

In [2]:

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root_path = 'C:/Users/rodri/Dropbox/Malawi/SIEG2021 (1)/2022 July'
folder_fig = root_path+'/Figures'

import numpy as np
import pandas as pd
import os
import warnings

# Suppress warnings in the entire notebook
warnings.filterwarnings("ignore")

pd.options.display.max_columns = None
pd.options.display.max_rows = None

os.chdir('C:/Users/rodri/Dropbox/JMP/python')
from data_functions_albert import remove_outliers, gini
os.chdir(root_path)

save=False

dollar_MWK = 1030.36
pd.options.display.float_format = '{:,.2f}'.format
percentiles=[0.5]

#income
inc = pd.read_csv(root_path+'/Data/Clean data/Phase 3 - Consumption, Transfers, Inco
#consumption
cons = pd.read_csv(root_path+'/Data/Clean data/Phase 3 - Consumption, Transfers, Inc
#Roster
roster = pd.read_csv(root_path+'/Data/Clean data/Phase 1 - Roster/roster_22.csv')

foodtra = pd.read_csv(root_path+'/Data/Clean data/Phase 3 - Consumption, Transfers,

data = pd.merge( inc, roster, on='hhid', how='left')
data = pd.merge(data, cons, on='hhid', how='inner')
data = pd.merge(data, foodtra, on='hhid', how='left')

del data['cashtrans_yes.1'], data['k_farm.1']

data.rename(columns={'hh_area_plots': 'land_area', 'hh_value_plots': 'land_value'},

data['inctotal_cap'] = data['inctotal']/data['hh_size'].replace(0,np.mean(data['hh_s
data['ctotal_cap'] = data['ctotal']/data['hh_size'].replace(0,np.mean(data['hh_size'
data['inctotal_trans_cap'] = data['inctotal_trans']/data['hh_size'].replace(0,np.mea
data['land_area_cap'] = data['land_area']/data['hh_size'].replace(0,np.mean(data['hh
data['y_net_cap'] = data['y_net']/data['hh_size'].replace(0,np.mean(data['hh_size'])

data['ln_inc'] = np.log(data['inctotal']).replace([-np.inf, np.inf], np.nan)
data['ln_c'] = np.log(data['ctotal']).replace([-np.inf, np.inf], np.nan)
data['ln_land'] = np.log(data['land_area']).replace([-np.inf, np.inf], np.nan)
data['ln_inctrans'] = np.log(data['inctotal_trans']).replace([-np.inf, np.inf], np.
data['ln_agric'] = np.log(data['y_net']).replace([-np.inf, np.inf], np.nan)

data['ln_inc_cap'] = np.log(data['inctotal_cap']).replace([-np.inf, np.inf], np.nan
data['ln_c_cap'] = np.log(data['ctotal_cap']).replace([-np.inf, np.inf], np.nan)
data['ln_land_cap'] = np.log(data['land_area_cap']).replace([-np.inf, np.inf], np.na
data['ln_inctrans_cap'] = np.log(data['inctotal_trans_cap']).replace([-np.inf, np.i

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data['ln_agric_cap'] = np.log(data['y_net_cap']).replace([-np.inf, np.inf], np.nan)

data['rank_inctotal'] = data['inctotal'].rank(pct=True)
data['rank_landarea'] = data['land_area'].rank(pct=True)
data['rank_landvalue'] = data['land_value'].rank(pct=True)
data['rank_ctotal'] = data['ctotal'].rank(pct=True)

data['rank_wtotal'] = data['wtotal'].rank(pct=True)
data['wtotal_cap'] = data['wtotal']/data['hh_size']
data['ln_w'] = np.log(data['wtotal']+np.abs(np.min(data['wtotal']))).replace([-np.in
data['ln_w_cap'] = np.log(data['wtotal_cap']+np.abs(np.min(data['wtotal']))).replace

print( '
print('=====')
print(' HOUSEHOLD DATASET WAVE JULY 2022')
print('merges roster, income_weath, consumption, and hhtransfers datasets')
print( '
print('=====')
if save==True:
    data.to_csv(root_path+'/Data/Clean data/hhdata22.csv', index=False)

print('Data saved: hhdata22.csv')
print('Dataset contains the following variables:')
print(data.columns.tolist())

### SUMMARY

print( '
print('=====')
print(' SOCIODEMOGRAPHIC CHARACTERISTICS')
print('=====')
print( '
print(data[['hh_size', 'head_gender', 'head_marital', 'head_age', 'interviewed_19', 'h

## Summary CIW 2019
data[['inctotal']] = data[['inctotal']].replace([0, 0.00], np.nan)

data_sum = data[['ctotal', 'inctotal', 'wtotal', 'land_area', 'ctotal_cap', 'inctotal_cap

data_sum[['ctotal', 'inctotal', 'wtotal', 'ctotal_cap', 'inctotal_cap', 'wtotal_cap']] =

sum_cwi = pd.DataFrame((data_sum[['ctotal', 'inctotal', 'wtotal', 'land_area']])).mean(a
sum_cwi = sum_cwi.append(pd.DataFrame((data_sum[['ctotal', 'inctotal', 'wtotal', 'land_

varlog_df = pd.DataFrame(((data_sum[['ln_c', 'ln_inc', 'ln_w', 'ln_land']])).var(axis=0)

sum_cwi = sum_cwi.append(dict(zip(sum_cwi.columns, np.array(varlog_df.iloc[:,0]).T)))

print( '
print('=====')
print(' CONSUMPTION. INCOME, AND WEALTH')
print('=====')
print( '
print(sum_cwi)

sum_cwi_cap = pd.DataFrame((data_sum[['ctotal_cap', 'inctotal_cap', 'wtotal_cap', 'land
sum_cwi_cap = sum_cwi_cap.append(pd.DataFrame((data_sum[['ctotal_cap', 'inctotal_cap'

varlog_df = pd.DataFrame(((data_sum[['ln_c_cap', 'ln_inc_cap', 'ln_w_cap', 'ln_land_cap

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sum_cwi_cap = sum_cwi_cap.append(dict(zip(sum_cwi_cap.columns, np.array(varlog_df.il

print(sum_cwi_cap)
print('Mean, median, and log-variance')

# In per capita terms

### Summary I
income = data[['inctotal', 'y_net', 'wlabor_inc', 'ganyu_inc', 'business_profits', 'o

sum_inc = (income.loc[:, income.columns != 'hhid']/dollar_MWK).describe(percentiles=[

obs_inc = sum_inc.iloc[0,:]
shares_employ = obs_inc/238

shares_gdp = data[['inctotal', 'y_net', 'wlabor_inc', 'ganyu_inc', 'business_profits'

shares_gdp = shares_gdp/shares_gdp[0]

shares = pd.concat([shares_gdp, shares_employ], axis=1).T

print( '
print('=====')
print(' INCOME AND EMPLOYMENT SHARES')
print('=====')
print( '
print(shares)

## no one grew: cassava, sugarcane,pearlmillet

sum_agric = ((data[['y_agric', 'y_maize', 'y_groundnut', 'y_groundbean', 'y_sweetpot

obs_agric = sum_agric.iloc[0,:]
shares_crops = obs_agric/238

shares_agric = data[['y_agric', 'y_maize', 'y_groundnut', 'y_groundbean', 'y_sweetpot

shares_agric = shares_agric/shares_agric[0]

shares = pd.concat([shares_agric, shares_crops], axis=1).T

print( '
print('=====')
print(' AGRICULTURAL SHARES')
print('=====')
print( '
print(shares)

###Summary C

print( '
print('=====')
print(' CONSUMPTION')
print('=====')
print( '

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c_summary = ((data[['ctotal','c_food','c_food_purch','c_food_ownprod', 'c_nonfood',
print(c_summary)

print( '
print('=====')
print(' WEALTH')
print('=====')
print( '

### Summary W
wealth = data[['wtotal','land_value','k_farm', 'hhlivestock', 'housing', 'hh_assets']

sum_w = wealth.describe(percentiles=[0.5])

var_list = ['wtotal','land_value','k_farm', 'hhlivestock', 'housing', 'hh_assets']
gini_stat= np.empty((1, len(var_list)))

for i,state in enumerate(var_list):
    gini_stat[:,i] = gini(wealth[state].dropna().values)
data_gini = pd.DataFrame(gini_stat, columns=var_list)
data_gini.reset_index(inplace=True)
data_gini['index'] = 'gini'
sum_w.reset_index(inplace=True)
sum_w = sum_w.append(data_gini, ignore_index=True)
print(sum_w)

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HOUSEHOLD DATASET WAVE JULY 2022
merges roster, income_weath, consumption, and hhtransfers datasets

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Data saved: hhdata22.csv
Dataset contains the following variables:
['hhid', 'rightselland', 'chiefpreventsell', 'chiefpreventbequeat', 'cashtrans_ye
s', 'govcoupon', 'inctotal', 'inctotal_trans', 'y_net', 'y_agric', 'y_maize', 'y_gro
undnut', 'y_pigeonpeas', 'total_kg_maize', 'total_kg_groundnut', 'total_kg_pigeonpea
s', 'y_cassava', 'y_soyabean', 'y_sorghum', 'y_fingermillet', 'y_cotton', 'y_tanapos
i', 'y_groundbean', 'y_nkhwani', 'y_sugarcane', 'y_sweetpotatoe', 'sold_agric', 'sol
d_insiders_agric', 'store_agric', 'land_area', 'hh_ratio_value_rent', 'hh_p_acre plo
ts', 'area_cultivated', 'k_farm', 'labor_N', 'labor_h', 'hh_labor_hours', 'hired_men
_L', 'hired_women_L', 'hired_kids_L', 'interm', 'fertilizerkg', 'p_fert', 'value_fer
tilizer', 'spendseeds', 'spendpesticides', 'wlabor_inc', 'wlabor_supply', 'ganyu_ye
s', 'ganyu_inc', 'ganyu_supply', 'business_revenue', 'business_costs', 'business_pro
fits', 'business_profits2', 'NGO_yes', 'gov_yes', 'remittances_yes', 'other_inc', 'c
ashtrans_value', 'NGO_trans', 'gov_trans', 'remittances', 'wtotal', 'housing', 'hh_a
ssets', 'land_value', 'hhlivestock', 'shocks', 'shock_flood', 'shock_drought', 'shoc
k_ldnslide', 'shock_covid', 'shock_adultill', 'shock_kidill', 'shock_death_earner',
'shock_death_oothermemb', 'shock_inp_p', 'shock_out_p', 'shock_pests', 'shock_lvstk',
'shock_theft', 'shock_theft_agric', 'shock_business', 'shock_unemp', 'shock_wage_dec
r', 'shock_other', 'wave', 'invillage_19', 'interviewed_19', 'oldhhid', 'interviewee
name', 'head_name', 'village', 'subvillage', 'key_landmark', 'mosque_church', 'hh_si
ze', 'hh_phone', 'head_gender', 'head_marital', 'head_age', 'head_nickname', 'head_e
duc', 'head_religion', 'head_female', 'head_married_mono', 'head_married_poly', 'hea
d_nevermarried', 'head_divorced', 'head_widowed', 'head_separated', 'head_christia
n', 'head_noeduc', 'spouse_educ', 'ethnic', 'mlanguage', 'village_born', 'village_ye
ars', 'chief_related', 'chief_relation', 'elder_yes', 'elders_related', 'elders_rela
tion', 'head_belowprimary4', 'head_belowprimary7', 'head_belowsecond3', 'head_second
ary', 'head_educ_countin', 'gps_lat', 'gps_long', 'c_food', 'c_food_purch', 'c_food_
ownprod', 'c_nonfood', 'c_housing', 'c_clothes', 'c_education', 'c_health', 'c_funer
alout', 'c_funeralin', 'c_weddingout', 'c_weddingin', 'ctotal', 'transfers1_net', 't
ransfers2_net', 'transfers3_net', 'inctotal_cap', 'ctotal_cap', 'inctotal_trans_ca
p', 'land_area_cap', 'y_net_cap', 'ln_inc', 'ln_c', 'ln_land', 'ln_inctrans', 'ln_ag
ric', 'ln_inc_cap', 'ln_c_cap', 'ln_land_cap', 'ln_inctrans_cap', 'ln_agric_cap', 'r

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ank_inctotal', 'rank_landarea', 'rank_landvalue', 'rank_ctotal', 'rank_wtotal', 'wto
tal_cap', 'ln_w', 'ln_w_cap']
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SOCIODEMOGRAPHIC CHARACTERISTICS

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	hh_size	head_gender	head_marital	head_age	head_married_mono \
count	272.00	272.00	272.00	272.00	272.00
mean	4.47	1.35	2.06	43.51	0.62
std	1.87	0.48	1.56	17.58	0.49
min	1.00	1.00	1.00	10.00	0.00
25%	3.00	1.00	1.00	29.00	0.00
50%	4.00	1.00	1.00	40.50	1.00
75%	6.00	2.00	4.00	53.25	1.00
max	11.00	2.00	6.00	90.00	1.00

	head_married_poly	head_divorced	head_belowprimary4 \
count	272.00	272.00	272.00
mean	0.09	0.19	0.18
std	0.28	0.40	0.39
min	0.00	0.00	0.00
25%	0.00	0.00	0.00
50%	0.00	0.00	0.00
75%	0.00	0.00	0.00
max	1.00	1.00	1.00

	head_belowprimary7
count	272.00
mean	0.28
std	0.45
min	0.00
25%	0.00
50%	0.00
75%	1.00
max	1.00

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CONSUMPTION, INCOME, AND WEALTH

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	ctotal	inctotal	wtotal	land_area
0	872.61	297.89	1,160.30	2.21
1	764.34	148.93	863.29	1.50
2	0.43	1.48	0.55	0.53

  

	ctotal_cap	inctotal_cap	wtotal_cap	land_area_cap
0	214.02	76.45	301.22	0.57
1	173.88	34.80	186.83	0.38
2	0.40	1.51	0.38	0.57

Mean, median, and log-variance

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INCOME AND EMPLOYMENT SHARES

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	inctotal	y_net	wlabor_inc	ganyu_inc	business_profits	other_inc
0	1.00	0.30	0.05	0.42	0.22	0.22
count	1.12	1.06	0.03	0.43	0.26	0.55

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AGRICULTURAL SHARES

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	y_agric	y_maize	y_groundnut	y_groundbean	y_sweetpotatoe \
0	1.00	0.74	0.15	0.01	0.00
count	1.06	1.06	0.40	0.05	0.02

  

	y_pigeonpeas	y_nkhwani
0	0.05	0.01

count 0.62 0.03

=====
CONSUMPTION
=====

Table with 7 columns: count, cttotal, c\_food, c\_food\_purch, c\_food\_ownprod, c\_nonfood, c\_housing. Rows include mean, std, min, 50%, and max values.

Table with 4 columns: count, c\_clothes, c\_education, c\_health. Rows include mean, std, min, 50%, and max values.

=====
WEALTH
=====

Table with 7 columns: index, wtotal, land\_value, k\_farm, hhlivestock, housing, hh\_assets. Rows include count, mean, std, min, 50%, max, and gini values.

In [ ]: