# ANOVAforMRMCLimitsofAgreement

December 12, 2020

result.validMRMCVarEstimate.nReader

Result of simulation study to validate and characterize the MRMC limits of agreement estimates

#### **Description**

Result of simulation study to validate and characterize the MRMC limits of agreement estimates

# Usage

result.validMRMCVarEstimate.nReader

result.validMRMCVarEstimate.nCase

 $result.valid {\tt MRMCVarEstimate.parameter}$ 

#### Format

A dataframe with the following columns:

nReader Number of Readers

nCase Number of Cases

alpha\_R Reader related parameter

sigma\_C Case related parameter

WR\_var\_MCmean Monte Carlo mean of variance estimation for the WRBM difference score

WR\_var\_MCvar Monte Carlo variance of variance estimation for the WRBM difference score

True\_WR\_var Theoretical value for WRBM variance derived from the simulation model

WR\_var\_relative\_bias relative bias for WRBM variance

WR CV coefficient of variation for WRBM variance

BR\_var\_MCmean Monte Carlo mean of variance estimation for the BRBM difference score

BR\_var\_MCvar Monte Carlo variance of variance estimation for the BRBM difference score

True\_BR\_var Theoretical value for BRBM variance derived from the simulation model

BR\_var\_relative\_bias relative bias for BRBM variance

BR\_CV coefficient of variation for BRBM variance

An object of class data. frame with 11 rows and 14 columns.

An object of class data. frame with 25 rows and 14 columns.

#### **Details**

There are three parameter sets for this simulation:

**result.validMRMCVarEstimate.nReader** Different number of readers  $J=3,4,...,10,K=100,\alpha_R=10,\sigma_C^2=1$ 

**result.validMRMCVarEstimate.nCase** Different number of cases  $K=50,60,...,150, J=5, \alpha_R=10, \sigma_C^2=1$ 

**result.validMRMCVarEstimate.parameter** Different reader and case variabilities  $(\alpha_R, \sigma_C^2) \in \{3, 4, 6, 11, 21\} \times \{0.1, 0.2, 0.4, \frac{2}{3}, 1\}, J = 5, K = 100$ 

In all the parameter sets,  $\beta_R = \beta_{\tau R} = 1$ ,  $\alpha_R = \alpha_{\tau R}$ ,  $\sigma_C^2 = \sigma_{\tau C}^2$ . For each parameter setting, we simulate 1000 trials.

result.validSimulation.alphaR

Result of simulation study to verify the simulation is consistent with the derived theoretical values

#### **Description**

Result of simulation study to verify the simulation is consistent with the derived theoretical values

# Usage

```
result.validSimulation.alphaR result.validSimulation.sigmaC
```

# Format

A dataframe with the following column:

sigma\_C Case related parameter

alpha\_R Reader related parameter

WR\_mean Monte Carlo mean of the WRBM difference score

WR\_var Monte Carlo variance of the WRBM difference score

BR\_mean Monte Carlo mean of the BRBM difference score

BR\_var Monte Carlo variance of the BRBM difference score

True\_WR\_var Theoretical value for WRBM variance derived from the simulation model

True\_BR\_var Theoretical value for BRBM variance derived from the simulation model

WR\_var\_relative\_bias relative bias for WRBM variance

**BR\_var\_relative\_bias** relative bias for BRBM variance

An object of class data. frame with 20 rows and 10 columns.

#### **Details**

There are two parameter sets for this simulation:

**result.validSimulation.alphaR** In this study, we fixed the case-related parameter but change the reader related fixed the case related parameters  $\sigma_C^2 = \sigma_{\tau C}^2 = 1$  and allowed the reader related parameter  $\alpha_R (= \alpha_{\tau R})$  to range from 2 to 20.

**result.validSimulation.sigmaC** In this study, we fixed the reader related parameters  $\alpha_R = \alpha_{\tau R} = 10$  and allowed the case related parameter  $\sigma_C^2 (= \sigma_{\tau C}^2)$  to range from 0.1 to 2, incrementing by 0.1.

In both parameter sets,  $\beta_R = \beta_{\tau R} = 1$ , so that the reader variability is only affected by  $\alpha_R$ . We simulated 100,000 trials with each trial having 4 measurements from 2 readers for a single case under 2 modalities. The Monte Carlo estimates of the variances are the sample variances of the 100,000 independent WRBM and BRBM differences.

validateMRMCVarEstimation

Validate and characterize the MRMC limits of agreement estimates

# **Description**

This function simulates MRMC study with nReader readers and nCase cases. From the output of laWRBM anova and laBRBM anova, we exact the variance estimations for WRBM and BRBM differences. It outputs the Monte Carlo mean of the variance estimation across all the simulation trials. The relative bias of the Monte Carlo mean comparing to the theoretical value of the variance and the coefficient of variation is also calculated.

### Usage

```
validateMRMCVarEstimation(
    nR.list,
    nC.list,
    alpha_R.list,
    sigma_C.list,
    nTrials = 1000
)
```

# **Arguments**

```
nR.list A r-dimension array for the number of readers in each MRMC study nC.list A c-dimension array for the number of cases in each MRMC study alpha_R.list A a-dimension array for the reader related parameter sigma_C.list A b-dimension array for the case related parameter nTrials Number of MRMC simulations. Default is 1000
```

#### **Details**

Let  $\hat{V}_{BR}^{12}$  denote the estimation of the variance for the BRBM difference for each simulated MRMC study, then the relative bias is defined as relativeBias  $(\hat{V}_{BR}^{12}) = (\sum \hat{V}_{BR}^{12}/nTrials - V_{BR}^{12})/V_{BR}^{12}$ , and the coefficient of variantion is defined as CV  $(\hat{V}_{BR}^{12}) = sd(\hat{V}_{BR}^{12})/V_{BR}^{12}$ , where sd() denote sample standard deviation.

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#### Value

A dataframe with r \* c \* a \* b rows. Each column is as following:

nReader Number of Readers

nCase Number of Cases

alpha\_R Reader related parameter

sigma\_C Case related parameter

WR\_var\_MCmean Monte Carlo mean of variance estimation for the WRBM difference score

WR\_var\_MCvar Monte Carlo variance of variance estimation for the WRBM difference score

True\_WR\_var Theoretical value for WRBM variance derived from the simulation model

WR\_var\_relative\_bias relative bias for WRBM variance

WR CV coefficient of variation for WRBM variance

BR\_var\_MCmean Monte Carlo mean of variance estimation for the BRBM difference score

BR\_var\_MCvar Monte Carlo variance of variance estimation for the BRBM difference score

True\_BR\_var Theoretical value for BRBM variance derived from the simulation model

BR\_var\_relative\_bias relative bias for BRBM variance

BR\_CV coefficient of variation for BRBM variance

#### **Examples**

```
library(iMRMC)
nR.list <- c(5)
nC.list <- c(100)
sigma_C.list <- c(1)
alpha_R.list <- c(10)
#result <- validateMRMCVarEstimation(nR.list, nC.list, alpha_R.list, sigma_C.list)</pre>
```

validateSimulation

Verify the simulation is consistent with the derived theoretical values

# **Description**

This function simulates independent WRBM and BRBM difference scores and outputs the Monte Carlo mean and variance across the trials for each parameter setting. The relative bias of the Monte Carlo variance comparing to the theoretical value of the variance is also calculated.

#### Usage

```
validateSimulation(alpha_R.list, sigma_C.list, nTrials = 1e+05)
```

# Arguments

```
alpha_R.list A a-dimension array for the reader related parameter sigma_C.list A b-dimension array for the case related parameter nTrials Number of independent trials. Default is 100000
```

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#### **Details**

Let  $\tilde{V}_{BR}^{12}$  denote the Monte Carlo variance of the BRBM difference, then the relative bias is defined as relativeBias  $(\tilde{V}_{BR}^{12}) = (\tilde{V}_{BR}^{12} - V_{BR}^{12})/V_{BR}^{12}$ .

#### Value

A dataframe with a \* b rows. Each column is as following:

sigma\_C Case related parameter

alpha\_R Reader related parameter

WR\_mean Monte Carlo mean of the WRBM difference score

WR\_var Monte Carlo variance of the WRBM difference score

BR\_mean Monte Carlo mean of the BRBM difference score

BR\_var Monte Carlo variance of the BRBM difference score

True\_WR\_var Theoretical value for WRBM variance derived from the simulation model

True\_BR\_var Theoretical value for BRBM variance derived from the simulation model

WR\_var\_relative\_bias relative bias for WRBM variance

BR\_var\_relative\_bias relative bias for BRBM variance

# **Examples**

```
library(iMRMC)
sigma_C.list <- c(1)
alpha_R.list <- c(10)
#result <- validateSimulation(alpha_R.list, sigma_C.list)</pre>
```

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