Presentation of new R functions for forecast verification

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please go to https://github.com/sieste/ic3-workshop and download all files into a new directory

This workshop

- https://github.com/sieste/ic3-workshop
- ▶ hands-on session
- presentation of some new R functions
- interpretation of the output

Overview of available verification packages in R

- verification
 - developed at NCAR
 - ▶ methods: Brier, CRPS, ROC, reliability diagram, rank histogram
- ▶ s2dverification
 - currently developed at IC3
 - methods: ACC, RMSSS, plotting!

New contributions

- ensemble verification
- uncertainty estimates
- ► comparative verification
- ▶ ... work in progress

```
source("R/toydata.r")

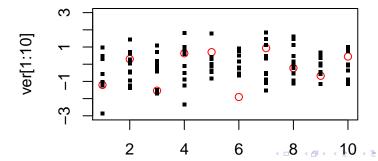
## Loading required package: boot

source("R/rankhist.r")
source("R/rel-diag.r")
source("R/ensemble-scores.r")
```

Gaussian toy data

- ▶ Gaussian ensemble data with mean mu.ens and stdev sd.ens
- ▶ Gaussian verification with mean mu.ver and stdev sd.ver
- number of samples N
- number of ensemble members K

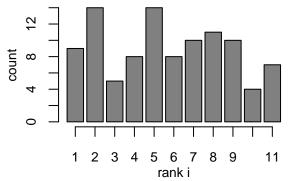
Gaussian toy data



Rank histogram

- $ightharpoonup r_i$: rank of the verification in the ordered ensemble
- e.g. ver=1.5, ens= $\{1,2,3\}$: rank = 2
- histogram over the observed ranks

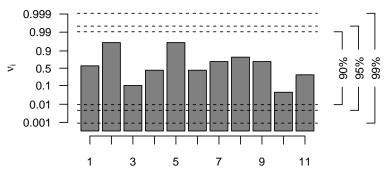
```
source("R/rankhist.r")
rh <- with(toydata, rankhist(ens, ver))
PlotRankhist(rh, mode="raw")</pre>
```



Rank histogram on probability paper

- ▶ H_0 : the individual rank counts are $\sim Binomial(N, \frac{1}{K+1})$
- ightharpoonup plot the cumulative likelihood of the observed rank counts under H_0

PlotRankhist(rh, mode="prob.paper")



Rank histogram significance tests

- ▶ Pearson χ^2 -test
- ▶ Jolliffe-Primo χ^2 -decomposition

```
rh.tests <- rankhist.tests(rh)
print(rh.tests)</pre>
```

```
## pearson.chi2 jp.slope jp.convex
## test.statistic 11.3200 1.0890 0.7601
## p.value 0.3331 0.2967 0.3833
```

Fair Brier Score for binary ensemble forecasts

▶ j ... verification, 1 = yes, 0 = no▶ i ... number of ensemble members that predict the event
▶ $Br(i,j) = (j-\frac{i}{K})^2 - \frac{i(K-i)}{K^2(K-1)}$ source("R/ensemble-scores.r")

tau <-1 # exceedance threshold
with(toydata, mean(fairbrier(ens, ver, tau)))

[1] 0.1024

Fair continuously ranked probability score for ensemble forecasts

- ▶ fair Brier Score integrated over all possible thresholds
- $crps(e, y) = \langle |y e_i| \rangle \frac{1}{2K(K-1)} \langle |e_i e_j| \rangle$

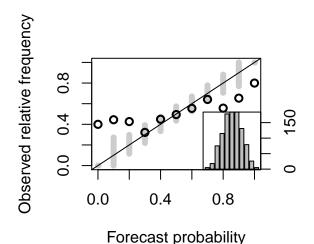
```
source("R/ensemble-scores.r")
fcrps <- with(toydata, faircrps(ens, ver))
mean(fcrps)</pre>
```

Reliability diagram

```
N <- 1000
mu <- runif(N)
toydata <- GenerateToyData(N=N, mu.ens=mu, mu.ver=mu)
tau <- .5
i <- with(toydata, rowSums(ens > tau))
j <- with(toydata, 1 * (ver > tau))
```

Reliability diagram

```
source("R/rel-diag.r")
rd <- rel.diag(probs=i/K, ver=j, nbins=11, plot=TRUE)</pre>
```



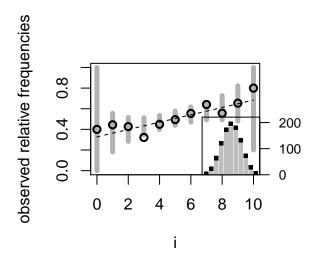
Reliability diagram

print(rd)

```
##
      p.avgs cond.probs cbar.lo cbar.hi
## 1
         0.0
                  0.4000
                          0.0000
                                  0.0000
                  0.4444
                         0.0000
                                  0.2727
## 2
         0.1
## 3
         0.2
                  0.4267
                                  0.2973
                         0.1147
## 4
         0.3
                  0.3217
                         0.2188
                                   0.3922
## 5
         0.4
                  0.4494
                         0.3332
                                  0.4727
##
         0.5
                  0.4946
                         0.4350
                                  0.5741
## 7
         0.6
                  0.5543
                          0.5247
                                   0.6706
## 8
         0.7
                  0.6412
                          0.6229
                                   0.7778
##
   9
         0.8
                  0.5570
                          0.7059
                                   0.8816
## 10
         0.9
                  0.6538
                          0.7692
                                   1.0000
## 11
         1.0
                  0.8000
                          1.0000
                                   1.0000
```

Fair reliability diagram

frd <- fair.rel.diag(i=i, j=j, K=K, plot=TRUE, plot.refin=TRUE)</pre>



Fair reliability diagram

print(frd)

```
cond.probs HO.line cbar.lo cbar.hi
##
##
   1
              0.4000
                      0.3271
                              0.0000
                                       1.0000
                      0.3627
                              0.1818
##
              0.4444
                                       0.5557
##
   3
              0.4267
                      0.3984
                                       0.5161
                              0.2833
##
              0.3217
                      0.4340
                                       0.5136
                              0.3486
##
              0.4494
                      0.4696
                              0.3986
                                       0.5360
##
              0.4946
                      0.5053
                              0.4391
                                       0.5773
##
  7
              0.5543
                      0.5409
                              0.4702
                                       0.6183
##
   8
              0.6412
                      0.5765
                              0.4917
                                       0.6593
##
   9
       8
              0.5570
                      0.6122
                              0.4921
                                       0.7273
##
   10
              0.6538
                      0.6478
                              0.4800
                                       0.8214
##
   11
      10
              0.8000
                      0.6834
                               0.2000
                                       1.0000
```

Comparative ensemble verification

- ▶ We want to address the question: Is the forecast ens better than a reference forecast ens.ref at predicting the same verification ver?
- Our hindcast dataset now has 3 members

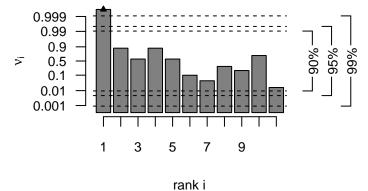
Comparison of two imperfect ensemble forecasts

- ens and ver as before
- additionally: ens.ref, a benchmark ensemble, to which the performance of ens is compared

Good statistical tests should find that both ensembles are unreliable (biased), and that ens is more reliable than ens.ref.

Rank histogram analysis of ens.ref

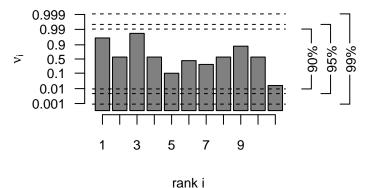
```
rh.ref <- with(toydata2, rankhist(ens.ref, ver))
PlotRankhist(rh.ref, mode="prob.paper")</pre>
```



Rank histogram analysis of ens.ref

Rank histogram analysis of ens

```
rh <- with(toydata2, rankhist(ens, ver))
PlotRankhist(rh, mode="prob.paper")</pre>
```



Rank histogram analysis of ens

```
rh.tests <- rankhist.tests(rh)
print(rh.tests)

## pearson.chi2 jp.slope jp.convex
## test.statistic 13.9600 3.84400 0.2051
## p.value 0.1748 0.04992 0.6506</pre>
```

Comparison of rank histograms: AnalyzeRankhistDifference

rh.comp is a matrix that summarizes the rank histogram comparison:

```
print(t(as.matrix(rh.comp)))
```

```
##
            pearson.chi2 jp.slope jp.convex
## score.diff
                 19.580 13.580 6.343
## p.value
                  0.060 0.010
                                   0.150
## Q0.01
                -38.150 -1.894 -15.417
## Q0.05
                -15.191 -1.055 -5.049
## Q0.1
                 -3.388 3.681 -1.023
## Q0.9
                 59.906 22.357 31.172
## Q0.95
                 81.686 24.675 39.391
## Q0.99
                116.853 31.923
                                  52.397
```

Similar for the specific scores for slope and convexity:

p-values:

▶ bootstrap quantiles:

Comparison of fair Brier Scores

Analysis of fair Brier score difference

```
fbr.comp <- with(toydata3,
  AnalyzeFairBrierDifference(ens, ens.ref, ver,
                              tau=.5, n.boot=100))
print(as.matrix(fbr.comp))
                        [,1]
##
## fair.brier.diff
                    0.01956
## p.value
                    0.20000
                   -0.03849
## Q0.01
## Q0.05
                   -0.02274
                   -0.01193
## Q0.1
                    0.05360
## Q0.9
                    0.05712
## Q0.95
## Q0.99
                    0.06434
```

Comparison of fair crps

```
fcrps.comp <- with(toydata3,</pre>
  AnalyzeFairCrpsDifference(ens=ens, ens.ref=ens.ref,
                             ver=ver, n.boot=100))
print(as.matrix(fcrps.comp))
                      [,1]
##
   fair.crps.diff 0.12023
## p.value
                  0.00000
## Q0.01
                  0.04691
## Q0.05
                  0.05940
                  0.06810
## Q0.1
## Q0.9
                  0.18077
                  0.19096
## Q0.95
## Q0.99
                   0.20236
```

Some actual data

- tropical sea surface temperature data
- ▶ 51 years
- ▶ 10 lead times (10 years)
- ▶ 8 different ensembles
- ▶ up to 10 members each

```
load("R/SST-raw.Rdata")
print(dim(SST.trop))

## [1] 51 10 8 10
print(dim(SST.trop.obs))

## [1] 51 10
```

Some actual data

the available models

A specific analysis

set parameters

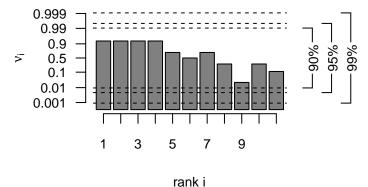
```
lead <- 7
model <- "hadcm3_ff"
model.ref <- "hadcm3_an"
dates <- 1:40
members <- 1:10
members.ref <- 1:10</pre>
```

▶ get data:

```
ens <- SST.trop[dates, lead, model, members]
ens.ref <- SST.trop[dates, lead, model.ref, members.ref]
ver <- SST.trop.obs[dates, lead]</pre>
```

Rank histogram of ens.ref

```
rh.ref <- rankhist(ens.ref,ver)
PlotRankhist(rh.ref, mode="prob.paper")</pre>
```



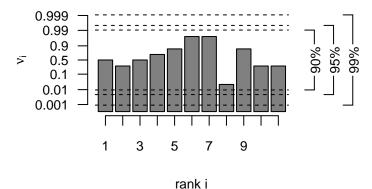
Rank histogram of ens.ref

```
rankhist.tests(rh.ref)
```

```
## pearson.chi2 jp.slope jp.convex
## test.statistic 13.3500 1.122e+01 0.002885
## p.value 0.2048 8.081e-04 0.957167
```

Rank histogram of ens

```
rh <- rankhist(ens,ver)
PlotRankhist(rh, mode="prob.paper")</pre>
```



Rank histogram of ens

```
rankhist.tests(rh)
```

```
## pearson.chi2 jp.slope jp.convex
## test.statistic 13.3500 0.0625 3.9490
## p.value 0.2048 0.8026 0.0469
```

Rank histogram comparison

```
rh.comp <- AnalyzeRankhistDifference(ens, ens.ref,
                                  ver, n.boot=100)
print(t(as.matrix(rh.comp)))
##
            pearson.chi2 jp.slope jp.convex
## score.diff
                   0.000 11.1600 -3.946
                   0.490 0.0000
                                     0.840
## p.value
## Q0.01
                 -26.554 -0.6922 -13.215
## Q0.05
                 -18.727 0.8324 -6.823
## Q0.1
                 -6.655 3.6180 -4.015
## Q0.9
                28.655 18.4220 4.903
                  32.532 20.4600 7.525
## Q0.95
## Q0.99
                  50.072 26.1248 14.706
```

Brier Score comparison

```
br.comp <- AnalyzeFairBrierDifference(ens, ens.ref, ver,</pre>
                                        tau=mean(ver),
                                        n.boot=100)
print(as.matrix(br.comp))
                          [,1]
##
## fair.brier.diff -3.383e-18
## p.value
                  5.100e-01
## Q0.01
                    -8.678e-02
## Q0.05
                    -6.133e-02
                    -4.478e-02
## Q0.1
                    4.006e-02
## Q0.9
                     5.353e-02
## Q0.95
## Q0.99
                     6.488e-02
```

Crps comparison

```
crps.comp <- AnalyzeFairCrpsDifference(ens, ens.ref,</pre>
                                       ver, n.boot=100)
print(as.matrix(crps.comp))
##
                      [,1]
## fair.crps.diff 0.013999
## p.value
               0.010000
                0.002906
## Q0.01
## Q0.05
               0.005512
## Q0.1
               0.006312
## Q0.9
                0.020968
## Q0.95
                0.023224
## Q0.99
                 0.026772
```