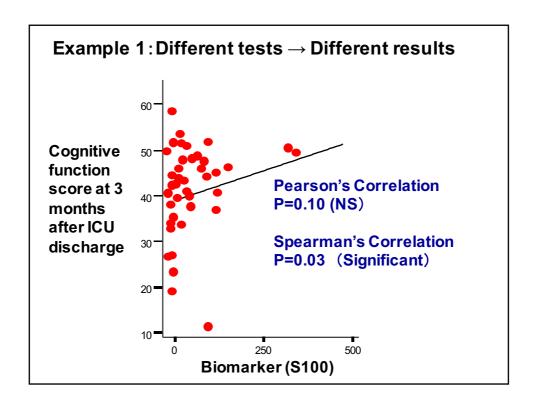
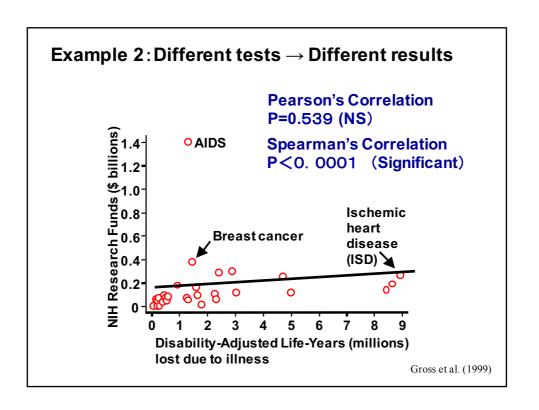
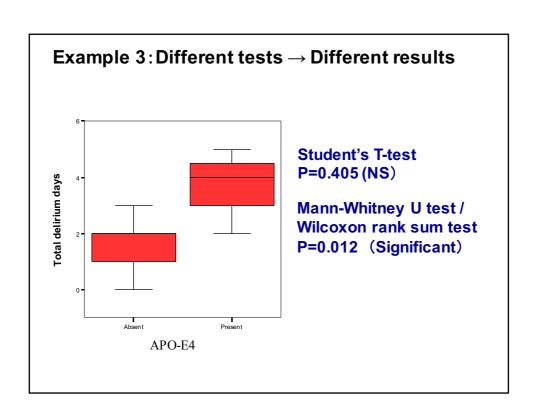
# **Selecting Proper Statistical Tests for Evidence Based Medicine**

## Overview:

- 1.1 Why the selection of valid statistical tests is important?
- 1.2 What are the factors to be considered for test selection?
- 1.3 Can you select now? (Tutorials)







## MED101x Introduction to Applied Biostatistics

## The Scandal of Poor Medical Research

Douglas G. Altman. British Medical Journal, 1994.

What should we think about a doctor who uses the wrong treatment, either willfully or through ignorance, or who uses the right treatment wrongly (such as by giving the wrong dose of a drug)? Most people would agree that such behavior was unprofessional, arguably unethical, and certainly unacceptable.

What then would we think about researchers who use the wrong techniques (either willfully or in ignorance), use the right techniques wrongly, misinterpret their results, report their results selectively, cite the literature selectively, and draw unjustified conclusions? We should be appalled. Yet numerous studies of the medical literature, in both general and specialist journals, have shown that all of the above phenomena are common. This is surely a scandal.

## 2 Factors to be considered for test selection



Understanding A Statistician's Mind

# MED101x Introduction to Applied Biostatistics

1.Univariate	Q 2, D ifference	Q3,	Q 4. Q 5	Q 6. No. of	Q 7.sampl	
l utivariab le	/Correlatio	Paired / related	Type of outcome (Normality)	groups	e size	Valid Tests
Univariate Q1	D ifference	nt nt (un-paired)	Continuous Normal)	2 >2		Student's t-test O ne-way ANOVA
			Continuous Non-normal)/	2		M ann-W hitney U test
			0 rdered categorical	>2		Kruska HW allis H test
				2	< 20	Fisher's exact test
			Nom nai	≧2	≧20	Chi-square test
			Time to Event			Log-Rank test Kaplan-Meierplot)
				2		Paied-t test
			Continuous <del>Marma</del> N			Repeated measured ANOVA
		Q U	04 5	$\cap$ 6		ked effect Regression
			Continuous Q4,5	Q6	<mark>⊨ Q7</mark>	iboxon signed-rank test
			0 rdered cat			riedm an test
			Nominal	2		M cN em ar's test
	Correlation		Continuous Normal)			Pearson's correlation ()
			Continuous Non-normal)/ordered			Spearm an's correlation (s)
			Nominal (2 levels)	2		Spearm an/Kappa (Agrreem ent)
		Indepdende	Continuous Normal residulas)			Linear Regression
			Continuous Non-normal residulas)			Linear Regression*
		nt	o lucicu categolical			0 rdered Logistic Regression
		(un-paired)	Nominal (2 levels)			Binary Logistic Regression
u Itivariab le		parou,	(>Z)			Multinomial Logistic Regression
m u itivai lab le			Time to Event			Cox Proportional Hazard Regression
			Continuous Normal residulas)			Linear Mixed Effect Regression
			Continuous Non-normal residulas)			Linear Mixed Effect Regression*
			0 rdered categorical			Generalized Estimation Equation (GEE
			Nominal (2 levels)			Generalized Estimation Equation (GEE

	Q2.					
Q1,Univariate Mutivariable	Difference /Correlatio	Q3, Paired / related	Q4, Q5 Type of outcome (Normality)	Q6, No. of groups	Q7,sampl e size	Valid Tests
		Indepdende nt (un-paired)	Continuous (Normal)	2		Student's t-test
			· · ·	>2		One-way ANOVA
			Continuous (Non-normal)/	2		Mann-Whitney U test
Univariate	Difference		Ordered categorical	>2	<del> </del>	Kruskal-Wallis H test
			Nominal	2	<20	Fisher's exact test
				≥2	≥20	Chi-square test
			Time to Event			Log-Rank test(Kaplan-Meier plot)
		Dependent (paired)	Continuous (Normal)	2		Paied-t test
				>2		Repeated measured ANOVA
				, -		Mixed effect Regression
			Continuous (Non-normal)/	2		
			Ordered categorical	>2		
			Nominal	2	ļ	
	Correlation		Continuous (Normal)			
			Continuous (Non-normal)/ordered		<b>.</b>	
			Nominal (2 levels)	2		
			Continuous (Normal residulas)			
		Indepdende	Continuous (Non-normal residulas)			
		nt	Ordered categorical		Log-Rank test (Kaplan-Meier  Paied-t test Repeated measured ANOVA Mixed effect Regression Wilcoxon signed-rank test Friedman test McNemar's test Pearson's correlation (r) Snearman/Kappa (Agrression Linear Regression Linear Regression Binary Logistic Regression Multinomial Logistic Regression	
		(un-paired)	Nominal (2 levels)			
Multivariable			(>2)			
idici variabio			Time to Event			Cox Proportional Hazard Regression
		(paired)	Continuous (Normal residulas)			
			Continuous (Non-normal residulas)			Linear Mixed Effect Regression*
			Ordered categorical		1	Generalized Estimation Equation (GEE)
			Nominal (2 levels) izing residuals		<u> </u>	Generalized Estimation Equation (GFF)

# Question 1 — Univariate?

Which type of test do you need: Univariate or Multivariable?

Univariate - Unadjusted Analysis Multivariable - Adjusted Analysis

- -Are there confounders?
- Need for adjustment?

# Question 1 — Univariate? (cont'd)

Randomization prevents confounding. Thus, in general, confounders are more problematic in observations studies than RCT's.

If you want to adjust for confounders, then you need to perform **regression analysis**.

RCT -> Probably OK with univariate analysis Observational studies -> Need to use Regression

ORIGINAL CONTRIBUTION

## Aspirin Use and All-Cause Mortality Among Patients Being Evaluated for Known or Suspected Coronary Artery Disease

A Propensity Analysis

Patricia A. Gum, MD	
Maran Thamilarasan, MD	
Junko Watanabe, MD	
Eugene H. Blackstone, MD	
Michael S. Lauer, MD	

Context Although aspirin has been shown to reduce cardiovascular morbidity and short-term mortality following acute myocardial infarction, the association between its use and long-term all-cause mortality has not been well defined.

**Objectives** To determine whether aspirin is associated with a mortality benefit in stable patients with known or suspected coronary disease and to identify patient characteristics that predict the maximum absolute mortality benefit from aspirin.

Population? Patients with an echo for possible coronary disease

Exposure?
Use of Aspirin at the baseline visit

Control? No use of Aspirin at the baseline visit

Long term mortality (median FU of 3.1 years)

## Results

Table 2. Cox Proportional Hazards Analyses of Time to Death Among Patients Using Aspirin

Model	Hazard Ratio (95% CI)	<i>P</i> Value
Unadjusted	1.08 (0.85-1.39)	.50
Adjusted for age and sex	0.75 (0.58-0.96)	.02
Adjusted for age, sex, and history of CAD	0.57 (0.44-0.74)	<.001
Multivariable adjusted†	0.67 (0.51-0.87)	.002
Adjusted for age and sex among prespecified strata Normal LV function	0.75 (0.56-1.01)	.06
Abnormal LV function	0.54 (0.34-0.84)	.006
No history of prior CABG surgery	0.74 (0.54-1.08)	.06
History of prior CABG surgery	0.56 (0.35-0.89)	.01
The state of the s		

<sup>\*</sup>CI indicates confidence interval; CAD, coronary artery disease; LV, left ventricular; and CABG, coronary artery bypass

<sup>\*\*</sup>Cl Indicates confidence interval; Cuto hay a ricely disease, and the properties of the properties

## MED101x Introduction to Applied Biostatistics

❖ People who use aspirin had reasons to use aspirin, they were sicker and had poorer prognosis. Therefore this may mask the effect of aspirin as its effect is mixed with poorer patients prognosis. This is called "Confounding".

	Aspirin	No Aspirin	P
Variable	(n = 2310)	(n = 3864)	Value
mographics			
Age, mean (SD), y	62 (11)	56 (12)	<.001
Men, No. (%)	1779 (77)	2167 (56)	<.001
nical history	()	()	
Diabetes, No. (%)	388 (17)	432 (11)	<.001
Hypertension, No. (%)	1224 (53)	1569 (41)	<.001
Tobacco use, No. (%)	234 (10)	500 (13)	.001
Prior coronary artery disease, No. (%)	1609 (70)	778 (20)	<.001
Prior coronary artery bypass graft, No. (%)	689 (30)	240 (6)	<.001
Prior percutaneous coronary intervention, No. (%)	667 (29)	148 (4)	<.001
Prior Q-wave MI, No. (%)	369 (16)	285 (7)	<.001
Atrial fibrillation, No. (%)	27 (1)	55 (1)	.04
Congestive heart failure, No. (%)	127 (6)	178 (5)	.12
dication use			
Digoxin use, No. (%)	171 (7)	216 (6)	.004
β-Blocker use, No. (%)	811 (35)	550 (14)	<.001
Diltiazem/verapamil use, No. (%)	452 (20)	405 (10)	<.001
Nifedipine use, No. (%)	261 (11)	283 (7)	<.001
Lipid-lowering therapy, No. (%)	775 (34)	380 (10)	<.001
ACE inhibitor use, No. (%)	349 (15)	441 (11)	<.001
rdiovascular assessment and exercise capacity Body mass index, mean (SD), kg/m²	29 (5)	30 (7)	<.001
Ejection fraction, mean (SD), %	50 (9)	53 (7)	<.001
Resting heart rate, mean (SD), beats/min	74 (13)	79 (14)	<.001
Resting blood pressure, mean (SD), mm Hg Systolic	141 (21)	138 (20)	<.001
Diastolic	85 (11)	86 (11)	.04
Purpose of test to evaluate chest pain, No. (%)	300 (13)	468 (12)	.31
Mayo Risk Index ≥1, No. (%)†	2021 (87)	2517 (65)	<.001
Peak exercise capacity, mean (SD), METs Men	8.6 (2.4)	9.1 (2.6)	<.001
Women	6.6 (2.0)	7.3 (2.1)	<.001
Heart rate recovery, mean (SD), beats/min	28 (11)	30 (12)	<.001

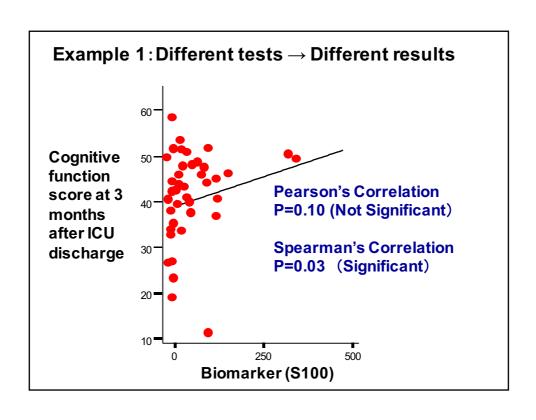
# Question 1 — Univariate? (cont'd)

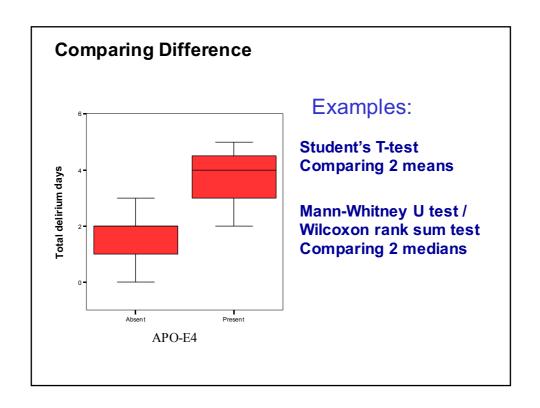
Randomization prevents confounding. Thus, in general, confounders are more problematic in observations studies than RCT's.

If you want to adjust for confounders, then you need to perform **regression analysis**.

# Question 2 -Difference?

- Do you want to test for a difference between groups or want to test for correlation between variables?
- -Comparing mean (or median) of two groups?
- -Correlation between two variables in one group?





# Question 3 - Paired?

 Were the groups paired or unpaired / (dependent or independent)?

Are you measuring more than once from one sample?

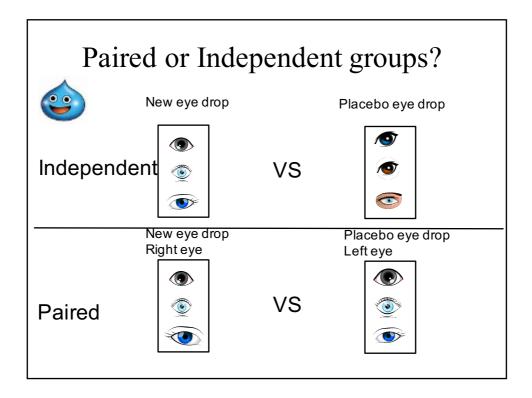
## **Examples:**

Student t-test comparing 2 independent means.

(Comparing outcome between intervention and control groups)

Paired t-test comparing 2 related means.

(Comparing outcome before and after an intervention).



# Question 4 - Outcome Type?

- What is the level of measurement for outcome variable?
  - -Continuous (Interval)? Ex. Blood pressure, BMI, Weight
  - Discrete/Categorical/Factor?

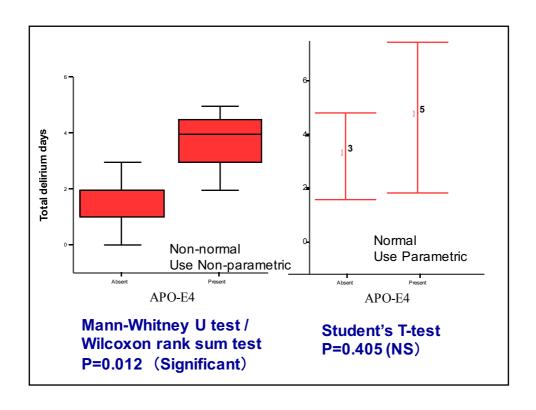
-Nominal? 2 levels (Binary, dichotomous) ex. Died / Survived

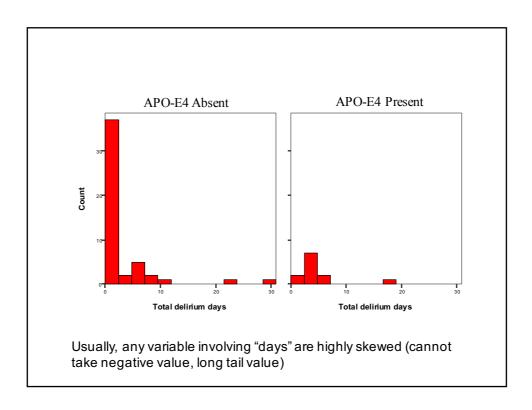
>2 levels. Ex. Disease Type (cancer, DM, cardiovascular)

-Ordinal?

> 2 levels. Ex. Disease severity (1: Mild, 2: Moderate, 3: Severe)
Disease score (0: normal, 10: abnormal)

# • If an outcome variable is continuous, is it normally distributed? If your histogram forms a bell-shaped curve, assume that it is normal; otherwise, assume that it is non-normal. Histogram Normal Non-normal / skewed





## Parametric Tests are valid only when...

Student t-test, ANOVA are valid only when outcome variable is normally distributed within a group.

Paired t-test (for example comparing BP before after an intervention) is valid only when within-patient difference in outcome variable (e.g.BP) is normally distributed.

Linear regression is valid only when residuals (difference between observed and predicted values) are normally distributed.

Pearson-correlation analysis is valid only when both (outcome and exposure) variables are normally distributed.

Non-Parametric Tests are always valid regardless of distribution of data

# Statistical Methods Recommendation by New England Journal of Medicine

The basis for these guidelines is described in Bailar JC III, Mosteller F. Guidelines for statistical reporting in articles for medical journals: amplifications and explanations. Ann Intern Med 1988;108:266-73. Exact methods should be used as extensively as possible in the analysis of categorical data. For analysis of measurements, nonparametric methods should be used to compare groups when the distribution of the dependent variable is not normal.

This page can be found at <a href="http://authors.nejm.org/Misc/NewMs.asp#statistics">http://authors.nejm.org/Misc/NewMs.asp#statistics</a>

# Question 6 - #groups?

- How many groups are there for the independent (predictor) variable?
  - 2 levels?
  - 3 or More?

## **Examples**:

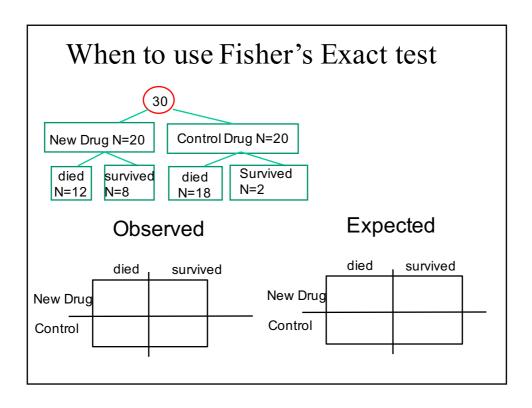
Student t-test comparing 2 group means ANOVA comparing 3 or more group means

# Question 7 - Sample Size?

What is the total sample size?

## **Examples:**

Greater than total N=20, use Chi-square test Greater than 20 and less than 40 and an expected # in a cell < 5, use Fisher's exact test



## Selection of Regression

Only depends on the following 2 things:

- type of outcome variable
- Whether data are paired or not (Repeated or not).

Not repeated Repeated

Continuous Linear Mixed effect

Binary Logistic GEE

Time to Events Cox MULCOX

### Flow-chart for popularly used statistical tests Q 1,U n ivariate /M utivariab le Q4,Q5 Type of outcome (Normality) Q 7,sample size D ifference /C orre latio Q 6, No. of Paired, related Valid Tests Student's t-tes Continuous Nomal 0 ne-way ANOV Continuous Non-nomal) 0 rdered categorical (un-paired) Nom inal Chi-square test Log-Rank test Kaplan-Meierpbt) Time to Event D ifference Univariate Continuous Normal) ixed effect Regression (paired) rdered categorical Pearson's correlation 🚯 Corre latio Spearman's correlation (s) Linear Regression\* Ordered Logistic Regression Binary Logistic Regression Multinom al Logistic Regression Cox Proportional Hazard Regression Continuous Non-normal residula M u Itivariab le inear Mixed Effect Regression Linear Mixed Effect Regression Generalized Estimation Equation (GEE) Generalized Estimation Equation (GEE) (baired) Created based on Publishing Your Medical Research Paper, by Daniel Byrne, Williams and W

1.3 Tutorials for selecting valid statistical tests

## Example 1

 Comparing ventilator free days between patients who were randomized to daily awakening and breathing trial vs daily breathing trial among ventilated patients in medical ICU: A prospective randomized study.

```
Q1: (Univariate?) Univariate 

Mutivariable 

Linear regression

Q2: (Difference?) Difference

Q3: (Paired?) Unpaired

Q4: (Type?) Continuous

Q5: (Normality?) Normal 

Non-Normal

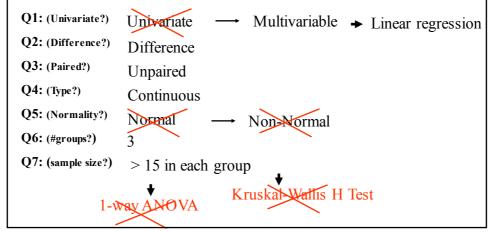
Q6: (#groups?) 2

Q7: (sample size?) > 30 in each group

Student: I-test Mann-Whitney U Test
```

# Example 2

• Cytokine responses of peripheral blood mononuclear cells (PBMC) from HIVseronegative adults with prior extra pulmonary TB were compared with responses from persons with prior pulmonary tuberculosis and latent *M. tuberculosis* infection in a case-control study. Antas, *Journal of Allergy and Clinical Immunology*. 2006.



## Example 3

• We want to estimate the relationship between two numerical measures: Bio-marker value for S100 and patient's cognitive scores measured at 3 months after ICU discharge among patients in medical ICU.

```
Q1: (Univariate?)
                 Univariate → Multivariable → Linear regression
Q2: (Difference?)
                 Correlation
Q3: (Paired?)
                 NA
Q4: (Type?)
                 Continuous
Q5: (Normality?)
                 Normal
Q6: (#groups?)
                 1 group
Q7: (sample size?)
                  > 30 in each group
         Pearson's r
                                  Spearman's o
         Correlation coefficient
                                  Rank Correlation coefficient
```

## Example 4

Martinez-Picado et. al. compared proportion of patients with HIV infection who had viral surge between alternation of antiretroviral drug regimens and standard regimens. A Randomized, Controlled Trial. Annals of Internal Medicine. 2003

→ Multivariable → Logistic regression Q1: (Univariate?) Univariate -Q2: (Difference?) Difference Q3: (Paired?) Unpaired **Q4:** (Type?) Nominal Q5: (Normality?) NA **Q6:** (#groups?) < 20 Q7: (sample size?) Fisher's Exact test

## Example 5

A researcher wants to evaluate the effect of a new diet on weight loss by comparing patient's weight before and after the diet program.

Q1: (Univariate?) Univariate Q2: (Difference?) Difference Q3: (Paired?) Paired **Q4:** (Type?) Continuous Q5: (Normality?)

Normal Non-Normal

**Q6:** (#groups?)

Q7: (sample size?) > 30 in each group