

2PL Demo

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Understanding MIRT

2PL Model Simulate data using Monte Carlo

```
# person trait level, e.g., math skill
theta <- rnorm(n = 1000, mean = 0, sd = 1)

# Item parameters (a, b, c) for 4 items
a <- c(0.5, 1, 1.5, 2)
b <- c(0, -1, 1, 0)
c <- c(0.2, 0.15, 0.25, 0.2)

n.persons <- length(theta)
n.items <- length(a)
response.data <- matrix(NA, n.persons, n.items)

for (i in 1:n.persons) {
  for (j in 1:n.items) {
    p <- c[j] + (1 - c[j]) / (1 + exp(-(a[j] * theta[i] - b[j]))) #2PL model
    u <- runif(n = 1, min = 0, max = 1)
    if (u < p) {
      response.data[i, j] <- 1
    } else {
      response.data[i, j] <- 0
    }
  }
}
colnames(response.data) <- c("I1", "I2", "I3", "I4")
```

- a, b, c : item parameters.
- θ , `theta`: personal trait level.
- p : probability of positive response (gets answer right)
- u : random number between 0 and 1, (uniformly distributed)
- if $u < p$ then `response.data` = 1: individual gets answer correct.
- if $u > p$ then `response.data` = 0: individual gets answer wrong.

```
start_time = Sys.time() #start time
mirt.out <- mirt::mirt(data = response.data, model = 1, itemtype = "2PL", storeEMhistory=TRUE) # 2PL mo
```

```
## Iteration: 1, Log-Lik: -2596.883, Max-Change: 0.07289Iteration: 2, Log-Lik: -2595.099, Max-Change: 0
```

```
end_time = Sys.time() # end time
time_to_conv = end_time - start_time
time_to_conv
```

Time difference of 0.1188409 secs

- `mirt.out <- mirt::mirt()` function fits a maximum likelihood (posterior) factor analysis model to the data (dichotomus/polytomous)

Output of `mirt.out`:

- ‘M-step’ = indicates that the Broyden-Fletcher-Goldfarb-Shanno (BFGS) algorithm used for optimization in M-step if the EM algorithm.
- ‘EM acceleration’ = indicates that ramsay’s acceleration method is used to speed up the convergence of the EM algorithm.
- ‘Number of rectangular quadrature’ = the number of quadrature points used in the numerical integration of the likelihood function.
- ‘Latent density type’ = the distribution of the latent trait (F1 in this case)
- ‘log-likelihood’ = how well the model fits the data
- ‘Estimated parameters’ = number of estimated parameters
- ‘AIC’ = Akaike information criterion (measure of model fit, takes into number of parameters)
- BIC/SABIC = Bayesian information criterion/Sample-size Adjusted BIC (measure of model fit, takes into number of parameters)
- $G2(3) = 1.1, p = 0.7772$: results of the likelihood ratio test comparing the fitted model against saturated model
- $RMSEA = 0, CFI = NaN, TLI = NaN$: additional measures of model fit, the Root Mean Square Error of Approximation (RMSEA), the Comparative Fit Index (CFI), and the Tucker-Lewis Index (TLI)