

Geospatial Data in R

Geospatial Data in R *And Beyond!*

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Spatial Data in R

Historical perspective



Before R

- The S Language (Chambers, Becker, Wilks)
 - Born 1976
 - Released 1984
 - Reborn 1988 (The New S Language)
- S-Plus
 - Commercialised version, 1988

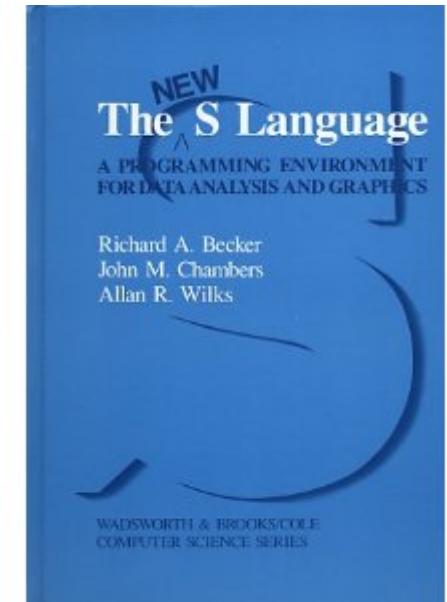
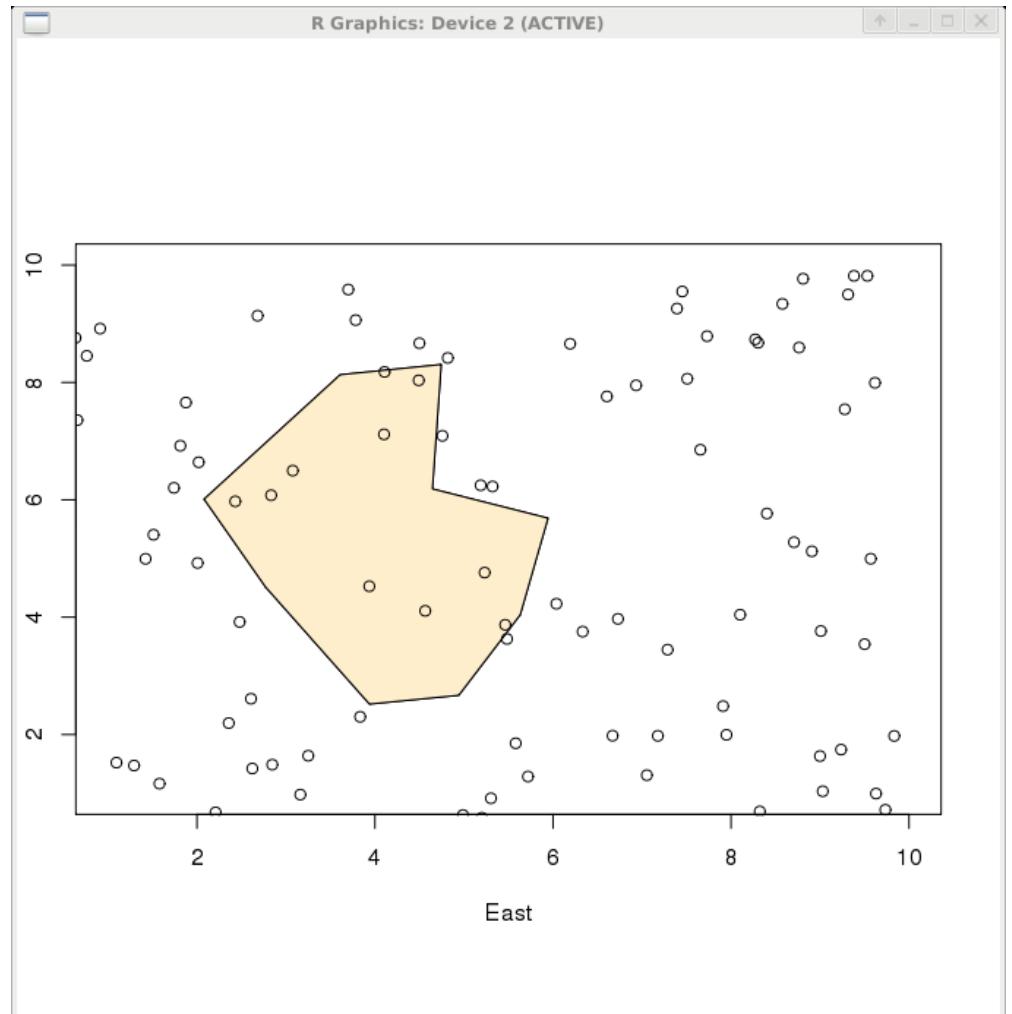


Image from amazon.com

points, polygon, lines

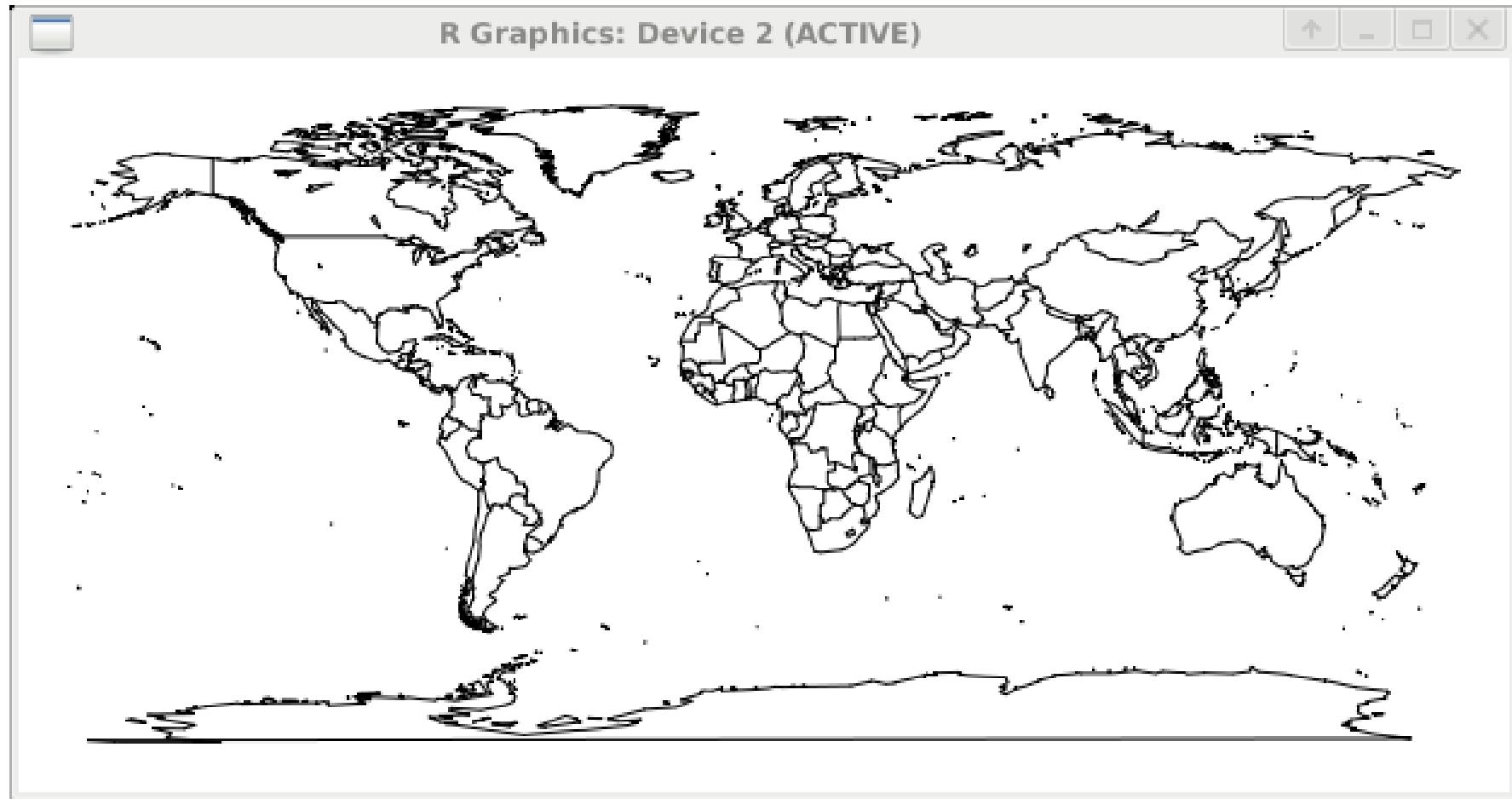
- “Roll your own” maps
- `plot(xy, asp=1)`
- `polygon(xy, col=...)`
- `points(xy, pch=...)`
- `lines(xy, lwd=2, ...)`





First Maps

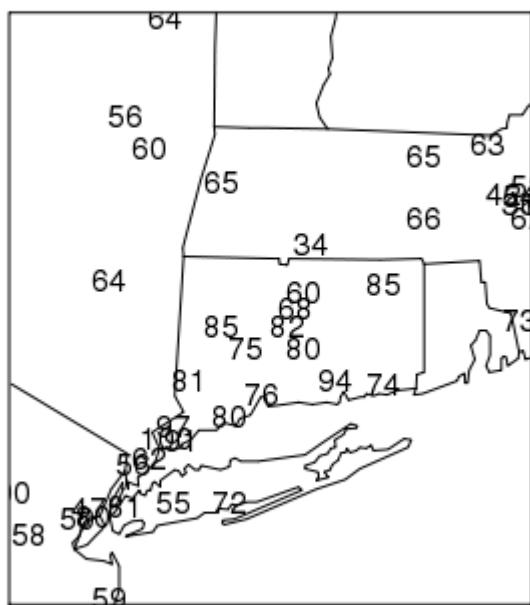
- The map package: Becker and Wilks
- **library(maps); map()**



R Graphics: Device 2 (ACTIVE)

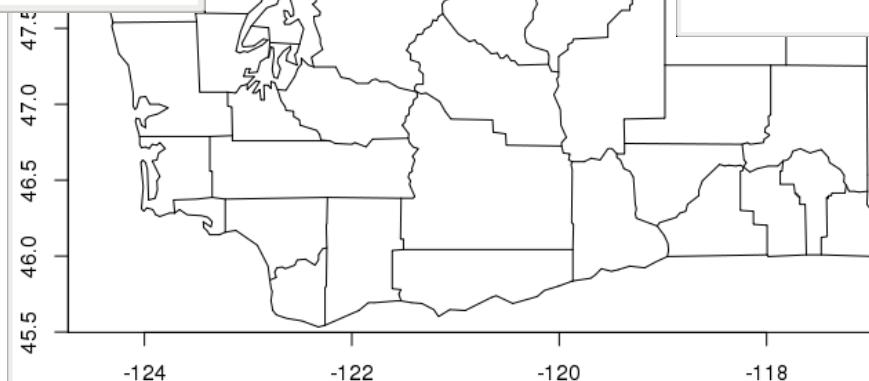
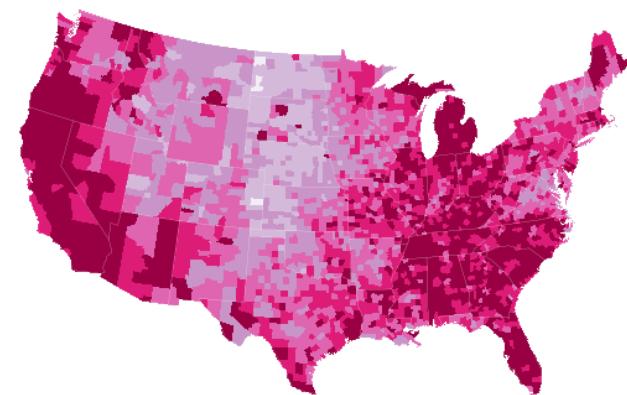
R Graphics: Device 2 (ACTIVE)

map (database, regions, . . .)

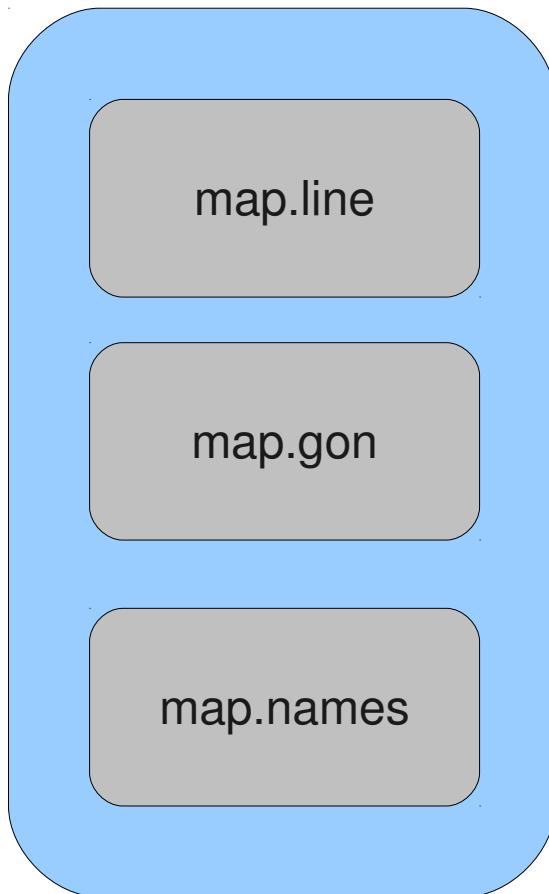


R Graphics: Device 2 (ACTIVE)

Washington State

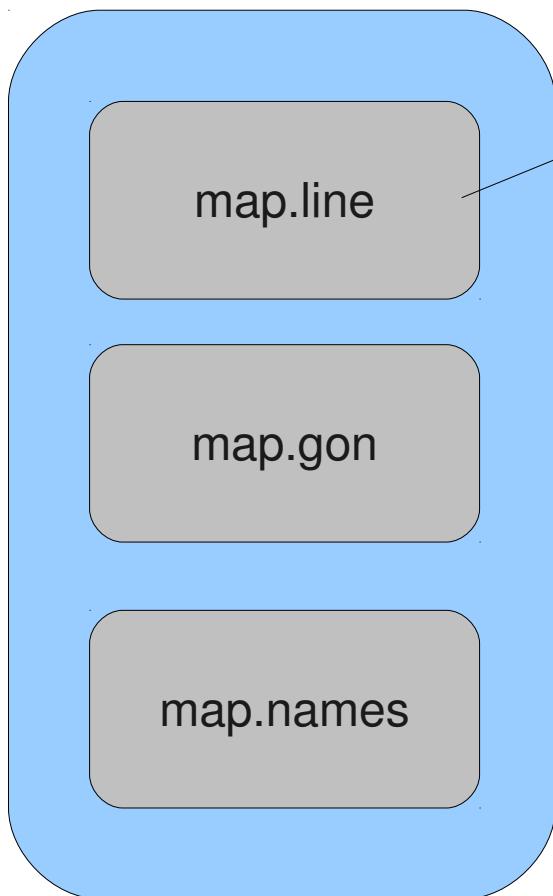


map database format



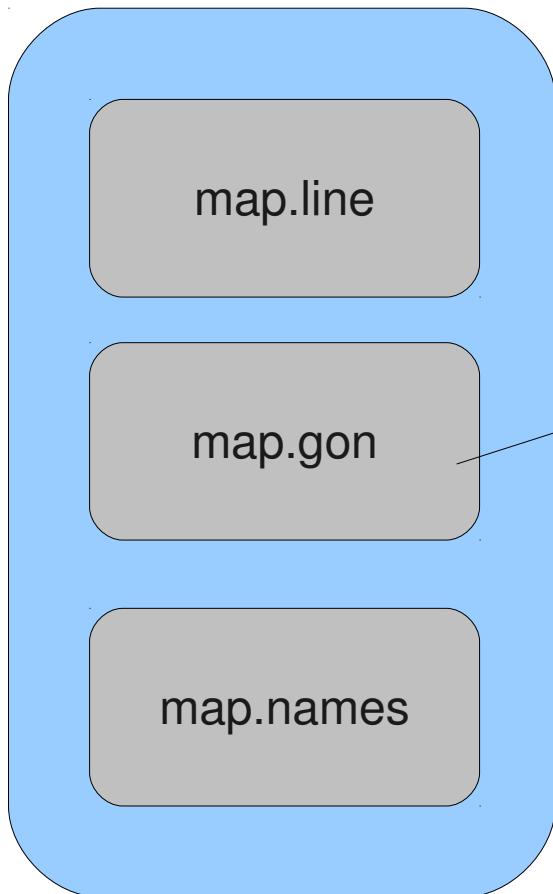
- To use your data with the maps package, first create these three files...

map database format



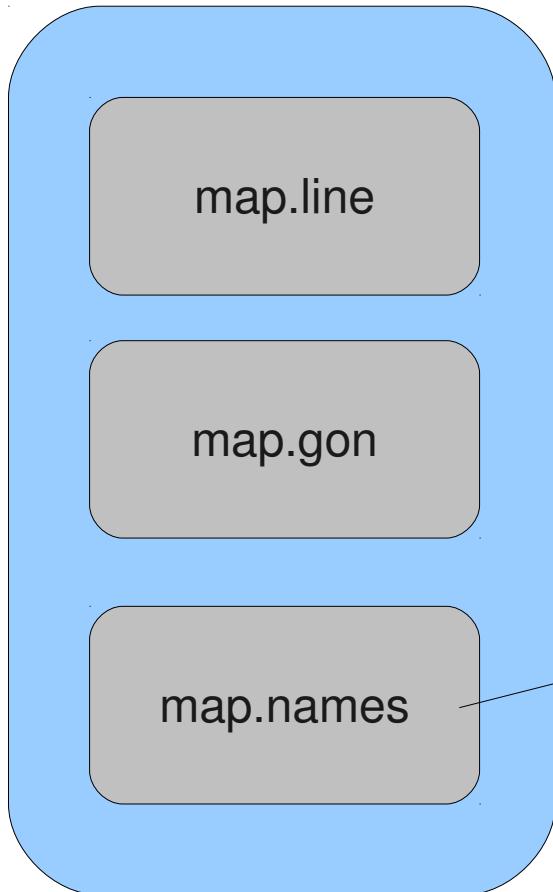
2 1
 4.07789 5.58567 4.04461 5.52577 3.91483
 5.5291 3.86491 5.47585 3.79169 5.42926
 3.78171 5.33941 3.80168 5.2828
 3 3.86491 5.17634 3.88487 5.09314 3.99137
 5.07317 4.13447 5.14306 4.17773 5.16303
 4.28755 5.13973 4.35078 5.0
 765 4.36077 5.12642
 EOR
 3 1
 4.36077 5.12642 4.37075 5.12642 4.47392
 5.11643 4.5438 5.14306 4.71686 5.11643
 4.77343 5.15304 4.77676 5.1996
 3 4.69356 5.23624 4.61369 5.23624 4.57043
 5.26619 4.51052 5.22626 4.51385 5.28616
 4.49388 5.3627 4.46061 5.38
 932 4.46726 5.38932
 EOR

map data format



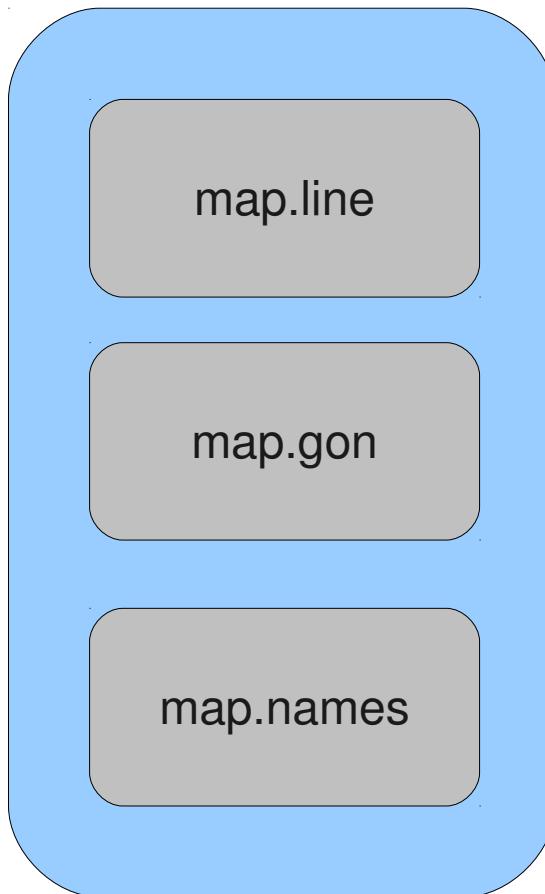
```
1 2 3 4
EOR
1 5 3 -6
EOR
2 -5
EOR
```

map data format

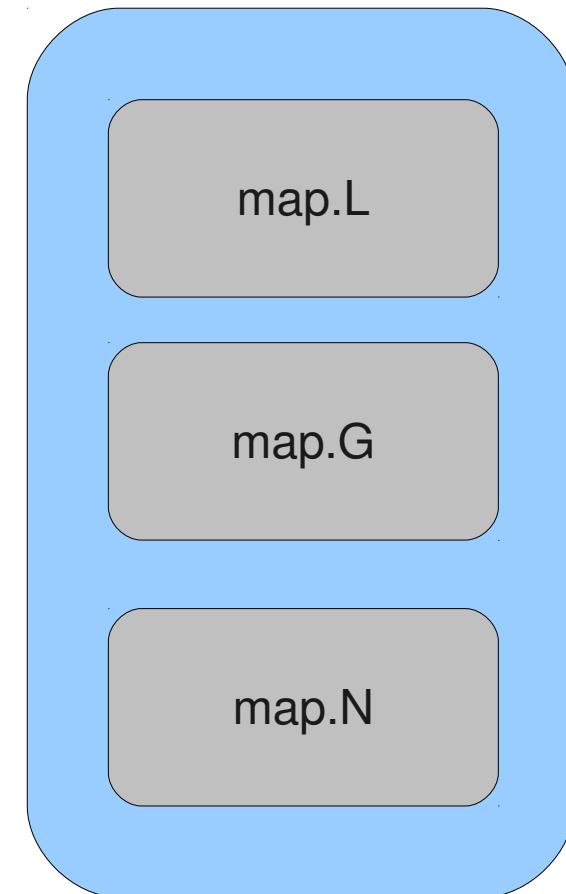


Scotland 1
Scotland 2
Scotland:Harris 3
Scotland:Harris 4
Scotland:Harris 5
Scotland:Harris 6

map data format



- Then process it with **mapgetl** and **mapgetg**



`maps` package now maintained by Ray Brownrigg

New Millennium

- GIS spreading
- Many more map file formats (shapefile, 1998)
- Maps on the web (UMN Mapserver, mid-90s)
- Spatial Data Standards:



R didn't get left behind!

- Perceived need for better maps
- Better spatial data handling
- Assorted little projects (my **Rmap!**)

Vector data - the sp classes

- 2004 : A Geospatial Odyssey
- More than just polygons!
 - Classes for points, lines, grids of points and polygons.
- Geometry + Data for each feature
- Bivand, Pebesma, Gomez-Rubio
- “Simple Features” OGC spec

Create Points

```
coords = cbind(x, y)
```

```
# create SpatialPoints  
sp = SpatialPoints(coords)
```

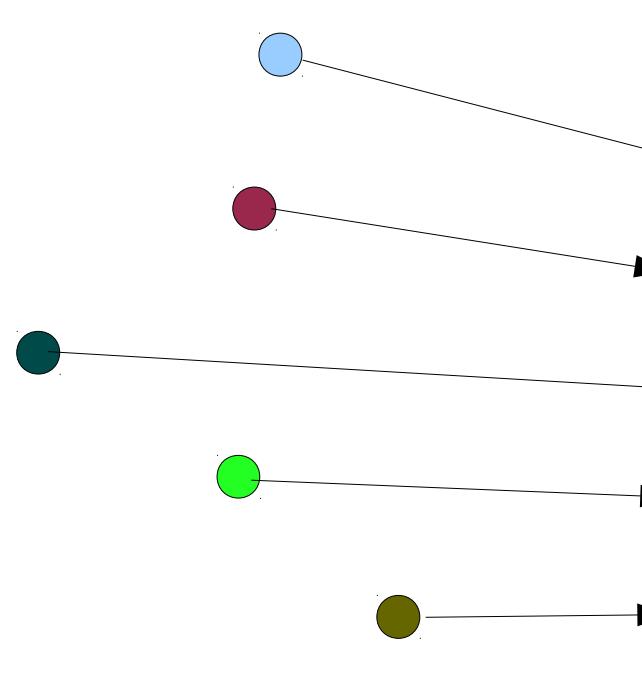
```
# create a SpatialPointsDataFrame  
spdf = SpatialPointsDataFrame(coords, data)  
spdf = SpatialPointsDataFrame(sp, data)
```

```
# promote an ordinary data frame  
coordinates(data) = cbind(x, y)  
coordinates(data) = ~lon + lat
```

```
# and demote back to data frame  
data = as.data.frame(data)  
# or  
data = data@data
```

Point Features

`SpatialPointsDataFrame`



ID	coordinate	Name	Pop
1	(x,y)	Alphaville	12000
2	(x,y)	Betaton	45255
3	(x,y)	Gamma Clty	5345
4	(x,y)	Delta VIIlage	7745
5	(x,y)	Epsilondon	77514

Manipulation

```
BigTowns = Towns[Towns$pop>10000, ]
SmallTowns = Towns[Towns$pop<5000, ]
```

Treat like a
data frame...

```
BigTowns = subset(Towns, pop>10000)
BigTowns = subset(Towns, Towns$pop>10000)
```

...almost (this fails)

this works

```
BigSmall = rbind(BigTowns, SmallTowns)
```

Retrieve coords and data

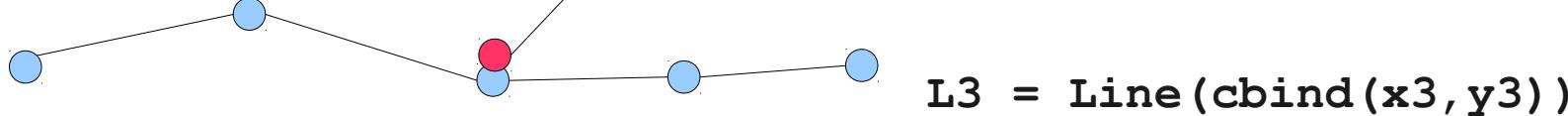
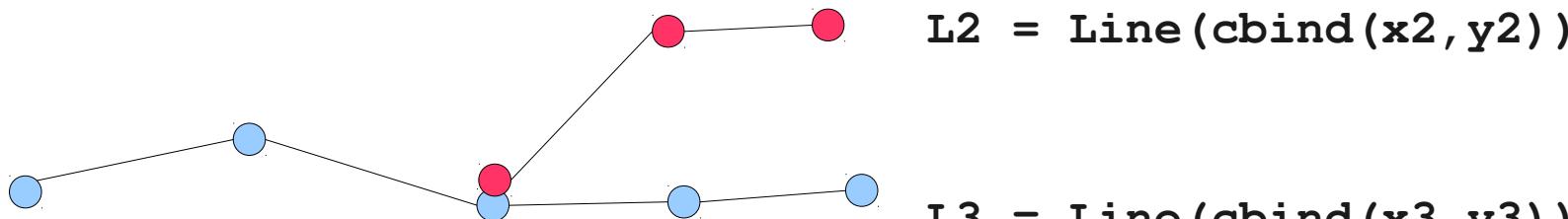
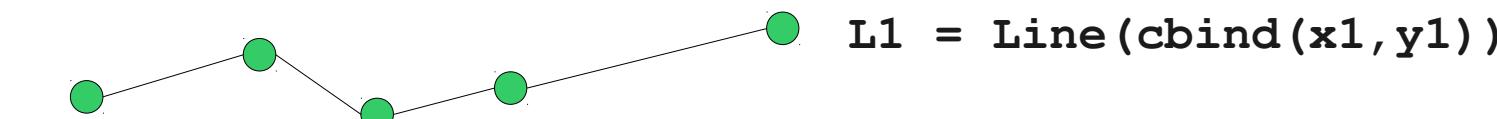
```
xy = coordinates(Towns)
```

```
TownData = Towns@data
```

```
TownData = as.data.frame(Towns)
```

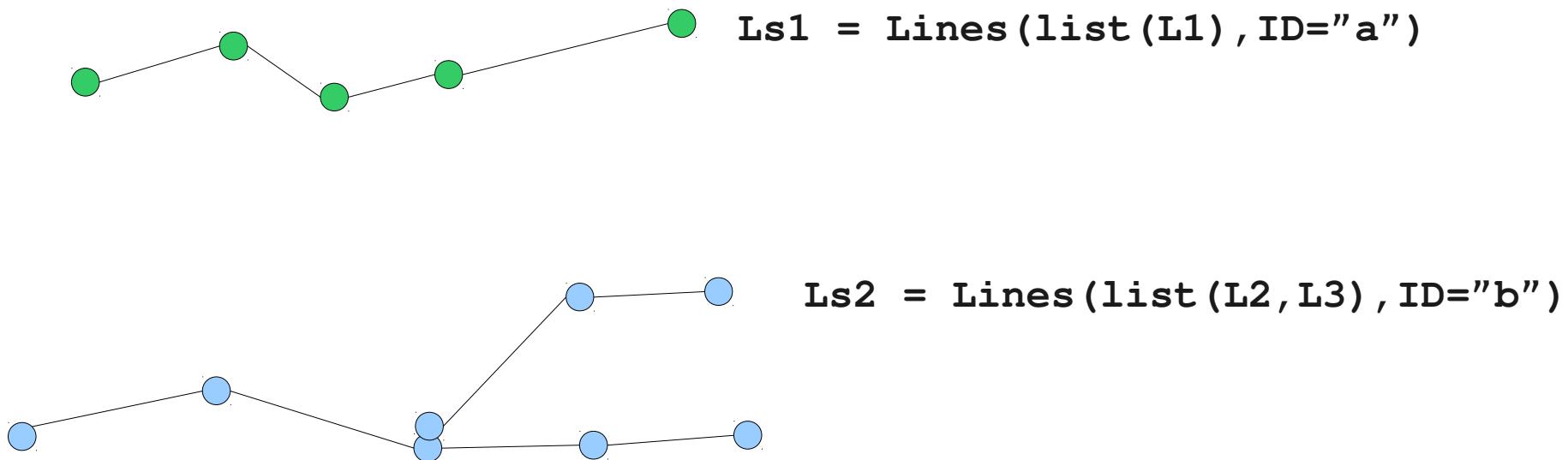
Create Line Objects

A **Line** is a single chain of points



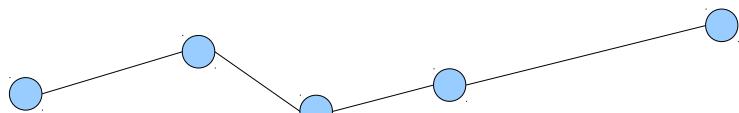
Create Lines Objects

A **Lines** is a list of chains with an ID

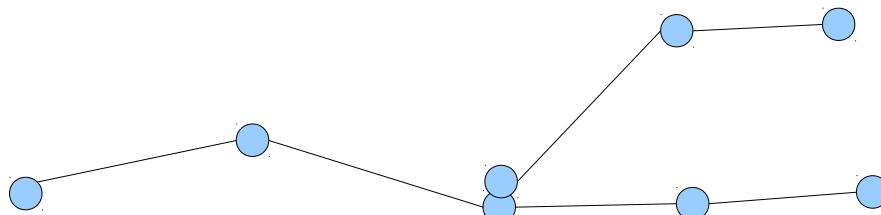


Create SpatialLines Object

A **SpatialLines** is a list of **Lines**

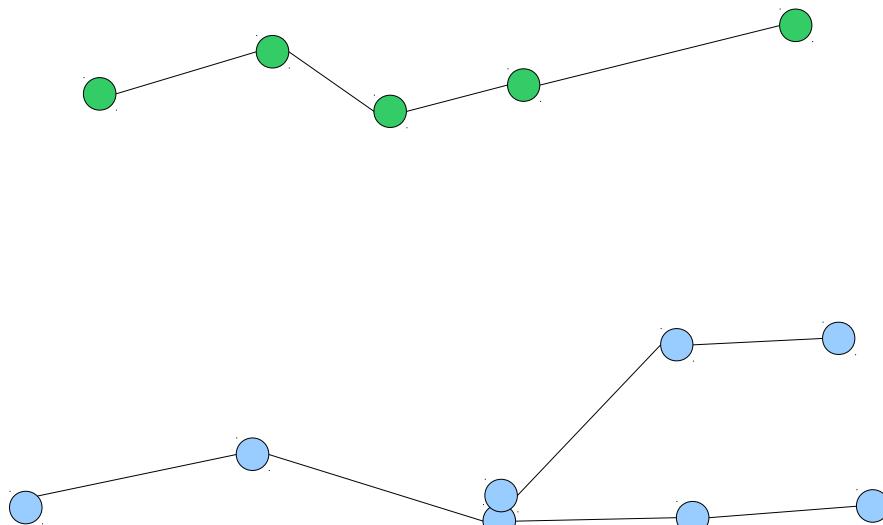


```
SL12 = SpatialLines(list(Ls1,Ls2))
```



Create SpatialLinesDataFrame

A `SpatialLinesDataFrame` is a `SpatialLines` with a matching Data Frame

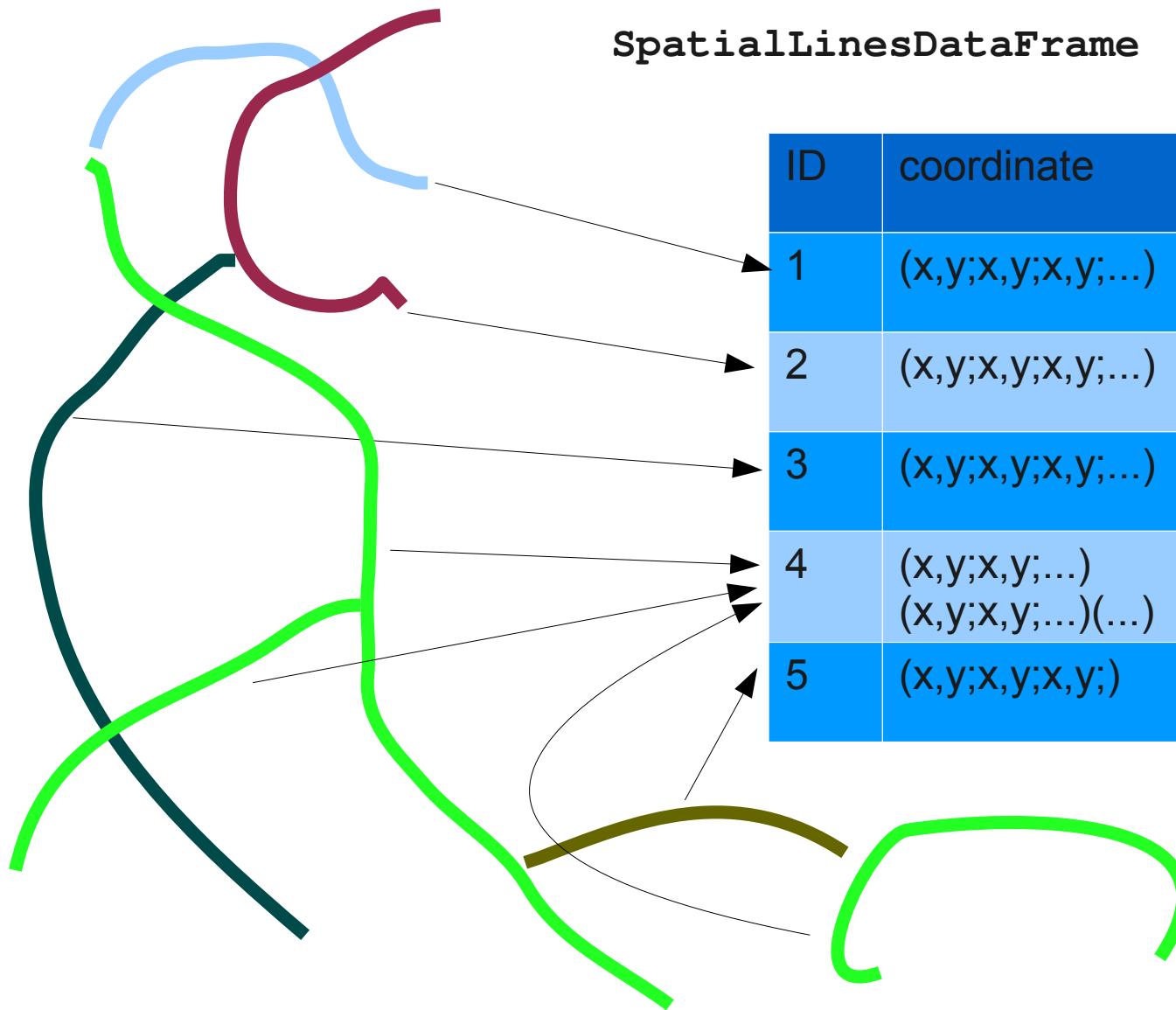


```
SLDF = SpatialLinesDataFrame(  
  SL12,  
  data.frame(  
    Z=c("road", "river"),  
    row.names=c("a", "b"))  
)
```

Matching on the `row.names`
to the `ID` values or optionally
by order

Line Features

SpatialLinesDataFrame



ID	coordinate	Name	Quality
1	(x,y;x,y;x,y;...)	River Alpha	Good
2	(x,y;x,y;x,y;...)	Bravo Beck	Good
3	(x,y;x,y;x,y;...)	Charlie River	Fair
4	(x,y;x,y;...) (x,y;x,y;...)(...)	Delta Drains	Poor
5	(x,y;x,y;x,y;)	Echo Stream	Good

Polygons

Now it starts to get complicated



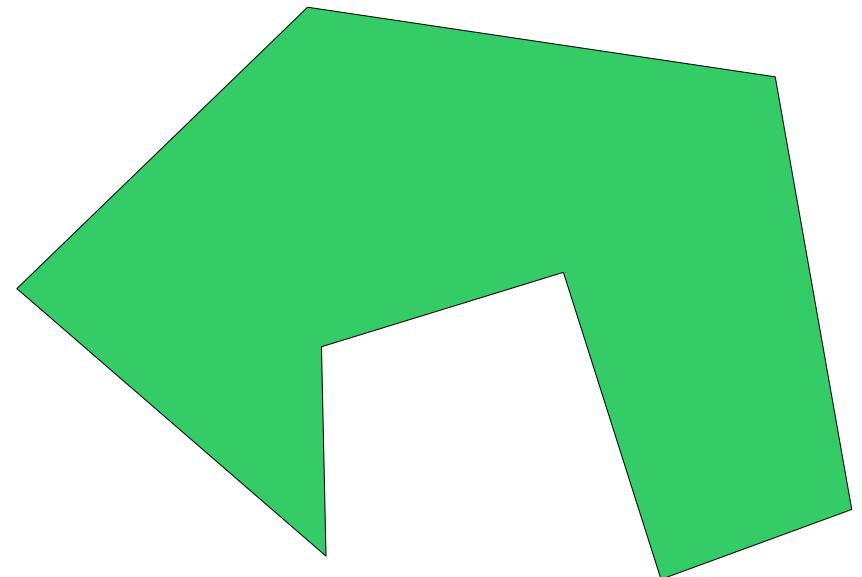
Single Ring Feature

Make sure your coordinates connect:

```
c1 = cbind(x1,y1)
r1 = rbind(c1,c1[1,])
```

A **Polygon** is a single ring

```
P1 = Polygon(r1)
```



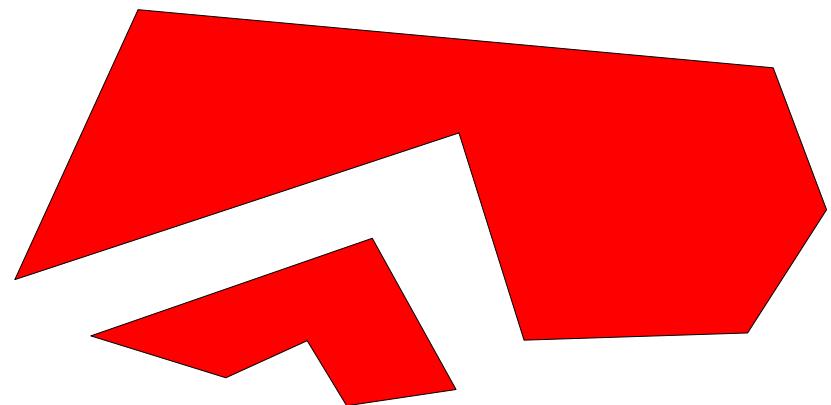
A **Polygons** is a list of **Polygon** objects with an **ID**

```
Ps1 = Polygons(list(P1, ID="a"))
```

Double Ring Feature

Make sure your coordinates connect:

```
c2a = cbind(x2a, y2a)
r2a = rbind(c2a, c2a[1, ])
c2b = cbind(x2b, y2b)
r2b = rbind(c2b, c2b[1, ])
```



Make two single rings:

```
P2a = Polygon(r2a)
P2b = Polygon(r2b)
```

Make a Polygons object with two rings

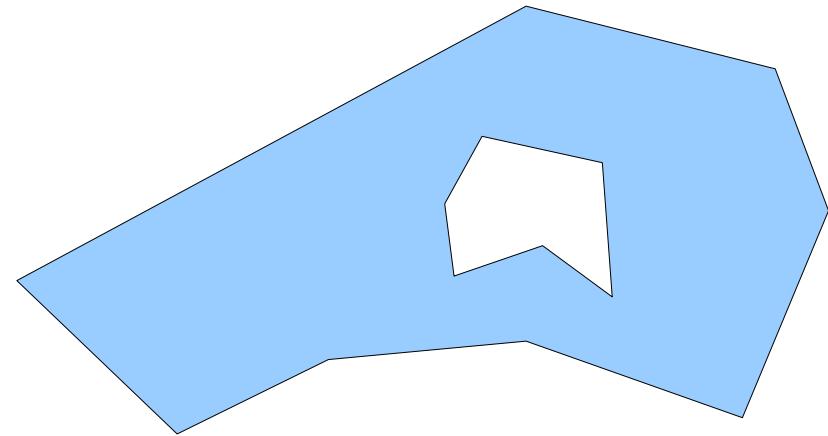
```
Ps2 = Polygons(list(P2a, P2b), ID="b") )
```

Feature with hole

```
hc = cbind(xh, yh)  
rh = rbind(hc, hc[1, ])
```

Make a holey **Polygon**:

```
H1 = Polygon(rh, hole=TRUE)
```

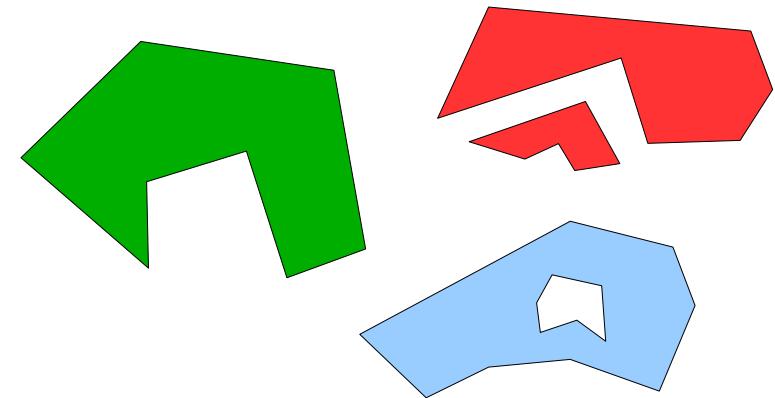


Make a **Polygons** object with two rings, one the hole

```
Ps3 = Polygons(list(P3, H1), ID="c"))
```

SpatialPolygonsDataFrame

```
SPs = SpatialPolygons(
  list(Ps1,Ps2,Ps3)
)
```



```
SPDF = SpatialPolygonsDataFrame(
  SPs,
  data.frame(
    info=c("single","double","hole"),
    row.names=c("a","b","c")
  )
)
```

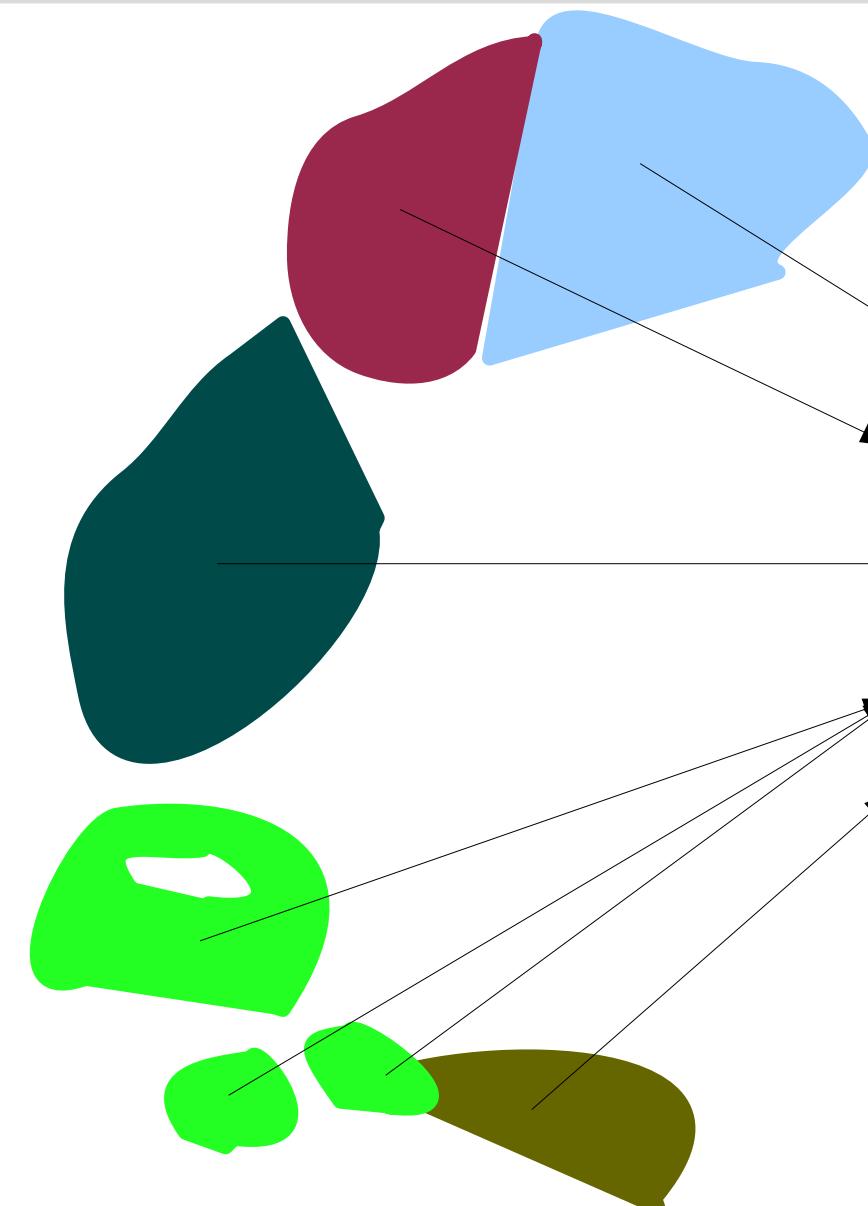
And back again

For ring j of feature i :

```
> SPDF@polygons[[i]]@Polygons[[j]]@coords
```

	x	y
[1,]	42.53623	76.85066
[2,]	42.57308	76.79820
[3,]	43.05132	74.56176
...		

Polygon Features

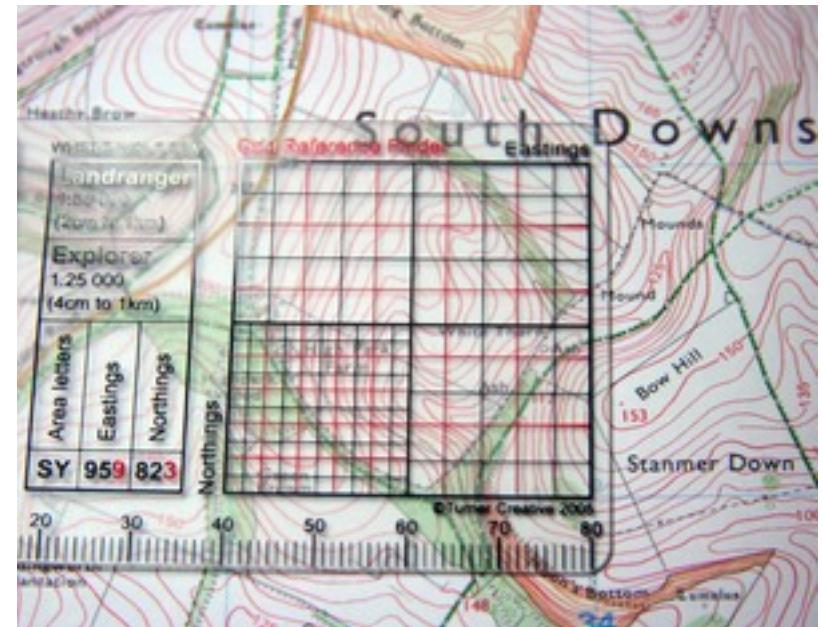


SpatialPolygonsDataFrame

ID	coordinates	Name	Pop.
1	(x,y;x,y;x,y;...)	Alphashire	12365
2	(x,y;x,y;x,y;...)	Bravoshire	45366
3	(x,y;x,y;x,y;...)	County Charlie	78725
4	(x,y;x,y;...) (x,y;x,y;...)(...)	Delta District	77439
5	(x,y;x,y;x,y;)	Echoville	34545

Coordinate Systems

- Because the Earth isn't flat
- But small parts of it are near enough
- So lots of coordinate systems exist
- Anything **Spatial*** can have one



CRS and proj4string

without:

```
sp = SpatialPoints(coords)
```

at construction time:

```
sp = SpatialPoints(  
    coords,  
    proj4string = CRS("+init=epsg:4326")  
)
```

After the event:

```
proj4string(sp) = CRS("+init=epsg:4326")
```

And for Spatial*DataFrames

```
spdf = SpatialPointsDataFrame(  
  sp,  
  data,  
  proj4string=CRS ("+init=epsg: 4326")  
)
```

```
proj4string(spdf)=CRS ("+init=epsg: 4326")
```

What are these proj4strings?

- International standard codes for coordinate reference systems
- **+init=epsg:4326** lat-long, GPS
- **+init=epsg:27700** UK Grid
- **+init=epsg:900913** “google”
- **+init=epsg:3857** Spherical Mercator
- **+proj=utm +zone=19 +north**

Search spatialreference.org for UTM zones

Transform Spatial* Objects

- Assigning a CRS doesn't change the numbers
- So assigning the wrong CRS really messes things up
- Convert from one to another with **spTransform**

If **sp** is **SpatialPoints** from my GPS:

```
proj4string(sp)=CRS ("+init=epsg:4326")
```

And to put on a UK National Grid map:

```
spUK = spTransform(sp, CRS ("+init=epsg:27700"))
```

- The coordinates are now in metres!

Raster Data – regular grids

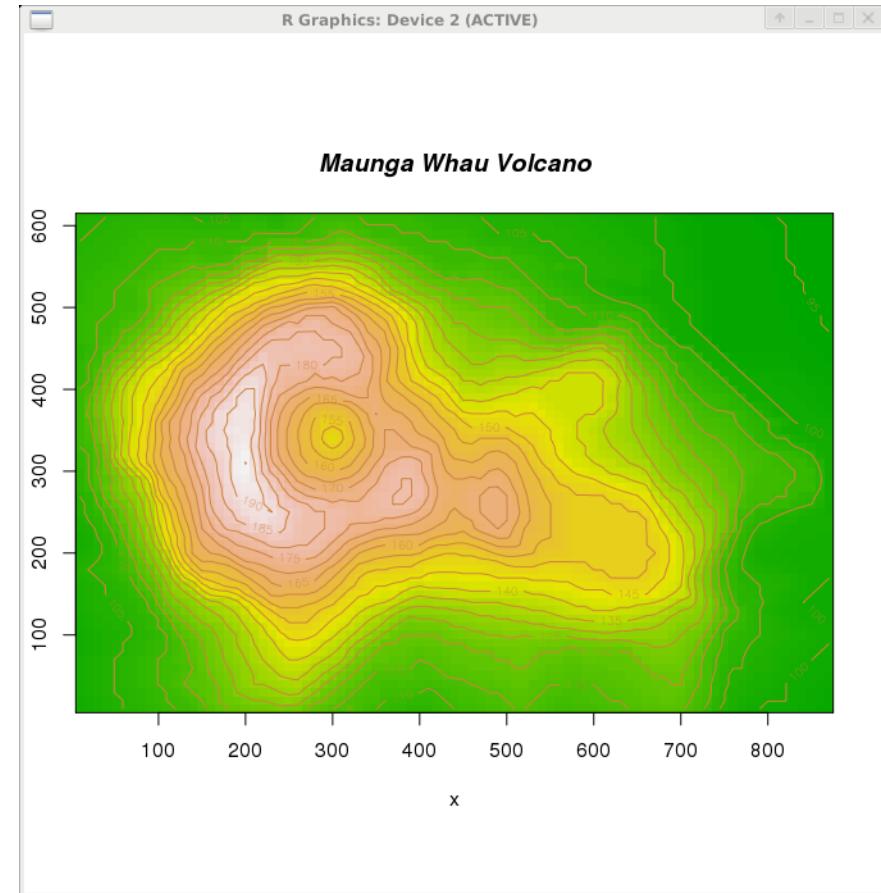
- Satellite Imagery
 - Photography
 - Radar Height
 - Other Sensors
- Count Data
 - Animals in quadrats
- Modelled Values
 - Climate Data
 - Interpolations



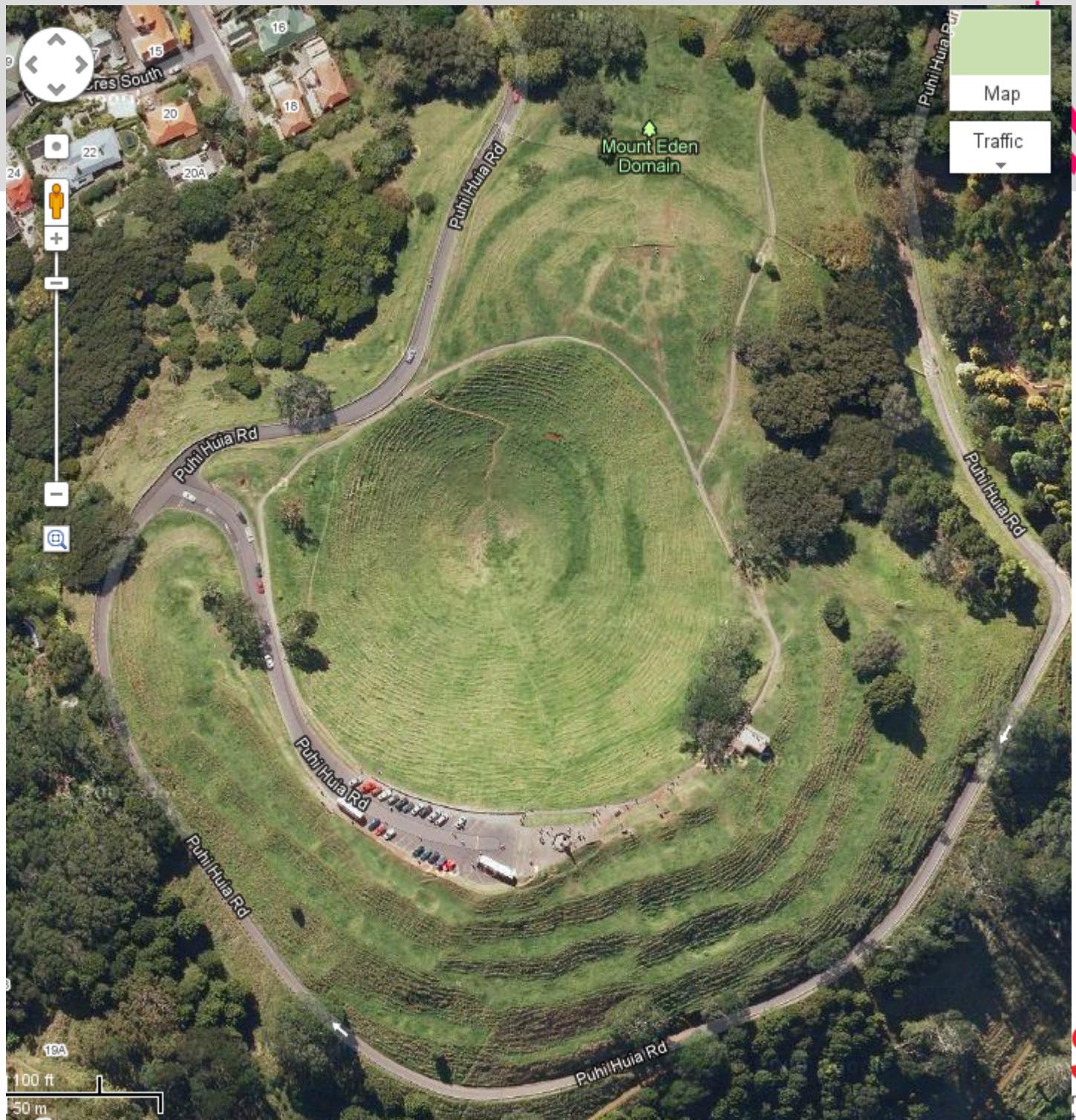


Rasters in R - history

- The **image** function
- Works on a matrix
- X and Y vectors define position



Volcano?





Volcano



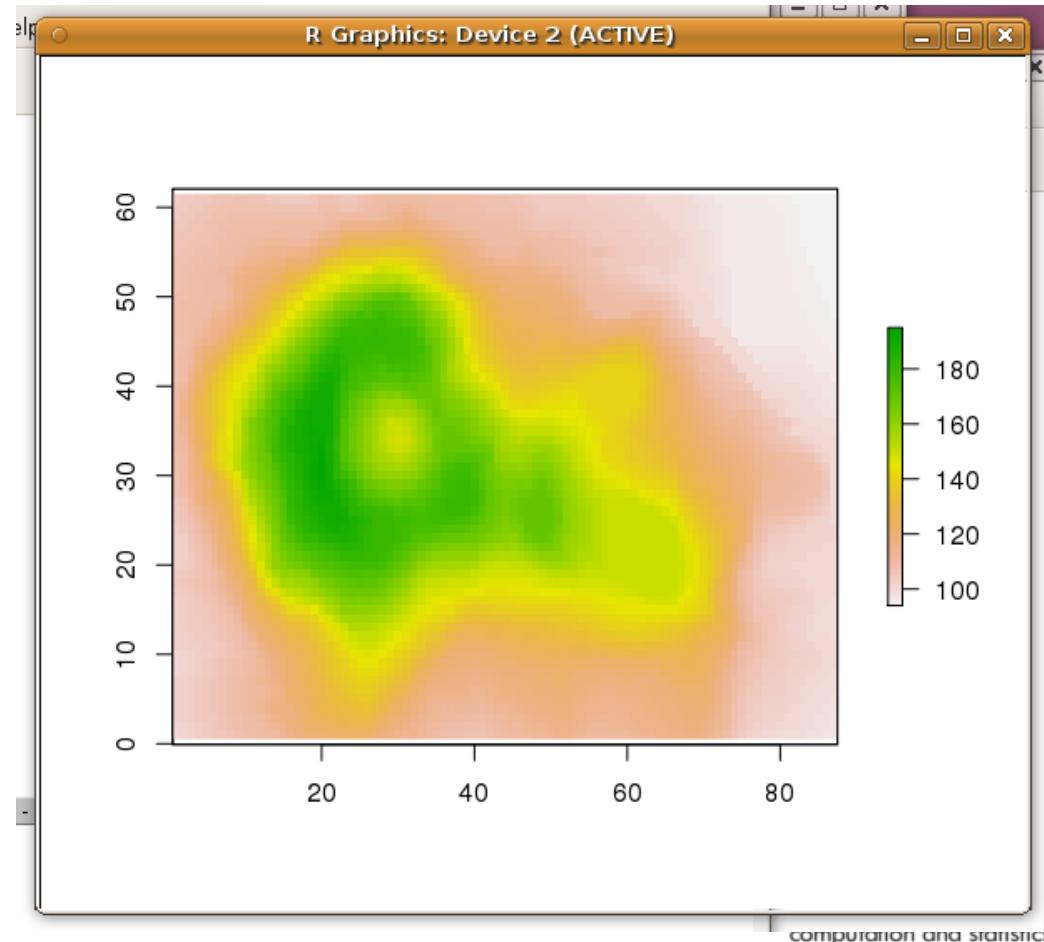


The raster package

- Create from vector x, vector y, matrix z

```
r = raster(list(x=x, y=y, z=z))
```

```
plot(r)
```

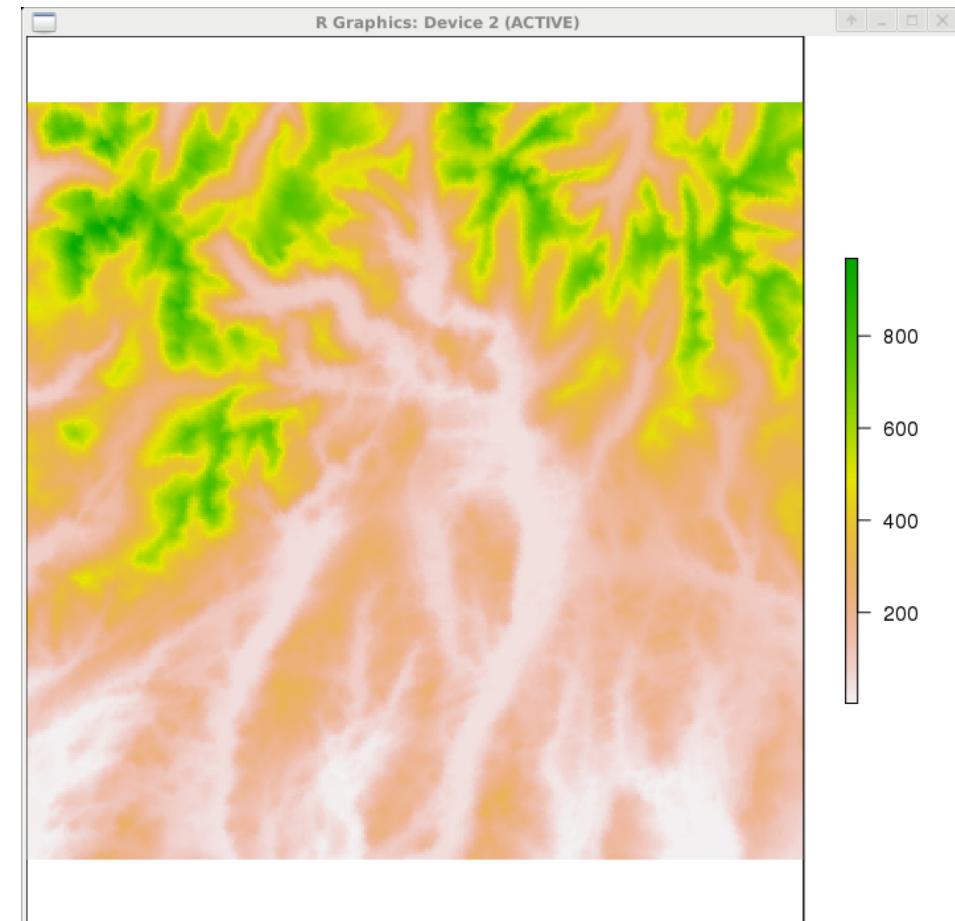


The raster package

- Read and plot standard raster files

```
dem = raster('cumbriaDem.tif')
```

```
plot(dem)
```

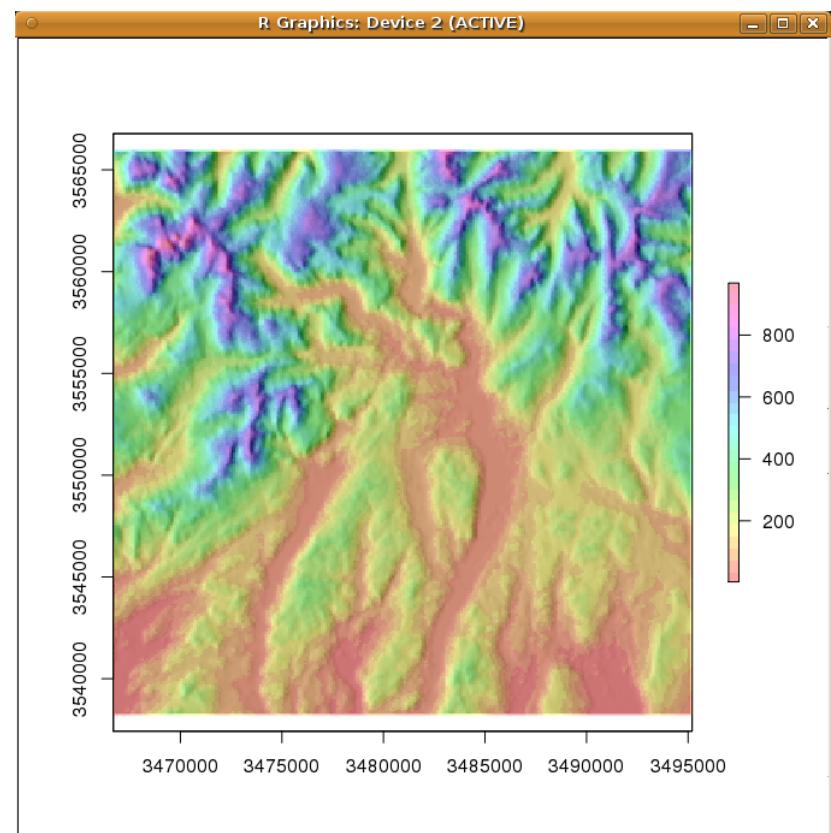




Raster ops

```
highGround = dem > 600
slope = terrain(dem, opt='slope')
aspect = terrain(dem, opt='aspect')
hill = hillShade(slope, aspect, 40, 270)
plot(hill, col=grey(0:100/100)
     , legend=FALSE)

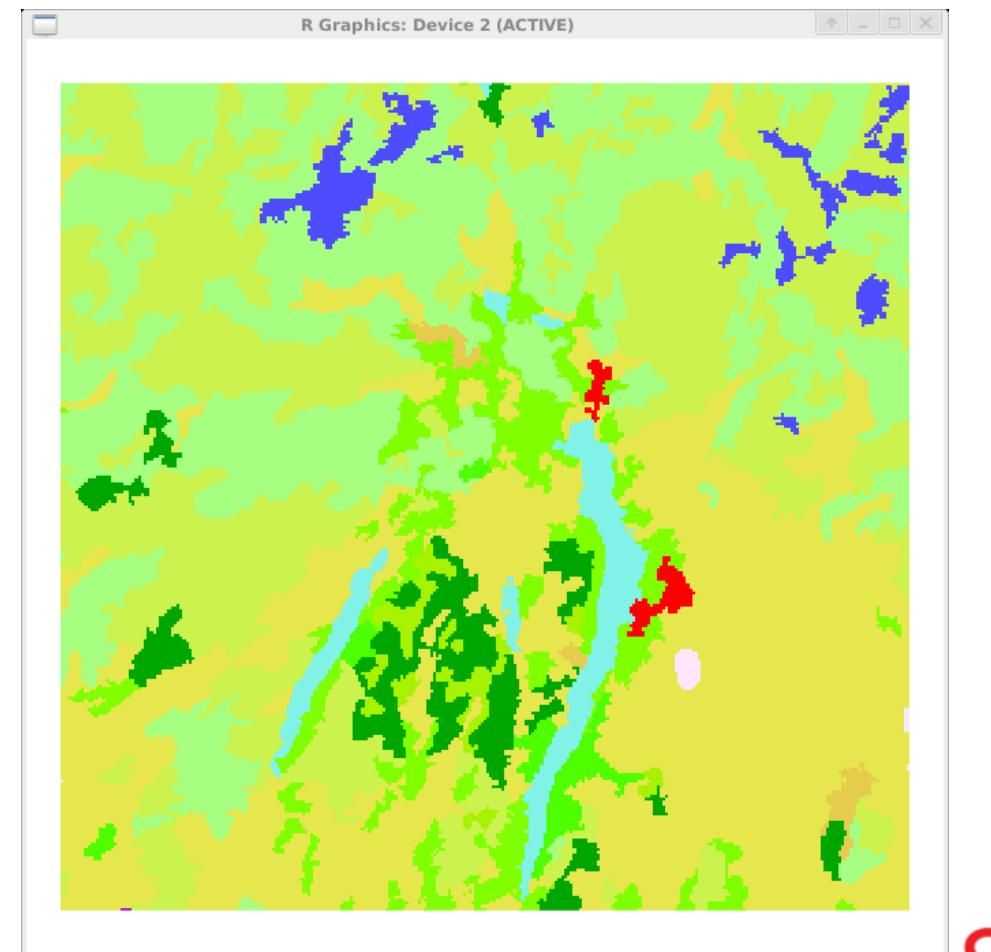
plot(dem,
      col=rainbow(25, alpha=0.35) ,
      add=TRUE)
```



Categorical rasters

- **use = raster('cumbria.tif'); plot(use)**

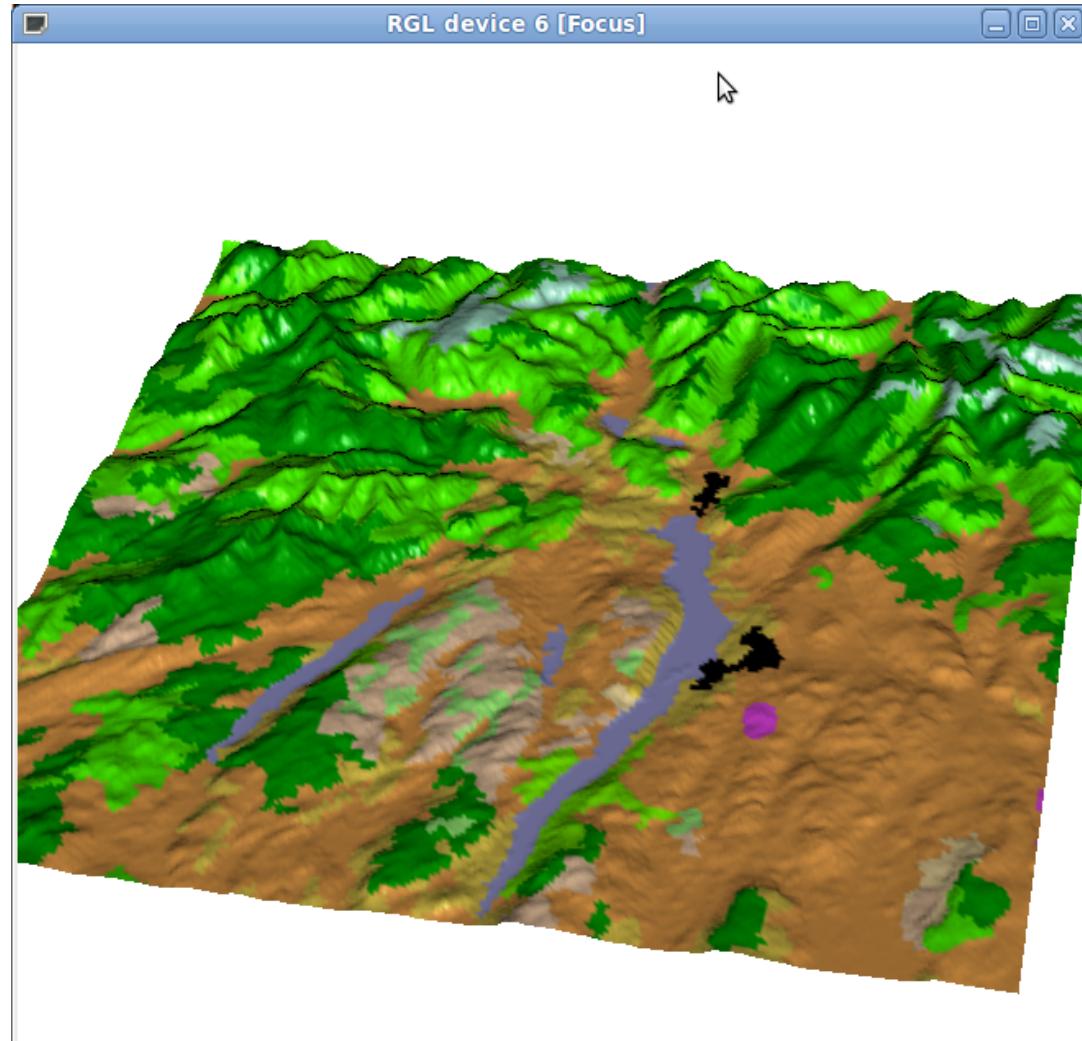
```
> as.matrix(table(values(use)))
   [,1]
2     325
7      4
11    121
18  24399
21    425
23   5087
24   3084
[etc]
```



3d visuals – rasterVis

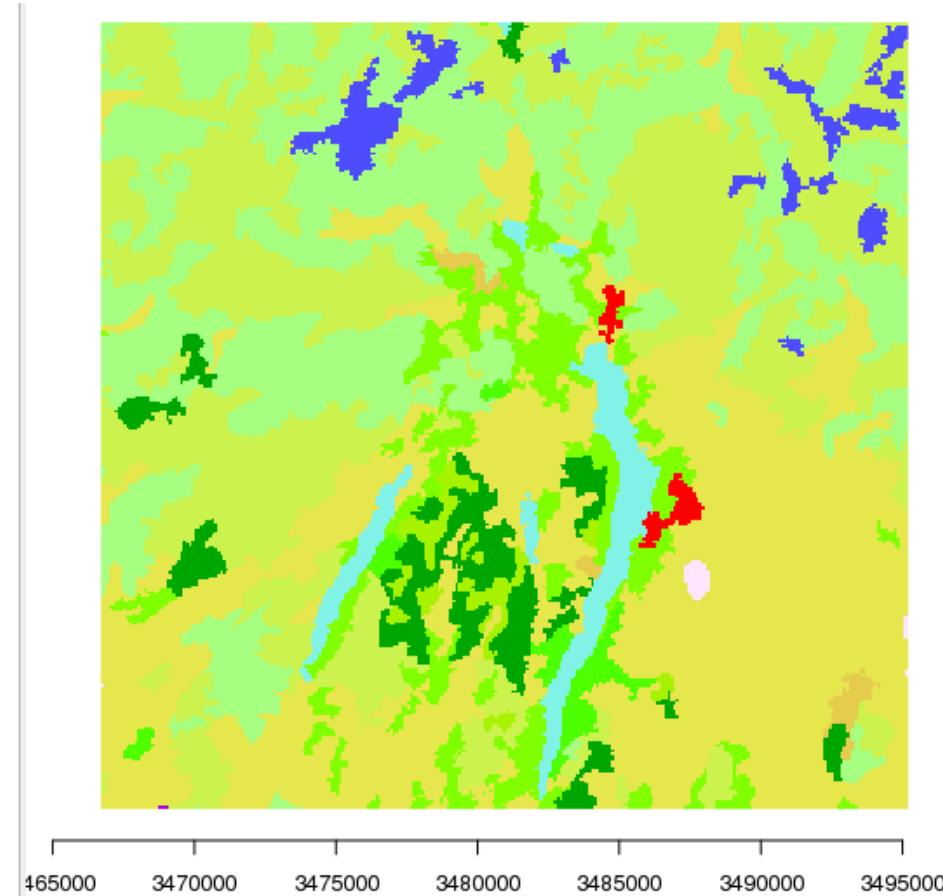
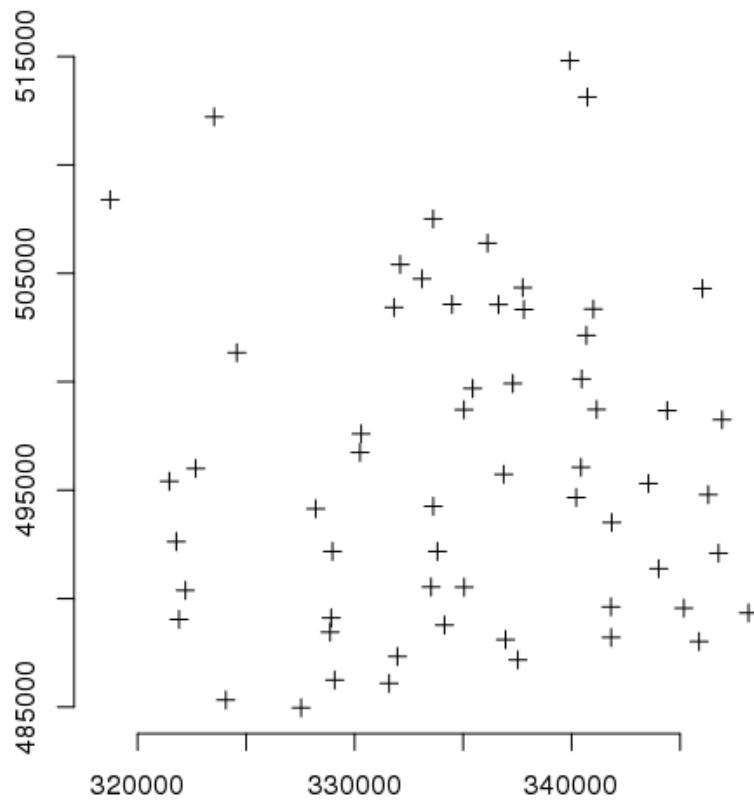
- Drape land-use over dem

```
plot3D(dem,  
       drape=use)
```



Projections and CRS

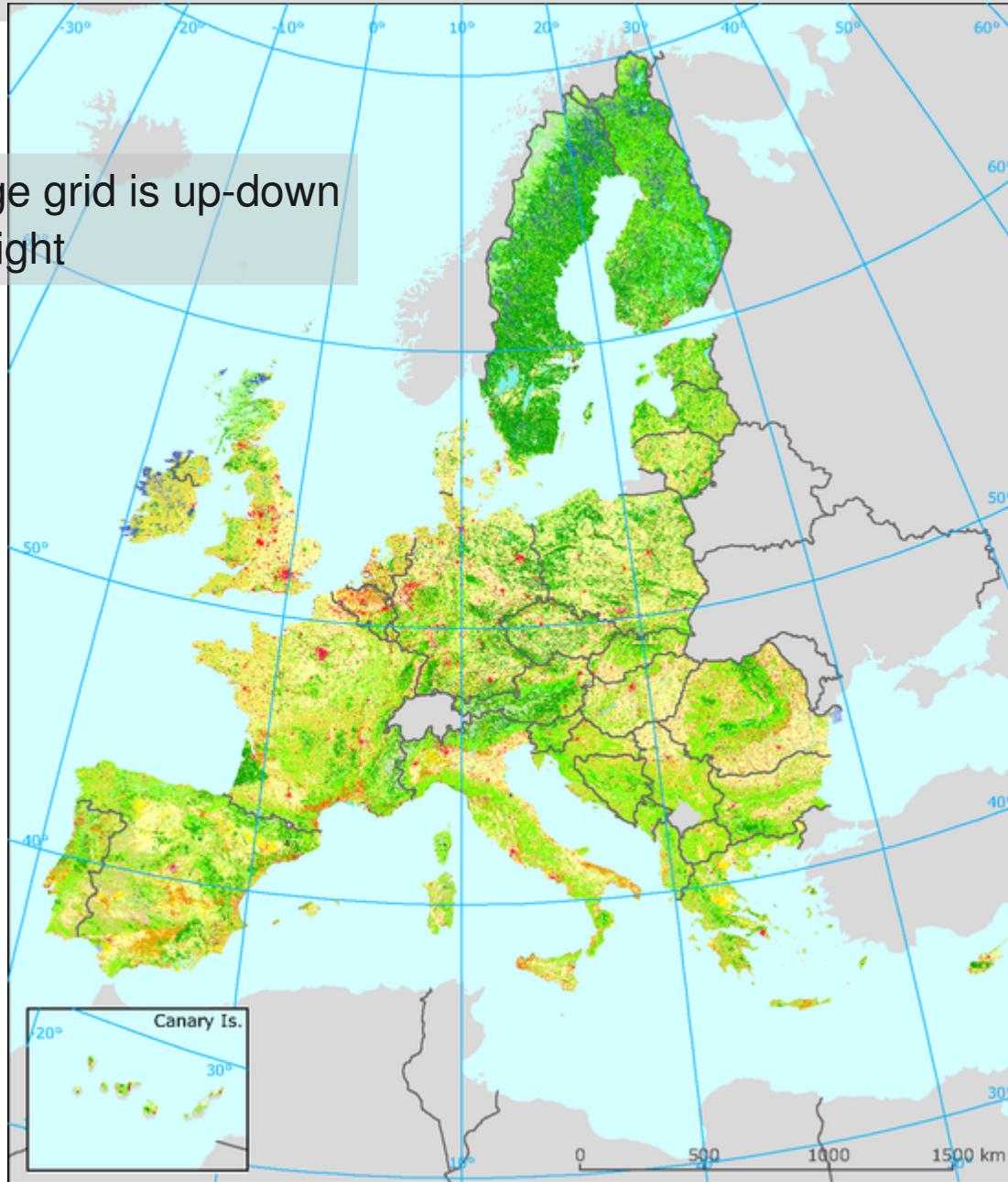
```
> use = raster('/data/CorineLandCover/cumbria.tif')
> plot(settlements)
> plot(use)
```





Projection

Image grid is up-down
left right



Corine land cover classes	
1. Artificial surfaces	3. Forest and semi-natural areas
1.1 Urban fabric	3.1 Forests
1.1.1 Continuous urban fabric	3.1.1 Broad-leaved forest
1.1.2 Discontinuous urban fabric	3.1.2 Coniferous forest
1.2 Industrial, commercial and transport areas	3.1.3 Mixed forest
1.2.1 Industrial areas	3.1.4 Sub-shrub and herbaceous vegetation associations
1.2.2 Road network networks and associated areas	3.1.5 Tree pruned
1.2.3 Water bodies	3.1.6 Woods and woodland
1.3 Mine, quarry and waste extraction areas	3.1.7 Invasive vegetation
1.3.1 Waste extraction sites	3.1.8 Thickets, scrubland areas
1.3.2 Deep pits	3.3 Open spaces with little or no vegetation
1.3.3 Construction sites	3.3.1 Dunes, dunes, and sand plains
1.4 Artificial, non-agricultural vegetated areas	3.3.2 Marshes
1.4.1 Green urban areas	3.3.3 Shrubry vegetation areas
1.4.2 Sport and leisure facilities	3.3.4 Barren areas
2. Agricultural areas	3.3.5 Glaciated periglacial areas
2.1 Arable land	4.1 Inland wetlands
2.1.1 Non-irrigated arable land	4.1.1 Inland waters
2.1.2 Irrigated arable land	4.1.2 Marshes
2.1.3 Rice fields	4.1.3 Inland lakes
2.2 Permanent crops	4.2 Coastal wetlands
2.2.1 Vineyards	4.2.1 Intertidal zones
2.2.2 Not trees monocultures	4.2.2 Tidal flats
2.2.3 Other crops	4.2.3 Tidal marshes
2.3 Pastures	5.1 Inland waters
2.3.1 Pastures	5.1.1 Water courses
2.4 Heterogeneous agricultural areas	5.1.2 Water bodies
2.4.1 Areas crop associated with permanent crops	5.2 Marine waters
2.4.2 Complex agricultural patterns	5.2.1 Coastal regions
2.4.3 Small-scale highly diversified agriculture	5.2.2 Islands
2.4.4 Agroforestry areas	5.2.3 Offshore areas

Projections

- rasters have a CRS

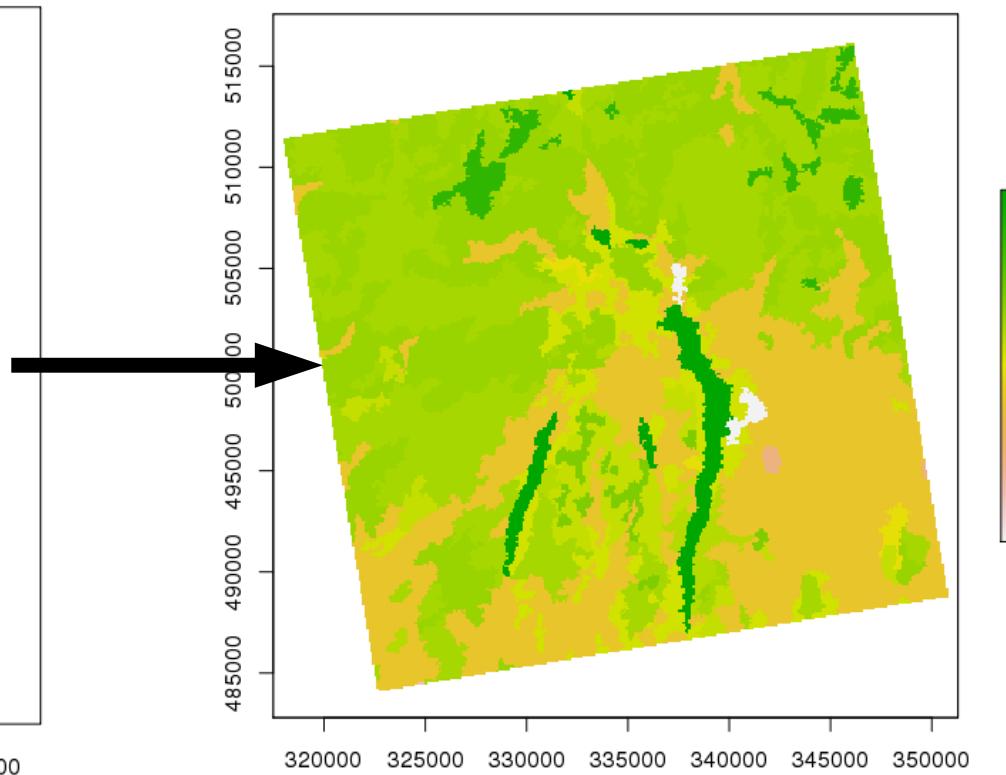
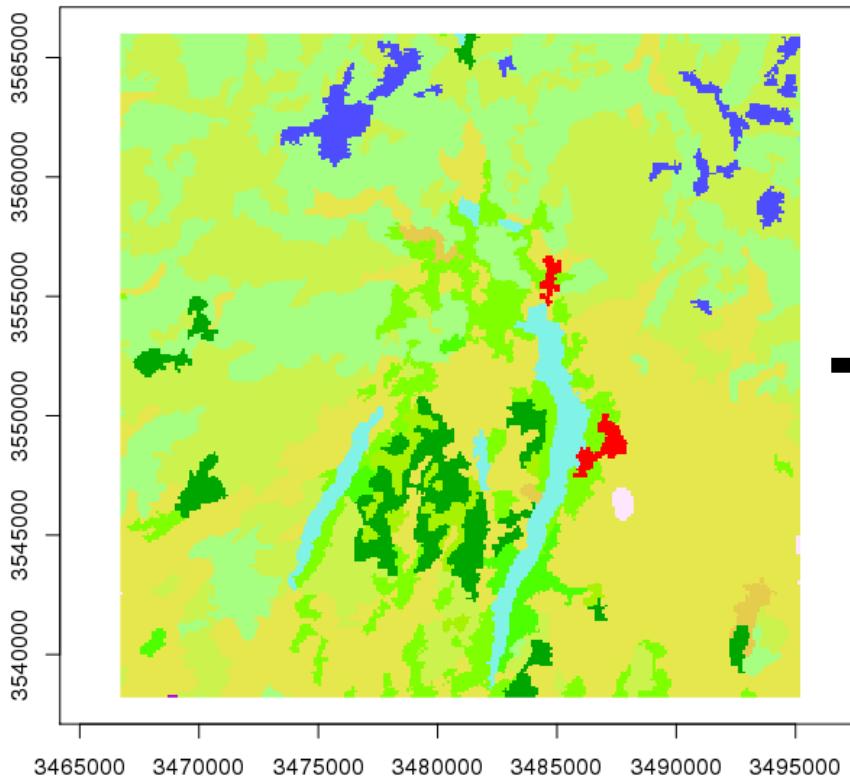
```
> use = raster('/data/CorineLandCover/cumbria.tif')
> use
class       : RasterLayer
dimensions   : 278, 285, 79230 (nrow, ncol, ncell)
resolution   : 100, 100 (x, y)
extent       : 3466700, 3495200, 3538200, 3566000
coord. ref. : +proj=laea +lat_0=52 +lon_0=10
                  +x_0=4321000 +y_0=3210000 +ellps=GRS80
                  +units=m +no_defs
values       : /data/CorineLandCover/cumbria.tif
min value    : 2
max value    : 41
> projection(use)
[1] "+proj=laea +lat_0=52 +lon_0=10 +x_0=4321000
+y_0=3210000 +ellps=GRS80 +units=m +no_defs"
```



Reproject Raster

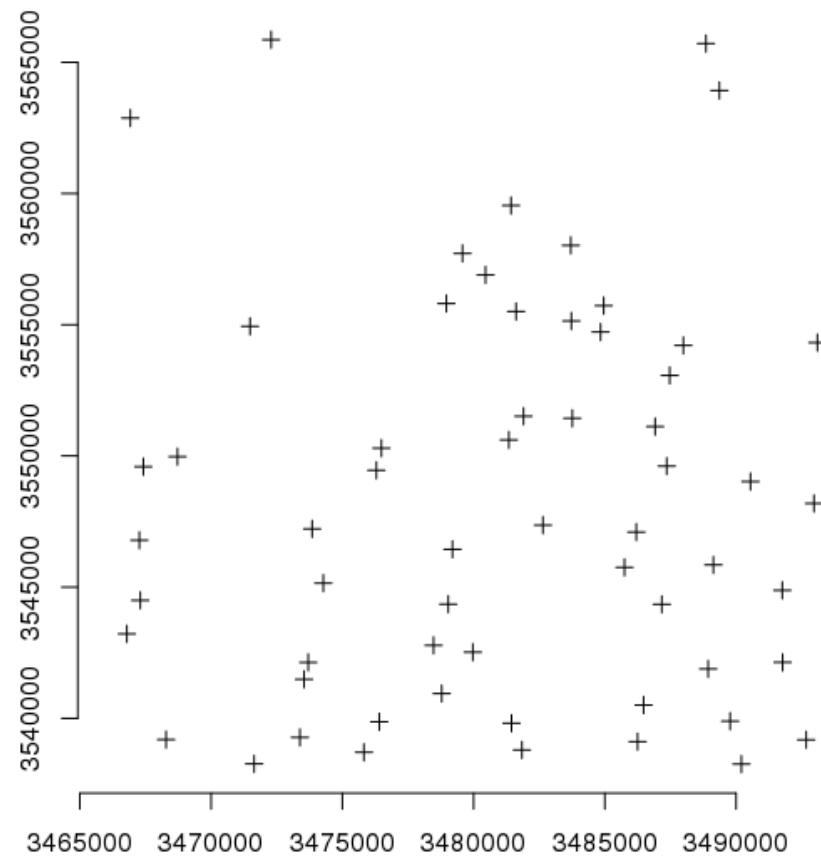
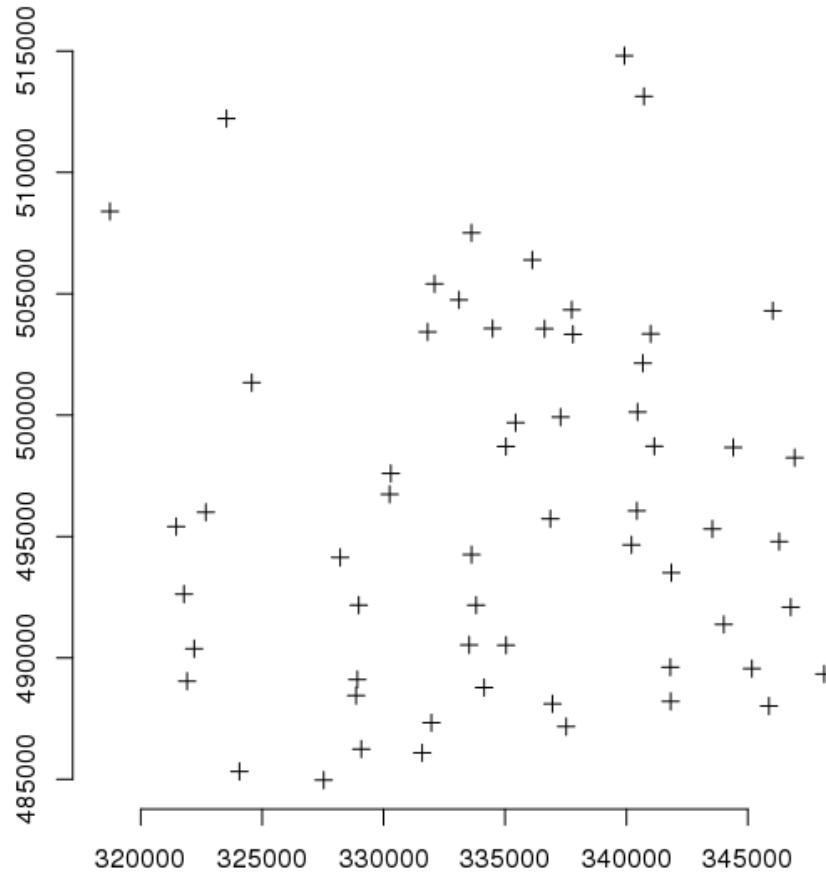
- Involves a 'warp'

- `useOS = projectRaster(use, res=100,
crs="+init=epsg:27700", method="ngb")`

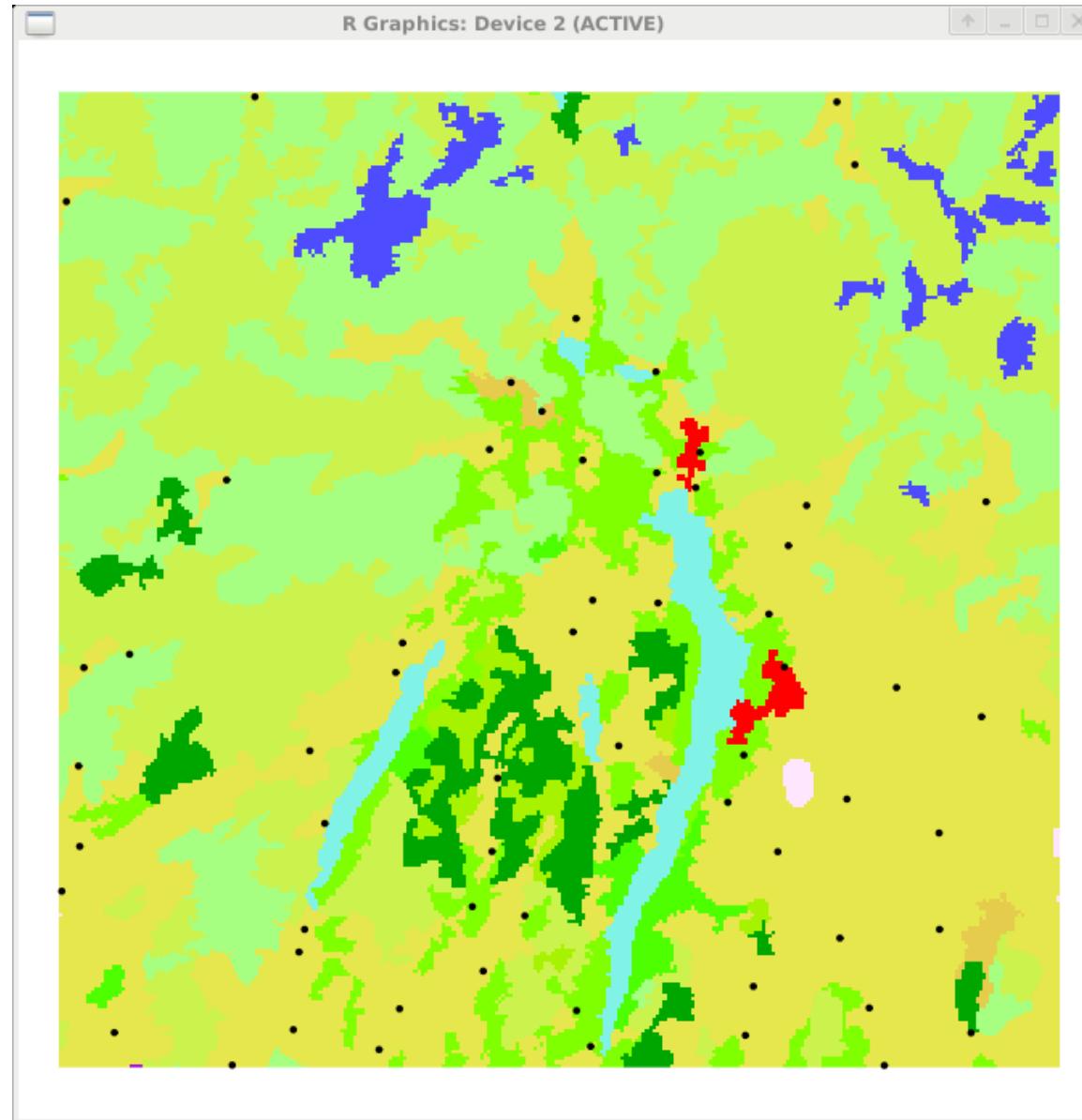


Better to transform points

```
> settlementsP = spTransform(settlements,
    CRS(projection(use)))
```



Points and Raster

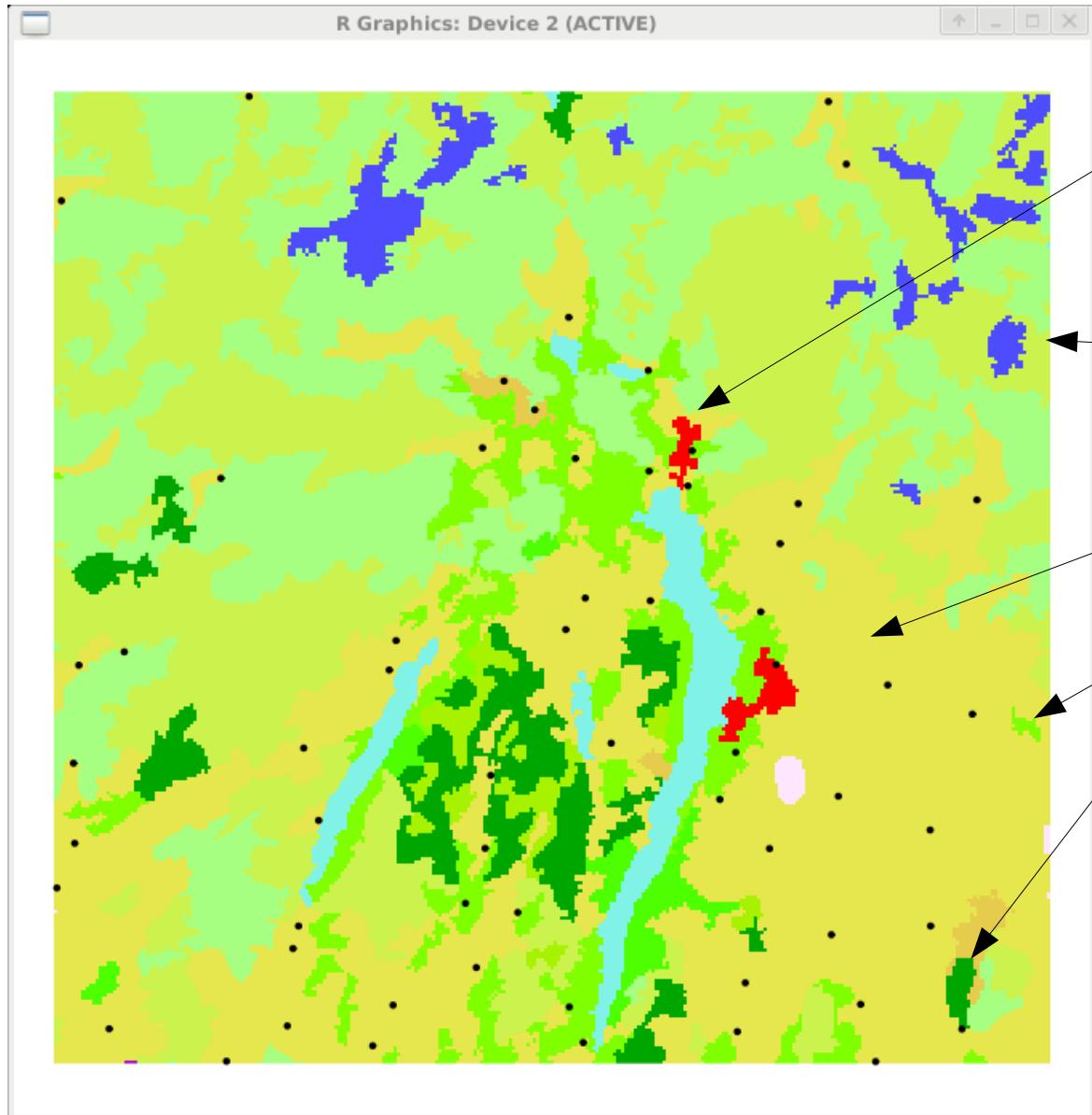


Now lets go to work

- Data created
- Coordinates corrected
- Lets do some analysis!



Settlements and Land Cover

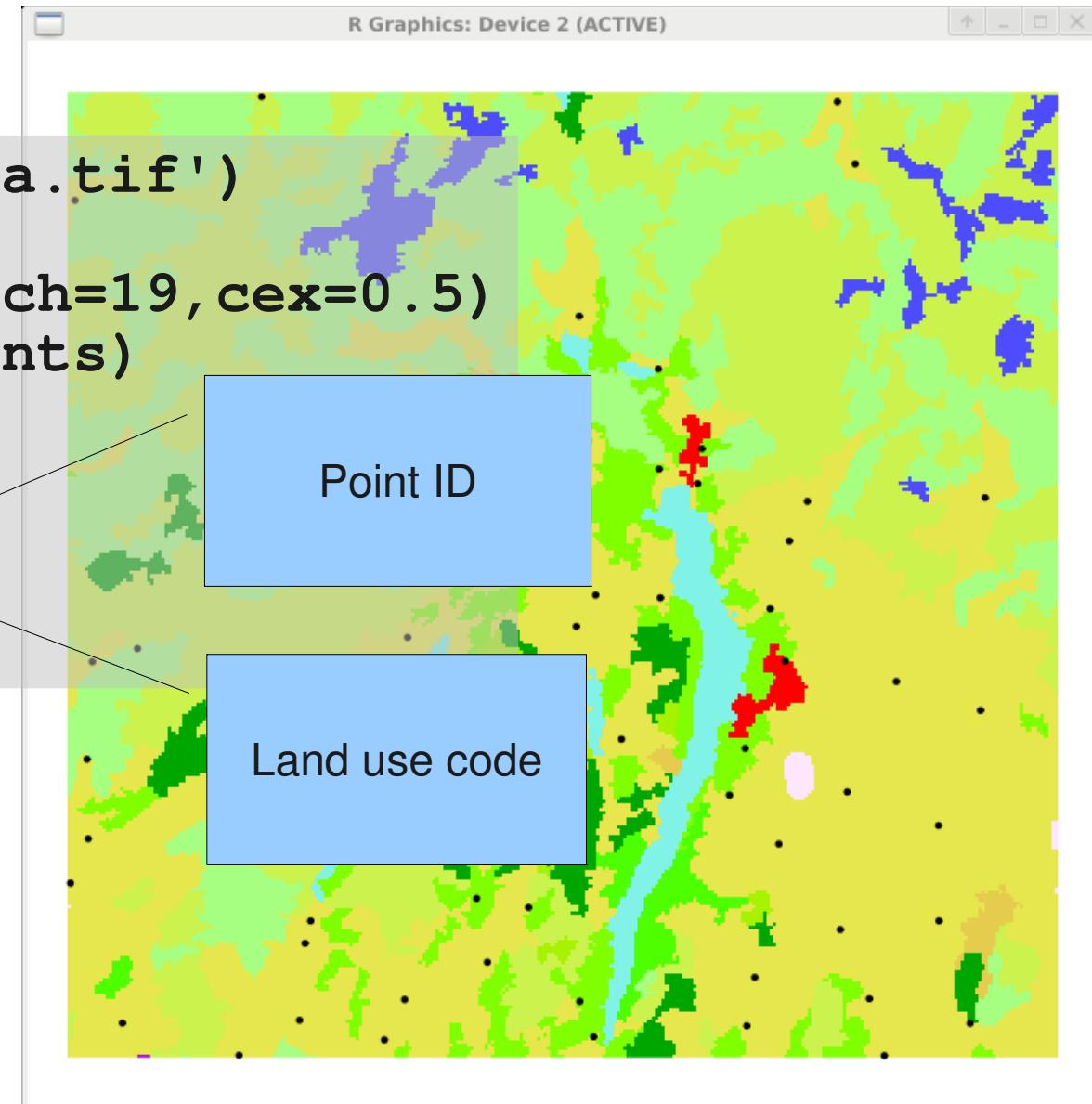


Land Use	
■	Continuous urban fabric
■	Discontinuous urban fabric
■	Industrial or commercial units
■	Road and rail networks and associated land
■	Port areas
■	Airports
■	Mineral extraction sites
■	Dump sites
■	Construction sites
■	Green urban areas
■	Sport and leisure facilities
■	Non-irrigated arable land
■	Permanently irrigated land
■	Rice fields
■	Vineyards
■	Fruit trees and berry plantations
■	Olive groves
■	Pastures
■	Annual crops associated with permanent crops
■	Complex cultivation patterns
■	Land principally occupied by agriculture, with sig
■	Agro-forestry areas
■	Broad-leaved forest
■	Coniferous forest
■	Mixed forest
■	Natural grasslands
■	Moors and heathland
■	Sclerophyllous vegetatio
■	Transitional woodland-sh
■	Beaches, dunes, sands
■	Bare rocks
■	Sparsely vegetated area
■	Burnt areas
■	Glaciers and perpetual s
■	Inland marshes
■	Peat bogs
■	Salt marshes
■	Salines
■	Intertidal flats
■	Water courses
■	Water bodies
■	Coastal lagoons
■	Estuaries
■	Sea and ocean
■	NODATA
■	UNCLASSIFIED LAND S
■	UNCLASSIFIED WATER
■	UNCLASSIFIED

Point-raster overlay

```
> use = raster('cumbria.tif')
> plot(use)
> points(settlements, pch=19, cex=0.5)
> extract(use, settlements)
```

1900	1912	1931	1937
18	18	18	18
1958	1964	1976	1980
23	18	18	23
[etc]			



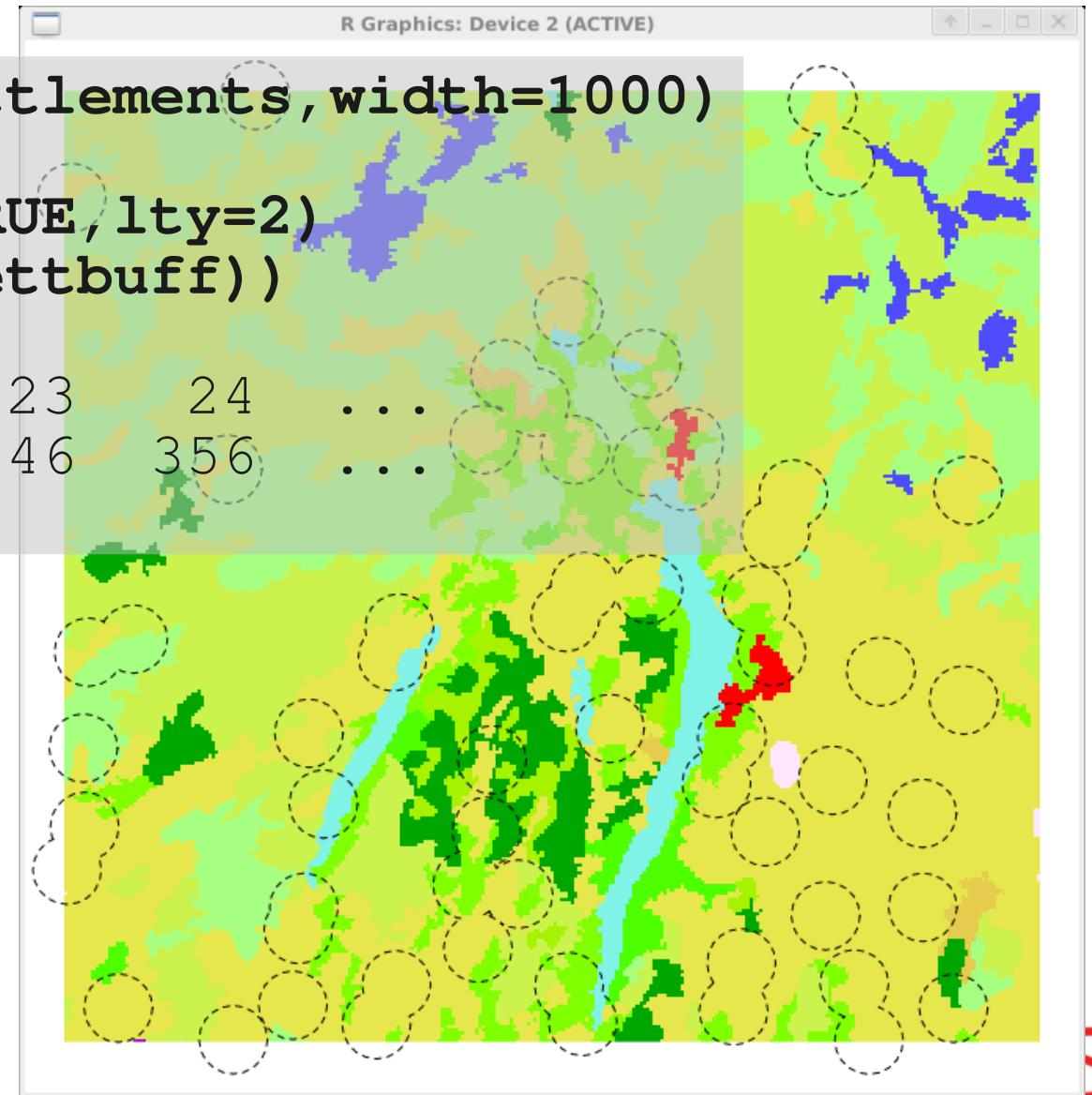
Buffering with rgeos

```
> settbuff=gBuffer(settlements,width=1000)
> plot(use)
> plot(settbuff,add=TRUE,lty=2)
> table(extract(use,settbuff))
```

2	11	18	21	23	24	...
243	5	9730	173	2346	356	...

category

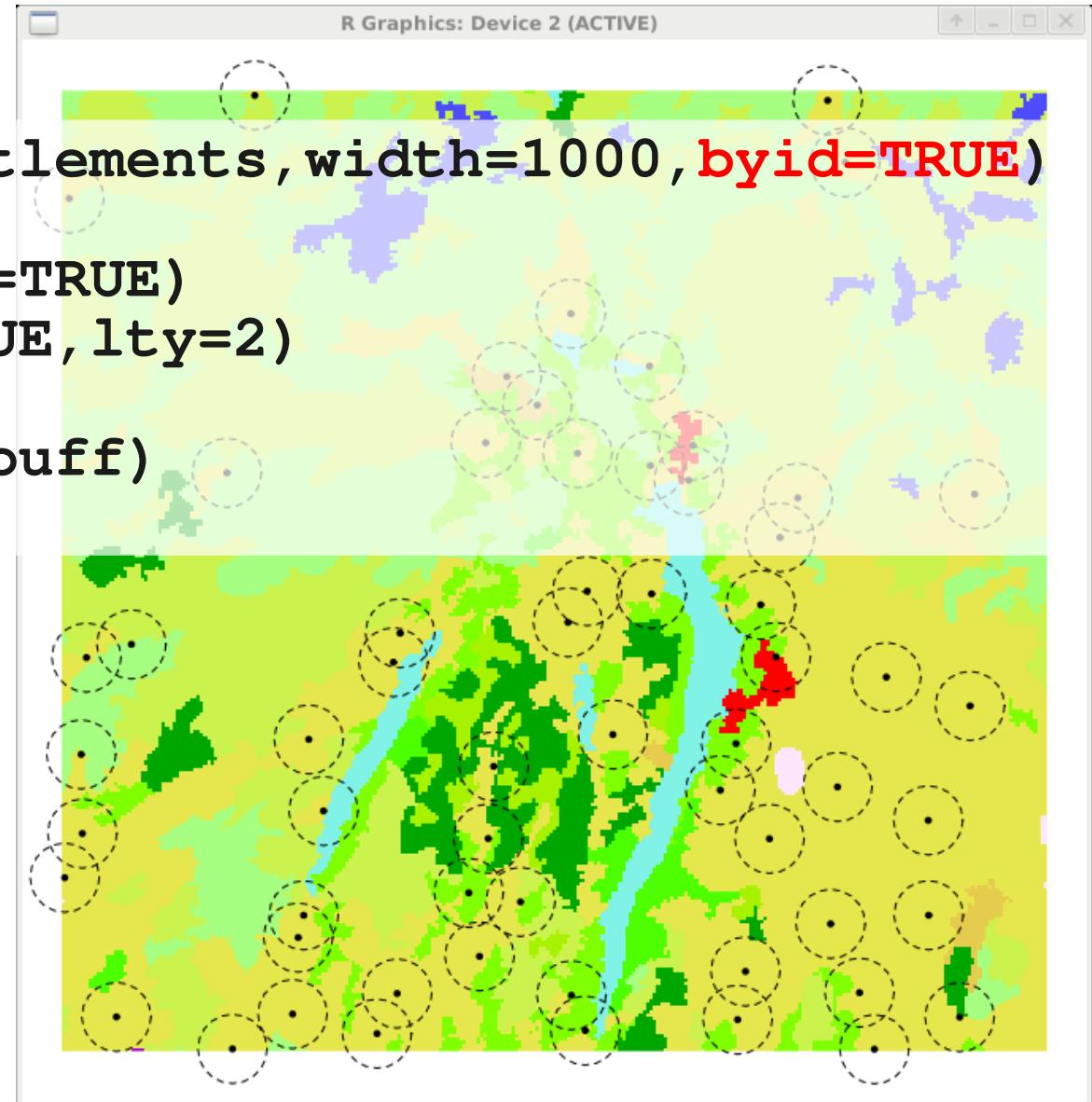
Number of grid cells



Per-feature buffers

```
> settbuff=gBuffer(settlements,width=1000,byid=TRUE)
> plot(use)
> plot(settlements,add=TRUE)
> plot(settbuff,add=TRUE,lty=2)

> e = extract(use,settbuff)
> et = lapply(e,table)
```



Extracting The Values

`e = extract(use, settbuff)`

Pixels under town 1 buffer

```
[ [1]
  [1] 18 18 18 18 18
  [6] 18 18 18 18 18
  [11] 26 26 26 18 18
  [16] 18
```

[...]

```
[ [2]
```

```
  [1] 25 25 18 18 18
  [6] 25 25 25 18 18
  [11] 18 18 18 18 18
  [16] 18 18 25 25 25
  [21] 18 18 18 18 18
```

...

Pixels under town 2 buffer

Tabulating the values

```
et = lapply(e, table)
```

Category count for town 55

```
...
[[55]]
  18   23   26   27   41
  60  112    1  109   32

[[56]]
  18   23   26   27   41
 180   51    7   36   33

[[57]]
  18   26   27
  71     3 124
...
```

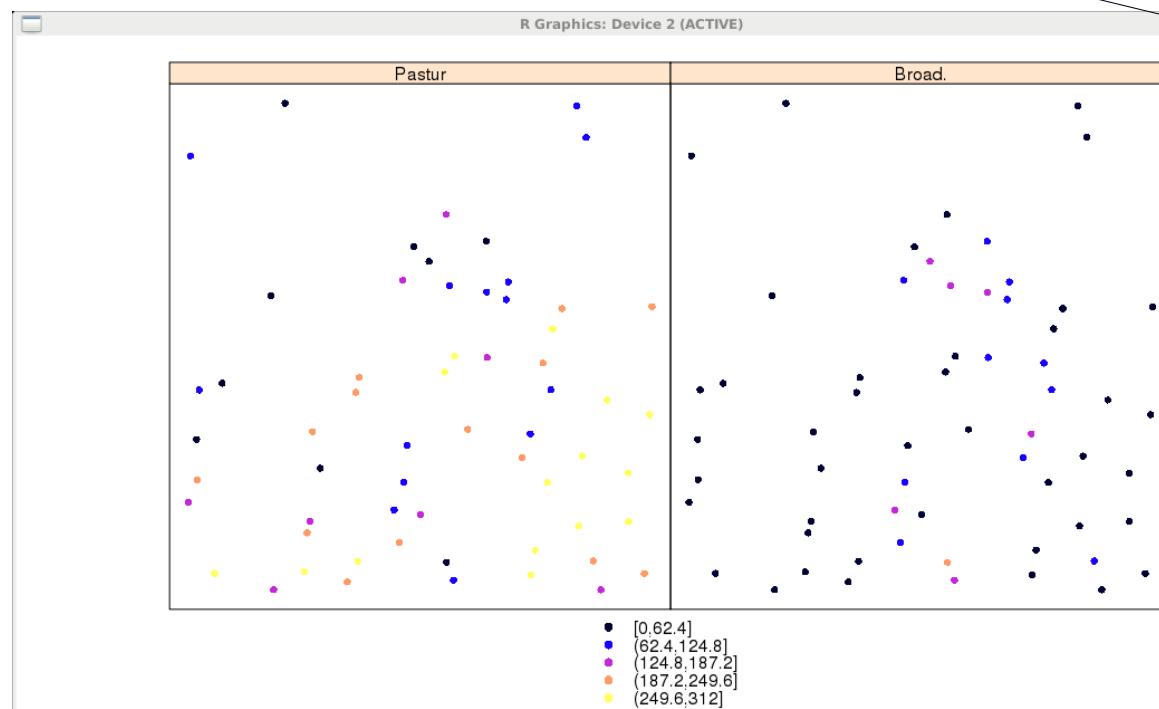
With a bit more massage and manipulation...

New SpatialPoints

```
> suse
```

	coordinates	Sport.	Pastur	Discon	Broad.	Conife	NAME
1	(3471640, 3538260)	0	141	0	0	0	Gawthwaite
2	(3468280, 3539190)	0	302	0	0	0	Grizebeck
3	(3475820, 3538700)	0	224	0	28	0	Colton
4	(3473380, 3539270)	0	293	0	0	0	Lowick
5	(3481840, 3538800)	0	73	0	158	1	Lakeside
6	(3476400, 3539870)	0	311	0	0	0	Oxen Park
7	(3490200, 3538250)	0	131	0	34	0	The Howe

```
> spplot(suse,c('Pastur','Broad.'))
```



Grid cell counts

Question

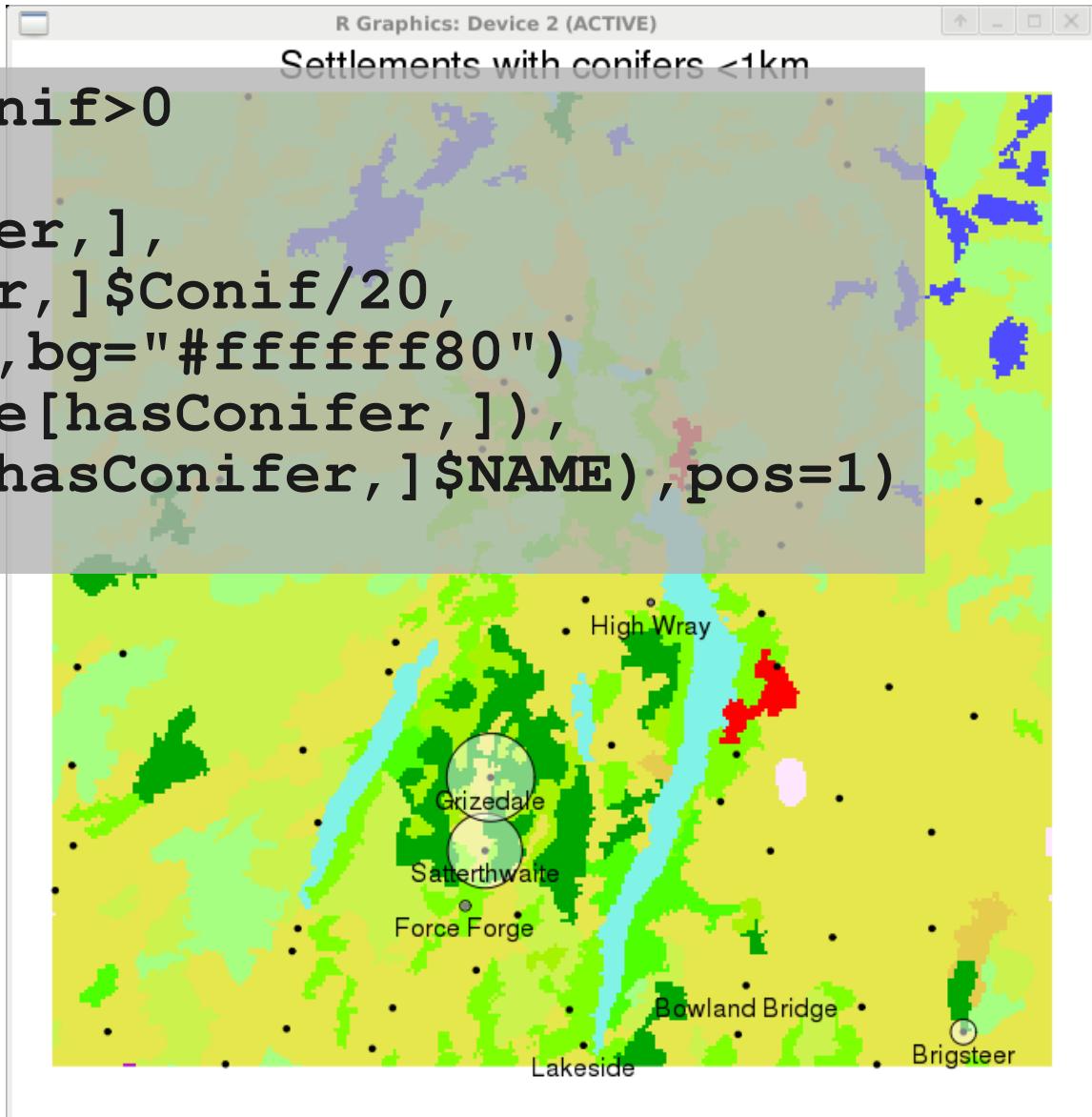
- Find all the towns that have any conifer forest within 1km
- Make a map showing how much conifer forest is around each town





Map presentation

```
> hasConifer = suse$Conif>0
> plot(use)
> points(suse[hasConifer, ],
  cex=suse[hasConifer, ]$Conif/20,
  pch=21,col="black",bg="#ffffffff80")
> text(coordinates(suse[hasConifer, ]),
  as.character(suse[hasConifer, ]$NAME),pos=1)
```



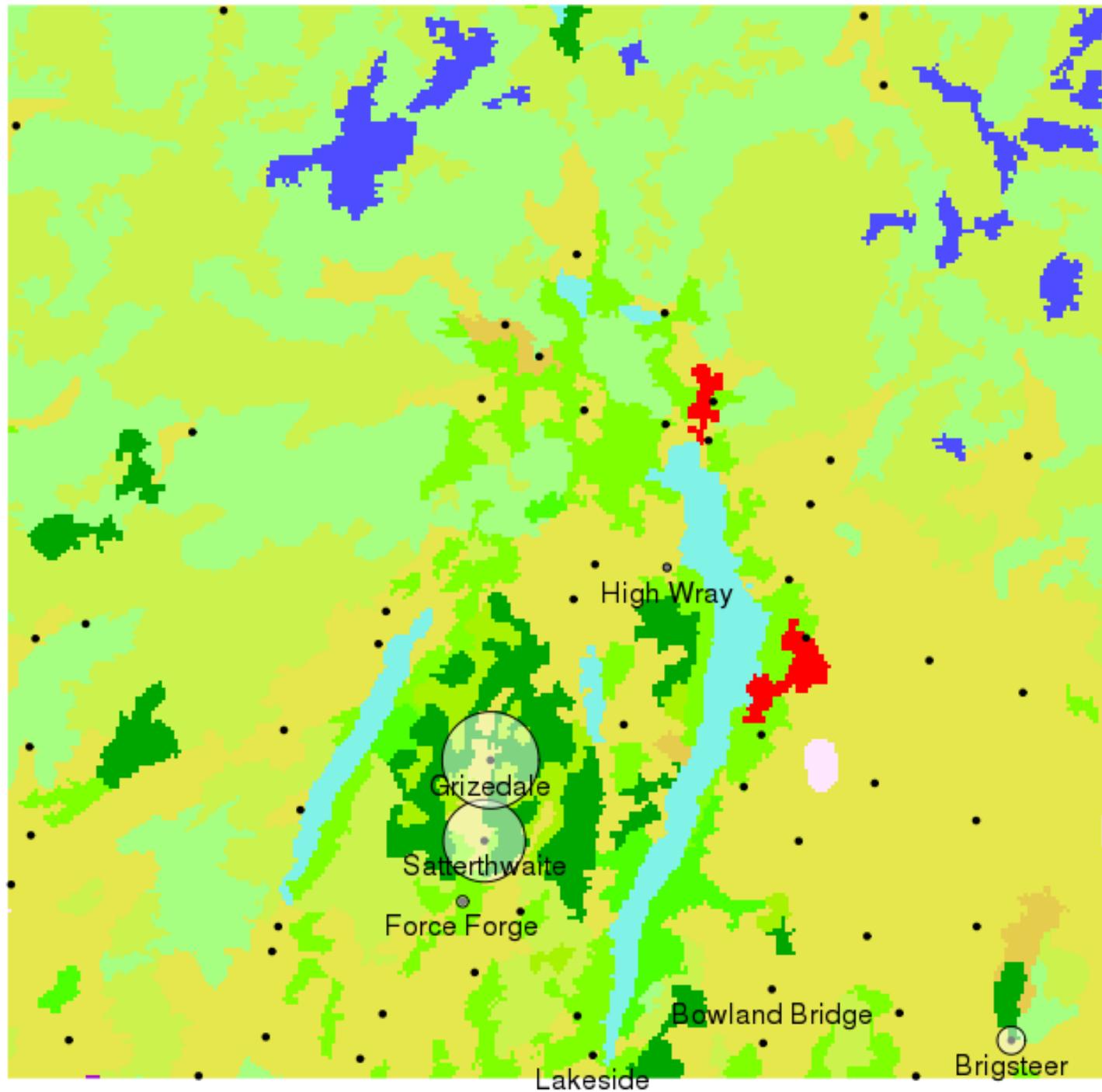
Automation

- Write it all up into a function

```
nearBy = function(  
  towns,  
  landUse,  
  code,  
  dist=1000,  
  thresh=0) {  
  
  ...  
}
```

```
mapNearBy(settlements, landUse,  
           22, dist=2000)
```

Settlements with conifers <1km



End of Part One

- Stay tuned for more!

