# Package CompSign

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**CompSign** is a package for yadayada... overlooked that mutational signatures are compositional in nature yadayada. The reference manual can be found here.

```
knitr::opts_chunk$set(cache = FALSE)
```

```
## This chunk was last ran in
timestamp()
## ##----- Wed Oct 24 09:51:46 2018 -----##
## install latest version
library(devtools)
devtools::install_github("lm687/CompSign")
## Skipping install of 'CompSign' from a github remote, the SHA1 (ae95502a)
has not changed since last install.
## Use 'force = TRUE' to force installation
library(CompSign)
library(compositions)
## Loading required package: tensorA
##
## Attaching package: 'tensorA'
## The following object is masked from 'package:base':
##
##
      norm
## Loading required package: robustbase
## Loading required package:
                              energy
## Loading required package: bayesm
## Welcome to compositions, a package for compositional data analysis.
## Find an intro with "? compositions"
## Attaching package: 'compositions'
```

```
## The following objects are masked from 'package:stats':
##
## cor, cov, dist, var
## The following objects are masked from 'package:base':
##
## %*%, scale, scale.default
```

#### 1 Summarise the signature matrix

```
## This chunk was last ran in
timestamp()

## ##----- Wed Oct 24 09:51:48 2018 -----##

add_together_matrix(sign_dummy)

## An object of class "sign"

## Slot "id":

## [1] "input_dummy"

##

## Slot "id_samples":

## [1] "sam1" "sam2" "sam3" "sam4"

##

## ## Slot "id_signatures":

## [1] "s1" "s2" "s3" "s4" "s5" "s6" "s7" "s8" "s9" "s10" "s11"

## [12] "s12" "s13" "s14" "s15" "s16" "s17" "s18" "s19" "s20" "s21" "s22"

## [23] "s23" "s24" "s25"

##

## ## Slot "count_matrix":
```

```
##
                         s2
                                   s3
                                             s4
## sam1 0.6764623 0.2729444 0.3436545 0.1002153 0.06561845 0.03094423
## sam2 0.4520095 0.4091145 0.1040464 0.3355779 0.68618118 0.44924840
## sam3 0.9363835 0.3928727 0.4027341 0.2753264 0.65121843 0.66822314
  sam4 0.5446427 0.1544315 0.2462713 0.3190422 0.43053817 0.77015991
##
               s7
                         s8
                                   s9
                                            s10
                                                       s11
                                                                 s12
## sam1 0.3532618 0.8480522 0.7084803 0.8895526 0.6296052 0.4025335
  sam2 0.8411673 0.7207060 0.8103149 0.1315620 0.8245883 0.5702180
## sam3 0.9850511 0.5855240 0.5503083 0.8914260 0.8225037 0.9071219
  sam4 0.7384766 0.2273273 0.7245602 0.1884538 0.2813367 0.4807784
##
               s13
                          s14
                                    s15
                                              s16
                                                          s17
## sam1 0.61211950 0.58389072 0.7192076 0.4294944 0.16047355 0.6105783
## sam2 0.01719359 0.16899129 0.7834863 0.1660177 0.45931066 0.1057023
## sam3 0.94706416 0.02446017 0.7819861 0.1789141 0.29158493 0.3087867
## sam4 0.51996204 0.42566369 0.6535919 0.7844009 0.02664156 0.6373981
               s19
                         s20
                                    s21
                                               s22
## sam1 0.30225377 0.9639017 0.79162225 0.02198544 0.08396833 0.6812706
## sam2 0.07004629 0.7094089 0.70866945 0.43777464 0.78886873 0.6356862
## sam3 0.27966923 0.8772923 0.04130497 0.82702003 0.41161577 0.2505033
## sam4 0.65060962 0.9220265 0.16407478 0.60034012 0.54354101 0.4179587
##
                s25
## sam1 0.193195837
## sam2 0.036536185
## sam3 0.607445674
## sam4 0.006969308
##
## Slot "modified":
## [1] TRUE
results_sumarise <- summarise(add_together_matrix(sign_dummy))
results_sumarise$General
## [1] "Object of class sign"
```

## 2 Linear model for numerical predictors

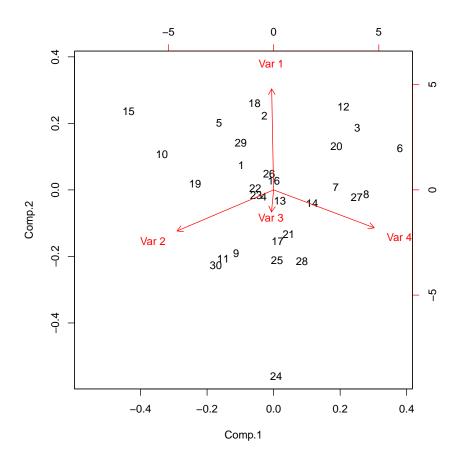
```
id_signatures= c('s1', 's2', 's3', 's4'), ## signature name.
                                count_matrix=MCMCpack::rdirichlet(30, c(1,1,1,1)),
                                df=data.frame(a=sample(1:1e4, 30), b=rep(10, 30)))
comp_lm(tmp_merged_compositional)
## [[1]]
## Response Y1 :
##
## Call:
## lm(formula = Y1 ~ as.matrix((x@df)[, indices_predictor]))
## Residuals:
##
      Min
               1Q Median
                                30
                                       Max
## -1.7305 -0.6656 -0.1989 0.6349 2.8453
##
## Coefficients: (1 not defined because of singularities)
##
                                             Estimate Std. Error t value
## (Intercept)
                                            2.308e-01 3.747e-01
                                                                   0.616
## as.matrix((x@df)[, indices_predictor])a -2.947e-05 7.314e-05 -0.403
## as.matrix((x@df)[, indices_predictor])b
                                                   NA
                                                              NA
##
                                           Pr(>|t|)
## (Intercept)
                                              0.543
## as.matrix((x@df)[, indices_predictor])a
                                              0.690
## as.matrix((x@df)[, indices_predictor])b
##
## Residual standard error: 1.134 on 28 degrees of freedom
## Multiple R-squared: 0.005766, Adjusted R-squared:
## F-statistic: 0.1624 on 1 and 28 DF, p-value: 0.69
##
##
## Response Y2 :
##
## Call:
## lm(formula = Y2 ~ as.matrix((x@df)[, indices_predictor]))
##
## Residuals:
##
      Min
               1Q Median
                                3Q
## -2.8634 -0.5298 0.2247 0.7899 1.9961
##
## Coefficients: (1 not defined because of singularities)
##
                                             Estimate Std. Error t value
## (Intercept)
                                            1.438e-01 3.704e-01 0.388
## as.matrix((x@df)[, indices_predictor])a -3.753e-05 7.231e-05 -0.519
## as.matrix((x@df)[, indices_predictor])b
                                                   NA
                                                              NA
                                                                      NΑ
                                           Pr(>|t|)
##
                                              0.701
```

## (Intercept)

```
## as.matrix((x@df)[, indices_predictor])a
                                              0.608
## as.matrix((x@df)[, indices_predictor])b
##
## Residual standard error: 1.121 on 28 degrees of freedom
## Multiple R-squared: 0.00953, Adjusted R-squared: -0.02584
## F-statistic: 0.2694 on 1 and 28 DF, p-value: 0.6078
##
##
## Response Y3 :
##
## Call:
## lm(formula = Y3 ~ as.matrix((x@df)[, indices_predictor]))
##
## Residuals:
   Min
               1Q Median
                                3Q
                                       Max
## -3.6174 -0.7001 0.2863 0.8532 2.6761
## Coefficients: (1 not defined because of singularities)
##
                                            Estimate Std. Error t value
                                           -4.560e-01 4.850e-01 -0.940
## (Intercept)
## as.matrix((x@df)[, indices_predictor])a 8.054e-06 9.468e-05
                                                                   0.085
## as.matrix((x@df)[, indices_predictor])b
                                                 NA
                                                           NA
                                                                      NA
                                          Pr(>|t|)
## (Intercept)
                                              0.355
## as.matrix((x@df)[, indices_predictor])a
                                              0.933
## as.matrix((x@df)[, indices_predictor])b
                                                 NA
##
## Residual standard error: 1.468 on 28 degrees of freedom
## Multiple R-squared: 0.0002584, Adjusted R-squared: -0.03545
## F-statistic: 0.007236 on 1 and 28 DF, p-value: 0.9328
```

## 3 Importing data

```
## This chunk was last ran in
timestamp()
## ##----- Wed Oct 24 09:51:49 2018 -----##
biplot(princomp(acomp(MCMCpack::rdirichlet(30, rep(1, 4)))))
```



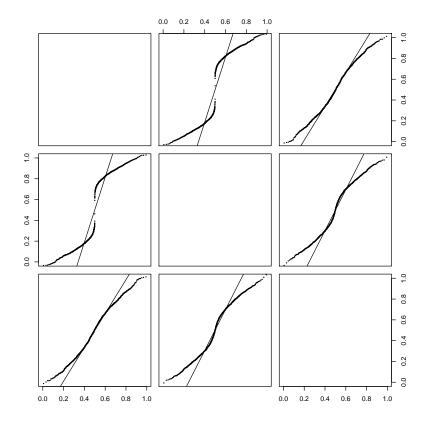
### 4 Other

1. Test for normality as follows:

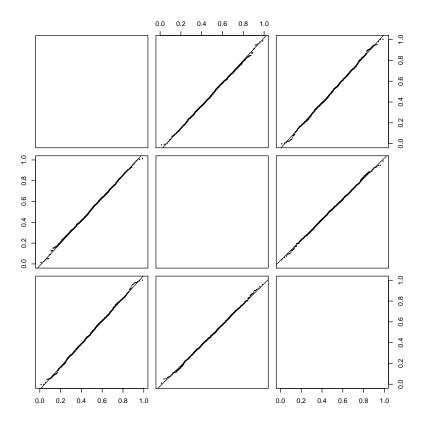
```
## This chunk was last ran in
timestamp()

## ##----- Wed Oct 24 09:51:49 2018 -----##

data(two_normal_pops)
par(mfrow=c(1,2))
qqnorm.acomp(acomp(two_normal_pops@count_matrix), pch=19, cex=0.2)
```



qqnorm.acomp(acomp(two\_normal\_pops@count\_matrix[1:1000,]), pch=19, cex=0.2)



#### 4.1 Testing hypotheses about two populations

We might have our samples split into two categories; e.g. sex. As in Aithison 1986, I follow a hierarchy of alternative hypotheses, from least to most complex.

Our first question is whether two populations have the same covariance and structure and center (i.e. if there is any distributional difference)

```
## This chunk was last ran in
timestamp()
## ##----- Wed Oct 24 09:51:50 2018 -----##
##TODO!!
```

The next is whether the populations have a different center:

```
## This chunk was last ran in
timestamp()
## ##----- Wed Oct 24 09:51:50 2018 -----##
## This dataset includes the two components above, as well as four others
## (a total of seven)
data("two_normal_pops_extended")
compare_populations(predictors = two_normal_pops_extended@count_matrix,
                    response = two_normal_pops_extended@df[,1])
## Loading required package: Compositional
##
## Attaching package: 'Compositional'
## The following object is masked from 'package:compositions':
##
##
## Error in Compositional::comp.test(x = predictors[, -1], ina = tmp_response,
: object 'result' not found
```

# 5 Data for 560 breast cancer patients

Data from 560 breast cancer patients is available as part of the document as well:

```
##continue in save_560BRCA_rda.R
```