Is the rate of biological ageing, as measured by age at diagnosis of cancer, socio-economically patterned?

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Abstract

Study objective: To investigate the hypothesis that biological ageing, as measured by age at diagnosis of some common cancers, is socio-economically patterned.

Design: A cross-sectional analysis of the association between an area based measure of material deprivation and age at diagnosis of four common cancers (breast, prostate, colorectal and lung cancers). A further analysis, restricted to breast and colorectal cancer, allowed us to adjust for stage and grade of cancer at diagnosis.

Setting: The Northern and Yorkshire cancer registry and information service, Northern and Yorkshire region, UK.

Participants: All individuals living in the Northern and Yorkshire region diagnosed with breast, prostate, colorectal or lung cancer in 1986-1995. All individuals living in the Northern and Yorkshire region diagnosed with breast or colorectal cancer in 1998-2000 with documented data on stage and grade of cancer at diagnosis.

Main results: There was evidence that greater material deprivation was associate with younger age at diagnosis of cancer in prostate (beta coefficient -0.073), colorectal (women: -0.042; men: -0.063) and lung cancer (women: -0.214; men: -0.161). The opposite relationship was found in women with breast cancer (0.149). Adjusting for stage and grade at incidence, where possible, had little effect on the magnitude of the beta coefficients.

Conclusions: Age at diagnosis of some common cancers appears to be socio-economically patterned with individuals from more deprived areas being diagnosed with prostate, colorectal and lung cancers earlier in life. The opposite was seen in women with breast cancer. In all cases the size of the effect was small. Further work is required to investigate the socio-economic distribution of more accurate measures of biological ageing.

Keywords: ageing, socio-economic position, cancer
Background

Socio-economic differences in health (SEDH) have been widely documented and appear to be universal.[1-4] However, the precise biological pathways linking socio-economic position (SEP) and health remain unclear.[5] Although differential exposure to specific risk and protective factors is often cited as one of the key determinants of SEDH,[3] the relationship between SEP and specific risk and protective factors is not straightforward and appears to vary across both time and space.[6] This challenges risk factor-specific explanations of SEDH – as soon as prevailing risk factors change, so too must associated explanations of SEDH.

An alternative explanation of SEDH is that more fundamental pathways to health and disease exist and that specific risk and protective factors feed into these fundamental pathways.[7] Whilst the socio-economic distribution of specific risk and protective factors may vary over time, the socio-economic distribution of overall risk appears to be fairly constant in developed countries. Sociologists have offered a number of explanations for why socio-economically deprived groups tend to experience the least healthy risk factor profiles in society.[8]

One possible fundamental, biological mechanism of SEDH is biological ageing (Figure 1).[7] This is the progressive decrease in physiological ability to meet demands that occurs over time[9] and is currently understood to be due to the accumulation of damage at the cellular level.[10] The rate of accumulation of cellular damage is determined by the balance between cellular damage occurring and the action of cellular defence and repair mechanisms.[10] Cellular damage includes mitochondrial damage, the accumulation of aberrant proteins within the cytosol, reactive oxygen species and the oxidative stress they generate and somatic mutations.[11] Cellular defence and repair mechanisms include antioxidant vitamins and enzymes and DNA checking and repair enzymes.

Unlike chronological ageing, there is considerable variation between individuals in the rate of biological ageing.[12] Both genetic and environmental factors play a role[1314] and numerous environmental factors, known to be socio-economically patterned, are now known
to influence the rate of cellular damage accumulation, including cigarette smoking,[15] diet[16] and exposure to radiation.[17]

Furthermore, there is good reason to believe that biological ageing and the accumulation of cellular damage are important determinants of health. Population based research suggests that some crude markers of ageing, such as total and healthy life expectancy, are socio-economically patterned.[1819] However, no research, to date, has attempted to investigate the socio-economic distribution of more accurate measures of biological ageing. Cancers are due, at the most proximate level, to acquired or inherited mutations in genes that control growth – specifically, oncogenes and tumour suppressor genes.[20] As genetic mutations are one, well recognised, type of cellular damage,[11] the development of cancer can be understood as being closely related to the processes involved in biological ageing. Chronological age of development of cancer can, therefore, be used as a comparative measure of biological ageing.

We investigated the effect of socio-economic position (SEP) on the age of development of some common cancers using data from a population based cancer registry.

Methods
The Northern and Yorkshire Cancer Registration and Information Service (NYCRIS) is one of nine regional cancer registries in England which aims to collect data on all incident cancers as they occur. We used data held by NYCRIS to investigate the effects, if any, of SEP on age at diagnosis of cancer.

Variables of interest
Socio-economic position was measured using Townsend Deprivation Scores (TDS)[4] of the enumeration district of residence at date of diagnosis of cancer calculated from full seven digit postcode data and standardised to the NYCRIS region as a whole using data from the 1991 census. A positive TDS represents more material deprivation whilst a negative score represents less material deprivation.
Age at diagnosis of cancer was calculated from incident date and date of birth. As is routine in UK cancer registries, incident date was defined as the first available (not necessarily earliest) date from: date of pathologically confirmed diagnosis, date of first hospital attendance during which a clinical diagnosis was made, date of diagnosis by a GP and date of death.

Cancer site was assigned using three figure codes from the 10th revision of the International Classification of Diseases (ICD-10) as extrapolated from clinical records and recorded in the NYCRIS database. We restricted the analysis to the four commonest non-skin cancer sites: breast (ICD-10 C50), colorectal (ICD-10 C18, C19 or C20), lung (ICD-10 C33 or C34) and prostate cancer (ICD-10 C61).

Sample size and study years
As no previous work in this area has been reported, we were unable to perform sample size or power calculations. Instead, we adopted a pragmatic approach aiming to use as large a sample size as possible. This was, however, restricted by our measure of SEP, TDS, which are calculated from decennial UK census data. In order to avoid using data from more than one census, we restricted ourselves to 10 years of NYCRIS data and chose to centre our data collection years around a census year so that the maximum temporal extrapolation of census data would be five years. The most recent census year for which data could be extrapolated five years both backwards and forwards was 1991. We therefore chose 1986-1995 inclusive as our study years.

Further analysis to assess the effects of adjusting for stage and grade at diagnosis
Using the method described to determine incident date, and therefore age at diagnosis, does not necessarily result in comparable data for all individuals. In order to minimise any bias or error introduced by this method, we wished to adjust for stage and grade of cancer at diagnosis. However, these data were far from complete for the main study cohort and it was estimated that the only cohorts with more than 50% complete data on stage and grade were individuals registered with colorectal and breast cancers since 1998. We, therefore,
performed a further analysis, using similar methods as the main study, which investigated the effect of adjusting for stage and grade at diagnosis on the relationship between age and TDS at diagnosis. This analysis used data on colorectal and breast cancers registered between 1998 and 2000 inclusive.

**Exclusions**
Individuals were excluded from the analysis if: key data were missing, registration had been by death certification only (and therefore the date of incidence had been recorded as the date of death),[21] or the incident cancer was a second primary. Men with breast cancer were excluded due to the rarity of this condition.

A proportion of individuals inherit a specific genetic defect which puts them at high risk of developing a cancer early in life.[22] We felt that such individuals formed a distinct group and should not be included in the present analyses because of our focus on environmental factors. As cancer registries do not routinely record information on inherited susceptibilities to cancer, we excluded the youngest 25% of individuals, at diagnosis, from each cancer and gender specific group as a conservative approach to removing all those whose cancer may include a substantial inherited component.

**Analysis**
Data was analysed in gender and cancer site specific groups throughout. Linear regression techniques were used to assess the ability of TDS at diagnosis to predict age at diagnosis of cancer both before, and in the case of the 1998-2000 cohort, after adjustment for stage and grade of cancer at diagnosis. Neither age nor TDS was normally distributed in any of the gender and cancer site specific groups. As simple transformation did not result in appreciably more normal distributions of these variables, bootstrapping methods, with 1000 repetitions, were used to generate confidence intervals for beta coefficients derived from linear regression techniques.[23] All analyses were performed in Intercooled Stata version 8.0.
Results

The 1986-1995 cohort

Between January 1st 1986 and December 31st 1995, 144,627 cases of lung, colorectal, prostate and breast cancer were registered with NYCRIS. Of these, 39,301 (27.2%) met one of the exclusion criteria, including the age cut off, leaving 105,326 individuals in the analysis.

Table 1 shows the distribution of age and TDS at diagnosis in the 1986-1995 cohort before and after exclusion of the youngest 25% of individuals at diagnosis. In the cohort included in the analysis, women with breast cancer were youngest (median age 67.84) and women with colorectal cancer oldest (median age 78.05) at diagnosis. In addition, women with breast cancer were least deprived (median TDS -0.50) and women with lung cancer most deprived (median TDS 1.28) at diagnosis. Whilst exclusion of the youngest 25% at diagnosis led to an overall increase in the affluence of the cohorts with lung cancer, the opposite was seen in the other groups.

[Table 1 about here]

The results of simple linear regression analysis of the ability of TDS at diagnosis to predict age at diagnosis are shown in Table 2. The 95% confidence intervals of the beta coefficients of TDS at diagnosis excluded unity in all cases. The negative beta coefficients seen in relation to prostate (beta coefficient -0.073), colorectal (women: -0.042; men: -0.063) and lung cancer (women: -0.214; men: -0.161) suggest that individuals living in more deprived areas tended to have these cancers diagnosed earlier in life than those living in more affluent areas. These coefficients equate to a variation in age at diagnosis of cancer of between 0.21 (colorectal cancer, women) and 1.07 years (lung cancer, women) across the inter-quartile range of TDS.

Amongst women with breast cancer, the beta coefficient of TDS at diagnosis was positive (0.149) suggesting that women living in more deprived areas tended to have breast cancer
diagnosed later in life than those living in more affluent areas. This equates to a variation of 0.72 years in age at diagnosis across the inter-quartile range of TDS.

[Table 2 about here]

The 1998-2000 cohort
There were 25,314 cases of colorectal and breast cancer registered with NYCRIS between 1st January 1998 and 31st December 2000. Of these, 11,445 (45.2%) were excluded from the analysis – almost half of these exclusions were made on the grounds of missing stage or grade data. There was evidence that those excluded from the analysis on the grounds of missing stage or grade data were from more deprived neighbourhoods than those included (p<0.0001) and were older (p<0.0001). Table 3 shows the distribution of age and TDS at incidence in the 1998-2000 cohort before and after exclusion of the youngest 25% of individuals at diagnosis. As before, women with breast cancer tended to be youngest (median age 63.30) and women with colorectal cancer tended to oldest (median age 76.11) at diagnosis. Individuals included in the 1998-2000 cohort tended to be more affluent and slightly younger at cancer incidence than those in comparable groups in the main 1986-1995 cohort.

[Table 3 about here]

Table 4 shows the results of linear regression analyses of the ability of TDS at diagnosis to predict age at diagnosis both before and after adjustment for stage and grade at diagnosis. There was less evidence of statistically significant relationships between TDS and age at diagnosis in this cohort compared to the 1986-1995 cohort. However, adjustment for stage and grade at diagnosis had little effect on the magnitude of the beta coefficients in all three site and gender specific groups in the 1998-2000 cohort.

[Table 4 about here]

Discussion
This is the first study that has assessed the socio-economic distribution of either age at diagnosis of cancer or other possible markers of biological ageing. Using data from more
than 100 000 individuals, we found evidence of a small, negative, effect of TDS on age at diagnosis of prostate, lung and colorectal cancers. Individuals living in more materially deprived areas tended to have these cancers diagnosed earlier in life with a difference of between 0.21 and 1.07 years in age at diagnosis of cancer across the inter-quartile range of TDS. The opposite relationship was found amongst women with breast cancer with women living in more deprived areas tending to have breast cancer diagnosed later in life than those living in less deprived areas. In a second analysis, adjusting for stage and grade of cancer at incidence had very little effect on the magnitude of the effect of TDS on age at diagnosis for breast and colorectal cancers.

Although routinely collected data, such as used here, is not always of high quality, there is reason to believe that the UK cancer registries, and NYCRIS in particular, collect data of a consistently high standard.[21] For example, in 1994 NYCRIS collected contemporaneous data from more than 93% of individuals who died with cancer in the region – compared with a UK average of only 88%.[24] In addition, only 497 (0.3% of the total number of registrations) individuals were excluded from the main cohort in the present study due to missing data. We are, therefore, confident that our analysis, at least of the main 1986-1995 cohort, used high quality data with good ascertainment and completeness. In contrast, substantial exclusions were made from the 1998-2000 cohort due to missing data on stage or grade of cancer. We are, therefore, less confident of the quality of this data and cannot be sure that adjusting for stage and grade of cancer in this cohort provided us with a clear picture of actual effect of this procedure in a more complete cohort.

It is not necessarily desirable or appropriate to apply area based measures of SEP, such as TDS, to individuals as done here. The process involves both the assumption that small areas are relatively socio-economically homogeneous and that there may be significant temporal extrapolation of data.[2526] However, the limitations of the data collected by NYCRIS meant that an area based measure of SEP were all that was available to us. A wide variety of
previous work has applied TDS to individuals[27-31] and we made the pragmatic decision to use TDS as an individual measure of SEP in order to make use of this large and rich dataset. We used age at diagnosis of cancer as a proxy for age at development of cancer. Any socio-economic variations in the length of time between cancer development and diagnosis therefore have the potential to confound the relationships seen. However, the research evidence concerning socio-economic variations in diagnostic delay is far from consistent[32-35] and adjustment for stage and grade at incidence (two key markers of cancer progression and, therefore, diagnostic delay) in the present study had very little effect on the magnitude of the effect of TDS on age at diagnosis, suggesting that variations in timeliness of diagnosis did not explain the results seen. Given the quality of this data we cannot, however, be confident of this conclusion. However, recent research has reported an inverse association between deprivation and proximity to health services[36] suggesting that more deprived individuals are actually more likely to live in close proximity to treatment centres than more affluent ones. It is hard to define, even theoretically, exactly when a cancer occurs. Using the date at which a cancer is clinically detectable is one possible method but this would require a large prospective screening study with a short time span between screens. Our analyses have provided preliminary data in this area without recourse to such a lengthy and resource intensive study.

In addition to using date of diagnosis as a proxy for date of development of cancer, we also used it as a proxy for biological ageing. Although we believe that age at diagnosis of cancer is a useful comparative clinical marker of the rate of biological ageing, the pathways between biological ageing and diagnosis of cancer are long and complex and the cellular damage responsible for cancer is not necessarily the same cellular damage responsible for biological ageing. The results presented here can only begin to untangle the relationship, if any, between SEP and biological ageing.
Because our hypothesis focuses on environmental risk and protective factors, we felt that it was important to exclude individuals with cancers that had a substantial inherited component. This was done, somewhat arbitrarily, by excluding the youngest 25% of each gender and site specific group on the basis that inherited cancers are likely to occur earlier in life than sporadic ones, with inevitable knock on effects on the age and TDS distributions of the cohorts used in the analyses.[22] However, the division between cancers with a large inherited component and those that are primarily sporadic is not clear cut and not all of the 25% of cancers occurring at the youngest ages will necessarily have a predominant inherited component, nor will all cancers with a strong inherited component necessarily occur within this age group.

Further, despite the statistical significance of the influence of TDS on age at diagnosis found, the effect sizes seen were small with a standard deviation change in TDS at diagnosis resulting in less than a year change in age at diagnosis of cancer. The inaccuracies in our measure of biological ageing and the effects of unknown confounders may contribute to the minimal effect size seen and further research investigating socio-economic variations of more accurate markers of biological ageing is warranted.

We have identified two possible explanations for the main finding that age at diagnosis of prostate, colorectal and lung cancer decreases with increasing deprivation of the area of residence. Firstly, it is possible that the cancers that individuals across the socio-economic spectrum develop are biologically different in some way that leads to variations in age of onset. It has been highlighted that not all cancers, even of the same tissue, involve the same disease process.[37] Given the different environmental risks and hazards, and therefore carcinogens, that individuals across the socio-economic spectrum are exposed to, it is possible that, despite being classified under the same ICD heading in the NYCRIS database, not all cancers that we have analysed are comparable. Although such biological variations have been documented according to age,[38-40] we are not aware of any evidence that the biological nature of individual cancer types varies by SEP. An alternative explanation of our findings is
that the rate of accumulation of damage at the cellular level, and thus biological ageing, is socio-economically patterned due to socio-economic patterning of exposure to carcinogenic risk factors (Figure 1).

Neither of the preceding explanations can be applied to the results seen amongst women with breast cancer. The one key difference in the diagnosis of breast cancer in British women, compared to the other cancers studied, is the presence of the UK national breast cancer screening programme. This programme invites all women aged between 50 and 64 for triennial mammography. Although highly successful in detecting early breast cancer and reducing morbidity and mortality,[41] uptake of breast screening in the UK is known to be socio-economically patterned with more affluent women tending to attend more regularly than less affluent women.[4243] It is, therefore, possible that the tendency we found for more affluent women to have breast cancer diagnosed at younger ages is due to their more regular attendance for breast cancer screening and resultant more timely diagnosis of breast cancer. As we did not have information of who did and did not attend for screening, we were unable to explore this possibility further. However, the plausibility of this explanation further calls into question the accuracy of our adjustment for stage and grade of cancer at diagnosis in the 1998-2000 cohort.

Conclusions
By providing evidence that the age at diagnosis of some common cancers is socio-economically patterned, this study provides preliminary evidence that the rate of biological ageing may also be socio-economically patterned. Furthermore, we found little evidence that socio-economic variations in diagnostic delay – as measured by stage or grade of cancer at diagnosis – were confounding the relationships seen. To further investigate the relationship between SEP and biological ageing, studies using longitudinal measures of individual socio-economic circumstances across the lifecourse and measures of the actual process of biological ageing at the cellular level, rather than the clinical outcomes of it, will be required.
Acknowledgements
Many thanks to Caroline Brook and Cheryl Craigs at NYCRIS for methodological suggestions and help with abstracting data and to Nigel Unwin at Newcastle University for critical comments on a previous draft of the manuscript. JA is supported by the BUPA/Faculty of Public Health Medicine Joint Research Fellowship (2001-2004).

Key points
• Rate of biological ageing may mediate the relationship between socio-economic position and health and age at diagnosis of cancer may be a comparative marker of biological ageing with earlier age at diagnosis being indicative of faster biological ageing.

• More deprived individuals tended to develop colorectal, lung and prostate cancer earlier in life than more affluent individuals. The opposite trend was seen in relation to breast cancer.

Policy implications
• If the rate of biological ageing is confirmed as mediating the relationship between socio-economic position and health, targeting interventions which reduce the rate of biological ageing at deprived populations may help reduce socio-economic variations in health.
Table 1 – Distribution of age and Townsend Deprivation Score at diagnosis of cancer (1986-1995 cohort)

<table>
<thead>
<tr>
<th></th>
<th>Before exclusion of youngest 25% at diagnosis</th>
<th>After exclusion of youngest 25% at diagnosis</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>N</td>
<td>Age</td>
</tr>
<tr>
<td></td>
<td>Median</td>
<td>IQR$^\dagger$</td>
</tr>
<tr>
<td>Prostate cancer</td>
<td>17 107</td>
<td>74.80</td>
</tr>
<tr>
<td>Breast cancer, women</td>
<td>34 265</td>
<td>62.63</td>
</tr>
<tr>
<td>Colorectal cancer, women</td>
<td>16 872</td>
<td>74.29</td>
</tr>
<tr>
<td>Colorectal cancer, men</td>
<td>18 053</td>
<td>70.30</td>
</tr>
<tr>
<td>Lung cancer, women</td>
<td>18 116</td>
<td>69.78</td>
</tr>
<tr>
<td>Lung cancer, men</td>
<td>35 273</td>
<td>70.03</td>
</tr>
</tbody>
</table>

$^\dagger$inter-quartile range; $^\dagger$Townsend Deprivation Score

Table 2 – Linear regression models of the ability of Townsend Deprivation Score to predict age at diagnosis of cancer (1986-1995 cohort)

<table>
<thead>
<tr>
<th></th>
<th>beta coefficient (95% CI*)</th>
<th>t (p-value)</th>
<th>adjusted $r^2$</th>
<th>Change in age/TDS IQR$^+$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Prostate cancer</td>
<td>-0.073 (-0.106 to -0.041)</td>
<td>-4.446 (&lt;0.0001)</td>
<td>0.001</td>
<td>-0.36 years</td>
</tr>
<tr>
<td>Breast cancer</td>
<td>0.149 (0.109 to 0.193)</td>
<td>7.018 (&lt;0.0001)</td>
<td>0.002</td>
<td>0.72 years</td>
</tr>
<tr>
<td>Colorectal cancer, women</td>
<td>-0.042 (-0.081 to -0.004)</td>
<td>-2.070 (0.039)</td>
<td>&lt;0.001</td>
<td>-0.21 years</td>
</tr>
<tr>
<td>Colorectal cancer, men</td>
<td>-0.063 (-0.100 to -0.027)</td>
<td>-3.367 (0.001)</td>
<td>0.001</td>
<td>-0.33 years</td>
</tr>
<tr>
<td>Lung cancer, women</td>
<td>-0.214 (-0.252 to -0.177)</td>
<td>-11.350 (&lt;0.0001)</td>
<td>0.009</td>
<td>-1.07 years</td>
</tr>
<tr>
<td>Lung cancer, men</td>
<td>-0.161 (-0.185 to -0.136)</td>
<td>-12.886 (&lt;0.0001)</td>
<td>0.006</td>
<td>-0.82 years</td>
</tr>
</tbody>
</table>

*95% confidence intervals (bias corrected) of beta derived from bootstrapping with 1000 repetitions; $^+$change in age at diagnosis of cancer across the interquartile range in Townsend Deprivation Score from lower (most affluent) quartile to upper (most deprived quartile)
Table 3 – Distribution of age and Townsend Deprivation Score at diagnosis of cancer (1998-2000 cohort)

<table>
<thead>
<tr>
<th></th>
<th>Before exclusion of youngest 25% at diagnosis</th>
<th>After exclusion of youngest 25% at diagnosis</th>
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<tbody>
<tr>
<td></td>
<td>N</td>
<td>Median</td>
</tr>
<tr>
<td><strong>Age</strong></td>
<td></td>
<td></td>
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<tr>
<td>Breast cancer, women</td>
<td>9920</td>
<td>58.50</td>
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<td>Colorectal cancer, women</td>
<td>3707</td>
<td>72.67</td>
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<tr>
<td>Colorectal cancer, men</td>
<td>4861</td>
<td>70.38</td>
</tr>
</tbody>
</table>

†inter-quartile range; ‡Townsend Deprivation Score;

Table 4 – Linear regression models of the ability of Townsend Deprivation Score to predict age at diagnosis of cancer before and after adjustment for, stage and grade at diagnosis (1998-2000 cohort)

<table>
<thead>
<tr>
<th></th>
<th>Unadjusted</th>
<th>Adjusted for stage and grade at diagnosis of cancer</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>beta (95% CI*)</td>
<td>t (p-value)</td>
</tr>
<tr>
<td>Breast cancer</td>
<td>0.227 (0.159 to 0.306)</td>
<td>6.069 (&lt;0.0001)</td>
</tr>
<tr>
<td>Colorectal cancer, women</td>
<td>0.009 (-0.074 to 0.096)</td>
<td>0.151 (0.880)</td>
</tr>
<tr>
<td>Colorectal cancer, men</td>
<td>-0.056 (-0.119 to 0.012)</td>
<td>-1.617 (0.106)</td>
</tr>
</tbody>
</table>

*95% confidence intervals (bias corrected) of beta derived from bootstrapping with 1000 repetitions; ‡change in age at diagnosis of cancer across the interquartile range in Townsend Deprivation Score from lower (most affluent) quartile to upper (most deprived quartile)
Figure 1 – The hypothesised pathway linking socio-economic position, environmental risk and protective factors, biological ageing and health status
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