

10-multispectral\_sensor

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# Contents

<b>1</b>	<b>About</b>	<b>5</b>
1.1	.....	5
<b>2</b>		<b>7</b>
<b>3</b>	<b>(PART*) Base R</b>	<b>9</b>
<b>4</b>	<b>Base R you have to know</b>	<b>11</b>
4.1	.....	11
4.2	R? .....	11
4.3	Vector( ) .....	12
4.4	Lists( ) .....	14
4.5	Matrices( ) .....	14
4.6	Data Frame( ) .....	18
	<b>All about tidyverse</b>	<b>21</b>
4.7	.....	21
4.8	ggplot2 .....	21
4.9	Reference links .....	21
<b>5</b>	<b>(PART*) bookdown</b>	<b>23</b>
	<b>how to write a book</b>	<b>25</b>

<b>6</b>	<b>Raster</b>	<b>27</b>
6.1	Resolution . . . . .	27
6.2	CRS . . . . .	29
6.3	Manipulation in R . . . . .	29
6.4	Reference . . . . .	29
<b>7</b>	<b>Multispectral sensors</b>	<b>31</b>

# Chapter 1

# About

shuxin R Arcgis Qgis ENVI SNAP

R

githu

, ,

emo

## 1.1



## Chapter 2

shuxin R





## Chapter 3

### (PART\*) Base R



# Chapter 4

## Base R you have to know

### 4.1

### 4.2 R?

1992                  Ross Ihaka Robert Gentleman



Figure 4.1: *Ross Ihaka and Robert Gentleman, the creators of R.*

R                          R                          R                          R                          The

R Base Package

B R Python stata R

### 4.3 Vector( )

Vector R

1 5 5

```
x <- c(1,2,3,4,5)
x
#> [1] 1 2 3 4 5
```

c() 1 2 3 4 5 <- x 1 2 3 4 5 x R  
R c() google :, R

```
x <- c(1:5)
x
#> [1] 1 2 3 4 5
```

vector typeof()

```
typeof(x)
#> [1] "integer"
```

length,length()

```
length(x)
#> [1] 5
```

R seq()

```
seq(1, 9, 0.5)
#> [1] 1.0 1.5 2.0 2.5 3.0 3.5 4.0 4.5 5.0 5.5 6.0 6.5 7.0 7.5
#> [15] 8.0 8.5 9.0
```

1 9 0.5 3 ?? Console  
??seq  
Help :: :: package::function cli an  
generation  
Description Usage,seq(...) seq (), A  
vector

```
# Vector of logical values
log_values <- c(TRUE, FALSE, TRUE, FALSE)

log_values
#> [1] TRUE FALSE TRUE FALSE
```

# R

```
fruits <- c("banana", "apple", "orange", "mango", "lemon", "50")
fruits
#> [1] "banana" "apple" "orange" "mango" "lemon" "50"
```

[] brackets, fruits "banana" "mango"

```
fruits[c(1,4)]
#> [1] "banana" "mango"
```

```
fruits[1:4]
#> [1] "banana" "apple" "orange" "mango"
```

"banana"

```
fruits[-1]
#> [1] "apple" "orange" "mango" "lemon" "50"
```

sort,

```
fruits <- c("banana", "apple", "orange", "mango", "lemon")
numbers <- c(13, 3, 5, 7, 20, 2)

sort(fruits) # Sort a string
#> [1] "apple" "banana" "lemon" "mango" "orange"
sort(numbers) # Sort numbers
#> [1] 2 3 5 7 13 20
```

## 4.4 Lists( )

R list()

```
thislist <- list(
  a = c("apple", "banana", "cherry"),
  b = c(1,2,5,6,7,9),
  c = c(TRUE, FALSE, TRUE)
)
# Print the list
thislist
#> $a
#> [1] "apple" "banana" "cherry"
#>
#> $b
#> [1] 1 2 5 6 7 9
#>
#> $c
#> [1] TRUE FALSE TRUE
```

```
typeof(thislist)
#> [1] "list"
```

```
length(thislist)
#> [1] 3
```

## 4.5 Matrices( )

(column) (row) matrix()

```
# Create a matrix
thismatrix <- matrix(c(1,2,3,4,5,6), nrow = 3, ncol = 2)

# Print the matrix
thismatrix
#>      [,1] [,2]
#> [1,]    1    4
#> [2,]    2    5
#> [3,]    3    6
```

NOTE: `c()`

```
thismatrix <- matrix(c("apple", "banana", "cherry", "orange"), nrow = 2, ncol = 2)

thismatrix
#>      [,1]      [,2]
#> [1,] "apple"  "cherry"
#> [2,] "banana" "orange"
```

Access Matrix Items You can access the items by using `[ ]` brackets. The first number “1” in the bracket specifies the row-position, while the second number “2” specifies the column-position:

```
thismatrix <- matrix(c("apple", "banana", "cherry", "orange"), nrow = 2, ncol = 2)

thismatrix[1, 2]
#> [1] "cherry"
```

The whole row can be accessed if you specify a comma after the number in the bracket:

```
thismatrix <- matrix(c("apple", "banana", "cherry", "orange"), nrow = 2, ncol = 2)

thismatrix[2,]
#> [1] "banana" "orange"
```

The whole column can be accessed if you specify a comma before the number in the bracket:

```
thismatrix <- matrix(c("apple", "banana", "cherry", "orange"), nrow = 2, ncol = 2)

thismatrix[,2]
#> [1] "cherry" "orange"
```

Access More Than One Row More than one row can be accessed if you use the `c()` function:

```
thismatrix <- matrix(c("apple", "banana", "cherry", "orange", "grape", "pineapple", "pear", "melon"),
nrow = 2, ncol = 4)

thismatrix[c(1,2),]
#>      [,1]      [,2]      [,3]      [,4]
#> [1,] "apple"  "orange" "pear"   "melon"
#> [2,] "banana" "grape"  "pineapple" "cherry"
```

Access More Than One Column More than one column can be accessed if you use the `c()` function:

```
thismatrix <- matrix(c("apple", "banana", "cherry", "orange", "grape", "pineapple", "pear"), nrow=3, byrow=TRUE)

thismatrix[, c(1,2)]
#>      [,1]      [,2]
#> [1,] "apple"  "orange"
#> [2,] "banana" "grape"
#> [3,] "cherry" "pineapple"
```

Add Rows and Columns Use the `cbind()` function to add additional columns in a Matrix:

```
thismatrix <- matrix(c("apple", "banana", "cherry", "orange", "grape", "pineapple", "pear"), nrow=3, byrow=TRUE)

newmatrix <- cbind(thismatrix, c("strawberry", "blueberry", "raspberry"))

# Print the new matrix
newmatrix
#>      [,1]      [,2]      [,3]      [,4]
#> [1,] "apple"  "orange"  "pear"   "strawberry"
#> [2,] "banana" "grape"   "melon" "blueberry"
#> [3,] "cherry" "pineapple" "fig"   "raspberry"
```

Use the `rbind()` function to add additional rows in a Matrix:

```
thismatrix <- matrix(c("apple", "banana", "cherry", "orange", "grape", "pineapple", "pear"), nrow=3, byrow=TRUE)

newmatrix <- rbind(thismatrix, c("strawberry", "blueberry", "raspberry"))

# Print the new matrix
newmatrix
#>      [,1]      [,2]      [,3]
#> [1,] "apple"  "orange"  "pear"
#> [2,] "banana" "grape"   "melon"
#> [3,] "cherry" "pineapple" "fig"
#> [4,] "strawberry" "blueberry" "raspberry"
```

Remove Rows and Columns Use the `c()` function to remove rows and columns in a Matrix:

```
thismatrix <- matrix(c("apple", "banana", "cherry", "orange", "mango", "pineapple"), nrow=3, byrow=TRUE)
```



```
#Remove the first row and the first column
thismatrix <- thismatrix[-c(1), -c(1)]

thismatrix
#> [1] "mango"      "pineapple"
```

Check if an Item Exists To find out if a specified item is present in a matrix, use the %in% operator:

```
thismatrix <- matrix(c("apple", "banana", "cherry", "orange"), nrow = 2, ncol = 2)

"apple" %in% thismatrix
#> [1] TRUE
```

Number of Rows and Columns Use the dim() function to find the number of rows and columns in a Matrix:

```
thismatrix <- matrix(c("apple", "banana", "cherry", "orange"), nrow = 2, ncol = 2)

dim(thismatrix)
#> [1] 2 2
```

Matrix Length Use the length() function to find the dimension of a Matrix:

```
thismatrix <- matrix(c("apple", "banana", "cherry", "orange"), nrow = 2, ncol = 2)

length(thismatrix)
#> [1] 4
```

Combine two Matrices Again, you can use the rbind() or cbind() function to combine two or more matrices together:

```
# Combine matrices
Matrix1 <- matrix(c("apple", "banana", "cherry", "grape"), nrow = 2, ncol = 2)
Matrix2 <- matrix(c("orange", "mango", "pineapple", "watermelon"), nrow = 2, ncol = 2)

# Adding it as a rows
Matrix_Combined <- rbind(Matrix1, Matrix2)
Matrix_Combined
#>      [,1]      [,2]
#> [1,] "apple"  "cherry"
#> [2,] "banana" "grape"
#> [3,] "orange" "pineapple"
```

```
#> [4,] "mango" "watermelon"

# Adding it as a columns
Matrix_Combined <- cbind(Matrix1, Matrix2)
Matrix_Combined
#>      [,1]      [,2]      [,3]      [,4]
#> [1,] "apple" "cherry" "orange" "pineapple"
#> [2,] "banana" "grape" "mango" "watermelon"
```

## 4.6 Data Frame( )

`data.frame()`

```
# Create a data frame
Data_Frame <- data.frame (
  Training = c("Strength", "Stamina", "Other"),
  Pulse = c(100, 150, 120),
  Duration = c(60, 30, 45)
)

# Print the data frame
Data_Frame
#>   Training Pulse Duration
#> 1 Strength   100      60
#> 2  Stamina   150      30
#> 3   Other   120      45
```

Use the `summary()` function to summarize the data from a Data Frame:

```
summary(Data_Frame)
#>   Training      Pulse      Duration
#> Length:3      Min.   :100.0   Min.   :30.0
#> Class :character 1st Qu.:110.0   1st Qu.:37.5
#> Mode  :character Median :120.0   Median :45.0
#>          Mean   :123.3   Mean   :45.0
#>          3rd Qu.:135.0   3rd Qu.:52.5
#>          Max.   :150.0   Max.   :60.0
```

`[]`   `[[ ]]`   `$`

```
Data_Frame[1]
#>   Training
#> 1 Strength
#> 2 Stamina
#> 3   Other

Data_Frame[["Training"]]
#> [1] "Strength" "Stamina" "Other"

Data_Frame$Training
#> [1] "Strength" "Stamina" "Other"
```

```
rbind()
```

```
# Add a new row
New_row_DF <- rbind(Data_Frame, c("Strength", 110, 110))

# Print the new row
New_row_DF
#>   Training Pulse Duration
#> 1 Strength   100       60
#> 2 Stamina   150       30
#> 3   Other   120       45
#> 4 Strength   110      110
```

```
cbind()
```

```
# Add a new column
New_col_DF <- cbind(New_row_DF, Steps = c(1000, 6000, 2000, 5000))

# Print the new column
New_col_DF
#>   Training Pulse Duration Steps
#> 1 Strength   100       60 1000
#> 2 Stamina   150       30 6000
#> 3   Other   120       45 2000
#> 4 Strength   110      110 5000
```

```
rbind()      R
```

```
Data_Frame1 <- data.frame (
  Training = c("Strength", "Stamina", "Other"),
  Pulse = c(100, 150, 120),
  Duration = c(60, 30, 45)
```

```

)

Data_Frame2 <- data.frame (
  Training = c("Stamina", "Stamina", "Strength"),
  Pulse = c(140, 150, 160),
  Duration = c(30, 30, 20)
)

New_Data_Frame <- rbind(Data_Frame1, Data_Frame2)
New_Data_Frame
#>   Training Pulse Duration
#> 1 Strength  100      60
#> 2 Stamina  150      30
#> 3   Other  120      45
#> 4 Stamina  140      30
#> 5 Stamina  150      30
#> 6 Strength 160      20

```

`cbind()`      R

```

Data_Frame3 <- data.frame (
  Training = c("Strength", "Stamina", "Other"),
  Pulse = c(100, 150, 120),
  Duration = c(60, 30, 45)
)

Data_Frame4 <- data.frame (
  Steps = c(3000, 6000, 2000),
  Calories = c(300, 400, 300)
)

New_Data_Frame1 <- cbind(Data_Frame3, Data_Frame4)
New_Data_Frame1
#>   Training Pulse Duration Steps Calories
#> 1 Strength  100      60  3000      300
#> 2 Stamina  150      30  6000      400
#> 3   Other  120      45  2000      300

```

# All about tidyverse

## 4.7

## 4.8 ggplot2

## 4.9 Reference links

- [The R Graph Gallery](#)
- [A ggplot2 Tutorial for Beautiful Plotting in R](#)



## Chapter 5

### (PART\*) bookdown





# how to write a book

<https://www.youtube.com/watch?v=9i0ElncHGRg&t=905s>



# Chapter 6

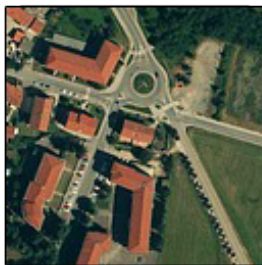
## Raster

### 6.1 Resolution

What is resolution of a satellite image? How can we understand it? Generally speaking we can say there are **three** different resolutions for a satellite image in Geoscience.

#### 6.1.1 Spatial resolution

Spatial resolution is the detail in pixels of an image. High spatial resolution means more detail and a smaller grid cell size. Whereas, lower spatial resolution means less detail and larger pixel size. Overall, spatial resolution describes the quality of an image and how detailed objects are in an image. If the grid cells are smaller, this means the spatial resolution has more detail with more pixels.



High Spatial Resolution



Medium Spatial Resolution



Low Spatial Resolution

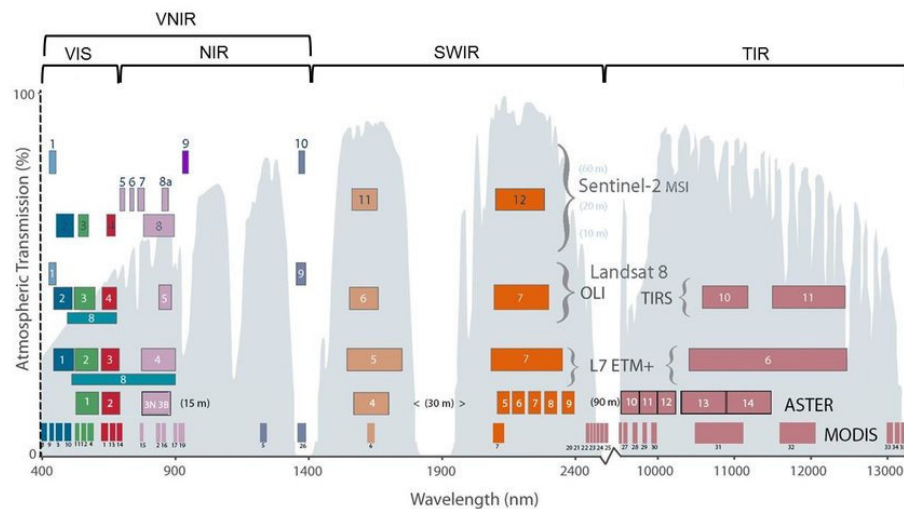
### 6.1.2 Temporal resolution

Same definition of temporal resolution can be applied to polar orbiting satellites. But defining it more precisely, temporal resolution for a polar orbiting satellite is the amount of time that the satellite takes to revisit and recapture a particular site. It is also commonly referred to as a satellite's revisit period.

Mission	Number of satellites	Temporal resolution (single satellite)	Temporal resolution (constellation)
SENTINEL-1	2	12 days	6 days
SENTINEL-2	2	10 days	5 days
LANDSAT 7	1	16 days	16 days
WorldView-3	1	1 day	1 day
Terra	1	16 days	16 days

### ### Spectral resolution

Spectral resolution is determined by the width of each band in a wavelength. The more bands in an image, the more complex the color will be.



## 6.2 CRS

A Coordinate reference system (CRS) defines, with the help of coordinates, how the two-dimensional, projected map is related to real locations on the earth. There are two different types of coordinate reference systems: **Geographic Coordinate Systems** and **Projected Coordinate Systems**.

## 6.3 Manipulation in R

How can we manipulate remote sensing imagery in R after we have a general understanding of raster data? Here I want to introduce to you a package: **terra**, which is a tweaked version or even a more powerful package than the well-known one: **raster**. As for the reason why the author choose to rebuild some functions, you can check it here.

## 6.4 Reference

- <http://modern-rstats.eu/index.html#note-to-the-reader>



## Chapter 7

# Multispectral sensors

In this section I'll introduce you to three well-known high-resolution multispectral sensor satellites in this section: MODIS, Landsat, and Sentinel-2.