

Burden of cardiovascular diseases associated with fine particulate matter in Beijing, China: an economic modelling study

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ABSTRACT

Objective To evaluate the economic and humanistic burden associated with cardiovascular diseases that were attributable to fine particulate matter ($\leq 2.5 \mu\text{g}/\text{m}^3$ in aerodynamic diameter; $\text{PM}_{2.5}$) in Beijing.

Methods This study used a health economic modelling approach to compare the actual annual average $\text{PM}_{2.5}$ concentration with the $\text{PM}_{2.5}$ concentration limit ($35 \mu\text{g}/\text{m}^3$) as defined by the Chinese Ambient Air Quality Standard in terms of cardiovascular disease outcomes in Beijing adult population. The outcomes included medical costs, quality-adjusted life-years (QALYs) and net monetary loss (NML). Beijing annual average $\text{PM}_{2.5}$ concentration was around $105 \mu\text{g}/\text{m}^3$ during 2013–2015. Therefore, we estimated the differences in cardiovascular outcomes of Beijing adults between exposure to the $\text{PM}_{2.5}$ concentration of $105 \mu\text{g}/\text{m}^3$ and exposure to the concentration of $35 \mu\text{g}/\text{m}^3$. According to WHO estimates, the hazard ratios of coronary heart disease and stroke associated with the increase of $\text{PM}_{2.5}$ concentration from 35 to $105 \mu\text{g}/\text{m}^3$ were 1.15 and 1.29, respectively.

Results The total 1-year excess medical costs of cardiovascular diseases associated with $\text{PM}_{2.5}$ pollution in Beijing was US\$147.9 million and the total 1-year QALY loss was 92 574 in 2015, amounting to an NML of US\$2281.8 million. The expected lifetime incremental costs for a male Beijing adult and a female Beijing adult were US\$237 and US\$163, the corresponding QALY loss was 0.14 and 0.12, and the corresponding NML was US\$3514 and US\$2935.

Conclusions $\text{PM}_{2.5}$ -related cardiovascular diseases imposed high economic and QALY burden on Beijing society. Continuous and intensive investment on reducing $\text{PM}_{2.5}$ concentration is warranted even when only cardiovascular benefits are considered.

INTRODUCTION

Long-term exposure to ambient fine particulate matter ($\leq 2.5 \mu\text{g}/\text{m}^3$ in aerodynamic diameter; $\text{PM}_{2.5}$) has been linked to increased morbidity and mortality.^{1–2} WHO Global Burden of Disease (GBD) project estimated that over 3.2 million premature deaths worldwide and approximately 2.0 million

Key questions

What is already known?

- There is ample evidence that $\text{PM}_{2.5}$ air pollution increases cardiovascular disease risk.

What are the new findings?

- The total one-year excess medical costs of cardiovascular diseases associated with $\text{PM}_{2.5}$ pollution in Beijing was US\$ 147.9 million and the total one-year QALY loss was 92 574 in 2015, amounting to an NML of US\$ 2281.8 million.

What do the new findings imply?

- Continuous and intensive investment on reducing $\text{PM}_{2.5}$ concentration may be warranted even when only cardiovascular benefits are considered.

premature deaths in East Asia were attributable to ambient air $\text{PM}_{2.5}$ pollution in 2010.³ The impact of $\text{PM}_{2.5}$ pollution in China is particularly substantial because of the severe pollution level and the population size,⁴ and has drawn increasing concerns in recent years.^{5,6}

An important reason high $\text{PM}_{2.5}$ concentration leads to excess morbidity and mortality is that it increases the risks of cardiovascular diseases.^{7,8} In fact, a previous study on Chinese population in Hong Kong found statistically significant association between $\text{PM}_{2.5}$ increase and both coronary heart disease (CHD) and stroke.¹ Hence, it is important to estimate the corresponding economic burden of cardiovascular diseases attributable to $\text{PM}_{2.5}$. Such estimates can provide the foundation for policymaking in the aspects of cardiovascular disease management and environment protection. Beijing, the capital city of China and one of the largest metropolitan areas in the world, is located in an area (Beijing-Tianjin-Hebei) having the highest annual average $\text{PM}_{2.5}$ concentration in China.⁹ The

hazy weather in Beijing has often been featured in global media reports.^{10 11} However, the economic and humanistic burden of cardiovascular diseases in Beijing associated with PM_{2.5} has not been specifically documented in literature so far.

By focusing on a specific disease area, economic modelling allows estimates of direct medical costs and quality-adjusted life-year (QALY) loss. A QALY is a life year weighted by utility that quantifies individuals' or societies' preference for a health state.¹² Utility is assessed on a scale of which 0 represents death and 1 represents full health. Accordingly, the decrement of utility is disutility.¹³ Reliable and thorough quantification of the disease-specific burden can be critical for policy-makers to understand the priority of resource allocation to pollution reduction and cardiovascular disease treatment. As such, the current study aimed to evaluate both excess direct medical costs and QALY loss associated with cardiovascular diseases that were attributable to PM_{2.5} pollution in Beijing using a health economic approach.

METHODS

Model

This study used an economic modelling approach that was borrowed from health economic evaluations. In a health economic evaluation, a model simulates alternative courses of action to compare interventions in terms of healthcare costs and health outcomes.¹⁴ The current study did not aim to evaluate a specific medical or policy intervention. Instead, it compared cardiovascular disease outcomes in terms of healthcare costs and QALY of Beijing adult population when exposed to alternative levels of ambient air PM_{2.5} pollution. Otherwise, it was the same as a health economic evaluation. The model was constructed from the societal perspective.

We created a patient-level discrete-time microsimulation model that allowed age-varying event probabilities and the occurrence of competing events in each cycle.¹⁵⁻¹⁷ The model advanced at a yearly cycle and simulated three cardiovascular events in each cycle: acute myocardial infarction (AMI), stroke and unstable angina. Several previous cardiovascular modelling studies in both Chinese and non-Chinese settings used the same set of events and showed that these three events imposed sizeable burden on healthcare systems.^{18 19} In addition to the cardiovascular events, the model simulated background mortality as an independent event in each cycle to be consistent with the literature.¹⁸⁻²¹ When one or multiple events happened in a cycle, the model calculated corresponding costs and QALY loss, re-evaluated the mortality rate, and updated disease history. Each simulated individual started at 18 years old without any prevalent cardiovascular events, iterated through cycles, and ended at either death or turning 90 years old. The costs and QALYs up to death or 90 years old were accrued across cycles accordingly. To estimate the incremental economic burden and QALY loss associated with

cardiovascular diseases that were attributable to PM_{2.5} air pollution, we conducted the simulation using the actual annual average pollution level in the years 2013–2015 and a relatively low pollution level which was the maximum reasonable PM_{2.5} concentration as defined by air quality standards. Then we compared the individual average lifetime costs and QALY results from the two alternative courses. A flow chart of the model structure is shown in [figure 1](#), additional information of the modelling methodology has been provided elsewhere.²² Each course of simulation modelled 10 000 individuals.

Cardiovascular disease risk calculation and data inputs

Our model used a cardiovascular risk equation from a previous epidemiology study in China (online supplemental materials part 1).²³ This equation took as inputs a set of risk factors including age, systolic blood pressure (SBP), total cholesterol (TC), body mass index (BMI), smoking and diabetes. The output of the equation was a predicted 10-year integrated probability of AMI and stroke. To use the risk prediction in yearly cycles, the 10-year probability was transferred into its 1-year equivalent using the declining exponential approximation of life expectancy (DEALE) method (online supplemental materials part 2).²⁴ The 1-year probability of AMI was then calculated using the 1-year integrated probability of AMI and stroke and the relative incidence of AMI to the composite of AMI and stroke (online supplementary materials part 3).²³ The 1-year probability of stroke was calculated similarly. The 1-year probability of unstable angina was determined using the age-specific ratio of unstable angina incidence to stroke incidence in China (online supplemental material part 4 and online supplemental table S1).²⁵ For each of the events in each cycle, a random number between 0 and 1 from a uniform distribution was generated to be compared with the probability of the given event, which was judged to be incurred if the random number was smaller than the probability and absent otherwise. Probabilities were updated at the beginning of each cycle. We used Beijing-specific data for SBP, TC and BMI,^{26 27} and used overall Chinese population data for smoking and diabetes (online supplemental materials part 5 and online supplemental table S2) because Beijing-specific overall adult smoking information and diabetes prevalence and incidence information were absent in the literature.²⁸

For each simulated individual, BMI was sampled once from the distribution of Beijing male BMI at the age of 43.4 years or female BMI at the age of 44.3 years.²⁶ Following that, the BMI at a certain age in the model was calculated using the relationship between BMI and age in China that was previously estimated in the literature (online supplemental materials part 6).^{26 29} SBP was calculated using a similar approach. To enable such calculation, we estimated the relationship between SBP and age using data from the 2013 wave of the China Health and Retirement Longitudinal Study (online supplemental materials part 7).³⁰ TC input values by age group

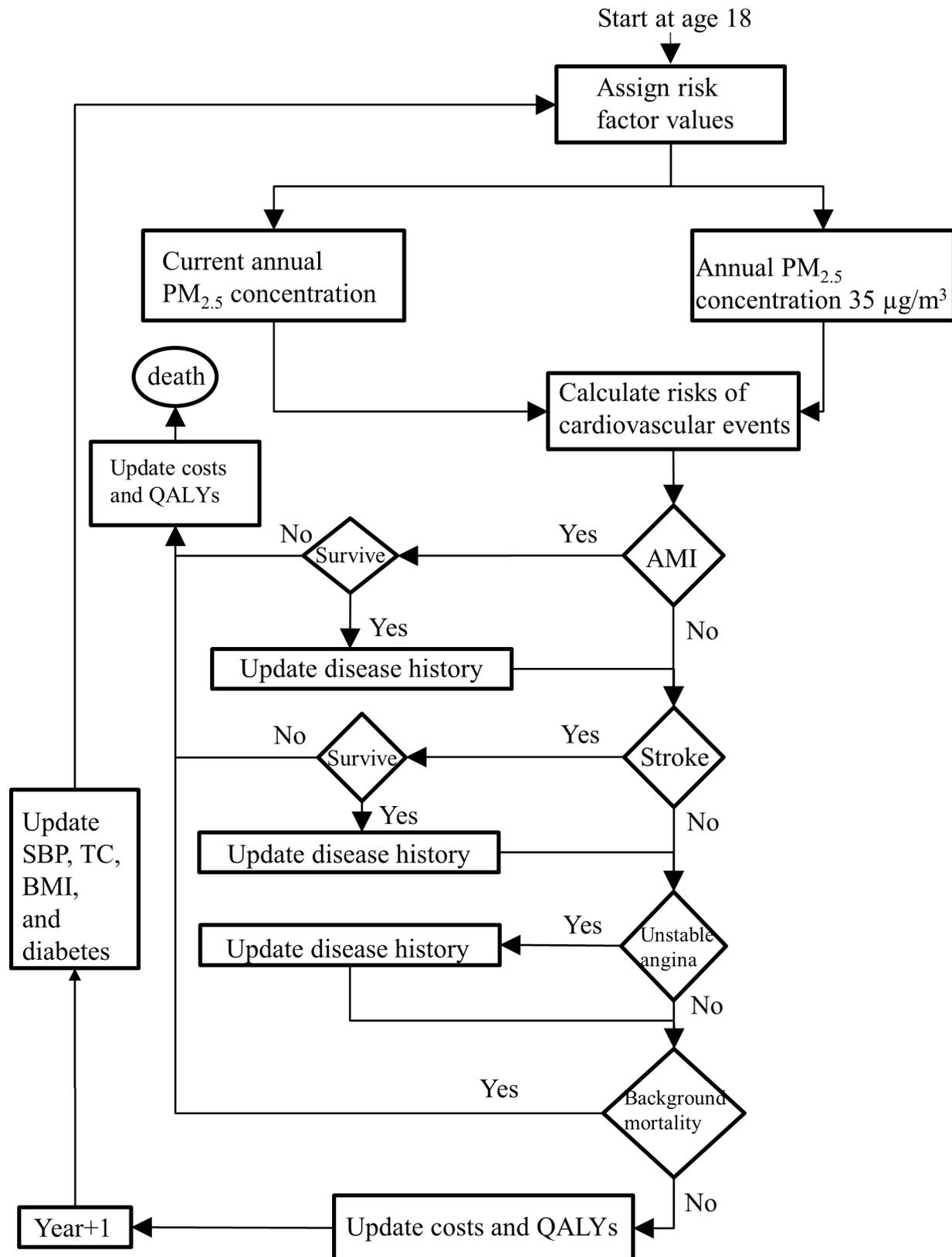


Figure 1 Flow chart of the discrete time microsimulation model. The simulation of individuals started at 18 years old, progressed at 1-year increment, and ended at death or 90 years old. AMI, acute myocardial infarction; BMI, body mass index; PM 25, particulate matter; QALY, quality-adjusted life years; SBP, systolic blood pressure; TC, total cholesterol.

were extracted from an epidemiology study.²⁷ Smoking was determined only once for the lifetime at the beginning of the simulation of each individual.²⁸ The model

also determined whether a given individual had diabetes at the age of 18 using the diabetes prevalence data for Chinese children.³¹ For those who did not have diabetes

at the beginning cycle, their diabetes status was evaluated again in the next cycle using age-specific diabetes incidence data.²⁸ The diabetes re-evaluation process was iterated in all subsequent cycles for those who did not have diabetes up to the previous cycle. [Table 1](#) lists data inputs of risk factors.

PM_{2.5} air pollution level and effect of PM_{2.5} on cardiovascular event risk

The calculated event risks using the risk equation and data inputs described in the section above represented the risks as the status quo. Studies in the literature suggested that the annual average concentration of PM_{2.5} was in the range of 100–105 µg/m³ in Beijing from 2013 to 2015.^{32,33} The Chinese Ambient Air Quality Standards (CAAQS) set the grade II standard of annual average PM_{2.5} concentration at 35 µg/m³.^{4,34} As such, the current analysis evaluated the excess economic burden and QALY loss associated with cardiovascular diseases that could be avoided if the annual average concentration of PM_{2.5} in Beijing was decreased from 105 to 35 µg/m³. To accomplish this, it was necessary to apply the cardiovascular effect of PM_{2.5} to cardiovascular event risk calculation. We used the hazard ratios (HR) of CHD and stroke associated with PM_{2.5} increase estimated by WHO using an integrated exposure response (IER) approach, the latest estimates of which available on the GBD official website were from the 2010 GBD.³⁵ Per the WHO IER estimates, the HRs of CHD and stroke were 1.15 and 1.29 if the ambient PM_{2.5} concentration was increased from 105 to 35 µg/m³. In the simulation course of low pollution, the reciprocals of HRs were multiplied by the status quo CHD and stroke event risks.

Mortality

Background mortality data by age were obtained from WHO country-specific life tables.³⁶ In the model, mortality was affected by an AMI or stroke event in both short and long terms. In the short term, an individual who had an AMI or stroke experienced 28-day excess mortality.^{19,37} The 28-day excess mortality rate was age varying.¹⁹ In the long term, an individual who had any prevalent AMI or stroke or both experienced excess mortality subsequently which were two times of background mortality if new AMI or stroke events did not happen.³⁸ Mortality inputs are listed in [table 2](#).

Outcome measures and inputs

The outcomes of interest in the current study were direct medical costs and QALYs. Direct medical costs included hospitalisation costs, first-year long-term costs, and subsequent year outpatient visit costs ([table 2](#)). AMI and stroke survivors were assumed to require one office visit each year in their residual lifetime, whereas individuals with a history of unstable angina required one office visit each year up to a duration of sequelae that varied with the age of angina occurrence and the sex of the individual (online supplemental materials part 8 and online

supplemental table S3).³⁷ Costs were first inflated to 2015 Chinese currency using the China Healthcare Component Consumer Price Indices and then converted to 2015 US\$ using currency conversion rate in the same year.^{39,40} The calculation of QALYs used both age-specific utility weights and event-related disutility weights ([table 2](#)). The source of age-specific utility weights was from a national health services survey in China using EuroQol five dimensions questionnaire.⁴¹ The 1-month acute disutility weights associated with AMI and stroke and the disutility weight associated with angina were obtained from WHO GBD project.⁴² Post-AMI and poststroke disutility weights were obtained from a quality of life study on Chinese population with chronic conditions and WHO GBD project, respectively.⁴³ In addition, the aforementioned duration of angina sequelae also applied to angina disutility.

We summarised the costs and QALYs using two distinct metrics. First, we calculated the 1-year total excess cardiovascular disease costs and QALY loss of the entire Beijing adult population that were attributable to PM_{2.5} pollution in 2015 (illustrated in online supplemental figure S1). To calculate this, we extracted yearly average incremental costs and QALY loss of each age group for male and female population, multiplied these results with the population of each corresponding age–sex group (online supplemental materials part 9 and online supplemental table S4) in Beijing,^{44,45} and summed the results across age–sex groups. The outcomes were not discounted because they were only evaluated in a 1-year time frame. Second, we simulated the lifetime courses of an 18-year-old Beijing male and female to calculate the individual expected lifetime excess cardiovascular disease costs and QALY loss attributable to PM_{2.5} pollution of male and female adults, respectively. The individual expected lifetime estimates resemble the concept of healthy life expectancy, they represent the average incremental costs and QALY loss that an 18-year-old individual is expected to have in the residual lifetime using current information on all age groups. The calculation of expected lifetime outcome values used an annual discount rate of 3%.^{46,47}

In addition, we calculated the net monetary loss (NML) associated with cardiovascular diseases that were attributable to PM_{2.5} pollution for both metrics. The NML was the sum of direct medical costs and monetised QALY loss. It is common to monetise QALYs in health economic analysis to better understand cost–benefit.⁴⁸ Per WHO recommendations, the willingness to pay (WTP) to monetise each QALY was three times of China gross domestic product (GDP) per capita in 2014.^{49,50} Consequently, the WTP for a QALY in the current study was US\$23 051.

Alternative scenarios and sensitivity analyses

An additional analysis examined the changes in results using an alternative scenario of reducing the annual average PM_{2.5} concentration to 15 µg/m³ because both the CAAQS grade I standard and the United States National Ambient Air Quality Standards secondary

Table 1 Risk factor inputs

Risk factor type	Value			References
SBP (mm Hg)	Distribution	Mean	SD	26
SBP of 43.4 years old men	Normal	134.1	18.2	
SBP of 44.3 years old women	Normal	126.1	19.2	
Men TC (mmol/L)				27
Age 18–19	3.38			
Age 20–24	4.01			
Age 25–29	4.26			
Age 30–34	4.49			
Age 35–39	4.69			
Age 40–44	4.93			
Age 45–49	5.07			
Age 50–54	4.98			
Age 55–59	4.96			
Age 60–64	4.91			
Age 65–69	5.08			
Age 70–74	5.05			
Age 75–79	5.01			
Age≥80	4.91			
Women TC (mmol/L)				27
Age 18–19	3.77			
Age 20–24	4.01			
Age 25–29	4.17			
Age 30–34	4.25			
Age 35–39	4.4			
Age 40–44	4.63			
Age 45–49	4.87			
Age 50–54	5.22			
Age 55–59	5.32			
Age 60–64	5.44			
Age 65–69	5.57			
Age 70–74	5.6			
Age 75–79	5.58			
Age≥80	5.41			
BMI (kg/m ²)	Distribution assumption	Mean	SD	26
BMI of 43.4 years old men	Normal	25.5	3.7	
BMI of 44.3 years old women	Normal	24.5	3.8	
Diabetes prevalence of 18 years old men	0.11%			31
Male 1-year diabetes probabilities				28
Age 20–29	0.25%			
Age 30–39	0.26%			
Age 40–49	0.61%			
Age 50–59	0.44%			
Age 60–69	0.27%			
Age 70–79	0.38%			

Continued

Table 1 Continued

Risk factor type	Value	References
Age \geq 80	0.38%	
Diabetes prevalence of 18 years old women	0.11%	31
Female 1-year diabetes probabilities		28
Age 20–29	0.11%	
Age 30–39	0.18%	
Age 40–49	0.44%	
Age 50–59	0.59%	
Age 60–69	0.74%	
Age 70–79	0.17%	
Age \geq 80	0.17%	
Male smoking percentage	58.20%	28
Female smoking percentage	3.40%	28

BMI, body mass index; SBP, systolic blood pressure; TC, total cholesterol.

standard set the annual average PM_{2.5} concentration limit at 15 $\mu\text{g}/\text{m}^3$.^{34 51} According to WHO estimates, the HRs of CHD and stroke events associated with the increase of PM_{2.5} concentration from 15 to 105 $\mu\text{g}/\text{m}^3$ are 1.27 and 1.74, respectively.³⁵

In addition, one-way sensitivity analyses were conducted for the individual lifetime analysis. Specifically, key input parameters including hospitalisation costs associated with AMI and stroke, 28-day mortality associated with AMI and stroke, and the HR of PM_{2.5} exposure, smoking rates, and diabetes prevalence/incidence rates were increased or decreased by 15% in each sensitivity analysis. More, the annual discount rate was adjusted to 5% and 1% in a set of sensitivity analyses. Even more, we decreased the population background mortality at each age by 10% to reflect longer life expectancy in future as a sensitivity analysis. Finally, the base-case QALY results were monetised using once instead of three times the GDP per capita to estimate the NMLs under the alternative WTP.⁵² One-way sensitivity analyses were conducted for male and female individuals separately and the results were summarised using NML.

The model was programmed using Excel 2013 VBA (Microsoft).

Patient and public involvement

It was not appropriate or possible to involve patients or the public in the design, or conduct, or reporting, or dissemination plans of our research.

RESULTS

Table 3 lists 1-year excess costs and QALY loss in Beijing population associated with cardiovascular diseases that were attributable to PM_{2.5} pollution in 2015. The excess direct costs in male and female Beijing population were US\$87.9 million and US\$60.1 million, respectively, summing to total excess direct costs of US\$147.9 million. The QALY loss in male and female Beijing population

was 44 186 and 48 388. Consequently, the total QALY loss was 92 574. Accordingly, the NML in male, female, and the entire Beijing population was US\$1106.4 million, US\$1175.4 million, and US\$2281.8 million, respectively. Furthermore, the total direct costs of cardiovascular diseases in Beijing was US\$1056.9 million when calculated using our model outputs (not listed in the tables). Hence, about 14% of direct costs of cardiovascular diseases in Beijing was attributable to PM_{2.5} pollution.

Table 4 presents expected lifetime excess costs and QALY loss for male and female individuals at the age of 18. The expected incremental costs for a male Beijing adult and female Beijing adult were US\$237 and US\$163. The corresponding QALY loss was 0.14 and 0.12. Consequently, the NML for an average male and an average female Beijing adult was US\$3514 and US\$2935.

As expected, the excess costs and QALY loss using 15 $\mu\text{g}/\text{m}^3$ as the standard PM_{2.5} concentration level were greater than using 35 $\mu\text{g}/\text{m}^3$ as the standard PM_{2.5} concentration level. In this alternative scenario, 1-year excess direct costs in male, female and the entire Beijing population were US\$149.6 million, US\$115.2 million, and US\$264.8 million, respectively. The QALY loss in male, female and the entire Beijing population was 81 040, 71 394, and 152 434, respectively. Hence, the corresponding NML in the three population groups was US\$2017.6 million, US\$1760.9 million, and US\$3778.5 million. In addition, the expected lifetime excess costs, QALY loss and NML for a male Beijing adult and female Beijing adult using the alternative scenario were US\$364 and US\$275, 0.24 and 0.19, and US\$5924 and US\$4763.

One-way sensitivity analyses results are shown in online supplemental figure S1. In summary, increasing or decreasing discount rate by 2% had the most prominent impact on the expected individual lifetime NML estimates. Changing the effect of pollution on cardiovascular events by $\pm 15\%$ also changed the expected individual lifetime NML estimates substantially. By contrast, $\pm 15\%$ changes

Table 2 Mortality, utility and costs inputs

Input type	Value		References
	Men	Women	
Annual background mortality			³⁶
Age 18–19	0.009%	0.007%	
Age 20–24	0.012%	0.011%	
Age 25–29	0.013%	0.011%	
Age 30–34	0.014%	0.013%	
Age 35–39	0.020%	0.017%	
Age 40–44	0.035%	0.028%	
Age 45–49	0.059%	0.043%	
Age 50–54	0.099%	0.070%	
Age 55–59	0.176%	0.119%	
Age 60–64	0.307%	0.211%	
Age 65–69	0.527%	0.373%	
Age 70–74	0.894%	0.653%	
Age 75–79	1.484%	1.128%	
Age 80–84	2.600%	2.063%	
Age 85–89	4.223%	3.512%	
28 day post-AMI mortality	Men	Women	¹⁹
35–44	12%	18%	
45–54	21%	23%	
55–64	29%	27%	
65–74	33%	43%	
75–84	48%	51%	
28 day poststroke mortality	Men	Women	¹⁹
35–44	25%	18%	
45–54	18%	14%	
55–64	12%	15%	
65–74	20%	20%	
75–84	45%	45%	
Age-specific utility	Men	Women	⁴¹
15–19	0.898	0.896	
20–24	0.888	0.882	
25–29	0.878	0.867	
30–34	0.86	0.848	
35–39	0.848	0.832	
40–44	0.834	0.815	
45–49	0.814	0.792	
50–54	0.793	0.772	
55–59	0.774	0.752	
60–64	0.751	0.728	
65–69	0.725	0.702	
70–74	0.701	0.685	
75–79	0.684	0.669	
80–84	0.662	0.655	
85–89	0.661	0.643	

Continued

Table 2 Continued

Input type	Value		References
	Men and women		
One-month acute disutility			⁴²
AMI	0.439		
Stroke	0.920		
Postevent long-term disutility			^{42 43}
AMI	0.107		
Stroke	0.266		
Unstable angina	0.124		
Costs (in 2018 CN ¥)	Men and women		
	AMI and unstable angina	Stroke	
Hospitalisation costs (2015 US\$)	US\$2630	US\$1977	⁶⁴
First year long-term costs (2015 US\$)	US\$602	US\$369	¹⁹
Office visit costs in subsequent years (2015 US\$)	US\$66	US\$77	⁶⁵

AMI, acute myocardial infarction.

in hospitalisation costs, 28-day mortality associated with AMI and stroke, smoking rates and diabetes prevalence/incidence rates did not have substantial impacts on the expected individual lifetime NML estimates. Similarly, decreasing background mortality by 10% caused modest changes in results. When once the GDP per capita was used to monetise the QALYs (online supplementary materials part 10, online supplemental table S5), the NML among men, women and the entire adult population for 1 year was US\$421.4 million, US\$437.8 million and US\$859.2 million. The expected lifetime NML for a male Beijing and female Beijing adult under the alternative WTP was US\$1329 and US\$1087, respectively.

DISCUSSION

We found that PM_{2.5}-related cardiovascular diseases imposed high direct and indirect economic burden on both the entire Beijing population and each Beijing individual during the lifetime course. Consequently, the total economic loss associated with cardiovascular diseases due to PM_{2.5} was substantial. The NML that was attributable to PM_{2.5}-related cardiovascular diseases in the base-case scenario accounted for 0.68% of Beijing GDP.⁵³ Our results also suggested that QALY loss contributed more than direct medical costs by a sizeable margin to the NML associated with PM_{2.5} pollution in Beijing. In addition, PM_{2.5} pollution accounted for 14% of direct medical costs of cardiovascular diseases in Beijing. More, the NML further increased by over 60% if the standard PM_{2.5} concentration level was 15 µg/m³ instead of 35 µg/m³. Hence, it is important to note that additional economic

Table 3 Excess costs and QALY loss in Beijing population associated with cardiovascular diseases that are attributable to PM_{2.5} ambient air pollution in 2015

	Excess costs* (million)	QALY loss†	NML‡ (million)
Male			
Age 18–19	US\$0.3	30	US\$1.0
Age 20–24	US\$0.7	371	US\$9.2
Age 25–29	US\$2.8	2598	US\$62.7
Age 30–34	US\$2.2	1395	US\$34.4
Age 35–39	US\$4.0	2795	US\$68.4
Age 40–44	US\$4.5	2942	US\$72.3
Age 45–49	US\$6.5	3043	US\$76.7
Age 50–54	US\$8.3	2610	US\$68.5
Age 55–59	US\$6.1	3881	US\$95.6
Age 60–64	US\$12.5	4160	US\$108.4
Age 65–69	US\$5.2	4697	US\$113.5
Age 70–74	US\$7.0	4774	US\$117.0
Age 75–79	US\$12.2	4730	US\$121.3
Age 80–84	US\$10.1	4699	US\$118.5
Age 85–89	US\$5.4	1461	US\$39.1
Male total	US\$87.9	44 186	US\$1106.4
Female			
Age 18–19	US\$0.1	1	US\$0.1
Age 20–24	US\$0.4	175	US\$4.4
Age 25–29	US\$0.6	246	US\$6.3
Age 30–34	US\$0.7	415	US\$10.3
Age 35–39	US\$0.6	359	US\$8.9
Age 40–44	US\$0.7	581	US\$14.1
Age 45–49	US\$1.9	701	US\$18.1
Age 50–54	US\$4.5	3215	US\$78.6
Age 55–59	US\$8.0	4458	US\$110.7
Age 60–64	US\$5.7	4169	US\$101.8
Age 65–69	US\$7.9	4837	US\$119.4
Age 70–74	US\$8.3	5610	US\$137.6
Age 75–79	US\$7.2	9891	US\$235.2
Age 80–84	US\$6.4	7853	US\$187.4
Age 85–89	US\$7.0	5878	US\$142.5
Female total	US\$60.1	48 388	US\$1175.4
Total	US\$147.9	92 574	US\$2281.8

*Excess costs were the difference between the costs of the population (or the population in a certain age group) when exposed to the PM_{2.5} concentration of 105 µg/m³ and the costs of the same population when exposed to the PM_{2.5} concentration of 35 µg/m³.

†QALY loss was the difference between the QALYs of the population (or the population in a certain age group) when exposed to the PM_{2.5} concentration of 35 µg/m³ and the QALYs of the same population when exposed to the PM_{2.5} concentration of 105 µg/m³.

‡NML was the sum of excess costs and monetised QALY loss. NML, net monetary loss; PM, particulate matter; QALY, quality-adjusted life-years.

benefit should be expected if PM_{2.5} concentration can be further reduced to below 15 µg/m³.

All Chinese cities have a target of controlling PM_{2.5} concentration at or below 35 µg/m³ by 2030.⁵⁴ Beijing spends around US\$3 billion annually to reduce air pollution in the recent years.^{55 56} Our results suggest that the incremental net monetary benefit from decreased cardiovascular diseases alone can offset over two-thirds of the spending if the PM_{2.5} concentration can be reduced to 35 µg/m³ and offsets the total spending if the PM_{2.5} concentration can be reduced to 15 µg/m³. The economic benefit of reduced air pollution will be substantially greater if the avoided loss in future is considered. Even more, reduced pollution can result in cost savings in other disease areas. As such, the cumulative monetary benefit will likely outweigh the investment eventually, if not immediately. Therefore, continuing the current efforts or even expanding the efforts to aggressively reduce the PM_{2.5} concentration should be considered.

The current study has several strengths and weaknesses. Using a health economic modelling approach, the current study could estimate cardiovascular disease burden associated with long-term exposure to PM_{2.5} pollution in terms of costs and QALYs. We know of one study that estimated chronic kidney disease-related disability-adjusted life year burden that was attributable to ambient air pollution using population-level reweighting.⁵⁷ However, we know of no other studies that used economic simulations to approach such issues. The economic modelling approach may be appealing to researchers by affording the opportunity to evaluate both population-level disease burden in a period and individual residual lifetime disease burden imposed by PM_{2.5} pollution. The population estimates gave a snapshot of the economic burden in 2015, whereas the individual expected lifetime estimates took into account the longitudinal excess costs and QALY loss. More, the model allowed exploration of various scenarios, which was not possible using retrospective data analysis. In fact, the simulation tool that we developed can be adapted to explore other scenarios than the ones already examined should it be necessary for interested parties to change input parameters. Finally, we chose microsimulation over the common practice of Markov cohort state-transition model. Compared with the latter, the former allows multiple events in one cycle and keeps track of disease history of each simulated individual.^{16 58} Failure to account for multiple events per cycle may lead to downward bias of the burden of ambient air pollution since a fraction of the adverse outcomes might not be reflected in the model.

Potential limitations should be considered when interpreting our results. Economic modelling has an innate limitation of making assumptions about data and course of action, and sensitivity analyses cannot examine model robustness to each assumption. One of the strong assumptions we made was that the population under investigation had been continuously exposed to PM_{2.5} pollution to experience the adverse effects and would still be exposed in their residual lifetime.

Table 4 Expected lifetime excess costs and QALY loss of an 18-year-old Beijing male individual and an 18-year-old Beijing female individual associated with cardiovascular diseases that are attributable to PM_{2.5} ambient air pollution in 2015

	Current PM _{2.5} concentration (105 µg/m ³)	PM _{2.5} concentration at 35 µg/m ³	Difference
Male			
Expected lifetime costs associated with cardiovascular disease	US\$1692	US\$1455	US\$237
Expected lifetime QALYs	23.49	23.63	-0.14
Net monetary benefit	US\$539 802	US\$543 316	-US\$3514
Female			
Expected lifetime costs associated with cardiovascular disease	US\$1247	US\$1084	US\$163
Expected lifetime QALYs	23.49	23.61	-0.12
Net monetary benefit	US\$540 259	US\$543 194	-US\$2935

.PM, particulate matter; QALY, quality-adjusted life-years.;

In other words, we assumed the population of interest was static. Numerous studies have considered exposures of at least 9–12 months in duration as long term.^{59 60} However, the demographic structure is in fact highly dynamic and a small fraction of the population may migrate between low pollution and high pollution regions even within a 9-month period. This assumption affected all results of the current study including population 1-year estimates and individual expected lifetime estimates. Another noteworthy assumption was associating PM_{2.5} cardiovascular effects with annual average PM_{2.5} concentration. We used the annual average to be compliant with available effect estimates in the literature. However, the actual long-term effect of PM_{2.5} pollution on cardiovascular diseases may not necessarily be represented by annual average concentration. The model was additionally based on several assumptions which are listed in online supplementary materials part 11 online supplemental table S6. In addition, the current study examined the economic burden associated with long-term exposure to PM_{2.5}, but not short-term exposure. Short-term exposure also has cardiovascular effects.⁸ Unfortunately, it is not technically feasible to model both long-term effects and short-term effects in one model since the simulation advances at a chosen time unit (yearly increment in the current study) and the time unit cannot accommodate both long-term and short-term effects. More, despite that focusing on cardiovascular diseases provided insights into certain aspects, it also limited the understanding of overall burden. Even more, the current study investigated the cardiovascular disease outcomes associated with PM_{2.5}, but not other air pollutants such as PM_{10-2.5}. We focused on PM_{2.5} because of the relatively well-established epidemiology evidence on PM_{2.5} and the fact that the combined health effect of air pollutants on cardiovascular events is not well understood at present.⁷ Although some evidence suggests that the finest PM_{2.5} has a stronger cardiovascular effect than the relatively coarse PM_{10-2.5}, the effect of PM_{10-2.5} is nontrivial.⁸ Therefore, it is reasonable to expect that the economic burden of cardiovascular diseases

is even greater than that estimated in the current study if the effects of other air pollutants are considered. Finally, the IER estimates in the analysis were not necessarily up to date. However, the relatively recent estimates of 2016 GBD were only presented for concentration intervals, failing the need of calculating HRs based on exact concentrations.

To our best knowledge, the current study represents the first attempt to estimate both economic and humanistic burden associated with PM_{2.5}-related cardiovascular diseases of any area. Thus, the current study is not directly comparable to results from existing studies in the literature. However, previous studies estimated that the total direct and indirect economic loss due to PM_{2.5} pollution accounted for 0.76%–2.75% of Beijing GDP.^{11 61} By including QALY loss in the calculation, we estimated that the NML due to cardiovascular diseases alone associated with PM_{2.5} pollution accounted for 0.68% of Beijing GDP. Although the estimates are not directly comparable, they provide face validity to each other. These numbers also resonate the estimates in the UK where the burden of air pollution in monetary terms was equivalent to 1.5% of GDP.⁶² Beyond monetary terms and cardiovascular diseases, *Bowe et al* has shown PM_{2.5} caused about 6.6 million disability-adjusted life years globally per annum due to chronic kidney diseases. These estimates, put together with the present findings, suggest that much remains to be done to reduce the disease burden of ambient air pollution.

Future research should accomplish two immediate tasks in the near term. First, similar models should be developed to estimate direct costs and QALY loss of other diseases to which the contribution of PM_{2.5} is strongly supported by epidemiology evidence including but not limited to chronic obstructive pulmonary disease, lung cancer and acute lower respiratory infections.⁶³ This exercise will help to provide additional insights on the disease-specific economic burden of PM_{2.5} pollution. Second, analyses of the economic burden of cardiovascular diseases caused by PM_{2.5} pollution should expand to other regions in China as well as other low-income

and middle-income countries that have similarly high levels of air pollution. In 2015, 165 out of 190 priority pollution monitoring cities in China did not meet the CAAQS Grade II standard of annual average PM_{2.5} concentration,⁹ which highlights the importance of expanding the analysis to other areas since the total economic burden of cardiovascular diseases caused by PM_{2.5} pollution in the broader population is expected to be much higher than that of Beijing alone.

CONCLUSIONS

PM_{2.5}-related cardiovascular diseases lead to substantial excess medical costs and QALY loss in Beijing adult population. Intensive and effective initiatives to reduce the PM_{2.5} concentration to 35 µg/m³ or even lower levels should be considered by policymakers to reduce the burden of cardiovascular diseases in Beijing.

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Supplementary materials

A Health Economic Approach to Estimate the Burden of Cardiovascular Diseases
Associated with Fine Particulate Matter: The Case of Beijing and Its Implications

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Part 1. Cardiovascular event risk equation

The cardiovascular risk equation was extracted from a previous epidemiology study in China.¹

10-year integrated probability of AMI and stroke for Chinese male=

$$1 - 0.9835^{\exp(\sum \beta_i X_i)},$$

where $\sum \beta_i X_i = 0.0656 \times age_i - 0.5488 \times I(SBP_i < 120) + 0.4011 \times I(130 \leq SBP_i \leq 139) + 0.8073 \times I(139 < SBP_i \leq 159) + 1.7041 \times I(159 < SBP_i \leq 179) + 2.5327 \times I(SBP_i > 179) + 0.2864 \times I(BMI_i \geq 24) - 0.007 \times I(3.62 \leq TC_i < 5.17) + 0.304 \times I(TC_i \geq 5.17) + 0.7082 \times SMOKING_i + 0.0651 \times DIABETES_i - 3.5951$.

10-year integrated probability of AMI and stroke for Chinese female=

$$1 - 0.9948^{\exp(\sum \beta_i X_i)},$$

where $\sum \beta_i X_i = 0.0851 \times age_i - 0.8267 \times I(SBP_i < 120) + 0.2294 \times I(130 \leq SBP_i \leq 139) + 0.7970 \times I(139 < SBP_i \leq 159) + 1.3674 \times I(159 < SBP_i \leq 179) + 1.8483 \times I(SBP_i > 179) + 0.6762 \times I(BMI_i \geq 24) - 0.0866 \times I(3.62 \leq TC_i < 5.17) + 0.2654 \times I(TC_i \geq 5.17) + 0.4699 \times SMOKING_i + 0.9599 \times DIABETES_i - 3.8780$.

In the equations above, $I(\cdot)$ is an indicator function of whether the condition in the parentheses is true, SBP_{*i*} is the systolic blood pressure (mm Hg) of the simulated individual, TC_{*i*} is the total cholesterol (mmol/L), BMI is the body mass index (kg/m²), SMOKING_{*i*} is an indicator of smoking, and DIABETES_{*i*} is an indicator of having diabetes.

Part 2. The DEALE method and transferring 10-year probability to 1-year probability

The DEALE method assumes an exponential distribution of time to event and allows algebraic calculation of 1-year probability from 10-year probability.² Specifically, the following formulas were used:

$$r_{1year} = -\left(\frac{1}{10}\right) \cdot \ln[1 - (p_{10year})],$$

$$p_{1year} = 1 - e^{-r_{1year}}$$

where r_{1year} is the 1-year rate, p_{1year} is the 1-year probability, and p_{10year} is the 10-year probability.

Part 3. Proportions of AMI and stroke

The respective proportions of AMI and stroke out of total AMI and stroke events were calculated using numbers from the cardiovascular risk equation estimation study¹. The numbers of AMI and stroke events were 66 and 158 among men, and were 39 and 108 among women. Using these numbers for calculation, the proportions of AMI and stroke were 29.46% and 70.54% among men, and were 26.53% and 73.57% among women. The proportions were not age-varying because the source data did not allow age-varying data extraction.

As such, the 1-year probability of AMI was calculated as $p_{AMI} = p_{AMI\&stroke} \times 29.46\%$ and the 1-year probability of stroke was calculated as $p_{stroke} = p_{AMI\&stroke} \times 70.54\%$ for men. The probabilities for women were calculated using the same approach.

Part 4. Angina to stroke ratio

We calculated angina-stroke incidence ratio by age group (Table S1) using age group specific angina and stroke prevalence data.³

Table S1. Angina-stroke incidence ratio

	Self-reported prevalence (%)		Incidence in the corresponding period (%)		Incidence ratio
	stroke	angina	stroke	angina	Angina/stroke
<59	1.5	4.4	1.5	4.4	2.93
60-69	3.4	9.0	1.9	4.6	2.42
70-79	5.3	13.1	1.9	4.1	2.16
>80	7.1	14.5	1.8	1.4	0.78

Part 5. Diabetes prevalence to incidence conversion

The diabetes prevalence rate of Chinese children is 0.11%.⁴ The exact prevalence rate of 18 years old Chinese people was not available. Diabetes prevalence rates of Chinese adults were obtained from a Chinese epidemiology study.⁵ We calculated 10-year incidence rates of diabetes (Table S2) using prevalence rates and transferred 10-year incidence rates to 1-year probabilities using the aforementioned DEALE method. Age group 70-79 incidence rate was extrapolated to age group 80-89.

Table S2. Diabetes prevalence rates and incidence rates of Chinese population

	Prevalence rate	10-year incidence rate	rate	1-year probability
male				
18-19	0.00108	NA	NA	NA
20-29	0.0261	0.02502	0.002534	0.25%
30-39	0.0521	0.026	0.002634	0.26%
40-49	0.1111	0.059	0.006081	0.61%
50-59	0.1547	0.0436	0.004458	0.44%
60-69	0.1813	0.0266	0.002696	0.27%
70-79	0.2185	0.0372	0.003791	0.38%
80-89	NA	0.0372 ^a	0.003791 ^a	0.38% ^a
female				
18-19	0.00108	NA	NA	NA
20-29	0.0124	0.01132	0.001138	0.11%
30-39	0.0301	0.0177	0.001786	0.18%
40-49	0.073	0.0429	0.004385	0.44%
50-59	0.1309	0.0579	0.005964	0.59%
60-69	0.2028	0.0719	0.007462	0.74%
70-79	0.2197	0.0169	0.001704	0.17%
80-89	NA	0.0169 ^a	0.001704 ^a	0.17% ^a

^a: Extrapolated from the adjacent age group.

Part 6. Relationship between BMI and age

The relationship between BMI and age is as following.⁶

Male: BMI = BMI at 43.4 years old + 0.002 × 43.4² - 0.22 × 43.4 + 0.22 × age - 0.002 × age²

Female: BMI = BMI at 44.3 years old + 0.002 × 44.3² - 0.22 × 44.3 + 0.22 × age - 0.002 × age²

Part 7. Relationship between SBP and age

The relationship between SBP and age is as following:

Male: SBP = SBP at 43.4 years old + 0.64 × (age - 43.4)

Female: SBP = SBP at 44.3 years old + 0.64 × (age - 44.3).

The following regression equation was used to estimate the linear relationship between SBP and age of Chinese using the 2013 wave of the China Health and Retirement Longitudinal Study (CHARLS):

$$SBP = \alpha + \beta_1 age + \beta_2 I_{male} + \beta_3 I_{married} + \beta_4 BMI + \beta_5 I_{dd} + \beta_6 I_{smoke} + \beta_7 GH$$

where I_{male} is an indicator for male, $I_{married}$ is an indicator for currently married, I_{dd} is an indicator for drink daily or more often, and GH is self-reported general health (1=excellent, 5=poor). More information on CHARLS can be found at <http://charls.pku.edu.cn/en>.

Part 8. Angina sequelae duration

To be consistent with literature, the duration of angina sequelae varied across age and gender groups (Table S3).⁷

Table S3. Angina sequelae duration (years)

age	male	female
18-19	26	30
20-24	26	30
25-29	26	30
30-34	16	20
35-39	16	20
40-44	16	20
45-49	11	13
50-54	11	13
55-59	11	13
60-64	8	9
65-69	8	9
70-74	5	5
75-79	5	5
80-84	3	3
85-89	3	3

Part 9. Beijing population by age and sex group

The Beijing population by age group⁸ and the Beijing population age-sex pyramid⁹ were used to calculate the population in each age group for each sex (Table S4).

Table S4. Beijing population in 2014 by age-sex group

age	% female population in the age group out of total Beijing population	% male population in the age group out of total Beijing population	% female in this age group	% male in this age group	2014 population in age group	2014 female population	2014 male population
15-19	2.53	2.85	46.97	53.03	846,000	397,343	448,657
20-24	6.16	6.66	48.07	51.93	2,237,000	1,075,329	1,161,671
25-29	5.58	5.93	48.50	51.50	2,436,000	1,181,486	1,254,514
30-34	4.27	4.68	47.71	52.29	2,297,000	1,095,895	1,201,105
35-39	3.91	4.46	46.69	53.31	1,682,000	785,319	896,681
40-44	3.76	4.39	46.16	53.84	1,836,000	847,454	988,546
45-49	3.76	4.24	47.00	53.00	1,708,000	802,780	905,220
50-54	3.33	3.44	49.17	50.83	1,615,000	794,163	820,837
55-59	2.96	3.00	49.70	50.30	1,513,000	751,887	761,113
60-64	1.95	1.90	50.56	49.44	1,093,000	552,609	540,391
65-69	1.51	1.32	53.40	46.60	715,000	381,790	333,210
70-74	1.44	1.32	52.17	47.83	495,000	258,223	236,777
75-79	1.00	1.02	49.40	50.60	456,000	225,253	230,747
80-84	0.64	0.66	49.14	50.86	293,000	143,991	149,009
85-89	0.42	0.37	53.34	46.66	164,000	87,473	76,527

Part 10. Monetize QALYs using an alternative WTP threshold

Table S5. Population and individual expected lifetime NML using the once the GDP per capita WTP threshold

	Male	Female	Total
Population 1-year NML	\$427,396,864	\$431,837,990	\$859,234,855
Individual expected lifetime NML	\$1,329	\$1,087	NA

NA, not applicable.

Part 11. Model assumptions

Table S6. Summary of assumptions

Assumption category	Assumption
Population	The population have been exposed long enough to experience the full hazard of PM _{2.5} pollution.
Risk calculation	Incidence ratios of angina to stroke in all groups in Beijing are the same as those in entire China
Risk calculation	Smoking status of an individual doesn't change
Risk calculation	When the data for the age group 80-89 are not available, the data from 70-79 are used (diabetes). Similarly, the data for age group 80-84 (angina sequelae duration) or 75-84 (AMI and stroke 28-day mortality) are used for age group 85-89 when the relevant data are absent.
Risk calculation	The relationships between SBP and age and between BMI and age of overall Chinese population are portable to Beijing population.
Risk calculation	Diabetes prevalence and incidence data for overall Chinese population are portable to Beijing population.
Risk calculation	Risk factors of cardiovascular diseases are not correlated.
Risk calculation	Without risk factors, the baseline hazard function of cardiovascular disease of Beijing population is the same as that of overall Chinese population.
Risk calculation	The CVD risk equation from the reference represents average individual risk at the current pollution level
Costs and QALY inputs	Unstable angina requires hospitalization.
Effect of pollution	The long-term effect of PM _{2.5} pollution is represented by annual average concentration.

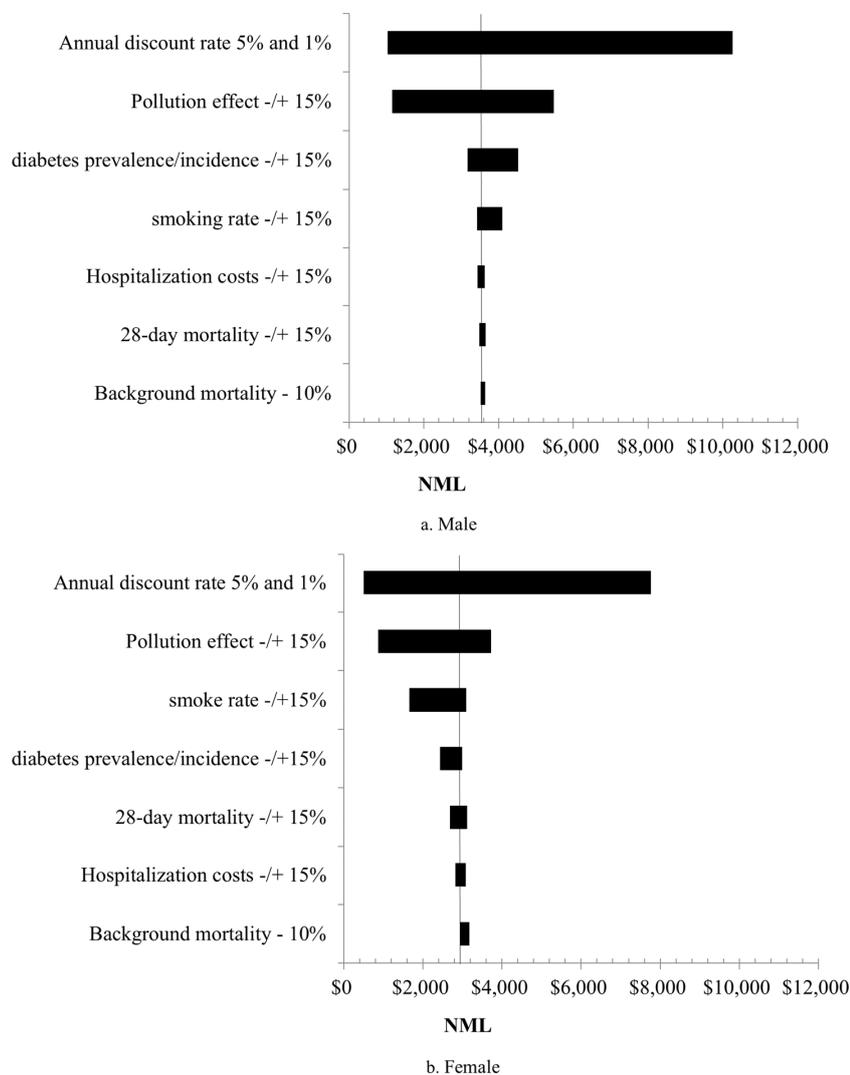


Fig S1. One-way sensitivity analyses for individual male (a) and female (b). The expected individual lifetime net monetary loss (NML) results were sensitive to changes in the annual discount rate and pollution effect.

Supplementary Materials References

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