APPENDIX

## the spline functions were defined for boundary knots at ages 20 and 65 years and W(a) was defined to be 1 at age 50 years; for D(t), the spline functions were defined with boundary knots at 1945 and 1980 and D(t) was defined to be 1 in 1970; internal knots were assumed to be at the mid-point between the

 $u = \tau + 0.5$  to  $u = a.^{\infty}$  The dose potential functions W(a) and D(t) were specified as exponentials of natural splines with one

internal knot and two parameters to be estimated. For W(a),

Clements, Berry, Shi, et al.

The five parameters to be estimated included the log of the constant  $\beta$  and four parameters for the spline functions log(W) and log(D). We assumed that k was fixed at 3.5. The asbestos half-life H was given a value of 15 years. For incidence data, the latency period \u03c4 was assumed to be 5 years. As in Model 1 of Hodgson et al, the diagnosis fraction  $D_x(t)$  was assumed to be almost complete (98%) in 1997, with a 5% annual percentage trend in the missing cases, such that

boundary knots.

 $D_r(t) = 1-0.02 \times 1.05^{1997-t}$ . The models were fitted using the mle() function in the R statistical package,22 which uses a quasi-Newton approach to maximise the Poisson log-likelihood. The design matrices for the spline functions were calculated using the ns() function in R. We assessed goodness of fit by performing a likelihood ratio test using the residual deviance with the residual degrees of freedom. For interval estimation for predictions, we used the

bootstrap, re-sampling from Pearson's residuals.23 Following

Friedl, we standardised the residuals by dividing by the square

root of (1-model degrees of freedom/number of observations)

and scaled the residuals to have zero mean.24

## AGE AND CALENDAR YEAR MODEL Following Hodgson et al,6 we assumed: (1) that the mesothe-

lioma rate for a cohort aged a at time t exposed to asbestos at time u in the past was proportional to the earlier asbestos dose (that is, rate(a,t|u) was proportional to dose(a-u,t-u)); (2) that

dose was multiplicative by age and time, that is, dose(a,t) was proportional to W(a)D(t), for dose potentials by age ( = W) and by time (=D); (3) that the rate was proportional to time from exposure to malignant conversion raised to some power k together with an effect due to clearance of asbestos fibres from the lung, where the half-life is represented by H and  $\tau$  is the average latency time from malignant conversion of the cancer to clinical detection; and (4) that the observed number of cases followed a Poisson distribution with mean μ<sub>st</sub>. To model the mesothelioma rate, we averaged the rates across all of the times

since exposure from  $\tau$  to a. The predicted number of cases  $\mu_{at}$  was

calculated by weighting the mesothelioma rate by the person-

years at risk estimated by the population  $(=P_{at})$  and scaled by the

fraction of correctly diagnosed cases at year  $t = D_x(t)$ . Taking  $\beta$ 

as a constant, the Poisson regression model was:

$$\mu_{at} = \left[\beta \int_{\tau}^{a} D\left(t - u\right) W\left(a - u\right) \left(u - \tau\right)^{k} O.5^{(u - \tau)/H} du\right] P_{at} D_{x}(t)$$

We modelled the rates for the mid-point of the five-year age intervals (for example, for ages 40-44 years, a = 42.5). The mean rate  $\mu_{ot}$  was numerical integrated using Simpson's rule, with functional evaluations at single-year increments for