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Scand J Work Environ Health 1992;18(2):105-112

<https://doi.org/10.5271/sjweh.1604>

Issue date: 01 Apr 1992

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by [Kauppinen TP](#), [Mutanen PO](#), [Seitsamo JT](#)

**Affiliation:** Institute of Occupational Health, Helsinki, Finland.

The following articles refer to this text: [2013;39\(6\):599-608](#);  
[2017;43\(5\):415-425](#)

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## Magnitude of misclassification bias when using a job-exposure matrix

by Timo P Kauppinen, PhD, Pertti O Mutanen, MSc, Jorma T Seitsamo, MSocSc<sup>1</sup>

KAUPPINEN TP, MUTANEN PO, SEITSAMO JT. Magnitude of misclassification bias when using a job-exposure matrix. *Scand J Work Environ Health* 1992;18:105–12. A job-exposure matrix constructed in Southampton, the United Kingdom, was assessed with the use of 1205 occupational histories collected for a case-referent study on primary liver cancer in Finland. The odds ratios calculated on the basis of the matrix and an assessment by occupational hygienists were compared. The ability of the matrix to detect excess risk was generally satisfactory for chemical and physical agents to which about 10% or more of the studied population was exposed. If only probable exposure at a high level was assigned to the category of the exposed, an excess risk was usually detectable also when the prevalence of exposure was below 10%. This assessment indicates that the British job-exposure matrix is an acceptably valid screening tool also outside the United Kingdom, provided that the misclassification characteristics of the matrix and the significant differences in exposure between the countries are taken into account.

**Key terms:** prevalence of exposure, sensitivity, specificity.

The job-exposure matrix provides a rapid means with which to convert coded occupational titles into potential exposures in epidemiologic studies. Especially in large population-based studies, it is an easy and systematic method with which to penetrate from proxy variables of exposure, such as industry or occupation, directly to agents and other risk factors. In addition, a job-exposure matrix does not distinguish between diseased and nondiseased subjects, and thus the differential information bias is diminished that could occur if exposure were assessed with the traditional person-by-person approach. However, there are several methodological problems in the use of job-exposure matrices (1). These matrices do not usually take into account the variability of exposure within occupational classes between different workplaces, countries, or calendar periods and therefore may result in false positive and negative exposure assignments for a considerable proportion of the subjects. This misclassification of exposure, even though nondifferential as to the disease status of the subjects, tends to mute the observed risk estimates towards unity. The inaccuracy of the job-exposure matrix approach is diminished if the time and even the factory dimensions are incorporated into the matrices, but an increase in accuracy is often followed by a decrease in general applicability. Therefore, the question of the reliability of the results is always warranted when job-exposure matrices are used to analyze exposure, especially if the matrix used is a simple cross-tabulation of occupational titles and agents.

The validity of job-exposure matrices has been evaluated mainly in comparisons of results provided by the matrix with those of some other method of exposure assessment (2–5). The results given by two different matrices have also been compared (6). Still another approach is to calculate the effect of misclassification on the basis of estimates given to the factors which determine the magnitude of the misclassification bias. The advantage of this method, which is used in the present study, is that it does not require an excess risk for testing because the true risk can be set by assumption. All agents in the matrix can therefore be assessed for misclassification bias.

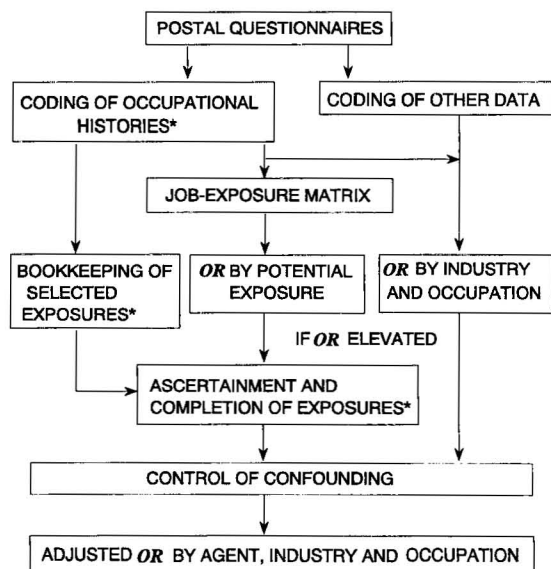
In this paper we report the magnitude of the misclassification effect on risk estimates when using a British job-exposure matrix in connection with a population-based case-referent study on primary liver cancer in Finland (7, 8). We also attempted to elucidate some general principles of the feasibility of the job-exposure matrix approach and the validity of the results in retrospective studies on chronic health hazards.

### Material and methods

The material for testing the job-exposure matrix consisted of chronological occupational histories of 344 cases (primary liver cancer) and 861 referents (stomach cancer or coronary infarction) collected with postal questionnaires from next-of-kin. The mean age of this population was 68.0 years. The schedule of exposure assessment of the study is presented in figure 1, and it has been described in more detail elsewhere (8). The occupational histories were coded according to the British industrial and occupational classifications to allow for the use of a job-exposure matrix constructed

<sup>1</sup> Institute of Occupational Health, Helsinki, Finland.

Reprint requests to: Dr T Kauppinen, Institute of Occupational Health, Topeliuksenkatu 41 A A, SF-00250 Helsinki, Finland.



**Figure 1.** Exposure assessment procedure of the national population-based case-referent study on primary liver cancer in Finland. Operations marked with a star were carried out by occupational hygienists. (JEM=job-exposure matrix, OR=odds ratio)

in the Medical Research Council's Environmental Epidemiology Unit in Southampton, the United Kingdom (5). The British job-exposure matrix is based on combinations of industrial and occupational classes that have been cross-tabulated with 50 chemical agents or other exposure factors. A crude classification of the calendar period (before 1950, after 1950), probability of exposure (low = 1, high = 2), and level of exposure (low = 1, high = 2) are also incorporated into the matrix. A computer program was constructed to convert the occupational histories of the subjects into exposure indices for 50 factors. For example, the assignment to class 1122 for agent X would mean that the person is considered to have had a low probability and level of exposure before 1950, but a high probability and level of exposure after 1950.

The inclusion of probability and level of exposure in the matrix enables the persons to be rated into semi-quantitative classes according to the potential of exposure. One extreme way to use this matrix would be to consider only those exposed persons who have a high probability and level of exposure (exposure = "high" potential exposure in the job-exposure matrix). The other extreme is to consider everyone exposed whose probability of exposure is nonzero according to the matrix (exposure = sum of "low" and "high" potential exposure = "total" potential exposure). The first approach is strict as to the exposure assignment because it accepts only relatively certain exposure, and thus it maximizes the specificity of the exposure assessment at the cost of sensitivity. The second approach is liberal in accepting all potentially exposed persons and thus

maximizes sensitivity at the cost of specificity. In this paper we consider both of these approaches.

The method to assess the British job-exposure matrix was based on the theory of misclassification and its effect on the risk estimates. (See, eg, references 9–12.) It may be proved that the observed odds ratio (OR\*) depends on the true odds ratio (OR), the prevalence of exposure, the sensitivity of exposure assessment, and the specificity of exposure assessment according to the following formula (modified from reference 10):

$$OR^* = \frac{[(Se)(OR)(Pr) + (1 - Sp)(1 - Pr)][(1 - Se)(Pr) + (Sp)(1 - Pr)]}{[(Se)(Pr) + (1 - Sp)(1 - Pr)][(1 - Se)(OR)(Pr) + (Sp)(1 - Pr)]}$$

where OR\* = the observed (biased) odds ratio, OR = the true (unbiased) odds ratio, Pr = the prevalence of exposure (ie, probability of exposure in the reference group during the follow-up period), Se = sensitivity of exposure assessment (ie, probability that a truly exposed person will be classified as exposed), Sp = specificity of exposure assessment (ie, probability that a truly unexposed person will be classified as unexposed).

We decided to fix the OR at two values, OR = 5 and OR = 2, representing situations in which the risk is high and moderate, respectively. The prevalence of the studied population's exposure to the 50 agents was estimated on the basis of the exposure indices provided by the job-exposure matrix in the following manner. For example, if the job-exposure matrix gave "high" potential exposure (H) for 10% and "low" potential exposure (L) for 50% of the persons under study, the fractions of truly exposed persons ( $f_H$  and  $f_L$ , say, 90 and 10%, respectively) in these classes (H and L) were estimated by one or several industrial hygienists. The estimation was based on the survey of the occupational titles and work periods of the potentially exposed persons and on various documents concerning the occurrence and level of exposure in Finland. The value of  $f_H$  varied rather widely by agent and ranged from 30 to 100% (mean 86%, most frequent value 90%). The corresponding variation of  $f_L$  was 0–90% (mean 18%, most frequent value 10%). Because occupational histories are often incompletely reported, a small proportion of exposures usually remains unrecognized by the job-exposure matrix. We assumed for each agent that 10% of the truly exposed persons would be incorrectly assigned to the class of the unexposed ("total" sensitivity of the job-exposure matrix = 90%) and should therefore be added to the exposed group to achieve the best estimate of the prevalence of exposure. The formula to calculate prevalence thus became  $Pr = 1.11(f_H H + f_L L)$ , where the multiplier 1.11 (= 1/0.9) was due to the incompleteness of the occupational histories. The estimate of the prevalence of exposure in our example would thus be  $Pr = 1.11[(0.9)(0.1) + (0.1)(0.5)] = 0.16$  (or 16%).

For 14 agents it was also possible to check the value of the estimated prevalence against the value of the ascertained exposure in the study population as assessed by a team of occupational hygienists. The agents checked were aromatic amines, asbestos, beryllium, cadmium, chlorophenols, chromium, detergents, dyestuffs, electromagnetic fields, ethylene oxide, lead, organic solvents, other inorganic dusts, and welding fumes. The two estimates of prevalence were usually in good agreement, and, if there was a discrepancy, the estimated prevalence was corrected to correspond with the assessment of the team.

The values of sensitivity and specificity can be directly calculated with the use of their definitions and the estimated values of the truly exposed ( $Pr$ ) and unexposed ( $1 - Pr$ ) persons. Sensitivity is the proportion of correctly classified exposed persons of all the exposed persons. Specificity is the same proportion for the unexposed persons. In the situation in which only high potential exposure is considered as exposure (subscript H referring to high potential exposure),  $Se_H = f_H H / Pr$  and  $Sp_H = [1 - (Pr) - (1 - f_H)H] / (1 - Pr)$ . When exposure encompasses all potential exposure assigned by the job-exposure matrix (subscript T referring to total potential exposure),  $Se_T = 0.9$  (=90% by assumption due to incompleteness of the occupational histories) and  $Sp_T = [1 - (Pr) - (1 - f_H)H - (1 - f_L)L] / (1 - Pr)$ . Because the values of prevalence, sensitivity, and specificity are often different for men and women, they were calculated separately for both genders.

With the values of OR,  $Pr$ ,  $Se_H$  and  $Sp_H$  (or  $Se_T$ , and  $Sp_T$ ) it is possible to assess how much the OR\* is biased towards unity due to nondifferential misclassification of exposure. The bias is independent of the size of the study. We also wanted to study the effect of misclassification on the statistical significance of the result. The statistical significance (confidence interval) depends on the size of the study, and therefore we assumed a study of 250 cases and 500 referents (all men or alternatively all women). The lower confidence limits for OR and OR\* were calculated according to the test-based principle (13) except for values taken directly from the original study (in table 3 in the Results section) for which Cornfield's method was used (14).

## Results

The estimated values of the prevalence of exposure and the sensitivity and specificity of the job-exposure matrix are presented in table 1 for the men and women separately. Physical factors such as physical stress, outdoor occupation, and cold were common in the studied population. Surrogates of biological agents, such as contact with animals or the public, were also relatively common for both genders. The most frequently occurring chemical agents were "other" organic dusts (eg, hay dust, flour dust, fodder dust), grain dust, diesel fuel or exhaust (eg, from tractors)

and nitrates (eg, from fertilizers), reflecting the high proportion of persons who had at some time during their occupational history been farmers or farmers' wives. Most of the industrial agents were rarer, their prevalence being typically < 5% and often even < 1%. The women were exposed significantly less frequently than the men. The average prevalence was 9.0 (median 2.3)% for the men and 7.1 (median 0.8)% for the women. The respective values for chemical agents were 6.2 (median 1.4)% for the men and 4.1 (median 0.7)% for the women. However, some agents such as detergents, textile dust, and contact with the public or with animals were more common among the women than among the men. No men were assigned to the high potential exposure category for 18 out of 50 agents, and no women for 25 out of 50 agents. In most cases these were agents that were rare ( $Pr < 1\%$ ).

The sensitivity and specificity of the job-exposure matrix depended on the agent and varied widely. When only high potential exposure was taken into account, sensitivity was, on the average, low (43% for the men and 45% for the women) and specificity was high (99% for the men and 98% for the women). The use of total potential exposure as exposure resulted in higher sensitivity (90% by assumption concerning the incompleteness of the occupational histories) and lower specificity (83% for the men and 87% for the women).

The effect of misclassification on the OR\* is presented in table 2. When the "high" approach was used, the OR\* dropped by an average of 30–35% (from 2 to 1.7 and from 5 to 3.6). The effect was approximately the same for both the men and the women. The drop was much greater for the "total" approach, on the average 80% (from 2 to 1.2 and from 5 to 1.8).

The magnitude of misclassification bias as a function of prevalence for the high and total approaches is illustrated in figures 2 and 3. With the total approach OR\* decreased and the misclassification bias increased when the agent became rarer. With the high approach OR\* decreased when the agent became more common. When the prevalence was below 10%, the misclassification bias was relatively small, and the restricting factor was not misclassification, but rather the lack of heavily exposed cases when the agent was rare.

Solvents and welding fumes gave statistically significantly elevated values in the original study and thus allowed a direct comparison between the job-exposure matrix method and the assessment of a team of occupational hygienists. The results of this comparison are presented in table 3. The observed pattern for the OR\* was similar for both methods, but the occupational hygienic analysis provided somewhat higher OR\* values than the analysis with the job-exposure matrix.

## Discussion

The British job-exposure matrix used in a Finnish case-referent study on liver cancer turned out to be rela-

**Table 1.** Estimated prevalences of exposure and the sensitivities and specificities of the British job-exposure matrix (JEM) for the men and women in the Finnish case-referent study on primary liver cancer. (Pr = prevalence of exposure in the reference group, Se<sub>H</sub> = sensitivity of the JEM when only "high" potential exposure was considered as exposure, Sp<sub>H</sub> = specificity of the JEM when only "high" potential exposure was considered as exposure, Se<sub>T</sub> = sensitivity of the JEM when all ("total") potential exposure was considered as exposure, Sp<sub>T</sub> = specificity of the JEM when all ("total") potential exposure was considered as exposure, NA = "high" potential exposure approach not applicable because nobody was classified as exposed by the JEM)

Agent	Men					Women				
	Pr (%)	Se <sub>H</sub> (%)	Sp <sub>H</sub> (%)	Se <sub>T</sub> (%)	Sp <sub>T</sub> (%)	Pr (%)	Se <sub>H</sub> (%)	Sp <sub>H</sub> (%)	Se <sub>T</sub> (%)	Sp <sub>T</sub> (%)
Physical stress	56	77	89	90	51	54	71	91	90	37
Outdoor occupation	50	88	90	90	69	44	89	92	90	88
Other organic dusts	41	2	100	90	93	48	2	100	90	91
Grain dust	37	4	100	90	94	45	4	100	90	92
Contact with animals	36	89	94	90	90	46	84	92	90	46
Cold	30	4	100	90	62	16	2	100	90	62
Diesel fuel and fumes	28	28	99	90	43	2.6	7	100	90	58
Nitrates, nitrites	20	NA	NA	90	77	4.8	NA	NA	90	60
Wood dust	18	83	98	90	84	4.9	87	100	90	98
Other inorganic dusts	13	19	100	90	89	2.9	22	100	90	98
Contact with the public	11	61	99	90	90	21	72	98	90	87
Asbestos	11	NA	NA	90	75	1.6	NA	NA	90	97
Synthetic adhesives	11	61	97	90	66	2.5	20	100	90	83
Natural adhesives	9.3	65	97	90	74	2.1	7	100	90	84
Soot, tar, mineral oils	8.6	48	99	90	64	1.8	26	100	90	90
Organic solvents	7.4	41	100	90	84	3.0	46	100	90	94
Heat	6.7	26	100	90	95	4.3	NA	NA	90	64
Paints and pigments	5.2	48	100	90	79	0.9	66	100	90	98
Welding fumes	3.6	44	100	90	93	0.1	NA	NA	90	100
Antiknock agents	3.4	NA	NA	90	97	0.3	NA	NA	90	100
Polycyclic aromatic hydrocarbons	3.0	27	100	90	82	0.8	NA	NA	90	93
Lead and its compounds	2.6	49	100	90	77	0.5	39	100	90	95
Benzene	2.5	41	99	90	87	1.3	19	100	90	91
Chromium and chromates	2.5	4	100	90	58	0.6	NA	NA	90	90
Degreasing agents	2.3	31	100	90	87	1.5	52	100	90	95
Formaldehyde	2.3	NA	NA	90	81	0.9	NA	NA	90	93
Electromagnetic fields	2.2	32	100	90	88	0.2	NA	NA	90	99
Waxes and polishes	1.4	NA	NA	90	88	1.8	NA	NA	90	85
Cutting fluids	1.4	70	100	90	97	0.1	NA	NA	90	99
Detergents	1.3	21	100	90	92	22	82	84	90	64
Printing inks	1.2	73	100	90	98	0.7	79	100	90	92
Chlorophenols	1.1	NA	NA	90	49	0.5	NA	NA	90	54
Dyestuffs	1.1	28	100	90	94	0.8	NA	NA	90	94
Coal dust	1.1	64	99	90	97	0.1	NA	NA	90	99
Textile dust	1.1	43	100	90	95	15	20	100	90	96
Aromatic amines	1.0	52	99	90	91	0.4	29	100	90	95
Cadmium and its compounds	1.0	17	100	90	93	0.3	NA	NA	90	97
Styrene	1.0	NA	NA	90	92	0.3	NA	NA	90	98
Herbicides	0.9	NA	NA	90	58	0.2	90	100	90	57
Ultraviolet light except sunlight	0.9	NA	NA	90	93	0.8	NA	NA	90	93
Soldering fumes	0.8	NA	NA	90	94	0.1	NA	NA	90	99
Arsenic and its compounds	0.6	NA	NA	90	51	0.7	25	100	90	53
Mercury and its compounds	0.5	NA	NA	90	76	0.1	NA	NA	90	92
Acrylonitrile	0.2	NA	NA	90	98	0.2	NA	NA	90	99
Carbon tetrachloride	0.2	NA	NA	90	85	0.6	79	100	90	94
Epoxy resins	0.2	30	100	90	90	<0.1	NA	NA	90	96
Ionizing radiation	0.2	NA	NA	90	99	0.2	NA	NA	90	98
Beryllium and its compounds	0.1	NA	NA	90	98	0.1	NA	NA	90	99
Polychlorinated biphenyls	0.1	NA	NA	90	93	0.1	NA	NA	90	91
Ethylene oxide	<0.1	NA	NA	90	98	<0.1	NA	NA	90	98

tively powerful in revealing elevated risk. The ability of the matrix to detect the true excess was strongly influenced by the prevalence of exposure and by the way the matrix was used (figures 2 and 3). The misclassification bias between the individual agents of the matrix varied widely, from nearly unbiased to very biased. The misclassification characteristics of the individual agents can be identified from table 2.

The job-exposure matrix indicated the excess risk for agents that were common (prevalence about 10% or more) fairly well, whereas rare agents (prevalence about 1% or less) were difficult or impossible to de-

tect because of the lack of exposed cases with the high approach or because of strong misclassification bias with the total approach. For the moderately common agents (prevalence 1–10%), the high approach indicated an excess (OR\* at least 1.5) for most of the agents when the OR was 2, but the total approach generally failed. When the OR was 5, both approaches indicated an excess, but the high approach was much less biased (OR\* about 4 versus OR\* about 2 with the total approach).

When a study is of the size of ours, the values are statistically significant when the prevalence of exposure

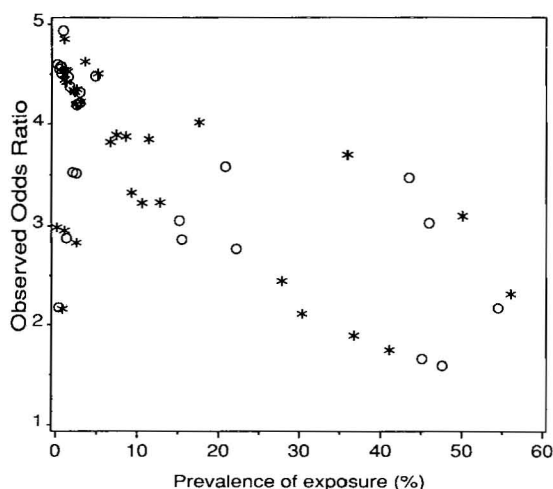
**Table 2.** Observed (biased) odds ratio from the British job-exposure matrix (JEM) when the true odds ratio (OR) was assumed to equal two or five. Statistical significance at the 5% level (two-sided test) of the results is calculated for a study population of 250 cases and 500 referents (all men or all women).<sup>a</sup> (High = only high potential exposure assigned by the JEM considered as exposure, total = all (total) potential exposure assigned by the JEM considered as exposure)

Agent	Men				Women			
	High		Total		High		Total	
	OR = 2	OR = 5	OR = 2	OR = 5	OR = 2	OR = 5	OR = 2	OR = 5
Physical stress	1.5*	2.3*	1.4*	2.1*	1.5*	2.7*	1.3+	1.8*
Outdoor occupation	1.7*	3.1*	1.5*	2.6*	1.7*	3.5*	1.7*	3.3*
Other organic dusts	1.4+	1.8+	1.8*	3.6*	1.3+	1.6+	1.7+	3.4*
Grain dust	1.4+	1.9+	1.8*	3.7*	1.3+	1.7+	1.8*	3.5*
Contact with animals	1.8*	3.7*	1.7*	3.5*	1.7*	3.0*	1.4*	2.0*
Cold	1.5+	2.1*	1.4*	2.4*	1.6+	2.9*	1.3+	2.0*
Diesel fuel and fumes	1.6*	2.4*	1.3+	1.9*	1.9+	4.2+	1.0+	1.2+
Nitrates, nitrites	NA	NA	1.5*	2.7*	NA	NA	1.1+	1.4*
Wood dust	1.8*	4.0*	1.5*	2.9*	1.9*	4.5*	1.7*	3.7*
Other inorganic dusts	1.7+	3.2*	1.5*	3.0*	1.9+	4.2*	1.6+	3.2*
Contact with the public	1.8*	3.9*	1.5*	3.0*	1.8*	3.6*	1.6*	3.2*
Asbestos	NA	NA	1.3+	2.1*	NA	NA	1.3	2.2*
Synthetic adhesives	1.6*	3.2*	1.2+	1.8*	1.7+	3.5*	1.1+	1.5*
Natural adhesives	1.6*	3.3*	1.2+	1.9*	1.7	3.5+	1.1	1.4*
Soot, tar, mineral oils	1.8	3.9*	1.2+	1.7*	1.9	4.4*	1.1	1.5*
Organic solvents	1.8*	3.9*	1.3+	2.2*	1.9+	4.3*	1.3+	2.3*
Heat	1.8+	3.8*	1.6*	3.2*	NA	NA	1.1+	1.4*
Paints and pigments	1.9+	4.5*	1.2+	1.7*	2.0	4.9*	1.3	2.2*
Welding fumes	2.0+	4.6*	1.3+	2.3*	NA	NA	1.2	1.8
Antiknock agents	NA	NA	1.5+	3.0*	NA	NA	1.5	3.0+
Polycyclic aromatic hydrocarbons	1.9+	4.2*	1.1+	1.5*	NA	NA	1.1	1.4+
Lead and its compounds	1.9+	4.3*	1.1+	1.4*	1.9	4.5+	1.1	1.3+
Benzene	1.5+	2.9*	1.2+	1.6*	1.5	2.9+	1.1	1.5*
Chromium and chromates	1.9+	4.2+	1.0+	1.2+	NA	NA	1.0	1.2+
Degreasing agents	1.9+	4.3*	1.1+	1.6*	1.9	4.5*	1.2	1.8*
Formaldehyde	NA	NA	1.1+	1.4*	NA	NA	1.1	1.4+
Electromagnetic fields	1.9	4.3*	1.1	1.6*	NA	NA	1.1	1.4
Waxes and polishes	NA	NA	1.1	1.4+	NA	NA	1.1	1.4*
Cutting fluids	1.9	4.5*	1.3	2.3*	NA	NA	1.1	1.4
Detergents	1.9	4.4+	1.1	1.5*	1.5*	2.8*	1.1+	1.4*
Printing inks	1.9	4.5*	1.4	2.4*	NA	NA	1.1	1.3+
Chlorophenols	NA	NA	1.0	1.1+	NA	NA	1.0	1.0+
Dyestuffs	2.0	4.9*	1.1	1.6*	NA	NA	1.1	1.4+
Coal dust	1.5	3.0*	1.2	1.9*	NA	NA	1.1	1.4
Textile dust	1.9	4.5*	1.2	1.7*	1.7+	3.1*	1.8*	4.0*
Aromatic amines	1.3	2.2+	1.1	1.3+	1.3	2.2+	1.1	1.3+
Cadmium and its compounds	1.9	4.4+	1.1	1.5+	NA	NA	1.1	1.4+
Styrene	NA	NA	1.1	1.4+	NA	NA	1.1	1.4+
Herbicides	NA	NA	1.0	1.1+	1.9	4.6	1.0	1.0
Ultraviolet light except sunlight	NA	NA	1.1	1.4+	NA	NA	1.1	1.4+
Soldering fumes	NA	NA	1.1	1.4+	NA	NA	1.1	1.4
Arsenic and its compounds	NA	NA	1.0	1.0+	1.9	4.5+	1.0	1.0+
Mercury and its compounds	NA	NA	1.0	1.1+	NA	NA	1.0	1.0
Acrylonitrile	NA	NA	1.1	1.4	NA	NA	1.1	1.4
Carbon tetrachloride	NA	NA	1.0	1.0	1.9	4.6*	1.1	1.3+
Epoxy resins	1.5	3.0	1.0	1.0	NA	NA	1.0	1.0
Ionizing radiation	NA	NA	1.1	1.4	NA	NA	1.1	1.4
Beryllium and its compounds	NA	NA	1.0	1.2	NA	NA	1.0	1.2
Polychlorinated biphenyls	NA	NA	1.0	1.0	NA	NA	1.0	1.0
Ethylene oxide	NA	NA	1.0	1.0	NA	NA	1.0	1.1

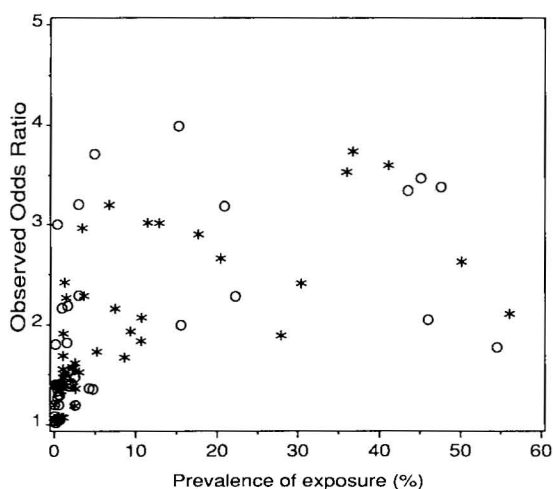
<sup>a</sup> Notations: \* = both true and observed OR statistically significant; + = true OR statistically significant, observed OR not; no notations = neither of OR values statistically significant; NA = high approach not applicable because nobody was classified as exposed by the JEM.

exceeds about 10% when OR = 2, or about 1% when OR = 5. The difference between the high and total approaches in this respect is small, the total approach reaching statistical significance slightly more frequently. When the study is smaller than the example study, the OR\* is not affected, but the confidence interval tends to be wider. Consequently fewer statistically significant results are observed and also the applicability of the high approach may decrease. The situation is the contrary for a study that is larger than the example study.

This assessment of misclassification bias suggests that the British job-exposure matrix is an acceptably valid tool with which to survey risks for certain agents. However, the results should be considered as highly biased if the agent is so rare that no one can be assigned to the category of a high probability and level of exposure with the job-exposure matrix. In the present study, an OR as high as 5 could not be detected by the job-exposure matrix for the following agents: ethylene oxide, polychlorinated biphenyls, beryllium, ionizing radiation, acrylonitrile, mercury, solder



**Figure 2.** Observed odds ratios by prevalence of exposure when the true odds ratio was assumed to be 5 and only "high" potential exposure assigned by the job-exposure matrix was considered as exposure. (stars = men, circles = women)



**Figure 3.** Observed odds ratios by prevalence of exposure when the true odds ratio was assumed to be 5 and all ("total") potential exposure assigned by the job-exposure matrix was considered as exposure. (stars = men, circles = women)

fumes, ultraviolet light, styrene, chlorophenols, waxes/polishes, and formaldehyde. For these agents, the OR\* was, in all cases, < 1.5 because of misclassification or in calculable because of a lack of subjects in the category of a high probability and level of exposure. This result cannot be generalized because, in a larger study or in a population with different prevalences of exposure, the agents not working properly in the analysis with the job-exposure matrix would probably have been different from those listed. In addition, because of the population characteristics of the present study,

**Table 3.** Primary liver cancer and exposure to solvents and welding fumes according to the job-exposure matrix and assessment by a team of occupational hygienists. (OR = odds ratio, 95% CI = 95% confidence interval, High = high probability and level of exposure, Total = all potential exposure, Heavy = at least 10 years of a high level of exposure, Any = any time at any level of exposure)

Agent and exposure	Exposed cases (N)	OR	95% CI
<i>Analysis by the job-exposure matrix</i>			
Organic solvents (chlorinated and nonchlorinated)			
High			
Men	5	0.8	0.3–2.1
Women	4	2.4	0.7–9.0
Both	9	1.1	0.5–2.4
Total			
Men	36	0.9	0.6–1.3
Women	19	2.0	1.1–3.7
Both	55	1.0	0.7–1.5
Degreasing agents (chlorinated hydrocarbons)			
High			
Men	1	—	
Women	3	3.5	0.7–18
Both	4	1.4	0.4–4.6
Total			
Men	20	0.7	0.4–1.1
Women	14	1.6	0.8–3.1
Both	34	0.9	0.6–1.3
Welding fumes			
High			
Men	6	4.2	1.2–14
Total			
Men	18	1.1	0.6–1.9
<i>Assessment by a team of occupational hygienists</i>			
Organic solvents (mainly nonchlorinated hydrocarbons for the men but chlorinated hydrocarbons for the women)			
Any			
Men	7	0.6	0.3–1.3
Women	7	3.4	1.3–8.6
Both	14	1.0	0.6–2.0
Welding fumes			
Heavy			
Men	5	13.5	2.1–88
Any			
Men	6	1.4	0.5–3.5

the job-exposure matrix was generally less powerful for the women than for the men (lower prevalence of exposure for the women).

The assumption that the occupational histories are 90% complete as to the exposures ( $Se_T = 0.9$ ) may seem arbitrary and debatable because the correct figure is probably not 90%, and it is likely to vary by agent. According to our experience, however, the most relevant exposures (high level, long term) are revealed with high probability in these kinds of studies, and the underreporting concerns mainly short-term jobs and

exposures which generally contribute less than the long-term exposures to the risk of chronic diseases. Therefore the figure of 90% may be too high as an estimate of all lifetime exposures, but it is still a reasonable estimate of long-term exposures relevant to the magnitude of the risk. In addition, the misclassification bias is, in most cases, not affected much by  $Se_T$ . The relative weight of specificity as compared with sensitivity can be seen in tables 1 and 2. For example, the  $OR^*$  for dyestuffs is almost unbiased (4.9 versus 5) even though  $Se_H$  is only 28%. The  $OR^*$  of cadmium ( $Se_H = 0.17$ ;  $Sp_H = 1.00$ ;  $OR^* = 4.4$ ) is much less biased than that of aromatic amines ( $Se_H = 0.52$ ;  $Sp_H = 0.99$ ;  $OR^* = 2.2$ ) even though the specificity for cadmium is only 1% higher. The assumption concerning the completeness of the occupational histories is thus not likely to change the conclusions of this study.

The preferable way to use the grading of exposure with this job-exposure matrix is the "high" strategy, with which only a high probability and level of exposure is considered to be exposure. This strategy is likely to work well when the prevalence of exposure is approximately 10% or less, except when there are no heavily exposed cases. However, use of the total approach also provides additional information about the plausibility of the observed risk. If the total and high approaches show an increasing or a flat exposure-response relationship, the result can be considered to be in keeping with occupational etiology. A decreasing exposure-response relationship indicates that the observed risk is probably caused by chance or bias. The job-exposure matrix can also be used as a preliminary survey tool to trigger a more-detailed assessment of exposure, as done in the Finnish study on liver cancer (see figure 1). In that study, an  $OR^*$  of at least 1.5 on the basis of at least five exposed cases and a flat or increasing exposure-response relationship were used as criteria for the reassessment of exposure carried out by occupational hygienists.

The differences in the exposure patterns between the United Kingdom and Finland turned out to be small in most cases. The agreement of the Finnish hygienists with the British job-exposure matrix was good in the category of a high probability and level of exposure. Agreement was worse for some agents, especially in the lower categories of exposure. Examples of such agents are cold, asbestos, chlorophenols, aromatic amines, and arsenic. The poor agreement can be explained in some cases by true differences in exposure, such as the colder climate of Finland (cold). Another reason for the differences is the assessment criteria, that is, the limits between negligible, low, and high probability and those between a low and high level of exposure. It is likely that partially different criteria have been used in the construction of the job-exposure matrix in the United Kingdom as compared with its assessment in Finland. Sometimes it is difficult even to specify the criteria unequivocally because the job-

exposure matrix omits relevant factors influencing exposure. For example, it can be argued that the probability of exposure to asbestos in construction work is both high (in the case of long employment) and low (in the case of short employment). The results of this assessment must therefore be considered as no more than approximate estimates derived from a comparison of the job-exposure matrix with a team of occupational hygienists basing the assessment on the available occupational histories, and on the knowledge of the occurrence and level of exposure in their own country.

The economic structure of the country, the time period considered, and the gender and age distributions of the studied population mainly influence the prevalences of exposure. The bias due to misclassification is therefore different in every study and for every job-exposure matrix. The Finnish population used in the present assessment was a retrospective one whose occupational exposures dated mainly from the 1920s to the 1970s. Over 50% of the work force was employed in agriculture in the 1920s, but since then this percentage has dropped to about 8% (late 1980s). The share of industry has grown slowly from below 20% to about 25% during the same period. The current main branches of industry are metals and metal products (37% of the employees in industry); wood, paper, and publishing (26%); food and beverage (12%); textile and garment (11%); and chemical, rubber, and plastics (8%). According to the statistics, the economic structure of Australia, New Zealand, Japan, the United States, Canada, and many countries in northern, western and middle Europe does not differ very much from that of Finland (15). The misclassification effects of the British job-exposure matrix, if applied to retrospective populations in these countries, would probably be close to those reported in this paper. However, if this job-exposure matrix is used outside the United Kingdom, it is worth checking the exposure codes assigned by the job-exposure matrix for those industrial and occupational categories which are common in the studied population. Incorrect assignments due to national differences in these categories may strongly influence the sensitivity, specificity, and prevalence, and thus also the misclassification bias and the observed odds ratios. This kind of checking is not laborious and could improve the validity of the results by decreasing the misclassification of exposure.

In summary, misclassification and the corresponding bias in odds ratios when the British job-exposure matrix was used depended strongly on two factors, (i) the prevalence of exposure and (ii) the inclusion criterion for exposure. The ability of the matrix to detect excess risk was satisfactory for agents with a prevalence of about 10% or more. If only probable exposure to a high level was considered to be true exposure, the misclassification bias due to the use of the job-exposure matrix was relatively small, and an excess risk was usually detectable also when the prevalence was from

1 to 10%. In large studies, even rarer agents can be surveyed, provided that there are subjects who are exposed to a high level with a high probability. It is therefore recommended that the true exposure be restricted to exposure of high probability and high level. However, supplementing this approach by a separate analysis including, for example, all or all remaining categories of potential exposure could provide additional information on the plausibility of the observed risk. Applying this job-exposure matrix to a population from another country probably does not bias the results seriously, provided that the assignments of the job-exposure matrix for those industries and occupations that are common in the studied population are checked and corrected when necessary.

## Acknowledgments

We are very grateful to Dr B Pannett and Dr D Coggon, Medical Research Council, Environmental Epidemiology Unit, Southampton, United Kingdom, for kindly providing their job-exposure matrix for our study; to Ms R Riala, Dr M Hietanen, and Mr A Anttila, Institute of Occupational Health, Helsinki, for their valuable contribution to the exposure assessment; and to Ms T Kaustia for her linguistic revision of the manuscript.

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Received for publication: 7 May 1991