

INTRODUCTION TO SPATIAL DATA ANALYSIS

WORKING WITH RASTER DATA IN R

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COURSE MATERIALS

<https://exeter-data-analytics.github.io/>

Exeter Data Analytics Hub

Site for workshops organised at the University of Exeter

About

The Exeter Data Analytics Hub is a team of academics based at the University of Exeter across Exeter and Penryn campuses, who offer a range of workshops in the field of statistics, data science, machine learning, programming and more. We provide technical and analytical support for early career researchers based in at the University of Exeter.

Workshops

Penryn (partly funded by the Doctoral College)

- Python for scientific research
- Introduction to R
- Advanced Visualisation and Data Wrangling in R [Data] [Slides]
- Open Science and Reproducible Research in R
- R Scripting and R Markdown [Data and scripts]
- Statistical Modelling in R [Data] [Slides]
- Machine Learning [Data] [Slides]
- Spatial Data Analysis

Streatham

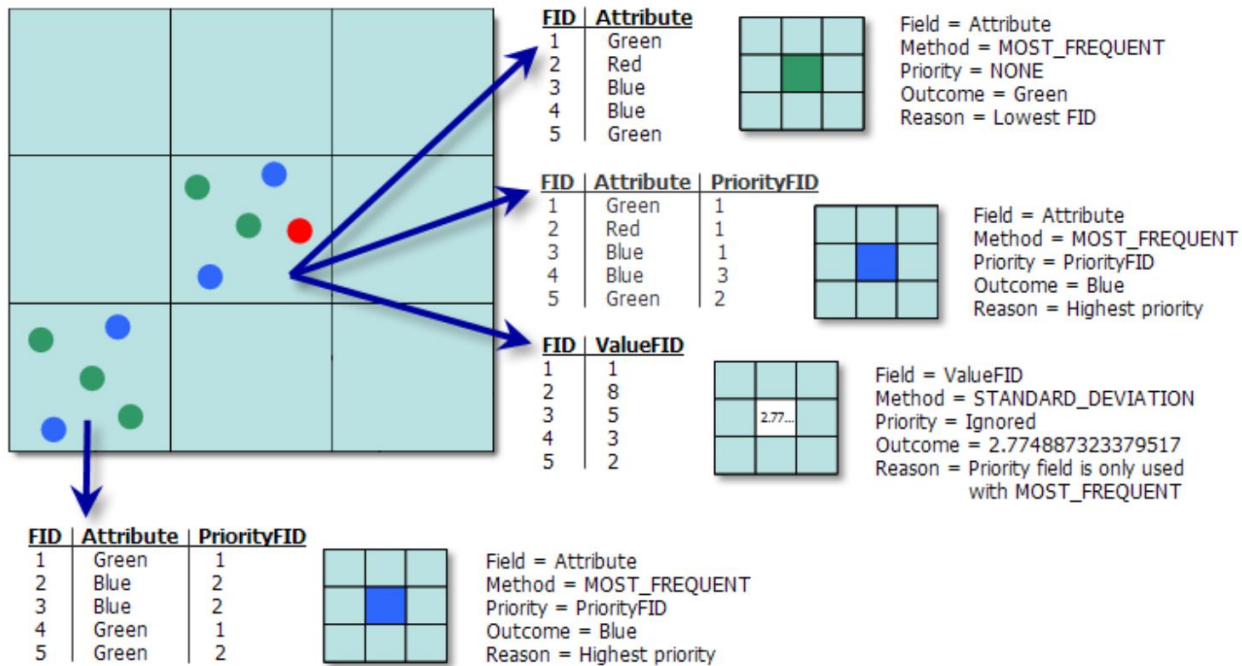
- ImageJ GUI
- ImageJ Macros
- Introduction to Python
- Python for Data Analysis

REAL WORLD TO RASTER MODEL

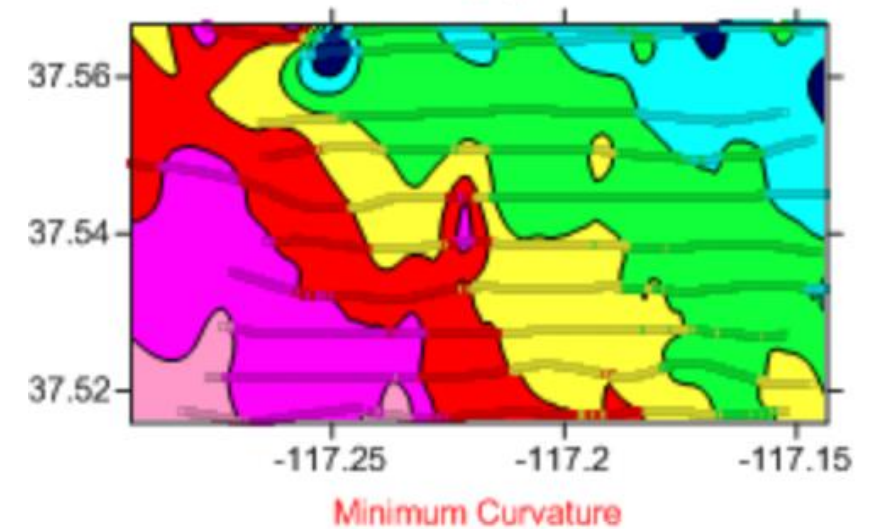


RASTER DATA COLLECTION

Cell assignment



Gridding/Interpolation/Spline



Picture Elements or pixels



LEARNING BY HANDS-ON PRACTICAL SESSIONS

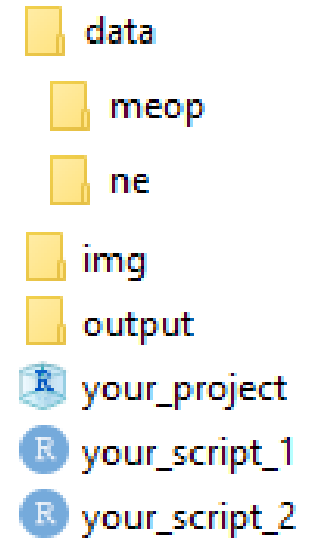


Contents

- Import spatial data in *R* (raster)
- Generating synthetic data
- Real examples
- Raster operations
- Raster manipulation
- Visualize maps
- Save data
- Time to explore other methods or your own data

SET UP YOUR PROJECT

1. Create a new project from Rstudio
2. Download the Data
3. Unzip and store its contents in a folder named `data/` within your new project.
4. Create folders `output/` and `img/`
5. Install all required packages



RASTER DATA FORMATS

Getting raster data in is easy and can be done using the following commands

```
r1 <- raster()  
s1 <- stack()
```



The `raster()` tool is used for single layer files and the `stack()` tool for raster datasets with multiple layers, or for combining single layer rasters into a multilayer raster.

IMPORT RASTER DATA

4.2.1 Importing a raster layer

A single raster layer is easily imported. Here we load an example from the system repository.

```
r1 <- raster(system.file("external/test.grd", package="raster"))
```


IMPORT RASTER DATA

4.2.2 Importing a raster stack

Rasters can be stacked and this is particularly useful for RGB layers in a raster. For example, we can illustrate this using the R logo:

```
sRLogo <- stack(system.file("external/rlogo.grd", package="raster"))  
nlayers(sRLogo)
```

PLOTTING RASTER DATA QUICKLY

There are three layers in the `sRLogo` representing the red, green and blue channels (RGB). These can be plotted individually.

```
image(sRLogo, y = 1)
```

```
image(sRLogo$green)
```

```
image(sRLogo[[3]])
```

PLOTTING RASTER DATA QUICKLY

These data may be more useful if plotted as an RGB plot. R isn't great at this but for a quick visualisation we can use `plotRGB` :

```
plotRGB(sRLogo, stretch = "lin")
```



There are few tricks to `plotRGB` worth exploring such as changing the stretch:

```
plotRGB(sRLogo, stretch = "hist")
```

And changing the order of the bands - useful when RGB are not in the correct order:

PLOTTING RASTER DATA QUICKLY

4.2.2 Importing a raster stack

Rasters can be stacked and this is particularly useful for RGB layers in a raster. For example, we can illustrate this using the R logo:

```
sRLogo <- stack(system.file("external/rlogo.grd", package="raster"))  
nlayers(sRLogo)
```


CREATING SYNTHETIC DATA

Sometimes it can be useful to generate a random synthetic dataset. This could be for adding random noise to a dataset or for simple testing.

```
rSynth <- raster()  
set.seed(0)  
values(rSynth) <- runif(ncell(rSynth), 0, 100)
```

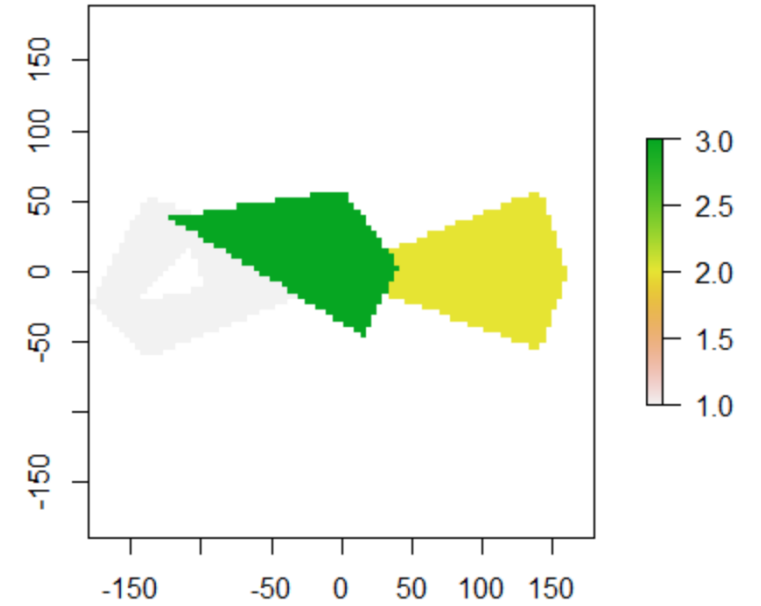
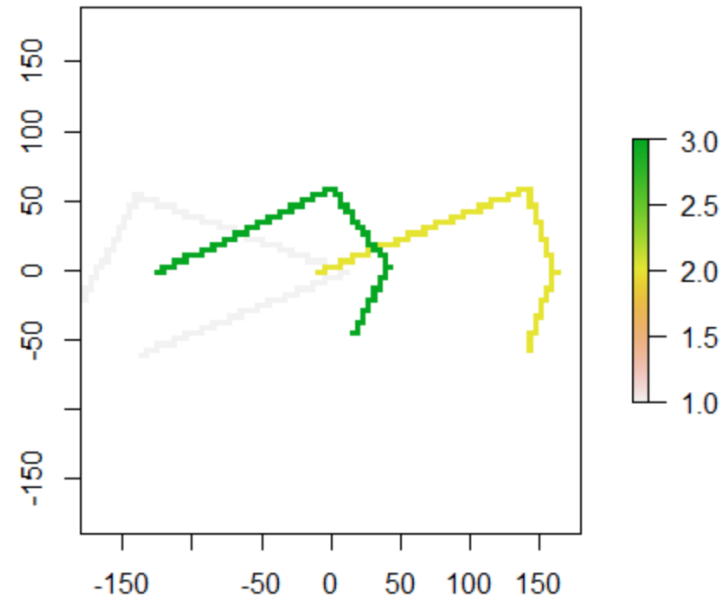
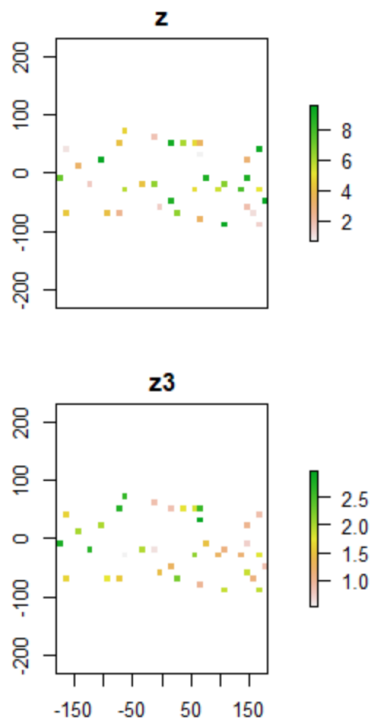
The success of this can be checked in a number of ways:

```
hasValues(rSynth)  
length(values(rSynth))  
ncell(rSynth)
```

Similarly, a synthetic raster stack can be generated by simply manipulating the:

```
sSynth <- stack(rSynth, rSynth*2, sqrt(rSynth))  
plot(sSynth[[2]])
```

RASTERISING VECTOR DATA



REAL EXAMPLES — ELEVATION DATA

```
cwl <- readOGR("data/ukdata/england_ct_2001.shp")
```

```
## OGR data source with driver: ESRI Shapefile  
## Source: "C:\Git\spatial_data_analysis\data\ukdata\england_ct_2001.shp", layer: "england_ct_2001"  
## with 18 features  
## It has 2 fields
```

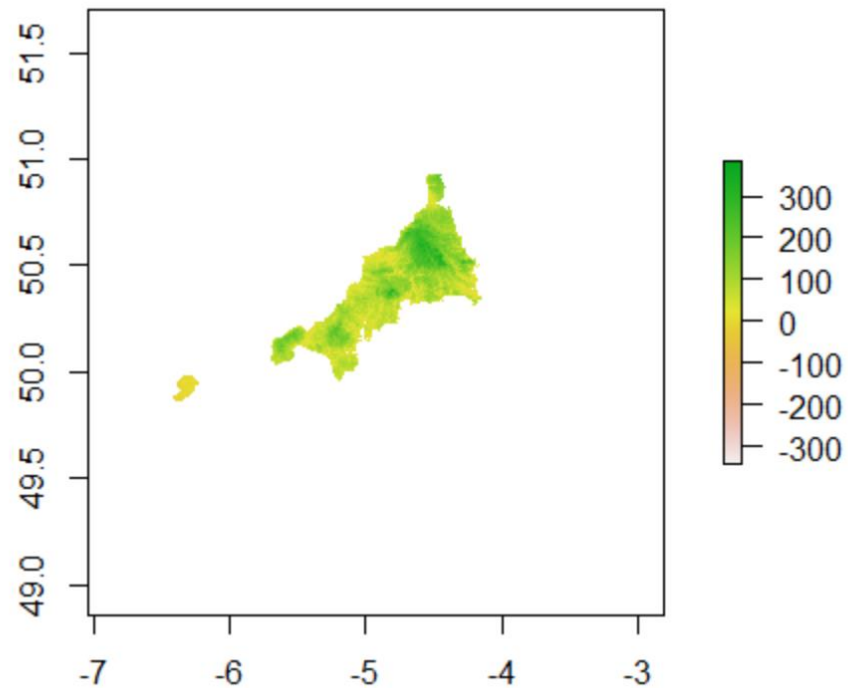
```
wgs84 = '+proj=longlat +datum=WGS84'  
cwl_WGS84 <- spTransform(cwl, CRS(wgs84))
```

The elevation data is accessed using the `elevatr` package. more details can be found [here](#)

```
elev_cwl_WGS84 <- get_elev_raster(cwl_WGS84, z=9)
```

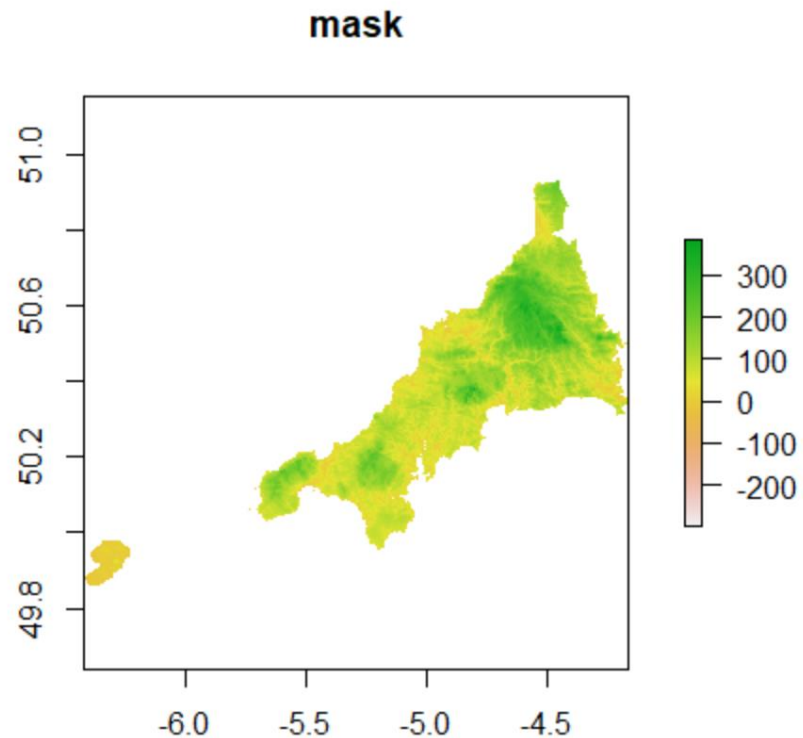
RASTER OPERATIONS — MASK

```
mask_elev <- mask(elev_cwl_WGS84, cwl_WGS84)  
plot(mask_elev)
```



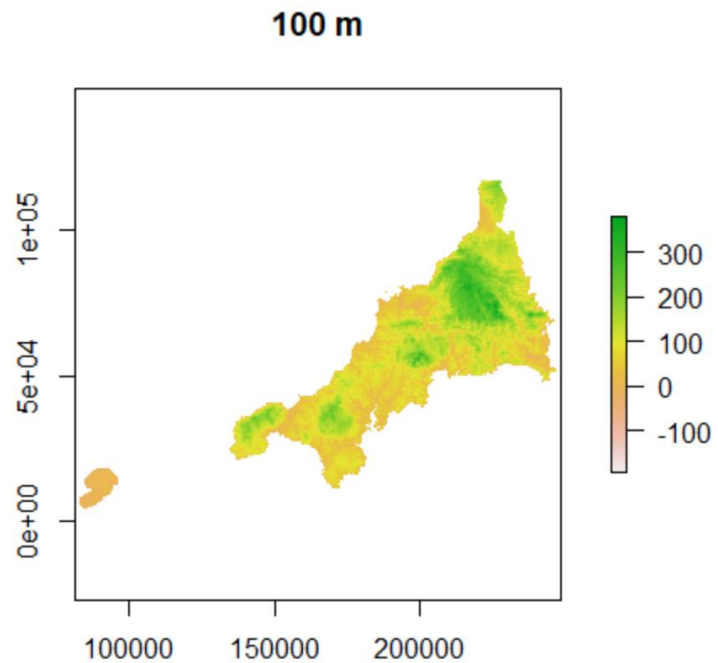
RASTER OPERATIONS — CROP

```
e <- extent(cwl_WGS84)
crop_mask_elev <- crop(mask_elev, e)
plot(crop_mask_elev, main="mask")
```



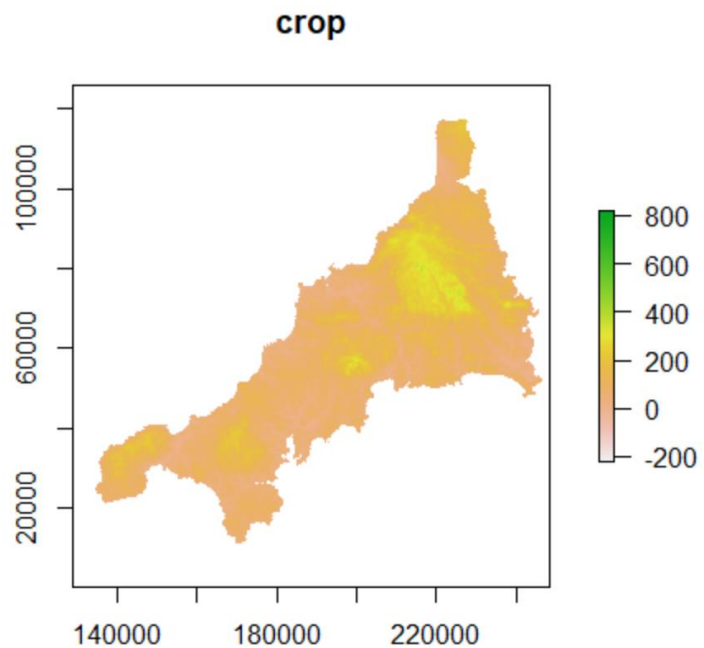
RASTER OPERATIONS — RESAMPLE

```
tmpR <- raster(nrow=1271,ncol=1669)
crs(tmpR) <- crs(cwl)
extent(tmpR) <- c(81600,248500,-2600,124500)
elev_100m <- resample(elev_cwl_BNG, tmpR)
plot(elev_100m,main="100 m")
```



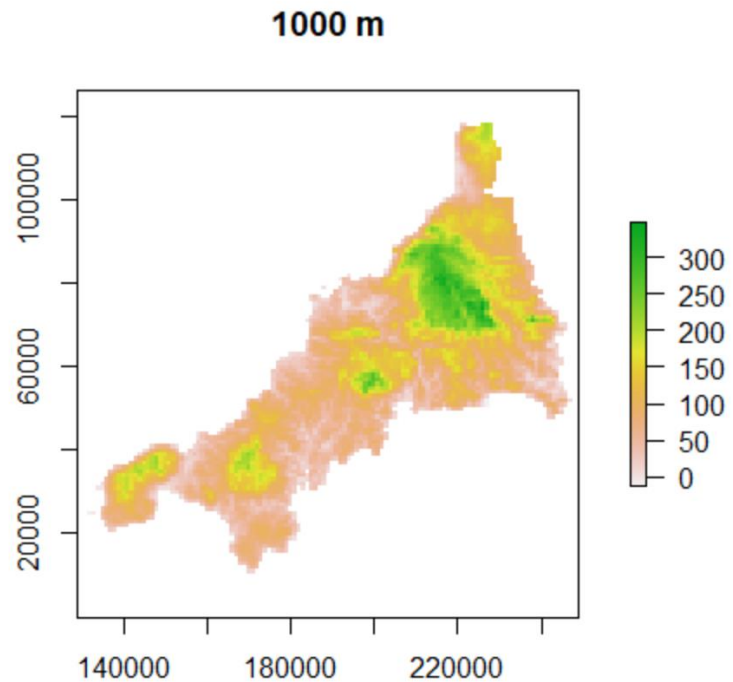
RASTER OPERATIONS — EXTRACT

```
# For example you could try this... but give the locator() tool a go
e_crop <- c(129000, 248500, 1600, 124500) # xmin, xmax, ymin, ymax
# Now crop
crop_elev_100m <- crop(elev_100m, e_crop)
plot(crop_elev_100m, main="crop")
```



RASTER OPERATIONS — AGGREGATE

```
### Aggregate to a lower res raster  
elev_1000m <- aggregate(crop_elev_100m, 10, fun=mean, expand=TRUE, na.rm=TRUE)  
plot(elev_1000m, main = "1000 m")
```



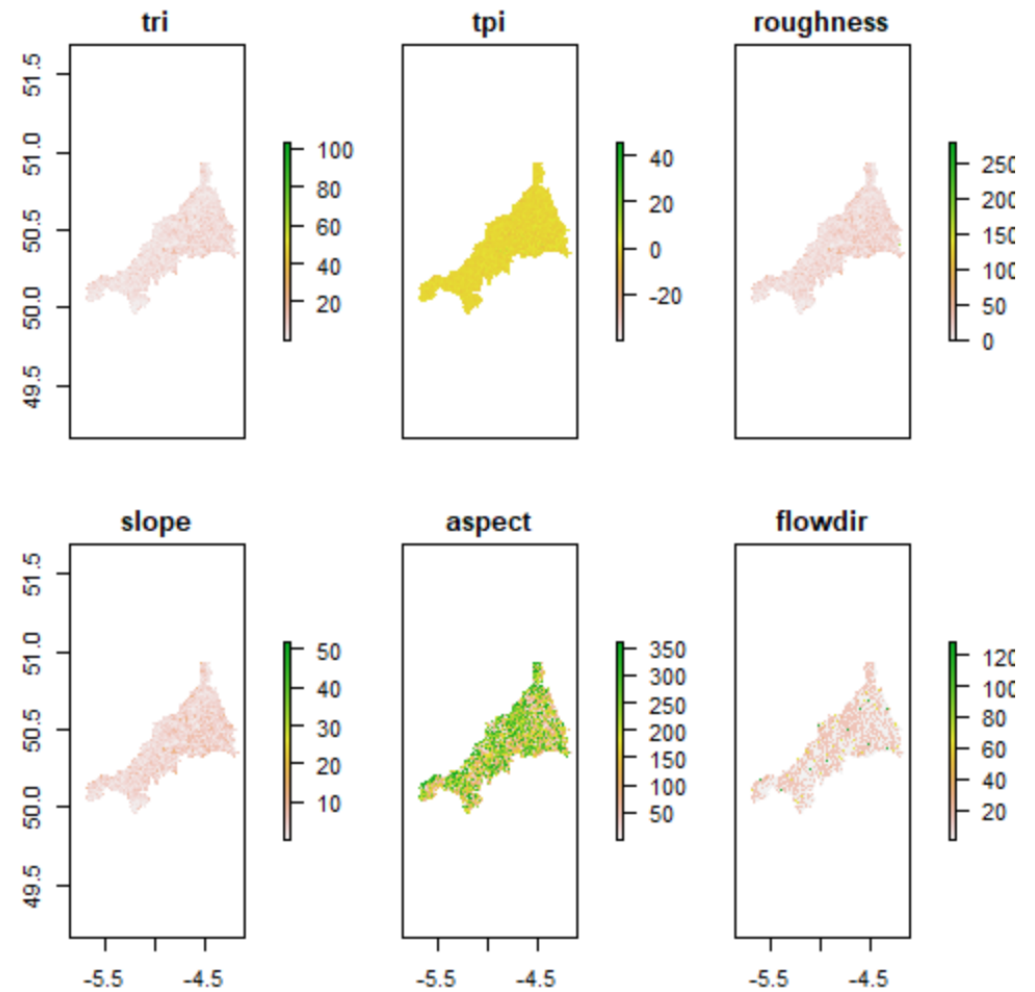
RASTER MANIPULATION

The `cellStats` function is useful for finding broad statistics about a layer e.g. sum, mean, min, max, sd, 'skew' and 'rms' of which the latter two must be supplied as a character value (with quotes)

```
cellStats(elev_100m_WGS84, stat='mean')
```

RASTER MANIPULATION

```
models <- terrain(elev_100m_WGS84, opt="M",  
class=models)
```



RASTER MANIPULATION — HILLSHADE

We can also make a hillshade raster and overlay the elevation data. We introduce a way of subsetting the raster stack/brick here using `subset` but you could also use `slope <- models$slope` or other means we introduced at the beginning of this section.

```
slope <- subset(models, "slope") # select slope
aspect <- subset(models, "aspect") # select aspect
hill <- hillShade(slope, aspect, 45, 270)
plot(hill, col = grey(0:100/100), legend = FALSE)
plot(elev_100m_WGS84, col = rainbow(25, alpha=0.35), add=TRUE)
```

RASTER VISUALISATION

Maps can be quickly created using `leaflet` and basemaps from OpenStreetMap or tiles from MapBox imagery can be imported to enhance the map.

```
pal <- colorNumeric(c("#FFFFCC", "#41B6C4", "#0C2C84"), domain=c(0,500, na.rm=T),
                    na.color = "transparent", reverse = T)

leaflet() %>%
  addTiles() %>% # OpenStreetMap base map
  addRasterImage(elev_100m_WGS84, colors = pal, opacity = 0.8) %>%
  addLegend(pal = pal, values = c(0, 500),
            title = "Cornwall Elevation")
```


SAVING RASTER FILES

Now we can write our rasters as layers, stacks or write layers from stacks...

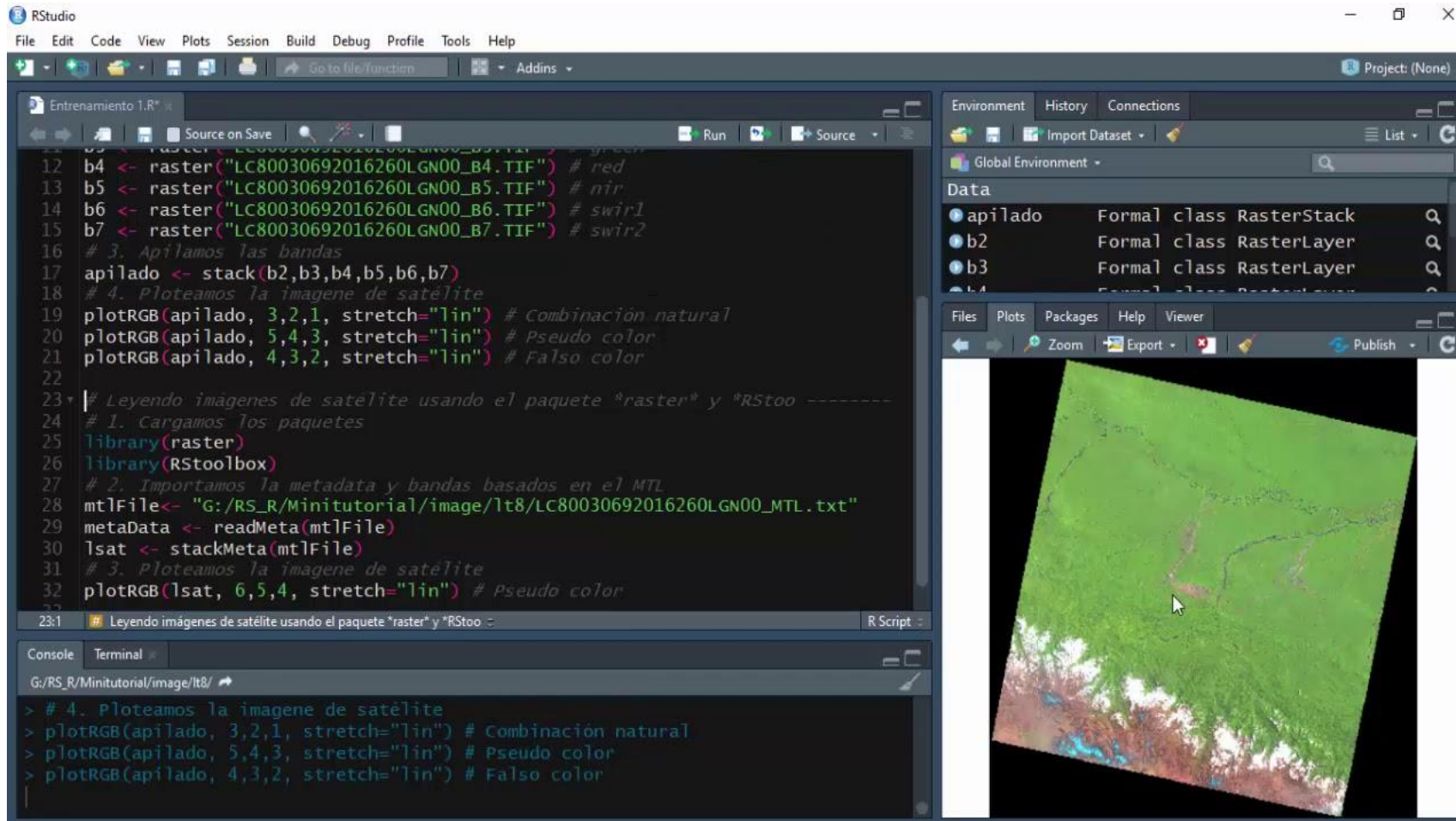
```
# A raster layer
writeRaster(slope, filename="output/slope.tif", format="GTiff", overwrite=TRUE)
# A layer from a raster stack/brick
writeRaster(models[[1]], filename="output/tri", format="GTiff", overwrite=TRUE)
# A whole stack/brick
writeRaster(models, filename="output/models", format="GTiff", overwrite=TRUE)
# A whole raster stack/brick by layer
writeRaster(models, filename=paste("output/", names(models), sep = ""), format="GTiff", bylayer=T, overwrite=TRUE)
```

Or you can write it to KML with a set colour palette

```
myPal <- colorRampPalette(brewer.pal(9, 'Blues'), alpha=TRUE) # palette
KML(elev_100m_WGS84, "output/elev_100m_WGS84.kml", col = myPal(100), overwrite = TRUE)
```


ADDITIONAL APPLICATIONS

REMOTE SENSING



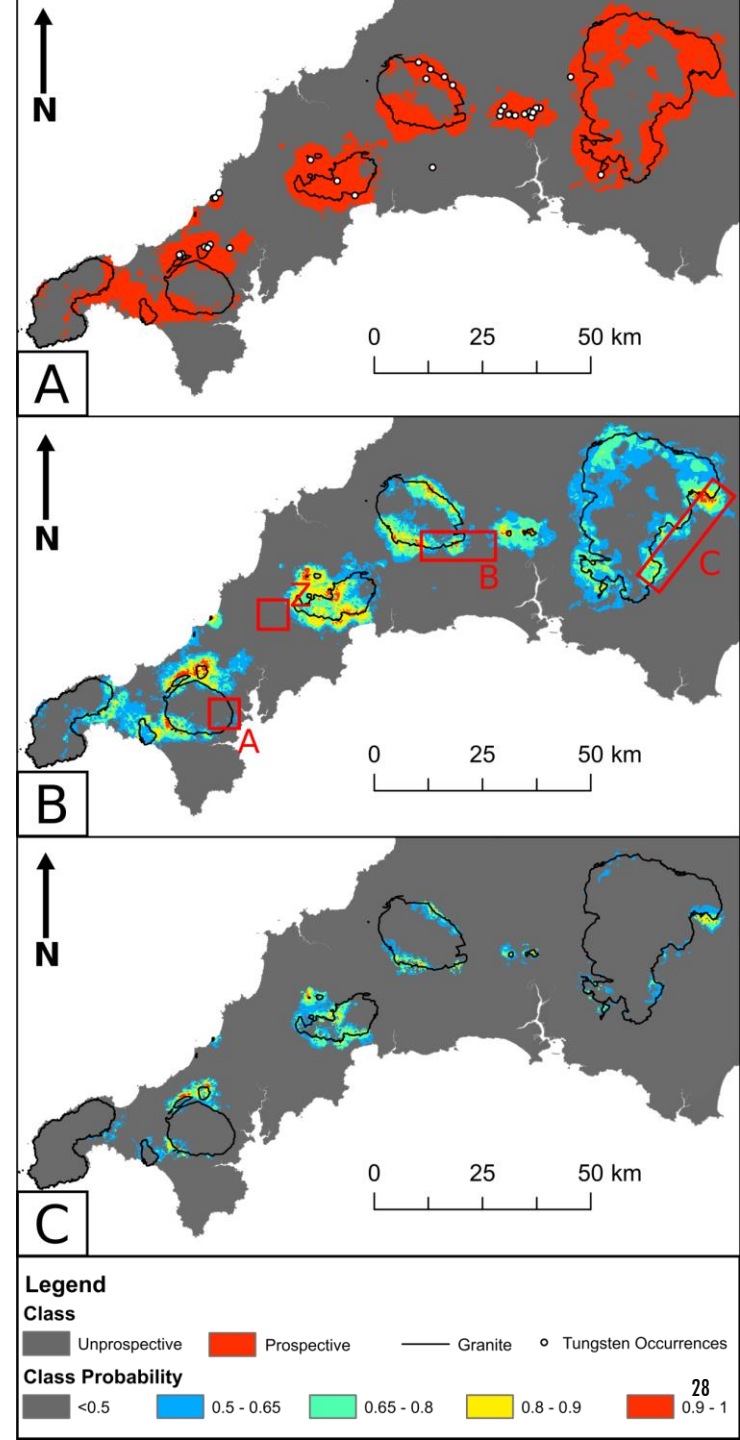
```
Entrenamiento 1.R  
12 b4 <- raster("LC80030692016260LGN00_B4.TIF") # red  
13 b5 <- raster("LC80030692016260LGN00_B5.TIF") # nir  
14 b6 <- raster("LC80030692016260LGN00_B6.TIF") # swir1  
15 b7 <- raster("LC80030692016260LGN00_B7.TIF") # swir2  
16 # 3. Apilamos las bandas  
17 apilado <- stack(b2,b3,b4,b5,b6,b7)  
18 # 4. Ploteamos la imagen de satélite  
19 plotRGB(apilado, 3,2,1, stretch="lin") # Combinación natural  
20 plotRGB(apilado, 5,4,3, stretch="lin") # Pseudo color  
21 plotRGB(apilado, 4,3,2, stretch="lin") # Falso color  
22  
23 # Leyendo imágenes de satélite usando el paquete "raster" y "RStoo -----  
24 # 1. Cargamos los paquetes  
25 library(raster)  
26 library(RStoolbox)  
27 # 2. Importamos la metadata y bandas basados en el MTL  
28 mtlFile <- "G:/RS_R/Minitutorial/image/lt8/LC80030692016260LGN00_MTL.txt"  
29 metaData <- readMeta(mtlFile)  
30 lsat <- stackMeta(mtlFile)  
31 # 3. Ploteamos la imagen de satélite  
32 plotRGB(lsat, 6,5,4, stretch="lin") # Pseudo color  
33  
23:1 Leyendo imágenes de satélite usando el paquete "raster" y "RStoo - R Script -  
Console Terminal  
G:/RS_R/Minitutorial/image/lt8/  
> # 4. Ploteamos la imagen de satélite  
> plotRGB(apilado, 3,2,1, stretch="lin") # Combinación natural  
> plotRGB(apilado, 5,4,3, stretch="lin") # Pseudo color  
> plotRGB(apilado, 4,3,2, stretch="lin") # Falso color
```

Landsat in R

Sentinel-2 in R

MLA APPLICATIONS

```
#### Random Forest Prospectivity ####
#### Manipulate to training data to dataframe
df.Wtr <- as.data.frame(Wtr@data)
# Change response col name to match training function which is "class"
colnames(df.Wtr) <- "class"
# Check
View(df.Wtr)
# Stick it back into the spatial points file
Wtr@data <- df.Wtr
#### Random Forest Model Training
# Model for classification
mod1 <- rasterRFmodel(Wr.mg, Wtr, cp="raw", imp=T, ntree=20000)
# Model for Probability
mod2 <- rasterRFmodel(Wr.mg, Wtr, cp="prob", imp=T, ntree=20000)
#### Random Forest Prediction
# Predict model for classification
rfp1.mod1 <- rasterRFPredict(r=Wr.mg, mod=mod1, cp="raw", imp=T, ntree=20000)
# For some reason the script won't write this to raster layer, only to a brick
# so convert to raster layer
rfp1.mod1 <- rfp1.mod1[[1]]
# Predict model for probability
rfp1.mod2 <- rasterRFPredict(r=Wr.mg, mod=mod2, cp="prob", imp=T, ntree=20000)
#### Model Confidence for prospective class (Class 2)
uncert.mod2 <- rasterUncert(rfp1.mod2)
rfconf <- rfp1.mod2[[2]] - uncert.mod2/(nlayers(rfp1.mod2))
rfconfnorm <- (rfconf - cellStats(rfconf, "min")) /
              (cellStats(rfconf, "max") - cellStats(rfconf, "min"))
plot(rfconfnorm, main = "RF confidence")
```



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- DM and JD are thanked for creating the Spatial Data Analytics repo

