



stats::free1way(): Semiparametrically Efficient Population and Permutation Inference in Distribution-free Stratified K-sample Oneway Layouts

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Neck Pain Reduction by Laser Therapy



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The effect of 300 mW, 830 nm laser on chronic neck pain: A double-blind, randomized, placebo-controlled study

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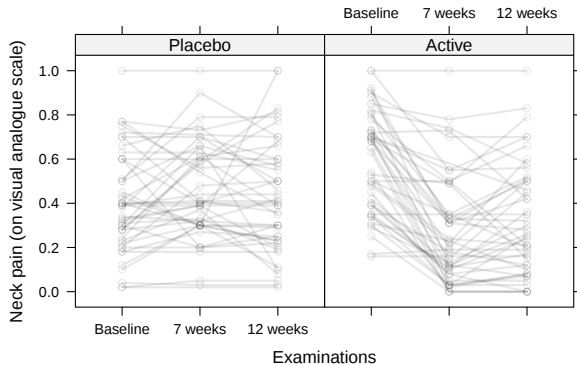
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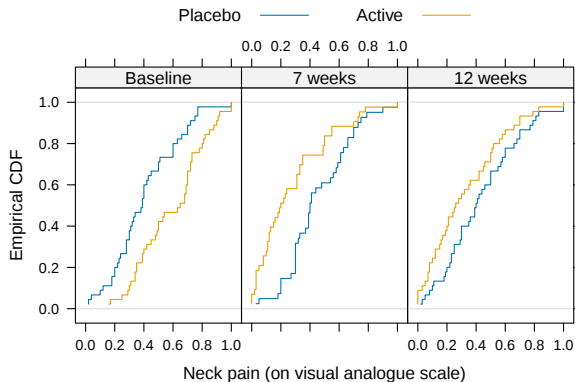
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Contains a discussion about how to analyse VAS and motivated research accumulating in **ordinalCont**

Neck Pain Reduction by Laser Therapy: Joint



Neck Pain Reduction by Laser Therapy: Marginal



stats::wilcox.test()

```
> wilcox.test(vas ~ laser, data = pain,  
+             subset = time == "12 weeks",  
+             exact = TRUE, conf.int = TRUE)
```

Wilcoxon rank sum exact test

```
data:  vas by laser  
W = 1260, p-value = 0.046  
alternative hypothesis: true location shift is not equal to 0  
96 percent confidence interval:  
 0.00 0.22  
sample estimates:  
difference in location  
      0.11
```

Wilcoxon 1945

INDIVIDUAL COMPARISONS BY RANKING METHODS

FRANK WILCOXON
American Cyanamid Co.

For 12-week ranked VAS measurements $R_i = \text{rank}(Y_i)$ with $i = 1, \dots, n$ for placebo and $i = n + 1, \dots, N$ for active laser, the rank sum statistic is

$$\sum_{i=1}^n R_i \propto \underbrace{\frac{1}{N-n} \sum_{i=n+1}^N R_i - \frac{1}{n} \sum_{i=1}^n R_i}_{t\text{-test on ranks}}$$

testing the CDF hypothesis

$$H_0 : F(y \mid \text{placebo}) = F(y \mid \text{active}) \quad \forall y \in (0, 1)$$

Constructing Confidence Sets Using Rank Statistics

DAVID F. BAUER*

$$F(y \mid \text{active}) = F(y - \beta \mid \text{placebo})$$

Hodges-Lehmann estimate and confidence interval for location-shift

Problems

- location shift inappropriate for bounded or discrete distributions
- test not necessarily optimal against such alternatives
- model diagnostics?
- `power.wilcox.test()`?
- stratification?

THE POWER OF RANK TESTS¹

BY E. L. LEHMANN

Stanford University and University of California, Berkeley

$$F(y \mid \text{active}) = m(F(y \mid \text{placebo}), \delta)$$

maybe $m(p, \delta) = p^\delta$ (“Lehmann alternative”)?

STATISTICS IN MEDICINE, VOL. 12, 2257–2271 (1993)

SAMPLE SIZE CALCULATIONS FOR ORDERED CATEGORICAL DATA

JOHN WHITEHEAD

Department of Applied Statistics, University of Reading, Earley Gate, P.O. Box 238, Reading RG6 2AL, U.K.

$$m(p, \delta) = \text{expit}(\text{logit}(p) - \delta)$$

aka “proportional odds (PO) model” with δ a log-odds ratio

NB: `wilcox.test()` is the nonparametric (or empirical) score test in a PO model for binary, ordered, continuous outcomes.

Distribution-free Models in Oneway Layouts

For

- strata (S , for $b = 1, \dots, B$),
- treatment groups (G , for $k = 2, \dots, K$), and
- outcome values y based on
- effect parameters $\delta_2, \dots, \delta_K$:

$$\underbrace{F(y \mid G = k, S = b)}_{\text{Treatment } k \text{ CDF in } b} = m \left(\underbrace{F(y \mid G = 1, S = b)}_{\text{Placebo CDF in } b, \text{unspecified}}, \underbrace{\delta_k}_{k \text{ vs. Placebo}} \right)$$

`stats::free1way()`, termed after SAS' `NPAR1WAY`, since R 4.6.0 for semiparametrically efficient population and permutation inference in distribution-free stratified K -sample oneway layouts.

Implementation in stats

Enhances: `wilcox.test()`, `kruskal.test()`, `friedman.test()`,
`mantelhaen.test()`, `prop.test()`, `mcnemar.test()`,
MASS::polr(), tram::Polr, and (as a role model) rms::orm(), ...

Fast joint optimisation of empirical likelihood wrt effect and nuisance parameters by Newton-Raphson with sparse analytic Hessians, courtesy by Frank Harrell.

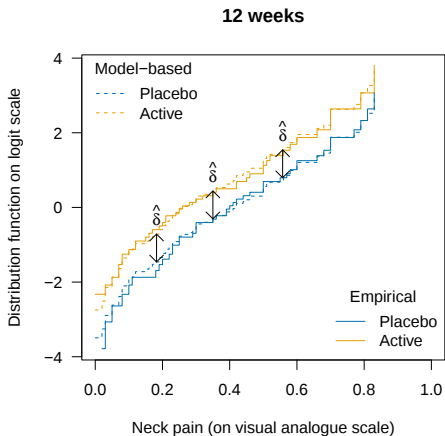
Applicable to binary, ordinal, discrete, continuous, and potentially (independently) right-censored observations.

Special Models

- $m_L(p, \delta) = p^{\exp(-\delta)}$ (Lehmann alternatives, loglog link),
- $m_{PH}(p, \delta) = 1 - (1 - p)^{\exp(-\delta)}$ (proportional hazards, cloglog link)
- $m_{PO}(p, \delta) = \text{expit}(\text{logit}(p) - \delta)$ (proportional odds, logit link),
and
- $m_{Cd}(p, \delta) = \Phi(\Phi^{-1}(p) - \delta)$ (generalised Cohen's d , probit link)

Against these alternatives, tests implemented in `free1way()` are locally most powerful.

Pain Reduction: Model Diagnostics



Pain Reduction: Parameter Estimation

```
> f <- free1way(vas ~ laser, data = pain,  
+             subset = time == "12 weeks",  
+             exact = TRUE)  
> ### ### empirical log-likelihood  
> logLik(f)  
[1] -338.214  
  
> ### estimated log-odds ratio + SE  
> c(coef(f), sqrt(vcov(f)))  
laserActive  
-0.746923    0.372346
```

Pain Reduction: Permutation Score Inference

```
> summary(f, test = "Permutation")
```

2-sample Wilcoxon test against proportional odds
alternatives

```
data:  vas by laser (Placebo, Active)
```

```
W = -2.015, p-value = 0.046
```

```
alternative hypothesis: true log-odds ratio is not equal to 0
```

```
> confint(f, test = "Permutation")
```

```
2.5 %      97.5 %
```

```
laserActive -1.48199 -0.0119882
```

Pain Reduction: Empirical Likelihood Inference

```
> summary(f, test = "Rao")
```

2-sample Wilcoxon test against proportional odds
alternatives

data: vas by laser (Placebo, Active)

Rao chi-squared = 4.062, df = 1, p-value = 0.0439

alternative hypothesis: true log-odds ratio is not equal to 0

```
> confint(f, test = "Rao")
```

2.5 % 97.5 %

laserActive -1.47361 -0.0203546

test = "LRT" and test = "Wald" also available.

Pain Reduction: Power Assessment

```
> power.free1way.test(delta = coef(f),  
+                      power = .8, sig.level = .05)
```

2-sample Wilcoxon test against proportional odds
alternatives

```
      n = 88  
Total sample size = 88 (Control) + 88 (laserActive) = 176  
      power = 0.801078  
      sig.level = 0.05  
Standard error = 0.266241  
log-odds ratio = -0.746923
```

NOTE: 'n' is sample size in control group

Pain Reduction: Repeated Measurements

patient = block, aka `friedman.test()` in each group:

```
> pain$tc <- with(pain, interaction(laser, time))
> levels(pain$tc)[1:2] <- "Baseline"
> coef(m <- free1way(vas ~ tc | id, data = pain))
```

tcPlacebo.7 weeks	tcActive.7 weeks	tcPlacebo.12 weeks
1.144270	-5.623352	0.407917
tcActive.12 weeks		
-4.352816		

Pain Reduction: Repeated Measurements

```
> library("multcomp")
> K <- matrix(c(-1, 1, 0, 0,
+              0, 0, -1, 1), byrow = TRUE, nrow = 2)
> rownames(K) <- c("7 weeks: Active - Placebo",
+                  "12 weeks: Active - Placebo")
> confint(glht(m, linfct = K))$confint
```

	Estimate	lwr	upr
7 weeks: Active - Placebo	-6.76762	-8.72808	-4.80716
12 weeks: Active - Placebo	-4.76073	-6.57525	-2.94622

```
attr("conf.level")
[1] 0.95
attr("calpha")
[1] 2.18344
```

Resources

Visit

- `stats::free1way`
- `stats::rfree1way`
- `stats::power.free1way.test`
- `stats::ppplot`

and add-on package **free1way.docreg** at CRAN near you:

- all the models
- likelihood and inference procedures
- optimisation algorithm
- literate pure R implementation of ~ 2500 lines
- more examples
- regression tests