

TEST FOR TREND WITH A MULTINOMIAL OUTCOME

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1. INTRODUCTION

Consider a study in which a multinomial outcome with K possible unordered values is measured in subjects belonging to one of G ordered groups. The size of each group, n_i , is defined by the study design, and will be treated as fixed. Let $\mathbf{p}_i = (p_{i1}, \dots, p_{iK})^\top$ denote the probabilities of the multinomial outcomes in the i th group. The hypothesis of interest is to evaluate the homogeneity of these probabilities across the groups with a targeted alternative of a trend in at least one of the categories. Formally, we consider testing $H_0 = \bigcap_{j=1}^K H_{0j}$ versus $H_1 = \bigcup_{j=1}^K H_{1j}$, where

$$\begin{aligned} H_{0j} &: p_{1j} = \dots = p_{Gj} \\ H_{1j} &: p_{1j} \leq \dots \leq p_{Gj} \text{ or } p_{1j} \geq \dots \geq p_{Gj} \text{ with at least one inequality} \end{aligned} \tag{1}$$

The test is based on the following result:

Theorem 1. *Let $\mathcal{J} \subset \{1, \dots, K\}$, then under $H_{0\mathcal{J}} = \bigcap_{j \in \mathcal{J}} H_{0j}$ as $N \rightarrow \infty$*

$$W_{\mathcal{J}} = \sum_{j \in \mathcal{J}} (1 - p_{\cdot j}) T_j^2 + \left(\sum_{j \in \mathcal{J}} p_{\cdot j} \right) T_{\mathcal{J}}^2 \xrightarrow{d} \chi_d^2, \tag{2}$$

where $d = \min(|\mathcal{J}|, K - 1)$, $T_{\mathcal{J}} = [\sum_{i=1}^G \sum_{j \in \mathcal{J}} n_{ij}(c_i - \bar{c})] / \sqrt{p_{\cdot \mathcal{J}}(1 - p_{\cdot \mathcal{J}})s^2}$ denotes the Cochran-Armitage trend test statistic for testing for marginal trend in $p_{i\mathcal{J}} = \sum_{j \in \mathcal{J}} p_{ij}$, $i = 1, \dots, G$.

2. IMPLEMENTING THE OVERALL TEST

The main `multiCA.test` function is a generic, with methods for a matrix and formula input.

```
"../R/aaa-generics.R" 2a≡
```

```
#'Multinomial Cochran-Armitage trend test
#
#'\code{multiCA.test} performs a multinomial generalization of the
#'\code{Cochran-Armitage} trend test.
#
#
#'\@export
#'\@param x a two-dimensional matrix of event counts with the outcomes as rows and ordered groups as columns
#'\@param \dots other arguments
#'\@return a list with two components
#'\item{overall}{an object of class "htest" with the results of the overall test}
#'\item{individual}{a vector with adjusted p-values for individual outcomes}
#'\@author Aniko Szabo
#'\@references Szabo, A. (2016) Test for trend with a multinomial outcome.
#'\@keywords nonparametric
#'\@examples
#
#'\data(stroke)
#'\## using formula interface
#'\multiCA.test(Type ~ Year, weights=Freq, data=stroke)
#
#'\## using matrix interface and testing only the first 3 outcomes
#'\strk.mat <- xtabs(Freq ~ Type + Year, data=stroke)
#'\multiCA.test(strk.mat, outcomes=1:3)
#
#'\@name multiCA.test

multiCA.test <- function(x,...) UseMethod("multiCA.test")

◇
```

The actual calculation of the test statistic, overall and unadjusted individual p-values is encapsulated in an internal function that operates on a matrix. No error control is provided here.

```
"../R/multiCA.R" 2b≡
```

```
#'\@keywords internal

.multiCA.test <- function(x, scores, outcomes){
  K <- nrow(x)
  full <- length(outcomes) == K #full test

  nidot <- apply(x, 2, sum)
  n <- sum(nidot)

  cbar <- sum(nidot * scores)/n

  s2 <- sum(nidot * (scores - cbar)^2)
  pdot <- prop.table(rowSums(x))[outcomes]
  nonz <- (pdot > 0)

  if (!any(nonz)) return(1)

  X <- x[outcomes, ,drop=FALSE] %*% (scores - cbar)

  #individual tests
  CAT <- X[nonz]^2 / (pdot[nonz] * (1-pdot[nonz])) / s2
```

```

CAT.p.value <- pchisq(CAT, df=1, lower.tail=FALSE)

#overall test
if (full || sum(pdot) >= 1){
  Tt <- ( sum(X[nonz]^2 / pdot[nonz])) / s2
} else {
  Tt <- (sum(X)^2 / (1-sum(pdot)) + sum(X[nonz]^2 / pdot[nonz])) / s2
}

df <- length(outcomes) - full
p.value <- pchisq(Tt, df=df, lower.tail=FALSE)

res <- list(statistic = Tt, parameter = df, p.value = p.value,
            indiv.statistics = CAT, indiv.p.value = CAT.p.value)
return(res)
}
◇

```

File defined by [2b](#), [3](#), [4a](#), [6c](#), [8a](#), [9](#).

Defines: [.multiCA.test](#) [3](#), [6a](#), [9](#), [10a](#), [12ac](#).

The default method uses a two-dimensional contingency matrix with the outcomes as rows and ordered groups as columns.

"../R/multiCA.R" 3≡

```

#'@rdname multiCA.test
#'@method multiCA.test default
#'@param scores non-decreasing numeric vector of the same length as the number of ordered groups. Default
#'@param outcomes integer or character vector defining the set of outcomes (by row index or row name) over
#'@param p.adjust.method character string defining the correction method for individual outcome p-values.
#'@export

multiCA.test.default <- function(x, scores=1:ncol(x), outcomes=1:nrow(x),
  p.adjust.method=c("none","closed.set","holm-Shaffer"),...){
  if (!is.matrix(x)) {
    cat(str(x))
    stop("x should be a two-dimensional matrix")
  }
  if (length(scores) != ncol(x)) stop("The length of the score vector should equal the number of columns")

  testres <- .multiCA.test(x=x, scores=scores, outcomes=outcomes)

  Tt <- c(W = testres$statistic)
  df <- c(df = testres$parameter)

  p.value <- testres$p.value
  null.value <- 0
  names(null.value) <- sprintf("slope for outcomes %s", deparse(substitute(outcomes)))

  res <- list(statistic = Tt, parameter = df, p.value = p.value,
             method="Multinomial Cochran-Armitage trend test",
             alternative="two.sided",
             null.value=null.value,
             data.name = deparse(substitute(x)))
  class(res) <- "htest"

  ⟨ Calculate adjusted p-values 5a ⟩

```

```

    return(list(overall = res, individual = indiv.res))
  }
  ◇

```

File defined by 2b, 3, 4a, 6c, 8a, 9.

Defines: multiCA.test.default Never used.

Uses: .multiCA.test 2b.

The formula interface converts data into the appropriate contingency matrix for use with the default method.

The code is based on t.test.formula.

"../R/multiCA.R" 4a≡

```

#'@rdname multiCA.test
#'@method multiCA.test formula
#'@param formula a formula of the form \code{outcome ~ group} where \code{outcome} is a factor representing
#'@param data an optional matrix or data frame containing the variables in the formula \code{formula}. B
#'@param subset an optional vector specifying a subset of observations to be used.
#'@param na.action a function which indicates what should happen when the data contain NAs. Default
#'@param weights an integer-valued variable representing the number of times each \code{outcome} - \code{group}
#'@export

```

```

multiCA.test.formula <- function(formula, data, subset, na.action, weights, ...){
  if (missing(formula) || (length(formula) != 3L) || (length(attr(terms(formula)[-2L]),
    "term.labels")) != 1L))
    stop("'formula' missing or incorrect")
  m <- match.call(expand.dots = FALSE)
  if (is.matrix(eval(m$data, parent.frame()))){
    m$data <- as.data.frame(data)
  }
  m[[1L]] <- quote(stats::model.frame)
  m$... <- NULL
  mf <- eval(m, parent.frame())
  responsevar <- attr(attr(mf, "terms"), "response")
  response <- mf[[responsevar]]
  weightvar <- which(names(mf)=="(weights)")
  w <- if(length(weightvar) > 0) mf[[weightvar]] else rep(1L, nrow(mf))
  g <- factor(mf[, -c(responsevar, weightvar)])

  tab <- xtabs(w ~ response + g)
  multiCA.test(tab, ...)
}
  ◇

```

File defined by 2b, 3, 4a, 6c, 8a, 9.

Defines: multiCA.test.formula Never used.

"../tests/testthat/test_overall.R" 4b≡

```

context("Multinomial CA test")
test_that("Overall test works on stroke data", {
  data(stroke)
  res0 <- multiCA.test(Type ~ Year, weights=Freq, data=stroke, p.adjust="none")
  expect_equivalent(res0$overall$statistic, 40.06580869)

  strk.mat <- xtabs(Freq ~ Type + Year, data=stroke)
  res1 <- multiCA.test(strk.mat, p.adjust="none")
  expect_equal(res0$overall[c("statistic", "parameter", "p.value")],
    res1$overall[c("statistic", "parameter", "p.value")])
  expect_equivalent(res0$individual, res1$individual)
}

```

```

res2 <- multiCA.test(strk.mat, outcomes=1:5, p.adjust="none")
expect_equal(res1$overall[c("statistic","paramater", "p.value")],
             res2$overall[c("statistic","paramater", "p.value")])
expect_equivalent(res1$individual, res2$individual)
})

```

File defined by [4b](#), [5c](#), [6b](#).

3. MULTIPLE TESTING ADJUSTED INFERENCE FOR INDIVIDUAL OUTCOMES

⟨ Calculate adjusted p-values 5a ⟩ \equiv

```

if (missing(p.adjust.method)){
  if (length(outcomes)<=3) p.adjust.method <- "closed.set"
  else p.adjust.method <- "holm-Shaffer"
} else {
  p.adjust.method <- match.arg(p.adjust.method)
}

full.set <- (length(outcomes) == nrow(x))
if (p.adjust.method=="none") {
  indiv.res <- testres$indiv.p.value
} else if (p.adjust.method=="closed.set") {
  ⟨ Closed set adjustment 6a ⟩
} else if (p.adjust.method=="holm-Shaffer") {
  ⟨ Holm-Shaffer adjustment 5b ⟩
}
attr(indiv.res, "method") <- p.adjust.method

```

Fragment referenced in [3](#).

3.1. Holm-Shaffer approach. Shaffer's modification of Holm's adjustment involves multiplying the ordered p-values by t_s , the maximum number of possibly true hypotheses, given that at least $s - 1$ hypotheses are false. In our case the logical restriction means that if there is at least one false null hypothesis, then no more than $K - 2$ null hypotheses could be true. So

$$p_{(j)}^{HS} = \max_{s \leq j} (\min(t_s p_{(s)}, 1))$$

$$\text{where } t_s = \begin{cases} K - s + 1, & s \neq 2 \\ K - 2, & s = 2 \end{cases}$$

⟨ Holm-Shaffer adjustment 5b ⟩ \equiv

```

s <- seq_along(testres$indiv.p.value)
if (full.set) s[2] <- 3
o <- order(testres$indiv.p.value)
ro <- order(o)
indiv.res <- pmin(1, cummax((length(outcomes) - s + 1L) * testres$indiv.p.value[o]))[ro]

```

Fragment referenced in [5a](#).

"../tests/testthat/test_overall.R" 5c \equiv

```

test_that("Holm-Shaffer consistent with Holm", {
  data(stroke)
  res0 <- multiCA.test(Type ~ Year, weights=Freq, data=stroke, p.adjust="none")
  expect_equal(attr(res0$individual, "method"), "none")

  res1 <- multiCA.test(Type ~ Year, weights=Freq, data=stroke, p.adjust="holm-Shaffer")
  expect_equal(attr(res1$individual, "method"), "holm-Shaffer")
  expect_equivalent(sort(p.adjust(res0$individual, method="holm"))[-2],
    sort(res1$individual)[-2])

  res0a <- multiCA.test(Type ~ Year, weights=Freq, data=stroke, p.adjust="none",
    outcomes=1:4)
  res1a <- multiCA.test(Type ~ Year, weights=Freq, data=stroke, p.adjust="holm-Shaffer",
    outcomes=1:4)
  expect_equivalent(sort(p.adjust(res0a$individual, method="holm")),
    sort(res1a$individual))
})

```

File defined by [4b](#), [5c](#), [6b](#).

3.2. Closed set adjustment. In a closed testing procedure an elementary hypothesis H_{0j} is rejected if and only if all composite hypotheses $H_{0\mathcal{J}}$, where $j \in \mathcal{J}$ are rejected. The process can be rewritten using adjusted p-values for $H_{0j}, j = 1, \dots, K$:

$$p_j^* = \max_{\mathcal{J}: j \in \mathcal{J}} p(\mathcal{J}), \quad (3)$$

where $p(\mathcal{J}) = P(W_j \geq \chi^2_{|\mathcal{J}|})$ is the unadjusted p-value for testing $H_{0\mathcal{J}}$. From the logical constraints sets \mathcal{J} of cardinality $K - 1$ do not need to be considered.

$\langle \text{Closed set adjustment 6a} \rangle \equiv$

```

mytest <- function(hypotheses){
  .multiCA.test(x, scores, hypotheses)$p.value
}
indiv.res <- .p.adjust.closed(mytest, outcomes, remove=full.set)

```

Fragment referenced in [5a](#).

Uses: [.multiCA.test](#) [2b](#).

"../tests/testthat/test_overall.R" 6b≡

```

test_that("Closed set works with 3 outcomes", {
  data(stroke)
  strk.mat <- xtabs(Freq ~ Type + Year, data=stroke)
  res0 <- multiCA.test(strk.mat[1:3,], p.adjust="none")
  res1 <- multiCA.test(strk.mat[1:3,], p.adjust="closed.set")
  expect_equivalent(pmax(res0$individual, res0$overall$p.value),
    res1$individual)
})

```

File defined by [4b](#), [5c](#), [6b](#).

The actual adjustment calculation is based on code from `cherry::closed`, removing the $K - 1$ element sets if the full set of hypotheses is being tested.

"../R/multiCA.R" 6c≡

```

#' @importFrom bitops bitAnd
#' @keywords internal
.bit2boolean <- function (x, N)
{
  base <- 2^(1:N - 1)
  bitAnd(x, base) != 0
}

#' @param test function that performs the local test. The function should accept a subvector of the hypotheses
#' @param hypotheses identifiers of the collection of elementary hypotheses.
#' @param remove logical indicator of whether hypotheses of length N-1 should be removed
#' @param ... additional parameters to the 'test' function
#' @return numeric vector of adjusted p-values for each hypothesis
#' @keywords internal
.p.adjust.closed <- function (test, hypotheses, remove=FALSE, ...)
{
  N <- length(hypotheses)
  Nmax <- log2(.Machine$integer.max + 1)
  if (N > Nmax)
    stop("no more than ", Nmax, " hypotheses supported in full closed testing.\n Use a shortcut-based test")
  closure <- 1:(2^N - 1)
  base <- 2^(1:N - 1)
  offspring <- function(x) {
    res <- bitAnd(x, closure)
    res[res != 0]
  }
  lengths <- rowSums(sapply(base, function(bs) bitAnd(closure, bs) != 0))

  idx <- sort.list(lengths, decreasing = TRUE)
  closure <- closure[idx]
  lengths <- lengths[idx]
  if (remove) closure <- closure[lengths != (N-1)]

  adjusted <- numeric(2^N - 1)
  for (i in closure) {
    if (adjusted[i] < 1) {
      localtest <- test(hypotheses[.bit2boolean(i,N)], ...)
      if (localtest > adjusted[i]) {
        offs <- offspring(i)
        adjusted[offs] <- pmax(adjusted[offs], localtest)
      }
    }
  }

  out <- adjusted[base]
  names(out) <- hypotheses
  return(out)
}

```

File defined by [2b](#), [3](#), [4a](#), [6c](#), [8a](#), [9](#).

4. POWER AND SAMPLE SIZE CALCULATION

The calculation is based on the following result: Let $\nu_i = n_{i\cdot}/N$ denote the proportion of subjects in group i .

Theorem 2. Under H_a , the asymptotic distribution of W is approximately $\chi^2_{K-1}(\lambda)$ with non-centrality parameter

$$\lambda = N s_\nu^2 \sum_{j=1}^K \frac{\beta_j^2}{p_{\cdot j}}, \quad (4)$$

where $s_\nu^2 = \sum_{i=1}^G \nu_i (c_i - \bar{c})^2 = s^2/N$ and $\beta_j = [\sum_{i=1}^G \nu_i (p_{ij} - p_{\cdot j})(c_i - \bar{c})] / s_\nu^2$ is the slope of p_{ij} , $i = 1, \dots, G$ regressed on c_i with weights ν_i .

A non-centrality parameter calculation function can be useful by itself. It calculates the non-centrality parameter for a chi-square distribution that achieves the target power at a given significance level.

"../R/multiCA.R" 8a≡

```
#' Non-centrality parameter for chi-square distribution
#
#' Calculates the non-centrality parameter for a chi-square distribution for a given
#' quantile. This is often needed for sample size calculation for chi-square based tests.
#
#'@details The function is modeled after the SAS function CNONCT. If \code{p} is larger
#' than the cumulative probability of the central chi-square distribution at \code{x}, then
#' there is no solution and NA is returned.
#
#'@param x a numeric value at which the distribution was evaluated
#'@param p a numeric value giving the cumulative probability at \code{x}
#'@param df an integer giving the degrees of freedom of the chi-square variable
#'@examples
#' (ncp <- cnonct(qchisq(0.95, df=10), 0.8, df=10))
#' ## check
#' pchisq(qchisq(0.95, df=10), df=10, ncp=ncp) ## 0.8
#'@export

cnonct <- function(x, p, df){

  if (pchisq(x, df=df) < p) return(NA)

  f <- function(ncp){pchisq(x, df=df, ncp=pmax(0,ncp)) - p}

  res <- uniroot(f, interval=c(0, 100), extendInt="downX", tol=.Machine$double.eps^0.5)
  res$root
}
◇
```

File defined by [2b](#), [3](#), [4a](#), [6c](#), [8a](#), [9](#).

Defines: [cnonct](#) [8b](#), [9](#).

"../tests/testthat/test_power.R" 8b≡

```
context("Power calculations")

test_that("non-centrality calculation works", {
  x <- qchisq(0.75, df=10)
  expect_equal(cnonct(x, df=10, p=0.75), 0)
  expect_equal(cnonct(x, df=10, p=0.9), NA)
  expect_equal(pchisq(x, df=10, ncp=cnonct(x, p=0.6, df=10)), 0.6)
})
◇
```

File defined by [8b](#), [10a](#), [12ac](#).

Uses: [cnonct](#) [8a](#).


```
"../R/multiCA.R" 9≡
```

```
#' Power calculations for the multinomial Cochran-Armitage trend test
#'
#' Given the probabilities of outcomes, compute the power of the overall multinomial
#' Cochran-Armitage trend test or determine the sample size to obtain a target power.
#'
#'@details
#' The distribution of the outcomes can be specified in two ways: either the full matrix of #' outcome pr
#'
#' @param N integer, the total sample size of the study. If \code{NULL} then \code{power} needs to be spe
#' @param power target power. If \code{NULL} then \code{N} needs to be specified.
#' @param pmatrix numeric matrix of hypothesized outcome probabilities in each group, with #' the outcom
#' @param p.ave numeric vector of average probability of each outcome over the groups
#' weighted by \code{n.prop}.
#' @param p.start,p.end numeric vectors of the probability of each outcome for the
#' first / last ordered group
#' @param slopes numeric vector of the hypothesized slope of each outcome when regressed
#' against the column \code{scores} wiht weights \code{n.prop}
#' @param scores non-decreasing numeric vector of the same length as the number of ordered groups
#' giving the trend test scores. Defaults to linearly increasing values.
#' @param n.prop numeric vector describing relative sample sizes of the ordered groups.
#' Will be normalized to sum to 1. Defaults to equal sample sizes.
#' @param G integer, number of ordered groups
#' @param sig.level significance level
#' @return object of class "power.htest"
#'
#' @examples
#' power.multiCA.test(power=0.8, p.start=c(0.1,0.2,0.3,0.4), p.end=c(0.4, 0.3, 0.2, 0.1),
#'                    G=5, n.prop=c(3,2,1,2,3))
#'
#' ## Power of stroke study with 100 subjects per year and observed trends
#' data(stroke)
#' strk.mat <- xtabs(Freq ~ Type + Year, data=stroke)
#' power.multiCA.test(N=900, pmatrix=prop.table(strk.mat, margin=2))
#' @export

power.multiCA.test <- function(N=NULL, power=NULL, pmatrix=NULL, p.ave=NULL, p.start=NULL,
                              p.end=NULL, slopes=NULL, scores=1:G, n.prop=rep(1, G),
                              G=length(p.ave), sig.level=0.05){
  if (sum(sapply(list(N, power), is.null)) != 1)
    stop("exactly one of 'N', and 'power' must be NULL")
  if (!is.numeric(sig.level) || any(0 > sig.level | sig.level > 1))
    stop("'sig.level' must be numeric in [0, 1]")

  (Calculate p.ave and slopes from specification 10b)

  df <- K - 1
  crit <- qchisq(sig.level, df=df, lower.tail=FALSE)
  ncp0 <- sum(slopes^2 / p.ave) * s2
  if (missing(power)){
    ncp <- ncp0 * N
    power <- pchisq(crit, df=df, ncp=ncp, lower.tail=FALSE)
  }
  else {
    ncp <- cnonct(crit, p=1-power, df=df)
    N <- ncp / ncp0
  }
}
```

```

res <- structure(list(n = N, n.prop = n.prop, p.ave=p.ave, slopes = slopes, G = G,
  sig.level = sig.level, power = power,
  method = "Multinomial Cochran-Armitage trend test"),
  class = "power.htest")
res
}

```

File defined by 2b, 3, 4a, 6c, 8a, 9.

Defines: power.multiCA.test 10a, 12ac.

Uses: .multiCA.test 2b, cnonct 8a.

"../tests/testthat/test_power.R" 10a≡

```

test_that("calculated power is independent of the input format", {
  pmat <- rbind(seq(0.1, 0.4, length=5),
    seq(0.2, 0.3, length=5),
    seq(0.3, 0.1, length=5),
    seq(0.4, 0.2, length=5))
  res0 <- power.multiCA.test(N=100, pmatrix=pmat)
  expect_equal(res0, power.multiCA.test(N=100, p.start=pmat[,1], p.end=pmat[,5], G=5))
  expect_equal(res0, power.multiCA.test(N=100, p.start=pmat[,1], p.ave=rowMeans(pmat),
    G=5))
  expect_equal(res0, power.multiCA.test(N=100, p.end=pmat[,5], p.ave=rowMeans(pmat),
    G=5))
  expect_equal(res0, power.multiCA.test(N=100, p.ave=rowMeans(pmat),
    slopes=pmat[,2]-pmat[,1], G=5))
  expect_equal(res0, power.multiCA.test(N=100, p.start=pmat[,1],
    slopes=pmat[,2]-pmat[,1], G=5))
  expect_equal(res0, power.multiCA.test(N=100, p.end=pmat[,5],
    slopes=pmat[,2]-pmat[,1], G=5))
})

test_that("Power is computed correctly", {
  pmat <- rbind(seq(0.1, 0.4, length=5),
    seq(0.2, 0.3, length=5),
    seq(0.3, 0.1, length=5),
    seq(0.4, 0.2, length=5))
  res0 <- power.multiCA.test(N=100, pmatrix=pmat)
  expect_equal(100, power.multiCA.test(power=res0$power, pmatrix=pmat)$n)
  expect_equal(0.1, power.multiCA.test(N=100, p.ave=c(0.5, rep(0.1, 5)),
    slopes=rep(0,6), G=6, sig.level=0.1)$power)
})

```

File defined by 8b, 10a, 12ac.

Uses: .multiCA.test 2b, power.multiCA.test 9.

When `slopes` is not specified, then a linear trend for each outcome is assumed:

$$p_{ij} = \bar{p}_j + \beta_j(c_i - \bar{c})$$

{ Calculate p.ave and slopes from specification 10b } ≡

```

if (!is.null(pmatrix)){
  K <- nrow(pmatrix)
  G <- ncol(pmatrix)
  if (!isTRUE(all.equal(colSums(pmatrix), rep(1, G),
    check.attributes=FALSE, use.names=FALSE)))
    stop("pmatrix should have column sums of 1.")
}

```

```

    { Get cbar and s2 11 }
    slopes <- as.vector(pmatrix %*% (n.prop * (scores-cbar))) / s2
    p.ave <- as.vector(pmatrix %*% n.prop)
  }
  else {
    if (sum(sapply(list(p.ave, slopes, p.start, p.end), is.null)) != 2)
      stop("Either pmatrix, or exactly two of 'p.ave', 'slopes', 'p.start', and 'p.end' must be specified")

    if (!is.null(p.ave) & !is.null(slopes)){
      if (length(p.ave) != length(slopes))
        stop("p.ave and slopes should have the same length")
      K <- length(p.ave)
      { Get cbar and s2 11 }
    }
    else if (!is.null(p.ave) & !is.null(p.start)){
      if (length(p.ave) != length(p.start))
        stop("p.ave and p.start should have the same length")
      K <- length(p.ave)
      { Get cbar and s2 11 }
      slopes <- (p.start - p.ave) / (scores[1] - cbar)
    }
    else if (!is.null(p.ave) & !is.null(p.end)){
      if (length(p.ave) != length(p.end))
        stop("p.ave and p.end should have the same length")
      K <- length(p.ave)
      { Get cbar and s2 11 }
      slopes <- (p.end - p.ave) / (scores[G] - cbar)
    }
    else if (!is.null(p.start) & !is.null(p.end)){
      if (length(p.start) != length(p.end))
        stop("p.start and p.end should have the same length")
      K <- length(p.start)
      { Get cbar and s2 11 }
      slopes <- (p.end - p.start) / (scores[G] - scores[1])
      p.ave <- p.start - slopes * (scores[1] - cbar)
    }
    else if (!is.null(p.start) & !is.null(slopes)){
      if (length(p.start) != length(slopes))
        stop("p.start and slopes should have the same length")
      K <- length(p.start)
      { Get cbar and s2 11 }
      p.ave <- p.start - slopes * (scores[1] - cbar)
    }
    else if (!is.null(p.end) & !is.null(slopes)){
      if (length(p.end) != length(slopes))
        stop("p.end and slopes should have the same length")
      K <- length(p.end)
      { Get cbar and s2 11 }
      p.ave <- p.end - slopes * (scores[G] - cbar)
    }
  }

  { Check validity of p.ave and slopes 12b }
}

```

Fragment referenced in 9.

{ Get cbar and s2 11 } \equiv

```

if (missing(G)){
  if (!missing(scores)) G <- length(scores)
  else if (!missing(n.prop)) G <- length(n.prop)
  else stop("The number of groups G needs to be specified explicitly or implicitly through the dimensions")
}
if (sum(n.prop) != 1) n.prop <- n.prop/sum(n.prop)
cbar <- weighted.mean(scores, w=n.prop)
s2 <- sum(n.prop * (scores-cbar)^2)

```

Fragment referenced in [10b](#).

"../tests/testthat/test_power.R" 12a≡

```

test_that("G is properly identified", {
  expect_error(power.multiCA.test(N=100, p.start=c(0.1, 0.9), p.end=c(0.8, 0.2)),
    "G needs to be specified")
  expect_equal(power.multiCA.test(N=100, p.start=c(0.1, 0.9), p.end=c(0.8, 0.2),
    n.prop=rep(1,4))$G, 4)
  expect_equal(power.multiCA.test(N=100, p.start=c(0.1, 0.9), p.end=c(0.8, 0.2),
    scores=1:4)$G, 4)
})

test_that("Scaling of n.prop does not matter", {
  expect_equal(
    power.multiCA.test(N=100, p.start=c(0.1, 0.9), p.end=c(0.8, 0.2), G=6,
      n.prop=rep(1,6)),
    power.multiCA.test(N=100, p.start=c(0.1, 0.9), p.end=c(0.8, 0.2), G=6,
      n.prop=rep(2,6)))
})

```

File defined by [8b](#), [10a](#), [12ac](#).

Uses: `.multiCA.test` [2b](#), `power.multiCA.test` [9](#).

To ensure a valid setup, slopes should add up to 0, and all of the p_{ij} 's implied by a linear trend should be between 0 and 1.

< Check validity of p.ave and slopes 12b > ≡

```

if (!isTRUE(all.equal(sum(slopes), 0, check.attributes=FALSE, use.names=FALSE)))
  stop("Implied or specified values of slopes should sum to 0.")
if (!isTRUE(all.equal(sum(p.ave), 1, check.attributes=FALSE, use.names=FALSE)))
  stop("Implied or specified values of p.ave should sum to 1.")
check <- outer(1:K, 1:G, function(j,i)p.ave[j] + slopes[j]*(scores[i]-cbar))
if (!all(check >= 0) || !(all(check <=1)))
  stop("The parameters do not define a valid probability matrix")

```

Fragment referenced in [10b](#).

"../tests/testthat/test_power.R" 12c≡

```

test_that("p.ave and slopes are checked for validity", {
  expect_error(power.multiCA.test(N=100, p.start=c(0.1, 0.3), p.end=c(0.8, 0.2), G=3),
    "slopes should sum to 0")
  expect_error(power.multiCA.test(N=100, p.ave=c(0.1, 0.8), slopes=c(0.1, -0.1), G=4),
    "p.ave should sum to 1")
  expect_error(power.multiCA.test(N=100, p.ave=c(0.1, 0.9), slopes=c(0.1, -0.1), G=4),

```

```

    "valid probability matrix")
  expect_error(power.multiCA.test(N=100, p.ave=c(0.4, 0.6), slopes=c(0.1, 0.1), G=3),
    "slopes should sum to 0")
})

```

◇
 File defined by [8b](#), [10a](#), [12ac](#).
 Uses: `.multiCA.test` [2b](#), `power.multiCA.test` [9](#).

5. FILES

"../R/aaa-generics.R" Defined by [2a](#).
 "../R/multiCA.R" Defined by [2b](#), [3](#), [4a](#), [6c](#), [8a](#), [9](#).
 "../tests/testthat/test_overall.R" Defined by [4b](#), [5c](#), [6b](#).
 "../tests/testthat/test_power.R" Defined by [8b](#), [10a](#), [12ac](#).

6. MACROS

⟨ Calculate adjusted p-values [5a](#) ⟩ Referenced in [3](#).
 ⟨ Calculate p.ave and slopes from specification [10b](#) ⟩ Referenced in [9](#).
 ⟨ Check validity of p.ave and slopes [12b](#) ⟩ Referenced in [10b](#).
 ⟨ Closed set adjustment [6a](#) ⟩ Referenced in [5a](#).
 ⟨ Get cbar and s2 [11](#) ⟩ Referenced in [10b](#).
 ⟨ Holm-Shaffer adjustment [5b](#) ⟩ Referenced in [5a](#).

7. IDENTIFIERS

`.multiCA.test`: [2b](#), [3](#), [6a](#), [9](#), [10a](#), [12ac](#).
`cnonct`: [8a](#), [8b](#), [9](#).
`multiCA.test.default`: [3](#).
`multiCA.test.formula`: [4a](#).
`power.multiCA.test`: [9](#), [10a](#), [12ac](#).