Environmental Impacts on Mortality in the US

2010 - 2019

Shuai Zhou
Global Development, Cornell University

In collaboration with Chuan Liao, Ziqing Wei, and Guangqing Chi

Annual Meeting of the Rural Sociological Society

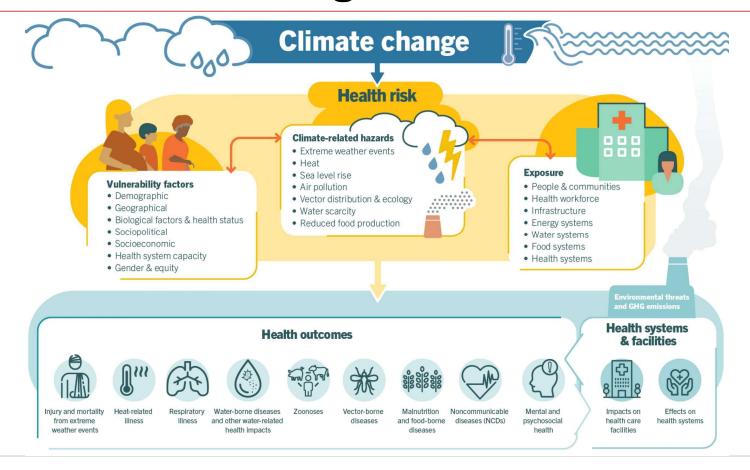
July 27th, 2024



Outline

- 1. Background
- 2. Knowledge gaps
- 3. Research objectives
- 4. Methods
- 5. Results
- 6. Discussion

Environmental change burdens health



Source: WHO

Existing studies emphasized SDOH

Five components of social determinants of health (SDOH):

- Economic stability (Murray, 2003; Sommers et al., 2017)
- Education access and quality (Vable et al., 2020)
- Health care access and quality (Herman et al., 2011)
- Neighborhood and built environment (Penney et al., 2015)
- Social and community context (Clair et al., 2021)

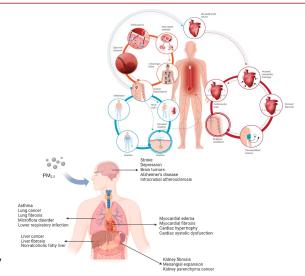




Source: OASH

Related work showed environmental impact

- Rural mortality penalty exists, with such penalty being highest in high-poverty rural communities (Cosby et al., 2019)
- Rising temperature increased mortality in both rural (James, 2014; Rhubart & Santos, 2023) and urban settings (Lowe, 2016; Cleland et al., 2023)
- Air pollution caused cardiovascular and respiratory diseases among people exposed for both short (Ye et al., 2022) and prolonged periods (Raaschou-Nielsen et al., 2023)
- Green space tended to attenuate all-cause and cardiovascular mortality (Coutts et al., 2010)





Source: The Lancet; The Innovation



What we do...

- 1. Investigate the nonlinear environmental impacts on cause- and age-specific mortality
- 2. Test whether rural mortality penalty persists across cause- and age-specific mortality



Measure mortality rates using the IHME data

- Cause-specific mortality rates
 - 1. Chronic respiratory (CR) diseases
 - 2. Cardiovascular (CV) diseases
- Age-specific mortality rates

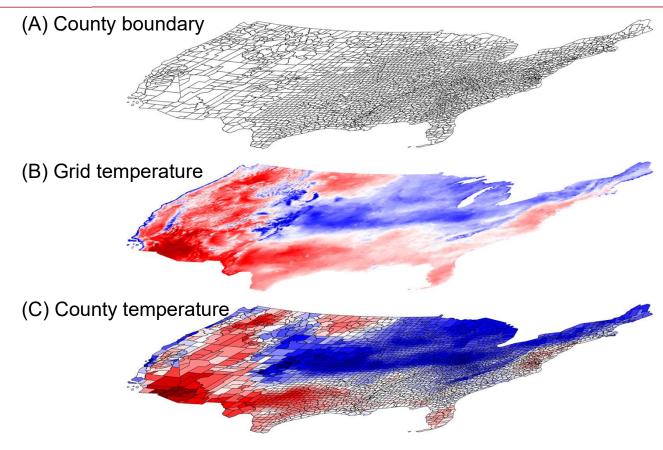
1.
$$0-1$$

$$2. 1 - 14$$

$$3. 15 - 64$$



Measure environmental factors from grid data





Fixed-effects modelling

$$Y_{i,t} = \beta_0 + \beta X_{i,t} + \gamma_i + \delta_t + \varepsilon_{i,t}$$

 $Y_{i,t}$ are the mortality of county *i* at time *t*

 $X_{i,t}$ is a matrix of environmental, socioeconomic, and demographic factors

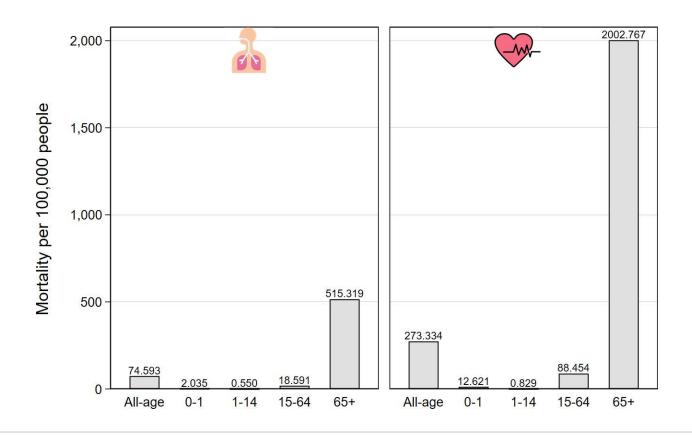
 β_0 and β are intercept and coefficient estimates, respectively

 γ_i and δ_t are county and year fixed-effects, respectively

 $\varepsilon_{i,t}$ is the error term

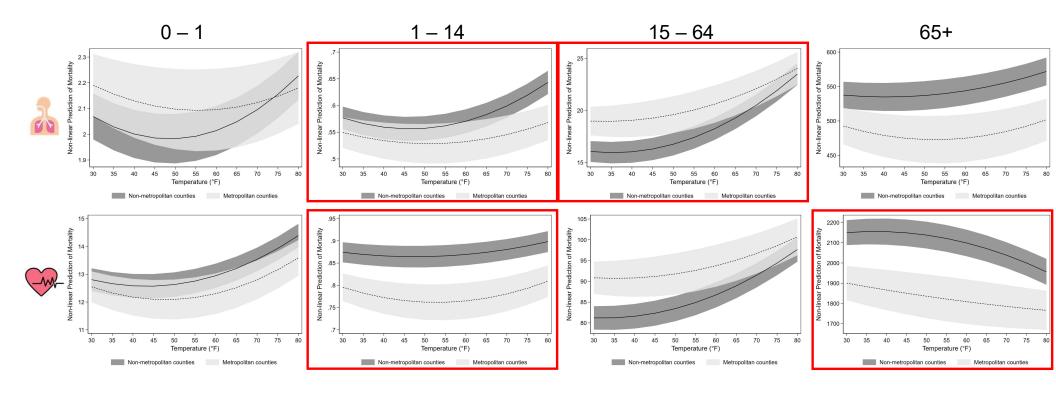


Mortality shows age/cause specific patterns

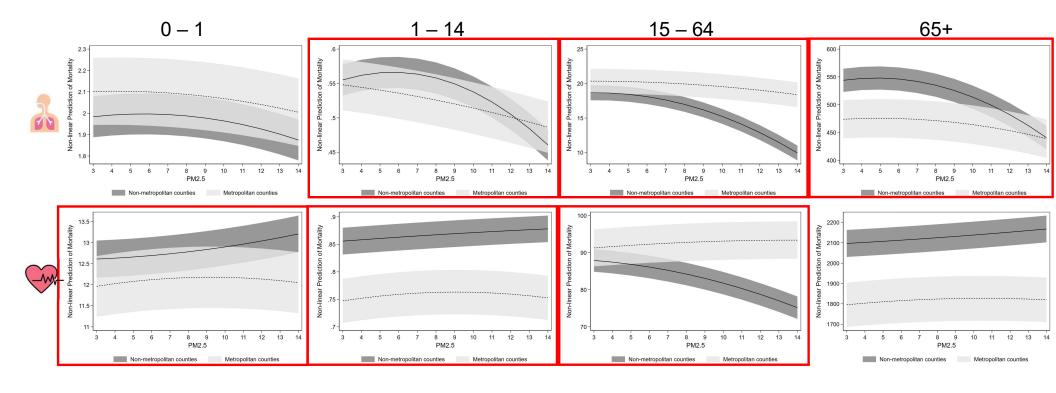




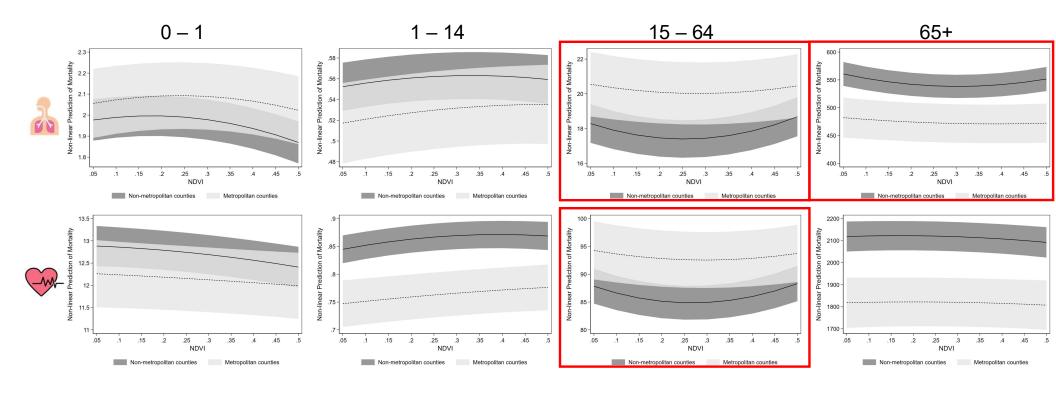
Temperature



PM2.5



NDVI



Takeaways

- 1. Temperature showed non-linear effects and a rural mortality penalty for teenagers and adults, but not for the elderly
- 2. Air quality lowered CR-related mortality but increased CV-related mortality among infants and teenagers in rural counties and working-age adults in urban counties
- 3. Greenness coverage influences mortality, but not as much as temperature and air quality did

Implications for research and policy-making

- 1. Rural mortality penalty exists but not uniformly across all ages or causes of death
- 2. Policy-making should prioritize addressing the unique health challenges specific to different age groups and regions
- 3. Additional unobserved behavioral, pathological, and context-specific factors deserve further investigation
 - Exposure avoidance
 - Early warning system
 - Advanced medical treatment
 - Pathological differences between CR and CV diseases

Thank you

Questions? Email Dr. Shuai Zhou at sz675@cornell.edu



Appendix



Data sources

Variable	Data source
Dependent variable	
Cause-specific mortality rates	Institute for Health Metrics and Evaluation
Environmental factors	
Temperature	Parameter-elevation Regressions on Independent Slopes Model
PM _{2.5}	Atmospheric Composition Analysis Group
NDVI	National Center for Atmospheric Research
Sociodemographic factors	
Household income	American Community Survey
% College graduate	American Community Survey
% Health insurance	American Community Survey
County metropolitan status	The US Office of Management and Budget

Descriptive statistics

	N	mean	SD	Min	Max
CR _{AII}	33,924	74.593	19.003	18.146	187.262
CR ₀₋₁	33,924	2.035	0.593	0.615	5.075
CR ₁₋₁₄	33,924	0.550	0.181	0.174	1.531
CR ₁₅₋₆₄	33,924	18.591	7.922	3.912	76.537
CR ₆₅₊	33,924	515.319	133.552	161.624	1,335.552
CVAII	33,924	273.334	54.834	73.103	638.873
CV ₀₋₁	33,924	12.621	2.979	3.262	25.581
CV ₁₋₁₄	33,924	0.829	0.236	0.302	2.062
CV ₁₅₋₆₄	33,924	88.454	32.708	20.834	269.208
CV ₆₅₊	33,924	2,002.767	386.100	652.433	4,912.635
Temperature	34,199	54.945	8.406	31.747	78.809
PM2.5	34,199	7.343	1.696	2.968	14.016
NDVI	34,199	0.239	0.059	0.060	0.523
Income	34,192	65.675	16.103	30.949	184.416
College	34,193	49.951	10.903	6.283	93.302
Insurance	34,193	85.298	6.055	57.300	97.900
Metropolitan county	34,199	0.373	0.484	0.000	1.000



Regression table

$ \begin{array}{c c c c c c c c c c c c c c c c c c c $	0	1589 X	NACOTO:	90900	92300	1570100	10100	No.	1400000	20.000	CONTROL CONTRO
Temperature		(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
Temperature	2	CR_{All}	CV_{All}	CR_{0-1}	CV_{0-1}	CR_{1-14}	CV_{1-14}	CR ₁₅₋₆₄	CV ₁₅₋₆₄	CR ₆₅₊	CV ₆₅₊
Temperature2	Environmental factors										
PM2.5 1.835*** 0.943*** 0.002*** 0.002*** 0.003** 0.502*** 0.041 12.146*** 4.113 PM2.52 -0.025*** -0.073**** -0.002*** -0.000*** -0.007*** -0.071*** -1.266*** 0.139 NDVI 27.337*** -9.702 0.435*** -1.892 -0.135** -0.228*** 21.252*** 63.64*** 34.6655*** -276.673 NDVI2 46.061*** 11.358 -1.222*** -1.392 -0.135** -0.228*** 21.252*** 63.64*** 34.6655*** -276.673 Interactions Temperature * Metro -0.137 -0.478 0.008 -0.022 0.002** -0.001** 0.111* 0.113 -2.121* -12.770**** PM2.5 * Metro 1.165**** 0.983 -0.012 0.105*** -0.000*** -0.001*** -0.002 0.015 0.127**** PM2.5 * Metro 1.185*** 0.933 -0.012 0.105*** -0.006*** 0.060**** 0.457** -7.735*** 7.845 <td>Temperature</td> <td>-0.558***</td> <td>0.266</td> <td>-0.024***</td> <td>-0.117***</td> <td></td> <td>-0.003***</td> <td>-0.268***</td> <td>-0.469***</td> <td>-1.947***</td> <td>8.066***</td>	Temperature	-0.558***	0.266	-0.024***	-0.117***		-0.003***	-0.268***	-0.469***	-1.947***	8.066***
PM2.52 NDVI -0.025**** -0.073**** -9.702 0.03*** -0.288 0.090** -0.178*** -1.081**** -1.266*** -1.266*** -1.266*** -2.10.966*** -2.10.900* 0.139 NDVI -27.337*** -9.702 0.435*** -0.288 0.090** -0.135** -10.814*** -33.923*** -210.966*** -210.966*** -276.673 NDVI? 46.061*** -13.58 1.222*** -1.392 -0.135** -0.228*** -0.228*** -0.225*** -0.215*** -0.225** -0.225**	Temperature ²	0.006***	-0.005*	0.000***	0.001***		0.000***	0.004***	0.007***	0.024***	-0.109***
NDVI NDVI A6.061** 11.358 -1.222** -1.392 -0.135** -0.228** 21.252*** 63.614*** 346.655*** 93.090 NDVI A6.061** 11.358 -1.222** -1.392 -0.135** -0.228*** 21.252*** 63.614*** 346.655*** -276.673 NDVI NDVI Metro -0.137 -0.478 0.008 -0.022 0.002* -0.0004** 0.111* 0.113 -2.121* -12.770*** NDVI NDVI Metro -1.165*** 0.983 -0.012 0.105*** -0.000** 0.000** -0.000** 0.005** -0.406*** 0.457* -7.735*** 7.845 NDVI Metro 0.138*** 0.03 0.001* -0.009** 0.001** -0.000** 0.000** 0.060*** 0.060*** 0.052** 0.820*** -0.710** NDVI Metro 18.058** 19.244 0.059 -0.162 -0.001 -0.084 5.367* 17.002** -35.008** -246.318** 112.999 NDVI Metro -0.083*** -0.137*** -0.000 0.005** -0.0000** -0.000** -0.000** -0.000** -0.000** -0.065** -2.46.318** 112.999 NDVI NDVI -0.084 -0.083** -0.137** -0.001** -0.094** -0.001** -0.009** -0.005** -0.005** -0.000** -0.005** -0.000	PM2.5	1.835***	0.943**	0.020***	-0.002	0.017***	0.003**	0.502***	0.041	12.146***	4.113
NDVI2 A6.061*** 11.358	$PM2.5^{2}$	-0.205***	-0.073***		0.003**	-0.002***		-0.076***	-0.071***	-1.266***	0.139
Temperature Metro -0.137 -0.478 0.008 -0.022 0.002* -0.004** 0.111* 0.113 -2.121* -12.770*** Temperature Metro 0.001 0.004 -0.000* 0.000 0.000** 0.000** 0.001** -0.001** -0.002 0.015 0.127*** PM2.5 * Metro -1.165*** 0.983 -0.012 0.105*** -0.002** 0.005** -0.406*** 0.457* -7.735*** 7.845 PM2.52 * Metro 0.138*** 0.003 0.001** -0.009*** 0.001*** -0.000*** 0.060*** 0.052*** 0.820*** -0.710** NDV1 * Metro 18.058** 19.244 0.059 -0.162 -0.001 -0.084 5.367* 17.002** 133.751** -27.376 NDV1 * Metro -36.855** -34.648 0.193 1.116 0.047 0.175 -11.720** -35.008** -246.318** 112.999 Timcome -0.083*** -0.137*** -0.000 0.005*** -0.000*** -0.000*** -0.040*** -0.065*** -0.402*** -0.688*** College -0.071*** -0.297*** -0.004*** -0.003** -0.000*** -0.001*** -0.009*** -0.100*** -0.101*** -0.433*** -2.327*** Insurance -0.018 -0.231*** -0.000 -0.003* -0.000** -0.000** -0.000*** -0.100*** -0.101*** -0.433*** -2.939*** Age-specific population Yes Ye	NDVI	-27.337***	-9.702			0.090**	0.178***	-10.814***	-33.923***	-210.969***	93.090
Temperature * Metro	NDVI ²	46.061***	11.358	-1.222***	-1.392	-0.135**	-0.228***	21.252***	63.614***	346.655***	-276.673
Temperature ² * Metro	Interactions										
Temperature ² * Metro	Temperature * Metro	-0.137	-0.478	0.008	-0.022	0.002*	-0.004**	0.111*	0.113	-2.121*	-12.770***
PM2.5 * Metro -1.165*** 0.983 -0.012 0.105*** -0.020*** 0.005** -0.406*** 0.457* -7.735*** 7.845 PM2.5² * Metro 0.138*** 0.003 0.001* -0.009*** 0.001*** -0.000*** 0.060*** 0.052*** 0.820*** -0.710** NDVI * Metro 18.058** 19.244 0.059 -0.162 -0.001 -0.084 5.367* 17.002** 133.751** -27.376 NDVI2 * Metro -36.855** -34.648 0.193 1.116 0.047 0.175 -11.720** -35.008** -246.318** 112.999 Controls Income -0.083*** -0.137*** -0.000 0.005*** -0.000*** 0.000*** -0.040*** -0.065*** -0.402*** -0.688*** College -0.071*** -0.000*** -0.003** -0.001*** -0.001*** -0.009** -0.100*** -0.617*** -2.327*** Insurance -0.018 -0.231*** -0.000 -0.03** -0.000** -0.00*** 0.01***		0.001	0.004	-0.000*	0.000	-0.000**	0.000**	-0.001**	-0.002	0.015	0.127***
NDVI * Metro		-1.165***	0.983	-0.012	0.105***	-0.020***	0.005**	-0.406***	0.457*	-7.735***	7.845
NDVI² * Metro -36.855** -34.648 0.193 1.116 0.047 0.175 -11.720** -35.008** -246.318** 112.999 Controls Income -0.083*** -0.137*** -0.000 0.005*** -0.000*** 0.000*** -0.040*** -0.065*** -0.402*** -0.688*** College -0.071*** -0.297*** -0.004*** -0.023*** -0.001*** -0.001*** -0.009** -0.100*** -0.617*** -2.327*** Insurance -0.018 -0.231*** -0.000 -0.003* -0.000*** -0.000*** 0.101*** -0.433*** -2.939*** Age-specific population Yes Yes <t< td=""><td>PM2.52 * Metro</td><td>0.138***</td><td>0.003</td><td>0.001*</td><td>-0.009***</td><td>0.001***</td><td>-0.000***</td><td>0.060***</td><td>0.052***</td><td>0.820***</td><td>-0.710**</td></t<>	PM2.52 * Metro	0.138***	0.003	0.001*	-0.009***	0.001***	-0.000***	0.060***	0.052***	0.820***	-0.710**
Controls Income -0.083*** -0.137*** -0.000 0.005*** -0.000*** 0.000*** -0.040*** -0.065*** -0.402*** -0.688*** College -0.071*** -0.297*** -0.004*** -0.023*** -0.001*** -0.009** -0.100*** -0.617*** -2.327*** Insurance -0.018 -0.231*** -0.000 -0.003* -0.000** -0.000*** 0.078*** 0.101*** -0.433*** -2.939*** Age-specific population Yes Yes <td>NDVI * Metro</td> <td>18.058**</td> <td>19.244</td> <td>0.059</td> <td>-0.162</td> <td>-0.001</td> <td>-0.084</td> <td>5.367*</td> <td>17.002**</td> <td>133.751**</td> <td>-27.376</td>	NDVI * Metro	18.058**	19.244	0.059	-0.162	-0.001	-0.084	5.367*	17.002**	133.751**	-27.376
Income -0.083*** -0.137*** -0.000 0.005*** -0.000*** 0.000*** -0.040*** -0.065*** -0.402*** -0.688*** College -0.071*** -0.297*** -0.004*** -0.023*** -0.001*** -0.009** -0.100*** -0.617*** -2.327*** Insurance -0.018 -0.231*** -0.000 -0.003* -0.000** -0.000*** 0.078*** 0.101*** -0.433*** -2.939*** Age-specific population Yes	NDVI ² * Metro	-36.855**	-34.648	0.193	1.116	0.047	0.175	-11.720**	-35.008**	-246.318**	112.999
College -0.071*** -0.297*** -0.004*** -0.023*** -0.001*** -0.009** -0.100*** -0.617*** -2.327*** Insurance -0.018 -0.231*** -0.000 -0.003* -0.000** -0.000*** 0.078*** 0.101*** -0.433*** -2.939*** Age-specific population Yes Yes </td <td>Controls</td> <td></td>	Controls										
Insurance -0.018 -0.231*** -0.000 -0.003* -0.000*** -0.000*** 0.078*** 0.101*** -0.433*** -2.939*** Age-specific population Yes	Income	-0.083***	-0.137***	-0.000	0.005***	-0.000***	0.000***	-0.040***	-0.065***	-0.402***	-0.688***
Age-specific population Yes	College	-0.071***	-0.297***	-0.004***	-0.023***	-0.001***	-0.001***	-0.009**	-0.100***	-0.617***	-2.327***
Year effect Yes Yes <th< td=""><td>Insurance</td><td>-0.018</td><td>-0.231***</td><td>-0.000</td><td>-0.003*</td><td>-0.000**</td><td>-0.000***</td><td>0.078***</td><td>0.101***</td><td>-0.433***</td><td>-2.939***</td></th<>	Insurance	-0.018	-0.231***	-0.000	-0.003*	-0.000**	-0.000***	0.078***	0.101***	-0.433***	-2.939***
County effect Yes <	Age-specific population	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Constant 101.016*** 328.359*** 2.680*** 16.932*** 0.731*** 1.015*** 19.665*** 100.148*** 664.701*** 2,446.170*** Observations 33,839 33,839 33,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 30,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 30,830 30,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 33,830 30,830 33,830 30,760 30,76 3,076	Year effect	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations 33,839 33,839 33,830 3	County effect	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations 33,839 33,839 33,830 3	Constant	101.016***	328.359***	2.680***	16.932***	0.731***	1.015***	19.665***	100.148***	664.701***	2,446.170***
R-squared 0.353 0.389 0.369 0.504 0.463 0.657 0.596 0.386 0.351 0.540 Number of counties 3,078 3,078 3,076 3,076 3,076 3,076 3,076 3,076 AIC 157,997 225,600 -76,944 26,158 -174,677 -169,649 86,908 156,934 287,240 365,462											
Number of counties 3,078 3,078 3,076 3,076 3,076 3,076 3,076 3,076 3,076 AIC 157,997 225,600 -76,944 26,158 -174,677 -169,649 86,908 156,934 287,240 365,462	Observations	33,839	33,839	33,830	33,830	33,830	33,830	33,830	33,830	33,830	33,830
AIC 157,997 225,600 -76,944 26,158 -174,677 -169,649 86,908 156,934 287,240 365,462	R-squared	0.353	0.389	0.369	0.504	0.463	0.657	0.596	0.386	0.351	0.540
	Number of counties	3,078	3,078	3,076	3,076	3,076	3,076	3,076	3,076	3,076	3,076
BIC 158,225 225,828 -76,717 26,386 -174,450 -169,422 87,136 157,161 287,468 365,690	AIC	157,997	225,600	-76,944	26,158	-174,677	-169,649	86,908	156,934	287,240	365,462
	BIC	158,225	225,828	-76,717	26,386	-174,450	-169,422	87,136	157,161	287,468	365,690

***p < 0.001, **p < 0.01, *p < 0.05. AIC = Akaike's information criterion; BIC = Schwartz's Bayesian information criterion

