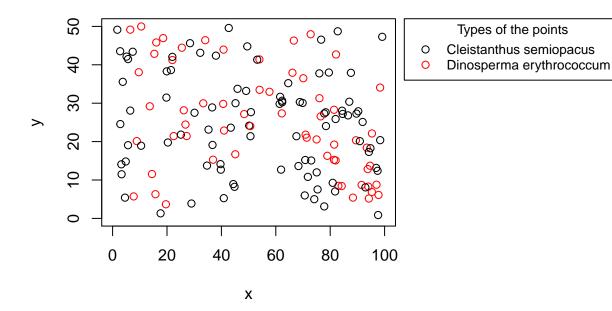
Queensland rainforest

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.

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,

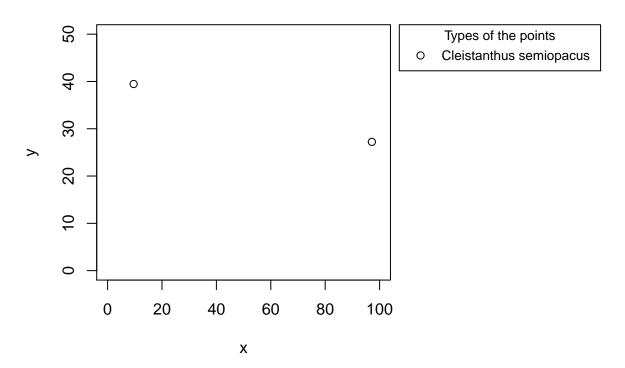
```
fit <- ppjsdm::gibbsm(configuration, window = window)

#> log_lambda_1 log_lambda_2 alpha_1_1 alpha_1_2 alpha_2_2

#> -3.978784 -4.431216 -3.936413 8.070406 -3.636435
```

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.

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)</pre>
```

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)</pre>
                                              alpha_1_2
#> log_lambda_1 log_lambda_2
                                alpha_1_1
                                                           alpha 2 2
#> -4.22694003 -4.85656953
                                             0.02636396
                                                          0.49504414
                               0.16730695
summary(fit)
#>
#> Call:
#> glm(formula = as.formula(gibbsm_data$formula), family = binomial(),
#>
       data = as.data.frame(gibbsm_data$data))
#>
#> Deviance Residuals:
#>
                 1Q
                      Median
                                            Max
                             -0.3920
#> -0.6315 -0.5781
                    -0.5220
                                         2.2930
#>
#> Coefficients:
#>
                Estimate Std. Error z value Pr(>|z|)
#> log_lambda_1 -4.22694
                            0.21507 -19.654 < 2e-16 ***
#> log_lambda_2 -4.85657
                            0.24659 -19.695
                                             < 2e-16 ***
#> alpha_1_1
                 0.16731
                            0.14357
                                      1.165
                                             0.24388
#> alpha_1_2
                 0.02636
                            0.11261
                                       0.234
                                             0.81489
```

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

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

Points in the configuration

