Applied Epidemiology I: Tables and interpreting results

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February 4, 2021

Acknowledgements

This course material is based on my learning from Anastasia Lam's teachings in last year's Applied Epidemiology I lab sessions, and readings from *Epidemiology* by Gordis [1], *A First Course in Probability and Statistics* by Goldsman and Goldsman [2], *Principles of Biostatistics* by Pagano and Gauvreau [3], and *Biostatistics I* by Gabriel and Frumento [4]. I especially want to thank Marlene Stratmann for reviewing the slides and Prof. Paul Dickman for providing me with suggestions to improving the teaching.

Outline

- **1** How to deal with missing values?
- 2 Good epidemiologists
- Tables

Bad example
Basics of making tables
One-way tables
Two by two tables
Stata tool for Epidemiology

- A Basic Epidemiology terms Rate vs. proportion Risk, risk difference, risk ratio Odds. odds ratio
- Interpreting results
 Principles
 Ratio > or < 1</p>
 More examples
 Avoid dehumanising language
- 6 Calculate ratios using Stata Risk ratio Odds ratio Incidence rate ratio

use

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 - Missing completely at random
 - Missing not at random

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 - Missing completely at random
 - Missing not at random
- I don't know.

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 - (2) with the last measured value. [5]
 - For categorical/binary variables, you might assume the values have the same distribution as those non-missing. But...
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 - Sensitivity analysis. [5]
 - Be aware of what you are doing !
 - Whatever you do is merely assumption to your data.
 - You have to
 - 1. illustrate your assumption in the methods part
 - 2. explain the influence of the assumption in the discussion part
 - 3. make a flow chart to say what you kicked out (recommended).

Good epidemiologists

Migeal A Hernán ^a, "Epidemiologists are trained to generate, integrate, analyse, interpret data."

"And description is the foundation."

^aProfessor at Harvard University T.H. Chan Public Health School

Tables: Bad example

What is the problem here?

Table 5
Simulation results for using full data, CRs only, and proposed method under four missing mechanisms

	Bias ^a		$Variance^{b}$		95% CI°	
Method	$(\hat{\beta}_W)$	(\hat{eta}_X)	(\hat{eta}_W)	(\hat{eta}_X)	(\hat{eta}_W)	$(\hat{\beta}_X)$
		(M.1) P(R	= 1) = 0	0.66		
Full	0.01346	0.02229	0.04008	0.03685	0.955	0.950
Comp	0.03062	-0.003561	0.1149	0.06732	0.960	0.955
Impu	0.01431	0.021	0.04088	0.05169	0.980	0.975
	(I)	A.2) logit P	R(R=1)	= 2Y		
Full	0.007908	-0.02116	0.03838	0.03624	0.975	0.925
Comp	0.01945	0.07096	0.107	0.06581	0.960	0.950
Impu	0.006966	0.01597	0.04227	0.05226	0.975	0.985
	(I)	A.3) logit P	(R=1)	=2X		
Full	0.007908	-0.02116	0.03838	0.03624	0.975	0.925
Comp	0.01225	0.0589	0.08856	0.06818	0.980	0.975
Impu	0.009563	-0.04699	0.03865	0.04923	0.985	0.970
	(M.	4) logit $P(I$	R = 1) =	X + Y		
Full	0.01346	0.02229	0.04008	0.03685	0.955	0.950
Comp	0.02404	1.613	0.1102	0.08202	0.955	0.580
Impu	0.01814	0.08289	0.0578	0.06075	0.955	0.970

 $^{{}^{8}\}text{Bias} = (\hat{\beta} - \beta_{0})/\beta_{0}.$

^bSimulation variance.

^cConfidence interval using jackknife standard error.

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 - (Graph headings are placed below.)

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Table 1: Baseline characteristics of colon cancer patients diagnosed during 1981-1990, Sweden.

		< 50	50-59	60-69	70-79	≥80	All
Colon cancer							
1981-1990	Patient size (n)	1 148	2 485	6 227	9 381	5 442	24 683
	Female (%)	51.74	52.43	50.09	51.61	59.28	53.01
	Proportion of censoring ¹ (%)	34.41	15.69	3.12	0.22	0.09	4.07

We use cancer data still.
 sysuse cancer, clear
 keep if drug == 1 | drug == 2

- We use cancer data still.
 sysuse cancer, clear
 keep if drug == 1 | drug == 2
- One-way table of frequencies with mean and sd of age
 table died, contents(freq mean age sd age)

1 if patient died		Freq.	mean(age)	sd(age)
	0 1	9 25	55.1111 56.88	5.487359 6.227091

- One-way table of frequencies
 - . tabulate died

1 if patient			
died	Freq.	Percent	Cum.
0	9	26.47	26.47
1	25	73.53	100.00
Total	34	100.00	

- Create table I of baseline characteristics using table1_mc
- This command is useful. Play it on your own!
- See help table1_mc
- ssc install table1_mc, replace
- table1_mc, vars(age conts)

Data are presented as median (IQR).

 2 by 2 table for drug and died with relative frequency by column or row

. tabulate died drug, col row

1 : £ |

1 11	I		
patient	Drug type	(1=placebo)	
died	1 0	1	Total
	+		+
0	1 8	1	9
	88.89	11.11	100.00
	57.14	5.00	26.47
	+		+
1	1 6	19	J 25
	1 24.00	76.00	100.00
	42.86	95.00	73.53
	+		+
Total	l 14	20	l 34
	41.18	58.82	100.00
	100.00	100.00	100.00

 2 by 2 table with chi-square test and fisher's exact test

```
. tabulate died drug, col row chi2 exact
     1 if |
  patient | Drug type (1=placebo)
                                  Total
              88.89 11.11 |
                               100.00
              57.14
                       5.00 l
       1 I
                        19 I
              24.00 76.00 I
                               100.00
              42.86
                       95.00 I
                                 73.53
   Total |
              14 20 |
                                     34
              41.18 58.82 I
                              100.00
             100.00 100.00 |
                               100.00
        Pearson chi2(1) = 11.5039 Pr = 0.001
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                                      0.001
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- How to interpret the results?
- Chi-square test: testing the association between two binary variables.
- Using placebo has association with that the patients died or not.

• 2 by 2 tables straitified by sex

. bysort sex: tab died drug, col row chi2

 \rightarrow sex = 0

1 if patient died	•	(1=placebo)	Total
0	6	1	7
	85.71	14.29	100.00
	75.00	9.09	36.84
1	2	10	12
	16.67	83.33	100.00
	25.00	90.91	63.16
Total	8	11	19
	42.11	57.89	100.00
	100.00	100.00	100.00

Pearson chi2(1) = 8.6466 Pr = 0.003

-> sex = 1 1 if patient died	Drug type	(1=placebo) 1	Total
0	2	0	2
	100.00	0.00	100.00
	33.33	0.00	13.33
1	4	9	13
	30.77	69.23	100.00
	66.67	100.00	86.67
Total	6	9	15
	40.00	60.00	100.00
	100.00	100.00	100.00

Tables: Stata tool for Epidemiology

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- How to use Stata to generate risk ratios and odds ratios?
- A useful tool in Stata's default function can be found at
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```
Stata/IC 16.1 File Edit View Data Graphics Statistics User Window Help
```

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 Stata/IC 16.1 File Edit View Data Graphics Statistics User Window Help
- But before demonstrating how this works, a recapture on basic epi terms!

Rate

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- Period prevalence: **over a time period**, no. of diseased no. of persons

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- Period prevalence: over a time period, no. of diseased no. of persons
- Survival (proportion/probability) rate:

no. of alive persons (since diagnosis)
no. of initially disease-free persons (since diagnosis)

Quizs

1. What is the key difference between rate and proportion?

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- 3. Is risk a rate or a proportion?

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- Risk difference: the difference of the probabilities of an event between the exposed group and non-exposed group
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- Caution! Relative risk could be either risk ratio or rate ratio!

	Female (Exposed)	Male (Unexposed)	Total
shiba (Case)	2	2	4
guinea pig (Noncase)	2	1	3
Total	4	3	7

Epidemiologists love two by two tables!

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- Risk difference between females having shiba and males having shiba = $\widehat{p_F} \widehat{p_M} = 2/4 2/3 = -0.16667$
- Interpretation: Females have 16.67 % lower risk of having shiba 16.67 less per 100 subjects of having shiba than males.

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- Risk ratio between females having shiba and males having shiba

$$=\widehat{p_F} \div \widehat{p_M} = 2/4 \div 2/3 = 0.75.$$

 Interpretation: The risk of females having shiba is 0.75

Odds: the ratio between those having and not having an outcome.

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 E.g., the OR is 0.5, which indicates that there is a 50% decrease in the odds of having an outcome among the exposed compared to the unexposed.

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- Why is there no odds difference?

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 The odds of having shiba among females is

$$\widehat{odds_F} = \frac{p(\text{having shiba}|\text{female})}{p(\text{having guinea pig}|\text{female})}$$

$$= \frac{(2/4)}{(2/4)} = 1$$

 The odds of having shiba among males is 2 (calculation ignored).

	Female (Exposed)	Male (Unexposed)	Total
shiba	2	2	4
(Case) guinea pig	2	1	3
(Noncase) Total	4	3	7

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$$= \frac{(2/4)}{(2/4)} = 1$$

 The odds of having shiba among males is 2 (calculation ignored).

- OR of having shiba (females to males)
- OR = $\frac{Odds_f}{Odds_m} = \frac{1}{2}$
- Interpretation: there is a 50% decrease in the odds of having shiba among females compared to males. Higher odds of shiba ownership among males than females!
- It seems that females instead love guinea pigs more.

Interpreting results: Principles

- When describing a ratio, it can ideally be illustrated by
 - 1. Exposed group
 - 2. Ratio (exact value, higher or lower percentage)
 - 3. Outcome
 - 4. Unexposed

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- Example:
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 - 1. Exposed group
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 - 4. Unexposed
- Example:
 - 1. Females have a risk ratio of 0.75 having shiba compared to males.
 - Females have a 50% decrease in the odds of having shiba compared to males.

Interpreting results: Ratio > or < 1

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- As ratio < 1,
 - $(1 RR/OR) \times 100\%$
 - E.g., RR = 0.75, $(1 0.75) \times 100\% = 25\%$
 - 25% lower risk

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- As ratio < 1,
 - $(1 RR/OR) \times 100\%$
 - E.g., RR = 0.75, $(1 0.75) \times 100\% = 25\%$
 - 25% lower risk
- As ratio > 1.
 - $(RR/OR 1) \times 100\%$
 - E.g., OR = 2.05, $(2.05 1) \times 100\% = 105\%$
 - 105% higher odds
 - The odds is 2 times higher.
 - Twice the odds

Diabetes Is a Risk Factor for Pulmonary Tuberculosis: A Case-Control Study from Mwanza, Tanzania (Faurholt-Jepsen, 2011)

	OR (95% C.I.)
	Model 2
	sex, y ² Model 1 + AGP ³
V negative (n = 770)	
Glucose intolerance status ¹	
normal glucose tolerance	ref.
IFG/IGT	2.65 (1.00;7.06)
diabetes	4.23 (1.54;11.57)
Glucose intolerance status ¹ normal glucose tolerance IFG/IGT	2.65 (1.00;7.06)

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	OR (95% C.I.)	OR (95% C.I.) Model 1 Adjusted for age, sex, socio-demography ²	OR (95% C.I.)	
			Model 2	
	Unadjusted		Model 1 + AGP ³	
HIV negative (n = 770)				
Glucose intolerance status ¹				
normal glucose tolerance	ref.	ref.	ref.	
IFG/IGT	2.26 (1.50;3.41)	2.34 (1.52;3.61)	2.65 (1.00;7.06)	
diabetes	2.15 (1.35;3.42)	2.14 (1.32;3.46)	4.23 (1.54;11.57)	

1. People with diabetes had a higher odds of TB (OR 2.15, 95% CI: 1.35-3.42) relative to people without diabetes.

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		Model 2
Unadjusted	Adjusted for age, sex, socio-demography ²	Model 1 + AGP ³
ref.	ref.	ref.
2.26 (1.50;3.41)	2.34 (1.52;3.61)	2.65 (1.00;7.06)
2.15 (1.35;3.42)	2.14 (1.32;3.46)	4.23 (1.54;11.57)
	Unadjusted ref. 2.26 (1.50;3.41)	Model 1 Adjusted for age, sex, socio-demography 2 ref. ref. 2.26 (1.50;3.41) 2.34 (1.52;3.61)

- 1. People with diabetes had a higher odds of TB (OR 2.15, 95% CI: 1.35-3.42) relative to people without diabetes.
- Having diabetes was associated with more than a
 2-fold increase (OR: 2.15, 95% CI: 1.35; 3.42) in the odds of TB compared to not having diabetes.

Bidirectional association between physical activity and symptoms of anxiety and depression: the Whitehall II study (Azevedo Da Silva, 2012)

 $\begin{tabular}{ll} \textbf{Table 3} Cross-sectional associations between physical activity at recommended levels and anxiety and/or depression symptoms at phase 1 (1985–1988) (N = 9,309) \\ \end{tabular}$

	OR (CI 95 %)	P value
Anxiety symptoms		
Model 1		
Physical activity		
Yes	0.71 (0.54, 0.91)	0.01
No	1 (reference)	
Model 2		
Physical activity		
Yes	0.71 (0.55, 0.93)	0.01
No	1 (reference)	
Depression symptoms		
Model 1		
Physical activity		
Yes	0.63 (0.48, 0.81)	< 0.001
No	1 (reference)	
Model 2		
Physical activity		
Yes	0.63 (0.49, 0.82)	0.001
No	1 (reference)	

Bidirectional association between physical activity and symptoms of anxiety and depression: the Whitehall II study (Azevedo Da Silva, 2012)

Table 3 Cross-sectional associations between physical activity at recommended levels and anxiety and/or depression symptoms at phase 1 (1985–1988) (N = 9.309)

	OR (CI 95 %)	P value
Anxiety symptoms		
Model 1		
Physical activity		
Yes	0.71 (0.54, 0.91)	0.01
No	1 (reference)	
Model 2		
Physical activity		
Yes	0.71 (0.55, 0.93)	0.01
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Depression symptoms		
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1. Patients who conducted recommended levels of physical activity had a 29% lower odds of anxiety (OR: 0.71, 95% CI: 0.54-0.91) and a 37% lower odds of depression (OR: 0.63, 95% CI: 0.48-0.81) relative to those who did not.

Interpreting results: More examples

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Our results showed that individuals who practiced recommended levels of physical activity were less likely to have anxiety
 (OR: 0.71, 95% CI: 0.54-0.91) and depression
 (OR: 0.63, 95% CI: 0.48-0.81) in comparison with those who did not.

What is the problem of the following sentences?

1. Patients that have diabetes have a higher risk of developing cardiovascular disease.

What is the problem of the following sentences?

- 1. Patients that have diabetes have a higher risk of developing cardiovascular disease.
- 2. Our study suggests diabetic patients control their blood sugar level.

What is the problem of the following sentences?

- 1. Patients that have diabetes have a higher risk of developing cardiovascular disease.
- 2. Our study suggests diabetic patients control their blood sugar level.
- 3. Each observation in this Stockholm Public Health cohort was recruited through cross-sectional surveys done in 2002, 2006, and 2010.

What is the problem of the following sentences?

- 1. Patients that have diabetes have a higher risk of developing cardiovascular disease.
- 2. Our study suggests diabetic patients control their blood sugar level.
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- 4. The normal individuals (the control group) showed a lower risk of developing lung cancer.

What is the problem of the following sentences?

1. Patients that who have diabetes have a higher risk of developing cardiovascular disease.

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Calculate ratios using Stata: Risk ratio

• Finally we come back to Stata again!

Calculate ratios using Stata: Risk ratio

- Finally we come back to Stata again!
- cs case exposed
 - . cs died drug

	Drug type [1 Exposed	-	 Tot	al
Cases Noncases		6 8	 	25 9
Total	20	14	 	34
Risk	.95	.4285714	.73529	941
	Point estimate		 [95%	Conf. Interval]
Risk difference Risk ratio Attr. frac. ex. Attr. frac. pop	2.216667 .5488722			.7976911 631 4.092525 .043 .7556521
		chi2(1) =	11.50 F	Pr>chi2 = 0.0007

Calculate ratios using Stata: Odds ratio

cs case exposed, or
 cs died drug, or

	0 01	[1=placebo] Unexposed	 Total		
Cases Noncases	19 1	6 8			
Total	20	14	34		
Risk	.95	.4285714	.7352941		
	Point	estimate	 [95% Cor	nf. Interval]	
Risk difference	.5	5214286	.245166	6 .7976911	
Risk ratio	1 2.	216667	1.200631	1 4.092525	
Attr. frac. ex.	1 .5	.5488722		3 .7556521	
Attr. frac. pop	1 .4	171429	1		
Odds ratio	l 25	.33333	3.189793	3.	(Cornfield
•	+	chi2(1) =	11.50 Pr>c	chi2 = 0.0007	

Calculate ratios using Stata: Incidence rate ratio

ir case exposed studytime

. ir died drug studytime

Incidence-rate co	mparison				
	Drug type [1=placebo]			
	Exposed	Unexposed	Total		
1 if patient die	•	6	25		
Months to death	180	209	389		
	-+ 		-+ 		
Incidence rate	.1055556	.0287081	.0642674		
	Point	estimate	[95% Conf.	Interval]	
Inc. rate diff.	.07	68474	-+ .0241182	.1295766	
Inc. rate ratio	3.676852		1.411772	11.24864	(exact)
Attr. frac. ex.	.7280282		.2916701	.9111003	(exact)
Attr. frac. pop	.5533014		1		
	+				
Mid p-values for	tests of inci	dence-rate	difference:		

Adj Pr(Exposed 1 if patient die <= 19) = 0.9985 (lower one-sided)

Adj Pr(Exposed 1 if patient die >= 19) = 0.0015 (upper one-sided)

Two-sided p-value = 0.0031

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