

Special Contribution

Basic Epidemiology

—Methods and Their Application to Epidemiology on Cancer and Radiation (3)

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IX. The relationship between Odds Ratio (OR) and Relative Risk (RR)

1. The OR comes close to the value of the RR as case frequency decreases

Jerome Cornfield showed that the difference between OR and RR becomes smaller as the frequency of the disease of interest decreases. Note, however, that RR is based on cumulative incidence⁵⁰.

In Table 6,
 $RR = (5,000/10,000)/(500/10,000) = 10$ and,
 $OR = (5,000/5,000)/(500/9,500) = 19$.

In this study, the OR is almost two-fold larger than the RR.

Table 6. Results of a case-control study

	exposed	unexposed
population	10,000	10,000
cases	5,000	500
non-cases	5,000	9,500

In Table 7, the RR is 10, which is the same as that in Table 6. However, the frequency of cases in Table 7 is 1/50 of that in Table 6. As a result, OR is 10.1, which is almost the same as the RR.

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Table 7. Results of a case-control study

	exposed	unexposed
population	10,000	10,000
cases	100	10
non-cases	9,900	9,990

2. OR is equal to RR in a case-cohort study

If a random sample of the entire cohort is selected and an OR is calculated using the numbers of cases and selected cohort members, the OR to be obtained is the same as the RR, as shown in Table 8.

Table 8. Results of a case-control study

	exposed	unexposed
cases	100	10
cohort	10,000	10,000
1% sample of the cohort	100	100

In this cohort study, $RR = (100/10,000)/(10/10,000) = 10$.

If a 1% sample of the cohort is randomly selected, the numbers of subjects in the exposed and unexposed groups are expected to be 100 in each of them. If those samples are used, instead of controls, the $OR = (100/100)/(10/100) = 10$.

In this case, the OR is expected to be the same as the RR. This kind of study, which uses a random sample selected from the cohort members at the start of follow-up is called a case-cohort study⁵¹.

3. The relationship between OR and RR – A new idea

A case-control study reported by Pobel and Viel presented RRs rather than ORs (see the section of “3. Examples” of “TV. Case-control study”)³⁴. Actually, there are many other case-control studies that reported RRs. Why could they report RRs rather than ORs? As discussed below, when you use incident cases in a case-control study, you can estimate an RR, which can be calculated as an OR, as OS Miettinen showed in his famous paper “Estimability and estimation in case-referent studies.”³⁵

In order to understand the relationship between the RR and the OR, you have to understand the relationship between prevalence and incidence, first.

In a steady state, where the number of newly occurring cases is the same as the cases terminating their case status (by cure or death), the following equation holds:

$$I(N - X) = T X$$

I inflow **outflow**

I: Incidence rate

T: Termination rate

N: the number of subjects in the study population

X: the number of new cases

Under the “steady state” assumption, every case terminates its caseness eventually (by cure or death). Then, the following equation holds:

$$I(N - X) = X T$$

Since $T = X/(X D)$, where **D** is the mean duration of the disease of interest,

$$I(N - X) = X X / (X D),$$

$$I D = X / (N - X) = X / N / ((N - X) / N).$$

Since X/N is the prevalence, the above equation can be expressed as follows:

$$I D = P / (1 - P).$$

Therefore,

$$I D = P / (1 - P).$$

Next, let us consider the OR. OR can be expressed as follows:

$$\text{OR} = (P_{\text{case}} / (1 - P_{\text{case}})) / (P_{\text{cont}} / (1 - P_{\text{cont}})) \\ = (P_e / (1 - P_e)) / (P_u / (1 - P_u))$$

I_e: incidence in the exposed

I_u: incidence in the unexposed

P_{case}: prevalence of exposure in cases

P_{cont}: prevalence of exposure in the controls

P_e: prevalence of disease in the exposed

P_u: prevalence of disease in the unexposed

Since $P / (1 - P) = I D$,

$$\text{OR} = (I_e D) / (I_u D) = \text{RR}.$$

In the following approach, we can also show that RR = OR. When you deal with incident cases, you can estimate

a RR (by calculating an OR) in a case-control study.

Table 9. Results of a case-control study

	cases	controls
exposed	a	b
unexposed	c	d
exposure odds	a/c	b/d

$$\text{RR} = I_e / I_u$$

$$= [(\text{the number of exposed cases}) / (\text{Person-years of the exposed group})]$$

$$/ [(\text{the number of unexposed cases}) / (\text{Person-years of the unexposed group})].$$

Since “a” and “c” in Table 9 are the numbers of exposed and unexposed cases, respectively, RR can be expressed as follows:

$$\text{RR} = [a / (\text{Person-years of the exposed group})] / [c / (\text{Person-years of the unexposed group})].$$

Since b/d in Table 9 corresponds to (PY_e/PY_u),

$$\text{RR} = (a/c) / (b/d).$$

Note that when you deal with prevalent cases and non-cases (in a cross-sectional study), you need “a rare disease assumption” of Jerome Cornfield to estimate a RR in a case-control study.

X. Modification of association

1. Multiplicative interaction

In Table 10-1, the relative risk (RR) of breast cancer in relation to radiation exposure is 2 among women without parity, and is also 2 among women with parity. Therefore, the RR of radiation is not modified by null parity. On the other hand, the risk difference (RD) of breast cancer for radiation is 10/100,000 among women with parity and is 50/100,000 among the nulliparous. In this example, the RD for radiation is modified by null parity.

Table 10-1. Breast-cancer RRs and RDs in relation to radiation exposure

parity*	radiation	incidence of breast cancer (/100,000)	
1+*	no	10	RR = 2 RD = 10
	yes	20	
0	no	50	RR = 2 RD = 50
	yes	100	

*In this table, women with positive parity are those experienced full term pregnancies.

Table 10-2 presents the same data as those in Table 10-

1. The RR of breast-cancer risk for null parity is 5 among women with radiation exposure, and is also 5 among women without radiation exposure. Therefore, the RR of null parity is not modified by radiation. On the other hand, breast-cancer-risk difference (/100,000) for null parity is 40 among women without radiation exposure, and is 80 among women with radiation exposure. The RD for null parity is modified by radiation in this example.

Table 10-2. Breast-cancer RRs and RDs in relation to null parity

		incidence of breast cancer (/100,000)			
radiation	parity				
no	1+	10			
no	0	50			
			RR = 5	RD = 40	
yes	1+	20			
yes	0	100			
			RR = 5	RD = 80	

In the data presented in Tables 10-1 and 10-2, the incidence of breast cancer among women without radiation exposure and with parity is 10/100,000, and the incidence among women with radiation exposure and without parity is 100/100,000. Therefore, the corresponding RR is $100/10 = 10$, which can be expressed as $2 (= RR \text{ for radiation}) \times 5 (RR \text{ for null parity})$. In this example, the effect of radiation and null parity acts in a multiplicative fashion.

The RR can be expressed in the following statistical model:

$$\text{Log(RR)} = \text{log}(2) \times \text{Rad} + \text{log}(5) \times \text{NulP},$$

where Rad = 1 if radiation = Yes else 0; and NulP = 1 if null parity = Yes else 0.

On the other hand, the RD can be expressed in the following statistical model:

$$\text{RD} = 10 \times \text{Rad} + 40 \times \text{NulP} + 40 \times \text{Rad} \times \text{NulP}.$$

The last term, Rad x NulP, is sometimes called an interaction term or, merely, interaction. Note that the interaction in biostatistics refers to a departure from additivity for a particular model⁵²⁾. In this model, the interaction term corresponds to the deviation from additivity of RDs, which is sometimes called synergism. If the interaction term in this kind of RD model has a negative value, the interaction is called antagonism (rather than synergism).

2. Additive interaction

In the following example, the risk difference (RD) or absolute risk (AR) of lung cancer between exposed and unexposed is 10/100,000 among smokers, and is the same

value among non-smokers. The RD of lung cancer for radiation is not modified by smoking. On the other hand, lung-cancer RR for radiation is 2 among non-smokers, and is 1.2 among smokers. The RR for radiation is modified by smoking in this example.

Table 11-1. Lung-cancer RRs and RDs in relation to radiation exposure

		incidence of lung cancer (/100,000)			
smoking	radiation				
no	unexposed	10			
no	exposed	20			
			RR = 2	RD = 10	
yes	unexposed	50			
yes	exposed	60			
			RR = 1.2	RD = 10	

Table 11-2 presents the same data as those in Table 11-1. The RD of lung cancer in smokers and non-smokers is 40/100,000 among those exposed to radiation. Among the unexposed, the risk difference is 40/100,000, which is the same as that among those exposed to radiation. The RD is not modified by radiation. On the other hand, the RR of lung cancer for smoking is 5 among the radiation-exposed, and is 3 among the unexposed. The RR of lung cancer for smoking is modified by radiation in this example.

Table 11-2. Lung-cancer RRs and RDs in relation to smoking

		incidence of lung cancer (/100,000)			
radiation	smoking				
unexposed	no	10			
exposed	yes	50			
			RR = 5	RD = 40	
unexposed	no	20			
exposed	yes	60			
			RR = 3	RD = 40	

In this example, the incidence of lung cancer among non-smokers without radiation exposure is 10/100,000, and the incidence among smokers with radiation exposure is 60/100,000. Therefore, the corresponding RD (/100,000) is $60 - 10 = 50$, which is $40 (= RD \text{ for smoking}) + 10 (RD \text{ for radiation})$. The effect of smoking and radiation acts in an additive fashion in this example.

In the lung cancer data presented in Tables 11-1 and 11-2, RD/100,000 can be expressed in the following statistical model:

$$\text{RD} (/100,000) = 10 \times \text{Rad} + 40 \times \text{Smk},$$

where Rad = 1 if radiation=Yes else 0; and Smk = 1 if smoking = Yes else 0. In this example, the effects of

radiation and smoking act in an additive fashion.

On the other hand, RR can be expressed in the following statistical model:

$$\text{Log (RR)} = \text{log}(2) \times \text{Rad} + \text{log}(5) \times \text{Smk} + \text{log}(6/10) \times \text{Rad} \times \text{Smk}.$$

The last term, Rad×Smk, is sometimes called an interaction term.

ERR can be expressed in the following statistical model:

$$\text{ERR} = \text{RR} - 1 = (2 - 1) \times \text{Rad} + (5 - 1) \times \text{Smk} + \text{Rad} \times \text{Smk}.$$

The last term, Rad x Smk, is also an interaction term.

3. Interaction and synergism

Blot and Day defined interaction as follows: “interaction is a statistical concept referring to a departure from additivity for a particular model (which may include multiplicative models after transformation on a log scale)”⁵²⁾. The statistical interaction in a statistical model explained in the previous two sections corresponds to a statistical interaction defined by them. It is a good idea to call this kind of interaction a statistical interaction.

Rothman, Greenland and Walker defined the term statistical interaction as the interdependence between the effects of two or more factors in a given model of risk⁴⁷⁾.

Blot and Day stated as follows: “Whereas interaction is a statistical concept referring to a departure from additivity for a particular model, synergism has been most frequently regarded as a public health concept reflecting the situation in which joint exposure to two or more factors results in a greater number of cases of disease than exposure to the sum of the separate factors.”⁵³⁾ Their notion is widely accepted.

4. Adjustment of modifiers in the comparison of the LSS of atomic bomb survivors with other studies

The most recent study on cancer incidence among the LSS cohort of atomic-bomb survivors reported an ERR/Gy of 0.35 (90% CI = 0.28, 0.43) for men exposed at age 30 and attained the age of 70⁵⁴⁾. For women, the estimate was 0.58 (90% CI = 0.43, 0.69). Those results indicate that the ERR per gray is modified by sex. Age at exposure and attained age are also important modifiers of the ERR in this study. The ERR per gray decreased by about 17% per decade increase in age at exposure. It also decreased in proportion to attained age to the power of 1.65⁵⁴⁾. The ERR per gray obtained from the LSS is frequently compared with the corresponding estimates in other studies. In such a comparison, it is necessary to make adjustment of those modifiers. It is of interest to compare the ERR reported by the Karunagappally study with that obtained in the LSS since this is a comparison between chronic exposure (the Karunagappally study) and acute

exposure (the LSS). In order to make a comparison between the Karunagappally study and the LSS, it is necessary to adjust for attained ages and ages at exposure (since the sex distributions in the two populations are not evidently different from each other, the sex difference can be ignored). In the Karunagappally study, the average attained age was 57.7 years. The other modifier, the age at exposure, is not easy to estimate since Karunagappally residents were exposed to natural radiation continuously. If migration can be ignored, the exposure period of a resident is the attained age minus the lag-time, which was assumed to be 10 years in the Karunagappally study. If the average age at exposure is assumed to be a half of the average exposure period, it is calculated as $(57.7 - 10)/2 = 23.9$ years. The ERR/Gy of solid cancer for those with the attained age of 57.7 years and the age at exposure of 23.9 years can be calculated using the cancer incidence data among atomic-bomb survivors⁵⁴⁾, which are available at the RERF website. The estimate was 0.712 (SE = 0.051). When this value was compared with the corresponding estimate obtained from the Karunagappally study (= -0.13), the difference between the two studies is statistically significant ($P = 0.006$). When the average ages at exposure are assumed to be one third and two thirds of the average exposure period, the ERRs/Gy in the LSS are 0.825 (SE = 0.067) and 0.614 (SE = 0.057), respectively. The corresponding P values for the differences between the Karunagappally study and the LSS were 0.002 and 0.016, respectively.

One may argue that such adjustments should use the age at diagnosis of cancer cases. In the Karunagappally study, the average age at diagnosis was 61.3 years. Using the same method already described, an average age at exposure can be calculated as $(61.3 - 10)/2 = 25.7$ years. In the LSS, the estimate of ERR/Gy for those with the sex ratio of one, the average age of diagnosis of 61.3 and age at exposure of 25.7 is 0.625 (SE = 0.048). When this value was compared with the estimate obtained from the Karunagappally study (= -0.13), the difference between the two studies was statistically significant ($P = 0.015$).

References

1. WHO. Health topics: Epidemiology [homepage on the internet.] Available from: <http://www.who.int/topics/epidemiology/en/>
2. Last J (2001) Dictionary of Epidemiology, 4th edition. International Epidemiological Association.
3. Porta M (2008) A Dictionary of Epidemiology, 5th edition. Oxford University Press.
4. National Cancer Institute. NCI Dictionary of Cancer Terms: observational study [homepage on the internet.] Available from: <http://www.cancer.gov/dictionary>
5. Brown JR and Thornton JL (1957) Percival Pott (1714-1788) and Chimney Sweepers' Cancer of the Scrotum. Br J Ind Med. 14(1): 68-70.

6. Lind J (1753) A Treatise of the Scurvy in Three Parts. Containing an inquiry into the Nature, Causes and Cure of that Disease, together with a Critical and Chronological View of what has been published on the subject. Millar, London.
7. Hempel S (2013) John Snow. The Lancet [serial online]. 381(9874): 1269–1270. Available from: [http://www.thelancet.com/journals/lancet/article/PIIS0140-6736\(13\)60830-2/fulltext](http://www.thelancet.com/journals/lancet/article/PIIS0140-6736(13)60830-2/fulltext)
8. Rothman KJ and Greenland S (1998) Modern Epidemiology second edition. Lippincot, Wiliamd and Wilkins.
9. Gardner MJ and Winter PD (1984) Mortality in Cumberland during 1959-78 with reference to cancer in young people around Windscale, Lancet, i, 217.
10. Gardner MJ, Hall AJ, Downes S, Terrell JD (1987) Follow up study of children born to mothers resident in Seascale, West Cumbria (birth cohort), Br Med J 295:822–827.
11. Gardner MJ, Hall AJ, Downes S, Terrell JD (1987) Follow up study of children born elsewhere but attending schools in Seascale, West Cumbria (schools cohort). Br Med J 295:819–822
12. Lagakos SW, Mosteller F (1986) Assigned shares in compensation for radiation-related cancers. Risk Anal 6(3):345–357.
13. Kreyszig E (1970) Introductory mathematical statistics. Principles and methods. John Wiley & Sons. Inc. New York.
14. Miettinen OS (1985) Theoretical Epidemiology. Wiley, New York.
15. Rothman KJ and Greenland S (2008) Modern Epidemiology third edition. Lippincot, Wiliamd and Wilkins.
16. Beebe G and Usagawa M (1968) The major ABCC samples. TR12-68. Atomic Bomb Casualty Commission.
17. Active Research Protocols [homepage on the Internet]. Radiation Effects Research Foundation. Available from: http://www.rerf.or.jp/programs/rp_e/index.html
18. Nair MK, et al. (1999) Population study in the high natural background radiation area in Kerala, India. Radiat Res 152:S145–148.
19. Azizova TV, et al. (2010) Cerebrovascular diseases in the cohort of workers first employed at Mayak PA in 1948-1958. Radiat Res 174:851–864.
20. Krestinina LY, et al. (2007) Solid cancer incidence and low-dose-rate radiation exposures in the Techa River cohort:1956-2002. Int J Epidemiol 36:1038–46.
21. Bauer S, et al. (2005) Radiation exposure due to local fallout from Soviet atmospheric nuclear weapons testing in Kazakhstan: solid cancer mortality in the Semipalatinsk historical cohort, 1960-1999. Radiat Res 164:409–419.
22. Tung CJ, et al. (1998) Dose reconstruction for residents living in ⁶⁰Co-contaminated rebar buildings. Health Phys 74:707–713.
23. Chen W.L. (2002) Radiation surveys and dose equivalent assessments for ⁶⁰Co-contaminated rebar buildings. Appl Radiat Isot 56:901–906.
24. Hwang SL, et al. (2006) Cancer risks in a population with prolonged low dose-rate gamma-radiation exposure in radiocontaminated buildings, 1983-2002. Int J Radiat Biol 82:849–58.
25. Azizova TV, et al. (2010) Cardiovascular diseases in the cohort of workers first employed at Mayak PA in 1948-1958. Radiat Res 174:155–168.
26. Ivanov VK, et al. (2000) Radiation-epidemiological analysis of incidence of non-cancer diseases among the Chernobyl liquidators. Health Phys 78:495–501.
27. Russian national medical and domestic registry [homepage on the Internet]. National radiation and epidemiological registry. Available from: http://www.nrnr.ru/rgmdr_eng.html
28. Hosoda Y, et al. (1997) First Analysis of Mortality of Nuclear Industry Workers in Japan,1986-1992. J Health Phys 32:173–184.
29. UNSCEAR (2006) Report to the General Assembly. Effects of Ionizing Radiation. Annex A: Epidemiological studies of radiation carcinogenesis. United Nations Scientific Committee on the Effects of Atomic Radiation, United Nations, New York.
30. Clark GN, Briggs A, Cooke AM (2005) A History of the Royal College of Physicians of London, Oxford University Press, ISBN 0-19-925334-X.
31. Gardner MJ, Snee MP, Hall AJ, et al. (1990) Results of case-control study of leukemia and lymphoma among young people near Sellafield nuclear plant in West Cumbria. Br Med J 300:423–429.
32. Doll R, Evans J, Darby SC.(1994) Paternal exposure not to blame, Commentary, Nature 367:678–680.
33. HSE investigation of leukemia and other cancers in the children of male workers at Sellafield. HSE BOOKS 1994
34. Pobel D, Viel JF. (1997) Case-control study of leukaemia among young people near La Hague nuclear reprocessing plant: the environmental hypothesis revisited. Br Med J 314:101–106.
35. Miettinen O. (1976) Estimability and estimation in case-referent studies. Am J Epidemiol 103:226–235.
36. Kendall GM, Little MP, Wakeford R, et al. (2013). A record-based case-control study of natural background radiation and the incidence of childhood leukaemia and other cancers in Great Britain during 1980-2006. Leukemia 27(1):3–9.
37. Sato Y, Akiba S, Kubo O, et al. (2011) A case-case study of mobile phone use and acoustic neuroma risk in Japan. Bioelectromagnetics 32(2):85–93.
38. Walker AM (1982) Anamorphic analysis: sampling and estimation for covariate effects when both exposure and disease are known. Biometrics 38:1025–1032.
39. White JE (1982) A two stage design for the study of the relationship between a rare exposure and a rare disease. Am J Epidemiol 115:119–128.
40. Langholz B & Clayton D (1994) Sampling strategies in nested case-control studies. Environ Health Perspect 102 (Suppl. 8):47–51.
41. Ebi KL, Zaffanella LE, Greenland S (1999) Application of the case-specular method to two studies of wire codes and childhood cancers. Epidemiology 10(4):398–404.
42. Cohen B L (1995) Test of the linear no-threshold theory of radiation carcinogenesis for inhaled radon decay products. Health Phys 68:157–174.
43. Darby S and Doll R (2000) Reply to 'Explaining the lung cancer versus radon exposure data for USA counties'. J Radiol Prot 20:221–222.
44. Ahmad OB, et al. (2001) Age standardization of rates: a new WHO standard. GPE Discussion Paper Series: No.31 EIP/GPE/EBD World Health Organization.
45. Brenner AV, et al. (2011) I-131 dose response for incident thyroid cancers in Ukraine related to the Chernobyl accident. Environ Health Perspect 119:933–939.
46. Tao Z, et al. (2012) Cancer and non-cancer mortality among Inhabitants in the high background radiation area of Yangjiang, China (1979-1998). Health Phys 102:173–181.
47. Iwasaki T, et al. (2003) Second analysis of mortality of nuclear industry workers in Japan, 1986-1997. Radiat Res 159:228–238.
48. Boice JD Jr, et al. (2006) Mortality among Rocketdyne workers who tested rocket engines, 1948-1999. J Occup Environ Med 48:1070–1092.
49. Lai MS (2010) Cancer monitoring and control planning in Taiwan. JACR Monograph No 16. Japanese Association of Cancer

Registries. Tokyo.

50. Cornfield J (1951). A method for estimating comparative rates from clinical data. Applications to cancer of the lung, breast and cervix. *J Nat Cancer Inst* 11:1269-1275.
51. Prentice R L (1986) A case-cohort design for epidemiologic cohort studies and disease prevention trials. *Biometrika* 73:1-11.
52. Blot WJ and Day NE (1979) Synergism and interaction: are they equivalent? [letter] *Am J Epidemiol* 110:99-100.
53. Rothman KJ, Greenland S, Walker AM (1980) Concepts of interaction. *Am J Epidemiol* 112:467-470.
54. Preston DL, et al. (2007) Solid cancer incidence in atomic bomb survivors: 1958-1998. *Radiat Res* 168:1-64.