5 main steps to the analysis:

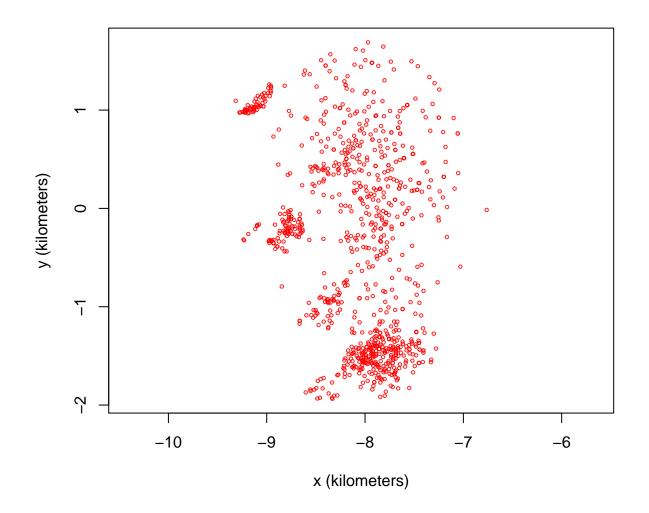
1. import data and convert to telemetry format, then plot by position and time

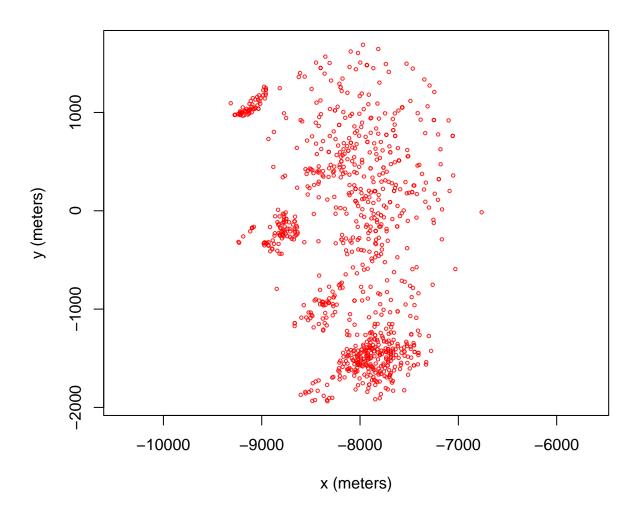
```
library('ctmm') # using the github version (0.6.1)
library('dplyr')
library('lubridate')
library('ggplot2')
theme_set(theme_bw())

tapirs <-
    read.csv('.../data/1_ATLANTICFOREST_11.csv') %>%
    as.telemetry(timeformat = '%Y-%m-%d %H:%M:%S')

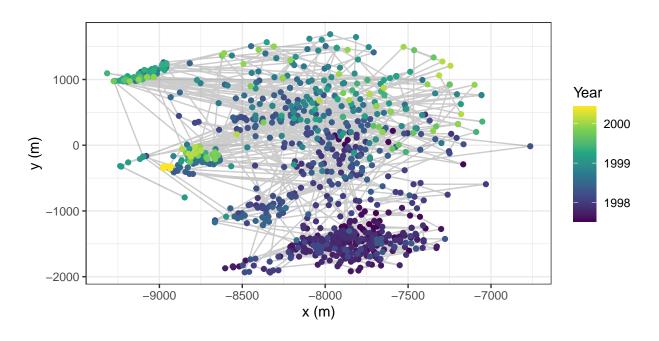
joana <- tapirs[['AF_01_JOANA']] # take the first tapir

plot(joana) # default units; no information on errors, so assume DOP = 1</pre>
```

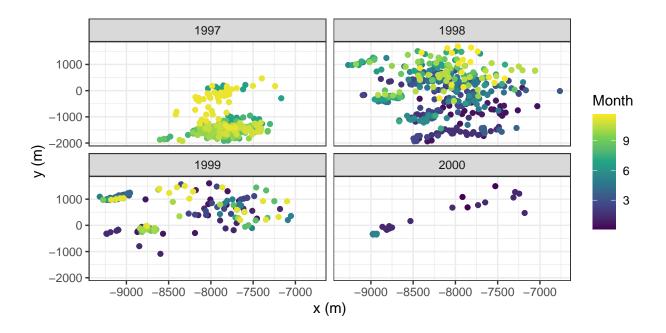




```
# non-stationary movement (plot is not north-oriented)
ggplot(joana, aes(x, y)) +
  geom_path(alpha = 0.2) +
  geom_point(aes(color = decimal_date(timestamp))) +
  scale_color_viridis_c('Year', breaks = c(1998:2000)) +
  labs(x = 'x (m)', y = 'y (m)')
```

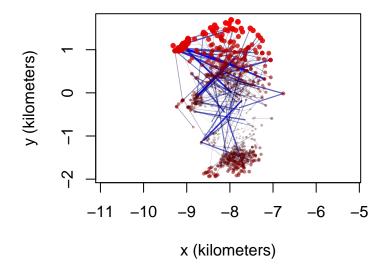


```
# facet plot to see seasonal detail
ggplot(joana, aes(x, y)) +
  facet_wrap(~ year(timestamp)) +
  geom_point(aes(color = (decimal_date(timestamp) %% 1) * 12)) +
  scale_color_viridis_c('Month', breaks = c(3, 6, 9)) +
  labs(x = 'x (m)', y = 'y (m)')
```

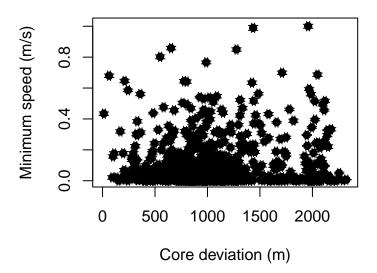


2. check for outliers

diagn <- outlie(joana) # movement is not stationary</pre>

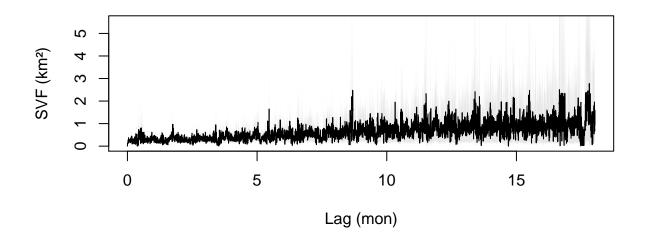


plot(diagn, units = FALSE) # very narrow CIs, nonstationary movement

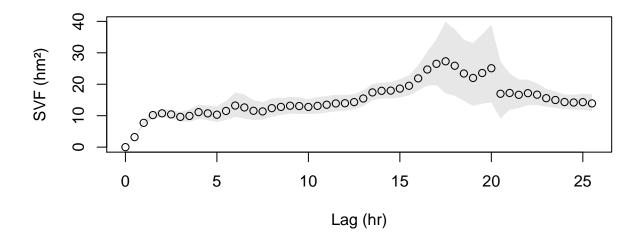


3. Variograms

```
svf <- variogram(joana) # estimate semi-variance function
head(warnings(), n = 2) # some warnings occur occasionally
# NULL
plot(svf) # plots 50% max lag by default, > 50% is "usually garbage"
```



plot(svf, fraction = 1e-3) # zoom in on the variance at shortest lags



4. fit and select movement model

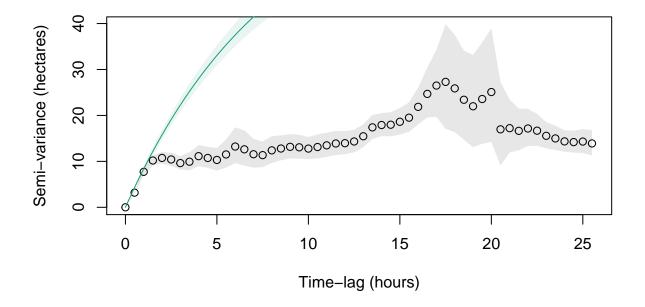
```
variogram.fit(svf, name = 'theta0', fraction = 0.0003)
fitted.mods <- ctmm.select(joana, CTMM = theta0, verbose = TRUE, cores = 0)</pre>
```

Negligible difference between OU and OUF models (both anisotropic and isotropic), but choose anisotropic since the tapir's movement is not directionally uniform.

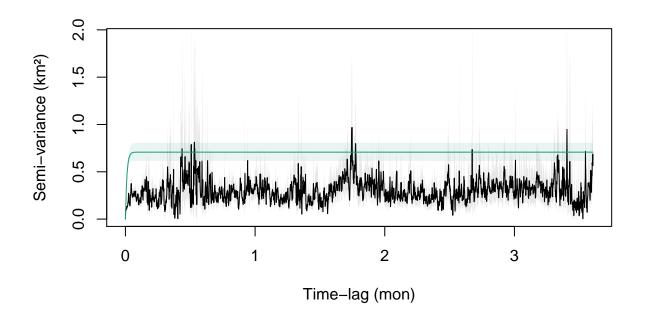
```
summary(fitted.mods)
                         <U+0394>AICc
                                             <U+0394>RMSPE (m) DOF[area]
#
# OU anisotropic
                      0.000000
                                       153.35132 216.7769
# OU isotropic
                      1.480760
                                       123.15327 248.7336
# OUF anisotropic
                                       149.91268 220.0352
                      2.015385
# OUF isotropic
                      3.488707
                                       120.41786 252.1144
# OUf anisotropic
                    465.039453
                                         0.00000 408.7265
                                        70.34647 955.3603
# IID anisotropic 1955.215875
OUa <- fitted.mods$`OU anisotropic` # Ornstein-Uhlenbeck
```

5. calculate distance and speed estimates

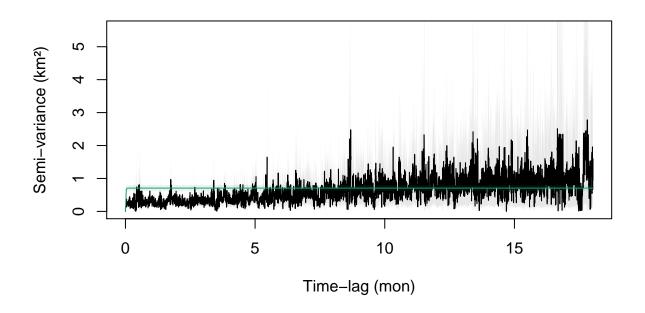
```
# plot semivariance function against empirical variogram
plot(svf, CTMM = OUa, col.CTMM = '#1b9e77', fraction = 0.001)
```

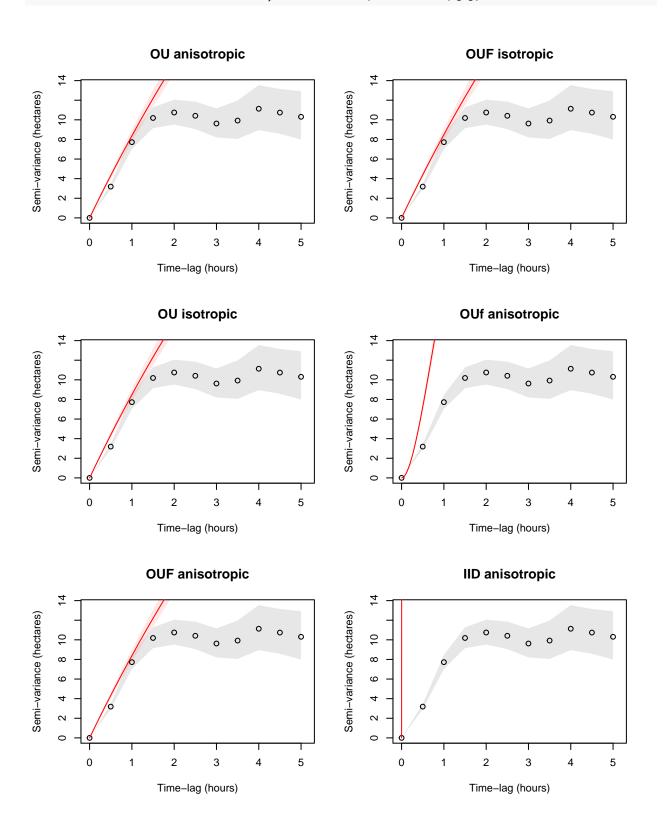


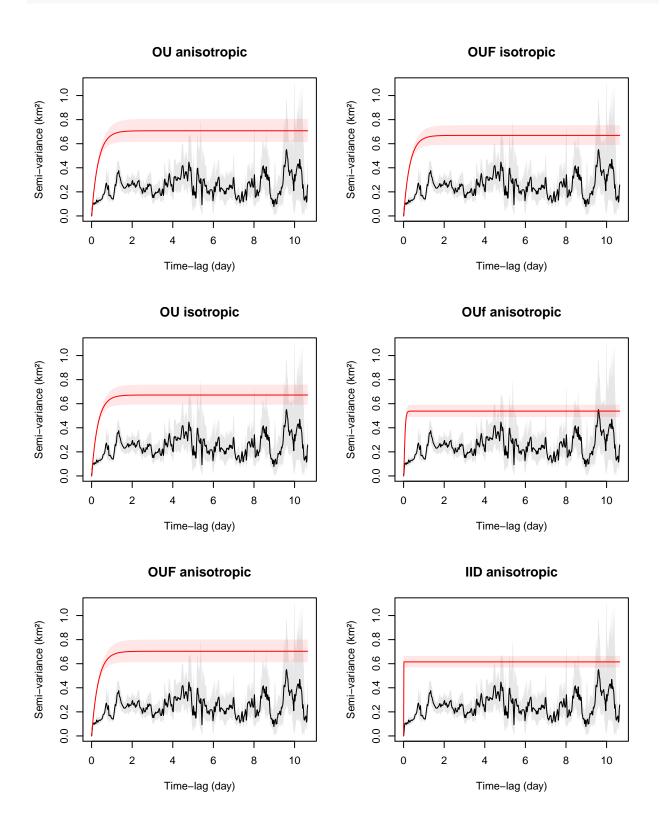
```
plot(svf, CTMM = OUa, col.CTMM = '#1b9e77', fraction = 0.1)
```

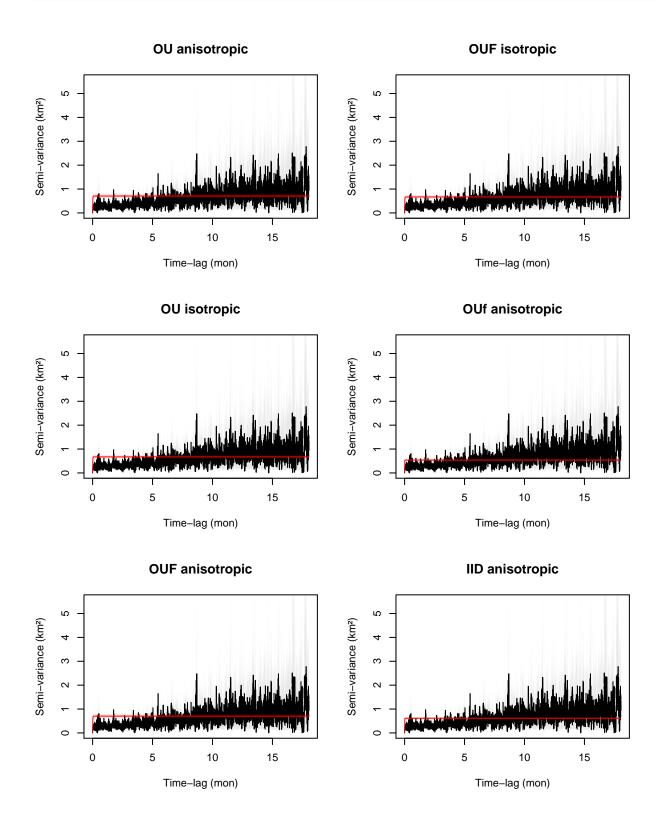


plot(svf, CTMM = OUa, col.CTMM = '#1b9e77', fraction = 0.5)









```
layout(1)
summary(OUa) # parameter and estimates with CIs
# $name
# [1] "OU anisotropic"
#
# $DOF
# mean area speed
# 182.3084 216.7769 0.0000
# $CI
#
                             low est high
# area (square kilometers) 11.558513 13.265792 15.08896
# t[position] (hours) 6.692213 7.939544 9.41936
speed(OUa) # OU model cannot make speed estimates (OUF can)
# Warning in speed.ctmm(OUa): Movement model is fractal.
                    low est high
est <- akde(joana, CTMM = OUa)</pre>
plot(est) # 95% quantile of home range distribution with 95% CIs
plot(joana, add = TRUE) # datapoints
```

