

# Traffic-Related Air Pollution and All-Cause Mortality during Tuberculosis Treatment in California

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**BACKGROUND:** Ambient air pollution and tuberculosis (TB) have an impact on public health worldwide, yet associations between the two remain uncertain.

**OBJECTIVE:** We determined the impact of residential traffic on mortality during treatment of active TB.

**METHODS:** From 2000–2012, we enrolled 32,875 patients in California with active TB and followed them throughout treatment. We obtained patient data from the California Tuberculosis Registry and calculated traffic volumes and traffic densities in 100- to 400-m radius buffers around residential addresses. We used Cox models to determine mortality hazard ratios, controlling for demographic, socioeconomic, and clinical potential confounders. We categorized traffic exposures as quintiles and determined trends using Wald tests.

**RESULTS:** Participants contributed 22,576 person-years at risk. There were 2,305 deaths during treatment for a crude mortality rate of 1,021 deaths per 10,000 person-years. Traffic volumes and traffic densities in all buffers around patient residences were associated with increased mortality during TB treatment, although the findings were not statistically significant in all buffers. As the buffer size decreased, fifth-quintile mortality hazards increased, and trends across quintiles of traffic exposure became more statistically significant. Increasing quintiles of nearest-road traffic volumes in the 100-m buffer were associated with 3%, 14%, 19%, and 28% increased risk of death during TB treatment [first quintile, referent; second quintile hazard ratio (HR) = 1.03 [95% confidence interval (CI): 0.86, 1.25]; third quintile HR = 1.14 (95% CI: 0.95, 1.37); fourth quintile HR = 1.19 (95% CI: 0.99, 1.43); fifth quintile HR = 1.28 (95% CI: 1.07, 1.53), respectively; *p*-trend = 0.002].

**CONCLUSIONS:** Residential proximity to road traffic volumes and traffic density were associated with increased all-cause mortality in patients undergoing treatment for active tuberculosis even after adjusting for multiple demographic, socioeconomic, and clinical factors, suggesting that TB patients are susceptible to the adverse health effects of traffic-related air pollution. <https://doi.org/10.1289/EHP1699>

## Introduction

Tuberculosis (TB) has an impact on health worldwide, with an estimated 9.6 million people developing active TB and 1.5 million people dying from the disease in 2014 (WHO 2015). TB mortality rates remain high despite adequate treatment (Fielder et al. 2002; Pascopella et al. 2014), and the effects of environmental factors such as ambient air pollution on TB outcomes remain uncertain. The majority of the world's population is exposed to unhealthy levels of ambient air pollution, with approximately 89% living in areas where fine particulate matter <2.5 μm in diameter (PM<sub>2.5</sub>) exceeds World Health Organization (WHO) air quality standards (Brauer et al. 2012), and it is estimated that ambient air pollution contributes to

approximately 3.3 million premature deaths each year (Lelieveld et al. 2015).

It has long been observed that certain inhaled toxicants are associated with pulmonary infection: Published studies as far back as 100 y ago reported associations between tobacco smoking and TB (Webb 1918). There is now ample evidence to suggest that active tobacco smokers are at increased risk for TB infection, progression to active TB, and worse treatment outcomes including mortality (Bates et al. 2007; Horne et al. 2012; Jee et al. 2009; Lin et al. 2007; Maciel et al. 2013; Slama et al. 2007). This evidence has led the WHO to affirm tobacco smoking as a significant TB risk factor (WHO 2016).

Tobacco smoke is composed of a number of chemical compounds that are also found in traffic emissions (CDC 2010; Gentner et al. 2013; Karjalainen et al. 2014), raising concerns that traffic-related air pollution (TRAP) could also be associated with adverse TB outcomes. Ambient air pollution is an established risk factor for community-acquired pneumonia (Chiu et al. 2009; Neupane et al. 2010; Zanobetti et al. 2000), yet studies linking ambient air pollution to other pulmonary infections such as TB are limited. The effects of ambient air pollution on TB incidence have been investigated in a few studies with mixed findings. In a nested case–control study using a northern California managed-care database, nitrogen dioxide (NO<sub>2</sub>) was associated with increased odds of active TB (Smith et al. 2016); in a community-based cohort study in Taipei, Taiwan, both PM<sub>2.5</sub> and NO<sub>2</sub> were associated with increased risk of active TB (Lai et al. 2016); and in Beijing and Hong Kong, China, outdoor PM<sub>2.5</sub> was associated with seasonal changes in TB incidence (You et al. 2016). However, in a time-series study in Seoul, South Korea, spikes in sulfur dioxide (SO<sub>2</sub>) but not in particulate

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matter <10 µm in diameter (PM<sub>10</sub>), NO<sub>2</sub>, or ozone were associated with temporally related increases in TB incidence (Hwang et al. 2014). The effects of ambient air pollution on TB treatment outcomes remain uncertain, and there has been only one published study to date that has examined the association of ambient air pollution with mortality during TB treatment; this study found that annual mean estimates of PM<sub>2.5</sub> were associated with increased mortality (Peng et al. 2017). Additionally, there has been only one published study to evaluate traffic proximity as a TB risk factor. This study did not find a statistically significant association between distance to the nearest major road or freeway and smear positivity among patients with pulmonary TB in Los Angeles, California (Jassal et al. 2013). However, the study's small sample size and cross-sectional design preclude firm conclusions linking these traffic metrics to TB outcomes, thereby warranting a larger study with a prospective cohort study design from which causal inferences can be made.

We used proximity to road traffic volumes and proximity to traffic density as measures of traffic-related air pollution instead of individual air pollutant models because our traffic proximity models capture the mix of pollutants emitted from motor vehicles as opposed to models for individual pollutants that might provide a less-complete representation of traffic exposure. For instance, in a study evaluating the association of traffic exposures with childhood incident asthma, the effect of NO<sub>2</sub> on incident asthma, which had a hazard ratio (HR) of 2.17 [95% confidence interval (CI): 1.18, 4.00], was attenuated to 1.37 (95% CI: 0.69, 2.71) when adjusting for traffic exposure using a line source dispersion model that included distance to roadways and vehicle counts (McConnell et al. 2010), suggesting that traffic proximity was a key factor for determining TRAP health effects. Furthermore, in another study comparing different TRAP exposure metrics, traffic density was found to be "reasonably consistent with the more sophisticated metrics" (Batterman et al. 2014).

Both TB and ambient air pollution are significant public health issues in California. Nearly 25% of all U.S. TB cases occur in California (Salinas et al. 2016), and >80% of Californians live in counties with unhealthy ambient air concentrations of PM<sub>2.5</sub>, ozone, or both (ALA 2016). Additionally, the predominant source of air pollution in California is from traffic emissions (ARB 2017). We hypothesized that proximity to road traffic volumes and traffic density would be associated with increased mortality in patients undergoing TB treatment and that the effects would be mediated through clinical markers of TB severity. To test these hypotheses, we designed and implemented a large cohort study of TB patients reported to the State of California and followed longitudinally by California Department of Public Health TB clinics throughout their TB treatment.

## Methods

### Study Cohort

All reported pediatric and adult TB cases in California between 1 January 2000 and 31 December 2012 were eligible for inclusion. A TB case was specified as a person of any age with clinically diagnosed or microbiologically confirmed (or both) active TB as defined by the Centers for Disease Control and Prevention (CDC). Clinical diagnostic criteria included symptoms, physical exam, and radiographic findings consistent with TB along with an appropriate response to treatment. Acceptable microbiologic confirmation included culture or nucleic acid amplification testing (or both) positive for *Mycobacterium tuberculosis* (*Mtb*). Patients were excluded from survival analyses if traffic data were unavailable at residential addresses, if residential addresses were either not available or not geocodable to street-level resolution, if

patients died or moved out of California before treatment was initiated, or if treatment dates were unavailable.

### Traffic Exposure Assessment

Residential street addresses were obtained at the time of TB diagnosis and were later transformed into geocoded coordinates using browser-based geocoding software developed by the California Environmental Health Tracking Program (CEHTP) at the California Department of Public Health (CDPH). This software matched addresses to geographic coordinates using Tele Atlas® (TomTom Telematics), Navteq® (Nokia Here), and TIGER® (U.S. Census Bureau) reference data sets for 2010 and 2011. For homeless patients, we used the shelter address or the street intersection of the patient's most recent sleeping location. Geocoding accuracy was quantified using a score based on how well the input address text elements matched the same elements of the geocoding reference database. The score ranged from 0 to 100, with 100 representing an exact match.

Traffic data from 2004 were obtained from the California Department of Transportation (Caltrans) Highway Performance Monitoring System, with data available for roads functionally classified as collectors, arterials, freeways, expressways, and interstates, collectively referred to as "major" roads in this study. Traffic data were not available for small local roads. Because the distance from traffic source for maximum health impact was uncertain (Puett et al. 2014), we calculated traffic exposures within four circular zones (buffers) of radius 100–400 m around each participant's residential address using CEHTP Traffic Spatial Linkage Service software (CDPH). Within each of these buffers, we evaluated traffic volume [average number of vehicles (vh) traversing a road segment per 24 h] at the nearest major road and at the highest-trafficked road, and traffic density was calculated as the sum of length-adjusted major road segment traffic counts within each buffer per hour (vehicle-kilometers/hour). We selected these traffic indicators to best characterize distance to exposure source as well as peak and overall exposure concentrations. Left-censored traffic data due to the absence of major roads in a buffer were considered missing.

### Outcome Assessment

Patients were followed longitudinally throughout TB treatment by one of 61 local health department TB programs. Baseline and follow-up demographic, socioeconomic, and clinical data were recorded by trained local health officials into the Report of Verified Case of Tuberculosis (RVCT), a CDC-developed data collection tool. RVCT data were then entered into a CDPH central data repository (the California TB Registry). Our preselected primary outcome was all-cause mortality during active TB treatment. Follow-up time began on the first day of TB treatment and ended on the date of death (event) or was right-censored on the date of adequate completion of TB therapy, cessation of therapy for other reasons, moving out of state, loss to follow-up, or the end of the study, whichever occurred first.

### Ethics Approval

The study was approved by the institutional review boards at the University of California, San Francisco and the California Health and Human Services Agency Committee for the Protection of Human Subjects. Informed consent was not required because the study relied on existing public health records rather than on direct patient contact, and all personal identifiers were removed from the database before analysis.

## Statistical analysis

We fit Cox proportional hazards models to determine HRs for mortality during TB treatment predicted by each traffic metric in 100-to-400-m buffers around residential addresses. We transformed traffic volumes and traffic density into quintiles to better adhere to log-linearity of hazard function assumptions, and we tested for linear trends across quintiles using Wald tests (Vittinghoff et al. 2012), reporting calculated *p*-values as *p*-trends. Traffic exposure associations with mortality were considered statistically significant for *p*-trends <0.05 in at least two of the four buffers tested. In forming quintiles, we chose exposure cut points that would split participants into five equal groups. Individual-level demographic, socioeconomic, and clinical covariates were selected *a priori* as potential confounders and were included in the final multivariable models if associated with mortality in univariate Cox models with a *p*-value <0.2. In alternative models, we additionally adjusted for two group-level variables: census block group median annual household income from the 2006–2010 American Community Survey (U.S. Census Bureau 2016) and census-tract tobacco smoking prevalence estimates (Ortega Hinojosa et al. 2014). We tested for effect modification by defining interaction terms between traffic density and each of the following: age, sex, race, ethnicity, region of residence, enrollment year, bacteriologic confirmation, directly observed TB treatment (DOT), census-tract tobacco smoking estimates, and HIV; we fitted separate multivariable Cox models with each of these interaction terms and reported effect modification for a *p*-interaction ≤0.05 in at least one buffer and a consistent pattern of effect modification in all buffers. We performed mediation analyses to determine if the effects of traffic on mortality were mediated through TB severity, and we performed multiple sensitivity analyses to test the robustness of our findings across alternative Cox models. Standard diagnostics and assumption checks were performed on all models. Time-varying covariate models were not employed; instead, covariates with time dependency were assigned fixed measurement times and values. The comorbidities diabetes, end-stage renal disease, and non-HIV immunosuppression were reported only from 2010–2012. In subgroup analyses including these comorbidities, we transformed traffic exposures into dichotomous rather than five-level categorical variables as in the other analyses owing to the small number of deaths in each subgroup. We evaluated the effects of nearest-road traffic volumes on secondary outcomes including culture conversion and successful treatment completion using unadjusted Kaplan–Meier survival analyses and testing for statistical significance between fifth- and first-quintile exposures using log-rank tests. Statistical analyses were performed using Stata/SE 13.1 (StataCorp).

## Results

Of 36,511 patients with active TB reported to the State of California, we excluded 2,558 patients (7.0%) for whom we were unable to perform geocoding and traffic linkage with street-level accuracy (Figure 1). We excluded an additional 1,078 patients (3.0%) who moved from California, died before treatment was initiated, or for whom treatment records were incomplete. Our final cohort of 32,875 patients with active TB contributed 22,576 person-years at risk {median follow-up of 224 d [interquartile range (IQR) 184–293] per patient}. There were 2,305 deaths during follow-up for a crude mortality rate of 1,021 deaths per 10,000 person-years. Most right-censored events were due to either treatment completion or moving out of California, and only 2.8% were lost to follow-up or had unspecified outcomes. Sixty percent of included patients were male, 44% were Asian, 38%

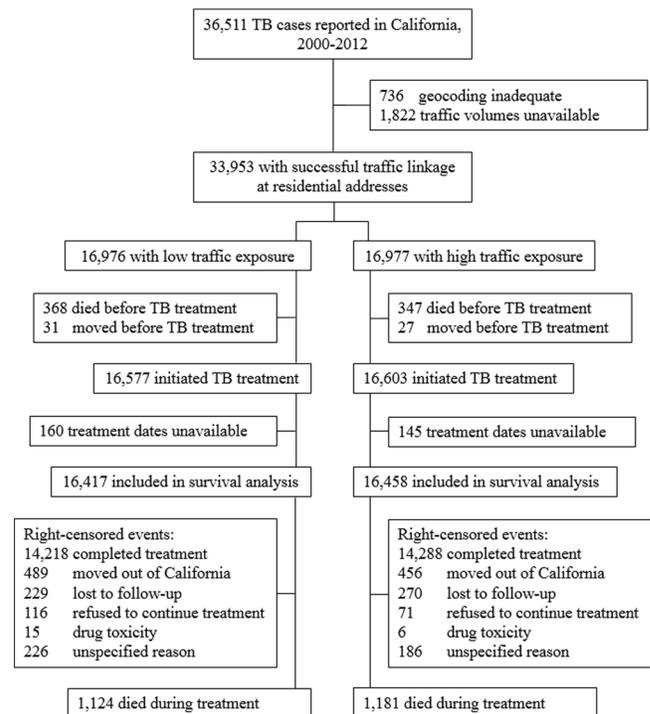


Figure 1. Study flow chart. TB, tuberculosis.

were Hispanic, 77% were foreign-born, and 5% were HIV-infected (Table 1). Socioeconomic hardship was evident, with 53% unemployed, 5.8% homeless, and 13% using recreational drugs, excess alcohol, or both within the year before enrollment. Pulmonary TB was diagnosed in 80% of participants (24% of these cases were cavitory), extrapulmonary TB was diagnosed in 29% of participants, and 79% of TB cases were microbiologically confirmed. Patients living in high-traffic-density neighborhoods were younger and more likely to be male, Asian, black, foreign-born, a recent immigrant, from southern California or the San Francisco Bay Area, living within the city limits, living in more densely populated and lower-income block groups, unemployed, homeless, a recreational drug/excess alcohol user, and HIV-infected compared with those with low traffic density exposure. A number of covariates were associated with increased mortality during TB treatment in unadjusted analysis, including being unemployed, having both pulmonary and extrapulmonary TB, having TB meningitis, having miliary TB, having microbiologically confirmed TB, and having the following comorbidities: diabetes, HIV, non-HIV immunosuppression, and end-stage renal disease (see Table S1).

Median (IQR) traffic volumes (vehicles/day) were 12,610 (5,000–24,800) vh/d at the nearest major road and 29,000 (17,600–45,000) vh/d at the highest-trafficked road in the 400-m buffer (Table 2). Median (IQR) traffic densities (vehicles-kilometer/hour) in the 100-to-400-m buffers were 106 (39.6–230), 356 (138–746), 820 (337–1,647), and 1,525 (690–3,009) vh·km/h, respectively. In sensitivity analyses, excluded patients were similar to included patients with the following exceptions: Those excluded were more likely to reside outside the city limits [257/3,608 (7.1%) vs. 687/32,816 (2.09%)], in lower population density census block groups [mean ± standard deviation (SD) 2,979 ± 4,656 persons/km<sup>2</sup> vs. 5,672 ± 6,198 persons/km<sup>2</sup>], in less traffic-dense neighborhoods [1,066 ± 2,724 vh·km/h in the 400-m buffer vs. 2,977 ± 4,193 vh·km/h], in the Central Valley [800/3,636 (22%) vs. 4,385/32,875 (13%)], and in higher income census block groups [69,467 ± 35,039 USD vs. 55,284 ± 28,847 USD].

**Table 1.** Characteristics by traffic density,  $n = 32,875$ .

Characteristic	Traffic density, $n$ (%) <sup>a</sup>		<i>p</i> -Value
	Low ( $n = 16,417$ ) <sup>b</sup>	High ( $n = 16,458$ ) <sup>b</sup>	
Age at diagnosis, median (IQR), y	47.7 (30.8–64.5)	46.2 (30.5–62.2)	<0.001
Male sex	9,570 (58.3)	10,012 (60.8)	<0.001
Race and ethnicity			<0.001
Asian	7,058 (43.0)	7,252 (44.1)	
Hispanic	6,525 (39.8)	6,075 (36.9)	
White, non-Hispanic	1,559 (9.50)	1,452 (8.82)	
Black, non-Hispanic	1,100 (6.70)	1,508 (9.16)	
Native American/Alaskan or Pacific Islander	144 (0.88)	137 (0.83)	
Unknown	31 (0.19)	34 (0.21)	
Foreign born	12,442/16,362 (76.0)	12,731/16,419 (77.5)	0.001
Recent immigrant	3,852/16,362 (23.5)	4,175/16,419 (25.4)	<0.001
Region			<0.001
Southern California	8,840 (53.9)	9,722 (59.1)	
Central Valley	3,106 (18.9)	1,279 (7.77)	
North Coast and Mountain	393 (2.39)	172 (1.05)	
San Francisco Bay Area	3,159 (19.2)	4,760 (28.9)	
Central Coast	919 (5.60)	525 (3.19)	
Residence within the city limits	15,823/16,376 (96.6)	16,306/16,440 (99.2)	<0.001
Population density, mean $\pm$ SD, persons/km <sup>2c</sup>	4,196 $\pm$ 3,452	7,144 $\pm$ 7,779	<0.001
Median annual household income, median (IQR), USD <sup>c</sup>	53,832 (38,230–74,836)	45,722 (31,680–65,214)	<0.001
Estimated census tract percent smoking prevalence, mean $\pm$ SD	14.9 $\pm$ 5.11	14.1 $\pm$ 5.46	<0.001
Unemployed	8,124/15,824 (51.3)	8,680/16,082 (54.0)	<0.001
Homeless	568/16,339 (3.48)	1,325/16,375 (8.09)	<0.001
Substance abuse <sup>d</sup>	1,776/16,193 (11.0)	2,370/16,263 (14.6)	<0.001
HIV-infected	648 (3.95)	985 (5.98)	<0.001
Microbiologically confirmed TB	12,868/16,409 (78.4)	12,995/16,451 (79.0)	0.21
Pulmonary TB	13,050 (79.5)	13,146 (79.9)	0.39
Extrapulmonary TB	4,754 (29.0)	4,886 (29.7)	0.15
Cavitary TB	3,221/16,399 (19.6)	3,145/16,443 (19.1)	0.24
Miliary TB	279 (1.70)	287 (1.74)	0.76
MDR-TB	194/12,664 (1.53)	180/12,891 (1.40)	0.37
Treatment	$n = 16,316$	$n = 16,358$	<0.001
Self-administered for all doses	2,857 (17.5)	3,020 (18.5)	
DOT for all doses	8,856 (54.3)	10,145 (62.0)	
DOT and self-administered doses	4,603 (28.2)	3,193 (19.5)	
Days to culture conversion, median (IQR)	53 (32–78)	53 (33–80)	0.33
Treatment duration, <sup>e</sup> median (IQR), days	244 (188–301)	248 (190–302)	0.37

Note: DOT, directly observed tuberculosis (TB) treatment; IQR, interquartile range; MDR-TB, multidrug-resistant TB, resistant to both isoniazid and rifampicin; SD, standard deviation; USD, U.S. dollars.

<sup>a</sup>Traffic density is defined as the sum of length-adjusted road segment traffic volumes in the 400-m buffer around residential addresses, using the median value of 1,525  $\text{vh} \cdot \text{km/h}$  as the low-high cutoff. Column values represent “Number (%)” unless indicated otherwise.

<sup>b</sup>The denominator is 16,417 for low traffic density and 16,458 for high traffic density unless otherwise indicated.

<sup>c</sup>Obtained from American Census Survey 2006–2010 census block group data (U.S. Census Bureau 2016).

<sup>d</sup>Excess alcohol and/or recreational drug use (oral, inhaled, or injected) within one year before TB diagnosis.

<sup>e</sup>Treatment duration among those who completed treatment.

Geocoding accuracy was slightly higher for patients enrolled between 2010 and 2011 (geocoding score of  $95 \pm 14$ ) than for those enrolled in other years of the study (geocoding score of  $93 \pm 19$ ;  $p < 0.001$ ). Patients reported with TB between 2010 and 2011 were less likely to be excluded because of an inability to geocode the residential address [ $n = 30/4,697$  (0.63%)] than those reported with TB in other years of the study [ $n = 706/31,079$  (2.2%)].

Traffic volumes and traffic densities were associated with increased mortality during TB treatment, with statistically significant *p*-trends for increasing HRs across quintiles for the majority of the 12 scenarios tested (Table 3; see also Table S1). For instance, participants exposed to the second quintile of nearest-road traffic volumes within the 100-m buffer had a 3% increased risk of death compared with those exposed to lowest-quintile traffic volumes [HR = 1.03 (95% CI: 0.86, 1.25)] (Figure 2). This risk increased incrementally to a 28% risk of death for the highest quintile of traffic volume exposure [HR = 1.28 (95% CI: 1.07, 1.53)], a trend that was statistically significant (*p*-trend = 0.002) (Figure 2). Furthermore, as the buffer size decreased, fifth-quintile mortality hazards increased, and trends across quintiles became more statistically significant. For example, although the

highest-trafficked road in the 400-m buffer was associated with increased mortality, the fifth-quintile HR was only 1.10 (95% CI: 0.95, 1.27), and there was not a significant trend for increasing mortality across quintiles. In contrast, fifth-quintile traffic volumes of the highest-trafficked road in the 100-m buffer were associated with a 29% increased hazard of death [HR = 1.29 (95% CI: 1.07, 1.55)], with a significant trend across traffic exposure quintiles (*p*-trend < 0.001).

Among potential confounders, age and recent immigration augmented mortality hazards, whereas sex, employment status, HIV status, population density, and excess alcohol and/or recreational drug use attenuated mortality hazards. In alternative Cox models, we adjusted for group-level covariates in addition to individual-level covariates (Table 4) and found that the addition of census block group household income did not attenuate HRs. In contrast, adjusting for census-tract smoking estimates augmented mortality hazards such that exposure to the highest quintile (Q5) of nearest-road traffic volumes in the 100-m buffer was associated with a 41% increased risk of death during TB treatment [HR = 1.41 (95% CI: 1.16, 1.75)] compared with lowest-quintile (Q1) traffic volume exposures. In additional alternative Cox models, the findings did not change significantly with exclusion of

**Table 2.** Traffic statistics by buffer and quintile.

Exposure	<i>n</i>	Mean	Median	IQR	Minimum	Maximum
Traffic volumes of nearest road (vh/24 h)						
100-m buffer						
Q1	3,866	2,077	2,082	1,210–2,830	32	3,900
Q2	3,815	6,087	5,960	4,940–7,290	3,906	8,680
Q3	3,865	12,242	12,000	10,245–14,100	8,686	16,500
Q4	3,815	21,283	21,314	18,700–23,650	16,521	26,500
Q5	3,830	54,292	35,056	30,000–45,000	26,508	382,000
All	19,191	19,160	12,000	4,940–23,600	32	382,000
200-m buffer						
Q1	5,621	2,110	2,110	1,210–2,910	32	3,950
Q2	5,573	6,212	6,060	5,000–7,410	3,956	8,930
Q3	5,622	12,595	12,475	10,500–14,450	8,938	16,827
Q4	5,580	21,971	22,000	19,400–24,473	16,840	27,700
Q5	5,573	60,158	36,929	30,795–48,200	27,725	382,000
All	27,969	20,563	12,400	5,000–24,400	32	382,000
300-m buffer						
Q1	6,212	2,119	2,130	1,210–2,920	32	3,960
Q2	6,246	6,249	6,100	5,000–7,500	3,961	9,000
Q3	6,246	12,785	12,646	10,700–14,700	9,010	17,000
Q4	6,157	22,227	22,285	19,700–24,700	17,004	27,879
Q5	6,201	61,321	37,208	31,000–49,275	27,887	382,000
All	31,062	20,898	12,600	5,000–24,700	32	382,000
400-m buffer						
Q1	6,523	2,134	2,130	1,210–2,990	32	4,000
Q2	6,434	6,283	6,126	5,000–7,500	4,001	9,030
Q3	6,486	12,799	12,673	10,702–14,700	9,035	17,000
Q4	6,457	22,272	22,300	19,700–14,800	17,004	27,900
Q5	6,463	61,578	37,300	31,246–49,700	27,903	382,000
All	32,363	20,985	12,610	5,000–24,800	32	382,000
Traffic volumes of highest-trafficked road (vh/24 h)						
100-m buffer						
Q1	3,912	2,682	2,600	1,481–3,860	32	5,000
Q2	3,821	8,080	7,960	6,470–9,610	5,020	11,500
Q3	3,856	15,748	15,700	13,600–17,700	11,530	20,300
Q4	3,951	25,099	24,789	22,600–27,740	20,301	30,300
Q5	3,758	74,191	41,853	35,000–61,000	30,358	382,000
All	19,298	24,876	15,700	6,390–27,500	32	382,000
200-m buffer						
Q1	5,684	3,956	3,958	2,290–5,600	32	7,500
Q2	5,701	11,842	11,900	9,500–14,089	7,503	16,500
Q3	5,642	20,969	21,100	18,800–23,300	16,504	25,200
Q4	5,643	30,623	30,000	27,725–33,400	25,231	37,800
Q5	5,651	106,348	54,600	42,500–164,000	37,839	382,000
All	28,321	34,677	21,000	9,480–33,300	32	382,000
300-m buffer						
Q1	6,286	5,633	5,500	3,260–8,000	32	10,809
Q2	6,322	16,253	16,400	13,666–18,841	10,810	21,300
Q3	6,332	25,612	25,600	23,450–27,700	21,303	29,865
Q4	6,260	35,908	35,540	32,356–39,400	29,892	44,004
Q5	6,308	138,578	124,000	52,718–210,000	44,012	382,000
All	31,508	44,410	25,600	13,700–39,400	32	382,000
400-m buffer						
Q1	6,566	7,929	7,890	4,530–11,690	59	14,600
Q2	6,607	20,247	20,499	17,600–22,806	14,623	25,000
Q3	6,639	29,249	29,120	27,142–31,269	25,009	34,000
Q4	6,514	41,116	40,333	37,208–45,300	34,022	52,000
Q5	6,549	168,797	165,000	100,000–229,000	52,200	382,000
All	32,875	53,332	29,000	17,600–45,000	32	382,000
Traffic density (vh · km/h)						
100-m buffer						
Q1	3,861	15.3	14.8	8.33–14.8	0.008	30.6
Q2	3,860	51.1	49.8	39.6–62.7	30.6	74.8
Q3	3,878	107	106	88.8–125	74.8	146
Q4	3,852	204	200	172–230	146	276

**Table 2.** (Continued.)

Exposure	<i>n</i>	Mean	Median	IQR	Minimum	Maximum
Q5	3,847	704	419	333–633	277	8,998
All	19,298	216	106	39.6–230	0.008	8,998
200-m buffer						
Q1	5,656	51.4	50.6	26.5–75.4	0.004	104
Q2	5,675	175	173	138–211	104	256
Q3	5,689	361	356	305–413	256	477
Q4	5,659	657	646	553–749	477	905
Q5	5,642	2,594	1,496	1,109–3,335	905	22,020
All	28,321	766	356	138–746	0.004	22,020
300-m buffer						
Q1	6,266	126	123	61.5–190	0.17	262
Q2	6,316	425	423	335–507	262	610
Q3	6,328	829	819	711–947	610	1,087
Q4	6,318	1,453	1,416	1,241–1,650	1,087	1,964
Q5	6,280	5,532	3,939	2,533–7,614	1,964	34,642
All	31,508	1,671	820	337–1,647	0.17	34,642
400-m buffer						
Q1	6,547	266	263	127–402	0.09	542
Q2	6,578	837	827	687–988	542	1,159
Q3	6,590	1,533	1,522	1,334–1,729	1,160	1,954
Q4	6,581	2,649	2,547	2,233–3,008	1,955	3,759
Q5	6,579	9,590	8,226	5,098–12,449	3,760	48,305
All	32,875	2,977	1,525	690–3,009	0.09	48,305

Note: IQR, interquartile range; Q1–Q5, quintiles 1 through 5, where Q5 is the highest quintile of traffic exposure; vh, vehicle.

cases that were not microbiologically confirmed or with exclusion of patients who lived at multiple addresses during treatment.

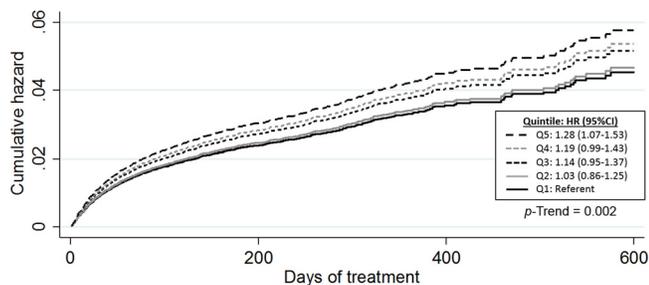
Region of residence, directly observed therapy, and census-tract smoking prevalence modified the effects of traffic exposure on mortality (Table 5). For instance, those exposed to the highest quintile of traffic density in the 100-m buffer had an 88% increased mortality hazard in the Central Valley and a 19% increased mortality hazard in southern California, but no increased hazard in the San Francisco Bay Area (*p*-interaction = 0.03). Those who self-administered some or all of their TB medication experienced an increased traffic mortality hazard compared with those who received only directly observed therapy, and traffic mortality

hazards were smaller for TB patients living in high-smoking census tracts compared with those living in low-smoking census tracts. Additionally, traffic effects were slightly higher in most buffers for patients >65 y old, in patients enrolled from 2004–2009, and in patients with diabetes (Table 5; see also Table S2); traffic effects were slightly lower for those with HIV and other forms of immunosuppression, but these interactions did not reach statistical significance except for other forms of immunosuppression in the 100-m buffer. We tested multidrug-resistant (MDR) TB and cavitation as possible confounders. Although both types of TB were associated with mortality (see Table S1), adjusting for these variables did not significantly change HRs in multivariable analyses

**Table 3.** Adjusted mortality hazard ratios (95% confidence intervals) for traffic metrics in 100–400-m buffers around residential addresses.

Exposure	100-m buffer	200-m buffer	300-m buffer	400-m buffer
Traffic volumes of nearest road (vh/24 h)				
<i>n</i>	18,396	26,814	29,753	30,985
Q1 (HR)	1.00	1.00	1.00	1.00
Q2 [HR (95%CI)]	1.03 (0.86, 1.25)	1.02 (0.87, 1.18)	1.01 (0.87, 1.17)	0.99 (0.85, 1.14)
Q3 [HR (95%CI)]	1.14 (0.95, 1.37)	1.13 (0.98, 1.32)	1.12 (0.97, 1.29)	1.10 (0.96, 1.26)
Q4 [HR (95%CI)]	1.19 (0.99, 1.43)	1.19 (1.03, 1.39)	1.20 (1.04, 1.38)	1.15 (1.00, 1.33)
Q5 [HR (95%CI)]	1.28 (1.07, 1.53)	1.19 (1.02, 1.38)	1.15 (1.00, 1.33)	1.13 (0.98, 1.30)
<i>p</i> -Trend	0.002	0.004	0.006	0.01
Traffic volumes of highest-trafficked road (vh/24 h)				
<i>n</i>	18,500	27,156	30,185	31,480
Q1 (HR)	1.00	1.00	1.00	1.00
Q2 [HR (95%CI)]	1.07 (0.88, 1.29)	1.22 (1.05, 1.42)	1.22 (1.06, 1.41)	1.21 (1.05, 1.40)
Q3 [HR (95%CI)]	1.29 (1.08, 1.55)	1.29 (1.10, 1.50)	1.30 (1.12, 1.50)	1.11 (0.96, 1.28)
Q4 [HR (95%CI)]	1.29 (1.07, 1.54)	1.32 (1.14, 1.54)	1.22 (1.05, 1.42)	1.14 (0.98, 1.32)
Q5 [HR (95%CI)]	1.29 (1.07, 1.55)	1.24 (1.06, 1.45)	1.15 (0.99, 1.34)	1.10 (0.95, 1.27)
<i>p</i> -Trend	<0.001	0.005	0.10	0.46
Traffic density (vh · km/h)				
<i>n</i>	18,500	27,156	30,185	31,480
Q1 (HR)	1.00	1.00	1.00	1.00
Q2 [HR (95%CI)]	1.02 (0.85, 1.23)	1.25 (1.08, 1.46)	1.11 (0.96, 1.29)	1.07 (0.93, 1.24)
Q3 [HR (95%CI)]	1.07 (0.89, 1.29)	1.25 (1.08, 1.46)	1.30 (1.12, 1.49)	1.23 (1.07, 1.42)
Q4 [HR (95%CI)]	1.33 (1.12, 1.59)	1.31 (1.13, 1.52)	1.17 (1.01, 1.36)	1.12 (0.97, 1.30)
Q5 [HR (95%CI)]	1.18 (0.98, 1.41)	1.17 (1.00, 1.37)	1.13 (0.97, 1.31)	1.06 (0.91, 1.23)
<i>p</i> -Trend	0.005	0.04	0.09	0.36

Note: CI, confidence interval; HR, hazard ratio; *p*-trend, the *p*-value for the trend across quintiles of traffic exposure. Q1–Q5, quintiles 1 through 5, where Q5 is the highest quintile of traffic exposure; vh, vehicle. Adjusted for age, sex, race, ethnicity, foreign birth, recent immigration within 5 y before tuberculosis (TB) diagnosis, population density, region of residence, unemployment within 1 y before TB diagnosis, homeless within one year before TB diagnosis, excess alcohol and/or recreational drug use within one year before TB diagnosis, and HIV infection.



**Figure 2.** Cumulative hazards of death by nearest-road traffic volume quintiles in 100-m buffers around residential addresses. Adjusted for age, sex, race, ethnicity, foreign birth, recent immigration within 5 y before tuberculosis (TB) diagnosis, population density, unemployment within one year before TB diagnosis, homelessness within one year before TB diagnosis, excess alcohol and/or recreational drug use within one year before TB diagnosis, and HIV infection. CI, confidence interval; HR, hazard ratio.

(see Table S4). In addition, we tested both variables as potential effect modifiers in the association between traffic density and mortality and did not find statistically significant interactions in any of the buffers (see Table S5). In mediation analyses, the effects of traffic exposure on mortality did not appear to be mediated through TB severity. Although surrogates for TB severity (smear positivity, having both pulmonary and extrapulmonary TB, TB meningitis, and miliary TB) were all significantly associated with increased mortality (see Table S1), traffic exposures were not statistically significantly associated with TB severity, and the addition of TB severity variables into Cox models did not attenuate mortality HRs. We also evaluated diabetes and end-stage renal disease as intermediaries on the causal pathway from TRAP exposure to mortality during TB treatment and found that although each comorbidity was associated with increased mortality (see Table S1), we did not find a statistically significant association between traffic density and increased risk for diabetes or end-stage renal disease.

In Kaplan–Meier survival analyses, the median time to culture conversion was 54 d (IQR, 32–83), and the median time to successful completion of treatment was 256 d (IQR, 190–309); these times did not differ significantly between lowest- and highest-quintile nearest-road traffic volumes in the 100-m buffer (see Table S3, Figure S1, and Figure S2). The 2-month culture conversion rates were 56% overall, 55% for patients exposed to lowest-quintile nearest-traffic volumes (in the 100-m buffer), and 57% for those exposed to highest-quintile traffic. The 9-month successful treatment completion rates were 54% overall, 55% for patients exposed to lowest-quintile traffic, and 53% for patients living near the highest-quintile traffic volumes.

## Discussion

To our knowledge, this is the first study to investigate the effects of traffic proximity on mortality during active TB treatment in a large, longitudinally followed patient cohort. We found that patients residing in high-traffic neighborhoods in California were at increased risk of death during TB treatment compared with those residing in low-traffic neighborhoods; we also found that there were statistically significant trends of increasing mortality across increasing quintiles of traffic exposure. Our findings were robust across several analytical models and after controlling for multiple demographic, socioeconomic, and clinical variables.

Our findings are consistent with a growing body of evidence linking traffic proximity and traffic-related air pollution with all-cause, cardiopulmonary, and cancer mortality (Beelen et al. 2008b; Cesaroni et al. 2012, 2013; Hart et al. 2011; Jerrett et al.

**Table 4.** Adjusted mortality hazard ratios (95% confidence intervals) for alternative Cox models using nearest-road traffic volume exposures, 100-m buffer.

Exposure: Traffic volumes of nearest road (vh/24 h)	Alternative Cox model					
	Cox model with only individual-level covariates <sup>a</sup>	Add group-level household income <sup>b</sup> to adjusted model	Add group-level smoking estimates <sup>c</sup> to adjusted model	Exclude non-microbiologically confirmed cases <sup>d</sup>	Exclude patients who moved within California <sup>e</sup>	Left-censored values: Assign zero instead of missing <sup>f</sup>
n	18,396	18,375	14,478	14,404	17,480	32,832
Q1 (HR)	1.00	1.00	1.00	1.00	1.00	1.00
Q2 [HR (95% CI)]	1.03 (0.86, 1.25)	1.03 (0.86, 1.24)	1.13 (0.91, 1.40)	1.05 (0.87, 1.28)	1.05 (0.87, 1.27)	1.00
Q3 [HR (95% CI)]	1.14 (0.95, 1.37)	1.14 (0.95, 1.36)	1.23 (1.00, 1.52)	1.16 (0.96, 1.40)	1.16 (0.96, 1.40)	0.90 (0.79, 1.02)
Q4 [HR (95% CI)]	1.19 (0.99, 1.43)	1.19 (0.99, 1.43)	1.29 (1.05, 1.59)	1.20 (1.00, 1.46)	1.21 (1.01, 1.46)	1.01 (0.90, 1.14)
Q5 [HR (95% CI)]	1.28 (1.07, 1.53)	1.27 (1.06, 1.52)	1.41 (1.16, 1.75)	1.28 (1.06, 1.54)	1.28 (1.07, 1.54)	1.10 (0.98, 1.23)
p-Trend	0.002	0.003	<0.001	0.004	0.002	0.04

Note: CI, confidence interval; HR, hazard ratio; p-Trend, the p-value for the trend across quintiles of traffic exposure; Q1–Q5, quintiles 1 through 5, where Q5 is the highest quintile of traffic exposure; vh, vehicle.

<sup>a</sup>Adjusted for individual-level covariates: age, sex, race, ethnicity, foreign birth, recent immigration within 5 y prior to tuberculosis (TB) diagnosis, population density, region of residence, unemployment within one year before TB diagnosis, homelessness within one year before TB diagnosis, excess recreational drug and/or alcohol use within one year before TB diagnosis, and HIV infection.

<sup>b</sup>Group-level covariate: census block group median annual household income.

<sup>c</sup>Group-level covariate: census tract smoking prevalence estimates.

**Table 5.** Adjusted mortality hazard ratios (95% confidence intervals) for highest quintile traffic density in each buffer, stratified by effect modifier.

Population	<i>n</i>	100-m buffer [HR (95% CI)]	200-m buffer [HR (95% CI)]	300-m buffer [HR (95% CI)]	400-m buffer [HR (95% CI)]
Entire cohort	31,480	1.18 (0.98, 1.41)*	1.17 (1.00, 1.37)*	1.13 (0.97, 1.31)	1.06 (0.91, 1.23)
Region					
San Francisco Bay Area	7,731	0.88 (0.61, 1.26)	0.99 (0.73, 1.36)	1.18 (0.87, 1.61)	0.93 (0.69, 1.25)
Central Valley	3,902	1.88 (1.09, 3.24)*	1.78 (1.12, 2.84)*	1.53 (1.02, 2.31)	1.27 (0.87, 1.87)
Southern California	17,935	1.19 (0.94, 1.50)*	1.13 (0.92, 1.39)	1.04 (0.85, 1.26)	1.07 (0.87, 1.30)
<i>p</i> -Interaction	—	0.03	0.19	0.52	0.85
Exclusive DOT					
No	13,098	1.21 (0.86, 1.70)*	1.35 (1.00, 1.82)*	1.47 (1.10, 1.96)*	1.39 (1.06, 1.83)*
Yes	18,207	1.12 (0.91, 1.40)	1.03 (0.86, 1.24)	0.93 (0.78, 1.12)	0.85 (0.71, 1.01)
<i>p</i> -Interaction	—	0.21	0.29	0.01	0.005
Smoking prevalence					
Low	12,861	1.38 (1.04, 1.84)*	1.48 (1.16, 1.89)*	1.41 (1.11, 1.79)*	1.28 (1.01, 1.62)*
High	12,241	1.14 (0.83, 1.55)	1.12 (0.86, 1.44)	1.03 (0.81, 1.31)	0.93 (0.73, 1.17)
<i>p</i> -Interaction	—	0.40	0.003	0.08	0.32
Age, y					
0–64	24,154	1.05 (0.77, 1.41)	1.11 (0.91, 1.53)	0.94 (0.73, 1.21)	0.87 (0.68, 1.11)
≥65	7,326	1.24 (0.98, 1.55)*	1.17 (0.96, 1.42)	1.19 (0.99, 1.43)*	1.14 (0.96, 1.37)
<i>p</i> -Interaction	—	0.63	0.84	0.45	0.47
Enrollment Year					
2000–2003	11,392	1.01 (0.77, 1.34)	1.04 (0.81, 1.33)	1.05 (0.83, 1.34)	0.98 (0.77, 1.24)
2004–2006	7,140	1.68 (1.14, 2.48)*	1.40 (1.01, 1.95)*	1.29 (0.94, 1.75)	1.01 (0.76, 1.36)
2007–2009	6,793	0.99 (0.67, 1.46)	1.31 (0.95, 1.82)	1.37 (0.99, 1.88)	1.35 (0.99, 1.85)
2010–2012	6,155	1.33 (0.83, 2.12)	1.05 (0.73, 1.50)	0.90 (0.65, 1.25)	1.00 (0.73, 1.38)
<i>p</i> -Interaction	—	0.07	0.69	0.84	0.44
HIV					
No	29,929	1.26 (1.04, 1.52)*	1.19 (1.01, 1.40)*	1.15 (0.98, 1.34)*	1.06 (0.91, 1.23)
Yes	1,551	0.60 (0.32, 1.14)	0.95 (0.55, 1.63)	0.88 (0.51, 1.51)	1.09 (0.59, 1.99)
<i>p</i> -Interaction	—	0.053	0.31	0.51	0.74

Note: CI, confidence interval; DOT, directly observed tuberculosis (TB) treatment; HR, hazard ratio. Adjusted for age, sex, race, ethnicity, foreign birth, recent immigration within 5 y before TB diagnosis, population density, region, unemployment 1 y before TB diagnosis, homeless within one year before TB diagnosis, excess alcohol and/or recreational drug use within one year before TB diagnosis, and HIV infection. The HR represents the highest quintile traffic density compared with the lowest quintile traffic density in each effect modifier level and buffer. \*A statistically significant trend (*p*-trend < 0.05) exists across quintiles.

2005, 2013; Thurston et al. 2015). Our study expands this evidence by focusing on a specific and potentially vulnerable cohort of patients using individual-level exposure and outcome data. Mortality during TB treatment remains high, as was evident in our study cohort and throughout the world (de Meer and van Geuns 1992; Fielder et al. 2002; Sterling et al. 2006). Although many factors have been implicated in these poor TB patient outcomes (Oursler et al. 2002), the effects of environmental factors such as ambient air pollution have remained uncertain. Our findings provide preliminary evidence that residential proximity to traffic could be an important modifiable risk factor for poor TB outcomes, and further studies investigating the effects of specific traffic-related pollutants are needed to confirm these observations.

Traffic effects were greatest and most statistically significant in the smaller buffers around residential addresses. Our observations are consistent with roadside air quality monitoring studies, which have found progressive decay of ambient pollutant concentrations down to background levels within 300–600 m, with the rate of decay over distance from the road dependent on vegetation barriers, wind direction, and type of pollutant (Karner et al. 2010; Nayeb Yazdi et al. 2015; Zhu et al. 2002). However, between-buffer comparisons should be made with caution in our study. Low-traffic neighborhood roads were not measured in this analysis and likely accounted for a larger proportion of missing data in smaller buffers, thus leading to potential differential misclassification of exposure between buffers.

There was a statistically significant dose–response relationship across many of the 12 exposure scenarios tested, as indicated by *p*-trends < 0.05. However, the traffic effect on mortality typically plateaued in the third or fourth quintile and often dropped slightly in the fifth quintile. It is possible that the traffic effect on mortality is not linear but rather plateaus with higher levels of exposure. The fifth-quintile exposure level also encompassed a

much larger exposure range than the other levels (Table 2), and this heterogeneity in exposure in the fifth quintile could have affected the results for this group.

Several interactions were evident from our analyses. Traffic effects were more pronounced in southern California and the Central Valley, where the background air quality is the poorest in the state (ALA 2016), and were also greater in patients living in low-smoking-prevalence census tracts. Evidence of smoking effect modification has been observed in several ambient air pollution health effects cohort studies, with some studies reporting augmented air pollution effects in smokers and others reporting diminished effects (Beelen et al. 2008a, 2008b; Blount et al. 2013; Puett et al. 2014). To our knowledge, there have been no studies examining the effect modification of smoking in the association between air pollution and TB outcomes, and more research is needed in this area using individual-level data because our smoking analysis was limited by aggregate smoking estimates. Traffic mortality hazards were significantly lower in those exclusively receiving DOT for the duration of their treatment than in those who received at least some self-administered dosing, suggesting that substandard adherence to therapy may accentuate the effects of traffic exposures in patients with active TB. We are unaware of other studies that have evaluated medication adherence and air pollution interactions, and further research is needed to corroborate these findings because air pollution and medication adherence may both be important risk factors for poor treatment outcomes among vulnerable groups with respiratory disease (Canino et al. 2006). Traffic effects were slightly higher in those ≥ 65 y old and in those with diabetes, raising concern that advanced age and certain chronic diseases could increase susceptibility to the harmful effects of air pollution. Further research is needed in these vulnerable populations.

Our observational study is hypothesis-generating for future studies. Through what biologic mechanisms did TRAP contribute

to mortality in patients undergoing TB treatment? It is possible that exposure to TRAP has deleterious effects on immunologic responses to *Mtb*, a theory supported by *in vitro* and animal studies. For instance, TB-infected rats chronically exposed to diesel exhaust particles (DEP) carried a higher mycobacterial burden than nonexposed controls (Hiramatsu et al. 2005), and *Mtb*-stimulated human peripheral blood monocytes exposed to DEP demonstrated suppression of key host antimycobacterial immune responses (Sarkar et al. 2012).

Our study has several strengths. TB patients were followed longitudinally with demographic, clinical, treatment, and mortality data collected prospectively by trained health care providers at the time of clinic visits, decreasing the likelihood of recall bias and reporting errors. This was a large cohort study focusing on a specific disease process for the entire state of California, where TB notification rates are estimated at >99% (Curtis et al. 2001), limiting the risk for potential selection bias. Medical fees were not a significant barrier to receipt of necessary TB care, which is covered by private health insurance, Medi-Cal, and county Department of Public Health programs, thereby also reducing loss to follow-up secondary to inability to pay for treatment.

Our study also has several limitations. Exposures with time dependency, namely traffic volumes, traffic density, and residential addresses, were assigned fixed measurement times and values, likely leading to exposure misclassification. For instance, we calculated traffic volumes and traffic densities from 2004 Caltrans data, reflecting traffic conditions near the midpoint of our study. This was a reasonable approach because spatial traffic gradients tend to remain stable over time, with a relatively homogenous increase in traffic volumes as population increases (Beelen et al. 2007, 2008a). Any exposure misclassification introduced would likely be nondifferential, biasing results toward the null. Indeed, traffic effects appeared to be smaller in patients enrolled before 2004 and after 2009, although this effect modification did not reach statistical significance. Geocoding was not performed concurrently with patient enrollment, but rather during the analysis, using 2010–2011 geocoding reference data sets. This difference in timing introduced slight inaccuracies in geocoding the addresses of patients enrolled before or after the reference period, but the overall accuracy of the match remained high (score 93/100) with the exclusion of only 2% of TB patients owing to an inability to geocode addresses. Traffic exposure estimates were limited to residential addresses and did not account for microenvironments such as household air pollution, indoor and outdoor pollution at work or school locations, or commuting patterns. We did not estimate exposures to specific traffic-related air pollutants such as PM<sub>2.5</sub>, NO<sub>2</sub>, or ozone. Any misclassifications of exposure were likely randomly distributed, resulting in an underestimate of true adverse effects of TRAP on TB treatment outcomes. We only considered baseline home addresses and did not exclude those who moved to different addresses within California during follow-up. However, exclusion of movers in alternative Cox models did not alter mortality hazards. Determining the hazards of traffic exposures on cause-specific mortality would have been informative, but these data were not available. The differences in baseline characteristics by traffic density are likely a reflection of inherent socioeconomic differences present in those living in highly polluted areas rather than a reflection of selection bias given the similarities in baseline characteristics between those excluded and those included in the final survival analyses. The California TB Registry provided a wealth of covariates but no individual-level direct measures of poverty or tobacco smoking. Poverty could contribute to poor outcomes at multiple stages in the TB management cascade, from late presentation for diagnosis to lack of access to medical care for

comorbidities to increased exposure to other risk factors. To address these limitations, in alternative models (Table 4), we adjusted for census block group household income and census-tract smoking prevalence, and we found that adjusting for group-level smoking augmented the effects of traffic exposure on mortality. However, to avoid misinterpretation of group-level data that could give rise to the ecological fallacy (Haneuse and Bartell 2011), we only employed aggregate data in alternative models. Traffic data were not available for small local roads, which likely led to an underestimation of traffic densities at residential addresses and to the exclusion of patients who did not live near major roads. Excluded patients were similar to those included except for factors that appeared to be related to rural living, and of the included patients, 98% lived within the city limits. As such, our findings are not generalizable to TB patients living outside of towns and cities.

## Conclusion

TB patients living in high-traffic neighborhoods were at increased risk for mortality during treatment even after controlling for demographic, socioeconomic, and clinical factors. Our findings suggest that TB patients are susceptible to the adverse health effects of traffic-related air pollution and that traffic exposure might be an important modifiable risk factor for poor TB treatment outcomes. These findings should be confirmed with additional studies to determine the effects of traffic-related pollutants, and of traffic measures in other settings, on tuberculosis outcomes.

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