

Spatial R Exercise 3: Spatial Dataset Interactions

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Chapter 1

Data interactions

In this exercise, we'll look at using different types of geospatial data to try and answer some questions about the past. In this case, we're interested how micromammal communities of southern Africa reflect

1.1 Adding packages

First, as usual, we'll start by loading the packages we need to work with spatial data: `sf` and `terra`.

```
#Load packages  
require("sf")
```

```
## Loading required package: sf
```

```
## Linking to GEOS 3.9.1, GDAL 3.3.2, PROJ 7.2.1; sf_use_s2() is TRUE
```

```
require("terra")
```

```
## Loading required package: terra
```

```
## terra 1.5.21
```


Chapter 2

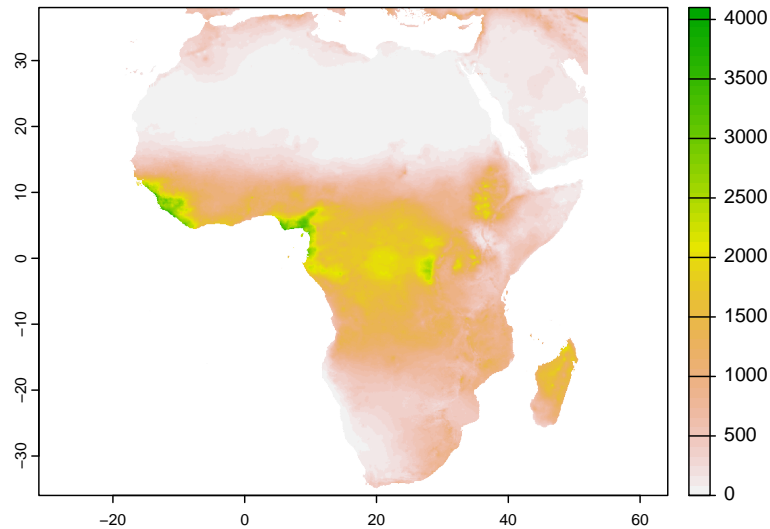
Spatial subsetting: Vector to Raster

Here we'll work on subsetting one spatial dataset with another. In this case, we want to use a vector dataset to set the limits of a raster dataset.

2.1 Load vector and raster data

Let's use `rast` to load some data that we'll be using to assess contemporary seasonality in southern Africa.

```
#Load annual precipitation data for all of Africa  
annRain<-rast("africaANNPPT.tif")  
#Plot it  
plot(annRain)
```



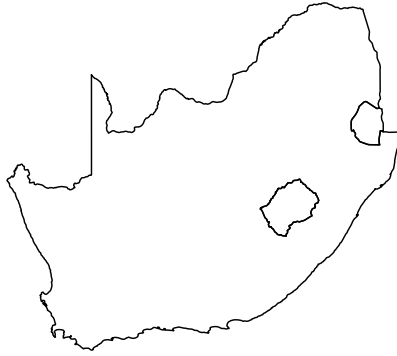
This precipitation data comes from the Terraclimate product, giving modeled monthly spatial averages from 1958-present. Here, the values for each year have been summed, and then these annual values are then averaged to give average annual precipitation.

However, this particular raster covers all of Africa, which is way more data that we need. We'll import southern Africa borders shapefile to look at how we might subset it.

```
saBorders<-st_read("south_africa_border.shp")
```

```
## Reading layer `south_africa_border' from data source
##   `C:\Users\bdav_\Dropbox\Teaching\Spatial R Short Course\Bookdown\Exercise3\Exerci
##   using driver `ESRI Shapefile'
## Simple feature collection with 3 features and 94 fields
## Geometry type: POLYGON
## Dimension:     XY
## Bounding box:  xmin: 16.46998 ymin: -34.82195 xmax: 32.89308 ymax: -22.12645
## Geodetic CRS:  WGS 84
```

```
plot(st_geometry(saBorders))
```

2.2 Crop raster with vector data

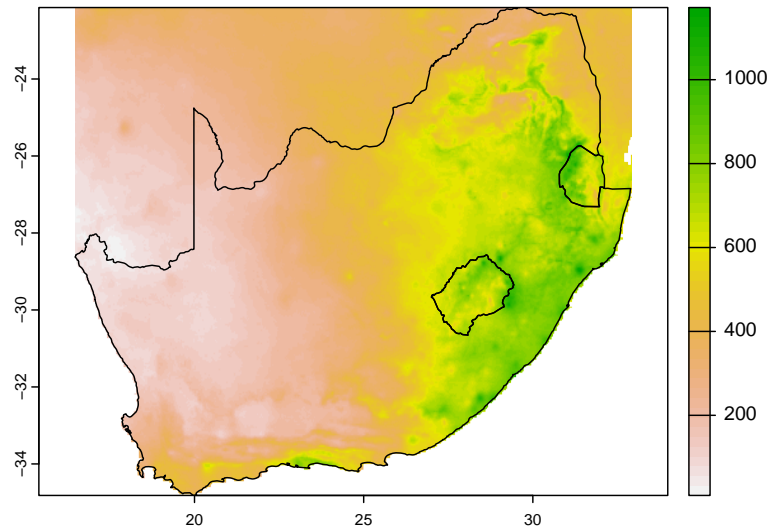
Now that we have a raster and a vector dataset, we'll use the latter to spatially subset the former. To do this, we will need the `crop` function.

```
#Crop rainfall data to southern Africa  
saAnnRain<-crop(annRain,saBorders)
```

The first argument to the `crop` function is the raster we want crop (`annRain`). The second is the object we are using to set the new margins (`saBorders`).

Now we can take a look...

```
#Plot new raster and borders  
plot(saAnnRain)  
plot(st_geometry(saBorders),add=T)
```



Hmmm... this looks OK, but maybe not exactly what we wanted since the rainfall values extend beyond the area of interest. The main reason for this is that we used an `sf` object as the cropping object. When `terra` tries to do this, it can only crop to the extent of the object.

In order to crop more closely, we need to turn our `saBorders` object into a `terra` `SpatVector` object, which is a format `terra` uses to represent vector data. We can use the `vect` function for this:

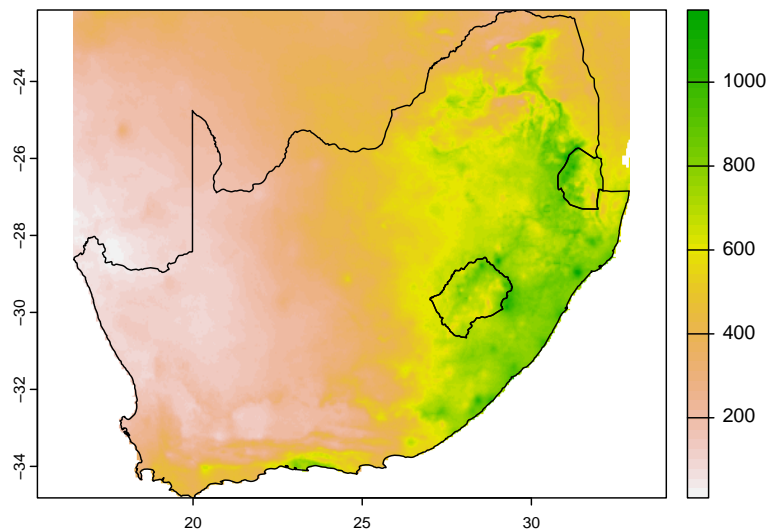
```
#Turn sf object into a SpatVector dataset
saBordersSF<-vect(saBorders)
saBordersSF
```

```
## class      : SpatVector
## geometry   : polygons
## dimensions  : 3, 94 (geometries, attributes)
## extent     : 16.46998, 32.89308, -34.82195, -22.12645 (xmin, xmax, ymin, ymax)
## coord. ref. : lon/lat WGS 84 (EPSG:4326)
## names      : featurecla scalerank LABELRANK SOVEREIGNT SOV_A3 ADMO_DIF
## type       : <chr> <int> <int> <chr> <chr> <int>
## values     : Admin-0 country 0 2 South Africa ZAF 0
##             Admin-0 country 0 4 eSwatini SWZ 0
##             Admin-0 country 0 6 Lesotho LSO 0
## LEVEL      TYPE ADMIN ADMO_A3 (and 84 more)
## <int>      <chr> <chr> <chr>
```

```
##      2 Sovereign country South Africa      ZAF
##      2 Sovereign country   eSwatini       SWZ
##      2 Sovereign country    Lesotho      LSO
```

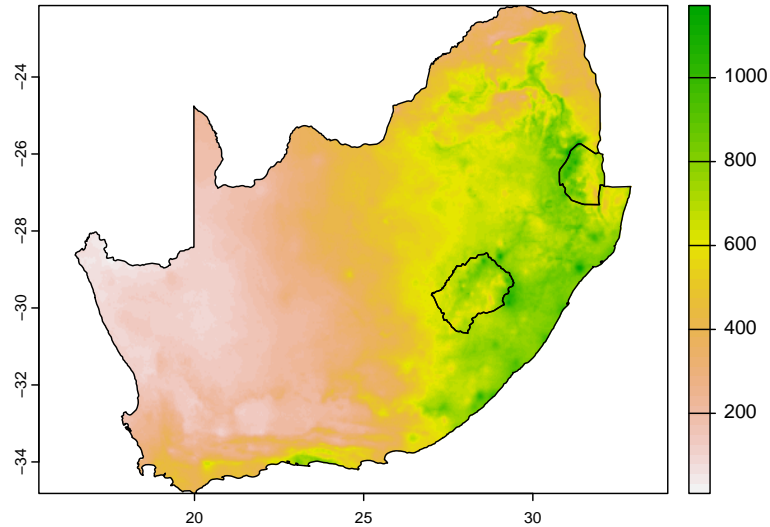
OK, now we have the borders as a `SpatVector`. We can plug this in and see what we get:

```
#Crop rain data again with SpatVector
saAnnRain<-crop(annRain,saBordersSF)
plot(saAnnRain)
plot(st_geometry(saBorders),add=T)
```



Still not quite there! The last step is that we need add the `mask` argument to the `crop` function and set it to `true`. This tells `crop` to take any values outside the cropping object and convert them to `NA` values.

```
#Crop rain data again with SpatVector
saAnnRain<-crop(annRain,saBordersSF,mask=T)
plot(saAnnRain)
plot(st_geometry(saBorders),add=T)
```



There we go! That looks much better. Let's use `global` to get a sense of the average and standard deviation on this new raster.

```
#Get some stats
global(saAnnRain, "mean")
```

```
##              mean
## africaAnnPPT NaN
```

```
global(saAnnRain, "sd")
```

```
##              sd
## africaAnnPPT NaN
```

Ack! When we see NaN, that stands for “not a number”. That means R tried to do something mathematically impossible. The problem here is that, since we used `mask` to carve out our raster, there are now lots of NA values in the raster, and R can't calculate statistics when these values are present. We can use `na.rm` to get rid of them.

```
#Get some stats
global(saAnnRain, "mean", na.rm=TRUE)
```

```
##                               mean
## africaAnnPPT 439.047
```

```
global(saAnnRain, "sd", na.rm=TRUE)
```

```
##                               sd
## africaAnnPPT 228.5005
```

2.3 Try it yourself!

- It's very easy to make histograms out of raster data. See if you can make a histogram out of the rainfall data for southern Africa and then all of Africa. How do they compare?
- Can you create a boolean raster for places in southern Africa that get more than the mean rainfall?

Chapter 3

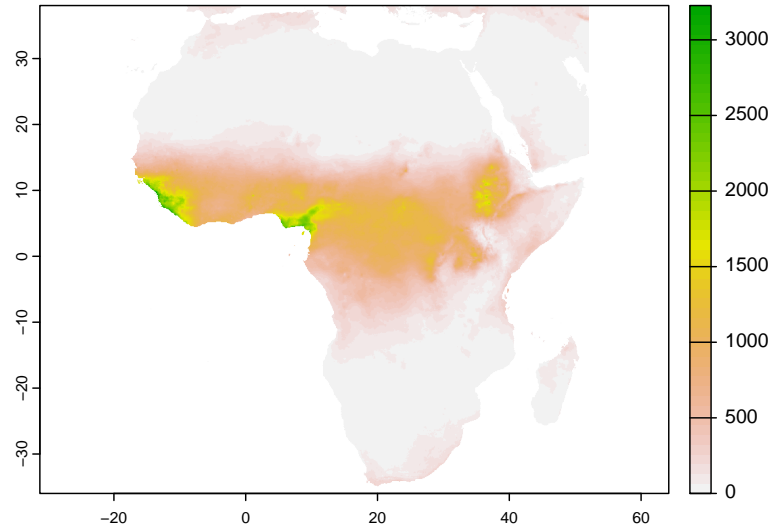
Spatial subsetting: Raster to Raster

This time, we're going to use a raster data to perform a spatial subset on another raster. This should be a little more straightforward since we aren't trying to make **sf** and **terra** talk to each other.

3.1 Add more raster data

Here we'll add another Terraclimate-derived dataset.

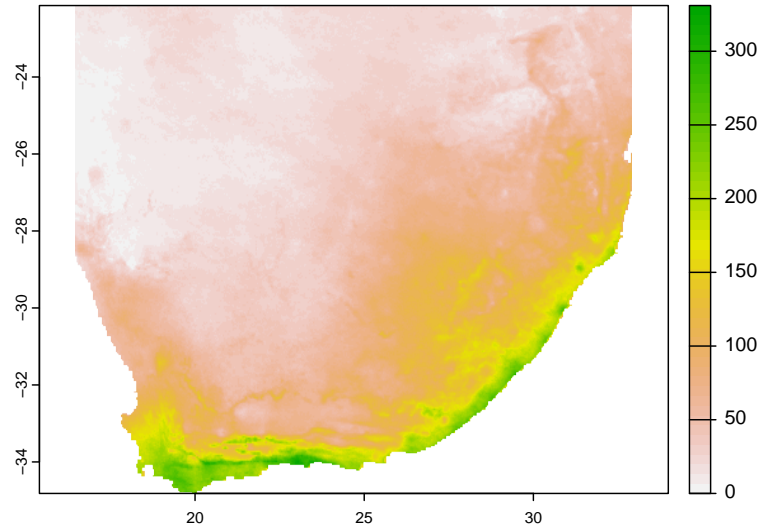
```
#Import winter rainfall  
winRain<-rast("africaWinPPT.tif")  
  
#Plot it  
plot(winRain)
```



This is just like the annual average rainfall data, but only for the winter months (June, July, and August).

Since our annual rainfall data has already been cropped, we should be able to use that data to crop our new dataset.

```
#Crop winter rain using annual rain  
saWinRain<-crop(winRain,saAnnRain,mask=TRUE)  
  
#Plot it  
plot(saWinRain)
```

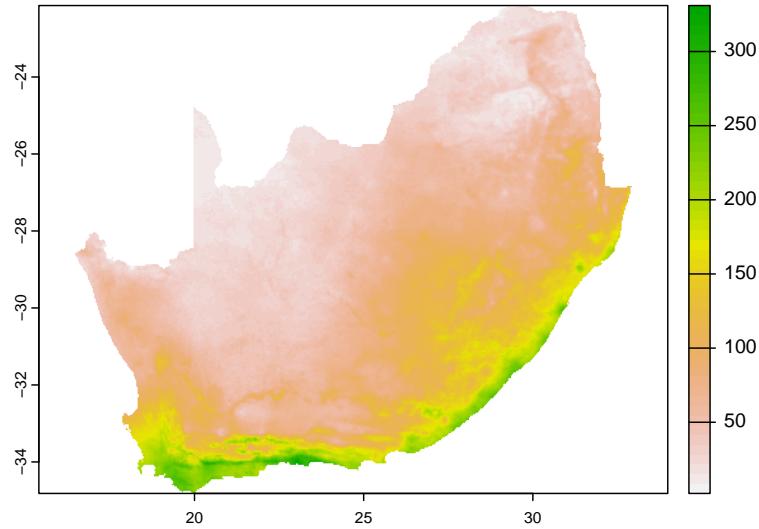



This again! The problem here is that the `mask` argument here only applies to a `SpatVector` object like we made in the previous section. If we want to use a raster to mask another raster, we need to call the `mask` function:

```
#Crop it again
saWinRain<-crop(winRain,saAnnRain)

#Mask the winter rain data with the annual rain data
saWinRain<-mask(saWinRain,saAnnRain)

#Plot it
plot(saWinRain)
```

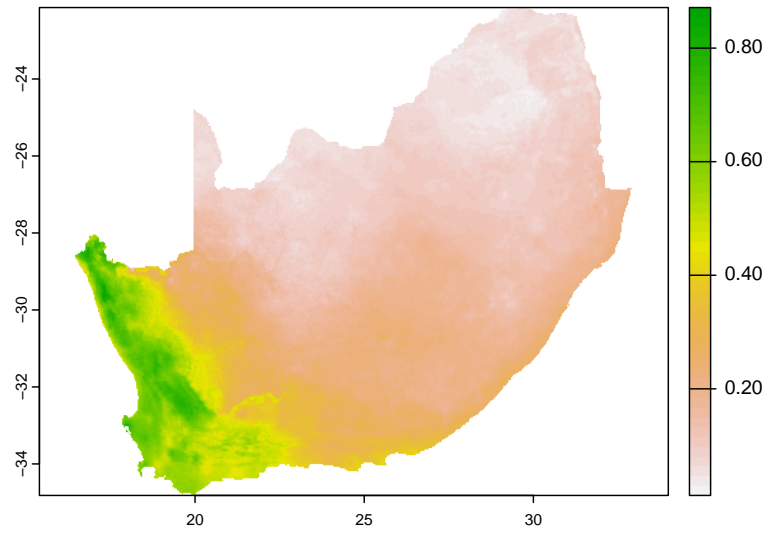


OK, great!

3.2 Raster algebra revisited

OK, for this last step, we want to know what percentage of an area's rainfall occurs in the winter. This will give us a sense of seasonality across southern Africa. To get this all we have to do is divide the winter rainfall values by the total annual values:

```
#Get winter percentages  
winRainPercent<-saWinRain/saAnnRain  
  
#Plot it  
plot(winRainPercent)
```



Cool! Hopefully from this map you can see southern Africa's very distinctive winter and summer rainfall zones.

Chapter 4

Spatial subsetting: Vector to Vector

The last spatial subsetting we'll do is using one vector to subset another.

4.1 Convert table data to vector data

First let's bring in some table data with spatial coordinates:

```
micro<-read.csv("micromammals.csv")
head(micro)
```

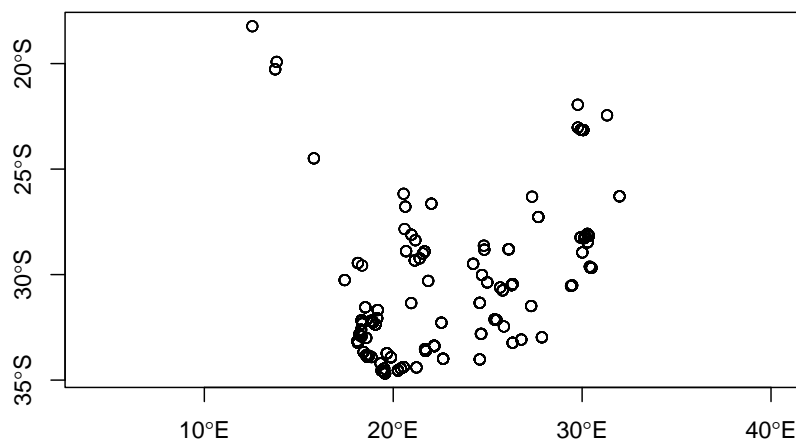
##		Site	Abbreviation	Latitude	Longitude	CODE	ORDER
## 1	100 Elliott Street	Kokstad	ESK	-30.54	29.41	CCYA	SORICOMORPHA
## 2	100 Elliott Street	Kokstad	ESK	-30.54	29.41	CFLA	SORICOMORPHA
## 3	100 Elliott Street	Kokstad	ESK	-30.54	29.41	DINC	RODENTIA
## 4	100 Elliott Street	Kokstad	ESK	-30.54	29.41	MMIN	RODENTIA
## 5	100 Elliott Street	Kokstad	ESK	-30.54	29.41	MMUS	RODENTIA
## 6	100 Elliott Street	Kokstad	ESK	-30.54	29.41	MNAT	RODENTIA
##	FAMILY	SUBFAMILY	GENUS	SPECIES			COMMONNAME
## 1	Soricidae	Crocidurinae	Crocidura	cyanea		Reddish-gray	Musk Shrew
## 2	Soricidae	Crocidurinae	Crocidura	flavescens		Greater Red	Musk Shrew
## 3	Muridae	Murinae	Dasymys	incomtus	s.l.		Common Dasymys
## 4	Muridae	Murinae	Mus	minutoides		Southern African	Pygmy Mouse
## 5	Muridae	Murinae	Mus	musculus			House Mouse
## 6	Muridae	Murinae	Mastomys	natalensis			Natal Mastomys

These are modern micromammal occurrences that were used in this paper:

Faith, J. Tyler, Brian M. Chase, and D. Margaret Avery. 2019. “Late Quaternary Micromammals and the Precipitation History of the Southern Cape, South Africa.” *Quaternary Research* 91 (2): 848–60. <https://doi.org/10.1017/qua.2018.105>.

The information includes the site, the coordinates, and the species encountered. In order to make use of it in our analysis, we need to turn it into vector data.

```
microSF<-st_as_sf(micro,coords=c("Longitude","Latitude"),crs=4326)
plot(st_geometry(microSF),axes=T)
```



4.2 Cropping vector to vector

The way to crop one vector dataset with another is to use the `st_crop` function.

```
#Crop micromammal dataset to southern Africa
micromammals<-st_crop(microSF,saBorders)
```

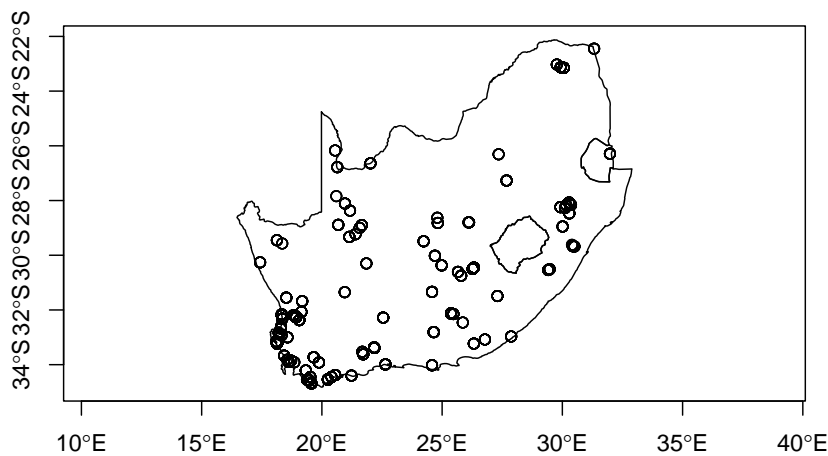
```
## Warning: attribute variables are assumed to be spatially constant throughout all
## geometries
```

Note that warning, but don't worry too much about it. This is **terra** reminding you that, by subsetting this way, you are assuming that the attributes are spatially constant.

Imagine if you were using this function to crop a polygon that represented grassland vegetation. This wouldn't be problematic, because the cropped polygon would still be grassland. However, now imagine that you used it to crop the polygon for Lesotho, which contains a population estimate. That population estimate would no longer be valid for the cropped section of Lesotho.

Here, we're not really worried about this because we're pruning points, so there isn't any cutting of individual features. Let's take a look.

```
#Plot borders and micromammals  
plot(st_geometry(saBorders), axes=T)  
plot(st_geometry(micromammals), add=T)
```



Looks good!

4.3 Try it yourself

- Create a plot that shows the borders, but instead of just the micromammal point locations, plot them by genus (hint: this may involve square brackets).
- Change the symbol shape with the `pch` argument (it takes an integer and defaults to 1).

Chapter 5

Spatial Sampling

For this final step, we're going to use our micromammal vector data to sample our raster data on rainfall seasonality.

5.1 Extract point values

To accomplish this, we need to use `extract`, which we've seen before. In that instance, we used a matrix of XY values. Here, we want to use the locations of our micromammals. To use it here with our `SpatRaster` object, we need to make it into a `SpatVector` object with `vect`.

```
winRainMM<-extract(winRainPercent,vect(micromammals))
winRainMM
```

```
##      ID africaWinPPT
## 1      1    0.17195767
## 2      2    0.17195767
## 3      3    0.17195767
## 4      4    0.17195767
## 5      5    0.17195767
## 6      6    0.17195767
## 7      7    0.17195767
## 8      8    0.17195767
## 9      9    0.17195767
## 10     10    0.17195767
## 11     11    0.14122682
## 12     12    0.14122682
## 13     13    0.14122682
```

## 14	14	0.14122682
## 15	15	0.14122682
## 16	16	0.14122682
## 17	17	0.14122682
## 18	18	0.14122682
## 19	19	0.14122682
## 20	20	0.14122682
## 21	21	0.14122682
## 22	22	0.14122682
## 23	23	0.14122682
## 24	24	0.13150289
## 25	25	0.13150289
## 26	26	0.13150289
## 27	27	0.13150289
## 28	28	0.13150289
## 29	29	0.13150289
## 30	30	0.13150289
## 31	31	0.13150289
## 32	32	0.13150289
## 33	33	0.09735974
## 34	34	0.09735974
## 35	35	0.09735974
## 36	36	0.09735974
## 37	37	0.09735974
## 38	38	0.09735974
## 39	39	0.63436123
## 40	40	0.63436123
## 41	41	0.63436123
## 42	42	0.63436123
## 43	43	0.63436123
## 44	44	0.63436123
## 45	45	0.63436123
## 46	46	0.63436123
## 47	47	0.63436123
## 48	48	0.63436123
## 49	49	0.63436123
## 50	50	0.63436123
## 51	51	0.63436123
## 52	52	0.63436123
## 53	53	0.63436123
## 54	54	0.63436123
## 55	55	0.63436123
## 56	56	0.63436123
## 57	57	0.63436123
## 58	58	0.63436123
## 59	59	0.63436123

## 60	60	0.63436123
## 61	61	0.63436123
## 62	62	0.63436123
## 63	63	0.63436123
## 64	64	0.63436123
## 65	65	0.63436123
## 66	66	0.63436123
## 67	67	0.63436123
## 68	68	0.63436123
## 69	69	0.63436123
## 70	70	0.60000000
## 71	71	0.60000000
## 72	72	0.60000000
## 73	73	0.60000000
## 74	74	0.60000000
## 75	75	0.60000000
## 76	76	0.60000000
## 77	77	0.60000000
## 78	78	0.60000000
## 79	79	0.60000000
## 80	80	0.60000000
## 81	81	0.60000000
## 82	82	0.60000000
## 83	83	0.70895522
## 84	84	0.70895522
## 85	85	0.70895522
## 86	86	0.70895522
## 87	87	0.70895522
## 88	88	0.70895522
## 89	89	0.70895522
## 90	90	0.70895522
## 91	91	0.70895522
## 92	92	0.70895522
## 93	93	0.70895522
## 94	94	0.70895522
## 95	95	0.70895522
## 96	96	0.70895522
## 97	97	0.70895522
## 98	98	0.70895522
## 99	99	0.70895522
## 100	100	0.70895522
## 101	101	0.53664303
## 102	102	0.53664303
## 103	103	0.53664303
## 104	104	0.53664303
## 105	105	0.53664303

##	106	106	0.53664303
##	107	107	0.53664303
##	108	108	0.53664303
##	109	109	0.18820577
##	110	110	0.18820577
##	111	111	0.18820577
##	112	112	0.18820577
##	113	113	0.18820577
##	114	114	0.18820577
##	115	115	0.18820577
##	116	116	0.17977528
##	117	117	0.17977528
##	118	118	0.17977528
##	119	119	0.05405405
##	120	120	0.05405405
##	121	121	0.05405405
##	122	122	0.05405405
##	123	123	0.05405405
##	124	124	0.05405405
##	125	125	0.57986871
##	126	126	0.57986871
##	127	127	0.57986871
##	128	128	0.57986871
##	129	129	0.57986871
##	130	130	0.57986871
##	131	131	0.57986871
##	132	132	0.57986871
##	133	133	0.57986871
##	134	134	0.57986871
##	135	135	0.09001637
##	136	136	0.09001637
##	137	137	0.09001637
##	138	138	0.09001637
##	139	139	0.09001637
##	140	140	0.09001637
##	141	141	0.09001637
##	142	142	0.09001637
##	143	143	0.09001637
##	144	144	0.09001637
##	145	145	0.09001637
##	146	146	0.09001637
##	147	147	0.09001637
##	148	148	0.09001637
##	149	149	0.09001637
##	150	150	0.09001637
##	151	151	0.09001637

##	152	152	0.09001637
##	153	153	0.09001637
##	154	154	0.13785047
##	155	155	0.13785047
##	156	156	0.13785047
##	157	157	0.13785047
##	158	158	0.13785047
##	159	159	0.13785047
##	160	160	0.13785047
##	161	161	0.13785047
##	162	162	0.13785047
##	163	163	0.13785047
##	164	164	0.13785047
##	165	165	0.63348416
##	166	166	0.63348416
##	167	167	0.63348416
##	168	168	0.63348416
##	169	169	0.63348416
##	170	170	0.63348416
##	171	171	0.63348416
##	172	172	0.63348416
##	173	173	0.63348416
##	174	174	0.63348416
##	175	175	0.63348416
##	176	176	0.60606061
##	177	177	0.60606061
##	178	178	0.60606061
##	179	179	0.60606061
##	180	180	0.60606061
##	181	181	0.60606061
##	182	182	0.60606061
##	183	183	0.60606061
##	184	184	0.46351931
##	185	185	0.46351931
##	186	186	0.46351931
##	187	187	0.46351931
##	188	188	0.46351931
##	189	189	0.46351931
##	190	190	0.46351931
##	191	191	0.46351931
##	192	192	0.46351931
##	193	193	0.46351931
##	194	194	0.46351931
##	195	195	0.46351931
##	196	196	0.46351931
##	197	197	0.26937269

##	198	198	0.26937269
##	199	199	0.26937269
##	200	200	0.26937269
##	201	201	0.26937269
##	202	202	0.26937269
##	203	203	0.26937269
##	204	204	0.26937269
##	205	205	0.26937269
##	206	206	0.49870130
##	207	207	0.49870130
##	208	208	0.49870130
##	209	209	0.49870130
##	210	210	0.49870130
##	211	211	0.49870130
##	212	212	0.49870130
##	213	213	0.49870130
##	214	214	0.49870130
##	215	215	0.49870130
##	216	216	0.49870130
##	217	217	0.49870130
##	218	218	0.49870130
##	219	219	0.49870130
##	220	220	0.49870130
##	221	221	0.49870130
##	222	222	0.49870130
##	223	223	0.49870130
##	224	224	0.49870130
##	225	225	0.49870130
##	226	226	0.49870130
##	227	227	0.49870130
##	228	228	0.49870130
##	229	229	0.49870130
##	230	230	0.49870130
##	231	231	0.49870130
##	232	232	0.49870130
##	233	233	0.49870130
##	234	234	0.49870130
##	235	235	0.49870130
##	236	236	0.49870130
##	237	237	0.49870130
##	238	238	0.49870130
##	239	239	0.49870130
##	240	240	0.49870130
##	241	241	0.56511628
##	242	242	0.56511628
##	243	243	0.56511628

##	244	244	0.56511628
##	245	245	0.56511628
##	246	246	0.56511628
##	247	247	0.56511628
##	248	248	0.56511628
##	249	249	0.56511628
##	250	250	0.56511628
##	251	251	0.56511628
##	252	252	0.56511628
##	253	253	0.69082126
##	254	254	0.69082126
##	255	255	0.69082126
##	256	256	0.69082126
##	257	257	0.69082126
##	258	258	0.69082126
##	259	259	0.69082126
##	260	260	0.69082126
##	261	261	0.69082126
##	262	262	0.69082126
##	263	263	0.69082126
##	264	264	0.69082126
##	265	265	0.69082126
##	266	266	0.69082126
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## 1455 1455 0.18181818
## 1456 1456 0.18181818
## 1457 1457 0.18181818
## 1458 1458 0.13903743
## 1459 1459 0.13903743
## 1460 1460 0.13903743
## 1461 1461 0.13903743
## 1462 1462 0.13903743
## 1463 1463 0.13903743
## 1464 1464 0.13903743
## 1465 1465 0.13903743
## 1466 1466 0.13903743
## 1467 1467 0.13903743
## 1468 1468 0.13903743
## 1469 1469 0.13903743
## 1470 1470 0.13903743
## 1471 1471 0.13903743
## 1472 1472 0.57048458
## 1473 1473 0.57048458
## 1474 1474 0.57048458
## 1475 1475 0.57048458
## 1476 1476 0.57048458
## 1477 1477 0.57048458
## 1478 1478 0.57048458
## 1479 1479 0.57048458
## 1480 1480 0.57048458
## 1481 1481 0.57048458
## 1482 1482 0.57048458
```

Great! This is a dataframe with our winter rainfall percentage data. Notice it takes it's name from the original winter rainfall dataset that we read in earlier.

Now we just want to tack this on to our micromammal dataset. To do this, we'll first turn the table above into a vector of values, and then use `cbind` to add that vector as a new column.

```
#Add this column back to the micromammal data
```

```
WR<-winRainMM$africaWinPPT
micromammals<-cbind(micromammals,WR)
micromammals
```

```
## Simple feature collection with 1482 features and 10 fields
## Geometry type: POINT
## Dimension: XY
## Bounding box: xmin: 17.43 ymin: -34.69 xmax: 31.98 ymax: -22.45
## Geodetic CRS: WGS 84
## First 10 features:
```

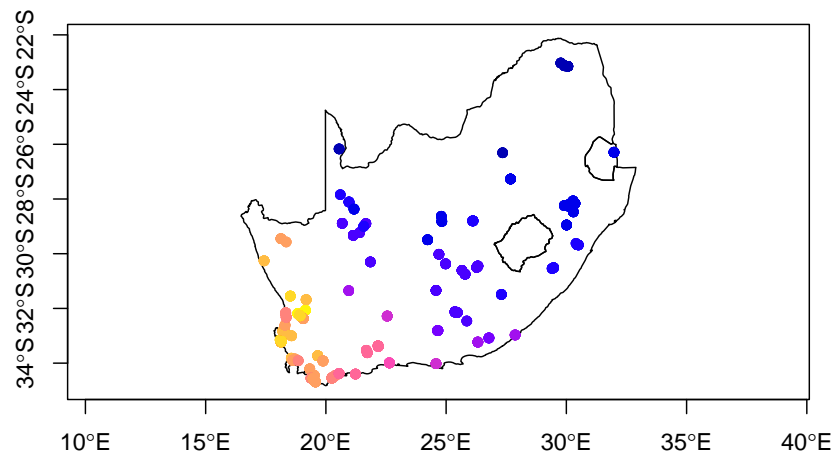
##	Site	Abbreviation	CODE	ORDER	FAMILY
## 1	100 Elliott Street Kokstad	ESK CCYA	SORICOMORPHA	Soricidae	
## 2	100 Elliott Street Kokstad	ESK CFLA	SORICOMORPHA	Soricidae	
## 3	100 Elliott Street Kokstad	ESK DINC	RODENTIA	Muridae	
## 4	100 Elliott Street Kokstad	ESK MMIN	RODENTIA	Muridae	
## 5	100 Elliott Street Kokstad	ESK MMUS	RODENTIA	Muridae	
## 6	100 Elliott Street Kokstad	ESK MNAT	RODENTIA	Muridae	
## 7	100 Elliott Street Kokstad	ESK MVAR	SORICOMORPHA	Soricidae	
## 8	100 Elliott Street Kokstad	ESK OIRR	RODENTIA	Muridae	
## 9	100 Elliott Street Kokstad	ESK RPUM	RODENTIA	Muridae	
## 10	100 Elliott Street Kokstad	ESK SINF	SORICOMORPHA	Soricidae	

```
## SUBFAMILY GENUS SPECIES COMMONNAME WR
## 1 Crocidurinae Crocidura cyanea Reddish-gray Musk Shrew 0.1719577
## 2 Crocidurinae Crocidura flavescens Greater Red Musk Shrew 0.1719577
## 3 Murinae Dasymys incommis s.l. Common Dasymys 0.1719577
## 4 Murinae Mus minutoides Southern African Pygmy Mouse 0.1719577
## 5 Murinae Mus musculus House Mouse 0.1719577
## 6 Murinae Mastomys natalensis Natal Mastomys 0.1719577
## 7 Myosoricinae Myosorex varius Forest Shrew 0.1719577
## 8 Otomyinae Otomys irroratus Southern African Vlei Rat 0.1719577
## 9 Murinae Rhabdomys pumilio Xeric Four-striped Grass Rat 0.1719577
## 10 Crocidurinae Suncus infinitimus Least Dwarf Shrew 0.1719577
## geometry
## 1 POINT (29.41 -30.54)
## 2 POINT (29.41 -30.54)
## 3 POINT (29.41 -30.54)
## 4 POINT (29.41 -30.54)
## 5 POINT (29.41 -30.54)
## 6 POINT (29.41 -30.54)
## 7 POINT (29.41 -30.54)
## 8 POINT (29.41 -30.54)
```

```
## 9 POINT (29.41 -30.54)
## 10 POINT (29.41 -30.54)
```

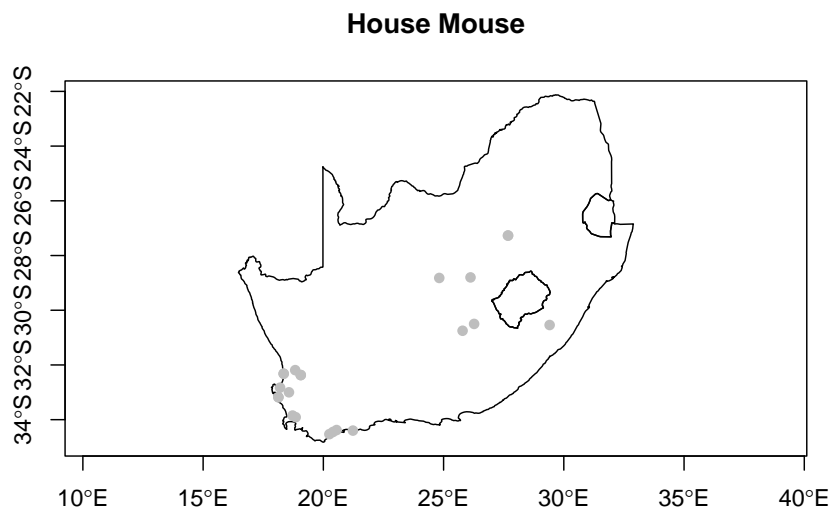
We did it! Now we can start to look at the seasonality of different species niches. First, let's plot the data:

```
#Plot borders and micromammals by their winter rainfall percentage
plot(st_geometry(saBorders),axes=T)
plot(micromammals['WR'],add=T,pch=16)
```



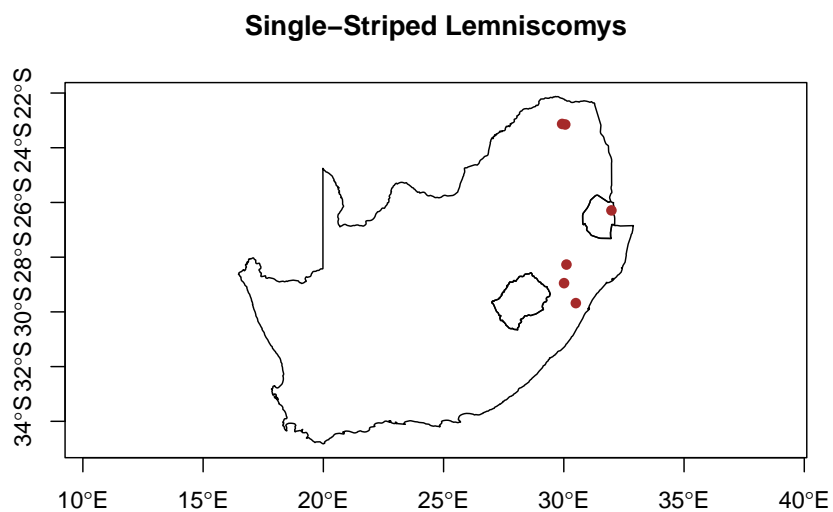
We can use `subset` to look at specific micromammals. Here's the house mouse:

```
#Plot house mouse rainfall
houseMouse<-subset(micromammals,COMMONNAME=="House Mouse")
plot(st_geometry(saBorders),axes=T,main="House Mouse")
plot(st_geometry(houseMouse),add=T,pch=16,col="gray")
```



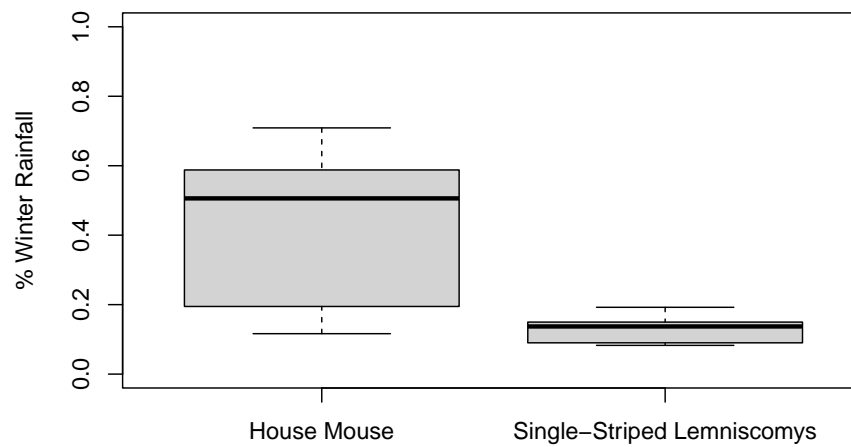
And here's the single striped lemniscomys:

```
lemniscomys<-subset(micromammals,COMMONNAME=="Single-Striped Lemniscomys")
plot(st_geometry(saBorders),axes=T,main="Single-Striped Lemniscomys")
plot(st_geometry(lemniscomys),add=T,pch=16,col="brown")
```



Finally, we can use this data to compare these species. Here's a `boxplot` for these two species.

```
boxplot(houseMouse$WR,lemniscomys$WR,names=c("House Mouse","Single-Striped Lemniscomys"),ylim=c(0.0,1.0))
```



Chapter 6

Bring it all together

- Can you subset the borders data to include only Lesotho?
- Can you then spatially subset the annual and winter rainfall data? Can you get the percentage for Lesotho? The subsetting methods are up to you, but make sure you plot both with axes.
- Can you get a histogram of Lesotho annual rainfall?