Package 'FMP'

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_	otion Estimates Filtered Monotonic Polynomial IRT Models as decribed by Liang and Browne (2015) <doi:10.3102 1076998614556816="">.</doi:10.3102>									
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eap	Compute eap trait estimates for FMP and FUP models									
Descrip Con	ption mpute eap trait estimates for items fit by filtered monotonic polynomial IRT models.									
Usage										
ea	ap(data, bParams, NQuad = 21, priorVar = 2, mintheta = -4, maxtheta = 4)									

2 eap

Arguments

data N(subjects)-by-p(items) matrix of 0/1 item response data.

bParams A p-by-9 matrix of FMP or FUP item parameters and model designations. Columns

1 - 8 hold the (possibly zero valued) polynomial coefficients; column 9 holds the

value of k.

NQuad Number of quadrature points used to calculate the eap estimates.

priorVar Variance of the normal prior for the eap estimates. The prior mean equals 0.

mintheta, maxtheta

NQuad quadrature points will be evenly spaced between mintheta and maxtheta

Value

```
eap trait estimates.
```

Author(s)

Niels Waller

```
## this example demonstrates how to calculate
## eap trait estimates for a scale composed of items
## that have been fit to FMP models of different
## degree
NSubjects <- 2000
## Assume that
## items 1 - 5 fit a k=0 model,
## items 6 - 10 fit a k=1 model, and
## items 11 - 15 fit a k=2 model.
 itmParameters <- matrix(c(</pre>
 # b0
        b1 b2 b3
                            b4 b5, b6, b7, k
  -1.05, 1.63, 0.00, 0.00, 0.00, 0, 0, 0,
                                                0, #1
  -1.97, 1.75, 0.00, 0.00, 0.00, 0,
                                    0, 0,
                                                0, #2
  -1.77, 1.82, 0.00, 0.00, 0.00, 0,
                                    0, 0,
                                                0, #3
  -4.76, 2.67, 0.00, 0.00, 0.00, 0,
                                    0, 0,
                                                0, #4
  -2.15, 1.93, 0.00, 0.00, 0.00, 0,
                                      0, 0,
                                                0, #5
  -1.25, 1.17, -0.25, 0.12, 0.00, 0,
                                      0, 0,
                                                1, #6
  1.65, 0.01, 0.02, 0.03, 0.00, 0,
                                      0, 0,
                                                1, #7
  -2.99, 1.64, 0.17, 0.03, 0.00, 0,
                                       0, 0,
                                                1, #8
  -3.22, 2.40, -0.12, 0.10, 0.00, 0,
                                      0, 0,
                                                1, #9
  -0.75, 1.09, -0.39, 0.31, 0.00, 0,
                                       0, 0,
                                                1, #10
  -1.21, 9.07, 1.20,-0.01,-0.01, 0.01, 0, 0,
                                                2, #11
  -1.92, 1.55, -0.17, 0.50, -0.01, 0.01, 0, 0,
                                                2, #12
  -1.76, 1.29, -0.13, 1.60, -0.01, 0.01, 0, 0,
                                                2, #13
  -2.32, 1.40, 0.55, 0.05, -0.01, 0.01, 0, 0,
                                                2, #14
  -1.24, 2.48, -0.65, 0.60, -0.01, 0.01, 0,
                                                2),#15
  15, 9, byrow=TRUE)
```

[#] generate data using the above item parameters

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erf

Utility fnc to compute the components for an empirical response function

Description

Utility function to compute empirical response functions.

Usage

```
erf(theta, data, whichItem, min = -3, max = 3, Ncuts = 12)
```

Arguments

theta Vector of estimated latent trait scores.
data A matrix of binary item responses.

whichItem Data for an erf will be generated for whichItem.

min Default = -3. Minimum value of theta.

max Default = 3. Maximum value of theta.

Ncuts Number of score groups for erf.

Value

probs A vector (of length Ncuts) of bin response probabilities for the empirical re-

sponse function.

centers A vector of bin centers.

Ni Bin sample sizes.

se.p Standard errors of the estimated bin response probabilities.

Author(s)

Niels Waller

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Examples

```
NSubj <- 2000
#generate sample k=1 FMP data
b <- matrix(c(</pre>
   #b0
                      b3
                              b4
                                  b5 b6 b7 k
         b1
                b2
  1.675, 1.974, -0.068, 0.053, 0,
                                 0, 0, 0, 1,
  1.550, 1.805, -0.230, 0.032, 0,
                                  0,
                                      0,
  1.282, 1.063, -0.103, 0.003,
                              0,
                                  0,
                                      0,
  0.704, 1.376, -0.107, 0.040,
                              0,
                                  0,
                                      0,
 1.417, 1.413, 0.021, 0.000,
                              0,
                                  0,
                                      0,
 -0.008, 1.349, -0.195, 0.144, 0,
                                  0,
                                      0,
 0.512, 1.538, -0.089, 0.082, 0, 0,
                                     0,
 0.122, 0.601, -0.082, 0.119, 0, 0, 0,
                                          0, 1,
 1.801, 1.211, 0.015, 0.000, 0, 0, 0,
                                         0, 1,
 -0.207, 1.191, 0.066, 0.033, 0, 0, 0,
                                         0, 1,
 -0.215, 1.291, -0.087, 0.029, 0, 0, 0,
                                         0, 1,
 0.259, 0.875, 0.177, 0.072, 0, 0, 0,
                                          0.1.
 -0.423, 0.942, 0.064, 0.094, 0, 0, 0, 0, 1,
 0.113, 0.795, 0.124, 0.110, 0, 0, 0, 0, 1,
 1.030, 1.525, 0.200, 0.076, 0, 0, 0, 0, 1,
 0.140, 1.209, 0.082, 0.148, 0, 0, 0, 0, 1,
 0.429, 1.480, -0.008, 0.061, 0, 0, 0, 0, 1,
 0.089, 0.785, -0.065, 0.018, 0, 0, 0, 0, 1,
 -0.516, 1.013, 0.016, 0.023, 0, 0, 0, 0, 1,
 0.143, 1.315, -0.011, 0.136, 0, 0, 0, 0, 1,
 0.347, 0.733, -0.121, 0.041, 0, 0, 0, 0, 1,
 -0.074, 0.869, 0.013, 0.026, 0, 0, 0, 0, 1,
 0.630, 1.484, -0.001, 0.000, 0, 0, 0, 0, 1),
  nrow=23, ncol=9, byrow=TRUE)
theta <- rnorm(NSubj)</pre>
data<-genFMPData(NSubj = NSubj, bParam = b, theta = theta, seed = 345)$data
erfItem1 <- erf(theta, data, whichItem = 1, min = -3, max = 3, Ncuts = 12)
plot( erfItem1$centers, erfItem1$probs, type="b",
     main="Empirical Response Function",
     xlab = expression(theta),
     ylab="Probability",
     cex.lab=1.5)
```

FMP

Estimate the coefficients of a filtered monotonic polynomial IRT model

Description

Estimate the coefficients of a filtered monotonic polynomial IRT model.

Usage

```
FMP(data, thetaInit, item, startvals, k, eps = 1e-06)
```

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Arguments

data N(subjects)-by-p(items) matrix of 0/1 item response data. thetaInit Initial theta (θ) surrogates (e.g., calculated by svdNorm).

item ltem number for coefficient estimation.

startvals Start values for function minimization. Start values are in the gamma metric

(see Liang & Browne, 2015)

k Order of monotonic polynomial = 2k+1 (see Liang & Browne, 2015). k can

equal 0, 1, 2, or 3.

eps Step size for gradient approximation, default = 1e-6. If a convergence failure

occurs during function optimization reducing the value of eps will often produce

a converged solution.

Details

As described by Liang and Browne (2015), the filtered polynomial model (FMP) is a quasi-parametric IRT model in which the IRF is a composition of a logistic function and a polynomial function, $m(\theta)$, of degree 2k+1. When k=0, $m(\theta)=b_0+b_1\theta$ (the slope intercept form of the 2PL). When k=1, 2k+1 equals 3 resulting in $m(\theta)=b_0+b_1\theta+b_2\theta^2+b_3\theta^3$. Acceptable values of k=0,1,2,3. According to Liang and Browne, the "FMP IRF may be used to approximate any IRF with a continuous derivative arbitrarily closely by increasing the number of parameters in the monotonic polynomial" (2015, p. 2) The FMP model assumes that the IRF is monotonically increasing, bounded by 0 and 1, and everywhere differentiable with respect to theta (the latent trait).

Value

b Vector of polynomial coefficients.

gamma Polynomial coefficients in gamma metric (see Liang & Browne, 2015).

FHAT Function value at convergence.

counts Number of function evaluations during minimization (see optim documentation

for further details).

AIC Pseudo scaled Akaike Information Criterion (AIC). Candidate models that pro-

duce the smallest AIC suggest the optimal number of parameters given the sample size. Scaling is accomplished by dividing the non-scaled AIC by sample

size.

BIC Pseudo scaled Bayesian Information Criterion (BIC). Candidate models that

produce the smallest BIC suggest the optimal number of parameters given the sample size. Scaling is accomplished by dividing the non-scaled BIC by sample

size.

convergence = 0 indicates that the optimization algorithm converged; conver-

gence=1 indicates that the optimization failed to converge.

Author(s)

Niels Waller

References

Liang, L. & Browne, M. W. (2015). A quasi-parametric method for fitting flexible item response functions. *Journal of Educational and Behavioral Statistics*, 40, 5–34.

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```
## Not run:
## In this example we will generate 2000 item response vectors
## for a k = 1 order filtered polynomial model and then recover
## the estimated item parameters with the FMP function.
k <- 1 # order of polynomial
NSubjects <- 2000
## generate a sample of 2000 item response vectors
## for a k = 1 FMP model using the following
## coefficients
b <- matrix(c(</pre>
  #b0
          b1
                  b2
                         b3 b4 b5 b6 b7 k
  1.675, 1.974, -0.068, 0.053, 0, 0, 0, 0, 1,
  1.550, 1.805, -0.230, 0.032, 0,
                                  0,
                                       0.
                                          0, 1,
                                      0, 0, 1,
  1.282, 1.063, -0.103, 0.003, 0, 0,
 0.704, 1.376, -0.107, 0.040, 0, 0,
                                       0, 0, 1,
 1.417, 1.413, 0.021, 0.000, 0, -0.008, 1.349, -0.195, 0.144, 0,
                                   0,
                                       0,
                                           0, 1,
                                   0,
                                       0.
                                           0, 1,
                                  0,
  0.512, 1.538, -0.089, 0.082, 0,
                                       0,
                                           0, 1,
 0.122, 0.601, -0.082, 0.119, 0, 0,
                                       0,
 1.801, 1.211, 0.015, 0.000, 0, 0,
                                       0,
                                           0, 1,
 -0.207, 1.191, 0.066, 0.033, 0, 0,
                                       0,
                                          0, 1,
 -0.215, 1.291, -0.087, 0.029, 0, 0, 0,
                                          0, 1,
 0.259, 0.875, 0.177, 0.072, 0, 0, 0, 0, 1,
 -0.423, 0.942, 0.064, 0.094, 0, 0, 0, 0, 1,
 0.113, 0.795, 0.124, 0.110, 0, 0, 0,
                                           0, 1,
  1.030, 1.525, 0.200, 0.076, 0, 0, 0,
                                           0, 1,
  0.140, 1.209, 0.082, 0.148, 0, 0, 0,
                                           0, 1,
  0.429, 1.480, -0.008, 0.061, 0, 0, 0,
 0.089, 0.785, -0.065, 0.018, 0, 0, 0, 0, 1,
 -0.516, 1.013, 0.016, 0.023, 0, 0, 0, 0, 1,
 0.143, 1.315, -0.011, 0.136, 0, 0, 0, 0, 1,
 0.347, 0.733, -0.121, 0.041, 0, 0, 0, 0, 1,
 -0.074, 0.869, 0.013, 0.026, 0, 0,
                                      0,
                                          0, 1,
 0.630, 1.484, -0.001, 0.000, 0, 0, 0,
                                          0, 1),
  nrow=23, ncol=9, byrow=TRUE)
ex1.data<-genFMPData(NSubj = NSubjects, bParams = b, seed = 345)$data
## number of items in the data matrix
NItems <- ncol(ex1.data)
# compute (initial) surrogate theta values from
# the normed left singular vector of the centered
# data matrix
thetaInit <- svdNorm(ex1.data)</pre>
## earlier we defined k = 1
  if(k == 0) {
           startVals <- c(1.5, 1.5)
```

```
bmat <- matrix(0, NItems, 6)</pre>
          colnames(bmat) <- c(paste("b", 0:1, sep = ""), "FHAT", "AIC", "BIC", "convergence")</pre>
  if(k == 1) {
            startVals <- c(1.5, 1.5, .10, .10)
            bmat <- matrix(0, NItems, 8)</pre>
         colnames(bmat) <- c(paste("b", 0:3, sep = ""), "FHAT", "AIC", "BIC", "convergence")</pre>
  }
  if(k == 2) {
            startVals <- c(1.5, 1.5, .10, .10, .10, .10)
            bmat <- matrix(0, NItems, 10)</pre>
         colnames(bmat) <- c(paste("b", 0:5, sep = ""), "FHAT", "AIC", "BIC", "convergence")</pre>
  if(k == 3) {
            startVals <- c(1.5, 1.5, .10, .10, .10, .10, .10, .10)
            bmat <- matrix(0, NItems, 12)</pre>
         colnames(bmat) <- c(paste("b", 0:7, sep = ""), "FHAT", "AIC", "BIC", "convergence")</pre>
  }
# estimate item parameters and fit statistics
  for(i in 1:NItems){
    out <- FMP(data = ex1.data, thetaInit, item = i, startvals = startVals, k = k)</pre>
    Nb <- length(out$b)
    bmat[i,1:Nb] <- out$b</pre>
    bmat[i,Nb+1] <- out$FHAT</pre>
    bmat[i,Nb+2] <- out$AIC</pre>
    bmat[i,Nb+3] <- out$BIC</pre>
    bmat[i,Nb+4] <- out$convergence</pre>
  }
# print output
print(bmat)
## End(Not run)
```

FMPMonotonicityCheck Utility function for checking FMP monotonicity

Description

Utility function for checking whether candidate FMP coefficients yield a monotonically increasing polynomial.

Usage

```
FMPMonotonicityCheck(b, lower = -20, upper = 20)
```

Arguments

```
b A vector of 8 polynomial coefficients (b) for m(\theta)=b_0+b_1\theta+b_2\theta^2+b_3\theta^3+b_4\theta^4+b_5\theta^5+b_6\theta^6+b_7\theta^7.
```

lower, upper Theta bounds for monotonicity check.

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Value

minDeriv Minimum value of the derivative for the polynomial.

Author(s)

Niels Waller

Examples

```
## A set of candidate coefficients for an FMP model.
## These coefficients fail the test and thus
## should not be used with genFMPdata to generate
## item response data that are consistent with an
## FMP model.
b <- c(1.21, 1.87, -1.02, 0.18, 0.18, 0, 0, 0)
FMPMonotonicityCheck(b)</pre>
```

FUP

Estimate the coefficients of a filtered unconstrained polynomial IRT model

Description

Estimate the coefficients of a filtered unconstrained polynomial IRT model.

Usage

```
FUP(data, thetaInit, item, startvals, k)
```

Arguments

data N(subjects)-by-p(items) matrix of 0/1 item response data. thetaInit Initial theta surrogates (e.g., calculated by svdNorm).

item item number for coefficient estimation.start values for function minimization.

k order of monotonic polynomial = 2k+1 (see Liang & Browne, 2015).

Value

b Vector of polynomial coefficients.FHAT Function value at convergence.

counts Number of function evaluations during minimization (see optim documentation

for further details).

AIC Pseudo scaled Akaike Information Criterion (AIC). Candidate models that pro-

duce the smallest AIC suggest the optimal number of parameters given the sample size. Scaling is accomplished by dividing the non-scaled AIC by sample

size.

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BIC Pseudo scaled Bayesian Information Criterion (BIC). Candidate models that

produce the smallest BIC suggest the optimal number of parameters given the sample size. Scaling is accomplished by dividing the non-scaled BIC by sample

size.

convergence = 0 indicates that the optimization algorithm converged; conver-

gence=1 indicates that the optimization failed to converge.

.

Author(s)

Niels Waller

References

Liang, L. & Browne, M. W. (2015). A quasi-parametric method for fitting flexible item response functions. *Journal of Educational and Behavioral Statistics*, 40, 5–34.

```
## Not run:
NSubjects <- 2000
## generate sample k=1 FMP data
b <- matrix(c(</pre>
    #b0
          b1
                 b2
                        b3
                                b4
                                    b5 b6 b7 k
  1.675, 1.974, -0.068, 0.053, 0, 0, 0, 0, 1,
  1.550, 1.805, -0.230, 0.032,
                               0,
                                   0,
                                       0,
                                           0, 1,
  1.282, 1.063, -0.103, 0.003,
                               0, 0,
                                       0,
                                           0, 1,
                                       0,
  0.704, 1.376, -0.107, 0.040, 0, 0,
                                           0, 1,
  1.417, 1.413, 0.021, 0.000, 0, 0,
                                       0,
                                           0, 1,
 -0.008, 1.349, -0.195, 0.144, 0, 0,
                                       0,
                                           0, 1,
  0.512, 1.538, -0.089, 0.082, 0, 0,
                                           0, 1,
  0.122, 0.601, -0.082, 0.119, 0, 0,
                                       0,
                                            0, 1,
 1.801, 1.211, 0.015, 0.000, 0, 0, 0,
 -0.207, 1.191, 0.066, 0.033, 0, 0, 0,
                                           0, 1,
 -0.215, 1.291, -0.087, 0.029, 0, 0, 0,
                                           0, 1,
 0.259, 0.875, 0.177, 0.072, 0, 0,
                                       0,
                                           0, 1,
 -0.423, 0.942, 0.064, 0.094, 0,
                                   0,
                                       0,
                                           0, 1,
   0.113, \ 0.795, \quad 0.124, \ 0.110, \quad 0, \\
                                   0,
                                       0.
                                           0, 1,
  1.030, 1.525, 0.200, 0.076, 0,
                                   0.
                                       0.
                                           0, 1,
  0.140, 1.209, 0.082, 0.148, 0,
                                   0,
                                       0,
                                           0, 1,
  0.429, 1.480, -0.008, 0.061,
                               0,
                                   0,
                                            0, 1,
  0.089, 0.785, -0.065, 0.018,
                               0,
                                    0,
                                       0,
                               0,
 -0.516, 1.013, 0.016, 0.023,
                                   0,
                                       0,
  0.143, 1.315, -0.011, 0.136, 0,
                                   0,
                                       0,
  0.347, 0.733, -0.121, 0.041, 0,
                                   0,
                                       0,
                                           0, 1,
 -0.074, 0.869, 0.013, 0.026, 0, 0,
                                       0,
                                           0, 1,
  0.630, 1.484, -0.001, 0.000,
                               0, 0,
                                       0,
  nrow=23, ncol=9, byrow=TRUE)
# generate data using the above item parameters
ex1.data<-genFMPData(NSubj = NSubjects, bParams = b, seed = 345)$data
NItems <- ncol(ex1.data)
```

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```
# compute (initial) surrogate theta values from
# the normed left singular vector of the centered
# data matrix
thetaInit <- svdNorm(ex1.data)</pre>
# Choose model
k <-1 \# order of polynomial = 2k+1
# Initialize matrices to hold output
if(k == 0) {
  startVals <- c(1.5, 1.5)
  bmat <- matrix(0,NItems,6)</pre>
  colnames(bmat) <- c(paste("b", 0:1, sep = ""), "FHAT", "AIC", "BIC", "convergence")</pre>
}
if(k == 1) {
  startVals <- c(1.5, 1.5, .10, .10)
  bmat <- matrix(0,NItems,8)</pre>
  colnames(bmat) <- c(paste("b", 0:3, sep = ""), "FHAT", "AIC", "BIC", "convergence")</pre>
}
if(k == 2) {
  startVals <- c(1.5, 1.5, .10, .10, .10, .10)
  bmat <- matrix(0,NItems,10)</pre>
  colnames(bmat) <- c(paste("b", 0:5, sep = ""),"FHAT", "AIC", "BIC", "convergence")</pre>
if(k == 3) {
  startVals <- c(1.5, 1.5, .10, .10, .10, .10, .10, .10)
  bmat <- matrix(0,NItems,12)</pre>
  colnames(bmat) <- c(paste("b", 0:7, sep = ""), "FHAT", "AIC", "BIC", "convergence")</pre>
# estimate item parameters and fit statistics
for(i in 1:NItems){
  out<-FUP(data = ex1.data,thetaInit = thetaInit, item = i, startvals = startVals, k = k)</pre>
  Nb <- length(out$b)
  bmat[i,1:Nb] <- out$b</pre>
  bmat[i,Nb+1] <- out$FHAT</pre>
  bmat[i,Nb+2] <- out$AIC</pre>
  bmat[i,Nb+3] <- out$BIC</pre>
  bmat[i,Nb+4] <- out$convergence</pre>
# print results
print(bmat)
## End(Not run)
```

gen4PMData

Generate item response data for 1, 2, 3, or 4-parameter IRT models

Description

Generate item response data for or 1, 2, 3 or 4-parameter IRT Models.

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Usage

Arguments

NSubj the desired number of subject response vectors.

abcdParams a p(items)-by-4 matrix of IRT item parameters: a = discrimination, b = difficulty,

c = lower asymptote, and d = upper asymptote.

D Scaling constant to place the IRF on the normal ogive or logistic metric. Default

= 1.702 (normal ogive metric)

seed Optional seed for the random number generator.

theta Optional vector of latent trait scores. If theta = NULL (the default value) then

gen4PMData will simulate theta from a normal distribution.

thetaMN Mean of simulated theta distribution. Default = 0.
thetaVar Variance of simulated theta distribution. Default = 1

Value

data N(subject)-by-p(items) matrix of item response data.

theta Latent trait scores.

seed Value of the random number seed.

Author(s)

Niels Waller

```
## Generate simulated 4PM data for 2,000 subjects
# 4PM Item parameters from MMPI-A CYN scale
```

```
Params<-matrix(c(1.41, -0.79, .01, .98, #1
                 1.19, -0.81, .02, .96, #2
                 0.79, -1.11, .05, .94, #3
                 0.94, -0.53, .02, .93, #4
                 0.90, -1.02, .04, .95, #5
                 1.00, -0.21, .02, .84, #6
                 1.05, -0.27, .02, .97, #7
                 0.90, -0.75, .04, .73, #8
                 0.80, -1.42, .06, .98, #9
                 0.71, 0.13, .05, .94, #10
                 1.01, -0.14, .02, .81, #11
                 0.63, 0.18, .18, .97, #12
                 0.68, 0.18, .02, .87, #13
                 0.60, -0.14, .09, .96, #14
                 0.85, -0.71, .04, .99, #15
                 0.83, -0.07, .05, .97, #16
                 0.86, -0.36, .03, .95, #17
                 0.66, -0.64, .04, .77, #18
                 0.60, 0.52, .04, .94, #19
```

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```
0.90, -0.06, .02, .96, #20
0.62, -0.47, .05, .86, #21
0.57, 0.13, .06, .93, #22
0.77, -0.43, .04, .97),23,4, byrow=TRUE)

data <- gen4PMData(NSubj=2000, abcdParams = Params, D = 1.702, seed = 123, thetaMN = 0, thetaVar = 1)$data

cat("\nClassical item difficulties for simulated data")
print( round( apply(data,2,mean),2) )</pre>
```

genFMPData

Generate item response data for a filtered monotonic polynomial IRT model

Description

Generate item response data for the filtered polynomial IRT model.

Usage

```
genFMPData(NSubj, bParams, theta = NULL, thetaMN = 0, thetaVar = 1, seed)
```

Arguments

NSubj	the desired number of subject response vectors.
bParams	a p(items)-by-9 matrix of polynomial coefficients and model designations. Columns 1 - 8 hold the polynomial coefficients; column 9 holds the value of k.
theta	A user-supplied vector of latent trait scores. Default theta = NULL.
thetaMN	If theta = NULL genFMPdata will simulate random normal deviates from a population with mean thetaMN and variance thetaVar.
thetaVar	If theta = NULL genFMPData will simulate random normal deviates from a population with mean thetaMN and variance thetaVar.
seed	initial seed for the random number generator.

Value

theta	theta values used for data generation
data	N(subject)-by-p(items) matrix of item response data.
seed	Value of the random number seed.

Author(s)

Niels Waller

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Examples

```
# The following code illustrates data generation for
# an FMP of order 3 (i.e., 2k+1)
# data will be generated for 2000 examinees
NSubjects <- 2000
## Example item paramters, k=1 FMP
b <- matrix(c(</pre>
   #b0
         b1
                       b3
                               b4
                                   b5 b6 b7 k
  1.675, 1.974, -0.068, 0.053, 0,
                                   0, 0,
  1.550, 1.805, -0.230, 0.032, 0,
                                   0,
                                       0,
                                          0, 1,
  1.282, 1.063, -0.103, 0.003, 0, 0,
                                       0,
                                          0, 1,
 0.704, 1.376, -0.107, 0.040, 0, 0,
                                      0,
                                          0, 1,
 1.417, 1.413, 0.021, 0.000, 0, 0,
                                      0,
                                          0, 1,
 -0.008, 1.349, -0.195, 0.144, 0, 0, 0,
                                          0, 1,
 0.512, 1.538, -0.089, 0.082, 0, 0, 0,
                                           0, 1,
 0.122, 0.601, -0.082, 0.119, 0, 0,
                                           0, 1,
 1.801, 1.211, 0.015, 0.000, 0, 0, 0,
                                          0, 1,
 -0.207, 1.191, 0.066, 0.033, 0, 0, 0, 0, 1,
 -0.215, 1.291, -0.087, 0.029, 0, 0, 0, 0, 1,
 0.259, 0.875, 0.177, 0.072, 0, 0, 0, 0, 1,
 -0.423, 0.942, 0.064, 0.094, 0, 0, 0, 0, 1,
 0.113, 0.795, 0.124, 0.110, 0, 0, 0, 0, 1,
 1.030, 1.525, 0.200, 0.076, 0, 0, 0, 0, 1,
 0.140, 1.209, 0.082, 0.148, 0, 0, 0, 0, 1,
                                  0, 0, 0, 1,
 0.429, 1.480, -0.008, 0.061, 0,
 0.089, 0.785, -0.065, 0.018, 0,
                                          0, 1,
                                  0,
                                      0,
 -0.516, 1.013, 0.016, 0.023, 0.143, 1.315, -0.011, 0.136,
                              0,
                                   0,
                                       0,
                                           0, 1,
                               0,
                                   0,
                                       0,
                                           0, 1,
                                      0,
 0.347, 0.733, -0.121, 0.041, 0,
                                  0,
                                           0, 1,
 -0.074, 0.869, 0.013, 0.026, 0, 0,
                                      0,
                                          0, 1,
 0.630, 1.484, -0.001, 0.000, 0, 0, 0, 0, 1),
 nrow=23, ncol=9, byrow=TRUE)
# generate data using the above item paramters
data<-genFMPData(NSubj = NSubjects, bParams=b, seed=345)$data</pre>
```

Plot item response functions for polynomial IRT models.

irf

Description

Plot model-implied (and possibly empirical) item response function for polynomial IRT models.

Usage

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Arguments

data N(subjects)-by-p(items) matrix of 0/1 item response data.

bParams p(items)-by-9 matrix. The first 8 columns of the matrix should contain the FMP

or FUP polynomial coefficients for the p items. The 9th column contains the value of k for each item (where the item specific order of the polynomial is

2k+1).

item The IRF for item will be plotted.

plotERF A logical that determines whether to plot discrete values of the empirical re-

sponse function.

thetaEAP If plotERF=TRUE, the user must supply previously calculated eap trait estimates

to thetaEAP.

minCut, maxCut If plotERF=TRUE, the program will (attempt to) plot NCuts points of the em-

pirical response function between trait values of minCut and maxCut Default

minCut = -3. Default maxCut = 3.

NCuts Desired number of bins for the empirical response function.

Author(s)

Niels Waller

NSubjects <- 2000

Examples

```
NItems <- 15
itmParameters <- matrix(c(</pre>
# b0 b1
              b2 b3
                           b4 b5,
                                   b6, b7, k
 -1.05, 1.63, 0.00, 0.00, 0.00, 0,
                                    0, 0,
                                               0, #1
                                     0, 0,
 -1.97, 1.75, 0.00, 0.00, 0.00, 0,
                                               0, #2
                                     0, 0,
 -1.77, 1.82, 0.00, 0.00, 0.00, 0,
                                               0, #3
 -4.76, 2.67, 0.00, 0.00, 0.00, 0,
                                     0, 0,
                                               0, #4
 -2.15, 1.93, 0.00, 0.00, 0.00, 0,
                                     0, 0,
                                               0, #5
 -1.25, 1.17, -0.25, 0.12, 0.00, 0,
                                     0, 0, 1, #6
 1.65, 0.01, 0.02, 0.03, 0.00, 0,
                                      0, 0,
                                              1, #7
 -2.99, 1.64, 0.17, 0.03, 0.00, 0,
                                      0, 0,
                                              1, #8
 -3.22, 2.40, -0.12, 0.10, 0.00, 0,
                                      0, 0,
                                               1, #9
 -0.75, 1.09, -0.39, 0.31, 0.00, 0,
                                      0, 0,
                                               1, #10
 -1.21, 9.07, 1.20,-0.01,-0.01, 0.01, 0, 0,
                                               2, #11
 -1.92, 1.55, -0.17, 0.50, -0.01, 0.01, 0, 0,
                                               2, #12
 -1.76, 1.29, -0.13, 1.60, -0.01, 0.01, 0, 0,
                                               2, #13
 -2.32, 1.40, 0.55, 0.05, -0.01, 0.01, 0, 0,
                                               2, #14
 -1.24, 2.48, -0.65, 0.60, -0.01, 0.01, 0, 0,
                                               2),#15
 15, 9, byrow=TRUE)
ex1.data<-genFMPData(NSubj = NSubjects, bParams = itmParameters,</pre>
                   seed = 345)$data
## compute initial theta surrogates
thetaInit <- svdNorm(ex1.data)</pre>
## For convenience we assume that the item parameter
```

estimates equal their population values. In practice,

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```
## item parameters would be estimated at this step.
itmEstimates <- itmParameters</pre>
## calculate eap estimates for mixed models
thetaEAP <- eap(data = ex1.data, bParams = itmEstimates, NQuad = 21,
                priorVar = 2,
                mintheta = -4, maxtheta = 4)
## plot irf and erf for item 1
irf(data = ex1.data, bParams = itmEstimates,
    item = 1,
    plotERF = TRUE,
    thetaEAP)
## plot irf and erf for item 12
irf(data = ex1.data, bParams = itmEstimates,
    item = 12,
    plotERF = TRUE,
    thetaEAP)
```

restScore

Plot an ERF using rest scores

Description

Plot an empirical response function using rest scores.

Usage

```
restScore(data, item, NCuts)
```

Arguments

data N(subjects)-by-p(items) matrix of 0/1 item response data.

item Generate a rest score plot for item item.

NCuts Divide the rest scores into NCuts bins of equal width.

Value

A restscore plot with 95% confidence interval bars for the conditional probability estimates.

item The item number.

bins A vector of bin limits and bin sample sizes.
binProb A vector of bin conditional probabilities.

Author(s)

Niels Waller

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Examples

```
NSubj <- 2000
#generate sample k=1 FMP data
b <- matrix(c(</pre>
    #b0
          b1
                 b2
                       b3
                               b4 b5 b6 b7 k
  1.675, 1.974, -0.068, 0.053, 0, 0, 0, 0, 1,
  1.550, 1.805, -0.230, 0.032, 0, 0,
                                       0,
  1.282, 1.063, -0.103, 0.003, 0, 0, 0, 0, 1,
  0.704, 1.376, -0.107, 0.040, 0, 0, 0, 0, 1,
 1.417, 1.413, 0.021, 0.000, 0, 0, 0, 0, 1,
 -0.008, 1.349, -0.195, 0.144, 0, 0, 0, 0, 1,
 0.512, 1.538, -0.089, 0.082, 0, 0, 0, 0, 1,
 0.122, 0.601, -0.082, 0.119, 0, 0, 0, 0, 1,
 1.801, 1.211, 0.015, 0.000, 0, 0, 0, 0, 1,
 -0.207, 1.191, 0.066, 0.033, 0, 0, 0, 0, 1,
 -0.215, 1.291, -0.087, 0.029, 0, 0, 0, 1,
 0.259, 0.875, 0.177, 0.072, 0, 0, 0, 0, 1,
 -0.423, 0.942, 0.064, 0.094, 0, 0, 0, 0, 1,
 0.113, 0.795, 0.124, 0.110, 0, 0, 0, 0, 1,
  1.030, 1.525, 0.200, 0.076, 0, 0, 0, 0, 1,
  0.140, 1.209, 0.082, 0.148, 0, 0, 0, 0, 1,
   0.429, \ 1.480, \ -0.008, \ 0.061, \quad 0, \quad 0, \quad 0, \quad 0, \quad 1, \\
 0.089, 0.785, -0.065, 0.018, 0, 0,
                                       0, 0, 1,
 -0.516, 1.013, 0.016, 0.023, 0.143, 1.315, -0.011, 0.136,
                               0,
                                   0,
                                           0, 1,
                               0,
                                   0,
  0.347, 0.733, -0.121, 0.041, 0, 0,
                                       0,
                                           0, 1,
 -0.074, 0.869, 0.013, 0.026, 0, 0,
                                       0,
                                           0, 1,
 0.630, 1.484, -0.001, 0.000, 0, 0, 0,
  nrow=23, ncol=9, byrow=TRUE)
data<-genFMPData(NSubj = NSubj, bParam = b, seed = 345)$data</pre>
## generate a rest score plot for item 12.
## the grey horizontal lines in the plot
## respresent pseudo asymptotes that
## are significantly different from the
## (0,1) boundaries
restScore(data, item = 12, NCuts = 9)
```

svdNorm

Compute theta surrogates via normalized SVD scores

Description

Compute theta surrogates by calculating the normalized left singular vector of a (mean-centered) data matrix.

Usage

```
svdNorm(data)
```

Arguments

data

N(subjects)-by-p(items) matrix of 0/1 item response data.

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Value

the normalized left singular vector of the mean centered data matrix. svdNorm will center the data automatically.

Author(s)

Niels Waller

```
NSubj <- 2000
```

```
## example item parameters for sample data: k=1 FMP
b <- matrix(c(</pre>
   #b0
                 b2
                       b3
                                   b5 b6 b7 k
          b1
                               b4
  1.675, 1.974, -0.068, 0.053, 0, 0, 0, 0, 1,
  1.550, 1.805, -0.230, 0.032, 0, 0, 0, 0, 1,
 1.282, 1.063, -0.103, 0.003, 0, 0,
 0.704, 1.376, -0.107, 0.040, 0, 0, 0, 0, 1,
 1.417, 1.413, 0.021, 0.000, 0, 0, 0, 0, 1,
 -0.008, 1.349, -0.195, 0.144, 0, 0, 0, 0, 1,
 0.512, 1.538, -0.089, 0.082, 0, 0, 0, 0, 1,
 0.122, 0.601, -0.082, 0.119, 0, 0, 0, 0, 1,
 1.801, 1.211, 0.015, 0.000, 0, 0, 0, 0, 1,
 -0.207, 1.191, 0.066, 0.033, 0, 0, 0, 0, 1,
 -0.215, 1.291, -0.087, 0.029, 0,
                                          0, 1,
                                   0,
                                      0,
 0.259, 0.875, 0.177, 0.072, 0,
                                           0, 1,
                                   0,
                                      0,
 -0.423, 0.942, 0.064, 0.094, 0.113, 0.795, 0.124, 0.110,
                               0,
                                   0,
                                       0,
                                           0, 1,
                               0,
                                   0,
                                       0,
                                           0, 1,
 1.030, 1.525, 0.200, 0.076,
                               0,
                                   0,
                                       0,
 0.140, 1.209, 0.082, 0.148, 0,
                                  0,
                                       0,
 0.429, 1.480, -0.008, 0.061, 0, 0,
                                       0,
                                           0, 1,
 0.089, 0.785, -0.065, 0.018, 0, 0,
                                       0,
                                          0, 1,
 -0.516, 1.013, 0.016, 0.023, 0, 0,
                                      0,
                                          0, 1,
 0.143, 1.315, -0.011, 0.136, 0, 0, 0, 0, 1,
 0.347, 0.733, -0.121, 0.041, 0, 0, 0, 0, 1,
 -0.074, 0.869, 0.013, 0.026, 0, 0, 0, 0, 1,
 0.630, 1.484, -0.001, 0.000, 0, 0, 0, 0, 1),
  nrow=23, ncol=9, byrow=TRUE)
# generate data using the above item paramters
data<-genFMPData(NSubj=NSubj, bParam=b, seed=345)$data</pre>
# compute (initial) surrogate theta values from
# the normed left singular vector of the centered
# data matrix
thetaInit<-svdNorm(data)</pre>
```

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