

BOOST YOUR CONFIDENCE (INTERVALS) WITH SAS

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Background

Confidence Interval Definition

- **DEFINITION:** An interval around a statistic that contains the true underlying value of the statistic (the population parameter) a certain amount of the time

Confidence Interval Definition

- Example: a survey of 50 SAS programmers finds that the average IQ is 130 ± 10
- If we did 100 surveys, the average IQ should be between 120 and 140 in 95 of them

Confidence Interval Definition

- **95% confidence interval** bounded by the upper 95% confidence **limit** and the lower 95% confidence **limit**
- 95% just conventional. Can have for e.g.:
 - 90% CIs (narrower)
 - 99% CIs (wider)
- CI for any level (95% etc) is narrower if based on more observations

General Formula

- Can make CIs around almost any statistic you calculate, for example...
- **Data summaries** (1 var) such as:
 - Categorical variables: proportion
 - Continuous variables: mean
- Statistics resulting from **hypothesis tests** (2+ vars):
 - Correlation, regression slope
 - Relative risk, odds ratios

Using Confidence Intervals

Why Use CIs?

- Gives audience idea of impact of chance
- Gives reasonable bounds on your result(s)
- Can check if your data are compatible with a certain value
 - (From data summary): Does 95% confidence interval of IQ in SAS programmers include 100?
 - (From hypothesis testing): Is occurrence of schizophrenia higher in SAS programmers or SPSS users? (Does relative risk = 1.00?)

Why Not Use CIs?

- Some organizations consider their figures to be a census, not a sample
- Increases statistical work for staff
- Some data users may find the concept confusing

Overlapping CIs vs Hypothesis Tests

- Example: Want to compare prevalence of schizophrenia in SAS vs SPSS users
- You could:
 - (A) Calculate schizophrenia prevalence and 95% CI for each group and see if overlap
OR
 - (B) Calculate prevalence ratio of one group vs another, and see if includes 1.00

Overlapping CIs vs Hypothesis Tests

- Answer: (B) usually better
- Why? More statistical power
- Why even consider the first approach?
 - Easier to do if already have data summaries published
 - Can't anticipate all comparisons readers will want to make

Confidence Intervals For Data Summaries

CIs for Means

- Based on normal distribution
- Where your study sample is large (# of subjects > 30), the sampling distribution → normal, and you can use:

$$\begin{aligned} \text{CI} &= \text{obsd mean} \pm 1.96 \times \text{standard error (of mean)} \\ &= \text{obsd mean} \pm 1.96 \times \text{standard dev'n} / \text{sqrt}(n) \end{aligned}$$

$$= \bar{x} \pm 1.96 \frac{s}{\sqrt{n}}$$

CI's for Means: Using SAS

- Let SAS do it!
- Recall: Using SAS for simple desc stats:

```
proc means;  
  var <variable name>;
```

Results

Results

- Means: USING SAS FOR CONFIDENC

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The MEANS Procedure

Analysis Variable : i_bwt_v

N	Mean	Std Dev	Minimum	Maximum
100	3305.77	484.8643850	1644.00	4340.00

CI's for Means: Using SAS

- To get confidence limits, request “clm” as PROC MEANS option
- (Request “mean” to get the mean printed out too)

```
proc means mean clm;  
  var <variable name>;
```


Results

- Results
- Means: USING SAS FOR CONFIDENC

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The MEANS Procedure

Analysis Variable : i_bwt_v

Mean	Lower 95% CL for Mean	Upper 95% CL for Mean
3305.77	3209.56	3401.98

CIs for Proportions

- Based on binomial distribution
- Where your study sample is large (# with and without the characteristic > 5), the sampling distribution \rightarrow normal, and you can use:
 - = obsd proportion \pm 1.96 X SE of prop'n

$$= p \pm 1.96 \sqrt{\frac{p(1-p)}{n}}$$

CIs for Proportions: Using SAS

- To get confidence limits, request “binomial” as TABLE option in PROC FREQ:

```
proc freq;  
  table m_edu2g_v / binomial;
```

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The FREQ Procedure

m_edu2g_v	Frequency	Percent	Cumulative Frequency	Cumulative Percent
< High School	35	35.00	35	35.00
High School +	65	65.00	100	100.00

Binomial Proportion for
m_edu2g_v = < High School

Proportion	0.3500
ASE	0.0477
95% Lower Conf Limit	0.2565
95% Upper Conf Limit	0.4435

Exact Conf Limits

95% Lower Conf Limit	0.2573
95% Upper Conf Limit	0.4518

Test of H0: Proportion = 0.5

ASE under H0	0.0500
Z	-3.0000
One-sided Pr < Z	0.0013
Two-sided Pr > Z	0.0027

Sample Size = 100

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- Freq: USING SAS FOR CONFIDENCE

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- Freq: USING SAS FOR CONFIDENCE

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CIs for Rates of Rare Outcomes

- Follows Poisson distribution
- Standard error based on number of cases

CIs for Rates of Rare Outcomes

PREVALENCE BASED ON < 30 CASES

- Find upper and lower 95% conf limits for # cases:
 - In table (e.g. poisson.xls)
 - Using PEPI program POISSON
- Multiply both by (10000) and divide by # population or live births to express as birth prevalence

	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R
1	COUNT	LFACTOR	UFACTOR															
2	0	0.0000	3.6889															
3	1	0.0253	5.5716															
4	2	0.2422	7.2247															
5	3	0.6187	8.7673															
6	4	1.0899	10.2416															
7	5	1.6235	11.6683															
8	6	2.2019	13.0595															
9	7	2.8144	14.4227															
10	8	3.4538	15.7632															
11	9	4.1154	17.0848															
12	10	4.7954	18.3904															
13	11	5.4912	19.6820															
14	12	6.2006	20.9616															
15	13	6.9220	22.2304															
16	14	7.6539	23.4896															
17	15	8.3954	24.7402															
18	16	9.1454	25.9830															
19	17	9.9031	27.2186															
20	18	10.6679	28.4478															
21	19	11.4392	29.6709															
22	20	12.2165	30.8884															
23	21	12.9993	32.1007															
24	22	13.7873	33.3083															
25	23	14.5800	34.5113															
26	24	15.3773	35.7101															
27	25	16.1787	36.9049															
28	26	16.9841	38.0960															
29	27	17.7932	39.2836															
30	28	18.6058	40.4678															
31	29	19.4218	41.6488															
32	30	20.2409	42.8269															
33	31	21.0630	44.0020															

CIs for Rates of Rare Outcomes

EXAMPLE

- # cases of anophthalmia in CA 1983-1986 = 18
- Prevalence = $(18 * 10,000 / 452,287) = 0.40$
- Looking in table, lower 95% CL is 10.67 and upper 95% CL is 28.45 **for # cases**
- To express as prevalence CLs, multiply both by 10,000 and divide by 452,287 live births
- Lower 95% CL = 0.24, upper 95% CL = 0.63
- Thus we say prev = 0.40 cases per 10,000 live births, 95% CI = 0.24 - 0.63

CIs for Rates of Rare Outcomes

PREVALENCE BASED ON 30+ CASES

- Considered large # cases (more or less)
- Poisson \rightarrow normal distribution
- Can use normal approximation in SAS code
- Several equations for doing this, yielding similar results

CIs for Rates of Rare Outcomes

- SAS code for obs with few cases:
Combine with Poisson lower and upper limits for cases (get lfactor and ufactor for the observed # cases):

```
proc sort data=b1; by count;

proc sort data=lib2.poisson out=poisson;
  by count;

data c1 prob2;
  merge b1(in=b) poisson(in=p);
  by count;
  if b;
```

CIs for Rates of Rare Outcomes

- SAS code: Calculate CIs for obs with many cases too

```
data c2;  
  set c1;  
  calcrate = count * 10000 / births;  
  if count le 30 then do;  
    calclocl = lfactor * 10000 / births;  
    calcucl = ufactor * 10000 / births;  
  end;  
  else if count > 30 then do;  
    calclocl = ((count / births) - (1.96 * sqrt(count) / births)) * 10000;  
    calcucl = ((count / births) + (1.96 * sqrt(count) / births)) * 10000;  
  end;  
  rate = round(calcrate,.01);  
  lcl = round(calclocl,.01);  
  ucl = round(calcucl,.01);  
  rename count=cases;
```

Reminder

- To compare groups (e.g. whether rates are statistically different), can calculate 95% confidence intervals and see if they overlap
- Better to do hypothesis testing

Confidence Intervals
For Statistics From
Hypothesis Tests / Measures
of Association

Types of Hypothesis Tests / Measures of Association

Indep Var	Dep Var	Approach (SAS PROC)
Categorical	Categorical	Contingency tables (FREQ)
Cat. / Cont.	Categorical	Logistic reg. (LOGISTIC)
Categorical	Cont.	ANOVA (GLM); if 2 levels then T-test (TTEST or GLM)
Cont.	Cont.	Linear reg. (REG or GLM)
Cat. / Cont.	Rates	Poisson reg. (GENMOD)

Contingency Tables Using SAS

- Recall: Using SAS to produce the basic 2x2 table:

```
proc freq;  
  tables <indep var> * <outcome var>;
```

- To get odds ratios and their CIs, request measures of association:

```
proc freq;  
  tables <indep var> * <outcome var>  
  / measures;
```


Contingency Tables Using SAS

QUESTION: Is mother's education associated with % low birth weight babies?

- Independent variable = m_edu2g_v
- Outcome variable = lbw
- If no statistically significant association, 95% CI will include 1.00

```
proc freq;  
  tables m_edu2g_v * lbw / measures;
```

Results

- Freq: USING SAS FOR CONFIDENCE

The FREQ Procedure

Table of m_edu2g_v by lbw

m_edu2g_v	lbw		Total
	Low bwt	Normal bwt	
< High School	10769 2.65 8.64 31.40	113937 28.05 91.36 30.63	124706 30.70
High School +	23525 5.79 8.36 68.60	258010 63.51 91.64 69.37	281535 69.30
Total	34294 8.44	371947 91.56	406241 100.00

Frequency Missing = 539

Statistics for Table of m_edu2g_v by lbw

Statistic	Value	ASE
Gamma	0.0180	0.0061
Kendall's Tau-b	0.0046	0.0016
Stuart's Tau-c	0.0024	0.0008
Somers' D C R	0.0028	0.0010
Somers' D R C	0.0077	0.0026
Pearson Correlation	0.0046	0.0016
Spearman Correlation	0.0046	0.0016

Results

- Results
- Freq: USING SAS FOR CONFIDENCE

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The FREQ Procedure

Statistics for Table of m_edu2g_v by lbw

Estimates of the Relative Risk (Row1/Row2)

Type of Study	Value	95% Confidence Limits	
Case-Control (Odds Ratio)	1.0366	1.0122	1.0616
Cohort (Co11 Risk)	1.0335	1.0112	1.0562
Cohort (Co12 Risk)	0.9969	0.9949	0.9990

Effective Sample Size = 406241
Frequency Missing = 539

Logistic Regression: Using SAS

- Can use PROC LOGISTIC in SAS; nice since it will exponentiate the slope (b) and its 95% confidence interval
- Basic syntax:

```
proc logistic;  
  model <outcome var> = <independent var>;
```

- For low birthweight example:

```
proc logistic;  
  model lbw = m_edu2g_v;
```

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The LOGISTIC Procedure

Testing Global Null Hypothesis: BETA=0

Test	Chi-Square	DF	Pr > ChiSq
Likelihood Ratio	8.7032	1	0.0032
Score	8.7372	1	0.0031
Wald	8.8380	1	0.0030

Analysis of Maximum Likelihood Estimates

Parameter	DF	Estimate	Standard Error	Wald Chi-Square	Pr > ChiSq
Intercept	1	-2.3226	0.0213	11912.0530	<.0001
m_edu2g_v	1	-0.0362	0.0122	8.8380	0.0030

Odds Ratio Estimates

Effect	Point Estimate	95% Wald Confidence Limits
m_edu2g_v	0.964	0.942 0.988

Association of Predicted Probabilities and Observed Responses

Percent Concordant	21.8	Somers' D	0.008
Percent Discordant	21.0	Gamma	0.018
Percent Tied	57.2	Tau-a	0.001
Pairs	1275550418	c	0.504

Logistic Regression: Using SAS

- (One way) to get correct odds ratio:
declare independent var to be a
classification (categorical) var:

```
proc logistic;  
  class m_edu2g_v;  
  model lbw = m_edu2g_v;
```

Results

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Model Fit Statistics

Criterion	Intercept Only	Intercept and Covariates
AIC	235157.79	235151.08
SC	235168.70	235172.91
-2 Log L	235155.79	235147.08

Testing Global Null Hypothesis: BETA=0

Test	Chi-Square	DF	Pr > ChiSq
Likelihood Ratio	8.7032	1	0.0032
Score	8.7372	1	0.0031
Wald	8.8380	1	0.0030

Type 3 Analysis of Effects

Effect	DF	Wald Chi-Square	Pr > ChiSq
m_edu2g_v	1	8.8380	0.0030

Analysis of Maximum Likelihood Estimates

Parameter	DF	Estimate	Standard Error	Wald Chi-Square	Pr > ChiSq
Intercept	1	-2.3768	0.00608	152684.523	<.0001
m_edu2g_v < High School	1	0.0181	0.00608	8.8380	0.0030

Odds Ratio Estimates

Effect	Point Estimate	95% Wald Confidence Limits
m_edu2g_v < High School vs High School +	1.037	1.012 1.062

Logistic Regression: Using SAS

- Comparing results from PROC FREQ and PROC LOGISTIC:

SAS Proc	OR	95% CI
FREQ	1.0366	1.0122 – 1.0616
LOGISTIC	1.037	1.012 – 1.062

Logistic Regression

With Multiple Predictor Variables

- Like other regression, the slope (b) is adjusted for all other independent variables in the model
- SAS takes both cont and categorical vars
 - SAS assumes ind vars are continuous
 - If categorical, list in CLASS statement and SAS creates dummy vars automatically

```
proc logistic;  
  class <categorical independent vars>;  
  model <dependent var> = <independent vars>;
```

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Thanks