

DELTA SCORING FUNCTION REFERENCE

D. Atanasov

October 15, 2021

1 Introduction

This document describes the the set of functions and the way of their usage for obtaining the DELTA SCORING approach for test evaluations.

2 Instalation

To install place the folder in the MATLAB path and rename it to +deltaScoring.

3 Ussage

Here a example of a simple ussage of the package.

Suppose the raw dichotomous item response id placed in variable itemScore.

To estimate the items delta by bootstrapping procedure

```
[ItemDelta, estimatedDeltaSE] = ...  
    deltaScoring.estimate.itemDeltaBootstrap(  
    itemScore);
```

is called. The resulted item deltas and the corresponding standard error of estimate are returned in variables ItemDelta, estimatedDeltaSE.

The classical person D-scores are calculated usind the response paterns in itemScore and already calculated item deltas.

```

personDscores = ...
    deltaScoring.scoring.dScore(ItemDelta,
        itemScore, opt.Dscore_method);

```

Here `opt` is a structure containing the options for the considered delta scoring model. It can be generated by

```
opt = deltaScoring.scoring.Options;
```

Here and after the default will be the model RFM2. If a RFM3 model is required this can be stated in

```
opt.model = 3;
```

and the corresponding options should be passed to the functions.

The item properties location b and shape s can be obtained by

```

[params, CI, ~, Results] = ...
deltaScoring.estimate.logitDeltaFit(itemScore,
    personDscores, opt);

```

where *params* contains the matrix with corresponding parameters for any item in the test $[b, s]$. The first column corresponds to the location parameter b while the second represents the shape s . If the model is RFM3, the guessing parameter is in the third column.

The matrix *CI* contains the 95% confidence interval of the estimated values. *Results* contains additional fitting parameters (for example MAD is available in *Results.MAD*).

The persons true scores can be calculated by

```

personTrueScores = ...
deltaScoring.scoring.trueScore(ItemDelta,
    ItemParameters, personDscores, opt);

```

and the SE

```

personTrueScoresSE = ...
deltaScoring.scoring.trueScoreSE(ItemDelta,
    ItemParameters, personDscores, opt);

```

A latent version of the location and shape parameters (together with their SE) can be obtained by

```

[LatentParams, LatentSE] = deltaScoring.estimate.
    ML_RFM_params(itemScore, personDscores, opt);

```

The corresponding MAD is obtained by

```
LatentMAD = deltaScoring.item.MAD(LatentParams ,
    Results.observationsLogitDelta ,opt);
```

where *Results.observationsLogitDelta* is calculated with *deltaScoring.estimate.logitDeltaFit* above and contains the proportion of observed correct answers for the values on the D-score scale.

The corresponding latent values of the person D-scores (and SE) can be obtained by

```
[latentScore, ~, latentScoreSE]= ...
deltaScoring.estimate.ML_RFM_scores( itemScore,
    LatentParams, opt)
```

The estimations of latent item parameters and person D-scores can be iterated until a convergence is reached

```
OldLatentParams = zeros(size(LatentParams));
while max(abs(OldLatentParams - LatentParams)) > eps
    OldLatentParams = LatentParams;
    [LatentParams,LatentSE] = ...
        deltaScoring.estimate.ML_RFM_params(
            itemScore, personDscores, opt);
    [latentScore, ~, latentScoreSE]= ...
        deltaScoring.estimate.ML_RFM_scores(
            itemScore,LatentParams, opt)
end
```

Equating of different tests can be reached by functions located in *deltaScoring.equating*. Here an example of equating of classical (nonlatent) parameters of the test will be presented. The equating is based on calculation of two constants *A* and *B* which represents the change of scale for the test equating. These constants can be calculated on the base of item deltas of the target test *targetDeltas*, item deltas (already calculated above) and few common items between the two tests (stated in the variable *CommonItems*)

```
[A,B] = deltaScoring.equating.constants(
    targetDeltas, ItemDelta, CommonItems);
equatedItemDeltas = deltaScoring.equating.rescale
    (ItemDelta,A,B);
equatedDscores = deltaScoring.scoring.dScore(
    ItemDelta,itemScore,opt.Dscore_method);
```

4 Function reference

`deltaScoring.assembly.multipleTest`

```
multipleTest(nOfTest, nOfItems, itemDeltas,
             varargin)
```

Returnt the item indexes from the itemParams
which compose a number of tests test.

nOfTest
nOfItems - number of items in teh test.
itemDeltas - list (column) of estimated item deltas

Optional parameters: ['Name',value] pairs
The approach is based on Linear Optimal Test Design
Uses singleTest.

`deltaScoring.assembly.singleTest`

Returnt the item indexes from the itemParams
which compose a test.

nOfItems - number of items in teh test.
itemDeltas - list (column) of estimated item deltas

Optional parameters: ['Name',value] pairs
The approach is based on Linear Optimal Test Design

`deltaScoring.dif.conditionalDIF`

```
conditionalDIF(focal_params,reference_params,o)
```

probabilities for correct item performance between

focal and reference group, based on parameters of the items for the two groups.

o - deltaScoring.scoring.Options

deltaScoring.dif.ESonDIF

```
[onFocal, onReference] = ESonDIF(focal_params,
    reference_params,o)
Calculates the effect size on focal and reference
group for DIF
```

INPUT:

```
    focal_params      - item parameters,
                        estimated on focal group
    reference_params  - item parameters,
                        estimated on reverence group
    o                  - options
```

deltaScoring.dif.functioning

Calculates different characteristics, corresponding to the DIF

deltaScoring.dif.Mantel_Haenszel

```
[a_MH, log_a_MH_SE, da_MH, z, p, MH, type, against]=
    Mantel_Haenszel(response, score, groups, reference)
Calculates Mantel-Haenszel statistics
```

deltaScoring.dif.MPDonDIF

```
[onFocal, onReference] = MPDonDIF(dif)
dif = pcr_f - pcr_r;
```

deltaScoring.dif.paramsForGroups

```
[focal_params,reference_params,deltasF,deltasR] =  
  paramsForGroups(itemResponse,group,latent,o)  
Calculates the item parameters for the focal and  
  reference group
```

```
group 0 - reference, 1 - focal  
params by default are latent
```

deltaScoring.dif.plotConditionalDIF

```
plotConditionalDIF(dif,opt,visible)  
Plots conditional DIF for the Focal and reference  
  group
```

deltaScoring.dif.plotICC

```
plotICC(focal_params,reference_params,o,visible)  
Plots ICC for the Focal and reference group
```

deltaScoring.dif.plotTSC

```
plotTSC(focal_params,reference_params,o,visible)  
Plots test score for the Focal and reference group
```

deltaScoring.dif.testing

```
[DIF, DIFT, HA, HB, HAT, HBT, Results] = testing(  
  focal_params, reference_params, o)  
Tests DIF having estimated item parameters for  
  focal  
and reference group.
```

DIF - indicator for a specific item
 0 - no DIF
 1 - against focal
 2 - against reference
 DIFT - DIF on test level
 HA - Hypothesis nder approach A. Cell array of
 structures
 HB - Hypothesis nder approach B. Cell array of
 structures
 HAT on test level
 HBT on test level
 Results - structure with detailed results

deltaScoring.equating.constants

[A,B] = constants(Base_test_deltas,New_test_deltas,
 common_items)

 Y X
 Calculates the rescaling constants, based on common
 items
 between two test.

INPUT:

Base_test_deltas - item deltas of the base
 test
 New_test_deltas - item deltas of the new
 test
 common_items - two columns
 [base_test_item_id new_test_item_id]

OUTPUT:

A and B

deltaScoring.equating.dscores_common

NOT IN USE

deltaScoring.equating.dscore_rfm

```
[Dscore_equated] = dscore_rfm(X_params, X_rescaled,
                               Dscores, type, o)
Calculates equated latent D-Score based on the
latent parameters
```

INPUT:

```
X_params      - latent parameters of the test
X_rescaled    - rescaled parameters of the test
                after equating
Dscores       - persons D-score
type          - default value is m1
o             - options
```

OUTPUT:

```
Dscore_equated - equated D-score
```

deltaScoring.equating.dscoreOnSubtest

NOT IN USE

deltaScoring.equating.rescale

```
res = rescale(deltas,A,B)
Rescale the item deltas of a test according
to rescaling constants A and B.
```

deltaScoring.equating.rescale_rfm

```
[X_params_rescaled, opts] = rescale_rfm(X_params,
                                          Y_params,common_items,method_type,o)
```


Calculates the rescaled latent item parameters of test X on the scale of the base test Y.

INPUT:

X_params - parameters of test X
Y_params - parameters of test Y
common_items - common items; two columns
 [base_test_item_id
 new_test_item_id]
method_type - Rescaling of the shape
 parameter s has two options
 direct [
 default]
 |
 tough_a

OUTPUT:

X_params_rescaled
opts - Structure with
 bA, bB, sA, sB

deltaScoring.estimate.EM_RFM

Function [pars, ability] = irt.
ItemParametersEstimate_EM_3PL(data, o)
estimates the parameters of the item
characteristic
curves under the IRT model using the EM
algorithm.

Input:

data - Dichotomous item response
o - scoring.Options (optional)

Output:

pars - Item parameters
[difficulty, discrimination, guessing]

deltaScoring.estimate.itemDeltaBootstrap

deltaScoring.estimate.latentLklh

```
latentLklh(xi,itemResponse,deltaScores,o)
Calculates person likelihood on a specific test
  with a specific item response
For internal use in estimations
```

deltaScoring.estimate.logitDeltaFit

deltaScoring.estimate.logitDeltaPlot

```
h = logitDeltaPlot(GF,observedLogitDelta,o)
Plots the fit and the estimated logistics curve
Returns the figure object
```

INPUT:

```
GF - output from logitDeltaFit
observedLogitDelta - from Results of
  logitDeltaFit
o - options
dScale
```

deltaScoring.estimate.ML_RFM_params

```
[pars,se] = ML_RFM_params( itemResponse ,
  deltaScores, o)
Estimates the latent parameters of the items base
  on
RFM model, using JML approach.
```

INPUT:

itemResponse - dichotomous item response
deltaScores - person D-scores
o - options

OUTPUT:

pars - estimated parameter values
se - standard errors of the
estimates

deltaScoring.estimate.ML_RFM_scores

```
[scores,se, see]=ML_RFM_scores( itemResponse,  
itemParams, o)
```

Estimates the latent parameters for person
abilities, based on
RFM model, using JML approach.

INPUT:

itemResponse - dichotomous item response
itemParams - person D-scores
o - options

OUTPUT:

scores - person latent D-scores
se - standard errors of the
estimates
see - analitical solution for se

deltaScoring.generate.guttman

```
guttman(NofPersons,itemParams,options,reverse)  
Generates a item response according Guttman concept
```

deltaScoring.generate.itemResponse

```
[res, out] = itemResponse(Persons,itemParams,
    options,env)
```

Generates a item response for a number of Persons
over a set
of items, defined by their item Parameters.

env is a structure, containing additional
information about cheating and guessing.

deltaScoring.item.characteristicsFromParameters

```
res = characteristicsFromParameters(item_params,o)
```

Calculates the item characteristics from item
parameters

INPUT:

```
    item_params - item parameters
    o           - options
```

OUTPUT

```
res - [location, discrimination]
```

deltaScoring.item.icc

```
res = icc(itemParameters,o)
```

Plots ICC curves under given item parameters

INPUT:

```
    itemParameters
    o - options
```

OUTPUT:

```
res - figure handle
```

deltaScoring.item.MAD

```
res = MAD(params,observedLogitDelta,o)
Calculates the Mean Absolute Difference between
opserved probability for correct response and
predicted, according RFM model
```

INPUT:

```
params           - item parameters
observedLogitDelta - observed PCR
o                - options
```

OUTPUT:

```
res - MAD values
```

deltaScoring.item.parametersFromCharacteristics

```
res = parametersFromCharacteristics(location,
discrimination,o)
Calculates the item parameters from item
characteristics
```

INPUT:

```
location
discrimination
o          - options
```

OUTPUT

```
res - [b,s]
```

deltaScoring.person.aberrant

```
res = aberrant(itemParams,itemDeltas,Dscores,
itemResponse,options)
Find aberrant person behaviour according
to the "quantile method" according to
```

An Examination of Different Methods of Setting
Cutoff Values in Person Fit Research
Amin Mousavi, Ying Cui & Todd Rogers

INPUT:

itemParams - item parameters
itemDeltas - item deltas
Dscores - persons D-scores
itemResponse - dichotomous item response

OUTPUT:

res - aberrant indicator 0/1

deltaScoring.person.fitHStatistics

fitHStatistics(item_response)
Calculates the H statistics for the
dichotomous item response

deltaScoring.person.fitIndexZ

fitIndexZ(lklh,Elklh,Vlklk)
Inputs are from person.likelihood

based on
D. Dimitrov, R. Smith. Adjusted Rasch Person-Fit
Statistics. J. of
Applied measurement. 2006

deltaScoring.person.fitMSE

[Outfit, Infit] = fitMSE(item_response,
expected_item_score)
Calculates the Outfit and Infit of the MSE fit

deltaScoring.person.fitU

```
fitU( params, Dscore, responses, o)
Calculates U statistics
```

deltaScoring.person.fitUD1

```
fitUD1(deltas, Dscore, responses)
Calculates UD1 Statistics
```

deltaScoring.person.fitUD2

```
fitUD2( deltas, Dscore, responses, params, o)
Calculates UD2 statistics
```

deltaScoring.person.likelihood

```
[res, expected, variance] = likelihood(dScores,
    item_parameters,item_response,o)
Calculates the likelihood for a specific response
    pattern in item_response from a
person with ability in dScores, over a set of items
    in item_parameters.
```

```
based on
D. Dimitrov, R. Smith. Adjusted Rasch Person-Fit
    Statistics. J. of
Applied measurement. 2006
```

deltaScoring.person.plotConditionalSE

```
plotConditionalSE(dScores,SE)
Plots conditional standard error
```

deltaScoring.person.responseProbability

```
res = responseProbability(delta,item_parameters,o)
Calculates the probability for correct response
from a
person with ability delta over the set of items
with
parameters, defined in item_parameters.
```

INPUT

```
delta - Person ability
item_parameters - Item parameters [b,s,c]
o - Delta Scoring Options
```

deltaScoring.poly.ccr

```
ccr(itemParameters,observedLogitDelta,o)
Plots Cathegory Characteristic Curves for the
Polytomous items
```

INPUT:

```
itemParameters - row vector of values of the
difficulty
parameter for item grades.
observedLogitDelta - person ability value
o - options
```

deltaScoring.poly.itemPerformance

```
res = itemPerformance(itemParameters,delta,o)
Calculates the probability for correct performance
for the polytomous item
```

INPUT:

itemParameters - row vector of values of the
difficulty
parameter for item grades.
delta - person ability value
o - options

OUTPUT:

res - probability for correct performance

deltaScoring.poly.logitDeltaPlot

h = logitDeltaPlot(GF,observedLogitDelta,o)
Plots the fit and the estimated logistics curve
Returns the figure object

INPUT:

GF - output from logitDeltaFit
observedLogitDelta - from Results of
logitDeltaFit
o - options
dScale

OUTPUT:

h - figure handle

deltaScoring.scoring.dScore

Returns the so called d-Score for a person
with a given response vector over a set
of items with precalculated deltas;

INPUT:

itemDeltas - item delta values
response - 0/1 item response
t - type

```
total / relative_to_n /
relative_to_d [default]
```

deltaScoring.scoring.dScoreSE_IRT

Returns the so called d-Score SE for a person
with a given ability theta (on a logit
scale)
over a set of items with IRT parameters [b,a,c];

left here only for convenience.

deltaScoring.scoring.itemPersonMap

deltaScoring.scoring.observedLogitDelta

observedLogitDelta(ItemResponse, Dscore,o)
Calculates the proportions of the correct scores
for different values of the dScale

INPUT:

ItemResponse - dichotomous item response 0/1
Dscore - estimated person's dScore
o - options

OUTPUT:

res - proportion of correct answers on
dScale

deltaScoring.scoring.Options

option = Options(varargin)
Defines the options for DELTA SCORING
Default Values

```

      NofSamplesForBootstrapping: 1000
sampleProportionForBootstrapping: 0.1000
      estTypeForBootstrapping: mode
      dScale: [21x1 double]
      Models: {1x3 cell}
      ModelNames: {RFM1 RFM2
      RFM3}
      ModelFixedParams: [1x1 struct]
Model_coefficients: {b s c d}
      model: 2
      type: raw
skipObservedOnPlot: 2
      aberrantQuantile: 0.7000
      EM: [1x1 struct]
      StartingPoint: [0.5000 1
      0.1000]
      Lower: [0.0100 0.2000
      0]
      Upper: [0.9900 5
      0.5000]
      RFM_params_method: 'constrained'

```

deltaScoring.scoring.PCR

```

res = PCR(params,delta,o)
Probability for correct response

```

```

INPUT:
  params - logistics parameres
  delta  - delta values
  o      - options
          mmodel

```

deltaScoring.scoring.poly2dih

```
[DIHscores,Poly,Org] = poly2dih(Response)
Convert polytomous to dihotomous item response
```

INPUT:

Response - polytomous item response

OUTPUT:

DIHscores - dihotomous ite response
Poly - indicator for polytomous items
Org - Labels, etc. for poly items

deltaScoring.scoring.scaledScore

```
res=scaledScore(scores,t)
Scales the score according to type t
t = 'range' scales in range 0..100
```

deltaScoring.scoring.trueScore

```
res = trueScore(itemDeltas, parameters, dScore, o)
Calculates the true-score measure for person's
dScore
on a set of items with delta scores in itemDeltas,
logistic parameters of the items and the persons
dScore
```

INPUT:

itemDeltas - item's delta scores
parameters - item's logistics parameters
dScore - persons dScore
o - Options (defaults scoring.Options)

deltaScoring.scoring.trueScoreSE

```
res = trueScoreSE(itemDeltas, parameters, dScore, o
)
```

Calculates the true-score SE measure for person's
dScore
on a set of items with delta scores in itemDeltas,
logistic parameters of the items and the persons
dScore

INPUT:

- itemDeltas - item's delta scores
- parameters - item's logistics parameters
- dScore - persons dScore
- o - Options (defaults scoring.Options)