## Simulate MRMC Data

Here we demonstrate how to simulate Multi-reader Multi-case (MRMC) data using functions in the iMRMC package.

sim.NormalIG.Hierarchical is a function that simulates MRMC agreement data (no binary truth state returned). The model for the sim.NormalIG.Hierarchical function is described here:

• S. Wen and B. D. Gallas, "Three-Way Mixed Effect ANOVA to Estimate MRMC Limits of Agreement," *Statistics in Biopharmaceutical Research*}\*, **14**, pp. 532–541, 2022, https://www.doi.org/10.1080/19466315.2022.2063169.

sim.gRoeMetz is a function that simulates MRMC ROC data. We note here that the model for sim.gRoeMetz does not yield readers with different average/expected levels of agreement with one another. This limitation was the reason for developing the model for the sim.NormalIG.Hierarchical simulation. The model for the sim.gRoeMetz function is described here:

• B. D. Gallas and S. L. Hillis, "Generalized Roe and Metz ROC model: analytic link between simulated decision scores and empirical AUC variances and covariances," *J Med Img*, **1**, no. 3, p. 031006, 2014, https://www.doi.org/10.1117/1.JMI.1.3.031006.

## 1. Using the new hierarchical model to simulate the MRMC agreement data

The following is how to use the sim.NormalIG.Hierachcial function to simulate MRMC agreement data (with no truth state). First, we initialize a list of simulation parameters using the sim.NormalIG.Hierarchical.config function. There are many simulation parameters. All have default values that can be changed in the function call. There are some parameters returned by the function that are equal for convenience and might need to be changed before simulating the data. Below, we demonstrate specifying non-default numbers of readers and cases.

Here are all the simulation parameters:

"alpha R.A"

"C scale"

## [11] "sigma C.A"

## [16] "beta R.B"

"sigma C.B"

"tauC scale"

"alpha R.B"

"tauRCE scale"

"beta R.A"

"RC scale"

Next, we set the random seed for reproducible simulations (if needed) and simulate the MRMC agreement data based on the config list of parameters. The data frame for 5 readers and 100 cases for 2 modalities has 1000 rows (5x100x2). The 4 columns specify the reader ID, the case ID, the modality ID, and the score, or observation.

```
# Set the simulation kind (L'Ecuyer), seed (=1), and stream (=2)
init.lecuyerRNG()
# simulate MRMC study
dFrame.newH <- sim.NormalIG.Hierarchical(config)</pre>
# check the first and last few lines of the simulated dataframe
print(dFrame.newH[1:6, ])
##
     caseID readerID modalityID
                                    score
## 1
     Case1 reader1
                          testA -1.190073
## 2
     Case2 reader1
                          testA -1.035331
## 3
     Case3 reader1
                          testA -1.831626
     Case4 reader1
                          testA 1.592661
## 5
     Case5 reader1
                          testA 0.067100
## 6
     Case6 reader1
                          testA -2.992642
print(dFrame.newH[(nrow(dFrame.newH) - 5):nrow(dFrame.newH), ])
##
         caseID readerID modalityID
                                         score
## 995
         Case95 reader5
                              testB 2.5922272
## 996
         Case96 reader5
                              testB -0.9025843
## 997
         Case97 reader5
                              testB 0.4175832
## 998
         Case98 reader5
                              testB -1.3322086
## 999
         Case99 reader5
                              testB 0.9106875
## 1000 Case100 reader5
                              testB -0.1559130
```

## 2. Using sim.gRoeMetz in iMRMC package to simulate MRMC ROC data

The following demonstrates how to use the sim.gRoeMetz function to simulate MRMC ROC data (with truth state). Again, the process starts by creating a config list of default simulation parameters that can be changed in the function call.

Here are the simulation parameters for the sim.gRoeMetz simulation function. Note that there are some parameters that are set equal for convenience and might need to be changed before simulating the data.

```
print(names(config))
```

```
[1] "modalityID.A" "modalityID.B" "nR"
                                                        "nC.neg"
                                                                        "nC.pos"
  [6] "mu.neg"
                                                        "var_rc.neg"
##
                        "var r.neg"
                                        "var c.neg"
                                                                        "mu.pos"
## [11] "var r.pos"
                        "var_c.pos"
                                        "var rc.pos"
                                                        "mu.Aneg"
                                                                        "var_r.Aneg"
## [16] "var_c.Aneg"
                        "var_rc.Aneg"
                                        "mu.Apos"
                                                        "var_r.Apos"
                                                                        "var_c.Apos"
## [21] "var rc.Apos"
                        "mu.Bneg"
                                        "var r.Bneg"
                                                        "var c.Bneg"
                                                                        "var rc.Bneg"
## [26] "mu.Bpos"
                        "var r.Bpos"
                                        "var c.Bpos"
                                                        "var rc.Bpos"
```

We again set the seed for reproducible simulations (if needed) before we simulate the data. As we see below, the simulated data starts with truth state of each case. This is followed by the reading scores from each of the readers. Since we simulate 50 positive cases and 50 negative cases, there are 100 rows for the truth and 100x5x2 = 1000 rows for the scores from 5 readers for 2 modalities. Therefore, the data frame has 1100 rows total.

```
# simulate MRMC study
set.seed(1, kind = "L'Ecuyer-CMRG")
dFrame.gRM <- sim.gRoeMetz(config)</pre>
# check the first and last few lines of the simulated dataframe
print(dFrame.gRM[1:6, ])
##
     readerID
                caseID modalityID score
## 1
        truth negCase1
                            truth
## 2
        truth negCase2
                            truth
                                       0
        truth negCase3
## 3
                            truth
                                       0
## 4
        truth negCase4
                            truth
                                       0
## 5
        truth negCase5
                                       0
                            truth
## 6
        truth negCase6
                                       0
                            truth
print(dFrame.gRM[(nrow(dFrame.gRM) - 5):nrow(dFrame.gRM), ])
##
        readerID
                    caseID modalityID
                                            score
## 1095 reader5 posCase45
                                testB
                                        2.9342657
## 1096 reader5 posCase46
                                testB
                                        1.0639768
## 1097 reader5 posCase47
                                        0.9885389
                                testB
## 1098
        reader5 posCase48
                                testB -0.5130517
## 1099
        reader5 posCase49
                                testB 0.9254049
## 1100
        reader5 posCase50
                                testB 2.4164279
```

To combine the truth data and the reader scores and change the data to a dataframe with 5 columns (readerID, caseID, modalityID, score, truth) we can use the function undoIMRMCdf in iMRMC pacakge.

```
dFrame.gRM.2 <- undoIMRMCdf(dFrame.gRM)

# check the first and last few lines of the simulated dataframe
print(dFrame.gRM.2[1:6, ])</pre>
```

```
##
      caseID readerID modalityID
                                       score truth
## 1 negCase1
              reader1
                           testA 0.36795249
                           testB 0.05607467
## 2 negCase1 reader2
                                                 0
## 3 negCase1
              reader3
                           testB -0.46914121
                                                 0
## 4 negCase1 reader4
                           testA -1.38686722
                                                 Ω
## 5 negCase1 reader1
                           testB -0.49895246
## 6 negCase1 reader5
                           testA -1.64836669
```

```
print(dFrame.gRM.2[(nrow(dFrame.gRM.2) - 5):nrow(dFrame.gRM.2), ])
```

```
## 995 posCase9 reader5 testA 0.9233879 1
## 996 posCase9 reader4 testB 4.1614474 1
## 997 posCase9 reader4 testA 3.0874475 1
## 998 posCase9 reader2 testB 3.4945769 1
## 999 posCase9 reader1 testA 2.2259825 1
## 1000 posCase9 reader5 testB 1.3017587 1
```