

ANIMOVE

```
PI.x=diag(c(0, 0.001, 0.001))
PI.y=diag(c(0, 0.001, 0.001))
```

```
displayPar(mov.model=~1, err.model=list(x=~errX, y=~errY), drift.noise=0.001,
data=nfsNew, fixPar=c(NA, 1, NA, 1, NA, NA, NA, NA))
```

```
t <- crwMLE(mov.model=~1, err.model=list(x=~errX, y=~errY), drift.noise=0.001,
data=nfsNew, coord=c("longitude", "latitude"), polar.coord=0,
Time.name="Time", initial.state=initial.drift,
fixPar=c(NA, 1, NA, 1, NA, NA, NA, NA),
control=list(maxit=2000, trace=1, REPORT=10),
```





June 2024 **Movement data in R**

Area centered analysis: space use

Home range

A “home range” is the area that an individual utilizes to maintain its own energetic and behavioural demands.

Definition

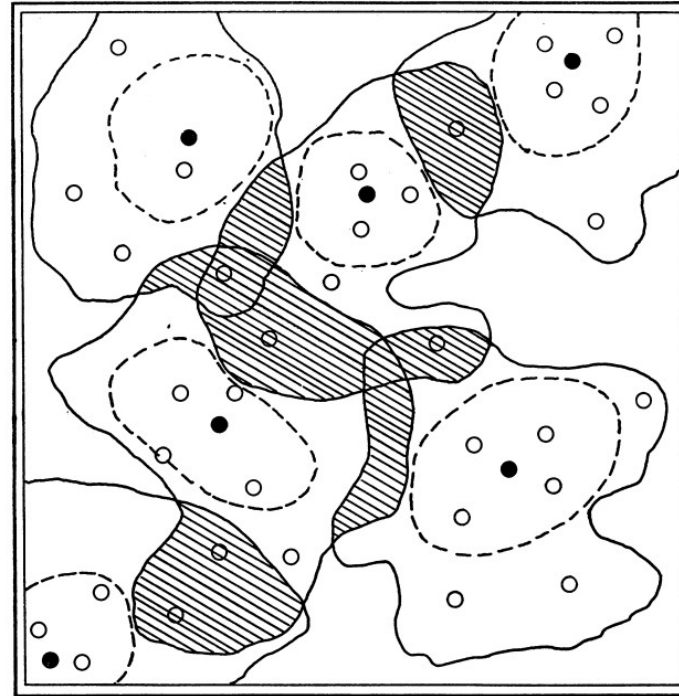
Burt 1943[#]:

- “[...] I would restrict the home range to that area traversed by the individual in its normal activities of food gathering, mating, and caring for young [...].”
- migratory animals have different home ranges during summer and winter, their migratory route is not part of a home range.
- “[...] young adolescent animals often do a bit of wandering in search of a home region. During this time they do not have a home, nor [...] a home range.”

General understanding: “home range” is the area that an individual utilizes to maintain its own energetic and behavioral demands, due to technical and/or logistic limitations, the term “home range” nonetheless remains widely used and is often understood in a loose reference to the area a tagged individual was observed in during the time of the study period.

[#]Burt, W. H. (1943). Territoriality and home range concepts as applied to mammals. Journal of mammalogy, 24(3), 346-352.

How to quantify “home range”?



— HOME RANGE BOUNDARY ▨ NEUTRAL AREA
- - - TERRITORIAL BOUNDARY ● NESTING SITE
BLANK--UNOCCUPIED SPACE ○ REFUGE SITE

FIG. 1. Theoretical quadrat with six occupants of the same species and sex, showing territory and home range concepts as presented in text.

Burt, J Mammal 1943

From observation to movement

Trade-off between how much for how long.

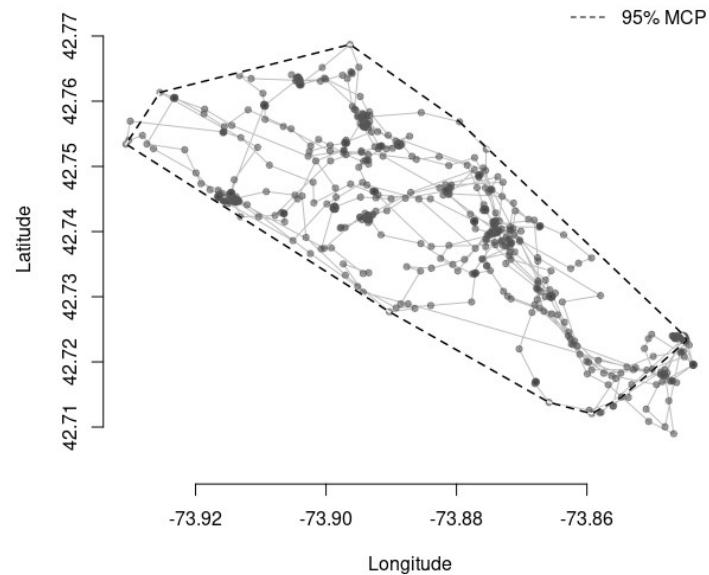
Temporal resolution

Temporal coverage



MCP (Minimum Convex Polygon)

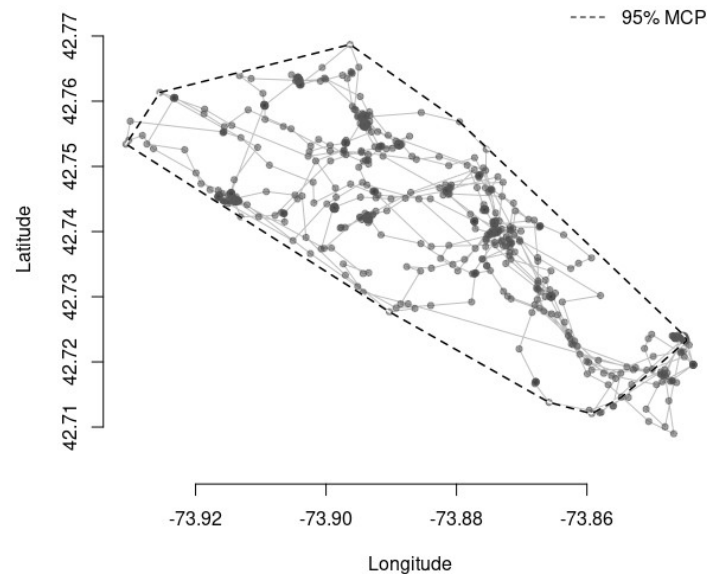
- * polygon with the minimum area containing e.g. 95% of the points (MCP95)



MCP (Minimum Convex Polygon)

* polygon with the minimum area containing e.g. 95% of the points (MCP95)

- + simple approach
- assumes independence between observations
- does not account for errors
- includes areas that maybe do not make sense
- intensity of space use is lost

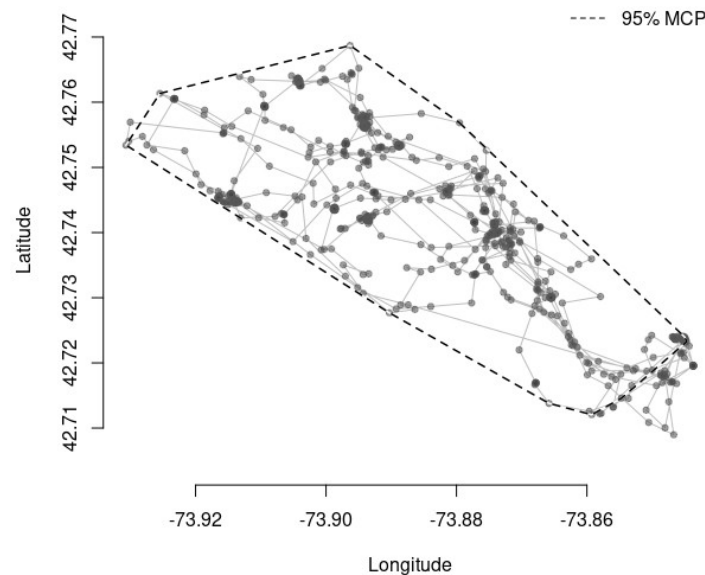


MCP (Minimum Convex Polygon)

* polygon with the minimum area containing e.g. 95% of the points (MCP95)

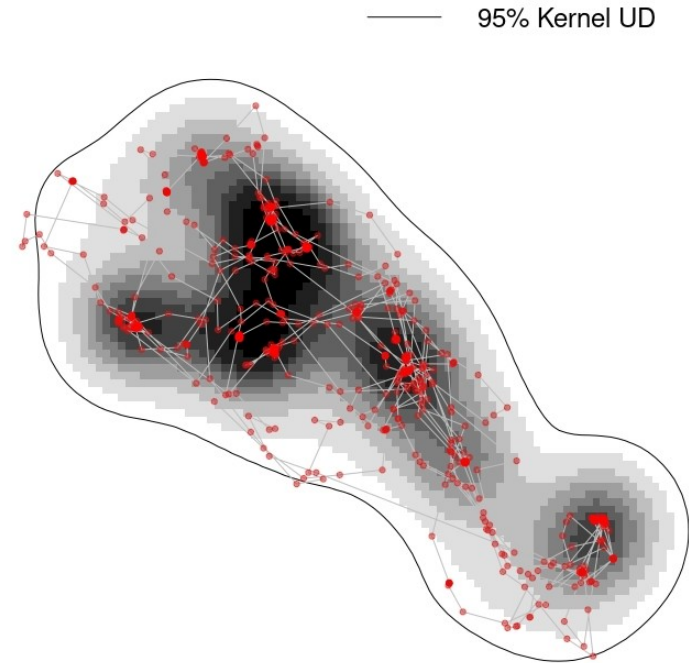
- + simple approach
- assumes independence between observations
- does not account for errors
- includes areas that maybe do not make sense
- intensity of space use is lost

Note!: always reproject locations into equidistant projection before calculating MCP



KDE (Kernel-density estimation)

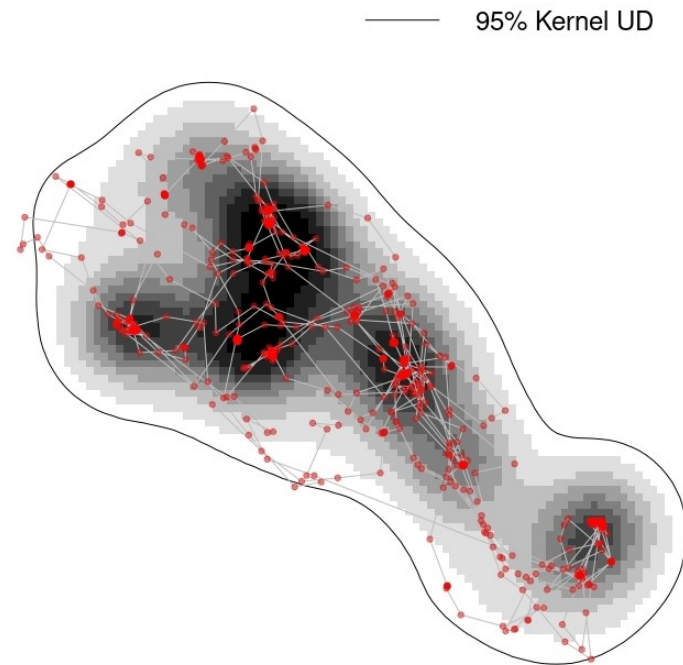
- * calculates density of locations. Estimates the probability of encountering the animal given the independence of sampling.



KDE (Kernel-density estimation)

* calculates density of locations. Estimates the probability of encountering the animal given the independence of sampling.

- assumes independence between observations and regular sampling. Auto-correlation results in underestimating the area, often dramatically
- results highly dependent on the chosen h value

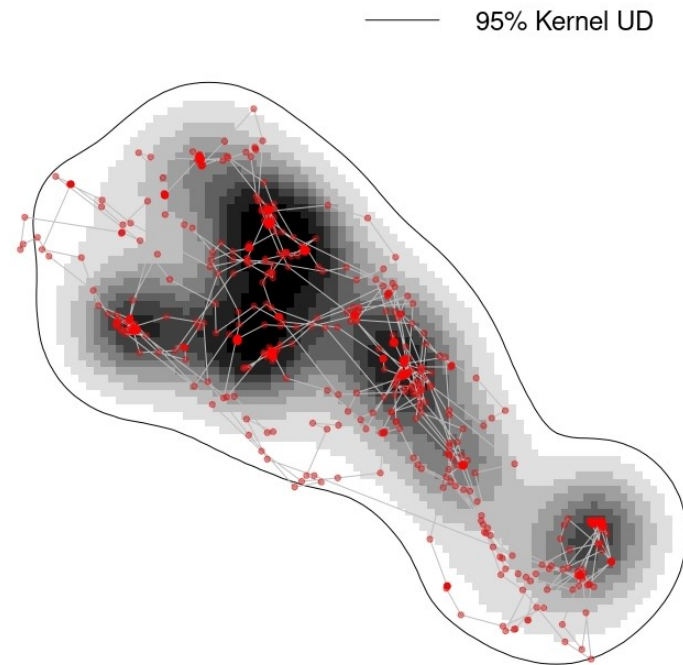


KDE (Kernel-density estimation)

* calculates density of locations. Estimates the probability of encountering the animal given the independence of sampling.

- assumes independence between observations and regular sampling. Auto-correlation results in underestimating the area, often dramatically
- results highly dependent on the chosen h value

These assumptions were not such a problem when data was scarce.



LoCoH (Local Convex-Hull)

- * builds local MCP (Hulls) for each location with k neighbors, or in r radius, or in a sum of distances

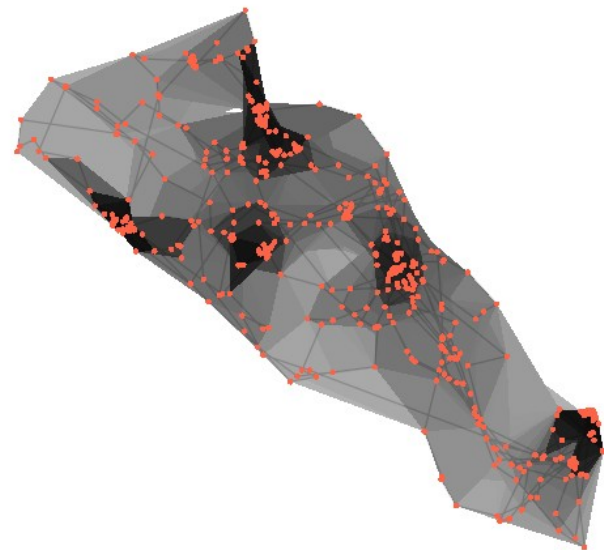
k-LoCoH



LoCoH (Local Convex-Hull)

- * builds local MCP (Hulls) for each location with k neighbors, or in r radius, or in a sum of distances
 - + it takes into account that points are somewhat related
 - + it excludes the areas that are not used
 - choosing k , r or a values in non-trivial
 - assumes independence between observations and regular sampling

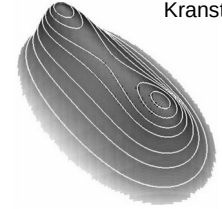
k-LoCoH



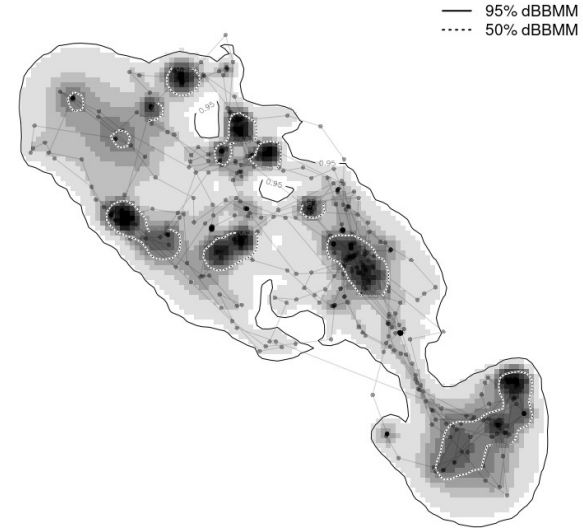
dBBMM (dynamic Brownian Bridge Movement Model)

- * calculates the probability landscape for the transition between any two known consecutive locations given the amount of time it had available (assumes conditional random (Brownian) motion between locations)
- * "dynamic" because it allows the variance to change along the trajectory (see Lesson 4)

Horne, Ecology 2007
Kranstauber, J Anim Ecol 2012



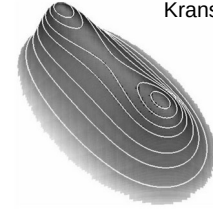
Dynamic brownian bridge



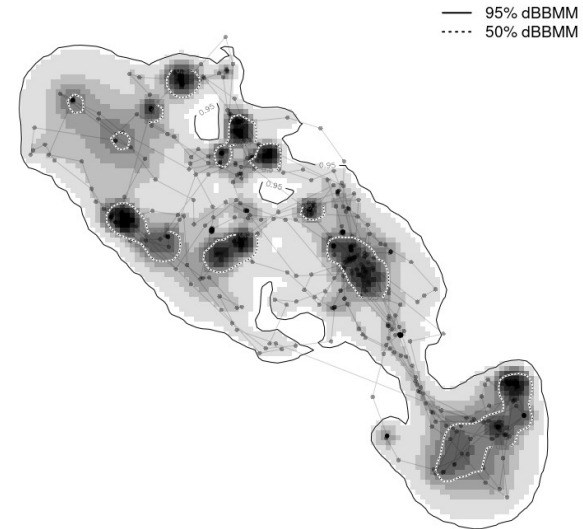
dBBMM (dynamic Brownian Bridge Movement Model)

- * calculates the probability landscape for the transition between any two known consecutive locations given the amount of time it had available (assumes conditional random (Brownian) motion between locations)
- * "dynamic" because it allows the variance to change along the trajectory (see Lesson 4)
 - + does not assume independence between locations
 - + irregular sampling is not a problem

Horne, Ecology 2007
Kranstauber, J Anim Ecol 2012



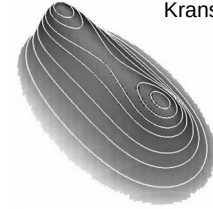
Dynamic brownian bridge



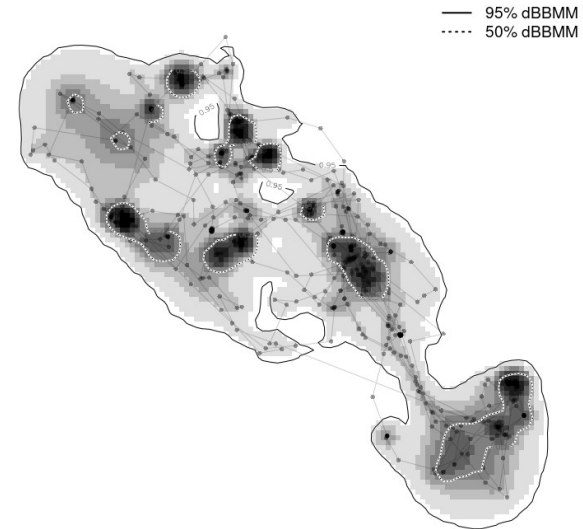
dBBMM (dynamic Brownian Bridge Movement Model)

- * calculates the probability landscape for the transition between any two known consecutive locations given the amount of time it had available (assumes conditional random (Brownian) motion between locations)
- * "dynamic" because it allows the variance to change along the trajectory (see Lesson 4)
 - + does not assume independence between locations
 - + irregular sampling is not a problem
- ! obtained area is highly scale dependent, it changes with sampling rate and with error of locations

Horne, Ecology 2007
Kranstauber, J Anim Ecol 2012



Dynamic brownian bridge



Space use

Range distribution[§] :

Occurrence distribution[§] :

[§]Fleming, et al. (2015). Rigorous home range estimation with movement data: a new auto-correlated kernel density estimator. Ecology, 96(5), 1182-1188.

Space use

Range distribution[§] :

- Lifetime space requirements of an animal (\sim Home range)
- Provides a metric that should be comparable across individuals
- Ideally not affected by sample size and study duration (not the case for MPC, LoCoh and KDE)
- MCP, LoCoH, and KDE assume independent observations, but teleportation is not possible
 - Targeted by MPC, LoCoH, KDE, **AKDE**(ctmm)

Occurrence distribution[§] :

[§]Fleming, et al. (2015). Rigorous home range estimation with movement data: a new auto-correlated kernel density estimator. Ecology, 96(5), 1182-1188.

Space use

Range distribution[§] :

- Lifetime space requirements of an animal (\sim Home range)
- Provides a metric that should be comparable across individuals
- Ideally not affected by sample size and study duration (not the case for MPC, LoCoh and KDE)
- MCP, LoCoH, and KDE assume independent observations, but teleportation is not possible
- Targeted by MPC, LoCoH, KDE, **AKDE**(ctmm)

Occurrence distribution[§] :

- Estimates where an animal was located during the observation period
- Area cannot be compared across individuals (it would be comparing "lack of knowledge")
- Sensitive to sampling frequency but robust to data irregularity
- Targeted by dBBMM (move), occurrence (ctmm)

[§]Fleming, et al. (2015). Rigorous home range estimation with movement data: a new auto-correlated kernel density estimator. Ecology, 96(5), 1182-1188.