

# Ghana Agricultural Profits Analysis Steps and Explanations

Code ▾

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## Read data

After reading through the documentation, we determined these data files contain the data we want to use to analyze the influencing factors of Ghana agricultural profits

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```
# list of data file pairs to read, first is subfolder, second is file name
datafiles <- list(c("aggregates/", "agg2"),
                 c("", "sec8b"),
                 c("", "sec0a"),
                 c("", "sec1"),
                 c("", "sec2a"),
                 c("", "sec8a2"),
                 c("", "sec8a3"),
                 c("", "sec8c1"),
                 c("", "sec8c2"),
                 c("community/", "cs2"))
invisible(sapply(datafiles, readRawData))
```

## Calculate agricultural profit per area unit

First we need to convert different area unit into a same one. Looking at count of each unit, acre is the major one, so we decide to map other units to acre. According to G4USERSG.pdf, we create a map between each unit to a multiplier of it when it's converted to acre (the "Other" unit is treated as NA). Then join the map onto `sec8b` and multiply with original land size to get unified land size (values greater than  $1e+100$  are treated as NA). We noticed that some households have multiple entries of land size info and among them there is NA value. So we convert NA to 0 so we can sum up land size for each household. Eventually, household with land size less than 0.5 acre are filtered out because that's probably not a real farm.

Area unit	To Acre
1 Acre	1 Acre
1 Poles	1 Acre

1 Ropes

1/9 Acre

Other

NA

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```
# map land size units to acre with a multiplier (s8bq4bm)
# the "Other" unit is treated as NA
table(sec8b$s8bq4b)
```

1	2	3
7725	1477	1354
4	1.74980057982641e+100	
13	1	

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```
land_size_unit_map <- data.frame("s8bq4b" = as.double(c(1:4)),
                                "s8bq4bm" = c(1, 1, 1/9, NA))
# convert all land size to acre
land_size_info <- sec8b %>%
  select(clust, nh, s8bq4a, s8bq4b) %>%
  mutate(s8bq4a = ifelse(s8bq4a > 1e+100, NA, s8bq4a),
         s8bq4b = ifelse(s8bq4b > 1e+100, NA, s8bq4b)) %>%
  left_join(land_size_unit_map) %>%
  # s8bq4ac is corrected s8bq4a, which is land size in acre
  mutate(s8bq4ac = s8bq4a * s8bq4bm) %>%
  # convert NA to 0 to calculate sum of land size for each household
  replace(., is.na(.), 0) %>%
  group_by(clust, nh) %>%
  summarise(landSize = sum(s8bq4ac)) %>%
  # remove household with small land size
  filter(landSize >= 0.5)
```

Joining, by = "s8bq4b"

Then we use agricultural income minus depreciation of farm equipment (`agri1 - hhagdepn`) as household profit, and divide it by household land size to get profit per area unit. `agg2` does not have NA value so no need to clean up.

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```
# identify count of NAs in agg2
colSums(is.na(agg2))
```

```

  clust      nh      agril      agrilc      agri2      agri2c      hhagdepn
    0         0         0         0         0         0         0

```

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```

# check correlation between agricultural income and corrected agricultural income
cor(agg2$agril, agg2$agrilc)

```

```
[1] 1
```

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```

# resolve warnings during join
attr(agg2$clust, "label") <- "Enumeration Area number"
attr(agg2$nh, "label") <- "Household ID"
# calculate household agri profit per acre
# because the above correlations are both 1,
# use only "agril" to calculate agricultural profit
hh_profit_info <- agg2 %>%
  select(clust, nh, agril, hhagdepn) %>%
  filter(agril != 0) %>%
  inner_join(land_size_info, by = c("clust", "nh")) %>%
  mutate(profit = (agril - hhagdepn) / landSize) %>%
  select(-agril, -hhagdepn, -landSize)
attr(hh_profit_info$profit, "label") <- "HH agri profit"

```

## Tidy household basic information

For basic household information, we are interested in language of respondent (values greater than 1e+100 are treated as “Unknown” language), ecological zone number and 3 different locality classifications. All variables are converted to factor type so they could be used as dummy variables in linear model.

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```
# household basic information
hh_basic_info <- sec0a %>%
  select(region, district, eanum, clust, nh, reslan, ez:loc3) %>%
  mutate(reslan = cleanAndFactorize(reslan, 1e+100, 99,
                                    levels = as.character(c(1:8, 99)),
                                    labels = c("English", "Akan", "Ewe", "GaAdangbe",
                                                "Dagbani", "Hausa", "Nzema", "Other",
                                                "Unknown")),
         ez = factor(ez,
                     levels = as.character(c(1:3)),
                     labels = c("Coastal", "Forest", "Savannah")),
         loc2 = factor(loc2,
                       levels = as.character(c(1,2)),
                       labels = c("Urban", "Rural")),
         loc5 = factor(loc5,
                       levels = as.character(c(1:5)),
                       labels = c("Accra", "OtherUrban", "RuralCoastal",
                                   "RuralForest", "RuralSavannah")),
         loc3 = factor(loc3,
                       levels = as.character(c(1:3)),
                       labels = c("Accra", "OtherUrban", "Rural"))
  )
```

## Tidy household member information

We picked some data about household members such as sex, age and education and aggregate them on household level to analyze the effect of variety of household members on agricultural profit.

Select household members sex and age information and calculate mean, max and min age of household. The relation ( rel ) data is used to filter by head of household later.

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```
# household member information
hbm_info <- sec1 %>%
  select(clust, nh, pid, sex, agey, rel) %>%
  group_by(clust, nh) %>%
  mutate(female = sex == 2,
         age = agey,
         avgAge = mean(agey),
         maxAge = max(agey),
         minAge = min(agey)) %>%
  select(-sex, -agey)
```

For education data, we categorize educational qualification ( `s2aq3` ) into 5 levels: “None”, “BasicEducation”, “SecondaryEducation”, “TertiaryEducation” and “Other”, based on information from the two sources below. For missing value in `s2aq3` , it’s because these people never attended school ( `s2aq1 == 2` ). In this case we set education to None.

Education System in Ghana (<https://www.scholaro.com/pro/Countries/Ghana/Education-System>)

Education in Ghana Wikipedia ([https://en.wikipedia.org/wiki/Education\\_in\\_Ghana](https://en.wikipedia.org/wiki/Education_in_Ghana))

Educational Qualification	Education Level
01 None	None
02 MSLC?BECE	BasicEducation
03 Voc/Comm	SecondaryEducation
04 ‘O’ Level	SecondaryEducation
05 SSS	SecondaryEducation
06 ‘A’ Level	SecondaryEducation
07 T/T Cert. B	TertiaryEducation
08 T /T Cert. A	TertiaryEducation
09 Nursing	TertiaryEducation
10 Tech/Prof Cert.	TertiaryEducation
11 Tech/Prof Dip	TertiaryEducation
12 Bachelor	TertiaryEducation
13 Masters	TertiaryEducation
14 Doctorate	TertiaryEducation
96 Other	Other

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```
# household member education information
# map education qualification to education level
educ_level_map <- data.frame("s2aq3" = c(1:14, 96),
                             "s2aq3l" = c(1, 2, rep(3, 4), rep(4, 8), 5))

# resolve warnings
attr(educ_level_map$s2aq3, "label") <- "Highest educ qualification"
attr(educ_level_map$s2aq3, "format.stata") <- "%10.0g"

# Check NAs
colSums(is.na(sec2a))
```

nh	pid	s1q23	clust	s2aq1	s2aq2	s2aq3	s2aq4	s2aq5a
0	0	0	0	0	0	0	0	0
s2aq5b	s2aq6	s2aq7	s2aq8	s2aq9	s2aq10	s2aq11	s2aq12	s2aq13
0	0	0	0	0	0	0	0	0
s2aq13a	s2aq14	s2aq15	s2aq16	s2aq17				
0	0	0	0	0				

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```
hbm_educ <- sec2a %>%
  select(clust, nh, pid, s2aq1, s2aq3) %>%
  # if never attended school (s2aq1 == 2), set education to None (s2aq3 = 1)
  mutate(s2aq3 = ifelse(s2aq1 == 2, 1, s2aq3)) %>%
  left_join(educ_level_map) %>%
  mutate(educ = factor(s2aq3l,
                      levels = as.character(c(1:5)),
                      labels = c("None", "BasicEducation", "SecondaryEducation",
                                "TertiaryEducation", "Other"))) %>%
  select(-s2aq1, -s2aq3, -s2aq3l)
```

Joining, by = "s2aq3"

Then we join household member info and education info, and use data from head of household (rel == 1) as a representative of the household.

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```
hh_head_info <- hbm_info %>%
  inner_join(hbm_educ, by=c("clust", "nh", "pid")) %>%
  filter(rel == 1) %>% # head of household
  select(-pid, -rel)
```

## Tidy agricultural characteristics information

We included agricultural characteristics information in our analysis as well, including Livestock and Fishing ( sec8a2 ), Agricultural equipment ( sec8a3 ) and Crops ( sec8c1 and sec8c2 ). We would like to see if any certain types of livestock, equipment and/or crop would contribute a lot to profit.

For livestock data, we found that there are too many missing values (value greater than 1e+100) in its unit of measure ( s8aq22b ). So we decided to spread livestock count by livestock type so that each livestock type is a column with the count as its value, and assume same livestock type has same unit of measure so we can compare across them. We also created a column for the count of different livestock types each household owns.

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```
table(sec8a2$s8aq22b)
```

1	2	3
32	41	22
4	1.74980057982641e+100	
30	4716	

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```
# spread livestock count and count livestock type
# since there are too many missing value in livestock unit of measure (s8aq22b),
# we assume same livestock type has same unit,
# and we only take livestock count (s8aq22a) into consideration
hh_livestock_info <- sec8a2 %>%
  select(clust, nh, livstcd, s8aq22a) %>%
  group_by(clust, nh) %>%
  mutate(livstcdTypeCount = n()) %>%
  spread(key = livstcd,
         value = s8aq22a,
         fill = 0,
         sep = "")
summary(hh_livestock_info)
```

clust	nh	livstcdTypeCount	livstcd1
Min. :4002	Min. : 1.00	Min. :1.00	Min. :0.0000
1st Qu.:4395	1st Qu.: 6.00	1st Qu.:1.00	1st Qu.:0.0000
Median :4589	Median :11.00	Median :2.00	Median :0.0000
Mean :4580	Mean :10.79	Mean :2.13	Mean :0.1205
3rd Qu.:4855	3rd Qu.:16.00	3rd Qu.:3.00	3rd Qu.:0.0000
Max. :4999	Max. :25.00	Max. :7.00	Max. :9.0000

livstcd2	livstcd3	livstcd4	livstcd5
Min. : 0.0000	Min. : 0	Min. : 0	Min. : 0.0000
1st Qu.: 0.0000	1st Qu.: 0	1st Qu.: 0	1st Qu.: 0.0000
Median : 0.0000	Median : 0	Median : 1	Median : 0.0000
Mean : 0.8517	Mean : 464	Mean : 1178	Mean : 0.6881
3rd Qu.: 0.0000	3rd Qu.: 3	3rd Qu.: 5	3rd Qu.: 0.0000
Max. :90.0000	Max. :1050000	Max. :2670000	Max. :36.0000

livstcd6	livstcd7	livstcd8	livstcd9
Min. :0.000	Min. : 0.0	Min. : 0.000	Min. : 0.00000
1st Qu.:0.000	1st Qu.: 1.0	1st Qu.: 0.000	1st Qu.: 0.00000
Median :0.000	Median : 6.0	Median : 0.000	Median : 0.00000
Mean :0.022	Mean : 146.1	Mean : 2.227	Mean : 0.07787
3rd Qu.:0.000	3rd Qu.: 12.0	3rd Qu.: 0.000	3rd Qu.: 0.00000
Max. :8.000	Max. :130000.0	Max. :150.000	Max. :40.00000

livstcd10	livstcd11	livstcd12
Min. : 0.000	Min. : 0.00000	Min. : 0.00000
1st Qu.: 0.000	1st Qu.: 0.00000	1st Qu.: 0.00000
Median : 0.000	Median : 0.00000	Median : 0.00000
Mean : 9.321	Mean : 0.06599	Mean : 0.06247
3rd Qu.: 0.000	3rd Qu.: 0.00000	3rd Qu.: 0.00000
Max. :13000.000	Max. :100.00000	Max. :50.00000

Similarly we process the agricultural equipment data.

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```
# spread agricultural equipment count and count agricultural equipment type
hh_equip_info <- sec8a3 %>%
  select(clust, nh, eqcdwn, s8aq34) %>%
  replace(., . > 1e+100, 0) %>%
  filter(s8aq34 != 0) %>%
  group_by(clust, nh) %>%
  mutate(equipTypeCount = n()) %>%
  spread(key = eqcdwn,
         value = s8aq34,
         fill = 0,
         sep = "")
summary(hh_equip_info)
```



clust	nh	equipTypeCount	eqcdwn21
Min. :4002	Min. : 1.00	Min. :1.000	Min. :0.000000
1st Qu.:4125	1st Qu.: 6.00	1st Qu.:1.000	1st Qu.:0.000000
Median :4539	Median :11.00	Median :1.000	Median :0.000000
Mean :4481	Mean :10.71	Mean :1.142	Mean :0.003823
3rd Qu.:4795	3rd Qu.:16.00	3rd Qu.:1.000	3rd Qu.:0.000000
Max. :4992	Max. :25.00	Max. :4.000	Max. :1.000000

eqcdwn22	eqcdwn31	eqcdwn51	eqcdwn61
Min. :0.00000	Min. :0.000	Min. : 0.0000	Min. :0.0000
1st Qu.:0.00000	1st Qu.:0.000	1st Qu.: 0.0000	1st Qu.:0.0000
Median :0.00000	Median :0.000	Median : 0.0000	Median :0.0000
Mean :0.03287	Mean :0.013	Mean : 0.2378	Mean :0.0107
3rd Qu.:0.00000	3rd Qu.:0.000	3rd Qu.: 0.0000	3rd Qu.:0.0000
Max. :2.00000	Max. :2.000	Max. :160.0000	Max. :1.0000

eqcdwn62	eqcdwn63	eqcdwn64	eqcdwn65
Min. : 0.0000	Min. : 0.0000	Min. :0.0000	Min. : 0.000
1st Qu.: 0.0000	1st Qu.: 0.0000	1st Qu.:0.0000	1st Qu.: 1.000
Median : 0.0000	Median : 0.0000	Median :0.0000	Median : 3.000
Mean : 0.2791	Mean : 0.6017	Mean :0.0107	Mean : 3.993
3rd Qu.: 0.0000	3rd Qu.: 0.0000	3rd Qu.:0.0000	3rd Qu.: 4.000
Max. :130.0000	Max. :236.0000	Max. :7.0000	Max. :800.000

For crop data, we noticed that some households have multiple entries of same crop type. So we need to group by household and crop type and sum up across same crop type.

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```
# spread household harvested crop count and count harvested crop type
# some households have multiple entries of same crop type, for example:
(sec8c1 %>% select(clust, nh, cropcd, s8cq3a))[c(377, 378),]
```

clust <dbl>	nh <dbl>	cropcd <dbl>	s8cq3a <dbl>
4049	18	1	10.0
4049	18	1	0.5

2 rows

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```
# so need to sum up quantity of same crop type
hh_crop_info <- sec8c1 %>%
  select(clust, nh, cropcd, s8cq3a) %>%
  replace(., . > 1e+100, 0) %>%
  group_by(clust, nh, cropcd) %>%
  # sum up quantity of same crop type
  summarise(s8cq3ac = sum(s8cq3a)) %>%
  group_by(clust, nh) %>%
  mutate(cropTypeCount = n()) %>%
  spread(key = cropcd,
         value = s8cq3ac,
         fill = 0,
         sep = "")
summary(hh_crop_info)
```

clust	nh	cropTypeCount		cropcd0	
Min. :4002	Min. : 1.00	Min. :1.000	Min. : 0.000000	Min. : 0.000000	Min. : 0.000000
1st Qu.:4392	1st Qu.: 6.00	1st Qu.:1.000	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000000
Median :4572	Median :11.00	Median :2.000	Median :0.000000	Median :0.000000	Median :0.000000
Mean :4560	Mean :10.83	Mean :2.012	Mean :0.003218	Mean :0.003218	Mean :0.003218
3rd Qu.:4819	3rd Qu.:16.00	3rd Qu.:3.000	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000000
Max. :4999	Max. :25.00	Max. :7.000	Max. :12.000000	Max. :12.000000	Max. :12.000000

cropcd1		cropcd2		cropcd3	
Min. : 0.000	Min. : 0.00000	Min. : 0.00000	Min. :0.0000000	Min. : 0.00000	Min. :0.0000000
1st Qu.: 0.000	1st Qu.: 0.00000	1st Qu.: 0.00000	1st Qu.:0.0000000	1st Qu.: 0.00000	1st Qu.:0.0000000
Median : 0.000	Median : 0.00000	Median : 0.00000	Median :0.0000000	Median : 0.00000	Median :0.0000000
Mean : 3.781	Mean : 0.06846	Mean : 0.06846	Mean :0.0008045	Mean : 0.00000	Mean :0.0008045
3rd Qu.: 0.000	3rd Qu.: 0.00000	3rd Qu.: 0.00000	3rd Qu.:0.0000000	3rd Qu.: 0.00000	3rd Qu.:0.0000000
Max. :800.000	Max. :227.80000	Max. :227.80000	Max. :3.0000000	Max. : 0.00000	Max. :3.0000000

cropcd4		cropcd5		cropcd6	
Min. : 0.000	Min. :0.000000	Min. :0.000000	Min. : 0.00000	Min. : 0.00000	Min. : 0.00000
1st Qu.: 0.000	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.: 0.00000	1st Qu.: 0.00000	1st Qu.: 0.00000
Median : 0.000	Median :0.000000	Median :0.000000	Median : 0.00000	Median : 0.00000	Median : 0.00000
Mean : 9.701	Mean :0.002145	Mean :0.002145	Mean : 0.03433	Mean : 0.03433	Mean : 0.03433
3rd Qu.: 0.000	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.: 0.00000	3rd Qu.: 0.00000	3rd Qu.: 0.00000
Max. : 999.990	Max. :6.000000	Max. :6.000000	Max. :125.00000	Max. :125.00000	Max. :125.00000

cropcd8		cropcd9		cropcd10	
Min. : 0.0000	Min. : 0.000000	Min. : 0.000000	Min. : 0.000	Min. : 0.000	Min. : 0.000
1st Qu.: 0.0000	1st Qu.: 0.000000	1st Qu.: 0.000000	1st Qu.: 0.000	1st Qu.: 0.000	1st Qu.: 0.000
Median : 0.0000	Median : 0.000000	Median : 0.000000	Median : 0.000	Median : 0.000	Median : 0.000
Mean : 0.5093	Mean : 0.004023	Mean : 0.004023	Mean : 6.393	Mean : 6.393	Mean : 6.393
3rd Qu.: 0.0000	3rd Qu.: 0.000000	3rd Qu.: 0.000000	3rd Qu.: 0.000	3rd Qu.: 0.000	3rd Qu.: 0.000
Max. :999.0000	Max. :15.000000	Max. :15.000000	Max. :300.000	Max. :300.000	Max. :300.000

cropcd11		cropcd12		cropcd13		cropcd14	
Min. : 0.00000	Min. : 0.00000	Min. : 0.00000	Min. : 0.00000	Min. : 0.00000	Min. : 0.00000	Min. : 0.000	Min. : 0.000
1st Qu.: 0.00000	1st Qu.: 0.00000	1st Qu.: 0.00000	1st Qu.: 0.00000	1st Qu.: 0.00000	1st Qu.: 0.00000	1st Qu.: 0.000	1st Qu.: 0.000

Median : 0.00000	Median : 0.00000	Median : 0.0000	Median : 0.000
Mean : 0.01341	Mean : 0.02105	Mean : 0.1077	Mean : 1.671
3rd Qu.: 0.00000	3rd Qu.: 0.00000	3rd Qu.: 0.0000	3rd Qu.: 0.000
Max. :50.00000	Max. :61.00000	Max. :302.0000	Max. :132.000
cropcd15	cropcd16	cropcd17	cropcd18
Min. : 0.0000	Min. : 0.00000	Min. : 0.000	Min. : 0.000
1st Qu.: 0.0000	1st Qu.: 0.00000	1st Qu.: 0.000	1st Qu.: 0.000
Median : 0.0000	Median : 0.00000	Median : 0.000	Median : 0.000
Mean : 0.1483	Mean : 0.01341	Mean : 2.198	Mean : 0.341
3rd Qu.: 0.0000	3rd Qu.: 0.00000	3rd Qu.: 0.000	3rd Qu.: 0.000
Max. :200.0000	Max. :50.00000	Max. :999.000	Max. :375.000
cropcd19	cropcd20	cropcd21	
Min. : 0.000	Min. : 0.000000	Min. : 0.000000	
1st Qu.: 0.000	1st Qu.: 0.000000	1st Qu.: 0.000000	
Median : 0.000	Median : 0.000000	Median : 0.000000	
Mean : 0.458	Mean : 0.005363	Mean : 0.009118	
3rd Qu.: 0.000	3rd Qu.: 0.000000	3rd Qu.: 0.000000	
Max. :990.000	Max. :20.000000	Max. :10.000000	
cropcd22	cropcd23	cropcd24	cropcd25
Min. : 0.00	Min. : 0.0	Min. : 0.0000	Min. : 0.0000
1st Qu.: 1.00	1st Qu.: 0.0	1st Qu.: 0.0000	1st Qu.: 0.0000
Median : 4.00	Median : 0.0	Median : 0.0000	Median : 0.0000
Mean : 21.69	Mean : 1.2	Mean : 0.7434	Mean : 0.1372
3rd Qu.: 11.00	3rd Qu.: 0.0	3rd Qu.: 0.0000	3rd Qu.: 0.0000
Max. :1029.00	Max. :180.0	Max. :40.0000	Max. :102.0000
cropcd26	cropcd27	cropcd28	
Min. : 0.000000	Min. : 0.00000	Min. : 0.0000	
1st Qu.: 0.000000	1st Qu.: 0.00000	1st Qu.: 0.0000	
Median : 0.000000	Median : 0.00000	Median : 0.0000	
Mean : 0.007375	Mean : 0.01555	Mean : 0.9546	
3rd Qu.: 0.000000	3rd Qu.: 0.00000	3rd Qu.: 0.0000	
Max. :13.000000	Max. :35.00000	Max. :200.0000	
cropcd29	cropcd31	cropcd32	
Min. : 0.0000	Min. :0.000000	Min. : 0.0000	
1st Qu.: 0.0000	1st Qu.:0.000000	1st Qu.: 0.0000	
Median : 0.0000	Median :0.000000	Median : 0.0000	
Mean : 0.2605	Mean :0.001073	Mean : 0.4868	
3rd Qu.: 0.0000	3rd Qu.:0.000000	3rd Qu.: 0.0000	
Max. :343.7500	Max. :4.000000	Max. :700.0000	
cropcd33	cropcd34	cropcd35	
Min. : 0.00000	Min. : 0.00000	Min. : 0.00000	
1st Qu.: 0.00000	1st Qu.: 0.00000	1st Qu.: 0.00000	
Median : 0.00000	Median : 0.00000	Median : 0.00000	
Mean : 0.03252	Mean : 0.06704	Mean : 0.04545	
3rd Qu.: 0.00000	3rd Qu.: 0.00000	3rd Qu.: 0.00000	
Max. :20.00000	Max. :250.00000	Max. :50.00000	

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```
# spread household harvested root count and count harvested root type
hh_root_info <- sec8c2 %>%
  select(clust, nh, rootcd, s8cq21a) %>%
  replace(., . > 1e+100, 0) %>%
  group_by(clust, nh, rootcd) %>%
  # sum up quantity of same root type
  summarise(s8cq21ac = sum(s8cq21a)) %>%
  group_by(clust, nh) %>%
  mutate(rootTypeCount = n()) %>%
  spread(key = rootcd,
        value = s8cq21ac,
        fill = 0,
        sep = "")
summary(hh_root_info)
```

clust	nh	rootTypeCount	rootcd0
Min. :4002	Min. : 1.00	Min. : 1.000	Min. :0
1st Qu.:4395	1st Qu.: 6.00	1st Qu.: 3.000	1st Qu.:0
Median :4572	Median :11.00	Median : 6.000	Median :0
Mean :4562	Mean :10.86	Mean : 6.511	Mean :0
3rd Qu.:4819	3rd Qu.:16.00	3rd Qu.: 9.000	3rd Qu.:0
Max. :4999	Max. :25.00	Max. :18.000	Max. :0

rootcd5	rootcd6	rootcd7	rootcd8
Min. : 0.000	Min. : 0.0	Min. : 0.0000	Min. : 0.000
1st Qu.: 0.000	1st Qu.: 0.0	1st Qu.: 0.0000	1st Qu.: 0.000
Median : 0.000	Median : 0.0	Median : 0.0000	Median : 0.000
Mean : 1.747	Mean : 2.8	Mean : 0.3084	Mean : 3.587
3rd Qu.: 0.000	3rd Qu.: 2.0	3rd Qu.: 0.0000	3rd Qu.: 0.000
Max. :225.000	Max. :143.0	Max. :80.0000	Max. :550.000

rootcd9	rootcd11	rootcd14	rootcd16
Min. : 0.0000	Min. : 0.0000	Min. :0	Min. : 0.0000
1st Qu.: 0.0000	1st Qu.: 0.0000	1st Qu.:0	1st Qu.: 0.0000
Median : 0.0000	Median : 0.0000	Median :0	Median : 0.0000
Mean : 0.1723	Mean : 0.6951	Mean :0	Mean : 0.4183
3rd Qu.: 0.0000	3rd Qu.: 0.0000	3rd Qu.:0	3rd Qu.: 0.0000
Max. :210.0000	Max. :700.0000	Max. :0	Max. :280.0000

rootcd18	rootcd19	rootcd20	rootcd21
Min. : 0.000	Min. : 0.000	Min. : 0.0000	Min. : 0.0000
1st Qu.: 0.000	1st Qu.: 0.000	1st Qu.: 0.0000	1st Qu.: 0.0000
Median : 1.000	Median : 0.000	Median : 0.0000	Median : 0.0000
Mean : 2.565	Mean : 3.729	Mean : 0.4482	Mean : 0.2046
3rd Qu.: 3.000	3rd Qu.: 0.000	3rd Qu.: 0.2500	3rd Qu.: 0.0000
Max. :83.000	Max. :800.000	Max. :30.0000	Max. :100.0000

rootcd22	rootcd25	rootcd26	rootcd27
Min. : 0.00000	Min. : 0.0000	Min. : 0.000	Min. : 0.00000

1st Qu.: 0.00000	1st Qu.: 0.0000	1st Qu.: 0.000	1st Qu.: 0.0000
Median : 0.00000	Median : 0.0000	Median : 0.000	Median : 0.0000
Mean : 0.03964	Mean : 0.7617	Mean : 1.626	Mean : 0.6471
3rd Qu.: 0.00000	3rd Qu.: 0.0000	3rd Qu.: 0.000	3rd Qu.: 0.0000
Max. :100.00000	Max. :120.0000	Max. :375.000	Max. :80.0000
rootcd29	rootcd30	rootcd31	rootcd33
Min. : 0.000	Min. : 0.000	Min. : 0.0000	Min. : 0.0000
1st Qu.: 0.000	1st Qu.: 0.000	1st Qu.: 0.0000	1st Qu.: 0.0000
Median : 0.000	Median : 0.000	Median : 0.0000	Median : 0.0000
Mean : 1.897	Mean : 2.534	Mean : 0.4419	Mean : 0.1937
3rd Qu.: 2.000	3rd Qu.: 3.000	3rd Qu.: 0.0000	3rd Qu.: 0.0000
Max. :108.000	Max. :120.000	Max. :21.0000	Max. :63.0000
rootcd34	rootcd35	rootcd36	
Min. : 0.000	Min. : 0.0000	Min. : 0.0000	
1st Qu.: 0.000	1st Qu.: 0.0000	1st Qu.: 0.0000	
Median : 0.000	Median : 0.0000	Median : 0.0000	
Mean : 3.025	Mean : 0.5926	Mean : 0.9104	
3rd Qu.: 0.000	3rd Qu.: 0.0000	3rd Qu.: 0.0000	
Max. :550.000	Max. :150.0000	Max. :50.0000	

Do a full join to combine all 4 trunk of data together. Since not every household have entries in all 4 type of agricultural characteristics, we assume the missing value simply means they just don't have that type and replace it with 0.

[Hide](#)

```
hh_agri_info <- hh_livestock_info %>%
  full_join(hh_equip_info, by=c("clust", "nh")) %>%
  full_join(hh_crop_info, by=c("clust", "nh")) %>%
  full_join(hh_root_info, by=c("clust", "nh")) %>%
  replace(., is.na(.), 0)
```

## Tidy community information

We chose market and transport information from the community data and convert them to logical value.

[Hide](#)

```
hh_comm_info <- cs2 %>%
  select(region, district, eanum, s2q19, s2q23) %>%
  mutate(market = ifelse(s2q19 == 1, TRUE, FALSE),
         transport = ifelse(s2q23 == 1, TRUE, FALSE)) %>%
  select(-s2q19, -s2q23)
```

## Combine all information

Combine household basic info, household member info (using head as representative), community info and household agricultural characteristics info together by inner join, because if a household does not have all 4 info, we don't want to use its data in our model fitting. From the combined data, we also created two subset for rural and urban areas. We are interested to see if the two areas have different influencing factors on profit.

[Hide](#)

```
hh_all_info <- hh_basic_info %>%
  inner_join(hh_head_info, by=c("clust", "nh")) %>%
  inner_join(hh_comm_info, by=c("region", "district", "eanum")) %>%
  inner_join(hh_agri_info, by=c("clust", "nh"))
hh_profit <- hh_profit_info %>%
  inner_join(hh_all_info, by=c("clust", "nh")) %>%
  select(-region, -district, -eanum, -clust, -nh)
# filter out rural and urban data from hh_profit
hh_profit_rural <- hh_profit %>%
  filter(loc2 == "Rural")
# remove factor type column with only one value
hh_profit_rural <- Filter(function(x) !isSingleValueFactorColumn(x), hh_profit_rural)
hh_profit_urban <- hh_profit %>%
  filter(loc2 == "Urban")
# remove factor type column with only one value
hh_profit_urban <- Filter(function(x) !isSingleValueFactorColumn(x), hh_profit_urban)
```

## Fit model and regression diagnostics

Now we fit linear regression models with different variables and do regression diagnostics.

### Unrestricted model

First we fit an unrestricted model. Looking at its summary, there are some significant variables. Looking at null hypothesis test result, there are correlated variables in this model. So we should fit a restricted model.

Following the model summary is the histogram of standardized residuals and plot of residuals vs fitted values.

[Hide](#)

```
hh_profit_model_ur <- lm(profit ~ .,
                        data = hh_profit)
checkCorrVarAndTestHnull(hh_profit_model_ur)
```

```
===== model summary =====
```

```
Call:
lm(formula = profit ~ ., data = hh_profit)
```

```
Residuals:
```

Min	1Q	Median	3Q	Max
-4146784	-238140	-117148	47055	20645824

Coefficients: (6 not defined because of singularities)

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	2.397e+05	2.756e+05	0.870	0.384553
reslanAkan	-3.865e+04	6.138e+04	-0.630	0.528985
reslanEwe	-1.034e+05	8.011e+04	-1.290	0.197057
reslanGaAdangbe	2.459e+05	1.149e+05	2.139	0.032504 *
reslanDagbani	-1.891e+05	1.084e+05	-1.745	0.081089 .
reslanHausa	-1.629e+05	1.280e+05	-1.273	0.203271
reslanOther	-2.358e+05	8.377e+04	-2.815	0.004908 **
reslanUnknown	-1.120e+05	7.880e+04	-1.421	0.155413
ezForest	2.186e+05	2.824e+05	0.774	0.438984
ezSavannah	1.021e+05	5.465e+04	1.867	0.061931 .
loc2Rural	7.353e+04	2.560e+05	0.287	0.773985
loc5RuralCoastal	NA	NA	NA	NA
loc5RuralForest	-1.728e+05	2.799e+05	-0.617	0.537106
loc5RuralSavannah	NA	NA	NA	NA
loc3Rural	NA	NA	NA	NA
femaleTRUE	1.523e+04	3.363e+04	0.453	0.650717
age	3.905e+03	1.600e+03	2.440	0.014744 *
avgAge	-4.050e+02	3.579e+03	-0.113	0.909914
maxAge	-3.166e+03	1.842e+03	-1.718	0.085804 .
minAge	-2.154e+03	2.423e+03	-0.889	0.374255
educBasicEducation	-3.320e+04	3.289e+04	-1.009	0.312885
educSecondaryEducation	7.350e+04	8.948e+04	0.821	0.411469
educTertiaryEducation	-8.837e+04	7.737e+04	-1.142	0.253494
educOther	-1.876e+05	1.750e+05	-1.072	0.283873
marketTRUE	-9.510e+04	4.119e+04	-2.309	0.021017 *
transportTRUE	-2.059e+04	3.026e+04	-0.681	0.496191
livstcdTypeCount	-6.877e+03	1.269e+04	-0.542	0.587965
livstcd1	-1.155e+04	3.835e+04	-0.301	0.763335
livstcd2	-2.666e+03	5.866e+03	-0.454	0.649505
livstcd3	-2.455e-01	5.336e-01	-0.460	0.645512
livstcd4	-1.318e-01	2.959e-01	-0.445	0.656044
livstcd5	4.406e+04	6.197e+03	7.110	1.40e-12 ***
livstcd6	1.450e+05	4.777e+04	3.037	0.002411 **
livstcd7	-1.248e+01	6.172e+00	-2.022	0.043273 *
livstcd8	-3.450e+03	2.906e+03	-1.187	0.235234
livstcd9	1.871e+04	1.298e+04	1.442	0.149521
livstcd10	-8.227e+03	3.768e+03	-2.184	0.029059 *
livstcd11	-1.127e+05	2.187e+05	-0.515	0.606245
livstcd12	-3.972e+02	1.658e+04	-0.024	0.980891
equipTypeCount	-6.038e+04	2.862e+04	-2.109	0.034989 *
eqcdwn21	1.021e+05	3.091e+05	0.330	0.741238
eqcdwn22	-6.542e+04	1.325e+05	-0.494	0.621535

eqcdownd31	-5.400e+04	2.466e+05	-0.219	0.826694	
eqcdownd51	-7.935e+02	4.891e+03	-0.162	0.871115	
eqcdownd61	-2.109e+06	2.767e+05	-7.623	3.18e-14	***
eqcdownd62	3.385e+03	6.850e+03	0.494	0.621172	
eqcdownd63	-7.457e+02	5.806e+03	-0.128	0.897819	
eqcdownd64	7.207e+04	9.454e+04	0.762	0.445890	
eqcdownd65	-3.892e+02	9.763e+02	-0.399	0.690164	
cropTypeCount	-1.186e+04	1.683e+04	-0.705	0.480902	
croppcd0	NA	NA	NA	NA	
croppcd1	-1.932e+02	4.620e+02	-0.418	0.675833	
croppcd2	-2.091e+03	3.470e+03	-0.603	0.546816	
croppcd3	-1.497e+04	2.621e+05	-0.057	0.954461	
croppcd4	7.705e+01	1.996e+02	0.386	0.699475	
croppcd5	4.081e+05	2.472e+05	1.651	0.098897	.
croppcd6	-1.580e+05	2.705e+05	-0.584	0.559116	
croppcd8	-1.482e+03	4.839e+02	-3.062	0.002216	**
croppcd9	-8.346e+03	5.246e+04	-0.159	0.873590	
croppcd10	1.783e+01	6.821e+02	0.026	0.979143	
croppcd11	-8.952e+04	1.756e+04	-5.099	3.60e-07	***
croppcd12	3.008e+03	1.290e+04	0.233	0.815664	
croppcd13	-1.600e+04	4.106e+04	-0.390	0.696752	
croppcd14	-3.387e+02	1.705e+03	-0.199	0.842497	
croppcd15	1.995e+03	4.547e+03	0.439	0.660795	
croppcd16	1.486e+04	1.571e+04	0.946	0.344249	
croppcd17	2.534e+02	4.916e+02	0.516	0.606229	
croppcd18	5.804e+02	1.810e+03	0.321	0.748524	
croppcd19	-2.968e+02	2.927e+03	-0.101	0.919234	
croppcd20	1.824e+04	3.999e+04	0.456	0.648278	
croppcd21	3.078e+04	7.429e+04	0.414	0.678686	
croppcd22	-4.025e+01	1.668e+02	-0.241	0.809310	
croppcd23	2.136e+03	2.485e+03	0.859	0.390242	
croppcd24	-4.521e+03	6.434e+03	-0.703	0.482287	
croppcd25	9.963e+04	8.311e+03	11.988	< 2e-16	***
croppcd26	1.108e+04	5.549e+04	0.200	0.841796	
croppcd27	-5.238e+03	2.079e+04	-0.252	0.801116	
croppcd28	1.381e+03	2.157e+03	0.640	0.521981	
croppcd29	-6.901e+03	3.997e+03	-1.727	0.084298	.
croppcd31	1.468e+05	1.959e+05	0.749	0.453815	
croppcd32	-1.836e+02	7.815e+02	-0.235	0.814233	
croppcd33	-1.174e+04	2.763e+04	-0.425	0.670842	
croppcd34	1.720e+03	3.145e+03	0.547	0.584485	
croppcd35	-3.955e+02	1.244e+04	-0.032	0.974643	
rootTypeCount	2.498e+03	4.209e+03	0.593	0.552936	
rootcd0	NA	NA	NA	NA	
rootcd5	9.402e+02	1.310e+03	0.718	0.473066	
rootcd6	2.582e+03	1.718e+03	1.503	0.133013	
rootcd7	2.045e+04	7.834e+03	2.610	0.009097	**



rootcd8	5.665e+02	6.516e+02	0.869	0.384716
rootcd9	-1.315e+03	3.763e+03	-0.350	0.726726
rootcd11	-9.676e+01	1.366e+03	-0.071	0.943541
rootcd14	NA	NA	NA	NA
rootcd16	1.392e+03	2.704e+03	0.515	0.606768
rootcd18	2.340e+04	3.453e+03	6.779	1.41e-11 ***
rootcd19	-3.822e+01	4.430e+02	-0.086	0.931262
rootcd20	4.535e+04	1.056e+04	4.294	1.81e-05 ***
rootcd21	-1.873e+02	4.863e+03	-0.039	0.969273
rootcd22	-1.023e+03	7.899e+03	-0.129	0.896983
rootcd25	-1.450e+03	4.088e+03	-0.355	0.722916
rootcd26	-1.152e+03	1.450e+03	-0.794	0.427080
rootcd27	-1.669e+04	3.276e+03	-5.095	3.68e-07 ***
rootcd29	6.320e+02	3.090e+03	0.205	0.837959
rootcd30	1.534e+02	2.828e+03	0.054	0.956734
rootcd31	1.078e+03	1.041e+04	0.103	0.917595
rootcd33	1.408e+04	8.392e+03	1.677	0.093536 .
rootcd34	1.647e+02	8.711e+02	0.189	0.850032
rootcd35	-3.864e+03	2.366e+03	-1.633	0.102498
rootcd36	1.993e+04	5.531e+03	3.603	0.000319 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 782200 on 3510 degrees of freedom

Multiple R-squared: 0.137, Adjusted R-squared: 0.1119

F-statistic: 5.461 on 102 and 3510 DF, p-value: < 2.2e-16

===== model alias =====

Model :

```
profit ~ reslan + ez + loc2 + loc5 + loc3 + female + age + avgAge +
  maxAge + minAge + educ + market + transport + livstcdTypeCount +
  livstcd1 + livstcd2 + livstcd3 + livstcd4 + livstcd5 + livstcd6 +
  livstcd7 + livstcd8 + livstcd9 + livstcd10 + livstcd11 +
  livstcd12 + equipTypeCount + eqcdownd21 + eqcdownd22 + eqcdownd31 +
  eqcdownd51 + eqcdownd61 + eqcdownd62 + eqcdownd63 + eqcdownd64 +
  eqcdownd65 + cropTypeCount + cropcd0 + cropcd1 + cropcd2 +
  cropcd3 + cropcd4 + cropcd5 + cropcd6 + cropcd8 + cropcd9 +
  cropcd10 + cropcd11 + cropcd12 + cropcd13 + cropcd14 + cropcd15 +
  cropcd16 + cropcd17 + cropcd18 + cropcd19 + cropcd20 + cropcd21 +
  cropcd22 + cropcd23 + cropcd24 + cropcd25 + cropcd26 + cropcd27 +
  cropcd28 + cropcd29 + cropcd31 + cropcd32 + cropcd33 + cropcd34 +
  cropcd35 + rootTypeCount + rootcd0 + rootcd5 + rootcd6 +
  rootcd7 + rootcd8 + rootcd9 + rootcd11 + rootcd14 + rootcd16 +
  rootcd18 + rootcd19 + rootcd20 + rootcd21 + rootcd22 + rootcd25 +
  rootcd26 + rootcd27 + rootcd29 + rootcd30 + rootcd31 + rootcd33 +
  rootcd34 + rootcd35 + rootcd36
```

Complete :

	(Intercept)	reslanAkan	reslanEwe	reslanGaAdangbe		
loc5RuralCoastal	1	0	0	0		
loc5RuralSavannah	-1	0	0	0		
loc3Rural	0	0	0	0		
croppcd0	0	0	0	0		
rootcd0	0	0	0	0		
rootcd14	0	0	0	0		
	reslanDagbani	reslanHausa	reslanOther	reslanUnknown		
loc5RuralCoastal	0	0	0	0		
loc5RuralSavannah	0	0	0	0		
loc3Rural	0	0	0	0		
croppcd0	0	0	0	0		
rootcd0	0	0	0	0		
rootcd14	0	0	0	0		
	ezForest	ezSavannah	loc2Rural	loc5RuralForest	femaleTRUE	
loc5RuralCoastal	-1	-1	0	0	0	
loc5RuralSavannah	1	1	1	-1	0	
loc3Rural	0	0	1	0	0	
croppcd0	0	0	0	0	0	
rootcd0	0	0	0	0	0	
rootcd14	0	0	0	0	0	
	age	avgAge	maxAge	minAge	educBasicEducation	
loc5RuralCoastal	0	0	0	0	0	
loc5RuralSavannah	0	0	0	0	0	
loc3Rural	0	0	0	0	0	
croppcd0	0	0	0	0	0	
rootcd0	0	0	0	0	0	
rootcd14	0	0	0	0	0	
	educSecondaryEducation	educTertiaryEducation	educOther			
loc5RuralCoastal	0	0	0			
loc5RuralSavannah	0	0	0			
loc3Rural	0	0	0			
croppcd0	0	0	0			
rootcd0	0	0	0			
rootcd14	0	0	0			
	marketTRUE	transportTRUE	livstcdTypeCount	livstcd1		
loc5RuralCoastal	0	0	0	0		
loc5RuralSavannah	0	0	0	0		
loc3Rural	0	0	0	0		
croppcd0	0	0	0	0		
rootcd0	0	0	0	0		
rootcd14	0	0	0	0		
	livstcd2	livstcd3	livstcd4	livstcd5	livstcd6	livstcd7
loc5RuralCoastal	0	0	0	0	0	0
loc5RuralSavannah	0	0	0	0	0	0
loc3Rural	0	0	0	0	0	0

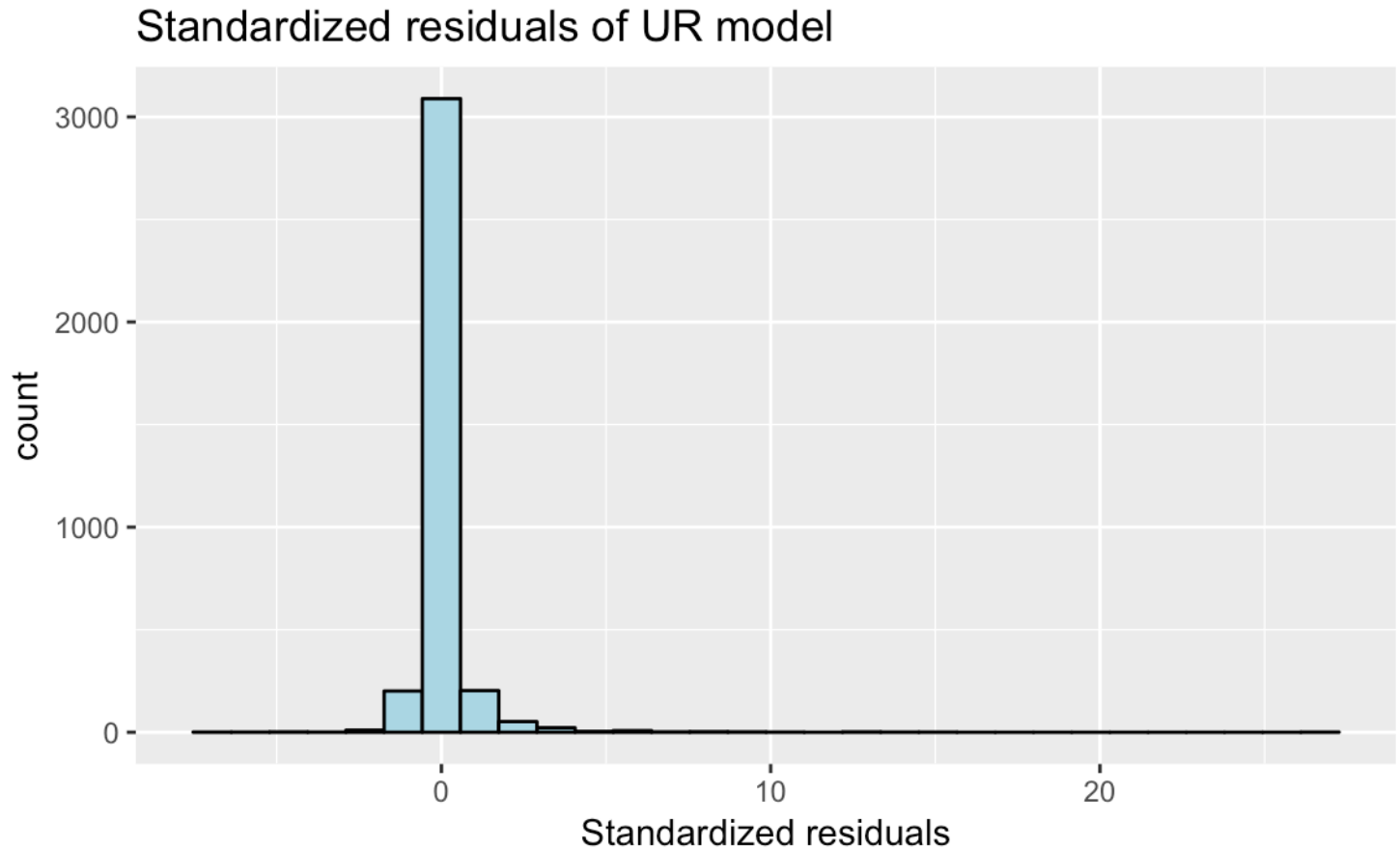
crogcd0	0	0	0	0	0	0
rootcd0	0	0	0	0	0	0
rootcd14	0	0	0	0	0	0
	livstcd8	livstcd9	livstcd10	livstcd11	livstcd12	
loc5RuralCoastal	0	0	0	0	0	
loc5RuralSavannah	0	0	0	0	0	
loc3Rural	0	0	0	0	0	
crogcd0	0	0	0	0	0	
rootcd0	0	0	0	0	0	
rootcd14	0	0	0	0	0	
	equipTypeCount	eqcdwn21	eqcdwn22	eqcdwn31	eqcdwn51	
loc5RuralCoastal	0	0	0	0	0	
loc5RuralSavannah	0	0	0	0	0	
loc3Rural	0	0	0	0	0	
crogcd0	0	0	0	0	0	
rootcd0	0	0	0	0	0	
rootcd14	0	0	0	0	0	
	eqcdwn61	eqcdwn62	eqcdwn63	eqcdwn64	eqcdwn65	
loc5RuralCoastal	0	0	0	0	0	
loc5RuralSavannah	0	0	0	0	0	
loc3Rural	0	0	0	0	0	
crogcd0	0	0	0	0	0	
rootcd0	0	0	0	0	0	
rootcd14	0	0	0	0	0	
	cropTypeCount	crogcd1	crogcd2	crogcd3	crogcd4	crogcd5
loc5RuralCoastal	0	0	0	0	0	0
loc5RuralSavannah	0	0	0	0	0	0
loc3Rural	0	0	0	0	0	0
crogcd0	0	0	0	0	0	0
rootcd0	0	0	0	0	0	0
rootcd14	0	0	0	0	0	0
	crogcd6	crogcd8	crogcd9	crogcd10	crogcd11	crogcd12
loc5RuralCoastal	0	0	0	0	0	0
loc5RuralSavannah	0	0	0	0	0	0
loc3Rural	0	0	0	0	0	0
crogcd0	0	0	0	0	0	0
rootcd0	0	0	0	0	0	0
rootcd14	0	0	0	0	0	0
	crogcd13	crogcd14	crogcd15	crogcd16	crogcd17	crogcd18
loc5RuralCoastal	0	0	0	0	0	0
loc5RuralSavannah	0	0	0	0	0	0
loc3Rural	0	0	0	0	0	0
crogcd0	0	0	0	0	0	0
rootcd0	0	0	0	0	0	0
rootcd14	0	0	0	0	0	0
	crogcd19	crogcd20	crogcd21	crogcd22	crogcd23	crogcd24
loc5RuralCoastal	0	0	0	0	0	0

loc5RuralSavannah	0	0	0	0	0	0
loc3Rural	0	0	0	0	0	0
cropcd0	0	0	0	0	0	0
rootcd0	0	0	0	0	0	0
rootcd14	0	0	0	0	0	0
	cropcd25	cropcd26	cropcd27	cropcd28	cropcd29	cropcd31
loc5RuralCoastal	0	0	0	0	0	0
loc5RuralSavannah	0	0	0	0	0	0
loc3Rural	0	0	0	0	0	0
cropcd0	0	0	0	0	0	0
rootcd0	0	0	0	0	0	0
rootcd14	0	0	0	0	0	0
	cropcd32	cropcd33	cropcd34	cropcd35	rootTypeCount	rootcd5
loc5RuralCoastal	0	0	0	0	0	0
loc5RuralSavannah	0	0	0	0	0	0
loc3Rural	0	0	0	0	0	0
cropcd0	0	0	0	0	0	0
rootcd0	0	0	0	0	0	0
rootcd14	0	0	0	0	0	0
	rootcd6	rootcd7	rootcd8	rootcd9	rootcd11	rootcd16
loc5RuralCoastal	0	0	0	0	0	0
loc5RuralSavannah	0	0	0	0	0	0
loc3Rural	0	0	0	0	0	0
cropcd0	0	0	0	0	0	0
rootcd0	0	0	0	0	0	0
rootcd14	0	0	0	0	0	0
	rootcd18	rootcd19	rootcd20	rootcd21	rootcd22	rootcd25
loc5RuralCoastal	0	0	0	0	0	0
loc5RuralSavannah	0	0	0	0	0	0
loc3Rural	0	0	0	0	0	0
cropcd0	0	0	0	0	0	0
rootcd0	0	0	0	0	0	0
rootcd14	0	0	0	0	0	0
	rootcd26	rootcd27	rootcd29	rootcd30	rootcd31	rootcd33
loc5RuralCoastal	0	0	0	0	0	0
loc5RuralSavannah	0	0	0	0	0	0
loc3Rural	0	0	0	0	0	0
cropcd0	0	0	0	0	0	0
rootcd0	0	0	0	0	0	0
rootcd14	0	0	0	0	0	0
	rootcd34	rootcd35	rootcd36			
loc5RuralCoastal	0	0	0			
loc5RuralSavannah	0	0	0			
loc3Rural	0	0	0			
cropcd0	0	0	0			
rootcd0	0	0	0			
rootcd14	0	0	0			

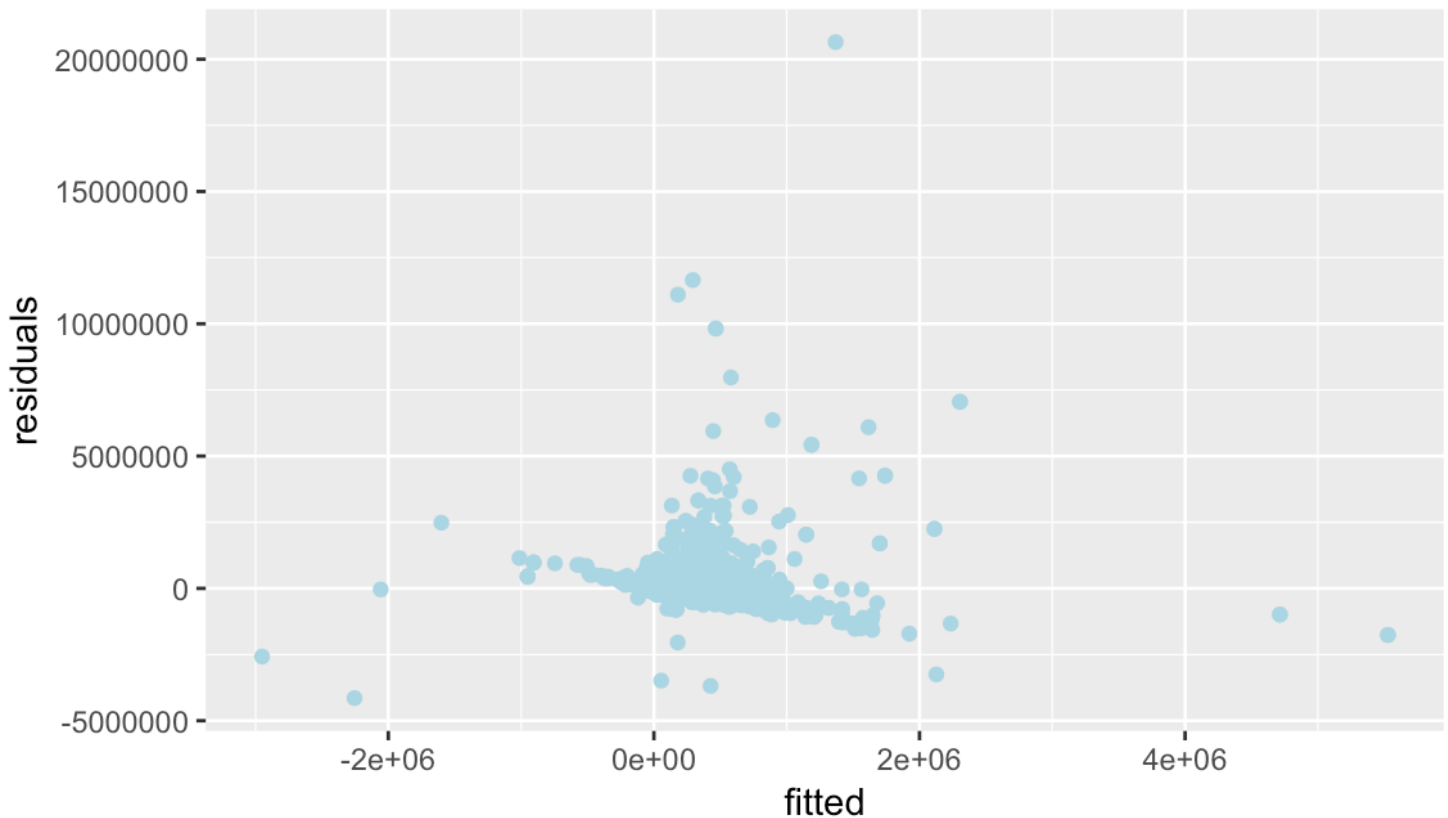
There are correlated variables. See above model alias.

Hide

```
suppressWarnings(plotResiduals(hh_profit, hh_profit_model_ur, "lightblue", "UR"))
```



## Homoskedasticity of UR model



## Restricted model with significant variables (R1)

We pick the significant variables from the above unrestricted model and fit a restricted model.

[Hide](#)

```
hh_profit_model_r1 <- lm(profit ~ reslan + ez + age + market + livstcd5 + livstcd6 +
livstcd7 + livstcd10 +
                        equipTypeCount + eqcdwn61 + cropcd5 + cropcd8 + cropcd11
+ cropcd25 + cropcd29 +
                        rootcd7 + rootcd18 + rootcd20 + rootcd27 + rootcd33 + root
cd36,
                        data = hh_profit)
checkCorrVarAndTestHnull(hh_profit_model_r1)
```

===== model summary =====

Call:

```
lm(formula = profit ~ reslan + ez + age + market + livstcd5 +
  livstcd6 + livstcd7 + livstcd10 + equipTypeCount + eqcdwn61 +
  cropcd5 + cropcd8 + cropcd11 + cropcd25 + cropcd29 + rootcd7 +
  rootcd18 + rootcd20 + rootcd27 + rootcd33 + rootcd36, data = hh_profit)
```

## Residuals:

Min	1Q	Median	3Q	Max
-4230445	-234684	-129489	33517	20721029

## Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	2.166e+05	7.787e+04	2.781	0.00545	**
reslanAkan	4.694e+03	5.728e+04	0.082	0.93469	
reslanEwe	-4.304e+04	7.487e+04	-0.575	0.56544	
reslanGaAdangbe	2.707e+05	1.092e+05	2.480	0.01319	*
reslanDagbani	-1.701e+05	1.048e+05	-1.624	0.10448	
reslanHausa	-1.750e+05	1.245e+05	-1.405	0.16015	
reslanOther	-2.324e+05	8.018e+04	-2.899	0.00377	**
reslanUnknown	-6.183e+04	7.529e+04	-0.821	0.41159	
ezForest	4.855e+04	3.988e+04	1.218	0.22347	
ezSavannah	6.429e+04	4.706e+04	1.366	0.17194	
age	1.676e+02	8.861e+02	0.189	0.84997	
marketTRUE	-9.375e+04	3.902e+04	-2.403	0.01632	*
livstcd5	4.293e+04	5.847e+03	7.341	2.60e-13	***
livstcd6	1.508e+05	4.643e+04	3.247	0.00118	**
livstcd7	-9.917e+00	5.336e+00	-1.859	0.06316	.
livstcd10	-8.339e+03	3.712e+03	-2.246	0.02476	*
equipTypeCount	-5.785e+04	2.463e+04	-2.349	0.01890	*
eqcdwn61	-2.123e+06	2.644e+05	-8.031	1.30e-15	***
cropcd5	3.344e+05	2.348e+05	1.424	0.15441	
cropcd8	-1.430e+03	4.700e+02	-3.043	0.00236	**
cropcd11	-9.308e+04	1.711e+04	-5.441	5.65e-08	***
cropcd25	1.002e+05	8.164e+03	12.278	< 2e-16	***
cropcd29	-5.640e+03	3.735e+03	-1.510	0.13115	
rootcd7	2.022e+04	7.682e+03	2.632	0.00852	**
rootcd18	2.408e+04	3.248e+03	7.413	1.53e-13	***
rootcd20	4.457e+04	1.023e+04	4.356	1.36e-05	***
rootcd27	-1.617e+04	3.195e+03	-5.062	4.36e-07	***
rootcd33	1.436e+04	7.675e+03	1.872	0.06133	.
rootcd36	2.221e+04	5.141e+03	4.320	1.60e-05	***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 779000 on 3584 degrees of freedom

Multiple R-squared: 0.1259, Adjusted R-squared: 0.1191

F-statistic: 18.44 on 28 and 3584 DF, p-value: &lt; 2.2e-16

===== hypothesis test =====

Linear hypothesis test

Hypothesis:

reslanAkan = 0

```

reslanEwe = 0
reslanGaAdangbe = 0
reslanDagbani = 0
reslanHausa = 0
reslanOther = 0
reslanUnknown = 0
ezForest = 0
ezSavannah = 0
age = 0
marketTRUE = 0
livstcd5 = 0
livstcd6 = 0
livstcd7 = 0
livstcd10 = 0
equipTypeCount = 0
eqcdwn61 = 0
cropcd5 = 0
cropcd8 = 0
cropcd11 = 0
cropcd25 = 0
cropcd29 = 0
rootcd7 = 0
rootcd18 = 0
rootcd20 = 0
rootcd27 = 0
rootcd33 = 0
rootcd36 = 0

```

Model 1: restricted model

Model 2: profit ~ reslan + ez + age + market + livstcd5 + livstcd6 + livstcd7 +  
 livstcd10 + equipTypeCount + eqcdwn61 + cropcd5 + cropcd8 +  
 cropcd11 + cropcd25 + cropcd29 + rootcd7 + rootcd18 + rootcd20 +  
 rootcd27 + rootcd33 + rootcd36

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	3612	2.4883e+15				
2	3584	2.1750e+15	28	3.1327e+14	18.436	< 2.2e-16 ***

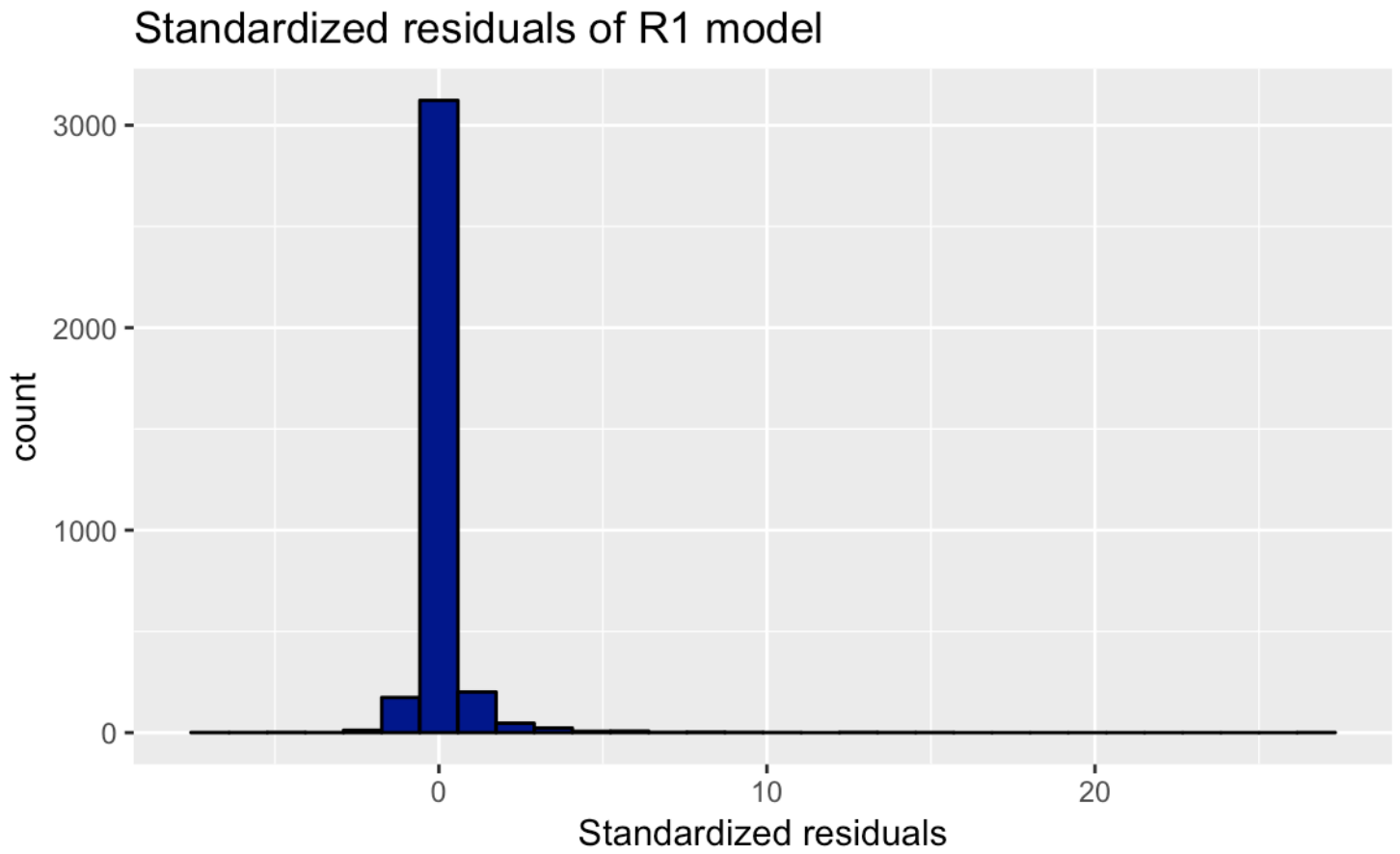
---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

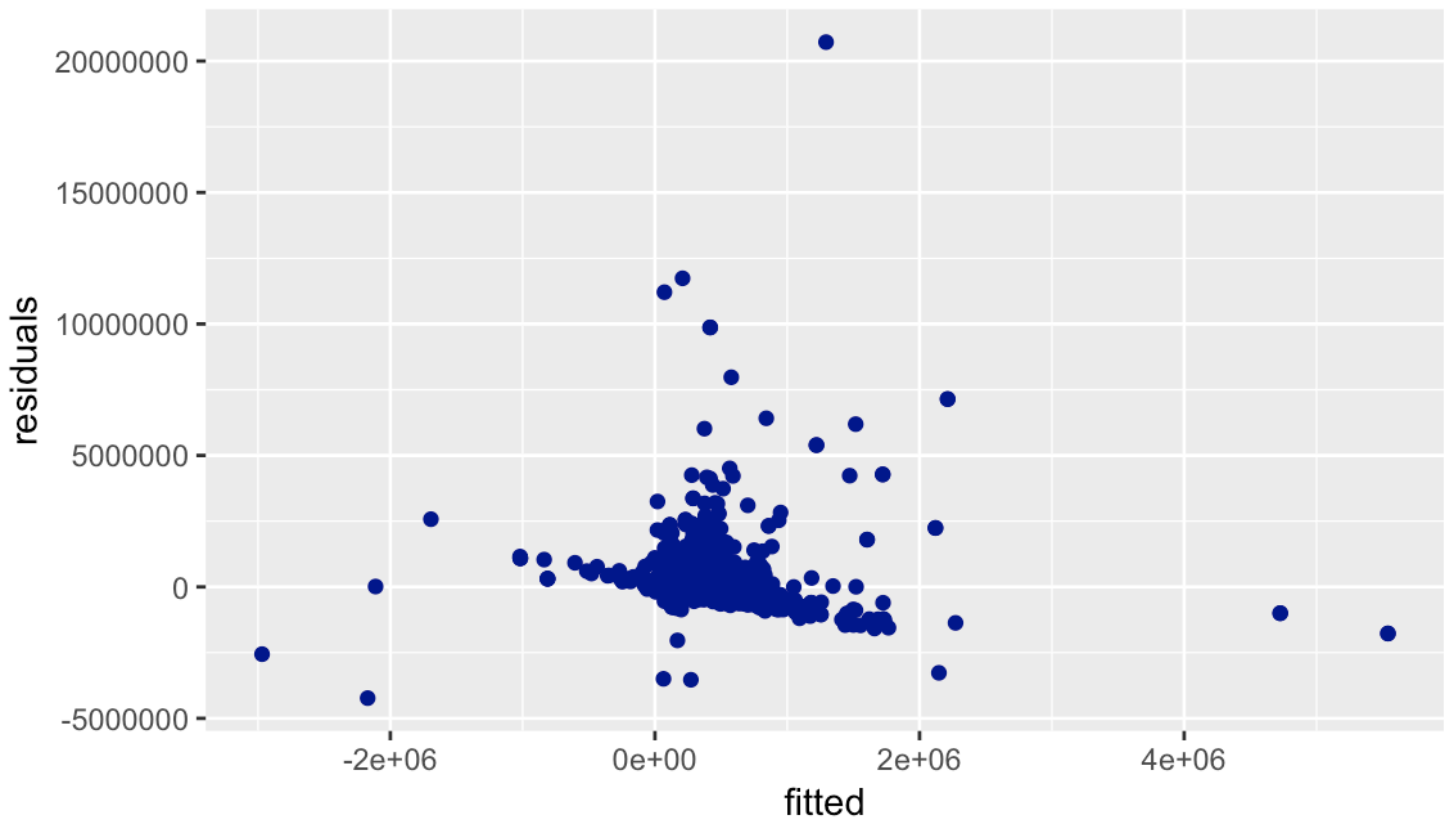
[Hide](#)

```
suppressWarnings(plotResiduals(hh_profit, hh_profit_model_r1, "darkblue", "R1"))
```





## Homoskedasticity of R1 model



## Restricted model with top features from agricultural characteristics information (TOP)

We are also interested to see how will the model perform if fitted only with top 15 features from agricultural characteristics information. These top features are the variables that have higher correlation with profit, no matter it's positive or negative correlation.

[Hide](#)

```
hh_profit_agri <- hh_profit[, -c(2:14)] # get agricultural characteristics variables
all_correlations <- findAbsoluteCorrelation(hh_profit_agri)
```

```
the standard deviation is zero
the standard deviation is zero
the standard deviation is zero
```

[Hide](#)

```
hh_profit_agri_topFeatures <- hh_profit[c("profit", all_correlations$colName[1:15])]
hh_profit_model_topfeatures <- lm(profit ~ ., data = hh_profit_agri_topFeatures)
checkCorrVarAndTestHnull(hh_profit_model_topfeatures)
```

```
===== model summary =====
```

Call:

```
lm(formula = profit ~ ., data = hh_profit_agri_topFeatures)
```

Residuals:

Min	1Q	Median	3Q	Max
-4356876	-223572	-143826	9718	20824127

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	226865.6	21550.4	10.527	< 2e-16 ***
cropcd25	84128.0	7257.6	11.592	< 2e-16 ***
rootcd18	23465.0	3380.7	6.941	4.60e-12 ***
rootcd20	44255.6	10409.3	4.252	2.18e-05 ***
rootTypeCount	2448.2	3722.7	0.658	0.51082
eqcdwn61	-1969761.3	270480.3	-7.282	4.01e-13 ***
rootcd36	17450.2	5337.4	3.269	0.00109 **
livstcd5	32012.9	5377.6	5.953	2.89e-09 ***
equipTypeCount	-59648.8	23358.8	-2.554	0.01070 *
rootcd6	2489.4	1662.2	1.498	0.13432
rootcd7	16346.7	7718.2	2.118	0.03425 *
livstcd10	-8578.5	3735.6	-2.296	0.02171 *
rootcd30	276.4	2479.5	0.111	0.91124
rootcd5	1477.8	1296.6	1.140	0.25448
rootcd8	430.4	600.4	0.717	0.47355
livstcd2	-6932.1	5149.1	-1.346	0.17830

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 788200 on 3597 degrees of freedom

Multiple R-squared: 0.102, Adjusted R-squared: 0.09828

F-statistic: 27.25 on 15 and 3597 DF, p-value: < 2.2e-16

===== hypothesis test =====

Linear hypothesis test

Hypothesis:

cropcd25 = 0

rootcd18 = 0

rootcd20 = 0

rootTypeCount = 0

eqcdwn61 = 0

rootcd36 = 0

livstcd5 = 0

equipTypeCount = 0

rootcd6 = 0

rootcd7 = 0

```

livstcd10 = 0
rootcd30 = 0
rootcd5 = 0
rootcd8 = 0
livstcd2 = 0

```

Model 1: restricted model

Model 2: profit ~ cropped25 + rootcd18 + rootcd20 + rootTypeCount + eqcdwn61 +  
 rootcd36 + livstcd5 + equipTypeCount + rootcd6 + rootcd7 +  
 livstcd10 + rootcd30 + rootcd5 + rootcd8 + livstcd2

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	3612	2.4883e+15				
2	3597	2.2344e+15	15	2.5387e+14	27.246	< 2.2e-16 ***

---  
 Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

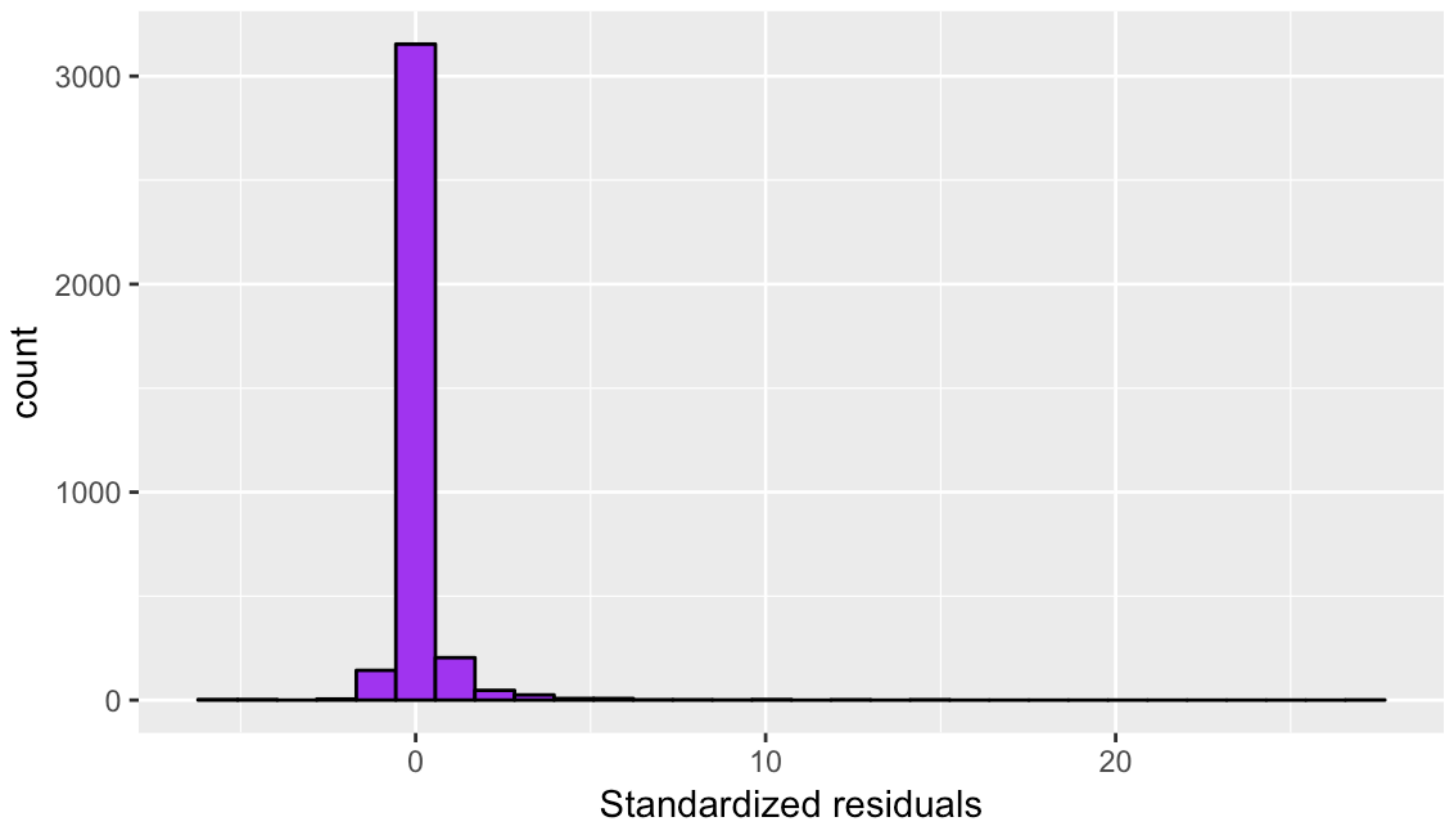
[Hide](#)

```

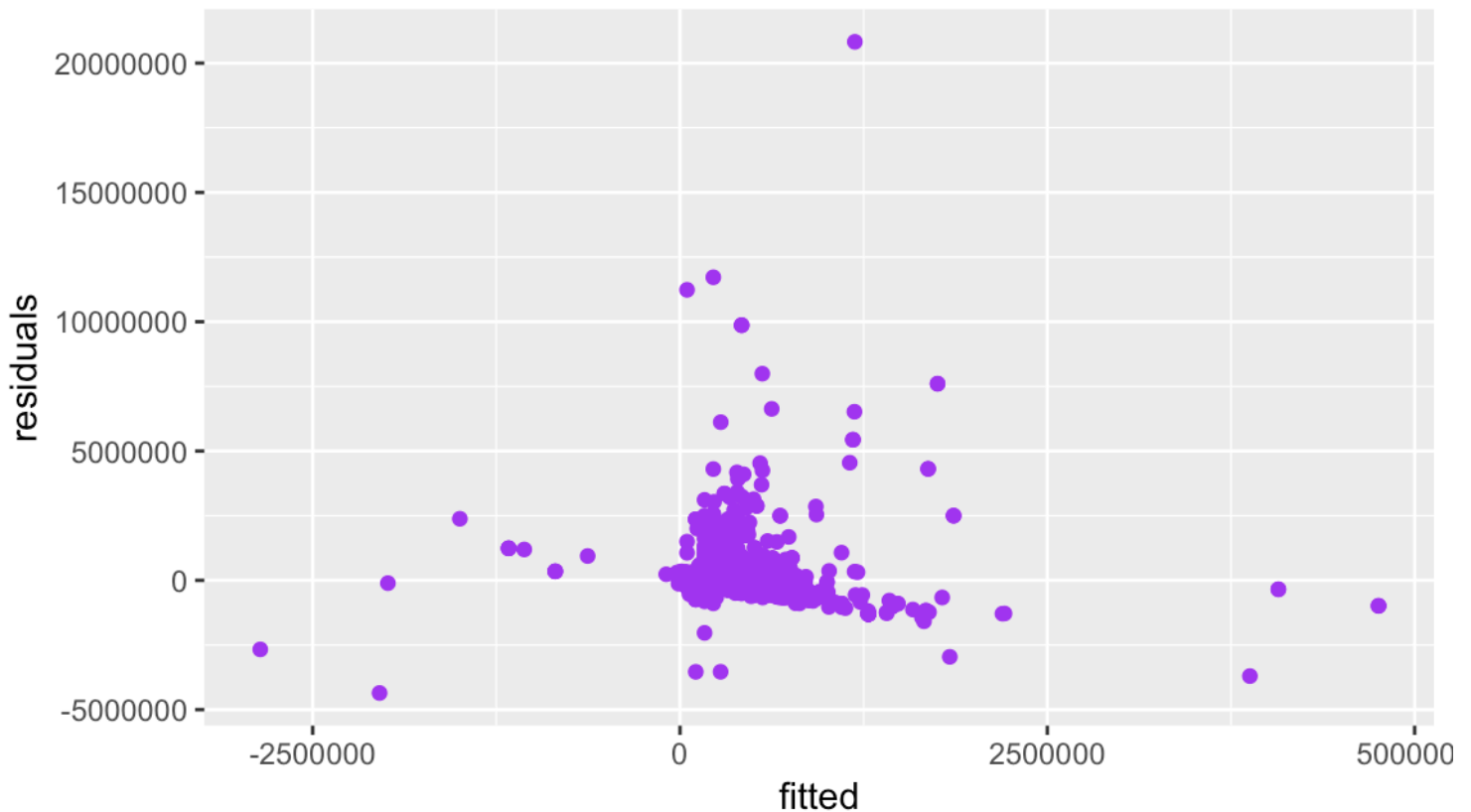
suppressWarnings(plotResiduals(hh_profit, hh_profit_model_topfeatures, "purple", "TOP
"))

```

## Standardized residuals of TOP model



## Homoskedasticity of TOP model



## Unrestricted model for rural area (RURAL)

Fit unrestricted model for rural area profit.

Hide

```
hh_profit_model_rural <- lm(profit ~ .,
                             data = hh_profit_rural)
checkCorrVarAndTestHnull(hh_profit_model_rural)
```

===== model summary =====

Call:

```
lm(formula = profit ~ ., data = hh_profit_rural)
```

Residuals:

Min	1Q	Median	3Q	Max
-4146964	-238546	-118235	46546	20643791

Coefficients: (5 not defined because of singularities)

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	3.165e+05	9.745e+04	3.248	0.001174 **
reslanAkan	-3.980e+04	6.191e+04	-0.643	0.520400

reslanEwe	-9.813e+04	8.080e+04	-1.214	0.224663	
reslanGaAdangbe	2.437e+05	1.157e+05	2.107	0.035192	*
reslanDagbani	-1.920e+05	1.091e+05	-1.761	0.078401	.
reslanHausa	-1.617e+05	1.288e+05	-1.256	0.209224	
reslanOther	-2.378e+05	8.432e+04	-2.820	0.004826	**
reslanUnknown	-1.122e+05	7.966e+04	-1.408	0.159184	
ezForest	4.293e+04	4.459e+04	0.963	0.335781	
ezSavannah	9.988e+04	5.515e+04	1.811	0.070223	.
loc5RuralForest	NA	NA	NA	NA	
loc5RuralSavannah	NA	NA	NA	NA	
femaleTRUE	1.203e+04	3.416e+04	0.352	0.724752	
age	3.770e+03	1.622e+03	2.324	0.020190	*
avgAge	-2.234e+02	3.628e+03	-0.062	0.950902	
maxAge	-3.088e+03	1.867e+03	-1.654	0.098229	.
minAge	-2.242e+03	2.456e+03	-0.913	0.361347	
educBasicEducation	-3.337e+04	3.331e+04	-1.002	0.316476	
educSecondaryEducation	7.066e+04	9.065e+04	0.780	0.435738	
educTertiaryEducation	-8.169e+04	7.886e+04	-1.036	0.300339	
educOther	-1.880e+05	1.760e+05	-1.068	0.285462	
marketTRUE	-1.099e+05	4.196e+04	-2.620	0.008844	**
transportTRUE	-1.863e+04	3.045e+04	-0.612	0.540606	
livstcdTypeCount	-7.444e+03	1.281e+04	-0.581	0.561133	
livstcd1	-1.234e+04	3.862e+04	-0.320	0.749338	
livstcd2	-2.516e+03	5.901e+03	-0.426	0.669861	
livstcd3	-2.526e-01	5.366e-01	-0.471	0.637801	
livstcd4	-1.312e-01	2.975e-01	-0.441	0.659209	
livstcd5	4.394e+04	6.286e+03	6.989	3.30e-12	***
livstcd6	1.448e+05	4.803e+04	3.014	0.002599	**
livstcd7	-1.244e+01	6.207e+00	-2.004	0.045190	*
livstcd8	-3.226e+03	2.936e+03	-1.099	0.271995	
livstcd9	1.853e+04	1.305e+04	1.420	0.155793	
livstcd10	-8.260e+03	3.788e+03	-2.180	0.029292	*
livstcd11	-1.117e+05	2.199e+05	-0.508	0.611625	
livstcd12	-3.790e+02	1.667e+04	-0.023	0.981864	
equipTypeCount	-6.493e+04	2.889e+04	-2.247	0.024675	*
eqcdwn21	1.085e+05	3.108e+05	0.349	0.726991	
eqcdwn22	-6.254e+04	1.333e+05	-0.469	0.638887	
eqcdwn31	-5.073e+04	2.480e+05	-0.205	0.837927	
eqcdwn51	-7.668e+02	4.917e+03	-0.156	0.876087	
eqcdwn61	-2.103e+06	2.783e+05	-7.557	5.27e-14	***
eqcdwn62	3.453e+03	6.887e+03	0.501	0.616111	
eqcdwn63	-6.506e+02	5.839e+03	-0.111	0.911284	
eqcdwn64	7.366e+04	9.505e+04	0.775	0.438424	
eqcdwn65	-3.703e+02	9.817e+02	-0.377	0.706041	
cropTypeCount	-1.197e+04	1.696e+04	-0.706	0.480214	
cropcd0	NA	NA	NA	NA	
cropcd1	-1.807e+02	4.647e+02	-0.389	0.697401	

cropcd2	-2.185e+03	3.490e+03	-0.626	0.531245	
cropcd3	-1.369e+04	2.636e+05	-0.052	0.958589	
cropcd4	8.770e+01	2.011e+02	0.436	0.662844	
cropcd5	4.061e+05	2.486e+05	1.634	0.102416	
cropcd6	-1.595e+05	2.720e+05	-0.586	0.557587	
cropcd8	-1.475e+03	4.866e+02	-3.032	0.002450	**
cropcd9	-8.029e+03	5.274e+04	-0.152	0.879010	
cropcd10	1.812e+02	7.170e+02	0.253	0.800458	
cropcd11	-8.957e+04	1.765e+04	-5.074	4.10e-07	***
cropcd12	3.217e+03	1.298e+04	0.248	0.804330	
cropcd13	-1.618e+04	4.128e+04	-0.392	0.695136	
cropcd14	-3.311e+02	1.714e+03	-0.193	0.846893	
cropcd15	1.923e+03	4.573e+03	0.420	0.674180	
cropcd16	1.499e+04	1.579e+04	0.950	0.342383	
cropcd17	2.698e+02	4.950e+02	0.545	0.585786	
cropcd18	6.021e+02	1.820e+03	0.331	0.740859	
cropcd19	-3.004e+02	2.943e+03	-0.102	0.918688	
cropcd20	1.923e+04	4.020e+04	0.478	0.632395	
cropcd21	3.164e+04	7.476e+04	0.423	0.672133	
cropcd22	-4.868e+01	1.678e+02	-0.290	0.771793	
cropcd23	2.084e+03	2.499e+03	0.834	0.404478	
cropcd24	-4.612e+03	6.471e+03	-0.713	0.476056	
cropcd25	9.962e+04	8.357e+03	11.921	< 2e-16	***
cropcd26	1.044e+04	5.579e+04	0.187	0.851625	
cropcd27	-4.896e+03	2.091e+04	-0.234	0.814870	
cropcd28	1.346e+03	2.169e+03	0.621	0.534910	
cropcd29	-6.881e+03	4.019e+03	-1.712	0.086926	.
cropcd31	1.466e+05	1.970e+05	0.744	0.456695	
cropcd32	-1.818e+02	7.857e+02	-0.231	0.816998	
cropcd33	-1.161e+04	2.779e+04	-0.418	0.675999	
cropcd34	1.733e+03	3.162e+03	0.548	0.583810	
cropcd35	-3.535e+02	1.251e+04	-0.028	0.977460	
rootTypeCount	3.166e+03	4.278e+03	0.740	0.459246	
rootcd0	NA	NA	NA	NA	
rootcd5	9.504e+02	1.318e+03	0.721	0.470781	
rootcd6	2.612e+03	1.729e+03	1.511	0.130914	
rootcd7	2.034e+04	7.901e+03	2.574	0.010096	*
rootcd8	5.491e+02	6.555e+02	0.838	0.402265	
rootcd9	-9.208e+03	1.021e+04	-0.902	0.367376	
rootcd11	-1.110e+02	1.374e+03	-0.081	0.935580	
rootcd14	NA	NA	NA	NA	
rootcd16	1.442e+03	2.719e+03	0.531	0.595771	
rootcd18	2.329e+04	3.478e+03	6.695	2.51e-11	***
rootcd19	-4.070e+01	4.454e+02	-0.091	0.927205	
rootcd20	4.520e+04	1.063e+04	4.254	2.16e-05	***
rootcd21	-2.646e+02	4.892e+03	-0.054	0.956864	
rootcd22	-1.128e+03	7.942e+03	-0.142	0.887095	

```

rootcd25      -1.697e+03  4.113e+03  -0.412  0.680027
rootcd26      -1.180e+03  1.459e+03  -0.809  0.418485
rootcd27      -1.682e+04  3.342e+03  -5.033  5.08e-07 ***
rootcd29       6.127e+02  3.111e+03   0.197  0.843884
rootcd30      -2.119e+02  2.880e+03  -0.074  0.941367
rootcd31       1.178e+03  1.048e+04   0.112  0.910460
rootcd33       1.425e+04  8.667e+03   1.644  0.100187
rootcd34       1.766e+02  8.761e+02   0.202  0.840252
rootcd35      -3.915e+03  2.379e+03  -1.646  0.099954 .
rootcd36       1.968e+04  5.578e+03   3.528  0.000425 ***

```

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 786400 on 3456 degrees of freedom

Multiple R-squared: 0.1377, Adjusted R-squared: 0.1127

F-statistic: 5.518 on 100 and 3456 DF, p-value: < 2.2e-16

===== model alias =====

Model :

```

profit ~ reslan + ez + loc5 + female + age + avgAge + maxAge +
  minAge + educ + market + transport + livstcdTypeCount + livstcd1 +
  livstcd2 + livstcd3 + livstcd4 + livstcd5 + livstcd6 + livstcd7 +
  livstcd8 + livstcd9 + livstcd10 + livstcd11 + livstcd12 +
  equipTypeCount + eqcdown21 + eqcdown22 + eqcdown31 + eqcdown51 +
  eqcdown61 + eqcdown62 + eqcdown63 + eqcdown64 + eqcdown65 +
  cropTypeCount + cropcd0 + cropcd1 + cropcd2 + cropcd3 + cropcd4 +
  cropcd5 + cropcd6 + cropcd8 + cropcd9 + cropcd10 + cropcd11 +
  cropcd12 + cropcd13 + cropcd14 + cropcd15 + cropcd16 + cropcd17 +
  cropcd18 + cropcd19 + cropcd20 + cropcd21 + cropcd22 + cropcd23 +
  cropcd24 + cropcd25 + cropcd26 + cropcd27 + cropcd28 + cropcd29 +
  cropcd31 + cropcd32 + cropcd33 + cropcd34 + cropcd35 + rootTypeCount +
  rootcd0 + rootcd5 + rootcd6 + rootcd7 + rootcd8 + rootcd9 +
  rootcd11 + rootcd14 + rootcd16 + rootcd18 + rootcd19 + rootcd20 +
  rootcd21 + rootcd22 + rootcd25 + rootcd26 + rootcd27 + rootcd29 +
  rootcd30 + rootcd31 + rootcd33 + rootcd34 + rootcd35 + rootcd36

```

Complete :

	(Intercept)	reslanAkan	reslanEwe	reslanGaAdangbe
loc5RuralForest	0	0	0	0
loc5RuralSavannah	0	0	0	0
cropcd0	0	0	0	0
rootcd0	0	0	0	0
rootcd14	0	0	0	0
	reslanDagbani	reslanHausa	reslanOther	reslanUnknown
loc5RuralForest	0	0	0	0
loc5RuralSavannah	0	0	0	0
cropcd0	0	0	0	0



rootcd0	0	0	0	0			
rootcd14	0	0	0	0			
	ezForest	ezSavannah	femaleTRUE	age	avgAge	maxAge	minAge
loc5RuralForest	1	0	0	0	0	0	0
loc5RuralSavannah	0	1	0	0	0	0	0
cropcd0	0	0	0	0	0	0	0
rootcd0	0	0	0	0	0	0	0
rootcd14	0	0	0	0	0	0	0
	educBasicEducation	educSecondaryEducation					
loc5RuralForest	0	0					
loc5RuralSavannah	0	0					
cropcd0	0	0					
rootcd0	0	0					
rootcd14	0	0					
	educTertiaryEducation	educOther	marketTRUE	transportTRUE			
loc5RuralForest	0	0	0	0			
loc5RuralSavannah	0	0	0	0			
cropcd0	0	0	0	0			
rootcd0	0	0	0	0			
rootcd14	0	0	0	0			
	livstcdTypeCount	livstcd1	livstcd2	livstcd3	livstcd4		
loc5RuralForest	0	0	0	0	0		
loc5RuralSavannah	0	0	0	0	0		
cropcd0	0	0	0	0	0		
rootcd0	0	0	0	0	0		
rootcd14	0	0	0	0	0		
	livstcd5	livstcd6	livstcd7	livstcd8	livstcd9	livstcd10	
loc5RuralForest	0	0	0	0	0	0	
loc5RuralSavannah	0	0	0	0	0	0	
cropcd0	0	0	0	0	0	0	
rootcd0	0	0	0	0	0	0	
rootcd14	0	0	0	0	0	0	
	livstcd11	livstcd12	equipTypeCount	eqcdwn21	eqcdwn22		
loc5RuralForest	0	0	0	0	0		
loc5RuralSavannah	0	0	0	0	0		
cropcd0	0	0	0	0	0		
rootcd0	0	0	0	0	0		
rootcd14	0	0	0	0	0		
	eqcdwn31	eqcdwn51	eqcdwn61	eqcdwn62	eqcdwn63		
loc5RuralForest	0	0	0	0	0		
loc5RuralSavannah	0	0	0	0	0		
cropcd0	0	0	0	0	0		
rootcd0	0	0	0	0	0		
rootcd14	0	0	0	0	0		
	eqcdwn64	eqcdwn65	cropTypeCount	cropcd1	cropcd2	cropcd3	
loc5RuralForest	0	0	0	0	0	0	
loc5RuralSavannah	0	0	0	0	0	0	

cropcd0	0	0	0	0	0	0	
rootcd0	0	0	0	0	0	0	
rootcd14	0	0	0	0	0	0	
	cropcd4	cropcd5	cropcd6	cropcd8	cropcd9	cropcd10	cropcd11
loc5RuralForest	0	0	0	0	0	0	0
loc5RuralSavannah	0	0	0	0	0	0	0
cropcd0	0	0	0	0	0	0	0
rootcd0	0	0	0	0	0	0	0
rootcd14	0	0	0	0	0	0	0
	cropcd12	cropcd13	cropcd14	cropcd15	cropcd16	cropcd17	
loc5RuralForest	0	0	0	0	0	0	
loc5RuralSavannah	0	0	0	0	0	0	
cropcd0	0	0	0	0	0	0	
rootcd0	0	0	0	0	0	0	
rootcd14	0	0	0	0	0	0	
	cropcd18	cropcd19	cropcd20	cropcd21	cropcd22	cropcd23	
loc5RuralForest	0	0	0	0	0	0	
loc5RuralSavannah	0	0	0	0	0	0	
cropcd0	0	0	0	0	0	0	
rootcd0	0	0	0	0	0	0	
rootcd14	0	0	0	0	0	0	
	cropcd24	cropcd25	cropcd26	cropcd27	cropcd28	cropcd29	
loc5RuralForest	0	0	0	0	0	0	
loc5RuralSavannah	0	0	0	0	0	0	
cropcd0	0	0	0	0	0	0	
rootcd0	0	0	0	0	0	0	
rootcd14	0	0	0	0	0	0	
	cropcd31	cropcd32	cropcd33	cropcd34	cropcd35		
loc5RuralForest	0	0	0	0	0		
loc5RuralSavannah	0	0	0	0	0		
cropcd0	0	0	0	0	0		
rootcd0	0	0	0	0	0		
rootcd14	0	0	0	0	0		
	rootTypeCount	rootcd5	rootcd6	rootcd7	rootcd8	rootcd9	
loc5RuralForest	0	0	0	0	0	0	
loc5RuralSavannah	0	0	0	0	0	0	
cropcd0	0	0	0	0	0	0	
rootcd0	0	0	0	0	0	0	
rootcd14	0	0	0	0	0	0	
	rootcd11	rootcd16	rootcd18	rootcd19	rootcd20	rootcd21	
loc5RuralForest	0	0	0	0	0	0	
loc5RuralSavannah	0	0	0	0	0	0	
cropcd0	0	0	0	0	0	0	
rootcd0	0	0	0	0	0	0	
rootcd14	0	0	0	0	0	0	
	rootcd22	rootcd25	rootcd26	rootcd27	rootcd29	rootcd30	
loc5RuralForest	0	0	0	0	0	0	

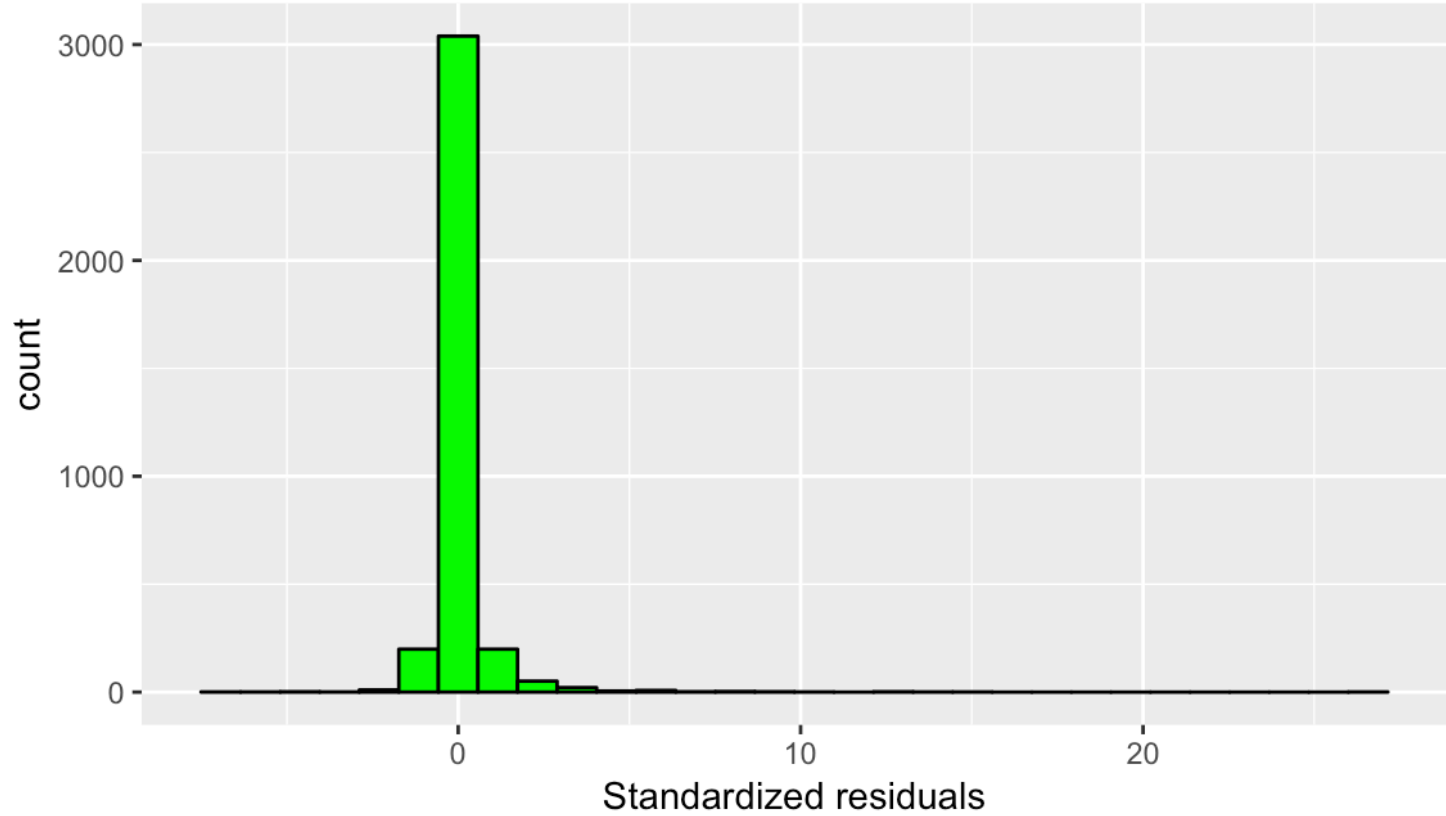
loc5RuralSavannah	0	0	0	0	0	0
cropcd0	0	0	0	0	0	0
rootcd0	0	0	0	0	0	0
rootcd14	0	0	0	0	0	0
	rootcd31	rootcd33	rootcd34	rootcd35	rootcd36	
loc5RuralForest	0	0	0	0	0	
loc5RuralSavannah	0	0	0	0	0	
cropcd0	0	0	0	0	0	
rootcd0	0	0	0	0	0	
rootcd14	0	0	0	0	0	

There are correlated variables. See above model alias.

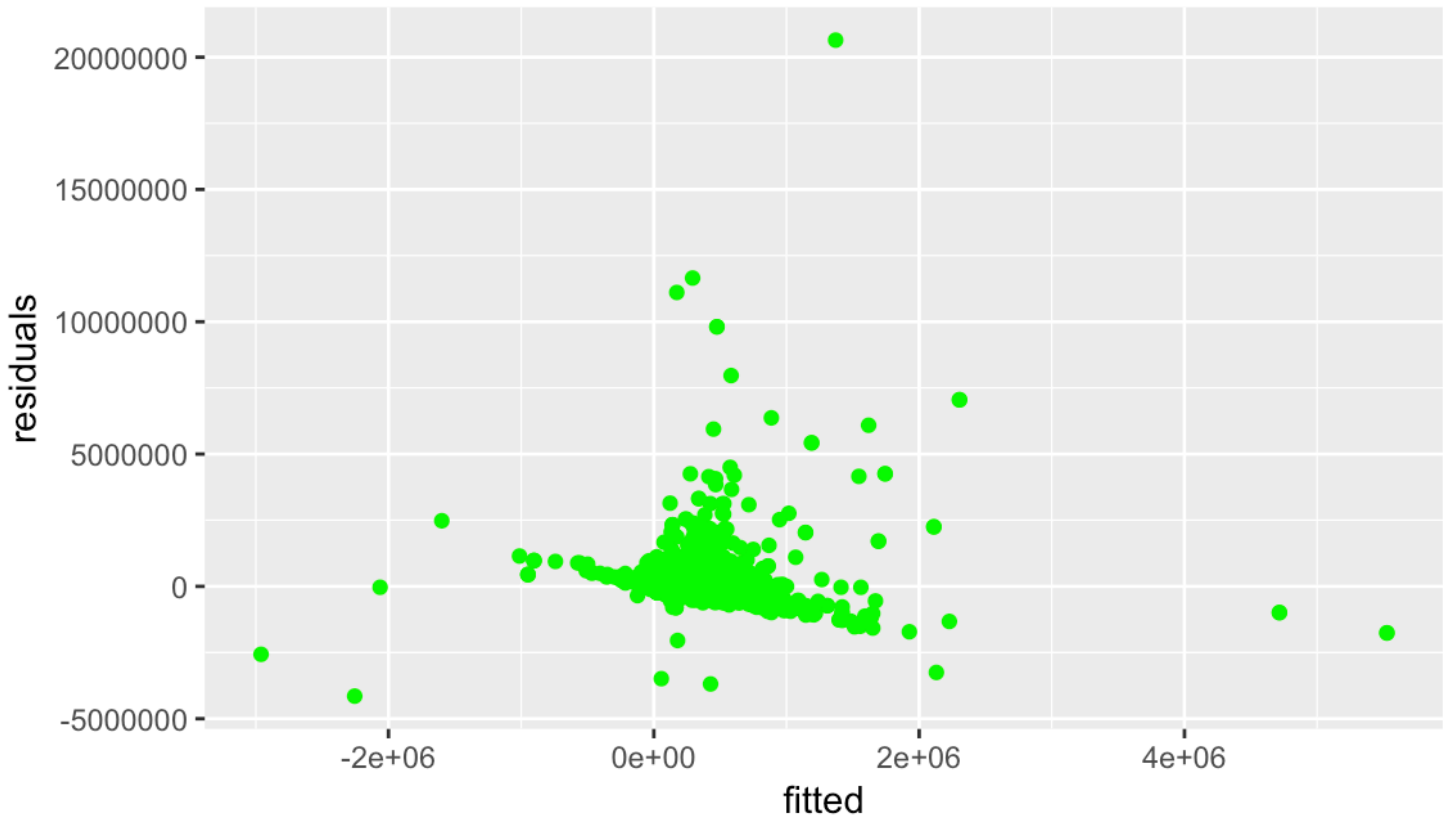
Hide

```
suppressWarnings(plotResiduals(hh_profit_rural, hh_profit_model_rural, "green", "RURAL"))
```

Standardized residuals of RURAL model



## Homoskedasticity of RURAL model



## Unrestricted model for urban area

Fit unrestricted model for urban area profit. We notice that the adjusted R squared is really low, and there are a lot of correlated variable. By looking at data size, we see that there are less observations than variables for urban are profit data. So we don't have enough data to fit a model.

[Hide](#)

```
hh_profit_model_urban <- lm(profit ~ .,
                             data = hh_profit_urban)
checkCorrVarAndTestHnull(hh_profit_model_urban)
```

```
===== model summary =====
```

Call:

```
lm(formula = profit ~ ., data = hh_profit_urban)
```

Residuals:

Min	1Q	Median	3Q	Max
-609766	-62750	0	43147	709516

Coefficients: (53 not defined because of singularities)

Estimate	Std. Error	t value	Pr(> t )
----------	------------	---------	----------

(Intercept)	-3.969e+05	1.048e+06	-0.379	0.712
reslanAkan	1.276e+05	7.214e+05	0.177	0.863
reslanUnknown	2.854e+05	8.071e+05	0.354	0.730
ezSavannah	-9.612e+05	6.879e+05	-1.397	0.188
femaleTRUE	-1.599e+04	2.718e+05	-0.059	0.954
age	1.074e+04	1.302e+04	0.825	0.425
avgAge	-1.654e+04	3.730e+04	-0.443	0.665
maxAge	6.680e+03	1.790e+04	0.373	0.716
minAge	4.451e+02	2.880e+04	0.015	0.988
educBasicEducation	3.686e+05	3.712e+05	0.993	0.340
educSecondaryEducation	2.229e+05	8.876e+05	0.251	0.806
educTertiaryEducation	9.755e+02	5.419e+05	0.002	0.999
marketTRUE	1.075e+06	5.401e+05	1.989	0.070
transportTRUE	NA	NA	NA	NA
livstcdTypeCount	-8.311e+05	8.671e+05	-0.959	0.357
livstcd1	NA	NA	NA	NA
livstcd2	NA	NA	NA	NA
livstcd3	9.080e+04	1.196e+05	0.759	0.462
livstcd4	5.326e+04	1.320e+05	0.403	0.694
livstcd5	1.049e+05	6.914e+04	1.517	0.155
livstcd6	NA	NA	NA	NA
livstcd7	7.646e+04	7.178e+04	1.065	0.308
livstcd8	8.764e+04	9.539e+04	0.919	0.376
livstcd9	NA	NA	NA	NA
livstcd10	NA	NA	NA	NA
livstcd11	NA	NA	NA	NA
livstcd12	NA	NA	NA	NA
equipTypeCount	7.001e+05	1.150e+06	0.609	0.554
eqcdown21	NA	NA	NA	NA
eqcdown22	NA	NA	NA	NA
eqcdown31	NA	NA	NA	NA
eqcdown51	NA	NA	NA	NA
eqcdown61	NA	NA	NA	NA
eqcdown62	NA	NA	NA	NA
eqcdown63	NA	NA	NA	NA
eqcdown64	NA	NA	NA	NA
eqcdown65	-2.093e+05	1.025e+06	-0.204	0.842
cropTypeCount	-9.111e+04	4.412e+05	-0.206	0.840
cropcd0	NA	NA	NA	NA
cropcd1	-6.547e+04	1.327e+05	-0.493	0.631
cropcd2	NA	NA	NA	NA
cropcd3	NA	NA	NA	NA
cropcd4	NA	NA	NA	NA
cropcd5	NA	NA	NA	NA
cropcd6	NA	NA	NA	NA
cropcd8	NA	NA	NA	NA
cropcd9	NA	NA	NA	NA

cropcd10	3.242e+03	6.012e+03	0.539	0.600
cropcd11	NA	NA	NA	NA
cropcd12	NA	NA	NA	NA
cropcd13	NA	NA	NA	NA
cropcd14	NA	NA	NA	NA
cropcd15	NA	NA	NA	NA
cropcd16	NA	NA	NA	NA
cropcd17	-1.669e+02	9.282e+04	-0.002	0.999
cropcd18	NA	NA	NA	NA
cropcd19	NA	NA	NA	NA
cropcd20	NA	NA	NA	NA
cropcd21	NA	NA	NA	NA
cropcd22	-1.723e+04	1.136e+04	-1.517	0.155
cropcd23	NA	NA	NA	NA
cropcd24	NA	NA	NA	NA
cropcd25	NA	NA	NA	NA
cropcd26	NA	NA	NA	NA
cropcd27	NA	NA	NA	NA
cropcd28	9.335e+05	2.474e+06	0.377	0.713
cropcd29	NA	NA	NA	NA
cropcd31	NA	NA	NA	NA
cropcd32	NA	NA	NA	NA
cropcd33	NA	NA	NA	NA
cropcd34	NA	NA	NA	NA
cropcd35	NA	NA	NA	NA
rootTypeCount	1.883e+04	1.323e+05	0.142	0.889
rootcd0	NA	NA	NA	NA
rootcd5	3.974e+04	9.861e+04	0.403	0.694
rootcd6	1.564e+05	2.891e+05	0.541	0.598
rootcd7	-4.601e+04	2.293e+05	-0.201	0.844
rootcd8	2.544e+04	9.786e+04	0.260	0.799
rootcd9	NA	NA	NA	NA
rootcd11	2.299e+07	6.715e+07	0.342	0.738
rootcd14	NA	NA	NA	NA
rootcd16	-1.143e+07	3.387e+07	-0.337	0.742
rootcd18	-1.380e+04	1.851e+05	-0.075	0.942
rootcd19	-1.535e+05	7.176e+05	-0.214	0.834
rootcd20	2.374e+05	3.858e+05	0.615	0.550
rootcd21	NA	NA	NA	NA
rootcd22	NA	NA	NA	NA
rootcd25	9.814e+04	2.739e+05	0.358	0.726
rootcd26	-1.006e+05	1.687e+05	-0.597	0.562
rootcd27	2.846e+05	8.460e+05	0.336	0.742
rootcd29	1.115e+04	1.757e+05	0.063	0.950
rootcd30	5.394e+03	1.116e+05	0.048	0.962
rootcd31	NA	NA	NA	NA
rootcd33	-3.586e+04	3.993e+04	-0.898	0.387

```

rootcd34          NA          NA          NA          NA
rootcd35          NA          NA          NA          NA
rootcd36      -2.105e+04  1.195e+05  -0.176      0.863
---

```

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 475100 on 12 degrees of freedom

Multiple R-squared: 0.7237, Adjusted R-squared: -0.2664

F-statistic: 0.7309 on 43 and 12 DF, p-value: 0.7818

===== model alias =====

Model :

```

profit ~ reslan + ez + female + age + avgAge + maxAge + minAge +
  educ + market + transport + livstcdTypeCount + livstcd1 +
  livstcd2 + livstcd3 + livstcd4 + livstcd5 + livstcd6 + livstcd7 +
  livstcd8 + livstcd9 + livstcd10 + livstcd11 + livstcd12 +
  equipTypeCount + eqcdown21 + eqcdown22 + eqcdown31 + eqcdown51 +
  eqcdown61 + eqcdown62 + eqcdown63 + eqcdown64 + eqcdown65 +
  cropTypeCount + cropcd0 + cropcd1 + cropcd2 + cropcd3 + cropcd4 +
  cropcd5 + cropcd6 + cropcd8 + cropcd9 + cropcd10 + cropcd11 +
  cropcd12 + cropcd13 + cropcd14 + cropcd15 + cropcd16 + cropcd17 +
  cropcd18 + cropcd19 + cropcd20 + cropcd21 + cropcd22 + cropcd23 +
  cropcd24 + cropcd25 + cropcd26 + cropcd27 + cropcd28 + cropcd29 +
  cropcd31 + cropcd32 + cropcd33 + cropcd34 + cropcd35 + rootTypeCount +
  rootcd0 + rootcd5 + rootcd6 + rootcd7 + rootcd8 + rootcd9 +
  rootcd11 + rootcd14 + rootcd16 + rootcd18 + rootcd19 + rootcd20 +
  rootcd21 + rootcd22 + rootcd25 + rootcd26 + rootcd27 + rootcd29 +
  rootcd30 + rootcd31 + rootcd33 + rootcd34 + rootcd35 + rootcd36

```

Complete :

```

      (Intercept) reslanAkan reslanUnknown ezSavannah femaleTRUE
transportTRUE    1          0          0          0          0
livstcd1         0          0          0          0          0
livstcd2         0          0          0          0          0
livstcd6         0          0          0          0          0
livstcd9         0          0          0          0          0
livstcd10        0          0          0          0          0
livstcd11        0          0          0          0          0
livstcd12        0          0          0          0          0
eqcdown21        0          0          0          0          0
eqcdown22        0          0          0          0          0
eqcdown31        0          0          0          0          0
eqcdown51        0          0          0          0          0
eqcdown61        0          0          0          0          0
eqcdown62        0          0          0          0          0
eqcdown63        0          0          0          0          0
eqcdown64        0          0          0          0          0

```

croprd0	0	0	0	0	0
croprd2	0	0	0	0	0
croprd3	0	0	0	0	0
croprd4	0	0	0	0	0
croprd5	0	0	0	0	0
croprd6	0	0	0	0	0
age avgAge maxAge minAge educBasicEducation					
transportTRUE	0	0	0	0	0
livstcd1	0	0	0	0	0
livstcd2	0	0	0	0	0
livstcd6	0	0	0	0	0
livstcd9	0	0	0	0	0
livstcd10	0	0	0	0	0
livstcd11	0	0	0	0	0
livstcd12	0	0	0	0	0
eqcdwn21	0	0	0	0	0
eqcdwn22	0	0	0	0	0
eqcdwn31	0	0	0	0	0
eqcdwn51	0	0	0	0	0
eqcdwn61	0	0	0	0	0
eqcdwn62	0	0	0	0	0
eqcdwn63	0	0	0	0	0
eqcdwn64	0	0	0	0	0
croprd0	0	0	0	0	0
croprd2	0	0	0	0	0
croprd3	0	0	0	0	0
croprd4	0	0	0	0	0
croprd5	0	0	0	0	0
croprd6	0	0	0	0	0
educSecondaryEducation educTertiaryEducation marketTRUE					
transportTRUE	0			0	0
livstcd1	0			0	0
livstcd2	0			0	0
livstcd6	0			0	0
livstcd9	0			0	0
livstcd10	0			0	0
livstcd11	0			0	0
livstcd12	0			0	0
eqcdwn21	0			0	0
eqcdwn22	0			0	0
eqcdwn31	0			0	0
eqcdwn51	0			0	0
eqcdwn61	0			0	0
eqcdwn62	0			0	0
eqcdwn63	0			0	0
eqcdwn64	0			0	0
croprd0	0			0	0



croprd2	0	0	0
croprd3	0	0	0
croprd4	0	0	0
croprd5	0	0	0
croprd6	0	0	0
livstcdTypeCount livstcd3 livstcd4 livstcd5 livstcd7 livstcd8			
transportTRUE	0	0	0
livstcd1	0	0	0
livstcd2	0	0	0
livstcd6	0	0	0
livstcd9	0	0	0
livstcd10	0	0	0
livstcd11	0	0	0
livstcd12	0	0	0
eqcdwn21	0	0	0
eqcdwn22	0	0	0
eqcdwn31	0	0	0
eqcdwn51	0	0	0
eqcdwn61	0	0	0
eqcdwn62	0	0	0
eqcdwn63	0	0	0
eqcdwn64	0	0	0
croprd0	0	0	0
croprd2	0	0	0
croprd3	0	0	0
croprd4	0	0	0
croprd5	0	0	0
croprd6	0	0	0
equipTypeCount eqcdwn65 cropTypeCount cropcd1 cropcd10			
transportTRUE	0	0	0
livstcd1	0	0	0
livstcd2	0	0	0
livstcd6	0	0	0
livstcd9	0	0	0
livstcd10	0	0	0
livstcd11	0	0	0
livstcd12	0	0	0
eqcdwn21	0	0	0
eqcdwn22	0	0	0
eqcdwn31	0	0	0
eqcdwn51	0	0	0
eqcdwn61	0	0	0
eqcdwn62	0	0	0
eqcdwn63	0	0	0
eqcdwn64	0	0	0
cropcd0	0	0	0
cropcd2	0	0	0

croprd3	0	0	0	0	0	0
croprd4	0	0	0	0	0	0
croprd5	0	0	0	0	0	0
croprd6	0	0	0	0	0	0
	croprd17	croprd22	croprd28	rootTypeCount	rootcd5	rootcd6
transportTRUE	0	0	0	0	0	0
livstcd1	0	0	0	0	0	0
livstcd2	0	0	0	0	0	0
livstcd6	0	0	0	0	0	0
livstcd9	0	0	0	0	0	0
livstcd10	0	0	0	0	0	0
livstcd11	0	0	0	0	0	0
livstcd12	0	0	0	0	0	0
eqcdwn21	0	0	0	0	0	0
eqcdwn22	0	0	0	0	0	0
eqcdwn31	0	0	0	0	0	0
eqcdwn51	0	0	0	0	0	0
eqcdwn61	0	0	0	0	0	0
eqcdwn62	0	0	0	0	0	0
eqcdwn63	0	0	0	0	0	0
eqcdwn64	0	0	0	0	0	0
croprd0	0	0	0	0	0	0
croprd2	0	0	0	0	0	0
croprd3	0	0	0	0	0	0
croprd4	0	0	0	0	0	0
croprd5	0	0	0	0	0	0
croprd6	0	0	0	0	0	0
	rootcd7	rootcd8	rootcd11	rootcd16	rootcd18	rootcd19
transportTRUE	0	0	0	0	0	0
livstcd1	0	0	0	0	0	0
livstcd2	0	0	0	0	0	0
livstcd6	0	0	0	0	0	0
livstcd9	0	0	0	0	0	0
livstcd10	0	0	0	0	0	0
livstcd11	0	0	0	0	0	0
livstcd12	0	0	0	0	0	0
eqcdwn21	0	0	0	0	0	0
eqcdwn22	0	0	0	0	0	0
eqcdwn31	0	0	0	0	0	0
eqcdwn51	0	0	0	0	0	0
eqcdwn61	0	0	0	0	0	0
eqcdwn62	0	0	0	0	0	0
eqcdwn63	0	0	0	0	0	0
eqcdwn64	0	0	0	0	0	0
croprd0	0	0	0	0	0	0
croprd2	0	0	0	0	0	0
croprd3	0	0	0	0	0	0

cropcd4	0	0	0	0	0	0	0
cropcd5	0	0	0	0	0	0	0
cropcd6	0	0	0	0	0	0	0
	rootcd25	rootcd26	rootcd27	rootcd29	rootcd30	rootcd33	
transportTRUE	0	0	0	0	0	0	
livstcd1	0	0	0	0	0	0	
livstcd2	0	0	0	0	0	0	
livstcd6	0	0	0	0	0	0	
livstcd9	0	0	0	0	0	0	
livstcd10	0	0	0	0	0	0	
livstcd11	0	0	0	0	0	0	
livstcd12	0	0	0	0	0	0	
eqcdwn21	0	0	0	0	0	0	
eqcdwn22	0	0	0	0	0	0	
eqcdwn31	0	0	0	0	0	0	
eqcdwn51	0	0	0	0	0	0	
eqcdwn61	0	0	0	0	0	0	
eqcdwn62	0	0	0	0	0	0	
eqcdwn63	0	0	0	0	0	0	
eqcdwn64	0	0	0	0	0	0	
cropcd0	0	0	0	0	0	0	
cropcd2	0	0	0	0	0	0	
cropcd3	0	0	0	0	0	0	
cropcd4	0	0	0	0	0	0	
cropcd5	0	0	0	0	0	0	
cropcd6	0	0	0	0	0	0	
	rootcd36						
transportTRUE	0						
livstcd1	0						
livstcd2	0						
livstcd6	0						
livstcd9	0						
livstcd10	0						
livstcd11	0						
livstcd12	0						
eqcdwn21	0						
eqcdwn22	0						
eqcdwn31	0						
eqcdwn51	0						
eqcdwn61	0						
eqcdwn62	0						
eqcdwn63	0						
eqcdwn64	0						
cropcd0	0						
cropcd2	0						
cropcd3	0						
cropcd4	0						

```
cropcd5      0
cropcd6      0
[ reached getOption("max.print") -- omitted 31 rows ]
```

There are correlated variables. See above model alias.

[Hide](#)

```
# check data size
dim(hh_profit_urban)
```

```
[1] 56 94
```

## Restricted model with only education and local characteristic information (R2, R3)

We also want to see how education and local characteristic info alone influence profit. Looking at hypothesis test result, loc5 and loc3 are correlated with other variables. So remove them and fit model again. Looking at the summary, there are not many significant variables, also the R squared is very low.

[Hide](#)

```
hh_profit_model_r2 <- lm(profit ~ educ + ez + loc2 + loc5 + loc3 + market + transport
,
                        data = hh_profit)
checkCorrVarAndTestHnull(hh_profit_model_r2)
```

```
===== model summary =====
```

Call:

```
lm(formula = profit ~ educ + ez + loc2 + loc5 + loc3 + market +
    transport, data = hh_profit)
```

Residuals:

Min	1Q	Median	3Q	Max
-6686068	-259175	-173421	3700	21648271

Coefficients: (3 not defined because of singularities)

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	341148.54	271694.77	1.256	0.20933
educBasicEducation	604.85	31312.81	0.019	0.98459
educSecondaryEducation	20078.89	88658.86	0.226	0.82085
educTertiaryEducation	-114169.37	78495.67	-1.454	0.14590
educOther	-245371.94	181893.76	-1.349	0.17743
ezForest	79473.13	294500.56	0.270	0.78729
ezSavannah	-25449.92	44857.89	-0.567	0.57051

```

loc2Rural          -31347.21  266430.71  -0.118  0.90635
loc5RuralCoastal    NA          NA      NA      NA
loc5RuralForest     -36.17  292751.08   0.000  0.99990
loc5RuralSavannah  NA          NA      NA      NA
loc3Rural           NA          NA      NA      NA
marketTRUE          -110715.56  41813.82  -2.648  0.00814 **
transportTRUE       -23157.10  30171.81  -0.768  0.44283

```

```

---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

Residual standard error: 828500 on 3602 degrees of freedom
Multiple R-squared:  0.006371, Adjusted R-squared:  0.003612
F-statistic: 2.309 on 10 and 3602 DF, p-value: 0.01058

```

```

===== model alias =====

```

Model :

```

profit ~ educ + ez + loc2 + loc5 + loc3 + market + transport

```

Complete :

```

              (Intercept) educBasicEducation educSecondaryEducation
loc5RuralCoastal      1             0             0
loc5RuralSavannah  -1             0             0
loc3Rural             0             0             0
              educTertiaryEducation educOther ezForest ezSavannah
loc5RuralCoastal      0             0          -1          -1
loc5RuralSavannah      0             0           1           1
loc3Rural             0             0           0           0
              loc2Rural loc5RuralForest marketTRUE transportTRUE
loc5RuralCoastal      0             0             0             0
loc5RuralSavannah      1            -1             0             0
loc3Rural             1             0             0             0

```

There are correlated variables. See above model alias.

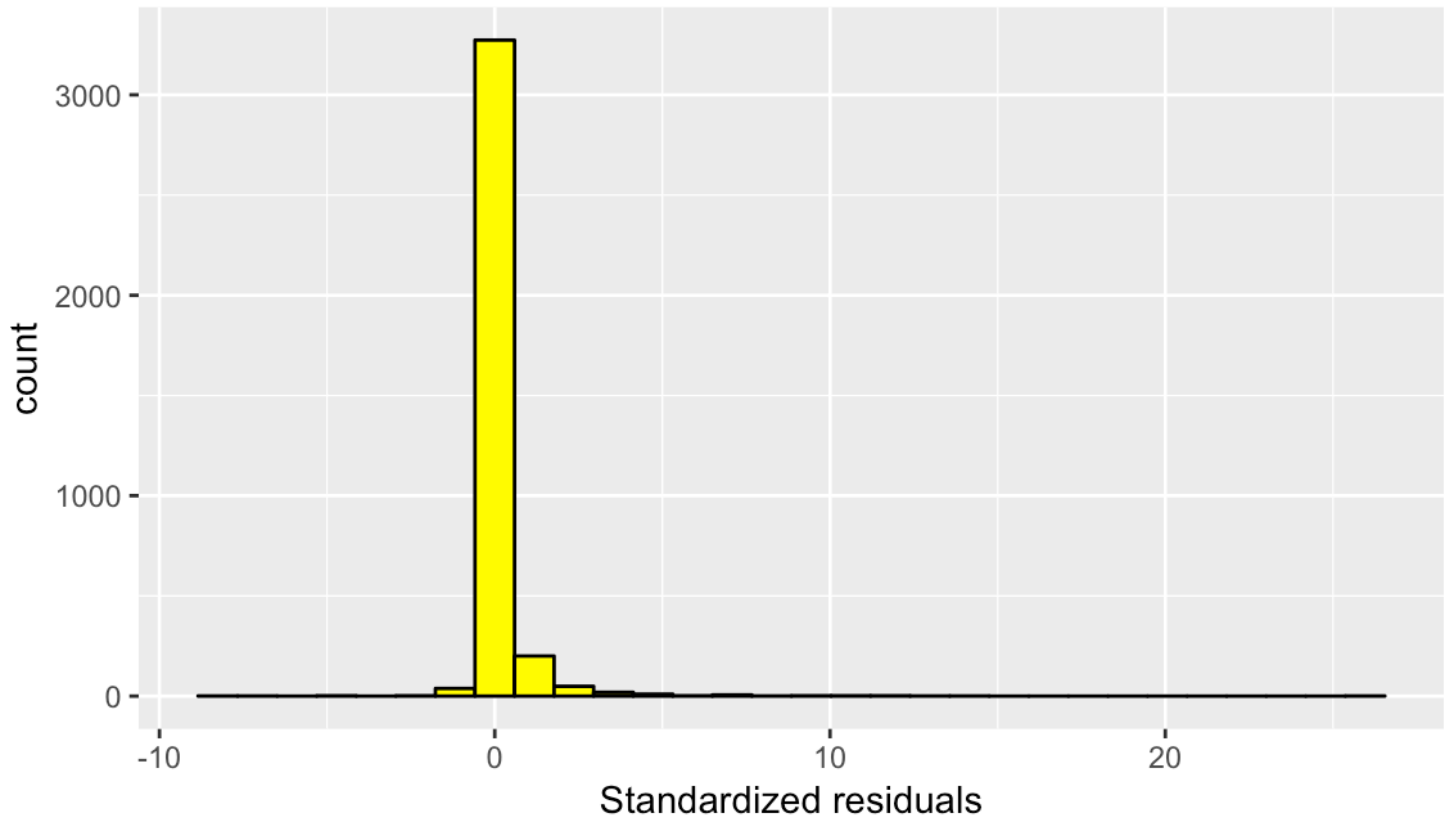
Hide

```

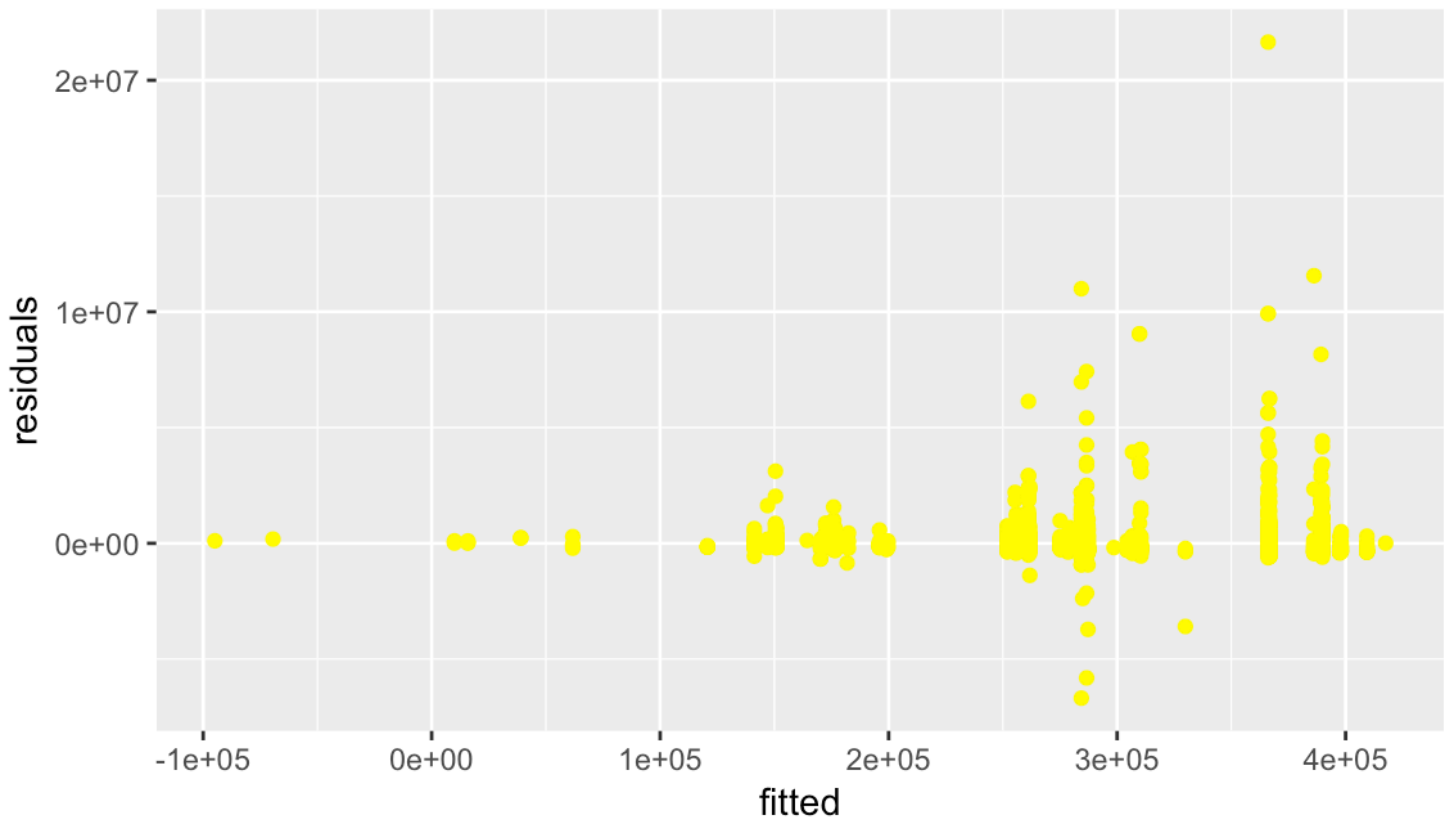
suppressWarnings(plotResiduals(hh_profit, hh_profit_model_r2, "yellow", "R2"))

```

## Standardized residuals of R2 model



## Homoskedasticity of R2 model


[Hide](#)

```
# remove loc5 and loc3 and fit model again
hh_profit_model_r3 <- lm(profit ~ educ + ez + loc2 + market + transport,
                        data = hh_profit)
checkCorrVarAndTestHnull(hh_profit_model_r3)
```

```
===== model summary =====
```

Call:

```
lm(formula = profit ~ educ + ez + loc2 + market + transport,
    data = hh_profit)
```

Residuals:

Min	1Q	Median	3Q	Max
-6686068	-259175	-173421	3700	21648271

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	341178.5	122474.1	2.786	0.00537 **
educBasicEducation	604.7	31295.3	0.019	0.98458
educSecondaryEducation	20079.0	88642.2	0.227	0.82081

```
educTertiaryEducation  -114169.3    78481.0   -1.455   0.14583
educOther               -245371.9   181867.3   -1.349   0.17736
ezForest                79437.1    39472.3    2.012   0.04424 *
ezSavannah            -25450.3    44729.0   -0.569   0.56940
loc2Rural              -31377.0   112961.1   -0.278   0.78121
marketTRUE             -110716.1    41581.9   -2.663   0.00779 **
transportTRUE          -23157.1    30167.4   -0.768   0.44276
```

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 828400 on 3603 degrees of freedom

Multiple R-squared: 0.006371, Adjusted R-squared: 0.003889

F-statistic: 2.567 on 9 and 3603 DF, p-value: 0.006086

===== hypothesis test =====

Linear hypothesis test

Hypothesis:

educBasicEducation = 0

educSecondaryEducation = 0

educTertiaryEducation = 0

educOther = 0

ezForest = 0

ezSavannah = 0

loc2Rural = 0

marketTRUE = 0

transportTRUE = 0

Model 1: restricted model

Model 2: profit ~ educ + ez + loc2 + market + transport

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	3612	2.4883e+15				
2	3603	2.4724e+15	9	1.5852e+13	2.5668	0.006086 **

---

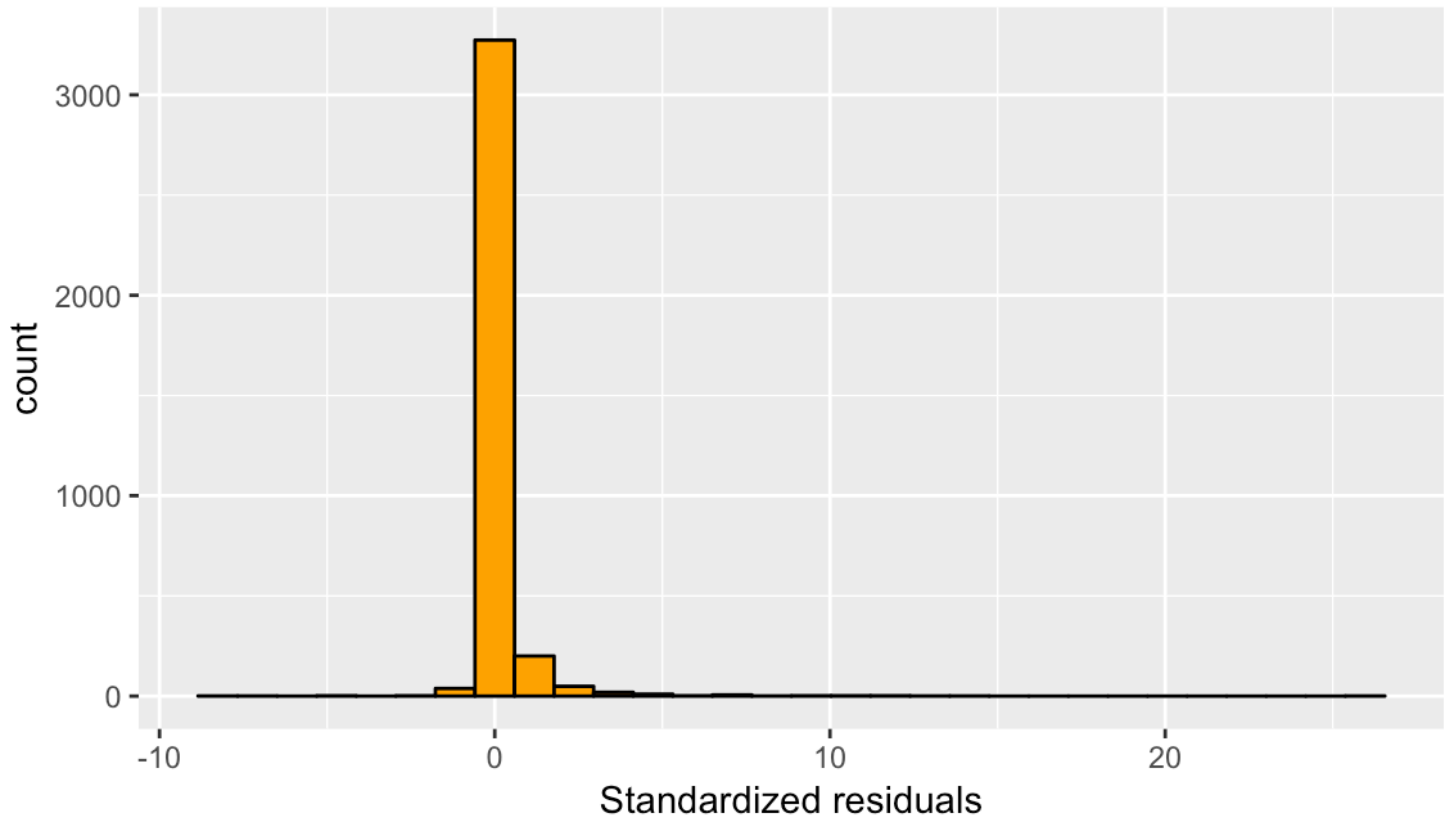
Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Hide

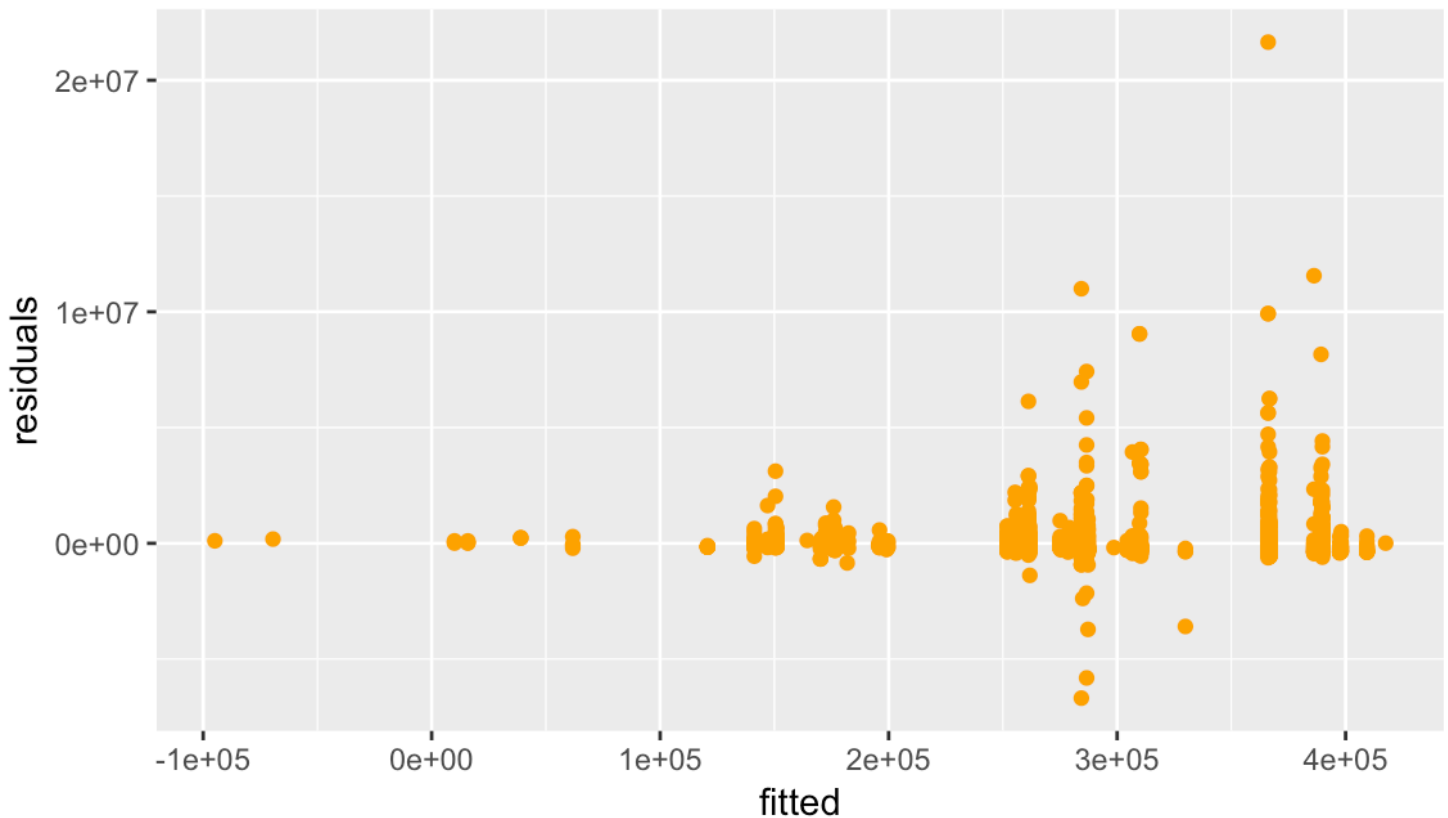
```
suppressWarnings(plotResiduals(hh_profit, hh_profit_model_r3, "orange", "R3"))
```



## Standardized residuals of R3 model



## Homoskedasticity of R3 model



## Restricted model with educ \* age (R4)

Although the education in the above model is not significant, we are curious to see if educ and age together would become significant. So we fit a model with `educ * age`. However, looking at the summary it's still not significant.

[Hide](#)

```
hh_profit_model_r4 <- lm(profit ~ educ * age + female + ez + loc2 + market + transport,
                        data = hh_profit)
checkCorrVarAndTestHnull(hh_profit_model_r4)
```

```
===== model summary =====
```

Call:

```
lm(formula = profit ~ educ * age + female + ez + loc2 + market +
    transport, data = hh_profit)
```

Residuals:

Min	1Q	Median	3Q	Max
-6693082	-260609	-173405	4360	21635925

## Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	363514.37	136714.01	2.659	0.00787	**
educBasicEducation	-43926.23	108243.80	-0.406	0.68491	
educSecondaryEducation	206525.08	346672.42	0.596	0.55139	
educTertiaryEducation	-221002.20	358817.33	-0.616	0.53799	
educOther	19470.37	903976.83	0.022	0.98282	
age	-75.88	1130.08	-0.067	0.94647	
femaleTRUE	-35363.11	33340.55	-1.061	0.28891	
ezForest	78569.77	39620.05	1.983	0.04743	*
ezSavannah	-32145.59	45435.95	-0.707	0.47931	
loc2Rural	-37651.00	113301.97	-0.332	0.73968	
marketTRUE	-106856.37	41927.77	-2.549	0.01086	*
transportTRUE	-19935.82	30302.98	-0.658	0.51066	
educBasicEducation:age	866.44	2429.91	0.357	0.72143	
educSecondaryEducation:age	-5229.05	8809.74	-0.594	0.55285	
educTertiaryEducation:age	2087.10	7643.66	0.273	0.78483	
educOther:age	-5491.26	17510.71	-0.314	0.75385	

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 828900 on 3597 degrees of freedom

Multiple R-squared: 0.006882, Adjusted R-squared: 0.002741

F-statistic: 1.662 on 15 and 3597 DF, p-value: 0.0515

===== hypothesis test =====

Linear hypothesis test

## Hypothesis:

```
educBasicEducation = 0
educSecondaryEducation = 0
educTertiaryEducation = 0
educOther = 0
age = 0
femaleTRUE = 0
ezForest = 0
ezSavannah = 0
loc2Rural = 0
marketTRUE = 0
transportTRUE = 0
educBasicEducation:age = 0
educSecondaryEducation:age = 0
educTertiaryEducation:age = 0
educOther:age = 0
```

Model 1: restricted model

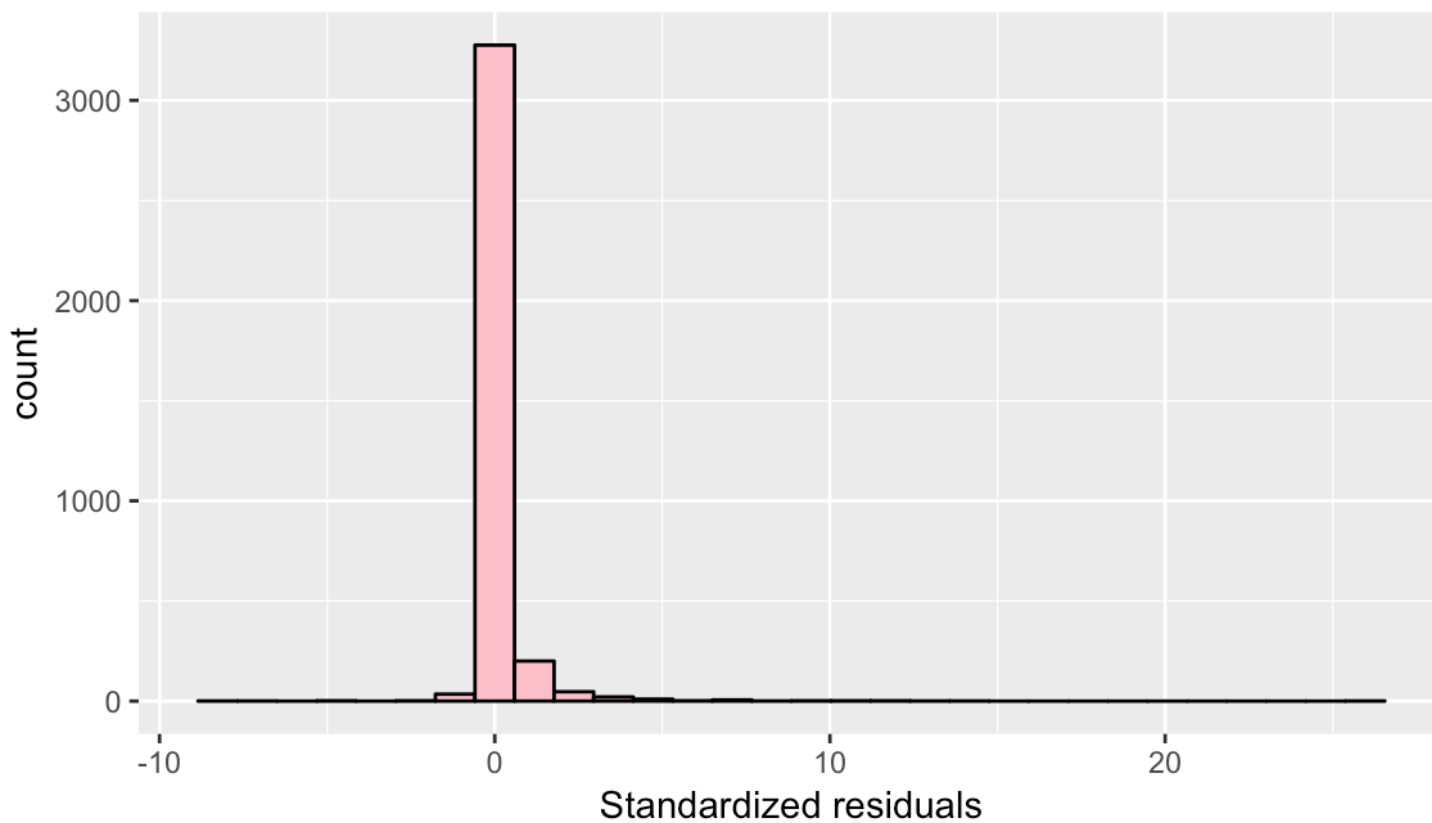
Model 2: profit ~ educ \* age + female + ez + loc2 + market + transport

```
Res.Df      RSS Df Sum of Sq      F Pr(>F)
1    3612 2.4883e+15
2    3597 2.4711e+15 15 1.7125e+13 1.6618 0.0515 .
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

[Hide](#)

```
suppressWarnings(plotResiduals(hh_profit, hh_profit_model_r4, "pink", "R4"))
```

## Standardized residuals of R4 model



## Homoskedasticity of R4 model

