



MULTIVARIATE CUMULANTS IN R

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ABSTRACT. We give algorithms and **R** code to compute all multivariate cumulants up to order p of m variables.

1. INTRODUCTION

Multivariate cumulants are of increasing importance in various areas of statistics and data analysis. An excellent introduction to cumulants is McCullagh and Kolassa [2009]. More details, and many deeper and far-reaching results, are in Speed [1983]; McCullagh [1984, 1987]; Kamanzi-wa-Binyavanga [2009]; Peccati and Taqqu [2011]. Applications in signal processing are in Hinich [1994]; Mendel [1991]. Applications in independent component analysis are in Lim and Morton [2008]; Morton and Lim [2009]; Abrar and Nandi [2009].

2. COMPUTATION

In most publications the discussion of cumulants is algebraic and combinatorial. Although the algebraic results can certainly be used to actually compute multivariate cumulants, the translation into algorithms and code is not entirely straightforward. In many cases

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the formulas suggest symbolic computation in packages such as Maple or Mathematica and they are geared towards numerical computation. An exception are Smith [1995] and Balakrishnan et al. [1998], where numerical recursions are given to compute cumulants from moments (and vice versa).

In this paper we present an algorithm that does use combinatorics, because we sum over all partitions of a finite set to compute a cumulant from raw multivariate moments. We leave the heavy combinatorial lifting, however, to the `setparts()` function for set partitioning [West and Hankin, 2007], implemented in the `R` package `partitions` [Hankin, 2006]. Our cumulant calculations, using the partitions given by the `setparts()` function, are illustrated nicely in Table A-1 of Mendel [1991, page 297]. The key formula is

$$(1) \quad \kappa(x_1, \dots, x_p) = \sum_{\text{partitions}} (-1)^{q-1} (q-1)! \prod_{r=1}^q m_x(I_r),$$

where the summation is over all partitions of $\{1, 2, \dots, p\}$ and q is the number of subsets in the partition. The subsets in the partition are I_r , and $m_x(I_r)$ is the raw (product) moment of the variables in subset I_r .

3. CODE

The Appendix has the code, written in `R` [R Development Core Team, 2012]. The code requires, in addition to the `partitions` package, also the `ap1` package [De Leeuw and Yajima, 2011] for the functions `ap1Encode()` and `ap1Select()`. The `ap1` package makes it easy to deal with arrays of arbitrary dimension.

3.1. `raw_moments_upto_p`. Given an $n \times m$ multivariate data matrix this function computes the $\underbrace{(m+1) \times \dots \times (m+1)}_{p \text{ times}}$ symmetric array of multivariate moments and product moments around zero up to order p . It uses the `outer()` function recursively, and

does not require anything outside base `R`. There are $m + 1$ rows, columns, slices, ... , because we include a variable with all elements equal to one. Thus the array contains the m moments of order one, the m^2 moments of order two, ... , the m^p moments of order p .

3.2. **`cumulants_from_raw_moments()`**. This function takes the raw moment array and makes an array of the same size with the cumulants up to order p , relying heavily on the `cumulant_from_raw_moments()` function discussed below. It uses `ap1Encode()` to move rapidly through the array.

3.3. **`cumulants_upto_p()`**. This is a simple function to compute the cumulants directly from the data matrix. It calls the previous two functions.

3.4. **`first_four_cumulants()`**. Another utility function that returns a list of the first four cumulants

3.5. **`one_cumulant_from_raw_moments()`**. Most of the computation goes on here. The function takes as arguments an index vector indicating which cumulant should be computed and the array with raw moments. Because of our convention to add a column of ones to the data the index vector `c(1,2,1,4)` actually leads to the cumulant $\kappa(1,3)$. Eliminate the ones from the index vector and subtract one from the remaining elements.

3.6. **`four_cumulants_direct()`**. This function also returns the first four cumulant arrays, but it uses completely written out versions of formula (1) for orders two, three, and four. It has much less array manipulation and indexing, but it cannot be used for higher-order cumulant arrays. It is quite a bit faster than the function `first_four_cumulants()` that selects the cumulants out of the much larger array.

4. EXAMPLE

We generate data by using $X = YB'$, where B is a fixed orthonormal matrix and Y has independent standard normal random numbers raised to a given power. The function `make_artificial()` can also generate factor analysis examples, using `err=TRUE`, but we do not discuss these in this paper.

```

1 source("/Users/deLeeuw/Dropbox/cumulants/cumulants.R")
2
3 make_artificial<-function(n, m, p, pow, err = FALSE, seed =
4   12345) {
5   set.seed (seed)
6   mp <- m * p
7   np <- n * p
8   nm <- n * m
9   x <- matrix (rnorm(np), n, p) ^ pow
10  if (err) {
11    x <- cbind (x, matrix (rnorm (nm), n, m) ^ pow)
12  }
13  x <- apply (x, 2, function (z) z - mean (z))
14  x <- sqrt (n) * qr.Q (qr (x))
15  b <- qr.Q ( qr (matrix (sample (1 : mp, mp), m, p)))
16  if (err) {
17    b <- cbind (b, diag (sample (1 : m, m) / m))
18  }
19  return (list (x = x, b = b, y = tcrossprod (x, b)))
20 }
21 arti <- make_artificial (n = 1000, m = 9, p = 4, pow = 2)
22 x <- arti $ y
23 print (system.time (cumuold <- four_cumulants_direct (x)))
24 print (system.time (cumunew <- first_four_cumulants (x)))
25 print (max (abs (cumuold $ c2 - cumunew $ c2)))
26 print (max (abs (cumuold $ c3 - cumunew $ c3)))
27 print (max (abs (cumuold $ c4 - cumunew $ c4)))

```

The output when running this file is

```
1 > source("cumutry.R")
2   user  system elapsed
3   0.408   0.057   0.426
4   user  system elapsed
5   5.592   0.063   5.592
6 [1] 1.554312e-15
7 [1] 5.551115e-17
8 [1] 5.995204e-15
```

We see that in this example the direct method is 13 times faster than the more general method that uses `setparts()`.

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APPENDIX A. CODE

```
1 require("partitions")
2 require("apl")
3
4 raw_moments_upto_p <- function (x, p = 4) {
5   n <- nrow (x)
6   m <- ncol (x)
7   if (p == 1) {
8     return (c (1, apply (x, 2, mean)))
9   }
10  y <- array (0, rep (m + 1, p))
11  for (i in 1 : n) {
12    xi <- c (1, x[i, ])
13    z <- xi
14    for (s in 2:p) {
15      z <- outer (z, xi)
16    }
17    y <- y + z
18  }
19  return (y / n)
20 }
21
22 cumulants_from_raw_moments <- function (raw) {
23   dimr <- dim (raw)
24   nvar <- dimr[1]
25   cumu <- array (0, dimr)
26   nele <- prod (dimr)
27   ldim <- length (dimr)
28   spp <-< rep (list (0), ldim)
29   qpp <-< rep (0, ldim)
30   rpp <-< rep (list (0), ldim)
31   for (i in 1 : ldim) {
32     spp[[i]] <-< setparts (i)
33     qpp[[i]] <-< factorial (i)
34     if (i %% 2) {
35       qpp[[i]] <-< -qpp[[i]]
```



```

36     }
37     rpp[[i]] <-> apSelect (raw, c (rep (list (1 : nvar
      ), i), rep (list (1 : 1), ldim - i)))
38   }
39   qpp <-> c(1, qpp)
40   for (i in 2 : nele) {
41     ind <-> apEncode (i, dimr)
42     cumu[i] <-> one_cumulant_from_raw_moments (ind, raw)
43   }
44   return (cumu)
45 }
46
47 cumulants_upto_p <-> function (x, p = 4) {
48   return (cumulants_from_raw_moments (raw_moments_upto_p
      (x, p)))
49 }
50
51 first_four_cumulants <-> function (x) {
52   cumu <-> cumulants_upto_p (x)
53   nsel <-> dim (cumu)[1]
54   return (list (c1 = cumu[1, 1, 1, nsel],
55     c2 = cumu[1, 1, nsel, nsel],
56     c3 = cumu[1, nsel, nsel, nsel],
57     c4 = cumu[nsel, nsel, nsel, nsel]))
58 }
59
60 one_cumulant_from_raw_moments <-> function (jnd, raw) {
61   jnd <-> jnd [which (jnd != 1)] - 1
62   nnd <-> length (jnd)
63   ndr <-> dim (raw)[1]
64   nrt <-> length (dim (raw))
65   raw <-> rpp[[nnd]]
66   nvar <-> ndr - 1
67   nraw <-> max (1, length (dim (raw)))
68   sp <-> spp [[nraw]]
69   nbell <-> ncol (sp)
70   sterm <-> 0
71   for (i in 1 : nbell) {
72     ind <-> sp[, i]

```

```

73     und <- unique (ind)
74     term <- qpp[length (und)]
75     for (j in und) {
76         knd <- jnd[which (ind == j)] + 1
77         lnd <- c (knd, rep (1, nrow - length (knd)))
78         term <- term * raw [matrix (lnd, 1, nrow)]
79     }
80     sterm <- sterm + term
81 }
82 return (sterm)
83 }
84
85 four_cumulants_direct <- function (x) {
86     n <- nrow (x)
87     m <- ncol (x)
88     mm <- 1 : m
89     nn <- 1 : n
90     r1 <- colSums (x) / n
91     r2 <- crossprod (x) / n
92     r3 <- array (0, c (m, m, m))
93     r4 <- array (0, c (m, m, m, m))
94     for (i in nn) {
95         r3 <- r3 + outer (outer (x [i, ], x [i, ]), x [i,
96             ])
97         r4 <- r4 + outer (outer (x [i, ], x [i, ]), outer (
98             x [i, ], x [i, ]))
99     }
100     r3<-r3 / n
101     r4<-r4 / n
102     c2<-r2 - outer (r1, r1)
103     c3<-r3
104     for (i in mm) for (j in mm) for (k in mm) {
105         s3 <- r3 [i, j, k]
106         s21 <- r2 [i, j] * r1 [k] + r2 [i, k] * r1 [j] + r2
107             [j, k] * r1 [i]
108         s111 <- r1 [i] * r1 [j] * r1 [k]
109         c3 [i, j, k] <- s3 - s21 + 2 * s111
110     }
111     c4<-r4

```

```

109   for (i in mm) for (j in mm) for (k in mm) for (l in mm)
      {
110     s4 <- r4 [i, j, k, l]
111     s31 <- r3 [i, j, k] * r1 [l] + r3 [i, j, l] * r1 [k]
          + r3 [i, k, l] * r1 [j] + r3 [j, k, l] * r1 [i]
          ]
112     s22 <- r2 [i, j] * r2 [k, l] + r2 [i, k] * r2 [j, l]
          + r2 [j, k] * r2 [i, l]
113     s211 <- r2 [i, j] * r1 [k] * r1 [l] + r2 [i, k] *
          r1 [j] * r1 [l] + r2 [i, l] * r1 [k] * r1 [j] +
          r2 [j, k] * r1 [i] * r1 [l] + r2 [j, l] * r1 [i]
          * r1 [k] + r2 [k, l] * r1 [i] * r1 [j]
114     s1111 <- r1 [i] * r1 [j] * r1 [k] * r1 [l]
115     c4 [i, j, k, l] <- s4 - s31 - s22 + 2 * s211 - 6 *
          s1111
116   }
117   return (list(c1 = r1, c2 = c2, c3 = c3, c4 = c4))
118 }

```

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