## 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()
## ##---- Tue Oct 30 15:28:24 2018 -----##
## install latest version
library(devtools)
devtools::install_github("lm687/CompSign")
## Skipping install of 'CompSign' from a github remote, the SHA1 (2d73ef33)
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()

## ##----- Tue Oct 30 15:28:25 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
## sam1 0.4260863 0.2755039 0.08276189 0.6208283 0.8857211 0.5549314
## sam2 0.1411205 0.8640290 0.73175191 0.1036485 0.2420634 0.8307967
## sam3 0.4049222 0.4724351 0.09832678 0.7830958 0.1972511 0.7619885
## sam4 0.3090491 0.4183054 0.91512885 0.5695176 0.6805209 0.6588088
##
               s7
                         s8
                                   s9
                                            s10
                                                        s11
## sam1 0.7470576 0.3506052 0.8857402 0.3951667 0.78707440 0.2070237
  sam2 0.4229675 0.3203786 0.5956866 0.9977689 0.65447704 0.8519154
## sam3 0.2811029 0.5965693 0.8706372 0.4038557 0.23177664 0.7540944
  sam4 0.7962489 0.4228749 0.4271666 0.5252904 0.01935426 0.3826507
##
              s13
                         s14
                                   s15
                                              s16
                                                         s17
## sam1 0.6481574 0.52636785 0.8357955 0.25466215 0.9000820 0.8057467
## sam2 0.5874839 0.04076431 0.5073318 0.61878639 0.5855561 0.7308849
## sam3 0.5922528 0.70681435 0.9099575 0.23085795 0.8690805 0.28333331
## sam4 0.1860448 0.30120701 0.7998575 0.03621599 0.1804691 0.6519621
              s19
                        s20
                                  s21
                                            s22
## sam1 0.2507280 0.3409191 0.6532946 0.2509273 0.96859098 0.3243763
## sam2 0.4975619 0.3582334 0.9787369 0.2323121 0.49452011 0.9581426
## sam3 0.6919547 0.5571275 0.8511707 0.9243823 0.27759639 0.8038591
## sam4 0.1117317 0.9430649 0.3228245 0.9384450 0.01049231 0.2159923
##
## sam1 0.8020041
## sam2 0.4620849
## sam3 0.8629497
## sam4 0.9033385
##
## 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

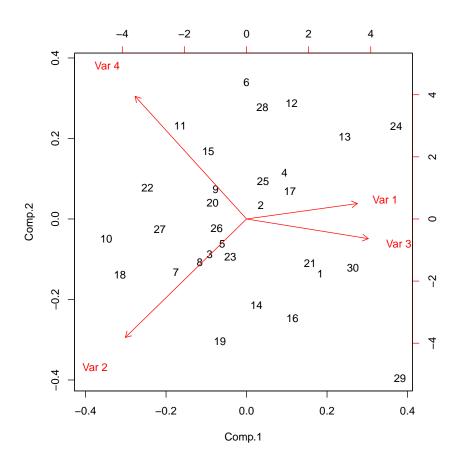
```
id_signatures= c('s1', 's2', 's3', 's4'), ## signature name.
                                count_matrix=MCMCpack::rdirichlet(30, c(1,1,1,1)),
                                df=data.frame(a=sample(1:1e4, 30), b=rep(10, 30)))
comp_lm(tmp_merged_compositional)
## [[1]]
## Response Y1 :
##
## Call:
## lm(formula = Y1 ~ as.matrix((x@df)[, indices_predictor]))
## Residuals:
      Min
               10 Median
                                30
                                       Max
## -3.1173 -0.7555 0.1894 0.6292 2.5870
##
## Coefficients: (1 not defined because of singularities)
##
                                             Estimate Std. Error t value
## (Intercept)
                                            4.905e-01 4.600e-01
## as.matrix((x@df)[, indices_predictor])a -4.835e-05 7.609e-05 -0.635
## as.matrix((x@df)[, indices_predictor])b
                                                   NA
                                                              NA
##
                                           Pr(>|t|)
## (Intercept)
                                              0.295
## as.matrix((x@df)[, indices_predictor])a
                                              0.530
## as.matrix((x@df)[, indices_predictor])b
##
## Residual standard error: 1.185 on 28 degrees of freedom
## Multiple R-squared: 0.01421, Adjusted R-squared: -0.02099
## F-statistic: 0.4038 on 1 and 28 DF, p-value: 0.5303
##
##
## Response Y2 :
##
## Call:
## lm(formula = Y2 ~ as.matrix((x@df)[, indices_predictor]))
##
## Residuals:
##
      Min
                  1Q
                     Median
                                    3Q
## -1.86633 -0.66484 -0.09362 0.95891 2.05713
##
## Coefficients: (1 not defined because of singularities)
##
                                             Estimate Std. Error t value
## (Intercept)
                                           -1.111e-01 4.314e-01 -0.257
## as.matrix((x@df)[, indices_predictor])a 5.307e-05 7.136e-05
                                                                   0.744
## as.matrix((x@df)[, indices_predictor])b
                                                   NA
                                                              NA
                                                                      NΑ
                                           Pr(>|t|)
##
                                              0.799
```

## (Intercept)

```
## as.matrix((x@df)[, indices_predictor])a
                                             0.463
## as.matrix((x@df)[, indices_predictor])b
##
## Residual standard error: 1.111 on 28 degrees of freedom
## Multiple R-squared: 0.01937, Adjusted R-squared: -0.01565
## F-statistic: 0.5531 on 1 and 28 DF, p-value: 0.4633
##
##
## Response Y3 :
##
## Call:
## lm(formula = Y3 ~ as.matrix((x@df)[, indices_predictor]))
##
## Residuals:
##
       Min
                 1Q
                     Median
                                   3Q
                                           Max
## -1.98099 -0.55176 -0.07108 0.65533 1.88851
## Coefficients: (1 not defined because of singularities)
##
                                            Estimate Std. Error t value
                                          -7.433e-01 3.974e-01 -1.870
## (Intercept)
## as.matrix((x@df)[, indices_predictor])a 1.424e-04 6.574e-05
                                                                  2.166
## as.matrix((x@df)[, indices_predictor])b
                                                 NA
                                                            NA
                                                                     NA
                                          Pr(>|t|)
## (Intercept)
                                             0.072 .
## as.matrix((x@df)[, indices_predictor])a
                                             0.039 *
## as.matrix((x@df)[, indices_predictor])b
                                                NA
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.024 on 28 degrees of freedom
## Multiple R-squared: 0.1435, Adjusted R-squared: 0.1129
## F-statistic: 4.693 on 1 and 28 DF, p-value: 0.03896
```

### 3 Importing data

```
## This chunk was last ran in
timestamp()
## ##----- Tue Oct 30 15:28:26 2018 -----##
biplot(princomp(acomp(MCMCpack::rdirichlet(30, rep(1, 4)))))
```



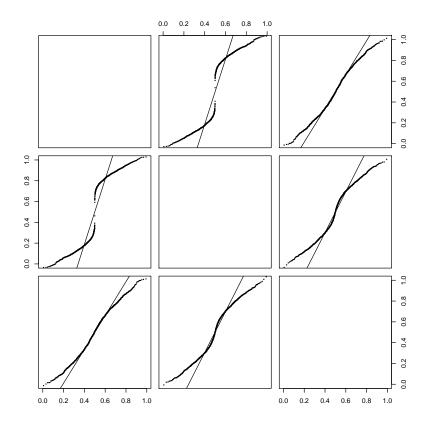
## 4 Other

1. Test for normality as follows:

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

## ##----- Tue Oct 30 15:28:27 2018 -----##

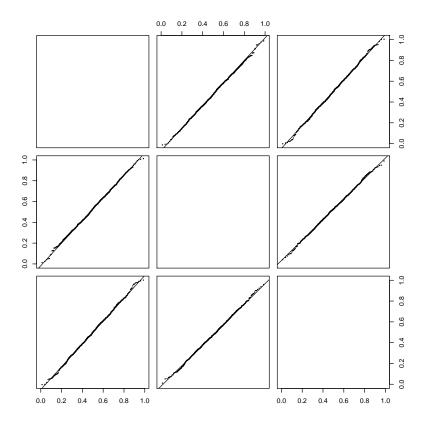
data(two_normal_pops)
par(mfrow=c(1,2))
qqnorm.acomp(acomp(two_normal_pops@count_matrix), pch=19, cex=0.2)
```



qqnorm.acomp(acomp(two\_normal\_pops@count\_matrix[1:1000,]), pch=19, cex=0.2, plot.it=FAI
## Warning in plot.window(...): "plot.it" is not a graphical
parameter
## Warning in plot.xy(xy, type, ...): "plot.it" is not a graphical
parameter
## Warning in title(...): "plot.it" is not a graphical parameter
## Warning in plot.window(...): "plot.it" is not a graphical
parameter
## Warning in plot.xy(xy, type, ...): "plot.it" is not a graphical
parameter
## Warning in title(...): "plot.it" is not a graphical parameter
## Warning in axis(side = side, at = at, labels = labels, ...):
"plot.it" is not a graphical parameter
## Warning in plot.xy(xy.coords(x, y), type = type, ...): "plot.it"
is not a graphical parameter

```
## Warning in plot.window(...): "plot.it" is not a graphical
parameter
## Warning in plot.xy(xy, type, ...): "plot.it" is not a graphical
parameter
## Warning in title(...): "plot.it" is not a graphical parameter
## Warning in axis(side = side, at = at, labels = labels, ...):
"plot.it" is not a graphical parameter
## Warning in plot.xy(xy.coords(x, y), type = type, ...): "plot.it"
is not a graphical parameter
## Warning in plot.window(...): "plot.it" is not a graphical
parameter
## Warning in plot.xy(xy, type, ...): "plot.it" is not a graphical
parameter
## Warning in title(...): "plot.it" is not a graphical parameter
## Warning in axis(side = side, at = at, labels = labels, ...):
"plot.it" is not a graphical parameter
## Warning in plot.xy(xy.coords(x, y), type = type, ...): "plot.it"
is not a graphical parameter
## Warning in plot.window(...): "plot.it" is not a graphical
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## Warning in title(...): "plot.it" is not a graphical parameter
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## Warning in plot.xy(xy, type, ...): "plot.it" is not a graphical
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## Warning in title(...): "plot.it" is not a graphical parameter
## Warning in plot.xy(xy.coords(x, y), type = type, ...): "plot.it"
is not a graphical parameter
## Warning in plot.window(...): "plot.it" is not a graphical
parameter
## Warning in plot.xy(xy, type, ...): "plot.it" is not a graphical
parameter
## Warning in title(...): "plot.it" is not a graphical parameter
## Warning in axis(side = side, at = at, labels = labels, ...):
"plot.it" is not a graphical parameter
## Warning in plot.xy(xy.coords(x, y), type = type, ...): "plot.it"
is not a graphical parameter
```

```
## Warning in plot.window(...): "plot.it" is not a graphical
parameter
## Warning in plot.xy(xy, type, ...): "plot.it" is not a graphical
parameter
## Warning in title(...): "plot.it" is not a graphical parameter
## Warning in plot.xy(xy.coords(x, y), type = type, ...): "plot.it"
is not a graphical parameter
## Warning in plot.window(...): "plot.it" is not a graphical
parameter
## Warning in plot.xy(xy, type, ...): "plot.it" is not a graphical
parameter
## Warning in title(...): "plot.it" is not a graphical parameter
## Warning in axis(side = side, at = at, labels = labels, ...):
"plot.it" is not a graphical parameter
## Warning in axis(side = side, at = at, labels = labels, ...):
"plot.it" is not a graphical parameter
```



## 5 Clustering of samples

#### 5.1 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()
## ##----- Tue Oct 30 15:28:27 2018 -----##
##TODO!!
```

The next is whether the populations have a different center:

```
## This chunk was last ran in
timestamp()
## ##----- Tue Oct 30 15:28:27 2018 -----##
## This dataset includes the two components above, as well as four others
## (a total of seven)
data("two_normal_pops_extended")
## Data from the Landscape... paper
data("Breast560")
wrapper_compare_populations <- function(predictors, response, ...){</pre>
  if(length(unique(response)) == 2){
    tmp <- compare_populations(predictors, response, ...)</pre>
    tmp <- tmp$info[1:2]</pre>
    tmp
x <- do.call('rbind', lapply(1:ncol(metadata(Breast560)),</pre>
       function(k){
         wrapper_compare_populations(predictors = count_matrix(Breast560),
                                      response = metadata(Breast560)[,k])
       ))
## Loading required package: Compositional
##
## Attaching package: 'Compositional'
## The following object is masked from 'package:compositions':
##
##
      alr
```

```
## test p-value
## [1,] 223.6681 6.334800e-26
## [2,] 237.6260 6.270514e-29
## [3,] 237.4362 2.457445e-43
## [4,] 78.3811 9.584122e-12
```

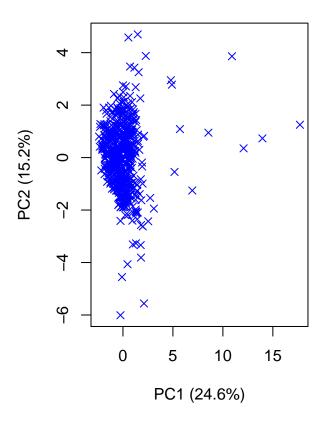
#### 6 Data for 560 breast cancer patients

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

```
data("Breast560")
metadata(Breast560)[1:4,1:5]
##
           donor_gender donor_age_at_diagnosis donor_age_at_last_follow.up
## PD10010
                female
                                             56
                                                           no_data_supplied
## PD10011
                 female
                                             75
                                                           no_data_supplied
## PD10014
                 female
                                             64
                                                           no_data_supplied
## PD11326
                 female
                                             38
                                                                          47
            specimen_type donor_vital_status
## PD10010 tumour_primary
## PD10011 tumour_primary
                                        alive
## PD10014 tumour_primary
                                    deceased
## PD11326 tumour_primary
                                        alive
count_matrix(Breast560)[1:4,1:5]
##
            Signature.1 Signature.2 Signature.3 Signature.5 Signature.6
## PD10010 0.0013656127 0.0002299146 0.0009201629 0.001224842
                                                                         0
## PD10011 0.0020433984 0.0000000000 0.0025729837 0.005577513
                                                                         0
## PD10014 0.0010016166 0.0000000000 0.0060192540 0.005093551
                                                                         0
## PD11326 0.0009765135 0.0005635162 0.0059827947 0.002310769
```

Not sure if this is correct

```
source("~/Documents/CantabPhD/CDA_in_Cancer/code/functions/various_functions.R")
plotPCA(ilr(count_matrix(Breast560)), pch=4, col='blue')
## Loading required package: ggplot2
```



#### 6.1 (ongoing) test for equality

```
comp.test(x = count_matrix(Breast560),
                            ina = as.numeric(as.factor(metadata(Breast560)$final.ER)),
                            test = "james", R = 0)
## $note
## [1] "James test"
##
## $mesoi
                X1
                          X2
                                             X4
##
                                    ХЗ
## Sample 1 0.583829 -1.336900 -0.1776487 1.467427 -0.5977703 0.3018865
## Sample 2 1.410313 1.626698 -1.0779649 1.773800 -0.0129786 0.7226163
                        X8 X9 X10 X11
                 Х7
```

```
## Sample 1 0.9095453 0.59145173 1.170214 0.9935606 0.9807331
## Sample 2 1.2939339 0.04855843 1.241930 1.0600895 1.0444659
##
## $info
## test p-value correction
## 2.797486e+02 6.279768e-51 1.046173e+00
## corrected.critical
## 2.058360e+01
```

# 7 Data for 12k TCGA samples, with ovarian cancer-derived CNA signatures

```
timestamp()
## ##----- Tue Oct 30 15:28:28 2018 -----##
data("CNA_12K_TCGA")
dim(metadata(CNA_12K_TCGA))
## [1] 10899 37
dim(count_matrix(CNA_12K_TCGA))
## [1] 10899 7
```