



# **CAMIS**

Comparing

**A**nalysis

Method

Implementations in

**S**oftware









Data Visualisation & Open Source Technology

# Disclaimer

This presentation reflects the view of the PHUSE CAMIS Working Group and does not necessarily represent the position of PAREXEL or GSK

# **Why Document Comparisons**



#### Lyn

- Has been in industry for 20+ years
- Prefers SAS



#### Christina

- Newer to industry
- Learned object oriented programming in school and prefers R







# **Why Document Comparisons**

#### Example 87.4 Hodges-Lehmann Estimation

(View the complete code for this example.)

rristina



This example uses the SAS data set *Persy, page sta.* A line that make the difference in medians (a common misconception) but rather the median of the difference between a sample from x and a sample from y.

If exact p-values are available, an exact confidence interval is obtained by the algorithm described in Bauer (1972), and the Hodges-Lehmann estimator is employed. Otherwise, the returned confidence interval and poi estimate are based on normal approximations. These are continuity-corrected for the interval but *not* the esti (as the correction depends on the alternative).

With small samples it may not be possible to achieve very high confidence interval coverages. If this happens warning will be given and an interval with lower coverage will be substituted.

then w (and w if applicable) are valid the function new always returns also in the conf. int. - TRITE of







#### How to Avoid Finding Differences at Critical Times

- Make sure to have <u>clear requirements</u>
- Check CAMIS

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 Consider defining tolerances rather than looking for exact match between prod and QC





#### What does CAMIS Cover?

Methods		R	SAS	Python	Comparison
Summary Statistics	Rounding	<u>R</u>	SAS	<u>Python</u>	R vs SAS
	Summary statistics	<u>R</u>	SAS	<u>Python</u>	R vs SAS
	Skewness/Kurtosis	<u>R</u>	SAS	<u>Python</u>	R vs SAS
General Linear Models	One Sample t-test	<u>R</u>	SAS	<u>Python</u>	R vs SAS
	Paired t-test	<u>R</u>	SAS	<u>Python</u>	R vs SAS
	Two Sample t-test	<u>R</u>	SAS	<u>Python</u>	R vs SAS
	ANOVA		SAS	<u>Python</u>	R vs SAS
	ANCOVA		SAS	Python	R vs SAS
	MANOVA		SAS	<u>Python</u>	R vs SAS
	Linear Regression	<u>R</u>	R SAS Python		R vs SAS
Generalized Linear Models	Logistic Regression	<u>R</u>	SAS	Python	R vs SAS
	Poisson/Negative Binomial Regression	<u>R</u>	SAS		R vs SAS
	Categorical Repeated Measures				
	Categorical Multiple Imputation				
Non-parametric Analysis	Wilcoxon signed rank	<u>R</u>	<u>SAS/</u> StatXac <u>t</u>		R vs SAS
	Mann-Whitney U/Wilcoxon rank sum	<u>R</u>	SAS		
	Kolmogorov-Smirnov test				
	Kruskall-Wallis test	<u>R</u>	SAS	<u>Python</u>	R vs SAS
	Friedman test	<u>R</u>	SAS		R vs SAS
	Jonckheere test	<u>R</u>	SAS		R vs SAS
	Hodges-Lehman Estimator	<u>R</u>	SAS		
Categorical Data Analysis	Binomial test	<u>R</u>	SAS	Python	
	McNemar's test	<u>R</u>	SAS		R vs SAS
	Marginal Homogeneity Tests	<u>R</u>			
	Chi-Square Association/Fishers exact	<u>R</u>	SAS	Python	R vs SAS
	Cochran Mantel Haenszel	<u>R</u>	SAS		R vs SAS
	Confidence Intervals for	R	SAS		R vs SAS

Repeated Measures	Linear Mixed Model (MMRM)	R SAS	<u> </u>	R vs SAS
	Linear Mixed Model (degrees of freedom)			
	Generalized Linear Mixed Model (MMRM)			
	Bayesian MMRM			
Multiple Imputation -	мсмс			
Continuous Data MAR	Linear regression	R SAS	<u>S</u>	
	Predictive Mean Matching	<u>R</u>		
Multiple Imputation - Continuous Data MNAR	Delta Adjustment/Tipping Point			
	Reference-Based Imputation/Joint Modelling	R SAS	<u> </u>	R vs SAS
Correlation	Pearson's/ Spearman's/ Kendall's Rank	R SAS	<u>Python</u>	R vs SAS
Survival Models	Kaplan-Meier Log-rank test and Cox-PH	R SAS	<u> </u>	R vs SAS
	Cause Specific Hazards	R SAS	3	R vs SAS
	Accelerated Failure Time	<u>R</u>		
	Weighted Log-rank test	<u>R</u>		
	Cumulative Incidence Functions	R SAS	<u> </u>	R vs SAS
	Tobit regression	R SAS	<u>S</u>	R vs SAS
	Restricted Mean Survival Time (RMST)	SAS	<u> </u>	
Sample size/ Power	Intro to Sample Size			<u>Summary</u>
calculations	Superiority Single timepoint	R SAS	<u> </u>	
	Equivalence Single timepoint	R SAS	<u> </u>	
	Non-Inferiority Single timepoint	R SAS	3	
	Average BioEquivalence	<u>R</u>		
	Cochran-Armitage Test For Trend	R StatX	ac	
	Group sequential designs	R Eas	<u>t</u>	East vs R
Causal inference/ Machine	Intro to Machine Learning			Summary

# **Differing Defaults**

#### Kaplan Meier

Analysis	Supported in R	Supported in SAS	Results Match	Notes
Kaplan Meier with confidence intervals using log- log method	Yes (using the option conf.type = "log-log")	Yes (Default)	Mostly	Survival estimates can disagree when last event is censored and survival estimate does not cross the percentile being estimated.
				2) Survival estimates at time X can disagree when the time X is after the last observed censored time
Kaplan Meier with confidence intervals using log method	Yes (Default)	Yes (using the option conftype=log)	Mostly	As above.

#### Cox – PH regression

Analysis	Supported in R	Supported in SAS	Results Match
Cox Proportional Hazards Model using breslow method for ties	Yes (using the option ties="breslow")	Yes (Default)	Yes
Cox Proportional Hazards Model using efron method for ties	Yes (Default)	Yes (using the option ties=efron)	Yes







# **Identifying Common Coding Mistakes**

Logistic Regression

#### **Summary of Common Mistakes in SAS**

- 1. Handling of missing data. Check SAS output that the number of missing values is as you expect. Make sure you have changed any NA results in the raw data to be missing, since SAS would consider NA as a valid category (a non-missing character result).
- 2. Make sure you consider continuous or categorical variables as you intended. Just because a variable is character or numeric in the dataset, doesn't mean SAS will treat it that way in the model. You have to use Class row to tell SAS which variables should be treated as character factors. You also have to use ref=' to tell SAS which is the reference category, otherwise SAS by default which use the last value of the variable alphabetically (e...g a categorical variable with 1, 2, 3 would default to 3 as the reference).
- 3. Be careful you are modelling the correct event (response vs non-response, or weight\_gain vs weight\_loss for example)
- 4. Be careful when interpreting any odds ratios that you have the factor of interest the correct way around (0 vs 1, or 1 vs 0)
- 5. If using proc logistic, be careful of how SAS creates its parameters used in the model as this determines how you can use the parameter estimates! It is often easiest to use param=glm so that the exp(maximum liklihood parameter estimate) = odds ratio. Check the class level information (Design variables) is as you would expect. See below for more detail on other parameterization methods.





# Identifying Packages Not Being Maintained

- McNemar's Chi Squared
- {epibasix} package
  - Categorised as 'High Risk' by {riskmetrics} package
  - Unknown by the author what equation they used to calculated the CIs

Analysis	Supported in R	Supported in SAS	Results Match	Notes
McNemar's Chi- Squared test	<u>Yes</u>	Yes	<b>V</b>	By default SAS doesn't include the continuity correction. In R use {stats} or {coin}
Cohen's Kappa CI	Yes	Yes	<b>V</b>	In R use {vcd}





# Finding Bugs and Getting them Corrected

- {RBesT} produces Clopper Pearson Cls
- Matched {cardx} and {Hmisc} except 2 cases
- 1. x = 0 (0% responders), in which case the lower limit does not match.
- 2. x = n (100% responders), in which case the upper limit does not match.

- Authors informed
- Package updated & re-released on CRAN as v1.8-0

CAMIS Improves reliability & confidence in R packages







# **Key Impacts**

- Documentation of Differences due to Methodology use
  - Continuity corrections
  - Different Defaults
  - Methods / Options not available in both software
- Common implementation mistakes and any unclear software documentation explained
- Bugs identified (and R packages updated & fixed!)
- CAMIS Is a Trusted Resource for Industry Guidance







• For a study comparing treatments A and B in patients. Apply the Wilcoxon rank-sum test to analyse the change from baseline FEV1 post-mannitol challenge 2. Report the Hodges-Lehmann estimate of the median difference between treatments (B vs A), corresponding 95% CI and the associated p-value

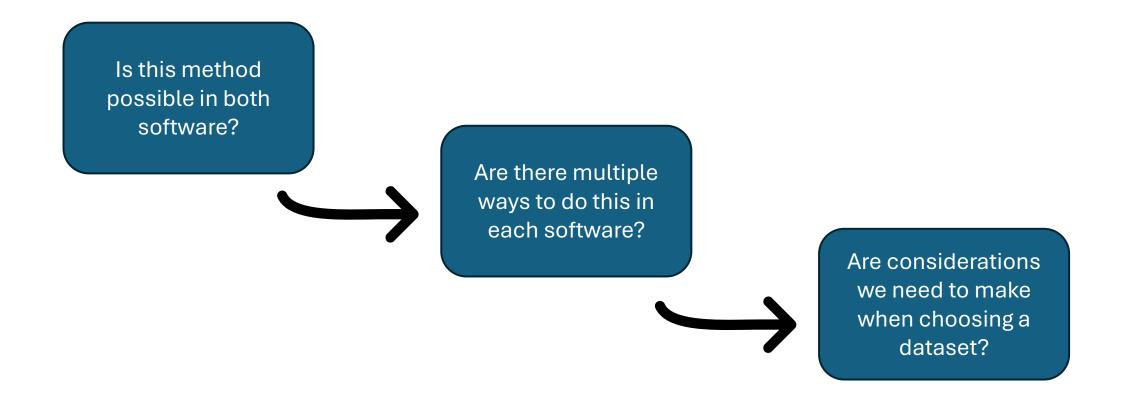




# **How to Compare**

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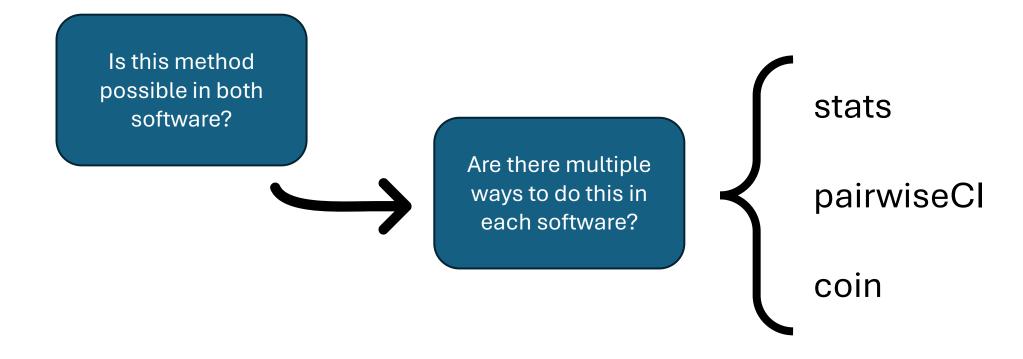
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# **Comparing Packages**

- Is this package being actively maintained?
- Is the author known in this field?
- Does one of the packages have better documentation?
- Look at community adoption?

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Can the packages handle different edge cases?





#### pairwiseCI: Confidence Intervals for Two Sample Comparisons

Calculation of the parametric, nonparametric confidence intervals for the difference or ratio of location parameters, nonparametric confidence interval for the Behrens-Fisher problem and for the difference, ratio and odds-ratio of binomial proportions for comparison of independent samples. Common wrapper functions to split data sets and apply confidence intervals or tests to these subsets. A by-statement allows calculation of CI separately for the levels of further factors. CI are not adjusted for multiplicity.

Version: 0.1-27

Depends:  $\underline{MCPAN}$ ,  $\underline{coin}$  ( $\geq 1.3-0$ )

Imports: graphics, stats, boot, MASS, mcprofile

Published: 2019-03-11

DOI: <u>10.32614/CRAN.package.pairwiseCI</u>

Author: Frank Schaarschmidt [aut, cre], Daniel Gerhard [aut]

Maintainer: Frank Schaarschmidt <schaarschmidt at biostat.uni-hannover.de>

License: GPL-2

NeedsCompilation: no

CRAN checks: <u>pairwiseCI results</u>





coin: Conditional Inference Procedures in a Permutation Test Framework

Conditional inference procedures for the general independence problem including two-sample, K-sample (non-parametric ANOVA), correlation, censored, ordered and multivariate problems described in <doi:10.18637/jss.v028.i08>.

Version: 1.4-3

Depends:  $R (\geq 3.6.0)$ , survival

Imports: methods, parallel, stats, stats4, utils, <u>libcoin</u> ( $\geq 1.0-9$ ), <u>matrixStats</u> ( $\geq 0.54.0$ ), <u>modeltools</u> ( $\geq$ 

0.2-9), mvtnorm ( $\geq 1.0-5$ ), multcomp

LinkingTo:  $\underline{\text{libcoin}} \ (\geq 1.0-9)$ 

Suggests:  $\underline{\text{xtable}}, \underline{\text{e1071}}, \underline{\text{vcd}}, \underline{\text{TH.data}} (\geq 1.0-7)$ 

Published: 2023-09-27

DOI: <u>10.32614/CRAN.package.coin</u>

Author: Torsten Hothorn [ [aut, cre], Henric Winell [ [aut], Kurt Hornik [ [aut], Mark A.

van de Wiel [aut], Achim Zeileis [aut]

Maintainer: Torsten Hothorn <a href="Torsten">Torsten Hothorn at R-project.org</a>

License: GPL-2

URL: <a href="http://coin.r-forge.r-project.org">http://coin.r-forge.r-project.org</a>

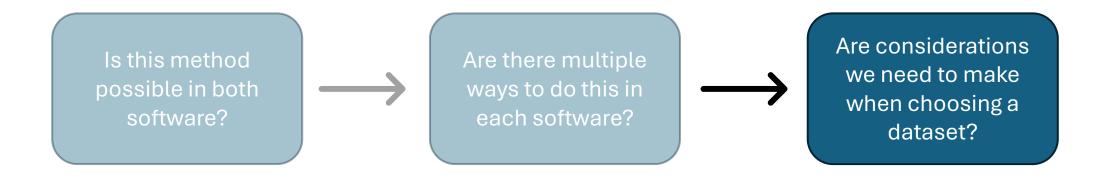




↓					
•	package	version	pkg_ref	pkg_score	
1	coin	1.4-3	<environment: 0x17addce58=""></environment:>	0.9106095	
2	pairwiseCl	0.1-27	<environment: 0x17b4b5240=""></environment:>	0.9758364	







 Because we have known issues with ties, we want to make sure to have a dataset that includes them

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	R {coin}	SAS
P-value (asymptotic)	0.009392	0.0094
CI (exact)	-0.76 – -0.1	-0.76000.1000
P-values (exact)	0.008181	0.0082
CI (asymptotic)	NA	-0.77 – -0.0900





• For a study comparing treatments A and B in patients. Apply the Wilcoxon rank-sum test to analyse the change from baseline FEV1 post-mannitol challenge 2. Report the Hodges-Lehmann estimate of the median difference between treatments (B vs A), corresponding 95% exact CI and the associated exact p-value (without continuity correction)





# **CAMIS Objectives**

- Understand and Document analysis result discrepancies across software
- Demonstrate the methodology through examples
- > Document in open GitHub repository
- For Grow the repository, increasing quality and quantity of information



Key Considerations When Understanding Differences in Statistical Methodology Implementations Across Programming Languages – An Introduction to the CAMIS Project

Min-Hua Jen, Brian Varney, Kyle Lee, Benjamin Arancibia, Mia Qi, Lyn Taylor, Christina Fillmore, Joseph Rickert, Mike Stackhouse, Michael Rimler









# Questions?

CAMIS website: <a href="https://psiaims.github.io/CAMIS/">https://psiaims.github.io/CAMIS/</a>

GitHub Repo: <a href="https://github.com/PSIAIMS/CAMIS/">https://github.com/PSIAIMS/CAMIS/</a>

Open issues: https://github.com/PSIAIMS/CAMIS/issues



#### The Global Healthcare Data Science Community

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