# Impact assessment of the mortality effects of longer-term exposure to air pollution: exploring cause-specific mortality and susceptibility 

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#### Abstract

In previous work, we have developed the use of actuarial life-table methods, using PC spreadsheets, for predicting the impacts of changes in pollution levels on longterm patterns of mortality. This approach is flexible, and allows the simulation of impacts that affect only some causes of death; to date, we have focused on effects on the group of non-malignant cardio-respiratory causes as well as on all-cause mortality.


Because recent work has suggested that risks of air pollution may be greater for cardiovascular than for respiratory causes, we present and compare new predictions for cardiovascular impacts. In addition, we have for the first time modelled differences in mortality risk in different notional strata of the population, to represent differences in individual frailty or susceptibility to pollution effects.

Previous results showed that the impact of $1.8 \%$ reduction in cardio-respiratory risk was of the order of $10 \%$ smaller than that of a $1 \%$ reduction in the all-cause effect, and our results show that a purely cardiovascular $2 \%$ reduction produces an impact almost $30 \%$ smaller than the all-cause.

Similar results were obtained when the population was stratified to simulate a frailty distribution; however, when the sensitivity to the effects of pollution was allowed to differ by stratum, with the sensitivity increasing with increasing frailty, then the differences across cause groups were much reduced.

These results are interpreted as being due to the different age distributions for different causes of death, and particularly to the tendency for cardiovascular deaths to be predominantly in older persons, which limits their cumulative effects on life expectancy.

The revised spreadsheets offer great flexibility for further predictions under complex assumptions.

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## 1 INTRODUCTION

Quantitative assessment of the health impacts of air pollution, and of the associated costs, has attained considerable importance in recent years. This is a discipline whose principal feature is prediction into the unknown (unknowable) future, based on present knowledge of past effects. To do this well in the context of air pollution requires, amongst other things:

- knowledge or assumptions regarding the distributions and characteristics of the population for whom predictions are constructed;
- reliable estimates of the strength of relationships between air pollution and health effects;
- knowledge or assumptions regarding future changes in pollution levels that would affect the target population;
- knowledge or assumptions regarding the patterns of impacts that the changes would produce;
- a consistent methodology for calculating and summarising the predicted effects of these impacts, given all the above assumptions (and, if possible, one that can be implemented easily).

To date, reliable information about strength of relationships for long-term health effects has been available only from a limited number of US studies. Extensive analyses of mortality data from these studies, compared across cities, have pointed to a relationship with levels of particulate air pollution, both for mortality generally and particularly for causes of death classified as cardiovascular or (perhaps) respiratory (Krewski et al, 2000). Coefficients from these analyses have been widely discussed, and have been used in risk assessments whose aim was to quantify total predicted impacts of reductions in particle pollution.

Some results for all-cause mortality were reported as part of a DoH-funded project, with a final report by Hurley et al (2000). The methodology for quantitative predictions described in that report was based on actuarial life-table methods, adapted for health impact assessment at the IOM. More recent developments at the IOM, also funded by DoH, included catering for cause-specific impacts, triggered by the epidemiological observation that, in the above studies, not all causes of death related to air pollution; and the implementation of the methodology for calculation through standard spreadsheet techniques (Miller and Armstrong, 2001; Miller and Hurley, 2003).

We describe here some extensions to previous work. The initial impact assessments for cause-specific impacts looked only at the cardio-respiratory group of causes, whereas published results (Krewski et al, 2000) suggest that cardiovascular causes show more evidence of a relationship with pollution than do respiratory causes. We have therefore created predictions for the impact of changes in the cardiovascular hazard rates alone. In addition, we have for the first time considered and modelled differences in mortality risk of separate strata within age- and sex-specific subcohorts.

## 2. OBJECTIVE

The work reported here had the following objectives:

1. to assess the sensitivity of impact predictions to assumptions on cause: specifically, to run predictions similar to those already carried out, but based on effects only on cardiovascular mortality hazard rates;
2. to examine the effects on predictions of simulating variation in individual risks (susceptibility) and in the strength of effects

## 3 METHODS

### 3.1 INPUT DATA

Mortality hazards for baseline scenarios were based on published age-, sex- and causespecific rates for England and Wales and for Scotland, for the year 1999. As before, these were summarised into six cause-groups, representing lung cancers, other cancers, cardiovascular causes, non-malignant respiratory causes, accidental causes, and other causes.

Figure 1 shows the age- and sex-specific average mortality hazard rates for cardiovascular causes and for non-malignant respiratory causes, for ages 30 and above. The last point on each curve is for ages $90+$. The plots for the Scottish data are shown in grey, and show somewhat greater variation, because of the smaller population size and numbers of deaths. The graph shows clearly the higher mortality rates in men, and that the gap is much greater for the cardiovascular causes. However, this gap decreases with increasing age. It is clear that there is higher Scottish mortality from cardiovascular causes, for both sexes. For respiratory causes, it appears that the mortality in Scotland is slightly higher under the age of about 70, but more similar to England and Wales at higher ages. Rates for the other causes are not shown.


Figure 2 Age-specific mortality hazard rates (1999) by sex and country for cardiovascular and non-malignant respiratory causes for ages 30 and ove

### 3.2 CARDIOVASCULAR MORTALITY

We have previously described in detail how life table methods can be used to make quantitative impact assessments, by comparing the future patterns of expected life-years within chosen combinations of age cohorts. A series of developments to the methodology, which uses simple arithmetic procedures, have been implemented using the standard PC spreadsheet package MS Excel. We have previously described how differential effects on specific causes of death may be incorporated (Miller and Hurley, 2003), and have obtained quantitative estimates from assumptions of change only in the cardio-respiratory group of causes (Miller and Armstrong, 2001). The set-up for these calculations included separate baseline mortality hazard rates for the two major cause-groups of cardiovascular and nonmalignant respiratory causes, and in the above work the hazards for both were subject to the same impact factor. It was therefore relatively straightforward to rerun these assessments for an impact affecting the cardiovascular rates only.

The calculations reported by Miller and Armstrong (2001) were designed to compare the effect of assuming an impact on all-cause mortality rates with that of an effect only on cardiorespiratory vascular mortality. The impact coefficients were derived from analyses of the American Cancer Society cohort study (Krewski et al, 2002), which had shown higher impact coefficients for cardio-vascular and cardio-respiratory causes than for all-cause mortality, suggesting strongly that these causes were more susceptible to pollution effects than the remainder of the principal cause groups. It was expected that differences in predicted impacts might arise if the age-specific pattern of rates differed by cause.

On the basis of published impact coefficients, Miller and Armstrong (2001) compared a $1 \%$ change in all-cause mortality with a $1.8 \%$ change in cardio-respiratory mortality. Details of analyses of cardiovascular mortality risks from the data of the American Cancer Society are in Table 20 of the HEI Special Report on the reanalysis, Part II (Krewski et al, 2000). These show that the excess risk of cardiovascular disease associated with the range of particulate exposures shown between the US cities involved was $36 \%$, while that for all causes was $18 \%$. Since these are in the ratio 2:1, we have compared the previous results with a $2 \%$ decrease in cardiovascular mortality rates. (Since the $36 \%$ difference across cities was associated with a range of $24.5 \mu \mathrm{~g} . \mathrm{m}^{-3}$ in $\mathrm{PM}_{2.5}$ concentrations, a $2 \%$ change in cardiovascular mortality - or a $1 \%$ change in all-cause mortality - corresponds to a reduction in $\mathrm{PM}_{2.5}$ of $1.4 \mu \mathrm{~g} . \mathrm{m}^{-3}$.)

### 3.3 VARIATION IN INDIVIDUAL RISKS

It is plausible (though very difficult to demonstrate) that any human age cohort comprises individuals with different, possible widely different, risks of mortality. In origin, those differences may be genetic, or induced by disease or accident, or by lifestyle; most are unidentified by present medical knowledge (else we would be better at predicting individual longevity). The end result, however, is that mortality hazards even for single-year age cohorts are almost certainly an average or composite over a wide but unquantified range of susceptibilities.

While we do not have any reliable markers of individual risk (or susceptibility or frailty), nevertheless we may, for the purposes of impact assessment, make and state our assumptions about what range of differences might underlie our averages. This could be expressed as some statistical distribution; given that it is to be used as input to an arithmetic procedure, it is more straightforward to imagine a stratification of the target population. It is then necessary to specify

- what proportions of the population lie in various strata, and
- how their mortality risks differ.

In consultation with the Project Officer, it was agreed that, for the first set of investigations, it was sufficient to construct a three-way stratification, to represent broad distinctions between a small sub-group of people who are at extreme risk (because they are very susceptible or already ill), a sub-group of those at somewhat increased risk, and the remainder of the population. Table 3.1 shows how differential baseline mortality hazards can be organised by creating factors to manipulate the hazard rates for particular causes, in this case cardiovascular.

This table has the structure needed to create differential factors, but the values it contains are just one of very many possible combinations. The values used split the population notionally into the three subgroups, assumed to differ in their baseline risk of cardiovascular mortality: susceptible, increased risk, and normal, in the percentage ratios 5:10:85. For all the causes of death except cardiovascular, the factors in the table are 1, implying that the subgroups all share the same hazard rate for each of these causes. For the cardiovascular, however, we have factors implying that the rates for the three sub-groups are in the ratios $10: 2: 1$. The factors are scaled so that their average weighted by the population percentage fractions is 1 (= $0.05 \times 6.4516+0.1 \times 1.2903+0.85 \times 0.6452$ ). Thus the groups combine to have the correct underlying baseline cardiovascular hazard rate. Both the proportions and the differential hazard factors are arbitrary, and can be varied to meet different population assumptions.

Table 3.1 Example of baseline population and hazard stratification factors

| Subgroup <br> type | Proportion <br> \% | Lung <br> Cancer | Differential baseline hazard factor by cause <br> Other <br> Cancer | Cardio- <br> vascular | Respiratory | Accidents | Other |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  |  |  |
| Susceptible | 5 | 1.0000 | 1.0000 | 6.4516 | 1.0000 | 1.0000 | 1.0000 |
| Increased risk | 10 | 1.0000 | 1.0000 | 1.2903 | 1.0000 | 1.0000 | 1.0000 |
| Normal | 85 | 1.0000 | 1.0000 | 0.6452 | 1.0000 | 1.0000 | 1.0000 |
|  |  |  |  |  |  |  |  |
| Total | 100 | 1.0000 | 1.0000 | 1.0000 | 1.0000 | 1.0000 | 1.0000 |

Table 3.2 Example of stratified hazard impact factors; impacts on cardiovascular causes only, constant across subgroups

| Subgroup <br> type | Proportion <br> \% | Lung <br> Cancer | Other <br> Cancer | Cardio- <br> vascular | Respiratory | Accidents | Other |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  |  |  |
| Susceptible | 5 | 1.0000 | 1.0000 | 0.9800 | 1.0000 | 1.0000 | 1.0000 |
| Increased risk | 10 | 1.0000 | 1.0000 | 0.9800 | 1.0000 | 1.0000 | 1.0000 |
| Normal | 85 | 1.0000 | 1.0000 | 0.9800 | 1.0000 | 1.0000 | 1.0000 |
|  |  |  |  |  |  |  |  |
| Total | 100 | 1.0000 | 1.0000 | 0.9800 | 1.0000 | 1.0000 | 1.0000 |

As in previous work, it is necessary to specify the impact factors that will produce altered hazard rates for comparison with the baseline.

With the stratification organised as above, the impact factors can differ across subgroups as well as cause group. An example is shown in Table 3.2. Thus, Table 3.1 represents a spectrum of baseline frailty in the population whereas Table 3.2 represents the change to be
applied to the baseline population e.g. as a result of air pollution reductions. Table 3.2 represents a $2 \%$ reduction in cardiovascular mortality rates in all three subgroups.

Table 3.3 shows how the factors can represent differential impacts by subgroup. This example has impacts on the cardio-respiratory groups of causes in the ratios 10:2:1, scaled to produce a weighted average of a $1.8 \%$ decrease in the hazards for these causes (rather than $2 \%$ for cardiovascular causes alone). In this example, a greater reduction in hazard (11.61\%) is shown for the susceptible group compared with the normal group (1.16\%). This is analogous to assuming that the strata differ not only in their baseline frailty but also in their sensitivity to air pollution. If the frailest group is also the most sensitive to air pollution the gains from a reduction in air pollution will be greatest in this group.

Table 3.3 Example of stratified hazard impact factors; impacts on cardio-respiratory cause groups, differing across subgroups

| Subgroup <br> type | Proportion <br> \% | Lung |  |  |  |  |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Cancer | Other <br> Cancer | Cardio- <br> vascular | Respiratory | Accidents | Other |  |  |
|  |  |  |  |  |  |  |  |
| Susceptible | 5 | 1.0000 | 1.0000 | 0.8839 | 0.8839 | 1.0000 | 1.0000 |
| Increased risk | 10 | 1.0000 | 1.0000 | 0.9768 | 0.9768 | 1.0000 | 1.0000 |
| Normal | 85 | 1.0000 | 1.0000 | 0.9884 | 0.9884 | 1.0000 | 1.0000 |
|  |  |  |  |  |  |  |  |
| Total | 100 | 1.0000 | 1.0000 | 0.9820 | 0.9820 | 1.0000 | 1.0000 |

The spreadsheets used for previous calculations were adapted for the stratified calculations, following some tidying up of their structure, and a triplicate set of spreadsheet files was set up to perform life-table calculations from the baseline hazard rates in the manner previously described. These took as inputs a spreadsheet of the 1999 age- and cause-specific rates for England and Wales and a separate sheet containing the stratification factors, from which were created stratum-specific sub-cohort numbers and mortality rates.

A second set of triplicate spreadsheet files was set up to perform life-table calculations for scenarios including impacts on the mortality rates. These took as input the stratified baseline hazards as above, plus a further sheet containing hazard impact factors.

The final spreadsheets were set up to recombine the results from the strata, and to compare the results from the impacted scenarios with those from the baseline. The final summary sheet was made self-documenting, by including a listing of the input tables of stratification and hazard impact factors. An example of the output is shown below, as Table 3.4.

As before, separate calculations were performed for males and females. For the present work, we have assumed the same stratification and impact factors for both sexes. As before, separate estimates have been made for England and Wales and for Scotland.

Table 3.4 Example of output spreadsheet summarising impacts for males and females, and recording input table of baseline stratification factors and stratified impact factors.


## 4 RESULTS

### 4.1 CARDIOVASCULAR EFFECTS

Table 4.1 summarises results of predictions on the effects of various impacts, for the population predicted to be alive at the start of the year 2005. We show the total size of these start populations broken down by sex and by country, and compare the life years gained by reducing the all-cause hazards for ages 30 and over by $1 \%$ from the start of 2005 onwards. We compare this with the effects of reducing cardio-respiratory hazards by $1.8 \%$, and cardiovascular hazards only by $2 \%$, over the same age range. As in previous work, the predicted gains from a $1 \%$ reduction in the all-cause hazards are similar in males and females, giving a total gain of 4.8 million life-years in England and Wales, and 0.46 million in Scotland. Scaled to the population sizes, the results are still similar across the sexes, and across countries.

Table 4.1 Results of simulations, comparing predictions of impact from 1\% reduction
in all-cause hazards with $1.8 \%$ reduction in cardio-respiratory hazards and $2 \%$ reduction on cardiovascular hazards alone. Reductions in hazard take place from start of 2005. Impacts summarised for populations alive at start 2005.

| Country | Impact | Value tabulated | Males | Females | Total |
| ---: | ---: | ---: | ---: | ---: | ---: |
|  |  |  |  |  |  |

The results from simulations with a $1.8 \%$ reduction in hazard rates for the cardio-respiratory group of causes show similar effects, but in all cases somewhat smaller total effects. The reduction in effect is similar across the sexes, but differs across countries; while the causespecific impact is $7.7 \%$ lower than the all-cause for England and Wales, the difference for Scotland is $11.6 \%$.

The impacts for a $2 \%$ reduction on cardiovascular mortality are even smaller. The total cause-specific impact was estimated at $29 \%$ lower than the all-cause effect in England and Wales, and $28 \%$ lower in Scotland.

Although the impact coefficients have been chosen to be comparable on the basis of the results from the cohort studies, it is clear that the predicted impacts differ considerably. This is precisely because the age distributions of the hazard rates for the different causes differ, and thus the impact of a proportional change will differ by cause. In particular, the cardiovascular and respiratory rates affect the older age groups. Thus, the impact of a change in a the hazards for, say, cardiovascular deaths will have a smaller impact on overall hazards at younger ages. Thus the shape of the survival curves for baseline and impacted cohorts will differ little at lower ages, and diverge significantly only at higher ages.

We have observed in earlier simulations that changes in rates in the young matter little, because the baseline rates are so low; and that changes in the old also have less impact than in the middle years of life, because the population sizes are smaller. The observations in the present simulations are consistent with this.

### 4.2 SIMULATING SUSCEPTIBILITY BY STRATIFICATION

Table 4.2 contains the results of predictions for a $1 \%$ reduction in all-cause hazards, along with $1.8 \%$ and $2 \%$ reductions in cardio-respiratory causes and in cardiovascular causes only respectively, for comparison with Table 4.1. The new predictions have been calculated for three separate strata within each age- and sex-specific subcohort, with numbers in the ratio 5:10:85. As detailed in Table 4.2a, these strata are assumed equal in their baseline hazards for all causes except cardiovascular, where the hazards are in the ratio 10:2:1.

It may be noticed that the predicted start populations in Table 4.2 differ slightly from those in Table 4.1. This is because for Table 4.2 the start populations are derived separately in each stratum by life-table calculations between 1999 and 2005 and then recombined. The small difference in start populations does not affect the interpretation of the observed differences.

Comparison of Table 4.2 with 4.1 shows that the predicted effects of a $1 \%$ reduction in all cause mortality were always slightly higher for the stratified simulations; but that the decreased impacts observed for the cause-specific hazard changes were even more pronounced in the stratified case. These findings applied to both countries and to both sexes.

Table 4.3 shows predictions for the same strata, but with the additional stratification of the hazard factors, as detailed in Table 4.3a. The all-cause impacts are slightly smaller than in Tables 4.1 and 4.2. However, there is now much less of a difference between the effects of an all-cause and cause-specific impacts. In England and Wales, the cardio-respiratory impact is almost the same as the all-cause, and in Scotland it is reduced by only about 3\%; in both countries the impact of the stratified cardiovascular effect is much greater than in the unstratified case, showing only a $15-17 \%$ reduction from the all-cause. For ease of comparison, the results from Tables 4.1 to 4.3 are shown together in Table 4.4

The results in Table 4.2 are consistent with what we know about impacts from earlier simulations representing simpler comparisons. The stratified population we have set up has a majority with rather lower than average cardiovascular mortality rates. They will therefore

Table 4.2 Results of simulations comparing predictions of impact from 1\% reduction in all-cause hazards and $1.8 \%$ and $2 \%$ reductions in cardio-respiratory hazards.
Predictions made separately for strata differing in baseline cardiovascular hazard, defined in table below.

| Country | Impact | Value tabulated | Males | Females | Total |
| :---: | ---: | ---: | ---: | ---: | ---: |
|  |  |  |  |  |  |

Table 4.2a Population and hazard stratification factors for Table 4.2

| Subgroup <br> type | Proportion <br> \% | Lung <br> Cancer | Differential baseline hazard factor by cause <br> Other <br> Cancer | Cardio- <br> vascular | Respiratory | Accidents | Other |
| :--- | :---: | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  |  |  |  |  |
| Susceptible | 5 | 1.0000 | 1.0000 | 6.4516 | 1.0000 | 1.0000 | 1.0000 |
| Increased risk | 10 | 1.0000 | 1.0000 | 1.2903 | 1.0000 | 1.0000 | 1.0000 |
| Normal | 85 | 1.0000 | 1.0000 | 0.6452 | 1.0000 | 1.0000 | 1.0000 |
| Total population |  | 1.0000 | 1.0000 | 1.0000 | 1.0000 | 1.0000 | 1.0000 |

experience a proportionally larger part of their mortality due to other causes with lower ages at death, and an all-cause impact here has greater effects on total YOLL than at later ages. The effects of a cardio-respiratory or cardio-specific impact remain lower, for the same reasons as before, but the difference is exaggerated by the stratification. In Table 4.3, the additional stratification of the impact factors restores reverses this reduction, to the extent that the effects of a cardio- or cardio-pulmonary impact are of the same order of magnitude as the equivalent all-cause. However, all of these variations are relatively small, and in fact all are of the same order of magnitude.

Table 4.3 Results of simulations, comparing predictions of impact from reduction in hazards differing across strata. Strata as for Table 4.2, hazard impact factors as in table below.

| Country | Impact | Value tabulated | Males | Females | Total |
| :--- | :--- | :--- | ---: | ---: | ---: |
|  |  | 2005 start population (m) | 26.46 | 26.98 | 53.44 |
|  |  | All causes | Life years gained (m) | 2.296 | 2.298 |
| England  |  | Life years gained per 100,000 | 8678 | 8519 | 8598 |
|  | Cardio- | Life years gained (m) | 2.350 | 2.268 | 4.618 |
|  | respiratory only | Life years gained per 100,000 | 8882 | 8407 | 8642 |
|  | Cardiovascular | Life years gained (m) | 2.000 | 1.830 | 3.829 |
|  | only | Life years gained per 100,000 | 7551 | 6785 | 7164 |
| Scotland |  | 2005 start population (m) | 2.502 | 2.631 | 5.134 |
|  |  | Cardio- | Life years gained (m) | 0.220 | 0.219 |
|  | respiratory only | Life years gained (m) | 0.439 |  |  |
|  | Cardiovascular | Life years gained per 100,000 | 8561 | 8051 | 8299 |
|  | only | Life years gained (m) | 0.191 | 0.178 | 0.369 |
|  |  | Life years gained per 100,000 | 7634 | 6792 | 7203 |

Table 4.3a Stratified impact factors for Table 4.3

| Subgroup <br> type | Proportion <br> \% | Lung <br> Cancer | Differential impact factor by cause <br> Other <br> Cancer | Cardio- <br> vascular | Respiratory | Accidents | Other |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| All causes |  |  |  |  |  |  |  |
| Susceptible | 5 | 0.9355 | 0.9355 | 0.9355 | 0.9355 | 0.9355 | 0.9355 |
| Increased risk | 10 | 0.9871 | 0.9871 | 0.9871 | 0.9871 | 0.9871 | 0.9871 |
| Normal | 85 | 0.9935 | 0.9935 | 0.9935 | 0.9935 | 0.9935 | 0.9935 |
| Total population |  | 0.9900 | 0.9900 | 0.9900 | 0.9900 | 0.9900 | 0.9900 |
| Cardio-respiratory only |  |  |  |  |  |  |  |
| Susceptible | 5 | 1.0000 | 1.0000 | 0.8839 | 0.8839 | 1.0000 | 1.0000 |
| Increased risk | 10 | 1.0000 | 1.0000 | 0.9768 | 0.9768 | 1.0000 | 1.0000 |
| Normal | 85 | 1.0000 | 1.0000 | 0.9884 | 0.9884 | 1.0000 | 1.0000 |
| Total population |  | 1.0000 | 1.0000 | 0.9820 | 0.9820 | 1.0000 | 1.0000 |
| Cardiovascular only |  |  |  |  |  |  |  |
| Susceptible | 5 | 1.0000 | 1.0000 | 0.8711 | 1.0000 | 1.0000 | 1.0000 |
| Increased risk | 10 | 1.0000 | 1.0000 | 0.9742 | 1.0000 | 1.0000 | 1.0000 |
| Normal | 85 | 1.0000 | 1.0000 | 0.9871 | 1.0000 | 1.0000 | 1.0000 |
| Total population |  | 1.0000 | 1.0000 | 0.9800 | 1.0000 | 1.0000 | 1.0000 |

Table 4.4 Summary of life years gained (m) across total population of England and Wales, estimated from different sets of assumptions and different impacts on mortality.

Data from Tables 4.1 to 4.3

| Assumptions | Mortality cause affected |  |  |
| :--- | ---: | ---: | ---: |
|  | All <br> Causes | Cardio- <br> respiratory | Cardio- <br> vascular |
| No stratification <br> (Table 4.1) | 4.784 | 4.415 | 3.39 |
| Stratified population <br> (Table 4.2) | 4.897 | 4.197 | 2.905 |
| Sensitivity variable by strata <br> (Table 4.3) | 4.595 | 4.618 | 3.829 |

## 5 ACKNOWLEDGMENTS

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