

SUPPLEMENTARY MATERIAL

The R code developed to represent the three mixture models considered.

(i)

```
# R Code for Exponential segments Model
```

```
#Adapted from: Zucchini, W. & MacDonald, I. L. (2009) Hidden Markov Models for Time Series - An Introduction Using R. Boca Raton, Chapman & Hall/CRC.
```

```
#
```

```
## Exponential segments using nlm
```

```
### Transform natural parameters to working parameters
```

```
glogit = function(p)
```

```
{ log(p/(1-sum(p))) }
```

```
glogitinv = function(p)
```

```
{ exp(p)/(1+sum(exp(p))) }
```

```
exp.mix.pn2pw = function(m,lambda)
```

```
{
```

```
tlambda = log(lambda)
```

```
params = c(tlambda)
```

```
params
```

```
}
```

```
### Transform working parameters to natural parameters
```

```
exp.mix.pw2pn = function(m,params) {
```

```
lambda = exp(params[1:m])
```

```
list(lambda=lambda) }
```

```
### Segment model
```

```
exp.mixlike = function(m,params,x, model_offset){
```

```
lambda = exp.mix.pw2pn(m,params)$lambda
```

```
allprobs <- matrix(NA,length(x),m+1)
```

```
allprobs[,1] <- ifelse(x <= model_offset[1],0,NA)
```

```
allprobs[,m+1] <- ifelse(x >= model_offset[m],dexp(x,lambda[m]),NA)
```

```
for (i in 2:(m)){
```

```
allprobs[,i] <- ifelse(x >= model_offset[i-1] & x <= model_offset[i], dexp(x,lambda[i-1]),NA)}
```

```
sums <- rowSums(allprobs,na.rm=T)
```

```
sums <- ifelse(sums==0,NA,sums)
```

```
foo = -sum(log(sums), na.rm=T)
```

```
foo
```

```
}
```

```
### Compute the parameters which minimize the deviance of the segments model
```

```
exp.mix.mle = function(x,m,lambda0, model_offset){
```

```
params0 = exp.mix.pn2pw(m,lambda0)
```

```
mod = nlm(exp.mixlike,params0,x=x,m=m, model_offset=model_offset)
```

```
pn = exp.mix.pw2pn(m,mod$estimate)
```

```
minusl = mod$minimum
```

```
np = length(params0)
```

```

AIC = 2*(minu+l+np)
n = length(x)
BIC = 2*minu+l+np*log(n)
list(lambda=pn$lambda, code=mod$code, minu=l, AIC=AIC, BIC=BIC)}

```

(ii)

```

# R Code for Gamma Mixture Model including offsets
# Adapted from: Zucchini, W. & MacDonald, I. L. (2009) Hidden Markov Models for Time Series - An
Introduction Using R. Boca Raton, Chapman & Hall/CRC.
#

```

```

### Gamma model using nlm
#### Transform natural parameters to working parameters
glogit = function(p)
{ log(p/(1-sum(p))) }
glogitinv = function(p)
{ exp(p)/(1+sum(exp(p))) }
gamma.mix.pn2pw = function(m,delta,alpha,theta)
{
tdelta = glogit(delta[1: (m-1)])
talpha = log(alpha)
ttheta = log(theta)
params = c(tdelta, talpha, ttheta )
params
}

```

```

#### Transform working parameters to natural parameters
gamma.mix.pw2pn = function(m,params) {
temp = glogitinv(params[1: (m-1)])
delta = c(temp,1-sum(temp))
alpha = exp(params[m: (2*m-1)])
theta = exp(params[(2*m): (3*m-1)] )
list(delta=delta, alpha=alpha, theta=theta) }

```

```

#### Compute the likelihood of mixture model
mixlike = function(m,params,x,model_offset){
delta = gamma.mix.pw2pn(m,params)$delta
alpha = gamma.mix.pw2pn(m,params)$alpha
theta = gamma.mix.pw2pn(m,params)$theta
allprobs <- matrix(NA,length(x),m)
for (i in 1:m){
allprobs[,i] <- dgamma(x, alpha[i], scale=theta[i])}
for (i in 2:m){
allprobs[,i] <- ifelse(x<model_offset[i],0,allprobs[,i])}
foo = -sum(log(allprobs%%delta))
foo
}

```

```

#### Compute the parameters which maximise the log-likelihood of the mixture model

```

```

gamma.mix.mle = function(x,m,delta0, alpha0, theta0, model_offset){
params0 = gamma.mix.pn2pw(m,delta0,alpha0,theta0)
mod     = nlm(mixlike,params0,x=x,m=m, model_offset=model_offset)
pn      = gamma.mix.pw2pn(m,mod$estimate)
minusl  = mod$minimum
np      = length(params0)
AIC     = 2*(minusl+np)
n       = length(x)
BIC     = 2*minusl+np*log(n)
list(delta=pn$delta,alpha=pn$alpha,theta=pn$theta,code=mod$code,minusl=minusl,AIC=AIC,BIC=BIC)}

```

(iii)

R Code for Log-normal Mixture Model

#Adapted from: Zucchini, W. & MacDonald, I. L. (2009) Hidden Markov Models for Time Series - An Introduction Using R. Boca Raton, Chapman & Hall/CRC.

#

Log-normal model using nlm

Transform natural parameters to working parameters

glogit = function(p)

```
{ log(p/(1-sum(p))) }
```

glogitinv = function(p)

```
{ exp(p)/(1+sum(exp(p))) }
```

lnorm.mix.pn2pw = function(m,delta,mu,sigma)

```
{
```

```
tdelta = glogit(delta[1: (m-1)])
```

```
tsigma = log(sigma)
```

```
params = c(tdelta, mu, tsigma )
```

```
params
```

```
}
```

Transform working parameters to natural parameters

lnorm.mix.pw2pn = function(m,params) {

```
temp = glogitinv(params[1: (m-1)])
```

```
delta = c(temp,1-sum(temp))
```

```
mu = params[m: (2*m-1)]
```

```
sigma = exp(params[(2*m): (3*m-1)] )
```

```
list(delta=delta, mu=mu, sigma=sigma) }
```

Compute the likelihood of mixture model

ln.mixlike = function(m,params,x){

```
delta = lnorm.mix.pw2pn(m,params)$delta
```

```
mu = lnorm.mix.pw2pn(m,params)$mu
```

```
sigma = lnorm.mix.pw2pn(m,params)$sigma
```

```
allprobs <- matrix(NA,length(x),m)
```

```
  for (i in 1:m){
```

```
    allprobs[,i] <- dlnorm(x, mu[i], sigma[i])}
```

```
foo = -sum(log(allprobs%*%delta))
```

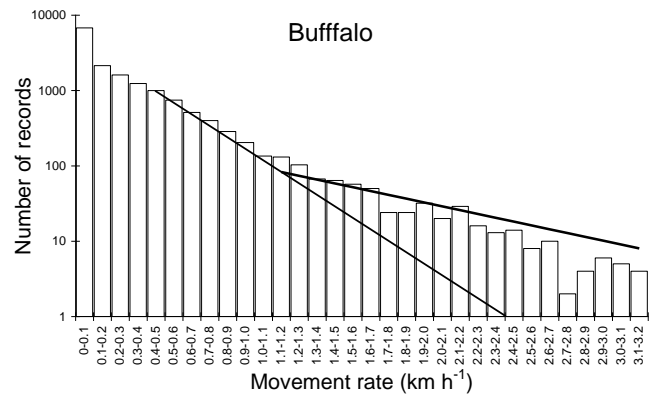
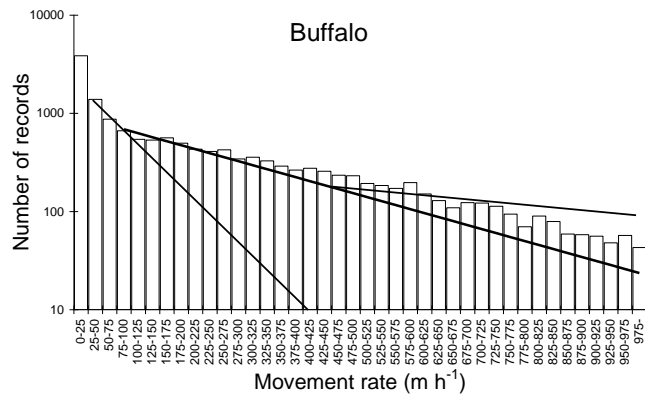
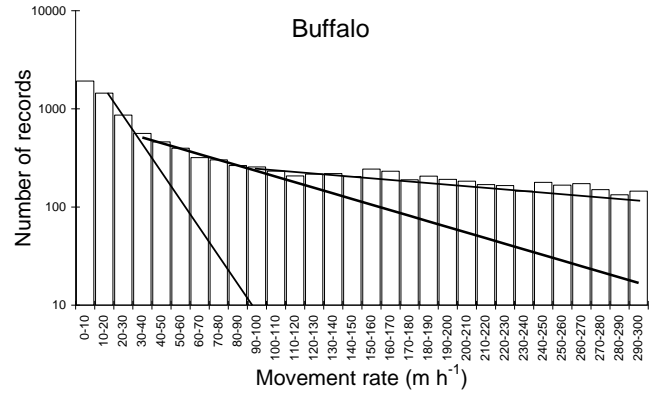
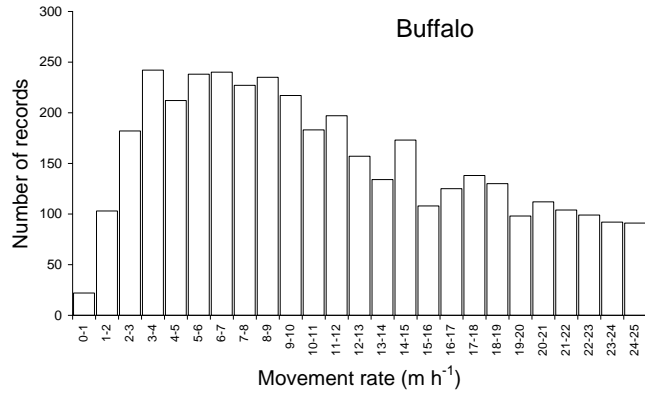
```
foo
```

```
}
```

```
### Compute the parameters which maximise the log-likelihood of the mixture model
lnorm.mix.mle = function(x,m,delta0, mu0, sigma0){
  params0 = lnorm.mix.pn2pw(m,delta0,mu0,sigma0)
  mod     = nlm(ln.mixlike,params0,x=x,m=m)
  pn      = lnorm.mix.pw2pn(m,mod$estimate)
  minusl  = mod$minimum
  np      = length(params0)
  AIC     = 2*(minusl+np)
  n       = length(x)
  BIC     = 2*minusl+np*log(n)
  list(delta=pn$delta,mu=pn$mu,sigma=pn$sigma,code=mod$code,minusl=minusl,AIC=AIC,BIC=BIC)}
```

Fig. S1. Log-frequency distributions of hourly movement rates over different bin resolutions, from amalgamated data over the entire year for all herds of (a) buffalo and (b) zebra.

(a) Buffalo



(b) Zebra

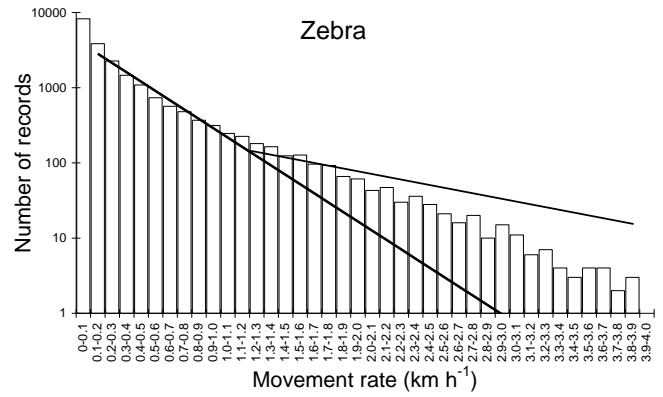
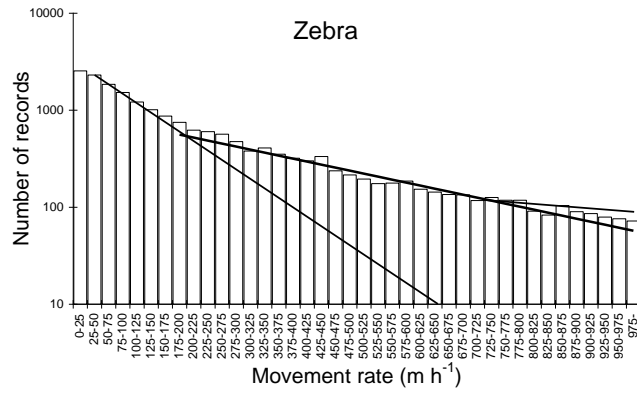
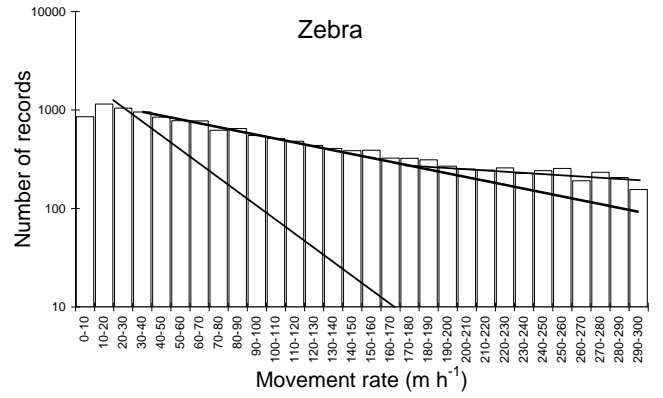
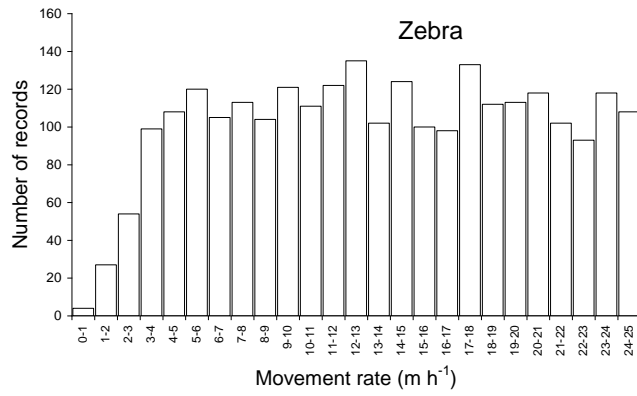
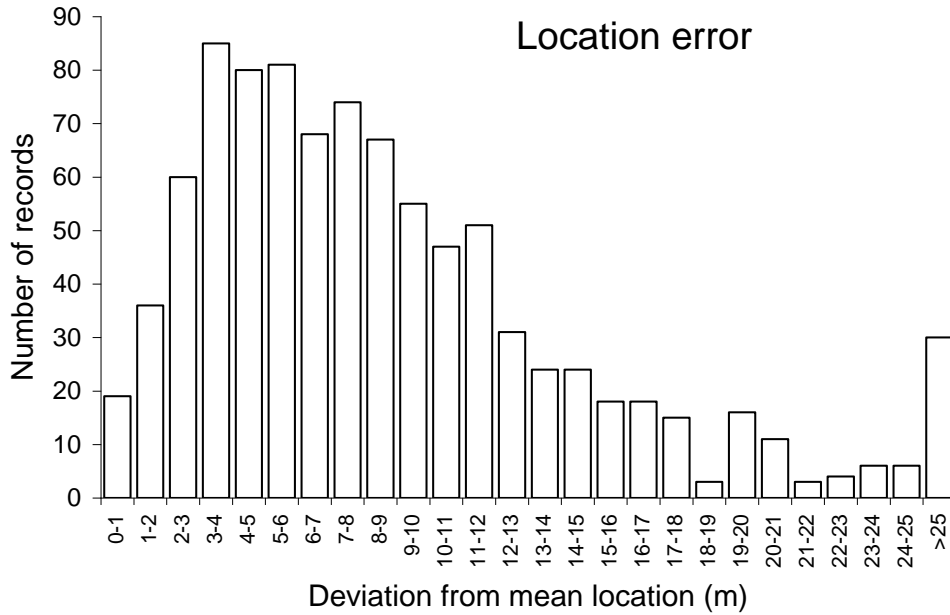


Fig. S2. Error distribution around the mean generated by stationary GPS collars left in trees or other locations. High errors were generally associated with high dilution of precision (PDOP) in the GPS locations

(a) Distribution of the GPS locations



(b) Distribution of hourly differences between successive GPS locations

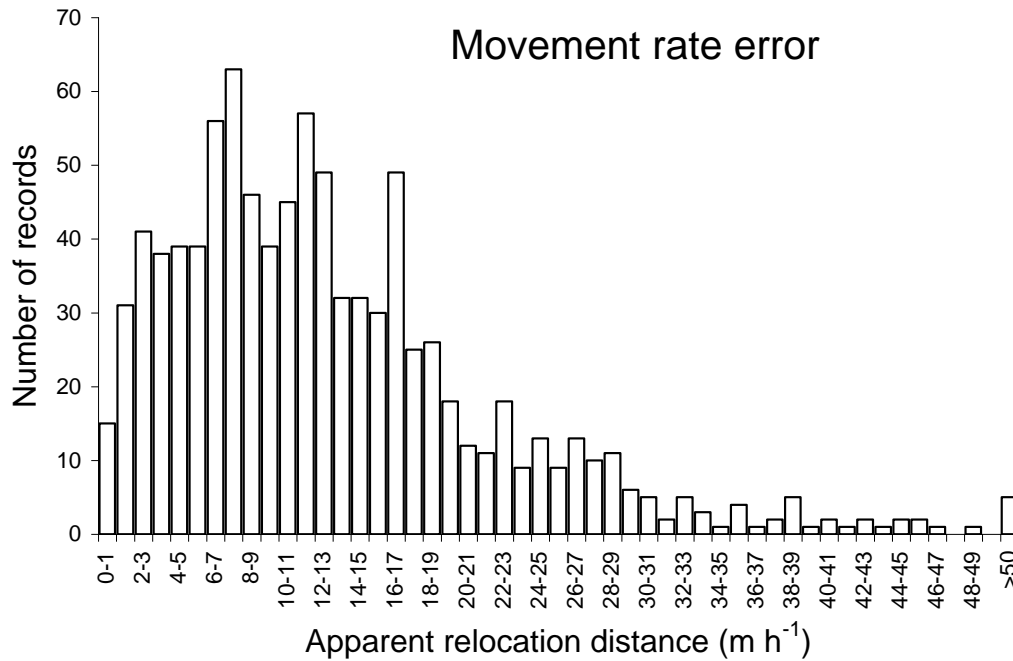
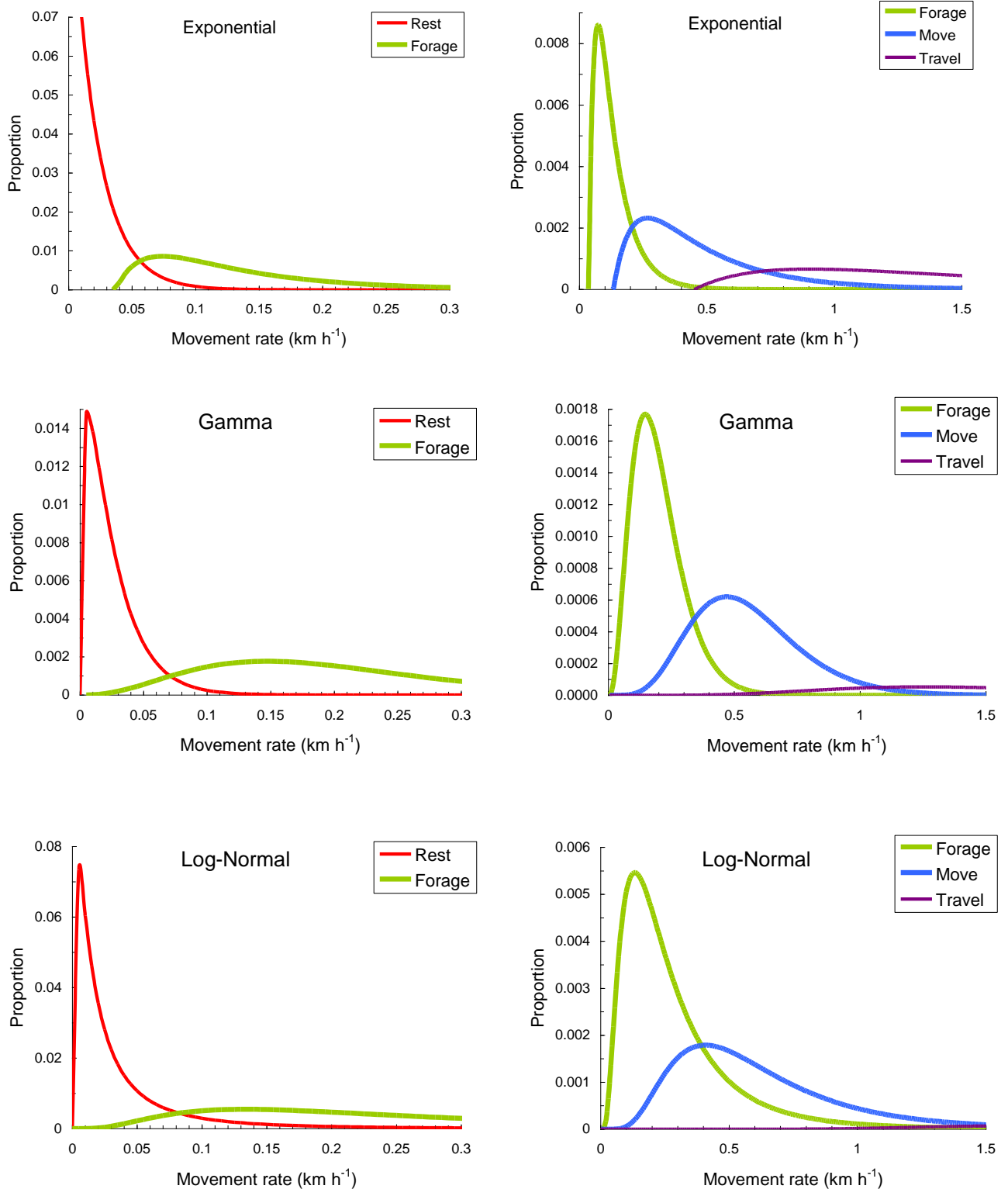


Fig. S3. Component distributions of alternative models fitted to the movement data, interpreted as representing resting, foraging, mixed movement and travelling activity states. For the gamma distribution mixture for zebra, dotted lines represent the split of the first component between resting and slow foraging for the five- component model. (a) Buffalo, (b) zebra.

(a) Buffalo



(b) Zebra

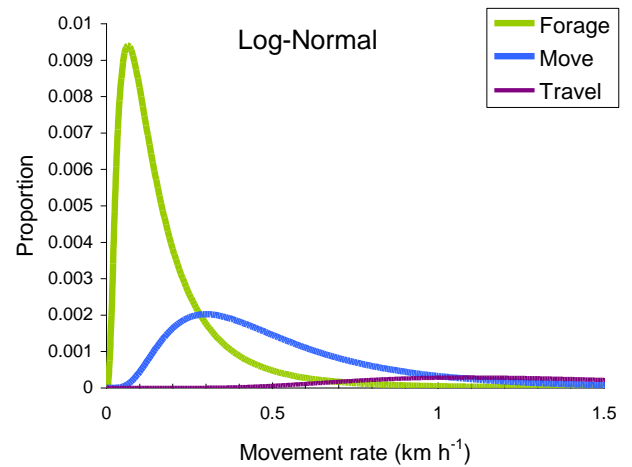
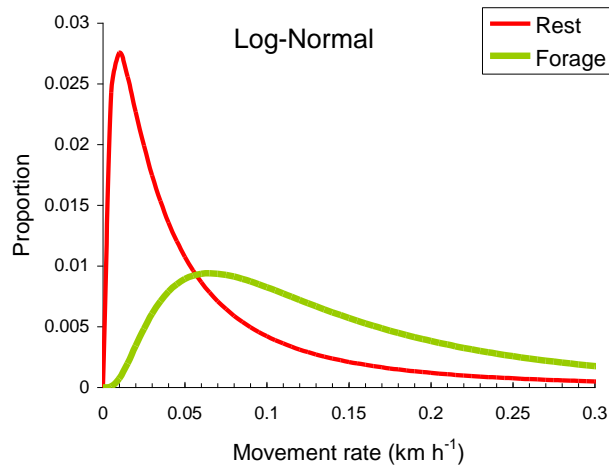
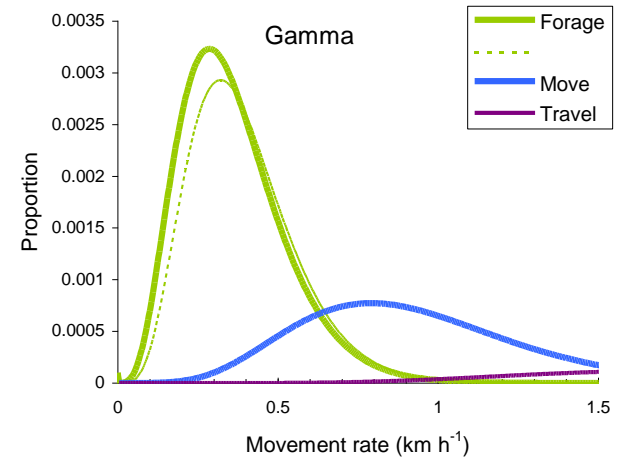
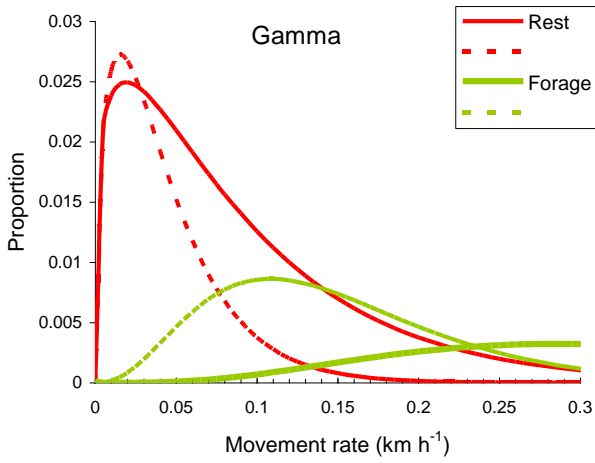
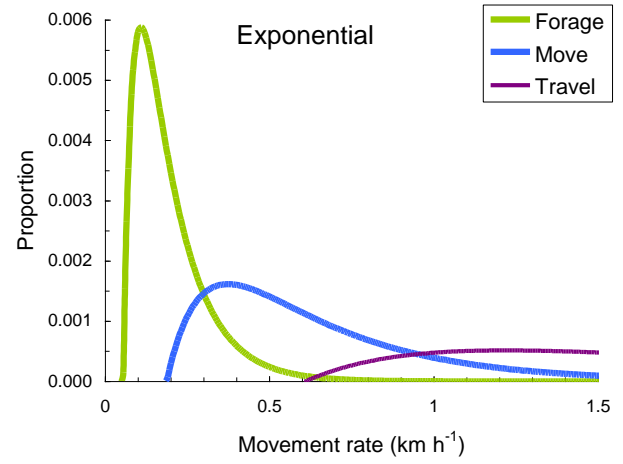
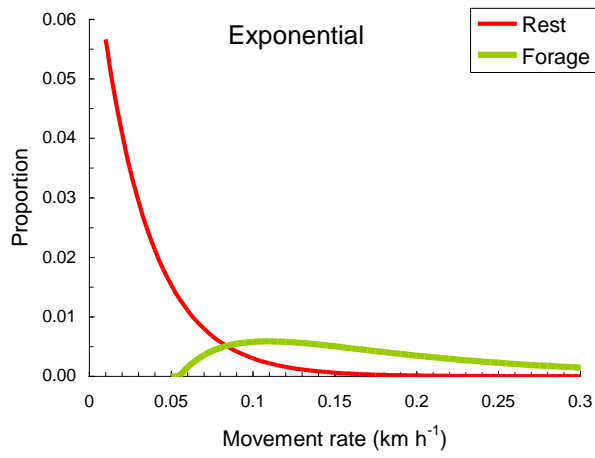
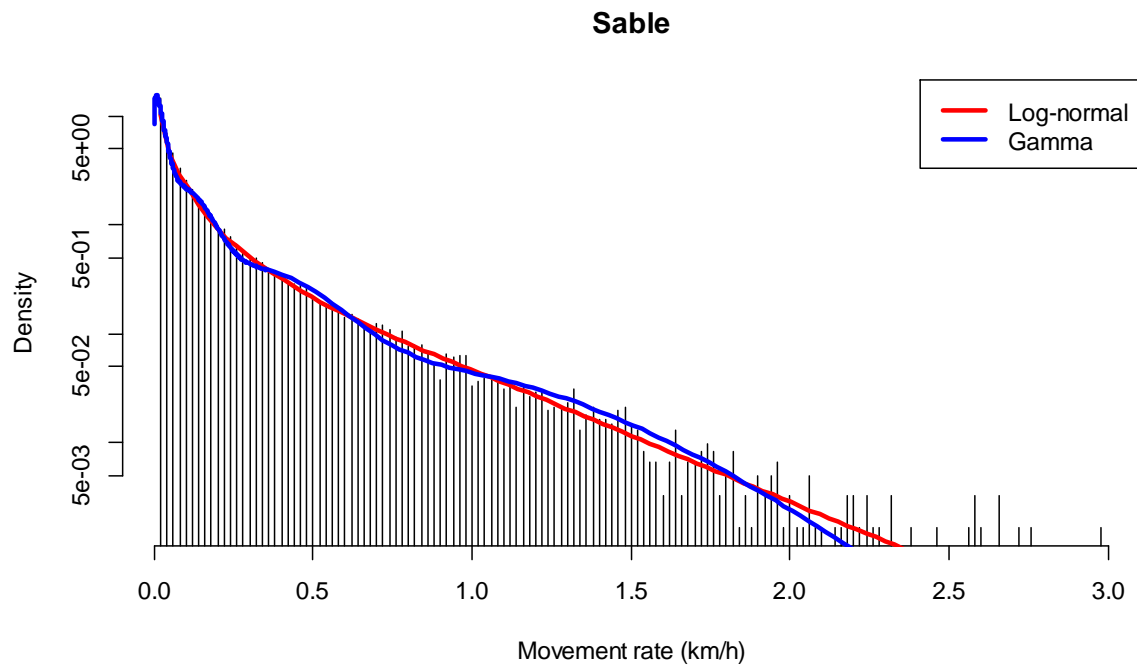
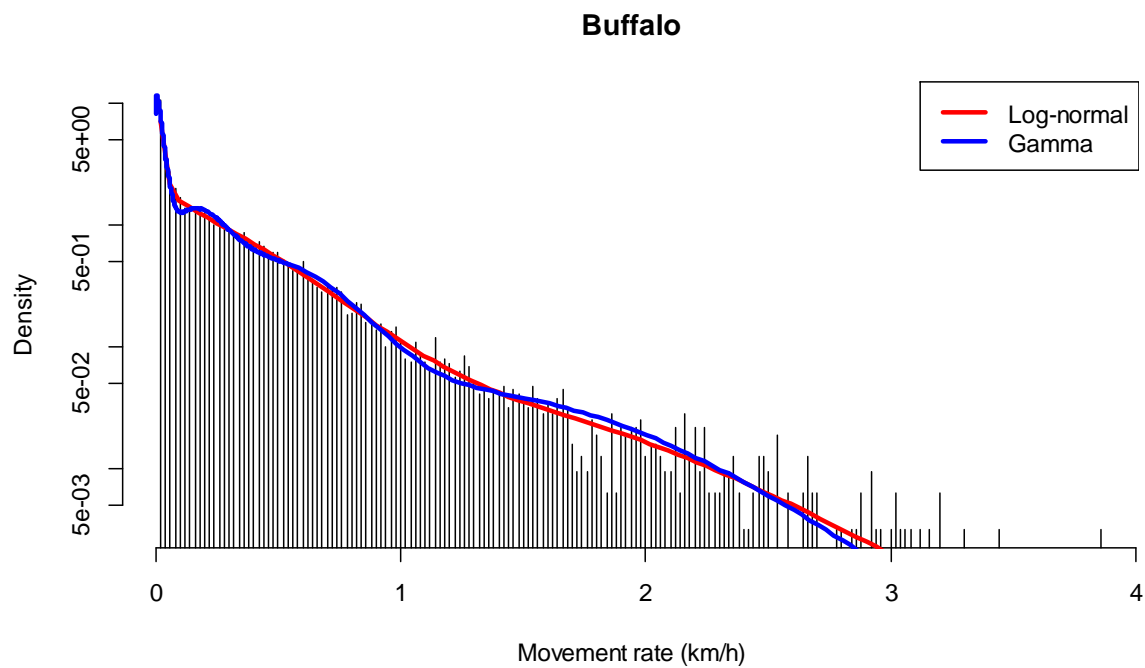


Fig. S4. Comparative fit of the alternative models to the amalgamated data distribution across herds and seasons of each ungulate species

(a) Sable



(b) Buffalo



(c) Zebra

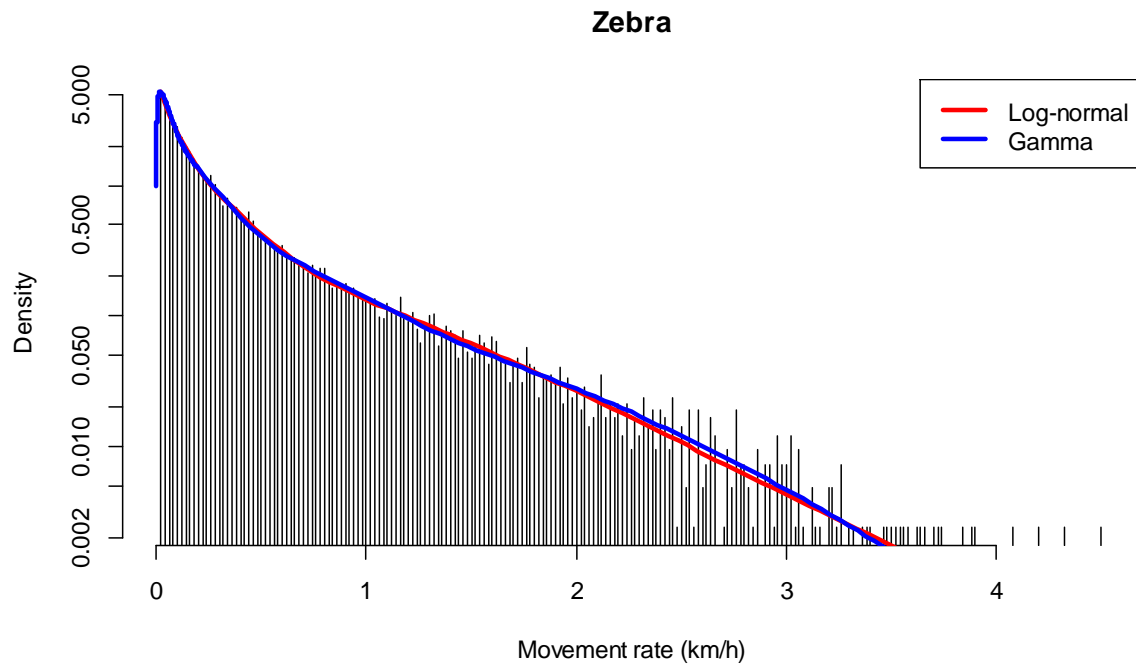
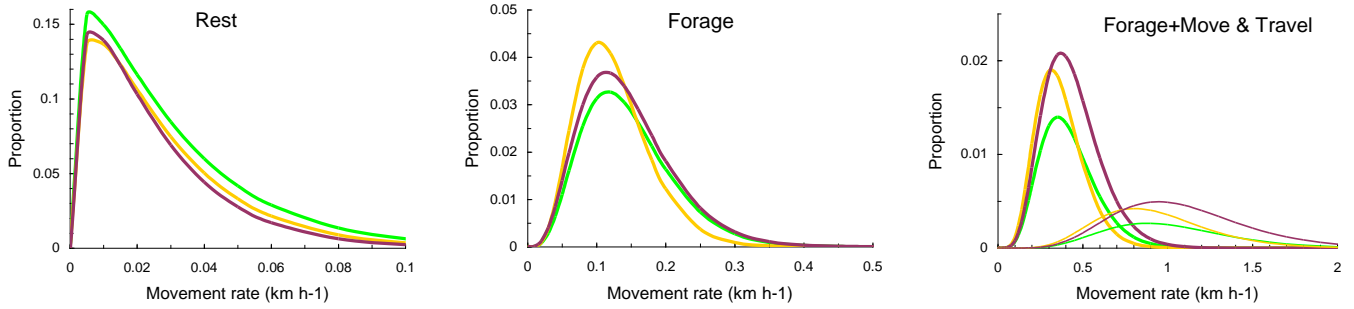
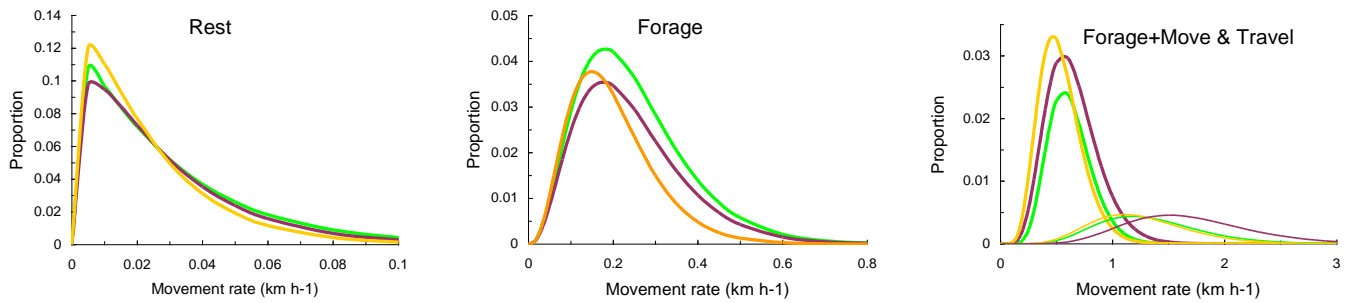


Fig. S5. Comparisons of the gamma functions representing the activity states among seasonal blocks. Green lines represents Wet season (December-March), gold line Early dry season (April-July), and purple line Late dry season (August-November).

(a) Sable



(b) Buffalo



(c) Zebra

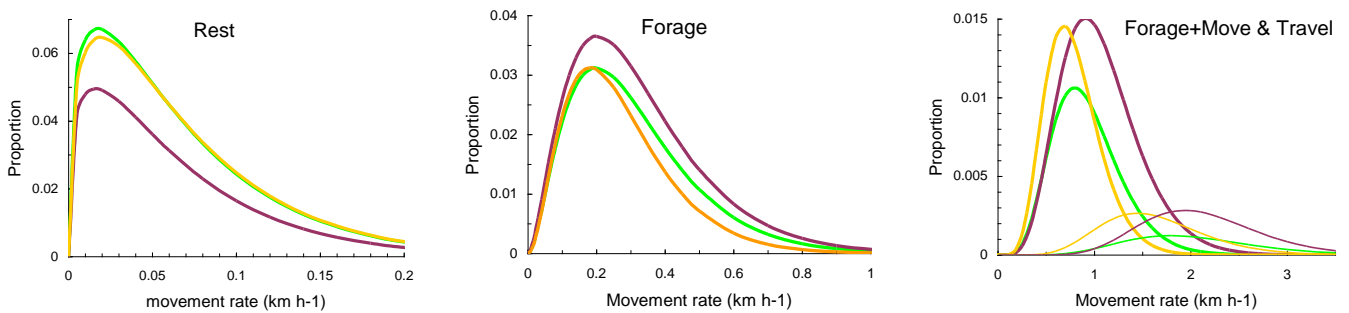
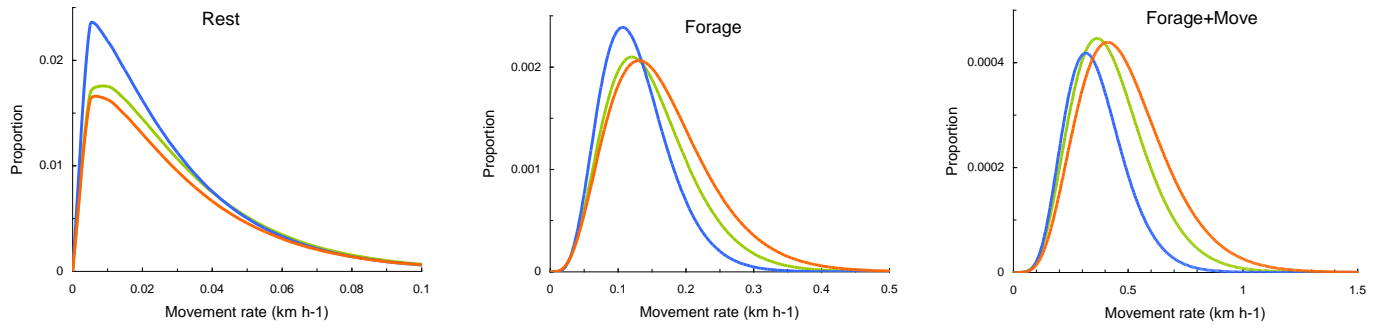


Fig. S6. Comparison of gamma functions representing the activity states derived among three different herds of sable and zebra

(a) Sable



(b) Zebra

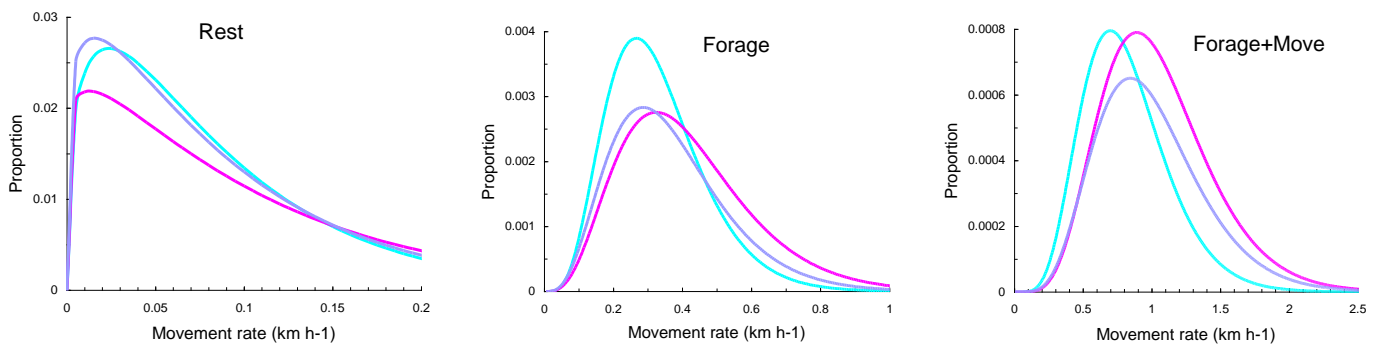


Table S1. Comparative parameter estimates for 4-component log-normal and gamma models derived from Microsoft Excel versus R programme

Species	Season	N	Model	States	Software	o1	o2	o3	p1	p2	p3	p4	alpha1	beta1	alpha2	beta2	alpha3	beta3	alpha4	beta4	Log Likelihood
Sable	Year	30453	Log-normal	4	Excel				0.616	0.196	0.163	0.025	-3.844	1.189	-2.229	0.594	-1.071	0.600	-0.090	0.417	34859
					R				0.618	0.194	0.163	0.025	-3.837	1.192	-2.228	0.591	-1.074	0.600	-0.091	0.418	34855
Sable	Wet	10291	Log-normal	4	Excel				0.668	0.113	0.181	0.038	-3.794	1.235	-2.557	0.520	-1.514	0.645	-0.372	0.482	13375
					R				0.672	0.111	0.179	0.038	-3.786	1.239	-2.553	0.517	-1.512	0.643	-0.372	0.482	13373
Sable	Early Dry	7439	Log-normal	3	Excel				0.550	0.305	0.138	0	-3.946	1.115	-2.256	0.628	-0.939	0.563	-	-	8885
					R				0.550	0.304	0.146	-	-3.946	1.116	-2.255	0.637	-0.904	0.621	-	-	8888
Sable	Late Dry	12725	Log-normal	4	Excel				0.593	0.245	0.138	0.024	-3.871	1.173	-1.925	0.647	-0.776	0.534	0.086	0.373	12776
					R				0.595	0.243	0.139	0.024	-3.867	1.176	-1.925	0.646	-0.778	0.534	0.085	0.374	12779
Sable	Year	30451	Gamma	4	Excel	0.035	0.11	0.35	0.573	0.255	0.134	0.037	1.311	0.021	5.062	0.027	6.494	0.063	7.619	0.135	34461
					R	0.035	0.11	0.35	0.573	0.255	0.134	0.037	1.311	0.021	5.059	0.027	6.492	0.063	7.618	0.135	34460
Sable	Wet	10291	Gamma	4	Excel	0.035	0.11	0.33	0.640	0.233	0.103	0.024	1.232	0.024	4.938	0.030	6.503	0.064	7.259	0.135	13242
					R	0.035	0.11	0.33	0.640	0.233	0.103	0.024	1.232	0.024	4.938	0.030	6.503	0.064	7.530	0.135	13241
Sable	Early Dry	7439	Gamma	4	Excel	0.035	0.11	0.33	0.560	0.273	0.131	0.035	1.353	0.020	5.555	0.023	7.010	0.052	7.937	0.116	8788
					R	0.035	0.11	0.33	0.560	0.273	0.132	0.035	1.353	0.020	5.557	0.023	7.010	0.052	7.937	0.116	8783
Sable	Late Dry	12725	Gamma	4	Excel	0.035	0.12	0.37	0.533	0.268	0.152	0.047	1.359	0.018	4.694	0.031	6.649	0.066	7.732	0.141	12621
					R	0.035	0.12	0.37	0.533	0.268	0.152	0.047	1.358	0.018	4.686	0.031	6.648	0.066	7.734	0.141	12626
Buffalo	Year	15734	Log-normal	4	Excel				0.424	0.336	0.223	0.017	-3.908	1.205	-1.495	0.722	-0.610	0.537	0.583	0.283	6519
					R				0.425	0.336	0.222	0.017	-3.903	1.208	-1.492	0.721	-0.608	0.535	0.582	0.283	6517
Buffalo	Wet	4861	Log-normal	4	Excel				0.440	0.399	0.139	0.022	-3.785	1.311	-1.402	0.715	-0.588	0.455	0.358	0.334	2169
					R				0.442	0.404	0.132	0.023	-3.777	1.316	-1.389	0.715	-0.582	0.443	0.347	0.336	2169
Buffalo	Early Dry	5269	Log-normal	3	Excel				0.398	0.290	0.311	0	-4.143	1.114	-1.720	0.790	-0.782	0.636	-	-	2813
					R				0.400	0.291	0.308		-4.136	1.118	-1.711	0.786	-0.777	0.635	-	-	2812
Buffalo	Late Dry	5605	Log-normal	4	Excel				0.422	0.272	0.280	0.027	-3.830	1.171	-1.510	0.703	-0.552	0.559	0.674	0.242	1623
					R				0.422	0.272	0.280	0.027	-3.828	1.172	-1.508	0.702	-0.581	0.558	0.674	0.242	1623

Buffalo	Year	15734	Gamma	4	Excel	0.045	0.18	0.55	0.397	0.346	0.209	0.048	1.220	0.022	3.846	0.059	7.918	0.076	8.532	0.177	6303
					R	0.045	0.18	0.55	0.397	0.346	0.209	0.048	1.220	0.022	3.846	0.059	7.915	0.076	8.532	0.177	6304
Buffalo	Wet	4861	Gamma	4	Excel	0.045	0.17	0.50	0.403	0.403	0.145	0.048	1.115	0.027	3.672	0.068	10.423	0.060	8.503	0.157	2107
					R	0.045	0.17	0.50	0.398	0.352	0.203	0.047	1.133	0.026	4.173	0.053	8.394	0.068	8.740	0.155	2110
Buffalo	Early Dry	5269	Gamma	4	Excel	0.041	0.16	0.47	0.399	0.327	0.225	0.049	1.265	0.018	3.963	0.050	7.948	0.068	8.409	0.149	2732
					R	0.041	0.16	0.47	0.399	0.327	0.225	0.049	1.264	0.018	3.964	0.050	7.949	0.068	8.409	0.150	2732
Buffalo	Late Dry	5606	Gamma	4	Excel	0.045	0.18	0.59	0.391	0.334	0.216	0.060	1.290	0.022	3.663	0.066	7.702	0.084	9.432	0.179	1558
					R	0.045	0.18	0.59	0.391	0.334	0.216	0.060	1.290	0.022	3.660	0.066	7.701	0.084	9.433	0.179	1559
Zebra	Year	21028	Log-normal	4	Excel				0.333	0.362	0.232	0.073	-3.177	1.226	-2.033	0.833	-0.807	0.629	0.276	0.430	4300
					R				0.334	0.361	0.232	0.073	-3.171	1.228	-2.031	0.833	-0.806	0.628	0.276	0.430	4300
Zebra	Wet	7416	Log-normal	4	Excel				0.363	0.413	0.175	0.049	-3.107	1.287	-2.064	0.925	-0.807	0.622	0.184	0.474	2760
					R				0.363	0.415	0.174	0.049	-3.105	1.288	-2.064	0.928	-0.807	0.623	0.184	0.474	2759
Zebra	Early Dry	5628	Log-normal	4	Excel				0.327	0.429	0.168	0.076	-3.178	1.244	-2.101	0.861	-0.917	0.634	0.018	0.465	2115
					R				0.328	0.429	0.167	0.076	-3.174	1.245	-2.099	0.861	-0.915	0.634	0.018	0.464	2115
Zebra	Late Dry	7986	Log-normal	4	Excel				0.194	0.328	0.361	0.117	-3.693	1.024	-2.195	0.755	-0.903	0.645	0.330	0.414	319
					R				0.201	0.311	0.370	0.117	-3.658	1.034	-2.209	0.737	-0.915	0.648	0.329	0.415	319
Zebra	Year	21028	Gamma	4	Excel	0.01	0.055	0.19	0.426	0.348	0.172	0.054	1.266	0.072	4.828	0.075	6.843	0.135	9.911	0.197	4232
					R	0.01	0.055	0.19	0.439	0.330	0.165	0.068	1.391	0.045	3.092	0.081	4.383	0.157	5.477	0.283	4263
Zebra	Wet	7416	Gamma	4	Excel	0.01	0.06	0.19	0.501	0.322	0.143	0.034	1.384	0.046	3.042	0.086	4.477	0.163	5.307	0.312	2753
					R	0.01	0.06	0.19	0.503	0.321	0.143	0.033	1.376	0.047	3.031	0.087	4.445	0.166	5.438	0.309	2752
Zebra	Early Dry	5628	Gamma	4	Excel	0.01	0.06	0.18	0.464	0.311	0.159	0.065	1.449	0.042	3.388	0.063	4.964	0.119	5.145	0.252	2108
					R	0.01	0.06	0.18	0.467	0.309	0.158	0.065	1.443	0.043	3.395	0.063	4.979	0.119	5.150	0.252	2108
Zebra	Late Dry	7986	Gamma	4	Excel	0.01	0.06	0.21	0.356	0.380	0.187	0.076	1.368	0.044	2.950	0.093	4.410	0.186	7.133	0.247	343
					R	0.01	0.06	0.21	0.356	0.380	0.189	0.075	1.368	0.044	2.947	0.093	4.382	0.188	7.189	0.246	343