

In [3]:

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# =====
# food transfers July 2022
# =====

import numpy as np
import pandas as pd
import os
import warnings
warnings.filterwarnings("ignore")

os.chdir('C:/Users/rodri/Dropbox/Malawi/SIEG2021 (1)/2022 July/Data/Clean data/Phase
percentiles = [0.05, 0.1, .25, .5, .75, 0.8, 0.9, 0.95, 0.99]
#July 14th 2022 MWK vs US dollar
dollar_MWK = 1030.36
pd.options.display.float_format = '{:,.2f}'.format

# =====
# Import data: Data from the field and conversion rates (ISA-LSMS price conversions)
# =====

# I followed Pau's code in 2019 to generate a dataset containing all the food trans
data = pd.read_stata("transfers.dta")

# Now, let's use this data to
# (1) convert the transfers to kgs and monetary units.
# (2) Provide summary stats of the transfers.
# (3) Create food transfers datasets
# (4) Create measures of total transfers_in transfers_out at household level to appe
#inreturn_sum = pd.value_counts(data['in.return'])

# (1) convert the transfers to kgs and monetary units. -----
data.replace('cassava', 'cassavatubers', inplace=True)
data.replace('potatocrips', 'potatocrisps', inplace=True)

list_items = list(data['good'].unique())
list_units = list(data['units'].unique())

## Other units
# hands/hand/handful conversion rate
data.loc[data['unit.other'].isin(['Hand', 'Hands']), 'transf_kg'] = 0.113 ##0.5 cup/
# Dove
data.loc[data['unit.other'].isin(['Dove']), 'transf_kg'] = 0.3 ## average weight of
# Pot
data.loc[data['unit.other'].isin(['Pot']), 'transf_kg'] = 0.5 ##from previous year
data_other = data.loc[data['units']=='other']

# Leandro-Raul kgs through price method

### NOTE ON CONVERSIONS =====
conversionkg_pivot = pd.read_csv('C:/Users/rodri/Dropbox/Malawi/Chied_Field_June_19/

#4. All units have at least one crop conversion. To fill the whole matrix I use the
conversionkg_pivot = conversionkg_pivot.apply(lambda x: x.fillna(x.median()),axis=1)

## import median price of goods
prices = pd.read_csv('C:/Users/rodri/Dropbox/Malawi/SIEG2021 (1)/2022 July/Data/Clea

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# there was not consumption of samosas in the village (not possible to compute price
# so I'll use the price for mandazidou
conversionkg_pivot['samosa'] = conversionkg_pivot['mandazidou']
prices.loc[40] = ['samosa', 1651.98]
prices.loc[41] = ['chips', 1174.4]
# chips/crisps
conversionkg_pivot['chips'] = conversionkg_pivot['potatocrisps']

data['units'] = data['units'].replace('other', 100)
data['units'] = pd.to_numeric(data['units'])
data['transf_kg'] = np.nan
data[['units']] = data[['units']].replace([np.nan, 23, 25, 0], 99)

## Convert to kgs
for i in range(0, len(data)):
    good = data['good'][i]
    if good != 'fert':
        data['transf_kg'][i] = data['quant'][i] * conversionkg_pivot.loc[int(data['uni

data.loc[data['good']=='fert', 'transf_kg'] = data.loc[data['good']=='fert', 'kg']

## Convert to MWK

# get a price from fertilizer for those that had to pay
print('for the food items I use the median consumption prices in the village')

print('for fertilizer I used the mean price reported in the agricultural production

p_fert = 14620.3665/50 # mean price 50kg fertilizer bag 14620.366563891745

new_row = {'good': 'fert', 'p_c': p_fert}
prices = prices.append(new_row, ignore_index=True)

data = data.merge(prices, on='good', how='left')
data['transf_MWK'] = data['transf_kg'] * data['p_c']
data['transf_dollar'] = data['transf_MWK'] / dollar_MWK

# for the moment dont remove outliers. Now we have fertilizer so outliers not so cle
...
def fun(x):
    q_99 = x.quantile(0.95)
    q_1 = x.quantile(0.00)
    return (x > q_99) | (x < q_1)

print('Summary transfers in the village')
print(data['transf_dollar'].describe(percentiles=[0.25, .5, .75, 0.95, 0.99]))

data['outlier'] = 1 * (data['transf_dollar'] > data['transf_dollar'].quantile(0.995))
data = data.loc[data['outlier'] == 0]

data = data.drop(columns='outlier')
...
print('=====')
print('Summary transfers in the village')
print('=====')
print(data['transf_dollar'].describe(percentiles=[0.1, .5, 0.9]))

print('=====')
print('Summary only FOOD transfers')
print('=====')
print(data.loc[data['good'] != 'fert', ['transf_kg', 'transf_dollar']].describe(percent

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print(data['good'].describe())

print('=====')
print('Summary only fertilizer')
print('=====')
print(data.loc[data['good']=='fert', ['transf_kg', 'transf_dollar', 'recipro.cashback'])

N_fert = len(data.loc[data['good']=='fert', 'good'])
N_payfert = sum(data['recipro.cashback']>0)
print('Proportion fert transfers household had to pay back: ', round(N_payfert/N_fert))

print('Comparison value given median price vs payback for those hhs that reported to')
print(data.loc[(data['good']=='fert') & (data['recipro.cashback']>0), ['transf_kg', ''])
print('There are some large discrepancies but avg is similar')

del data['kg']
# .
data.to_csv('transfers_in_kg_MWK_22.csv')

print('food+fertilizer transfers data in long format with values to kgs and MWK save')

### Create household level transfers dataset =====

all_count_bydirection = data[['direction', 'quant']].groupby(by='direction').count()
all_avg_bydirection = data[['direction', 'transf_kg', 'transf_MWK']].groupby(by='direction').mean()
print('=====')
print('transfers in vs out: number and average value')
print(all_count_bydirection)
print(all_avg_bydirection)
# in: 612, out:712
# Eliminate barter transfers: transfers were household was supposed to receive or give
# data['in.return'] = data['in.return'].astype(str)
# data = data[data['in.return']!='nan'] #there were 8 cases that reported return

# Only transfers that we could match in the village
data_match = data.loc[data['id']!=0]
count_bydirection = data_match[['direction', 'quant']].groupby(by='direction').count()
avg_bydirection = data_match[['direction', 'quant']].groupby(by='direction').mean()
print('=====')
print('transfers in vs out MATCHED in the village:')
print(count_bydirection)

print('Almost all transfers were matched! 97% of in-transfers matched, 97.6% of out-')

data_match_out = data_match.loc[data_match['direction']=='out']
data_match_in = data_match.loc[data_match['direction']=='in']
data_match_in.rename(columns={'hhid':'id', 'id':'hhid'}, inplace=True)

### Only transfers that we can cross-validate directions
data_directions_match = data_match_in.merge(data_match_out, on=['hhid', 'id'], how='inner')

### Only transfers that we can match cross-validate directions and the item
data_item_match = data_match_in.merge(data_match_out, on=['hhid', 'id', 'good'], how='inner')

print('=====')
print('Number of transfers cross-validated based on direction: coinciding hhid given')
print('num transf:', len(data_directions_match))
### Only transfers that we can match cross-validate directions and the item
data_item_match = data_match_in.merge(data_match_out, on=['hhid', 'id', 'good'], how='inner')

print('=====')

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print('Number of transfers cross-validated based on direction and food item')
print('num transf:',len(data_item_match))
print('the low number of matched transfers is reasonable in the sense that we are as
print('The number is slightly higher than in 2019 (119)')

### Import consumption dataset to create base ids
c22 = pd.read_csv('C:/Users/rodri/Dropbox/Malawi/SIEG2021 (1)/2022 July/Data/Clean d
data_final = c22[['hhid']]

# household level net transfers variables =====

# ----- variable 1: 'transfers1_net' ----- (no exploit network)
#Take only what households report to give and receive (not info about what other hou

print('=====')
print('household level net transfers variables')
print('=====')
print('net transfers only includes food transfers. Reasons: (1) most fertilizer trans

data = data.loc[data['good']!='fert']
data_given = data.loc[data['direction']=='out',['hhid','transf_MWK']].groupby(by='hh
data_given.columns = ['transfers1_out']

data_received = data.loc[data['direction']=='in',['hhid','transf_MWK']].groupby(by='
data_received.columns = ['transfers1_in']

transfers1 = data_received.merge(data_given,on='hhid', how='outer')

transfers1['transfers1_net'] = transfers1['transfers1_in'].fillna(0) - transfers1['t
data_final = data_final.merge(transfers1, on='hhid', how='left')

# -----variable 2: 'transfers2_net' ----- (restrictive)
# only those transfers that we could cross-validate---X reports giving food item to
# _x variables denote from direction in. _y variables denote from direction out.

# given the problem of the big span of time across surveys, I'd not use this measure

data_item_match[['id','hhid','good','transf_kg_x','transf_MWK_x','transf_kg_y','tran

data_item_match['transf_MWK_avg'] = data_item_match[['transf_MWK_x','transf_MWK_y']]

data_given = data_item_match[['id','transf_MWK_avg']].groupby(by='id').sum()
data_given.reset_index(inplace=True)
data_given.columns = ['hhid','transfers2_in']

data_received = data_item_match[['hhid','transf_MWK_avg']].groupby(by='hhid').sum()
data_received.reset_index(inplace=True)
data_received.columns = ['hhid','transfers2_out']

transfers2 = data_given.merge(data_received,on='hhid', how='outer')
transfers2['transfers2_net'] = transfers2['transfers2_in'].fillna(0) - transfers2['t

data_final = data_final.merge(transfers2, on='hhid', how='left')

# ----- variable 3: 'transfers3_net' ----- (extensive)
# what household X reports to receive plus what rest of households report to give to
# First, to avoid double-counting, eliminate the transfers that we could cross-valid
transfers3 = data.merge(data_item_match[['hhid','id','good','transf_MWK_avg']],on=['
transfers3 = transfers3[transfers3['transf_MWK_avg'].isnull()]

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# we lose 130 observations. Exactly the number of transfers we should cross-validate

data_given1 = transfers3.loc[transfers3['direction']=='out',['hhid','transf_MWK']]
data_given2 = transfers3.loc[transfers3['direction']=='in',['id','transf_MWK']]
data_given2.columns = ['hhid','transf_MWK']

data_given= data_given1.append(data_given2)
data_given= data_given.groupby(by='hhid').sum()
data_given.columns = ['transfers3_out']

data_received1 = transfers3.loc[transfers3['direction']=='in',['hhid','transf_MWK']]
data_received2 = transfers3.loc[transfers3['direction']=='out',['id','transf_MWK']]
data_received2.columns = ['hhid','transf_MWK']

data_received= data_received1.append(data_received2)
data_received= data_received.groupby(by='hhid').sum()
data_received.columns = ['transfers3_in']

transfers3 = data_received.merge(data_given,on='hhid', how='outer')
transfers3['transfers3_net'] = transfers3['transfers3_in'].fillna(0) - transfers3['t

data_final = data_final.merge(transfers3, on='hhid', how='left')
data_final.set_index('hhid', inplace=True)
data_final[['transfers1_net','transfers2_net','transfers3_net']]
data_final_year = data_final*4*12

data_final.to_csv('hhtransfers_week_22.csv')
data_final_year.to_csv('hhtransfers_year_22.csv')

print('=====')
print('saved datasets net food transfers hh level: hhtransfers_week_22.csv, hhtransf

data_final_dollars = data_final/dollar_MWK
print('=====')
print(data_final_dollars.describe(percentiles=[0.05,0.25,0.5,0.75,0.9,0.95]))

## transfers as gifts or as an exchange?
print('=====')
print('Comment: transfers as gifts or as an exchange? ')
print('=====')
print('93.29% of the food transfers reported nothing was given/received back for the
print('a common concern or criticism we have received is that, perhaps, these transf

print('Note: the number was computed with the data Albert first clean on transfers.

for the food items I use the median consumption prices in the village
for fertilizer I used the mean price reported in the agricultural production section
=====
Summary transfers in the village
=====
count    2,068.00
mean      1.34
std       13.53
min        0.00
10%        0.02
50%        0.20
90%        1.42
max       356.38
Name: transf_dollar, dtype: float64
=====

```

## Summary only FOOD transfers

```
=====
      transf_kg  transf_dollar
count    1,972.00        1,972.00
mean         1.97          1.12
std         26.32         13.76
min          0.00          0.00
10%          0.02          0.02
50%          0.43          0.19
90%          2.18          0.83
max     1,080.00        356.38
count         2087
unique         43
top      thobwa
freq         259
Name: good, dtype: object
=====
```

## Summary only fertilizer

```
=====
      transf_kg  transf_dollar  recipro.cashback
count     96.00        96.00          59.00
mean     20.81         5.91       8,055.93
std     18.75         5.32       6,875.59
min       1.00         0.28       1,500.00
10%       5.00         1.42       2,500.00
50%      15.00         4.26       7,000.00
90%      50.00        14.19      15,000.00
max     110.00        31.22      46,000.00
```

Proportion fert transfers household had to pay back: 0.61

Comparison value given median price vs payback for those hhs that reported to pay back for fertilizer transfer

```
      transf_kg  transf_MWK  recipro.cashback
count     59.00        59.00          59.00
mean     24.68      7,216.02      8,055.93
std     20.30      5,935.02      6,875.59
min       5.00      1,462.04      1,500.00
10%       5.00      1,462.04      2,500.00
50%      20.00      5,848.15      7,000.00
90%      50.00     14,620.37     15,000.00
max     110.00     32,164.81     46,000.00
```

There are some large discrepancies but avg is similar

food+fertilizer transfers data in long format with values to kgs and MWK saved: Phase 3/transfers/transfers\_in\_kg\_MWK.csv

```
=====
transfers in vs out: number and average value
      quant
```

direction

in 720

out 1271

```
      transf_kg  transf_MWK
```

direction

in 4.14 1,546.60

out 2.06 1,289.00

```
=====
transfers in vs out MATCHED in the village:
      quant
```

direction

in 705

out 1244

Almost all transfers were matched! 97% of in-transfers matched, 97.6% of out-transfers matched. booklet work very well. Enumerators did great job

```
=====
```

Number of transfers cross-validated based on direction: coinciding hhid giving with hhid from who you received---and equivalently for receiving.

num transf: 1269

```
=====
```

Number of transfers cross-validated based on direction and food item

num transf: 132

the low number of matched transfers is reasonable in the sense that we are asking for

r last 7 days while the span of the surveys was 2 months  
The number is slightly higher than in 2019 (119)

=====

household level net transfers variables

=====

net transfers only includes food transfers. Reasons: (1) most fertilizer transfers hh actually payed (2) consistent with previous data (3) timing is different

=====

saved datasets net food transfers hh level: hhtransfers\_week\_22.csv, hhtransfers\_year\_22.csv

=====

	transfers1_net	transfers2_net	transfers3_net
count	263.00	118.00	271.00
mean	-3.59	0.00	-1.62
std	31.05	3.46	47.95
min	-336.87	-14.19	-358.56
5%	-9.42	-4.86	-11.46
25%	-1.31	-0.34	-1.35
50%	-0.06	0.04	0.36
75%	0.88	0.41	2.69
90%	4.32	1.70	6.30
95%	6.60	3.62	10.47
max	20.31	14.11	341.05

=====

Comment: transfers as gifts or as an exchange?

=====

93.29% of the food transfers reported nothing was given/received back for the transfer. Neither give/receive back the food, or another food item, ganyu labor or other potential exchanges.

a common concern or criticism we have received is that, perhaps, these transfers have nothing to do with insurance or other motives, but they are just exchanges of goods in an economy where there is little use of cash. Our results reject this hypothesis. Whatever is going on with the transfers in the village, it is not direct exchange as in a barter economy.

Note: the number was computed with the data Albert first clean on transfers. check pdf Data/Summaries/old\_foodtransfers\_22\_summary

In [ ]: