
SPMODEL: SPATIAL MODELING IN R

A PREPRINT

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Abstract

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Keywords Spatial covariance · Linear Model · Autoregressive model

1 Introduction

Here we describe the general role of spatial modeling, discuss existing software, and argue spmodel is a valuable contribution.

2 Background and Usage

2.1 Spatial Linear Models

```
R> spmod <- splm(y ~ x, exdata, "exponential", xcoord, ycoord)
R> summary(spmod)
```

Call:

```
splm(formula = y ~ x, data = exdata, spcov_type = "exponential",
      xcoord = xcoord, ycoord = ycoord)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-2.17030	-0.46466	0.07753	0.56000	2.41177

Coefficients (fixed):

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-0.39465	0.38302	-1.030	0.303
x	-0.07311	0.07374	-0.991	0.322

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Pseudo R-squared: 0.009929

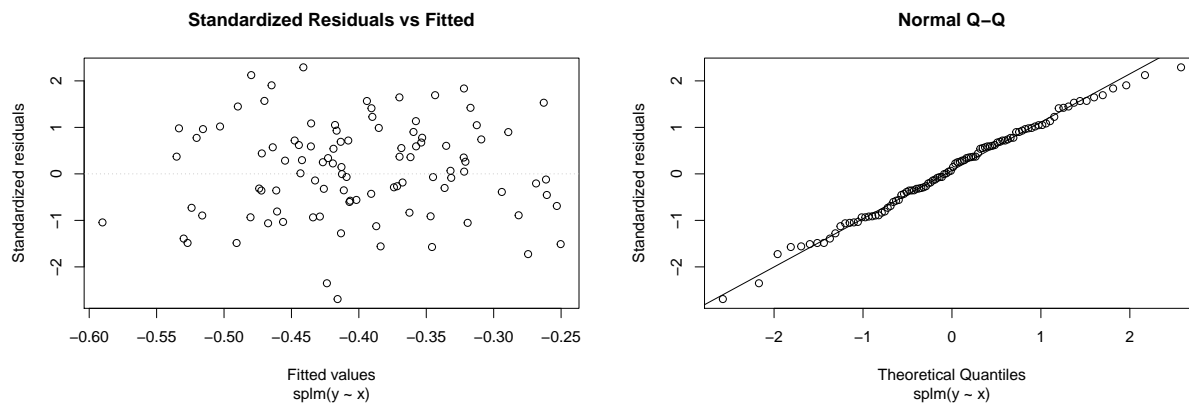
Coefficients (spatial covariance):

```
de    ie    range
0.6561 0.3874 1.3258
```

Spatial covariance type: exponential

- REML citations (Patterson and Thompson 1971; Harville 1977; Wolfinger, Tobias, and Sall 1994)
- SV-WLS citations (Cressie 1985, 1993)
- SV-CL citations (Curriero and Lele 1999)

```
R> plot(spmo, which = c(1, 2))
```



2.1.1 Anisotropy

```
R> spmo <- splm(y ~ x, exdata, "exponential", xcoord, ycoord, anisotropy = TRUE)
```

2.2 Prediction

```
R> spmo_preds <- predict(spmo, newexdata, interval = "prediction")
```

2.3 Model-fit Statistics

2.4 broom functions

```
R> tidy(spmo)
```

```
# A tibble: 2 x 5
  term      estimate std.error statistic p.value
<chr>      <dbl>    <dbl>    <dbl>    <dbl>
1 (Intercept) -0.391    0.364    -1.07    0.283
2 x          -0.0713   0.0738   -0.966   0.334
```

```
R> glance(spmo)
```

```
# A tibble: 1 x 9
  n      p npar value   AIC  AICc logLik deviance pseudo.r.squared
<int> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>   <dbl>         <dbl>
1  100    2    5  244.  256.  257.  -122.    98         0.00943
```

```
R> augment(spmo)
```

```
# A tibble: 100 x 11
      x      y xcoord ycoord group subgroup .fitted .resid .hat .cooksd .std.resid
  <dbl> <dbl> <dbl> <dbl> <fct> <fct>   <dbl> <dbl> <dbl> <dbl>   <dbl>
1 -1.12  0.602 -0.783  0.0723 1      1    -0.310  0.913  0.135  0.0642    1.05
2 -0.339  1.40 -0.847 -0.805 1      2    -0.366  1.77  0.0683  0.0882    1.67
3  2.67 -0.976 -0.0836  0.382 1      1    -0.581 -0.395  0.115  0.0500   -0.991
4  0.842 -0.981 -0.313  0.0923 1      2    -0.451 -0.531  0.0150  0.00740   -1.00
5 -0.283 -0.724  0.521  0.477 1      1    -0.370 -0.353  0.0483  0.00183   -0.282
6  0.438 -0.182  1.45  -0.703 1      2    -0.422  0.239  0.0644  0.00245    0.285
7  0.650  0.109  0.433  -1.18 1      1    -0.437  0.546  0.0287  0.00158    0.337
8 -0.315 -0.342 -0.0726 -2.59 1      2    -0.368  0.0262  0.0641  0.00238   -0.282
9  0.289 -2.59  1.52  -1.11 1      1    -0.411 -2.17  0.0217  0.0771   -2.70
10 0.315  0.383  1.21  1.05 1      2    -0.413  0.796  0.0452  0.0236    1.05
# ... with 90 more rows
```

```
R> augment(spmod, newdata = newexdata)
```

```
# A tibble: 10 x 6
      x xcoord ycoord group subgroup .fitted
  <dbl> <dbl> <dbl> <fct> <fct>   <dbl>
1  0.701 -0.857 -1.03  4      2      0.646
2 -1.45 -0.396  1.92  1      2     -0.954
3 -1.29 -2.05 -0.238 3      1     -0.233
4 -0.361  0.350 -0.243 2      2     -0.322
5 -0.802 -1.76 -1.98 1      2     -0.442
6  1.45 -1.04 -2.19 2      1     -0.504
7  0.218 -0.125  2.22 2      2     -0.972
8  0.480 -1.08  0.426 1      1     -0.0898
9  0.908  0.378 -2.29 4      1     -0.592
10 -0.883  1.57 -0.601 2      1     -0.173
```

2.5 Spatial Autoregressive Models

```
R> spmod <- spautorm(y ~ x, exdata_poly, "car")
```

2.6 Neighborhood Indexing for Big Data

```
R> spmod <- splm(y ~ x, exdata, "exponential", xcoord, ycoord, local = TRUE)
```

2.6.1 The local list

```
R> spmod <- splm(y ~ x, exdata, "exponential", xcoord, ycoord, local = list(parallel = TRUE))
```

2.7 Random Effects

```
R> spmod <- splm(y ~ x, exdata, "exponential", xcoord, ycoord, random = ~ group)
```

2.8 Partition Factors

```
R> spmod <- splm(y ~ x, exdata, "exponential", xcoord, ycoord, partition_factor = ~ group)
```

2.9 Initial Values and Known Values

```
R> spcov_params_init <- spcov_initial("exponential", de = 1, ie = 0, known = "ie")
R> spmod <- splm(
+   y ~ x,
+   exdata,
+   spcov_initial = spcov_params_init,
+   xcoord = xcoord,
```

```
+   ycoord = ycoord
+ )
```

2.10 Simulating Gaussian Random Variables

```
R> spcov_params_val <- spcov_params("exponential", de = 1, ie = 1, range = 2)
R> var <- sprnorm(spcov_params_val, data = exdata, xcoord = xcoord, ycoord = ycoord)
```

3 Discussion

Data and Code Availability

Acknowledgements

References

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