# Package 'TML'

July 29, 2024

```
Type Package
Title Tropical Geometry Tools for Machine Learning
Version 2.3.0
Description Suite of tropical geometric tools for use in machine learning applications. These meth-
     ods may be summarized in the following refer-
     ences: Yoshida, et al. (2022) <doi:10.2140/astat.2023.14.37>, Barn-
     hill et al. (2023) <doi:10.48550/arXiv.2303.02539>, Barn-
     hill and Yoshida (2023) <doi:10.3390/math11153433>, Alia-
     timis et al. (2023) <doi:10.1007/s11538-024-01327-
     8>, Yoshida et al. (2022) <doi:10.1109/TCBB.2024.3420815>, and Yoshida et al. (2019) <doi:10.1007/s11538-
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```

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bw.nn

Nearest neighbor bandwidth calculation

## **Description**

This function finds the bandwidth for an ultrametric based on the tropical distance of the nearest point. The function provides the bandwidth input to trop.KDE and was originally used in the KDETrees package.

## Usage

```
bw.nn(x, prop = 0.2, tol = 1e-06)
```

## Arguments

tol

matrix; dissimilarity matrix between points in a data set Х proportion of observations that defines neighborhood of a point prop tolerance for zero bandwidth check

#### Value

a vector of bandwidths for each tree (row) in x

#### Author(s)

Ruriko Yoshida < ryoshida @nps.edu>

#### References

Weyenberg, G., Huggins, P., Schardl, C., Howe, D. K., & Yoshida, R. (2014). kdetrees: Nonparametric Estimation of Phylogenetic Tree Distributions. In Bioinformatics.

https://github.com/grady/kdetrees/blob/master/R/bw.R

```
T1<-Sim_Trees15
T2<-Sim_Trees25
D <- rbind(T1, T2[1,])</pre>
M <- pw.trop.dist(D, D)</pre>
bw.nn(M)
```

4 cluster.ratio\_HC

cluster.ratio_HC	Ratio of within and between tropical measures for tropical hierarchi-
	cal clusters

## Description

Ratio of within and between cluster tropical measures for a set hierarchical clusters

## Usage

```
cluster.ratio_HC(A, V, method = mean)
```

## Arguments

A matrix of tropical points; rows are points

V list of clusters where each cluster is defined as a matrix method method to use for within cluster measure; mean or max

## Value

vector of ratios for each cluster

## Author(s)

David Barnhill <david.barnhill@nps.edu>

#### References

David Barnhill, Ruriko Yoshida (2023). Clustering Methods Over the Tropically Convex Sets.

```
har<-rbind(Sim_points[1:20,],Sim_points[51:70,])

V<-Tropical.HC.AGNES(har, method=mean)
inds<-V[[2]][[38]]
cluster.ratio_HC(har,inds,method=mean)</pre>
```

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cluster.ratio\_KM

Ratio of within and between tropical measures for k-means clusters

## **Description**

Ratio of within and between cluster tropical measures for k-means derived clusters

#### Usage

```
cluster.ratio_KM(A, C, method = mean)
```

## Arguments

A matrix of tropical points; rows are points

C number of clusters

method method to use for within cluster measure; mean or max

#### Value

vector of ratios for each cluster

## Author(s)

David Barnhill <david.barnhill@nps.edu>

#### References

David Barnhill, Ruriko Yoshida (2023). Clustering Methods Over the Tropically Convex Sets.

```
hars<-Sim_points
cls<-c(rep(1,50),rep(2,50),rep(3,50))
cl_pt<-cbind(hars,cls)

C<-3
cluster.ratio_KM(cl_pt,C,method=mean)</pre>
```

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convert.to.tree

Create a phylogenetic tree from an ultrametric

## Description

This function constructs a phylogenetic tree from an ultrametric.

#### Usage

```
convert.to.tree(n, L, u)
```

#### **Arguments**

n is the number of leaves

L is a vector of labels (strings) of leaves

u is an ultrametric

## Value

A phylogenetic tree of class phylo

## Author(s)

Ruriko Yoshida < ryoshida @nps.edu>

## **Examples**

```
um<-Sim_Trees21[1,]
ll <- 10
L <- LETTERS[1:10]
tr<-convert.to.tree(ll, L, um)</pre>
```

draw.tpolytope

Draw a 2-D or 3-D tropical polytope

## Description

This command draws a three dimensional tropical polytope

## Usage

```
draw.tpolytope.3d(D, col_lines, col_verts, plot = TRUE, tadd = max)
draw.tpolytope.2d(D, col_lines, col_verts, plot = TRUE, tadd = max)
```

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## Arguments

D matrix of vertices of a tropical polytope; rows are the vertices	ces
--	-----

col\_lines string; color to render the polytope.
col\_verts string; color to render the vertices.

plot logical; initiate new plot visualization or not.

tadd function; max indicates max-plus addition, min indicates min-plus addition. De-

faults to max

#### Value

2-D or 3-D rendering of a tropical polytope.

#### Author(s)

Ruriko Yoshida < ryoshida @nps.edu>

#### **Examples**

```
D <-matrix(c(0,0,0,0,0,1,0,0,0,1,0,0,0,0,1),4,4,TRUE)
col_lines<-'blue'
col_verts<-'red'
draw.tpolytope.3d(D,col_lines,col_verts,plot=TRUE)
draw.tpolytope.3d(D,col_lines,col_verts,plot=TRUE,tadd=min)

D <- matrix(c(0,-2,2,0,-2,5,0,2,1,0,1,-1),4,3,TRUE)
col_lines <- 'blue'
col_verts <- 'red'
draw.tpolytope.2d(D,col_lines,col_verts,plot=TRUE)
draw.tpolytope.2d(D,col_lines,col_verts,plot=TRUE)
draw.tpolytope.2d(D,col_lines,col_verts,plot=TRUE,tadd=min)</pre>
```

FWpoint.num.w.reg

Modified Fermat-Weber point numerical solver for ultrametrics

#### **Description**

Returns a modified Fermat-Weber point of N points using a gradient based numerical method This method is appropriate for points coming from ultrametrics. The algorithm tries to find a point that minimizes the sum of tropical distances from the samples, but also also tries to find a point that is as close as possible to the space of ultrametrics. The tradeoff between these two objectives is controlled by the penalty parameter. If penalty=0, the method is identical to FWpoint\_numerical; it finds the Fermat-Weber point, which may not be an ultrametric. If penalty is very large, the algorithm is trying to find the Fermat-Weber point in the space of ultrametrics.

#### Usage

```
FWpoint.num.w.reg(datamatrix, penalty = 0)
```

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## **Arguments**

datamatrix matrix of dimension N\*e, where N is the number of observations which lie in

R^e

penalty positive real number; the regularization rate

#### Value

vector; Fermat-Weber point approximation

## Author(s)

Georgios Aliatimis < g.aliatimis@lancaster.ac.uk>

#### References

Aliatimis, Georgios, Ruriko Yoshida, Burak Boyaci and James A. Grant (2023). Tropical Logistic Regression on Space of Phylogenetic Trees

## **Examples**

```
D = matrix(c(0,0,0,0,2,5,0,3,1),3,3,TRUE)
FWpoint.num.w.reg(D,1e4) # (0,2,5/3) not ultrametric
FWpoint.num.w.reg(D,1e4) # (0,5/3,5/3) ultrametric
```

FWpoint.numerical

Fermat-Weber point numerical solver

## **Description**

Returns the Fermat-Weber point of N points using a gradient based numerical method

#### Usage

```
FWpoint.numerical(datamatrix)
```

#### **Arguments**

datamatrix

matrix of dimension  $N^*e$ , where N is the number of observations which lie in  $R^*e$ .

#### Value

Fermat-Weber point approximation (vector in R^e)

## Author(s)

Georgios Aliatimis < g.aliatimis@lancaster.ac.uk>

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#### References

Aliatimis, Georgios, Ruriko Yoshida, Burak Boyaci and James A. Grant (2023). Tropical Logistic Regression on Space of Phylogenetic Trees

#### **Examples**

```
D = matrix(c(0,0,0,0,2,5,0,3,1),3,3,TRUE)
FWpoint.numerical(D)
```

HAR.TLineSeg

Uniformly sample from a max-plus tropical line segment

## Description

This function uses a hit-and-run sampler to uniformly sample from a max-plus tropical line segment

## Usage

```
HAR.TLineSeg(D1, D2, tadd = max)
```

#### **Arguments**

D1 point in the tropical projective torus
D2 point in the tropical projective torus

tadd string; max indicates max-plus addition, min indicates min-plus addition. De-

faults to max

#### Value

point on the line segment defined by D1 and D2

#### Author(s)

Ruriko Yoshida < ryoshida@nps.edu>

#### References

Yoshida, Ruriko, Keiji Miura and David Barnhill (2022). Hit and Run Sampling from Tropically Convex Sets.

```
D1 <-c(0,4,2)
D2 <- c(0,7,-1)
HAR.TLineSeg(D1, D2,tadd=max)
HAR.TLineSeg(D1, D2,tadd=min)
```

HAR.TLineSeg.centroid Gaussian-like Sampling on a max- or min-plus tropical line segment

## Description

This function samples points on a tropical line segment about a location parameter for a given scale parameter defined in terms of tropical distance

## Usage

```
HAR.TLineSeg.centroid(D1, D2, m, s, tadd = max)
```

## **Arguments**

D1	point in the tropical projective torus
D2	point in the tropical projective torus
m	location parameter
S	scale parameter
tadd	function; max indicates max-plus addition, min indicates min-plus addition. Defaults to max

## Value

point on the line segment defined by D1 and D2 sampled about mu

## Author(s)

David Barnhill <david.barnhill@nps.edu>

```
D1 <-c(0,4,2)
D2 <- c(0,7,-1)
m<-c(0,7,2)
s<-1
HAR.TLineSeg.centroid(D1, D2,m,s)
HAR.TLineSeg.centroid(D1, D2,m,s,tadd=min)
```

hyper\_3D

hyper 3	3D
---------	----

2D or 3D rendering of max-plus or min-plus tropical hyperplane

## Description

This function renders a 2D or 3D max-plus or min-plus tropical hyperplane

## Usage

```
draw.thyper(D, ext, min.ax, max.ax, plot = FALSE, tadd = max)
```

## Arguments

D	point in the tropical projective torus representing the apex of the hyperplane
ext	scalar; indicates how far the hyperplane should extend
min.ax	scalar; value applied to define the minimum limits of the axes of the plot
max.ax	scalar; value applied to define the maximum limits of the axes of the plot
plot	logical; if true produces a new plot otherwise overlays tropical hyperplane on existing plot
tadd	function; max indicates max-plus addition, min indicates min-plus addition. Defaults to max

## Value

2D or 3D rendering of max-plus or min-plus tropical hyperplane

## Author(s)

David Barnhill <david.barnhill@nps.edu>

```
# 2D Example
D <-c(0,0,0)
ext<-4
min.ax<- 5
max.ax<- 5
draw.thyper(D,ext,min.ax,max.ax,plot=TRUE)

# 3D Example
D <-c(0,0,0,0)
ext<-4
min.ax<- 5
max.ax<- 5
draw.thyper(D,ext,min.ax,max.ax,plot=TRUE)
draw.thyper(D,ext,min.ax,max.ax,plot=TRUE)
draw.thyper(D,ext,min.ax,max.ax,plot=TRUE)</pre>
```

max\_ins.ball

lung\_fish

Phylogenetic trees based on lung fish data

#### **Description**

1290 (non-equidistant) gene trees with 45 leaves originating from lung fish data in matrix form. Also we provide a vector of strings consisting of leaf labels for each species associated with the data set.

#### Usage

lung\_fish

#### **Format**

An object of class matrix (inherits from array) with 1290 rows and 45 columns.

#### Source

Liang D, Shen XX, Zhang P. One thousand two hundred ninety nuclear genes from a genome-wide survey support lungfishes as the sister group of tetrapods. Mol Biol Evol. 2013 Aug;30(8):1803-7. doi: 10.1093/molbev/mst072. Epub 2013 Apr 14. PMID: 23589454.

max\_ins.ball

Calculate the center point and radius of the maximum inscribed ball for a tropical simplex

#### **Description**

This function calculates the center point and radius of the maximum inscribed ball for a max- or min-plus tropical simplex

#### **Usage**

```
max_ins.ball(A, tadd = max)
```

#### **Arguments**

A matrix of points defining a tropical polytope; rows are the points

tadd function; max indicates max-plus addition, min indicates min-plus addition. De-

faults to max

#### Value

list containing the radius and center point of a maximum inscribed ball

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#### Author(s)

David Barnhill <david.barnhill@nps.edu>

#### References

Barnhill, David, Ruriko Yoshida and Keiji Miura (2023). Maximum Inscribed and Minimum Enclosing Tropical Balls of Tropical Polytopes and Applications to Volume Estimation and Uniform Sampling.

## **Examples**

```
P<-matrix(c(0,0,0,0,2,5,0,3,1),3,3,TRUE) max_ins.ball(P) max_ins.ball(P,tadd=min)
```

min\_enc.ball

Calculate a minimum enclosing ball for a tropical polytope

## Description

This function constructs a minimum enclosing ball for a set of points defining a tropical polytope.

#### Usage

```
min_enc.ball(A)
```

## **Arguments**

Α

matrix of points defining a tropical polytope. Rows are the points.

## Value

list containing center point and radius of minimum enclosing ball of P

#### Author(s)

David Barnhill <david.barnhill@nps.edu>

#### References

Barnhill, David, Ruriko Yoshida and Keiji Miura (2023). Maximum Inscribed and Minimum Enclosing Tropical Balls of Tropical Polytopes and Applications to Volume Estimation and Uniform Sampling.

```
P <-matrix(c(0,0,0,0,3,1,0,2,5),3,3,TRUE) min_enc.ball(P)
```

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normaliz.tree

Normalize a phylogenetic tree

## **Description**

This function normalizes the height of a phylogenetic tree

## Usage

```
normaliz.tree(D, h = 1)
```

## Arguments

D numeric vector; ultrametric equidistant tree

h desired height; defaults to 1

#### Value

normalized equidistant tree

## Author(s)

Ruriko Yoshida < ryoshida@nps.edu>

## **Examples**

```
D < -c(4,4,2)
normaliz.tree(D, h=1)
```

normalize

Normalize a point or set of points in the tropical projective torus

## Description

This function normalizes a point or set of points in the tropical projective torus by making the first coordinate zero

## Usage

```
normaliz.vector(D)
normaliz.vectors(D)
normaliz.polytope(D)
normaliz.ultrametrics(D)
```

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## **Arguments**

D

numeric vector in the tropical projective torus or a matrix of points in the tropical projective torus; for matrices, rows are the points

## Value

a single or set of normalized points with the first coordinate zero

#### Author(s)

Ruriko Yoshida < ryoshida @nps.edu>

## **Examples**

```
D <-c(8,4,2)
normaliz.vector(D)

P <-matrix(c(8,4,2,10,1,3,7,2,1),3,3,TRUE)
normaliz.vectors(P)

M<-matrix(c(2,2,2,3,6,4,2,4,7),3,3,TRUE)
normaliz.polytope(M)

M <- Sim_Trees15[1:3,]
normaliz.ultrametrics(M)</pre>
```

over\_bet\_HC

Tropical cluster betweeness measure for each cluster in a set of hierarchical clusters

## **Description**

This function calculates an overall betweenness measure based on tropical distance between a set of clusters derived from tropical hierarchical clustering

## Usage

```
over_bet_HC(A, V)
```

## Arguments

Α	matrix of tropical points; rows are points with the last column representing a
	numbered cluster assignment

V list of clusters defined as matrices derived from agglomerative or divisive hierarchical clustering

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#### Value

vector of betweenness cluster measures

#### Author(s)

David Barnhill <david.barnhill@nps.edu>

#### References

David Barnhill, Ruriko Yoshida (2023). Clustering Methods Over the Tropically Convex Sets.

#### **Examples**

```
har<-rbind(Sim_points[1:20,],Sim_points[51:70,])

V<-Tropical.HC.AGNES(har, method=mean)
inds<-V[[2]][[38]]
over_bet_HC(har,inds)</pre>
```

over\_bet\_KM

Tropical cluster betweeness measure for a each of a set of k-means derived set of clusters

## **Description**

This function calculates an overall betweenness measure between a set of clusters derived from tropical k-means clustering

#### Usage

```
over_bet_KM(A, C)
```

## Arguments

A matrix of tropical points; rows are points with the last column representing a numbered cluster assignment

C number of clusters

#### Value

betweenness cluster measure

#### Author(s)

David Barnhill <david.barnhill@nps.edu>

#### References

David Barnhill, Ruriko Yoshida (2023). Clustering Methods Over the Tropically Convex Sets.

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## **Examples**

```
hars<-Sim_points
cls<-c(rep(1,50),rep(2,50),rep(3,50))
cl_pt<-cbind(hars,cls)

C<-3
over_bet_KM(cl_pt,C)</pre>
```

Points.TLineSeg

Sample k equally spaced points on a max- or min-plus tropical line segment

## Description

This function calculates k equally spaced points on a tropical line segment

## Usage

```
Points.TLineSeg(D1, D2, k = 20, tadd = max)
```

faults to max

## Arguments

D1	point in the tropical projective torus
D2	point in the tropical projective torus
k	number of points
tadd	function; max indicates max-plus addition, min indicates min-plus addition. De-

## Value

matrix of k equally spaced points on a tropical line segment

## Author(s)

Ruriko Yoshida < ryoshida @nps.edu>

```
D1 <-c(0,4,2)
D2 <- c(0,7,-1)
Points.TLineSeg(D1, D2, k = 5)
Points.TLineSeg(D1, D2, k = 5,tadd=min)
```

18 pre.pplot.pro

pre.pplot.pro

Projections of points onto a tropical triangle

## **Description**

This function produces the a matrix of points projected onto a tropical triangle defined by the column space of a matrix

#### Usage

```
pre.pplot.pro(S, D)
```

## **Arguments**

- S matrix of points representing a tropical polytope; rows are the vertices
- D data points in the tropical projective torus

#### Value

matrix of points representing projections of the points in D (row vectors) onto S

## Author(s)

Ruriko Yoshida < ryoshida @nps.edu>

```
s <- 3 #number of vertices. Here it is a tropical triangle
d <- 3 ## dimension
N <- 100 ## sample size
D <- matrix(rep(0, N*d), N, d)
D[, 1] <- rnorm(N, mean = 5, sd = 5)
D[, 2] <- rnorm(N, mean = -5, sd = 5)
D[, 3] <- rnorm(N, mean = 0, sd = 5)

index <- sample(1:N, s)
S <- D[index,]</pre>
DD <- pre.pplot.pro(S, D)
```

prob.class 19

prob.class

Estimated probability for binary class assignment

#### **Description**

Estimates the probability that an observation x belongs to class 1.

#### Usage

```
prob.class(pars, x)
```

#### **Arguments**

pars vector of parameters, which can be decomposed as two normal vectors and two scaling parameters and has dimension 2\*e+2

x vector of dimension e

#### Value

real number

#### Author(s)

Georgios Aliatimis < g.aliatimis@lancaster.ac.uk>

#### References

Aliatimis, Georgios, Ruriko Yoshida, Burak Boyaci and James A. Grant (2023). Tropical Logistic Regression on Space of Phylogenetic Trees

```
library(ROCR)
T0 = Sim_Trees15
T1 = Sim\_Trees25
D = rbind(T0,T1)
Y = c(rep(0,dim(T0)[1]), rep(1,dim(T1)[1]))
N = length(Y)
set.seed(1)
train_set = sample(N,floor(0.8 * N)) ## 80/20 train-test split
pars <- trop.logistic.regression(D[train_set,],Y[train_set], penalty=1e4)</pre>
test_set = (1:N)[-train_set]
Y.hat <- rep(0, length(test_set))</pre>
for(i in 1:length(test_set))
                               Y.hat[i] <- prob.class(pars, D[test_set[i],])</pre>
Logit.ROC <- performance(prediction(Y.hat, Y[test_set]), measure="tpr", x.measure="fpr")
plot(Logit.ROC, lwd = 2, main = "ROC Curve for Logistic Regression Model")
print(paste("Logit.AUC=", performance(prediction(Y.hat, Y[test_set]), measure="auc")@y.values))
```

20 pw.trop.dist

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Project a point on the tropical projective torus onto a tropical polytope

#### **Description**

This function projects points in the tropical projective torus onto a max- or min-plus tropical polytope based on tropical distance

#### Usage

```
project.pi(D_s, D, tadd = max)
```

## Arguments

D\_s matrix where each row is a point defining a tropical polytope

D point to be projected onto D\_s

tadd function; max indicates max-plus addition, min indicates min-plus addition. De-

faults to max.

#### Value

projection of point D onto the tropical polytope defined by D\_s

## Author(s)

David Barnhill <david.barnhill@nps.edu>

## **Examples**

```
D_s <-matrix(c(0,0,0,0,2,5,0,3,1),3,3,TRUE)
D <- c(0,7,-1)
project.pi(D_s,D)
project.pi(D_s,D,tadd=min)</pre>
```

pw.trop.dist

Constructs the dissimilarity matrix for a set of ultrametrics

## **Description**

Constructs the dissimilarity matrix based on the tropical distance between points in a dataset

## Usage

```
pw.trop.dist(D1, D2)
```

rounding 21

## **Arguments**

D1	matrix of ultrametrics
D2	matrix of ultrametrics

## Value

matrix; dissimilarity matrix showing the tropical pairwise distance between each point

#### Author(s)

Ruriko Yoshida < ryoshida@nps.edu>

#### References

Weyenberg, G., Huggins, P., Schardl, C., Howe, D. K., & Yoshida, R. (2014). kdetrees: Nonparametric Estimation of Phylogenetic Tree Distributions. In Bioinformatics.

Yoshida, Ruriko, David Barnhill, Keiji Miura and Daniel Howe (2022). Tropical Density Estimation of Phylogenetic Trees.

https://github.com/grady/kdetrees/blob/master/R/dist.diss.R

## **Examples**

```
T1<-Sim_Trees15
T2<-Sim_Trees25
D <- rbind(T1, T2[1,])
pw.trop.dist(D, D)
```

rounding

Remove all tentacles from a tropical simplex

## **Description**

This function removes all tentacles from a tropical simplex. The remaining portion is a full-dimensional tropical polytope known as the trunk of the tropical polytope.

#### Usage

```
rounding(P)
```

## **Arguments**

Р

matrix of points defining a tropical simplex. Rows are the points

#### Value

matrix of points defining only the full-dimensional element (the trunk) of a tropical polytope; rows are points

22 sigmoid

#### Author(s)

David Barnhill <david.barnhill@nps.edu>

#### References

Barnhill, David, Ruriko Yoshida and Keiji Miura (2023). Maximum Inscribed and Minimum Enclosing Tropical Balls of Tropical Polytopes and Applications to Volume Estimation and Uniform Sampling.

## **Examples**

```
P<-matrix(c(0,-1,1,0,0,0,0,1,-1),3,3,TRUE)

BP<-min_enc.ball(P)

RP<-rounding(P)

BRP<-min_enc.ball(RP)
```

sigmoid

Sigmoid function

## Description

Returns the sigmoid function valuation

## Usage

```
sigmoid(x)
```

## Arguments

Х

real number

## Value

sigmoid function value at x

#### Author(s)

Georgios Aliatimis < g.aliatimis@lancaster.ac.uk>

#### References

Aliatimis, Georgios, Ruriko Yoshida, Burak Boyaci and James A. Grant (2023). Tropical Logistic Regression on Space of Phylogenetic Trees

```
sigmoid(0) # 0.5
```

Sim\_points 23

## Description

150 points generated using Gaussian-like Hit-and-Run sampling with three separate pairs of location and scale parameters

## Usage

Sim\_points

#### **Format**

 $Sim_points A 150 \times 3$  matrix where each row is a point in the: tropical projective torus

Sim_Trees1	Six data sets of phylogenetic trees data simulated from the Coalescant model.
------------	---

## Description

Six data sets of 1000 gene trees simulated from the Coalescant model based on a specified species with each data set possessing a ratio of species depth to effective population of 0.25, 0.5, 1, 2, 5, and 10.

## Usage

Sim\_Trees1025
Sim\_Trees105
Sim\_Trees11
Sim\_Trees12
Sim\_Trees15
Sim\_Trees110

Sim\_Trees2

#### **Format**

An object of class matrix (inherits from array) with 1000 rows and 45 columns. An object of class matrix (inherits from array) with 1000 rows and 45 columns. An object of class matrix (inherits from array) with 1000 rows and 45 columns. An object of class matrix (inherits from array) with 1000 rows and 45 columns. An object of class matrix (inherits from array) with 1000 rows and 45 columns. An object of class matrix (inherits from array) with 1000 rows and 45 columns.

Sim\_Trees2

Six data sets of phylogenetic trees data simulated from the Coalescant model.

## Description

Six data sets of 1000 gene trees simulated from the Coalescant model based on a specified species with each data set possessing a ratio of species depth to effective population of 0.25, 0.5, 1, 2, 5, and 10.

#### Usage

Sim\_Trees2025

Sim\_Trees205

Sim\_Trees21

Sim\_Trees22

Sim\_Trees25

Sim\_Trees210

#### **Format**

An object of class matrix (inherits from array) with 1000 rows and 45 columns. An object of class matrix (inherits from array) with 1000 rows and 45 columns. An object of class matrix (inherits from array) with 1000 rows and 45 columns. An object of class matrix (inherits from array) with 1000 rows and 45 columns. An object of class matrix (inherits from array) with 1000 rows and 45 columns. An object of class matrix (inherits from array) with 1000 rows and 45 columns.

tdets 25

tdets

Calculate the tropical determinant of a square matrix.

#### **Description**

This function calculates the tropical determinant (or singularity) of a square matrix

#### Usage

```
tdets(P, tadd = max)
```

## **Arguments**

P matrix of points defining a tropical polytope. Rows are the points

tadd function; max indicates max-plus addition, min indicates min-plus addition. De-

faults to max

## Value

list containing the value of the determinant and reordered matrix P

#### Author(s)

David Barnhill <david.barnhill@nps.edu>

#### **Examples**

```
P<-matrix(c(\emptyset,\emptyset,\emptyset,\emptyset,2,5,\emptyset,3,1),3,3,TRUE) tdets(P) tdets(P,tadd=min)
```

**TKmeans** 

K-means clustering over the tropical projective torus

## **Description**

This function performs k-means clustering over the tropical projective torus

## Usage

```
TKmeans(A, C, M)
```

#### Arguments

A matrix of points defining a tropical polytope; rows are the tropical points

C number of clusters

M maximum number of iterations of algorithm to find cluster centroids

TLineSeg

#### Value

list with matrix of observation classified by centroid; matrix of centroid coordinates; number of iterations used

## Author(s)

David Barnhill <david.barnhill@nps.edu>

#### References

David Barnhill, Ruriko Yoshida (2023). Clustering Methods Over the Tropically Convex Sets.

## **Examples**

```
P <-Sim_points
C<-3
M<-10
res<-TKmeans(P,C,M)
try<-res[[1]]
cen<-res[[2]]
plot(try[,2],try[,3],col=try[,4],asp=1)
plot(try[,2],try[,3],col=try[,4],asp=1,xlab='x2',ylab='x3')
points(cen[,2],cen[,3],col=c('purple','hotpink','orange'),pch=19)</pre>
```

TLineSeg

Construct a max- or min-plus tropical line segment between two points

## Description

This function constructs a max-plus tropical line segment between two points

## Usage

```
TLineSeg(D1, D2, tadd = max)
```

#### **Arguments**

D1 point in the tropical projective torus
D2 point in the tropical projective torus

tadd function; max indicates max-plus addition, min indicates min-plus addition. De-

faults to max

#### Value

list of points defining the tropical line segment

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#### Author(s)

Ruriko Yoshida < ryoshida @nps.edu>

#### **Examples**

```
D1 <-c(0,4,2)

D2 <- c(0,7,-1)

TLineSeg(D1, D2)

TLineSeg(D1, D2,tadd=min)
```

TML

Tropical Machine Learning in R

#### **Description**

TML provides a suite of tools for machine learning application on data over the tropical semiring

tree.to.vector

Phylogenetic tree to vector

#### **Description**

A tree is converted to a vector of pairwise distances between leaves. Distance between leaves is defined as the cophenetic distance between them. Normalization is applied so that the maximum distance in the vector output is 1.

#### Usage

```
tree.to.vector(tree, normalization = TRUE)
```

## Arguments

tree phylogenetic tree

normalization logical; normalize the tree if TRUE

## Value

vector of pairwise distances in R^(m choose 2), where m is the number of leaves

## Author(s)

Georgios Aliatimis < g.aliatimis@lancaster.ac.uk>

#### References

Aliatimis, Georgios, Ruriko Yoshida, Burak Boyaci, James A. Grant (2023). Tropical Logistic Regression on Space of Phylogenetic Trees

28 trop.bal.vert

#### **Examples**

```
tree <- ape::read.tree(text='((A:1, B:1):2, (C:1.5, D:1.5):1.5);')
tree.to.vector(tree)

trop.bal.vert

Calculate the minimum or entire generating vertex set of a tropical ball using a max- or min-plus algebra
```

## **Description**

This function calculates the coordinates of the minimum or entire vertex set of a tropical ball in terms of either a max- or min-plus algebra for a given a center point

## Usage

```
trop.bal.vert(x, d, tadd = max)
trop.bal.all_vert(x, d)
```

## **Arguments**

X	matrix where each row is a point defining a tropical polytope
d	radius of the tropical ball in terms of tropical distance
tadd	function; max indicates max-plus addition, min indicates min-plus addition, 'all'
	indicates all vertices. Defaults to 'max'

## Value

matrix of normalized tropical points defining the tropical ball. Rows are the points

## Author(s)

David Barnhill <david.barnhill@nps.edu>

#### References

Barnhill, David, Ruriko Yoshida and Keiji Miura (2023). Maximum Inscribed and Minimum Enclosing Tropical Balls of Tropical Polytopes and Applications to Volume Estimation and Uniform Sampling.

```
x <-c(0,3,7,5)
d <- 2
trop.bal.vert(x,d)
trop.bal.vert(x,d,tadd=min)
trop.bal.all_vert(x,d)</pre>
```

trop.dist 29

trop.dist

Compute the tropical distance

## Description

This function computes the tropical distance between two points in the tropical projective torus

## Usage

```
trop.dist(D1, D2)
```

## **Arguments**

D1 point in the tropical projective torus
D2 point in the tropical projective torus

## Value

tropical distance between D1 and D2

## Author(s)

Ruriko Yoshida < ryoshida @nps.edu>

## **Examples**

```
D1 <-c(0,4,2)
D2 <- c(0,7,-1)
trop.dist(D1, D2)
```

trop.FW

Calculate the tropical Fermat-Weber point

## Description

This function calculates the Fermat-Weber point for a tropical polytope

## Usage

```
trop.FW(A)
```

## Arguments

Α

matrix with normalized tropical points as rows

30 trop.hyper.dist

#### Value

numeric vector providing the tropical Fermat-Weber point for the tropical polytope

## Author(s)

David Barnhill <david.barnhill@nps.edu>

#### References

Lin, Bo and Ruriko Yoshida (2016). Tropical Fermat-Weber Points. SIAM J. Discret. Math. 32: 1229-1245.

## **Examples**

```
P <-matrix(c(0,0,0,0,2,5,0,3,1),3,3,TRUE)
trop.FW(P)
```

trop.hyper.dist

Calculate the tropical distance to a max-tropical hyperplane

## Description

Calculate the tropical distance to a max-tropical hyperplane

## Usage

```
trop.hyper.dist(0, x0, tadd = max)
```

#### **Arguments**

0 normal vector of a tropical hyperplane; numeric vector

x0 point of interest; numeric vector

tadd function; max indicates max-plus addition, min indicates min-plus addition. De-

faults to max

## Value

tropical distance to max-plus tropical hyperplane

#### Author(s)

David Barnhill <david.barnhill@nps.edu>

trop.logistic.regression 31

#### **Examples**

```
0 <-c(0,-1,-1)
x0 <- c(0,-2,-8)
trop.hyper.dist(0,x0)
trop.hyper.dist(0,x0,tadd=min)</pre>
```

```
trop.logistic.regression
```

Tropical Logistic Regression

## **Description**

Performs tropical logistic regression, by finding the optimal statistical parameters for the training dataset (D,Y), where D is the matrix of covariates and Y is the binary response vector

## Usage

```
trop.logistic.regression(D, Y, penalty = 0, model_type = "two_species")
```

#### **Arguments**

D	matrix of dimension N*e, where N is the number of observations which lie in
---	---

R^e

Y binary vector with dimension N, with each component corresponding to an ob-

servation

penalty scalar; positive real number

model\_type string; options are "two-species" (default), "one-species", "general"

## Value

vector; optimal model parameters (two normal vectors and two scaling factors)

#### Author(s)

Georgios Aliatimis < g.aliatimis@lancaster.ac.uk>

#### References

Aliatimis, Georgios, Ruriko Yoshida, Burak Boyaci and James A. Grant (2023). Tropical Logistic Regression on Space of Phylogenetic Trees.

32 trop.tri.plot.w.pts

#### **Examples**

#### Description

This function conducts tropical PCA to find the best fit tropical triangle given data defined in the tropical projective torus. It employs the vertex HAR with extrapolation sampler to sample points to determine the vertices of the tropical triangle.

#### Usage

```
trop.tri.plot.w.pts(S, D)
```

#### **Arguments**

- S inital set of vertices for the tropical triangle
- D matrix of data where each row is an observation in the tropical projective torus

#### Value

rendering of tropical triangle saved to current directory

#### Author(s)

Ruriko Yoshida < ryoshida@nps.edu>

trop. Volume 33

#### **Examples**

```
s <- 3 #number of vertices. Here it is a tropical triangle
d <- 3 ## dimension
N <- 100 ## sample size
V <- matrix(c(100, 0, 0, 0, 100, 0, 0, 0, 100, -100, 0, 0, 0, -100, 0, 0, 0, -100), 6, 3, TRUE)
D <- matrix(rep(0, N*d), N, d)
D[, 1] <- rnorm(N, mean = 5, sd = 5)
D[, 2] <- rnorm(N, mean = -5, sd = 5)
D[, 3] <- rnorm(N, mean = 0, sd = 5)
index <- sample(1:N, s)
S <- D[index,]
res <- tropical.PCA.Polytope(S, D, V, I = 1000,50)
DD <- pre.pplot.pro(res[[2]], res[[3]])
trop.tri.plot.w.pts(normaliz.ultrametrics(res[[2]]), DD)</pre>
```

trop.Volume

Estimate the volume of a tropical polytope

## **Description**

This function uses tropical HAR with a uniform target distribution to estimate the volume of a tropical polytope

## Usage

```
trop.Volume(B, P, x0, s, I, r)
```

#### **Arguments**

В	matrix of points defining a minimum enclosing ball for a polytope P; rows are the points
Р	matrix of points defining a tropical polytope; rows are the points
x0	initial point used for the HAR sampler
S	number of points to sample from the minimum enclosing ball
I	number of iterations for the HAR sampler
r	radius of the minimum enclosing tropical ball

#### Value

list containing ratio of points falling in P; volume of the tropical ball; volume estimate of P

#### Author(s)

David Barnhill <david.barnhill@nps.edu>

34 tropical.centroid

#### References

Barnhill, David, Ruriko Yoshida and Keiji Miura (2023). Maximum Inscribed and Minimum Enclosing Tropical Balls of Tropical Polytopes and Applications to Volume Estimation and Uniform Sampling.

## **Examples**

```
P <-matrix(c(0,0,0,0,3,1,0,2,5),3,3,TRUE)
BR<-min_enc.ball(P)
B<-trop.bal.vert(BR[[1]],BR[[2]])
x0<-c(0,1.5,.4)
S<-200
I<-50
R<-BR[[2]]
trop.Volume(B,P,x0,S,I,R)</pre>
```

tropical.centroid

Tropical centroid-based sampling about a center of mass

## **Description**

This function is a centroid-based HAR sampler about a center of mass denoted by a location parameter with scale parameter in terms of the tropical distance

## Usage

```
VE.HAR.centroid(D_s, x0, I = 1, m, s, tadd = max)
```

## **Arguments**

D_s	matrix of vertices of a tropical simplex; each row is a vertex
x0	initial point for sampler, numeric vector
I	number of states in Markov chain
m	location parameter; numeric vector indicating centroid
S	scale parameter; in terms of tropical distance
tadd	function; max indicates max-plus addition, min indicates min-plus addition. Defaults to max

## Value

next sampled point from the tropical polytope

## Author(s)

David Barnhill <david.barnhill@nps.edu>

tropical.centroid.MH 35

#### **Examples**

tropical.centroid.MH Centroid-based sampling using Metropolis filter

## Description

This function samples points on a tropical line segment about a location parameter for a given scale parameter defined in terms of tropical distance

## Usage

```
trop.centroid.MH(D, x0, m, s, n, I = 50, tadd = max) trop.centroid.MH.square(D, x0, m, s, n, I = 50, tadd = max)
```

#### **Arguments**

D	matrix of vertices of a tropical polytope; each row is a vertex
x0	initial point for sampler, numeric vector
m	location parameter; numeric vector
S	scale parameter; scalar
n	number of points to sample
I	states in Markov chain
tadd	function; max indicates max-plus addition, min indicates min-plus addition. Defaults to max

## Value

matrix of n sampled points where each point is a row

## Author(s)

David Barnhill <david.barnhill@nps.edu>

#### References

Yoshida, Ruriko, Keiji Miura and David Barnhill (2022). Hit and Run Sampling from Tropically Convex Sets.

Tropical.HC.AGNES

#### **Examples**

```
D1 <-matrix(c(0,0,0,0,10,0,0,10),3,3,TRUE)
D2 <-matrix(c(0,10,10,0,10,0,0,10),3,3,TRUE)
x0 <- c(0,0,0)
m1<-c(0,5,5)
m2<-c(0,-1,1)
s<-1
n<-10
trop.centroid.MH(D1, x0, m1, s, n, I=50)
trop.centroid.MH(D2, x0, m1, s, n, I=50,tadd=min)
trop.centroid.MH.square(D2, x0,m2, s, n, I=50,tadd=min)
```

Tropical.HC.AGNES

Agglomerative (AGNES) tropical hierarchical clustering

#### **Description**

This function performs agglomerative (AGNES) hierarchical clustering over the space of ultrametrics defining the space of equidistant trees

## Usage

```
Tropical.HC.AGNES(D, method = mean)
```

#### **Arguments**

D matrix of points defining a tropical polytope. Rows are the tropical points

method linkage method: mean, min, or max

#### Value

list of distances in when merges occur; list of indices of points in each cluster

#### Author(s)

David Barnhill <david.barnhill@nps.edu>

## References

David Barnhill, Ruriko Yoshida (2023). Clustering Methods Over the Tropically Convex Sets.

```
P <-Sim_points
Tropical.HC.AGNES(P, method=mean)
```

tropical.KDE 37

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+	ron	10	٦I	KDE
L	ιυυ	160	<b>11.</b>	NDE

Tropical Kernel Density Estimation of Phylogenetic Trees

## **Description**

This function calculates a non-parametric density estimate of a tree over the space of phylogenetic trees on m leaves. It mimics classical kernel density estimation by using a Gaussian kernel in conjunction with tropical distance.

#### Usage

```
tropical.KDE(D, n, sigma, h = 2)
```

#### **Arguments**

D matrix of phylogenetic tree observations as ultrametrics	D	matrix of phylogen	etic tree observation	ons as ultrametrics
--	---	--------------------	-----------------------	---------------------

n number of leaves for each tree

sigma bandwidth parameter based on tropical distance

h height of the tree

#### Value

list containing center point and radius of minimum enclosing ball of P

#### Author(s)

Ruriko Yoshida < ryoshida@nps.edu>

## References

Weyenberg, G., Huggins, P., Schardl, C., Howe, D. K., & Yoshida, R. (2014). kdetrees: Nonparametric Estimation of Phylogenetic Tree Distributions. In Bioinformatics.

Yoshida, Ruriko, David Barnhill, Keiji Miura and Daniel Howe (2022). Tropical Density Estimation of Phylogenetic Trees.

```
T1<-Sim_Trees15
T2<-Sim_Trees25
D <- rbind(T1, T2[1,])
T <- dim(D)[1]
X <- 1:T
M <- pw.trop.dist(D, D)
sigma <- bw.nn(M)
P_5 <- tropical.KDE(D, n, sigma, h = 2)
Q5 <- P_5[T]</pre>
```

38 tropical.PCA

tropical.PCA	Tropical principal component analysis (PCA) on over tropical projective torus
tropical.PCA	

#### **Description**

This function conducts tropical PCA to find the best fit tropical triangle given data defined in the tropical projective torus. It employs the vertex HAR with extrapolation sampler to sample points to determine the vertices of the tropical triangle.

#### Usage

```
tropical.PCA.Polytope(S, D, V, I = 1, k)
```

## Arguments

S	inital set of vertices for the tropical triangle
D	matrix of data where each row is an observation in the tropical projective torus
٧	matrix of vertices defining a polytope encompassing D
I	number of iterations to perform
k	number of iterations for the HAR sampler

#### Value

list with the sum of residuals

#### Author(s)

Ruriko Yoshida < ryoshida @nps.edu>

#### References

Page, Robert and others (2020), Tropical principal component analysis on the space of phylogenetic trees, Bioinformatics, Volume 36, Issue 17, Pages 4590–4598.

Yoshida, R., Zhang, L. & Zhang, X (2019). Tropical Principal Component Analysis and Its Application to Phylogenetics. Bull Math Biol 81, 568–597.

```
s <- 3 #number of vertices. Here it is a tropical triangle
d <- 3 ## dimension
N <- 100 ## sample size
V <- matrix(c(100, 0, 0, 0, 100, 0, 0, 100, -100, 0, 0, 0, -100, 0, 0, 0, -100), 6, 3, TRUE)
D <- matrix(rep(0, N*d), N, d)
D[, 1] <- rnorm(N, mean = 5, sd = 5)
D[, 2] <- rnorm(N, mean = -5, sd = 5)
D[, 3] <- rnorm(N, mean = 0, sd = 5)</pre>
```

Trop\_ball 39

```
index <- sample(1:N, s)
S <- D[index,]
DD <- pre.pplot.pro(S, D)
for(i in 1:N)
   DD[i, ] <- normaliz.vector(DD[i, ])

res <- tropical.PCA.Polytope(S, D, V, I = 1000,50)
DD <- pre.pplot.pro(res[[2]], res[[3]])
trop.tri.plot.w.pts(normaliz.ultrametrics(res[[2]]), DD)</pre>
```

Trop\_ball

Visualize a Tropical ball in 2D or 3D

## Description

This function constructs a visualization of a 2D or 3D tropical ball

## Usage

```
Trop_ball(
    v,
    d,
    a = 1,
    cls = "black",
    cent.col = "black",
    fil = TRUE,
    plt = TRUE,
    bord = "black"
)
```

## Arguments

V	center of tropical ball; numeric vector of length 3 or 4
d	radius of tropical ball
a	shading level; 1 is opaque
cls	string indicating color of interior of ball
cent.col	string indicating color of center point
fil	logical for 3D plots; if TRUE 2D facets of 3D ball fill in color of cls parameter
plt	logical; indicates plot a new object; defaults to TRUE; if FALSE, overlays the ball on existing plot
bord	string indicating color of border of ball (only for 2D plots)

## Value

2D or 3D visualization of tropical ball

40 trop\_wi\_dist

#### Author(s)

David Barnhill <david.barnhill@nps.edu>

#### **Examples**

```
v <-c(0,0,0)
d <- 2
Trop_ball(v,d,a=.1,cls='white',cent.col='black',fil=TRUE,plt=TRUE,bord='black')
v <-c(0,0,0,0)
d <- 2
Trop_ball(v,d,a=1,cls='red',cent.col='black',fil=FALSE,plt=TRUE)</pre>
```

trop\_wi\_dist

Tropical within-cluster measure

#### **Description**

This function calculates a within cluster measure by measuring the pairwise tropical distance between points in the cluster.

## Usage

```
trop_wi_dist(D1, method = mean)
```

## **Arguments**

D1 matrix of tropical points; rows are points

method function; metric to measure; mean is the average pairwise tropical distance; max

is the maximum pairwise tropical distance

#### Value

within cluster measure

#### Author(s)

David Barnhill <david.barnhill@nps.edu>

#### References

David Barnhill, Ruriko Yoshida (2023). Clustering Methods Over the Tropically Convex Sets.

```
D<-Sim_points
avg.m<-trop_wi_dist(D, method=mean)
max.m<-trop_wi_dist(D, method=max)</pre>
```

Ultrametrics.HAR 41

Ultrametrics.HAR

Hit-and-Run Sampler for the space of ultrametrics

## Description

This sampler samples a point in the space of ultrametrics where each point represents an equidistant tree on n leaves

## Usage

```
Ultrametrics. HAR(x0, n, I = 1, h = 1)
```

## Arguments

x0	an equidistant tree defined as ultrametric
n	number of leaves for the equidistant tree
I	number of states in the Markov chain
h	height of phylogenetic tree

## Value

point in the space of ultrametrics over n leaves

## Author(s)

Ruriko Yoshida < ryoshida @nps.edu>

## References

Yoshida, Ruriko, Keiji Miura and David Barnhill (2022). Hit and Run Sampling from Tropically Convex Sets.

```
x0 <-Sim_Trees15[1,]
n<-10
Ultrametrics.HAR(x0, n, I = 50, h = 1)</pre>
```

VE.HAR

VE.HAR	Vertex HAR with extrapolation (VHE) MCMC with uniform target distribution

## Description

This function samples points uniformly the space defined by a tropical simplex

## Usage

```
VE.HAR(D_s, x0, I = 1, tadd = max)
```

## Arguments

D_s	matrix of vertices of a tropical simplex; each row is a vertex
x0	initial point for sampler, numeric vector
I	number of states in Markov chain
tadd	function; max indicates max-plus addition, min indicates min-plus addition. Defaults to max

## Value

next sampled point from the tropical polytope

## Author(s)

David Barnhill <david.barnhill@nps.edu>

## References

Yoshida, Ruriko, Keiji Miura and David Barnhill (2022). Hit and Run Sampling from Tropically Convex Sets.

vector.to.equidistant.tree 43

```
vector.to.equidistant.tree
```

Vector to equidistant tree

## **Description**

A vector of pairwise distances is used to reconstruct the corresponding equidistant tree

## Usage

```
vector.to.equidistant.tree(vec)
```

## Arguments

vec

vector of pairwise distances in R^(m choose 2), where m is the number of leaves

#### Value

equidistant phylogenetic tree

## Author(s)

Georgios Aliatimis < g.aliatimis@lancaster.ac.uk>

## References

Aliatimis, Georgios, Ruriko Yoshida, Burak Boyaci and James A. Grant (2023). Tropical Logistic Regression on Space of Phylogenetic Trees

```
vec = c(1/3,1,1,1,1,1/3)
tree = vector.to.equidistant.tree(vec)
plot(tree)
```

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