

Essentials of

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ISRIC
World Soil Information

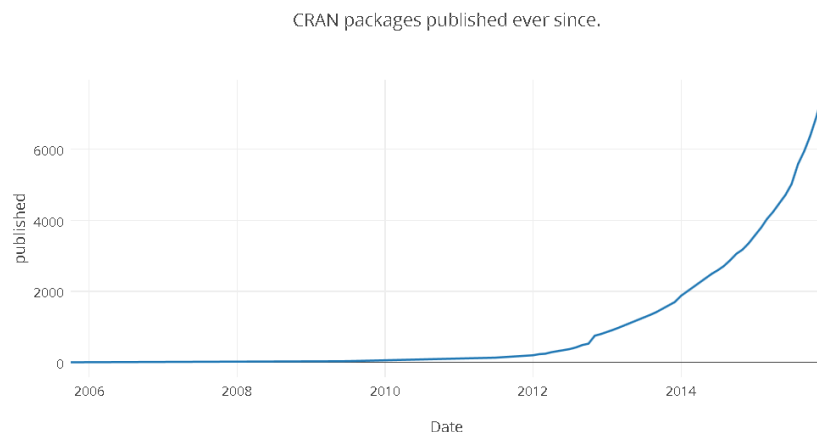
Content

- R and RStudio
- Setting-up an R session
- Data types
- Data structures
- Selecting subsets
- Functions
- Import/export of data
- Plotting
- Working with spatial data in R (sp and raster packages)



What is R?

- R is a free software environment for statistical computing and graphics
- R provides a wide variety of **data processing and analyses, statistical** (linear and nonlinear modelling, classical statistical tests, time-series analysis, classification, clustering, ...) and **graphical** techniques
- R is highly extensible:
 - Base functionality (comes with R)
 - Extension via '**packages**' (>10,000)
- Homepage: www.r-project.org





Why R?

- It is free and it runs on a variety of platforms
- Platform for (advanced) statistical data analyses
- State-of-the-art graphic capabilities
- Very large user community on the web; lots of resources
- Supports **reproducible/transparent** and **collaborative** research (GitHub)
- Connects with other software (SAGA GIS, GE, Python)
- R has a steep learning curve
- Thousands of packages, not always easy to find what you are looking for
- Sometimes cryptic error messages
- Not a GIS



Working with R

The screenshot displays the R GUI (64-bit) interface. The R Console window on the left shows the R version 3.2.0 (2015-04-16) and the R Foundation for Statistical Computing copyright notice. The R Editor window on the right shows a script file named 'D:\NSRIC\DSM_Drenthe\Geostatistical modelling\pSOM\1_classobs\4_pSOMprediction.r'. The script contains two main parts: 'PART 1: Create prediction grid' and 'PART 2: Prepare covariates'. Part 1 involves reading a spatial extent test area, sampling points, and creating a prediction grid. Part 2 involves loading a sheet with values for locations with covariate NA and preparing the covariates.

```
R version 3.2.0 (2015-04-16) -- "Full of Ingredients"
Copyright (C) 2015 The R Foundation for Statistical Computing
Platform: x86_64-w64-mingw32/x64 (64-bit)

R is free software and comes with ABSOLUTELY NO WARRANTY.
You are welcome to redistribute it under certain conditions.
Type 'license()' or 'licence()' for distribution details.

R is a collaborative project with many contributors.
Type 'contributors()' for more information and
'citation()' on how to cite R or R packages in publications.

Type 'demo()' for some demos, 'help()' for on-line help, or
'help.start()' for an HTML browser interface to help.
Type 'q()' to quit R.

> |
```

```
D:\NSRIC\DSM_Drenthe\Geostatistical modelling\pSOM\1_classobs\4_pSOMprediction.r - R Editor

#-----
#PART 1: Create prediction grid
#-----

# read spatial extent test area
area <- readOGR(dsn="D:/PROJECTEN/Actualisatie Drenthe/Workingdir", layer="studiegebied2011_clip")

# sample area: these are the prediction locations
p <- spsample(
  area,
  cellsize = resolution,
  type = "regular",
  offset = c(0.5, 0.5)
)

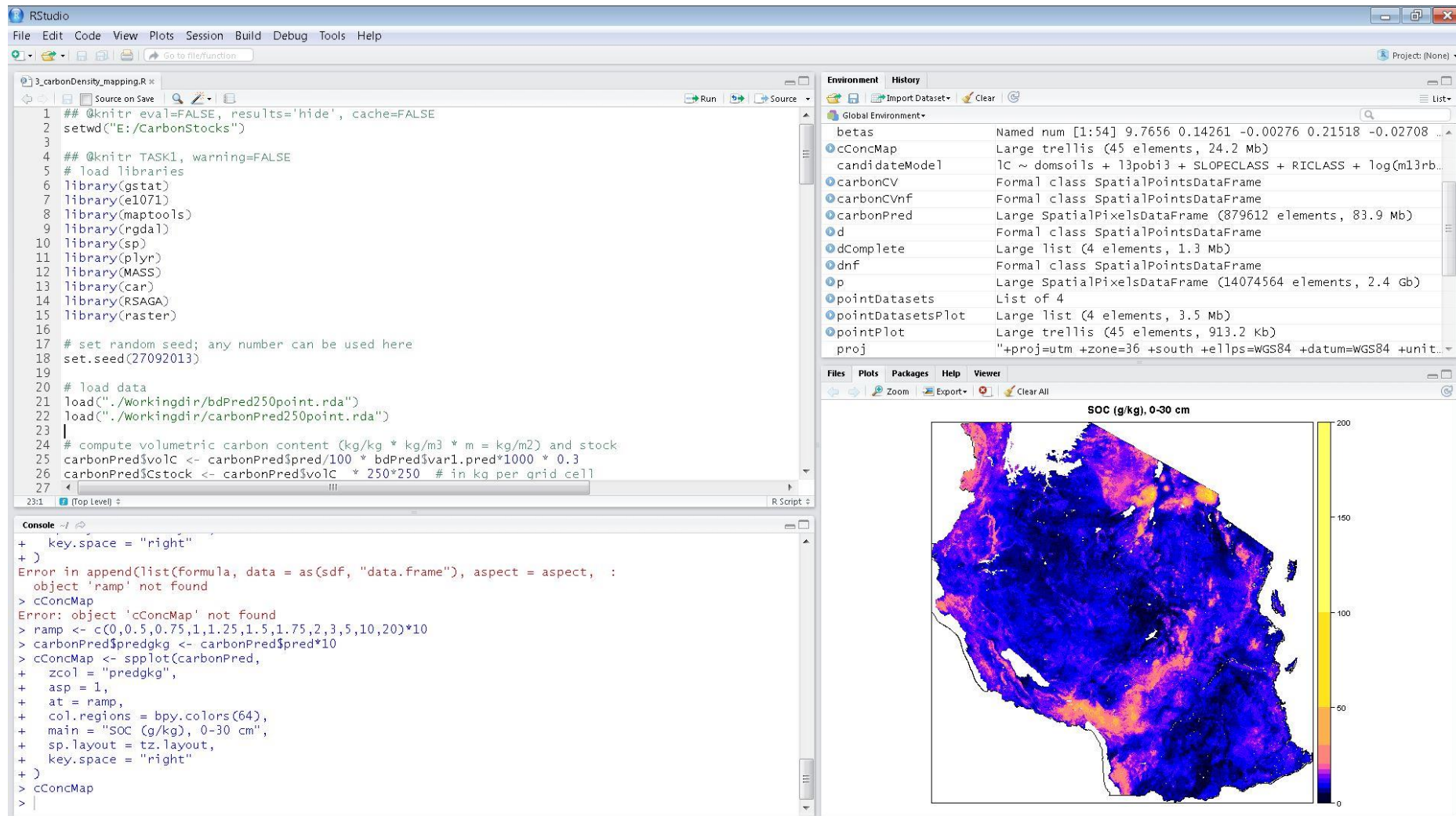
#make nodes 25m and 50m rasters align
p<-as.data.frame(p)
p$x1<-p$x1-3.582465
p$x2<-p$x2+0.891
coordinates(p) <- ~x1+x2

#-----
#PART 2: Prepare covariates
#-----

# load sheet with values for locations with covariate NA (few points outside province of Drenthe)
z <- odbcConnectExcel("D:/PROJECTEN/Actualisatie Drenthe/Spatial data/Covariates/covarvalues.xls")
missing.covar <- sqlFetch(z,"Sheet1")
close(z)
rm(z)

missing.covar$reclam3<-factor(missing.covar$reclam3, levels=c(levels(borings$reclam3)))
missing.covar$highmoor<-factor(missing.covar$highmoor, levels=c(levels(borings$highmoor)))
```

RStudio





First steps.....

- Tell R where to find and save your files: setting the **working directory** using the 'setwd' command:
 - `setwd("D:/SpringSchool/Rintro/workingdir")`
 - `setwd("D:\\SpringSchool\\Rintro\\workingdir")`
 - or use the RStudio file browser
- Load packages that are required for your analyses:
 - `library(gstat)`
 - `require(gstat)`packages should be **installed** on your computer
- Package repository: windows user profile
- Note: R is **case-sensitive**!



Setting up your session

The screenshot displays the RStudio interface with the following components:

- Script Editor:** Contains R code for setting the working directory, loading libraries, setting a random seed, loading data, and creating an object.
- Console:** Shows the execution output, including library loading messages and GDAL/PROJ version information.
- Environment Pane:** Displays the global environment with variables `carbonPred` and `d`.

```
# RStudio Script Editor content:
1 # set working directory
2 setwd("D:/ISRIC/DSM_Tanzania/CarbonStocks")
3
4 # load libraries
5 library(gstat)
6 library(sp)
7 library(maptools)
8 library(rgdal)
9
10 # set random seed
11 set.seed(06052016)
12
13 # load data from a previous session
14 load("./Workingdir/2a_ocPred1000point.rda")
15
16 # create an object
17 d <- 10
18
19
```

Console Output:

```
> setwd("D:/ISRIC/DSM_Tanzania/CarbonStocks")
>
> # load libraries
> library(gstat)
> library(sp)
> library(maptools)
Checking rgeos availability: TRUE
> library(rgdal)
rgdal: version: 1.2-5, (SVN revision 648)
Geospatial Data Abstraction Library extensions to R successfully loaded
Loaded GDAL runtime: GDAL 2.0.1, released 2015/09/15
Path to GDAL shared files: D:/R_LIBS/rgdal/gdal
Loaded PROJ.4 runtime: Rel. 4.9.2, 08 September 2015, [PJ_VERSION: 492]
Path to PROJ.4 shared files: D:/R_LIBS/rgdal/proj
Linking to sp version: 1.2-4
>
> # set random seed
> set.seed(06052016)
>
> # load data from a previous session
> load("./Workingdir/2a_ocPred1000point.rda")
>
> # create an object
> d <- 10
>
```

Environment Pane:

Variable	Value
carbonPred	Large SpatialPixelsDataFrame (879612 elements, 63.8 Mb)
d	10

- Set working directory
- Load packages
- Load data or R objects



R scripts and data objects

- R scripts are saved as “<name>.R” files
- R objects in the environment can be saved for future use.
- Saving:
 - entire environment: **save.image** function
 - couple of objects: **save** function
- Output saved as “<name>.rda” or “<name>.RDATA” files

```
Console D:/ISRIC/DSM_Tanzania/CarbonStocks/ ↗  
> #save data  
> save.image("./Workingdir/carbon.rda")  
> save(d, file = "./Workingdir/data.rda")  
> |
```

Basic data types

- Numeric

```
> x <- 10.1
> x
[1] 10.1
> class(x)
[1] "numeric"
```

- Integer

```
> y <- as.integer(x)
> y
[1] 10
> class(y)
[1] "integer"
```

- Logical

```
> x <- 1; y <- 2
> z <- x > y
> z
[1] FALSE
> class(z)
[1] "logical"
```

- Character

```
> z <- "Bas"
> z
[1] "Bas"
> class(z)
[1] "character"
```

vector

```
> lc <- c("arable", "grassland", "forest", "arable")
> lc
[1] "arable" "grassland" "forest" "arable"
> class(lc)
[1] "character"
```

- Factor: vector that can only contain **pre-defined** values, used to store **categorical** data.

function

```
> lc.f <- as.factor(lc)
> lc.f
[1] arable grassland forest arable
Levels: arable forest grassland
> class(lc.f)
[1] "factor"
```

```
> summary(lc)
  Length      Class      Mode 
      4  character character
> summary(lc.f)
  arable forest grassland 
      2       1         1
```



Vectors

- Sequence of elements of the **same** basic type (one-dimensional).

```
> c(6, 9, 2016)
[1] 6 9 2016
> c("spring", "school")
[1] "spring" "school"
```

- Vectors can be combined.

```
> a <- c(6, 9, 2016)
> b <- c("spring", "school")
>
> z <- c(a, b)

> z
[1] "6"      "9"      "2016"   "spring" "school"
> class(z)
[1] "character"
> length(z)
[1] 5
```

Vector arithmetic

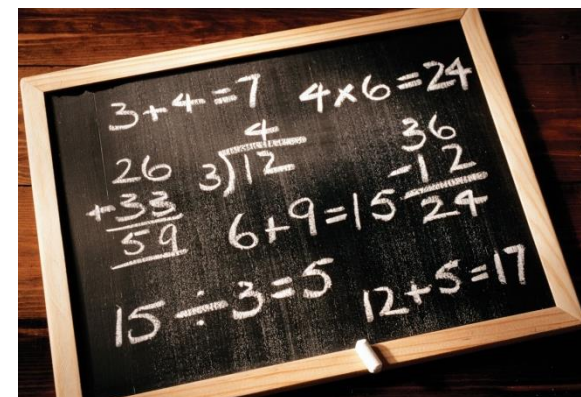
- Vectorized operations: most operations work on vectors with the same syntax as they work on scalars (no need for looping).

- Vector arithmetic:

```
> a <- c(1, 2, 3)
> b <- c(4, 5, 6)
>
> 5 * a
[1] 5 10 15
> a + b
[1] 5 7 9
```

- Recycling of vector elements:

```
> a <- c(1, 2, 3)
> b <- c(4, 5, 6, 7, 8, 9)
> a + b
[1] 5 7 9 8 10 12
> a - mean(a)
[1] -1 0 1
```



Other data structures

- **Matrix:** two-dimensional vector
- **List:** data structure that can hold any number of any types of other data structures
- **Data frame:** table



- Data frame is the **fundamental** data structure for statistical modelling in R.
- Data frame is a **table** with columns and rows (fields and records).



Data frame

- Columns can have different data types (numeric, integer, logical, character, factor)
- All columns must have the same length

```
> data(trees)
> str(trees)
'data.frame': 31 obs. of  3 variables:
 $ Girth : num  8.3 8.6 8.8 10.5 10.7 10.8 11 11 11.1 11.2 ...
 $ Height: num  70 65 63 72 81 83 66 75 80 75 ...
 $ Volume: num  10.3 10.3 10.2 16.4 18.8 19.7 15.6 18.2 22.6 19.9 ...
```

Selecting subsets

- Selection is done with '[...]'

- Vector:

```
> a
[1] 1 2 3
> a[1]
[1] 1
> a[-1]
[1] 2 3
> a[c(1,3)]
[1] 1 3
```

- Data frame:

```
> trees[1,]
  Girth Height Volume
1   8.3    70   10.3
> trees[,3]
[1] 10.3 10.3 10.2 16.4 18.8 19.7 15.6 18.2 22.6 19.9 24.2 21.0 21.4 21.3 19.1 22.2 33.8 27.4 25.7
[20] 24.9 34.5 31.7 36.3 38.3 42.6 55.4 55.7 58.3 51.5 51.0 77.0
> trees[1,3]
[1] 10.3

> trees$Girth
[1] 8.3 8.6 8.8 10.5 10.7 10.8 11.0 11.0 11.1 11.2 11.3 11.4 11.4 11.7 12.0 12.9 12.9 13.3 13.7 13.8 14.0
[22] 14.2 14.5 16.0 16.3 17.3 17.5 17.9 18.0 18.0 20.6
> trees$Height
[1] 70 65 63 72 81 83 66 75 80 75 79 76 76 69 75 74 85 86 71 64 78 80 74 72 77 81 82 80 80 80 87
> trees$Volume
[1] 10.3 10.3 10.2 16.4 18.8 19.7 15.6 18.2 22.6 19.9 24.2 21.0 21.4 21.3 19.1 22.2 33.8 27.4 25.7 24.9 34.5
[22] 31.7 36.3 38.3 42.6 55.4 55.7 58.3 51.5 51.0 77.0
```



Functions

- Data analyses and modelling is done through functions.
- Functions can be very simple:

```
> sqrt(25)
[1] 5
```

- More complex functions have multiple arguments (inputs):

```
# compute sample variogram
v <- variogram(carbon~1, data = d, width = 1)

# define variogram model: initial values for psill, range, nugget based on sample variogram
vm <- vgm(psill = 30, model = "Sph", range = 5000, nugget = 80)

# fit variogram model to the experimental variogram
vmf <- fit.variogram(v, model = vm)
```

- Arguments have specific requirements
- Access help: ?fit.variogram



Importing data

- Importing from tables:
 - csv: `read.csv()`
 - txt: `read.table()`
 - xls: `read.xls()` [requires package 'xls']
- Data is imported as a **data.frame** object

```
d <- read.csv("Soil.csv")  
d <- read.table("Soil.txt", sep="\t", header=TRUE)  
d <- read.xls("Soil.xls", sheetName = "All_Sam_Pit")
```



Exporting data

- Variety of exporting formats for tabular data:
 - `write.table()`
 - `write.csv()`
 - `write.xlsx()`

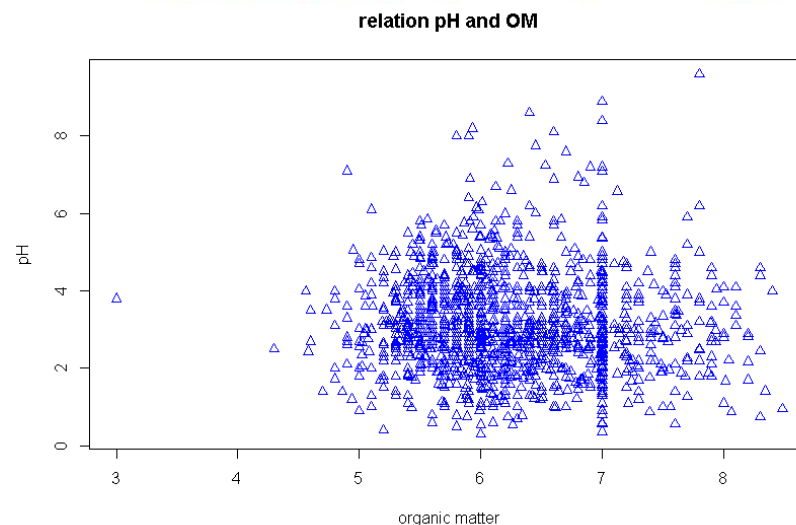
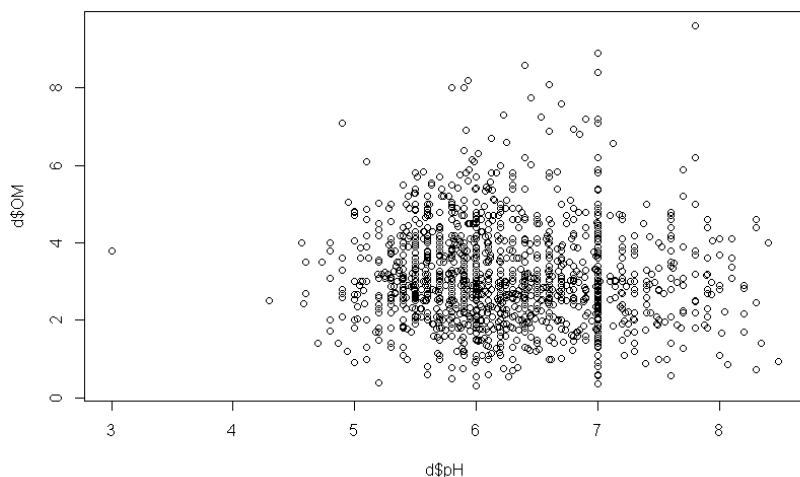
```
write.table(d, file = "soildata.txt", sep = ";")  
write.csv(d, file = "soildata.csv")  
write.xlsx(d, file="soildata.xlsx",sheetName="Nepal",col.names=TRUE, row.names=FALSE)
```

Plotting

- Large number of packages and functions for generating plots with basic functionality to ‘high-level’: e.g. lattice and ggplot.
- The basic function for plotting is ‘plot’.

```
plot(d$pH,d$OM)
```

```
plot(d$pH,d$OM, xlab="organic matter", ylab = "pH", main="relation pH and OM", pch=2, col="blue")
```



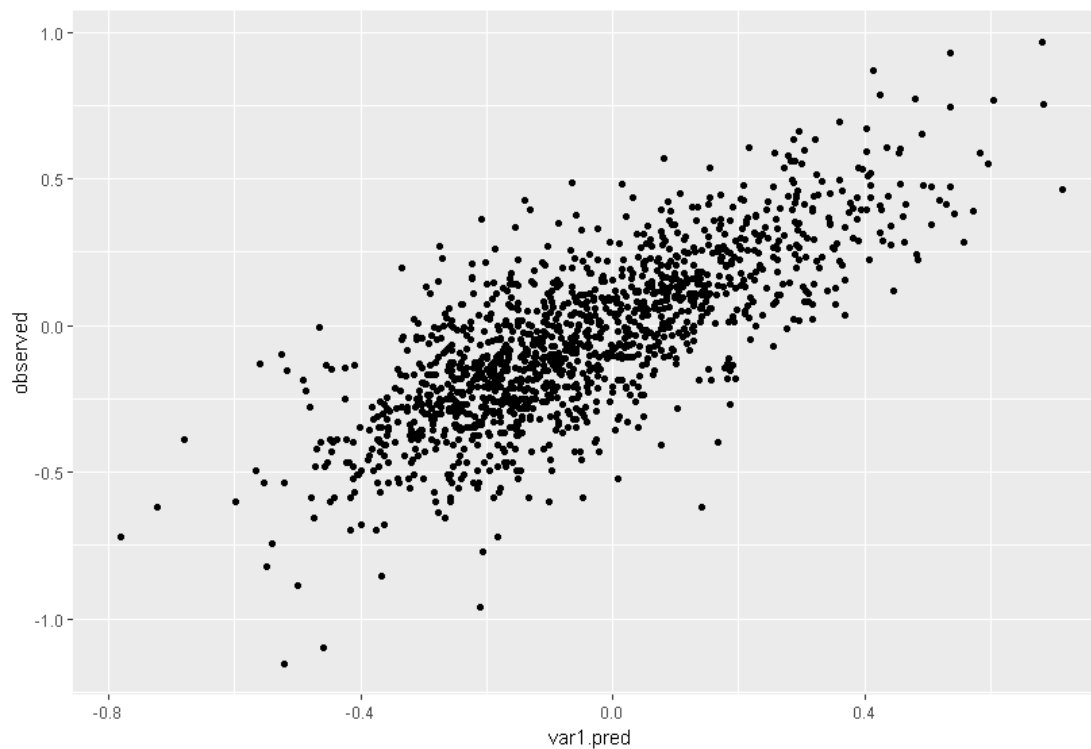


ggplot

- `library(ggplot2)`
- Build your plot layer by layer
- Building blocks:
 - **geom**: the geometric object that describes the type of plot that is produced.
 - **aes**: 'aesthetics', defines how variables in the data are mapped to visual properties.
 - **scales**: control the legend, plot layout.
 - **theme**: controls the appearance of all non-data components.
- <http://ggplot2.org/>

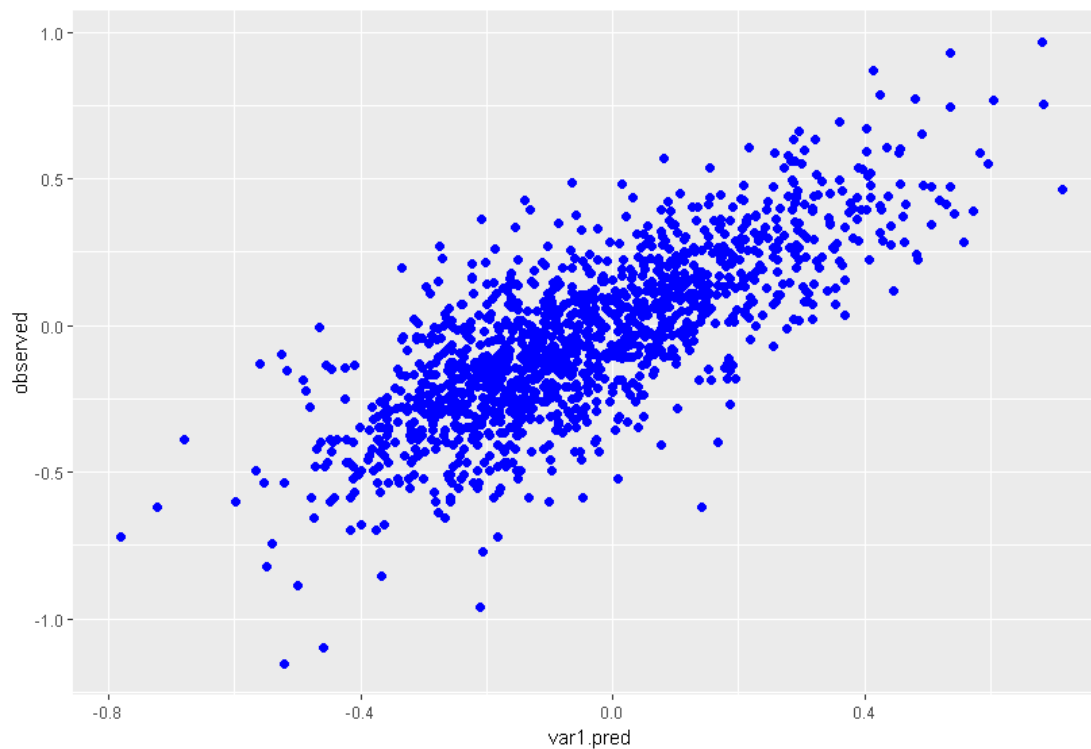
Plotting with ggplot

```
ggplot(data = d) +  
  geom_point(mapping = aes(x = var1.pred, y = observed))
```



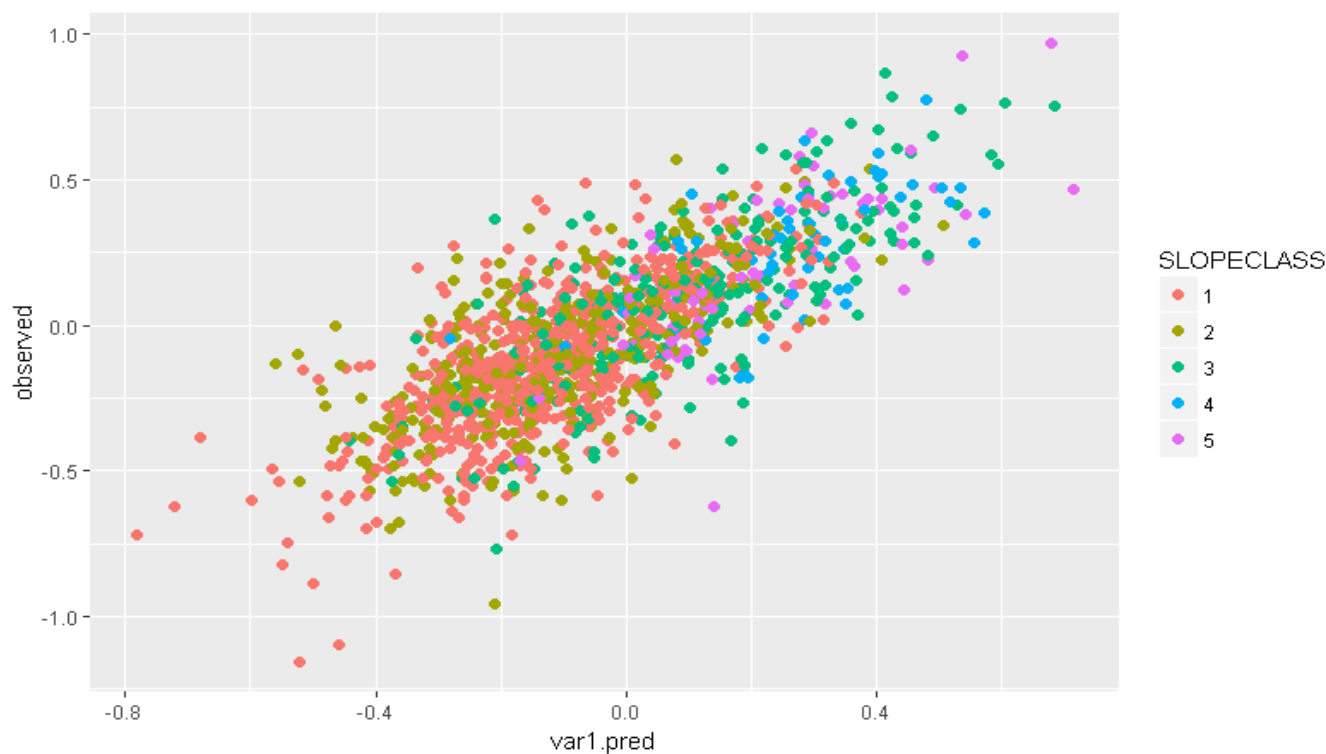
Plotting with ggplot

```
ggplot(data = d) +  
  geom_point(mapping = aes(x = var1.pred, y = observed), color="blue", size=2)
```



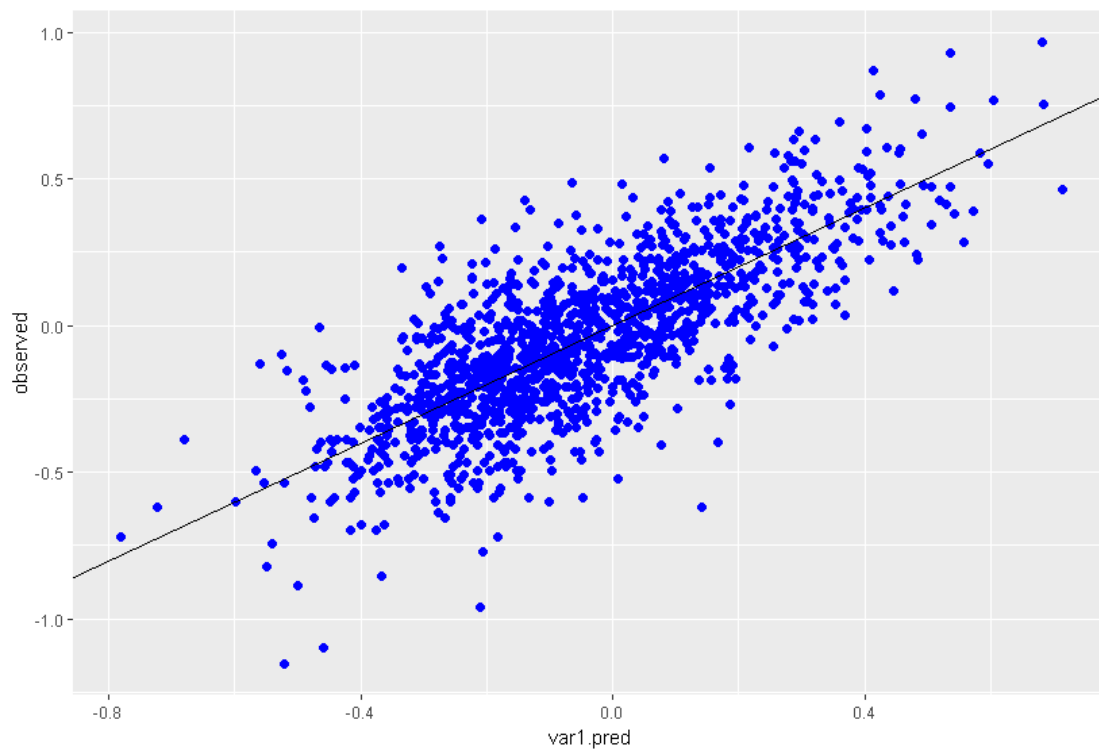
Plotting with ggplot

```
ggplot(data = d) +  
  geom_point(mapping = aes(x = var1.pred, y = observed, color=SLOPECLASS), size=2)
```



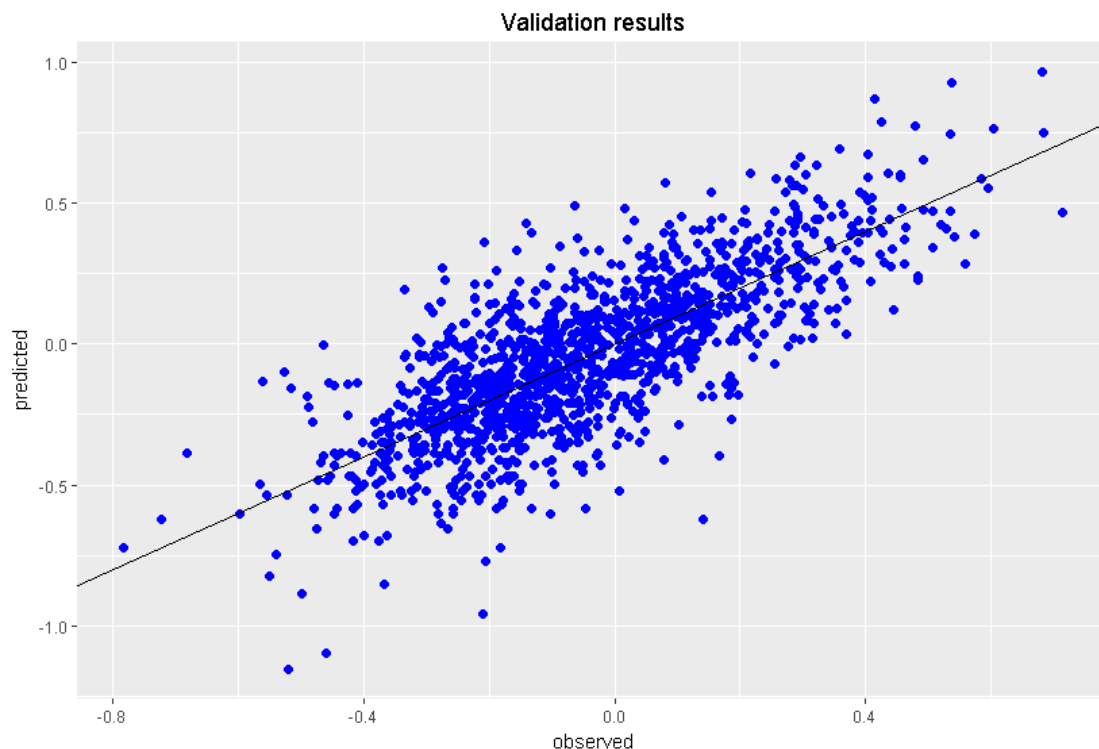
Plotting with ggplot

```
ggplot(data = d) +  
  geom_point(mapping = aes(x = var1.pred, y = observed), color="blue", size=2) +  
  geom_abline(intercept = 0, slope = 1)
```



Plotting with ggplot

```
ggplot(data = d) +  
  geom_point(mapping = aes(x = var1.pred, y = observed), color="blue", size=2) +  
  geom_abline(intercept = 0, slope = 1) +  
  scale_x_continuous(name = "observed") +  
  scale_y_continuous(name = "predicted") +  
  ggtitle("Validation results")+  
  theme(plot.title = element_text(hjust = 0.5))
```





Exporting plots

- Save plots to png, jpg or pdf

```
png("scatterPlot.png", width=550, height=500)  
plot(d$pH,d$OM, xlab="organic matter", ylab = "pH", main="relation pH and OM")  
dev.off()
```

```
scatter <- ggplot(data = d) +  
  geom_point(mapping = aes(x = var1.pred, y = observed))
```

```
png("scatterPlot.png", width=550, height=500)  
scatter  
dev.off()
```

```
jpeg("scatterPlot.jpg", width=550, height=500)  
scatter  
dev.off()
```

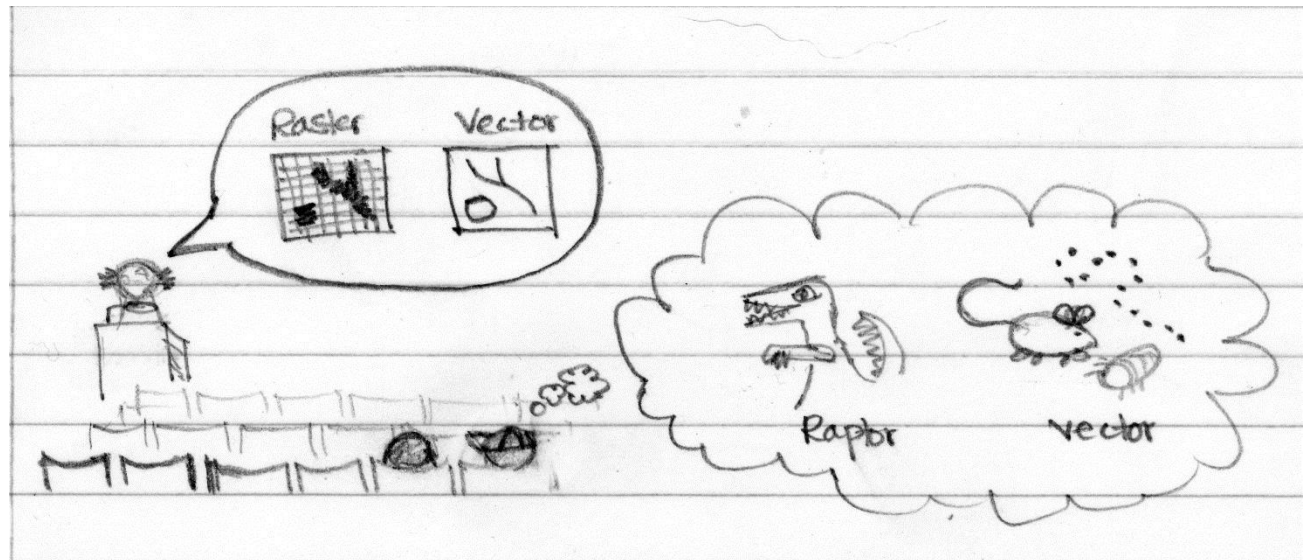
```
pdf("scatterPlot.pdf")  
scatter  
dev.off()
```



Working with spatial data in R

Content

- Spatial data classes (sp, raster)
- Importing/exporting spatial data
- Projections
- Plotting



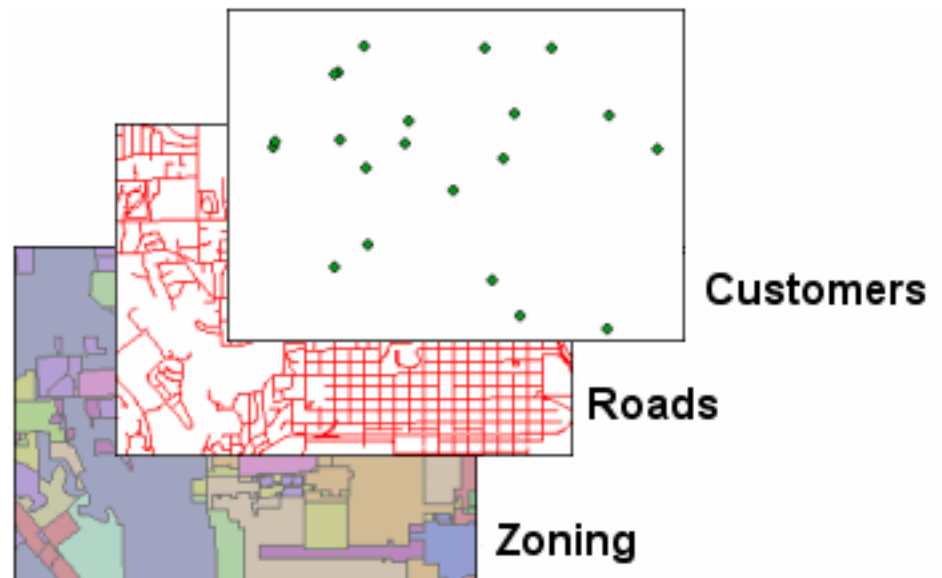


Spatial Data in R

- R offers a wide variety of packages and tools that can handle spatial data.
- Note: R is not a GIS.
- R is not so memory efficient.
- Relevant packages:
 - **sp**: handling spatial data
 - **raster**: reading/manipulating/writing spatial raster data
 - **rgdal**: reading/writing spatial data

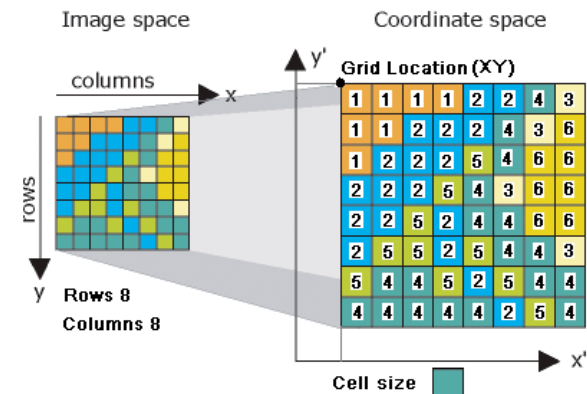
Spatial data classes and formats

- **Vector**: points, lines and polygons (areal).
- For storing data that has **discrete boundaries**, such as country borders, land parcels, and streets.
- Format: **shapefile**



Spatial data classes and formats

- **Raster**: surface divided into a regular grid of cells.
- For storing data that varies **continuously**, as in a satellite image, (surface of chemical concentrations, or an elevation surface).



- **Format**:
 - **GeoTiff**: allows embedding spatial reference information, metadata and color legends. It also supports internal compression
 - Ascii
 - ESRI Grid



Structures for spatial data

- Spatial data is nothing more than a **data frame** that has columns with **X and Y coordinates** (longitude and latitude).
- Example:

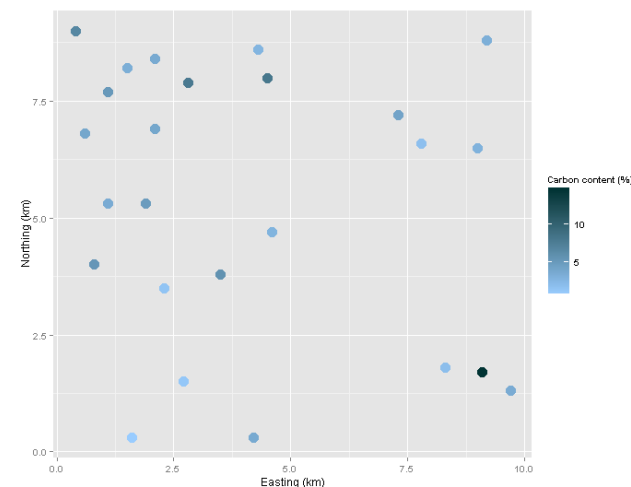
```
> str(d)
'data.frame': 25 obs. of  3 variables:
 $ x      : num  1.9 9.1 1.1 9.7 1.5 0.8 4.6 3.5 9 7.8 ...
 $ y      : num  5.3 1.7 5.3 1.3 8.2 4 4.7 3.8 6.5 6.6 ...
 $ carbon : num  4.8 14.7 3.5 3.5 3.3 5.3 2.8 5.7 2.9 1.8 ...
> head(d)
   x    y carbon
1 1.9 5.3   4.8
2 9.1 1.7  14.7
3 1.1 5.3   3.5
4 9.7 1.3   3.5
5 1.5 8.2   3.3
6 0.8 4.0   5.3
```

- Let's now take a look at R classes for spatial data: **sp** package

Spatial data classes I

- Convert a data frame to a **SpatialPointsDataFrame** object with the **coordinates** function.

```
> coordinates(d) <- ~x+y
> str(d)
Formal class 'SpatialPointsDataFrame' [package "sp"] with 5 slots
..@ data      : 'data.frame': 25 obs. of  1 variable:
.. ..$ carbon: num [1:25] 4.8 14.7 3.5 3.5 3.3 5.3 2.8 5.7 2.9 1.8 ...
..@ coords.nrs : int [1:2] 1 2
..@ coords     : num [1:25, 1:2] 1.9 9.1 1.1 9.7 1.5 0.8 4.6 3.5 9 7.8 ...
.. ..- attr(*, "dimnames")=List of 2
.. .. ..$ : chr [1:25] "1" "2" "3" "4" ...
.. .. ..$ : chr [1:2] "x" "y"
..@ bbox      : num [1:2, 1:2] 0.4 0.3 9.7 9
.. ..- attr(*, "dimnames")=List of 2
.. .. ..$ : chr [1:2] "x" "y"
.. .. ..$ : chr [1:2] "min" "max"
..@ proj4string:Formal class 'CRS' [package "sp"] with 1 slot
.. .. ..@ proj4args: chr NA
```

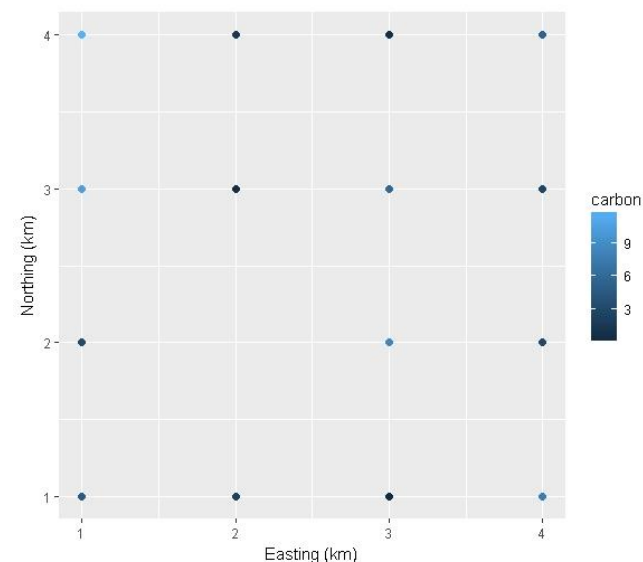


Spatial data classes I

- Data frame with data points at regular intervals.

```
> str(d)
'data.frame': 15 obs. of 3 variables:
 $ x      : num 1 2 3 4 1 3 4 1 2 3 ...
 $ y      : num 1 1 1 1 2 2 2 3 3 3 ...
 $ carbon: num 5 2.4 0.4 7.7 3.4 8.4 3.1 10.4 0.5 5.9 ...
```

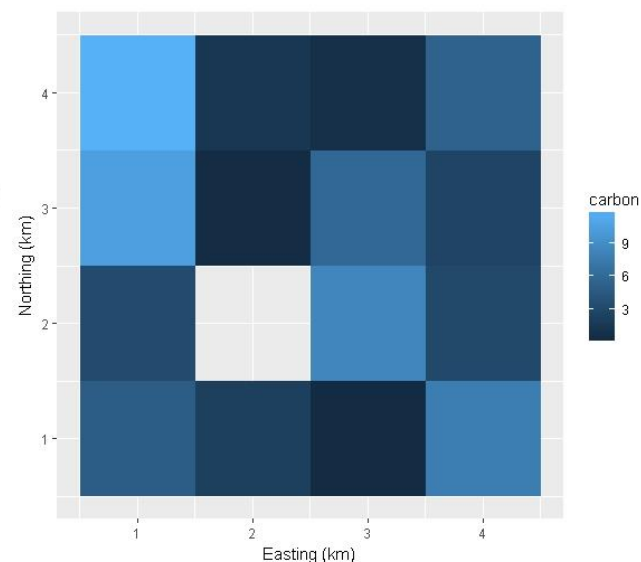
```
> coordinates(d) <- ~x+y
> str(d)
Formal class 'SpatialPointsDataFrame' [package "sp"] with 5 slots
 ..@ data      : 'data.frame': 15 obs. of 1 variable:
 .. ..$ carbon: num [1:15] 5 2.4 0.4 7.7 3.4 8.4 3.1 10.4 0.5 5.9 ...
 ..@ coords.nrs: int [1:2] 1 2
 ..@ coords    : num [1:15, 1:2] 1 2 3 4 1 3 4 1 2 3 ...
 .. ..- attr(*, "dimnames")=List of 2
 .. .. ..$ : chr [1:15] "1" "2" "3" "4" ...
 .. .. ..$ : chr [1:2] "x" "y"
 ..@ bbox      : num [1:2, 1:2] 1 1 4 4
 .. ..- attr(*, "dimnames")=List of 2
 .. .. ..$ : chr [1:2] "x" "y"
 .. .. ..$ : chr [1:2] "min" "max"
 ..@ proj4string: Formal class 'CRS' [package "sp"] with 1 slot
 .. .. ..@ projargs: chr NA
```



Spatial data classes II

- Create a grid object with the **gridded** function:
SpatialPixelsDataFrame

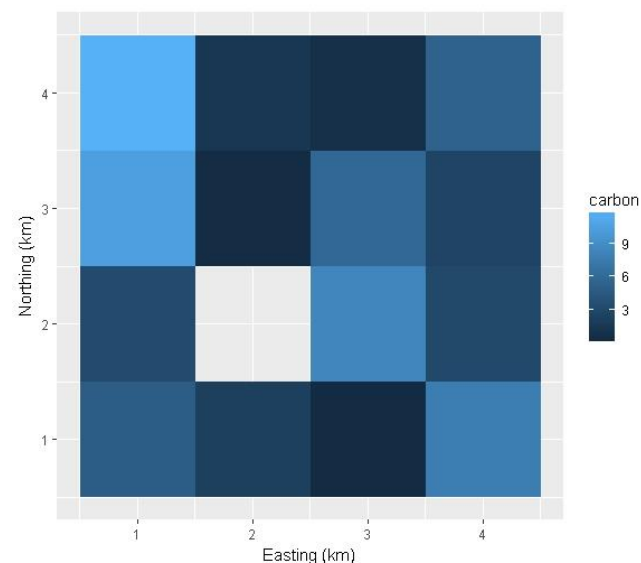
```
> gridded(d) <- TRUE
> str(d)
Formal class 'SpatialPixelsDataFrame' [package "sp"] with 7 slots
..@ data      : 'data.frame': 15 obs. of 1 variable:
.. ..$ carbon: num [1:15] 5 2.4 0.4 7.7 3.4 8.4 3.1 10.4 0.5 5.9 ...
..@ coords.nrs: num(0)
..@ grid      : Formal class 'GridTopology' [package "sp"] with 3 slots
.. ..@ cellcentre.offset: Named num [1:2] 1 1
.. ..@ attr(*, "names")= chr [1:2] "x" "y"
.. ..@ cellsize        : Named num [1:2] 1 1
.. ..@ attr(*, "names")= chr [1:2] "x" "y"
.. ..@ cells.dim       : Named int [1:2] 4 4
.. ..@ attr(*, "names")= chr [1:2] "x" "y"
..@ grid.index : int [1:15] 13 14 15 16 9 11 12 5 6 7 ...
..@ coords     : num [1:15, 1:2] 1 2 3 4 1 3 4 1 2 3 ...
.. ..- attr(*, "dimnames")=List of 2
.. .. ..$ : chr [1:15] "1" "2" "3" "4" ...
.. .. ..$ : chr [1:2] "x" "y"
..@ bbox       : num [1:2, 1:2] 0.5 0.5 4.5 4.5
.. ..- attr(*, "dimnames")=List of 2
.. .. ..$ : chr [1:2] "x" "y"
.. .. ..$ : chr [1:2] "min" "max"
..@ proj4string: Formal class 'CRS' [package "sp"] with 1 slot
.. ..@ projargs: chr NA
```



Spatial data classes III

- Create a full grid object with the **fullgrid** function:
SpatialGridDataFrame

```
> fullgrid(d) <- TRUE
> str(d)
Formal class 'SpatialGridDataFrame' [package "sp"] with 4 slots
..@ data      : 'data.frame': 16 obs. of 1 variable:
.. ..$ carbon: num [1:16] 11.8 1.5 0.9 5.5 10.4 0.5 5.9 2.8 3.4 NA ...
..@ grid      : Formal class 'GridTopology' [package "sp"] with 3 slots
.. ..@ cellcentre.offset: Named num [1:2] 1 1
.. ..@ cellsize         : Named num [1:2] 1 1
.. ..@ cells.dim        : Named int [1:2] 4 4
.. ..@ bbox             : num [1:2, 1:2] 0.5 0.5 4.5 4.5
.. ..@ proj4string      : Formal class 'CRS' [package "sp"] with 1 slot
.. ..@ proj4args        : chr NA
```



Spatial data classes IV

- The **sp** class for polygon data is the **SpatialPolygonsDataFrame**.

```
> str(p,3)
Formal class 'SpatialPolygonsDataFrame' [package "sp"] with 5 slots
..@ data      : 'data.frame': 16 obs. of  2 variables:
.. ..$ Dominant_S: Factor w/ 16 levels "CMe","CMg","CMo",...: 1 2 3 4 5 6 7 8 9 10 ...
.. ..$ Soil_ag   : Factor w/ 10 levels "CM","FL","GG",...: 1 1 1 1 1 2 3 4 5 6 ...
..@ polygons  :List of 16
.. ..$ :Formal class 'Polygons' [package "sp"] with 5 slots
.. ..$ :Formal class 'Polygons' [package "sp"] with 5 slots
.. ..$ :Formal class 'Polygons' [package "sp"] with 5 slots
.. ..$ :Formal class 'Polygons' [package "sp"] with 5 slots
.. ..$ :Formal class 'Polygons' [package "sp"] with 5 slots
.. ..$ :Formal class 'Polygons' [package "sp"] with 5 slots
.. ..$ :Formal class 'Polygons' [package "sp"] with 5 slots
.. ..$ :Formal class 'Polygons' [package "sp"] with 5 slots
.. ..$ :Formal class 'Polygons' [package "sp"] with 5 slots
.. ..$ :Formal class 'Polygons' [package "sp"] with 5 slots
.. ..$ :Formal class 'Polygons' [package "sp"] with 5 slots
.. ..$ :Formal class 'Polygons' [package "sp"] with 5 slots
.. ..$ :Formal class 'Polygons' [package "sp"] with 5 slots
.. ..$ :Formal class 'Polygons' [package "sp"] with 5 slots
.. ..$ :Formal class 'Polygons' [package "sp"] with 5 slots
.. ..$ :Formal class 'Polygons' [package "sp"] with 5 slots
..@ plotOrder : int [1:16] 9 1 14 4 5 13 8 11 2 6 ...
..@ bbox      : num [1:2, 1:2] -177746 2915230 618131 3379062
.. ..- attr(*, "dimnames")=List of 2
..@ proj4string:Formal class 'CRS' [package "sp"] with 1 slot
```

Spatial data classes VI

- The **raster** package comes with its own class for raster data: **RasterLayer**.

```
> str(r,2)
Formal class 'RasterLayer' [package "raster"] with 12 slots
 ..@ file      :Formal class '.RasterFile' [package "raster"] with 13 slots
 ..@ data      :Formal class '.SingleLayerData' [package "raster"] with 13 slots
 ..@ legend    :Formal class '.RasterLegend' [package "raster"] with 5 slots
 ..@ title     : chr(0)
 ..@ extent    :Formal class 'Extent' [package "raster"] with 4 slots
 ..@ rotated   : logi FALSE
 ..@ rotation  :Formal class '.Rotation' [package "raster"] with 2 slots
 ..@ ncols     : int 2360
 ..@ nrows     : int 1611
 ..@ crs       :Formal class 'CRS' [package "sp"] with 1 slot
 ..@ history   : list()
 ..@ z         : list()
```

- Multiple layers: **RasterStack** / **RasterBrick**

Spatial data class summary

- Spatial data classes for packages:
 - **sp:**
 - SpatialPointsDataFrame
 - SpatialPixelDataFrame
 - SpatialGridDataFrame
 - SpatialPolygonDataFrame
 - SpatialLinesDataFrame
 - **raster:**
 - RasterLayer (single layer)
 - RasterStack / RasterBrick (multiple layers)

Importing spatial data

- rgdal: **readOGR** (vector), **readGDAL** (raster)
- raster: **raster**

```
setwd("D:/ISRIC/DSM_Nepal/Data/Covariates/")  
  
# polygon (rgdal package)  
p <- readOGR(dsn="./shape", layer = "soter_domsoil")  
  
# raster (rgdal package)  
r <- readGDAL(fname="./raster/ORCDRC_M_sl3_1km_11.tif")  
  
# raster (raster package)  
r <- raster("./raster/ORCDRC_M_sl3_1km_11.tif")
```


Exporting spatial data

- rgdal: **writeOGR** (vector), **writeGDAL** (raster)
- raster: **writeRaster**

```
# rgdal
writeOGR(d, dsn="D:/", layer="sampleSites", driver="ESRI Shapefile")
writeGDAL(r["SOC"], "./SOC_Nepal.tif", drivename = "Gtiff", type = "Int16", mvFlag = "-99999")

# raster
writeRaster(r, "SOC_Nepal2.tif", format="GTiff")
```

Projections

- Once you have loaded your spatial data in R, you might need to tell R its geographic projection.
- Check the current projection: **proj4string** function (sp package).
- Setting a projection: **CRS** function (sp package).
- Reprojecting to another coordinate system: **spTransform** function (sp package) or **projectRaster** (raster package).

Projections

```
> utm.proj <- "+proj=utm +zone=45 +north +ellps=WGS84 +datum=WGS84 +units=m +no_defs"
> wgs84.proj <- "+proj=longlat +ellps=WGS84 +datum=WGS84 +no_defs"
>
> d <- read.csv("./Sample/Processed/Soil.csv")
> coordinates(d) <- ~x+y
> proj4string(d)
[1] NA
> proj4string(d) <- CRS(utm.proj)
> proj4string(d)
[1] "+proj=utm +zone=45 +ellps=WGS84 +datum=WGS84 +units=m +no_defs +towgs84=0,0,0"
> head(d@coords)
      x      y
1 80393.8 3138118
2 49817.4 3140610
3 82986.8 3127692
4 67575.2 3146391
5 63857.7 3127477
6 90005.3 3149250
> d <- spTransform(d, CRS(wgs84.proj))
> proj4string(d)
[1] "+proj=longlat +ellps=WGS84 +datum=WGS84 +no_defs +towgs84=0,0,0"
> head(d@coords)
      x      y
1 82.72255 28.30223
2 82.41068 28.31454
3 82.75265 28.20920
4 82.58911 28.37253
5 82.55837 28.20109
6 82.81638 28.40548
```

Projections

```
> utm.proj <- "+proj=utm +zone=45 +north +ellps=WGS84 +datum=WGS84 +units=m +no_defs"
> wgs84.proj <- "+proj=longlat +ellps=WGS84 +datum=WGS84 +no_defs"
>
> wgs84.epsg <- "+init=epsg:4326"
> utm.epsg <- "+init=epsg:32645"
>
> d <- read.csv("./Sample/Processed/Soil.csv")
> coordinates(d) <- ~x+y
> proj4string(d) <- CRS(utm.epsg)
> proj4string(d)
[1] "+init=epsg:32645 +proj=utm +zone=45 +datum=WGS84 +units=m +no_defs +ellps=WGS84 +towgs84=0,0,0"
>
> d <- read.csv("./Sample/Processed/Soil.csv")
> coordinates(d) <- ~x+y
> proj4string(d) <- CRS(utm.proj)
> proj4string(d)
[1] "+proj=utm +zone=45 +ellps=WGS84 +datum=WGS84 +units=m +no_defs +towgs84=0,0,0"
> |
```

Find projection definitions at:

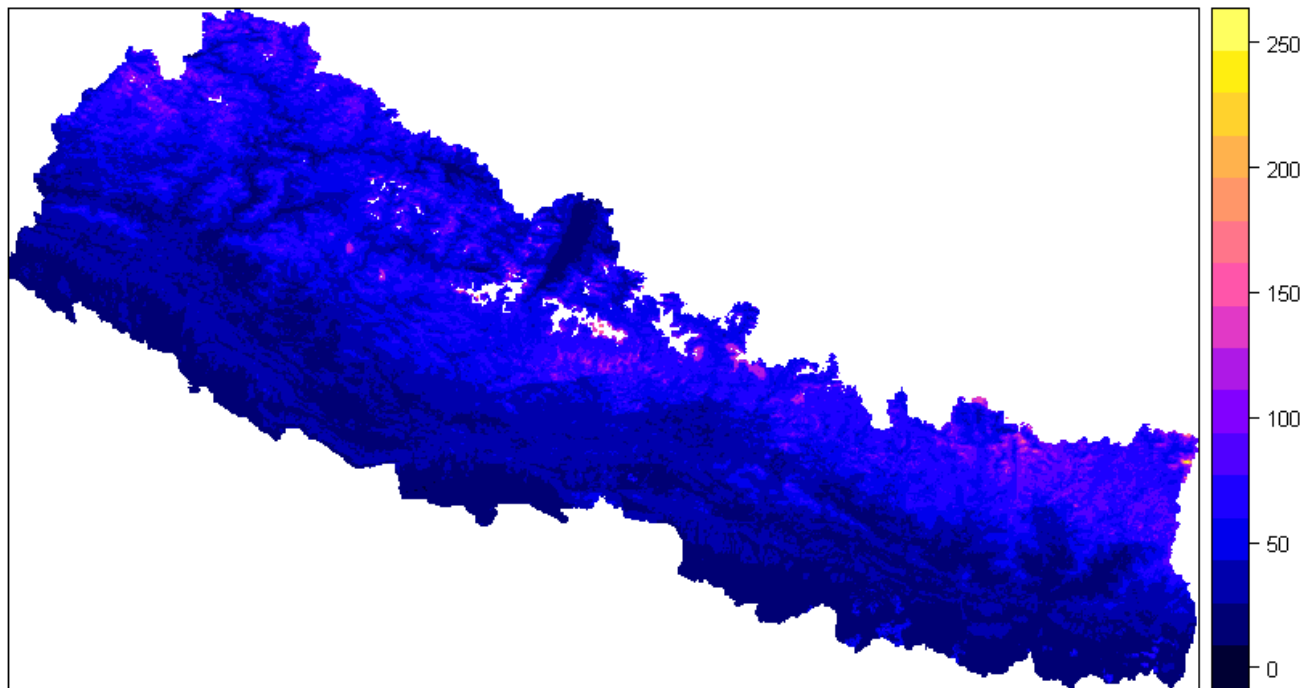
- <http://spatialreference.org/>
- <https://epsg.io/>

e.g. [UTM Zone 45 N](#)

Plotting

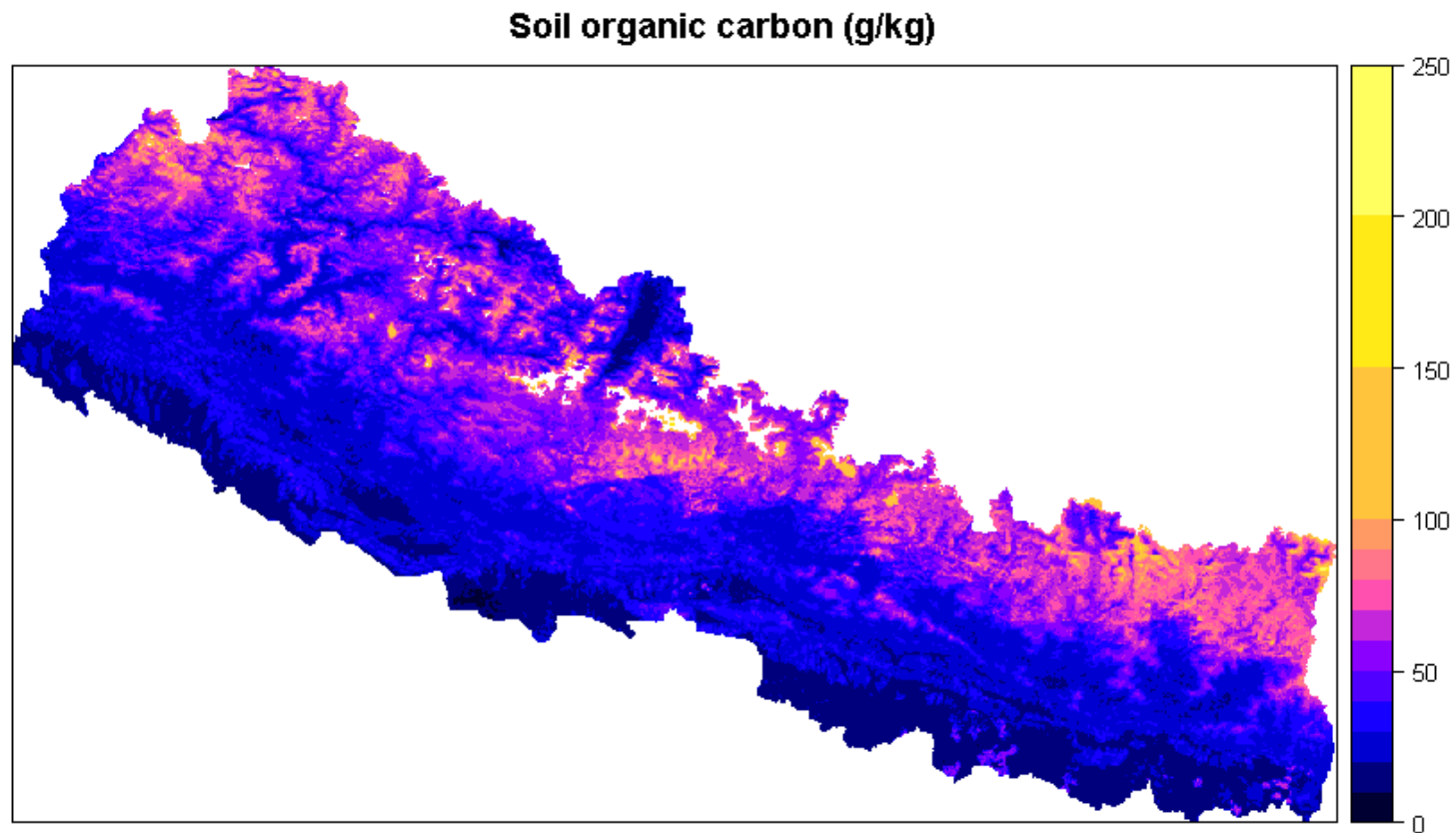
- sp package: **spplot**

```
spplot(r, zcol="SOC")
```



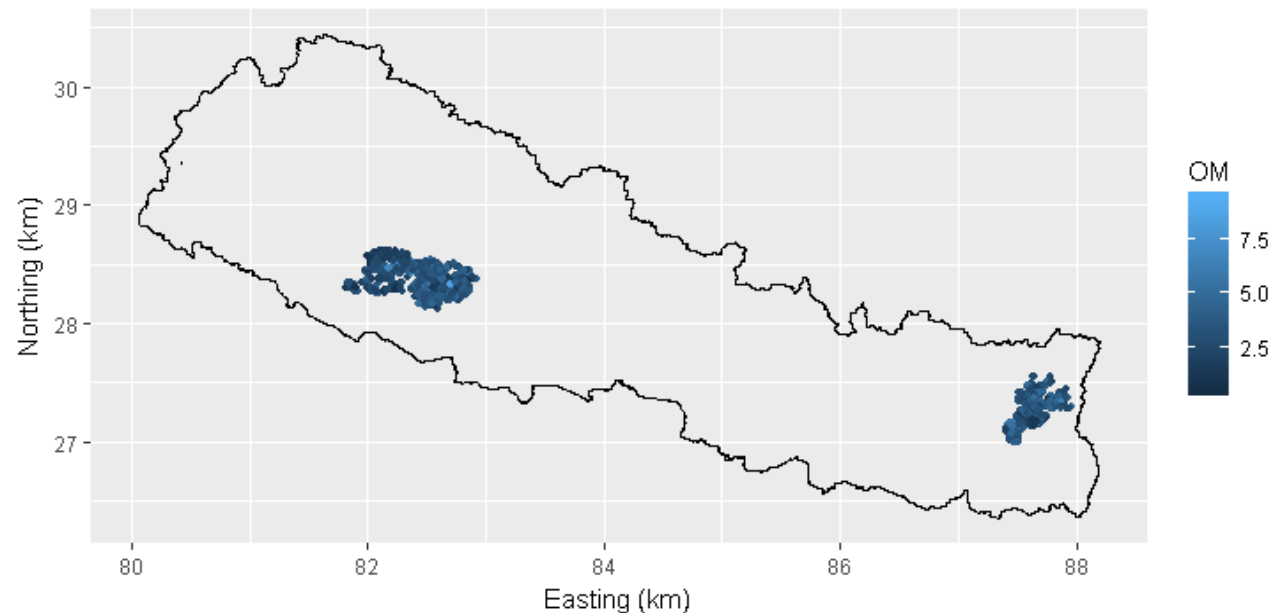
Plotting

```
ramp<-c(0,10,20,30,40,50,60,70,80,90,100,150,200,250)  
spplot(r, zcol = "SOC", at = ramp, main = "Soil organic carbon (g/kg)")
```



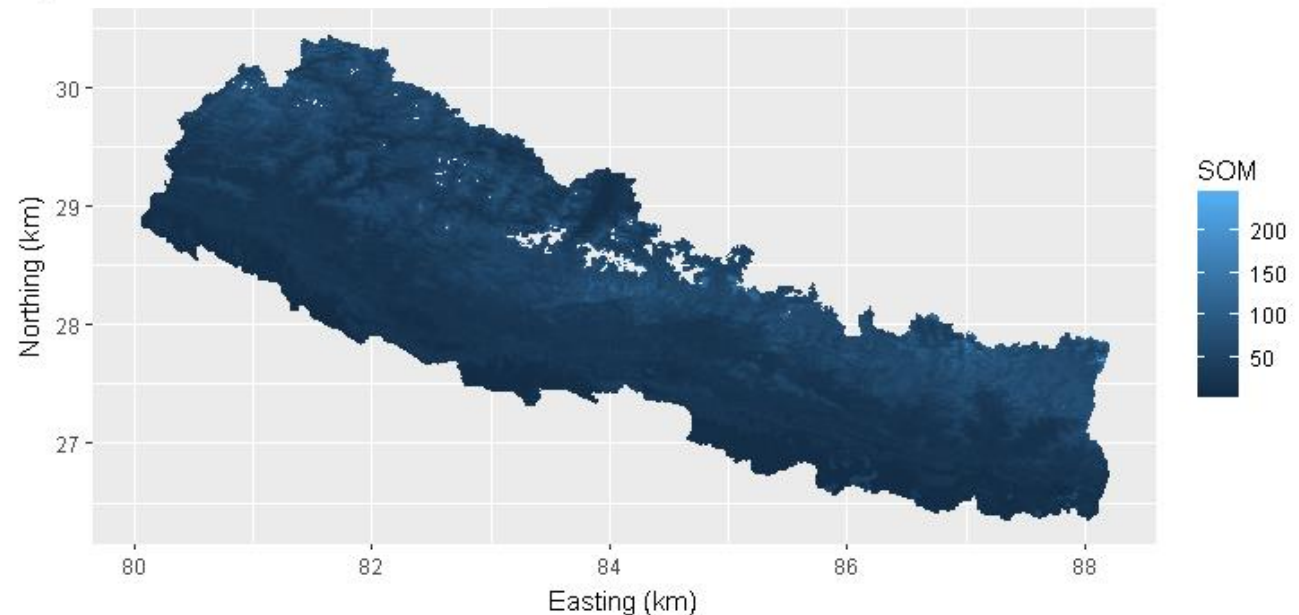
ggplot I

```
ggplot()+  
  geom_point(  
    data = d,  
    mapping = aes(x = x, y = y, colour = OM), size = 1  
  )+  
  geom_path(  
    data = boundary,  
    mapping = aes(x = long, y = lat, group = group)  
  )+  
  scale_x_continuous(name = "Easting (km)") +  
  scale_y_continuous(name = "Northing (km)") +  
  coord_equal(ratio = 1)
```



ggplot II

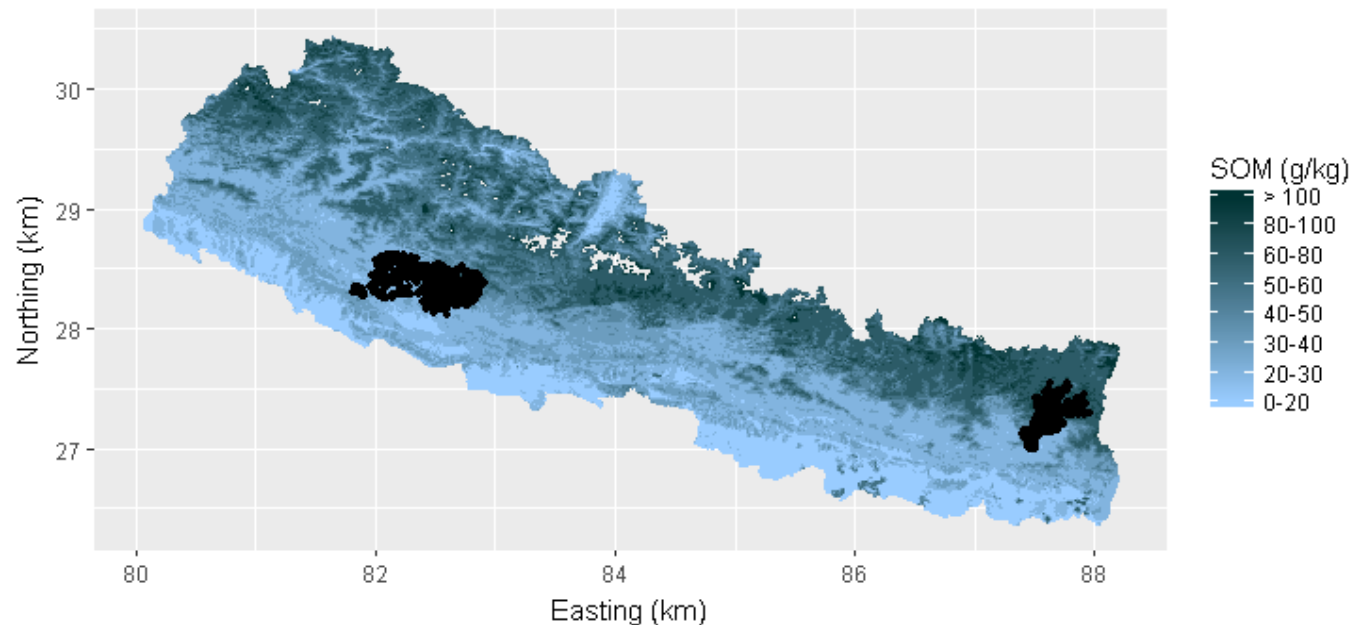
```
ggplot()+  
  geom_point(  
    data = d,  
    mapping = aes(x = x, y = y), size = 1  
  )+  
  geom_raster(  
    data = r,  
    mapping = aes(x = x, y = y, fill = SOM)  
  )+  
  scale_x_continuous(name = "Easting (km)") +  
  scale_y_continuous(name = "Northing (km)") +  
  coord_equal(ratio = 1)
```



ggplot III

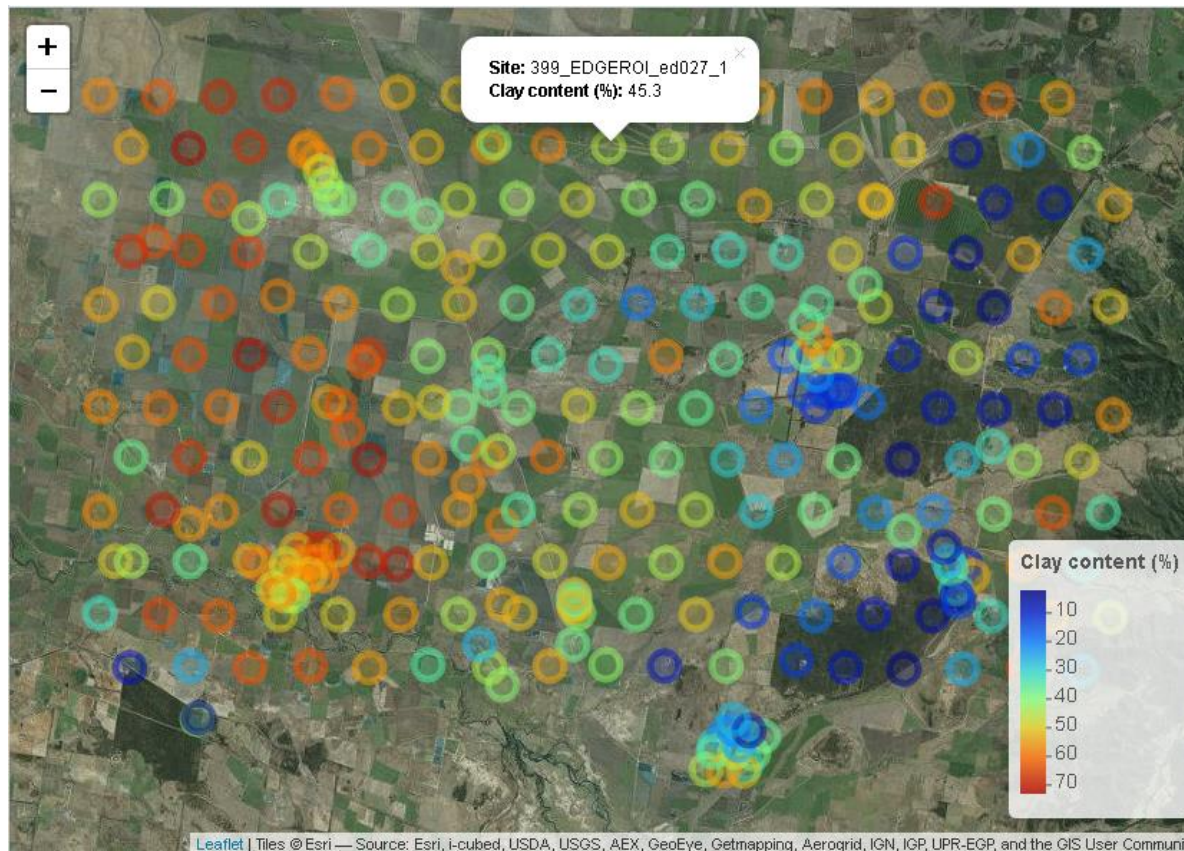
```
lab <- c("0-20", "20-30", "30-40", "40-50", "50-60", "60-80", "80-100", "> 100")
```

```
ggplot()+  
  geom_raster(  
    data = r,  
    mapping = aes(x = x, y = y, fill = klasse)  
  )+  
  geom_point(  
    data = d,  
    mapping = aes(x = x, y = y), size = 1  
  )+  
  scale_fill_continuous(name = "SOM (g/kg)", low = "#99CCFF", high = "#003333", breaks = c(1:8), labels = lab)+  
  scale_x_continuous(name = "Easting (km)")+  
  scale_y_continuous(name = "Northing (km)")+  
  coord_equal(ratio = 1)
```



Interactive maps

- Interactive maps can be generated with the **leaflet** package.



Resources

- [R tutor](#)
- [Rwiki](#)
- [Advanced R](#)
- [R for Data Science](#)
- [Making maps in R](#)
- [ggplot cheat sheet](#)

Problem solving:

- [StackOverflow](#)
- [R-FAQ](#)



Now lets practice. Have fun!!