

Spatial Modeling Assignment

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```
library(vegan)
```

```
## Warning: package 'vegan' was built under R version 3.5.3
```

```
## Loading required package: permute
```

```
## Warning: package 'permute' was built under R version 3.5.3
```

```
## Loading required package: lattice
```

```
## Warning: package 'lattice' was built under R version 3.5.3
```

```
## This is vegan 2.5-6
```

```
data(BCI)
## UTM Coordinates (in metres)
BCI_xy = data.frame(x = rep(seq(625754, 626654, by=100), each=5),
                    y = rep(seq(1011569, 1011969, by=100), len=50))
?BCI
```

```
## starting httpd help server ...
```

```
## done
```

```
head(BCI)
```

```

## Abarema.macradenia Vachellia.melanoceras Acalypha.diversifolia
## 1 0 0 0
## 2 0 0 0
## 3 0 0 0
## 4 0 0 0
## 5 0 0 0
## 6 0 0 0
## Acalypha.macrostachya Adelia.triloba Aegiphila.panamensis
## 1 0 0 0
## 2 0 0 0
## 3 0 0 0
## 4 0 3 0
## 5 0 1 1
## 6 0 0 0
## Alchornea.costaricensis Alchornea.latifolia Alibertia.edulis
## 1 2 0 0
## 2 1 0 0
## 3 2 0 0
## 4 18 0 0
## 5 3 0 0
## 6 2 1 0
## Allophylus.psilospermus Alseis.blackiana Amaioua.corymbosa
## 1 0 25 0
## 2 0 26 0
## 3 0 18 0
## 4 0 23 0
## 5 1 16 0
## 6 0 14 0
## Anacardium.excelsum Andira.inermis Annona.spraguei Apeiba.glabra
## 1 0 0 1 13
## 2 0 0 0 12
## 3 0 0 1 6
## 4 0 0 0 3
## 5 0 1 0 4
## 6 0 1 0 10
## Apeiba.tibourbou Aspidosperma.desmanthum Astrocaryum.standleyanum
## 1 2 0 0
## 2 0 0 2
## 3 1 0 1
## 4 1 1 5
## 5 0 1 6
## 6 0 1 2
## Astronium.graveolens Attalea.butyracea Banara.guianensis
## 1 6 0 0
## 2 0 1 0
## 3 1 0 0
## 4 3 0 0
## 5 0 0 0
## 6 1 1 0
## Beilschmiedia.pendula Brosimum.alicastrum Brosimum.guianense
## 1 4 5 0
## 2 5 2 0
## 3 7 4 0

```

## 4	5	3	0	
## 5	8	2	0	
## 6	6	2	0	
##	Calophyllum.longifolium	Casearia.aculeata	Casearia.arborea	
## 1	0	0	1	
## 2	2	0	1	
## 3	0	0	3	
## 4	2	0	2	
## 5	1	0	4	
## 6	2	0	1	
##	Casearia.commersoniana	Casearia.guianensis	Casearia.sylvestris	
## 1	0	0	2	
## 2	0	0	1	
## 3	1	0	0	
## 4	0	0	0	
## 5	1	0	0	
## 6	0	0	3	
##	Cassipourea.guianensis	Cavanillesia.platanifolia	Cecropia.insignis	
## 1	2	0	12	
## 2	0	0	5	
## 3	1	0	7	
## 4	1	0	17	
## 5	3	0	21	
## 6	4	0	4	
##	Cecropia.obtusifolia	Cedrela.odorata	Ceiba.pentandra	Celtis.schippii
## 1	0	0	0	0
## 2	0	0	1	0
## 3	0	0	1	0
## 4	0	0	0	2
## 5	1	0	1	2
## 6	0	0	0	0
##	Cespedesia.spathulata	Chamguava.schippii	Chimarrhis.parviflora	
## 1	0	0	0	
## 2	0	0	0	
## 3	0	0	0	
## 4	0	0	0	
## 5	0	0	0	
## 6	0	0	0	
##	Maclura.tinctoria	Chrysochlamys.eclipses	Chrysophyllum.argenteum	
## 1	0	0	4	
## 2	0	0	1	
## 3	0	0	2	
## 4	0	0	2	
## 5	0	0	6	
## 6	0	0	2	
##	Chrysophyllum.cainito	Coccoloba.coronata	Coccoloba.manzinellensis	
## 1	0	0	0	
## 2	0	0	0	
## 3	0	0	0	
## 4	0	1	0	
## 5	0	2	0	
## 6	0	0	0	
##	Colubrina.glandulosa	Cordia.alliodora	Cordia.bicolor	Cordia.lasiocalyx
## 1	0	2	12	8

## 2	0	3	14	6
## 3	0	3	35	6
## 4	0	7	23	11
## 5	0	1	13	7
## 6	0	1	7	6
##	Coussarea.curvigemma Croton.billbergianus Cupania.cinerea			
## 1	0	2	0	
## 2	0	2	0	
## 3	0	0	0	
## 4	1	11	0	
## 5	0	6	0	
## 6	2	0	0	
##	Cupania.latifolia Cupania.rufescens Cupania.seemannii			
## 1	0	0	2	
## 2	0	0	2	
## 3	0	0	1	
## 4	1	0	0	
## 5	0	0	3	
## 6	0	0	0	
##	Dendropanax.arboreus Desmopsis.panamensis Diospyros.artanthifolia			
## 1	0	0		1
## 2	3	0		1
## 3	6	4		1
## 4	0	0		1
## 5	5	0		0
## 6	2	0		0
##	Dipteryx.oleifera Drypetes.standleyi Elaeis.oleifera			
## 1	1	2	0	
## 2	1	1	0	
## 3	3	2	0	
## 4	0	0	0	
## 5	0	0	0	
## 6	0	0	0	
##	Enterolobium.schomburgkii Erythrina.costaricensis			
## 1	0	0		
## 2	0	0		
## 3	0	0		
## 4	0	0		
## 5	0	0		
## 6	0	3		
##	Erythroxylum.macrophyllum Eugenia.florida Eugenia.galalonensis			
## 1	0	0	0	
## 2	1	1	0	
## 3	0	0	0	
## 4	0	7	0	
## 5	0	2	0	
## 6	0	0	0	
##	Eugenia.nesiotica Eugenia.oerstediana Faramea occidentalis			
## 1	0	3	14	
## 2	0	2	36	
## 3	1	5	39	
## 4	0	1	39	
## 5	0	5	22	
## 6	0	2	16	

```

## Ficus.colubrinae Ficus.costaricana Ficus.insipida Ficus.maxima
## 1 0 0 0 1
## 2 1 0 0 0
## 3 0 0 0 0
## 4 0 0 0 0
## 5 0 0 0 0
## 6 0 0 0 0
## Ficus.obtusifolia Ficus.popenoei Ficus.tonduzii Ficus.trigonata
## 1 0 0 0 0
## 2 0 0 0 0
## 3 0 0 1 0
## 4 0 0 2 0
## 5 0 0 1 0
## 6 0 0 0 0
## Ficus.yoponensis Garcinia.intermedia Garcinia.madruno Genipa.americana
## 1 1 0 4 0
## 2 0 1 0 0
## 3 0 1 0 1
## 4 0 3 0 0
## 5 0 2 1 0
## 6 1 1 0 0
## Guapira.myrtiflora Guarea.fuzzy Guarea.grandifolia Guarea.guidonia
## 1 3 1 0 2
## 2 1 1 0 6
## 3 0 0 0 2
## 4 1 1 0 5
## 5 1 3 0 3
## 6 7 0 0 4
## Guatteria.dumetorum Guazuma.ulmifolia Guettarda.foliacea
## 1 6 0 1
## 2 16 0 5
## 3 6 0 1
## 4 3 1 2
## 5 9 0 1
## 6 7 0 0
## Gustavia.superba Hampea.appendiculata Hasseltia.floribunda
## 1 10 0 5
## 2 5 0 9
## 3 0 1 4
## 4 1 0 11
## 5 3 0 9
## 6 1 0 2
## Heisteria.acuminata Heisteria.concinna Hirtella.americana
## 1 0 4 0
## 2 0 5 0
## 3 0 4 0
## 4 0 6 0
## 5 1 4 0
## 6 1 8 0
## Hirtella.triandra Hura.crepitans Hieronyma.alchorneoides Inga.acuminata
## 1 21 0 0 0
## 2 14 0 2 0
## 3 5 0 0 0
## 4 4 0 0 0

```

## 5	6	0	0	0	
## 6	6	2	0	0	
##	Inga.cocleensis	Inga.goldmanii	Inga.laurina	Inga.semialata	Inga.nobilis
## 1	2	0	0	0	0
## 2	4	0	0	0	0
## 3	4	1	0	2	1
## 4	6	0	0	4	3
## 5	0	2	1	0	1
## 6	0	1	0	0	0
##	Inga.oerstediana	Inga.pezizifera	Inga.punctata	Inga.ruiziana	
## 1	0	0	3	0	
## 2	0	0	0	0	
## 3	0	0	0	0	
## 4	0	0	0	0	
## 5	0	0	0	0	
## 6	0	0	0	0	
##	Inga.sapindoides	Inga.spectabilis	Inga.umbellifera	Jacaranda.copaia	
## 1	2	0	0	6	
## 2	0	2	0	10	
## 3	3	0	0	9	
## 4	2	1	1	2	
## 5	5	0	0	3	
## 6	0	0	0	7	
##	Lacistema.aggregatum	Lacmellea.panamensis	Laetia.procera	Laetia.thamnia	
## 1	1	1	0	0	
## 2	0	0	1	1	
## 3	0	0	1	1	
## 4	1	2	0	0	
## 5	1	2	1	0	
## 6	2	1	0	0	
##	Lafoensia.punicifolia	Licania.hypoleuca	Licania.platypus		
## 1	0	0	0		
## 2	0	0	0		
## 3	0	0	0		
## 4	0	0	0		
## 5	0	1	0		
## 6	0	0	0		
##	Lindackeria.laurina	Lonchocarpus.heptaphyllus	Luehea.seemannii		
## 1	0	7	1		
## 2	0	7	0		
## 3	0	3	0		
## 4	0	9	0		
## 5	0	2	1		
## 6	0	1	1		
##	Macrocnemum.roseum	Maquira.guianensis.costaricana	Margaritaria.nobilis		
## 1	0	4	0		
## 2	0	3	0		
## 3	0	7	0		
## 4	0	7	0		
## 5	0	10	1		
## 6	0	4	0		
##	Marila.laxiflora	Maytenus.schippii	Miconia.affinis	Miconia.argentea	
## 1	1	2	0	2	
## 2	0	0	0	0	

```

## 3      0      0      0      1
## 4      0      1      1      0
## 5      0      0      0      1
## 6      0      1      0      0
##  Miconia.elata Miconia.hondurensis Mosannona.garwoodii Myrcia.gatunensis
## 1      0      0      1      1
## 2      0      0      0      0
## 3      0      0      0      0
## 4      0      0      0      0
## 5      0      0      1      0
## 6      0      0      1      0
##  Myrospermum.frutescens Nectandra.cissiflora Nectandra.lineata
## 1      0      0      0
## 2      0      1      0
## 3      0      2      0
## 4      0      2      0
## 5      0      2      0
## 6      2      0      0
##  Nectandra.purpurea Ochroma.pyramidale Ocotea.cernua Ocotea.oblonga
## 1      1      1      0      0
## 2      0      0      0      0
## 3      0      0      1      1
## 4      0      0      1      2
## 5      0      0      0      0
## 6      1      0      0      0
##  Ocotea.puberula Ocotea.whitei Oenocarpus.mapora Ormosia.amazonica
## 1      0      1      22      0
## 2      0      0      21      0
## 3      0      2      14      0
## 4      2      3      23      0
## 5      0      16      17      0
## 6      1      3      19      0
##  Ormosia.coccinea Ormosia.macrocalyx Pachira.quinata Pachira.sessilis
## 1      0      0      0      0
## 2      0      0      0      0
## 3      0      0      0      0
## 4      0      0      0      0
## 5      0      0      0      0
## 6      0      0      0      0
##  Perebea.xanthochyma Cinnamomum.triplinerve Picramnia.latifolia
## 1      0      0      0
## 2      1      0      0
## 3      0      1      1
## 4      0      0      0
## 5      1      1      0
## 6      0      0      0
##  Piper.reticulatum Platymiscium.pinnatum Platypodium.elegans
## 1      0      3      2
## 2      0      3      1
## 3      0      5      3
## 4      0      1      0
## 5      2      1      0
## 6      0      1      2
##  Posoqueria.latifolia Poulsenia.armata Pourouma.bicolor

```

## 1	0	24	5	
## 2	1	16	3	
## 3	0	28	0	
## 4	0	15	0	
## 5	0	25	1	
## 6	0	15	0	
##	Pouteria.fossicola	Pouteria.reticulata	Pouteria.stipitata	
## 1	0	5	0	
## 2	0	7	0	
## 3	0	3	1	
## 4	0	6	0	
## 5	0	5	0	
## 6	0	4	0	
##	Prioria.copaifera	Protium.costaricense	Protium.panamense	
## 1	13	5	2	
## 2	12	4	0	
## 3	12	1	2	
## 4	5	3	3	
## 5	3	7	2	
## 6	26	1	1	
##	Protium.tenuifolium	Pseudobombax.septenatum	Psidium.friedrichsthalianum	
## 1	11	0	0	
## 2	8	0	0	
## 3	3	0	0	
## 4	9	0	0	
## 5	3	0	0	
## 6	2	0	0	
##	Psychotria.grandis	Pterocarpus.rohrii	Quararibea.asterolepis	
## 1	0	1	11	
## 2	0	0	12	
## 3	0	0	15	
## 4	0	2	14	
## 5	0	1	9	
## 6	0	1	3	
##	Quassia.amara	Randia.armata	Sapium.broadleaf	Sapium.glandulosum
## 1	0	3	0	0
## 2	0	2	0	0
## 3	0	1	0	1
## 4	0	4	0	0
## 5	0	2	0	2
## 6	0	9	0	0
##	Schizolobium.parahyba	Senna.dariensis	Simarouba.amara	
## 1	0	0	14	
## 2	0	0	6	
## 3	0	0	16	
## 4	0	0	8	
## 5	0	0	7	
## 6	1	0	7	
##	Siparuna.guianensis	Siparuna.pauciflora	Sloanea.terniflora	
## 1	3	0	1	
## 2	2	0	0	
## 3	1	1	2	
## 4	2	0	2	
## 5	0	3	3	

	1	0	2	
## 6				
##	Socratea.exorrhiza	Solanum.hayesii	Sorocea.affinis	Spachea.membranacea
## 1	15	0	1	0
## 2	22	0	1	0
## 3	31	0	1	0
## 4	9	0	1	0
## 5	55	1	0	0
## 6	44	0	1	0
##	Spondias.mombin	Spondias.radlkoferi	Sterculia.apetala	
## 1	1	2	1	
## 2	1	0	2	
## 3	0	3	0	
## 4	1	3	0	
## 5	1	5	0	
## 6	0	0	1	
##	Swartzia.simplex.var.grandiflora	Swartzia.simplex.continentalis		
## 1		3		1
## 2		3		4
## 3		0		2
## 4		1		2
## 5		1		1
## 6		9		5
##	Symphonia.globulifera	Handroanthus.guayacan	Tabebuia.rosea	
## 1	0	1	1	
## 2	1	0	2	
## 3	1	1	1	
## 4	1	0	2	
## 5	2	0	3	
## 6	0	1	0	
##	Tabernaemontana.arborea	Tachigali.versicolor	Talisia.nervosa	
## 1	9	6	0	
## 2	5	1	0	
## 3	6	3	0	
## 4	10	3	0	
## 5	16	0	0	
## 6	11	1	0	
##	Talisia.princeps	Terminalia.amazonia	Terminalia.oblonga	
## 1	1	1	0	
## 2	0	0	0	
## 3	0	0	0	
## 4	0	0	0	
## 5	0	1	0	
## 6	0	1	0	
##	Tetragastris.panamensis	Tetrathylacium.johansenii	Theobroma.cacao	
## 1	5	0	1	
## 2	7	0	1	
## 3	10	0	0	
## 4	10	0	0	
## 5	7	0	1	
## 6	17	0	0	
##	Thevetia.ahouai	Tocoyena.pittieri	Trattinnickia.aspera	Trema.micrantha
## 1	0	0	3	0
## 2	0	1	1	0
## 3	0	0	1	0

```

## 4      0      0      0      2
## 5      0      0      2      1
## 6      0      0      0      0
## Trichanthera.gigantea Trichilia.pallida Trichilia.tuberculata
## 1      0      0      18
## 2      0      1      27
## 3      0      0      28
## 4      0      1      35
## 5      0      0      15
## 6      0      0      31
## Trichospermum.galeottii Triplaris.cumingiana Trophis.caucana
## 1      0      0      2
## 2      0      0      0
## 3      0      0      0
## 4      0      0      0
## 5      0      0      2
## 6      0      1      0
## Trophis.racemosa Turpinia occidentalis Unonopsis.pittieri
## 1      1      0      1
## 2      1      1      5
## 3      0      1      12
## 4      1      4      3
## 5      0      2      4
## 6      0      1      3
## Virola.multiflora Virola.sebifera Virola.surinamensis Vismia.baccifera
## 1      0      17      4      0
## 2      0      12      3      0
## 3      0      11      2      0
## 4      0      16      2      0
## 5      0      31      6      0
## 6      2      19      1      0
## Vochysia.ferruginea Xylopia.macrantha Zanthoxylum.ekmanii
## 1      0      1      3
## 2      0      0      4
## 3      0      0      8
## 4      0      0      13
## 5      0      0      3
## 6      0      0      1
## Zanthoxylum.juniperinum Zanthoxylum.panamense Zanthoxylum.setulosum
## 1      0      2      0
## 2      0      2      0
## 3      1      2      0
## 4      1      5      0
## 5      0      5      0
## 6      0      3      0
## Zuelania.guidonia
## 1      0
## 2      0
## 3      0
## 4      1
## 5      0
## 6      2

```

1. Examine if there is evidence of spatial dependence in a rare and a common species in the BCI tree dataset. *Aletris.blackiana* *Cordia.bicolor*. Pull a column or two and look etc.

```
colSums(BCI)
```

##	Abarema.macradenia	Vachellia.melanoceras
##	1	3
##	Acalypha.diversifolia	Acalypha.macrostachya
##	2	1
##	Adelia.triloba	Aegiphila.panamensis
##	92	23
##	Alchornea.costaricensis	Alchornea.latifolia
##	156	1
##	Alibertia.edulis	Allophylus.psilospermus
##	1	27
##	Alseis.blackiana	Amaioua.corymbosa
##	983	3
##	Anacardium.excelsum	Andira.inermis
##	22	28
##	Annona.spraguei	Apeiba.glabra
##	27	236
##	Apeiba.tibourbou	Aspidosperma.desmanthum
##	21	52
##	Astrocaryum.standleyanum	Astronium.graveolens
##	201	39
##	Attalea.butyracea	Banara.guianensis
##	33	1
##	Beilschmiedia.pendula	Brosimum.alicastrum
##	294	188
##	Brosimum.guianense	Calophyllum.longifolium
##	1	55
##	Casearia.aculeata	Casearia.arborea
##	23	100
##	Casearia.commersoniana	Casearia.guianensis
##	3	2
##	Casearia.sylvestris	Cassipourea.guianensis
##	54	87
##	Cavanillesia.platanifolia	Cecropia.insignis
##	19	264
##	Cecropia.obtusifolia	Cedrela.odorata
##	25	2
##	Ceiba.pentandra	Celtis.schippii
##	39	38
##	Cespedesia.spathulata	Chamguava.schippii
##	2	3
##	Chimarrhis.parviflora	Maclura.tinctoria
##	1	1
##	Chrysochlamys.eclipses	Chrysophyllum.argenteum
##	2	85
##	Chrysophyllum.cainito	Coccoloba.coronata
##	25	22
##	Coccoloba.manzinellensis	Colubrina.glandulosa
##	13	1
##	Cordia.alliodora	Cordia.bicolor
##	63	325
##	Cordia.lasiocalyx	Coussarea.curvigemma
##	364	55
##	Croton.billbergianus	Cupania.cinerea

##	98	1
##	Cupania.latifolia	Cupania.rufescens
##	12	4
##	Cupania.seemannii	Dendropanax.arboreus
##	47	88
##	Desmopsis.panamensis	Diospyros.artanthifolia
##	13	16
##	Dipteryx.oleifera	Drypetes.standleyi
##	33	285
##	Elaeis.oleifera	Enterolobium.schomburgkii
##	21	2
##	Erythrina.costaricensis	Erythroxylum.macrophyllum
##	26	18
##	Eugenia.florida	Eugenia.galalonensis
##	81	12
##	Eugenia.nesiotica	Eugenia.oerstediana
##	55	177
##	Faramea.occidentalis	Ficus.colubrinae
##	1717	1
##	Ficus.costaricana	Ficus.insipida
##	7	3
##	Ficus.maxima	Ficus.obtusifolia
##	4	7
##	Ficus.popenoei	Ficus.tonduzii
##	3	23
##	Ficus.trigonata	Ficus.yoponensis
##	5	6
##	Garcinia.intermedia	Garcinia.madrano
##	92	12
##	Genipa.americana	Guapira.myrtiflora
##	23	99
##	Guarea.fuzzy	Guarea.grandifolia
##	68	10
##	Guarea.guidonia	Guatteria.dumetorum
##	376	244
##	Guazuma.ulmifolia	Guettarda.foliacea
##	38	85
##	Gustavia.superba	Hampea.appendiculata
##	644	13
##	Hasseltia.floribunda	Heisteria.acuminata
##	229	7
##	Heisteria.concinna	Hirtella.americana
##	288	5
##	Hirtella.triandra	Hura.crepitans
##	681	101
##	Hieronyma.alchorneoides	Inga.acuminata
##	41	26
##	Inga.cocleensis	Inga.goldmanii
##	52	49
##	Inga.laurina	Inga.semialata
##	10	98
##	Inga.nobilis	Inga.oerstediana
##	67	2
##	Inga.pezizifera	Inga.punctata

##	20	10
##	Inga.ruiziana	Inga.sapindoides
##	5	76
##	Inga.spectabilis	Inga.umbellifera
##	14	14
##	Jacaranda.copaia	Lacistema.aggregatum
##	236	33
##	Lacmellea.panamensis	Laetia.procera
##	51	12
##	Laetia.thamnia	Lafoensia.punicifolia
##	27	5
##	Licania.hypoleuca	Licania.platypus
##	14	10
##	Lindackeria.laurina	Lonchocarpus.heptaphyllus
##	64	121
##	Luehea.seemannii	Macrocnemum.roseum
##	93	25
##	Maquira.guianensis.costaricana	Margaritaria.nobilis
##	167	2
##	Marila.laxiflora	Maytenus.schippii
##	10	21
##	Miconia.affinis	Miconia.argentea
##	8	70
##	Miconia.elata	Miconia.hondurensis
##	1	7
##	Mosannonna.garwoodii	Myrcia.gatunensis
##	15	5
##	Myrospermum.frutescens	Nectandra.cissiflora
##	7	33
##	Nectandra.lineata	Nectandra.purpurea
##	10	4
##	Ochroma.pyramidale	Ocotea.cernua
##	5	29
##	Ocotea.oblonga	Ocotea.puberula
##	36	22
##	Ocotea.whitei	Oenocarpus.mapora
##	184	788
##	Ormosia.amazonica	Ormosia.coccinea
##	1	5
##	Ormosia.macrocalyx	Pachira.quinata
##	3	1
##	Pachira.sessilis	Perebea.xanthochyma
##	9	21
##	Cinnamomum.triplinerve	Picramnia.latifolia
##	16	45
##	Piper.reticulatum	Platymiscium.pinnatum
##	9	61
##	Platypodium.elegans	Posoqueria.latifolia
##	43	15
##	Poulsenia.armata	Pourouma.bicolor
##	755	13
##	Pouteria.fossicola	Pouteria.reticulata
##	2	203
##	Pouteria.stipitata	Prioria.copaifera

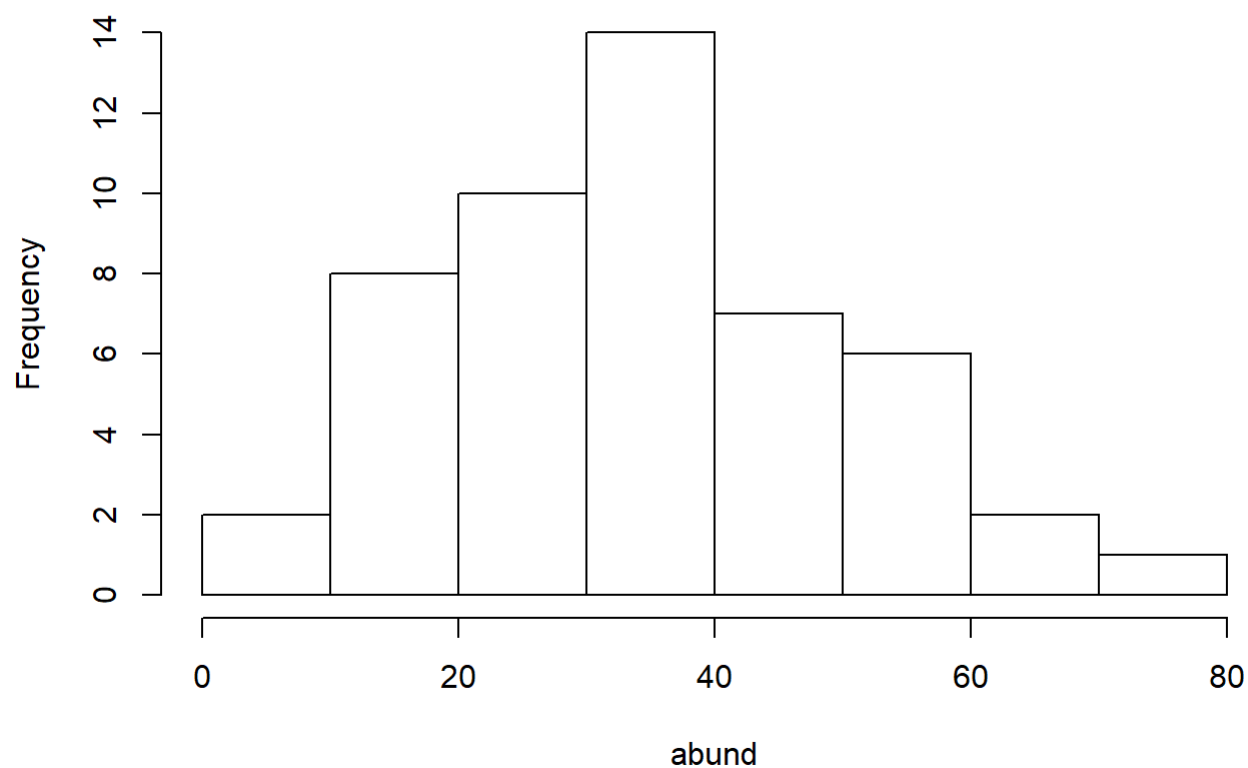
##	31	345
##	Protium.costaricense	Protium.panamense
##	111	50
##	Protium.tenuifolium	Pseudobombax.septenatum
##	381	8
##	Psidium.friedrichsthalianum	Psychotria.grandis
##	4	2
##	Pterocarpus.rohrii	Quararibea.asterolepis
##	80	724
##	Quassia.amara	Randia.armata
##	4	248
##	Sapium.broadleaf	Sapium.glandulosum
##	3	17
##	Schizolobium.parahyba	Senna.dariensis
##	2	1
##	Simarouba.amara	Siparuna.guianensis
##	289	13
##	Siparuna.pauciflora	Sloanea.terniflora
##	16	78
##	Socratea.exorrhiza	Solanum.hayesii
##	346	12
##	Sorocea.affinis	Spachea.membranacea
##	28	8
##	Spondias.mombin	Spondias.radlkoferi
##	29	63
##	Sterculia.apetala	Swartzia.simplex.var.grandiflora
##	26	218
##	Swartzia.simplex.continentalis	Symphonia.globulifera
##	118	26
##	Handroanthus.guayacan	Tabebuia.rosea
##	30	68
##	Tabernaemontana.arborea	Tachigali.versicolor
##	322	98
##	Talisia.nervosa	Talisia.princeps
##	1	3
##	Terminalia.amazonia	Terminalia.oblonga
##	28	43
##	Tetragastris.panamensis	Tetrathylacium.johansenii
##	379	7
##	Theobroma.cacao	Thevetia.ahouai
##	12	2
##	Tocoyena.pittieri	Trattinnickia.aspera
##	5	40
##	Trema.micrantha	Trichanthera.gigantea
##	15	2
##	Trichilia.pallida	Trichilia.tuberculata
##	82	1681
##	Trichospermum.galeottii	Triplaris.cumingiana
##	1	147
##	Trophis.caucana	Trophis.racemosa
##	33	32
##	Turpinia occidentalis	Unonopsis.pittieri
##	58	163
##	Virola.multiflora	Virola.sebifera

##	25	617
##	<i>Virola.surinamensis</i>	<i>Vismia.baccifera</i>
##	164	1
##	<i>Vochysia.ferruginea</i>	<i>Xylopia.macrantha</i>
##	12	143
##	<i>Zanthoxylum.ekmanii</i>	<i>Zanthoxylum.juniperinum</i>
##	149	45
##	<i>Zanthoxylum.panamense</i>	<i>Zanthoxylum.setulosum</i>
##	67	1
##	<i>Zuelania.guidonia</i>	
##	10	

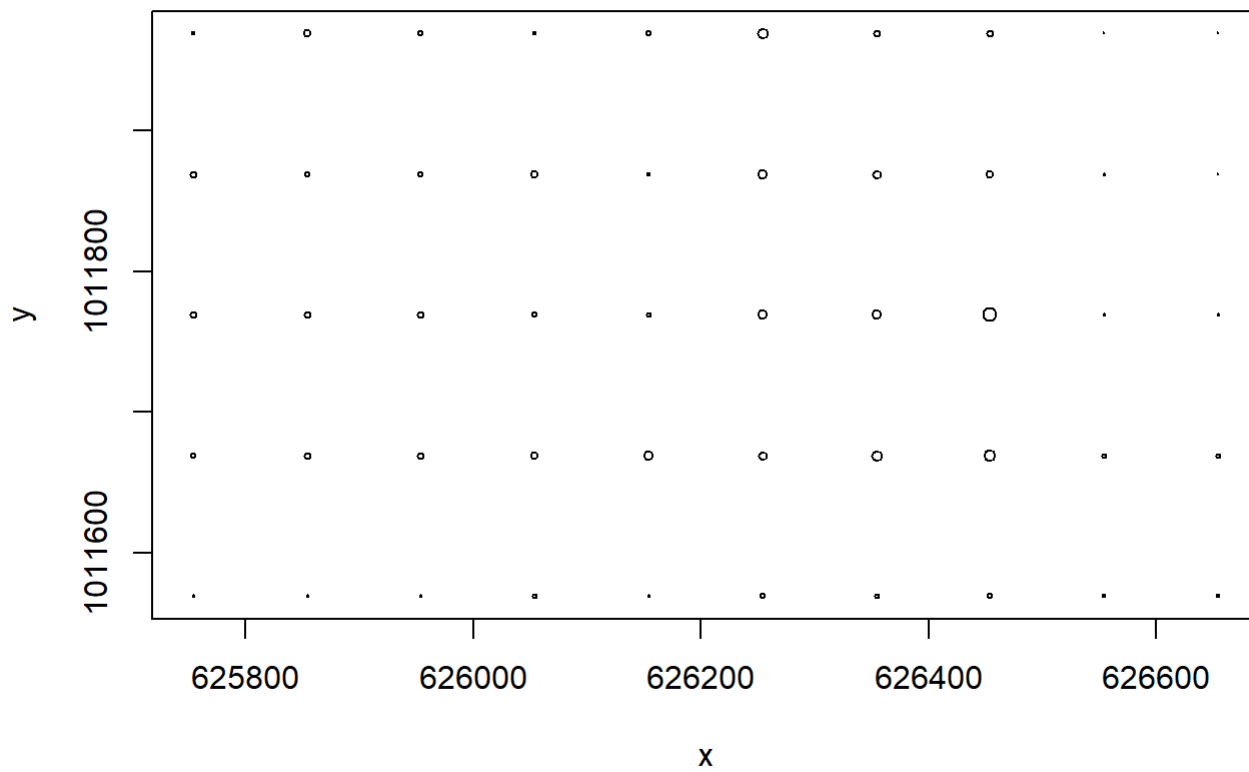
```
many <- BCI$Faramea occidentalis
few <- BCI$Zanthoxylum.panamense
many <- data.frame(many)
few <- data.frame(few)
```

```
abund <- apply(many, 1, sum)
hist(abund)
```

Histogram of abund



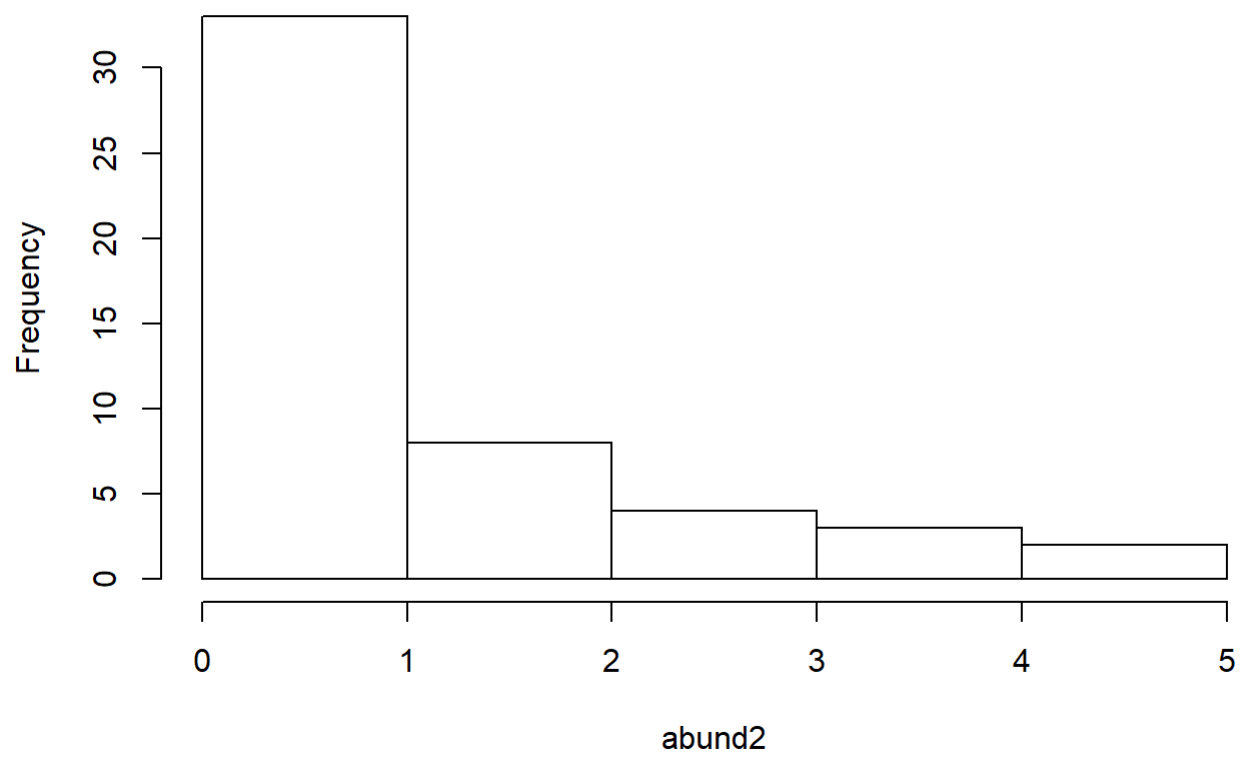
```
plot(BCI_xy, cex = abund/max(abund))
```

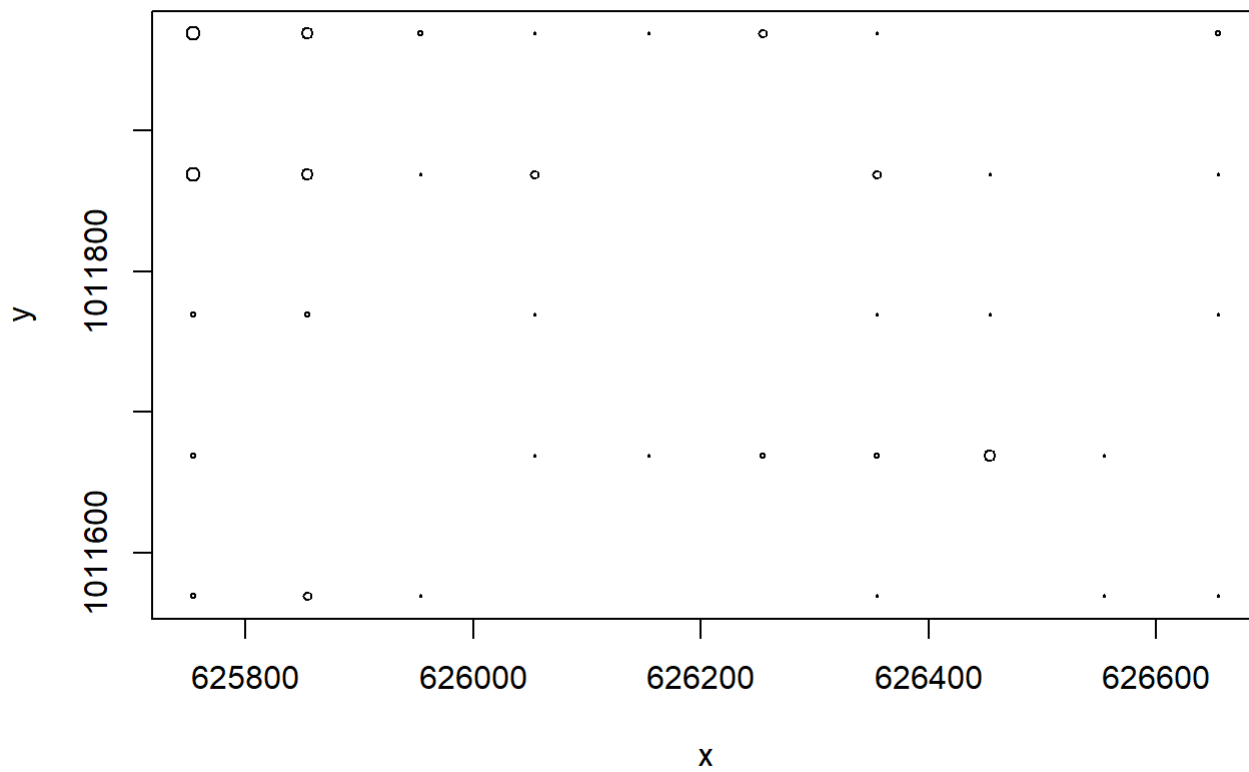
```
abund2 <- apply(few, 1, sum)

hist(abund2)
```

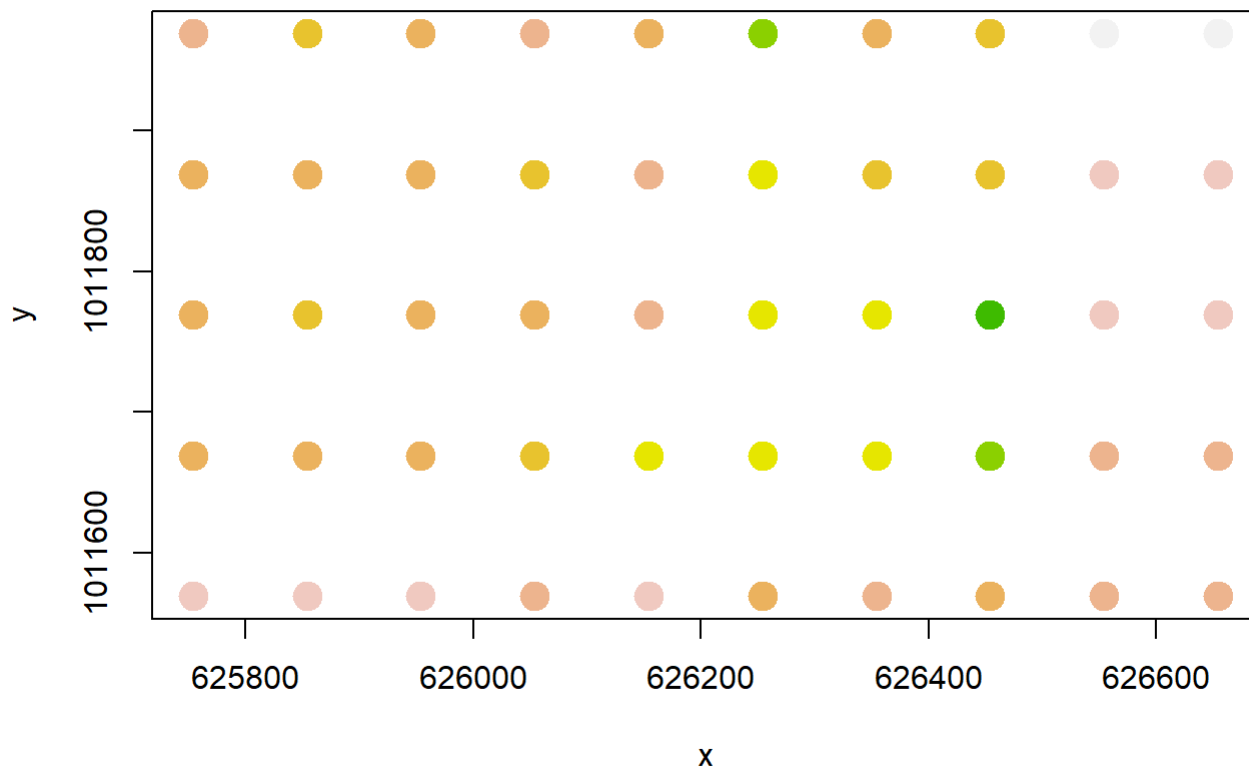
Histogram of abund2



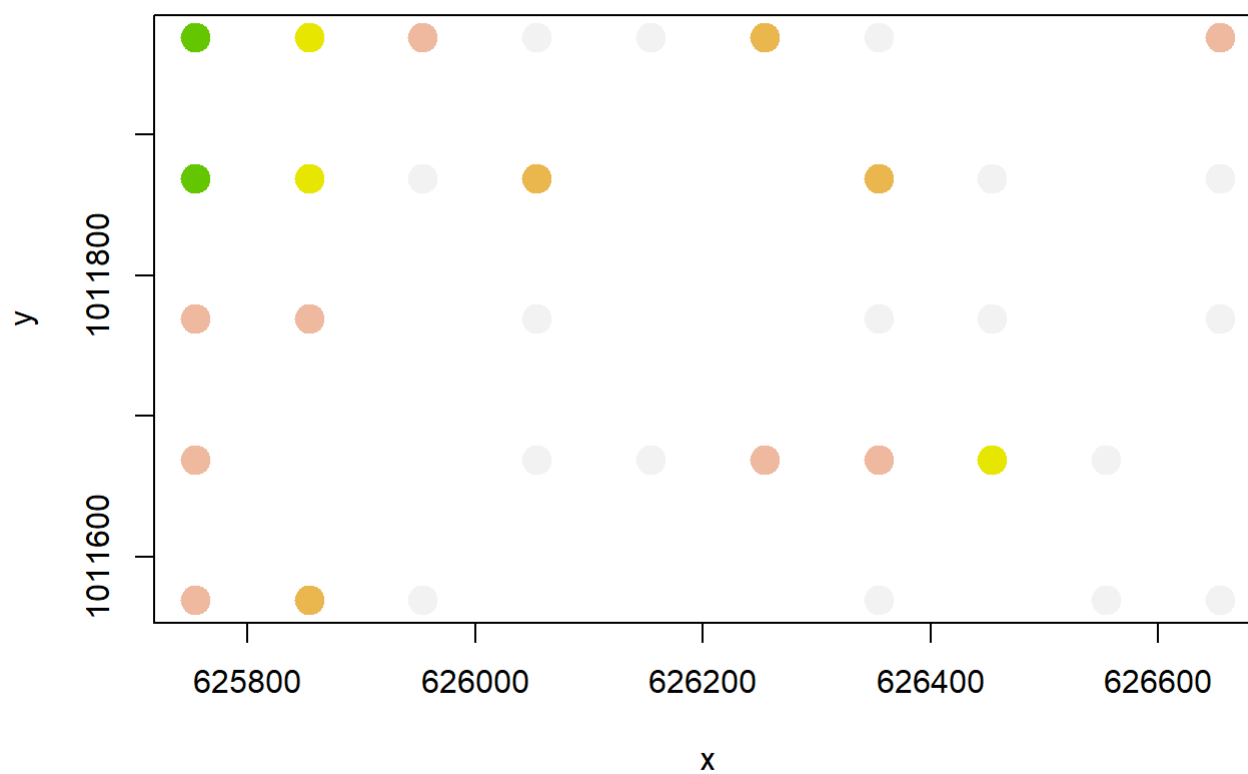
```
plot(BCI_xy, cex = abund2/max(abund2))
```



```
col_brks = hist(abund, plot=F)$breaks
col_indices = as.numeric(cut(abund, col_brks))
cols = rev(terrain.colors(length(col_brks)))
plot(BCI_xy, cex=2, pch=19, col=cols[col_indices])
```



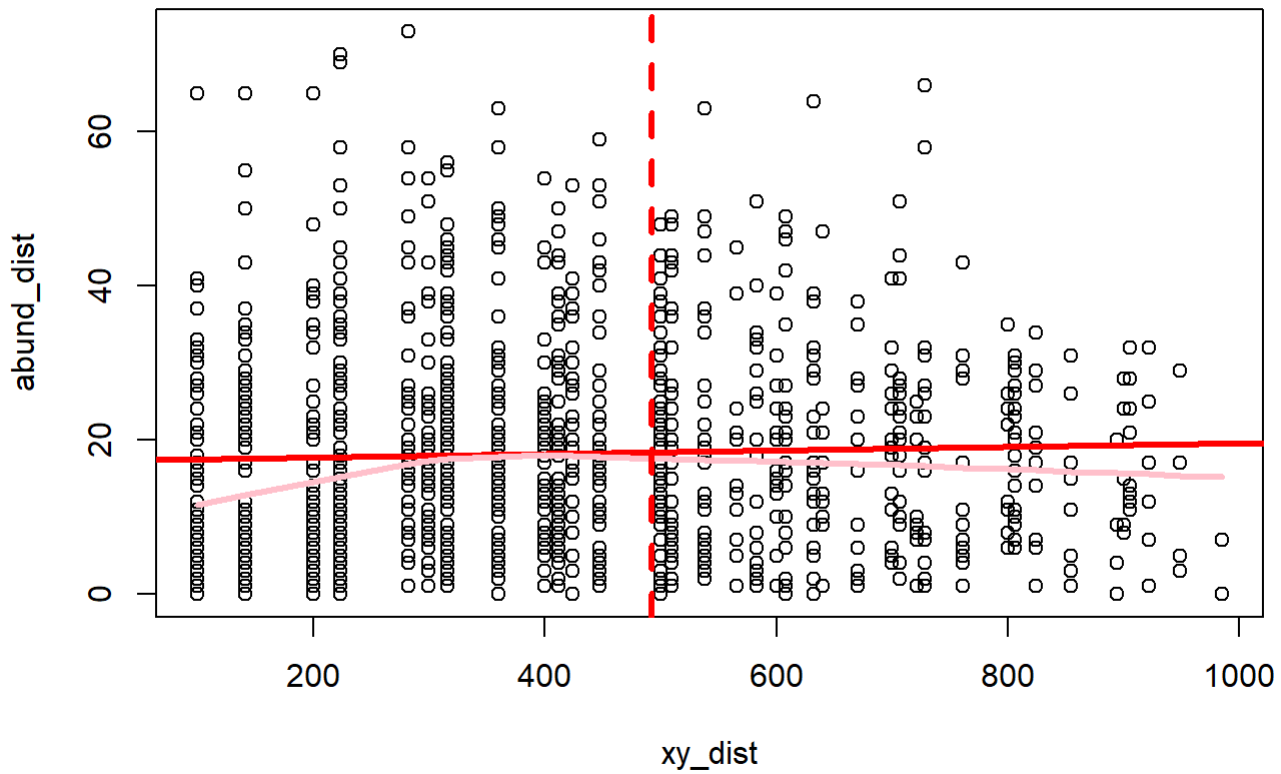
```
col_brks = hist(abund2, plot=F)$breaks
col_indices = as.numeric(cut(abund2, col_brks))
cols = rev(terrain.colors(length(col_brks)))
plot(BCI_xy, cex=2, pch=19, col=cols[col_indices])
```



```
abund_dist <- dist(abund)
xy_dist <- dist(BCI_xy)
abund2_dist <- dist(abund2)
```

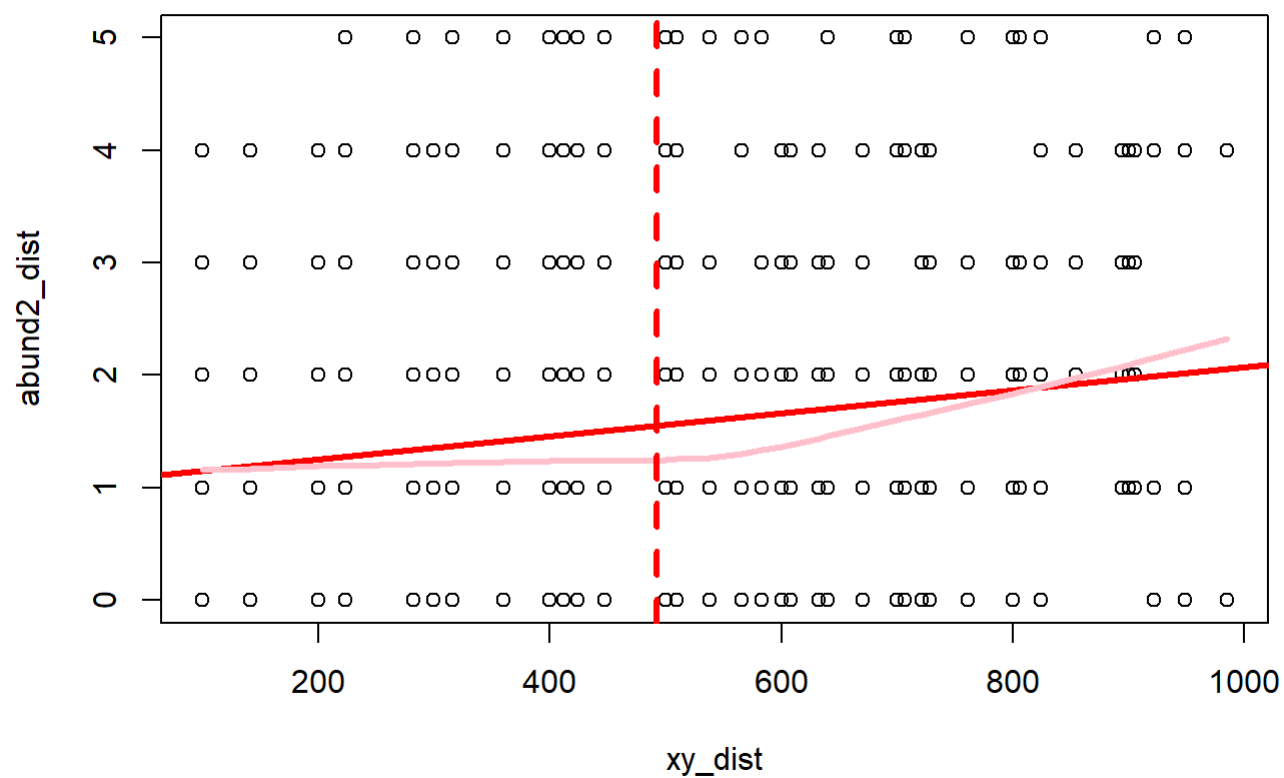
```
max_dist = max(xy_dist) / 2

# plot result
plot(xy_dist, abund_dist)
abline(lm(abund_dist ~ xy_dist), lwd=3, col='red')
lines(lowess(xy_dist, abund_dist), lwd=3, col='pink')
abline(v = max_dist, col='red', lwd=3, lty=2)
```



```
ax_dist = max(xy_dist) / 2

# plot result
plot(xy_dist, abund2_dist)
abline(lm(abund2_dist ~ xy_dist), lwd=3, col='red')
lines(lowess(xy_dist, abund2_dist), lwd=3, col='pink')
abline(v = max_dist, col='red', lwd=3, lty=2)
```



```
obs_cor <- cor(xy_dist, abund_dist)
obs_cor
```

```
## [1] 0.0355889
```

```
obs_cor2 <- cor(xy_dist, abund2_dist)
obs_cor2
```

```
## [1] 0.1656752
```

#measure of correlation between your distance and species matrix. Low is bad.

```
sr_mantel <- mantel(abund_dist, xy_dist)
sr_mantel
```

```
##
## Mantel statistic based on Pearson's product-moment correlation
##
## Call:
## mantel(xdis = abund_dist, ydis = xy_dist)
##
## Mantel statistic r: 0.03559
##      Significance: 0.237
##
## Upper quantiles of permutations (null model):
##      90%      95%     97.5%     99%
## 0.0645 0.0859 0.1050 0.1259
## Permutation: free
## Number of permutations: 999
```

This mantel statistic for the common species suggests there is a weak correlation between the species abundance matrix and the distance matrix. The p-value suggests this is not a statistically significant correlation

```
sr_mantel2 = mantel(abund2_dist, xy_dist)
sr_mantel2
```

```
##
## Mantel statistic based on Pearson's product-moment correlation
##
## Call:
## mantel(xdis = abund2_dist, ydis = xy_dist)
##
## Mantel statistic r: 0.1657
##      Significance: 0.005
##
## Upper quantiles of permutations (null model):
##      90%      95%     97.5%     99%
## 0.0641 0.0830 0.1018 0.1161
## Permutation: free
## Number of permutations: 999
```

The mantel statistic for the rare species suggests a low correlation between the species abundance matrix and the geographical distance matrix. However, the p-value of 0.003 suggests this is a statistically significant correlation.

```
corlog1 <- mantel.correlog(abund_dist, xy_dist)

corlog1
```



```
##
## Mantel Correlogram Analysis
##
## Call:
##
## mantel.correlog(D.eco = abund_dist, D.geo = xy_dist)
##
##      class.index      n.dist Mantel.cor Pr(Mantel) Pr(corrected)
## D.cl.1  136.870241 144.000000  0.041368    0.046    0.046 *
## D.cl.2  210.610723 376.000000  0.009563    0.349    0.349
## D.cl.3  284.351204 390.000000 -0.054005    0.024    0.072 .
## D.cl.4  358.091686 148.000000 -0.056782    0.010    0.040 *
## D.cl.5  431.832168 372.000000 -0.039064    0.115    0.230
## D.cl.6  505.572649 266.000000 -0.023481    0.143    0.345
## D.cl.7  579.313131 168.000000      NA      NA      NA
## D.cl.8  653.053613 100.000000      NA      NA      NA
## D.cl.9  726.794094 154.000000      NA      NA      NA
## D.cl.10 800.534576  88.000000      NA      NA      NA
## D.cl.11 874.275058  50.000000      NA      NA      NA
## D.cl.12 948.015539  24.000000      NA      NA      NA
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

This indicates that for the common species, there are two distance classes at which there is a statistically significant correlation between the matrixes.

```
corlog2 <- mantel.correlog(abund2_dist, xy_dist)
corlog2
```

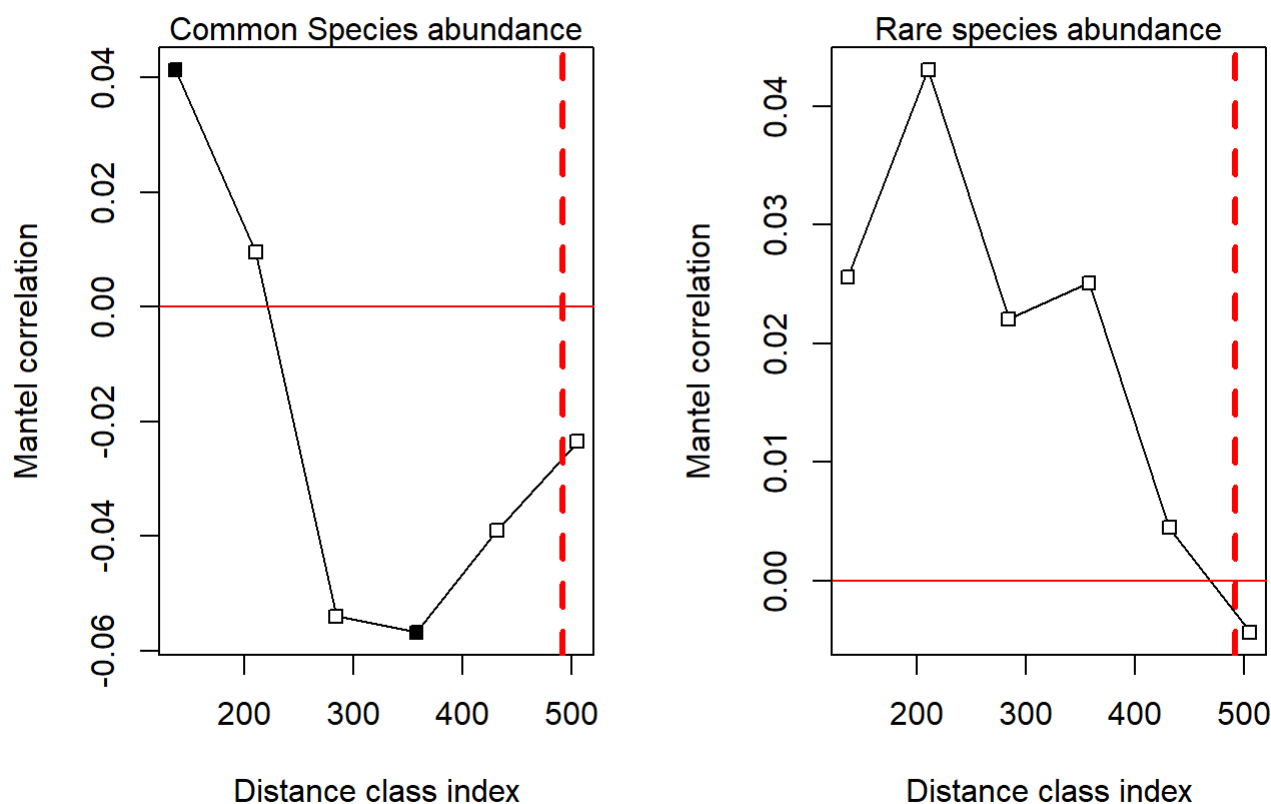
```
##
## Mantel Correlogram Analysis
##
## Call:
##
## mantel.correlog(D.eco = abund2_dist, D.geo = xy_dist)
##
##      class.index      n.dist Mantel.cor Pr(Mantel) Pr(corrected)
## D.cl.1  136.8702408 144.0000000  0.0256137    0.160    0.160
## D.cl.2  210.6107225 376.0000000  0.0430563    0.058    0.116
## D.cl.3  284.3512042 390.0000000  0.0220694    0.202    0.320
## D.cl.4  358.0916859 148.0000000  0.0251025    0.149    0.447
## D.cl.5  431.8321676 372.0000000  0.0044998    0.458    0.596
## D.cl.6  505.5726492 266.0000000 -0.0043089    0.465    0.916
## D.cl.7  579.3131309 168.0000000      NA      NA      NA
## D.cl.8  653.0536126 100.0000000      NA      NA      NA
## D.cl.9  726.7940943 154.0000000      NA      NA      NA
## D.cl.10 800.5345760  88.0000000      NA      NA      NA
## D.cl.11 874.2750577  50.0000000      NA      NA      NA
## D.cl.12 948.0155393  24.0000000      NA      NA      NA
```

For the rare species, it appears that no statistically significant correlation is found for any of the distances classes.

```

par(mfrow=c(1,2))
plot(corlog1)
mtext(side=3, 'Common Species abundance')
abline(v=max_dist, col='red', lwd=3, lty=2)
plot(corlog2)
mtext(side=3, 'Rare species abundance')
abline(v= max_dist, col='red', lwd=3, lty=2)

```



It appears there may be evidence of spatial dependence in the common species. According to the mantel correlation values a statistically significant relationship was found at a distance class of about 100, indicating a positive correlation between abundance and smaller distances. Also, there is a statistically significant relationship seen at a distance class of about 280, indicating a negative correlation between species abundance and larger distances.

However, for the rare species, no statistically significant correlations were found at any of the distance classes.

2. Build two generalized linear models to predict the abundance of the species *Drypetes standleyi* using the abundance of other tree species in the study site. Specifically examine the following species as predictor variables:

```
library(nlme)
```

```
all.dat <- data.frame(BCI, BCI_xy)
```

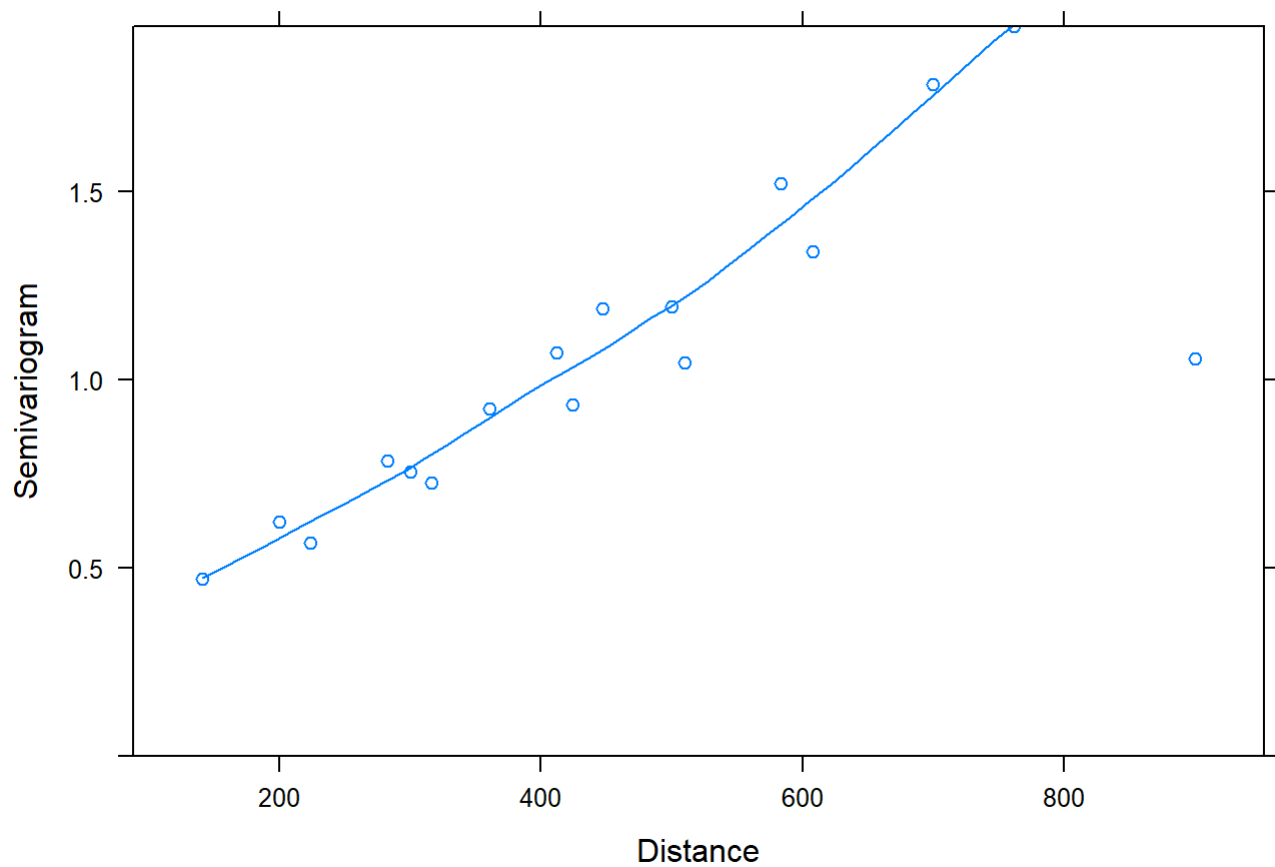
Model 1: only include a single species as a predictor variable

```
glm1 <- gls(Drypetes.standleyi~Quassia.amara, data=all.dat)
```

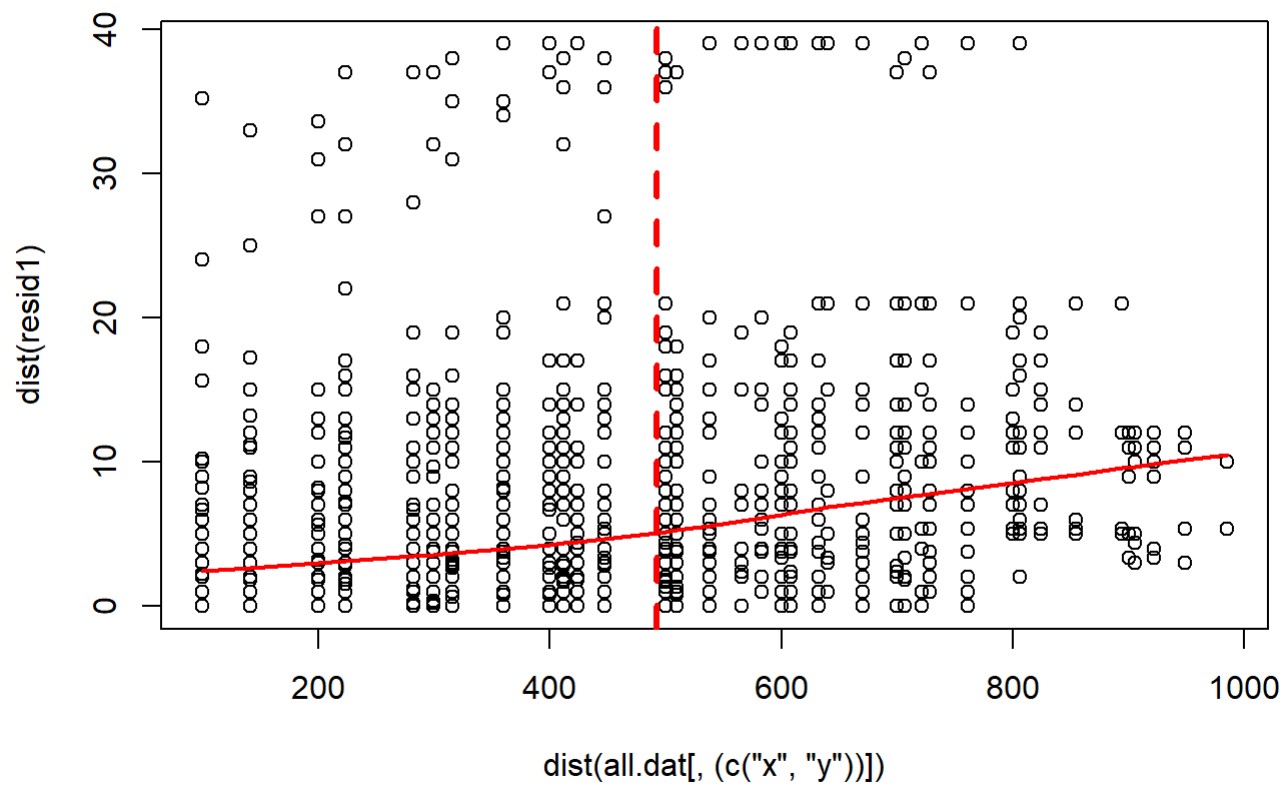
```
summary(glm1)
```

```
## Generalized least squares fit by REML
##   Model: Drypetes.standleyi ~ Quassia.amara
##   Data: all.dat
##           AIC      BIC    logLik
##   337.5287 343.1423 -165.7644
##
## Coefficients:
##              Value Std.Error  t-value p-value
## (Intercept)  4.962810  1.030804  4.814503   0e+00
## Quassia.amara 9.214876  2.304948  3.997867   2e-04
##
## Correlation:
##              (Intr)
## Quassia.amara -0.179
##
## Standardized residuals:
##           Min      Q1      Med      Q3      Max
## -0.6920362 -0.6920362 -0.4131473  0.2840748  4.7462963
##
## Residual standard error: 7.171316
## Degrees of freedom: 50 total; 48 residual
```

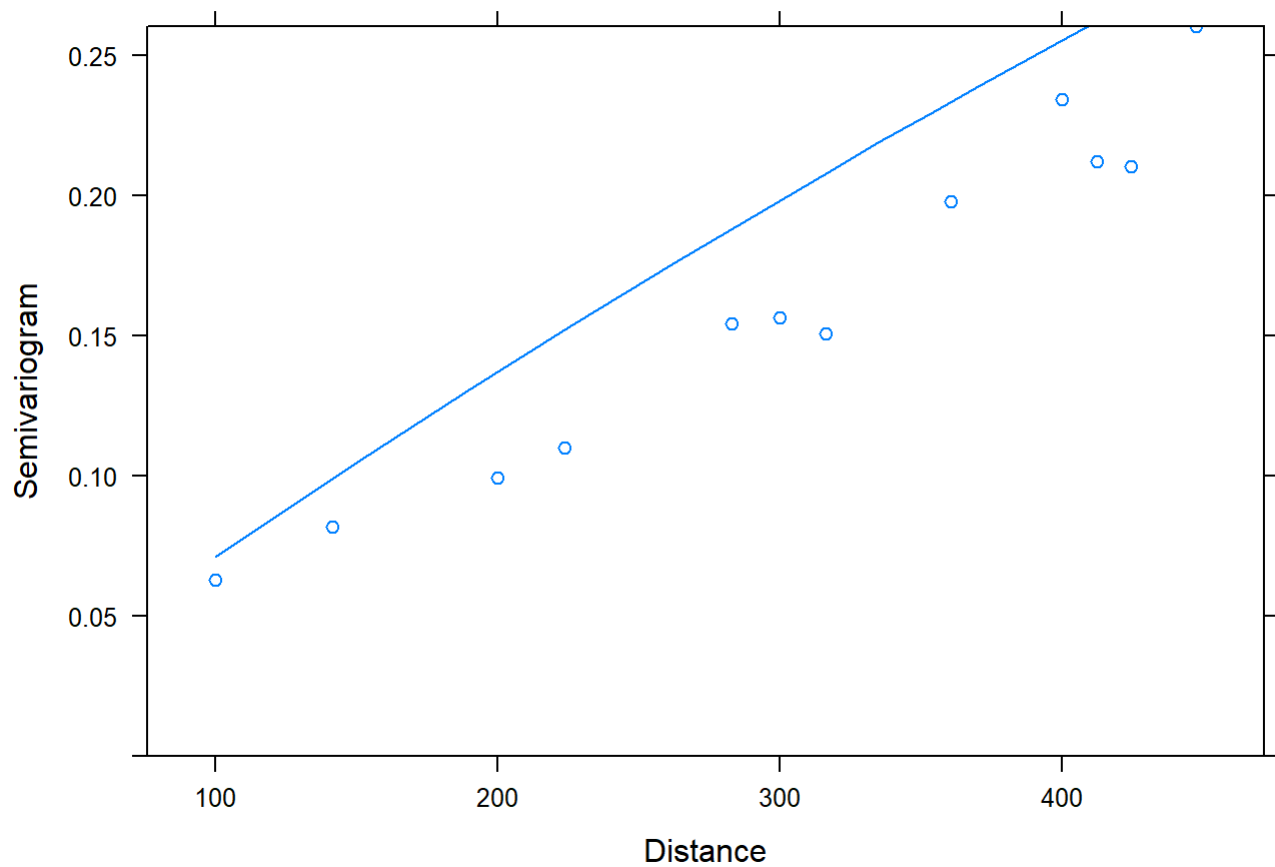
```
plot(Variogram(glm1, form= ~ x+y))
```



```
resid1 <- residuals(glm1)
plot(dist(all.dat[, (c('x', 'y'))]), dist(resid1))
lines(lowess(dist(all.dat[, c('x', 'y')]), dist(resid1)), col='red', lwd=2)
abline(v = max_dist, col='red', lwd=3, lty=2)
```

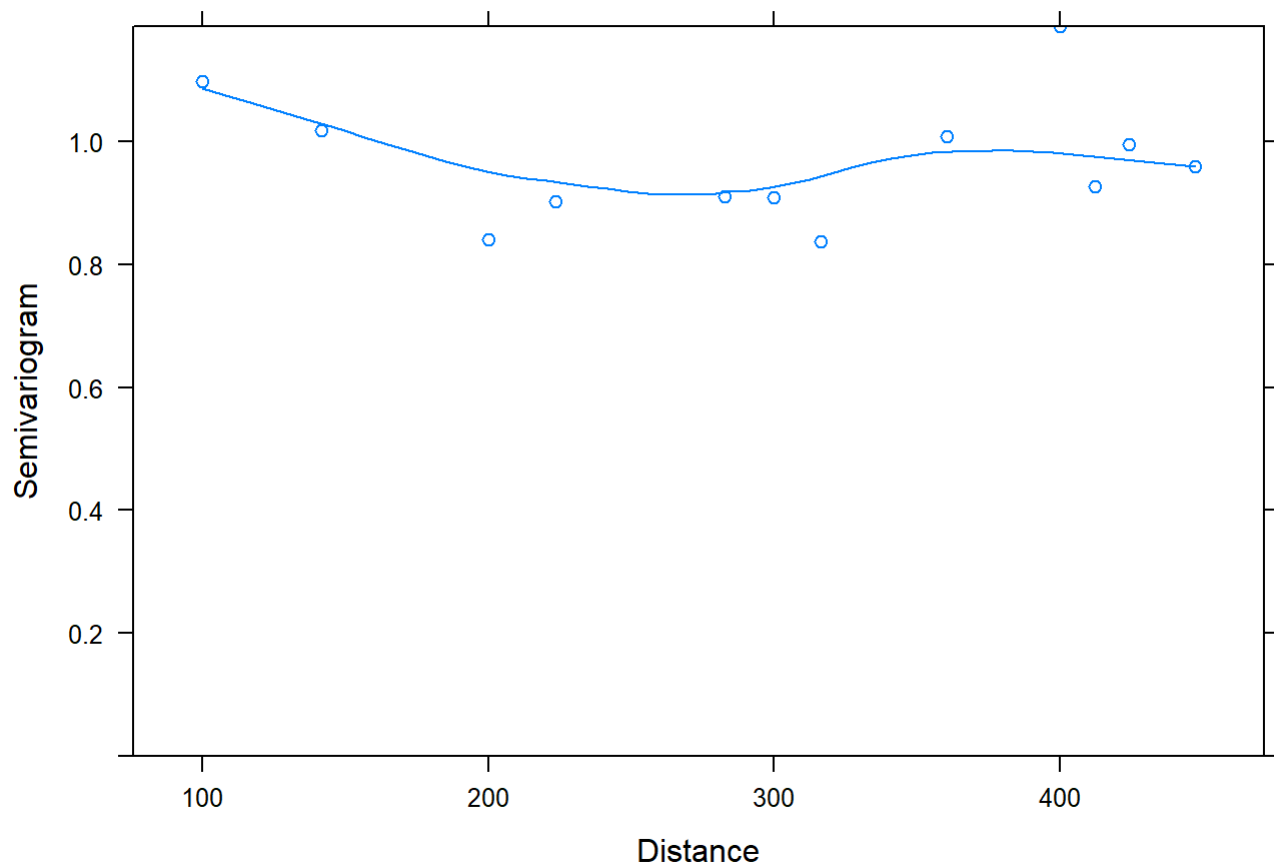


```
glm1.exp <- update(glm1, corr=corExp(form=~x + y))
plot(Variogram(glm1.exp, maxDist = max_dist))
```



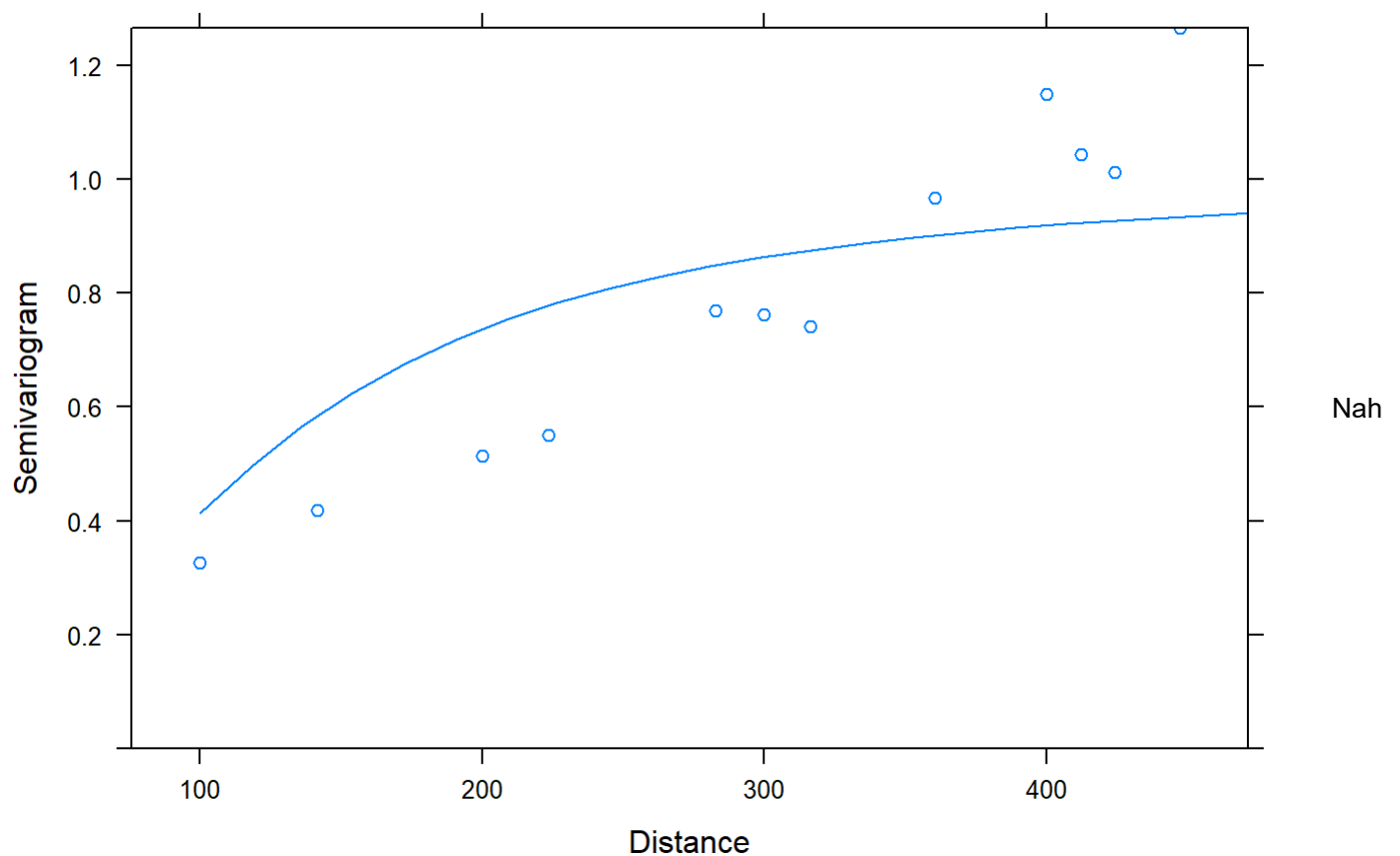
Not terrible. . .

```
plot(Variogram(glm1.exp, resType='normalized', maxDist = max_dist))
```



Some trend here

```
glm1.rat <- update(glm1, corr=corRatio(form=~x + y))  
plot(Variogram(glm1.rat, maxDist = max_dist))
```



```
anova(glm1, glm1.exp)
```

##	Model	df	AIC	BIC	logLik	Test	L.Ratio	p-value
##	glm1	1 3	337.5287	343.1423	-165.7644			
##	glm1.exp	2 4	301.4029	308.8877	-146.7014	1 vs 2	38.12584	<.0001

So both models with the spacial error term fits better than the one without. However, the exponential model fits the best. It appears there is a significant difference between the models

```
summary(glm1)
```



```
## Generalized least squares fit by REML
## Model: Drypetes.standleyi ~ Quassia.amara
## Data: all.dat
##      AIC      BIC    logLik
## 337.5287 343.1423 -165.7644
##
## Coefficients:
##              Value Std.Error  t-value p-value
## (Intercept)  4.962810  1.030804  4.814503  0e+00
## Quassia.amara 9.214876  2.304948  3.997867  2e-04
##
## Correlation:
##              (Intr)
## Quassia.amara -0.179
##
## Standardized residuals:
##      Min      Q1      Med      Q3      Max
## -0.6920362 -0.6920362 -0.4131473  0.2840748  4.7462963
##
## Residual standard error: 7.171316
## Degrees of freedom: 50 total; 48 residual
```

```
summary(glm1.exp)
```

```
## Generalized least squares fit by REML
## Model: Drypetes.standleyi ~ Quassia.amara
## Data: all.dat
##      AIC      BIC    logLik
## 301.4029 308.8877 -146.7015
##
## Correlation Structure: Exponential spatial correlation
## Formula: ~x + y
## Parameter estimate(s):
## range
## 1356.62
##
## Coefficients:
##              Value Std.Error  t-value p-value
## (Intercept)  9.794547 13.486551  0.7262455  0.4712
## Quassia.amara 3.763073  1.495159  2.5168384  0.0152
##
## Correlation:
##              (Intr)
## Quassia.amara -0.069
##
## Standardized residuals:
##      Min      Q1      Med      Q3      Max
## -0.5991027 -0.5991027 -0.4767688 -0.1250587  1.7864089
##
## Residual standard error: 16.34869
## Degrees of freedom: 50 total; 48 residual
```

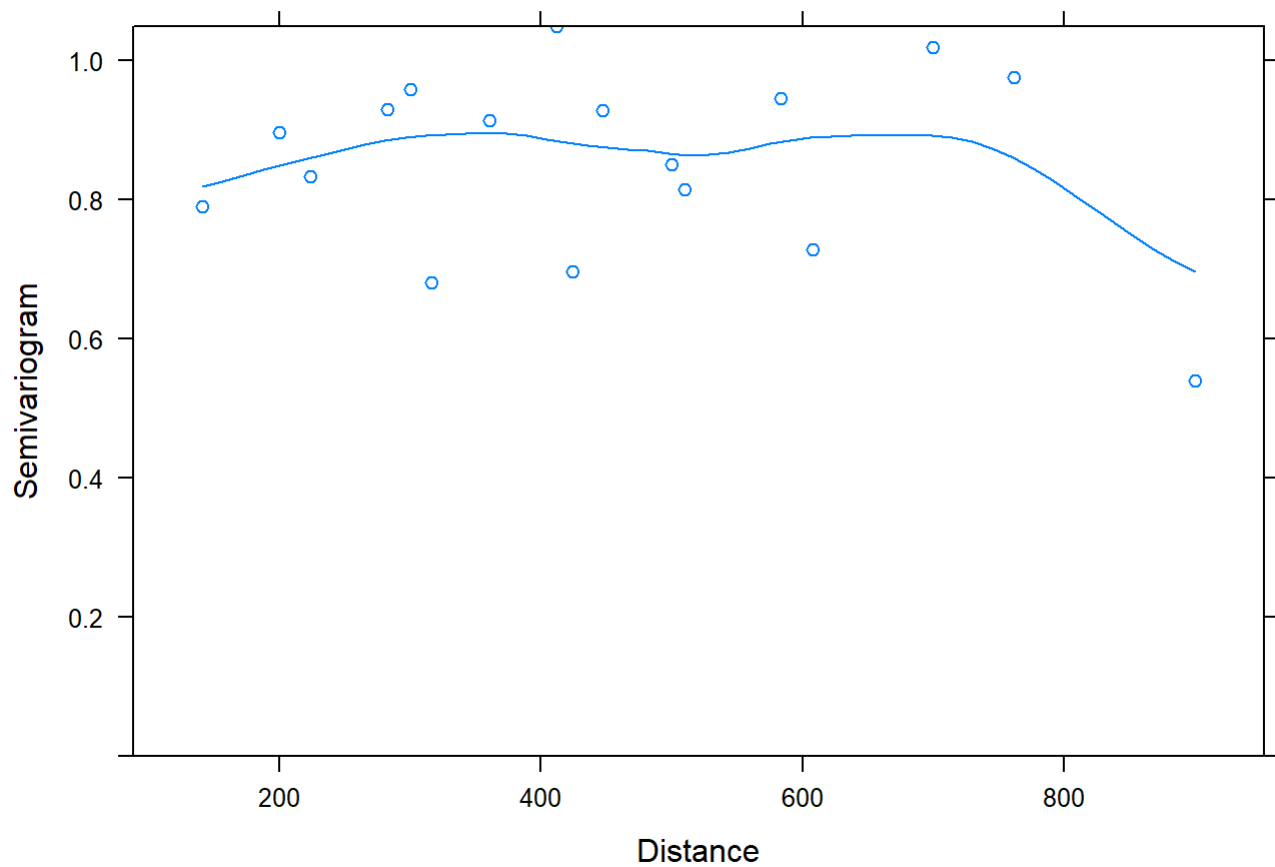
Model 2: include all of the species as predictor variables

```
glm2 <- gls(Drypetes.standleyi~Cordia.lasiocalyx + Hirtella.triandra + Picramnia.latifolia + Quassia.amara + Tabernaemontana.arborea + Trattinnickia.aspera + Xylopia.macrantha, data=all.dat)
```

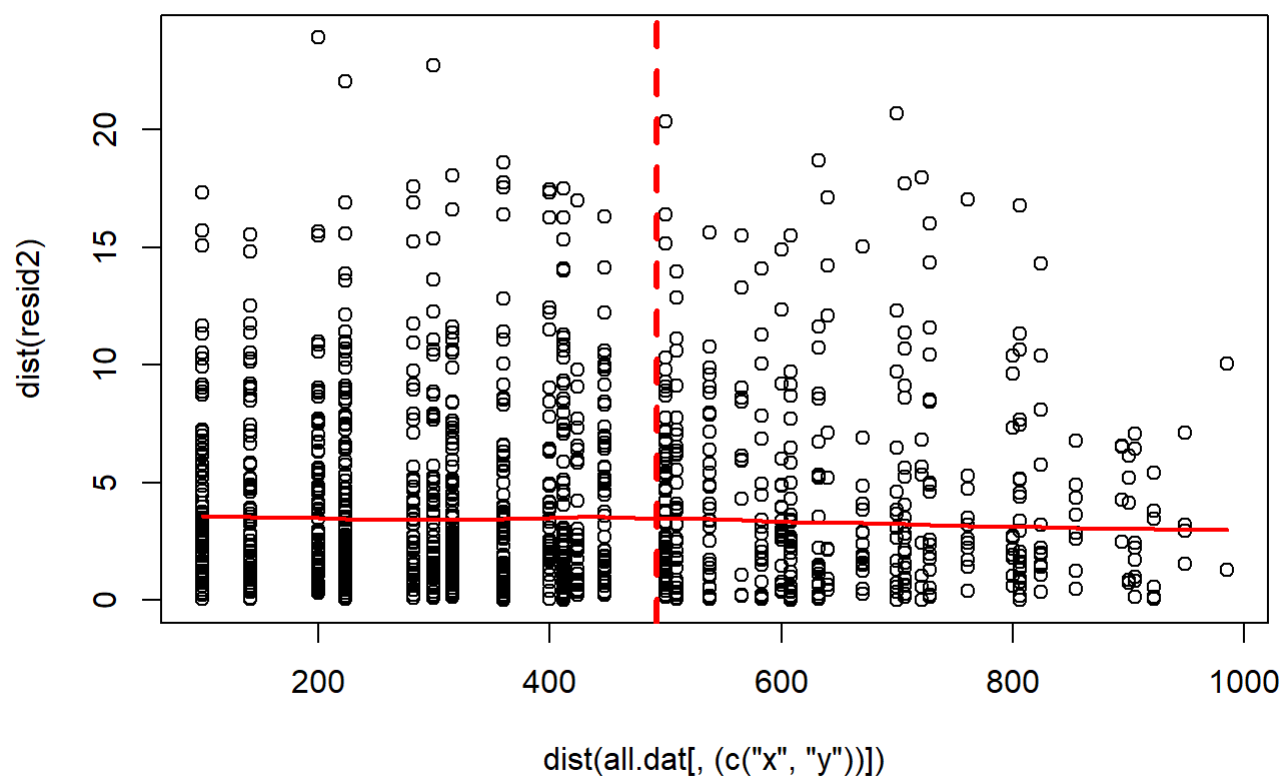
```
summary(glm2)
```

```
## Generalized least squares fit by REML
## Model: Drypetes.standleyi ~ Cordia.lasiocalyx + Hirtella.triandra + Picramnia.latifolia + Quassia.amara + Tabernaemontana.arborea + Trattinnickia.aspera + Xylopia.macrantha
## Data: all.dat
##      AIC      BIC    logLik
## 307.1163 322.7554 -144.5582
##
## Coefficients:
##              Value Std.Error   t-value p-value
## (Intercept)   -1.051752  2.1175346  -0.496687  0.6220
## Cordia.lasiocalyx    0.428920  0.2039316   2.103255  0.0415
## Hirtella.triandra    0.122279  0.0802638   1.523462  0.1351
## Picramnia.latifolia  0.662259  0.6358905   1.041468  0.3036
## Quassia.amara       4.085661  2.2842770   1.788602  0.0809
## Tabernaemontana.arborea -0.249725  0.1491192  -1.674667  0.1014
## Trattinnickia.aspera  1.349323  0.7147412   1.887848  0.0660
## Xylopia.macrantha    0.548832  0.1468772   3.736672  0.0006
##
## Correlation:
##              (Intr) Crd.ls Hrtll. Pcrmn. Qss.mr Tbrnm. Trtttn.
## Cordia.lasiocalyx   -0.618
## Hirtella.triandra   -0.212 -0.354
## Picramnia.latifolia  0.025 -0.019 -0.381
## Quassia.amara       0.163 -0.378  0.307 -0.302
## Tabernaemontana.arborea -0.708  0.245  0.163 -0.113  0.148
## Trattinnickia.aspera -0.139  0.187 -0.311  0.308 -0.708 -0.144
## Xylopia.macrantha   -0.140 -0.125  0.156 -0.463  0.314  0.279 -0.294
##
## Standardized residuals:
##      Min      Q1      Med      Q3      Max
## -1.87708765 -0.42701500 -0.04032793  0.23615609  3.38768871
##
## Residual standard error: 4.539713
## Degrees of freedom: 50 total; 42 residual
```

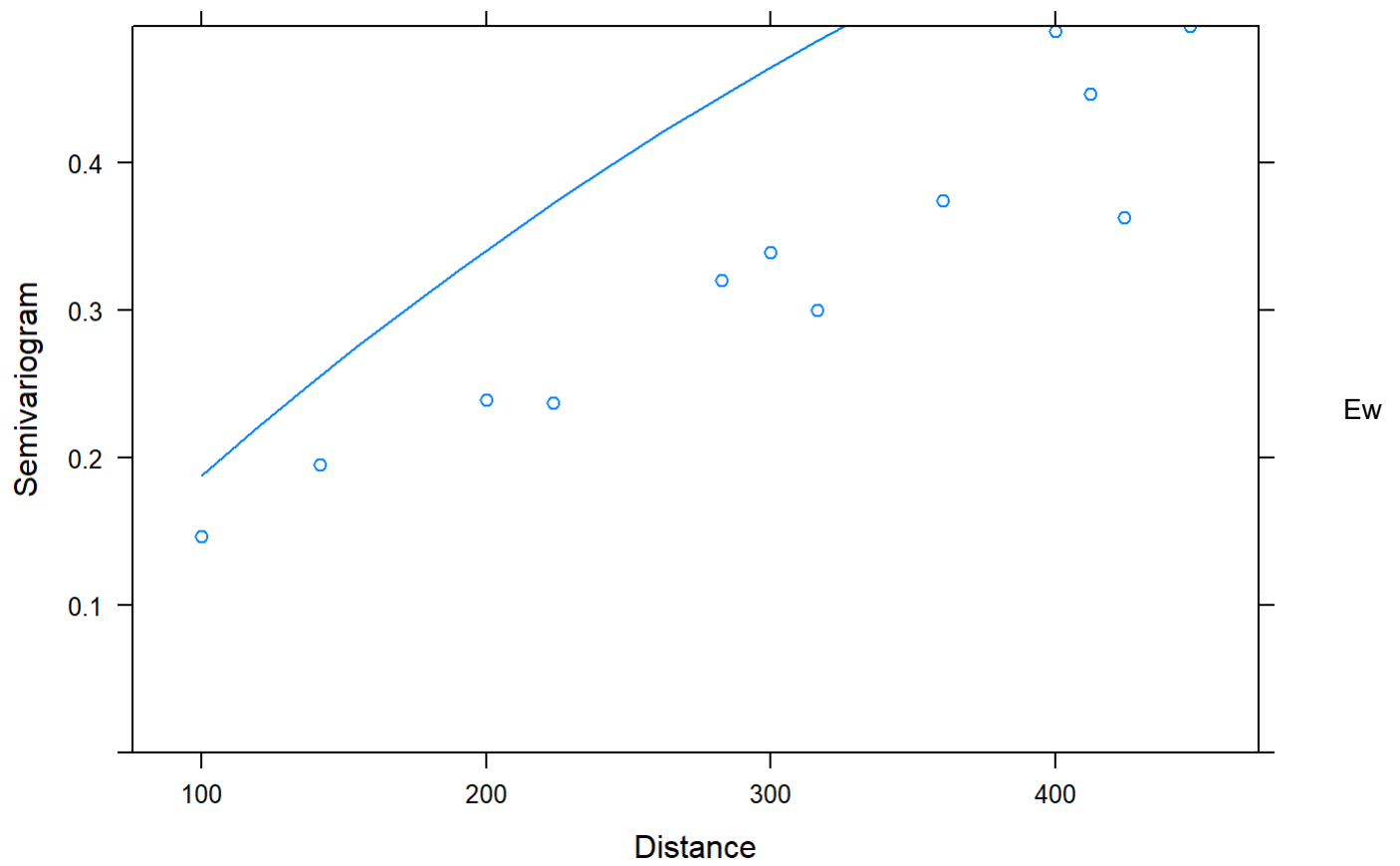
```
plot(Variogram(glm2, form = ~ x+y))
```



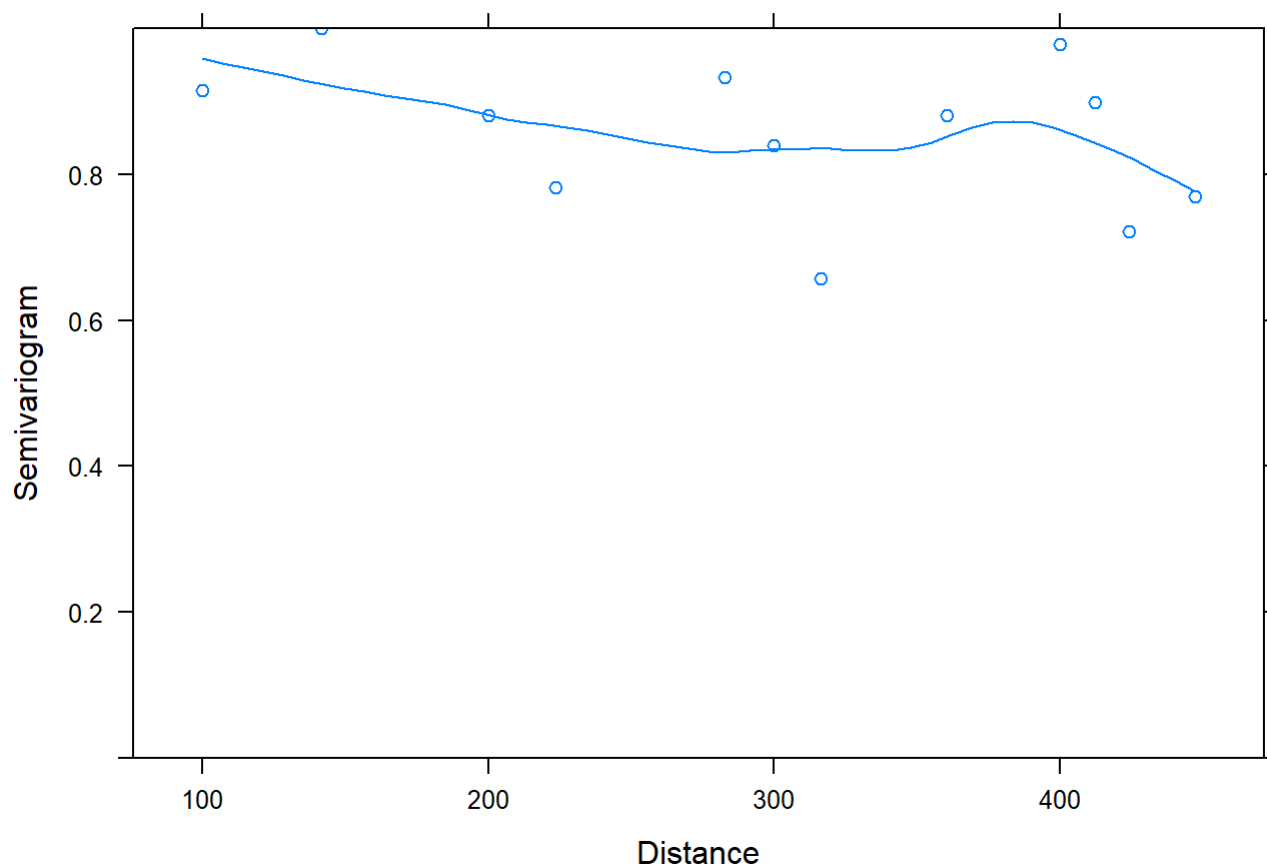
```
resid2 <- residuals(glm2)
plot(dist(all.dat[, c('x', 'y')]), dist(resid2))
lines(lowess(dist(all.dat[, c('x', 'y')]), dist(resid2)), col='red', lwd=2)
abline(v = max_dist, col='red', lwd=3, lty=2)
```



```
glm2.exp <- update(glm2, corr=corExp(form=~x + y))
plot(Variogram(glm2.exp, maxDist = max_dist))
```



```
plot(Variogram(glm2.exp, resType='normalized', maxDist = max_dist))
```



slightly better. suggests resids are kinda normal

```
anova(glm2, glm2.exp)
```

##	Model	df	AIC	BIC	logLik	Test	L.Ratio	p-value
##	glm2	1 9	307.1163	322.7554	-144.5582			
##	glm2.exp	2 10	301.6062	318.9829	-140.8031	1 vs 2	7.510175	0.0061

Anova indicates a significant difference between the models, and the spacial error term appears to have improved the model somewhat.

```
summary(glm2)
```

```
## Generalized least squares fit by REML
## Model: Drypetes.standleyi ~ Cordia.lasiocalyx + Hirtella.triandra + Picramnia.latifolia
## Data: all.dat
## AIC BIC logLik
## 307.1163 322.7554 -144.5582
##
## Coefficients:
## Value Std.Error t-value p-value
## (Intercept) -1.051752 2.1175346 -0.496687 0.6220
## Cordia.lasiocalyx 0.428920 0.2039316 2.103255 0.0415
## Hirtella.triandra 0.122279 0.0802638 1.523462 0.1351
## Picramnia.latifolia 0.662259 0.6358905 1.041468 0.3036
## Quassia.amara 4.085661 2.2842770 1.788602 0.0809
## Tabernaemontana.arborea -0.249725 0.1491192 -1.674667 0.1014
## Trattinnickia.aspera 1.349323 0.7147412 1.887848 0.0660
## Xylopia.macrantha 0.548832 0.1468772 3.736672 0.0006
##
## Correlation:
## (Intr) Crd.ls Hrtll. Pcrmn. Qss.mr Tbrnm. Trttn.
## Cordia.lasiocalyx -0.618
## Hirtella.triandra -0.212 -0.354
## Picramnia.latifolia 0.025 -0.019 -0.381
## Quassia.amara 0.163 -0.378 0.307 -0.302
## Tabernaemontana.arborea -0.708 0.245 0.163 -0.113 0.148
## Trattinnickia.aspera -0.139 0.187 -0.311 0.308 -0.708 -0.144
## Xylopia.macrantha -0.140 -0.125 0.156 -0.463 0.314 0.279 -0.294
##
## Standardized residuals:
## Min Q1 Med Q3 Max
## -1.87708765 -0.42701500 -0.04032793 0.23615609 3.38768871
##
## Residual standard error: 4.539713
## Degrees of freedom: 50 total; 42 residual
```

```
summary(glm2.exp)
```

```
## Generalized least squares fit by REML
## Model: Drypetes.standleyi ~ Cordia.lasiocalyx + Hirtella.triandra + Picramnia.latifolia
## Data: all.dat
## AIC BIC logLik
## 301.6062 318.9829 -140.8031
##
## Correlation Structure: Exponential spatial correlation
## Formula: ~x + y
## Parameter estimate(s):
## range
## 480.0567
##
## Coefficients:
## Value Std.Error t-value p-value
## (Intercept) 2.3485197 6.154919 0.381568 0.7047
## Cordia.lasiocalyx 0.1208390 0.179811 0.672033 0.5052
## Hirtella.triandra 0.0191759 0.098501 0.194677 0.8466
## Picramnia.latifolia 0.2014516 0.509196 0.395627 0.6944
## Quassia.amara 1.2792289 1.847570 0.692385 0.4925
## Tabernaemontana.arborea 0.0674943 0.133782 0.504511 0.6165
## Trattinnickia.aspera 1.8115374 0.525147 3.449582 0.0013
## Xylopia.macrantha 0.3388574 0.156874 2.160064 0.0365
##
## Correlation:
## (Intr) Crd.ls Hrtll. Pcrmn. Qss.mr Tbrnm. Trttn.
## Cordia.lasiocalyx -0.226
## Hirtella.triandra -0.309 -0.022
## Picramnia.latifolia 0.045 -0.066 -0.369
## Quassia.amara -0.059 -0.304 0.321 -0.142
## Tabernaemontana.arborea -0.240 -0.016 0.288 -0.221 0.112
## Trattinnickia.aspera -0.069 0.168 -0.237 0.212 -0.633 -0.041
## Xylopia.macrantha -0.056 -0.137 -0.063 0.109 0.290 0.102 -0.186
##
## Standardized residuals:
## Min Q1 Med Q3 Max
## -1.0051632 -0.5235683 -0.3176178 0.2208753 2.3746027
##
## Residual standard error: 8.628464
## Degrees of freedom: 50 total; 42 residual
```

The exponential spacial error model also seemed to improve this model somewhat.

Model 1: only include a single species as a predictor variable

Model 2: include all of the species as predictor variables

Did including the spatial error term have a large impact on the coefficients of the model?

```
coef(summary(glm1))
```



```
##              Value Std.Error   t-value    p-value
## (Intercept)  4.962810  1.030804  4.814503  1.511773e-05
## Quassia.amara 9.214876  2.304948  3.997867  2.194886e-04
```

```
coef(summary(glm1.exp))
```

```
##              Value Std.Error   t-value    p-value
## (Intercept)  9.794547 13.486551  0.7262455  0.47121500
## Quassia.amara 3.763073  1.495159  2.5168384  0.01523603
```

For the model testing only one predictor variable, addition of the spatial error term appeared to have an effect on the intercept, however it improved the std error and t-value, and still indicates a significant relationship (p-value) although perhaps not as strong as before.

```
coef(summary(glm2))
```

```
##              Value Std.Error   t-value    p-value
## (Intercept)  -1.0517523  2.11753456 -0.4966872  0.6219995234
## Cordia.lasiocalyx  0.4289202  0.20393162  2.1032551  0.0414742065
## Hirtella.triandra  0.1222789  0.08026383  1.5234616  0.1351381124
## Picramnia.latifolia  0.6622595  0.63589053  1.0414677  0.3036166790
## Quassia.amara  4.0856614  2.28427702  1.7886015  0.0808936542
## Tabernaemontana.arborea -0.2497250  0.14911920 -1.6746671  0.1014312033
## Trattinnickia.aspera  1.3493226  0.71474122  1.8878478  0.0659680951
## Xylopia.macrantha  0.5488320  0.14687723  3.7366715  0.0005567007
```

```
coef(summary(glm2.exp))
```

```
##              Value Std.Error   t-value    p-value
## (Intercept)  2.34851969  6.15491875  0.3815679  0.704705521
## Cordia.lasiocalyx  0.12083904  0.17981117  0.6720330  0.505242933
## Hirtella.triandra  0.01917586  0.09850099  0.1946768  0.846584598
## Picramnia.latifolia  0.20145161  0.50919625  0.3956267  0.694382984
## Quassia.amara  1.27922893  1.84756977  0.6923846  0.492506385
## Tabernaemontana.arborea 0.06749426  0.13378153  0.5045110  0.616539081
## Trattinnickia.aspera  1.81153745  0.52514691  3.4495822  0.001290775
## Xylopia.macrantha  0.33885742  0.15687377  2.1600643  0.036522530
```

In the model testing multiple species effects, we see some changes in the coefficients indicating far fewer species have a significant effect on the response variable.

Did including the spatial error terms significantly improve model fit (use function anova to carry out model comparison)?

Yes, p-values indicated significance. Including the spatial error terms did improve the model fit.

Explain why you did or did not observe a difference in the influence of adding the spatial error term between the two models.

Adding the spacial error term had an influence on the model, indicating that spacial dependance should be taken into consideration when building a model to predict the abundance of *Drypetes standleyi*. In this case, the model improved, indicating that the addition of a spacial error term fitted the model closer to the algorithm matching the relationship between the possible predictor variables in the dataset for *D. standleyi* abundance. In otherwords, it improved the model because spacial dependance likely has an effect on the abundance of *Drypetes standleyi*.