New Problems for an Old Design:

## Time-Series Analyses of Air Pollution and Health

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 <sup>2</sup> The Johns Hopkins Bloomberg School of Public Health Department of Biostatistics 615 N. Wolfe Street Baltimore, Maryland 21205 Since the early 1990s, numerous time-series studies have linked daily mortality counts to levels of particulate air pollution on the same or recent days (1;2). Studies of similar time-series design of morbidity indicators, hospitalization, and clinical status, for example, have provided complementary evidence for adverse effects of particulate air pollution on the public's health. The daily time-series studies of air pollution, together with findings of prospective cohort studies that indicate increased mortality associated with long-term exposure to air pollution, have motivated reassessment of air quality standards for particles in the United States and Europe.

The time-series studies of acute effects have largely been of similar design, involving analyses of data bases of daily counts of events, daily levels of particles and other air pollutants measured at central site monitors, and daily data on weather, a potential confounding factor. The analyses have typically controlled for weather, season, and other longer-term time-varying factors (e.g., trends of disease mortality) to assure that estimates of the effects of air pollutants, which may be associated with weather and season, are not confounded. Time-series studies estimate relative rates of mortality/morbidity, generally interpreted as percentage increase in mortality/morbidity per unit increase in the air pollutant levels. Regression models with non-linear functions of time and weather variables have been used for this purpose, including generalized additive models (GAM) with smoothing splines and generalized linear models (GLM) with natural cubic splines. Use of GAM became very popular in the mid-1990s with implementation using the S-Plus function *gam* (2). We used this software in extensive analyses of air pollution, mortality, and hospitalization in the National Morbidity, Mortality, and Air Pollution Study (NMMAPS) (3;4).

In this issue of *Epidemiology*, Ramsay and colleagues (2) point out that the S-Plus function *gam* uses a computational approximation which, in presence of large correlation between the non-linear functions included in the model (called concurvity), can underestimate the standard errors of the relative rates. We have recently identified and described another limitation of the S-Plus function *gam* (5). In an in-depth exploration of model sensitivity, we discovered that *gam* default convergence criteria (S-Plus version 3.4) were not sufficiently rigorous for these analyses; the result was an overestimation of the effect of particulate air pollution on mortality. In our initial exploration of the sensitivity of model findings to the details of model specification, we have found a complex interplay between the extent of smoothing of time-related confounding, the extent of concurvity, and the degree of bias in estimates (5).

Thus, studies using the *gam* function in S-Plus might have overestimated the magnitude of the risk to public health posed by air pollution, tending to provide risk coefficients that are biased upwards and estimated with overstated precision. In the NMMAPS analyses, for example, a pooled estimate based on the 90 largest U.S. cities was 0.41% increase in mortality per 10  $\mu$ g/m<sup>3</sup> increase in PM<sub>10</sub> (particulate matter less than 10  $\mu$  in aerodynamic diameter) with a posterior standard error of 0.05) with use of the standard *gam* convergence criteria; with use of substantially more strict convergence

criteria, the estimate dropped to 0.27 % per 10  $\mu$ g/m<sup>3</sup> increase in PM<sub>10</sub>, again with a posterior standard error of 0.05. For both the original and the revised analyses, there was strong evidence for an effect of air pollution on mortality; posterior probabilities for the PM<sub>10</sub> coefficient exceeding zero were essentially 1.0 for both analyses. Pooled estimates from multi-site time series are not affected by the underestimation of the standard errors in *gam*. Multi-site time series studies are analyzed by using hierarchical models that estimate the uncertainty in the pooled estimate by the sum of the within-city plus the between-city variance (total variance). Therefore, in hierarchical models, the underestimation of the within-city variance is balanced by the overestimation of the between-city variance, without effecting the total variance.

The community of air pollution researchers is now faced with the obligation of repeating analyses that have used the *gam* function and considering further methodologic issues, such as that described by Ramsay and colleagues (6). These methodologic issues are important when the air pollution effects are small and possibly confounded by varying processes, such as weather, which are correlated with pollution exposures. What are the alternative strategies for modeling daily time-series data? Of course, there is no "correct" model. We have compared GAMs to GLM with natural cubic splines for confounder adjustment (5). The pooled estimate obtained with GLM for the 90 NMMAPS cities (0.21% per 10  $\mu$ g/m<sup>3</sup> increase in PM<sub>10</sub>) is slightly lower than the estimate obtained with GAM and the updated convergence criteria. We caution against

selecting any particular model as "correct" and urge researchers to explore the sensitivity of findings to model selection.

These methodological issues in time-series analyses of air pollution data were identified as the U.S. Environmental Protection Agency was carrying out its process of evidence review for the National Ambient Air Quality Standard (NAAQS) for particulate matter. The process involves the compilation of all relevant evidence since the last review into a comprehensive document, the Criteria Document. In the most recent draft Criteria Document, the time-series studies, including NMMAPS, were covered extensively and considered as providing clear evidence of an adverse effect of particulate matter air pollution on human health (2). The Environmental Protection Agency is also using the effect estimates from the time-series studies in a quantitative risk assessment mandated by the Clean Air Act. The new analyses continue to provide strong evidence of an association between acute exposure to particles and mortality. However, the updated estimates of burden of disease and death due to acute exposure are smaller. It is important to remember that time series studies only quantify the effects of acute exposure and do not address the larger question of whether chronic exposure increases the risk of disease and death.

Many "lessons learned" might be listed based on the report by Ramsay and colleagues (6) and our recent findings (5). The difficulty of detecting the small signal of the effect of air pollution amidst the noise of the many other factors affecting mortality merits emphasis. To find this signal, we are analyzing large and complicated databases,

with models that inherently make assumptions. We are learning just how sensitive the model results are to these assumptions and finding that some of the tools that we have been using need to be improved for this application. Faster computers can now overcome software limitations easily. The S-Plus default convergence parameters have already been revised in the new S-Plus version, and substantially more stringent parameters can be used without much loss in computing time. In addition, revisions of GAM software implementations, allowing "exact" calculations of the standard errors, are underway. We have also learned again that a community of inquisitive researchers will continue to refine their work and replace less adequate with better approaches.

## Reference List

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