POSTSCRIPT

LETTERS

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Smoking verification and the risk of myocardial infarction

We read the paper by Godtfredsen et al with interest.1 The paper reported on the effect of smoking reduction on the incidence of myocardial infarction (MI) and found that although patients who stopped smoking had a decreased risk of MI, those who reportedly reduced their smoking did not. The conclusions drawn were that smoking reduction, rather than complete cessation, did not produce any benefit with regard to a reduction of risk of MI.

The major drawback to this study is that the information about smoking was totally reliant on self reported smoking habit. There is abundant evidence that patients who smoke, when questioned about a smoking habit, will under-report their cigarette consumption or deny smoking altogether. The more significant the effect smoking has, the greater the "social desirability bias", so increasing the likelihood of denial. To overcome this bias biochemical verification of smoking by measurement of nicotine metabolites, specifically cotinine, has become almost obligatory.

To improve the accuracy of information about smoking and to facilitate easier nicotine metabolite measurements we developed a six minute point of care test called SmokeScreen.2 The easy to use colorimetric urine test can provide qualitative, semi-quantitative, and quantitative measurements of nicotine intake. Using this test we undertook an audit of smoking habits of 154 new patients attending a large inner city hospital cardiology outpatient clinic, comparing the test identification of smoking with self completed questionnaire of current smoking habit. The results identified 112 (72.7%) patients as non-smokers, 30 (19.5%) as confessed smokers, and 12 (7.8%) as "smoking deceivers".

We followed this with another study of the same population (n = 85, 33 smokers and 52 never-smokers) to examine the interaction of smoking and risk factors associated with coronary artery disease, as assessed by a biochemical screen and a blood count. Interestingly, none of the parameters measured in the biochemical screen, such as cholesterol, HDL and triglycerides, urea and electrolytes, and liver function tests were associated with smoking habit or quantitative assessment of nicotine intake. Whereas white blood cell count was significantly higher in smokers (p = 0.002), in particular, neutrophils (p = 0.01) and eosinophils (p = 0.02). Lymphocytes, monocytes and basophils were higher but failed to reach significance. Quantitative assessment of nicotine intake of the smokers further revealed a positive correlation with white blood cell count (p = 0.0001), neutrophils (p = 0.001), eosinophils (p = 0.004) and lymphocytes (p = 0.02), with monocytes approaching significance (p = 0.7).

It would seem from this pilot study that smoking or the amount of tobacco consumed does not influence the biochemical risk factors for coronary artery disease, such as cholesterol and HDL. However, smoking does seem to increase many of the immune cells associated with both the formation and destabilisation of the atheromatous plaque. It would seem logical therefore that a reduction in nicotine intake would be accompanied by a reduced risk of MI, as supported by our quantitative findings. One reason for the poor association between smoking reduction and subsequent MI in the Godtfredsen et al study may be the inaccuracy of self report. We suggest that identification of smokers with the point of care test is a more valuable method of smoking assessment. Coupling this test with subsequent advice on smoking cessation could have a significant impact on reducing a major risk factor associated with coronary artery disease and decrease cardiovascular events and mortality.

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References

Authors’ reply
We appreciate the comments from Cope and Batterby on our paper reporting the association between smoking cessation and smoking reduction and subsequent risk of myocardial infarction. Specifically, they propose that the lack of a beneficial effect of reduced smoking—in contrast with smoking cessation—would be attributable to inaccuracy (under-reporting) of the self reported tobacco consumption. In addition, they raise the important question of which measurement method most accurately reflects tobacco exposure in the individual smoker.

We agree that nowadays almost every study of smoking habits apply one or more measurements of biochemical markers of smoking in addition to self report. It is also correct that in our paper the study participants are divided into the different smoking categories on basis of self reported smoking and changes in smoking. However, as mentioned in the discussion, measurements of expired carbon monoxide (CO) and serum cotinine were undertaken in one of the follow up examinations. We found increasing levels of CO (table 2) and cotinine (not shown) with increasing self reported tobacco consumption, indicating that underreporting of smoking alone cannot explain our results, but clearly misclassification cannot be ruled out in this observational study. Furthermore, a previous review and meta-analysis1 concluded that self reported smoking is an accurate measure of tobacco exposure in population based studies, whereas this is not the case in intervention and clinical studies. Our data were based on a sample of the general population; participants with known coronary heart disease before study entrance were excluded. In addition, information on smoking habits and changes in smoking were part of a large questionnaire initiated in the late 1970s and the 1980s, thus minimising the risk of “social desirability bias” in this study.

Cope and Batterby describe a pilot study using a urine cotinine test for measuring nicotine intake. There are various methods of validating tobacco intake including biochemical markers, and cotinine is one of the better because of its comparatively long half life and the possible linear relation with number of cigarettes smoked. However, cotinine is not very useful in smoking reduction studies as most of the participants in these trials are supplied with nicotine replacement medica-

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Fibrinogen, social position, and risk of heart disease

The report by Jousilahti and colleagues adds to growing evidence of a consistent association between serum inflammatory markers—particularly fibrinogen—and social position. The authors interpret their data as suggesting that the fibrinogen-social position link is not merely a reflection of the social patterning of prevalent disease, smoking, and obesity (all of which are associated with increased serum fibrinogen and lower social position) as a strong trend of increasing fibrinogen with decreasing social status survived statistical adjustment for these covariates. Fibrinogen, they conclude, is therefore a promising candidate for the “missing link” between social position and cardiovascular health.

The authors’ reasoning implicitly accepts that fibrinogen is a cause of coronary heart disease (CHD). However, this runs contrary to recently published evidence using the principle of “Mendelian randomisation” (the situation where a particular genetic polymorphism influences exposure level of a putative disease risk factor, and should in turn be related to increased risk of disease if the risk factor is indeed a cause). In fact we discussed this evidence in a recent commentary on psychosocial explanations of health inequalities. Plasma fibrinogen concentrations are related to a polymorphism in the β-fibrinogen gene, with presence of the “T” allele being associated with higher levels. Among controls of a recent large case-control study, fibrinogen increased by 0.12 g/l per T allele present. Comparing cases with controls, a 0.12 g/l higher fibrinogen was associated with a relative risk of CHD of 1.20 (95% CI 1.13 to 1.26). If increased fibrinogen actually causes heart disease then such a per allele relative risk of CHD should be seen. In fact the per-allele relative risk of CHD was 1.03 (0.96 to 1.10). People whose genotype would have subjected them to long-term raised plasma fibrinogen experienced no substantial increased risk of heart disease, suggesting that observed associations between fibrinogen and CHD risk are not causal. This finding is in keeping with evidence from randomised controlled trials that suggests that drugs lowering fibrinogen do not decrease the risk of CHD. Fibrinogen probably predicts cardiovascular events because of reverse causation (atherosclerosis is an inflammatory condition and raises circulating fibrinogen concentrations) and because of confounding—smoking, abstaining from alcohol, not exercising and being poor are all associated with raised fibrinogen and themselves increase the risk—or are markers for factors that increase the risk—of cardiovascular disease.

The data presented by Jousilahti and colleagues illustrate how easy it is to misattribute causality to associations in social epidemiology. Many factors are socially patterned and thus appear as possible candidates for a causal role in the processes that generate any disease outcome that is also socially patterned. Demonstrating apparent statistical independence of associations between such exposures and outcomes does little to infer their causal basis as it is often likely to reflect issues of residual confounding and measurement imprecision of correlated covariates. Strategies such as randomised control trials and Mendelian randomisation can help untangle these issues. Where feasible, they should be more widely adopted in epidemiology.

Air pollution and asthma in children

Lin and colleagues have published an interesting paper on asthma hospitalisation in children. They concluded that the study showed positive relations between gaseous pollutants and asthma hospitalisation in children and that the effects of certain specific gaseous pollutants were found to vary in boys and girls. There are several issues that should be discussed before these conclusions can be confidently accepted. The authors performed separate regression analyses for boys and girls, and compared the sexes by examining the odds ratios. Examination of their figure 1 suggests that the possibility that differences between the sexes might be chance fluctuations. It has been recommended that statistical tests for interaction, which directly examine the strength of evidence for the difference in treatment effect between subgroups, are the most useful approach for evaluating subgroup analyses. The most simple hypothesis is that there is no difference in susceptibility between boys and girls. It would thus be helpful to know if these apparent differences are statistically significant.

A consideration in performing hypothesis tests is the standard error of the coefficients. The authors have treated all hospitalisation events as independent. It is probable, however, that some children were admitted to hospital more than once during the course of the study. I have compiled a cohort of some 108 000 people from primary care and respiratory practices in the cities of Hamilton and Toronto, Ontario (Toronto was the setting for the study by Lin et al.). Hospitalisations for asthma, 1992–1999, were ascertained by linkage to the Provincial Hospital Discharge database. Ninety four children, aged 6–12, were hospitalised a total of 145 times, with a mean admission frequency of 1.5 times. Twenty two per cent of children were admitted more than once. Table 1 displays the distribution of numbers of asthma hospitalisations. The study of Lin et al. is thus, in a sense, a repeated measures longitudinal study. Failure to take account of the non-independence of events will lead to underestimation of the standard errors and the possibility of inappropriate rejection of the null hypotheses of no effect of pollutants or no difference between the sexes.

Table 1. Distribution of the numbers of hospitalisations for asthma, 1992–99, among 94 children aged 6–12, in a cohort from the cities of Hamilton and Toronto, Ontario

<table>
<thead>
<tr>
<th>Number of admissions</th>
<th>Number of children</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>73</td>
</tr>
<tr>
<td>3</td>
<td>10</td>
</tr>
<tr>
<td>4</td>
<td>5</td>
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<tr>
<td>6</td>
<td>1</td>
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<td>1</td>
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References


Authors’ reply

We appreciate Dr Finkelstein’s comments on our paper regarding sex differences in effects...
of air pollution on asthma hospitalisation and the possible impact that autocorrelations in the data would have on our risk estimates. In our study, the effects of certain gaseous pollutants on asthma hospitalisation were found to differ between boys and girls 6 to 12 years of age. Some of these differences were “statistically significant”, for example, the regression coefficient for six to seven day SO2 effect was significantly greater for girls than for boys (p < 0.001). Although individual results can be examined for statistical significance in this manner, we prefer to base conclusions on the broad risk patterns in the data that emerge after our analysis of a number of gaseous pollutants and exposure periods. Collectively, these results suggest a differential effect of gaseous air pollutants on asthma hospitalisation in girls as compared with boys. We agree that it is probable that some children would be admitted to hospital for asthma more than once during the study period, and consequently, some autocorrelation may exist in our admission series. Generalised estimation equations (GEE) can be used to address this issue if readmissions can be identified. Unfortunately, our data do not include personal identifiers needed to identify readmissions. However, the residuals of asthma hospitalisation count data did not display notable “intraclass” type correlation, and then it is not obvious that the repeated asthma hospital admissions have induced sizable additional variation. In addition, in a separate analysis using a different dataset from Vancouver in which asthma readmissions were identifiable, the results based on all admissions were similar to those based on first admission. This second analysis suggests that the effect of autocorrelation on our presented risk estimates within our asthma admission series is small.

![Table 1](image)

**Table 1** Effect of NHS smoking cessation services on the relative risk of smoking in people living in the most and least deprived quintiles using hypothetical smoking rates

<table>
<thead>
<tr>
<th>Deprivation quintile</th>
<th>Population*</th>
<th>Smoking rate, %</th>
<th>Number of smokers</th>
<th>Relative risk† (95% CI)</th>
<th>% Of population quitting*</th>
<th>Number of quitters</th>
<th>Number of smokers</th>
<th>Smoking rate, %</th>
<th>Relative risk (95% CI)*</th>
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</thead>
<tbody>
<tr>
<td>Least deprived</td>
<td>187064</td>
<td>25</td>
<td>467/66</td>
<td>1.00 (0.99 to 1.01)</td>
<td>0.05</td>
<td>85</td>
<td>466/61</td>
<td>24.95</td>
<td>0.99 (0.98 to 1.00)</td>
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<tr>
<td>Most deprived</td>
<td>336800</td>
<td>25</td>
<td>842/00</td>
<td>0.25</td>
<td>826</td>
<td>592/68</td>
<td>4.95</td>
<td>0.57 (1.56 to 1.58)</td>
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<tr>
<td>Least deprived</td>
<td>187064</td>
<td>5</td>
<td>593/55</td>
<td>1.58 (1.56 to 1.59)</td>
<td>0.25</td>
<td>826</td>
<td>167/574</td>
<td>47.75</td>
<td>1.57 (1.56 to 1.58)</td>
</tr>
<tr>
<td>Most deprived</td>
<td>336800</td>
<td>15</td>
<td>280/60</td>
<td>0.05</td>
<td>279/75</td>
<td>305/26</td>
<td>14.95</td>
<td>1.45 (2.56 to 2.62)</td>
<td></td>
</tr>
</tbody>
</table>

*Data from Lowey et al.; †sensitivity estimates including data from Lowey et al.; ‡data calculated by us; Relative risk of smoking in most deprived compared with least deprived quintile.

**Reference**


**Smoking cessation services may not reduce inequalities**

Given the oft quoted proposition that health promotion interventions frequently increase, rather than decrease, socioeconomic inequalities in health (SEIH), we were interested to read the article by Lowey et al.1 However, we believe that the title of this article is misleading, that the authors have not analysed the data to its fullest potential, and that a number of the conclusions drawn are unjustified. The authors do not present any data on the magnitude of inequalities in smoking in the areas studied either before or after the introduction of smoking cessation services. Without such data, conclusions relating to the impact of the intervention on inequalities in smoking rates by deprivation quintile cannot be drawn. As the authors note, the lack of data on overall quit rates in the different deprivation quintiles means that one cannot immediately determine the impact of the intervention on smoking rates and hence inequalities. However, it is possible to carry out sensitivity analyses based on estimated smoking rates to investigate the potential impact of the intervention on inequalities in smoking.

Table 1 shows the number of male smokers before and after the intervention, using data published in the paper and hypothetical pre-intervention smoking rates representing perfect equality (25% in both the least and most deprived quintiles), extreme inequality (5% in the least deprived, and 50% in the most deprived quintiles) and the degree of inequality quoted in the paper using data from the General Household Survey (15% in the least deprived and 39% in the most deprived quintiles). We have also calculated relative risks of smoking in the most deprived, compared with the least deprived, quintile before and after the intervention as a measure of inequality. In all three scenarios, it can be seen that the quit rates achieved by the intervention in people living in the most and least deprived quintiles have a very small effect on the relative risk of smoking in the least compared with the most deprived quintile, and hence on inequality.

From our further analysis of the published data, we conclude that the authors’ claims that “services are reducing inequalities between geographical areas” are not justified and hence the title of the article is misleading. Furthermore, the conclusion that “NHS smoking cessation services are successfully attracting significant number of people from deprived areas” seems of limited validity. Only 1.13% (3799 of 336 800) of people from the most deprived quintile actually accessed services. Even if 50% of people in the most deprived quintile were smokers, this still only represents 2.26% of smokers in this quintile.

We believe that the proposition that health promotion interventions may often increase, rather than decrease, overall SEIH is feasible, and worthy of further consideration. Analyses such as those attempted by Lowey et al., and completed by us, are essential for confirming or refuting the validity of this hypothesis. As we have shown, simple statistics can be used to quantify and compare the degree of inequality within a population. Authors should be careful to ensure that their claims are substantiated by the data they present and be prepared to extend analysis with sensitivity models where necessary to test relevant hypotheses.

**References**


**Authors’ reply**

Adams and White fall shy of refuting that smoking cessation services are reducing inequalities. However, using our data and...
speculative levels of smoking, they examine how smoking cessation services might have changed the relative risk of smoking between the least and most derived areas. Like Adams and White we are aware of data from the national household survey on prevalence of smoking according to deprivation. Furthermore, since then a smoking prevalence survey has been published for two of the primary care trusts (one relatively affluent, Bebington and West Wirral and one relatively deprived, Birkenhead and Wallasey) within our study area. Prevalence of smoking for these was measured at 13% and 25% respectively. Using these figures and those from the General Household Survey (15% smokers in least deprived areas and 39% in the most deprived) we have tested whether greater proportions of smokers are quitting from deprived areas. Data from our study showed that the proportions of males quitting from deprived areas are significantly higher (table 1). In our original paper we chose not to publish any such analyses as the actual prevalence of smoking across our study areas was not known. Instead, in our comments we acknowledged the urgency with which such data are needed.1

We would not dispute that currently changes in smoking prevalence resulting from smoking cessation services are relatively modest. However, this is to be expected as only a fraction of all smokers are currently accommodated by such services. Furthermore, disproportionate effects on areas of high deprivation are also reduced by a greater drop out rates of people recruited from those most deprived areas. Again this is an issue we have suggested is tackled as a matter of urgency.

In reality, smoking cessation services can never dramatically affect relative risks of smoking (between least and most deprived areas) while the number of people they see represent such small proportions of the smoking population. However, we have shown that modest investment in such services has had an impact disproportionately on more deprived populations. Consequently, greater investment in smoking cessation services (as part of a suite of interventions to reduce smoking) may even deliver the changes in relative risk sought by Adams and White.

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References

<table>
<thead>
<tr>
<th>Survey source</th>
<th>Grouping</th>
<th>Population (aged over 17 years)</th>
<th>Smoking rate (%)</th>
<th>Estimated smokers</th>
<th>Quit smoking</th>
<th>Smokers quitting (%)</th>
<th>χ²*</th>
<th>p Value</th>
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<td>General Household</td>
<td>Least deprived quintile</td>
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<td>28,060</td>
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<td>Most deprived quintile</td>
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<td>39</td>
<td>13,1352</td>
<td>826</td>
<td>0.63</td>
<td>43.2</td>
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<td>Birkenhead and Wallasey PCT</td>
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<td>387</td>
<td>2.28</td>
<td>5.94</td>
<td>0.01</td>
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*The χ² test compares the proportion quitting with the proportion not quitting between groupings.

Table 1 Estimated proportions of male smokers quitting by deprivation quintile