

# Migratory Responses to Environmental Variability in the United States

## A MULTILEVEL APPROACH

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# Background and knowledge gap

- Globally, previous studies primarily focused on environmental changes in the developing world
- In the US, studies on slow-onset environmental variabilities used aggregated data at the regional level or crude level

Gutmann et al. (2005): Great Plains region, 1930-1990

Poston et al. (2009): The entire US at the state level, 1995-2000

Feng et al. (2012): Corn belt region, 1970-2009

- There is a knowledge gap regarding the impact of slow-onset environmental variabilities on migration at the individual level in developed setting

# Research objectives

- Explore individuals' migratory responses to slow-onset environmental variabilities (precipitation, temperature, air quality, and environmental amenity)
- Examine the heterogeneous environmental impacts on migration across two demographic groups (age group 15-64 and age group 65+)

# Data

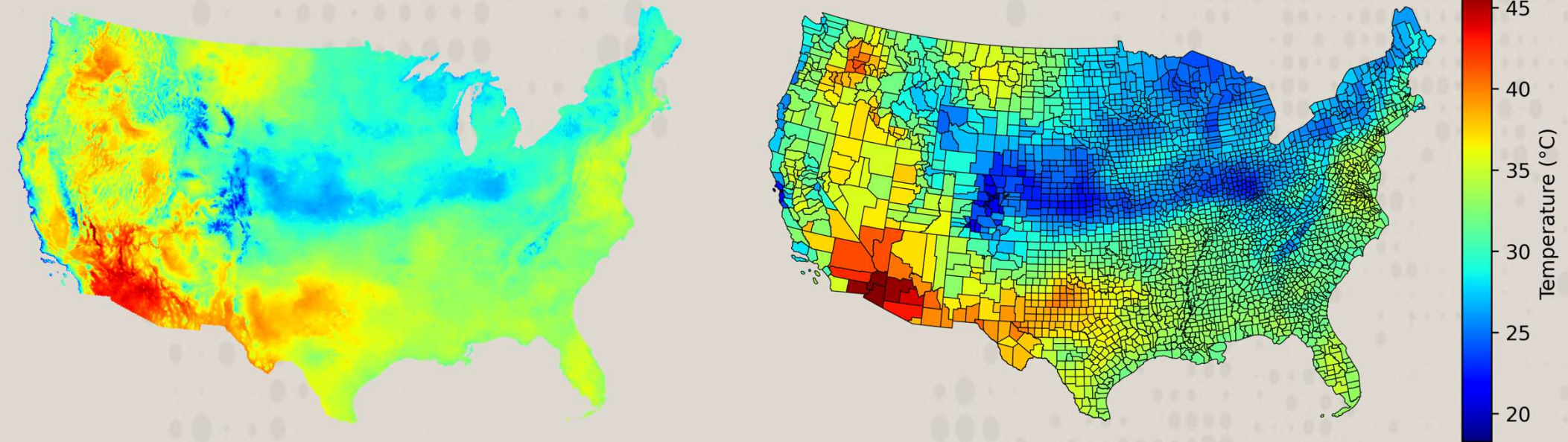
- ✓ The American Community Survey (ACS) Microdata
- ✓ The Parameter-elevation Regressions on Independent Slopes Model (PRISM)
- ✓ The Atmospheric Composition Analysis Group (ACAG)
- ✓ The National Oceanic and Atmospheric Administration (NOAA)

# Migration definition

Migrations are moves cross counties/Public Use Microdata Areas (PUMAs) between the ACS years



# Environmental variability (cont.)



# Environmental variability (cont.)

$$\text{Climate anomaly}_{i,t} = \frac{\text{Level}_{i,t} - \mu_i^{LR}}{\sigma_i^{LR}}$$

$\text{Level}_{i,t}$  = Annual average in county  $i$  at time  $t$  (2010-2020)

$\mu_i^{LR}$  = Long-run (30-year, 1980-2009) average in county  $i$

$\sigma_i^{LR}$  = Long-run (30-year, 1980-2009) standard deviation in county  $i$



# Two-level logistic regression

$$\text{Logit}(\Pr(Y_{ij} = 1)) \\ = \alpha_0 + \alpha_{0j} + \alpha_1 X_{1ij} + \dots + \alpha_k X_{kij} + \beta_1 Z_{1j} + \dots + \beta_m Z_{mj}$$

## Level-1 (individual) variables:

Age  
Personal income  
Gender  
Marital status  
Race  
Education

## Level-2 (county) variables:

Climate anomalies  
Household income  
Housing price  
Employment rate  
Homeownership  
Metro status

Note: Climate anomalies include anomalies in precipitation, temperature, PM2.5, and Normalized Difference Vegetation Index (NDVI).

# Descriptive statistics

	Mean	SD	Min	Max
<b><i>Dependent variable</i></b>				
Migration status	<b>0.42</b>	0.49	0	1
<b><i>Level-1 variables</i></b>				
Age	37.52	17.78	<b>15</b>	96
Personal income (\$1,000)	32.45	52.97	-14.10	1,378.00
Gender	0.50	0.50	0	1
Marital status	0.33	0.47	0	1
Race	0.59	0.49	0	1
Education	0.53	0.50	0	1

Note: N = 2,243,336.

Race distribution: Non-Hispanic White (59%), Non-Hispanic Black (14%), Hispanics (17%), Others (10%).

# Descriptive statistics (cont.)

	Mean	SD	Min	Max
<b><i>Level-2 variables</i></b>				
Precipitation anomaly	<b>0.09</b>	0.34	-0.84	1.55
Temperature anomaly	<b>0.06</b>	0.11	-0.40	0.48
NDVI anomaly	<b>-0.01</b>	0.17	-1.12	0.53
PM2.5 anomaly	<b>-1.07</b>	0.46	-2.12	1.13
Household income (\$1,000)	89.47	20.94	48.97	178.22
Housing price (\$1,000)	296.00	181.35	81.88	1,111.50
Employment rate	91.95	2.48	81.49	97.58
Homeownership	61.25	10.88	18.97	87.44
Metropolitan status	<b>0.99</b>	0.09	0	1

Note: N = 2,243,336.

# General models

## **Level-1 variables**

Age	-0.010***
Personal income	-0.001***
Gender, Male (Ref. = Female)	0.097***
Marital status, Married (Ref. = Unmarried)	-0.071***
Race, NHB (Ref. = NHW)	-0.209***
Race, Hispanics (Ref. = NHW)	-0.335***
Race, Others (Ref. = NHW)	0.022***
Education, College and above (Ref. = Below college)	0.170***

## **Level-2 variables**

Precipitation anomaly	0.017**
Temperature anomaly	0.075***
NDVI anomaly	-0.249***
PM2.5 anomaly	-0.006***

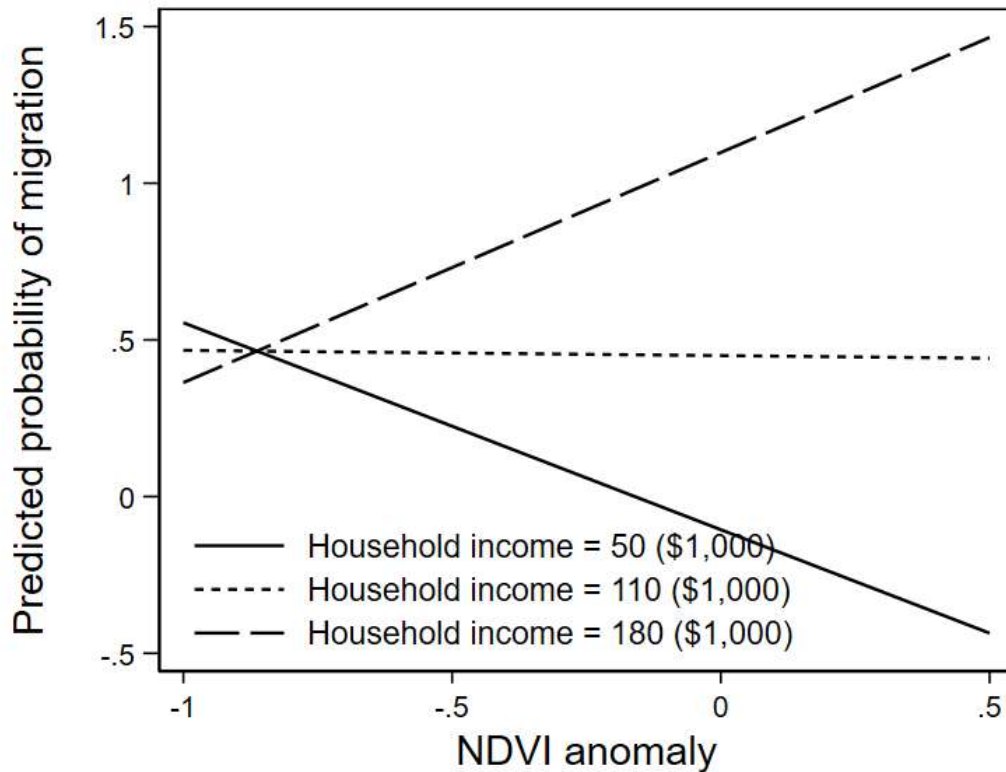
Note: \*\*\* p<0.001, \*\* p<0.01, \* p<0.05. NHB=Non-Hispanic Black, NHW=Non-Hispanic White. Level-2 sociodemographic factors and model diagnostics are not show.

# Age-specific models

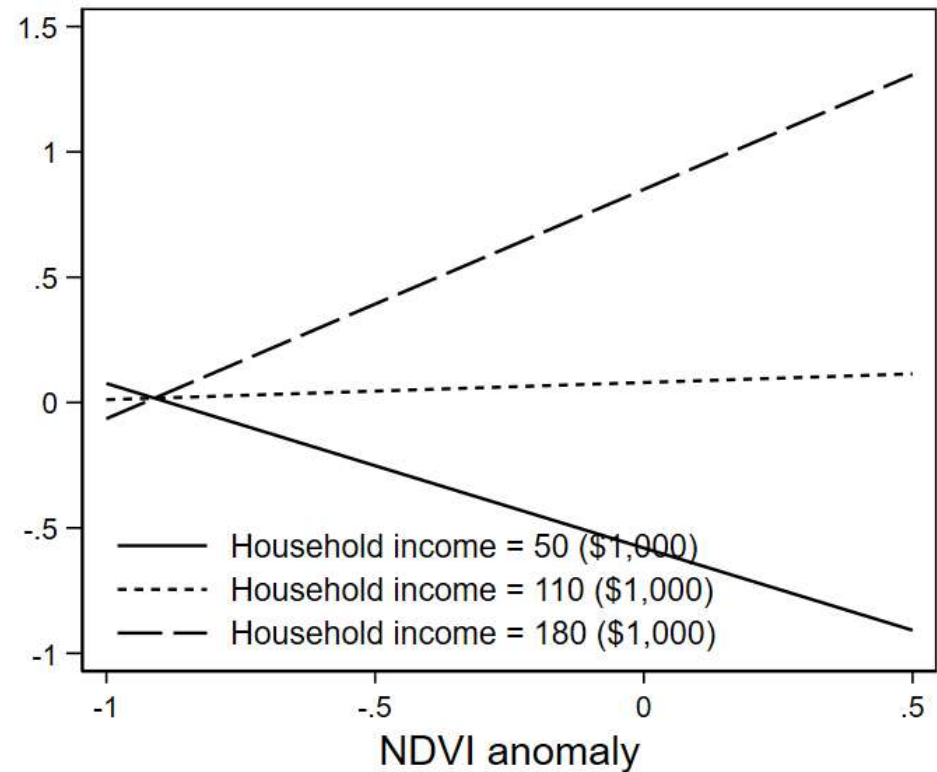
	Mig <sub>15-64</sub>	Mig <sub>65+</sub>
<b><i>Level-2 variables</i></b>		
Precipitation anomaly	0.016**	0.033
Temperature anomaly	0.093***	-0.127*
NDVI anomaly	-0.247***	-0.193***
PM2.5 anomaly	-0.092***	-0.001

*Note:* \*\*\* p<0.001, \*\* p<0.01, \* p<0.05. Level-1 variables, Level-2 sociodemographic factors and model diagnostics are not show.

# Climate-Income interaction

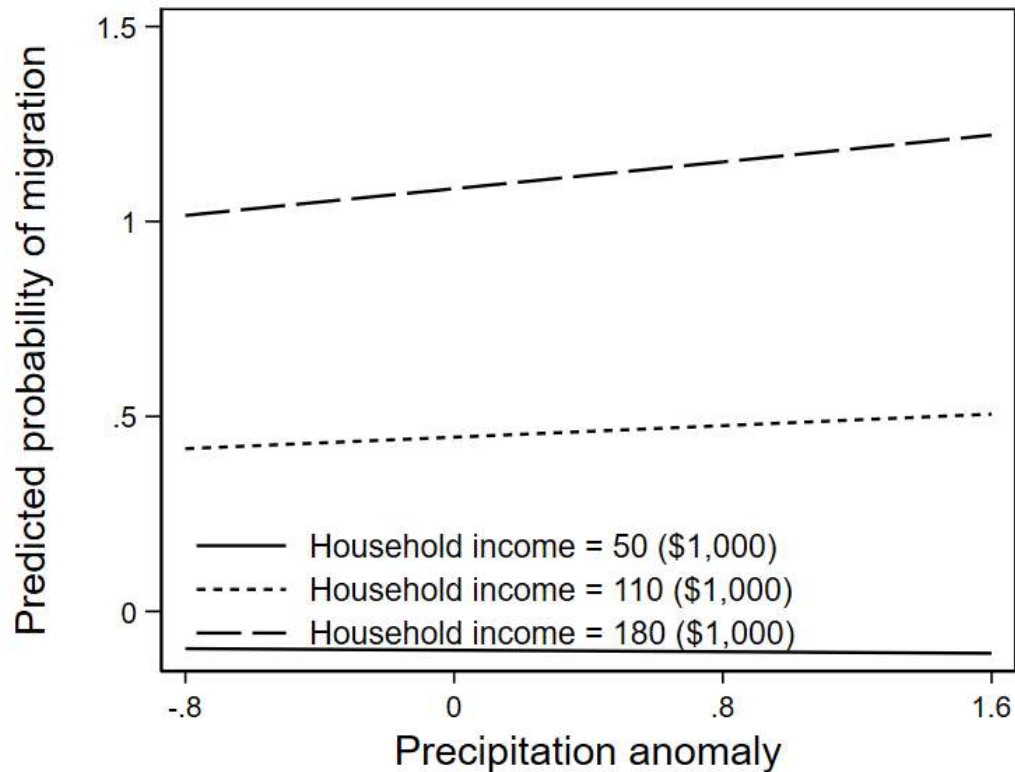


**Migration<sub>15-64</sub>**

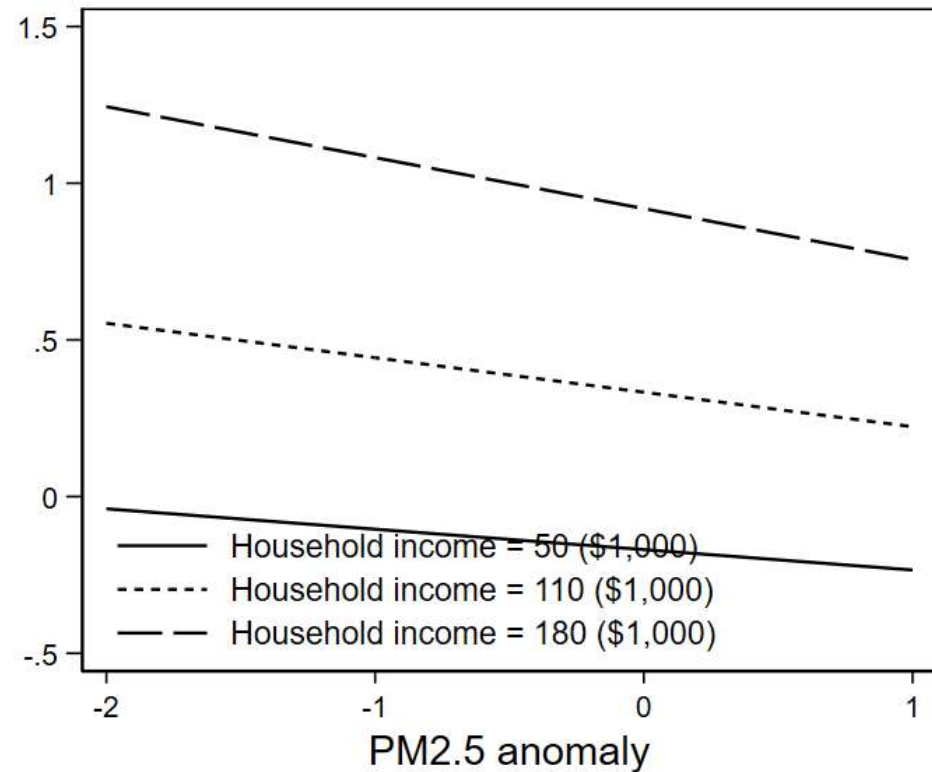


**Migration<sub>65+</sub>**

# Climate-Income interaction (cont.)



**Migration<sub>15-64</sub>**



**Migration<sub>15-64</sub>**

# Findings

- Being male, non-Hispanic white, and highly educated increased migration probability
- Precipitation and temperature anomalies generally increased migration probability, while PM2.5 and NDVI anomalies decreased migration probability
- The elder generation was responsive to temperature and environmental amenity, while the younger generation preferred places with environmental amenity, economic well-being, and affordable living costs



# Limitations

1. The ACS microdata from 2010 to 2020 were treated as cross-sectional rather than longitudinal data
2. Only 523 counties (~ 1/6 US counties) were identified through matching county and PUMA, among which 99% are metro counties



# Thank you

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# Appendixes

**Table 1:** Data sources, variables, and coding scheme

Variable	Coding scheme	Source
<b><i>Dependent variable</i></b>		
Migration status	Migrant = 1; Stayer = 0	ACS
<b><i>Level-1 variables</i></b>		
Age	Continuous variable	ACS
Personal income	Continuous variable	ACS
Gender	Female = 1; Male = 0	ACS
Marital status	Married = 1; Unmarried = 0	ACS
Race	Non-Hispanic White = 1; Non-Hispanic Black = 2; Hispanics = 3; Others = 4	ACS
Education	College and above = 1; Below college = 0	ACS
<b><i>Level-2 variables</i></b>		
Precipitation anomaly	Continuous variable	PRISM
Temperature anomaly	Continuous variable	PRISM
NDVI anomaly	Continuous variable	NCAR
PM2.5 anomaly	Continuous variable	ACAG
Household income	Continuous variable	ACS
Housing price	Continuous variable	ACS
Employment rate	Continuous variable	ACS
Homeownership	Continuous variable	ACS
Metropolitan status	Metro = 1; Nonmetro = 0	OMB

**Table 3:** Two-level logistic regression predicting migration status in the U.S., 2010-2020

	Model 1	Model 2	Model 3
<i>Level-1 variables</i>			
Age		-0.010***	-0.010***
Personal income		-0.000***	-0.001***
Gender, Male (Ref. = Female)		0.093***	0.097***
Marital status, Married (Ref. = Unmarried)		-0.078***	-0.071***
Race, NHB (Ref. = NHW)		-0.222***	-0.209***
Race, Hispanics (Ref. = NHW)		-0.326***	-0.335***
Race, Others (Ref. = NHW)		0.050***	0.022***
Education, College and above (Ref. = Below college)		0.172***	0.170***
<i>Level-2 variables</i>			
Precipitation anomaly			0.017**
Temperature anomaly			0.075***
NDVI anomaly			-0.249***
PM2.5 anomaly			-0.006***
Household income			0.010***
Housing price			0.001***
Employment rate			-0.047***
Homeownership			0.004***
Metro county (Ref. = Nonmetro county)			0.132***
Constant	0.170	0.513***	3.115***
Year effect	—	—	Controlled
County effect	—	—	Controlled
Observations	2,243,336	2,243,336	2,243,336
ICC	0.675	0.676	0.673
LR test	—	23,823***	20,932***

**Table 4:** Two-level logistic regression predicting age-specific migration status in the U.S., 2010-2020

	Mig <sub>15-64</sub>	Mig <sub>65+</sub>
<b><i>Level-1 variables</i></b>		
Age	-0.014***	-0.015***
Personal income	-0.001***	0.001***
Gender, Male (Ref. = Female)	0.104***	-0.007
Marital status, Married (Ref. = Unmarried)	-0.068***	0.083***
Race, NHB (Ref. = NHW)	-0.192***	-0.237***
Race, Hispanics (Ref. = NHW)	-0.331***	-0.227***
Race, Others (Ref. = NHW)	0.029***	-0.027
Education, College and above (Ref. = Below college)	0.170***	0.231***
<b><i>Level-2 variables</i></b>		
Precipitation anomaly	0.016**	0.033
Temperature anomaly	0.093***	-0.127*
NDVI anomaly	-0.247***	-0.193***
PM2.5 anomaly	-0.092***	-0.001
Household income	0.009***	0.011***
Housing price	0.001***	0.001***
Employment rate	-0.048***	-0.042***
Homeownership	0.006***	-0.020***
Metro county (Ref. = Nonmetro county)	0.160***	-0.128
Constant	3.123***	4.823***
Year effect	Controlled	Controlled
County effect	Controlled	Controlled
Observations	2,029,092	214,244

**Table 5:** Two-level logistic regression predicting migration status in the U.S. with interactions between environmental factors and household income, 2010-2020

	<i>Mig<sub>All</sub></i>	<i>Mig<sub>15-64</sub></i>	<i>Mig<sub>65+</sub></i>
<b><i>Level-1 variables</i></b>			
Age	-0.010***	-0.014***	-0.015***
Personal income	-0.001***	-0.001***	0.001***
Gender, Male (Ref. = Female)	0.097***	0.104***	-0.007
Marital status, Married (Ref. = Unmarried)	-0.071***	-0.068***	0.083***
Race, NHB (Ref. = NHW)	-0.209***	-0.192***	-0.237***
Race, Hispanics (Ref. = NHW)	-0.337***	-0.332***	-0.228***
Race, Others (Ref. = NHW)	0.021***	0.028***	-0.028
Education, College and above (Ref. = Below college)	0.170***	0.170***	0.230***
<b><i>Level-2 variables</i></b>			
Precipitation anomaly	-0.025	-0.04	0.152†
Temperature anomaly	0.125†	0.166*	-0.348
NDVI anomaly	-1.238***	-1.197***	-1.261***
PM2.5 anomaly	-0.015	-0.028	0.092
Household income	0.009***	0.008***	0.010***
Housing price	0.001***	0.001***	0.001***
Employment rate	-0.048***	-0.048***	-0.045***
Homeownership	0.004***	0.006***	-0.020***
Metro county (Ref. = Nonmetro county)	0.118***	0.146***	-0.137
<b><i>Interaction terms</i></b>			
Precipitation anomaly * Household income	0.001*	0.001**	-0.001
Temperature anomaly * Household income	-0.000	-0.001	0.002
NDVI anomaly * Household income	0.011***	0.011***	0.012***
PM2.5 anomaly * Household income	-0.001***	-0.001***	-0.001
Constant	3.288	3.275***	5.161***
Year effect	Controlled	Controlled	Controlled
County effect	Controlled	Controlled	Controlled
Observations	2,243,336	2,029,092	214,244
Numbers of county	523	523	523
ICC	0.670	0.668	0.675