

# Changes in Fine Particulate Matter (PM<sub>2.5</sub>) Levels in the United States: A Case Study of 1999 vs 2012

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## Abstract

This study compares PM<sub>2.5</sub> concentrations in the United States between 1999 and 2012 using national monitoring data. For clarity in descriptive trend estimation, analyses are restricted to valid measurements (observations with non-missing, non-negative PM<sub>2.5</sub> values). We describe national and state-level distributions, quantify changes in central tendency and dispersion, and identify states with the largest improvements or deteriorations. Results show a clear downward shift in median PM<sub>2.5</sub> between 1999 and 2012, with heterogeneous state-level patterns. We discuss implications for exposure assessment and highlight sensitivity checks to evaluate the impact of data exclusions.

**Keywords:** PM<sub>2.5</sub>, air pollution, data quality, United States, environmental epidemiology

## 1. Introduction

Airborne fine particulate matter with aerodynamic diameter  $\leq 2.5 \mu\text{m}$  (PM<sub>2.5</sub>) is a well-established risk factor for cardiovascular and respiratory disease [1]. Over the last two decades the United States has implemented a range of regulatory and technological interventions aimed at reducing emissions from industry, transportation, and power generation; monitoring networks have also expanded and evolved [2]. Quantifying how PM<sub>2.5</sub> distributions changed between 1999 and 2012 provides a useful overview of national progress and highlights spatial heterogeneity in exposure reductions.

This paper presents a descriptive comparison of PM<sub>2.5</sub> data from two benchmark years (1999 and 2012). To focus the analysis on reliable measurements, we restrict analyses to valid observations (non-missing, non-negative PM<sub>2.5</sub> values) and summarize changes at national and state levels. We present distributional comparisons (boxplots, densities), state-level trajectories, and maps, and we identify states with the largest percent changes. Where relevant, we flag limitations introduced by the data-cleaning choices and propose sensitivity checks [3].

### 1.1. Primary Objective

- Describe how PM<sub>2.5</sub> levels changed between 1999 and 2012 at national and state levels using valid monitoring observations.

### 1.2. Research Questions

1. Did median PM<sub>2.5</sub> decline nationally from 1999 to 2012?
2. Which states experienced the largest decreases or increases?
3. Are changes geographically structured (regional patterns)?
4. Do larger reductions occur where monitoring density changed?

### 1.3. Secondary Analyses

- Seasonal patterns: Compare monthly or seasonal distributions, if sufficient data are available.
- Robustness checks: (a) Include previously excluded negative values as NA (b) Exclude states with very few monitors and repeat key summaries

## 2. Methods

### 2.1. Data cleaning and selection

Data were sourced from the U.S. Environmental Protection Agency (EPA) Air Quality System (AQS) for the years 1999 and 2012. Observations with missing or negative PM<sub>2.5</sub> values were excluded from the primary analysis; these exclusions are reported in Table 1 (number and percent removed).

All subsequent analyses were performed on the cleaned dataset. Given the expected non-normal distribution of air pollution data, we selected the median as a robust measure of central tendency. To quantify the uncertainty in the change of medians between 1999 and 2012, we employed a percentile bootstrap procedure with  $R = 10,000$  resamples. The analysis was performed with a fixed random seed (`set.seed(2025)`) to ensure reproducibility. We calculated the 95% confidence interval (CI) for the difference in medians ( $\text{median}_{2012} - \text{median}_{1999}$ ) using the 2.5th and 97.5th percentiles of the bootstrap distribution.

## 3. Results

### 3.1. Data processing

The initial dataset contained 1,421,708 observations. The data cleaning process, summarized in Table 1, resulted in the exclusion of 7.9% of the data, primarily due to missing PM<sub>2.5</sub> values. The final analytical dataset comprised 1,308,884 valid observations.

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metric	value	percentage
total_rows	1421708	100 %
removed_missing	86350	6.074 %
removed_negative	26474	1.862 %
kept_rows	1308884	92.064 %

Table 1: Data cleaning summary (counts & % of total)

### 3.2. Change in National Median $PM_{2.5}$

A comparison of national  $PM_{2.5}$  distributions reveals a clear downward shift between 1999 and 2012. Figure 1 visually illustrates this change, showing a lower median and a tighter interquartile range in 2012 compared to 1999. Both distributions exhibit a strong right skew, with numerous high-concentration outliers.

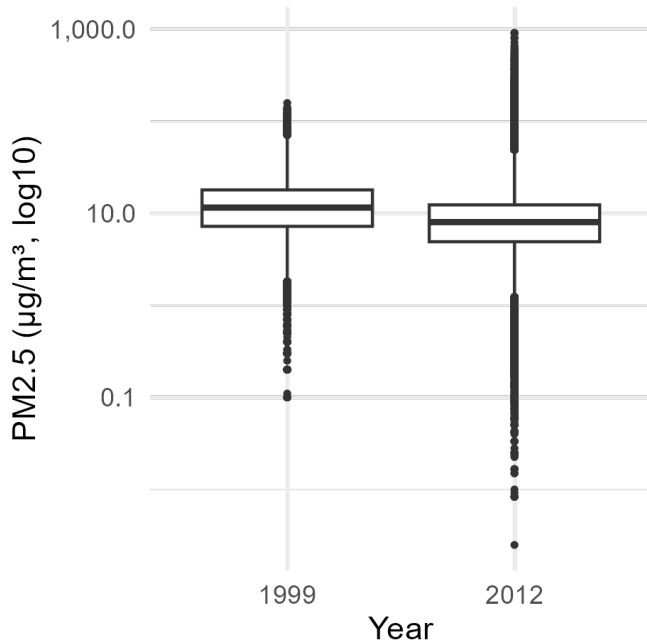


Figure 1: Distribution of  $PM_{2.5}$ : 1999 vs 2012

The precise change is quantified in Table 1. The national median  $PM_{2.5}$  concentration decreased from  $11.50 \mu\text{g m}^{-3}$  in 1999 to  $7.90 \mu\text{g m}^{-3}$  in 2012. Our primary inferential analysis focuses on the difference between these medians. The bootstrap estimate for this change (2012 – 1999) was  $-3.60 \mu\text{g m}^{-3}$ , with a 95% percentile bootstrap confidence interval of  $[-3.60, -3.50] \mu\text{g m}^{-3}$ . Because the confidence interval lies entirely below zero, this decline is statistically significant.

Statistic	1999	2012
Sample Size (n)	104,204	1,204,680
Median $PM_{2.5}$ ( $\mu\text{g}/\text{m}^3$ )	11.50	7.90
95% Bootstrap CI	[11.40, 11.54]	[7.90, 8.00]
<b>Median Difference (2012 - 1999): -3.60 (95% CI: [-3.60, -3.50])</b>		

Table 2: National  $PM_{2.5}$  statistics for 1999 and 2012.

### 3.3. State-Level Changes in $PM_{2.5}$

To investigate geographic heterogeneity in air quality improvements, we computed bootstrap confidence intervals for the median  $PM_{2.5}$  difference (2012 - 1999) for each state and territory with sufficient data. The results, visualized in the forest plot in Figure 2, demonstrate that the trend of decreasing  $PM_{2.5}$  is a near-universal national pattern.

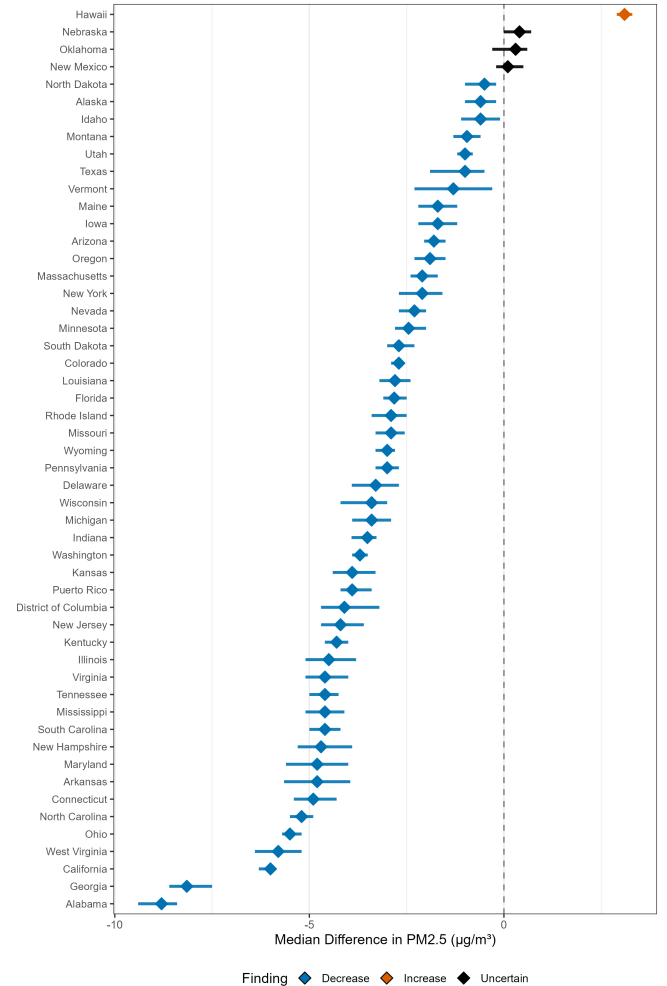


Figure 2: Change in median  $PM_{2.5}$  by state between 1999 and 2012. Diamonds represent the bootstrap point estimate for the median difference. The horizontal lines show the corresponding 95% bootstrap confidence interval. Blue indicates a statistically robust decrease (confidence interval is entirely below zero), orange indicates a robust increase, and black indicates a change that is not statistically robust (confidence interval overlaps zero). The plot includes all states and territories that met the minimum data criteria.

The plot highlights the magnitude and statistical robustness of these changes across the entire country. The vast majority of states show statistically robust declines, with their 95% confidence intervals falling entirely below zero. The most pronounced reductions occurred in southeastern and western states, including Alabama, Georgia, and California, which each saw median decreases exceeding  $5 \mu\text{g}/\text{m}^3$ . In contrast, a small number of states (notably Nebraska, Oklahoma, New Mexico, and Hawaii) had positive point estimates of change, indicating increases in median  $PM_{2.5}$ . However, only Hawaii's increase is

supported by the bootstrap interval; the other three states have confidence intervals that overlap zero and therefore are not distinguishable from no change.

## 4. Discussion

### 4.1. Principal findings

In this descriptive analysis of U.S. ambient PM<sub>2.5</sub> monitoring data, the national median concentration fell from 11.50  $\mu\text{g m}^{-3}$  in 1999 to 7.90  $\mu\text{g m}^{-3}$  in 2012. The bootstrap estimate of the median difference (2012 – 1999) was  $-3.60 \mu\text{g m}^{-3}$  with a 95% percentile bootstrap confidence interval of  $[-3.60, -3.50] \mu\text{g m}^{-3}$ , indicating a clear and statistically robust decline in the typical monitored PM<sub>2.5</sub> concentration over this period. State-level results show that the decline is broad-based: the vast majority of states have negative point estimates for the median difference and many have 95% bootstrap intervals that lie entirely below zero. A small number of states (for example Nebraska, Oklahoma, New Mexico) show positive point estimates whose intervals overlap zero; Hawaii is the one state in our set with an increase that is supported by its bootstrap interval.

### 4.2. Interpretation and context

The observed national decline in median PM<sub>2.5</sub> between 1999 and 2012 is consistent with long-term national trends reported by the EPA and with the timing of several regulatory and technological interventions that reduced emissions in the United States [2]. From a public-health perspective, reductions of the magnitude observed here are meaningful because ambient PM<sub>2.5</sub> is a well-established risk factor for cardiopulmonary disease and mortality [1]. Our focus on medians (rather than means) provides a robust summary of the typical monitored exposure and reduces sensitivity to extreme high values; this choice is supported by distributional diagnostics and by established robust methods in environmental statistics [3].

### 4.3. Heterogeneity across states and special cases

Although the national signal is clear, changes at the state level are heterogeneous. Several southeastern and western states show large, statistically robust decreases in median PM<sub>2.5</sub>, while a few states exhibit non-zero point estimates that are not statistically robust after accounting for sampling variability. Hawaii's relatively large percent increase in median PM<sub>2.5</sub> merits caution in interpretation: percent changes are unstable when baseline medians are small and monitoring coverage differed between years for some states.

### 4.4. Sources of uncertainty and limitations

Several limitations of the analysis must be acknowledged. First, the analytic dataset excludes missing and negative PM<sub>2.5</sub> values; these exclusions removed about 7.9% of the original records and could bias results if missingness or negative values are systematically related to pollution levels or monitor location. Second, monitor network changes (number of active monitors, siting changes, or different instruments) between 1999

and 2012 can influence observed changes in the monitored distribution; we examined monitor counts and interpreted state-level changes with this caveat in mind. Finally, our analysis is descriptive: observed changes are consistent with national emission reductions but do not, on their own, establish causal attribution to specific policies or technologies.

### 4.5. Robustness and next steps

The central finding (a large, robust decline in national median PM<sub>2.5</sub>) is robust to the percentile bootstrap procedure used to quantify uncertainty. Where percent changes or state-specific results appear unusual (for example very large percent increases driven by low baseline medians), we inspected raw counts and report those states as special cases. Remaining analytic steps that will strengthen inference—reported in the supplement—include (a) sensitivity to including previously excluded negative values (treated as missing), (b) alternative bootstrap schemes that resample monitors rather than observations.

## 5. Conclusions

Between 1999 and 2012 the national median PM<sub>2.5</sub> concentration in U.S. monitoring data fell by approximately 3.6  $\mu\text{g m}^{-3}$ , a change that is statistically robust under the percentile bootstrap procedure used here. The decline is broadly distributed across states, although the magnitude of change varies and a small number of states show increases or non-robust point estimates. These descriptive results document substantial progress in lowering typical monitored PM<sub>2.5</sub> levels during this period, while underscoring the importance of careful treatment of monitoring-network changes and small-sample states when interpreting geographic heterogeneity. Future work will address these issues more formally and examine potential drivers of the spatial patterning of changes.

## References

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