

Nonparametric Tests: Methods, Tests, and Effect Sizes

1. Nonparametric Tests: Design × Test × Effect Size

Design Type	Overall Test	Test Statistic	Overall Effect Size	Post Hoc Test	Pairwise Effect Size
Two Independent samples	Mann–Whitney U Test	U (or W)	Rank–biserial correlation(r_{tb})	–	–
	(Wilcoxon Rank–Sum Test)				
Multiple Independent Samples	Kruskal–Wallis Test	$H(\chi^2 \text{ approximation})$	Epsilon squared (ϵ^2 , $\eta^2[H]$)	Pairwise Mann – Whitney U tests	Rank–biserial correlation (r_{tb})
Two paired samples	Wilcoxon Signed–Rank Test	V (or Z)	Rank–biserial correlation (r_{tb})	–	–
Multiple paired samples	Friedman Test	χ^2_F	Kendall’ s W	Pairwise Wilcoxon signed–rank tests	Rank–biserial correlation (r_{tb})

2. Effect Size Definitions

2.1 Rank-Biserial Correlation (r_{rb})

Used for: Mann – Whitney U test, Wilcoxon signed-rank test.

Range: -1 to 1 , Interpretation: 0 = complete overlap, ± 1 = complete dominance.

Magnitude guidelines : Small: $|r| \approx 0.10$, Moderate: $|r| \approx 0.30$, Large: $|r| \geq 0.50$.

2.2 Epsilon Squared (ϵ^2)

Used for: Kruskal – Wallis test.

Range: 0 to 1 , Meaning: Proportion of rank variance explained by group membership.

Magnitude guidelines: Small: $\epsilon^2 \approx 0.01$, Moderate: $\epsilon^2 \approx 0.06$, Large: $\epsilon^2 \geq 0.14$.

2.3 Kendall's W

Used for: Friedman test.

Range: 0 to 1 , Meaning: Degree of agreement / consistency across conditions.

Interpretation: no systematic differences: $W = 0$, perfect agreement: $W = 1$.

3. Table-Note

3.1 Two Independent Samples

A Mann – Whitney U test was conducted to compare differences between two independent groups. Effect size was quantified using the rank-biserial correlation (r_{rb}) with 95% confidence intervals, calculated using the effectsize package in R.

3.2 Multiple Independent Samples

A Kruskal – Wallis test was performed to examine group differences across multiple independent samples. The overall effect size was estimated using epsilon squared (ϵ^2). Post hoc pairwise comparisons were conducted using Mann – Whitney U tests with Holm adjustment for multiple testing. Rank-biserial correlations with 95% confidence intervals were reported for all pairwise comparisons. All effect sizes were computed using the effectsize package in R.

3.3 Two Paired Samples

A Wilcoxon signed-rank test was conducted to assess differences between two paired conditions. Effect size was quantified using the rank-biserial correlation (r_{rb}) with 95% confidence intervals, computed using the effectsize package in R.

3.4 Multiple Paired Samples

A Friedman test was conducted to examine differences across repeated-measures conditions.

Kendall's W was reported as the overall effect size. Post hoc pairwise comparisons were performed using Wilcoxon signed-rank tests with Holm adjustment. Rank-biserial correlations with 95% confidence intervals were calculated for all pairwise comparisons using the effectsize package in R.

Appendix A

R-Script for Independent-Sample

```
# -----  
  
# Load required R packages  
  
# -----  
  
library(effectsize) # For effect size estimation and confidence intervals  
library(rstatix)    # For nonparametric tests and post hoc comparisons  
library(dplyr)      # For data manipulation (data wrangling)  
  
#####  
  
# PART 1: Two Independent Samples  
  
# Nonparametric Test: Mann – Whitney U Test  
  
# (also known as Wilcoxon Rank-Sum Test)  
  
#####  
  
# -----  
  
# Prepare dataset for two independent groups  
  
# Dependent variable: Sepal.Length  
  
# Grouping variable: Species (two independent groups)  
  
# -----  
  
data_two <- iris %>%  
  filter(Species %in% c("setosa", "versicolor")) %>%  
  droplevels() # Remove unused factor levels  
  
# -----  
  
# Mann – Whitney U Test  
  
# Purpose:  
  
#   Test whether the distributions of two independent groups differ  
  
# -----  
  
cat("=== Mann – Whitney U Test Results ===\n")
```

```

mw_test <- wilcox.test(
  Sepal.Length ~ Species,
  data = data_two,
  exact = FALSE,      # Normal approximation (large samples)
  conf.int = TRUE,    # Confidence interval for location shift
  conf.level = 0.95
)

# -----
# Effect size for Mann – Whitney U Test
# Effect size measure: Rank-biserial correlation (independent)
# -----

effect_two <- rank_biserial(
  Sepal.Length ~ Species,
  data = data_two,
  ci = 0.95,
  verbose = FALSE
)

# Effect size name: Rank-biserial correlation (r_rb)
# Range: -1 to 1
# Interpretation:
#   r = 0 → complete overlap between groups
# r = ± 1 → complete separation between groups
# Meaning:
#   Difference in probability that a randomly chosen observation from one group exceeds the other

```

```
#####
```

```
# PART 2: Multiple Independent Samples
```

```
# Nonparametric Test: Kruskal – Wallis Test
```

```
#####
```

```
# -----
```

```
# Prepare dataset for multiple independent groups
```

```
# Dependent variable: Sepal.Length
```

```
# Grouping variable: Species (three independent groups)
```

```
# -----
```

```
data_multi <- iris %>%
```

```
  select(Sepal.Length, Species) %>%
```

```
  droplevels()
```

```
# -----
```

```
# Kruskal – Wallis Test
```

```
# Purpose:
```

```
#   Test whether at least one group distribution differs
```

```
# -----
```

```
cat("=== Kruskal – Wallis Test Results ===\n")
```

```
kw_test <- kruskal.test(
```

```
  Sepal.Length ~ Species,
```

```
  data = data_multi
```

```
)
```

```
print(kw_test)
```

```
# -----
```

```
# Overall effect size for Kruskal – Wallis Test
```

```

# Effect size measure: Epsilon squared (  $\epsilon^2$  )

# -----

kw_effectsize <- kruskal_effsize(

  data = data_multi,

  Sepal.Length ~ Species,

  ci = 0.95

)

print(kw_effectsize)

# Effect size name: Epsilon squared (  $\epsilon^2$  )

# Often reported as  $\eta^2[H]$  or  $\epsilon^2$ 

# Range: 0 to 1

# Interpretation (common guidelines):

#    $\epsilon^2 \approx 0.01 \rightarrow$  small

#    $\epsilon^2 \approx 0.06 \rightarrow$  moderate

#    $\epsilon^2 \geq 0.14 \rightarrow$  large

# Meaning:

#   Proportion of rank variance explained by group membership

# -----

# Post hoc pairwise comparisons

# Test: Mann – Whitney U (Wilcoxon rank-sum)

# Multiple comparison correction: Holm adjustment

# -----

cat("\n=== Post hoc Pairwise Wilcoxon Tests (Holm-adjusted) ===\n")

pairwise_test <- data_multi %>%

  pairwise_wilcox_test(

    Sepal.Length ~ Species,

```

```

    p.adjust.method = "holm"
  )

print(pairwise_test)

# -----
# Effect sizes for pairwise comparisons
# Effect size measure: Rank-biserial correlation (independent)
# -----

# Generate all pairwise combinations of groups
group_pairs <- combn(levels(data_multi$Species), 2, simplify = FALSE)

pairwise_effectsize <- lapply(group_pairs, function(g) {

  # Extract observations for each independent group
  x <- data_multi$Sepal.Length[data_multi$Species == g[1]]
  y <- data_multi$Sepal.Length[data_multi$Species == g[2]]

  # Compute rank-biserial correlation
  es <- rank_biserial(
    x, y,
    ci = 0.95,
    verbose = FALSE
  )

  es_df <- as.data.frame(es)

  # Assemble results
  data.frame(

```



```

    group1 = g[1],
    group2 = g[2],
    rank_biserial = es_df[[1]],
    CI_low = es_df$CI_low,
    CI_high = es_df$CI_high
  )
}) %>%
  bind_rows()

```

pairwise_effectsize

```

# Effect size name: Rank-biserial correlation (independent)

# Interpretation:

#    $|r| \approx 0.10 \rightarrow$  small
#  $|r| \approx 0.30 \rightarrow$  moderate
#  $|r| \geq 0.50 \rightarrow$  large

# CI interpretation:

#   Narrow CI  $\rightarrow$  precise dominance estimate
# CI including 0  $\rightarrow$  weak or ambiguous group difference

```

Appendix B

R-Script for Paired-Sample

```
# -----  
# Load required R packages  
# -----  
library(effects) # For effect size estimation and confidence intervals  
library(rstatix) # For nonparametric tests and post hoc procedures  
library(dplyr)   # For data manipulation (data wrangling)  
library(tidyr)   # For reshaping data (wide <-> long)  
  
#####  
# PART 1: Two Paired Samples  
# Nonparametric Test: Wilcoxon Signed-Rank Test  
#####  
  
# -----  
# Read paired-sample dataset (wide format)  
# Each row represents one subject  
# Time2 and Time3 are two paired measurements  
# -----  
data_two <- read.csv(  
  "/Users/apple/Documents/教师/20-教学笔记/多元统计/假设检验/非参数检验-两个配对样本-案例数据集.csv"  
)  
  
# Select the two paired variables  
data_two <- data_two %>%  
  select(Time2, Time3)  
  
# -----
```

```

# Wilcoxon Signed-Rank Test

# Purpose:

#   Test whether the median of paired differences equals zero
# -----

w_test <- wilcox.test(
  data_two$Time2,
  data_two$Time3,
  paired = TRUE,          # Paired samples
  exact = FALSE,          # Normal approximation (large sample)
  conf.int = TRUE,        # Confidence interval for location shift
  conf.level = 0.95
)

print(w_test)

# -----

# Effect size for paired Wilcoxon test
# Effect size measure: Rank-biserial correlation (paired version)
# -----

effect_two <- rank_biserial(
  data_two$Time2,
  data_two$Time3,
  paired = TRUE,          # Paired-sample version
  ci = 0.95,              # 95% confidence interval
  verbose = FALSE
)

print(effect_two)

# Effect size name: Rank-biserial correlation (r_rb, paired)

```

```

# Interpretation:

#    $r = \pm 1 \rightarrow$  all paired differences in the same direction

#  $r = 0 \rightarrow$  symmetric differences

# Advantage:

#   Nonparametric

# Distribution-free

# Directly interpretable as dominance probability

#####

# PART 2: Multiple Paired Samples

# Nonparametric Test: Friedman Test

#####

# -----

# Read wide-format repeated-measures dataset

# Each row = one subject

# Each column (Time1, Time2, Time3) = one condition

# -----

data_wide <- read.csv(

  "/Users/apple/Documents/教师/20-教学笔记/多元统计/假设检验/非参数检验-多个配对样本

-iris.csv"

)

# Select repeated-measures variables

data_wide <- data_wide %>%

  select(Time1, Time2, Time3)

# -----

```

```

# Convert wide format to long format (required for Friedman test)

# ID = subject identifier

# -----

data_long <- data_wide %>%
  mutate(ID = row_number()) %>%      # Subject ID
  pivot_longer(
    cols = starts_with("Time"),
    names_to = "Time",                # Within-subject factor
    values_to = "Value"              # Measurement
  )

# -----

# Friedman Test

# Purpose:

#   Test whether distributions differ across  $\geq 3$  paired conditions

# -----

cat("=== Friedman Test Results ===\n")

friedman_test <- friedman.test(
  Value ~ Time | ID,
  data = data_long
)

print(friedman_test)

# -----

# Overall effect size for Friedman test

# Effect size measure: Kendall's W

# -----

cat("\n=== Overall Effect Size: Kendall's W ===\n")

```

```

friedman_es <- kendalls_w(
  Value ~ Time | ID,
  data = data_long,
  ci = 0.95
)

print(friedman_es)

# -----
# Post hoc pairwise comparisons
# Test: Wilcoxon signed-rank test (paired)
# P-value adjustment: Holm correction
# -----

cat("\n=== Post hoc Pairwise Wilcoxon Tests (Holm-adjusted) ===\n")

pairwise_test <- data_long %>%
  pairwise_wilcox_test(
    Value ~ Time,
    paired = TRUE,
    p.adjust.method = "holm"    # Holm or Bonferroni
  )

print(pairwise_test)

# -----
# Effect sizes for pairwise comparisons
# Effect size measure: Rank-biserial correlation (paired)
# -----

```

```

# Generate all pairwise combinations of conditions

time_levels <- unique(data_long$Time)

time_pairs <- combn(time_levels, 2, simplify = FALSE)

pairwise_effectsize <- lapply(time_pairs, function(tp) {

  # Extract paired observations for each condition
  x <- data_long %>% filter(Time == tp[1]) %>% pull(Value)
  y <- data_long %>% filter(Time == tp[2]) %>% pull(Value)

  # Compute paired rank-biserial correlation
  es <- rank_biserial(
    x, y,
    paired = TRUE,
    ci = 0.95,
    verbose = FALSE
  )

  es_df <- as.data.frame(es)

  # Assemble results
  data.frame(
    time1 = tp[1],
    time2 = tp[2],
    rank_biserial = es_df[[1]],
    CI_low = es_df$CI_low,
    CI_high = es_df$CI_high
  )
}) %>%

bind_rows()

```

pairwise_effectsize

Effect size name: Rank-biserial correlation (paired)

Interpretation:

$|r| \approx 0.10 \rightarrow$ small

$|r| \approx 0.30 \rightarrow$ moderate

$|r| \geq 0.50 \rightarrow$ large

CI meaning:

Reflects uncertainty of dominance probability

CI = $[-1, -1]$ or $[1, 1]$ indicates perfect separation