 exposur e Particulate Matter Air Pollution and Anosmia
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Abstract

IMPORTANCE Anosmia, the loss of the sense of smell, has profound implications for patient safety, well-being, and quality of life, and it is a predictor of patient frailty and mortality. Exposure to air pollution may be an olfactory insult that contributes to the development of anosmia.

OBJECTIVE To investigate the association between long-term exposure to particulate matter (PM) with an aerodynamic diameter of no more than 2.5 μm (PM 2.5) with anosmia.

DESIGN, SETTING, AND PARTICIPANTS This case-control study examined individuals who presented from January 1, 2013, through December 31, 2016, at an academic medical center in Baltimore, Maryland. Case participants were diagnosed with anosmia by board-certified otolaryngologists. Control participants were selected using the nearest neighbor matching strategy for age, sex, race/ethnicity, and date of diagnosis. Data analysis was conducted from September 2020 to March 2021.

EXPOSURES Ambient PM 2.5 levels.

MAIN OUTCOMES AND MEASURES Novel method to quantify ambient PM 2.5 exposure levels in patients diagnosed with anosmia compared with matched control participants.

RESULTS A total of 2690 patients were identified with a mean (SD) age of 55.3 (16.6) years. The case group included 538 patients with anosmia (20%), and the control group included 2152 matched control participants (80%). Most of the individuals in the case and control groups were women, White patients, had overweight (BMI 25 to <30), and did not smoke (women: 339 [63.0%] and 1355 [63.0%]; White patients: 318 [59.1%] and 1343 [62.4%]; had overweight: 179 [33.3%] and 653 [30.3%]; and did not smoke: 328 [61.0%] and 1248 [58.0%]). Mean (SD) exposure to PM 2.5 was significantly higher in patients with anosmia compared with healthy control participants at 12-, 24-, 36-, 60-month time points: 10.2 (1.6) μg/m³ vs 9.9 (1.9) μg/m³; 10.5 (1.7) μg/m³ vs 10.2 (1.9) μg/m³; 10.8 (1.8) μg/m³ vs 10.4 (2.0) μg/m³; and 11.0 (1.8) μg/m³ vs 10.7 (2.1) μg/m³, respectively. There was an association between elevated PM 2.5 exposure level and odds of anosmia in multivariate analyses that adjusted for age, sex, race/ethnicity, body mass index, alcohol or tobacco use, and medical comorbidities (12 mo: odds ratio [OR], 1.73; 95% CI, 1.28-2.33; 24 mo: OR, 1.72; 95% CI, 1.30-2.29; 36 mo: OR, 1.69; 95% CI, 1.30-2.19; and 60 mo: OR, 1.59; 95% CI, 1.22-2.08). The association between long-term exposure to PM 2.5 and the odds of developing anosmia was nonlinear, as indicated by spline analysis. For example, for 12 months of exposure to PM 2.5, the odds of developing anosmia at 6.0 μg/m³ was OR 0.79 (95% CI, 0.64-0.97); at 10.0 μg/m³, OR 1.42 (95% CI, 1.10-1.82); at 15.0 μg/m³, OR 2.03 (95% CI, 1.15-3.58).

CONCLUSIONS AND RELEVANCE In this study, long-term airborne exposure to PM 2.5 was associated with anosmia. Ambient PM 2.5 represents a potentially ubiquitous and modifiable risk factor for the loss of sense of smell.
Introduction

Anosmia, the loss of the sense of smell, has a substantial effect on overall well-being, quality of life, the experience of food, and the ability to detect environmental hazards, such as fire and toxins. Patients with disruptions in their ability to smell commonly experience weight loss, decreased social interaction, depression, and generalized anxiety. Moreover, olfactory function is one of the strongest predictors of mortality in older adults.  

Despite these concerns, anosmia is an overlooked public health problem. Although estimates vary, considerable portions of the general population have anosmia. In Sweden, more than 5.8% of adults in the general population have anosmia, while 13.7% of adults have anosmia in South Korea. In the US, the overall reported prevalence of anosmia ranges from 10% to 23% of the entire population, accounting for tens of millions of Americans. These dramatic statistics may underestimate the prevalence of anosmia, because patients may unknowingly experience subtle changes in olfactory function, and disruptions in olfaction may occur in more than 50% of healthy adults when detailed olfactory assessments are performed.

The causes of anosmia can be broadly subdivided into conductive (ie, physical barriers to odorants reaching the olfactory system, including allergic rhinitis or hay fever, nasal polyps, or rhinosinusitis) and sensorineural (ie, failure of the olfactory system to detect odorants, including viral infection, neurologic conditions, or head trauma). Beyond inflammatory sinonasal and neurocognitive diseases, air pollution may present an additional olfactory insult that contributes to the development of anosmia. Several studies have demonstrated an association between air pollution and olfaction. The unique positioning of the olfactory nerve in the nasal cavity, directly opposed to the external environmental exposures, places the olfactory system at particular risk from airborne pollutants.

One pollutant potentially associated with anosmia is ambient particulate matter (PM) with an aerodynamic diameter of no more than 2.5 μm (PM$_{2.5}$). This class of pollutant is associated with cardiovascular diseases, cognitive decline, and overall mortality. PM$_{2.5}$ contains a complex mixture of solids or liquid droplets containing organic compounds, metals, and dust particles that can be inhaled and directly contact the olfactory neurons that are located in the roof of the nasal cavity. Although exposure to PM$_{2.5}$ has been associated with olfactory dysfunction, few large-scale studies have specifically examined the association of PM air pollution with anosmia across all age groups and locations. Because nonvirally mediated anosmia clinically develops over longer periods, this study focused on investigating the association between long-term PM$_{2.5}$ air pollution and the risk of anosmia in a large outpatient-based case-control study of patients who visited the Johns Hopkins Hospital in Baltimore, Maryland.

Methods

Setting and Participants

This case-control study was approved by the Johns Hopkins University School of Medicine institutional review board with a waiver of informed consent. Consent was waived because, with the exception of zip codes, no patient-identifying information was collected. This study adhered to the Strengthening the Reporting of Observational Studies in Epidemiology (STROBE) reporting guideline with a completed checklist for case-control studies in epidemiology. Patients who were 18 years or older and were diagnosed with anosmia for the first time by an otolaryngologist within the Johns Hopkins Health System from January 1, 2013, to December 31, 2016, were included in this study. Selected patients did not have a diagnosis of chronic rhinosinusitis or nasal polyposis. The diagnosis was confirmed using relevant International Classification of Diseases, Ninth Revision and Tenth Revision (ICD-9 and ICD-10) diagnosis codes. The time of onset of anosmia was defined as the time at diagnosis. All patients underwent facial computed tomography (CT) scan or magnetic resonance imaging (MRI) and nasal endoscopy and were without evidence of sinonasal pathology.
Clinical characteristics, including demographic data, race/ethnicity, preexisting medical conditions, and socioeconomic status (SES), were extracted from the medical records of all case and control participants. We used the US Census Bureau’s American Community Survey to determine the median household income by patient’s residence zip code tabulation area (ZCTA). The median ZCTA household income was inflation-adjusted to 2016 US dollars. Control participants were selected from patients who visited the otolaryngology department based on the following criteria: alive during this time; without anosmia, chronic rhinosinusitis, or nasal polyposis; and with diseases that were not a priori related to anosmia or known risk factors for the diseases, such as smoking, alcohol consumption, diabetes, prior traumatic brain injury, or neurodegenerative disease. Four control participants per case were selected using the nearest neighbor matching strategy for age, sex, race/ethnicity, and date of anosmia diagnosis for each identified case.

Exposure Assessment

Ambient PM$_{2.5}$ exposure levels were estimated and validated based on previously published prediction models. Briefly, we used deep-learning neural networks that incorporated meteorological measurements, land-use terms, satellite-based measurements, and simulation outputs from a chemical transport model to estimate daily concentrations of PM$_{2.5}$ in unmonitored areas. We acquired air pollution monitoring data from the US Environmental Protection Agency (EPA) Air Quality System (AQS) (1928 monitors for PM$_{2.5}$). Data about daily air temperature and relative humidity were retrieved from North American Regional Reanalysis with grids that were approximately 32 km × 32 km. Satellite-based aerosol optical depths were retrieved from the Moderate Resolution Imaging Spectroradiometer (MODIS), using the Multi-Angle Implementation of Atmospheric Correction algorithm method. For vegetation coverage, we used the percentage of vegetation from the National Centers for Environmental Prediction North American Regional Reanalysis data and MODIS MOD13A2, a normalized difference vegetation index data product. We fit the neural network with monitoring data from the EPA AQS. We then estimated daily PM$_{2.5}$ concentrations from the year 2000 to 2016 for nationwide grids that were 3 km × 3 km. Cross-validation indicated that the models had a high accuracy across the entire study area. The national mean coefficients of determination ($R^2$) for PM$_{2.5}$ were 0.86, with a variation between 0.71 to 0.95; the mean square errors between the measurements and estimated daily values for PM$_{2.5}$ were 1.50 μg/m$^3$. We created various exposure metrics as appropriate to examine different windows of exposure, including 12-, 24-, 36- and 60-month mean PM$_{2.5}$ concentration before the diagnosis date. For each patient, we assigned a PM$_{2.5}$ exposure value from the nearest estimated 3 km × 3 km grid according to the zip code of the person’s residence address.

Statistical Analysis

Descriptive statistics for patient variables were calculated using mean (SD), or frequency count (percentage), as appropriate. Conditional logistic regression models were used to determine the association between long-term PM$_{2.5}$ exposure and risk of anosmia. We used a base model adjusted for age, sex, race/ethnicity, and state. In model 2, we further adjusted for body mass index (BMI), which was calculated as weight in kilograms divided by height in meters squared, current alcohol consumption status, and current smoking status, which may be associated with olfaction. In model 3, we added medical comorbidities (ie, medical history of hypertension, diabetes, chronic obstructive pulmonary disease [COPD], and asthma) as potential confounders of this association.

To evaluate nonlinear dose-response associations between PM$_{2.5}$ exposure and risk of anosmia, we modeled PM$_{2.5}$ air pollution exposure variables using restricted cubic splines with knots at the 10th, 50th, and 90th percentiles of the distribution of PM$_{2.5}$ exposure estimates. Statistical analyses were conducted using Stata version 16.0 (StataCorp) and R version 4.1 (R Project for Statistical Computing) from September 2020 to March 2021. P values were 2-sided, and $P < .05$ was considered statistically significant.
Results

A total of 2690 patients were identified with a mean (SD) age of 55.3 (16.6) years, and 1694 (63.0%) were women. Among the 538 case participants (20.0%) with anosmia, 339 (63.0%) were women, and the mean (SD) age at baseline was 54.8 (17.0) years. Among 2152 matched control participants (80.0%), 1355 (63.0%) were female patients with the mean (SD) age of 55.4 (16.5) years. Most of the individuals in the case and control groups were White patients, had overweight (BMI 25 to <30), and did not smoke (White patients: 318 [59.1%] and 1343 [62.4%]; had overweight: 179 [33.3%] and 653 [30.3%]; and did not smoke; 328 [61.0%] and 1248 [58.0%]) (Table 1). Patients with anosmia were more likely to consume alcohol at the time of enrollment, were more likely to live in an area with lower household income, and less likely to be diagnosed with hypertension or COPD compared with control participants (consume alcohol: 270 [50.2%] vs 814 [37.8%]; mean [SD] median household income: $75,927 [$32,319] vs $86,164 [$34,533]; hypertension: 162 [30.1%] vs 762 [35.4%]; COPD: 10 [1.9%] vs 80 [3.7%]). There was no difference in prevalence of diagnoses of diabetes, asthma, or environmental allergies between the 2 groups. Most patients lived in the northeastern United States (2555 of 2690 [95.0%]).

Mean (SD) PM$_{2.5}$ exposure levels were higher in patients with anosmia leading to the time of diagnosis compared with control participants at all measured estimates with a 12-, 24-, 36-, and 60-month mean concentration of 10.2 (1.6) μg/m$^3$ vs 9.9 (1.9) μg/m$^3$, 10.5 (1.7) μg/m$^3$ vs 10.2 (1.9) μg/m$^3$, and 10.8 (1.8) μg/m$^3$ vs 10.4 (2.0) μg/m$^3$ vs 10.2 (1.9) μg/m$^3$.

### Table 1. Demographic and Clinical Characteristics of Participants

| Characteristic            | No. (%)                      | P value$^a$ |
|---------------------------|------------------------------|-------------|
|                           | Patients with anosmia (n = 538) | Control participants (n = 2152) |
| Age, y                    | 54.8 (17.0)                  | 55.4 (16.5) | .43 |
| Male sex                  | 199 (37.0)                   | 797 (37.0)  | >.99|
| Female sex                | 339 (63.0)                   | 1355 (63.0) | .99 |
| Race/ethnicity            |                              |             | .19 |
| White                     | 318 (59.1)                   | 1343 (62.4) | .19 |
| African American          | 143 (26.6)                   | 556 (25.8)  | .19 |
| Hispanic/Latino           | 30 (5.6)                     | 80 (3.7)    | .19 |
| Other$^b$                 | 47 (8.7)                     | 173 (8.0)   | .19 |
| PM$_{2.5}$ exposure, mean (SD), μg/m$^3$ |            |             |     |
| 12-mo                     | 10.2 (1.6)                   | 9.9 (1.9)   | .003|
| 24-mo                     | 10.5 (1.7)                   | 10.2 (1.9)  | .001|
| 36-mo                     | 10.8 (1.8)                   | 10.4 (2.0)  | <.001|
| 60-mo                     | 11.0 (1.8)                   | 10.7 (2.1)  | .002|
| BMI                        |                              |             | .58 |
| Underweight, <18.5        | 17 (3.2)                     | 81 (3.8)    | .58 |
| Normal weight, 18.5 to <25 | 176 (32.7)                   | 734 (34.1)  | .58 |
| Overweight, 25 to <30     | 179 (33.3)                   | 653 (30.3)  | .58 |
| Obesity, ≥30              | 166 (30.9)                   | 684 (31.8)  | .58 |
| Current smoking status    |                              |             |     |
| Never smoked              | 328 (61.0)                   | 1248 (58.0) | .11 |
| Currently smokes          | 44 (8.2)                     | 242 (11.2)  | .11 |
| Formerly smoked           | 166 (30.9)                   | 662 (30.8)  | .11 |
| Current alcohol consumption| 270 (50.2)                   | 814 (37.8)  | <.001|
| Median household income, mean (SD), US $ | 75 927 (32 319) | 86 164 (34 533) | <.001|
| Comorbidity               |                              |             |     |
| Hypertension              | 162 (30.1)                   | 762 (35.4)  | .02 |
| Diabetes                  | 56 (10.4)                    | 260 (12.1)  | .32 |
| COPD                      | 10 (1.9)                     | 80 (3.7)    | .04 |
| Asthma                    | 50 (9.3)                     | 189 (8.8)   | .77 |

Abbreviations: BMI, body mass index (calculated as weight in kilograms divided by height in meters squared); COPD, chronic obstructive pulmonary disease; PM$_{2.5}$, particulate matter with an aerodynamic diameter of no more than 2.5 μm.

$^a$ Values were calculated using $\chi^2$ test for categorical variables and the Mann-Whitney U test for continuous variables.

$^b$ Includes Asian, American Indian, Alaska Native, Native Hawaiian, or other Pacific Islander.
µg/m^3, 10.8 (1.8) µg/m^3 vs 10.4 (2.0) µg/m^3, and 11.0 (1.8) µg/m^3 vs 10.7 (2.1) µg/m^3, respectively (Table 1). Multivariate modeling demonstrated a direct association between PM_{2.5} exposure levels and patients with anosmia across all models. Model 1 adjusted for age, sex, and race/ethnicity; model 2 also adjusted for BMI, current alcohol consumption status, and current smoking status; and model 3 further adjusted for a medical history of hypertension, diabetes, COPD, and asthma. In fully adjusted models (model 3), the odds ratios (ORs) for the development of anosmia associated with a 5-µg/m^3 increase in 12-, 24-, 36- and 60-month PM_{2.5} exposure were 1.73 (95% CI, 1.28-2.33), 1.72 (95% CI, 1.30-2.29), 1.69 (95% CI, 1.30-2.19), and 1.59 (95% CI, 1.22-2.08), respectively (Table 2). Increasing PM_{2.5} concentration was associated with an increased odds of anosmia in spline regression analyses, and the trend was consistent across different exposure periods (Figure). For example, for 12 months of exposure to PM_{2.5}, the odds of developing anosmia at 6.0 µg/m^3 was OR 0.79 (95% CI, 0.64-0.97); at 10.0 µg/m^3, OR 1.42 (95% CI, 1.10-1.82); at 15.0 µg/m^3, OR 2.03 (95% CI, 1.15-3.58).

**Discussion**

To our knowledge, this study found the strongest association to date between long-term exposure to air pollution and anosmia. We observed a dose-dependent association between increasing concentrations of PM_{2.5} exposure and anosmia that persisted over 5 years of PM_{2.5} exposure, even after adjusting for confounding factors. The current findings suggest that even small increases in ambient PM_{2.5} exposure may be associated with anosmia, which has broad public health ramifications in the setting of increasing global urbanization. This study benefits from many strengths, including a robust patient data set and the use of a novel control matching strategy. We have also used unique deep learning neural network modeling to accurately estimate PM_{2.5} exposure to demonstrate realistic clinical implications of air pollution on olfactory function.

An often-overlooked human sensory function, olfaction is vital to the perception and experience of human life. Olfactory impairments are intrinsically associated with the experience of food, eating-related quality of life, and malnutrition.\(^1\)\(^2\)\(^8\) In fact, most subjective gustatory deficits are a manifestation of olfactory loss.\(^3\)^\(^9\) Moreover, anosmia has been negatively associated with broad measures of quality of life, depression, anxiety, and cognitive impairment.\(^4\)^\(^3\)^\(^0\)-\(^3\)^\(^5\) Furthermore, in addition to inherent risks associated with the failure to detect toxins and environmental hazards, large population-based studies have demonstrated an association between olfactory disturbances and anosmia with measures of patient frailty and mortality.\(^5\)^\(^3\)^\(^6\) Because olfactory function declines with age and air pollution exposure is cumulative, our data are consistent with environmental determinants of chemosensory aging.\(^1\)^\(^2\)^\(^3\)\(^7\) Thus, air pollution may represent another ubiquitous risk factor for age-related sensory loss.

| Time exposed to PM_{2.5}, mo | OR (95% CI)\(^a\) | Model 1\(^b\) | Model 2\(^c\) | Model 3\(^d\) |
|-----------------------------|------------------|----------------|----------------|----------------|
| 12                          | 1.53 (1.16-2.02) | 1.68 (1.25-2.26) | 1.73 (1.28-2.33) |
| 24                          | 1.58 (1.21-2.07) | 1.69 (1.27-2.24) | 1.72 (1.30-2.29) |
| 36                          | 1.58 (1.23-2.03) | 1.66 (1.27-2.16) | 1.69 (1.30-2.19) |
| 60                          | 1.48 (1.15-1.90) | 1.56 (1.20-2.03) | 1.59 (1.22-2.08) |

Abbreviations: OR, odds ratio; PM_{2.5}, particulate matter with an aerodynamic diameter of no more than 2.5 µm.

\(^a\) ORs are based on 5-µg/m^3 increase in PM_{2.5} exposure.

\(^b\) Adjusted for age, sex, race/ethnicity, and state.

\(^c\) Additionally adjusted for body mass index, current alcohol consumption status, current smoking status, and median household income.

\(^d\) Additionally adjusted for medical history of hypertension, diabetes, chronic obstructive pulmonary disease, and asthma.
Determinants of olfactory function are multifactorial and can be broadly categorized into biology, individual experience, and environment. Although epigenetic and cultural differences in olfactory function have been described, the effect of environmental factors may be substantial.

Several studies have attempted to capture the direct effect of pollution and industrialization on olfaction. Two studies from Mexico compared olfactory ability of residents from geographically similar locations that differed drastically in their level of air pollution. In each of these studies, residents from the less polluted environment outperformed residents from the more polluted city. Similar observations have been observed in other countries; for example, individuals from Dresden, Germany, were found to perform significantly worse than those from the Bolivian rainforest or the Cook Islands in the South Pacific, which are 2 significantly less polluted areas. Although confounding differences exist, these studies support a role for environmental determinants in affecting olfaction.

The pathophysiologic mechanism of olfactory loss associated with PM$_{2.5}$ remains unclear. Evidence in the literature suggests that PM$_{2.5}$ may create sinonasal inflammation, which may compromise the odorsant's ability to reach the olfactory cleft. Alternatively, pollution levels may result in mucosal inflammation, which affects the olfactory cleft. Indeed, nasal biopsies from residents of Mexico City demonstrated dysplastic epithelial changes compared with patients from the less polluted Isla Mujeres in Mexico, implying that cellular changes may occur without overt

Figure. Odds Ratios (ORs) for Risks of Anosmia by the Level of Exposure to Particulate Matter With an Aerodynamic Diameter of No More Than 2.5 μm (PM$_{2.5}$) Concentration in Each Exposure Period

The dose-response curve was calculated using restricted cubic splines with knots at the 10th, 50th, and 90th percentiles of the distribution of 12-month PM$_{2.5}$ concentrations. The reference exposure level was set at the 10th percentile of the distribution of 12-month PM$_{2.5}$ concentrations (7.56 μg/m$^3$). ORs were adjusted for age, sex, race/ethnicity, state, body mass index, current alcohol consumption status, current smoking status, median household income of zip code of individual's residence, and medical history of hypertension, diabetes, chronic obstructive pulmonary disease, and asthma.
clinical manifestations. An alternative mechanism is that PM_{2.5} may cause direct insult to the olfactory neuroepithelium and olfactory bulb. There is also the possibility of direct nervous system insults, with increased levels of β-amyloid, cyclooxygenase-2, PM, and metals found in autopsies of patients with anosmia in both the olfactory bulb and frontal lobe, compared with control participants who experienced lower pollution levels. Additionally, the inhalation of ultrafine particles (PM <1 μm) may directly translocate along the olfactory nerve directly to the central nervous system. Overall, these results remain to be replicated and developed further in larger and more diverse human cohorts, different environments, and in animal models that can be manipulated.

The adverse effects of air pollution are pervasive and represent more serious implications for certain at-risk populations. The association between air pollution and more severe obstructive lower respiratory disease outcomes have been well-described, whereas more recent investigations have demonstrated the untoward effects of air pollution on the upper respiratory system. Although underlying respiratory disease may increase the relative risk of pollutant exposure, the associated health risks of air pollution are especially notable for lower-income, underserved, and minority communities, as they are often exposed to higher concentrations of potentially hazardous pollutants.

Although substantially less is known regarding the association of pollutants with olfactory dysfunction compared with other diseases, there is increasing awareness regarding the importance of olfaction. Recently, COVID-19 has thrust olfaction into the spotlight as olfactory disturbances appear to be both a cardinal symptom and, in some cases, a debilitating consequence of the ongoing global pandemic. The inability to detect hazards, such as gas leaks or fires, represents the immediate implications of disruptions in olfactory function. In contrast, increased levels of depression, dietary changes, and impaired cognition may be associated with effects on patient frailty and mortality. Nonetheless, prior studies have demonstrated a persistent association of olfactory dysfunction and mortality even after correcting for dementia. In the context of increasing global urbanization and an aging population, the pervasive association of air pollution with olfaction are likely to increase.

In this study, we developed a novel satellite-based model to estimate long-term exposure to PM_{2.5} with high spatial and temporal resolution. This model enabled an estimation of individual-level exposure and overcame the issue of spatial coverage associated with the use of data collected solely from ground monitoring stations. We also used a convolutional layer in the neural network to estimate PM_{2.5} by aggregating variable values from nearby grid cells or monitoring sites. This approach is versatile and more accurate in modeling complex pollutant exposure.

The findings of this current investigation present many avenues for future research, including individual and population studies to better understand mechanisms of PM_{2.5}-associated olfactory dysfunction. Also, air pollution is a mixture of pollutants, including PM_{10}, nitrogen dioxide, black carbon, and ozone, which uniquely contribute to patients' environment. However, air quality is often measured by individual components that may not reflect the actual effects of the mixture as a whole. It is also possible that individual components of the particulate matter, such as unique metals, may be associated with the prevalence of anosmia in this study population. Thus, further epidemiologic studies are required to examine the association of other components of air pollution, geographic regions, socioeconomic disparities, and personal activity on olfaction.

Limitations
This study has limitations. Because of the study's cross-sectional design, only prevalent anosmia case participants could be analyzed. Therefore, effect estimates are more likely to be associated with reverse causation and residual confounding. Additionally, although the air pollution exposure models had an excellent cross-validation performance, the model-estimated exposures are surrogates for personal exposure, which depend on daily activity patterns as well as workplace and commuting exposures. The models also failed to account for indoor air pollution and change in residential address during the study. Although personal monitoring would help to alleviate these potential...
Conclusions

In this cross-sectional study, long-term exposure to increasing concentrations of PM$_{2.5}$ exposure was associated with anosmia. This finding has broad implications for the association of a prevalent ambient air pollutant with a vital human sensory function.

ARTICLE INFORMATION

Accepted for Publication: April 2, 2021.
Published: May 27, 2021. doi:10.1001/jamanetworkopen.2021.11606

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Author Contributions: Dr Ramanathan had full access to all of the data in the study and takes responsibility for the integrity of the data and the accuracy of the data analysis. Drs Zhang and Rowan provided equal contributions for first authorship.

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Obtained funding: Ramanathan.

Administrative, technical, or material support: Rowan, Pinto, Ramanathan.

Supervision: Rowan, Ramanathan.

Conflict of Interest Disclosures: None reported.

Funding/Support: This work was supported by grant R01AI143731 from the National Institute of Allergy and Infectious Diseases and the National Institutes of Health.

Role of the Funder/Sponsor: The funder had no role in the design and conduct of the study; collection, management, analysis, and interpretation of the data; preparation, review, or approval of the manuscript; and decision to submit the manuscript for publication.

REFERENCES

1. Rowan NR, Soler ZM, Storck KA, et al. Impaired eating-related quality of life in chronic rhinosinusitis. Int Forum Allergy Rhinol. 2019;9(3):240-247. doi:10.1002/air.22242

2. Keller A, Malaspina D. Hidden consequences of olfactory dysfunction: a patient report series. BMC Ear Nose Throat Disord. 2013;13(1):8. doi:10.1186/1472-6815-13-8
3. Krusemark EA, Novak LR, Gitelman DR, Li W. When the sense of smell meets emotion: anxiety-state-dependent olfactory processing and neural circuitry adaptation. J Neurosci. 2013;33(39):15324-15332. doi:10.1523/JNEUROSCI.1835-13.2013

4. Kohli P, Soler ZM, Nguyen SA, Muus JS, Schlosser RJ. The association between olfaction and depression: a systematic review. Chem Senses. 2016;41(6):479-486. doi:10.1093/chemse/bjw061

5. Pinto JM, Wroblewski KE, Kern DW, Schumm LP, McClintock MK. Olfactory dysfunction predicts 5-year mortality in older adults. PLoS One. 2014;9(10):e107541. doi:10.1371/journal.pone.0107541

6. Hoffman HJ, Cruickshanks KJ, Davis B. Perspectives on population-based epidemiological studies of olfactory and taste impairment. Ann N Y Acad Sci. 2009;1170:514-530. doi:10.1111/j.1749-6632.2009.04597.x

7. Kang JW, Lee YC, Han K, Kim SW, Lee KH. Epidemiology of anosmia in South Korea: a nationwide population-based study. Sci Rep. 2020;10(1):3717. doi:10.1038/s41598-020-60678-z

8. Brämerson A, Johansson L, Ek L, Nordin S, Bende M. Prevalence of olfactory dysfunction: the Skövde population-based study. Laryngoscope. 2004;114(4):733-737. doi:10.1097/00005537-200404000-00026

9. Dong J, Pinto JM, Guo X, et al. The prevalence of anosmia and associated factors among U.S. black and white older adults. J Gerontol A Biol Sci Med Sci. 2017;72(8):1080-1086. doi:10.1093/gerona/glx081

10. Bhatnacharya N, Kepnes LJ. Contemporary assessment of the prevalence of smell and taste problems in adults. Laryngoscope. 2015;125(5):1102-1106. doi:10.1002/lary.24999

11. Rawal S, Hoffman HJ, Bainbridge KE, Huido-Medina TB, Duffy VB. Prevalence and risk factors of self-reported smell and taste alterations: results from the 2011-2012 US National Health and Nutrition Examination Survey (NHANES). Chem Senses. 2016;41(1):69-76. doi:10.1093/chemse/bjv057

12. Schlosser RJ, Desiato VM, Storck KA, et al. A community-based study on the prevalence of olfactory dysfunction. Am J Rhinol Allergy. 2020;34(5):661-670. doi:10.1177/1945892420922771

13. Yang J, Pinto JM. The epidemiology of olfactory disorders. Curr Otorhinolaryngol Rep. 2016;4(2):130-141. doi:10.1007/s40136-016-0120-6

14. Hummel T, Whitcroft KL, Andrews P, et al. Position paper on olfactory dysfunction. Rhinology. 2016;56(1):1-30.

15. Arnold C. Sensory overload? air pollution and impaired olfaction. Environ Health Perspect. 2019;127(6):62001. doi:10.1289/EHP3621

16. Hudson R, Arrida A, Martínez-Gómez M, Distel H. Effect of air pollution on olfactory function in residents of Mexico City. Chem Senses. 2006;31(1):79-85. doi:10.1093/chemse/bjv019

17. Guarnéros M, Hummel T, Martínez-Gómez M, Hudson R. Mexico City air pollution adversely affects olfactory function and intranasal trigeminal sensitivity. Chem Senses. 2009;34(9):819-826. doi:10.1093/chemse/bjp071

18. Al-Kindi SG, Brook RD, Biswal S, Rajagopalan S. Environmental determinants of cardiovascular disease: lessons learned from air pollution. Nat Rev Cardiol. 2020;17(10):656-672. doi:10.1038/s41569-020-0371-2

19. Ajmani GS, Suh HH, Pinto JM. Effects of ambient air pollution exposure on olfaction: a review. Environ Health Perspect. 2016;124(11):1683-1693. doi:10.1289/EHP136

20. Calderón-Garcidueñas L, Franco-Lira M, Henríquez-Roldán C, et al. Urban air pollution: influences on olfactory function and pathology in exposed children and young adults. Exp Toxicol Pathol. 2010;62(1):91-102. doi:10.1016/j.ett.2009.02.117

21. Ajmani GS, Suh HH, Wroblewski KE, et al. Fine particulate matter exposure and olfactory dysfunction among urban-dwelling older US adults. Environ Res. 2016;151:797-803. doi:10.1016/j.envres.2016.09.012

22. United States Census. American Community Survey. Accessed April 19, 2021. https://www.census.gov/acs/www/data/tables-and-tools/data-profiles/2016/

23. Hennessy S, Bilker WB, Berlin JA, Strom BL. Factors influencing the optimal control-to-case ratio in matched case-control studies. Am J Epidemiol. 1999;149(2):195-197. doi:10.1093/oxfordjournals.aje.a009786

24. Zhang Z, Wang J, Hart JE, et al. National scale spatiotemporal land-use regression model for PM2.5, PM10, and NO2 concentration in China. Atmos Environ. 2018;192:48-54. doi:10.1016/j.atmosenv.2018.08.046

25. Kalnay E, Kanamitsu M, Kistler R, et al. The NCEP/NCAR40-year reanalysis project. Bull Am Meteor Soc. 1996;77(3):437-472. doi:10.1175/1520-0477(1996)077<0437:TNYRP>2.0.CO;2

26. Superczynski SD, Kondragunta S, Lyapustin AI. Evaluation of the Multi-Angle Implementation of Atmospheric Correction (MAIAC) aerosol algorithm through intercomparison with VIIRS aerosol products and AERONET. J Geophys Res Atmos. 2017;122(5):3005-3022. doi:10.1002/2016JD025720

27. US Geological Survey. MOD13A2 v006. Accessed April 8, 2021. https://lpdaac.usgs.gov/products/mod13a2v006/
28. Van Regemorter V, Hummel T, Rosenzweig F, Mouraux A, Rombaux P, Huard C. Mechanisms linking olfactory impairment and risk of mortality. *Front Neurosci*. 2020;14:140. doi:10.3389/fnins.2020.00140

29. Hunt JD, Reiter ER, Costanzo RM. Etiology of subjective taste loss. *Int Forum Allergy Rhinol*. 2019;9(4):409-412. doi:10.1002/alr.22263

30. Croy I, Nordin S, Hummel T. Olfactory disorders and quality of life—an updated review. *Chem Senses*. 2014;39(3):185-194. doi:10.1093/chemse/bjt072

31. Chunlin I, Qazi J, Fermin CR, Wilson JH, Payne SC, Mattos JL. Association between olfactory and gustatory dysfunction and cognition in older adults. *Am J Rhinol Allergy*. 2019;33(2):170-177. doi:10.1177/1945892418824451

32. Bigman G. Age-related smell and taste impairments and vitamin D associations in the U.S. adults National Health and Nutrition Examination Survey. *Nutrients*. 2020;12(4):E984. doi:10.3390/nu12040984

33. Gopinath B, Russell J, Sue CM, Flood VM, Burlutsky G, Mitchell P. Olfactory impairment in older adults is associated with poorer diet quality over 5 years. *Eur J Nutr*. 2016;55(3):1081-1087. doi:10.1007/s00394-015-0921-2

34. Zang Y, Han P, Burghardt S, Knaapila A, Schriever V, Hummel T. Influence of olfactory dysfunction on the perception of food. *Eur Arch Otorhinolaryngol*. 2019;276(10):2811-2817. doi:10.1007/s00405-019-05558-7

35. Mady LJ, Schwarzbach HL, Moore JA, et al. The association of air pollutants and allergic and nonallergic rhinitis in chronic rhinosinusitis. *Int Forum Allergy Rhinol*. 2018;8(8):369-376. doi:10.1002/air.22060

36. Calderón-Garcidueñas L, Rodríguez-Alcaraz A, Villarreal-Calderón A, Lyght O, Janszen D, Morgan KT. Nasal epithelium as a sentinel for airborne environmental pollution. *Toxicol Sci*. 1998;46(2):352-364. doi:10.1093/toxsci/46.2.352

37. Oberdörster G, Sharp Z, Atudorei V, et al. Translocation of inhaled ultrafine particles to the brain. *Inhal Toxicol*. 2004;16(6-7):437-445. doi:10.1080/08958370409439597
50. Guarnieri M, Balmes JR. Outdoor air pollution and asthma. Lancet. 2014;383(9928):1581-1592. doi:10.1016/S0140-6736(14)60617-6

51. Khreis H, Kelly C, Tate J, Parslow R, Lucas K, Nieuwenhuijsen M. Exposure to traffic-related air pollution and risk of development of childhood asthma: a systematic review and meta-analysis. Environ Int. 2017;100:1-31. doi:10.1016/j.envint.2016.11.012

52. Doiron D, de Hoogh K, Probst-Hensch N, et al. Air pollution, lung function and COPD: results from the population-based UK Biobank study. Eur Respir J. 2019;54(1):1802140. doi:10.1183/13993003.02140-2018

53. Rice MB, Ljungman PL, Wilker EH, et al. Long-term exposure to traffic emissions and fine particulate matter and lung function decline in the Framingham heart study. Am J Respir Crit Care Med. 2015;191(6):656-664. doi:10.1164/rccm.201410-1875OC

54. Guo C, Zhang Z, Lau AKH, et al. Effect of long-term exposure to fine particulate matter on lung function decline and risk of chronic obstructive pulmonary disease in Taiwan: a longitudinal, cohort study. Lancet Planet Health. 2018;2(3):e114-e125. doi:10.1016/S2542-5196(18)30028-7

55. Hajat A, Hsia C, O’Neill MS. Socioeconomic disparities and air pollution exposure: a global review. Curr Environ Health Rep. 2015;2(4):440-450. doi:10.1007/s40572-015-0069-5

56. Fairburn J, Schüle SA, Dreger S, Hilz LK, Bolte G. Social inequalities in exposure to ambient air pollution: a systematic review in the WHO European region. Int J Environ Res Public Health. 2019;16(17):E3127. doi:10.3390/ijerph16173127

57. Doiron D, de Hoogh K, Probst-Hensch N, et al. Residential air pollution and associations with wheeze and shortness of breath in adults: a combined analysis of cross-sectional data from two large European cohorts. Environ Health Perspect. 2017;125(9):097025. doi:10.1289/EHP1353

58. Devanand DP, Lee S, Manly J, et al. Olfactory identification deficits and increased mortality in the community. Ann Neurol. 2015;78(3):401-411. doi:10.1002/ana.24447

59. Nordin S, Brämerson A, Bende M. Prevalence of self-reported poor odor detection sensitivity: the Skövde population-based study. Acta Otolaryngol. 2004;124(10):1171-1173. doi:10.1080/00016480410017468