Appendix Table 1. NCHS Urban-Rural Classification Scheme.

Appendix Table 2. Number and percent of counties with suicide rates >20 per 100,000, by urban-rural classification, 2005 and 2015.

Appendix Table 3. Model covariates, data sources, and specifications.

Appendix Figures 1–11. Annual model-based county-level estimates of suicide rates across the U.S., 2005–2015.

Appendix Figure 12. Deciles of model-based suicide rates in 2005 and 2015, U.S. Decile 1 corresponds to the lowest suicide rates, while Decile 10 corresponds to the highest.

Appendix Figure 13. Coefficients of variation (i.e., relative standard error/deviation) of countylevel suicide rates in the U.S., 2005 (top) and 2015 (bottom).

Appendix Figure 14. Percentage of counties with greater than 20%, 30% or 40% increases in model-based suicide rates over the time period from 2005–2015, by urban-rural classification.

Urban-rural	Definition
classification	
Metropolitan (urban)	
Large central metro (i.e., most urban)	Counties located in a metropolitan statistical area (MSA) of at least 1 million population and meet one of three conditions: (1) contain the entire population of the largest principal city of the MSA, (2) are completely contained within the largest principal city of the MSA, or (3) contain at least 250,000 inhabitants of any principal city of the MSA.
Large fringe metro (i.e., suburban)	Counties in MSAs of 1 million or more population that do not qualify as large central metro counties
Medium metro	Counties in MSAs of populations between 250,000 and 999,999
Small metro	Counties in MSAs of fewer than 250,000 population
Nonmetropolitan	
Micropolitan	Counties in micropolitan statistical areas (population 10,000–49,999)
Noncore (i.e., most rural)	Nonmetropolitan counties that do not qualify as micropolitan
	s determined using the NCHS Urban-Rural Classification Scheme for

Appendix Table 1. NCHS Urban-Rural Classification Scheme^a

^aUrbanization level was determined using the NCHS Urban-Rural Classification Scheme for Counties from 2006 (applied to data years 2005–2012) and 2013 (applied to data years 2013–2015).

NCHS, National Center for Health Statistics

Urban-rural classification		number unties	counti mode suicide	ber of es with I-based rates >20 00,000	model suicide	ies with -based rates>20 ,000, %
	2005	2015	2005	2015	2005	2015
Metropolitan (urban)						
Large central metro (most urban)	63	68	0	2	0.0	2.9
Large fringe metro (suburban)	354	368	5	53	1.4	14.4
Medium metro	332	372	9	83	2.7	22.3
Small metro	340	358	17	96	5.0	26.8
Non-metropolitan						
Micropolitan	694	641	49	181	7.1	28.2
Noncore (most rural)	1,357	1,333	175	600	12.9	45.0
Total	3,140	3,140	255	1,015	8.1	32.3

Appendix Table 2. Number and Percent of Counties With Suicide Rates >20 per 100,000,^a by Urban-Rural Classification,^b 2005 and 2015

^aBased on model-based suicide rates, with 20 corresponding roughly to the 90th percentile in 2005.

^bUrbanization level was determined using the NCHS Urban-Rural Classification Scheme for Counties from 2006 (applied to data years 2005–2012) and 2013 (applied to data years 2013–2015).

NCHS, National Center for Health Statistics

Covariate	Data source
Land area (square mile)	Area Resource Files, ¹ Decennial
	Census
Mean household size	Area Resource Files, ¹ Decennial
	Census
Housing unit density per square mile	Area Resource Files, ¹ Decennial
	Census
Median age	Area Resource Files, ¹ Decennial
	Census
Median household income	Area Resource Files, ¹ American
	Community Survey (ACS)
Median home value	Area Resource Files, ¹ ACS
Median gross rent	Area Resource Files, ¹ ACS
% Housing units with more than 1 person/room	Area Resource Files, ¹ ACS
Percent American Indian or Alaska Native	Area Resource Files, ¹ Decennial
	Census
Median per capita income	Area Resource Files, ¹ Regional
	Economic Information (REIS)
Percent Asian	Area Resource Files, ¹ Decennial
	Census
Percent non-Hispanic black	Area Resource Files, ¹ Decennial
-	Census
% Persons aged ≥ 25 years with ≥ 4 years college	Area Resource Files, ¹ ACS
Percent of females divorced	Area Resource Files, ¹ ACS
Percent female headed household	Area Resource Files, ¹ Decennial
	Census
Percent foreign born	Area Resource Files, ¹ ACS
Percent Hispanic	Area Resource Files, ¹ Decennial
	Census
% Persons aged ≥ 25 years with high school diploma	Area Resource Files, ¹ ACS
% Persons aged ≥ 25 years with less than high school	Area Resource Files, ¹ ACS
diploma	
Percent in owner-occupied housing	Area Resource Files, ¹ Decennial
	Census
Percent below the poverty level	Area Resource Files, ¹ Census SAIPE
Percent rural population	Area Resource Files, ¹ Decennial
	Census
Percent urban population	Area Resource Files, ¹ Decennial
	Census

Appendix Table 3. Model Covariates, Data Sources

Percent non-Hispanic white	Area Resource Files, ¹ Decennial
r creent non-rinspanie winte	Census
Percent workers in	Area Resource Files, ¹ ACS
agriculture/forestry/fishing/hunting/mine work	med Resource Thes, mes
Percent workers in construction	Area Resource Files, ¹ ACS
Percent workers in education/healthcare/social	Area Resource Files, ¹ ACS
assistance	
Percent workers in manufacturing	Area Resource Files, ¹ ACS
Percent workers in other industries	Area Resource Files, ¹ ACS
Unemployment rate, ≥ 16	Area Resource Files, ¹ Bureau of
	Labor Stats (BLS)
Veteran population estimate	Area Resource Files, ¹ Department of
	Veteran's Affairs
Number of violent crimes	Uniform Crime Reporting Program
	Data: County-Level Detailed Arrest
	and Offense Data ²
Number of property crimes	Uniform Crime Reporting Program
	Data: County-Level Detailed Arrest
	and Offense Data ²
Number of drug abuse violations-total	Uniform Crime Reporting Program
	Data: County-Level Detailed Arrest
	and Offense Data ²
All other offenses except traffic	Uniform Crime Reporting Program
	Data: County-Level Detailed Arrest
	and Offense Data ²
Estimated number of foreclosure starts divided by	Housing and Urban Development
number of mortgages times 100	Small Area Foreclosure Rates ³
U.S. Postal Service data from June 2008 on residential	Housing and Urban Development
addresses vacant 90-days or longer	Small Area Foreclosure Rates ³
Estimated high cost loan rate: percent of loans shown	Housing and Urban Development
to be high cost according to Home Mortgage	Small Area Foreclosure Rates ³
Disclosure Act data	
Illicit drug or alcohol abuse/dependence in past year	National Survey of Drug Use and
	Health Substate Estimates ⁴
Any mental illness in past year	National Survey of Drug Use and
	Health Substate Estimates ⁴
Nonmedical use of pain relievers in the past year	National Survey of Drug Use and
among individuals aged ≥ 12 years	Health Substate Estimates ⁴
Major depressive episode in the past year among	National Survey of Drug Use and
adults aged ≥ 18 years	Health Substate Estimates ⁴

Marijuana use in the past year among individuals aged ≥ 12 yearsNational Survey of Drug Use and Health Substate Estimates4Serious mental illness in the past year among adults aged ≥ 18 yearsNational Survey of Drug Use and Health Substate Estimates4Had serious thoughts of suicide in the past year among adults aged ≥ 18 yearsNational Survey of Drug Use and Health Substate Estimates4Illicit drug use in the past month among individuals aged ≥ 12 yearsNational Survey of Drug Use and Health Substate Estimates4Tobacco product use in the past month among individuals aged ≥ 12 yearsNational Survey of Drug Use and Health Substate Estimates4Treatment gap for alcohol – Needing but not receiving individuals aged ≥ 12 yearsNational Survey of Drug Use and Health Substate Estimates4Treatment for alcohol use in the past year among individuals aged ≥ 12 yearsNational Survey of Drug Use and Health Substate Estimates4Treatment for alcohol use in the past year among individuals aged ≥ 12 yearsNational Survey of Drug Use and Health Substate Estimates4
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Treatment gap for alcohol – Needing but not receiving treatment for alcohol use in the past year amongNational Survey of Drug Use and Health Substate Estimates ⁴
treatment for alcohol use in the past year among Health Substate Estimates ⁴
individuals aged ≥ 12 years
Treatment gap for drug use – Needing but not National Survey of Drug Use and
<i>receiving treatment for illicit drug use in the past year</i> Health Substate Estimates ⁴
among individuals aged ≥ 12 years
Percent of deaths with underlying cause of R99 code Mortality Data ⁵
(state-level): ill-defined and unknown cause of
mortality
Percent of deaths pending (state-level) Mortality Data ⁵
Model-based age adjusted death rates due to drug Model-Based Drug Poisoning
poisoning Estimates ⁶

MODEL SPECIFICATIONS

Hierarchical Bayesian models were implemented incorporating fixed covariates and spatiotemporal random effects. The posterior distributions were estimated via Integrated Nested Laplace Approximation (INLA) in R (<u>www.r-inla.org/</u>). These models borrow strength across both counties and years to produce smoothed annual county-level suicide rates. The model is described below, for i = 1, ..., m counties and t = 1, ..., T years:

logit $(p_{it}) = \alpha_0 + X_{it}' \boldsymbol{\beta} + u_i + v_i + \varphi_{1t} + \psi_{it}$

This model includes:

- a) A logit link function log $(p_{it}/(1-p_{it}))$; where, p_{it} is the probability of suicides in county *i* and time *t*.
- b) An overall intercept term α_0 . The intercept, α_0 was assigned a flat prior: $P(\alpha_0) \propto constant$, (where P indicates probability).
- c) A set of fixed effects, $X_{it}'\beta$, where X_{it} : is the *i* th row and *t* th column of the covariates matrix **X** and β is a vector of regression parameters. The β for fixed effects $(X_{it}'\beta)$ were assigned Normal priors. $\beta \sim N(0,100)$
- d) A spatial random effect, u_i , to account for county-level spatial dependence (e.g., clustering). This term was modeled using conditionally autoregressive priors (CAR)⁷ where weights were assigned to each county according to adjacency; neighboring counties receive a weight of one while non-neighboring counties receive a weight of zero. Delaunay triangulation was used to establish spatial weights. This method generates Voronoi triangles from county centroids, where nodes connected by a triangle edge are considered neighbors.⁸ Each county has at least one neighbor, and the number of neighbors is determined empirically based on the spatial distribution of the counties. The conditional precision of the spatial random effect was assigned $\tau_u \sim$ Gamma (1, 0.001) prior.
- e) A non-spatial random effect, v_i , to account for residual county-level variation that is not spatially dependent. This term was assigned a Normal prior, $v_i \sim N(0, \frac{1}{\tau_n})$, with

precision, τ_{ν} . The conditional precision of the non-spatial random effect was assigned $\tau_{\nu} \sim$ Gamma (1, 0.001) prior.

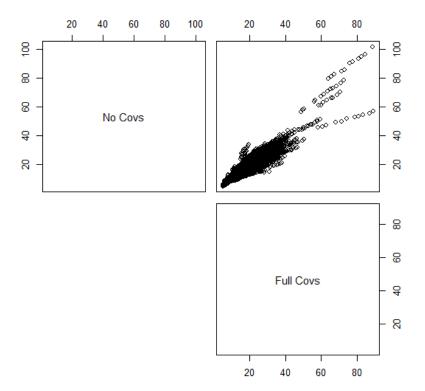
f) An overall temporal random effect, φ_{1t} , to account for temporal autocorrelation. This term was modeled via a first order random walk,⁹ where the values in a given year depend upon the values observed in the prior year plus a residual error term, $\varphi_{1t} \sim N(\varphi_{1,t-1}, \frac{1}{\tau_{\varphi_1}})$. The conditional precision of the random effect was assigned $\tau_{\varphi_1} \sim Gamma (1, 0, 001)$ prior

Gamma (1, 0.001) prior.

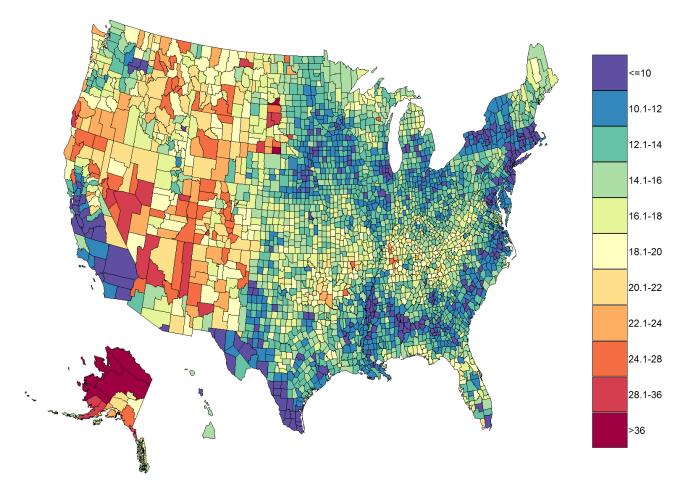
g) A space-time interaction term, ψ_{it} , which is a county- and year- specific random effect included to account for any residual spatiotemporal variation that was not captured by the spatial or temporal main effects. This term was assumed to be independently and identically distributed,^{10,11} $\psi_{it} \sim N(0, \frac{1}{\tau_{\psi}})$. The conditional precision of the random effect was assigned $\tau_{\psi} \sim$ Gamma (1, 0.001) prior.

The Deviance Information Criterion (DIC)¹² was used to determine the best fitting model among a larger set of models including the above terms and a variety of covariates and subsets of covariates. Sensitivity analyses were also conducted to examine the effect of different priors and different spatial weighting schemes. Residuals, posterior predicted means, and 95% credible

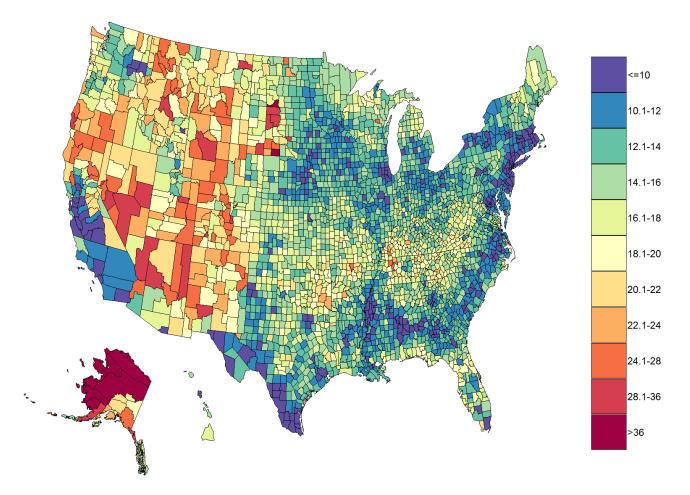
intervals were examined. Additionally, model-based estimates were aggregated by year, state, and urban-rural classification and compared with direct estimates. Posterior predictions from models with no covariates were highly correlated with the predictions from models with the full set of covariates described above ($R^2=0.88$), but the full model did have a substantially lower DIC. The below scatter plot shows the estimates from the model without covariates and the full model presented in the manuscript. More detail can be found in Khan et al.¹³



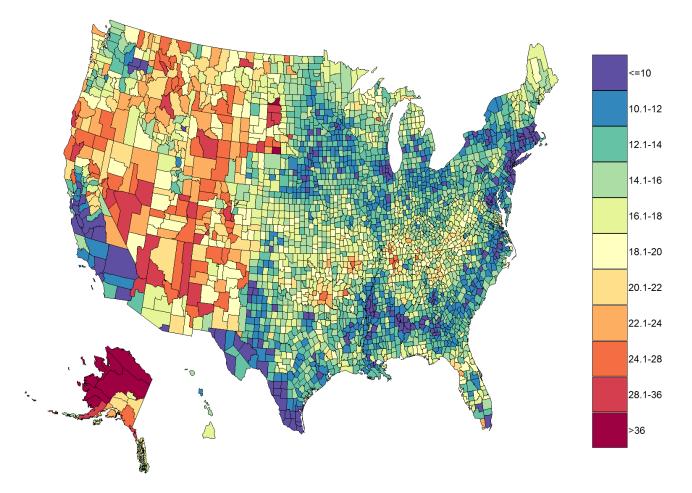
Appendix Figure 1. Model-based estimates of county-level suicide rates in the U.S., 2005.



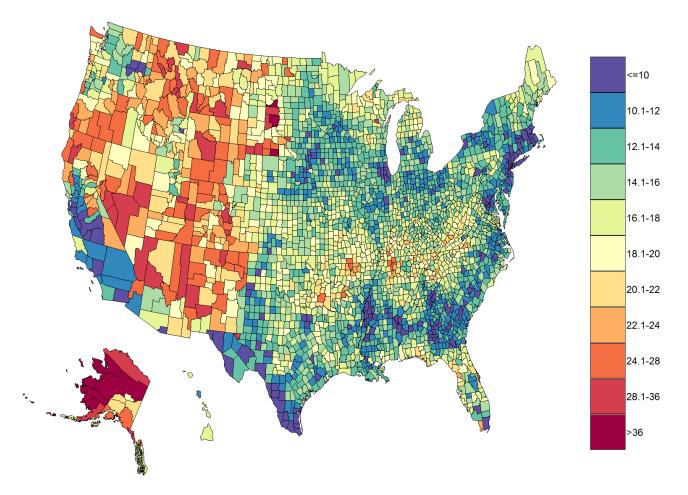
Appendix Figure 2. Model-based estimates of county-level suicide rates in the U.S., 2006.



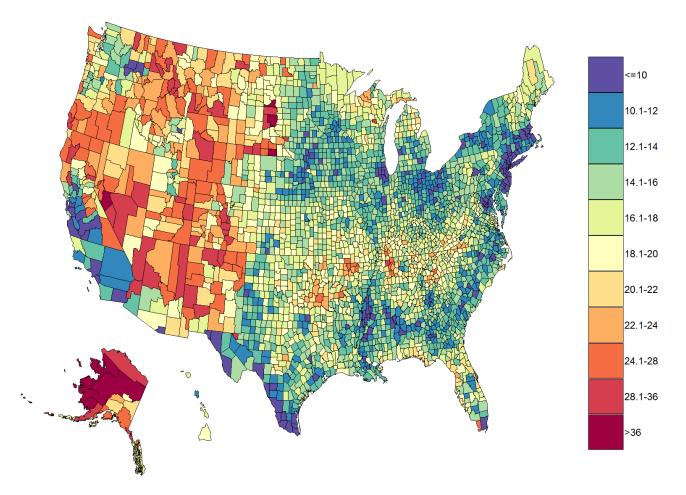
Appendix Figure 3. Model-based estimates of county-level suicide rates in the U.S., 2007.



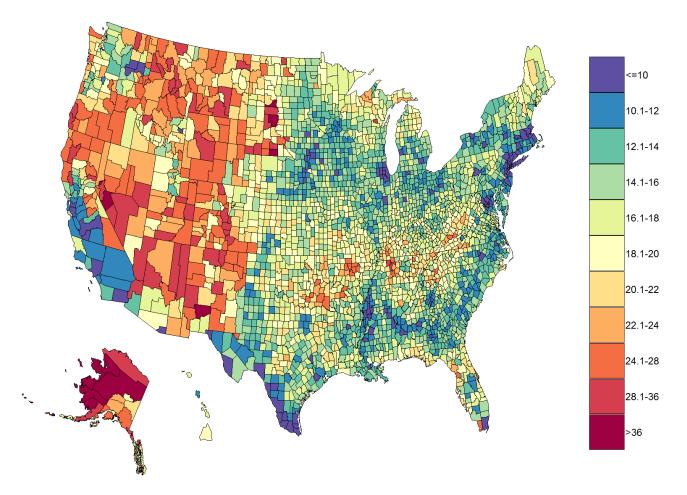
Appendix Figure 4. Model-based estimates of county-level suicide rates in the U.S., 2008.



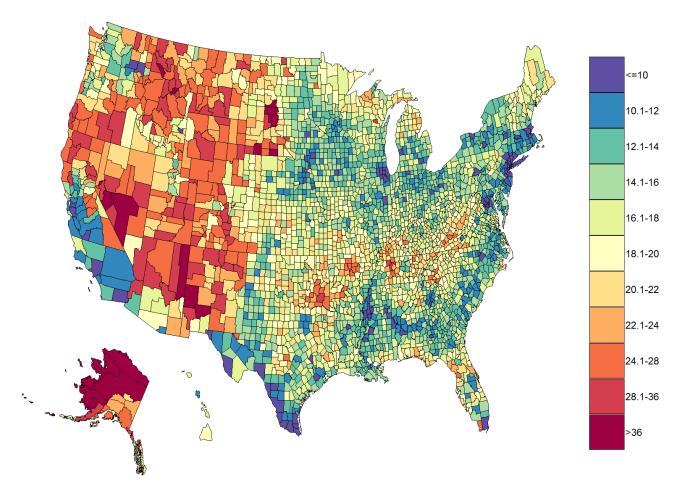
Appendix Figure 5. Model-based estimates of county-level suicide rates in the U.S., 2009.



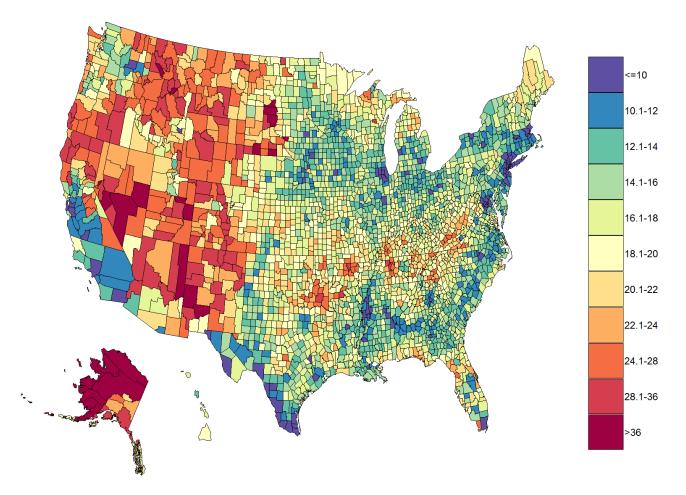
Appendix Figure 6. Model-based estimates of county-level suicide rates in the U.S., 2010.



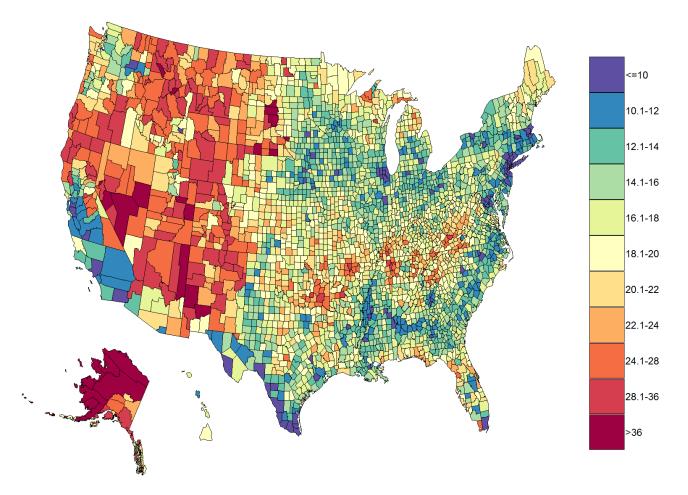
Appendix Figure 7. Model-based estimates of county-level suicide rates in the U.S., 2011.



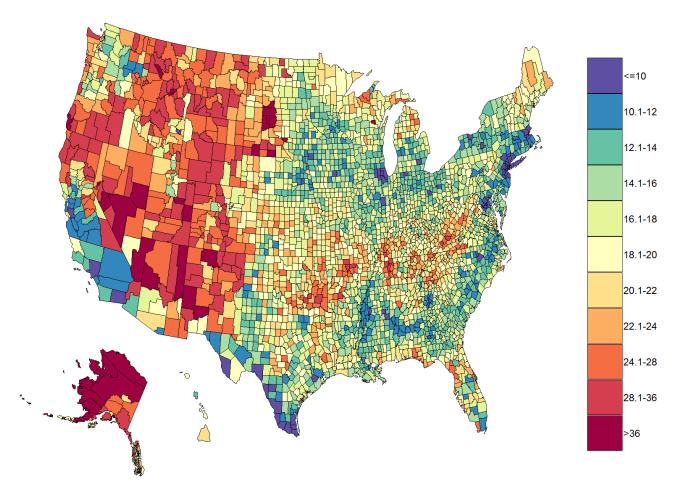
Appendix Figure 8. Model-based estimates of county-level suicide rates in the U.S., 2012.



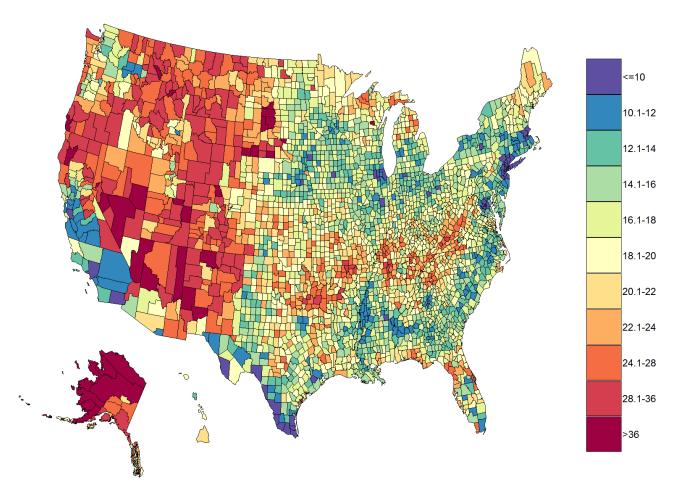
Appendix Figure 9. Model-based estimates of county-level suicide rates in the U.S., 2013.



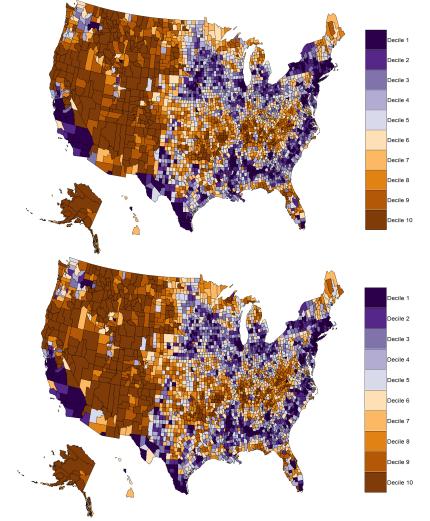
Appendix Figure 10. Model-based estimates of county-level suicide rates in the U.S., 2014.



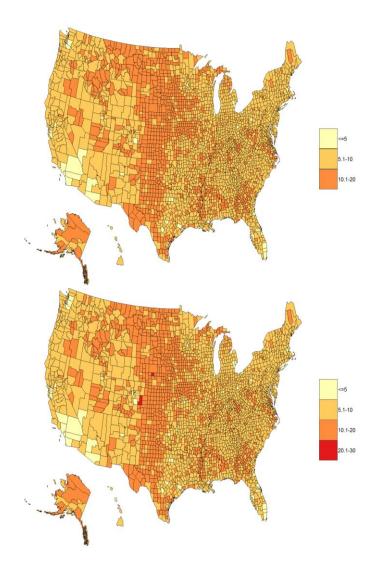
Appendix Figure 11. Model-based estimates of county-level suicide rates in the U.S., 2015.



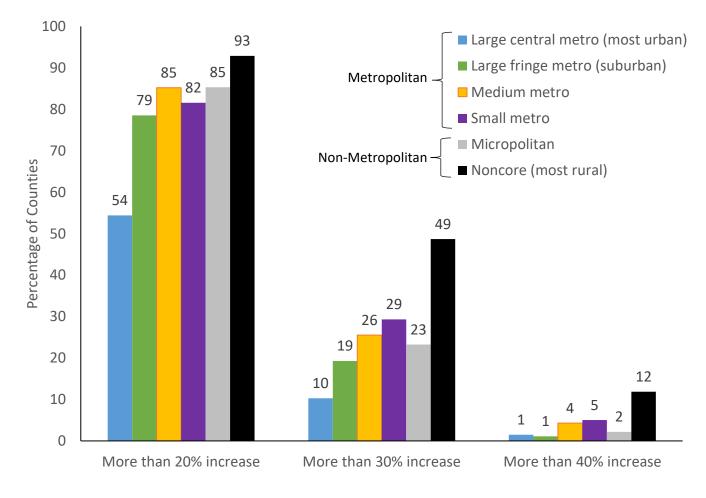
Appendix Figure 12. Deciles of model-based suicide rates in 2005 (top) and 2015 (bottom), U.S. Decile 1 corresponds to the lowest suicide rates, while Decile 10 corresponds to the highest.



Appendix Figure 13. Coefficients of variation (i.e., relative standard error/deviation) of countylevel suicide rates in the U.S., 2005 (top) and 2015 (bottom).



Appendix Figure 14. Percentage of counties with greater than 20%, 30% or 40% increases in model-based suicide rates from 2005 to 2015, by urban-rural classification.



Note: Categories of >20%, >30% and >40% increase are not mutually exclusive.

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