




# Latentcor: An R Package for Latent Correlation Estimation

Mingze Huang<sup>1, 2</sup>, Irina Gaynanova<sup>2</sup>, and Christian L. Muller<sup>1</sup>

<sup>1</sup> Department of Statistics, Texas A& M University <sup>2</sup> Department of Economics, Texas A& M University <sup>3</sup> Department of Statistics, University of Munich

DOI:

Software

- [Review](#) 
- [Repository](#) 
- [Archive](#) 

Submitted:

Published:

License

Authors of papers retain copyright and release the work under a Creative Commons Attribution 4.0 International License ([CC-BY](#)).

## Summary

The R package *latentcor* provides estimation for latent correlation with mixed data types (continuous, binary, truncated and ternary). Comparing to *MixedCCA*, which estimates latent correlation for canonical correlation analysis, our new package provides a standalone version for latent correlation estimation. Also we add new functionality for latent correlation between ternary/continuous, ternary/binary, ternary/truncated and ternary/ternary cases.

Compare to *MixedCCA*, memory footprint.

## Statement of need

Currently there is no standalone package dealing with latent correlation for mixed data type like we did in *latentcor*. The R package *stats* (Team & others, 2013) have some functionality to calculate different type of correlations (Pearson, Kendall and Spearman). The R package *pcaPP* (Croux, Filzmoser, & Fritz, 2013) provides a fast calculation for Kendall's  $\tau$ . The R package *MixedCCA* (Yoon, Carroll, & Gaynanova, 2020) have functionality for latent correlation estimation as an intermediate step for canonical correlation analysis on mixed data. No package deal with latent correlation across mixed data type.

## Usage

	Type	continuous
continuous		Liu, Lafferty, & Wasserman (2009)
binary		Fan, Liu, Ning, & Zou (2017)
truncated		Yoon, Carroll, & Gaynanova (2020)
ternary		Quan, Booth, & Wells (2018)

*Definition 1* Fan et al. (2017) considered the problem of estimating  $\Sigma$  for the latent Gaussian copula model based on Kendall's  $\tau$ . Given the observed data  $(X_{1j}, X_{1k}), \dots, (X_{nj}, X_{nk})$  for variables  $X_j$  and  $X_k$ , Kendall's  $\tau$  is defined as

$$\hat{\tau}_{jk} = \frac{2}{n(n-1)} \sum_{1 \leq i < i' \leq n} \text{sign}(X_{ij} - X_{i'j}) \text{sign}(X_{ik} - X_{i'k})$$

*Theorem 1* Let  $W_1 \in \mathcal{R}^{\infty}$ ,  $W_2 \in \mathcal{R}^{\epsilon}$ ,  $W_3 \in \mathcal{R}^{\beta}$ ,  $W_4 \in \mathcal{R}^{\Delta}$  be such that  $W = (W_1, W_2, W_3, W_4) \sim NPN(0, \Sigma, f)$  with  $p = p_1 + p_2 + p_3 + p_4$ . Let  $X = (X_1, X_2, X_3, X_4) \in \mathcal{R}^{\vee}$  satisfy  $X_j = W_j$  for  $j = 1, \dots, p_1$ ,  $X_j = I(W_j > c_j)$  for  $j = p_1 + 1, \dots, p_1 + p_2$ ,