Fithian et al. (2014) NSW

```
library(maptools)
#> Loading required package: sp
#> Checking rgeos availability: FALSE
        Note: when rgeos is not available, polygon geometry
                                                                 computations in maptools depend on gpcl
#>
        which has a restricted licence. It is disabled by default;
        to enable gpclib, type gpclibPermit()
library(ppjsdm)
library(raster)
library(sf)
#> Linking to GEOS 3.6.2, GDAL 2.2.3, PROJ 4.9.3
library(spatstat)
#> Loading required package: spatstat.data
#> Loading required package: nlme
#>
#> Attaching package: 'nlme'
#> The following object is masked from 'package:raster':
#>
       getData
#> Loading required package: rpart
#> spatstat 1.62-2
                         (nickname: 'Shape-shifting lizard')
#> For an introduction to spatstat, type 'beginner'
#> Attaching package: 'spatstat'
#> The following objects are masked from 'package:raster':
#>
       area, rotate, shift
remove(list = ls())
source("../R/get_nsw.R")
set.seed(1)
```

This vignette explains how to use the ppjsdm package with the NSW dataset from Fithian et al. (2014). We begin by loading the data with only the most prevalent species.

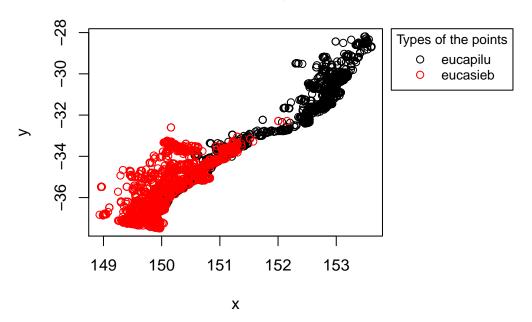
```
number_of_species <- 2 # Includes the most prevalent species from the plot

nsw <- get_nsw(prevalent = number_of_species)
configuration <- nsw$configuration
window <- nsw$window
covariates <- nsw$covariates</pre>
```

The point configuration is plotted below.

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

Points in the configuration



The matrix radii defined below models interaction radii within a species (on the diagonal), and between species (outside the diagonal).

```
radii <- matrix(0.1, number_of_species, number_of_species)</pre>
```

Fitting the model to the dataset is then quite easy.

```
fit <- ppjsdm::gibbsm(configuration,</pre>
                       window = window,
                       covariates = covariates,
                       model = "Geyer",
                       radius = radii,
                       use_glmnet = FALSE)
#>
                                     log_lambda_2
             log_lambda_1
                                                               alpha_1_1
                                                            2.462212e+00
#>
             6.141854e+00
                                    1.600987e+01
#>
                alpha_1_2
                                        alpha_2_2
                                                   unnamed\_covariate1\_1
                                                           -3.653586e-01
#>
             1.782363e-01
                                    3.007252e+00
#>
    unnamed\_covariate1\_2
                            unnamed\_covariate2\_1
                                                   unnamed\_covariate2\_2
#>
             1.340770e-01
                                   -1.431973e+00
                                                           -9.327638e-01
#>
    unnamed\_covariate3\_1
                            unnamed\_covariate3\_2
                                                   unnamed\_covariate4\_1
#>
            -6.614881e-02
                                   -1.846672e-01
                                                            4.371350e-03
#>
    unnamed\_covariate4\_2
                            unnamed\_covariate5\_1
                                                   unnamed\_covariate5\_2
#>
            -1.628112e-03
                                     1.241468e-01
                                                           -1.730611e-01
#>
    unnamed\_covariate6\_1
                            unnamed\_covariate6\_2
                                                   unnamed\_covariate7\_1
#>
             1.081237e-01
                                   -3.138805e-01
                                                            1.347198e+00
#>
    unnamed covariate7 2
                            unnamed covariate8 1
                                                   unnamed covariate8 2
#>
             5.839515e-01
                                    4.502081e-01
                                                            1.049417e+01
#>
    unnamed covariate9 1
                            unnamed covariate9 2 unnamed covariate10 1
#>
            -5.950067e-02
                                   -1.052719e-01
                                                           -2.725301e-02
#> unnamed_covariate10_2 unnamed_covariate11_1 unnamed_covariate11_2
```

```
#> -7.960943e-03 -1.405343e+00
                                                 -5.144668e+00
#> unnamed_covariate12_1 unnamed_covariate12_2 unnamed_covariate13_1
#> -1.757386e-02
                             -3.172272e-03
                                                 -3.300810e-02
#> unnamed_covariate13_2 unnamed_covariate14_1 unnamed_covariate14_2
#> 2.283882e-02
                             -9.260790e-05
                                                 -1.948034e-04
#> unnamed_covariate15_1 unnamed_covariate15_2
         -1.672047e-05
                             -6.903599e-05
summary(fit)
#>
#> Call:
#> NULL
#>
#> Deviance Residuals:
#> Min 1Q Median
                            3Q
#> -2.3029 -0.0907 -0.0420 -0.0152
                                    4.0224
#> Coefficients:
                        Estimate Std. Error z value Pr(>|z|)
#> log_lambda_1
                       6.142e+00 3.356e+00 1.830 0.067200 .
                      1.601e+01 3.315e+00 4.829 1.37e-06 ***
#> log_lambda_2
                       2.462e+00 1.248e-01 19.722 < 2e-16 ***
#> alpha 1 1
#> alpha 1 2
                       1.782e-01 4.194e-02
                                           4.250 2.14e-05 ***
                      3.007e+00 1.298e-01 23.175 < 2e-16 ***
#> alpha_2_2
#> unnamed_covariate1_1 -3.654e-01 9.221e-02 -3.962 7.42e-05 ***
#> unnamed_covariate1_2 1.341e-01 8.991e-02
                                           1.491 0.135905
#> unnamed_covariate2_1 -1.432e+00 7.339e-01 -1.951 0.051048 .
#> unnamed_covariate2_2 -9.328e-01 6.061e-01 -1.539 0.123837
#> unnamed_covariate3_1 -6.615e-02 4.183e-02 -1.582 0.113753
#> unnamed_covariate3_2 -1.847e-01 4.725e-02 -3.908 9.29e-05 ***
#> unnamed_covariate4_2 -1.628e-03 1.116e-03 -1.459 0.144593
#> unnamed_covariate5_1 1.241e-01 4.003e-02 3.101 0.001928 **
#> unnamed_covariate5_2 -1.731e-01 3.231e-02 -5.356 8.52e-08 ***
#> unnamed_covariate6_1 1.081e-01 1.347e-01 0.803 0.422080
#> unnamed_covariate6_2 -3.139e-01 1.438e-01 -2.183 0.029058 *
#> unnamed_covariate7_1 1.347e+00 9.677e-01 1.392 0.163855
#> unnamed_covariate7_2 5.840e-01 8.638e-01 0.676 0.499003
#> unnamed_covariate8_2 1.049e+01 1.248e+00 8.408 < 2e-16 ***
#> unnamed_covariate9_1 -5.950e-02 1.672e-02 -3.559 0.000372 ***
#> unnamed_covariate9_2 -1.053e-01 2.812e-02 -3.743 0.000182 ***
#> unnamed_covariate10_1 -2.725e-02 3.939e-03 -6.919 4.55e-12 ***
#> unnamed_covariate10_2 -7.961e-03  4.082e-03  -1.950  0.051149 .
#> unnamed_covariate11_1 -1.405e+00 5.655e-01 -2.485 0.012947 *
#> unnamed_covariate11_2 -5.145e+00 7.208e-01 -7.137 9.52e-13 ***
#> unnamed_covariate12_1 -1.757e-02 2.087e-03 -8.421 < 2e-16 ***
#> unnamed_covariate12_2 -3.172e-03 1.425e-03 -2.226 0.025998 *
#> unnamed_covariate13_1 -3.301e-02 1.316e-02 -2.509 0.012117 *
#> unnamed_covariate13_2 2.284e-02 1.443e-02 1.583 0.113372
#> unnamed_covariate14_1 -9.261e-05 3.400e-05 -2.724 0.006452 **
#> unnamed_covariate14_2 -1.948e-04 3.855e-05 -5.054 4.33e-07 ***
#> unnamed_covariate15_1 -1.672e-05 1.357e-05 -1.232 0.217826
#> unnamed_covariate15_2 -6.904e-05 1.419e-05 -4.866 1.14e-06 ***
```

```
#> ---

#> Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1

#>

#> (Dispersion parameter for binomial family taken to be 1)

#>

**Null deviance: 62010 on 24600 degrees of freedom

**Residual deviance: 11123 on 24565 degrees of freedom

**AIC: 11193

**>

**Number of Fisher Scoring iterations: 8
```