

Non-Parametric Analysis

27 Nov 2009

CPSY501

Dr. Sean Ho

Trinity Western University

Please download:

- ***relationships.sav***
- ***Field-Looks_Charis.sav***

Outline for today

- Non-Parametric **Between-subjects** Tests:
 - 2 groups: **Mann-Whitney U**
 - Many groups: **Kruskall-Wallis H**
 - **Factorial** variants
- Non-Parametric **Within-subjects** Tests:
 - 2 repetitions: **Wilcoxon Signed-rank**
 - Many repetitions: **Friedman's ANOVA**
- There exist other non-parametric options, too!

When To Use Non-Parametric

- Review: assumptions of ANOVA?
 - If met, parametric methods more power
 - ANOVA is pretty robust to non-normality
 - Try transforming data, or multi-level model
- Consider non-parametric alternatives if:
 - DV is ordinal, or
 - Violations are severe enough to affect result
- Non-parametric often use ranks, not raw scores
 - e.g., Spearman vs. Pearson correlation
 - Comparing medians rather than means

Mann-Whitney U Test

- DV: ordinal or scale, but non-parametric
- IV: dichotomous, between-subjects (2 groups)
- RQ: Is there a median difference between the two groups?
 - Non-parametric alternative to *t*-test
- What research designs might need this test?
- Data entry format:
 - 2 variables entered for each participant:
 - IV: group membership (0 or 1)
 - DV: outcome measure

Mann-Whitney U: SPSS

- **Dataset:** relationships.sav
- Analyze → Nonparametric → 2 Independent:
 - **Test Variable:** DV (quality of communication)
 - **Grouping Variable:** IV (having counselling)
 - Use “**Define Groups**” to choose **two** groups for comparison: **1** = no counsel; **2** = had counsel
 - **Test Type:** “**Mann-Whitney U**”
- Optional SPSS module for **Exact Tests**:
Computationally-expensive but more precise, especially for small / unbalanced designs

Mann-Whitney U: Output

“There was no significant effect of **Having Counselling** on **Quality of Communication**, $U = 311.00$, $p = .059$, $Mdn_H = 4.0$, $Mdn_N = 3.0$.”

- Use Analyze → Descriptives → Explore to get group **medians**

Test Statistics ^a	
	Quality of Communication
Mann-Whitney U	311.000
Wilcoxon W	1014.000
Z	-1.890
Asymp. Sig. (2-tailed)	.059

a. Grouping Variable: Had any counselling

Mann-Whitney U: Effect Size

- Effect size must be calculated manually, using the following formula:
 - $r = Z / \sqrt{N}$
 - e.g., $r = -1.89 / \sqrt{60} \approx -0.24499$
- Use existing research or Cohen's effect size "estimates":
 - "There is a small difference between the therapy and no therapy groups, $r = -.24$ "

Practise: Mann-Whitney U

- You try it!
- Is there a significant **difference** between spouses who report **communication** problems and spouses who have not (“**Com_prob**”), in terms of the level of **conflict** they experience (“**Conflict**”)?
- What is the **size** of the effect?

Kruskal-Wallis Test

- DV: ordinal or scale, but non-parametric
- IV: categorical, between-subjects (many groups)
- RQ: Is there a median difference amongst the groups?
 - Alternative to One-way ANOVA
- What research designs might need this test?
- Data entry format:
 - 2 variables entered for each participant:
 - IV: group membership
 - DV: outcome measure

Kruskal-Wallis: SPSS

- Analyze → Nonparametric → K Independent:
 - Test Variable: DV (level of conflict)
 - Grouping Variable: IV (type of counselling)
 - Test Type: “Kruskal-Wallis H”
 - Define Groups: specify highest and lowest group numbers to be compared
- Optional SPSS module for Exact Tests:
Computationally-expensive but more precise, especially for small / unbalanced designs

Kruskal-Wallis: Output

“Type of counselling has a significant effect on participants' level of conflict,
 $\chi^2(2) = 7.09$, $p = .029$.”

- Also report medians and post hoc results...

Test Statistics^{a,b}

	Level of Conflict
Chi-Square	7.094
df	2
Asymp. Sig.	.029

a. Kruskal Wallis Test

b. Grouping Variable: Type of Counselling

Kruskal-Wallis: Follow-up

- K-W test is omnibus: if significant, follow-up with Mann-Whitney tests on pairs of groups
 - Must do Bonferroni correction manually, or else Type 1 error inflates: lower the level of significance to $0.05 / (\# \text{ comparisons})$
- Try planned contrasts based on
 - Theory / literature,
 - Your research question; or
 - Ordering groups by mean ranks, and comparing each group to next highest group

Kruskal-Wallis: Effect Size

- Overall effect size not so useful; instead,
- Calculate effect size for each pair of groups that differs significantly (in Mann-Whitney follow-up)
 - $r = Z / \sqrt{n}$
 - Where the sample size n is the total number of participants in this pair of groups

Kruskal-Wallis: Reporting

- “Type of Counselling has a significant effect on participants' level of conflict, $\chi^2(2) = 7.09$, $p = .029$.”
- “Specifically, the No Counselling group had higher conflict scores, $Mdn_N = 4.0$, than did the Couples Counselling group, $Mdn_C = 3.0$, Mann-Whitney $U = 176.5$, $Z = -2.61$, $p = .009$, $r = -.37$.”
- Field uses “ H ” for the K-W: $H(2) = 7.09$
- Note: Bonferroni correction: $\alpha = .05 / 3 \approx .017$

Kruskal-Wallis: Non-significant

- If Kruskal-Wallis gives a **non-significant** result, but the research question behind the analysis is still “important”:
- The problem might be with **low power**, so
- **Descriptive** follow-up analyses can be helpful.
- See the illustration for **Friedman’s ANOVA** below for some clues.

Parametric vs. Non-Parametric

- There's nothing wrong with running **both** parametric and non-parametric tests!
- **Comparison** of **non-parametric** tests with the corresponding **ANOVA** may be able to lend more **confidence** in the overall adequacy of the patterns reported.
- Nonparametric analyses tend to have **less power** for well-distributed DVs, but they can be more **sensitive** to effects when the DV is, for instance, truly **bimodal**!

Param vs. Non-Param: ex.

- If we do the same analysis parametrically:
 - DV: Level of Conflict
 - IV: Type of Counselling
 - One-way ANOVA, with Levene's test and Bonferroni post-hoc
- Results are similar: $F(2, 57) = 4.05, p = .023$
 - The No Counselling group shows more conflict than the Couples Counselling group, $M_N = 3.87$ and $M_C = 3.04$
- Confirms results with non-parametric analysis

Factorial Between-Subjects

- SPSS doesn't have **factorial** between-subjects non-parametric analyses built-in!
- Try **creating** one IV that encodes all **cells**:
e.g., **2x3 factorial** → **6-level one-way**, use **K-W**
 - (-) **Follow-up** gets hard with lots of groups
 - (-) **Moderation**/interaction analysis is harder
 - (+) **Flexibility** in defining groups
 - (+) Target specific cells / **interactions**
- Or: Run **separate** K-W tests for each IV
- Or: Convert to **ranks** and use **log-linear** analysis

Factorial Analysis: Example

- Research question: How do Marital Status and Type of Counselling relate to Conflict Levels?
- Check cell sizes with Analyze → ... → Crosstabs:
 - The smallest cells (Individual Counselling) have 5-6 people per group, that is enough
- 2x3 factorial → convert to 1 IV with 6 levels:
 - Indiv. Counsel & Married; Indiv & Divorced; Couples Couns. & Marr; Couple & Div; etc.
 - Transform → Recode into Different, try the “If” conditions

Factorial Analysis: Output

- The **Kruskal-Wallis** test for the combined variable is **not significant**.
- This suggests that the **significant** effect for **Counselling Type** is **masked** when combined with **Marital Status**.

Test Statistics(a,b)

	Level of Conflict
Chi-Square	8.753
df	5
Asymp. Sig.	.119

a. Kruskal Wallis Test

b. Grouping Variable: Counselling Type & Marital Status

Factorial Analysis: Main effect

- The idea of a “masking effect” of Marital Status shows as well when we test that main effect alone.

Test Statistics(a)	
	Level of Conflict
Mann-Whitney U	337.000
Wilcoxon W	802.000
Z	-1.752
Asymp. Sig. (2-tailed)	.080
a. Grouping Variable: Marital Status	

Theory-Guided Interaction Test

- Divorced & No counselling group:
 - We might assume this group should have **high** conflict levels!
 - **Compare** with some of the other 5 groups using **Mann-Whitney** U tests
- This can be a “theoretically guided” replacement for **interaction** tests in non-parametric factorial analysis.
- The choice depends on **conceptual** relations between the IVs.

Kruskall-Wallis: Practise

- Does number of children (range 0-3) have a significant effect on quality of marital communication?

Wilcoxon Signed-rank Test

- DV: ordinal or scale, but non-parametric
- IV: within-subjects, with 2 repetitions
- RQ: Is there a median change between times?
 - Parallel to paired-samples *t*-test
- What research designs might need this test?
- Data entry format:
2 variables entered for each participant:
 - DV at first measurement
 - DV at second measurement

Wilcoxon Signed-rank: SPSS

- Analyze → Nonparametric → 2 Related Samples:
 - Test Pairs: DV before and DV after
 - Test Type: Wilcoxon
- Practise: does level of conflict decrease from pre-therapy (Pre-conf) to post-therapy (Conflict)?
- Multiple pairs may be specified in one analysis
 - But no built-in Bonferroni correction
- Exact tests also available with add-on module

Wilcoxon Signed-rank: Output

■ There was a significant **reduction** in level of conflict after therapy, $T = 4.5$, $p = .002$.

● Or: $Z = -3.09$, $p = .002$ (& calc. effect size)

Ranks

		N	Mean Rank	Sum of Ranks
Pre-therapy level of Conflict - Level of Conflict	Negative Ranks	1 ^a	4.50	4.50
	Positive Ranks	13 ^b	7.73	100.50
	Ties	46 ^c		
	Total	60		

a. Pre-therapy level of Conflict < Level of Conflict

b. Pre-therapy level of Conflict > Level of Conflict

c. Pre-therapy level of Conflict = Level of Conflict

	Pre-therapy level of Conflict - Level of Conflict
Z	-3.094^a
Asymp. Sig. (2-tailed)	.002

a. Based on negative ranks.

b. Wilcoxon Signed Ranks Test

Wilcoxon Signed: Effect Size

- Effect size must be calculated manually, using the following formula:
 - $r = Z / \sqrt{N}$
 - e.g., $r = -3.09 / \sqrt{120} \approx -0.28$
- N is total number of observations, not participants! (60 participants) * (2 observation)
- Use existing research or Cohen's effect size "estimates":
 - "The reduction in level of conflict after therapy was significant but small, $r = -.28$ "

Wilcoxon Signed-rank: Practise

- Do **levels of conflict** change significantly between **pre-therapy** (Pre_conf) and **1 year** after therapy (Follow_conf)?
- If so, calculate the **size** of the effect.
 - Participant **attrition** at time 3 (Follow_conf) affects the total number of observations!
- **Example** reporting: “There was a significant reduction in level of conflict after therapy, $T = 4.5$ [or: $Z = -3.09$], $p = .002$, $r = -.28$.”

Friedman's ANOVA

- DV: ordinal or scale, but non-parametric
- IV: within-subjects, several repetitions
- RQ: Is there a median change over time?
 - Parallel to repeated-measures ANOVA
- What research designs might need this test?
- Data entry format:
one variable for each repetition of the measure
 - DV at first measurement
 - DV at second measurement
 - DV at third measurement, etc.

Friedman's ANOVA: SPSS

- Analyze → Nonparametric → K Related Samples:
 - Test Variables: each repetition of DV
 - Test Type: "Friedman"
- No easy way in SPSS to do non-parametric mixed-design analysis
- Optional SPSS module for Exact Tests:
Computationally-expensive but more precise, especially for small / unbalanced designs

Friedman's ANOVA: Output

- “Levels of conflict changed significantly over time, $\chi^2(2, N = 57) = 9.07, p = .011$.”
- “Specifically...” [report *post hoc* results...]

Test Statistics^a

N	57
Chi-Square	9.065
df	2
Asymp. Sig.	.011

a. Friedman Test

Friedman's ANOVA: Follow-up

- If Friedman's is significant, **follow-up** with a series of **Wilcoxon Signed-ranks** tests:
- Either do **all** pairs of levels of the RM (**post-hoc**):
 - Remember manual **Bonferroni** correction:
 $\alpha = .05 / (\# \text{ comparisons})$
 - But Bonferroni is often too **conservative**
- Or target specific **planned** comparisons (better):
 - e.g., (**time1 vs. time2**), (**time2 vs. time3**), ...
k-levels of IV → only **k-1** comparisons
 - Useful if **power** is low, e.g. many **cells**:
control **Type II** error

Friedman's ANOVA: Example

- IV: Time (3 cells)
- Targeted follow-up analyses:
 - Run 3 Wilcoxon's signed-rank tests
 - Bonferroni correction: $\alpha = .05 / 3 \approx .017$
- Pre vs. Post: $Z = -3.09, p = .002, r = -.28$
- Pre vs. 1yr: $Z = -2.44, p = .015, r = -.22$
- Post vs. 1yr: not significant
- Conclusion: improvement after therapy is maintained at the follow-up assessment.

Friedman's ANOVA: Reporting

- “Levels of conflict significantly changed over time, $\chi^2 (2, N = 57) = 9.07, p = .011$.
- “Specifically, conflict decreased from pre-therapy levels at post-therapy observations, $Z = -3.09, p = .002, r = -.28$,
- “and levels remained below pre-therapy conflict levels one year later, $Z = -2.44, p = .015, r = -.22$.”

Friedman's: Non-significance

- If Friedman's is not significant:
 - Is it because of low power (Type II error)?
- Run a series of Wilcoxon Signed-ranks tests, but
 - Focus on effect sizes, not significance levels
- If the effect sizes are “moderate”, say $> .25$, then the results could be worth reporting.
- Enough detail should be reported to be useful for future meta-analyses.
- This could be true of any analysis; we just use Friedman's to illustrate

Friedman's ANOVA: Practise

- **Dataset:** Field-Looks_Charis.sav
 - “Looks or Personality” (Field text)
 - **Two** RM IVs in the dataset
- RQ: Is there a significant difference between participants' **judgements** of people who are of **average physical appearance**, but present as **dull** (“ave_none”); somewhat **charismatic** (“ave_some”), or as having **high charisma** (“ave_high”)?
- If so, conduct **follow-up** tests to identify where the specific differences lie.

Summary: Between-Subjects

Parametric

- *t*-test between groups (1 IV, dichotomous)
- One-way ANOVA (1 IV, many levels)
 - Post-hoc: *t*-tests
- Factorial ANOVA (several IVs)
 - Post-hoc: *t*-tests

Non-Parametric

- Mann-Whitney *U* test
- Kruskal-Wallis *H* test
 - Post-hoc: M-W
- K-W with encoding of cells in one IV
 - Post-hoc: M-W

Summary: Within-Subjects

Parametric

- Paired/related *t*-test (1 RM IV w/2 levels)
- Repeated-ms ANOVA (1 RM IV with many levels)
 - Post-hoc: paired *t*-tests between levels of RM

Non-Parametric

- Wilcoxon Signed-Rank
- Friedman's ANOVA
 - Post-hoc: Wilcoxon Signed-rank between levels of the RM