

# Association of Primary Care Physician Supply With Population Mortality in the United States, 2005-2015

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**IMPORTANCE** Recent US health care reforms incentivize improved population health outcomes and primary care functions. It remains unclear how much improving primary care physician supply can improve population health, independent of other health care and socioeconomic factors.

**OBJECTIVES** To identify primary care physician supply changes across US counties from 2005-2015 and associations between such changes and population mortality.

**DESIGN, SETTING, AND PARTICIPANTS** This epidemiological study evaluated US population data and individual-level claims data linked to mortality from 2005 to 2015 against changes in primary care and specialist physician supply from 2005 to 2015. Data from 3142 US counties, 7144 primary care service areas, and 306 hospital referral regions were used to investigate the association of primary care physician supply with changes in life expectancy and cause-specific mortality after adjustment for health care, demographic, socioeconomic, and behavioral covariates. Analysis was performed from March to July 2018.

**MAIN OUTCOMES AND MEASURES** Age-standardized life expectancy, cause-specific mortality, and restricted mean survival time.

**RESULTS** Primary care physician supply increased from 196 014 physicians in 2005 to 204 419 in 2015. Owing to disproportionate losses of primary care physicians in some counties and population increases, the mean (SD) density of primary care physicians relative to population size decreased from 46.6 per 100 000 population (95% CI, 0.0-114.6 per 100 000 population) to 41.4 per 100 000 population (95% CI, 0.0-108.6 per 100 000 population), with greater losses in rural areas. In adjusted mixed-effects regressions, every 10 additional primary care physicians per 100 000 population was associated with a 51.5-day increase in life expectancy (95% CI, 29.5-73.5 days; 0.2% increase), whereas an increase in 10 specialist physicians per 100 000 population corresponded to a 19.2-day increase (95% CI, 7.0-31.3 days). A total of 10 additional primary care physicians per 100 000 population was associated with reduced cardiovascular, cancer, and respiratory mortality by 0.9% to 1.4%. Analyses at different geographic levels, using instrumental variable regressions, or at the individual level found similar benefits associated with primary care supply.

**CONCLUSIONS AND RELEVANCE** Greater primary care physician supply was associated with lower mortality, but per capita supply decreased between 2005 and 2015. Programs to explicitly direct more resources to primary care physician supply may be important for population health.

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Primary care physicians are typically responsible for the prevention, diagnosis, management, and treatment of a wide array of conditions. When examining area-level differences in the availability of primary care services and average large-area health statistics (at the international and state levels, in particular), higher availability of primary care services within a health system has been correlated with lower all-cause and cause-specific mortality. However, these cross-sectional correlations generally did not control extensively for other health care and socioeconomic confounders.<sup>1-6</sup> Consequently, the association of incremental increases in primary care physician supply with population-wide health outcomes remains heavily debated,<sup>7-10</sup> and whether to invest in policies that specifically aim to increase primary care physician supply is an important question in health care reform across many high-income countries. In the United States, in the absence of such policies, market forces have reduced primary care supply relative to higher-income specialties.<sup>11-14</sup>

How primary care physician supply has changed at the local level during the past decade in the United States and the strength of the association between changes in primary care physician supply and mortality remain unclear. Addressing these uncertainties is important to understanding whether efforts to expand primary care physician supply have the potential to produce measurable population health improvements. Herein, we sought to test associations between population-level physician supply and population-level mortality indicators across the United States during 2005 to 2015.

## Methods

### Independent Variable

Primary care physicians were defined as the number of non-federally employed physicians younger than 75 years who were not hospital residents and whose major professional activity was outpatient care in general practice, family medicine, general internal medicine, or general pediatrics, per 100 000 population in each US county and the District of Columbia (N = 3142 counties).<sup>15</sup> Primary care physician counts were obtained from the American Medical Association Physician Masterfile for 2005, 2010, and 2015,<sup>16</sup> and population counts were obtained from the US Census Bureau.<sup>17</sup> Data were analyzed from March to July 2018. This study was approved by the Stanford University Institutional Review Board, which waived informed consent.

### Outcomes

Life-expectancy (the primary outcome) and cause-specific mortality (secondary outcomes) were derived from deidentified death records from the National Center for Health Statistics and population counts from the US Census Bureau, National Center for Health Statistics, and the Human Mortality Database to estimate age-standardized life expectancy at birth and cause-specific mortality in 2005, 2010, and 2014 (updated to 2015 with linear interpolation at the county level).<sup>18</sup> Five major categories of cause of death were considered: cardiovascular disease, cancer, infectious diseases, respiratory

## Key Points

**Question** What is the association between primary care physician density and population-level mortality?

**Findings** In this epidemiological study of US population data, every 10 additional primary care physicians per 100 000 population was associated with a 51.5-day increase in life expectancy. However, from 2005 to 2015, the density of primary care physicians decreased from 46.6 to 41.4 per 100 000 population.

**Meaning** Greater primary care physician supply was associated with improved mortality, but per capita primary care physician supply decreased between 2005 and 2015.

tract diseases, and substance use or injury (ie, deaths from alcohol use, drug use, self-harm, and interpersonal violence).

### Covariates

We considered covariates that may confound the association between primary care physician supply and mortality at the population level (Table 1). These covariates included the number of specialist physicians with principal activity as patient care, per 100 000 population (defined by primary specialty in the American Medical Association Masterfile)<sup>16</sup>; urban/rural designation<sup>17</sup>; percentage of people under the federal poverty threshold and median household income in 2015 US dollars<sup>19</sup>; educational attainment<sup>20</sup>; population age, sex, and race/ethnicity<sup>21</sup>; unemployment status<sup>22</sup>; percentage of individuals without health insurance<sup>23</sup>; number of hospital beds per 100 000 population<sup>24</sup>; percentage enrolled in Medicare<sup>25</sup>; indicators of geographic variation in inflation-adjusted costs of medical care<sup>26</sup>; age-adjusted percentage of adults who currently smoked tobacco and percentage of adults with current obesity<sup>27</sup>; number of days with maximum 8-hour average pollution concentration greater than the National Ambient Air Quality Standard<sup>28</sup>; and median home value in 2015 US dollars.<sup>20</sup>

### Statistical Analysis

#### Overview of Analytic Approach

Because primary care physician supply cannot be randomized, we undertook a series of analyses to examine the association between primary care supply and the study outcomes using observational data. Our approach was to use a primary analytic method and then test the robustness of the associations found using complementary methods that make different assumptions or interrogate different aspects of the question. *P* values were 2-sided, with a set to .05 to determine statistical significance. Analyses were performed in R version 3.4.3 (R Foundation for Statistical Computing), using the statistical code deposited at <https://sdr.stanford.edu>.

Owing to completeness of mortality surveillance at the county level and primary care physician supply being ultimately an area-level concept, our primary analysis was to examine whether changes in primary care and specialist density within a county were associated with changes in age-adjusted life expectancy and cause-specific mortality within

Table 1. Characteristics of the Study Sample of 3142 US Counties, 2005-2015

Characteristic	Mean (95% CI)			Within-County Change, 2005 to 2015 <sup>a</sup>
	2005	2010	2015	
<b>Independent Predictor Variables</b>				
Physicians per 100 000 population <sup>b</sup>				
Total	114.6 (0.0 to 425.7)	111.6 (0.0 to 434.9)	112.8 (0.0 to 461.3)	-1.9 (-64.0 to 67.1)
Primary care	46.6 (0.0 to 114.6)	44.0 (0.0 to 113.7)	41.4 (0.0 to 108.6)	-5.2 (-44.6 to 28.8)
Specialist	68.0 (0.0 to 326.7)	67.6 (0.0 to 327.5)	71.3 (0.0 to 356.2)	3.4 (-40.1 to 58.9)
Nonmetro area, %	67.3	73.0	73.0	5.7
Population in poverty, % <sup>c,d</sup>	15.3 (5.6 to 31.4)	16.8 (7.0 to 31.8)	16.3 (6.9 to 32.0)	1.0 (-4.2 to 5.6)
Median household income, 2015 \$US	54 038.4 (34 370.7 to 90 744.6)	53 068.2 (34 523.4 to 87 779.4)	48 600.6 (30 622.5 to 80 641.5)	-5448 (-15 011.1 to 6711.5)
Population with less than high school education, %	17.8 (6.4 to 37.7)	15.4 (5.5 to 32.3)	13 (3.2 to 29.2)	-4.8 (-16.7 to 4.3)
Population ≥65 y, %	14.8 (7.6 to 24.1)	15.8 (8.5 to 25.2)	18.0 (10.0 to 27.9)	3.1 (-1.4 to 7.9)
Population female, %	50.3 (45.1 to 53.0)	49.8 (43.3 to 53.6)	50.0 (43.9 to 52.8)	-0.3 (-3.1 to 1.5)
Population black, %	9.0 (0.0 to 53)	9.1 (0.2 to 53.1)	9.3 (0.3 to 53.2)	0.3 (-2.7 to 3.1)
Population Hispanic, %	7.1 (0.5 to 46.9)	8.1 (0.8 to 50.4)	9.2 (0.9 to 53.6)	2.1 (-0.1 to 7.5)
Unemployment rate, %	5.6 (2.8 to 12.3)	9.4 (3.7 to 17.5)	5.7 (2.5 to 12.4)	0.1 (-2.3 to 3.1)
Uninsured among persons aged <65 y, %	25.0 (13.0 to 40.4)	18.5 (9.3 to 30.7)	12 (4.7 to 23.2)	-13.0 (-25.4 to -4.0)
Hospital beds per 100 000 population	358.6 (0.0 to 1733.0)	324.3 (0.0 to 1462.3)	294.7 (0.0 to 1336.0)	-63.2 (-592.9 to 185.9)
Medicare enrollment, %	16.3 (7.0 to 26.4)	18.3 (8.7 to 28.2)	20.6 (9.8 to 32.0)	4.3 (-1.4 to 10.6)
Geographic variation in terms of per capita medical costs, 2015, \$US	8946.3 (5734.3 to 13 002.4)	9395.5 (7311.4 to 12 394.6)	9843.4 (7790.6 to 12 676.8)	898.4 (-2914.8 to 4608.3)
Adult tobacco smoking, %	17.9 (12.7 to 25.6)	21.3 (11.0 to 33.0)	21.3 (11.0 to 33.0)	3.4 (-5.9 to 12.2)
Adult obesity, %	27.5 (19.0 to 34.0)	30.7 (21.0 to 39.0)	32.1 (21.0 to 42.7)	4.5 (-3.0 to 13.0)
High pollution days, %	4.7 (0.0 to 24.3)	1.3 (0.0 to 9.3)	6.2 (0.0 to 34.6)	1.5 (-16.5 to 26.3)
Median home value 2015, \$US	158 526.7 (60 822.4 to 462 389.7)	143 960.4 (63 335.6 to 371 204.1)	129 446.3 (55 081.0 to 293 371.5)	-29 080.4 (-192 909.3 to 45 054.4)
<b>Dependent Outcome Variables</b>				
Life expectancy at birth, age-standardized years	76.8 (72.5 to 80.5)	77.7 (73.2 to 81.5)	77.8 (72.9 to 82.0)	1.0 (0.1 to 1.9)
Age-adjusted deaths per 100 000 population				
Cancer	214.1 (162.2 to 271.7)	204.2 (149.8 to 264.1)	206.9 (148.2 to 273.6)	-7.2 (-21.9 to 8.0)
Cardiovascular disease	317.2 (215.7 to 439.7)	278.4 (185.4 to 397.2)	277.6 (179.6 to 406.6)	-39.6 (-66.6 to -12.9)
Infectious diseases	38.8 (22.2 to 63.2)	34.6 (18.3 to 58.8)	34.8 (17.0 to 61.4)	-4.0 (-9.9 to 2.7)
Respiratory tract diseases	62.3 (38.1 to 93.1)	62.1 (35.5 to 95.0)	64.2 (34.3 to 102.2)	1.8 (-7.5 to 13.7)
Substance use or injury	30.4 (16.3 to 53.8)	33.3 (17.4 to 59.8)	35.5 (17.9 to 64.7)	5.1 (-1.3 to 15.1)
Interpersonal violence	5.3 (1.6 to 15.2)	4.9 (1.5 to 14.6)	5.0 (1.6 to 14.7)	-0.3 (-1.9 to 0.7)

<sup>a</sup> Owing to rounding of displayed values in the table, within-county differences may differ at the tenths place from a simple subtraction of the second and fourth columns.

<sup>b</sup> Physician counts are specific to the subset of physicians who list patient care

as their primary occupational activity.

<sup>c</sup> Absolute percentage.

<sup>d</sup> Determined at county level.

that county after controlling for the above characteristics. Because counties might not conform to care-seeking patterns, we repeated these analyses using alternative geographic levels—the primary care service area (N = 7144) and the hospital referral region (N = 306), which were developed based on geographic patterns of care. We also conducted additional sensitivity analyses as described below to test robustness.

**Primary Analysis Mixed-Effects Models**

For our area-level analyses, we used a linear mixed model to regress life expectancy against the independent variables within each county, allowing intercepts (baseline outcomes)

to vary among counties, time trends to vary across the study period, and slopes for the association between physician density and each outcome to vary among counties. A mixed model empirically estimates the within- vs between-county components of variation in each outcome. Standard errors were computed with an autoregressive correlation structure to account for serial correlation in outcomes across time within counties, with county population weights. We excluded from the analysis the less than 5% of counties with any missing variable. Further details of the mixed-effects models are provided in the eAppendix in the Supplement. To address individuals crossing county boundaries for medical care, we

repeated the modeling using geographically weighted regression with the latitude and longitude of county centroid to account for potential regional patterns of utilization and flow across boundaries.<sup>29,30</sup>

#### Instrumental Variable Analysis

To address the possibility for unmeasured confounding in our primary analyses, we conducted instrumental variable analyses. An instrumental variable is a factor that influences the outcome (mortality) only through its influence on the predictor variable of interest (primary care physician supply) but is not subject to reverse causality from the outcome or omitted variable bias. For example, the federal Public Service Loan Forgiveness program forgives some loan payments for physicians who enter into public service, commonly nonprofit community clinics; it is considered the major financial policy inducement for selecting primary care.<sup>31</sup> Although the loan forgiveness amount is fixed, the purchasing power of the forgiven amount varies widely by county (eg, influencing the ability to purchase a house). This instrument was found to be strong (first-stage  $F = 25.6$ ) for predicting primary care density, but not specialist density ( $F = 7.2$ ). We thus used changes in county purchasing power to predict changes in primary care physician supply (eFigures 5 and 6 in the [Supplement](#)) and in turn associate changes in supply with changes in mortality through this instrumental variable. Additional details on the instrumental variable approach are provided in the eAppendix in the [Supplement](#). We performed a robustness check on the instrumental variable analysis using near-far matching,<sup>32</sup> an analytic strategy that can strengthen the power of an instrument by matching counties that are similar in their key characteristics (Table 1) but different in their values of the instrumental variable to mimic a matched-pair randomized trial (eAppendix in the [Supplement](#)).

#### Individual-Level Analyses

Individual-level analyses with a second data source were performed to reduce the likelihood of ecological confounding using a national claims database linked to date of death from the Social Security Death Master File<sup>33</sup> (Optum Clinformatics Data Mart, 2003 through 2016; 1 505 554 individuals). We performed a survival analysis by estimating the Kaplan-Meier survival rate of participants, adjusted for censoring, to determine the association between area-level primary care physician supply and individual-level life expectancy. The outcome was restricted mean survival time, which is the area under the survival curve, conditional on exposure to area-level primary care physician density (equations are given in the eAppendix in the [Supplement](#)).<sup>34</sup> We adjusted for all area-level covariates in Table 1 as well as for individual-level age and sex. Characteristics of the individual participants in the sample are provided in eTable 1 in the [Supplement](#). A subgroup analysis was also conducted, focusing on individuals who moved between zip codes as a quasi-random exposure to changes in primary care physician density ([Supplement](#)).<sup>35</sup>

#### Falsification Testing

A falsification test was performed by regressing the independent variables against a dependent outcome variable that

would not be expected to have a significant association with primary care physician supply: mortality due to interpersonal violence (eg, murder).<sup>18</sup> This test examined whether unobserved factors, such as the propensity for physicians to move to desirable areas, which may have features correlated with lower mortality rates, would produce false associations between primary care physician density and improved outcomes.

In addition, we calculated the E value,<sup>36</sup> which estimates how strong unmeasured confounders (factors correlated with both primary care physician supply and life expectancy) would need to be to explain away the association between primary care physician density and life expectancy.

#### Sensitivity Analyses

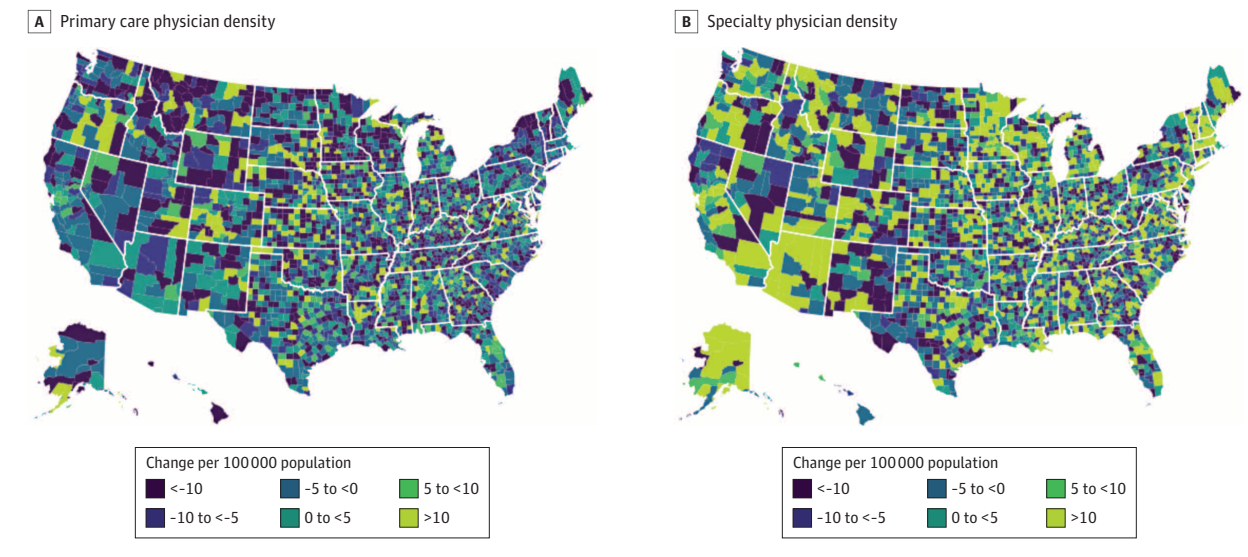
Our analyses were also repeated after including nurse practitioners and physician assistants with a national provider identifier registration who reported working in primary care per their Medicare registration information.<sup>16,37</sup> Nurse practitioners and physician assistants are not registered consistently or labeled consistently as primary care over time, and thus they were not included in the prespecified primary model because their inclusion could produce misclassification error and regression to the mean.

## Results

### Changes in Primary Care Physician and Specialist Density

The total number of primary care physicians increased from 196 014 in 2005 to 204 419 in 2015. However, owing to disproportionate losses of primary care physicians in some counties and population increases in general, mean primary care physician supply decreased from 46.6 per 100 000 population in 2005 (95% CI, 0.0-114.6 per 100 000 population) to 41.4 per 100 000 in 2015 (95% CI, 0.0-108.6 per 100 000 population) ([Figure 1](#) and Table 1). Primary care physician supply per 100 000 population had a skewed distribution; 296 counties had no primary care physicians in 2015, whereas 128 counties had more than 100 per 100 000 population. Primary care physician supply declined more in rural than in urban counties on average (-7.0 per 100 000 population vs -2.6 per 100 000 population) but with broad changes in both (eFigure 1 in the [Supplement](#)). Owing to small populations, rural counties can appear to have large variations in primary care physician density; thus, the absolute changes were also analyzed and varied from a loss of 32 to a gain of 37 primary care physicians, with a median loss of 1.0 physician per county. Urban county primary care physician changes ranged from a loss of 179 to a gain of 405 primary care physicians, with a median gain of 1.0 physician per county. Primary care physician supply in either density or absolute terms did not disproportionately decrease by county poverty level or racial/ethnic demographic features (eFigures 1 and 7 in the [Supplement](#)). Density decreased by 4.2 physicians per 100 000 population at the primary care service area level (95% CI, -31.1 to 18.8 per 100 000 population), and by 4.4 per 100 000 population at the hospital referral region level (95% CI, -32.1 to 19.8 per 100 000 population).

Figure 1. Changes in Density of Primary Care and Specialist Physicians in 3142 US Counties, 2005-2015



Concurrently, the number of specialist physicians increased from 699 989 in 2005 to 805 277 in 2015, corresponding to 68.0 per 100 000 population in 2005 to 71.3 per 100 000 population in 2015 (Figure 1, Table 1). In absolute terms, counties gained a mean (range) of 3.4 specialists (-359 to 1065 specialists) nationwide, but rural counties had no mean gain (eFigures 2 and 8 in the Supplement). Changes in primary care physician supply minimally correlated with changes in specialist physician supply (Pearson correlation coefficient, 0.14).

#### Associations Between Primary Care Physician Density and Life Expectancy

In adjusted regressions (Table 2), total physician supply, primary care physician supply, and specialist physician supply were associated with improved life expectancy.<sup>38</sup> In fully adjusted models that accounted for both primary care and specialist physician supply, an increase of 10 primary care physicians per 100 000 population was associated with a 51.5-day increase in life expectancy (95% CI, 29.5-73.5 days; 0.2% increase) (Table 2), and a similar increase of 10 specialist physicians per 100 000 population was associated with a 19.2-day increase in life expectancy (95% CI, 7.0-31.3 days).<sup>38</sup>

To contextualize these results, the association of primary care physician density and life expectancy (+33.1 days of life expectancy for a 2-SD increase in physician density) was approximately one-fifth the magnitude of the association between poverty and life expectancy (148.8 days for a 2-SD increase), and approximately two-thirds the magnitude of the association between tobacco and life expectancy (52.3 days for a 2-SD increase; eTable 2, Table 2).<sup>38</sup>

Analyses at alternative geographic levels revealed similar associations between primary care physician supply and life expectancy (Figure 2; increase of 51.5 days life expectancy per 10 additional physicians at the county level, an increase of 117.3

at the PCSA level, and an increase of 157.5 at the HRR level). The geographically weighted regression results also were consistent (Table 2; Figure 2).

#### Association Between Primary Care Physician Density and Cause-Specific Mortality

An increase of 10 primary care physicians per 100 000 population was associated with a reduction in cardiovascular mortality by 30.4 deaths per million (95% CI, -52.4 to -8.4; a 0.9% reduction), in cancer mortality by 23.6 deaths per million (95% CI, -35.0 to -12.3 deaths per million; 1.0% reduction), and in respiratory mortality by 8.8 deaths per million (95% CI, -15.3 to -2.2 deaths per million; a 1.4% reduction) after adjustment for covariates (Figure 3). Specialty physician supply was associated with significant reductions in cause-specific mortality for 2 groups: cardiologists and cardiovascular mortality (increase of 10 cardiologists associated with -49.4 deaths per million [95% CI, -76.8 to -22.0 deaths per million]) and pulmonologists and respiratory tract disease mortality (10 pulmonologists associated with -10.5 deaths per million [95% CI, -20.6 to -0.4 deaths per million]; Figure 3).

#### Instrumental Variable Analyses

An increase of 10 primary care physicians per 100 000 population was associated with an 88.9-day (95% CI, 15.6-162.2 days) increase in life expectancy in the instrumental variable analysis (Figure 2). The instrumental variable analysis also detected associations between primary care physician supply and lower cardiovascular, cancer, and respiratory mortality (eFigure 3 in the Supplement). Results were similar in the near-far matching analysis (eFigure 4 in the Supplement).

#### Individual-Level Analysis

In the individual-level analysis, survival time increased by 114.2 days (95% CI, 94.7-133.8 days) per decade of exposure to 10 more primary care physicians per 100 000 population

**Table 2. Results of Mixed-Effects Regressions Associating Physician Density and County-Level Covariates With Age-Standardized Life Expectancy at Birth in 3142 US Counties, 2005-2015**

Variable	Change in Age-Standardized Life Expectancy (95% CI)			
	Model 1 (Total Physician Density)	Model 2 (Primary Care Physician Density)	Model 3 (Specialist Density)	Model 4 (Primary Care Physician and Specialist Density)
Total physicians, per 100 000 population <sup>a</sup>	66.7 (47.5 to 85.8)	-NA	NA	NA
Covariate, per +10 physicians per 100 000 <sup>b</sup>	88.9	NA	NA	NA
Primary care physicians, per 100 000 population	NA	31.8 (17.7 to 45.9)	NA	33.1 (19.0 to 47.3)
Covariate, per +10 physicians per 100 000 <sup>b</sup>	NA	49.7	NA	51.5
Specialty physicians, per 100 000 population	NA	NA	23.3 (9.3 to 37.3)	20.6 (7.5 to 33.6)
Covariate, per +10 physicians per 100 000 <sup>b</sup>	NA	NA	21.7	19.2
Metro area, change to nonmetro area, d	-54.6 (-79.8 to -29.5)	-55.8 (-81.0 to -30.7)	-51.0 (-76.5 to -25.6)	-54.2 (-79.4 to -29.0)
Population in poverty, d	-149.7 (-172.2 to -127.2)	-146.6 (-169.0 to -124.3)	-152.4 (-175.2 to -129.7)	-148.8 (-171.2 to -126.4)
Population with less than high school education, d	-59.5 (-73.5 to -45.5)	-58.1 (-71.9 to -44.3)	-59.4 (-73.5 to -45.3)	-58.1 (-72.0 to -44.3)
Female, d	-20.4 (-34.9 to -5.8)	-19.7 (-34.2 to -5.2)	-18.9 (-33.6 to -4.2)	-20.1 (-34.6 to -5.6)
Black, d	-409.4 (-448.9 to -370.0)	-406.5 (-445.8 to -367.2)	-408.6 (-448.4 to -368.8)	-411.9 (-451.2 to -372.6)
Hispanic, d	185.9 (149.9 to 221.9)	185.2 (149.3 to 221.1)	185.1 (149.0 to 221.3)	185.3 (149.4 to 221.1)
Unemployment rate, d	-13.7 (-28.2 to 0.8)	-14.1 (-28.6 to 0.3)	-15.6 (-30.2 to -0.9)	-14.1 (-28.5 to 0.4)
Hospital beds, per 100 000 population, d	-3.9 (-24.1 to 16.4)	-0.2 (-20.2 to 19.8)	-0.9 (-21.1 to 19.3)	-2.3 (-22.4 to 17.8)
Medicare enrollment, d	106.4 (81.1 to 131.7)	108.6 (83.5 to 133.7)	111.0 (85.6 to 136.4)	107.8 (82.6 to 132.9)
Per capita medical cost variation, d	4.3 (-4.5 to 13.2)	5.0 (-3.8 to 13.8)	4.8 (-4.1 to 13.8)	5.0 (-3.8 to 13.8)
Adult tobacco smoking, d	-52.8 (-63.4 to -42.3)	-52.6 (-63.1 to -42.1)	-54.0 (-64.6 to -43.4)	-52.3 (-62.8 to -41.8)
Adult obesity, d	-39.1 (-49.3 to -28.9)	-39.2 (-49.4 to -29.0)	-40.3 (-50.7 to -30.0)	-39.3 (-49.5 to -29.1)
High pollution days, d	-21.2 (-28.7 to -13.8)	-21.1 (-28.5 to -13.7)	-22.3 (-29.8 to -14.8)	-21.1 (-28.5 to -13.7)
Median home value, d	-22.0 (-42.4 to -1.6)	-22.5 (-42.7 to -2.3)	-19.9 (-40.4 to 0.7)	-23.2 (-43.5 to -3.0)

Abbreviation: NA, not applicable.

<sup>a</sup> All means and SDs calculated after log transformation. Total physician coefficient corresponds to the change from mean minus 1 SD (61.0 physicians) to mean plus 1 SD (68.5 physicians); primary care, to the change from mean minus 1 SD (27.3 physicians) to mean plus 1 SD (33.7 physicians); specialty, to the change from mean minus 1 SD (21.3 physicians) to mean plus 1 SD (32.0 physicians); population in poverty, to the change from mean minus 1 SD (13.4% poverty) to mean plus 1 SD (16.4% poverty); population with less than high school education, to the change from mean minus 1 SD (13.1% less than high school) to mean plus 1 SD (16.3% less than high school); population female, to change from mean minus 1 SD (48.9% female) to mean plus 1 SD (51.0% female); population black, to change from mean minus 1 SD (1.4% black) to mean plus 1 SD (7.9% black); population Hispanic, to change from mean minus 1 SD (2.8% Hispanic) to mean plus 1 SD (7.8% Hispanic); unemployment rate to change from mean minus 1 SD (4.7% unemployed) to mean plus 1 SD (7.8%

unemployed); hospital beds, to change from mean minus 1 SD (32.6 hospital beds) to mean plus 1 SD (61.2 hospital beds); Medicare enrollment, to change from mean minus 1 SD (17.2% enrolled) to mean plus 1 SD (19.9% enrolled); per capita medical cost variation, to change from mean minus 1 SD (\$9268.5) to mean plus 1 SD (\$9270.9); adult tobacco smoking, to change from mean minus 1 SD (18.0% smoking) to mean plus 1 SD (20.6% smoking); adult obesity, to change from mean minus 1 SD (28.5% obesity) to mean plus 1 SD (30.9% obesity); high pollution days, to change from mean minus 1 SD (0% of days) to mean plus 1 SD (5.4% of days); and median home value, from change from mean minus 1 SD (\$126 825.2) to mean plus 1 SD (\$126 828.5).

<sup>b</sup> Continuous variables were log transformed and centered and scaled by 2 SDs, which allows coefficients for continuous covariates to reflect the change in the outcome variable given a change in the independent variable from its mean minus 1 SD to its mean plus 1 SD on the logged scale, correcting for right skew and enabling fair comparison of magnitudes among regression coefficients.<sup>38</sup>

(Figure 2). Results were also improved among a subgroup who moved between zip codes (eAppendix in the Supplement).

**Falsification Testing**

Increased primary care physician supply was not significantly associated with deaths from interpersonal violence. An increase of 10 primary care physicians per 100 000 population was associated with -0.5 death per million from violence (95% CI, -1.4 to 0.5 deaths per million).

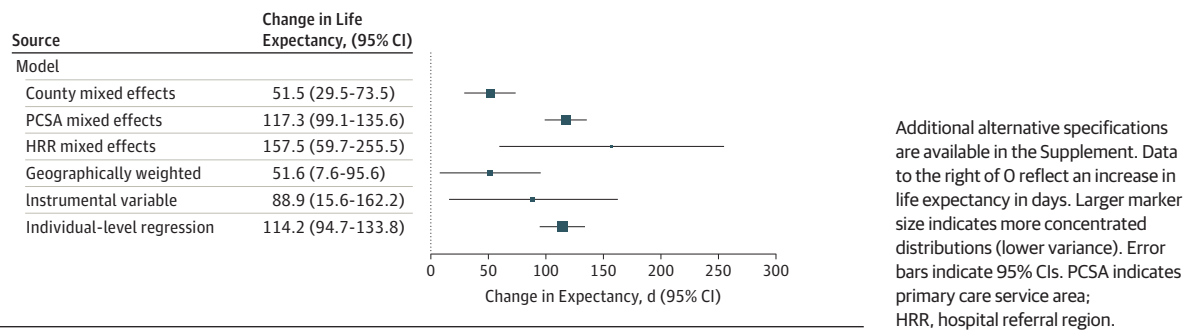
The E value for the association between primary care physician supply and life expectancy was 131.2 days. This means that unmeasured confounders correlated with both primary

care physician density and life expectancy would have to have strong associations with life expectancy (a 131.2-day increase in life expectancy is almost as great in magnitude as the association between poverty and life expectancy, Table 2) to explain away the observed association between primary care physician supply and life expectancy.<sup>38</sup>

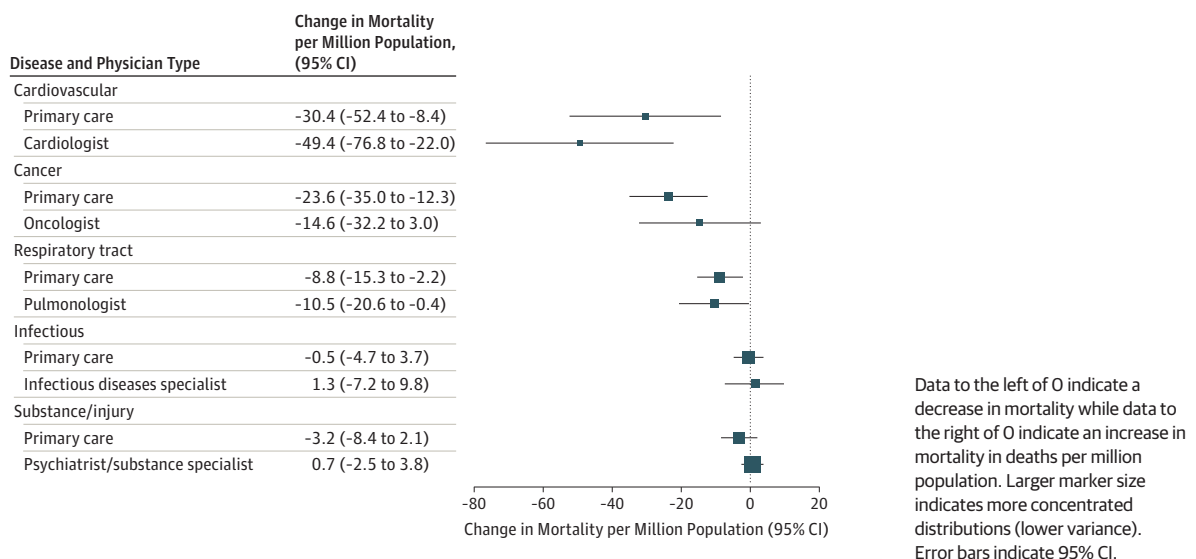
**Sensitivity Analyses**

When including primary care nurse practitioners and physician assistants in the analysis, an increase in primary care clinician supply (physician, nurse practitioner, or physician assistant) by 10 per 100 000 population showed consistent results

**Figure 2. Changes in Life Expectancy Associated With an Increase in 10 Primary Care Physicians per 100 000 Population Using Alternative Model Specifications**



**Figure 3. Changes in Cause-Specific Mortality Associated With an Increase in 10 Primary Care Physicians or 10 Specialist Physicians per 100 000 Population**



(increase in life expectancy, 36.4 days; 95% CI, -19.0 to 91.8 days), but the result was no longer statistically significant at the county level. Counties had a mean of 25.2 nurse practitioners and physician assistants in primary care per 100 000 population in 2005 (95% CI, 0.0-60.8 per 100 000 population) and 50.1 per 100 000 population in 2015 (0.0-141.1 per 100 000 population).

## Discussion

Although the total number of primary care physicians has increased in the United States, owing to disproportionate rural losses and general population size increases, the distribution of US primary care physicians per 100 000 population has changed, leading to a net loss in mean primary care physician supply at the county level. Greater primary care physician supply was associated with lower population mortality, suggesting that observed decreases in primary care physician supply may have important consequences for population health. These findings were consistent across several analytic specifica-

tions that varied the unit of analysis, level of analyses, and statistical assumptions underpinning the analysis.

The results of this study reinforce findings from earlier cross-sectional studies evaluating data from the 1990s, which suggested associations at the health care system and state levels between primary care physician density, overall life expectancy, cardiovascular disease deaths, and cancer deaths.<sup>1,3,8,10,39,40</sup> However, similar to another study,<sup>41</sup> our results are driven by changes in density over time within counties, which reduces confounding by features of areas that might have affected earlier research. Our study included a larger number of control variables than previous analyses, including specialist physician supply and numerous health care, socioeconomic, environmental, and demographic features omitted from earlier studies. The largest decreases in cause-specific mortality associated with increased primary care physician density were for cardiovascular disease, cancer, and respiratory tract disease, conditions with strong evidence of amenability to primary care management or with delayed mortality conditional on early screening through primary care.<sup>8,10,31,42-44</sup>

Many believe that a well-functioning health care system requires a solid foundation of primary care. However, persistent payment disparities between primary care and procedural specialties continue to erode the US primary care physician workforce.<sup>45,46</sup> Policy initiatives, such as Medicare's Accountable Care Organization programs, that attempt to focus on population health and spending and thereby rely on primary care physicians continue to gain traction, but these programs fail to explicitly direct more resources to primary care physician supply, instead relying on the usual Medicare fee schedule. The Centers for Medicare & Medicaid Service's Comprehensive Primary Care initiative invests more resources in primary care, and states such as Rhode Island and Oregon have substantially increased spending on primary care.<sup>47</sup> Whether these initiatives will encourage more graduating medical students to enter primary care remains to be seen. Other forms of investment, such as the National Health Services Corps, the Teaching Health Centers program, and Title VII programs, also offer the opportunity to increase the density of primary care physicians, especially in underserved areas.

### Limitations

Our study has important limitations. First, our main analysis was appropriately ecological, because we sought to identify relationships between population-level physician supply and population-level variations in mortality. However, to avoid the ecological fallacy, conclusions should not be drawn about individual-level effects of population-level associations. To help

mitigate this limitation, we conducted individual-level analyses using private insurance data. Although not nationally representative, the individuals in the insurance data were from all 50 states, Washington, DC, and Puerto Rico, covering 61% of US zip codes. Second, there remains the possibility for unobserved confounding, because we cannot randomize people to areas with varied primary care physician supply. We sought to address this with instrumental variable analyses, but the instrumental variable analysis has its own assumptions. Finally, our analysis focused on primary care physician supply relative to population size because this is a key focus of current policies.<sup>48-50</sup>

### Conclusions

Across a number of analytic approaches, greater primary care physician supply was associated with improved mortality outcomes. The decrease in primary care physician supply across US counties from 2005 to 2015 may have important population health implications. Future investigations should acquire data on the quality and comprehensiveness of primary care, types of primary care physician training and service delivery offerings, and effective access rather than just supply. In addition, future analyses should explore the dynamics of teamwork across primary care physicians and specialists in both traditional and alternative payment models to address how team-based approaches may affect mortality and other outcomes.

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