

Prospective/Retrospective Studies

In many studies, the population can be categorized according to two binary variables:

$$\underline{\underline{D}} = \begin{cases} 1 : & \text{presence of a disease (e.g., lung - cancer);} \\ 0 : & \text{absence of a disease.} \end{cases}$$

$$\underline{\underline{X}} = \begin{cases} 1 : & \text{exposure to a certain toxin (e.g., smoking);} \\ 0 : & \text{non - exposure to a certain toxin.} \end{cases}$$

		<i>D</i>	
		1	0
<i>X</i>	1	Y_{11}	Y_{10}
	0	Y_{01}	Y_{00}

There exist two sampling schemes to obtain data for the study,
Prospective and Retrospective.

Observational Study Designs: Case Control vs Cohort

Exposure

Disease

Wu, J.
Brainfarts



Can't use RR, can only use OR because researcher sets the prevalence within the study. Good for rare diseases. In rare diseases, OR approximates RR. In non-rare diseases, the direction of OR and RR are the same, but the actual number obtained for OR and RR are different. You CANNOT obtain a RR for this. It makes no sense to.



Case Control

retro

Exposure

Disease



RR and OR are both relevant for this. This is sometimes used to test out a new intervention/treatment.



Prospective Cohort



prop

Prospective sampling:

disease

		D		
		1	0	
<i>exposure</i> X	1	Y_1	$n_1 - Y_1$	m_1
	0	Y_0	$n_0 - Y_0$	m_0

- ▶ an exposed group is selected together with a non-exposure group.
- ▶ both groups are monitored over a prolonged period to compare the incidence of diseases in the two groups
- ▶ row totals are fixed, column totals are random.
- ▶ Example: randomized clinical trials

- Model: disease status is response; exposure (and other covariates) are predictors

- $Y_1 \sim \text{Bin}(m_1, \pi_1)$ $\leftarrow P(d=1 | x=1)$

- $Y_0 \sim \text{Bin}(m_0, \pi_0)$ $\leftarrow P(d=1 | x=0)$

- Logistic model:

$$\text{logit}(\pi_i) = \log \frac{\pi_i}{1 - \pi_i} = \beta_0 + \beta_1 X_i, \quad (X_i = 0, 1)$$

$$\beta_0 = \log \frac{\pi_0}{1 - \pi_0}, \quad \beta_1 = \log \frac{\pi_1 / (1 - \pi_1)}{\pi_0 / (1 - \pi_0)}$$

- We are interested in relative risk (RR) π_1 / π_0 and odds ratio (OR)

$$\frac{\pi_1 / (1 - \pi_1)}{\pi_0 / (1 - \pi_0)}$$

Retrospective sampling:

disease

		D	
		1	0
<i>exposure</i> X	1	Z_1	Z_0
	0		
		n_1	n_0

Covariate: D
Response: X

- ▶ Suitable when disease incidence rate is low and/or disease takes time to occur.
- ▶ Two groups of subjects are selected, one with disease, the other without disease (e.g., case-control study).
- ▶ Exposure information is obtained retrospectively.
- ▶ The difference in exposure history between cases and controls reveals disease-exposure relation.

Two ways to model retrospective data:

- ▶ Retrospective model: exposure history as response ((Z_1, n_1) and (Z_0, n_0)), disease status as predictor

$$Z_1 \sim \text{Bin}(n_1, \rho_1), \quad Z_0 \sim \text{Bin}(n_0, \rho_0)$$

Handwritten notes: $P(X=1|D=1)$ above ρ_1 ; $P(X=1|D=0)$ above ρ_0

$$\log \frac{\rho_i}{1 - \rho_i} = \alpha_0 + \alpha_1 D_i \quad (D_i = 0, 1)$$

$$\alpha_0 = \log \frac{\rho_0}{1 - \rho_0}, \quad \alpha_1 = \log \frac{\rho_1 / (1 - \rho_1)}{\rho_0 / (1 - \rho_0)}$$

- ▶ Can estimate $\mathbb{P}(\text{Exposure} | \text{Disease})$
- ▶ Cannot estimate $\mathbb{P}(\text{Disease} | \text{Exposure})$
- ▶ α_0 is the log odds of exposure given no disease (in the control group)
- ▶ α_1 is the log odds ratio of exposure between case and control groups



Invariance: The odds ratio of exposure in case vs control is equal to the odds ratio of disease in the exposed group vs the non-exposed group (Bayes rule)

$$\begin{aligned}\underline{OR} &= \frac{\text{odds}(\text{exposure}|\text{case})}{\text{odds}(\text{exposure}|\text{control})} \\ &= \frac{\text{odds}(\text{disease}|\text{exposure})}{\text{odds}(\text{disease}|\text{non-exposure})}\end{aligned}$$

- ▶ But retrospective data CANNOT be used to evaluate the relative risk of disease in the exposed group vs the non-exposed group.

$$RR = \frac{\mathbb{P}(\text{disease}|\text{exposure})}{\mathbb{P}(\text{disease}|\text{non-exposure})}$$

- ▶ Exception: for rare disease ($\mathbb{P}(\text{disease}) \approx 0$), $RR \approx OR$.

$$RR = \frac{\pi_1}{\pi_0} \quad OR = \frac{\pi_1(1-\pi_1)}{\pi_0(1-\pi_0)} \quad \leftarrow \begin{matrix} \pi_1 \approx 0 \\ \pi_0 \approx 0 \end{matrix}$$

	D		
	0	1	
X = 0	z_0	z_1	$m_0 \hat{=} z_0 + z_1$
X = 1	$n_0 - z_0$	$n_1 - z_1$	$m_1 \hat{=} n_0 + n_1 - z_0 - z_1$
	n_0	n_1	

$$\begin{array}{l} (z_0, m_0) \quad X=0 \\ (n_0 - z_0, m_1) \quad X=1 \end{array}$$

Pr of D given X & Selected

- Prospective model: despite the fact that data are collected retrospectively, we treat it as a prospective study.
- Treat disease status (D) as response, and exposure (X) as predictor (as in the prospective study), and fit a logistic model.
- Based on a variant of the invariance theory, the coefficient for X is the same as in the retrospective model, i.e., the desired log odds ratio!

Multiple Covariates

- ▶ We are interested in how the change of \mathbf{X} affects $\mathbb{P}(D)$

- ▶ Assume

$$\mathbb{P}(\underline{D = 1} | \underline{\mathbf{X}}) = \frac{\exp(\alpha + \mathbf{X}\beta)}{1 + \exp(\alpha + \mathbf{X}\beta)}$$

Handwritten notes: "Selected" with an arrow pointing to \mathbf{X} ; α and β are circled in blue.

- ▶ If prospective data are collected, we can fit a logistic regression model and estimate α and β .

- ▶ If we only have retrospective data, we can still use a prospective model.
- ▶ Essential assumption: the selection criteria of the case-control study is independent of covariates
- ▶ Using the prospective model on the retrospective data, we are really modeling $\mathbb{P}(D = 1|\mathbf{X}, S = 1)$
- ▶ From Bayes theorem, we can derive

$$\underline{\mathbb{P}(D = 1|\mathbf{X}, S = 1)} = \underline{\frac{\exp(\alpha^* + \mathbf{X}\beta)}{1 + \exp(\alpha^* + \mathbf{X}\beta)}}$$

- ▶ Thus we can estimate β , but not α (α^* is a nuisance parameter)
- ▶ $\exp(\beta)$ provides the odds ratios of disease corresponding to unit change in different covariates.

Recap

- ▶ Data may be collected prospectively or retrospectively.
- ▶ We can¹ always fit a prospective model to retrospective data by treating disease status as response, and having multiple covariates.
- ▶ OR is invariant.
- ▶ In general, RR can only be evaluated for prospective data.
- ▶ For rare diseases, $RR \approx OR$.

¹Sampling procedure must be independent of covariates.