

Queensland rainforest

```
library(ppjsdm)
library(spatstat)
#> Loading required package: spatstat.data
#> Loading required package: nlme
#> Loading required package: rpart
#>
#> spatstat 1.62-2      (nickname: 'Shape-shifting lizard')
#> For an introduction to spatstat, type 'beginner'
remove(list = ls())

source("../R/get_qld.R")

set.seed(1)
```

This vignette explains how to use the `ppjsdm` package with the Queensland rainforest dataset from CSIRO. We begin by loading the data with the most prevalent species.

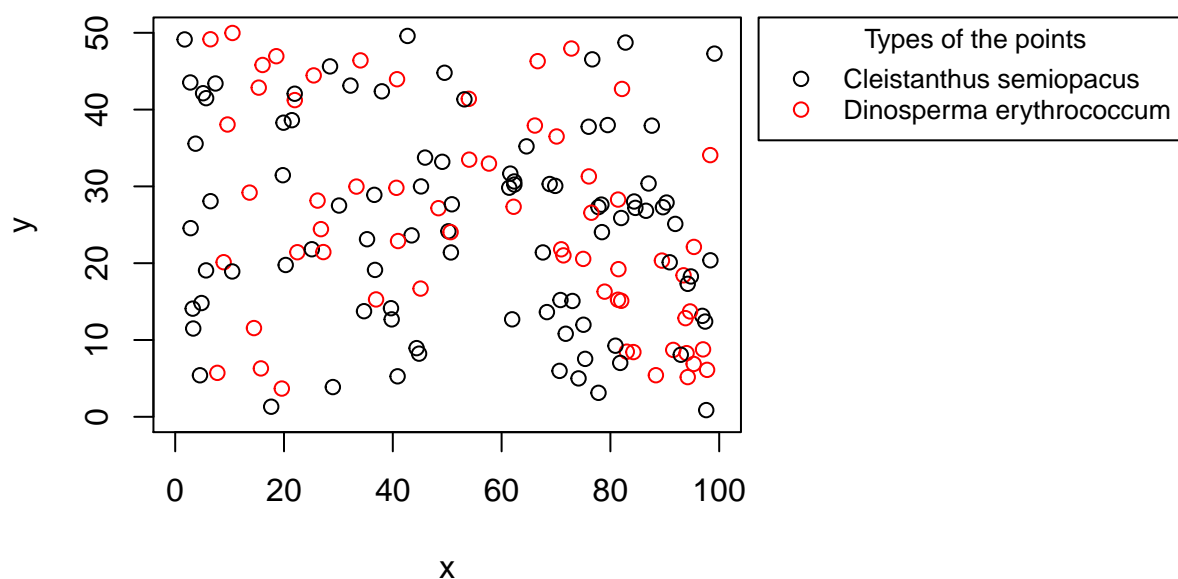
```
index_of_plot <- 3 # Between 1 and 20
year <- 2011 # Year of census
number_of_species <- 2 # Includes the most prevalent species from the plot

qld <- get_qld(index = index_of_plot,
               year = year,
               prevalent = number_of_species)
#> The chosen index corresponds to ep2.
configuration <- qld$configuration
window <- qld$window
```

The point configuration is plotted below.

```
par(mar = c(5, 4, 4, 13) + 0.1)
plot(configuration, window = window)
```

Points in the configuration



The function `gibbsm` fits a multivariate Gibbs point process to our dataset. For example,

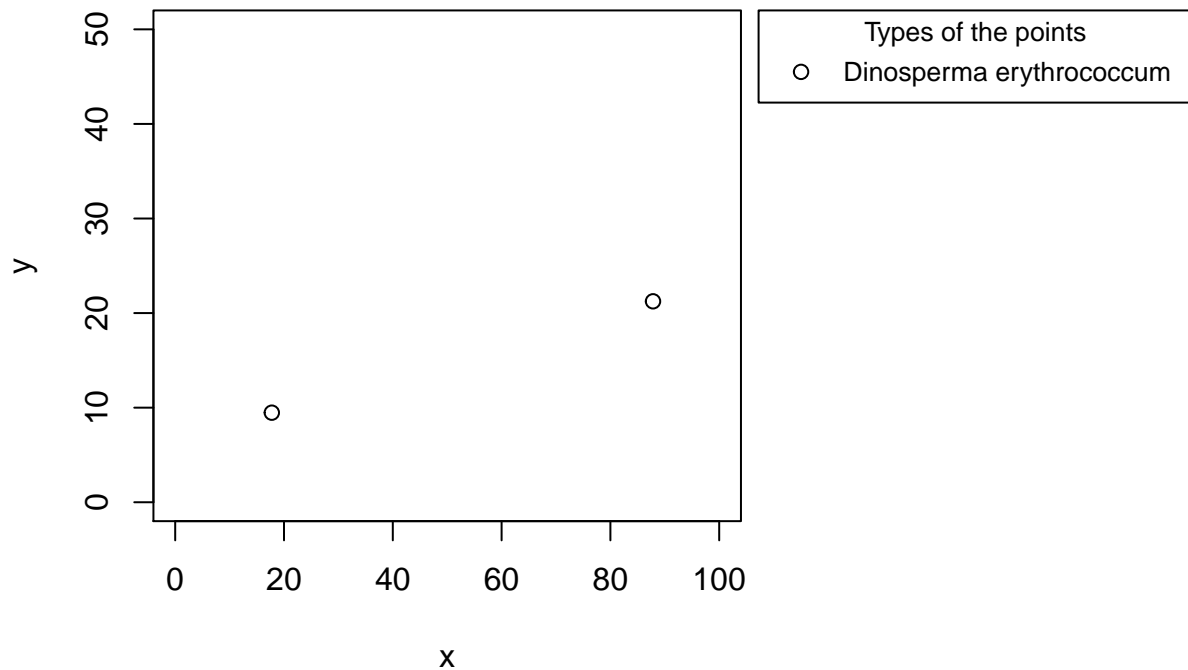
```
fit <- ppjsdm::gibbsm(configuration, window = window)
#> log_lambda_1 log_lambda_2 alpha_1_1 alpha_1_2 alpha_2_2
#> -3.970526 -4.430065 -4.710366 9.591091 -4.374614
```

By default, the function fits the model that was introduced in the ARC grant [TODO: Add reference]. This model has many drawbacks, the most important of which is that the model with the fitted values is degenerate, as can be seen by drawing from the model.

```
lambda <- exp(fit$coefficients[1:number_of_species])
alpha <- matrix(c(fit$coefficients[3],
                  fit$coefficients[4],
                  fit$coefficients[4],
                  fit$coefficients[5]), ncol = number_of_species, nrow = number_of_species)
draw <- ppjsdm::rmultigibbs(window = window,
                           alpha = alpha,
                           lambda = lambda,
                           types = levels(types(configuration)),
                           steps = 1000000)

par(mar = c(5, 4, 4, 13) + 0.1)
plot(draw, window = window)
```

Points in the configuration



The Geyer model is better suited to most situations, but the user needs to specify some additional parameters before the fitting may take place.

```
radii <- matrix(5, number_of_species, number_of_species)
```

The matrix `radii` models interaction radii within a species, and between species. An interaction radius of 5 gives good results in the fitting procedure.

```
fit <- ppjsdm::gibbsm(configuration, window = window, model = "Geyer", radius = radii)
#> log_lambda_1 log_lambda_2 alpha_1_1 alpha_1_2 alpha_2_2
#> -4.12136802 -4.71144863 0.12682397 -0.05099062 0.41186500
summary(fit)
#>
#> Call:
#> glm(formula = as.formula(gibbsm_data$formula), family = binomial(),
#> data = as.data.frame(gibbsm_data$data))
#>
#> Deviance Residuals:
#> Min 1Q Median 3Q Max
#> -0.8374 -0.6849 -0.6447 -0.5681 1.9735
#>
#> Coefficients:
#> Estimate Std. Error z value Pr(>|z|)
#> log_lambda_1 -4.12137 0.22632 -18.210 <2e-16 ***
#> log_lambda_2 -4.71145 0.25094 -18.775 <2e-16 ***
#> alpha_1_1 0.12682 0.14904 0.851 0.3948
#> alpha_1_2 -0.05099 0.11519 -0.443 0.6580
```

```

#> alpha_2_2      0.41187    0.17937    2.296    0.0217 *
#> ---
#> Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
#>
#> (Dispersion parameter for binomial family taken to be 1)
#>
#>      Null deviance: 3459.79  on 755  degrees of freedom
#> Residual deviance:  749.44  on 750  degrees of freedom
#> AIC: 759.44
#>
#> Number of Fisher Scoring iterations: 4

```

Note that in this case the model is not degenerate; indeed we can sample from it as follows.

```

lambda <- exp(fit$coefficients[1:number_of_species])
alpha <- matrix(c(fit$coefficients[3],
                  fit$coefficients[4],
                  fit$coefficients[4],
                  fit$coefficients[5]), ncol = number_of_species, nrow = number_of_species)
draw <- ppjsdm::rmultigibbs(window = window,
                           alpha = alpha,
                           lambda = lambda,
                           model = "Geyer",
                           radius = radii,
                           types = levels(types(configuration)),
                           steps = 1000000)

par(mar = c(5, 4, 4, 13) + 0.1)
plot(draw, window = window)

```

Points in the configuration

