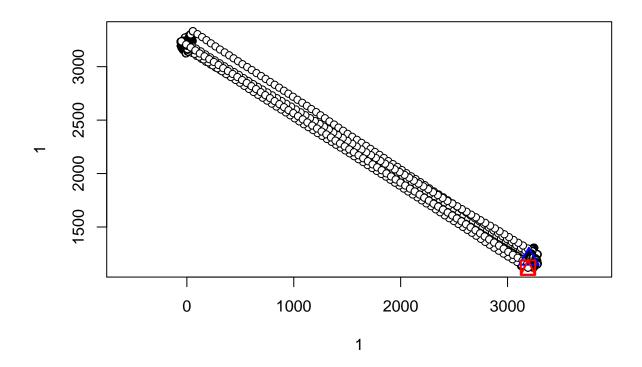
Vignette moveNT

Guillaume Bastille-Rousseau February 26, 2020

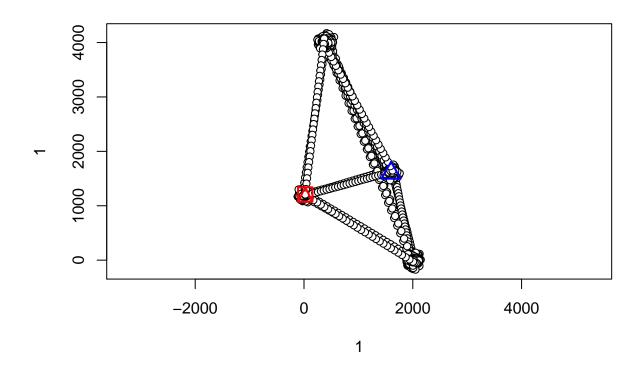
Simulating movement strategies - sim_mov

The function sim_mov generates movement trajectories including patches and movement between patches. Movement within patches can follow an Ornstein-Uhlenbeck process (based on simm.mou function from package adehabitatLT) or two-states movement model (based on simmData function from package moveHMM). Movement between patches is following a brownian bridge movement model (based on simm.bb function from package adehabitatLT). Generated outputs are of the class ltraj from package adehabitatlt.

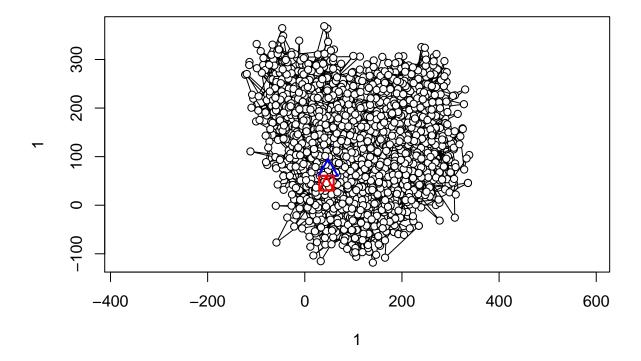
```
# Simulating migration with two-states model
mig<-sim_mov(type="2states", npatches=2, ratio=2, nswitch=25, ncore=150, grph=F)
mig
##
  ****** List of class ltraj *******
##
## Type of the traject: Type II (time recorded)
## * Time zone: GMT *
## Regular traject. Time lag between two locs: 1 seconds
##
## Characteristics of the bursts:
##
     id burst nb.reloc NAs
                                    date.begin
                                                          date.end
                  4350
                         0 1960-01-01 00:00:01 1960-01-01 01:12:30
## 1 id
           id
##
##
   infolocs provided. The following variables are available:
## [1] "out.Corri"
head(ld(mig))
##
                                      date
                                                      dx
                                                                    dy
            Х
## 1 3200.000 1200.000 1960-01-01 00:00:01 -0.0161782779 -0.0062107107
## 2 3199.984 1199.994 1960-01-01 00:00:02 -0.1817510724 -0.8524269926
## 3 3199.802 1199.141 1960-01-01 00:00:03 -6.6821985199 -7.8748645107
## 4 3193.120 1191.266 1960-01-01 00:00:04 -0.1277042358
                                                          0.6697123914
## 5 3192.992 1191.936 1960-01-01 00:00:05 -0.0007959994
## 6 3192.991 1191.937 1960-01-01 00:00:06 0.0001277452 0.0002266126
##
             dist dt
                              R2n abs.angle rel.angle id burst out.Corri
## 1 1.732944e-02 1 0.000000e+00 -2.775049
                                                    NA id
                                                             id
                                                                        2
## 2 8.715878e-01 1 3.003096e-04 -1.780867
                                            0.9941825 id
                                                             id
                                                                        2
                                                             id
                                                                        2
## 3 1.032789e+01 1 7.764347e-01 -2.274447 -0.4935800 id
## 4 6.817793e-01 1 1.236102e+02 1.759220 -2.2495191 id
                                                                        2
                                                                        2
## 5 1.312990e-03 1 1.141344e+02 2.222132 0.4629128 id
                                                             id
## 6 2.601387e-04 1 1.141287e+02 1.057484 -1.1646488 id
plot(mig)
```



Simulating multi-patches movement with Ornstein-Uhlenbeck process
patches<-sim_mov(nswitch=25, ncore=150, ratio=5, type="OU", npatches=5, grph=T)</pre>



Simulating sedentary movement
seden<-sim_mov(type="0U", npatches=10, spacecore=12, ratio=3, nswitch=150, ncore=20, grph=T)</pre>

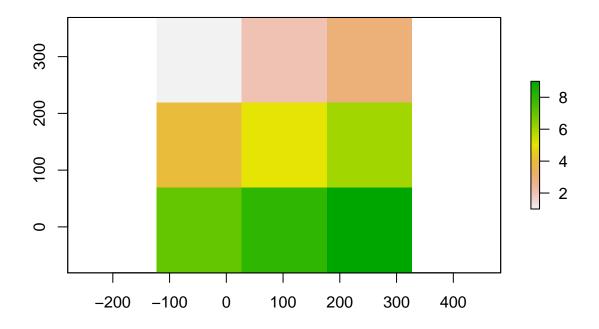


Converting movement to adjacency matrix - traj2adj

The function traj2adj converts a trajectory object of class ltraj to an adjacency matrix. This is done by overlapping a grid over the relocation data and tallying the number of transitions among each pixel. Users need to specify the grid size, which can be based on distance travelled. The function quant is a wrapper that allows to estimate quantiles of step length distribution from a ltraj object. Output produced by traj2adj is a list containing the adjacency matrix, the grid used (raster format), and a raster indicating pixel numbers that are occupied. These rasters are used by other functions such as adj2stack and clustnet.

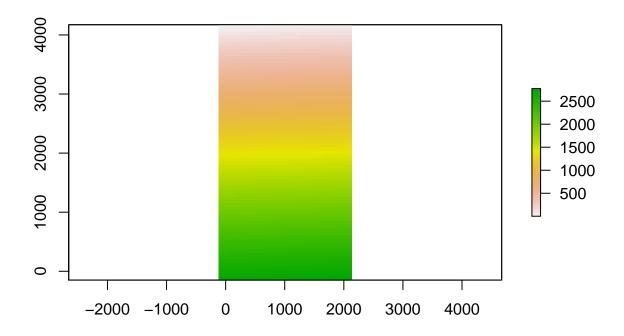
```
# Using sedentary movement and user specific grid-size
adj_seden<-traj2adj(seden, res=150) #Pixel size of 150m
adj_seden[[1]] # Adjency matrix</pre>
```

```
##
           [,1] [,2]
                       [,3] [,4] [,5] [,6] [,7]
                                                       [,8]
##
     [1,]
            138
                    25
                           0
                                46
                                       10
                                              0
                                                    0
                                                           0
                                                                 0
                                 7
                                                           0
     [2,]
             25
                    73
                          18
                                       31
                                             15
                                                    0
                                                                 0
##
                          75
##
     [3,]
              0
                    12
                                 0
                                       13
                                             62
                                                    0
                                                           0
                                                                 0
                                                   10
##
     [4,]
             48
                    10
                           0
                               183
                                       74
                                              0
                                                         13
                                                                 0
##
     [5,]
               8
                    25
                          14
                                81
                                     535
                                            93
                                                   14
                                                        133
                                                                 7
                                                    0
##
     [6,]
               0
                    24
                          56
                                 0
                                       99
                                           440
                                                           7
                                                                41
               0
                                                   36
##
     [7,]
                     0
                           0
                                11
                                       17
                                              0
                                                         49
                                                                 0
##
     [8,]
               0
                     0
                           0
                                10
                                     122
                                             15
                                                   52
                                                        783
                                                                37
##
     [9,]
               0
                     0
                           0
                                 0
                                        8
                                             40
                                                    0
                                                         38
                                                                71
```

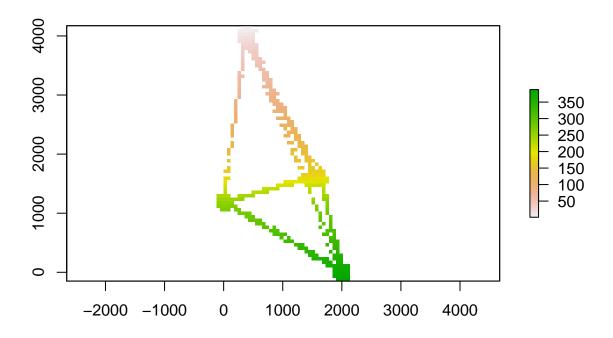


Using multi-patches movement and median distance travelled
adj_patches<-traj2adj(patches, res=quant(patches, p=0.5)) #Grid size based on median
dim(adj_patches[[1]]) # Size of the adjacency matrix</pre>

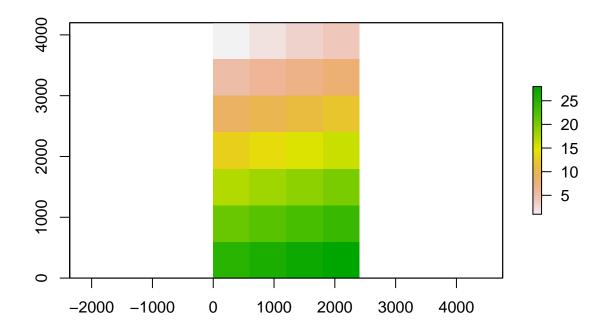
[1] 388 388
plot(adj_patches[[2]]) #Plot grid used



plot(adj_patches[[3]]) #Plot occupied pixels



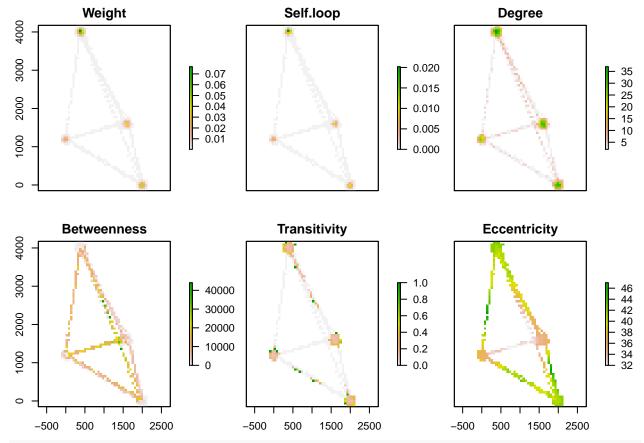
Using user defined grid
ras<-raster(nrows=10, ncols=10, xmn=0, ymn=0, xmx=6000, ymx=6000)
adj_patches2<-traj2adj(patches, res=quant(patches, p=0.5), grid=ras) #Grid size based on median
plot(adj_patches2[[2]]) #Crop version of the grid created</pre>



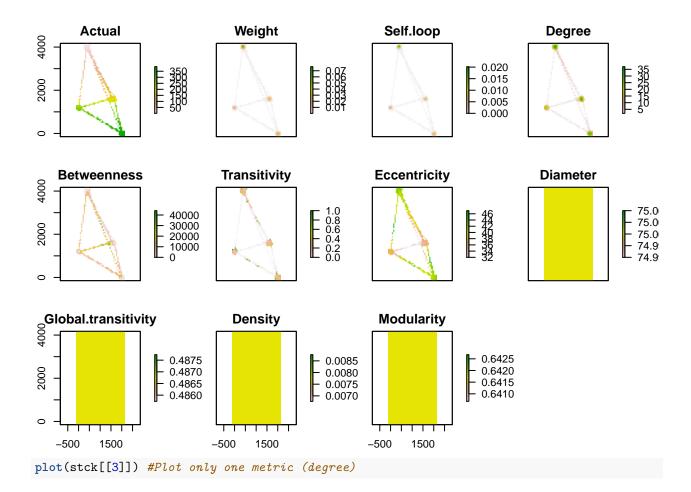
Calculation of network metrics - adj2stack

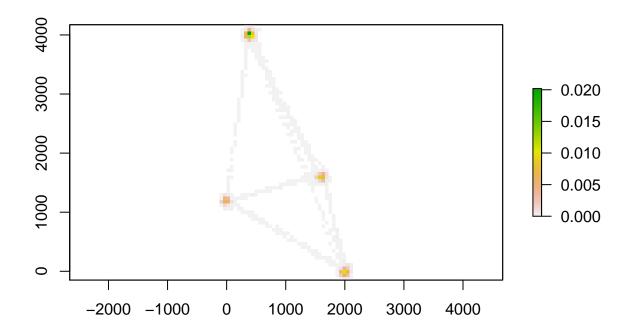
The function adj2stack takes the output of function traj2adj and calculates a series of node- and graph-level metrics. Each metric is stored as a individual raster and the output is a raster stack combining each metric. Graph-level metrics are also stored as a raster, each containing an unique value. The function graphmet extracts graph-level metrics. The function val extracts only the occupied cells (remove NA) in a raster and allows the calculation of statistics from node-level metrics.

Using multi-patches movement and median distance travelled stck<-adj2stack(adj_patches,grph=T) #Plot the node-level metrics at the same time



plot(stck) #Plot also the graph-level metrics (not really useful)





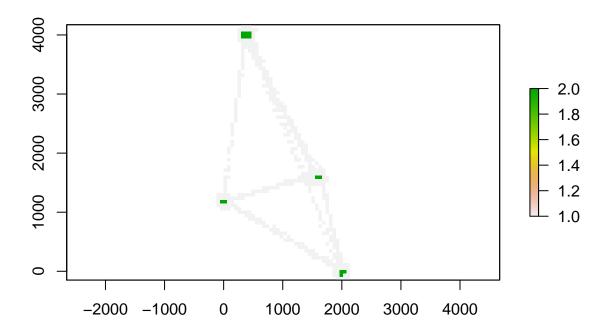
```
## Diameter Global.transitivity Density
## 75.000000000 0.486735246 0.007765257
## Modularity
## 0.641678214

cv(val(stck, 4)) #Extract coefficient of variation of node-level betweenness.
## [1] 119.7042
```

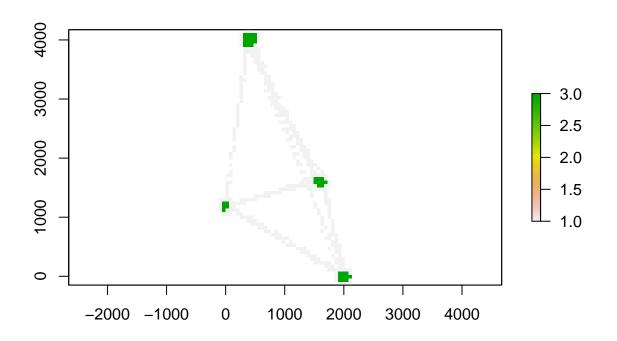
Clustering of node level metrics - clustnet

The function clustnet applies a normal mixture model to node-level metrics in order to cluster them into separate groups (default = 2). The function takes the output of function adj2stack with the user specifying the metric to cluster and the number of groups. Return a list containing output of function Mclust from package mclust and a raster displaying classification.

```
# Using multi-patches movement and median distance travelled
clust2<-clustnet(stck, id=3, nclust=2) # Clustering of degree in two groups</pre>
```

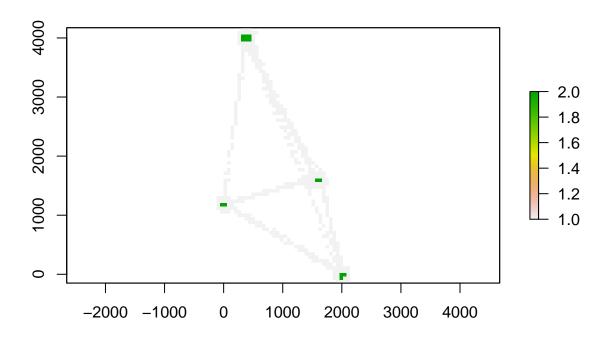


clust3<-clustnet(stck, id=4, nclust=3) #Clustering of betweenness in three groups</pre>



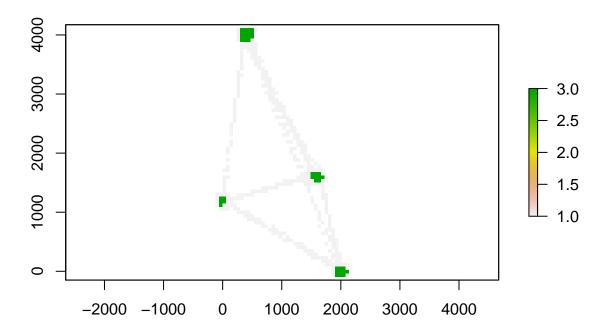
summary(clust2[[1]])

```
## Gaussian finite mixture model fitted by EM algorithm
##
## Mclust E (univariate, equal variance) model with 2 components:
##
   log-likelihood n df
                              BIC
##
                                       ICL
##
         2091.913 388 4 4159.982 4159.547
##
## Clustering table:
##
    1
## 375 13
plot(clust2[[2]])
```



summary(clust3[[1]])

```
## Gaussian finite mixture model fitted by EM algorithm
##
## Mclust E (univariate, equal variance) model with 3 components:
##
   log-likelihood n df
                               BIC
##
                                         ICL
##
        -1130.556 388 6 -2296.877 -2637.057
##
## Clustering table:
##
        2 3
## 349
       0 39
plot(clust3[[2]])
```

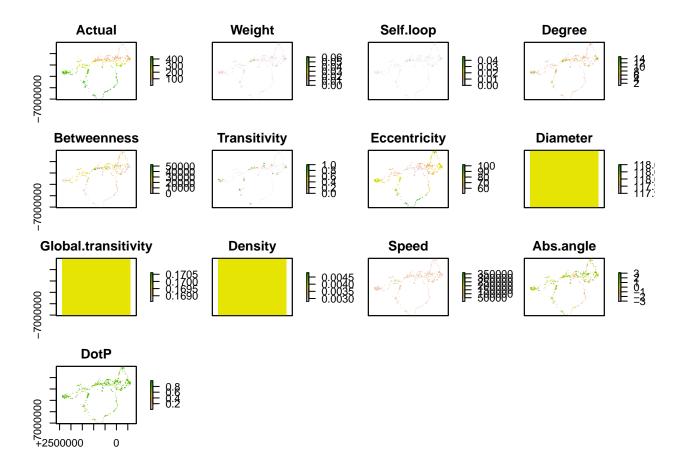


Looping over all individuals loop

The function *loop* is a wrapper of *traj2adj* and *adj2stack* applied to all individuals within a trajectory. The function will keep the same grid for all individuals. The user simply need to specify the trajectory object and the grid size. The loop function also adds additional movement properties regarding speed, absolute angle, and turning angle.

```
data(albatross) #Load a traj object from adehabitatLT
out1<-loop(albatross, res=35000)

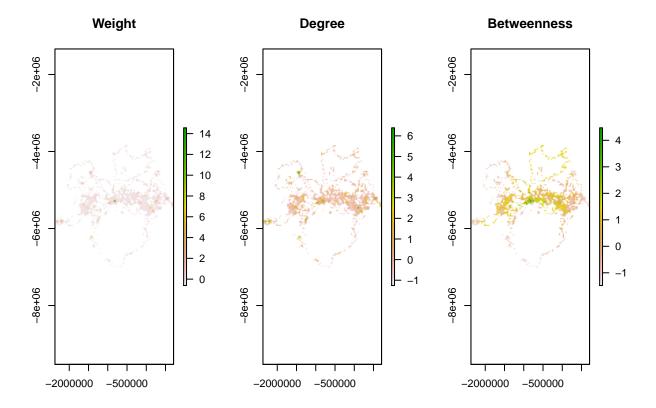
## balise.11378
## balise.16256
## balise.25070
## balise.8196
## balise.8337
plot(out1[[1]]) #Plot the first individual</pre>
```



Mosaic individual

Even if the function *loop* perform the analysis forevery individuals the outputs produced are at the individual-level. The function $mosaic_network$ can combine the different individual levels into a single raster representation. When multiple individuals overlap, the function apply a function (mean or max) to calculate a population-level value for that pixel. To use the function, the user needs to specify which variable to mosaic (using index), whether to scale the individual layers (recommended) and the function to apply. We recommend to use mean for degree and weight and max for the betweenness.

```
mean_weight<-mosaic_network(out1, index=2, sc=T, fun=mean) #Perform mean weight (not-interpolated)
#writeRaster(...
mean_degree<-mosaic_network(out1, index=4, sc=T, fun=mean) #Perform mean weight (not-interpolated)
max_between<-mosaic_network(out1, index=5, sc=T, fun=max) #Perform mean weight (not-interpolated)
par(mfrow=c(1,3))
plot(mean_weight, main="Weight")
plot(mean_degree, main="Degree")
plot(max_between, main="Betweenness")</pre>
```



Linear interpolation

As can be seen in the last plot produced, one of the limitation of the current approach is that it creates gaps in areas where no locations are observed (only pixels with gps locations in them have values). This can sometime limit interpretability or the visual appeal of the maps produced. To assist with this, we created a linear interpolation approach that can be applied at the individual level network calculation (i.e. after loop). The interpolation linearly interpolate each step (i.e. straight line) and assign the network metric of each starting location to the whole step. When multiples overlap in a pixel, a function is applied to summarize these steps (e.g. mean or max). This function will take an output from loop and performed the interpolation for five metrics (weight, degree, betweenness, speed, and turning angles). We recommend to take the mean for weight, degree, betweenness, and speed, the max for betweenness, and the dot-product for the turning angles (default).

```
data(albatross) #Load a traj object from adehabitatLT
out1<-loop(albatross, res=35000)

## balise.11378
## balise.16256
## balise.25070
## balise.8196
## balise.8337
out2<-interpolation(albatross, out1) #This is very slow, more than 5 minutes</pre>
```

```
1
                                                                             0%
                                                                            20%
                                                                            40%
                                                                            60%
                                                                            80%
                                                                        =| 100%
mean_mean_degree<-mosaic_network(out2, index=2, sc=T, fun=mean)</pre>
max_max_between<-mosaic_network(out2, index=3, sc=T, fun=max)</pre>
mean_mean_speed<-mosaic_network(out2, index=4, sc=T, fun=mean)</pre>
mean_dot_TA<-mosaic_network(out2, index=5, sc=T, fun=mean)</pre>
par(mfrow=c(2,2))
plot(mean_mean_degree, main= "Degree")
plot(max_max_between, main="Betweenness")
plot(mean_mean_speed, main="Speed")
plot(mean_dot_TA, main="Directionality")
```

