Predicted health impacts of urban air quality management

J Mindell, M Joffe

Study objective: The 1995 UK Environment Act required local authorities to review air quality and, where UK National Air Quality Strategy objectives (except ozone) are likely to be exceeded in 2005, to declare local air quality management areas and prepare action plans. This study modelled the impacts on health of reductions from current levels of \( \text{PM}_{10} \) to these objectives.

Design: The framework for conducting quantified health impact assessment assessed causality, then, if appropriate, examined the shape and magnitude of the exposure-response relations. The study modelled declines in pollution to achieve the objectives, then modelled the numbers of deaths and admissions affected if air pollution declined from existing levels to meet the objectives, using routine data.


Main results: Attaining the 2004 \( \text{PM}_{10} \) 24 hour objective in Westminster results in 1–21 lives no longer shortened in one year (annual deaths 1363). Reducing exceedences from 35 to seven almost doubles the estimates. The 2009 objective for the annual mean requires a substantial reduction in \( \text{PM}_{10} \), which would delay 8–20 deaths. About 20 respiratory and 14–20 circulatory admissions would be affected and around 5% of emergency hospital attendances for asthma by attaining the lower annual mean target. The effects of long term exposure to particulates may be an order of magnitude higher: models predict about 24 deaths are delayed by reaching the 2004 annual target \( 40 \mu g/m^3 \)\[^{\text{gravimetric}}\] and a hundred deaths by reducing annual mean \( \text{PM}_{10} \) to 20 \( \mu g/m^3 \)\[^{\text{gravimetric}}\].

Conclusions: Modelling can be used to estimate the potential health impacts of air quality management programmes.

T
he United Kingdom 1995 Environment Act required the development of a National Air Quality Strategy for the UK.\(^1,2\) This set standards for eight outdoor air pollutants: sulphur dioxide, lead, carbon monoxide, ozone, nitrogen dioxide, benzene, 1,3-butadiene, and particulates, measured as the mass of \( \text{PM}_{10} \) particles, collected through a size selective inlet that collects 50% of particles of 10 \( \mu \)m aerodynamic diameter, >95% of 5 \( \mu \)m particles and <5% of 20 \( \mu \)m diameter particles (hence \( \text{PM}_{10} \); particulate mass with less than 10 \( \mu \)m diameter). The act required each local authority to review air quality and, where target levels other than for ozone are likely to be exceeded in the year 2005, to declare a local Air Quality Management Area, and prepare an action plan.

Many of the highest levels in the UK of \( \text{PM}_{10} \) and nitrogen dioxide are found in central London.\(^3,4\) Westminster City Council, a central London borough with an estimated resident population of 216 244 in 2000 and an average annual mortality of 1363, has designated the whole borough an Air Quality Management Area as it will continue to exceed the UK National Air Quality Strategy targets for \( \text{PM}_{10} \) and nitrogen dioxide, although it will meet the objectives for the other pollutants.

We modelled the acute impacts on health of meeting the UK National Air Quality Strategy objectives, which require a reduction from the current levels of these two pollutants to the targets shown in table 1. We applied the method to Westminster, using a model that is generalisable to any area with raised levels of pollution.

METHOD

We developed a four stage framework for quantifying the potential health impacts of air pollution reduction (fig 1).

Literature review

We conducted a systematic literature review of epidemiological papers examining the effects of particulates and/or nitrogen dioxide, following the principles given by the World Health Organisation guidance\(^5\) both for searching for relevant publications and for appraising the studies. A comprehensive bibliographic search was conducted, using Medline (1966–2002); BIDS-Embase (1980–2002); Biological Abstracts (1991–1998) and the Health Management Information Consortium (HMIC) CD-ROM, which included “grey” literature. Other sources of information were expert reports commissioned by the government and publications cited in other references. Search terms are detailed by Mindell.\(^6\)

The search produced 410 studies of particulate matter, published by December 2001, of which 38 were of poor quality or irrelevant. An additional 52 studies were published in 2002. Assessment of the effects of air pollution is based on studies with good data collection, appropriate analysis, and in circumstances that can be extrapolated to pollution in London.\(^4\) Altogether 220 epidemiological studies were found on the effects of nitrogen dioxide, of which nine were of poor quality or irrelevant. An additional 27 studies were published in 2002.

Is there a causal association?

The published studies and reports were examined firstly for evidence of an association. Where one of the pollutants was associated with a specific effect on health, consideration was given to whether or not this was likely to be attributable to confounding. Where unconfounded associations were found, consideration was then given to the possibility that these were likely to be causal relations, by assessing the evidence using Bradford Hill’s viewpoints.\(^7\)

Abbreviations: APHEA, short term effects of air pollution on health: European approach; NHMAPS, national mortality and morbidity air pollution study; COPD, chronic obstructive pulmonary disease; TEO, tapered element oscillating microbalance; ACS, American Cancer Society study; GAM, generalised additive models

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The systematic review concluded that in the case of acute exposure to ambient particles, the evidence for a causal relation was sufficiently strong for the range of health outcomes given in table 2. Fewer studies have examined the long-term effects of air pollution; particulates have been associated with increased mortality in three American and one Dutch cohort studies, even after re-analysis adjusting for a wide range of socio-demographic and other potential confounders. The effect is likely to be causal.

For nitrogen dioxide, the evidence is much less clear. Although recent studies have generally shown associations between the ambient level of the gas and a number of cardiorespiratory outcomes, the evidence for causality is much weaker. The consensus is that the relation is probably not causal but that nitrogen dioxide is almost certainly acting as a proxy measure for another pollutant from traffic, probably fine particulates. We have therefore not presented results of health effects for reducing nitrogen dioxide.

**Estimate of the health effect of a unit change in exposure**

There are three elements to this: determining the shape of the relation, whether there is a threshold, and the magnitude of the effect.

Several studies have examined different models and have demonstrated a log-linear relation between average particulate concentration and relative risk. Most studies have assumed that the relation between relative risk and pollution concentration is linear, that is relative risk or percentage change in outcome is linear used here in this sense. Although some studies found the relation to be curvilinear, these are at concentrations well in excess of those now found in the UK; the WHO advised that in the range 20–200 μg/m³ PM10, the relation is linear.

No threshold (or at least none above about 20 μg/m³) has been found for acute effects on mortality in Madrid, Milan, Seattle, Birmingham (Alabama), London, or Atlanta, or for deaths from COPD or heart failure in Milan. For admissions for pneumonia or chronic obstructive airways diseases in older adults in Birmingham (Alabama), nor for IHD admissions in Detroit. Recent studies confirmed that a

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**Table 1** Air pollution in Westminster

<table>
<thead>
<tr>
<th>UK National Air Quality Strategy targets used in this study</th>
<th>UK target</th>
<th>Baseline levels in Westminster</th>
</tr>
</thead>
<tbody>
<tr>
<td>1: Annual mean PM$_{10}$ 30.8 μg/m$^3$[^{TEOM}]</td>
<td>Annual mean PM$_{10}$</td>
<td>36 μg/m$^3$ in 1996</td>
</tr>
<tr>
<td>(40 μg/m$^3$[^{gravimetric}])</td>
<td></td>
<td>38 μg/m$^3$ in 1997</td>
</tr>
<tr>
<td>2: Annual mean PM$_{10}$ 15.4 μg/m$^3$[^{TEOM}]</td>
<td>revised UK target for</td>
<td>24 hr mean PM$_{10}$ of 38.5 μg/m$^3$[^{TEOM}]</td>
</tr>
<tr>
<td>(20 μg/m$^3$[^{gravimetric}]), &lt; 35 exceedences</td>
<td>31.12.2004</td>
<td>(50 μg/m$^3$[^{gravimetric}]), exceeded:</td>
</tr>
<tr>
<td>3: 24 hour mean PM$_{10}$ 38.5 μg/m$^3$[^{TEOM}]</td>
<td>proposed UK target for</td>
<td>114 times in 1996</td>
</tr>
<tr>
<td>(50 μg/m$^3$[^{gravimetric}]), &lt; 7 exceedences</td>
<td>31.12.2009</td>
<td>130 times in 1997</td>
</tr>
<tr>
<td>4: 24 hour mean PM$_{10}$ 38.5 μg/m$^3$[^{TEOM}]</td>
<td>target for 31.12.2005</td>
<td>67 times in 1998</td>
</tr>
<tr>
<td>(50 μg/m$^3$[^{gravimetric}]), &lt;1 exceedences</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

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**Table 2** Strength of evidence for causal associations for particulates on health outcomes*

<table>
<thead>
<tr>
<th>Mortality</th>
<th>Moderate evidence</th>
<th>Strong evidence</th>
</tr>
</thead>
</table>
| All non-traumatic deaths—acute effects* | All respiratory disease. 15 | Emergency hospital admissions for all circulatory and all respiratory diseases and ischaemic heart disease at all ages. Emergency hospital admissions in people aged ≥65 for chronic obstructive pulmonary disease plus asthma and for pneumonia. Asthma admissions in children and younger adults. Accident and emergency attendances for asthma in children and younger adults.|**

*With example of key papers or reports.
linear model performed best for total and cardiorespiratory mortality; any threshold is \( \approx 15 \, \text{µg/m}^3 \) or \(< 5 \, \text{µg/m}^3 \) PM10. We therefore assumed there is no threshold for the effects of particulates, in line with World Health Organisation conclusions.

**Exposure-response coefficient**

We present models for those outcomes where the evidence for a causal relation with particulates is at least moderately strong in our view or where reports for government or international bodies have so concluded.

For each acute outcome-pollutant pair for which there was strong or moderate evidence of a causal relation, we chose the effect estimates from time-series studies that used untransformed pollution levels that were most relevant to current pollution in central London:

- results from the most recent time-series study in London for that pollutant and health outcome;
- results of pooled analyses from APHEA studies (short term effects of air pollution on health: European approach) in Western Europe, also a primary study, as the sources and levels of pollution were similar to London and the power of the pooled study was substantially greater than for most of the single studies available. Recently, results from the APHEA-2 studies, covering a population of more than 43 million people in 29 European cities have been published.
- estimates based on recent meta-analyses not subject to publication bias (the combined NMMAPS (national mortality and morbidity air pollution study)—APHEA meta-analysis) or conducted for the World Health Organisation, including those that were the basis of the 1998 COMEAP report on the burden of disease in the UK attributable to pollution.

Meta-analyses were not used as the sole effect estimate, nor did we conduct our own, because:

- many studies of ostensibly the same end points use different case definitions;
- many studies use different age groups;
- studies used different metrics of pollutants;
- each meta-analysis should contain only results from unipollutant models or with the same co-pollutant; and
- studies using single and cumulative lags should be analysed separately.

Although some authors have converted other measures of particulate matter into “PM10-equivalents” for their reviews or meta-analyses, we did not do this because of their well documented variability.

Where they exist, effect estimates that allow for the effects of other relevant pollutants would have been used, to permit addition of effects without double-counting. However, these were seldom available. Most studies have also found greater effects of pollution using cumulative rather than single day lag. Ideally, effect estimates that incorporate both co-pollutants and cumulative exposure would have been used but in practice, these are seldom available. Often, neither are available.

Although the three sources of effect estimates were chosen a priori, in practice, they were not available for all outcomes. Table 3 gives the range of effects estimates considered for acute effects of exposure to particulates, using the criteria above.

We also estimated the effect on all cause mortality of chronic PM10 exposure, although the paucity of good quality cohort studies and the different metrics used limited the effect estimates that could be selected. Most studies used fine particulates (PM2.5) rather than PM10 as their metric: where this was the case, we applied this effect estimate to the PM10 levels although this would be an overestimate because some of the PM10 is PM2.5-10.

We used:

- the observed effect in those with more than high school education in the reanalysis of the six cities study, to eliminate as far as possible many of the potential socio-economic confounders; 0.3% per 1 µg/m3 PM2.5;
- the all cause effect estimate from the 16 year follow up of the American Cancer Society cohort; 0.4% per 1 µg/m3 PM2.5;
- the variance weighted average of the 1995 ACS follow up and the six cities study calculated by Künzli et al in their three country study for the World Health Organisation; 0.43% per 1 µg/m3 PM10; and
- the effect estimate deemed “most likely” by COMEAP in their 2001 statement; 0.1% per 1 µg/m3 PM2.5.

We also estimated the effects on life expectancy of long term exposure to reduced particulate levels. The best estimate is that lifelong exposure to 10 µg/m3 PM10 shortens average life expectancy by 18.6 weeks per person across the whole population (13 days per 1 µg/m3 PM10). This assumes that the population is exposed to that pollution from birth until death or 105 years of age and that there is no delay before effects are seen. We applied the effects to the population now alive, with the full range of ages present, and made the assumption that the effects are spread evenly over the years of exposure. We assumed a life expectancy of 75 years.

Calculating the effect of a reduction in pollution

Effect estimates were calculated for a rise of 1 µg m\(^{-3}\) 24 h PM10 assuming a linear relation with no threshold. The 95% confidence intervals have been calculated where it was possible to do so.

We based our calculations for the effects of short term changes in pollution on the following formula:

\[
\Delta E = \beta \times \Delta C \times E
\]

where \( \Delta E = \text{change in number of outcome events} \), \( \beta = \text{exposure-response coefficient} \), \( \Delta C = \text{change in ambient concentration} \), and \( E = \text{baseline number of outcome events} \).

(This assumes a log-linear exposure-response function with no threshold, and is arithmetically equivalent to the formula given by Ostro and Chestnut.)

Change in ambient concentration

Air pollution measurements

We used data from three years (1996–1998) for the baseline for air pollutant levels because meteorological conditions can vary considerably from year to year. In particular, these change the peak levels and the number of days on which higher levels occur, even if emissions remain constant.

Continuous measurements of PM10 have been made in Westminster by a roadside tapered element oscillating...
<table>
<thead>
<tr>
<th>Diagnostic category</th>
<th>ICD-9 codes</th>
<th>ICD-10 codes</th>
<th>Age group</th>
<th>Annual mean number 1996–98</th>
<th>Effect estimates (percentage change in mortality per 1 μg/m³ PM&lt;sub&gt;10&lt;/sub&gt;)</th>
<th>Meta-analyses</th>
</tr>
</thead>
<tbody>
<tr>
<td>All non-traumatic deaths</td>
<td>&lt;800</td>
<td>A*-R*</td>
<td>All</td>
<td>1363</td>
<td>0.03% (−0.02 to +0.07)&lt;sup&gt;77&lt;/sup&gt;</td>
<td>0.06% (0.04–0.08) (lag 0–1)&lt;sup&gt;77&lt;/sup&gt; 0.16% (0.043–0.097) (distributed lag 0–40)&lt;sup&gt;14&lt;/sup&gt; 0.08% (distributed lag 0–40 with NO&lt;sub&gt;2&lt;/sub&gt;)&lt;sup&gt;14&lt;/sup&gt;</td>
</tr>
<tr>
<td>All respiratory deaths</td>
<td>460–519</td>
<td>J*</td>
<td>All</td>
<td>236</td>
<td>0.11% (0.04–0.20)&lt;sup&gt;16&lt;/sup&gt; +0.04% with NO&lt;sub&gt;2&lt;/sub&gt;&lt;sup&gt;16&lt;/sup&gt;</td>
<td>0.05% (−0.00 to +0.10)&lt;sup&gt;14&lt;/sup&gt; (WHO) 0.08% (0.03–0.12)&lt;sup&gt;10&lt;/sup&gt;</td>
</tr>
<tr>
<td>COPD and asthma deaths</td>
<td>490–496</td>
<td>J40–J47*</td>
<td>≥65</td>
<td>59</td>
<td>0.13% (0.03–0.25)&lt;sup&gt;14&lt;/sup&gt;</td>
<td>0.08% (0.01–0.16)&lt;sup&gt;10&lt;/sup&gt;</td>
</tr>
<tr>
<td>Circulatory admissions</td>
<td>390–459</td>
<td>I*</td>
<td>0–64</td>
<td>478</td>
<td>0.11% (0.04–0.18)&lt;sup&gt;16&lt;/sup&gt;</td>
<td>0.05% (−0.00 to +0.10)&lt;sup&gt;14&lt;/sup&gt; (WHO) 0.08% (0.03–0.12)&lt;sup&gt;10&lt;/sup&gt;</td>
</tr>
<tr>
<td>IHD admissions</td>
<td>410–414</td>
<td>I20-I25</td>
<td>0–4</td>
<td>204</td>
<td>0.13% (0.03–0.25)&lt;sup&gt;14&lt;/sup&gt;</td>
<td>0.08% (0.01–0.16)&lt;sup&gt;10&lt;/sup&gt;</td>
</tr>
<tr>
<td>Respiratory admissions</td>
<td>460–519</td>
<td>J*</td>
<td>≥65</td>
<td>317</td>
<td>0.08% (0.03–0.16)&lt;sup&gt;10&lt;/sup&gt;</td>
<td>0.04% (−0.03 to +0.12)&lt;sup&gt;10&lt;/sup&gt;</td>
</tr>
<tr>
<td>COPD and asthma admissions</td>
<td>490–496</td>
<td>J40–J47</td>
<td>≥65</td>
<td>238</td>
<td>0.03 (−0.08 to +0.15)&lt;sup&gt;16&lt;/sup&gt;</td>
<td>0.1% (0.04–0.15)&lt;sup&gt;16&lt;/sup&gt;</td>
</tr>
<tr>
<td>Lower respiratory infection admissions</td>
<td>466, 480–486</td>
<td>J12-J22</td>
<td>≥65</td>
<td>260</td>
<td>0.13% (0.02–0.25)&lt;sup&gt;16&lt;/sup&gt;</td>
<td>0.08% (−0.06 to +0.21) with NO&lt;sub&gt;2&lt;/sub&gt;</td>
</tr>
<tr>
<td>Asthma admissions</td>
<td>493</td>
<td>J45-J46</td>
<td>0–14</td>
<td>87</td>
<td>0.06% (−0.08 to +0.2)&lt;sup&gt;16&lt;/sup&gt;</td>
<td>0.12% (0.02–0.23)&lt;sup&gt;16&lt;/sup&gt;</td>
</tr>
<tr>
<td>Lower respiratory admissions</td>
<td>416, 480–486</td>
<td>J12-J22</td>
<td>≥65</td>
<td>92</td>
<td>0.14% (−0.01 to +0.3)&lt;sup&gt;16&lt;/sup&gt;</td>
<td>0.1% (0.08 to +0.1) with NO&lt;sub&gt;2&lt;/sub&gt; 0.11% (0.03–0.18)&lt;sup&gt;16&lt;/sup&gt;</td>
</tr>
<tr>
<td>Asthma attendances at A&amp;E</td>
<td>self reported</td>
<td>0–14</td>
<td>†</td>
<td>0.30% (0.09–0.50) cumulative&lt;sup&gt;35&lt;/sup&gt;</td>
<td>0.15% (−0.02 to +0.34) with NO&lt;sub&gt;2&lt;/sub&gt;&lt;sup&gt;35&lt;/sup&gt;</td>
<td>0.24% (0.07–0.42) cumulative&lt;sup&gt;35&lt;/sup&gt;</td>
</tr>
<tr>
<td>Asthma attendances at A&amp;E</td>
<td>self reported</td>
<td>15–64</td>
<td>†</td>
<td>0.30% (0.09–0.50) cumulative&lt;sup&gt;35&lt;/sup&gt;</td>
<td>0.15% (−0.02 to +0.34) with NO&lt;sub&gt;2&lt;/sub&gt;&lt;sup&gt;35&lt;/sup&gt;</td>
<td>0.24% (0.07–0.42) cumulative&lt;sup&gt;35&lt;/sup&gt;</td>
</tr>
</tbody>
</table>

†Events in Westminster residents; †data not available so calculations conducted per 100 visits at baseline. COPD, chronic obstructive pulmonary disease.
Figure 2  Three theoretical models for patterns of air pollution reduction. (A) 24 h PM$_{10}$ levels in Westminster, January 1996. (B) 24 h PM$_{10}$ levels in Westminster, January 1996 and effect of model B. (C) 24 h PM$_{10}$ levels in Westminster, January 1996 and effect of model A. (D) 24 h PM$_{10}$ levels in Westminster, January 1996 and effect of model C.

Table 4  Effects on premature deaths delayed by reducing PM$_{10}$ in Westminster from 1996–1998 roadside levels to UK standards

<table>
<thead>
<tr>
<th>Cause of death</th>
<th>Age</th>
<th>Source</th>
<th>Reductions to achieve PM$_{10}$ annual mean objective</th>
<th>Reductions to achieve 24 h PM$_{10}$ target for 31.12.2004</th>
<th>Reductions to achieve 24 h PM$_{10}$ target for 31.12.2009</th>
<th>Effect of reducing pollution to 0 µg/m$^3$</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td>40 µg/m$^3$ (gravimetric)</td>
<td>20 µg/m$^3$ (gravimetric)</td>
<td>50 µg/m$^3$ (gravimetric) with 35 permitted exceedences</td>
<td>50 µg/m$^3$ (gravimetric) with 7 permitted exceedences</td>
</tr>
<tr>
<td>Total non-traumatic</td>
<td>All</td>
<td>London</td>
<td>1.7</td>
<td>8.0</td>
<td>1.9</td>
<td>3.1</td>
</tr>
<tr>
<td></td>
<td></td>
<td>APHEA-2 (lag 0-1)</td>
<td>3.3</td>
<td>16</td>
<td>3.9</td>
<td>6.3</td>
</tr>
<tr>
<td></td>
<td></td>
<td>APHEA-2 (distributed lag 0-40)</td>
<td>8.9</td>
<td>43</td>
<td>10</td>
<td>17</td>
</tr>
<tr>
<td></td>
<td></td>
<td>APHEA-2 (distributed lag 0-40, with NO$_2$)</td>
<td>4.3</td>
<td>20</td>
<td>5.0</td>
<td>8.0</td>
</tr>
<tr>
<td></td>
<td></td>
<td>NMMAFS† (original)</td>
<td>2.3</td>
<td>11</td>
<td>2.7</td>
<td>4.3</td>
</tr>
<tr>
<td></td>
<td></td>
<td>NMMAFS (revised)</td>
<td>1.2</td>
<td>5.6</td>
<td>1.4</td>
<td>2.2</td>
</tr>
<tr>
<td></td>
<td></td>
<td>COMEAP</td>
<td>4.1</td>
<td>19.6</td>
<td>4.8</td>
<td>7.7</td>
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<td></td>
<td></td>
<td>Stieb et al</td>
<td>3.6</td>
<td>17</td>
<td>4.1</td>
<td>6.7</td>
</tr>
<tr>
<td></td>
<td></td>
<td>NMMAFS (original) pooled with APHEA-2</td>
<td>2.8</td>
<td>13</td>
<td>3.2</td>
<td>5.2</td>
</tr>
<tr>
<td>Respiratory</td>
<td>All</td>
<td>London</td>
<td>0.4</td>
<td>1.8</td>
<td>0.4</td>
<td>0.7</td>
</tr>
<tr>
<td>COPD and asthma</td>
<td>≥65</td>
<td>London</td>
<td>0.6</td>
<td>2.9</td>
<td>0.7</td>
<td>1.1</td>
</tr>
<tr>
<td>Total non-traumatic (chronic exposure)</td>
<td>All</td>
<td>16 yr ACS*</td>
<td>22</td>
<td>106</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

*Using effect estimate for PM$_{2.5}$; APHEA, short term effects of air pollution on health—a European approach; NMMAFS, national morbidity and mortality from air pollution study; COMEAP, committee on the medical effects of air pollution; COPD, chronic obstructive pulmonary disease; ACS, American Cancer study cohort; HEI, Health Effects Institute.
Modelling the declines in air pollution

Irrespective of the policy instruments proposed or the feasibility of achieving the targets, there are three general ways in which the hourly or daily pollution profile could change (fig 2).

### Table 5
Effects on admissions and consultations prevented or delayed by reducing PM$_{10}$ in Westminster from 1996–1998 roadside levels to UK standards

<table>
<thead>
<tr>
<th>Emergency hospital admissions</th>
<th>Age</th>
<th>Source</th>
<th>Reductions to achieve PM$_{10}$ annual mean objective</th>
<th>Reductions to achieve 24 h-PM$_{10}$ target for 31.12.2004</th>
<th>Reductions to achieve 24 h-PM$_{10}$ target for 31.12.2009</th>
<th>Effect of reducing pollution to 0 g/m$^3$</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td>40 µg/m$^3$</td>
<td>20 µg/m$^3$</td>
<td>50 µg/m$^3$ with 35 permitted exceedences</td>
<td>50 µg/m$^3$ with 7 permitted exceedences</td>
</tr>
<tr>
<td>Respiratory</td>
<td>All</td>
<td>London</td>
<td>4.3 20</td>
<td>5.0 20</td>
<td>8.0</td>
<td>4.5</td>
</tr>
<tr>
<td></td>
<td>COMEAP</td>
<td>4.3 20</td>
<td>5.0</td>
<td>8.0</td>
<td>4.5</td>
<td>4.9</td>
</tr>
<tr>
<td></td>
<td>APHEA-2</td>
<td>4.0 20</td>
<td>5.0</td>
<td>8.0</td>
<td>4.5</td>
<td>4.9</td>
</tr>
<tr>
<td></td>
<td>APHEA-2 (+NO$_2$)</td>
<td>0.0 20</td>
<td>0.0</td>
<td>0.1</td>
<td>0.3</td>
<td>0.3</td>
</tr>
<tr>
<td></td>
<td>15-64</td>
<td>London</td>
<td>0.5 2.5</td>
<td>0.6</td>
<td>1.0</td>
<td>4.5</td>
</tr>
<tr>
<td></td>
<td>APHEA-2</td>
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*Figures for accident and emergency department attendances are given as a percentage of the baseline attendance rate. COPD, chronic obstructive pulmonary disease; LRTI, lower respiratory tract infection; IHD, ischaemic heart disease.

- In Model A, all peaks above the standard (excluding permitted exceedences) have been eliminated. Levels on other days remain unchanged.
- In Model B assumes that the mean percentage decline in pollutant on the peak days, based on model A, also occurs on the remaining days.
- Model C uses the percentage reduction required to eliminate the highest peak (excluding permitted exceedences) to reduce each day’s pollution.

Models A and C provide lower and upper bounds, respectively; B models an intermediate condition. (Model A is the pattern that might be obtained by closing a city to motor vehicles when high pollutant levels are forecast; model C represents what might be seen in the long term, for example because of traffic reduction plus reduced trans-boundary pollution.) The target is not necessarily achieved by model B, as applying the average decline over all the exceedence days results in some of those days remaining above the standard but this is probably a more realistic estimate of the magnitude, if not the distribution, of the decline in pollutant. The simplest change is the reduction required to reach the target for the annual mean (model D in tables 2–4).

PM$_{10}$ levels to zero to compare with the Department of Health’s report on the burden of disease attributable to particulates.44

### Baseline number of outcome events

We used health outcome data for the same three calendar years (1996–1998) for the baseline; this reduces errors attributable to random fluctuations of small numbers of events. Mortality data came from the Office for National Statistics and emergency hospital admissions from Hospital...
Health and air quality management

Episode Statistics. No data were available on accident and emergency or general practice consultations locally, so results were calculated as a percentage of baseline age and disease specific consultations.

Creating the spreadsheets
Modelling the declines in air pollution
For each applicable combination of air quality target and pollution reduction model for each of the three baseline years, spreadsheets calculate the magnitude of change in ambient air pollution. Technical problems with the measuring devices can cause missing data for hours or days. Ignoring these underestimates the magnitude of the decline in pollution. We therefore imputed mean values to the missing data.

Health outcomes
For each air pollution reduction model, estimates of the number of lives no longer shortened or of emergency hospital admissions or consultations not occurring or not brought forward in a year were calculated automatically by multiplying the daily event number by the chosen effect estimate and the decline in pollutant summed over a year. In view of the uncertainties attached to both the effect estimates and the decline in pollution, we used a sensitivity analysis approach with three estimates for each where possible.

RESULTS
Short term effects
Attaining the 2004 24 h PM$_{10}$ objective at the Westminster roadside monitoring site, using the TEOM equivalent of 38.5 μg/m$^3$, results in one to four fewer deaths brought forward, based on the London studies, three to 21 based on the pooled APHEA-2 results from Western Europe, one to nine, based on Stieb et al’s 2002 meta-analysis, and one to 10 based on the World Health Organisation meta-analysis used by the Department of Health$^{44}$ (table 4). Reducing the permitted number of exceedences from 35 to the seven permitted in 2009 almost doubles the estimates. More respiratory and cardiovascular emergency admissions than deaths are delayed, because of the higher numbers of baseline events but the order of magnitude is similar (table 5). For example, about five respiratory and three to five cardiovascular admissions would be prevented or delayed by achieving the 2004 daily objective with an intermediate decline in pollution (model B). The fewer permitted exceedences for 2009 increases these to eight respiratory and six to eight cardiovascular admissions.

The annual mean PM$_{10}$ levels changed less than the number and extent of the exceedences of the 24 hour standard in the three baseline years. Achieving the annual mean objective for 2004 would delay between two and four deaths, using most effect estimates (table 4). More admissions are delayed or prevented by achieving these objectives (table 5). The 2009 objective for the annual mean requires a substantial reduction in PM$_{10}$, which would delay eight to 20 deaths, using the same effect estimates. About 20 respiratory and 14–20 circulatory admissions would be affected as well as around 5% of A&E attendances for asthma by achieving the lower annual mean target.

To check the models, we estimated the effects of reducing PM$_{10}$ pollution in Westminster to zero, to calculate the burden of disease attributable to particulates (tables 4 and 5). Deaths attributable to PM$_{10}$ ranged from 1.0% of the total, using the London exposure-effect coefficient, to 5.6%, using the distributed lag 0–40d APHEA-2 effect estimate (2.7% using the effect estimate that included nitrogen dioxide in the analysis). More than one in eleven deaths in Westminster from obstructive lung diseases in the 65–74 age group were attributable to PM$_{10}$. For emergency hospital admissions, the main groups affected by particulates are also adults with obstructive airways diseases: PM$_{10}$ caused almost 5% of emergency admissions in adults under 65 with asthma and up to 3.5% of admissions in those aged ≥65 with chronic obstructive pulmonary disease or asthma.

Long term effects
The effects on mortality of long term exposure to particulates may be an order of magnitude higher (table 5). Results using the 16 year follow up from the American Cancer Study are similar to those using Künzli and colleagues’ estimate for the long term effect of PM$_{10}$ on mortality, and slightly higher than using Krewski’s reanalysis of the six cities: about 20 deaths are delayed by reaching the 2004 annual target (40 μg/m$^3$ [gravimetric]) and a hundred deaths by reducing annual mean PM$_{10}$ to 20 μg/m$^3$ [gravimetric]. Calculations using the COMEAP “most likely” effect estimate are one quarter the magnitude, being closer to the acute effects of PM$_{10}$, particularly when a distributed lag is used for those acute effects.

The mean reduction in annual PM$_{10}$ for the roadside monitor is 4 μg/m$^3$ for the 2004 target and 19.5 μg/m$^3$ for the 2009 target. Ten year exposure to these lower levels would increase life expectancy by 7.3 and 34 days respectively, on average.$^{44}$

Choice of monitoring site
Particulate levels measured at the roadside in Westminster were higher than those at the background site in nearby Bloomsbury. The latter already met the 2004 annual mean target and almost met the 2004 24 hour target. To meet the 2009 targets, the decline required in pollution was one third to one half that required to achieve the same levels at the roadside in Westminster. The estimated health effects were lower by the same proportion (data not shown). Ten years exposure to the 2009 target rather than 1996–98 average levels at the Bloomsbury background site would increase average life expectancy by 19 days.$^{44}$

Policy implications

- Local authorities’ legal duty for air quality management was introduced to obtain health benefits; achievable health gain from reducing pollution is much less than the total health damage attributable to pollution.
- With no threshold of effect for particulates, more stringent air quality targets yield greater health benefits.
- Most air pollution in the UK comes from motor vehicles. Benefits to health and wellbeing of reducing traffic—reductions in injuries, greenhouse gas emissions, noise, and community severance from reduced car use accompanied by large positive health effects from increasing walking and cycling, are likely to be substantially greater than benefits from applying technical fixes to current travel patterns.
DISCUSSION

Attributable risk and achievable health gain

The comparison of our zero pollution findings with the Department of Health report\textsuperscript{44} demonstrates two main points. Our results using the same effect estimate suggests that 2.6% of deaths in Westminster were associated with baseline pollution levels. This compares well with the estimate for the Department of Health that 1.9% of urban deaths are attributable to PM\textsubscript{10} pollution,\textsuperscript{44} as PM\textsubscript{10} levels are higher in central London than almost anywhere else in the UK. Similarly, PM\textsubscript{10} in 1996–1998 accounted for 2.8% of all respiratory admissions, a little more than the report’s national estimate of 2.0%.

Secondly, the number of deaths or hospital admissions affected by the proposed reduction in air pollution is small compared with the total that is attributable to particulate air pollution. Attribution of a quantified burden of disease to a particular risk factor is quite different from quantifying the achievable health gain for a specified policy or target. This corresponds to a focus on the change in exposure rather than its level.\textsuperscript{56}

We have assumed that the observed effects of pollution are reversible by reduced exposure; this has been shown in a number of locations.\textsuperscript{57}

Choice of health outcome measures

Most HIA guidance advocates a holistic approach to health, including positive measures of good health.\textsuperscript{58} However, we were limited to using routine data, which are available only for disease measures, although they are not ideal outcome measures because few people are affected. As prevalence of mild disease is more common, more residents are likely to be affected by pollution exacerbating milder symptoms. While there is evidence that pollution affects lower respiratory symptoms and changes in lung function, we were not able to include these outcomes because of the lack of suitable routine data or of sufficient good quality information about general practitioner consultations.

The specific causes of mortality and morbidity were those identified by our systematic literature review as having at least moderate evidence for a causal relation with particulates. Few published meta-analyses have considered this first step,\textsuperscript{68} the 1998 COMEAP report being a notable exception.\textsuperscript{46} Omitting this crucial first step would have led us to report effects on mortality and hospital admissions of declines in levels of nitrogen dioxide. Our decision not to do so is in line with the most recent authoritative review.\textsuperscript{46}

Estimation of exposure-response functions

The choice of exposure-response coefficients depends on what is available. This is subject to change as new studies are published or methods are re-evaluated. The spreadsheets are designed so that these coefficients can be replaced easily.

International collaboration has resulted in studies using identical methods in a number of cities. In addition to examining local effects, pooled analyses are planned at the start; all study sites provide relevant information, regardless of their results. These calculations provide a much more robust estimate of the true effect.

Meta-analyses also combine results from a number of studies to give a more precise estimate. Expert reports have generally put most emphasis on meta-analyses but unlike pooled analyses, they are subject to publication bias so are more likely than pooled analyses to overestimate effects. Publication bias is the most probable explanation for the higher rate found in earlier American studies than the larger, more recent ones in the USA and in Europe.

The largest study is the American national morbidity, mortality, and air pollution study (NMMAPS), based on deaths in the 90 largest cities in the USA.\textsuperscript{58}\textsuperscript{59} Like many time series studies, it was analysed using generalised additive models (GAM). Investigators have found recently that when estimated regression coefficients are small and the default settings on the gam function of S-Plus software do not assure convergence of its iterative estimation procedure, the results overestimate the regression coefficients by 36%–42%.\textsuperscript{60} The effect estimate is more sensitive to the choice of method for adjustment for temporal effects than to the convergence criteria used.\textsuperscript{61} The greater the degree of concurvity (the non-parametric analogue of multicollinearity) the greater the overestimation of the regression coefficient.\textsuperscript{62} Concurvity also underestimates variance in the model and therefore the standard errors of linear predictors produced by the gam function are biased downwards, inflating type I errors.\textsuperscript{63} Stieb et al have shown a 23%–41% reduction in effect estimates from meta-analyses of primary PM\textsubscript{10} studies not analysed with GAM compared with meta-analyses of studies using GAM analyses.\textsuperscript{64}

Reanalysis of the NMMAPS data using generalised linear models (GLM) and natural cubic splines instead of GAM and smoothing splines to adjust for confounders halves the pooled estimate for non-traumatic mortality from +0.41% to +0.21% per 10 \(\mu\text{g/m}^3\) PM\textsubscript{10}; GAM used with stricter convergence criteria gives an estimate of 0.27%.\textsuperscript{65} The highest relative risk, for cardiorespiratory deaths, changed much less.\textsuperscript{66} Ramsay et al repeated Stieb et al’s meta-analysis of 31 studies of daily PM\textsubscript{10} and mortality.\textsuperscript{67} They found a slightly lower effect estimate (0.18% instead of 0.20% per 10 \(\mu\text{g/m}^3\) PM\textsubscript{10}) but the variance in the original meta-analysis was underestimated by 23%.\textsuperscript{68} These effect estimates are all within the same order of magnitude as those shown in table 3 (listed there per 1 \(\mu\text{g/m}^3\) PM\textsubscript{10}).

The effect of any reduction in the magnitude of the effect estimate once problems with GAM and with concurvity have been dealt with is to reduce the results by the same proportion. For example, the predicted number of deaths no longer delayed by reaching the annual mean target for 2009 therefore falls from 11 to 3.6 when the original is replaced with the revised estimate from NMMAPS (table 4). Corrected GAM analyses do not, however, affect the results from the pooled APHEA studies.\textsuperscript{69} This is because multisite time series are analysed using hierarchical models that estimate uncertainty in the pooled estimate using total variance.\textsuperscript{70} The WHO has reported that the unpublished meta-analysis from St George’s, London, of 26 studies that have not used GAM in analysis resulted in an effect estimate of 0.4% per 10 \(\mu\text{g/m}^3\) PM\textsubscript{10}, similar to the uncorrected NMMAPS result.\textsuperscript{44} The effect estimate using a 40 day distributed lag suggests that effect estimates from analyses using a few days’ lag are probably under-estimates.\textsuperscript{41}

Most studies of long term effects of particulates have used fine particulates not PM\textsubscript{10} as the metric; the WHO have recently recommended that a standard for PM\textsubscript{2.5} be established because of the increasing evidence that this fraction contains the principal harmful particles.\textsuperscript{71} Our models used PM\textsubscript{10} not only because that is the particulate measure used for the objectives in the current and proposed UK National Air Quality Strategy but also because PM\textsubscript{10} is now measured routinely but PM\textsubscript{2.5} is not. The results for the long term effects of PM\textsubscript{10} using PM\textsubscript{2.5} is an overestimate because some of the PM\textsubscript{10} is PM\textsubscript{2.5–10}; these coarser particles may also affect health\textsuperscript{72} but the evidence for that is less consistent.\textsuperscript{71}

Another contentious issue is whether effect estimates from uni-pollutant or multi-pollutant models should be used. It is advisable to use results from multi-pollutant models when examining the effect of more than one pollutant. As our
initial step indicated that the apparent effects of nitrogen dioxide are probably a proxy for effects of other traffic related pollutants and pollutants other than particulates and nitrogen dioxide are not of concern in central London, it is debatable whether it is appropriate to make such adjustments. In practice, few studies have reported such effect estimates. We have given results from the APHEA-2 distributed lag study for analyses including and excluding nitrogen dioxide.

**Sensitivity analyses**
Confidence intervals allow for random error but not for systematic error. The three estimates for the decline in pollution for each target (models A, B, and C), excluding annual mean, and the use of a number of exposure-response coefficients constitute built in sensitivity analyses. Analyses using pollution reductions based on each baseline year separately can be used for further sensitivity analyses (data available on request). Fortuitously, meteorological conditions resulted in particularly frequent exceedences of the PM$_{2.5}$ daily objective in 1996 and 1997 and unusually few in 1998. We have also calculated the effects based on air pollution measurements from two different sites.

There has been much media attention given to the annual fluctuations in peak levels and number of exceedences of these two pollutants. These are reflected in differing falls in pollution required to reach the hourly or 24 hour objectives using the different years as baseline. Use of mean values does not distinguish between longer exposures to lower levels and shorter exposures to higher levels of pollution. Abbey and Burchette used a number of metrics, namely the frequency of exceedences and the concentration for four predetermined cut off values of TSP in the well designed California Seventh Day Adventists cohort study. Excess concentration >200 µg/m$^3$ was the metric that correlated best with incidence of bronchitis, asthma or COPD and with change in severity of asthma.

**Validity of using fixed monitor measurements of air pollution**
Ambient pollution is only a proxy for personal exposure and dose, so studies relating levels of air pollution at a fixed site monitor to health outcomes probably underestimate the effect of pollution. However, it is important to use the same definition of exposure as that used in the studies from which the exposure-response functions have been derived. There is no reason to believe that the relation between residents’ exposure and the fixed site monitor is substantially different in central London from the studies we used. Those most affected by pollution are those who are already ill, who are unlikely to be highly mobile, so their spatial relation to the fixed site monitor is likely to be more consistent than for healthy people.

Comparing PM$_{10}$ levels at the two sites, hourly values, which vary more than 24 hour mean levels, were higher at the roadside monitor in Marylebone Road than the background site at Bloomsbury in 1999 but for many months, the levels varied almost in parallel. Since then, peak levels have fallen. In 1996 and 1998, 24 h-PM$_{10}$ was >50 µg/m$^3$ at the Bloomsbury background site more often than at the Marylebone roadside site. Crossroads sites, such as the Marylebone Road roadside monitor, are midway between a severe canyon site and a background site, picking up fresh emissions from the nearest street in certain wind directions and benefiting from the ventilation of the canyon down the side roads in other wind directions. They are therefore not as different from background measurements as might be imagined.

Roadside levels generally represent less well most people’s exposure to pollution but PM$_{10}$ and PM$_{2.5}$ levels outside individuals’ homes tracked fixed site PM$_{10}$ levels. Westminster City Council’s Air Quality Management Plan aims to bring roadside levels down to meet the standards. Background levels in Westminster would therefore fall to levels below the targets. As there is no observed threshold for effects on health and the exposure-response relation is approximately linear in this range, the effect is determined by the magnitude of the reduction not the absolute level reached. Therefore these calculations are likely to be reasonable estimates of the effects on deaths and admissions brought forward.

**Magnitude of results**
It is important to note that the figures we have calculated for the acute effects of pollutants are unlikely to represent a decline in annual mortality as most if not all of the affected people would have died within the year. The results shown are the number of people whose death in that year would have been brought forward had the pollution not declined. Ostro and Chestnut argue that a proportion of these deaths would have been so premature that a change in annual mortality does occur but this is disputed. It had been assumed that most of the people affected by air pollution are probably dying a few days, weeks, or months before they would have died without exposure to the air pollution but Zeger et al showed that the theory that deaths due to acute pollution exposure are due only to “harvesting” of the very frail is inconsistent with their analysis of the data, which showed that Poisson regression time-series analyses have underestimated the exposure-effect coefficient. For pneumonia, myocardial infarction, and all cause mortality, exposure to ambient air pollution is probably shortening life to a greater extent than was previously thought. This adds weight to our use of the larger, distributed lag effect estimates from APHEA-2. Additionally, the long term effects of chronic exposure to lower levels of pollution are generally held to be considerably larger than the acute effects. However, even over a lifetime’s exposure, the effects of these reductions on life expectancy is small compared with the average 12 years of life lost by smokers killed by tobacco.

**Impacts on inequalities in health**
Those at greatest risk from air pollution are those with pre-existing cardiorespiratory disease. Both cardiovascular and respiratory disease are more common in less affluent and less educated groups in the UK. In Canada, the highest prevalence of several risk factors for respiratory disease was in the area with the highest ambient pollution levels. In London there is a linear relation between levels of PM$_{10}$ and deprivation; with the greatest predicted falls in the areas of greatest deprivation. Reducing air pollution will therefore decrease inequalities both because exposure will be reduced most in deprived areas and because those who will benefit most are those with worse health, the very young and older people.

The spreadsheets we have developed are designed to model the most severe impacts on residents’ health of reductions from current levels of PM$_{10}$ in a local authority area to the UK National Air Quality Strategy targets. They have been piloted by three local authorities and a health authority in west London and have been used by 35 other local authorities in the UK. The same method could be applied to other pollutants. Models A, B, and C represent patterns of decline in pollutant levels but do not correspond directly to policy options to achieve such declines. The next step will be to combine the approach used in this study with one based on
actual sources and the ways in which emissions can be reduced. This paper has considered only the direct health gains of air pollution reduction, such as would be obtained by reducing emissions using technical fixes without changing behaviour. The benefits to health and wellbeing of changing travel patterns, with reductions in injuries, greenhouse gas emissions, noise and community severance from reduced car use accompanied by the large positive health effects from increasing walking and cycling, have the potential to increase much further the health benefits of air quality management.

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No place for modesty

“A leader is best when people barely know he exists; not so good when people obey and acclaim him; worst when they despise him. But of a good leader who talks little when his work is done, his aim fulfilled, they will say, ‘We did it ourselves.’”

Clearly John Ashton’s aphorism mirrors Lao Tzu’s thoughts on leadership, and is thus hard to argue against. However, I think public health practitioners should indeed be “bothered... if we don’t get the credit for our own ideas.” It is precisely this self-effacing stance that has led to the current situation where public health is grossly undervalued and under-resourced. It is hard enough trying to promote a positive—the disease outbreak that didn’t happen—we only make it worse to promote a negative—the disease outbreak under-resourced. It is hard enough trying to standardise both axes, by assigning life-quality its real scale, but it remains to be understood: SQALYs are the percentage of quality into a standard scale, where 0 = 0% = life expected to be lived was actually lived and 1 = 100% = life expectancy.

We can do the same with the dimension of time, assigning 0 to birth and 1 to lifetime expectancy. Continuing the example above, the dimensionless 1 is assigned to 80y, the dimensionless 0.625 to 50y (30y/80y) and the dimensionless 0.375 to 30y (30y/80y). Thus, fully standardised QALYs = SQALYs = 0.67x0.625x0.375 = 0.33. Disregarding age differences in summary measures of population health. Ann Rev Pub Health 2002;23:115–34.

Previous trials of dissemination were identified

Morrison et al examine the effect on questionnaire response of feeding back research findings to participants. As authors of the systematic review cited in their paper we would like to point out that we had in fact identified eight such randomised trials. The broad strategy under which these trials had been classified in our review was “non-monetary incentives”. We are currently updating our systematic review and have recently extended the search of databases to the beginning of 2003 and have contacted the authors of potentially eligible trials. A further two trials12 have been identified and the total of 10 trials have also been classified under the new strategy called “offer of research findings”. These trials include a total of 13 642 participants. The results of these trials are pooled in a random effects meta analysis the odds ratio for response with research findings is 0.92 (95% CI 0.75 to 1.11). Despite omitting to refer to these previous trials, Morrison et al were justified in conducting their trial: few of such trials have been health related and none has examined the effect of this intervention when participants are being resurveyed. However, even with the inclusion of their new trial in our systematic review, uncertainty about the effect of dissemination of research findings on questionnaire response remains.

The update to our systematic review now includes a total of 372 trials of methods to influence response to postal questionnaires. Although many conclusions remain unchanged our updated review presents the definitive account of the evidence for which strategies may be used to improve response to postal questionnaires. The updated review will appear in the Cochrane Library later this year.

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Standardised QALYs and DALYs are more understandable, avoid misleading units of measurement, and permit comparisons

QALYs and DALYs combine years of life and quality of life in a single measure.13 In Arnesen and Nord’s words: “DALYs and QALYs are complementary concepts. QALYs are years of healthy life lived; DALYs are years of healthy life lost. Both approaches multiply the number of years (x axis) by the quality of those years (y axis). QALYs use “utility weights” of health states; DALYs use “disability weights” to reflect the burden of the same states. For example, if the utility of deafness is 0.67, the disability weight of deafness is 1−0.67 = 0.33. Disregarding age weighting and discounting, and assuming lifetime expectancy of 80 years, a deaf man living 50 years represents 0.67x50 = 33.5 QALYs gained and 0.33×50=1.0×(80−50) = 46.5 DALYs lost.2 We can see that 33.5+46.5 = 80—that is, QALYs+DALYs = lifetime expectancy.

If we would like to be more accurate, we had to put QALYS = 0.67x50 = 33.5y and DALYS = 0.33×50=1.0×(80−50) = 46.5y. This means that the unit of measurement of QALYs and DALYS is years (y). As y is the unit of measurement of lifetime, using the same unit for the product “lifetime×quality” is confusing.

Saying that quality is rated on a scale from 0 to 1, we, in fact, have implicitly transformed the real but unknown scale of quality into a standard scale, where 0 denotes no quality at all and 1 the 100% of quality expected (lifeexpectancy). Therefore, QALYs and DALYS, combining actual years (axis x) and dimensionless quality (axis y), are, in fact, semi-standardised measures.

We can do the same with the dimension of time, assigning 0 to birth and 1 to lifetime expectancy. Continuing the example above, the dimensionless 1 is assigned to 80y, the dimensionless 0.625 to 50y (30y/80y) and the dimensionless 0.375 to 30y (30y/80y). Thus, fully standardised QALYS = SQALYS = 0.67x0.625x0.375 = 0.33. Disregarding age differences in summary measures of population health. Ann Rev Pub Health 2002;23:115–34.

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References
Terrorism and public health: a balanced approach to strengthening systems and protecting the public


The preface states: “we believe this is the first book that addresses terrorism from a public health perspective that is both comprehensive and balanced” (page xi). To a large extent, the book fulfils this promise, offering an informative, up to date, and highly readable summary of a broad range of public health issues that interface with the problem of international terrorism.

Part I has an introductory chapter followed by seven chapters examining public health challenges emerging after the September 11th attacks. Four of these chapters summarise events in New York City, one covers the anthrax epidemic, one covers public health problems in war-strapped Afghanistan, and one chapter offers an erudite and much needed account of the prospects for educating, informing, and mobilising the public. Much of the material in part I is based on firsthand experience, and it is packed with information and insights that are unlikely to be found elsewhere. Part II covers conventional, biological, chemical, nuclear, and radiological terrorist weapons. Attention often focuses on arms control and its political underpinnings, but clinical aspects are also covered (though in too little detail to provide an important reference for clinicians). Part III addresses terrorism related “challenges and opportunities,” with chapters aimed at public health systems, epidemiology, therapeutic interventions, research, environmental protection, civil liberties, roots of terrorism, and the promotion of international law. The comprehensiveness of the text suffers slightly from the lack of attention to methods of decontamination, structure and function of Incidence Command Systems, and the coordination of disaster services under the Federal Emergency Management Agency.

The text is well “balanced” in the manner intended by the authors in so far as it nicely situates the need for terrorism prevention and response capabilities within the context of other, potentially competing public health needs, and it balances these needs against the imperative to avoid “inauspicious or hazardous responses to threats of future terrorism.” On the other hand, there is little balance between competing viewpoints on ethical or policy issues. The book is structured by liberal cosmopolitan ideology—including numerous attacks on the Bush administration—with no attempt to fairly represent the range of credible, diverging opinions about the nature of justice or the intricacies of international collaboration and arms control.

Dawning answers: how the HIV/AIDS epidemic has helped to strengthen public health


The HIV epidemic is still a great threat to public health, and the complexity of the infection regarding both biological and social aspects has challenged our skills to prevent its spread. The book presents a historical analysis to inform current policy development and to forecast the future, and describes some very important lessons learned during more than two decades with the HIV epidemic.

HIV has influenced the development and understanding of the use of multiple surveillance methods, integrated case based and behavioural surveillance, active collaboration between different public health stakeholders, and confidentiality and anonymity have become important issues.

Although sex may well be the most pleasurable human activity it is also very tabooed. The HIV risk reduction thus entails difficult behaviour changes, and the involvement of community members in this public health activity has become crucial. The adoption of “grey area” behaviours among at risk populations has led to the need for structural and individual level intervention. The HIV epidemic has shown the necessity of understanding surveillance data in their social context, for example, sex for drug. At the same time the “All or nothing” thinking opened to the principle of harm reduction.

The HIV epidemic has shown the importance of translating research results into active intervention and routine service delivery. HIV has had an impact on the organisation of prevention and care services and the public health planners are urged to consider the entire healthcare system, using all data available. Legal aspects and ethical issues, such as human rights, especially in relation to testing policy, named reporting and partner notification are very well discussed in the book.

The nine chapters are mainly dealing with the situation in the USA, however, the history in most western countries is similar and the book is absolutely worth reading for those interested in public health and in the HIV/AIDS epidemic and policy. The public health challenges from infection diseases never stop.

G Trotter

E Smith

do: 10.1136/jech.2003.010553corr1

An editorial error occurred in this paper by Dr Malinauskienė and others (2004;58:131–5). The affiliation of Dr R Malinauskas is Department of Psychology, Lithuanian Academy of Physical Education (this was omitted from the article).

do: 10.1136/jech.2003.009589corr1

An editorial error occurred in this paper by Drs Mindell and Joffe (2004;58:103–13). In table 3, the age group for the first row of IHD admissions should be 0–64 (not 0–4).

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