

1 Risk factors among Black and White COVID-19 patients from a Louisiana Hospital System,
2 March, 2020 – August, 2021

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4 Qingzhao Yu¹, Wentao Cao¹, Diana Hamer², Norman Urbanek³, Susanne Straif-Bourgeois⁴,
5 Stephania Cormier^{5,6}, Tekeda Ferguson⁴, Jennifer Richmond-Bryant⁴

6

7 1. Biostatistics Program, Louisiana State University Health Sciences Center, New Orleans, LA
8 70112

9 2. Division of Academic Affairs, Our Lady of the Lake Regional Medical Center, Baton Rouge,
10 LA 70808

11 3. Department of Forestry and Environmental Resources, North Carolina State University,
12 Raleigh, NC 27695

13 4. Epidemiology Program, Louisiana State University Health Sciences Center, New Orleans,
14 LA 70112

15 5. Department of Biological Sciences, Louisiana State University, Baton Rouge, LA 70803

16 6. Pennington Biomedical Research Center, Baton Rouge, LA 70808

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18

19 Abstract

20

21 Objectives. To investigate relationships between race and COVID-19 hospitalizations, intensive
22 care unit (ICU) admissions, and mortality over time and which characteristics, may mediate
23 COVID-19 associations.

24

25 Methods. We analyzed hospital admissions, ICU admissions, and mortality among positive
26 COVID-19 cases within the ten-hospital Franciscan Ministries of Our Lady Health System
27 around the Mississippi River Industrial Corridor in Louisiana over four waves of the pandemic
28 from March 1, 2020 – August 31, 2021. Associations between race and each outcome were
29 tested, and multiple mediation analysis was performed to test if other demographic,
30 socioeconomic, or air pollution variables mediate the race-outcome relationships.

31

32 Results. Race was associated with each outcome over the study duration and during most waves.
33 Early in the pandemic, hospitalization, ICU admission, and mortality rates were greater among
34 Black patients, but as the pandemic progressed these rates became greater in White patients.
35 However, Black patients were still disproportionately represented in these measures. Age was a

36 significant mediator for all outcomes across waves, while comorbidity and emissions of
37 naphthalene and chloroprene acted as mediators for the full study period.

38
39 Conclusions. The role of race evolved throughout the pandemic in Louisiana, but Black patients
40 bore a disproportionate impact. Naphthalene and chloroprene air pollution partially explained the
41 long-term associations. Our findings imply that air pollution might contribute to the increased
42 COVID-19 hospitalizations and mortality among Black residents in Louisiana but likely do not
43 explain most of the effect of race.

44
45 **What is already known on this topic:** Early in the pandemic, there was evidence of disparities
46 in COVID-19 cases, hospitalizations, intensive care unit (ICU) admissions, and mortality due to
47 race. Studies were emerging to indicate that strength of these relationships was waning over
48 time.

49 **What this study adds:** This study tests relationships between race and hospitalizations, ICU
50 admissions, and mortality and finds that Black patients continue to be disproportionately
51 represented, although that inequity diminished over time. This study, the first to use multiple
52 mediation analysis to study COVID-19 associations, suggests that the relationship between race
53 and health outcome can be explained by mediators including age and, to a lesser extent,
54 comorbidity and air pollution.

55 **How this study might affect research, practice or policy:** This study supports the need for
56 healthcare resources to be available to Louisiana's communities of color, for policy to support
57 increased access to health care in the Industrial Corridor region, and for policy to support the
58 reduction of air pollution emissions to disproportionately impacting the health of the Industrial
59 Corridor's communities of color.

60
61

62 Introduction

63

64 Coronavirus Disease 2019 (COVID-19) severity and mortality have been associated with several
65 vulnerability factors, including comorbidities, environmental exposures, natural disasters,
66 sociodemographic factors, and residence in congregate settings[1,2]. During the first wave of
67 COVID-19 cases in the U.S., transmission among residents of congregate settings was
68 responsible for disease spread[3], while comorbidities among older residents likely elevated risk
69 of death[4]. The second wave of COVID-19 cases in the U.S. saw disproportionate numbers of
70 severe disease and deaths among Black, LatinX, Native American, and immigrant population
71 groups[2,5,6]. The third wave may have occurred in part due to asymptomatic transmission in
72 congregate settings including prisons and long-term care facilities, disproportionately impacting
73 Black and LatinX populations[2].

74

75 Soon after the start of the pandemic, some evidence emerged of an association between long-
76 term average air pollution concentrations and prevalence or severity of COVID-19. Notably,
77 significant associations were observed for long-term average particulate matter (PM) having
78 diameter smaller than 2.5 μm (PM_{2.5}) concentration with COVID-19 infection[7-9], COVID-19
79 prevalence[10], intensive care unit (ICU) admission[11,12], ventilator use[12], and
80 mortality[7,11-13]. Associations were also observed for long-term average diesel PM
81 concentration estimates for COVID-19 prevalence and mortality[7]; average nitrogen dioxide
82 (NO₂) concentrations for prevalence[9,10,14], hospitalization[12], ICU admission[12], ventilator
83 use[12], and mortality[12,14]; ozone (O₃) concentration for mortality[12]; and hazardous air
84 pollutant indices for respiratory and immunological hazard and mortality[15]. Chen et al.[12]
85 also calculated associations with hospitalization, ICU admission, ventilator use, and mortality for
86 1-month average concentrations of PM_{2.5} and NO₂. However, evidence was mixed, with some
87 studies showing no association for NO₂[11]; O₃[7,9,10,14]; or PM_{2.5}[14,15].

88

89 Although many studies suggested a relationship between air pollutant concentration and COVID-
90 19 outcomes, these studies primarily occurred early in the pandemic. Less is known about the
91 association between air pollutant exposure and COVID-19 over time. Sidell et al.[9] studied how
92 the relationship between air pollution and COVID-19 infection changed in a Southern California

93 cohort over four waves spanning March 1, 2020 through February 28, 2021. They observed
94 associations to persist for each wave and the entire duration of their study for both 1-month
95 average and 1-year average PM_{2.5} and NO₂ concentrations and between COVID-19 infection and
96 1-year average O₃ concentrations for the second, third, and fourth waves and entire study
97 duration. However, the magnitude of the associations declined over the third and fourth waves,
98 especially for PM_{2.5}. Uncertainties persist about the influence of air pollution on COVID-19
99 outcomes over the course of the pandemic.

100

101 Strategies to respond effectively to public health emergencies such as the COVID-19 pandemic
102 require understanding potential causal pathways for disease outcomes[16,17]. Mediation models
103 can be useful to test how conditions present in populations may influence disease status. For a
104 proposed causal pathway, a directed acyclic graph (DAG) may be used to represent the proposed
105 direct exposure-response pathway and mediating factors that comprise the total effect. For each
106 pathway, linear regression models would then be assigned to represent each pathway in the
107 DAG, and total effect includes the effect of each mediator and direct effect of the predictor.
108 Disparities in COVID-19 outcomes by race combined with evidence about the relationship
109 between COVID-19 and comorbidities, insurance status, and pollution exposure led to the
110 hypothesis that there is a causal pathway between race and COVID-19 mediated by
111 comorbidities, insurance status, and pollution exposure (Supplemental Figure A).

112

113 Louisiana parishes routinely score well below the national average for quality of life, morbidity,
114 and mortality indices such as low birthweight, child poverty, and median household income[18].
115 Based on most recently available data, Louisiana ranks 46th among the states in air quality given
116 by average daily PM_{2.5}, 47th in percent smokers among adults, and 45th in COVID-19 death rate.
117 For the period of March 1, 2020 – August 31, 2021, 37.7% of Louisiana’s COVID-19 deaths
118 occurred in people identifying as non-Hispanic Black (hereafter referred to as “Black
119 patients”)[19]; in 2020, that proportion was 41.7%, compared to 31.2% of Louisiana residents
120 identifying as Black[20]. A recent analysis connected disparities, systemic racism, economic
121 stress, and COVID-19 mortality[21].

122

123 Given the disproportionate impact of COVID-19 on communities of color in Louisiana and the
124 U.S., the goals of this research are to investigate the relationships of race and COVID-19
125 outcomes over time and to identify which characteristics, if any, may mediate associations of
126 race with COVID-19 hospitalizations, ICU admissions, and mortality, using individual-level data
127 from a Louisiana hospital system. We investigate factors including race, insurance status,
128 comorbidity, and pollutant exposure for four waves of COVID-19 between March 1, 2020 and
129 August 31, 2021. Identification of differences among these patient cohorts will enable improved
130 approaches to prevention and treatment of COVID-19 for each cohort and may identify structural
131 vulnerabilities for future pandemics.

132

133 Methods

134

135 We used the Franciscan Missionaries of Our Lady Health System (FMOLHS) COVID-19
136 registry to identify patients at ten locations. 13,454 patients ages eighteen years or older who
137 tested positive by a polymerase chain reaction (PCR) COVID-19 test were identified using the
138 Epic healthcare software between March 1, 2020 and August 31, 2021. This period is broken
139 down by waves: March 1 – June 10, 2020, June 11 – October 6, 2020, October 7, 2020 – June
140 30, 2021, and July 1 – August 31, 2021. These waves were chosen to minimize both cases and
141 mortality at the beginning and end of each period using the Johns Hopkins database for
142 Louisiana[22]. Patient-level variables included hospital department, COVID-19 test date,
143 COVID-19 test result, age, insurance status (private insurance, Medicaid, Medicare, and self-
144 pay), race, ethnicity, sex, admission date, discharge date, length of hospital stay, admission
145 status, ICU stay, ICU admission date, ICU discharge date, length of ICU stay, discharge
146 dispatch, body mass index (BMI), comorbidities, census tract, and census block group. To
147 minimize bias in the patient database, negative PCR tests were not included in the database
148 because tests were often obtained for non-medical reasons (e.g., work, travel, recreation).

149

150 Air pollution burden calculations were based on Mikati et al.[23]. Absolute burden for each
151 respiratory hazardous air pollutant was calculated by census tract as the weighted average of the
152 emissions over the block groups within each tract. Air pollutant emissions for the state of
153 Louisiana were obtained from the 2017 National Emissions Inventory[24], and data for the

154 census block groups and census tracts, including shape files and demographic characteristics,
155 were obtained from the 2015-2019 American Community Survey[25]. Air pollutants included
156 PM_{2.5} and hazardous air pollutants (HAPs) known to have respiratory health effects: 1,3-
157 dichloropropene, 2,4-toluene-diisocyanate, acetaldehyde, acrolein, acrylic acid, arsenic,
158 beryllium, cadmium, chlorine, chloroprene, chromium, diesel PM, formaldehyde,
159 hexamethylene-1,6-diisocyanate, hydrazine, hydrochloric acid, naphthalene, nickel, polycyclic
160 organic matter (POM), propylene, and triethylamine. Oil and gas wells and refineries, prevalent
161 naphthalene sources, and a neoprene plant, a chloroprene source, fall within the hospital service
162 area (Supplemental Figure B). We used the R software v4.0.5 for data organization (packages
163 *dplyr*, *tidyr*, *bit65*, and *data.table*) and for the merger of geographic data with air pollution
164 emissions data and output of shape files containing emissions burdens (packages *tigris*, *Hmisc*,
165 *sp*, and *rgdal*). Emissions burdens were then assigned to 12,031 individual COVID-19 patients in
166 the FMOLHS database for their census tract of residence. Bias minimization related to spatial
167 assignment of emissions burdens is described in Mikati et al.[23].

168
169 Differences in population characteristics, including air pollutant burden, were first illustrated
170 using summary statistics. Direct relationships of race with other demographic variables (age, sex,
171 BMI, comorbidities, insurance status) or with disease-related variables (hospital admission, ICU
172 admission, mortality) were tested via χ^2 or ANOVA for categorical or continuous variables,
173 respectively. Patient status was determined using hospital data for admission status, length of
174 hospital stay, ICU status, and length of ICU stay. P-value < 0.05 for the χ^2 or ANOVA test
175 signified a significant difference between Black and White COVID-19 patients. Because
176 comparisons were limited to these two racial groups, the final sample size was 11,331.

177
178 Associations of race with COVID-19 outcomes (hospital admissions, ICU admissions, mortality)
179 were tested via a multistep process. First, a cross-table was created and χ^2 testing was performed
180 to analyze the relationship between race and health outcome. Mediation analysis was then used
181 to test if a portion of the race-outcome relationship could be accounted for by an intermediate
182 variable[26-28]. Potential mediators and covariates in the association between race and health
183 effect were identified by testing for significant associations between each of the other variables
184 with race and health effect, respectively. Associations of each variable with both race and health

185 effect indicated that the variable is a potential mediator and so would be included in the
186 mediation analysis. Variables associated with just health effect but not with race were identified
187 as covariates and controlled in the mediation analysis. Significant mediators with the same sign
188 as the total effect were considered as part of the racial differences explained by the mediator,
189 while those with opposite sign suggested greater racial differences in effect considering the
190 mediator. The R package *mma* was used to perform the mediation analysis[29].

191
192 We confirmed each of the criteria listed under for the STrengthening the Reporting of
193 OBservational Studies in Epidemiology checklist for cross-sectional studies during completion
194 of this manuscript.

195
196 Results

197
198 Of the 11,331 patients, 5708 (50.4%) identified as non-Hispanic Black, and 5623 (49.6%)
199 identified as non-Hispanic White (Table 1). In comparison, 33.8% of the population of Louisiana
200 census tracts associated with patients' residential addresses (referred to hereafter as the "patient
201 population") identified as Black, and 58.8% identified as non-Hispanic White. Census tract
202 population data were available for 89% of patients. 6210 (54.8%) cases identified as Female, and
203 5119 (45.2%) identified as Male. On average, Black patients were 7.9 years younger than White
204 patients. Black patients had a higher average BMI, but average BMI for both groups was in the
205 obese range (BMI > 30). Length of hospital and ICU stays were both significantly higher among
206 White patients, although that difference diminished for Medicare recipients and those without
207 insurance. More Black patients had Medicaid or were uninsured, while more White patients had
208 private insurance or Medicare. Among the twenty-two pollutants tested, emissions burden was
209 significantly higher for Black patients in seventeen compounds and for White patients in three
210 compounds, with no significant difference for two pollutants.

211
212 For the study duration, hospital admissions were significantly higher among White patients,
213 while ICU admissions were significantly higher among Black patients. Compared to their share
214 of the patient population, Black patients were overrepresented among hospital admissions by
215 28%, among ICU admissions by 43%, and among total COVID-19 patients by 38% (Table 2).

216 Hospital and ICU admissions significantly exceeded the share of the population for Black
217 patients by 86% and 89%, respectively during the first wave and by 40% and 56%, respectively
218 during the second wave. By the third wave, the proportions of hospital and ICU admissions were
219 higher among White patients with a significant χ^2 , but the proportion of hospital and ICU
220 admissions among Black patients were 16% and 36% greater than the share of the population
221 identifying as Black.

222
223 Information regarding mortality (patients who expired while at the hospital or within 7 days of
224 discharge) was available for 11,032 (97.3%) cases (Table 2). For the study duration, the
225 proportion of those who died was significantly higher for White patients, based on the χ^2 test, but
226 the proportion of Black patients who died was still 25% greater than the proportion of Black
227 people in the Louisiana census tracts sending patients to FMOLHS. The proportion of patients
228 who died was nearly 65% for Black patients during the first wave, with the share of the patient
229 population that is Black overrepresented by 78% but was significantly higher for White patients
230 during the second and third waves and not significantly different in the fourth wave. During the
231 second wave, mortality among Black patients was still 28% higher than the share of patient
232 population identifying as Black.

233
234 Age and, with smaller contribution, comorbidities were significant mediators of the race-
235 hospitalization relationship (Figure 1) for the entire study period. Age and comorbidities were
236 also consistently significant mediators for each wave. Naphthalene and arsenic were significant
237 mediators of the race-hospitalization relationship for the duration of the study. Naphthalene was
238 not a significant mediator for any of the waves, and arsenic was only for the fourth wave. PM_{2.5}
239 and chromium added uncertainty to the race-hospital admissions relationship, because the
240 different sign of these mediation coefficients widened the confidence intervals around the total
241 effect. Cadmium added uncertainty to the race-hospitalization relationship for the third wave.
242 Several studies[7,11-13] found associations of PM_{2.5} with COVID-19 using data from the first
243 few months of the pandemic, but they either used a nationwide domain or studied different parts
244 of the country. Terrell and James[15] calculated a correlation of 0.21 for PM_{2.5} concentration
245 with COVID-19 mortality for Louisiana, and Xu et al.[30] noted for a study of COVID-19 in
246 Texas that PM_{2.5} concentrations were not associated with COVID-19 mortality.

247
248 The model for race-ICU admission for the entire study period (Figure 2) included a direct effect
249 that was larger than and opposite in sign to total effect, widening the confidence interval around
250 total effect to suggest uncertainty. This model also contained age as a mediator and, with smaller
251 magnitude, comorbidity and sex. Chloroprene and naphthalene emissions were both significant
252 mediators, while PM_{2.5} and chromium emissions appeared to widen the confidence intervals
253 around total effect. Age was a mediator of the race-ICU admission effect during each wave.
254 During the third wave, the total effect between race and ICU admission was near zero, but there
255 was a positive direct effect and positive indirect effect of PM_{2.5} emissions balanced by negative
256 indirect effects of age, cadmium emissions, and nickel emissions. The fourth wave produced a
257 large total effect for the race-ICU admission model that included a direct effect comprising more
258 than half of the total effect and indirect effects from age, insurance status, sex, and emissions of
259 POM.

260
261 Mediation analysis results indicate that, for the total duration and each wave, age was
262 consistently a significant mediator of the race-mortality relationship (Figure 3), consistent with
263 Cronin and Evans[31]'s finding of higher COVID-19 mortality for Black males and females for
264 every age group (0-44 y, 45-64 y, 65-74 y, 75+ y) with a greater effect of age than race or sex.
265 Sex and comorbidities had smaller indirect effects for the entire study period but were still
266 significant. Naphthalene was identified as a mediator of the race-mortality relationship for the
267 total duration, while hydrochloric acid added uncertainty to the assessment of mediation.
268 Naphthalene was identified as a potential mediator during the first wave but was not significant
269 and added uncertainty to that model. POM was a significant mediator of the race-mortality
270 relationship during the fourth wave. POM emerged as a potential mediator in the total duration
271 model but was of small magnitude.

272
273 There were some limitations specific to this dataset. We used data from one hospital system.
274 This selective population was not representative of all Louisiana COVID-19 hospitalizations and
275 could have imposed selection bias. The most recent HAP emission data were from 2017.
276 Additionally, vaccination status was not included in the dataset but could have affected severe
277 outcomes during the last two waves.

278

279 Mediation analysis showed a clear relationship between race and outcome at the beginning of the
280 pandemic, but race appeared less influential over time. Mediation analyses highlighted the
281 uncertainty in the race-outcome relationships across waves. Although several air pollutants were
282 associated with race, with higher emissions burdens among predominantly Black census tracts,
283 air pollution did not appear to consistently mediate the race-outcome relationship for most
284 waves. Uncertainties in the mediation analyses raise questions about unmeasured confounding.
285 VanderWeele[32] asserted four necessary assumptions for mediation analysis: 1) control for
286 confounding of the exposure-outcome relationship, 2) control for confounding of the mediator-
287 outcome relationship, 3) control for confounding of the exposure-mediator relationship, and 4)
288 no confounder of the mediator-outcome relationship is affected by the exposure. The first three
289 were accomplished through the process of checking for significant associations among the
290 exposure, potential mediator, and outcome. However, the final assumption is more difficult to
291 enforce for this study given that long-standing racialization may introduce other, uncontrolled
292 factors[33]. Similarly, it is difficult to ascertain whether any mediators were omitted from the
293 analysis. Additionally, exposure measurement error or exposure misclassification has the
294 potential to weaken the associations between the exposure and mediators. In the case of the HAP
295 burdens, Mikati et al.[23] sought to control this by testing different assignment radii and found
296 little difference. Use of census tract-level assignments also helps to localize the exposure
297 estimates.

298

299 Conclusions

300

301 A complicated picture of relationships between race and COVID-19 hospitalizations, ICU
302 admissions, and mortality emerges from these results. For the entire study period, hospitalization
303 and mortality rates among those diagnosed with COVID-19 were greater for White patients than
304 for Black patients, while ICU admission rates were higher for Black patients. These proportions
305 shifted towards White patients and were significant by late 2020. But, the proportion of those
306 diagnosed with COVID-19 as well as those hospitalized, admitted to the ICU, and died remained
307 disproportionately high for Black patients compared with the patients' residential areas, despite
308 the 10-year age difference between Black and White patients.

309

310 Among the population of hospitalized COVID-19 patients, most of the effect of race could be
311 explained by mediators. Age was the strongest mediator, accounting for the largest share of the
312 indirect effect. In each wave, the average age of Black patients was 8-9 years younger than the
313 average age of White patients. In fact, life expectancy for Black Louisiana residents is 3.4 years
314 shorter than for White Louisiana residents[34]. These factors make it difficult to disentangle the
315 effect of race from the effect of age.

316

317 Findings that naphthalene and chloroprene acted as mediators of race for ICU admissions and
318 that naphthalene acted as a mediator for hospitalizations and mortality were not surprising
319 because their burdens among Black patients were 8.9 and 4.5 times higher, respectively than for
320 White patients. Our results imply that policies to improve environmental conditions – especially
321 among Louisiana’s predominantly Black communities – may have lessened inequities in
322 COVID-19 impacts.

323

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331

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- 424

425 Table 1. Characteristics of the study population. When differences were statistically different, the
 426 characteristic and its highest value were shown in bold.

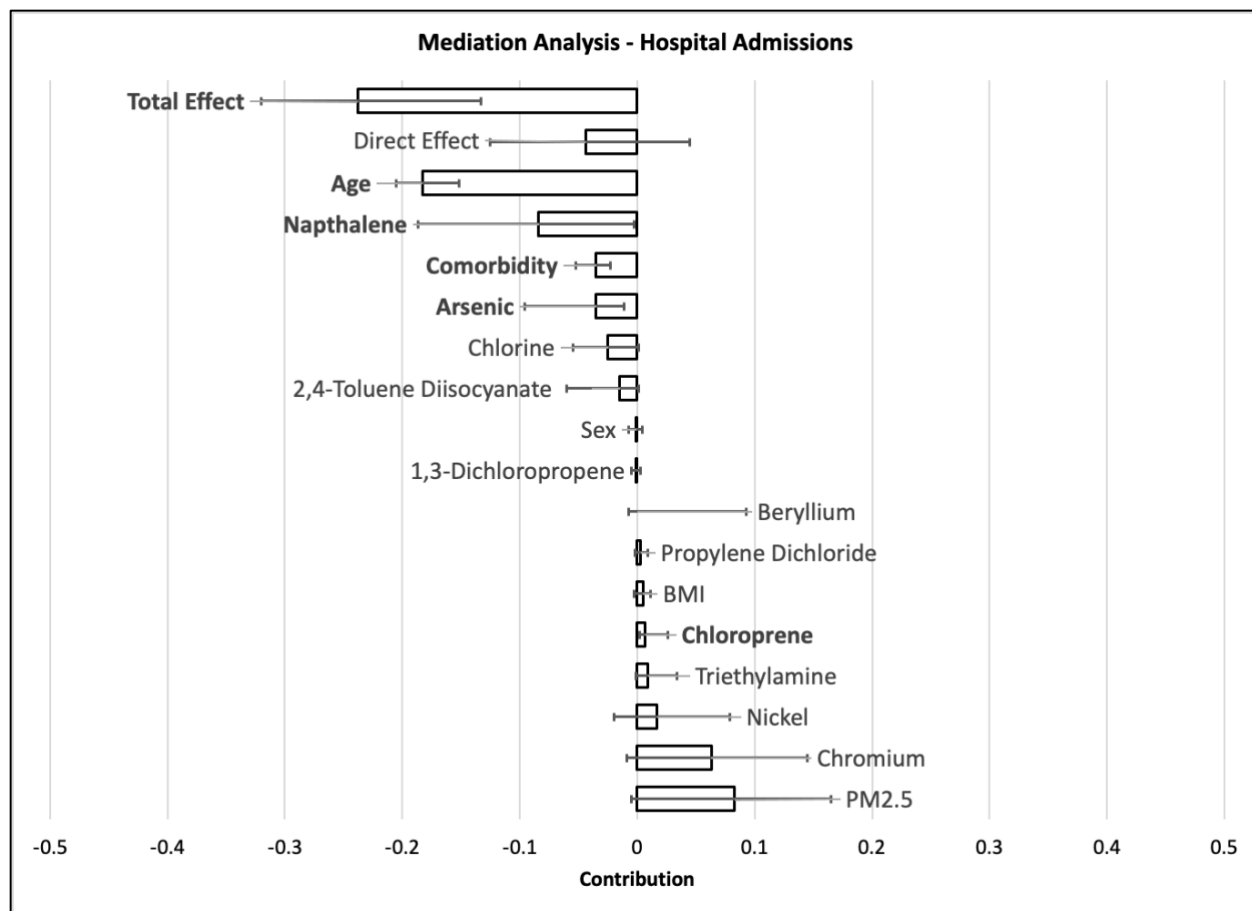
	Black	White	
N	5708	5623	
	mean	mean	p ⁺
Age	48.5	56.4	<2x10 ⁻¹⁶
Hospital length of stay (d)	5.70	6.82	2.40x10 ⁻⁶
ICU length of stay (d)	5.75	7.96	7.13x10 ⁻⁷
BMI	31.7	30.6	0.128
Hazardous Air Pollutants ⁺			
Dichloro:1,3-dichloropropene	0.104	0.233	1.28x10 ⁻¹⁰
2,4-toluene	0.0324	0.0143	9.10x10 ⁻⁸
Acetaldehyde	4120.	4640.	1.29x10 ⁻¹³
Acrolein	495.	272.	<2x10 ⁻¹⁶
Acrylic Acid	50.8	19.1	1.62x10 ⁻⁸
Arsenic	2.33	1.00	<2x10 ⁻¹⁶
Beryllium	0.338	0.109	<2x10 ⁻¹⁶
Cadmium	3.63	0.842	<2x10 ⁻¹⁶
Chlorine	3680	216.	<2x10 ⁻¹⁶
Chloroprene	231.	50.9	<2x10 ⁻¹⁶
Chromium	2.46	0.443	<2x10 ⁻¹⁶
Diesel PM	0.293	0.0481	<2x10 ⁻¹⁶
Formaldehyde	4430.	1830.	<2x10 ⁻¹⁶
HCl	21700	2720.	<2x10 ⁻¹⁶
Hexamethylene 6-diisocyanate	0.410	2.89	<2x10 ⁻¹⁶
Hydrazine	0.00231	0.0152	0.565
Naphthalene	2280.	255.	<2x10 ⁻¹⁶
Nickel	95.3	23.1	<2x10 ⁻¹⁶
PM2.5	83.0	16.1	<2x10 ⁻¹⁶
Polycyclic organic matter	0.0231	0.00793	1.47x10 ⁻⁵
Propylene	8.90	6.44	0.289
Triethylamine	46.6	19.0	<2x10 ⁻¹⁶
+ p-values were calculated using square root transformed data to normalize the data distribution			

427
 428

429 Table 2. Count tables for χ^2 analysis for each wave of the study and for the full study period.
 430 Equitable Black and equitable White indicate the ratio of the share of the population of patients
 431 in each group compared to the number of patients that should be in each group based on the
 432 proportion of each group in Louisiana census tracts sending patients to the FMOLHS.

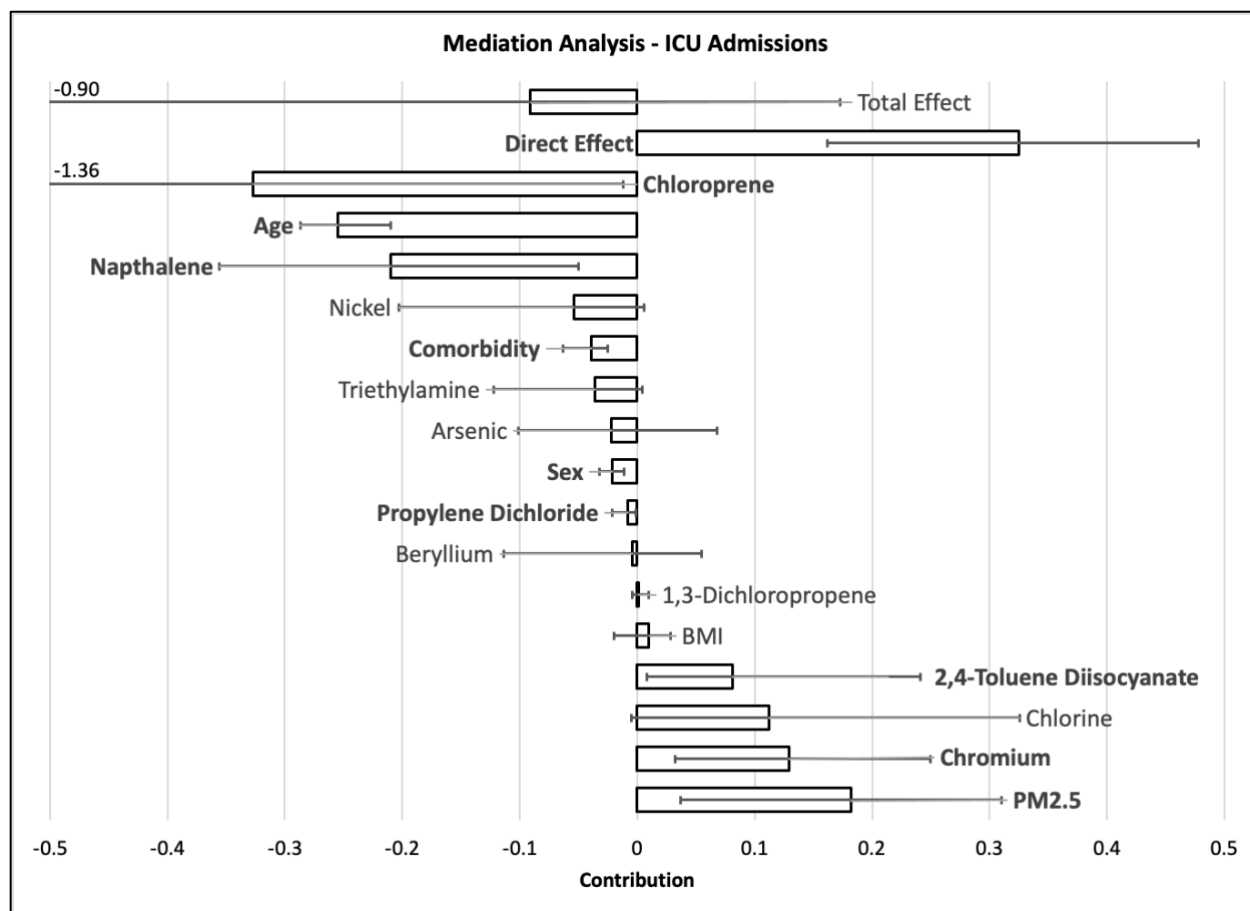
	HA	ICU	No HA	Death	No Death
First Wave: March 1, 2020-June 10, 2020					
Black	256	192	314	107	655
Equitable Black	1.86	1.89	2.10	1.78	1.99
White	121	86	96	58	245
Equitable White	0.51	0.49	0.37	0.55	0.43
p	0.0149			0.0475	
Second Wave: June 11, 2020-October 6, 2020					
Black	307	188	645	58	1082
Equitable Black	1.40	1.56	1.64	1.28	1.57
White	294	143	433	66	804
Equitable White	0.77	0.68	0.63	0.84	0.67
p	2.43×10^{-3}			0.0269	
Third Wave: October 7, 2020-June 30, 2021					
Black	619	287	1178	83	1921
Equitable Black	1.16	1.36	1.42	0.98	1.34
White	843	290	1089	148	2015
Equitable White	0.91	0.79	0.76	1.01	0.81
p	5.47×10^{-8}			1.85×10^{-4}	
Fourth Wave: July 1, 2021-August 31, 2021					
Black	439	59	1224	41	1590
Equitable Black	1.16	0.81	1.24	0.99	1.18
White	601	141	1486	72	2087
Equitable White	0.91	1.11	0.86	1.00	0.89
p	5.31×10^{-5}			0.169	
Full Study Period: March 1, 2020-August 31, 2021					
Black	1621	726	3361	289	5248
Equitable Black	1.28	1.43	1.42	1.25	1.38
White	1859	660	3104	344	5151
Equitable White	0.84	0.75	0.76	0.86	0.78
p	5.04×10^{-7}			0.0209	

433



434
435 Figure 1: Mediation analysis results for hospital admissions. Whiskers indicate the 95%
436 confidence interval around the mediation effect, with each tested mediator shown by a column.
437 Statistically significant effects are bolded.

438



439

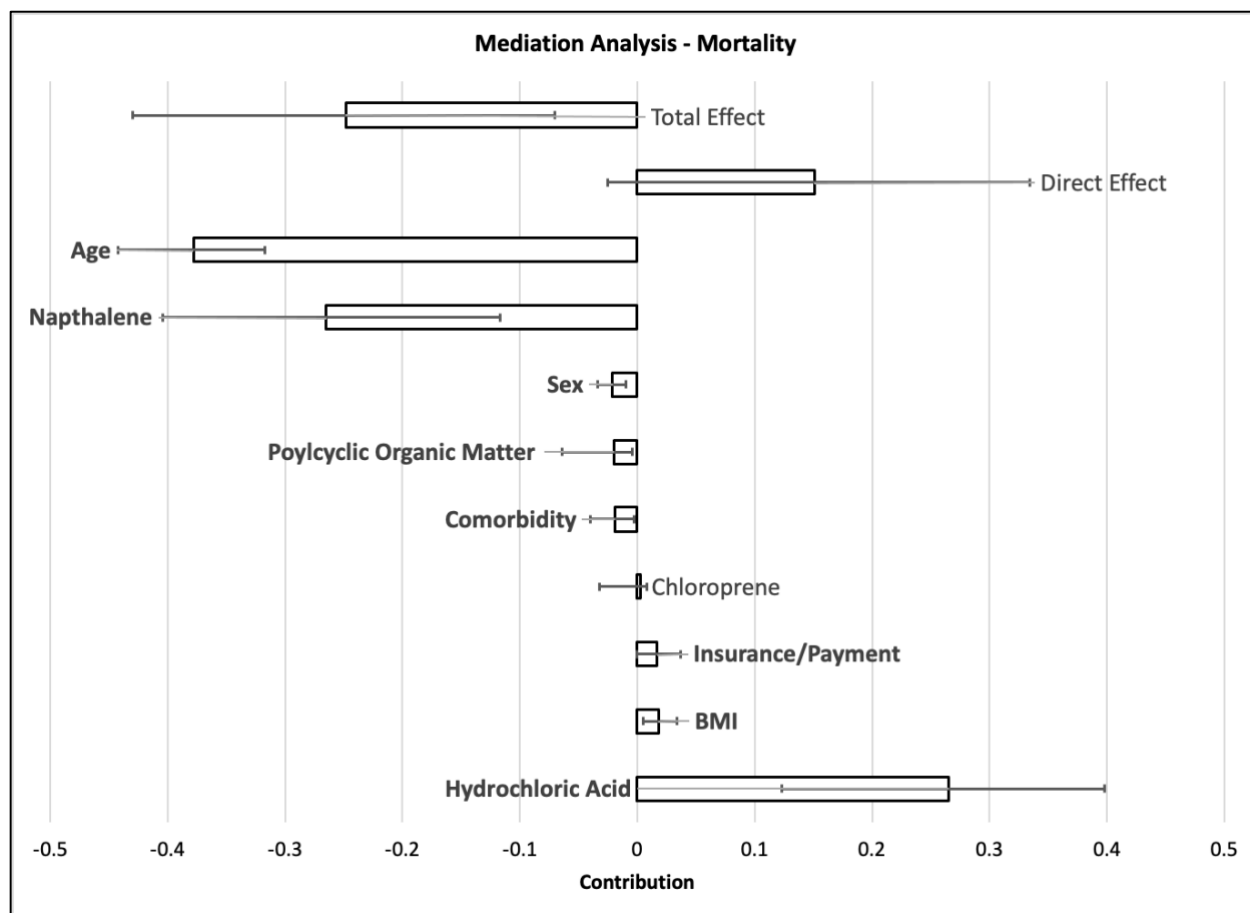
440 Figure 2: Mediation analysis results for ICU admissions. Whiskers indicate the 95% confidence

441 interval around the mediation effect, with each tested mediator shown by a column. Where the

442 lower confidence interval goes beyond the data range shown on the page, the lower bound is

443 provided numerically on the graph. Statistically significant effects are bolded.

444



445
446 Figure 3: Mediation analysis results for mortality. Whiskers indicate the 95% confidence interval
447 around the mediation effect, with each tested mediator shown by a column. Statistically
448 significant effects are bolded.