hhid", "rd")]
setkey(Bo, hhid, rd, bo)
Bo[, inkindAmount := (in_kind_amount_4_1) * (in_kind_price_4_1)]
```

```

Bo[is.na(inkindAmount), inkindAmount := 0]
Bo[is.na(cash_tk_4_1), cash_tk_4_1 := 0]
Bo[, cashAmount := cash_tk_4_1]
Bo[, totalSum := sum(cashAmount+inkindAmount), by = c("hhid", "rd")]
Bo[, purpose := pur_loan_4_1]
Bo[grepl("other", purpose), purpose := purpose_of_the_loan_specify_4_1]
Bo[grepl("cow|COW|cuw|cou ?bu|cow ?bu|cwo|gow|cokw|coe|coy|ci=ow", purpose),
  purpose := "buying cows"]
Bo[grepl("goa?t|goad|goot", purpose), purpose := "buying goats"]
Bo[grepl("shee|shepp", purpose), purpose := "buying sheep"]
Bo[grepl("boa?t|boad|ship", purpose), purpose := "buying a boat"]
Bo[grepl("land|lond|lnad", purpose), purpose := "buy/leasing in land"]
Bo[grepl("house", purpose), purpose := "buying a house"]
Bo[grepl("eremny|dowry", purpose), purpose := "ceremony, dowry"]
Bo[grepl("mach", purpose), purpose := "buying machines"]
Bo[grepl("buss?inn?es|trade", purpose), purpose := "business investment"]
table0(Bo[, loan_taken_from_4_1])

```

	0
799	1
Commercial Banks	Government Banks
1	1
Grameen Bank	Money lender
24	146
Non-relatives in village	Nonrelatives out of village
16	5
Relatives in village	Relatives out of village
540	15
Shop owner	Trader
694	70
co-operatives	other NGO's(specify)
8	2974
other(specify)	
67	

```

Bo[, lender := tolower(loan_taken_from_4_1)]
Bo[grepl("rela", lender), lender := "friends, relatives"]
Bo[grepl("mo", lender), lender := "money lenders"]
Bo[grepl("0", lender), lender := ""]
Bo[lender == "", lender := NA]
Bo[grepl("sho|tr", lender), lender := "shop owners, traders"]
Bo[grepl("gra|other|co-|ban", lender) | grepl("bra", loan_taken_from_specify_4_1),
  lender := "other NGO/MFI"]
Bo[grepl("GUK|guk|ugk", loan_taken_from_specify_4_1), lender := "GUK"]
table0(Bo[, lender])

```

GUK	friends, relatives	money lenders
552	576	146
other NGO/MFI	shop owners, traders	<NA>
2523	764	800

```
table0(Bo[grepl("other", lender), loan_taken_from_specify_4_1])
```

	.	brac	cow	buy
2520	1	1	1	


```
table0(Bo[grepl("other", lender), cashAmount])
```

0	300	400	500	600	1000	1500	1600	2000	2500	3000
10	1	1	3	2	4	5	1	9	1	10
3500	3800	4000	4500	5000	5600	6000	6500	7000	7500	7800
6	1	10	2	291	4	14	1	179	1	1
8000	9000	10000	11300	12000	13000	15000	16000	16800	18000	20000
39	2	25	1	9	2	1867	1	8	1	5
22000	70000	80000	150000							
1	1	1	3							

Append rd 1.

```
setkey(Bo, hhid, rd); setkey(Bo1, hhid, rd)
```

```
Bo13 ← rbind(Bo1, Bo, fill = T)
```

```
setkey(Bo13, hhid, rd, lender)
```

Merge treatment info.

```
tr01 ← reshape(tr0, direction = "long",
  idvar = c("gid", "hhid", "memstatus", "assignment", "arm"),
  varying = grepout("\\.\\d", colnames(tr0)))
```

```
setnames(tr01, "time", "rd")
```

```
setkey(Bo13, hhid, rd); setkey(tr01, hhid, rd)
```

```
Bot ← tr01[Bo13]
```

```
by(Bot[grepl("other|G", lender), cashAmount], Bot[grepl("other|G", lender), rd], destat)
```

```
Bot[grepl("other|G", lender), rd]: 1
  min 25\\% median 75\\%   max   mean     std 0s NAs   n
v.1. 500  1850   2500  6750 50000 7914.3 13634.9 0 2204 2218
-----
Bot[grepl("other|G", lender), rd]: 2
  min 25\\% median 75\\%   max   mean     std 0s NAs   n
v.1.  0 15000  15000 15000 150000 13085.9 5958  4  0 1285
-----
Bot[grepl("other|G", lender), rd]: 3
  min 25\\% median 75\\%   max   mean     std 0s NAs   n
v.1.  0  7000  15000 15000 150000 12690.9 6282.6 6  0 1790
```

```
Bot[, combined.lender := lender]
```

```
Bot[grepl("oth|GU", lender), combined.lender := "NGO/MFI"]
```

```
Bot[grepl("shop", lender), combined.lender := "money lenders"]
```

```
setwd(pathsave)
```

```
write.tablev(Bot, "borrowing_rd_1-3.prn")
```

Plot new loans in each rd. I will combine shop owners/traders with money lenders. I will also combine GUK and other NGO/MFI to NGO/MFI. We also omit zero borrowing from the histogram for clarity.

```
#Bot[, totalSum0 := totalSum]
```

```
#Bot[totalSum == 0, totalSum0 := NA]
```

```
library(ggplot2)
```

```
ggplot(data = subset(Bot, cashAmount > 0), aes(x = cashAmount, fill = combined.lender)) +
  geom_histogram(bins = 20) +
  scale_x_continuous(limits = c(0, 30000)) +
  scale_y_continuous(limits = c(0, 1000)) +
  ylab("Frequency") + xlab("Borrowing (Tk)") + labs(fill = "lenders") +
  facet_wrap(~ rd) +
  theme(axis.title.y = element_text(size = rel(.25), angle = 90),
```

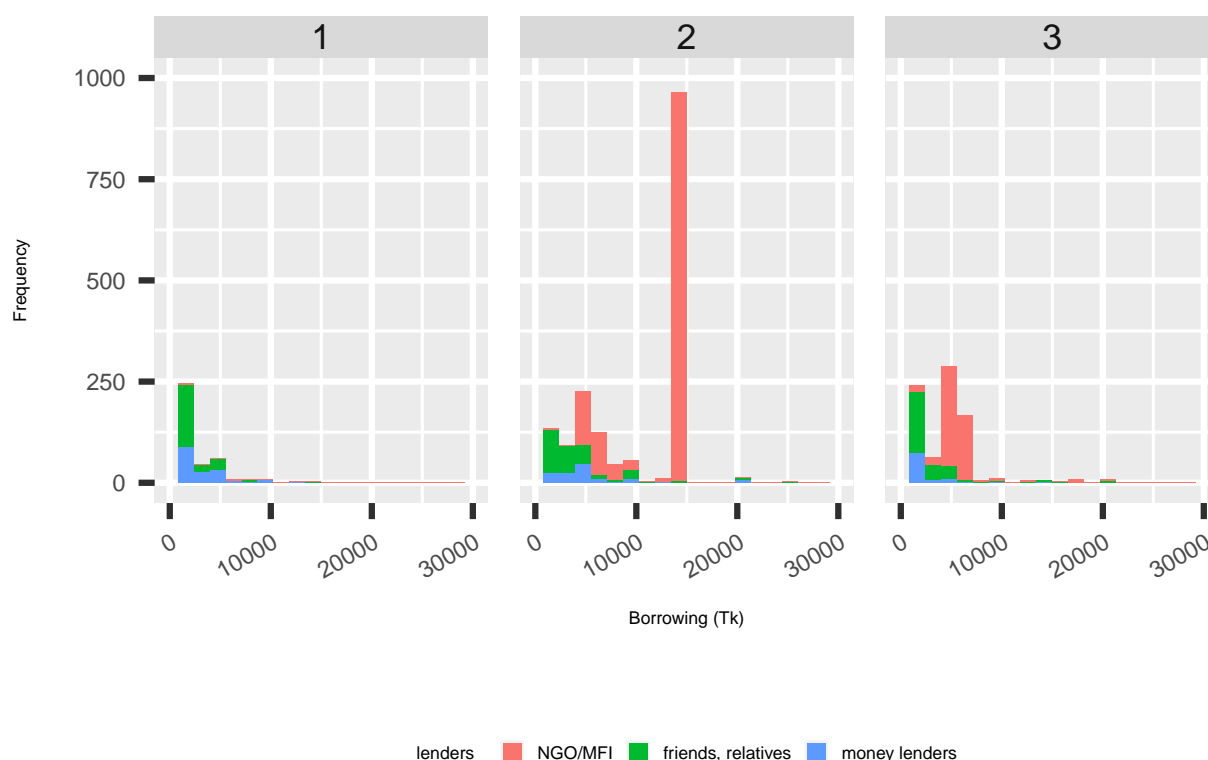


Figure 4 New loans

```
axis.title.x = element_text(size = rel(.25), angle = 0),
axis.text.x = element_text(size = rel(.5), angle = 30, hjust = 1),
axis.text.y = element_text(size = rel(.5), angle = 0),
legend.text = element_text(size=rel(.25)),
legend.position = "bottom",
legend.title = element_text(size = rel(.25)),
legend.key = element_rect(size = rel(.25)),
legend.key.size = unit(.15, "cm"),
strip.text = element_text(size=rel(.5)),
strip.text.x = element_text(margin = margin(.05, 0, .05, 0, "cm"))))
```

One can see that, in rd 1, there is virtually no borrowing from NGO/MFI among our subjects. This indicates that our study areas are relatively free from other non-indigenous financial intermediaries which allows us to estimate the impacts of our loans without much concerns of treatment contamination. In rd 2, borrowing from NGO/MFI increased rapidly as a result of our intervention. In rds 2 and 3, some individuals report smaller amount, which correspond to our traditional loan arm. It is hard to say that the loans from friends, relatives or money lenders have decreased after our intervention between rd 1 and rd 2.

VI Assets

Read files.

```
(fn.asset ← grepout("d/s.*14a|d/s.*14b|2/s.*15|3/s.*13", fn))
```

```
[1] "../1/combined/s14a.prn" "../1/combined/s14b.prn" "../2/section_15a.prn"
[4] "../2/section_15b.prn"   "../3/section_13a.prn"   "../3/section_13b.prn"
```

```

setwd(pathsource.mar)
As = copy(X[fn %in% fn.asset])
As <- lapply(As, function(x) if (any(grepl("^id$", colnames(x))))
             setnames(x, "id", "hhid") else x)
As <- lapply(As, function(x) x[!duplicated(x), ])
As <- lapply(As, function(x) x[!is.na(hhid), ])
invisible(lapply(As, setkey, hhid))
invisible(lapply(As[1:2], setkey, hhid, mid))

```

Separate into rds for rd-specific operations (to be merged back later).

```

As01 <- As[[2]][As[[1]]]
As02 <- As[3:4]
As03 <- As[5:6]

```

Rd 1.

```

As11 <- As01[, grepout("hhid|s14a", colnames(As01)), with = F]
setnames(As11, colnames(As11),
         gsub("s14a_(\\d)_1$", "item.\\1", colnames(As11)))
setnames(As11, colnames(As11),
         gsub("s14a_(\\d)_2$", "own.\\1", colnames(As11)))
setnames(As11, colnames(As11),
         gsub("s14a_(\\d)_3$", "value.\\1", colnames(As11)))
summary(As11[duplicated(As11), ])

```

hhid	item.1	own.1	value.1
Min. :7.01e+06	Length:5648	Length:5648	Min. :1200
1st Qu.:7.04e+06	Class :character	Class :character	1st Qu.:1500
Median :7.12e+06	Mode :character	Mode :character	Median :1800
Mean :1.44e+10			Mean :1700
3rd Qu.:9.81e+09			3rd Qu.:2000
Max. :9.91e+10			Max. :2000
			NA's :5644
item.2	own.2	value.2	item.3
Length:5648	Length:5648	Min. : NA	Length:5648
Class :character	Class :character	1st Qu.: NA	Class :character
Mode :character	Mode :character	Median : NA	Mode :character
		Mean :NaN	
		3rd Qu.: NA	
		Max. : NA	
		NA's :5648	
own.3	value.3	item.4	own.4
Length:5648	Min. : NA	Length:5648	Length:5648
Class :character	1st Qu.: NA	Class :character	Class :character
Mode :character	Median : NA	Mode :character	Mode :character
	Mean :NaN		
	3rd Qu.: NA		
	Max. : NA		
	NA's :5648		
value.4			
Min. : NA			
1st Qu.: NA			
Median : NA			
Mean :NaN			
3rd Qu.: NA			
Max. : NA			
NA's :5648			

```
As11[grepl("8207|8220|9416|212016", hhid) & item.1 != "", ]
```

	hhid	item.1	own.1	value.1	item.2	own.2	value.2	item.3
1:	9808148207	566	1	1200				NA
2:	9808148207	566	1	1200				NA
3:	9808148207	Tube well for drinking	Yes	2000				NA
4:	9808148207	Tube well for drinking	Yes	2000				NA
5:	9808148220	566	1	2000				NA
6:	9808148220	566	1	2000				NA
7:	9808148220	Tube well for drinking	Yes	1600				NA
8:	9808148220	Tube well for drinking	Yes	1600				NA
9:	9908169416	566	1	2000	567	1	200	
10:	9908169416	566	1	2000				NA
	own.3	value.3	item.4	own.4	value.4			
1:		NA		NA				
2:		NA		NA				
3:		NA		NA				
4:		NA		NA				
5:		NA		NA				
6:		NA		NA				
7:		NA		NA				
8:		NA		NA				
9:		NA		NA				
10:		NA		NA				

```

As11[grepl(8207, hhid), item.2 := .SD[,N, item.1]]
As11[grepl(8207, hhid), own.2 := .SD[,N, own.1]]
As11[grepl(8207, hhid), value.2 := .SD[,N, value.1]]
As11[grepl(8220, hhid), item.2 := .SD[,N, item.1]]
As11[grepl(8220, hhid), own.2 := .SD[,N, own.1]]
As11[grepl(8220, hhid), value.2 := .SD[,N, value.1]]
As11 ← As11[!duplicated(As11[, hhid]), ]
As12 ← As01[, grepout("hhid|s14b", colnames(As01)), with = F]
setnames(As12, colnames(As12),
  gsub("s14b_(\\d)_1$", "item.\\1", colnames(As12)))
setnames(As12, colnames(As12),
  gsub("s14b_(\\d)_2$", "own.\\1", colnames(As12)))
setnames(As12, colnames(As12),
  gsub("s14b_(\\d)_4$", "value.\\1", colnames(As12)))
setnames(As12, colnames(As12),
  gsub("s14b_(\\d)_3$", "ownership.\\1", colnames(As12)))
setnames(As12, colnames(As12),
  gsub("s14b_(\\d)_5$", "rental.\\1", colnames(As12)))
summary(As12[duplicated(As12), ])

```

hhid	item.1	own.1	ownership.1
Min. :7.01e+06	Length:4940	Length:4940	Min. : 1.0
1st Qu.:7.04e+06	Class :character	Class :character	1st Qu.: 75.2
Median :7.13e+06	Mode :character	Mode :character	Median :100.0
Mean :1.48e+10			Mean : 75.2
3rd Qu.:9.81e+09			3rd Qu.:100.0
Max. :9.91e+10			Max. :100.0
			NA's :4936
value.1	rental.1	item.2	own.2
Min. : 300	Min. : NA	Length:4940	Length:4940
1st Qu.: 345	1st Qu.: NA	Class :character	Class :character
Median : 380	Median : NA	Mode :character	Mode :character
Mean : 665	Mean :NaN		
3rd Qu.: 700	3rd Qu.: NA		
Max. :1600	Max. : NA		

```

NA's      :4936    NA's      :4940
ownership.2    value.2    rental.2    item.3
Min.      :100    Min.      :400    Min.      : NA    Length:4940
1st Qu.:100    1st Qu.:400    1st Qu.: NA    Class :character
Median :100    Median :400    Median : NA    Mode  :character
Mean      :100    Mean      :400    Mean      :NaN
3rd Qu.:100    3rd Qu.:400    3rd Qu.: NA
Max.      :100    Max.      :400    Max.      : NA
NA's      :4939    NA's      :4939    NA's      :4940
own.3        ownership.3    value.3    rental.3
Length:4940    Min.      : NA    Min.      : NA    Mode:logical
Class :character    1st Qu.: NA    1st Qu.: NA    NA's:4940
Mode  :character    Median : NA    Median : NA
Mean      :NaN    Mean      :NaN
3rd Qu.: NA    3rd Qu.: NA
Max.      : NA    Max.      : NA
NA's      :4940    NA's      :4940
item.4        own.4        ownership.4    value.4
Length:4940    Length:4940    Min.      : NA    Min.      : NA
Class :character    Class :character    1st Qu.: NA    1st Qu.: NA
Mode  :character    Mode  :character    Median : NA    Median : NA
Mean      :NaN    Mean      :NaN
3rd Qu.: NA    3rd Qu.: NA
Max.      : NA    Max.      : NA
NA's      :4940    NA's      :4940
rental.4
Mode:logical
NA's:4940

```

```

As12[grepl(8207, hhid), item.2 := .SD[,N, item.1]]
As12[grepl(8207, hhid), own.2 := .SD[,N, own.1]]
As12[grepl(8207, hhid), ownership.2 := .SD[,N, ownership.1]]
As12[grepl(8207, hhid), value.2 := .SD[,N, value.1]]
As12[grepl(8220, hhid), item.2 := .SD[,N, item.1]]
As12[grepl(8220, hhid), own.2 := .SD[,N, own.1]]
As12[grepl(8220, hhid), ownership.2 := .SD[,N, ownership.1]]
As12[grepl(8220, hhid), value.2 := .SD[,N, value.1]]
As12[grepl(8220, hhid), item.3 := .SD[,N, item.2]]
As12[grepl(8220, hhid), own.3 := .SD[,N, own.2]]
As12[grepl(8220, hhid), ownership.3 := .SD[,N, ownership.2]]
As12[grepl(8220, hhid), value.3 := .SD[,N, value.2]]
As12 <- As12[!duplicated(As12[, hhid]), ]
setnames(As11, colnames(As11), gsub("\\.(\\w)", ".1\\1", colnames(As11)))
setnames(As12, colnames(As12), gsub("\\.(\\w)", ".2\\1", colnames(As12)))
As11 <- reshape(As11, direction = "long", idvar = "hhid",
  varying = grepout("\\.\\d", colnames(As11)))
As12 <- reshape(As12, direction = "long", idvar = "hhid",
  varying = grepout("\\.\\d", colnames(As12)))
As1 <- rbind(As11, As12, fill = T)
As1[, time := NULL]
As1 <- As1[!(is.na(item) | item == ""), ]
As1[, assetNumber := 1:N, by = hhid]
setkey(As1, hhid, assetNumber)
As1[, totalSum := sum(value, na.rm = T), by = hhid]

```

Rd 2.

```

As21 <- As02[[1]]
setnames(As21, colnames(As21),

```

```

      gsub("sec21_item.*$", "item.1", colnames(As21)))
setnames(As21, colnames(As21),
      gsub("sec21_oth.*$", "specify.1", colnames(As21)))
setnames(As21, colnames(As21),
      gsub("cu.*$", "currentStatus.1", colnames(As21)))
setnames(As21, colnames(As21),
      gsub("dec.*$", "amount.1", colnames(As21)))
setnames(As21, colnames(As21),
      gsub("^ta.*$", "value.1", colnames(As21)))
setnames(As21, colnames(As21),
      gsub("^pu.*$", "lastYear.1", colnames(As21)))
summary(As21[duplicated(As21), ])

```

hhid	item.1	specify.1	currentStatus.1
Min. : NA	Length:0	Length:0	Length:0
1st Qu.: NA	Class :character	Class :character	Class :character
Median : NA	Mode :character	Mode :character	Mode :character
Mean :NaN			
3rd Qu.: NA			
Max. : NA			
amount.1	value.1	lastYear.1	
Min. : NA	Min. : NA	Length:0	
1st Qu.: NA	1st Qu.: NA	Class :character	
Median : NA	Median : NA	Mode :character	
Mean :NaN	Mean :NaN		
3rd Qu.: NA	3rd Qu.: NA		
Max. : NA	Max. : NA		

```

As22 = As02[[2]]
setnames(As22, colnames(As22),
      gsub("sec22_co.*$", "item.2", colnames(As22)))
setnames(As22, colnames(As22),
      gsub("sec22_oth.*$", "specify.2", colnames(As22)))
setnames(As22, colnames(As22),
      gsub("^cu.*$", "currentStatus.2", colnames(As22)))
setnames(As22, colnames(As22),
      gsub("^how.*$", "amount.2", colnames(As22)))
setnames(As22, colnames(As22),
      gsub(".*po.*$", "ownership.2", colnames(As22)))
setnames(As22, colnames(As22),
      gsub(".*taka.*$", "value.2", colnames(As22)))
setnames(As22, colnames(As22),
      gsub(".*rented.*$", "rental.2", colnames(As22)))
summary(As22[duplicated(As22), ])

```

hhid	item.2	specify.2	currentStatus.2
Min. : NA	Length:0	Length:0	Length:0
1st Qu.: NA	Class :character	Class :character	Class :character
Median : NA	Mode :character	Mode :character	Mode :character
Mean :NaN			
3rd Qu.: NA			
Max. : NA			
amount.2	ownership.2	value.2	rental.2
Min. : NA	Min. : NA	Min. : NA	Min. : NA
1st Qu.: NA	1st Qu.: NA	1st Qu.: NA	1st Qu.: NA
Median : NA	Median : NA	Median : NA	Median : NA
Mean :NaN	Mean :NaN	Mean :NaN	Mean :NaN
3rd Qu.: NA	3rd Qu.: NA	3rd Qu.: NA	3rd Qu.: NA

```
Max.      : NA      Max.      : NA      Max.      : NA      Max.      : NA
```

```
As21 <- reshape(As21, direction = "long", idvar = "hhid",
  varying = grepout("\\.\\d", colnames(As21)))
As22 <- reshape(As22, direction = "long", idvar = "hhid",
  varying = grepout("\\.\\d", colnames(As22)))
As2 <- rbind(As21, As22, fill = T)
As2[, time := NULL]
As2 <- As2[!(is.na(item) | item == ""), ]
As2[, assetNumber := 1:N, by = hhid]
setkey(As2, hhid, assetNumber)
As2[, totalSum := sum(value, na.rm = T), by = hhid]
```

Rd 3.

```
lapply(As03, colnames)
```

```
[[1]]
[1] "hhid" "sec21_item_code"
[3] "sec21_other_specifzz" "current_status"
[5] "decimal" "taka"
[7] "purchased_in_last_one_year"

[[2]]
[1] "hhid" "sec22_code"
[3] "sec22_others_specifz" "current_statut"
[5] "how_many" "sec22_portion_owned"
[7] "sec22_value_in_taka" "sec22_rented_amount_in_tk"
```

```
invisible(lapply(As03, function(x) setnames(x, grepout("code", colnames(x)), "item")))
invisible(lapply(As03, function(x) setnames(x, grepout("spec", colnames(x)), "specify")))
invisible(lapply(As03, function(x) setnames(x, grepout("curr", colnames(x)), "currentStatu")))
invisible(lapply(As03, function(x) setnames(x, grepout("taka", colnames(x)), "value")))
invisible(lapply(As03, function(x) setnames(x, grepout("deci|many", colnames(x)), "amount")))
setnames(As03[[1]], "purchased_in_last_one_year", "lastYear")
setnames(As03[[2]], c("sec22_portion_owned", "sec22_rented_amount_in_tk"),
  c("ownership", "rental"))
As3 <- rbindlist(As[5:6], fill = T)
As3 <- As3[!(is.na(item) | item == ""), ]
As3[, assetNumber := 1:N, by = hhid]
setkey(As3, hhid, assetNumber)
As3[, totalSum := sum(value, na.rm = T), by = hhid]
```

Bind all 3 rds together.

```
Aslist <- list(cbind(rd = 1, As1), cbind(rd = 2, As2), cbind(rd = 3, As3))
(As <- rbindlist(Aslist, fill = T))
```

	rd	hhid	item	own	value	ownership	rental
1:	1	7010102	Tube well for drinking	Yes	1500	NA	NA
2:	1	7010102	Hand pump	Yes	1500	100	NA
3:	1	7010102	Sickle/Dao/Axe/Spade	Yes	300	100	NA
4:	1	7010103	Tube well for drinking	Yes	700	NA	NA
5:	1	7010103	Hand pump	Yes	700	100	NA

22840:	3	99081912420	tube well for drinking	NA	1600	NA	NA
22841:	3	99081912420	mobile phone	NA	1400	NA	NA
22842:	3	99081912420	others	NA	400	NA	NA
22843:	3	99081912420	fishing net	NA	250	100	NA

22844:	3	99081912420	sickle/dao/axe/spade	NA	400	100	NA
	assetNumber	totalSum	specify	currentStatus	amount	lastYear	
1:	1	3300	NA	NA	NA	NA	
2:	2	3300	NA	NA	NA	NA	
3:	3	3300	NA	NA	NA	NA	
4:	1	1900	NA	NA	NA	NA	
5:	2	1900	NA	NA	NA	NA	

22840:	2	5950	bought in last year		1	yes	
22841:	3	5950	bought in last year		1	yes	
22842:	4	5950	bought in last year		1	yes	
22843:	5	5950	From previous year		2	NA	
22844:	6	5950	From previous year		2	NA	

```
setwd(pathsave)
```

```
write.tablev(As, "asset_holding-rd-1-3.prn")
```

```
Asset <- copy(As[!(rd > 1 & grepl("n", lastYear)), ])
```

```
Asset[, assetNumber := 1:N, by = c("hhid", "rd")]
```

```
Asset[, numberOfAssets := .N, by = c("hhid", "rd")]
```

```
Asset[, totalSum := sum(value, na.rm = T), by = c("hhid", "rd")]
```

```
as0 <- unique(Asset[, .(hhid, rd, totalSum)])
```

```
setkey(as0, hhid, rd)
```

```
as0[, rd := rd + 1]
```

```
as0 <- as0[rd < 4, ]
```

```
as1 <- copy(as0)
```

```
as1[, rd := rd + 1]
```

```
as1 <- as1[rd < 4, ]
```

```
setnames(as0, "totalSum", "prevSum.1")
```

```
setnames(as1, "totalSum", "prevSum.2")
```

```
as0[, prevassetNPV.1 := prevSum.1 * .95]
```

```
as1[, prevassetNPV.2 := prevSum.2 * .95^(2)]
```

```
setkey(as0, hhid, rd); setkey(as1, hhid, rd)
```

```
as01 <- as1[as0]
```

```
as01[is.na(prevassetNPV.2), prevassetNPV.2 := 0]
```

```
as01[, prevassetNPV := prevassetNPV.1 + prevassetNPV.2]
```

```
setkey(as01, hhid, rd); setkey(Asset, hhid, rd)
```

```
Asset01 <- merge(Asset, as01, by = c("hhid", "rd"), all = T)
```

```
Asset01[is.na(prevassetNPV), prevassetNPV := 0]
```

```
Asset01[, assetNPV := totalSum + prevassetNPV]
```

```
# merge with treatment info
```

```
setkey(Asset01, hhid, rd); setkey(tr01, hhid, rd)
```

```
Asset01t <- tr01[Asset01]
```

Drop rd 2 and 3 assets that were not bought in the lastYear to avoid double counting.

```
asset <- Asset01t[assetNumber == 1, ]
```

```
asset.ss <- subset(asset, assetNPV > 0 & !is.na(receivedCredit))
```

```
asset.cross <- tapply(asset.ss$assetNPV,
```

```
list(rd = asset.ss$rd, receivedCredit=asset.ss$receivedCredit), median)
```

```
asset.cross2 <- tapply(asset.ss$assetNPV,
```

```
list(rd = asset.ss$rd, receivedCredit=asset.ss$receivedCredit), mean)
```

```
vline.dat <- data.frame(rd = rep(1:3, 2), receivedCredit = repseq(c(F, T), 3))
```

```
vline.dat <- cbind(vline.dat, median = c(asset.cross), mean = c(asset.cross2))
```

```
library(ggplot2)
```

```
ggplot(data = subset(asset, assetNPV > 0 & !is.na(receivedCredit)),
```

```
aes(x = assetNPV, fill = arm)) +
```

```
geom_histogram(bins = 20) +
```

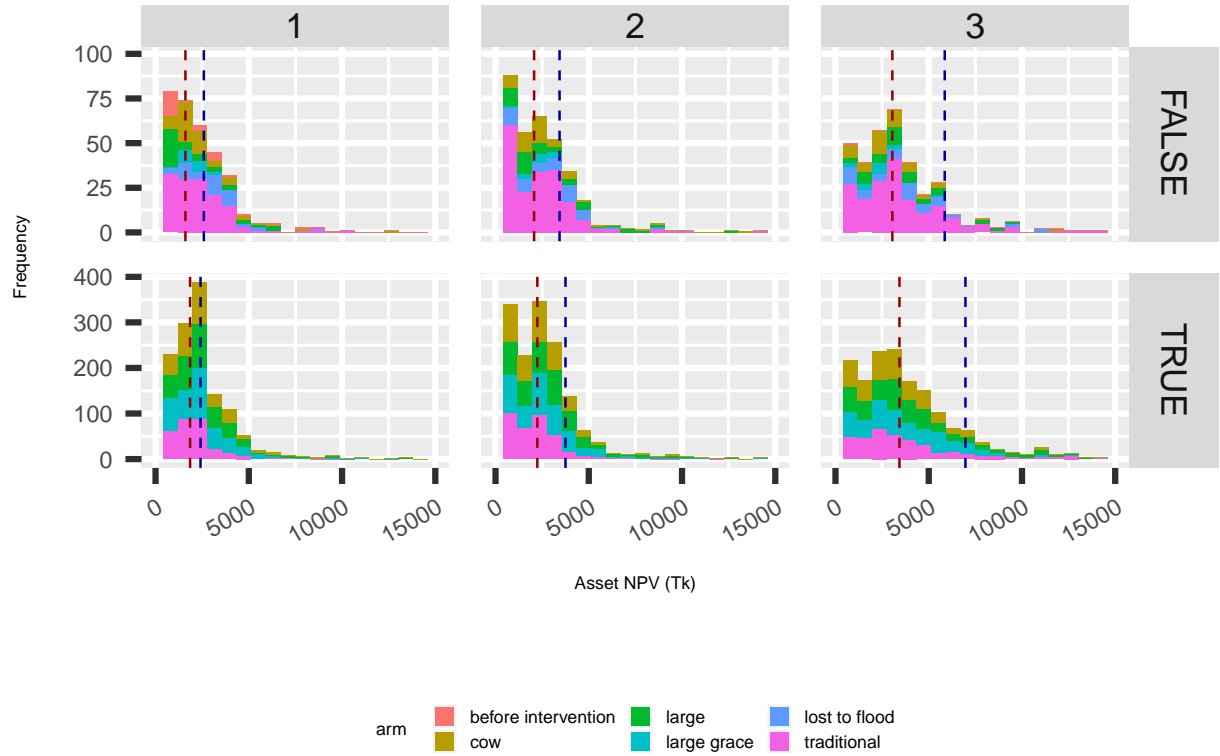



Figure 5 Assets by eventual treatment status

```
scale_x_continuous(limits = c(0, 15000)) +
#scale_y_continuous(limits = c(0, 1000)) +
geom_vline(aes(xintercept = median),
  colour="#990000", linetype="dashed", size = .2, data=vline.dat) +
geom_vline(aes(xintercept = mean),
  colour="#000099", linetype="dashed", size = .2, data=vline.dat) +
ylab("Frequency") + xlab("Asset NPV (Tk)") + labs(fill = "arm") +
facet_grid(receivedCredit ~ rd, scales = "free_y") +
theme(axis.title.y = element_text(size = rel(.25), angle = 90),
  axis.title.x = element_text(size = rel(.25), angle = 0),
  axis.text.x = element_text(size = rel(.5), angle = 30, hjust = 1),
  axis.text.y = element_text(size = rel(.5), angle = 0),
  legend.text = element_text(size=rel(.25)),
  legend.position = "bottom",
  legend.title = element_text(size = rel(.25)),
  legend.key = element_rect(size = rel(.25)),
  legend.key.size = unit(.15, "cm"),
  strip.text = element_text(size=rel(.5)),
  strip.text.x = element_text(margin = margin(.05, 0, .05, 0, "cm"))))
```

The histogram is created by imputing the NPV of household assets by assuming an annual 5% depreciation rate. We see that, at rd 1, there is no difference in mean of asset holding, while the medians are different. Interestingly, the median difference is preserved in the later rounds. In the meantime, means are not different in rd 1 yet they come to differ in later rounds. The subjects who actually received credits have higher mean asset holding. Given that the median differences are unchanged, this indicates that the upper half of the treated asset holders are getting better than the control.

To align dates of receiving credits for the subjects who did not, we use the median daysFrom-Start.

```
setkey(Asset01t, gid, hhid)
# surround with as.double becuse median function returns various types ....
# see SO: https://stackoverflow.com/questions/12125364/why-does-median-trip-up-data-table
Asset01t[, medianElapsedDaysOfGroup :=
  as.double(median(elapsed, na.rm = T)), by = gid]
Asset01t[, meanElapsedDaysOfGroup := mean(elapsed, na.rm = T), by = gid]
Asset01t[, elapsedGroupMedian := "early"]
Asset01t[medianElapsedDaysOfGroup -
  median(medianElapsedDaysOfGroup, na.rm = T) ≤ 0,
  elapsedGroupMedian := "late"]
Asset01t[, elapsedGroupMean := "early"]
Asset01t[meanElapsedDaysOfGroup -
  mean(meanElapsedDaysOfGroup, na.rm = T) ≤ 0,
  elapsedGroupMean := "late"]
Asset01t[, elapsedGroupMedian := factor(elapsedGroupMedian)]
Asset01t[, elapsedGroupMean := factor(elapsedGroupMean)]

asset ← Asset01t[assetNumber == 1, ]
asset.ss ← subset(asset, assetNPV > 0 & !is.na(gid) & !is.na(disbursed))
asset.cross ← tapply(asset.ss$assetNPV,
  list(rd = asset.ss$rd, disbursed=asset.ss$disbursed), median)
asset.cross2 ← tapply(asset.ss$assetNPV,
  list(rd = asset.ss$rd, disbursed=asset.ss$disbursed), mean)
vline.dat ← data.frame(rd = rep(1:3, 2), disbursed = repseq(c(F, T), 3))
vline.dat ← cbind(vline.dat, median = c(asset.cross), mean = c(asset.cross2))
library(ggplot2)
ggplot(data = asset.ss,
  aes(x = assetNPV, fill = arm)) +
  geom_histogram(bins = 20) +
  scale_x_continuous(limits = c(0, 15000)) +
  geom_vline(aes(xintercept = median),
    colour="#990000", linetype="dashed", size = .2, data=vline.dat) +
  geom_vline(aes(xintercept = mean),
    colour="#000099", linetype="dashed", size = .2, data=vline.dat) +
  ylab("Frequency") + xlab("Asset NPV (Tk)") + labs(fill = "arm") +
  facet_grid(disbursed ~ rd, scales = "free_y") +
  theme(axis.title.y = element_text(size = rel(.25), angle = 90),
    axis.title.x = element_text(size = rel(.25), angle = 0),
    axis.text.x = element_text(size = rel(.5), angle = 30, hjust = 1),
    axis.text.y = element_text(size = rel(.5), angle = 0),
    legend.text = element_text(size=rel(.25)),
    legend.position = "bottom",
    legend.title = element_text(size = rel(.25)),
    legend.key = element_rect(size = rel(.25)),
    legend.key.size = unit(.15, "cm"),
    strip.text = element_text(size=rel(.5)),
    strip.text.x = element_text(margin = margin(.05, 0, .05, 0, "cm")),
    strip.text.y = element_text(margin = margin(.05, 0, .05, 0, "cm")))
```

In this figure, we dropped observations without gid and disbursed. When intDate is NA (not interviewed), we cannot define disbursement for that round. We know disbursement took place before rd 3, so all assignment = treated have disbursed = T in rd 3.

```
asset ← Asset01t[assetNumber == 1, ]
```

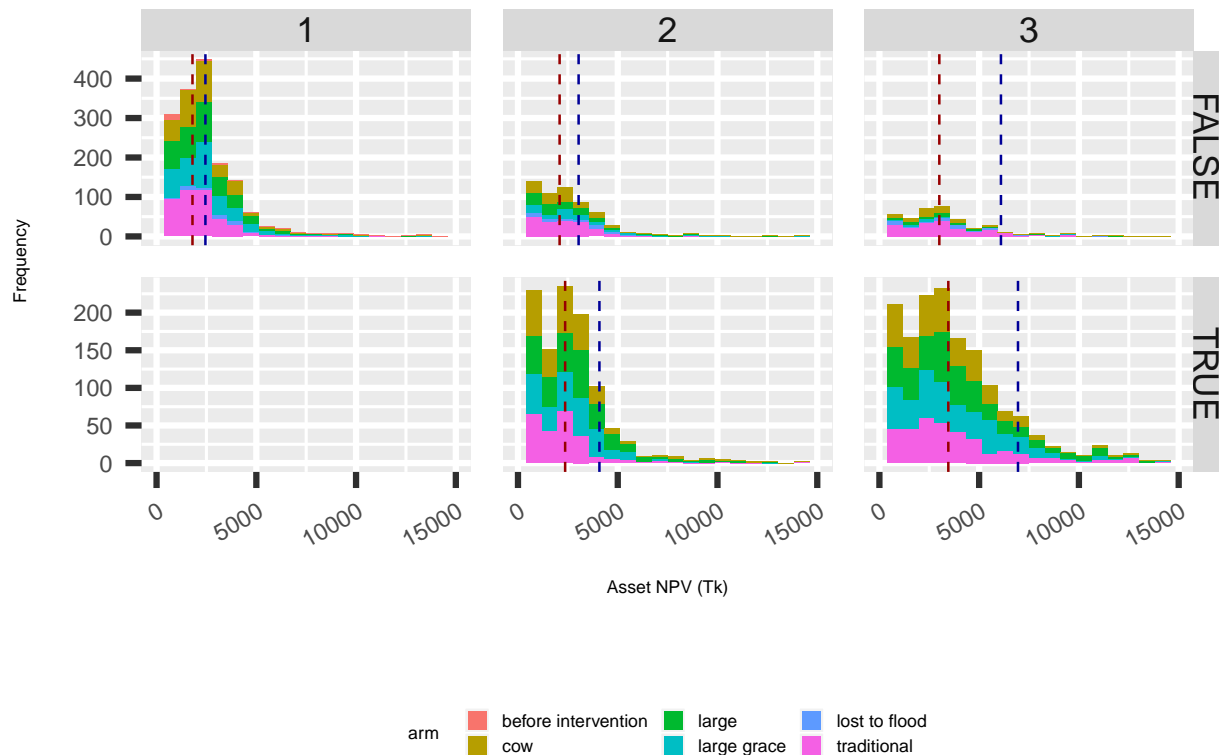


Figure 6 Assets by disbursement status

```
asset.ss <- subset(asset, assetNPV > 0 & !is.na(gid) & !is.na(elapsed) & receivedCredit)
asset.cross <- tapply(asset.ss$assetNPV,
  list(rd = asset.ss$rd, elapsedGroupMedian = asset.ss$elapsedGroupMedian), median)
asset.cross2 <- tapply(asset.ss$assetNPV,
  list(rd = asset.ss$rd, elapsedGroupMedian = asset.ss$elapsedGroupMedian), mean)
vline.dat <- data.frame(rd = rep(1:3, 2), elapsedGroupMedian = repseq(c("early", "late")))
vline.dat <- cbind(vline.dat, median = c(asset.cross), mean = c(asset.cross2))
library(ggplot2)
ggplot(data = asset.ss,
  aes(x = assetNPV, fill = arm)) +
  geom_histogram(bins = 20) +
  scale_x_continuous(limits = c(0, 15000)) +
  geom_vline(aes(xintercept = median),
    colour="#990000", linetype="dashed", size = .2, data=vline.dat) +
  geom_vline(aes(xintercept = mean),
    colour="#000099", linetype="dashed", size = .2, data=vline.dat) +
  ylab("Frequency") + xlab("Asset NPV (Tk)") + labs(fill = "arm") +
  facet_grid(elapsedGroupMedian ~ rd, scales = "free_y") +
  theme(axis.title.y = element_text(size = rel(.25), angle = 90),
    axis.title.x = element_text(size = rel(.25), angle = 0),
    axis.text.x = element_text(size = rel(.5), angle = 30, hjust = 1),
    axis.text.y = element_text(size = rel(.5), angle = 0),
    legend.text = element_text(size=rel(.25)),
    legend.position = "bottom",
    legend.title = element_text(size = rel(.25)),
    legend.key = element_rect(size = rel(.25)),
    legend.key.size = unit(.15, "cm"),
```

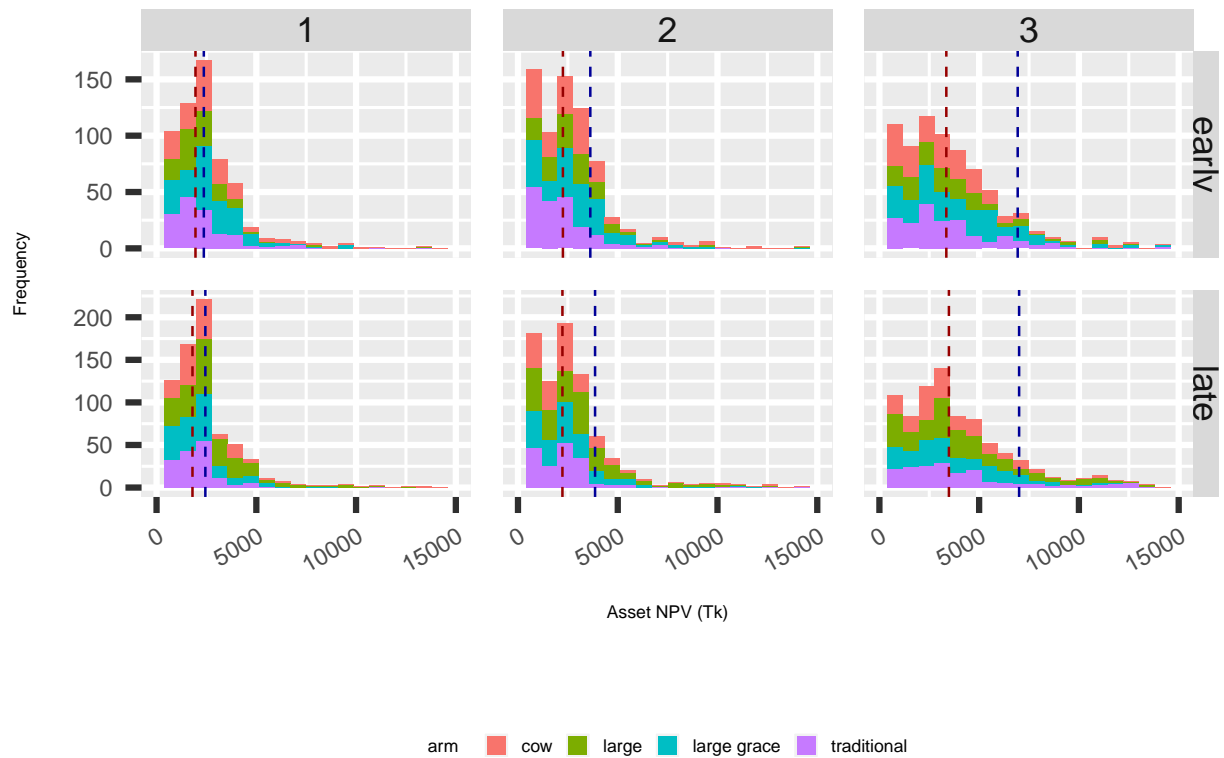


Figure 7 Assets by elapsed days from disbursement, ATT

```

strip.text = element_text(size=rel(.5)),
strip.text.x = element_text(margin = margin(.05, 0, .05, 0, "cm")),
strip.text.y = element_text(margin = margin(.05, 0, .05, 0, "cm"))

asset <- Asset01t[assetNumber == 1, ]
asset.ss <- subset(asset, assetNPV > 0 & !is.na(gid) & !is.na(elapsed) & receivedCredit)
library(ggplot2)
ggplot(data = asset.ss, aes(x = elapsed, y = assetNPV)) +
  #geom_jitter(aes(colour = arm, shape = arm), size = .05, width = .1) +
  geom_point(aes(colour = arm, shape = arm), size = .05) +
  scale_shape(solid = F) +
  #scale_y_continuous(limits = c(0, 25000)) +
  scale_y_log10() +
  xlab("elapsed day grouping") + ylab("Asset NPV (Tk)") + labs(fill = "arm") +
  facet_grid(arm ~ rd) +
# stat_smooth(method = "loess", size = .2, n = 150) +
  geom_smooth(method = "loess", size = .2) +
  theme(axis.title.y = element_text(size = rel(.25), angle = 90),
        axis.title.x = element_text(size = rel(.25), angle = 0),
        axis.text.x = element_text(size = rel(.5), angle = 30, hjust = 1),
        axis.text.y = element_text(size = rel(.5), angle = 0),
        legend.text = element_text(size=rel(.25)),
        legend.position = "bottom",
        legend.title = element_text(size = rel(.25)),
        legend.key = element_rect(size = rel(.25)),
        legend.key.size = unit(.15, "cm"),
        strip.text = element_text(size=rel(.5)),
        strip.text.x = element_text(margin = margin(.05, 0, .05, 0, "cm")),

```

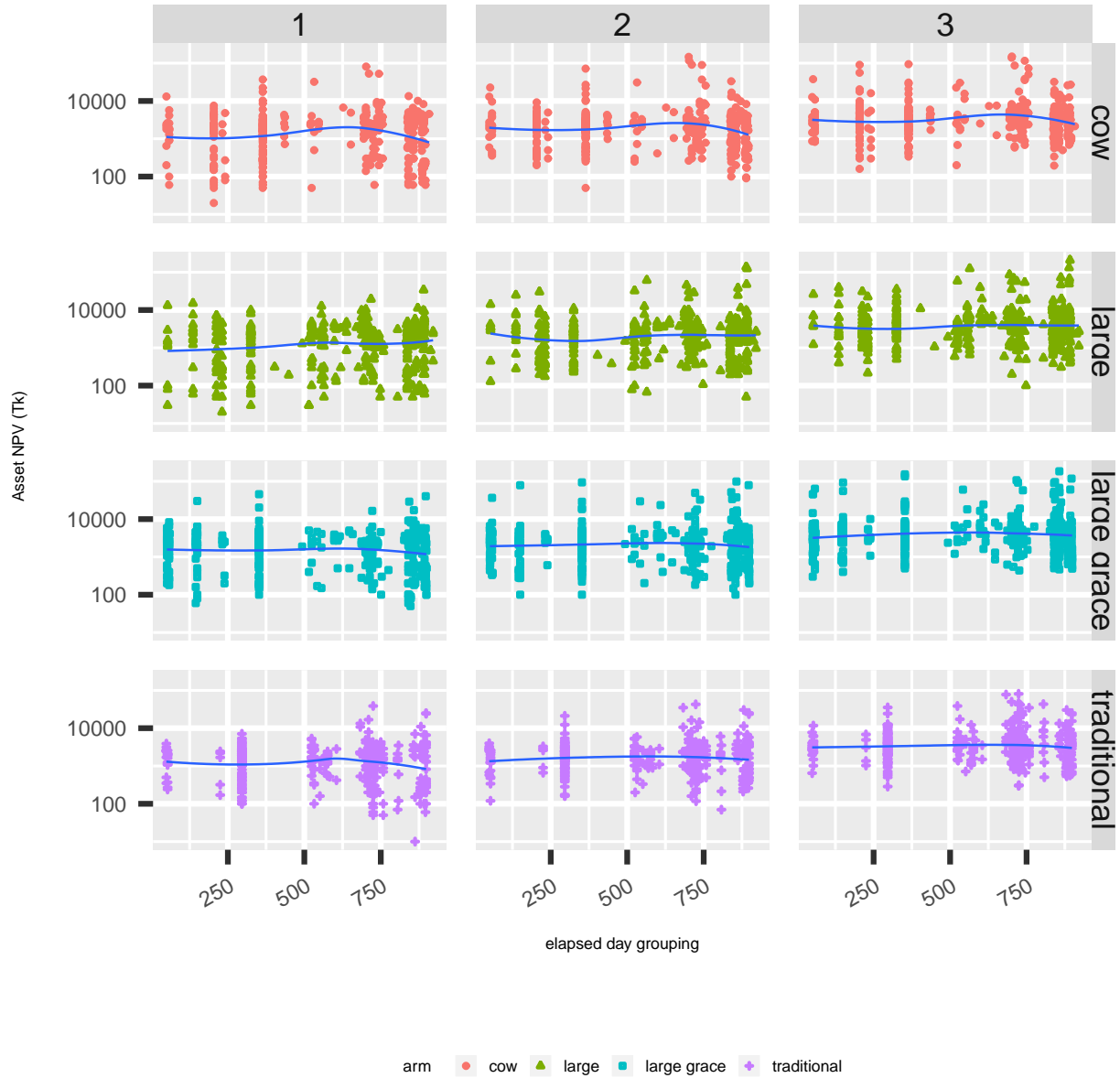


Figure 8 Assets by elapsed days from disbursement, ATT scatter plot

```
strip.text.y = element_text(margin = margin(.05, 0, .05, 0, "cm")))
```

In this scatter and loess plots, we put asset values against the treatment exposure, faceted by treatment arms. This aims to mimic ATT under a continuous treatment. The treatment exposure is defined by the elapsed days since receiving a credit. Since the treatment exposure is randomised, this is a statistically valid procedure to observe the treatment response without major confounding.

This plotting exercise leads one to consider the statistical model underlying the graphs. For an individual i 's outcome y_i , the treatment assignment $D_i = 0, 1$ may have an impact on the outcome. The standard Rubin causal model deals with a binary indicator variable for D_i . In our design, we vary the dates of intervention among the subjects. So what we randomly vary is the duration under treatment, or dose exposure, denoted with $D_i(t)$ where t is the calendar date of intervention. On average, there is about 1 year difference in t within a cluster of 20 subjects. Given that we randomise the calendar dates of starting the intervention, we can assume actual duration $t \in [t_0, t_1]$ is orthogonal to potential treatment response $y(t)$ for all t . Under the simplest setting, we follow [Imbens \(2000\)](#); [Hirano and Imbens \(2005\)](#); [Imai and van Dyk \(2004\)](#); [Egger and von Ehrlich \(2013\)](#) assume the

following conditional orthogonality in the continuous case. Denoting T as a random variable with its realisation written as t , we assume:

$$y(t) \perp T | \mathbf{x}.$$

Hirano and Imbens (2005) shows that this is equivalent to

$$\mathbf{x} \perp 1\{T = t\} | g(t, \mathbf{x})$$

where $g(t, \mathbf{x})$ is a generalised propensity score that gives the density of treatment at t given \mathbf{x} . This shows that one can estimate continuous treatment effect by first, estimating GPS g , second, estimate the conditional expectation of outcome as a function of g and \mathbf{x} :

$$\beta(t, g) = \mathcal{E}[y | T = t, G = g(t, \mathbf{x})],$$

and then average over g for a given t to obtain the dose-response function

$$\beta(t) = \mathcal{E}[\beta(t, g) | \mathbf{x}].$$

The approach is preceded by applied works related job training duration (Kluve et al., 2012).

We compare the effects of treatment exposure differences within the same group.

```
asset <- Asset01t[assetNumber == 1, ]
asset.ss <- subset(asset, assetNPV > 0 & !is.na(gid) & !is.na(elapsed))
setkey(asset.ss, rd, gid, assignment)
asset.ss[, avgElapsed := mean(elapsed, na.rm = T), by = c("rd", "gid", "assignment")]
asset.ss[, avgElapsed0 := avgElapsed[1], by = c("rd", "gid")]
asset.ss[, avgElapsed1 := avgElapsed[.N], by = c("rd", "gid")]
asset.ss[gid == 70650 & grepl("co", assignment), ]
```

	hhid	disburseDate	purchaseDate	memstatus	receivedCredit	exist	
1:	7065010	2014-12-09	2014-12-11	old	TRUE	123	
2:	9807065006	2014-12-09	2014-12-11	replacement	TRUE	123	
3:	9807065013	2014-12-09	2014-12-11	replacement	TRUE	123	
4:	9807065019	2014-12-09	2014-12-11	replacement	TRUE	123	
5:	7065010	2014-12-09	2014-12-11	old	TRUE	123	
6:	9807065006	2014-12-09	2014-12-11	replacement	TRUE	123	
7:	9807065013	2014-12-09	2014-12-11	replacement	TRUE	123	
8:	9807065019	2014-12-09	2014-12-11	replacement	TRUE	123	
9:	7065010	2014-12-09	2014-12-11	old	TRUE	123	
10:	9807065006	2014-12-09	2014-12-11	replacement	TRUE	123	
11:	9807065013	2014-12-09	2014-12-11	replacement	TRUE	123	
12:	9807065019	2014-12-09	2014-12-11	replacement	TRUE	123	
	inData	gid	i.memstatus	assignment	arm	i.receivedCredit	elapsed
1:	TRUE	70650	old	control	traditional	TRUE	296
2:	TRUE	70650	replacement	control	traditional	TRUE	296
3:	TRUE	70650	replacement	control	traditional	TRUE	296
4:	TRUE	70650	replacement	control	traditional	TRUE	296
5:	TRUE	70650	old	control	traditional	TRUE	296
6:	TRUE	70650	replacement	control	traditional	TRUE	296
7:	TRUE	70650	replacement	control	traditional	TRUE	296
8:	TRUE	70650	replacement	control	traditional	TRUE	296
9:	TRUE	70650	old	control	traditional	TRUE	296
10:	TRUE	70650	replacement	control	traditional	TRUE	296
11:	TRUE	70650	replacement	control	traditional	TRUE	296
12:	TRUE	70650	replacement	control	traditional	TRUE	296
	daysFromStart	rd	disbursed	purchased	intDate	daysSince2014	
1:	625	1	FALSE	FALSE	2012-11-08	-419	
2:	625	1	FALSE	FALSE	<NA>	NA	
3:	625	1	FALSE	FALSE	<NA>	NA	

4:	625	1	FALSE	FALSE	<NA>	NA
5:	625	2	FALSE	FALSE	2004-01-21	-3633
6:	625	2	FALSE	FALSE	2004-01-11	-3643
7:	625	2	FALSE	FALSE	2004-01-11	-3643
8:	625	2	FALSE	FALSE	2004-01-11	-3643
9:	625	3	TRUE	TRUE	2015-12-17	715
10:	625	3	TRUE	TRUE	2015-12-10	708
11:	625	3	TRUE	TRUE	2015-11-28	696
12:	625	3	TRUE	TRUE	2015-12-12	710
		item own	value	ownership	rental	assetNumber totalSum
1:	Tube well for drinking	Yes	700	NA	NA	1 1620
2:		566 1	1200	NA	NA	1 1200
3:		566 1	2000	NA	NA	1 2300
4:		566 1	1600	NA	NA	1 1900
5:	fishing net	NA	50	100	NA	1 350
6:	sickle/dao/axe/spade	NA	200	100	NA	1 200
7:	weeder	NA	100	100	NA	1 400
8:	sickle/dao/axe/spade	NA	150	100	NA	1 150
9:	sickle/dao/axe/spade	NA	200	100	NA	1 200
10:	sickle/dao/axe/spade	NA	300	100	NA	1 300
11:	tube well for drinking	NA	1200	NA	NA	1 1600
12:	sickle/dao/axe/spade	NA	300	100	NA	1 300
	specify	currentStatus	amount	lastYear	numberOfAssets	prevSum.2
1:	NA	NA	NA	NA	4	NA
2:	NA	NA	NA	NA	1	NA
3:	NA	NA	NA	NA	2	NA
4:	NA	NA	NA	NA	2	NA
5:	From previous year		1	NA	2	NA
6:	From previous year		2	NA	1	NA
7:	From previous year		1	NA	2	NA
8:	From previous year		1	NA	1	NA
9:	From previous year		2	NA	1	1620
10:	From previous year		3	NA	1	1200
11:	bought in last year		1	yes	3	2300
12:	From previous year		3	NA	1	1900
	prevassetNPV.2	prevSum.1	prevassetNPV.1	prevassetNPV	assetNPV	
1:	NA	NA	NA	0.00	1620.00	
2:	NA	NA	NA	0.00	1200.00	
3:	NA	NA	NA	0.00	2300.00	
4:	NA	NA	NA	0.00	1900.00	
5:	0.00	1620	1539.0	1539.00	1889.00	
6:	0.00	1200	1140.0	1140.00	1340.00	
7:	0.00	2300	2185.0	2185.00	2585.00	
8:	0.00	1900	1805.0	1805.00	1955.00	
9:	1462.05	350	332.5	1794.55	1994.55	
10:	1083.00	200	190.0	1273.00	1573.00	
11:	2075.75	400	380.0	2455.75	4055.75	
12:	1714.75	150	142.5	1857.25	2157.25	
	medianElapsedDaysOfGroup	meanElapsedDaysOfGroup	elapsedGroupMedian			
1:		737	693.795	early		
2:		737	693.795	early		
3:		737	693.795	early		
4:		737	693.795	early		
5:		737	693.795	early		
6:		737	693.795	early		
7:		737	693.795	early		
8:		737	693.795	early		
9:		737	693.795	early		
10:		737	693.795	early		
11:		737	693.795	early		
12:		737	693.795	early		

	elapsedGroupMean	avgElapsed	avgElapsed0	avgElapsed1
1:	early	296	296	828.9
2:	early	296	296	828.9
3:	early	296	296	828.9
4:	early	296	296	828.9
5:	early	296	296	828.9
6:	early	296	296	828.9
7:	early	296	296	828.9
8:	early	296	296	828.9
9:	early	296	296	828.9
10:	early	296	296	828.9
11:	early	296	296	828.9
12:	early	296	296	828.9

```
asset.ss[gid == 70204 & rd == 1, .(rd, gid, assignment, avgElapsed, avgElapsed0, avgElapsed1
```

	rd	gid	assignment	avgElapsed	avgElapsed0	avgElapsed1
1:	1	70204	control	517.222	517.222	837.1
2:	1	70204	control	517.222	517.222	837.1
3:	1	70204	control	517.222	517.222	837.1
4:	1	70204	control	517.222	517.222	837.1
5:	1	70204	control	517.222	517.222	837.1
6:	1	70204	control	517.222	517.222	837.1
7:	1	70204	control	517.222	517.222	837.1
8:	1	70204	control	517.222	517.222	837.1
9:	1	70204	control	517.222	517.222	837.1
10:	1	70204	treated	837.100	517.222	837.1
11:	1	70204	treated	837.100	517.222	837.1
12:	1	70204	treated	837.100	517.222	837.1
13:	1	70204	treated	837.100	517.222	837.1
14:	1	70204	treated	837.100	517.222	837.1
15:	1	70204	treated	837.100	517.222	837.1
16:	1	70204	treated	837.100	517.222	837.1
17:	1	70204	treated	837.100	517.222	837.1
18:	1	70204	treated	837.100	517.222	837.1
19:	1	70204	treated	837.100	517.222	837.1

```
asset.ss[, avgNPV := mean(assetNPV/1000, na.rm = T),
  by = c("rd", "gid", "assignment")]
asset.ss[, avgNPV0 := avgNPV[1], by = c("rd", "gid")]
asset.ss[, avgNPV1 := avgNPV[N], by = c("rd", "gid")]
asset.ss[, avgDiffElapsed := avgElapsed1 - avgElapsed0]
asset.ss[, avgDiffNPV := avgNPV1 - avgNPV0]
setkey(asset.ss, rd, gid, assignment)
dim(asset.sss <- asset.ss[!duplicated(asset.ss[, .(rd, gid, assignment)]), ])
```

```
[1] 543 49
```

```
library(ggplot2)
ggplot(data = asset.sss, aes(x = avgDiffElapsed, y = avgDiffNPV)) +
  geom_point(aes(colour = arm, shape = arm), size = .05) +
  scale_shape(solid = F) +
  xlab("difference in elapsed days") + ylab("difference in mean asset NPV (Tk '000)") +
  labs(fill = "arm") + facet_grid(~ rd) +
# stat_smooth(method = "loess", size = .2, n = 150) +
  geom_smooth(method = "loess", size = .2) +
  geom_hline(aes(yintercept = 0), colour = "#990000", linetype = "dashed", size = .2) +
  theme(axis.title.y = element.text(size = rel(.25), angle = 90),
```

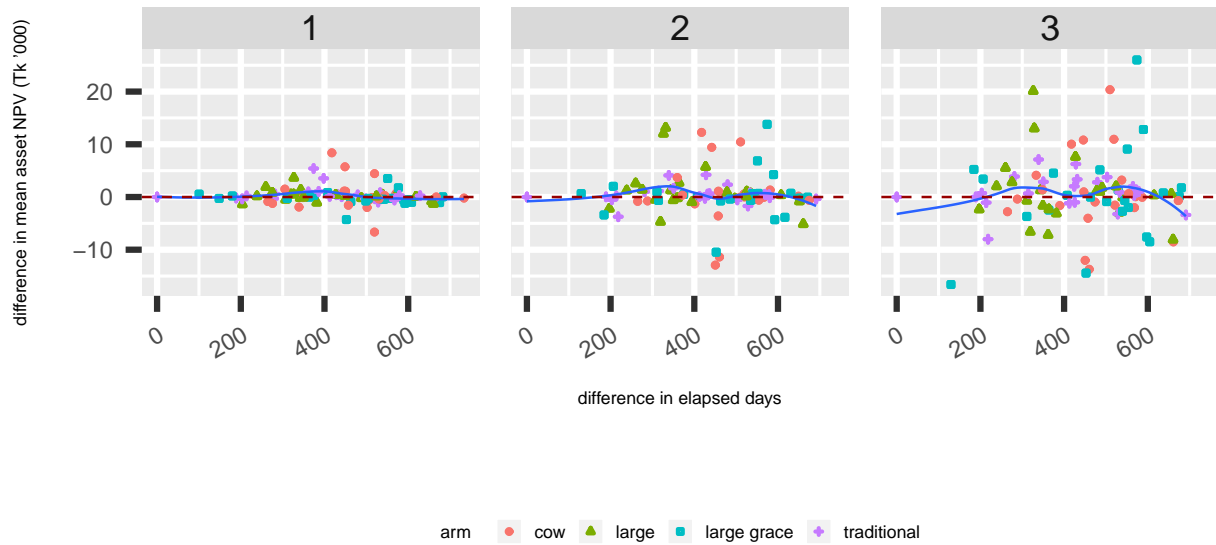



Figure 9 Assets by elapsed days from disbursement within a group

```
axis.title.x = element_text(size = rel(.25), angle = 0),
axis.text.x = element_text(size = rel(.5), angle = 30, hjust = 1),
axis.text.y = element_text(size = rel(.5), angle = 0),
legend.text = element_text(size=rel(.25)),
legend.position = "bottom",
legend.title = element_text(size = rel(.25)),
legend.key = element_rect(size = rel(.25)),
legend.key.size = unit(.15, "cm"),
strip.text = element_text(size=rel(.5)),
strip.text.x = element_text(margin = margin(.05, 0, .05, 0, "cm")),
strip.text.y = element_text(margin = margin(.05, 0, .05, 0, "cm"))

library(ggplot2)
ggplot(data = asset.sss, aes(x = avgDiffElapsed, y = avgDiffNPV)) +
  geom_point(aes(colour = arm, shape = arm), size = .05) +
  scale_shape(solid = F) +
  xlab("difference in elapsed days") + ylab("difference in mean asset NPV (Tk '000)") +
  labs(fill = "arm") + facet_grid(arm ~ rd, scale = "free_y") +
  # stat_smooth(method = "loess", size = .2, n = 150) +
  geom_smooth(method = "loess", size = .2) +
  geom_hline(aes(yintercept = 0), colour="#990000", linetype="dashed", size = .2) +
  theme(axis.title.y = element_text(size = rel(.25), angle = 90),
        axis.title.x = element_text(size = rel(.25), angle = 0),
        axis.text.x = element_text(size = rel(.5), angle = 30, hjust = 1),
        axis.text.y = element_text(size = rel(.5), angle = 0),
        legend.text = element_text(size=rel(.25)),
        legend.position = "bottom",
        legend.title = element_text(size = rel(.25)),
        legend.key = element_rect(size = rel(.25)),
        legend.key.size = unit(.15, "cm"),
        strip.text = element_text(size=rel(.5)),
        strip.text.x = element_text(margin = margin(.05, 0, .05, 0, "cm")),
        strip.text.y = element_text(margin = margin(.05, 0, .05, 0, "cm")))
```

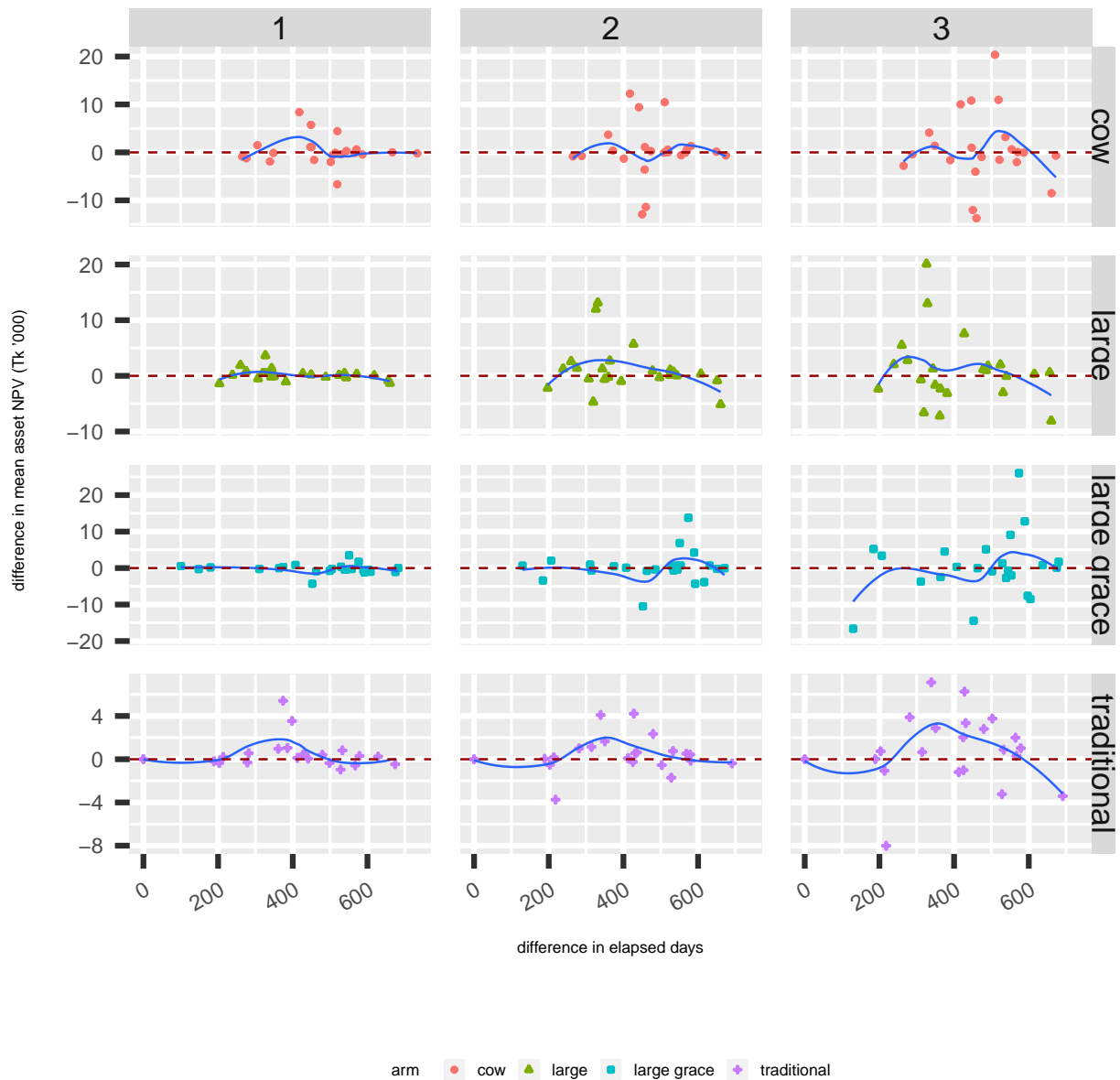


Figure 10 Assets by elapsed days from disbursement, within a group

VII livestock

```
(fn.lvstk ← grepout("d/s08|2/s.*9_|3/s.*_8", fn))
```

```
[1] "../1/combined/s08a.prn" "../1/combined/s08b.prn" "../2/section_9_1.prn"
[4] "../2/section_9_2.prn"   "../2/section_9_3.prn"   "../3/section_8.prn"
[7] "../3/section_8a.prn"   "../3/section_8b.prn"
```

```
setwd(pathsource.mar)
Ls = copy(X[fn %in% fn.lvstk])
Ls ← lapply(Ls, function(x) if (any(grepl("^id$", colnames(x))))
  setnames(x, "id", "hhid") else x)
Ls ← lapply(Ls, function(x)
```

```

x[!apply(is.na(x[, -grep("hh|mid|u_id", colnames(x)), with = F]) |
x[, -grep("hh|mid|u_id", colnames(x)), with = F] == "" |
x[, -grep("hh|mid|u_id", colnames(x)), with = F] == "No", 1, all), )
Ls1 <- lapply(Ls, a2b, a = NA, b = 0)
Ls2 <- lapply(Ls, a2b, a = "", b = 0)
Ls1[1:2] <- lapply(Ls1[1:2], setkey, hhid, mid)
Ls2[-(1:2)] <- lapply(Ls2[-(1:2)], setkey, hhid)
Ls1 <- merge(Ls1[[1]], Ls1[[2]], by = c("hhid", "mid"), all = T)
Ls2 <- merge(Ls2[[3]], Ls2[[4]], by = "hhid", all = T)
Ls2 <- merge(Ls2, Ls2[[5]], by = "hhid", all = T)
Ls2 <- Ls2[!duplicated(Ls2[, .(hhid, s17a_code)]), ]
Ls3 <- merge(Ls1[[6]], Ls1[[7]], by = "hhid", all = T)
Ls3 <- merge(Ls3, Ls1[[8]], by = "hhid", all = T)
Ls3 <- Ls3[!duplicated(Ls3[, .(hhid, s17a_code)]), ]

```

Rd 1.

```

# M: managing, L: leased in
Ls1[, ushiM := s8a-a-2 + s8a-a-3]
Ls1[, calfM := s8a-a-4]
Ls1[, yagiM := s8a-a-5 + s8a-a-6]
Ls1[, ushiL := s8a-b-8 + s8a-b-9]
Ls1[, calfL := s8a-b-10]
Ls1[, yagiL := s8a-b-11]
Ls1 <- a2b(Ls1, NA, 0)
destat(Ls1[, .(ushiM, calfM, yagiM, ushiL, calfL)])

```

	min	25\\%	median	75\\%	max	mean	std	0s	NAs	n
ushiM	0	0	0	1	6	0.4	0.7	1200	0	1780
calfM	0	0	0	0	4	0.3	0.6	1353	0	1780
yagiM	0	0	0	0	8	0.4	1.0	1395	0	1780
ushiL	0	0	0	0	3	0.1	0.4	1630	0	1780
calfL	0	0	0	0	2	0.1	0.3	1680	0	1780

```

cpr <- destat(Ls1[s8a-b-15 > 0, s8a-b-15])
cpr2 <- rbind(c(destat(Ls1[s8a-b-15 > 0, s8a-b-15])),
c(destat(Ls1[s8a-b-16 > 0, s8a-b-16])),
c(destat(Ls1[s8a-b-17 > 0, s8a-b-17])))
dimnames(cpr2) <- list(c("female calf", "male calf", "ox"),
colnames(cpr))

```

Price: female calf, male calf, ox.

```
cpr2
```

	min	25\\%	median	75\\%	max	mean	std	0s	NAs	n
female calf	5000	8000	10000	13500	31000	11244.4	4458.9	0	0	45
male calf	2000	10000	10000	12000	16000	10300.0	3221.0	0	0	25
ox	500	8000	10000	12000	30000	10549.3	5163.6	0	0	73

Let the price to be used as median price, and cow price is 15000. Lease share is 50%.

```
destat(Ls1[s8a-b-18 > 0, s8a-b-18])
```

	min	25\\%	median	75\\%	max	mean	std	0s	NAs	n
v.1.	15	50	50	50	60	49.6	4.3	0	0	71

```
destat(Ls1[s8a-b-24 > 0, s8a-b-24])
```

	min	25\\%	median	75\\%	max	mean	std	0s	NAs	n
v.1.	9000	9000	9000	9000	9000	9000	0	0	0	2

```
destat(Ls2[grepl("a", s17a_code) & s17a_4 > 0, s17a_4])
```

	min	25\\%	median	75\\%	max	mean	std	0s	NAs	n
v.1.	2	15000	17000	20000	60000	17169.1	4582.7	0	0	1266

```
destat(Ls3[grepl("cow", s17a_code) & s17a_4 > 0, s17a_4])
```

	min	25\\%	median	75\\%	max	mean	std	0s	NAs	n
v.1.	1	18000	20000	23000	50000	20240.5	5049.1	0	0	1705

```
Ls1[, cowValue := ushiM * 15000]
```

```
Ls1[, calfValue := calfM * 10000]
```

```
Ls1[, cowLValue := ushiL * 15000 * .5]
```

```
Ls1[, calfLValue := calfL * 10000 * .5]
```

Goats: Take prices from late rounds. 1900.

```
destat(Ls2[grepl("c", s17a_code) & s17a_4 > 0, s17a_4])
```

	min	25\\%	median	75\\%	max	mean	std	0s	NAs	n
v.1.	2	1500	1900	2800	18000	2161.8	1445.1	0	0	600

```
destat(Ls3[grepl("goa", s17a_code) & s17a_4 > 0, s17a_4])
```

	min	25\\%	median	75\\%	max	mean	std	0s	NAs	n
v.1.	2	1800	2400	3000	36000	2640.2	2530	0	0	810

```
Ls1[, yagiValue := yagiM * 1900]
```

```
Ls1[, yagiLValue := yagiL * 1900 * .5]
```

Total livestock value.

```
Ls1[, totalLivestockValue := cowValue + calfValue + yagiValue +  
cowLValue + calfLValue + yagiLValue]
```

```
setkey(Ls1, hhid)
```

Rd2.

```
Ls2 ← a2b(Ls2, NA, 0)
```

```
Ls2[, livestockValue := s17a_3 * s17a_4]
```

```
Ls2[grepl("sh", s17a_2), livestockValue := livestockValue * .5]
```

```
Ls2[grepl("0", s17a_2), livestockValue := 0]
```

```
Ls2[, livestockValue := sum(livestockValue, na.rm = T), by = hhid]
```

```
Ls2[, livestockSoldValue := s17a_9]
```

```
Ls2[, livestockSoldValue := sum(livestockSoldValue, na.rm = T),  
by = hhid]
```

```
Ls2[, livestockDCValue := (s17a_6 + s17a_7) * s17a_4]
```

```
Ls2[, livestockDCValue := sum(livestockDCValue, na.rm = T),  
by = hhid]
```

```
Ls2[, totalLivestockValue := sum(livestockValue + livestockSoldValue + livestockDCValue),
```

```
Ls2 ← Ls2[!duplicated(Ls2[, hhid]), ]
```

```
#Ls2[grepl("a", s17a_code) & s17a_8 > 0, .(s17a_8, s17a_9)]
```

Rd3.

```

Ls3 ← a2b(Ls3, NA, 0)
Ls3[, livestockValue := s17a_3 * s17a_4]
Ls3[grepl("sh", s17a_2), livestockValue := livestockValue * .5]
Ls3[grepl("0", s17a_2), livestockValue := 0]
Ls3[, livestockValue := sum(livestockValue, na.rm = T), by = hhid]
Ls3[, livestockSoldValue := s17a_9]
Ls3[, livestockSoldValue := sum(livestockSoldValue, na.rm = T),
  by = hhid]
Ls3[, livestockDCValue := (s17a_6 + s17a_7) * s17a_4]
Ls3[, livestockDCValue := sum(livestockDCValue, na.rm = T),
  by = hhid]
Ls3[, totalLivestockValue := sum(livestockValue + livestockSoldValue + livestockDCValue),
Ls3 ← Ls3[!duplicated(Ls3[, hhid]), ]

```

Merge.

```

ls ← rbind(cbind(rd = 1, Ls1[, .(hhid, totalLivestockValue)]),
  cbind(rd = 2, Ls2[, .(hhid, totalLivestockValue)]),
  cbind(rd = 3, Ls3[, .(hhid, totalLivestockValue)]))
ls ← ls[!duplicated(ls), ]
ls[, totalLivestockValue := totalLivestockValue/1000]
setkey(ls, hhid, rd); setkey(tr1l, hhid, rd)
lst ← tr0l[ls]

lstk.ss ← subset(lst, !is.na(gid) & !is.na(elapsed))
setkey(lstk.ss, rd, gid, assignment)
lstk.ss[, avgElapsed := mean(elapsed, na.rm = T), by = c("rd", "gid", "assignment")]
lstk.ss[, avgElapsed0 := avgElapsed[1], by = c("rd", "gid")]
lstk.ss[, avgElapsed1 := avgElapsed[.N], by = c("rd", "gid")]
lstk.ss[, avgLstkValue := mean(totalLivestockValue, na.rm = T),
  by = c("rd", "gid", "assignment")]
lstk.ss[, avgLstkValue0 := avgLstkValue[1], by = c("rd", "gid")]
lstk.ss[, avgLstkValue1 := avgLstkValue[.N], by = c("rd", "gid")]
lstk.ss[gid == 70204 & rd == 1, .(rd, gid, assignment, avgElapsed,
  avgElapsed0, avgElapsed1, avgLstkValue, avgLstkValue0, avgLstkValue1) ]

```

	rd	gid	assignment	avgElapsed	avgElapsed0	avgElapsed1	avgLstkValue
1:	1	70204	control	561.286	561.286	837.1	3.350
2:	1	70204	control	561.286	561.286	837.1	3.350
3:	1	70204	control	561.286	561.286	837.1	3.350
4:	1	70204	control	561.286	561.286	837.1	3.350
5:	1	70204	control	561.286	561.286	837.1	3.350
6:	1	70204	control	561.286	561.286	837.1	3.350
7:	1	70204	control	561.286	561.286	837.1	3.350
8:	1	70204	treated	837.100	561.286	837.1	9.105
9:	1	70204	treated	837.100	561.286	837.1	9.105
10:	1	70204	treated	837.100	561.286	837.1	9.105
11:	1	70204	treated	837.100	561.286	837.1	9.105
12:	1	70204	treated	837.100	561.286	837.1	9.105
13:	1	70204	treated	837.100	561.286	837.1	9.105
14:	1	70204	treated	837.100	561.286	837.1	9.105
15:	1	70204	treated	837.100	561.286	837.1	9.105
16:	1	70204	treated	837.100	561.286	837.1	9.105
17:	1	70204	treated	837.100	561.286	837.1	9.105
	avgLstkValue0		avgLstkValue1				
1:	3.35		9.105				
2:	3.35		9.105				
3:	3.35		9.105				
4:	3.35		9.105				

5:	3.35	9.105
6:	3.35	9.105
7:	3.35	9.105
8:	3.35	9.105
9:	3.35	9.105
10:	3.35	9.105
11:	3.35	9.105
12:	3.35	9.105
13:	3.35	9.105
14:	3.35	9.105
15:	3.35	9.105
16:	3.35	9.105
17:	3.35	9.105

```
lstk.ss[, avgDiffElapsed := avgElapsed1 - avgElapsed0]
lstk.ss[, avgDiffLstkValue := avgLstkValue1 - avgLstkValue0]
setkey(lstk.ss, rd, gid, assignment)
dim(lstk.sss <- lstk.ss[!duplicated(lstk.ss[, .(rd, gid, assignment)]), ])
```

```
[1] 540 28
```

We compare the effects of treatment exposure differences within the same group.

```
library(ggplot2)
ggplot(data = lstk.sss, aes(x = avgDiffElapsed, y = avgDiffLstkValue)) +
  geom_point(aes(colour = arm, shape = arm), size = .05) +
  scale_shape(solid = F) + scale_y_continuous() +
  xlab("difference in elapsed days") + ylab("difference in mean livestock value (Tk)") +
  labs(fill = "arm") + facet_grid(~ rd) +
  # stat_smooth(method = "loess", size = .2, n = 150) +
  geom_smooth(method = "loess", size = .2) +
  geom_hline(aes(yintercept = 0), colour = "#990000", linetype = "dashed", size = .2) +
  theme(axis.title.y = element_text(size = rel(.25), angle = 90),
        axis.title.x = element_text(size = rel(.25), angle = 0),
        axis.text.x = element_text(size = rel(.5), angle = 30, hjust = 1),
        axis.text.y = element_text(size = rel(.5), angle = 0),
        legend.text = element_text(size = rel(.25)),
        legend.position = "bottom",
        legend.title = element_text(size = rel(.25)),
        legend.key = element_rect(size = rel(.25)),
        legend.key.size = unit(.15, "cm"),
        strip.text = element_text(size = rel(.5)),
        strip.text.x = element_text(margin = margin(.05, 0, .05, 0, "cm")),
        strip.text.y = element_text(margin = margin(.05, 0, .05, 0, "cm")))
```

```
library(ggplot2)
ggplot(data = lstk.sss, aes(x = avgDiffElapsed, y = avgDiffLstkValue)) +
  geom_point(aes(colour = arm, shape = arm), size = .05) +
  scale_shape(solid = F) +
  xlab("difference in elapsed days") + ylab("difference in mean livestock value (Tk)") +
  labs(fill = "arm") + facet_grid(arm ~ rd) +
  # stat_smooth(method = "loess", size = .2, n = 150) +
  geom_smooth(method = "loess", size = .2) +
  geom_hline(aes(yintercept = 0), colour = "#990000", linetype = "dashed", size = .2) +
  theme(axis.title.y = element_text(size = rel(.25), angle = 90),
        axis.title.x = element_text(size = rel(.25), angle = 0),
        axis.text.x = element_text(size = rel(.5), angle = 30, hjust = 1),
        axis.text.y = element_text(size = rel(.5), angle = 0),
```

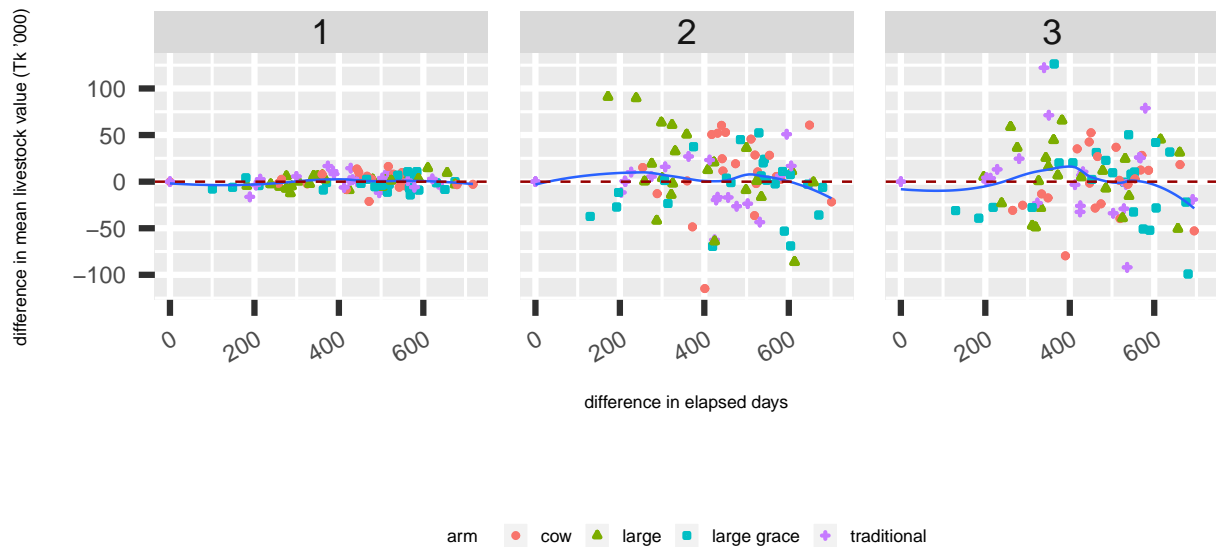


Figure 11 Livestock by elapsed days from disbursement within a group

```

legend.text = element_text(size=rel(.25)),
legend.position = "bottom",
legend.title = element_text(size = rel(.25)),
legend.key = element_rect(size = rel(.25)),
legend.key.size = unit(.15, "cm"),
strip.text = element_text(size=rel(.5)),
strip.text.x = element_text(margin = margin(.05, 0, .05, 0, "cm")),
strip.text.y = element_text(margin = margin(.05, 0, .05, 0, "cm"))

```

Add assets and livestock.

```

al.ss <- merge(asset.ss, lstk.ss,
  by = c("rd", "gid", "hhid", "assignment", "arm", "elapsed"), all = T)
al.ss[is.na(assetNPV), assetNPV := 0]
al.ss[is.na(totalLivestockValue), totalLivestockValue := 0]
al.ss[, val := (assetNPV/1000 + totalLivestockValue)]
setkey(al.ss, rd, gid, assignment)
al.ss[, avgElapsed := mean(elapsed, na.rm = T), by = c("rd", "gid", "assignment")]
al.ss[, avgElapsed0 := avgElapsed[1], by = c("rd", "gid")]
al.ss[, avgElapsed1 := avgElapsed[.N], by = c("rd", "gid")]
al.ss[, avgVal := mean(val, na.rm = T), by = c("rd", "gid", "assignment")]
al.ss[, avgVal0 := avgVal[1], by = c("rd", "gid")]
al.ss[, avgVal1 := avgVal[.N], by = c("rd", "gid")]
unique(al.ss[ gid == 70101 & rd == 1, .(avgElapsed0, avgElapsed1)])

```

```

  avgElapsed0 avgElapsed1
1:         688.3         892

```

```

al.ss[, avgDiffElapsed := avgElapsed1 - avgElapsed0]
al.ss[, avgDiffVal := avgVal1 - avgVal0]
dim(al.sss <- al.ss[!duplicated(al.ss[, .(rd, gid, assignment)]), ])

```

```
[1] 543  80
```

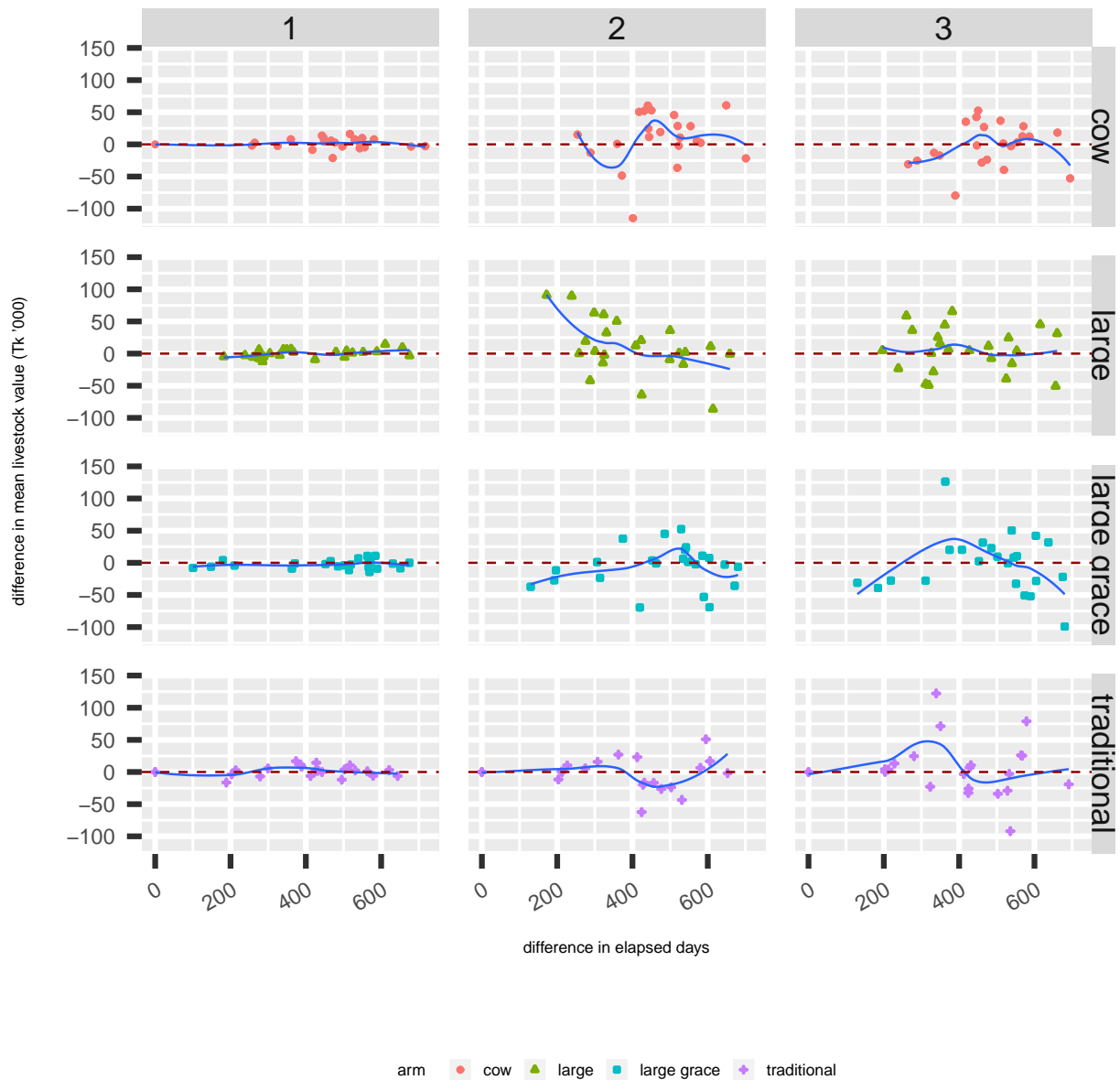


Figure 12 Livestock by elapsed days from disbursement, within a group

```
library(ggplot2)
ggplot(data = al.sss, aes(x = avgDiffElapsed, y = avgDiffVal)) +
  geom_point(aes(colour = arm, shape = arm), size = .05) +
  scale_shape(solid = F) +
  xlab("difference in elapsed days") + ylab("difference in mean value (Tk '000)") +
  labs(fill = "arm") + facet_grid(. ~ rd) +
  # stat_smooth(method = "loess", size = .2, n = 150) +
  geom_smooth(method = "loess", size = .2) +
  geom_hline(aes(yintercept = 0), colour = "#990000", linetype = "dashed", size = .2) +
  theme(axis.title.y = element_text(size = rel(.25), angle = 90),
        axis.title.x = element_text(size = rel(.25), angle = 0),
        axis.text.x = element_text(size = rel(.5), angle = 30, hjust = 1),
        axis.text.y = element_text(size = rel(.5), angle = 0),
        legend.text = element_text(size = rel(.25)),
        legend.position = "bottom",
```

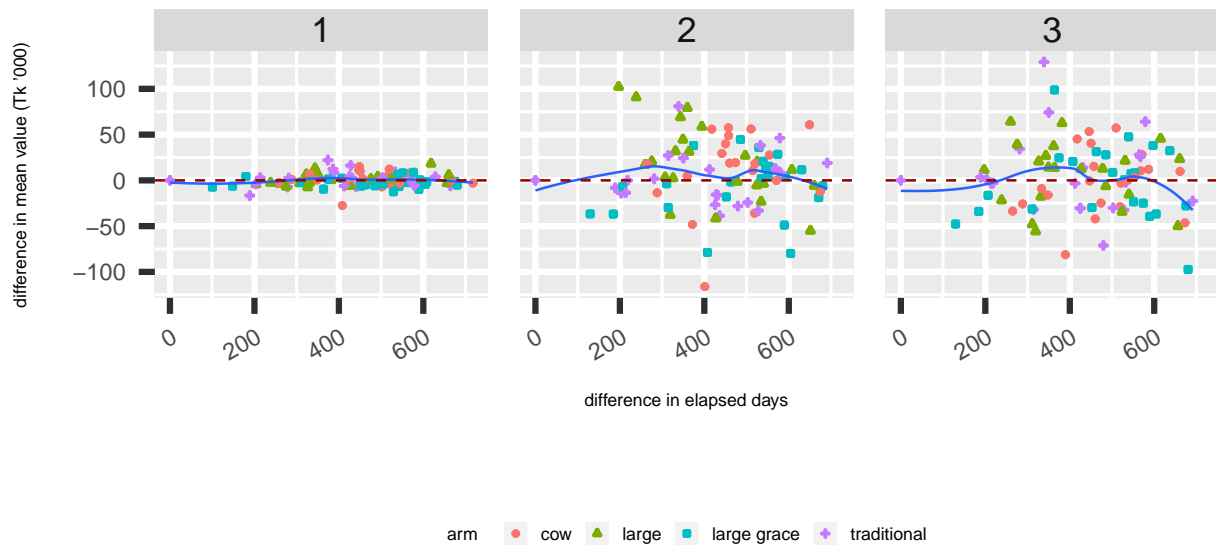



Figure 13 Total assets by elapsed days from disbursement within a group

```

legend.title = element_text(size = rel(.25)),
legend.key = element_rect(size = rel(.25)),
legend.key.size = unit(.15, "cm"),
strip.text = element_text(size=rel(.5)),
strip.text.x = element_text(margin = margin(.05, 0, .05, 0, "cm")),
strip.text.y = element_text(margin = margin(.05, 0, .05, 0, "cm")))

library(ggplot2)
ggplot(data = al.sss, aes(x = avgDiffElapsed, y = avgDiffVal)) +
  geom_point(aes(colour = arm, shape = arm), size = .05) +
  scale_shape(solid = F) +
  xlab("difference in elapsed days") + ylab("difference in mean value (Tk 1000)") +
  labs(fill = "arm") + facet_grid(arm ~ rd) +
# stat_smooth(method = "loess", size = .2, n = 150) +
  geom_smooth(method = "loess", size = .2) +
  geom_hline(aes(yintercept = 0), colour="#990000", linetype="dashed", size = .2) +
  theme(axis.title.y = element_text(size = rel(.25), angle = 90),
        axis.title.x = element_text(size = rel(.25), angle = 0),
        axis.text.x = element_text(size = rel(.5), angle = 30, hjust = 1),
        axis.text.y = element_text(size = rel(.5), angle = 0),
        legend.text = element_text(size=rel(.25)),
        legend.position = "bottom",
        legend.title = element_text(size = rel(.25)),
        legend.key = element_rect(size = rel(.25)),
        legend.key.size = unit(.15, "cm"),
        strip.text = element_text(size=rel(.5)),
        strip.text.x = element_text(margin = margin(.05, 0, .05, 0, "cm")),
        strip.text.y = element_text(margin = margin(.05, 0, .05, 0, "cm")))

```

Regressions. First, get roster files to obtain hh background.

```

setwd(pathsourcemar)
foldername <- list.dirs(path = ".", recursive = T, full.names = T)
foldername <- foldername[grepl("add|ori", foldername)]
fn1 <- unique(list.files(path = foldername, pattern = ".prn$",

```

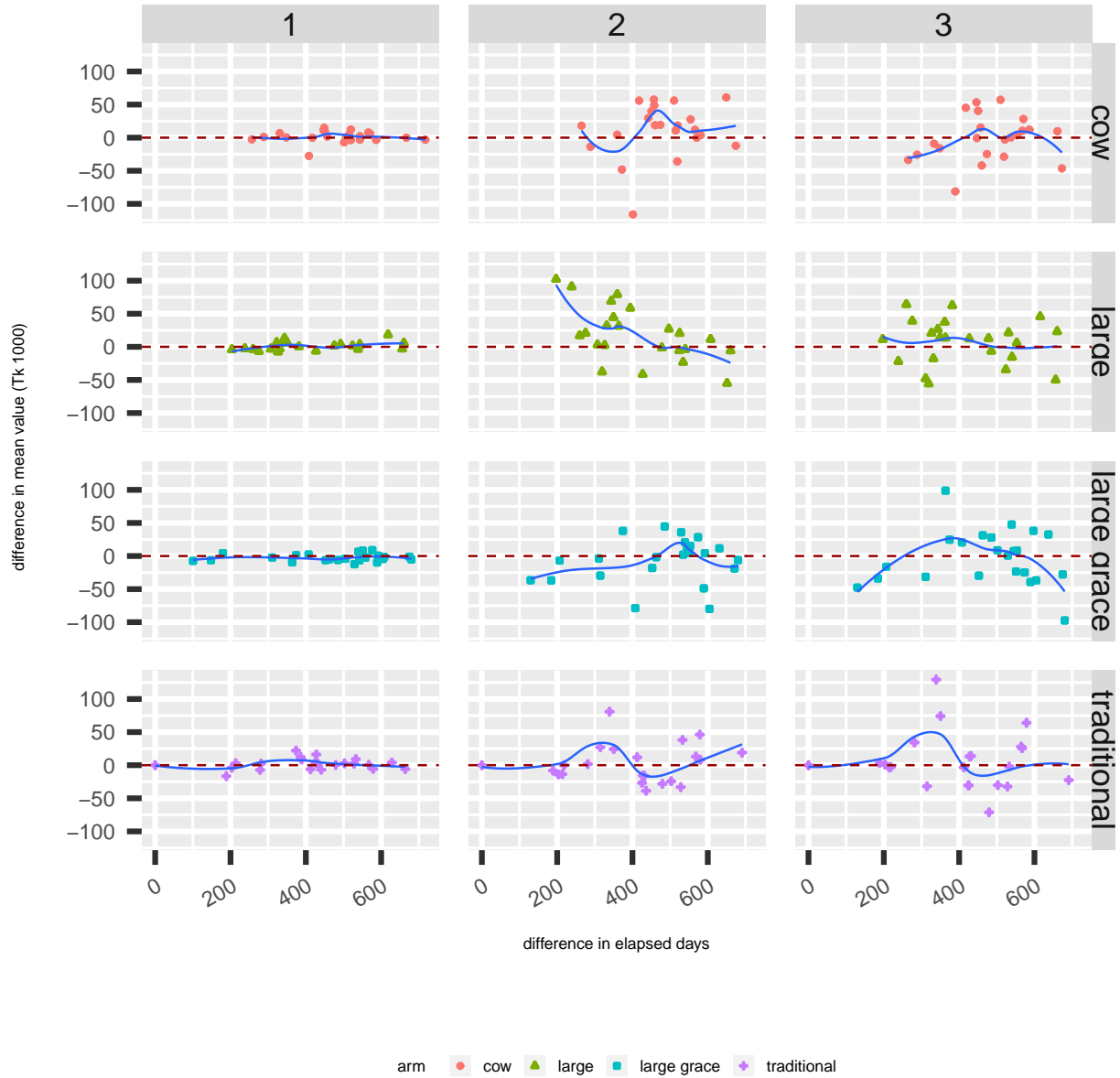


Figure 14 Total assets by elapsed days from disbursement, within a group

```

recursive = T, full.names = T))
fn.ros <- grepout("s1.p1|Se.*01", fn1)
Ro = lapply(fn.ros, fread, integer64 = "double")
ro1 <- rbindlist(Ro, fill = T, use.names = T)
ro1 <- ro1[!duplicated(ro1[, .(hhid, mid, memname)]), ]
ro1[, numAdults := sum(age_1 > 15 & age_1 ≤ 60, na.rm = T), by = hhid]
ro1[, numChildren := sum(age_1 ≤ 15, na.rm = T), by = hhid]
ro1[, numElderly := sum(age_1 > 60, na.rm = T), by = hhid]
ro1[, numDisabled := sum(grepl("Y|1", disability)), by = hhid]
ro1[, numMale := sum(grepl("M|1", sex)), by = hhid]
ro1[, numLiterate := sum(grepl("Can.*and", literacy) | grepl("4", lliteracy)), by = hhid]
ro1[, headLiterate :=
  (grepl("Can.*and", literacy) | grepl("4", lliteracy)) & grepl("He|1", rel_hhh),
  by = hhid]
ro1[, numLiterateMale :=

```

```

      sum((grepl("Can.*and", literacy) | grepl("4", lliteracy)) & grepl("M|1", sex)),
      by = hhid]
ro <- ro1[, .(hhid, numAdults, numChildren, numElderly,
      numDisabled, numMale, numLiterate, numLiterateMale, headLiterate)]
ro <- ro[!duplicated(ro), ]
setwd(pathsave)
write.tablev(ro, "rd1-roster-summary.prn")

```

Summarise at cluster level.

```

tr3 <- tr[, .(gid, hhid)]
setkey(ro1, hhid); setkey(tr3, hhid);
ros <- tr3[ro1]
ros[, size := .N, by = gid]
ros[, ratioAdults := sum(age_1 > 15 & age_1 ≤ 60, na.rm = T)/size, by = gid]
ros[, ratioChildren := sum(age_1 ≤ 15, na.rm = T)/size, by = gid]
ros[, ratioElderly := sum(age_1 > 60, na.rm = T)/size, by = gid]
ros[, ratioDisabled := sum(grepl("Y|1", disability))/size, by = gid]
ros[, ratioMale := sum(grepl("M|1", sex))/size, by = gid]
ros[, ratioLiterate := sum(grepl("Can.*and", literacy) | grepl("4", lliteracy))/size, by =
ros[, ratioHeadLiterate :=
      sum((grepl("Can.*and", literacy) | grepl("4", lliteracy)) &
      grepl("He|1", rel_hhh))/size, by = gid]
ros[, ratioLiterateMale :=
      sum((grepl("Can.*and", literacy) | grepl("4", lliteracy)) & grepl("M|1", sex))/si
      by = gid]
ro2 <- ros[, .(gid, size, ratioAdults, ratioChildren, ratioElderly,
      ratioDisabled, ratioMale, ratioLiterate, ratioLiterateMale, ratioHeadLiterate)]
ro2 <- ro2[!duplicated(ro2[, gid]), ]
ro2 <- ro2[!is.na(gid), ]

```

Merge with asset data.

```

setkey(alr.sss, gid, rd); setkey(ro2, gid)
alr.sss <- ro2[alr.sss]
dim(alr.sss <- alr.sss[!duplicated(alr.sss[, .(rd, gid)]), ])

```

```
[1] 276 89
```

```

setkey(alr.sss, gid, rd)
alr.sss[, exist := .N, by = gid]
dim(alr.sss <- alr.sss[exist == 3, ])

```

```
[1] 276 90
```

```

destat.alr <- destat(alr.sss[, .(elapsed,
      size, ratioChildren, ratioAdults, ratioDisabled, ratioMale, ratioLiterate,
      ratioLiterateMale, ratioHeadLiterate, avgDiffElapsed,
      avgDiffVal, avgVal, avgVal1, avgVal0, avgElapsed, avgElapsed1, avgElapsed0)])
destat.alr <- cbind(rownames(destat.alr), destat.alr)
setwd(pathsave)
ltab.alr <- latextab(destat.alr, headercolor = "blue!10",
      alternatcolor = "gray90", delimiterline = NULL,
      hleft = c("\\footnotesize\\hfill", rep("\\footnotesize\\hfil$", ncol(destat.alr)-1)),
      hcenter = c("2", rep("1.0", ncol(destat.alr)-1)),
      hright = c("", rep("$", ncol(destat.alr)-1)))
write.tablev(ltab.alr, "destat_alr_sss.tex", colnamestrue = F)
dalr.sss1 <- cbind(d.rd = 1, alr.sss[rd==1, .(gid, assignment, arm, elapsed,

```

```

      size , ratioChildren , ratioAdults , ratioDisabled , ratioMale , ratioLiterate ,
      ratioLiterateMale , ratioHeadLiterate , avgDiffElapsed)],
      alr.sss[rd==2, .(avgDiffVal , avgDiffLstkValue , avgVal , avgVal1 , avgVal0 , avgElapse
      alr.sss[rd==1, .(avgDiffVal , avgDiffLstkValue , avgVal , avgVal1 , avgVal0 , avgElapse
dalr.sss2 ← cbind(d.rd = 2, alr.sss[rd==1, .(gid , assignment , arm , elapsed ,
      size , ratioChildren , ratioAdults , ratioDisabled , ratioMale , ratioLiterate ,
      ratioLiterateMale , ratioHeadLiterate , avgDiffElapsed)],
      alr.sss[rd==3, .(avgDiffVal , avgDiffLstkValue , avgVal , avgVal1 , avgVal0 , avgElapse
      alr.sss[rd==2, .(avgDiffVal , avgDiffLstkValue , avgVal , avgVal1 , avgVal0 , avgElapse
dalr.sss ← rbind(dalr.sss1 , dalr.sss2)

l1 ← glm(avgDiffVal ~ avgDiffElapsed , data = dalr.sss)
l2 ← glm(avgDiffVal ~ avgDiffElapsed:arm, data = dalr.sss)
l3 ← glm(avgDiffVal ~ avgDiffElapsed:arm +
      size +ratioChildren +ratioAdults +ratioDisabled +ratioMale , data = dalr.sss)
l4 ← glm(avgDiffVal ~ avgDiffElapsed:arm +
      size +ratioChildren +ratioAdults +ratioDisabled +ratioMale +
      ratioLiterate + ratioLiterateMale + ratioHeadLiterate , data = dalr.sss)
linprob ← list(l1 , l2 , l3 , l4)
linest ← lapply(linprob , clx.regobj , Cluster = "gid")
linest ← lapply(linest , function(x) x[, -3])
linest ← tabs2latex(linest)
R2 ← round(asn(lapply(linprob ,
      function(x) 1-crossprod(summary(x)$deviance.res)/summary(x)$null.dev)), 3)
en ← asn(lapply(linprob , function(x) length(x$y)))
rn ← rownames(linest)
rn ← gsub("arm|^se.*", "", rn)
rn ← gsub(":", " * ", rn)
ltab ← rbind(as.matrix(cbind(rn , linest)), c("$R^{2}$", R2),
      c("$n$", en))
write.tablev(latextab(ltab , delimiterline = NULL, alternatecolor2 = "gray90",
      hleft = c("\\footnotesize", rep("\\scriptsize\\hfil$", ncol(ltab)-1)),
      hcenter = c(3.5 , rep(1.5 , ncol(ltab)-1)),
      hright = c("\\hfill", rep("$", ncol(ltab)-1)),
      adjustlineskip = "-.4ex"),
      paste0(pathsave , "asset_regression_alr_sss.tex"), colnamestrue = F)

l11 ← glm(avgDiffLstkValue ~ avgDiffElapsed , data = dalr.sss)
l12 ← glm(avgDiffLstkValue ~ avgDiffElapsed:arm, data = dalr.sss)
l13 ← glm(avgDiffLstkValue ~ avgDiffElapsed:arm +
      size +ratioChildren +ratioAdults +ratioDisabled +ratioMale , data = dalr.sss)
l14 ← glm(avgDiffLstkValue ~ avgDiffElapsed:arm +
      size +ratioChildren +ratioAdults +ratioDisabled +ratioMale +
      ratioLiterate + ratioLiterateMale + ratioHeadLiterate , data = dalr.sss)
l1linprob ← list(l11 , l12 , l13 , l14)
l1linest ← lapply(l1linprob , clx.regobj , Cluster = "gid")
l1linest ← lapply(l1linest , function(x) x[, -3])
l1linest ← tabs2latex(l1linest)
R2 ← round(asn(lapply(l1linprob ,
      function(x) 1-crossprod(summary(x)$deviance.res)/summary(x)$null.dev)), 3)
en ← asn(lapply(l1linprob , function(x) length(x$y)))
rn ← rownames(l1linest)
rn ← gsub("arm|^se.*", "", rn)
rn ← gsub(":", " * ", rn)
l1ltab ← rbind(as.matrix(cbind(rn , l1linest)), c("$R^{2}$", R2),
      c("$n$", en))

```

TABLE 2: DESCRIPTIVE STATISTICS OF ASSET REGRESSION DATA

	min	25%	median	75%	max	mean	std	0s	NAs	n
elapsed	49	296	352	556	892	400.8	204.6	0	0	276
size	65	82.8	89	97	171	91.8	17	0	0	276
ratioChildren	0.3	0.4	0.4	0.5	0.6	0.4	0.1	0	0	276
ratioAdults	0.4	0.5	0.6	0.6	0.7	0.6	0.1	0	0	276
ratioDisabled	0	0	0	0	0	0	0	168	0	276
ratioMale	0.4	0.5	0.5	0.5	0.6	0.5	0	0	0	276
ratioLiterate	0	0.2	0.3	0.4	0.5	0.3	0.1	0	0	276
ratioLiterateMale	0	0.1	0.2	0.2	0.3	0.2	0.1	0	0	276
ratioHeadLiterate	0	0	0	0	0.1	0	0	51	0	276
avgDiffElapsed	0	331.6	452.3	543.5	717.4	431.7	156.3	9	0	276
avgDiffVal	-116	-7.1	1	14.3	129.4	3	30.7	9	0	276
avgVal	1.2	15.2	43.2	87.6	192.4	56.4	48.3	0	0	276
avgVal1	0.7	15.3	54.8	88.8	230.4	59.4	48.4	0	0	276
avgVal0	1.2	15.2	43.2	87.6	192.4	56.4	48.3	0	0	276
avgElapsed	122.6	267.9	346.3	477.5	892	374.5	145.7	0	0	276
avgElapsed1	544.3	729	834.8	857.3	899	806.2	71.8	0	0	276
avgElapsed0	122.6	267.9	346.3	477.5	892	374.5	145.7	0	0	276

```

write.tablev(latexTab(lftab, delimiterline = NULL, alternatecolor2 = "gray90",
  hleft = c("\\footnotesize", rep("\\scriptsize\\hfil$", ncol(lftab)-1)),
  hcenter = c(3.5, rep(1.5, ncol(lftab)-1)),
  hright = c("\\hfill", rep("$", ncol(lftab)-1)),
  adjustlineskip = "-.4ex"),
  paste0(pathsave, "livestock_regression_alr_sss.tex"), colnamestrue = F)

```

References

- Egger, Peter H. and Maximilian von Ehrlich**, “Generalized propensity scores for multiple continuous treatment variables,” *Economics Letters*, 2013, 119 (1), 32 – 34.
- Hirano, Keisuke and Guido W. Imbens**, *The Propensity Score with Continuous Treatments*, John Wiley & Sons, Ltd,
- Imai, Kosuke and David A van Dyk**, “Causal Inference With General Treatment Regimes,” *Journal of the American Statistical Association*, 2004, 99 (467), 854–866.
- Imbens, Guido W.**, “The role of the propensity score in estimating dose-response functions,” *Biometrika*, 2000, 87 (3), 706.
- Kluve, Jochen, Hilmar Schneider, Arne Uhlendorff, and Zhong Zhao**, “Evaluating continuous training programmes by using the generalized propensity score,” *Journal of the Royal Statistical Society: Series A (Statistics in Society)*, 2012, 175 (2), 587–617.

TABLE 3: DID ESTIMATES OF ASSET IMPACTS

rn	(1)	(2)	(3)	(4)
(Intercept)	3.511 (4.794)	3.127 (4.744)	-177.737 (162.074)	-156.113 (162.405)
avgDiffElapsed	-0.006 (0.010)			
avgDiffElapsed * cow		-0.008 (0.011)	-0.006 (0.010)	-0.003 (0.011)
avgDiffElapsed * large		-0.003 (0.012)	-0.003 (0.011)	-0.003 (0.011)
avgDiffElapsed * large grace		-0.006 (0.013)	-0.009 (0.013)	-0.005 (0.013)
avgDiffElapsed * traditional		-0.004 (0.014)	-0.010 (0.014)	-0.008 (0.015)
size			-0.027 (0.107)	-0.007 (0.104)
ratioChildren			150.652 (165.090)	144.698 (161.753)
ratioAdults			202.275 (165.389)	191.880 (160.061)
ratioDisabled			-374.066** (173.656)	-364.441** (171.045)
ratioMale			17.408 (33.939)	-28.147 (50.547)
ratioLiterate				-33.128 (36.090)
ratioLiterateMale				105.570 (77.780)
ratioHeadLiterate				-6.597 (110.909)
R^2	0.001	0.001	0.022	0.028
n	184	184	184	184

- Notes: 1. Difference-in-differences estimates of asset accumulation against elapsed days.
2. large, large grace, cow are all time invariant and are interacted with a trend term.
3. *, **, *** indicate significance levels at 10%, 5%, 1%, respectively.

TABLE 4: DID ESTIMATES OF LIVESTOCK IMPACTS

rn	(1)	(2)	(3)	(4)
(Intercept)	3.760 (5.113)	3.312 (5.221)	-132.813 (243.731)	-107.738 (235.304)
avgDiffElapsed	-0.006 (0.012)			
avgDiffElapsed * cow		-0.009 (0.012)	-0.006 (0.013)	- 0.000 (0.014)
avgDiffElapsed * large		0.002 (0.014)	0.001 (0.013)	- 0.000 (0.015)
avgDiffElapsed * large grace		-0.005 (0.014)	-0.010 (0.016)	-0.007 (0.017)
avgDiffElapsed * traditional		-0.009 (0.019)	-0.022 (0.019)	-0.020 (0.021)
size			-0.004 (0.113)	0.043 (0.118)
ratioChildren			61.527 (241.252)	63.033 (231.535)
ratioAdults			163.485 (249.577)	171.256 (236.432)
ratioDisabled			-392.670** (199.920)	-413.142** (210.322)
ratioMale			45.227 (42.598)	-37.036 (54.860)
ratioLiterate				-61.047 (42.057)
ratioLiterateMale				174.673** (85.504)
ratioHeadLiterate				-115.088 (125.939)
R^2	0.001	0.003	0.033	0.045
n	141	141	141	141

- Notes: 1. Difference-in-differences estimates of asset accumulation against elapsed days.
2. large, large grace, cow are all time invariant and are interacted with a trend term.
3. *, **, * * * indicate significance levels at 10%, 5%, 1%, respectively.