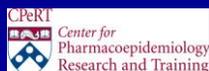


Statistical Test Selection in Epidemiologic Research

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“I was told there would be no math!”



- Chevy Chase
 ‘Spies Like Us’

Learning Objectives

- Understand variable characteristics to guide statistical test selection
- Learn to use, interpret correlation coefficients
- Understand variables influencing sample size
- Gain familiarity with use, interpretation of linear and logistic regression

Outline

- Constructing a research project
- Correlation / regression
- Linear regression
- Logistic regression

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- **Constructing a research project**
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- Linear regression
- Logistic regression

Constructing a Research Project

- Research question
- Variable characteristics
- Study design, sample size
- Statistical methods

Constructing a Research Project

- **Research question**
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3 Types of Questions:
Different Analyses

Research Question Type #1

- How much or common?
 - Design: Cross-sectional, cohort studies
 - Descriptive statistics:
 - Percentages, frequencies
 - Means (standard deviations)
 - Medians (interquartile ranges)
 - Prevalence (95% CIs)
 - Incidence (95% CIs)

Potential
Analyses

Research Question Type #2

- Are these groups different?
 - Design: Case-control, cohort, RCTs
 - T-test: difference in means
 - Wilcoxon rank-sum: difference in medians
 - Chi square, Fisher's exact: diff. in frequencies
 - ANOVA, Kruskal-Wallis: diff. in means, medians among ≥3 groups
 - Odds ratios, hazard ratios, relative risks

Potential
Analyses

Research Question Type #3

- Can certain variables predict outcome?
 - Design: Cohort study
 - Linear regression
 - Logistic regression
 - Survival analysis (Cox regression)

Potential
Analyses

Constructing a Research Project

- Research question
- **Variable characteristics**
- Study design, sample size
- Statistical methods

Variable Characteristics to Consider

- Categorical or continuous?
 - Continuous: Normal or not?
- How many independent variables?
- How many groups?

Variable Characteristics to Consider

- **Categorical or continuous?**
 - Continuous: Normal or not?
- How many independent variables?
- How many groups?

Categorical or Continuous?

- **Continuous: *Any value within a range***
 - Age (years)
 - Blood pressure (mm Hg)
 - Height (m)
 - Weight (kg)
 - CD4 cell count (cells/mm³)

Categorical or Continuous?

- **Categorical: *Discrete categories***

Nominal

Named categories with no order

- Blood type
- Medical specialty

Ordinal

Ordered categories, differences unequal

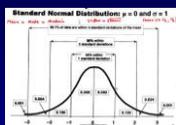
- NY Heart Class

Variable Characteristics to Consider

- **Categorical or continuous?**
 - Continuous: Normal or not?
- How many independent variables?
- How many groups?

Normal Distribution

- Continuous data
- Symmetrical, bell shaped
- Mean, median, mode all the same and located at the center
- Allows use of parametric tests (e.g., t-tests)
- If not, must use non-parametric tests



Variable Characteristics to Consider

- **Categorical or continuous?**
 - Continuous: Normal or not?
- **How many independent variables?**
- **How many groups?**

Depends on Clinical Question!

Constructing a Research Project

- Research question
- Variable characteristics
- **Study design, sample size**
- Statistical methods

Hypothesis Testing

- Develop hypothesis
- Test hypothesis:
 - Collect data, observe effect
 - H_0 = No effect or difference
 - H_a = Effect or difference
- How likely is it that effect occurs by chance
 - If very unlikely ($p < 0.05$), reject H_0

Effect Size, Significance, Power,

- **Effect Size**: Magnitude of effect being studied
 - Should represent clinically significant difference
- **Significance**: Probability of Type I error ($\alpha=0.05$)
- **Power**: Probability of detecting difference (80%)
- **Sample Size**: n required to show a difference at set values of effect size, power, and significance

Determination of Sample Size

- Why is this necessary?
 - To detect effect size (OR, RR, HR) as significant
 - Avoid false-positive, false-negative conclusions
 - Avoid enrolling too many patients
- When to determine sample size?
 - During preparation of all protocols (perform early!)
- How to calculate?
 - Stata
 - Other programs: PS - Power / Sample Size, nQuery

Variables Used to Calculate Sample Size

- Detectable (clinically meaningful) difference (d^*):
 - Magnitude of difference in proportions, means
- r: ratio of unexposed:exposed, controls:cases
- Power ($1 - \beta$):
 - Type II error (β) = prob. that there is no difference when one does exist (false-negative; set at 0.1, 0.2)
- Type I error (α):
 - Prob. of concluding that there is difference when one does not exist (false-positive; usually set at 0.05)

Variables Used to Calculate Sample Size

- p_1 (for proportions):
 - Proportion exposed who develop disease (cohort/ cross-sectional)
 - Proportion of cases exposed (case-control)
- p_0 (for proportions):
 - Proportion unexposed who develop disease (cohort/ cross-sectional)
 - Proportion of controls exposed (case-control)
- Standard deviation (σ) of continuous outcome

Calculation of Sample Size

- Primary outcome is variable for which you perform sample size calculation
 - If secondary outcomes important, ensure sample size is sufficient
- Typically, have more power to detect differences in continuous outcomes

Sample Size Calculation: Difference in Means

- Sample size for difference in means:

$$n = \frac{(Z_{\beta} + Z_{\alpha/2})^2 \sigma^2 (r + 1)}{(d^*)^2 r}$$

- Variables:
 - σ = standard deviation of outcome (σ^2 =variance)
 - $Z_{\alpha/2}$ = type I error of 0.05; value=1.96
 - Z_{β} = type II error; for 0.2 [80% power], value=0.84
 - $(Z_{\beta} + Z_{\alpha/2})^2 = 7.85$ for 80% power
 - $(Z_{\beta} + Z_{\alpha/2})^2 = 10.5$ for 90% power
 - r = ratio; d^* = detectable difference

Sample Size Calculation: Difference in Proportions

- Sample size for difference in proportions:

$$n = \frac{(Z_{\beta} + Z_{\alpha/2})^2 p_w (1 - p_w) (r + 1)}{(d^*)^2 r}$$

- Note: p_w = weighted average of p_1 and p_0

$$p_w = (p_1 + r p_0) / (1+r)$$

Variables Affecting Sample Size

- Detectable difference:
 - Smaller difference (effect size): \uparrow sample size
- Power:
 - \uparrow power (e.g., 80 \rightarrow 90%): \uparrow sample size
- Standard deviation of outcome (σ):
 - Smaller σ : \downarrow sample size
- P_0 :
 - Smaller p_0 : \uparrow sample size
- Significance level (α):
 - $\uparrow \alpha \rightarrow \downarrow$ sample size

Sample Size Calculations

- You must increase sample size to reflect:
 - Loss to follow up
 - Expected response rate
 - Lack of adherence, etc.
 } \downarrow Effective sample size
- Example:
 - Targeted number of exposed = 1,200 subjects
 - But only 70% expected to consent (30% refusal rate)
 - Adjust targeted number of exposed as follows:
 - $1,200 / 0.7 = 1,714$ exposed
 - So if 1:1 ratio, need $1,714 * 2 = 3,428$ total subjects

Constructing a Research Project

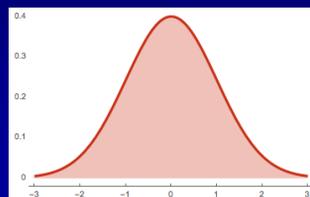
- Research question
- Variable characteristics
- Study design, sample size
- Statistical methods

Select Appropriate Statistical Test for Continuous Data

Purpose of Test	Normal Theory Test (Parametric)	Corresponding Non-Parametric Test
Compare paired data	Paired t-test	Wilcoxon signed-rank test
Compare 2 independent samples	Two-sample t-test	Wilcoxon rank-sum test (Mann-Whitney U test)
Compare ≥ 3 groups	One-way ANOVA	Kruskal-Wallis

Parametric Tests

- Continuous data
- Normally distributed



Non-Parametric Tests

- Not normally distributed continuous data
 - Small samples
- OR
- Categorical data
 - Nominal, ordinal
 - Dichotomous (Outcome vs. no outcome)

Select Appropriate Statistical Test for Each Question

Purpose of Test	Normal Theory Test (Parametric)	Corresponding Non-Parametric Test
Compare paired data	Paired t-test	Wilcoxon signed-rank test
Compare 2 independent samples	Two-sample t-test	Wilcoxon rank-sum test (Mann-Whitney U test)
Compare ≥ 3 groups	One-way ANOVA	Kruskal-Wallis

Categorical Data Analysis: Chi Square Analysis

- Answers: "Are these groups different?"
- Contingency tables evaluate relation between values of ≥ 2 categorical variables
 - Rows, columns are independent

Categorical Data Analysis: Fisher's and McNemar's Tests

- **Fisher's Exact Test:** Use if any value in a cell of table is < 5
- **McNemar's Test:** Use if data are from paired samples

Outline

- ❖ Constructing a research project
- ❖ **Correlation / regression**
- ❖ Linear regression
- ❖ Logistic regression

Correlation / Regression

- Examine relation between variables
- Correlation:
 - Tests significance of relation
- Regression:
 - Quantifies relationship, controlling for confounders

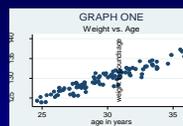
Correlation Coefficient

- Quantifies relationship between two variables
 - Correlation coefficient (“r”) ranges -1 to +1

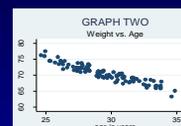
Value of r	Interpretation
$r = 0$	Two variables do not vary together at all
$0 < r < 1$	Two variables increase or decrease together
$r = 1.0$	Perfect correlation
$-1 > r > 0$	One variable increases as the other decreases
$r = -1.0$	Perfect negative or inverse correlation

- Often useful to graph data

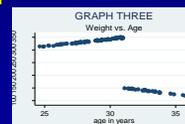
Correlations



Positive Correlation



Negative Correlation



Relation between weight and age is different for younger vs. older

Correlation Statistics

- Pearson’s correlation (most widely used):
 - Assumes normally distributed data
 - Compute pairwise correlation (“r”), p values
- Spearman’s rank-correlation coefficient:
 - One or more variables → not normally distributed
 - Less sensitive to effects of outlier data
 - Compute correlation (“r”), p values

Outline

- Constructing a research project
- Correlation / regression
- **Linear regression**
- Logistic regression

Linear Regression

- Allows you to relate continuous outcome (y) to one or more predictor variables (x_1, \dots, x_k)
 - Mean value of y is expressed as linear combination of x 's (x 's may be continuous or categorical)
 - Useful when have many potential confounders
- Have equation in the form:

$$y = \alpha + \beta * x$$

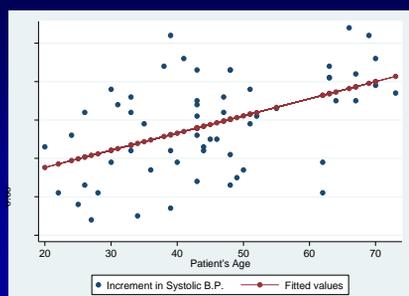
- Typically written as:

$$E(y) = \beta_0 + \beta_1 * x \rightarrow \text{fitted values}$$

Linear Regression

- Perform ordinary least squares regression of dependent variable y on independent variable x
- Estimates minimize squared distance between observed data and fitted values from model

Linear Regression: Determine Best Fit Line in Data



For every 1 unit (cm) ↑ in height, the pulmonary deadspace ↑ by 1.03 mL.

```

. regress deadspace height

```

Source	SS	df	MS	Number of obs = 15	
Model	5607.43156	1	5607.43156	F(1, 13) =	32.81
Residual	2221.50170	13	170.884732	Prob > F =	0.0001
Total	7828.93333	14	559.209524	R-squared =	0.7162
				Adj R-squared =	0.6944
				Root MSE =	13.072

deadspace	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
height	1.033323	.1803872	5.73	0.000	.6436202 1.423026
_cons	-82.4852	26.30147	-3.14	0.008	-139.3061 -25.66433

The constant is mean deadspace for a child without asthma.
 Mean deadspace for child without asthma = 83 mL
 Mean deadspace for child with asthma = 52.9 mL

```

. regress deadspace asthma

```

Source	SS	df	MS	Number of obs = 15	
Model	3388.05833	1	3388.05833	F(1, 13) =	9.92
Residual	4440.875	13	341.605769	Prob > F =	0.0077
Total	7828.93333	14	559.209524	R-squared =	0.4328
				Adj R-squared =	0.3891
				Root MSE =	18.483

deadspace	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
asthma	-30.125	9.565644	-3.15	0.008	-50.79032 -9.459683
_cons	83	6.985759	11.88	0.000	67.90819 98.09181

Multivariable Linear Regression

- Evaluate continuous outcome by linear relationship with independent variables
- Have >1 independent variable

Use of Multivariable Linear Regression

- Multiple factors may predict outcome
 - Blood pressure may be affected by weight, hormones, age, other factors
- Control for factors that can vary, but may confound, statistical analysis
 - E.g., age, sex, race, comorbidities
- Improve prediction

Outline

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- **Logistic regression**

Logistic Regression

- **Alternate form of regression**
 - Use when outcome is binary
 - Death versus survive
 - Acute MI versus no acute MI
 - One or more predictor variables

Logistic Regression

- Regression method for binary outcomes
- Useful for:
 - Continuous or discrete covariates
 - Adjusting for potential confounders
 - Evaluation of effect modifiers
- Provides OR (95% CI) of outcome for those with vs. without exposure of interest
-

Logistic Regression

- Determination of odds ratios (ORs) is based on maximum likelihood methods
 - Find coefficient values that maximize likelihood of obtaining observed data
- Output requested: β coefficients or ORs

Logistic Regression

```

. logit status admit_type
Iteration 0:   log likelihood = -100.00000
Iteration 1:   log likelihood = -92.170936
Iteration 2:   log likelihood = -92.528570
Iteration 3:   log likelihood = -92.524467
Iteration 4:   log likelihood = -92.524467

Logistic regression              Number of obs   =    200
                                LR chi2(1)         =    15.11
                                Prob > chi2        =    0.0001
                                Pseudo R2         =    0.0795

Log likelihood = -92.524467

+-----+-----+
| status |      Coef. |   Std. Err. |      z |   P>|z| | [95% Conf. Interval] |
+-----+-----+-----+-----+-----+
| admit_type | 2.184917 |   .7450489 |   2.93 |  0.003 |   .224676   3.645196 |
| _cons     | -3.238678 |   .7208382 |  -4.49 |  0.000 |  -4.601495  -1.825861 |
+-----+-----+-----+-----+-----+

. logistic atatus admit_type
Logistic regression              Number of obs   =    200
                                LR chi2(1)         =    15.11
                                Prob > chi2        =    0.0001
                                Pseudo R2         =    0.0795

Log likelihood = -92.524467

+-----+-----+
| status | Odds Ratio |   Std. Err. |      z |   P>|z| | [95% Conf. Interval] |
+-----+-----+-----+-----+-----+
| admit_type | 0.889907 |  6.623415 |   2.93 |  0.003 |  2.064004  30.28988 |
| _cons     | 0.392157 |  0.0282682 |  -4.49 |  0.000 |  0.0095473  .1610788 |
+-----+-----+-----+-----+

```

Issues to Consider with Logistic Regression

- Which variables to include
- Which fitting method to use
- Collinearity
- Effect modifiers:
 - Does alcohol use level alter relation between drugs and acute liver injury?

Final Important Considerations

- Know your data before analysis!
 - Look for missing data, develop plan to address
 - Graph variables, relationships between variables
- Collaboration with biostatistician is useful