

# LMM with Spatial data

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## spatial distribution

```
Animals <- data(koalas)
```

## explaining data

Var name	Detail
pprim ssite	Percentage of Primary trees in each sub-site
psec ssite	Percentage of Secondary trees in each sub-site
phss 1km	Percentage of the landscape within 1km respectively, of each subsite that is highly suitable
phss 2.5km	Percentage of the landscape within 2.5 KM , respectively, of each subsite that is highly suitable
phss 5km	Percentage of the landscape within 5km, respectively, of each subsite that is highly suitable
pm 1km	Percentage of the landscape within 1 KM respectively, of each subsite that is marginal suitable
pm 2.5km	Percentage of the landscape within 2.5 KM respectively, of each subsite that is marginal suitable
pm 5km	Percentage of the landscape within 5 KM respectively, of each subsite that is marginal suitable
pdens 1km	Density (patches/100 ha) of habitat at 1 KM
pdens 2.5km	Density (patches/100 ha) of habitat at 2.5 Km
pdens 5km	Density (patches/100 ha) of habitat
edens 1km	Density (m/ha) of habitat patch edges
edens 2.5km	Density (m/ha) of habitat patch edges
edens 5km	Density (m/ha) of habitat patch edges
rdens 1km	Density (m/ha) of paved roads
rdens 2.5km	Density (m/ha) of paved roads
rdens 5km	Density (m/ha) of paved roads

# Data expolaration and discriptive

The can be two major issue

1. Col linearity and/or
2. Spatial auto correlation.

## Collinear check

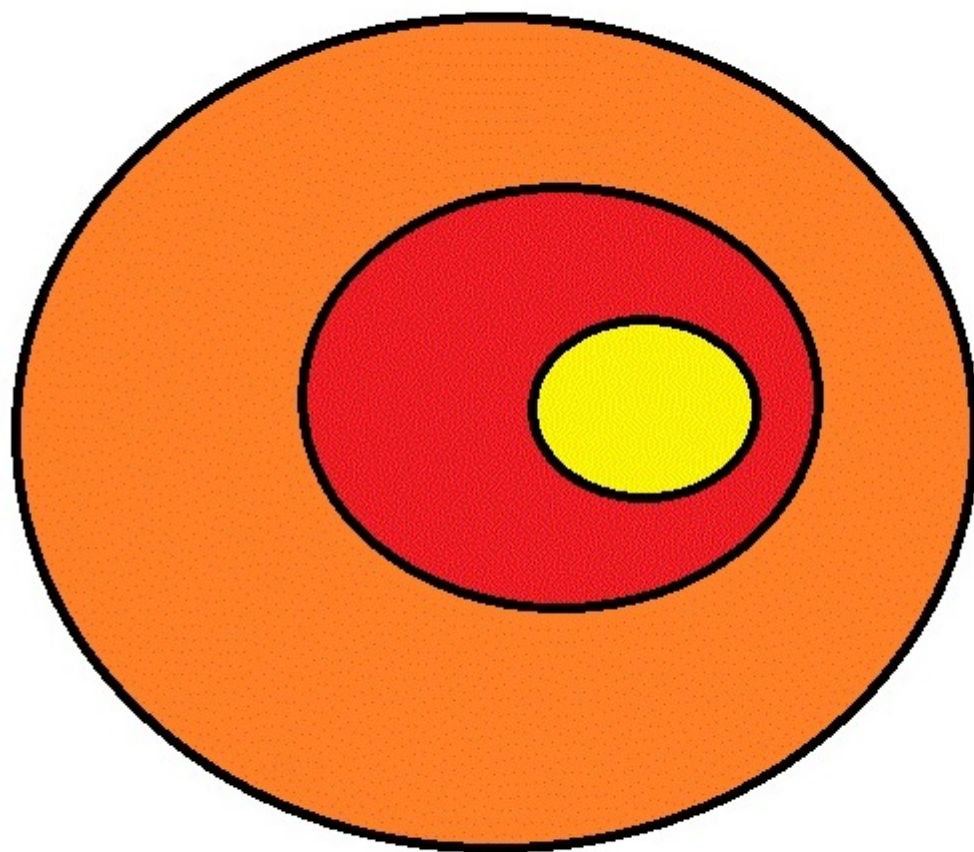
```
cor(Koalas[,6:22],method = "spearman")
```

```
##          pprim_ss site  psec_ss site  phss_5km  phss_2.5km  phss_1km
## pprim_ss site  1.000000000 -0.19120803 -0.04760121 -0.032780836 -0.020417112
## psec_ss site  -0.191208029 1.000000000 -0.10343151 -0.019252137 -0.046502154
## phss_5km      -0.047601206 -0.10343151  1.000000000  0.737515050  0.486706375
## phss_2.5km    -0.032780836 -0.01925214  0.73751505  1.000000000  0.740403325
## phss_1km      -0.020417112 -0.04650215  0.48670638  0.740403325  1.000000000
## pm_5km        -0.048185804 -0.24744572  0.22134042  0.041048407  0.171687372
## pm_2.5km      -0.083671041 -0.11792741  0.07360133 -0.045715484  0.008264348
## pm_1km        -0.027104365 -0.08654332 -0.11759707 -0.150455940 -0.220801736
## pdens_5km     -0.056547930  0.29468090 -0.45449481 -0.236224263 -0.247857240
## pdens_2.5km   -0.004534600  0.18902145 -0.44199460 -0.517430391 -0.501483157
## pdens_1km     -0.017110649  0.12294654 -0.18868017 -0.248650599 -0.453412507
## edens_5km     -0.159956521  0.30983775 -0.07457305 -0.007220872 -0.156310664
## edens_2.5km   -0.084139125  0.22307477 -0.17463724 -0.233201476 -0.392934799
## edens_1km     0.003885453  0.18287617 -0.18796229 -0.331345942 -0.508636124
## rdens_5km     -0.136356838  0.10771090  0.33350215  0.177156570 -0.027465879
## rdens_2.5km   -0.102642473  0.10068695  0.26134747  0.058231499 -0.074207327
## rdens_1km     -0.068976325  0.07343747  0.16980780  0.011513002 -0.023111583
##          pm_5km  pm_2.5km  pm_1km  pdens_5km  pdens_2.5km
## pprim_ss site -0.04818580 -0.083671041 -0.02710437 -0.05654793 -0.0045346
## psec_ss site  -0.24744572 -0.117927413 -0.08654332  0.29468090  0.1890214
## phss_5km       0.22134042  0.073601330 -0.11759707 -0.45449481 -0.4419946
## phss_2.5km     0.04104841 -0.045715484 -0.15045594 -0.23622426 -0.5174304
## phss_1km       0.17168737  0.008264348 -0.22080174 -0.24785724 -0.5014832
## pm_5km         1.00000000  0.587221746  0.24800403 -0.54598663 -0.3985728
## pm_2.5km       0.58722175  1.000000000  0.70879422 -0.35648307 -0.4194080
## pm_1km         0.24800403  0.708794217  1.00000000 -0.16858164 -0.1336637
## pdens_5km      -0.54598663 -0.356483070 -0.16858164  1.00000000  0.6214631
## pdens_2.5km    -0.39857280 -0.419407988 -0.13366371  0.62146309  1.0000000
## pdens_1km      -0.29777420 -0.253385544 -0.09933685  0.31095338  0.5422206
## edens_5km      -0.53281446 -0.428602762 -0.30420165  0.80090144  0.5515174
## edens_2.5km    -0.48947289 -0.402420683 -0.17982471  0.63474926  0.7377454
## edens_1km      -0.40283248 -0.285131240 -0.17488913  0.46422114  0.5964765
## rdens_5km      -0.24159421 -0.188272782 -0.09817424  0.17807288  0.1893438
## rdens_2.5km    -0.20496093 -0.228624776 -0.13984237  0.12040573  0.2391663
## rdens_1km      -0.21807094 -0.148326315 -0.13816014  0.07683865  0.1916224
##          pdens_1km  edens_5km  edens_2.5km  edens_1km  rdens_5km
## pprim_ss site -0.01711065 -0.159956521 -0.08413913  0.003885453 -0.13635684
## psec_ss site  0.12294654  0.309837751  0.22307477  0.182876165  0.10771090
## phss_5km      -0.18868017 -0.074573052 -0.17463724 -0.187962286  0.33350215
```

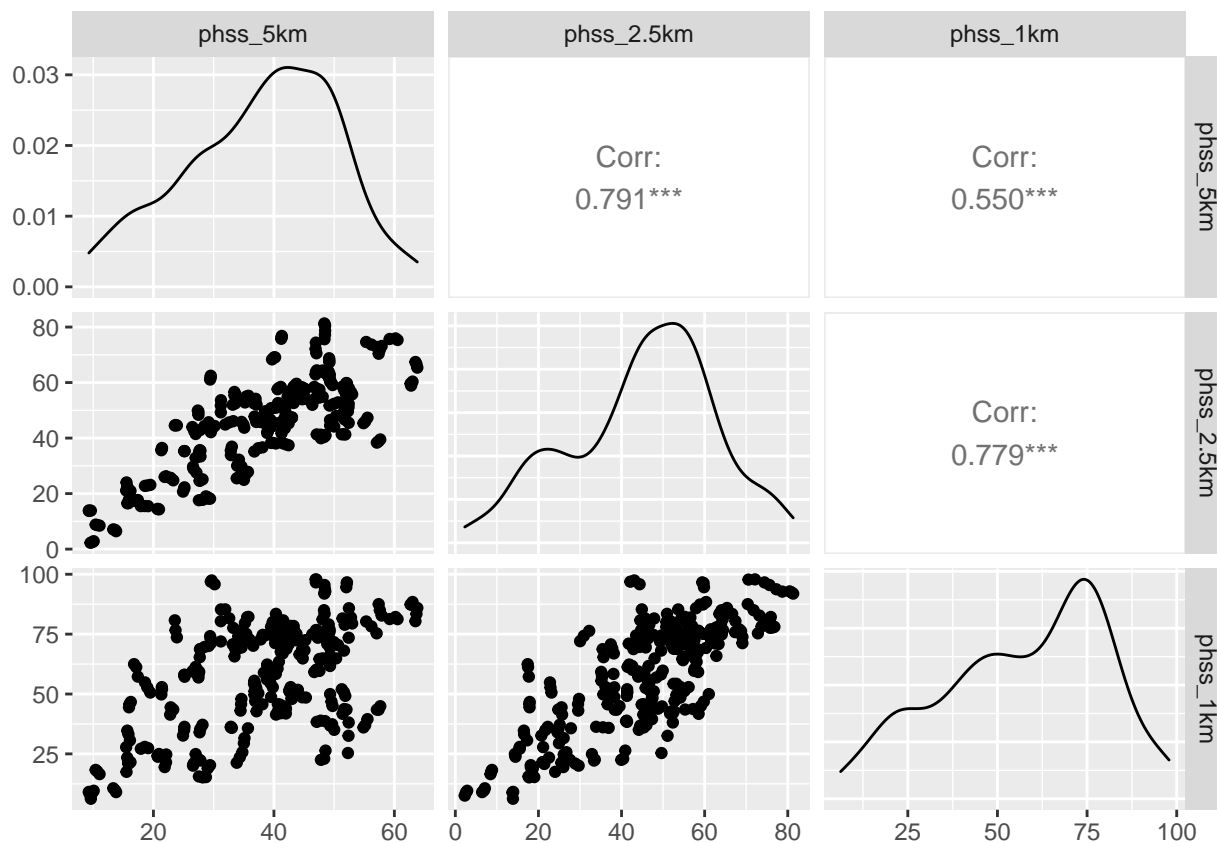
```
## phss_2.5km -0.24865060 -0.007220872 -0.23320148 -0.331345942 0.17715657
## phss_1km -0.45341251 -0.156310664 -0.39293480 -0.508636124 -0.02746588
## pm_5km -0.29777420 -0.532814459 -0.48947289 -0.402832483 -0.24159421
## pm_2.5km -0.25338554 -0.428602762 -0.40242068 -0.285131240 -0.18827278
## pm_1km -0.09933685 -0.304201652 -0.17982471 -0.174889125 -0.09817424
## pdens_5km 0.31095338 0.800901439 0.63474926 0.464221138 0.17807288
## pdens_2.5km 0.54222060 0.551517412 0.73774543 0.596476539 0.18934379
## pdens_1km 1.00000000 0.280339181 0.63902867 0.651218075 0.17236583
## edens_5km 0.28033918 1.000000000 0.74136688 0.492520941 0.51399634
## edens_2.5km 0.63902867 0.741366885 1.00000000 0.748572437 0.40945149
## edens_1km 0.65121807 0.492520941 0.74857244 1.000000000 0.28424550
## rdens_5km 0.17236583 0.513996339 0.40945149 0.284245500 1.00000000
## rdens_2.5km 0.18495223 0.435460921 0.37415201 0.281746295 0.90979343
## rdens_1km 0.15363443 0.336397239 0.30444877 0.253466246 0.72157159
##          rdens_2.5km  rdens_1km
## pprim_ssite -0.10264247 -0.06897632
## psec_ssite 0.10068695 0.07343747
## phss_5km 0.26134747 0.16980780
## phss_2.5km 0.05823150 0.01151300
## phss_1km -0.07420733 -0.02311158
## pm_5km -0.20496093 -0.21807094
## pm_2.5km -0.22862478 -0.14832632
## pm_1km -0.13984237 -0.13816014
## pdens_5km 0.12040573 0.07683865
## pdens_2.5km 0.23916632 0.19162241
## pdens_1km 0.18495223 0.15363443
## edens_5km 0.43546092 0.33639724
## edens_2.5km 0.37415201 0.30444877
## edens_1km 0.28174630 0.25346625
## rdens_5km 0.90979343 0.72157159
## rdens_2.5km 1.00000000 0.80657941
## rdens_1km 0.80657941 1.00000000
```

```
## An example of spatial colinear variable.
```

```
knitr::include_graphics("./imp_image/nested_spatial.jpg")
```



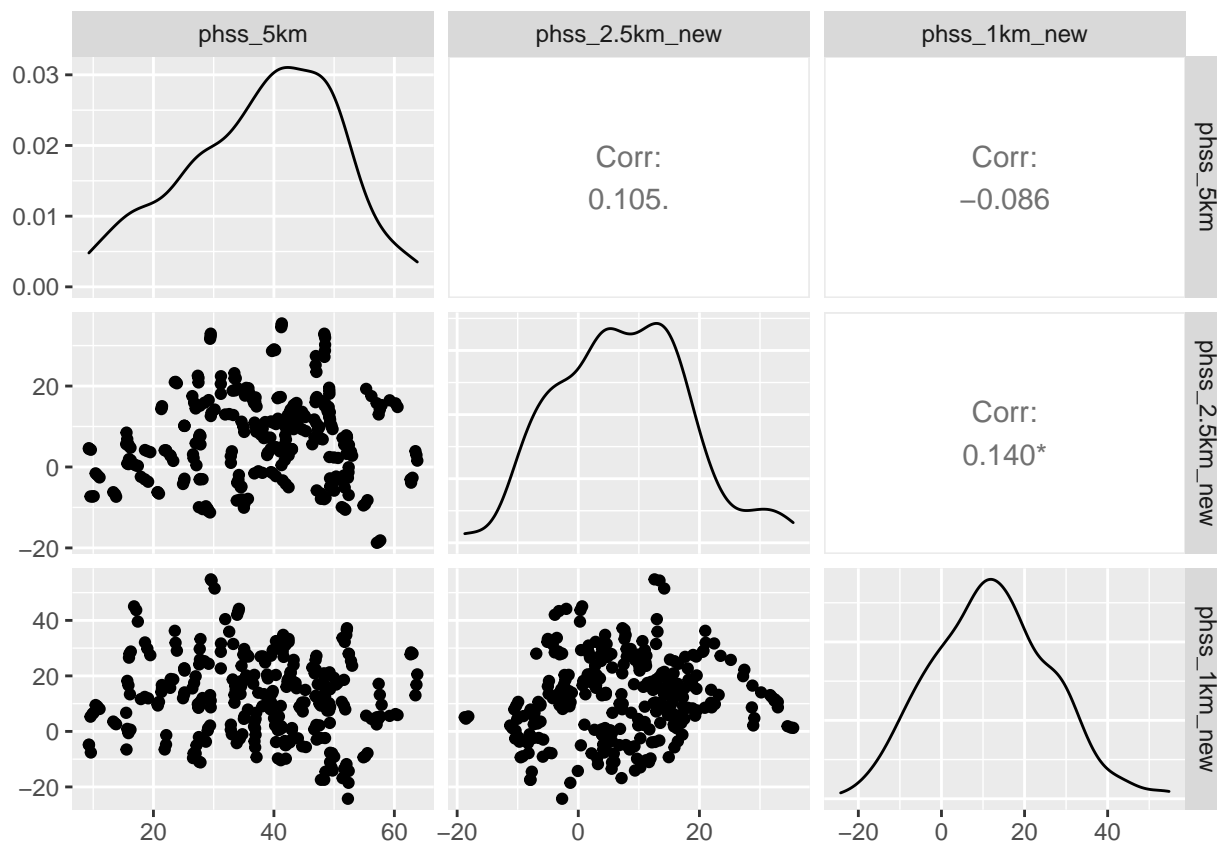
```
ggpairs(Koalas[,8:10])
```



We can see that similar variable at different distance are show co-linear behavior. This suggest that our data is spatial nested.

To overcome this variable we need to perform a calculation. This will divide out data such that new data will be 5 to 2.5 km, 2.5 to 1 KM

```
Koalas <- Koalas%>%mutate(phss_2.5km_new = phss_2.5km - phss_5km,
  phss_1km_new = phss_1km - phss_2.5km,
  pm_2.5km_new = pm_2.5km - pm_5km,
  pm_1km_new = pm_1km - pm_2.5km,
  pdens_2.5km_new = pdens_2.5km - pdens_5km,
  pdens_1km_new = pdens_1km - pdens_2.5km,
  edens_2.5km_new = edens_2.5km - edens_5km,
  edens_1km_new = edens_1km - edens_2.5km,
  rdens_2.5km_new = rdens_2.5km - rdens_5km,
  rdens_1km_new = rdens_1km - rdens_5km)
ggpairs(Koalas[,c(8,23,24)])
```



We can also check if co-linearity is impacting the model by using package called car and function call VCF

```
glm0 <- glm(presence~pprim_ssite+psec_ssite+
  phss_5km+phss_2.5km_new+phss_1km_new+
  pm_5km + pm_2.5km_new+pm_1km_new+
  pdens_5km+pdens_2.5km_new+pdens_1km_new+
  rdens_5km + rdens_2.5km_new +rdens_1km_new,
  data = Koalas, family = "binomial")

vif(glm0)
```

```
##      pprim_ssite      psec_ssite      phss_5km      phss_2.5km_new      phss_1km_new
##      1.121726      1.099841      3.196495      1.584206      1.495203
##      pm_5km      pm_2.5km_new      pm_1km_new      pdens_5km      pdens_2.5km_new
##      1.931660      1.575770      1.973593      2.474995      1.600666
##      pdens_1km_new      rdens_5km      rdens_2.5km_new      rdens_1km_new
##      1.273422      2.130524      1.676686      1.419444
```

```
glm0.old <- glm(presence~pprim_ssite+psec_ssite+
  phss_5km+phss_2.5km+phss_1km+
  pm_5km + pm_2.5km+pm_1km+
  pdens_5km+pdens_2.5km+pdens_1km+
  rdens_5km + rdens_2.5km +rdens_1km,
  data = Koalas, family = "binomial")
vif(glm0.old)
```

```
## pprim_ssite psec_ssite phss_5km phss_2.5km phss_1km pm_5km
## 1.121726 1.099841 5.725572 6.664211 3.872937 3.368125
## pm_2.5km pm_1km pdens_5km pdens_2.5km pdens_1km rdens_5km
## 4.530706 3.410261 3.290377 3.339844 1.653424 8.832575
## rdens_2.5km rdens_1km
## 9.329289 2.322678
```

```
glm0_2.5 <- glm(presence~pprim_ssite+psec_ssite+
  phss_2.5km_new+phss_1km_new+
  pm_2.5km_new+pm_1km_new+
  pdens_2.5km_new+pdens_1km_new+
  rdens_2.5km_new+rdens_1km_new,
  data = Koalas, family = "binomial")
vif(glm0_2.5)
```

```
## pprim_ssite psec_ssite phss_2.5km_new phss_1km_new pm_2.5km_new
## 1.081343 1.058178 1.488893 1.341732 1.475352
## pm_1km_new pdens_2.5km_new pdens_1km_new rdens_2.5km_new rdens_1km_new
## 1.459823 1.515581 1.240205 1.475457 1.387168
```

```
glm0_1 <- glm(presence~pprim_ssite+psec_ssite+
  phss_1km_new+
  pm_1km_new+
  pdens_1km_new+
  rdens_1km_new,
  data = Koalas, family = "binomial")
vif(glm0_1)
```

```
## pprim_ssite psec_ssite phss_1km_new pm_1km_new pdens_1km_new
## 1.053231 1.031041 1.341394 1.168002 1.202651
## rdens_1km_new
## 1.016711
```

```
## the value should be below 10, new study even less almost close to 1.5-2
```

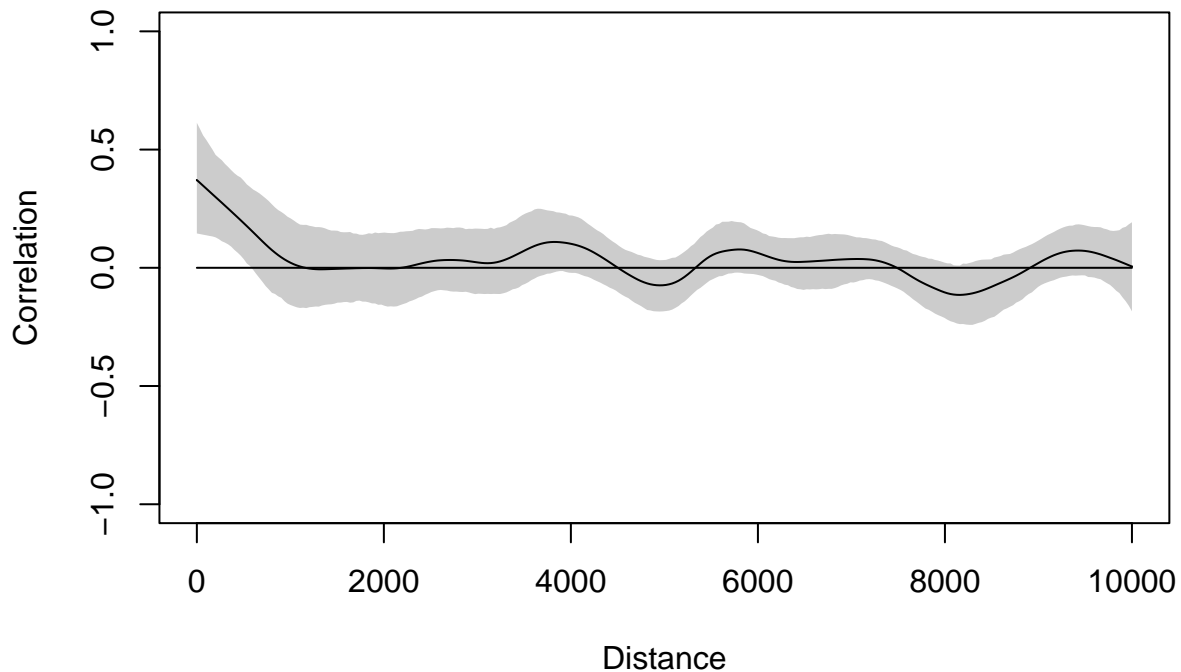
We can also test for data for 2.5 km and 1 km to check how co-linearity behave.

**spatial auto correaltion.**

```
library(ncf)
Correlog <- spline.correlog(x = Koalas[, "easting"],
  y = Koalas[, "northing"],
  z = Koalas[, "presence"], xmax = 10000)
```

```
## 100 of 1000 200 of 1000 300 of 1000 400 of 1000 500 of 1000 600 of 1000 700 of 1000 800
```

```
plot(Correlog)
```



## Model selection

```
glm_sec <- glm(presence~.,
              data = Koalas[,c("presence", "pprim_ssite", "psec_ssite",
                               "phss_5km", "phss_2.5km_new", "phss_1km_new",
                               "pm_5km", "pm_2.5km_new", "pm_1km_new",
                               "pdens_5km", "pdens_2.5km_new", "pdens_1km_new",
                               "rdens_5km", "rdens_2.5km_new", "rdens_1km_new",
                               "edens_5km", "edens_2.5km_new", "edens_2.5km_new.1")],
              family = "binomial")
Step_model <- stepAIC(glm_sec, trace = T, direction = "both")
```

```
## Start:  AIC=392.03
## presence ~ pprim_ssite + psec_ssite + phss_5km + phss_2.5km_new +
##           phss_1km_new + pm_5km + pm_2.5km_new + pm_1km_new + pdens_5km +
##           pdens_2.5km_new + pdens_1km_new + rdens_5km + rdens_2.5km_new +
##           rdens_1km_new + edens_5km + edens_2.5km_new + edens_2.5km_new.1
##
##
## Step:  AIC=392.03
## presence ~ pprim_ssite + psec_ssite + phss_5km + phss_2.5km_new +
##           phss_1km_new + pm_5km + pm_2.5km_new + pm_1km_new + pdens_5km +
##           pdens_2.5km_new + pdens_1km_new + rdens_5km + rdens_2.5km_new +
```



```

##      rdens_1km_new + edens_5km + edens_2.5km_new
##
##              Df Deviance    AIC
## - rdens_1km_new      1   358.04 390.04
## - pm_5km              1   358.07 390.07
## - phss_5km            1   358.08 390.08
## - pdens_5km           1   358.19 390.19
## - psec_ssite          1   358.36 390.36
## - rdens_5km           1   358.54 390.54
## - pdens_2.5km_new     1   358.74 390.74
## - edens_5km           1   359.09 391.09
## - pdens_1km_new       1   359.20 391.20
## - phss_1km_new        1   359.22 391.22
## - rdens_2.5km_new     1   359.28 391.28
## - edens_2.5km_new     1   359.67 391.67
## <none>                 358.03 392.03
## - phss_2.5km_new      1   360.56 392.56
## - pm_2.5km_new        1   361.55 393.55
## - pm_1km_new          1   362.56 394.56
## - pprim_ssite         1   379.05 411.05
##
## Step:  AIC=390.04
## presence ~ pprim_ssite + psec_ssite + phss_5km + phss_2.5km_new +
##           phss_1km_new + pm_5km + pm_2.5km_new + pm_1km_new + pdens_5km +
##           pdens_2.5km_new + pdens_1km_new + rdens_5km + rdens_2.5km_new +
##           edens_5km + edens_2.5km_new
##
##              Df Deviance    AIC
## - pm_5km              1   358.07 388.07
## - phss_5km            1   358.09 388.09
## - pdens_5km           1   358.19 388.19
## - psec_ssite          1   358.37 388.37
## - rdens_5km           1   358.54 388.54
## - pdens_2.5km_new     1   358.74 388.74
## - edens_5km           1   359.10 389.10
## - pdens_1km_new       1   359.20 389.20
## - phss_1km_new        1   359.22 389.22
## - edens_2.5km_new     1   359.67 389.67
## - rdens_2.5km_new     1   359.69 389.69
## <none>                 358.04 390.04
## - phss_2.5km_new      1   360.57 390.57
## - pm_2.5km_new        1   361.56 391.56
## + rdens_1km_new       1   358.03 392.03
## - pm_1km_new          1   362.79 392.79
## - pprim_ssite         1   379.15 409.15
##
## Step:  AIC=388.07
## presence ~ pprim_ssite + psec_ssite + phss_5km + phss_2.5km_new +
##           phss_1km_new + pm_2.5km_new + pm_1km_new + pdens_5km + pdens_2.5km_new +
##           pdens_1km_new + rdens_5km + rdens_2.5km_new + edens_5km +
##           edens_2.5km_new
##
##              Df Deviance    AIC
## - phss_5km            1   358.10 386.10

```

```

## - pdens_5km      1    358.22 386.22
## - psec_ssite     1    358.43 386.43
## - rdens_5km      1    358.55 386.55
## - pdens_2.5km_new 1    358.77 386.77
## - phss_1km_new   1    359.24 387.24
## - edens_5km      1    359.27 387.27
## - pdens_1km_new  1    359.30 387.30
## - edens_2.5km_new 1    359.70 387.70
## - rdens_2.5km_new 1    359.73 387.73
## <none>           358.07 388.07
## - phss_2.5km_new 1    360.89 388.89
## - pm_2.5km_new   1    361.56 389.56
## + pm_5km         1    358.04 390.04
## + rdens_1km_new  1    358.07 390.07
## - pm_1km_new     1    363.11 391.11
## - pprim_ssite    1    379.66 407.66
##
## Step: AIC=386.1
## presence ~ pprim_ssite + psec_ssite + phss_2.5km_new + phss_1km_new +
##           pm_2.5km_new + pm_1km_new + pdens_5km + pdens_2.5km_new +
##           pdens_1km_new + rdens_5km + rdens_2.5km_new + edens_5km +
##           edens_2.5km_new
##
##           Df Deviance   AIC
## - psec_ssite      1    358.45 384.45
## - pdens_5km       1    358.53 384.53
## - rdens_5km       1    358.55 384.55
## - pdens_2.5km_new 1    358.77 384.77
## - phss_1km_new    1    359.24 385.24
## - pdens_1km_new    1    359.36 385.36
## - edens_5km       1    359.52 385.52
## - rdens_2.5km_new 1    359.78 385.78
## - edens_2.5km_new 1    359.81 385.81
## <none>            358.10 386.10
## - phss_2.5km_new  1    360.96 386.96
## - pm_2.5km_new    1    361.60 387.60
## + phss_5km        1    358.07 388.07
## + pm_5km          1    358.09 388.09
## + rdens_1km_new    1    358.09 388.09
## - pm_1km_new      1    363.84 389.84
## - pprim_ssite     1    379.66 405.66
##
## Step: AIC=384.45
## presence ~ pprim_ssite + phss_2.5km_new + phss_1km_new + pm_2.5km_new +
##           pm_1km_new + pdens_5km + pdens_2.5km_new + pdens_1km_new +
##           rdens_5km + rdens_2.5km_new + edens_5km + edens_2.5km_new
##
##           Df Deviance   AIC
## - pdens_5km       1    358.87 382.87
## - rdens_5km       1    358.91 382.91
## - pdens_2.5km_new 1    359.15 383.15
## - phss_1km_new    1    359.58 383.58
## - pdens_1km_new    1    359.66 383.66
## - edens_5km       1    359.96 383.96

```

```

## - rdens_2.5km_new 1 359.98 383.98
## - edens_2.5km_new 1 360.03 384.03
## <none> 358.45 384.45
## - phss_2.5km_new 1 361.67 385.67
## - pm_2.5km_new 1 362.05 386.05
## + psec_ssite 1 358.10 386.10
## + pm_5km 1 358.41 386.41
## + phss_5km 1 358.43 386.43
## + rdens_1km_new 1 358.43 386.43
## - pm_1km_new 1 364.13 388.13
## - pprim_ssite 1 379.66 403.66
##
## Step: AIC=382.87
## presence ~ pprim_ssite + phss_2.5km_new + phss_1km_new + pm_2.5km_new +
## pm_1km_new + pdens_2.5km_new + pdens_1km_new + rdens_5km +
## rdens_2.5km_new + edens_5km + edens_2.5km_new
##
## Df Deviance AIC
## - rdens_5km 1 359.02 381.02
## - pdens_2.5km_new 1 359.76 381.76
## - phss_1km_new 1 359.79 381.79
## - pdens_1km_new 1 360.10 382.10
## - edens_5km 1 360.46 382.46
## - rdens_2.5km_new 1 360.71 382.71
## - edens_2.5km_new 1 360.84 382.84
## <none> 358.87 382.87
## - phss_2.5km_new 1 361.93 383.93
## - pm_2.5km_new 1 362.33 384.33
## + pdens_5km 1 358.45 384.45
## + psec_ssite 1 358.53 384.53
## + phss_5km 1 358.60 384.60
## + pm_5km 1 358.86 384.86
## + rdens_1km_new 1 358.86 384.86
## - pm_1km_new 1 364.17 386.17
## - pprim_ssite 1 379.73 401.73
##
## Step: AIC=381.02
## presence ~ pprim_ssite + phss_2.5km_new + phss_1km_new + pm_2.5km_new +
## pm_1km_new + pdens_2.5km_new + pdens_1km_new + rdens_2.5km_new +
## edens_5km + edens_2.5km_new
##
## Df Deviance AIC
## - pdens_2.5km_new 1 359.93 379.93
## - phss_1km_new 1 360.02 380.02
## - pdens_1km_new 1 360.17 380.17
## - edens_5km 1 360.59 380.59
## - edens_2.5km_new 1 360.97 380.97
## <none> 359.02 381.02
## - rdens_2.5km_new 1 362.01 382.01
## - phss_2.5km_new 1 362.05 382.05
## - pm_2.5km_new 1 362.54 382.54
## + psec_ssite 1 358.67 382.67
## + rdens_5km 1 358.87 382.87
## + pdens_5km 1 358.91 382.91

```

```

## + phss_5km          1    358.97 382.97
## + rdens_1km_new     1    359.01 383.01
## + pm_5km            1    359.02 383.02
## - pm_1km_new        1    364.21 384.21
## - pprim_ssitesite   1    379.87 399.87
##
## Step: AIC=379.93
## presence ~ pprim_ssitesite + phss_2.5km_new + phss_1km_new + pm_2.5km_new +
##           pm_1km_new + pdens_1km_new + rdens_2.5km_new + edens_5km +
##           edens_2.5km_new
##
##           Df Deviance    AIC
## - pdens_1km_new     1    360.71 378.71
## - phss_1km_new       1    360.80 378.80
## - edens_2.5km_new    1    361.03 379.03
## - edens_5km          1    361.90 379.90
## <none>                359.93 379.93
## - phss_2.5km_new     1    362.22 380.22
## - rdens_2.5km_new    1    362.61 380.61
## - pm_2.5km_new       1    362.70 380.70
## + pdens_2.5km_new    1    359.02 381.02
## + psec_ssitesite     1    359.54 381.54
## + pdens_5km          1    359.73 381.73
## + rdens_5km          1    359.76 381.76
## + phss_5km           1    359.89 381.89
## + pm_5km             1    359.92 381.92
## + rdens_1km_new      1    359.92 381.92
## - pm_1km_new         1    365.38 383.38
## - pprim_ssitesite    1    380.53 398.53
##
## Step: AIC=378.71
## presence ~ pprim_ssitesite + phss_2.5km_new + phss_1km_new + pm_2.5km_new +
##           pm_1km_new + rdens_2.5km_new + edens_5km + edens_2.5km_new
##
##           Df Deviance    AIC
## - phss_1km_new       1    361.15 377.15
## - edens_2.5km_new     1    361.17 377.17
## <none>                360.71 378.71
## - edens_5km          1    362.85 378.85
## - phss_2.5km_new     1    363.41 379.41
## - rdens_2.5km_new    1    363.60 379.60
## + pdens_1km_new      1    359.93 379.93
## - pm_2.5km_new       1    364.02 380.02
## + pdens_2.5km_new    1    360.17 380.17
## + psec_ssitesite     1    360.38 380.38
## + pdens_5km          1    360.48 380.48
## + rdens_5km          1    360.62 380.62
## + phss_5km           1    360.64 380.64
## + pm_5km             1    360.69 380.69
## + rdens_1km_new      1    360.71 380.71
## - pm_1km_new         1    365.54 381.54
## - pprim_ssitesite    1    380.87 396.87
##
## Step: AIC=377.15

```

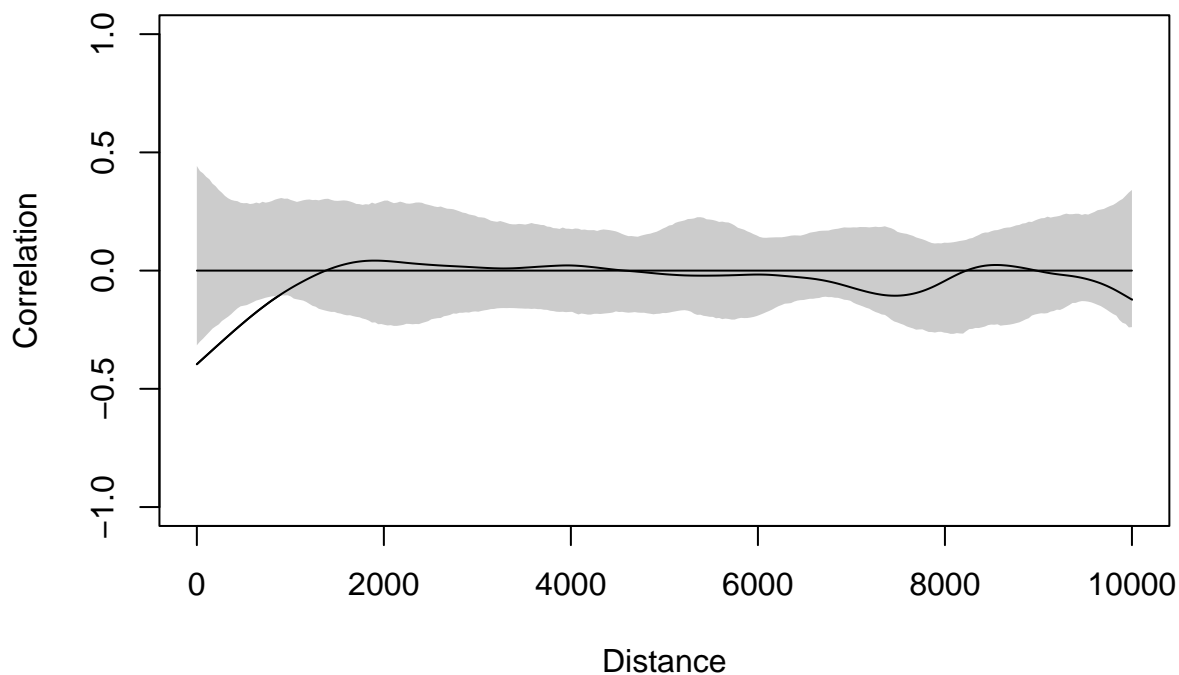
```
## presence ~ pprim_ssite + phss_2.5km_new + pm_2.5km_new + pm_1km_new +
##      rdens_2.5km_new + edens_5km + edens_2.5km_new
##
##           Df Deviance   AIC
## - edens_2.5km_new  1   361.83 375.83
## - edens_5km        1   362.99 376.99
## <none>              361.15 377.15
## - phss_2.5km_new   1   363.85 377.85
## - rdens_2.5km_new  1   364.06 378.06
## - pm_2.5km_new     1   364.30 378.30
## + pdens_2.5km_new  1   360.60 378.60
## + phss_1km_new     1   360.71 378.71
## + pdens_1km_new    1   360.80 378.80
## + psec_ssite       1   360.80 378.80
## + rdens_5km        1   361.00 379.00
## + pdens_5km        1   361.05 379.05
## + pm_5km           1   361.12 379.12
## + phss_5km         1   361.15 379.15
## + rdens_1km_new    1   361.15 379.15
## - pm_1km_new       1   365.57 379.57
## - pprim_ssite      1   380.99 394.99
##
## Step:  AIC=375.83
## presence ~ pprim_ssite + phss_2.5km_new + pm_2.5km_new + pm_1km_new +
##      rdens_2.5km_new + edens_5km
##
##           Df Deviance   AIC
## <none>              361.83 375.83
## - edens_5km        1   363.96 375.96
## - rdens_2.5km_new  1   364.60 376.60
## + edens_2.5km_new  1   361.15 377.15
## + edens_2.5km_new.1 1   361.15 377.15
## + phss_1km_new     1   361.17 377.17
## - pm_2.5km_new     1   365.54 377.54
## + psec_ssite       1   361.59 377.59
## + rdens_5km        1   361.64 377.64
## + pdens_5km        1   361.69 377.69
## + pdens_2.5km_new  1   361.79 377.79
## + phss_5km         1   361.80 377.80
## + rdens_1km_new    1   361.83 377.83
## + pm_5km           1   361.83 377.83
## + pdens_1km_new    1   361.83 377.83
## - pm_1km_new       1   365.89 377.89
## - phss_2.5km_new   1   366.02 378.02
## - pprim_ssite      1   381.18 393.18
```

#### Step\_model

```
##
## Call:  glm(formula = presence ~ pprim_ssite + phss_2.5km_new + pm_2.5km_new +
##      pm_1km_new + rdens_2.5km_new + edens_5km, family = "binomial",
##      data = Koalas[, c("presence", "pprim_ssite", "psec_ssite",
##      "phss_5km", "phss_2.5km_new", "phss_1km_new", "pm_5km",
##      "pm_2.5km_new", "pm_1km_new", "pdens_5km", "pdens_2.5km_new",
```

```
##      "pdens_1km_new", "rdens_5km", "rdens_2.5km_new", "rdens_1km_new",
##      "edens_5km", "edens_2.5km_new", "edens_2.5km_new"]])
##
## Coefficients:
##      (Intercept)      pprim_ssite    phss_2.5km_new      pm_2.5km_new
##      -1.6250136       0.0441092       0.0260414       0.0450319
##      pm_1km_new    rdens_2.5km_new      edens_5km
##      0.0235568      -0.0009315       0.0190902
##
## Degrees of Freedom: 299 Total (i.e. Null); 293 Residual
## Null Deviance:      398.4
## Residual Deviance: 361.8      AIC: 375.8

## 100 of 1000 200 of 1000 300 of 1000 400 of 1000 500 of 1000 600 of 1000 700 of 1000 800 of 1000
```



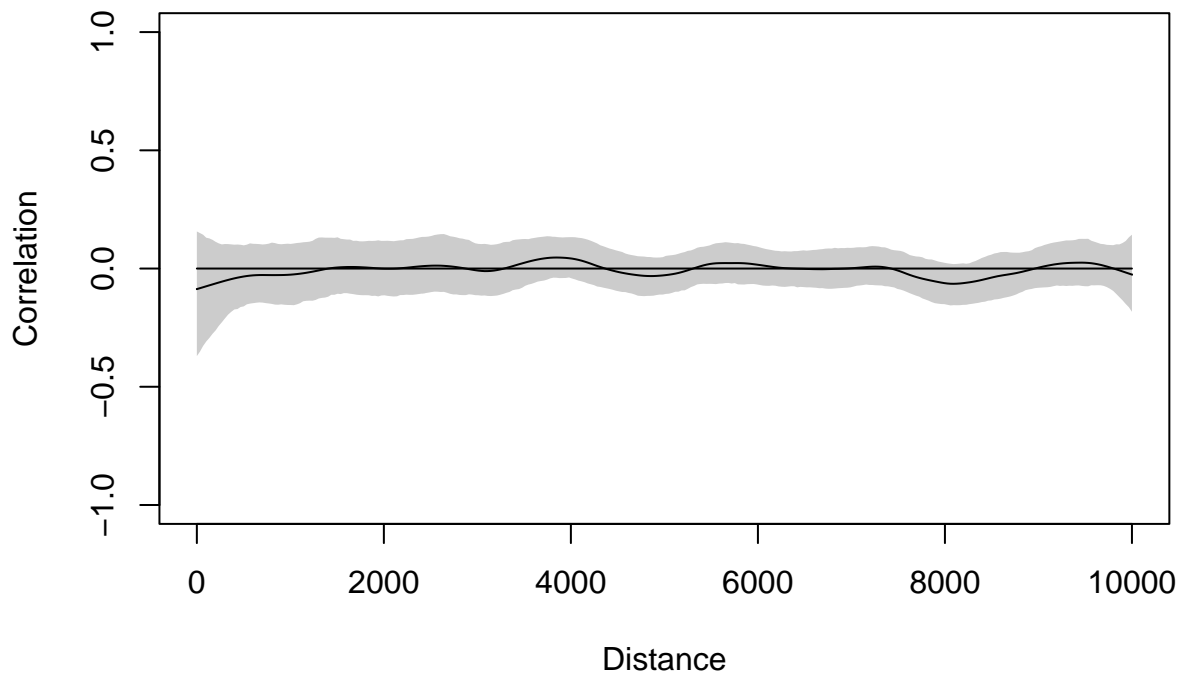
```
Best_model <- glmer(formula = presence ~ pprim_ssite + phss_2.5km_new + pm_2.5km_new +
  pm_1km_new + rdens_2.5km_new + edens_5km + (1|site), family = "binomial", data = Koalas)
```

```
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, : Model is nearly unidentifiable:
## - Rescale variables?
```

```
Best_auto <- spline.correlog(x = Koalas[, "easting"],
  y = Koalas[, "northing"],
  z = residuals(Best_model), xmax = 10000)
```

```
## 100 of 1000 200 of 1000 300 of 1000 400 of 1000 500 of 1000 600 of 1000 700 of 1000 800
```

```
plot(Best_auto)
```



```
summary(Best_model)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: presence ~ pprim_ssite + phss_2.5km_new + pm_2.5km_new + pm_1km_new +
##          rdens_2.5km_new + edens_5km + (1 | site)
## Data: Koalas
##
##      AIC      BIC  logLik deviance df.resid
##    361.3    391.0   -172.7    345.3     292
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.0124 -0.5170 -0.2981  0.6442  2.0660
##
## Random effects:
##   Groups Name            Variance Std.Dev.
##   site   (Intercept) 1.993      1.412
## Number of obs: 300, groups: site, 100
##
```

```
## Fixed effects:
##               Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -2.3409900  0.7867039  -2.976 0.002923 **
## pprim_ssite    0.0652716  0.0168439   3.875 0.000107 ***
## phss_2.5km_new 0.0343386  0.0209165   1.642 0.100653
## pm_2.5km_new   0.0583297  0.0386688   1.508 0.131441
## pm_1km_new     0.0328777  0.0198924   1.653 0.098376 .
## rdens_2.5km_new -0.0013848  0.0009209  -1.504 0.132656
## edens_5km      0.0288996  0.0219133   1.319 0.187230
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##               (Intr) pprm_s ph_2.5_ pm_2.5_ pm_1k_ r_2.5_
## pprim_ssite -0.453
## phss_2.5km_ -0.274  0.169
## pm_2.5km_nw -0.202  0.083  0.138
## pm_1km_new  -0.113  0.169  0.149  -0.347
## rdns_2.5km_ -0.084 -0.073  0.237  0.120  -0.047
## edens_5km   -0.897  0.240  0.001  0.123  0.052 -0.007
## optimizer (Nelder_Mead) convergence code: 0 (OK)
## Model is nearly unidentifiable: very large eigenvalue
## - Rescale variables?
```

```
pred_best <- ggpredict(Best_model,terms = "pprim_ssite",type = "re")
plot <- ggplot(data=pred_best,aes(x=x,y=predicted))
plot + geom_line() +
  geom_ribbon(aes(ymin=predicted-std.error,ymax=predicted+std.error),alpha=0.2)+
  geom_jitter(data = Koalas,aes(x=pprim_ssite,y=presence))
```



