

Matched Studies in Medical Research

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What is a Matched Pairs Design?

- Data consisted of observations of treatment outcome and control outcome on subjects that are paired
- Pairing is done in the hope that all other factors are the same within a pair.
- Comparison of treatment and controls is between like subjects



Examples

Biological Matching

- Diabetic Retinopathy LASER
 - Patient eyes randomized to treatment or not.
 - Event time to loss of vision
- Effects of Skin Graft HLA matching on burn patients
 - Patient's with extensive burns given grafts which are 8/8 match or mismatched HLA
 - Time to graft failure measured



Examples

Biological Examples

- Study of a surgical device to show tumor cells
 - Mice have tumor implanted in one flank
 - Mouse injected with radioactive iodine. Theory is tumor will pick up iodine have higher radioactivity count then opposite side.
 - Small pen like counter used to measure radioactive count
 - Experiment complicated by iodine absorption in thymus

Examples

Tests Based on Matched subjects

- Comparison of drug 6-MP with placebo (Freireich et al Blood 1963)
 - Multicenter trial of 6-MP as a remission maintenance therapy for children with acute leukemia
 - At each hospital patients in remission following prednisone therapy matched on disease status and one of pair randomized to 6-MP one to placebo
 - Study measured time to relapse



Tests Based on matched subjects

Retrospective studies

- Studies using retrospective large cohort samples
- Number of treated cases is small
- Number of control cases is large
- Each treated case is matched on some key risk factors to a treated case



Test Based on matched studies

Prospective studies

- Studies require a relatively homogenous population so it is easy to find a match
- Can match on only a few characteristics



Advantages of Design

- Allows comparison of like to like patients
- Allows additional data collection on smaller cohort of patients
- Simpler to understand



Disadvantages of design Retrospective Studies

- Don't use all the data
 - Cases without control deleted
 - In survival outcomes some pairs with censored outcomes are deleted
- Can not examine risk factors used to match subjects
- Outcome may depend on how you matched



Disadvantages of design Prospective Studies

- Logistics
 - Need to find match
 - What to do while waiting
 - How to randomize
 - Need similar measurement for each pair
- Dropouts
 - What to do with pair when there is a drop-out—
keep as solo, drop pair, find new match



Alternatives to Matched Designs

- Regression Adjusted Analysis
- Stratified Analysis
- Propensity Score Adjusted Designs
 - Fit Logistic regression model to chance a subject got treatment
 - Predicted probability is a propensity score
 - Stratify analysis, match on propensity score, use propensity in regression to make adjustment for risk factors



Example of Matched Pair Design Crossover Designs

- Two treatments A and B
- Patients randomized to one of two scenarios
 1. Treatment A -> washout -> Treatment B
 2. Treatment B -> washout -> Treatment A
- If there is no carryover effect (Effect of A in 1 same as effect in 2) then the crossover study is analyzed as a matched pairs using

Tests in Crossover Design

μ_j —Patient effect τ —Effect of Treatment A

λ_A (λ_B)—Carryover effect of A (B) in Period 1

	Period	Period	Crossover Difference	
	1	2	Trt A-Trt B	Sum
A/B	$\mu_j + \tau$	$\mu_j + \lambda_A$	$\tau - \lambda_A$	$2\mu_j + \tau + \lambda_A$
B/A	μ_k	$\mu_k + \tau + \lambda_B$	$\tau + \lambda_B$	$2\mu_k + \tau + \lambda_B$

- Comparison of sums in two arms tests for carryover effect—Independent two sample test
- No carryover use paired test on crossover differences
- Significant carryover effect use independent two sample test on period 1 data only

Advantages of Crossover Trials

No Carryover

- To obtain the same number of observations as a parallel design fewer patients need to be recruited
- To obtain the same power or precision as a parallel design fewer patients are needed



Disadvantages of Crossover Designs

- Dropouts
- Not reasonable for disease where the patient may deteriorate over time
- Complicated Analysis
- Period by treatment interactions
- Carryover effects
- For last two problems the data in the first period only is used

Approach 1 to Analysis of Paired Data

Data

$$(X_1, Y_1), \dots, (X_n, Y_n)$$

- Compute difference between individuals within a pair. Base tests on $d_i = (X_i - Y_i)$. Test if the d_i 's are sampled from a population centered at zero
- Examples of tests for continuous data
 - Paired t-test
 - Sign test
 - Sign Rank Test
 - McNemar's test



Approach 2 to Analysis of Paired Data

Data

$$(X_1, Y_1), \dots, (X_n, Y_n)$$

X has mean μ_X (M_X) Variance σ_X^2 Y has mean μ_Y (M_Y) Variance σ_Y^2

$$\text{Cov}(X, Y) = \sigma_{XY}$$

- Test based on $(M_X - M_Y)$
- Variance of $(M_X - M_Y) = \text{Var}[M_X] + \text{Var}[M_Y] - 2 \text{Cov}[M_X, M_Y]$
- Test Statistic

$$T = (M_X - M_Y) / (S_X^2/n + S_Y^2/n - 2 * S_{XY}/n)$$

Two Approaches with Normal Data

- $(M_x - M_y)$ = average values of the d's in approach 1
- $\text{Var}[M_x - M_y]$ = Variance of d's in approach 1
- Two tests give same result
- Note when $S_{xy} = 0$ the T test is not the usual two sample t-test in textbooks since that assumes equal variances

Affect of Incorrect Use of Unpaired t-test

- Paired samples of size 20
- Data Bivariate Normal (1,1), $\sigma_x=\sigma_y=1$, Correlation ρ , 100,000 samples

ρ	Unpaired	Paired	ρ	Unpaired	Paired
-0.9	0.157	0.048	0.9	0.000	0.049
-0.8	0.145	0.051	0.8	0.000	0.049
-0.7	0.133	0.050	0.7	0.001	0.050
-0.6	0.120	0.050	0.6	0.004	0.050
-0.5	0.108	0.051	0.5	0.007	0.049
-0.4	0.096	0.050	0.4	0.013	0.050
-0.3	0.087	0.051	0.3	0.021	0.051
-0.2	0.073	0.050	0.2	0.030	0.051
-0.1	0.061	0.049	0.1	0.039	0.050
0	0.051	0.050			

Comparison of Number of Patients needed- Paired vs. parallel design

- Assume testing mean difference =0 versus mean difference = Δ
- Two sided test with 5% type one error
- Data normal with standard deviations of 1
- Either use paired t-test for paired data test or an unpaired t-test with assumed equal variances for the parallel design
- Values from Proc Power in SAS



Comparison of Sample Sizes Needed

	Difference in Means = 0.5		Difference in Means=1.0	
Paired Design---Number of Pairs				
rho	80% power	90% power	80% power	90% power
-.5	97	129	26	34
-.3	84	112	23	30
-.1	72	95	20	26
0	65	87	18	24
.1	59	78	17	21
.3	46	61	14	17
.5	34	44	10	13
Parallel Design				
N per arm	64 per arm	86 per arm	17 per arm	23
N total	128 patients	172 patients	34 patients	46 patients

Examples

Biological Examples

- Study of a surgical device to show tumor cells
 - Mice have tumor implanted in one flank
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Number of radioactive counts in 60 seconds

control flank	Tumor flank	difference	Rank of diff	Sign
117	121	4	1	+
279	336	57	8	+
259	400	141	11	+
432	521	89	10	+
455	399	-56	7	-
601	798	197	12	+
29	43	14	4	+
58	69	11	3	+
93	114	21	5	+
88	156	68	9	+
132	174	42	6	+
159	169	10	2	+

Sign Test

- H_0 : Median Difference=0
 H_A : Median Difference > 0
- Test based on the number of positive differences $B = 11$ $n=12$
- Reject H_0 if B is too large
- p-value $\Pr[B > b_{\text{obs}} \mid p=1/2]$ with $B \sim \text{Binomial}$ or
- $\Pr[Z > (b_{\text{obs}} - (n/2)) / \{n/4\}^{1/2}]$ if n is large $Z \sim \text{Normal}[0,1]$

Here $p = \Pr[B \geq 11 \mid n=12, p=1/2] = 2p^{12} + p^{12} = .003174$



Wilcoxon Sign Rank Test

- H_0 : Median Difference=0
 H_A : Median Difference > 0
- Rank Absolute Values of Differences--- R_i rank of i^{th} pair
- Add up ranks associated with positive differences T^+
- Compute $E_0[T^+]=n(n+1)/4$, $\text{Var}_0[T^+]=\{n(n+1)(2n+1)\}/24$
- Standardized test statistic is $Z=\{T^+-E_0[T^+]\}/(\text{Var}_0[T^+])^{1/2}$
- $p\text{-value}=\Pr[Z>z]$, $Z\sim\text{Normal}(0,1)$

- In example $T^+=71$, $E_0[T^+]= 39$, $\text{Var}_0[T^+]=162.5$, $z=2.51$, $p=0.006$

Binary Data

McNemar Test

- Comparison of two skin creams
 - Put different cream on each arm
 - Measure yes or no did cream cure rash

		Cream A	
		yes	no
cream B	yes	75	45
	no	25	35

- Test based on $n=25+45=70$ discordant pairs

Binary Data

McNemar Test

- If no difference in treatments the chance of A yes, B No= chance of A no, B yes=1/2
- Test statistic based on $p = \{\text{Number A no, B yes}\} / n$
- Test statistic $Z = \{p - 1/2\} / \{.5/n^{1/2}\}$
- Here $p = 25/70 = 0.357$
- $Z = -2.39$
- $p = 2 * \Pr[Z > -2.39] = .0168$

Paired Survival Data

CTSI Supplemental Grant

- Paired data problems are more complex due to censoring
- Major complication is that in most techniques for comparison pairs where the patient with the smallest on study time is censored are omitted
- Coming soon an annotated bibliography of techniques on the CTSI webpage



Summary

- Paired data designs are a useful tool in medical studies
 - if they are analyzed by proper statistical techniques
 - if there is no expectation of studying variables patients are matched on
 - if the data is biologically matched
 - for crossover designs with no carryover effect

Summary

- Paired data designs may not be the best when they are drawn from large data bases
- Paired data designs require more logistical work than parallel data designs
- Paired data designs may suffer a loss of efficiency when patients drop out
- For many parameters point and interval estimation in paired designs is hard to do

Resources

- The **Clinical and Translation Science Institute (CTSI)** supports education, collaboration, and research in clinical and translational science: www.ctsi.mcw.edu
- The **Biostatistics Consulting Service** provides comprehensive statistical support www.mcw.edu/biostatistics.htm



Free Drop-In Consulting

- **MCW:** Tuesdays & Thursdays 1– 3 pm
 - Health Research Center, H2400
- **Froedtert:** Mondays, Wednesdays, Fridays 1 – 3 pm
 - Froedtert Pavilion, L772A- TRU offices
- **VA:** Every Monday, 9:30 – 10:30 am
 - VA Medical Center, Room 70-A 314-A
- **Marquette:** Every Tuesday, 8:30 – 10:30 am
 - School of Nursing-Clark Hall, Office of Research & Scholarship: 112D



Questions?

