

Methods in evolutionary ecology WS25/26

Stefanos Siozios

Michael Gerth

Table of contents

About	4
Quarto	4
How to find your way around	5
 I R	 6
1 First steps	7
1.1 Operators and functions	7
1.2 Data types	10
1.3 Exercises	11
 2 Data structures	 13
2.1 Vectors	13
2.1.1 Exercises	18
2.2 Matrices	18
2.2.1 Exercise	21
2.3 Data frames	21
2.3.1 Exercise	34
 3 Data, packages, and some more functions	 35
3.1 Setting up your working environment	35
3.2 Functions	36
3.3 Loops	37
3.4 Plots	38
3.5 Exercises	40
 4 Tidyverse	 41
4.1 What is the tidyverse?	41
4.2 Our data set for today	42
4.3 <code>filter()</code> for filtering data frames	43
4.4 The pipe <code>%>%</code> for combining commands	56
4.5 Sort by column with <code>arrange()</code>	67
4.6 Select columns with <code>select()</code>	107
4.7 Create new variables with <code>mutate()</code>	169
4.8 Exercise	180

4.9	<code>group_by()</code> and <code>summarise()</code> as powerful data exploration tools	181
4.10	More exercises	182
5	The <code>ggplot2</code> package	183
5.1	Very (!) brief introduction	183
5.2	Building up the plot	183
5.3	Faceting	187
5.4	Some common plot types	189
5.5	Fine tuning plots	193
5.6	Exercise	194
II	UNIX	195
6	Introduction to Linux Environment and Command Line.	196
6.1	The Unix / Linux environment	196
6.2	Why learn Unix command line?	196
6.3	Some terminology	197
6.4	Some important rules	197
6.5	Accessing your Terminal	199
7	Part 1. Your first Unix commands - Navigating the Unix File-System Structure	200
7.1	Finding out where you are	201
7.2	The command <code>ls</code>	202
7.3	Relative vs absolute path and getting help	203
7.4	Moving around	203
7.5	Exercise	204
8	Part 2. Working with Files and Directories	205
8.1	Creating new directories and files	205
8.2	Move or Rename a File or Directory	206
8.3	Delete (remove) files and directories	207
8.4	Viewing and inspecting Files	208
8.5	A text editor for the terminal	209
III	Phylogenetics	210
IV	Microbiome	211
	References	212

About

This script covers the computational and bioinformatics parts of the module “Methods in Evolutionary Ecology”. We will introduce you to **R** and **BASH**, two of the most widely used scripting languages, and make you familiar with navigating in a UNIX environment. These skills are important for any biologist, irrespective of the field you may want to specialise in in the future. Building upon your new knowledge, we will learn how to reconstruct phylogenies from sequencing data, how to work with genomic data, and how to characterise microbiomes. At the end of three weeks computational work, you will tackle a small computational group project, putting your new skills into practise.

The script is designed to cover the entire course content. While we will go you through all of the material together in detail during the course, the script should also enable you to work through the content on your own, e.g., to recap after the course has finished and as a reference and starting point for future computational endeavours.

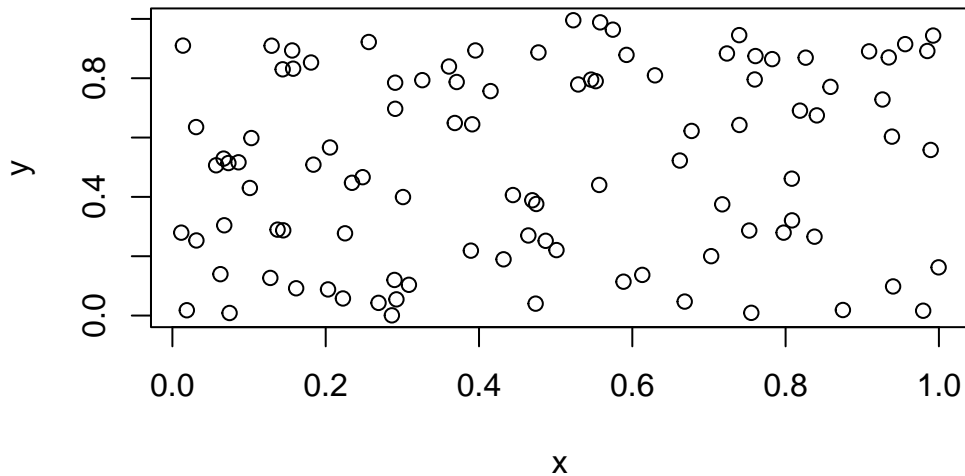
Quarto

The text is formatted using [Quarto](#), which comes with a number of benefits. It allows us to provide explanations as structured and nicely formatted regular text, and to include code blocks for all computational steps. When compiling Quarto documents, all of the code is run, which means that you not only see the code, but also the outputs it creates.

Here is an example:

This little block of **R** code generates 100 random coordinates and plots them. The code is shown below, together with the output the code has produced (in this case, a plot).

```
x <- runif(100)
y <- runif(100)
plot(x, y)
```



The code can conveniently be copied from the block into your own scripts.

Quarto supports many formats, we here provide the script as a webpage and a printable pdf. Writing Quarto documents is very simple and can be done using RStudio as an editor. The entire script is available for you on [github](#) – feel free to download it and modify it with your own comments, notes, and code. We will provide a short introduction to github and Quarto in the course.

How to find your way around

Simply use the navigation on the left to quickly access the different topics, or flip through the individual pages using the buttons at the bottom of the page. You may wish to download the pdf version of the script (click the pdf icon in the top left) which is ideal for printing. The script is organized by topics, rather than course days, because we will adapt the tempo according to your needs.

Please note: The script will very likely only be complete at the end of the course. We will still be modifying and correcting it throughout the three weeks you are with us. So make sure to check out the final version at the end of the course.

Part I

R

1 First steps

R is a statistical programming environment that has become a standard tool in the data and life sciences and many other fields. You may have used R already to run some statistics in a course you took in your studies, and this will be a likely use case for your remaining degree. However, R is much more: it can be used to analyse massive the datasets of the “omics”- age, build webpages, blogs, and interactive apps, and even for art!

Before taking full advantage of what the various R packages have to offer, we need to become familiar with its basic structure and commands. It pays off to invest a little effort in practicing the basics, because all R packages use the same syntax – a solid familiarity with base R thus allows you to explore the entire R universe independently.

1.1 Operators and functions

R can be used just like an arithmetic calculator. You are familiar with all of the basic syntax already, if you know how to use a calculator!

Some examples:

```
3 + 4
```

```
[1] 7
```

```
3 - 4
```

```
[1] -1
```

```
3 * 4
```

```
[1] 12
```

```
3 / 4
```

```
[1] 0.75
```

```
3 ^ 4 # power of
```

```
[1] 81
```

As with a regular calculator, there is operator precedence: power > multiplicative operations > additive operations:

```
(1 + 2) * 3
```

```
[1] 9
```

```
2^3 * 3
```

```
[1] 24
```

```
2^(3 * 3)
```

```
[1] 512
```

Square roots, exponentials, and logarithms also work just as with a calculator:

```
sqrt(9)
```

```
[1] 3
```

```
exp(3)
```

```
[1] 20.08554
```

```
log(3)
```

```
[1] 1.098612
```



```
log(exp(3)) # natural logarithm
```

```
[1] 3
```

```
log10(100) # logarithm to base 10
```

```
[1] 2
```

In order to “save” a value for use later on, you have to assign it to a variable! `<-` is the assignment operator you need to use for this (handy shortcut in RStudio is `ALT + -`).

```
x <- 3 + 4
```

Calling the variable will then print the result to the R console, and can be used in other calculations.

```
x
```

```
[1] 7
```

```
x + 10
```

```
[1] 17
```

You can call your variables whatever you want, but be careful: R will overwrite any variable if you tell it to, without a warning! You should also avoid giving your variables names that are already assigned to functions.

```
my_favourite_variable <- 100  
my_favourite_variable <- 50  
my_favourite_variable
```

```
[1] 50
```

All variables (among other things) are visible in the environment panel in RStudio (default: top right part of the screen).

“=” can also be used to assign variables but is discouraged, because the direction of the assignment is not immediately obvious. It is best practise to always start with the variable, followed by the assignment operator

1.2 Data types

You need to be familiar with at least three important data types in R: **logical**, **numeric**, and **character**. Data being stored in a different data type than required is one of the most frequent error messages you will encounter as an R beginner.

logical simply means true or false. R also understands the abbreviations T and F. To determine which types your data is in, you can use **mode** or **class**.

```
var1 <- TRUE
mode(var1)
```

```
[1] "logical"
```

numeric means numbers

```
var2 <- 10
class(10)
```

```
[1] "numeric"
```

A character is any form of text, a so called “string”. It must always be surrounded by quotation marks!

```
var3 <- "A so called string"
mode(var3)
```

```
[1] "character"
```

If in doubt, R will often convert or read in data as characters. Watch out for some common errors!

```
var4 <- "5"
var4
```

```
[1] "5"
```

```
is.numeric(var4)
```

```
[1] FALSE
```

```
var5 <- "TRUE"  
var5
```

```
[1] "TRUE"
```

```
is.logical(var5)
```

```
[1] FALSE
```

You can convert between types easily!

```
var6 <- as.numeric(var4)  
var6
```

```
[1] 5
```

```
class(var6)
```

```
[1] "numeric"
```

1.3 Exercises

- a) Sum the values of 1 to 5
- b) Create a variable v1 and assign it a character value
- c) Copy variable v1 to v2
- d) Compare the value of v1 against v2

Tip

Compare values and variables using the following operators

<	less than
<=	less than or equal to
>	greater than
>=	greater than or equal to

`==` equals
`!=` not equal

Please note, `=` and `==` do very different things! Don't mix them up.

2 Data structures

So far we've only looked at simple variables consisting of a single value or character. Typically, your data will be more complex. In R, there are three structures relevant for the data you will be working with.

2.1 Vectors

A **vector** is a number of elements of the same data type (`logical`, `numeric`, `character`). It can be generated by concatenating the elements using the function `c`.

```
vec1 <- c(T, F, T, F)
vec1
```

```
[1] TRUE FALSE TRUE FALSE
```

```
mode(vec1)
```

```
[1] "logical"
```

```
vec2 <- c(1, 2, 3, 4, 5)
vec2
```

```
[1] 1 2 3 4 5
```

```
mode(vec2)
```

```
[1] "numeric"
```

```
vec3 <- c("Spring", "Summer", "Autumn", "Winter")
vec3
```

```
[1] "Spring" "Summer" "Autumn" "Winter"
```

```
mode(vec3)
```

```
[1] "character"
```

Other ways to generate vectors are `rep` and `seq`. `rep` is used to repeat any number of elements any number of times.

```
rep(5, 10)
```

```
[1] 5 5 5 5 5 5 5 5 5 5
```

```
rep(vec3, 5)
```

```
[1] "Spring" "Summer" "Autumn" "Winter" "Spring" "Summer" "Autumn" "Winter"
[9] "Spring" "Summer" "Autumn" "Winter" "Spring" "Summer" "Autumn" "Winter"
[17] "Spring" "Summer" "Autumn" "Winter"
```

`seq` can be used to create numerical sequences.

```
seq(from = 0, to = 100, by = 5)
```

```
[1] 0 5 10 15 20 25 30 35 40 45 50 55 60 65 70 75 80 85 90
[20] 95 100
```

The command above is easy to read and understand for humans, which is good. R will also understand if you specify it as

```
seq(0, 100, 5)
```

```
[1] 0 5 10 15 20 25 30 35 40 45 50 55 60 65 70 75 80 85 90
[20] 95 100
```

As a shortcut for a common sequences, you can use

```
1:10
```

```
[1] 1 2 3 4 5 6 7 8 9 10
```

As mentioned above,, vectors can only combine elements of a single data type. Combining multiple different data types may result in some unwanted behaviour.

```
vec_mix1 <- c(5, TRUE, 65)
mode(vec_mix1)
```

```
[1] "numeric"
```

```
vec_mix2 <- c("blue", TRUE, "red")
mode(vec_mix2)
```

```
[1] "character"
```

In many cases you may wish to access a single element of a vector. You can do so using square brackets.

```
z <- c("order", "family", "genus", "species")
z[2]
```

```
[1] "family"
```

Similarly, you can access any combination of elements from the vector.

```
z[1:2]
```

```
[1] "order" "family"
```

```
i <- c(1, 3)
z[i]
```

```
[1] "order" "genus"
```

```
z[c(1, 1, 1, 4)]
```

```
[1] "order" "order" "order" "species"
```

```
z[-1]
```

```
[1] "family" "genus" "species"
```

The square brackets are also used if you need to change elements of the vector. Changes are made using the assignment operator which you already know.

```
x <- 1:5  
x
```

```
[1] 1 2 3 4 5
```

```
x[c(1, 4)] <- 10  
x
```

```
[1] 10 2 3 10 5
```

Which elements of a vector have certain characteristics? This is important for filtering/selecting in your dataset. You can combine different queries using logical operators.

```
x >= 5
```

```
[1] TRUE FALSE FALSE TRUE TRUE
```

```
x[x >= 5]
```

```
[1] 10 10 5
```

```
which(x >= 5)
```

```
[1] 1 4 5
```

```
z
```

```
[1] "order" "family" "genus" "species"
```

```
which(z == "genus")
```

```
[1] 3
```

```
z[z== "genus"]
```

```
[1] "genus"
```



```
z[z != "genus"]
```

```
[1] "order" "family" "species"
```

```
which(z== "genus" | z == "order")
```

```
[1] 1 3
```

Logical operators in R

	OR
&	AND
!	NOT

Conveniently, the elements of a vector can be named and accessed using the names. Let's first create a vector...

```
dmel <- c("Hexapoda", "Diptera", "Drosophilidae", "Drosophila", "Drosophila melanogaster")
dmel
```

```
[1] "Hexapoda"          "Diptera"
[3] "Drosophilidae"     "Drosophila"
[5] "Drosophila melanogaster"
```

... and then add names for each element

```
names(dmel) <- c("Class", "Order", "Family", "Genus", "Species")
dmel
```

	Class	Order	Family
	"Hexapoda"	"Diptera"	"Drosophilidae"
	Genus	Species	
	"Drosophila"	"Drosophila melanogaster"	

```
str(dmel)
```

```
Named chr [1:5] "Hexapoda" "Diptera" "Drosophilidae" "Drosophila" ...
- attr(*, "names")= chr [1:5] "Class" "Order" "Family" "Genus" ...
```

Now we can use the names to access the values

```
dmel[c("Class", "Species")]
```

Class	Species
"Hexapoda"	"Drosophila melanogaster"

```
dmel[names(dmel) == "Order"]
```

Order
"Diptera"

2.1.1 Exercises

- Create a vector consecutively numbering all days of the year 2026. Assign the correct weekday names for all elements of the vector.
- Use the vector to determine how many days in 2026 are weekend days.



Tip

If you struggle to assign the correct names, have a look at the help for `rep`.

2.2 Matrices

A **matrix** in R can be thought of as a two-dimensional vector. All elements must be of the same data type. There are various ways to create a matrix. For example, one can use the `matrix` function like this.

```
mat1 <- matrix(data = 1:12, nrow = 3, ncol = 4, byrow=T)
mat1
```

	[,1]	[,2]	[,3]	[,4]
[1,]	1	2	3	4
[2,]	5	6	7	8
[3,]	9	10	11	12

Alternatively, a vector can be transformed into a matrix

```
mat2 <- 1:12
dim(mat2) <- c(3, 4)
mat2
```

```
      [,1] [,2] [,3] [,4]
[1,]    1    4    7   10
[2,]    2    5    8   11
[3,]    3    6    9   12
```

Often you will want to combine multiple vectors into a matrix

```
dmel <- c("Hexapoda", "Diptera", "Drosophilidae", "Drosophila", "Drosophila melanogaster")
dhyd <- c("Hexapoda", "Diptera", "Drosophilidae", "Drosophila", "Drosophila hydei")
mat3 <- cbind(dmel, dhyd)
mat3
```

```
      dmel                      dhyd
[1,] "Hexapoda"                "Hexapoda"
[2,] "Diptera"                  "Diptera"
[3,] "Drosophilidae"            "Drosophilidae"
[4,] "Drosophila"                "Drosophila"
[5,] "Drosophila melanogaster" "Drosophila hydei"
```

```
mat4 <- rbind(dmel, dhyd)
mat4
```

```
      [,1]      [,2]      [,3]      [,4]
dmel "Hexapoda" "Diptera" "Drosophilidae" "Drosophila"
dhyd "Hexapoda" "Diptera" "Drosophilidae" "Drosophila"
      [,5]
dmel "Drosophila melanogaster"
dhyd "Drosophila hydei"
```

Just like vectors, matrix elements can have names

```
mat3
```

	dmel	dhyd
[1,]	"Hexapoda"	"Hexapoda"
[2,]	"Diptera"	"Diptera"
[3,]	"Drosophilidae"	"Drosophilidae"
[4,]	"Drosophila"	"Drosophila"
[5,]	"Drosophila melanogaster"	"Drosophila hydei"

```
colnames(mat3)
```

```
[1] "dmel" "dhyd"
```

```
rownames(mat3) <- c("Class", "Order", "Family", "Genus", "Species")
mat3
```

	dmel	dhyd
Class	"Hexapoda"	"Hexapoda"
Order	"Diptera"	"Diptera"
Family	"Drosophilidae"	"Drosophilidae"
Genus	"Drosophila"	"Drosophila"
Species	"Drosophila melanogaster"	"Drosophila hydei"

And just like with vectors, we can use square brackets to access and replace values. Because there are 2 dimensions, we need to provide 2 values (one for rows, one for columns, separated by ,).

```
mat3
```

	dmel	dhyd
Class	"Hexapoda"	"Hexapoda"
Order	"Diptera"	"Diptera"
Family	"Drosophilidae"	"Drosophilidae"
Genus	"Drosophila"	"Drosophila"
Species	"Drosophila melanogaster"	"Drosophila hydei"

```
mat3[1:3, 2]
```

Class	Order	Family
"Hexapoda"	"Diptera"	"Drosophilidae"

```
mat3[1:3, ]
```

	dmel	dhyd
Class	"Hexapoda"	"Hexapoda"
Order	"Diptera"	"Diptera"
Family	"Drosophilidae"	"Drosophilidae"

```
mat3[c("Class", "Species"), ]
```

	dmel	dhyd
Class	"Hexapoda"	"Hexapoda"
Species	"Drosophila melanogaster"	"Drosophila hydei"

2.2.1 Exercise

- create a matrix using with 20 rows & 5 columns, using 100 randomly generated numbers between 0 and 1000.
- replace all values in the 3rd column of this matrix that are larger than 500 with NA.

Tip

use the function `runif` to create random values

2.3 Data frames

Data frames are the R equivalent of spread sheets. Like matrices, they are two-dimensional, however they may combine different data types. Most biological data sets you will encounter will be data frames.

Lets create a data frame

```
# create some data
species <- rep(c("beech","ash","elm","maple", "sycamore"),40)
species
```

```

[1] "beech"      "ash"      "elm"      "maple"    "sycamore" "beech"
[7] "ash"      "elm"      "maple"    "sycamore" "beech"    "ash"
[13] "elm"      "maple"    "sycamore" "beech"    "ash"      "elm"
[19] "maple"    "sycamore" "beech"    "ash"      "elm"      "maple"
[25] "sycamore" "beech"    "ash"      "elm"      "maple"    "sycamore"
[31] "beech"    "ash"      "elm"      "maple"    "sycamore" "beech"
[37] "ash"      "elm"      "maple"    "sycamore" "beech"    "ash"
[43] "elm"      "maple"    "sycamore" "beech"    "ash"      "elm"
[49] "maple"    "sycamore" "beech"    "ash"      "elm"      "maple"
[55] "sycamore" "beech"    "ash"      "elm"      "maple"    "sycamore"
[61] "beech"    "ash"      "elm"      "maple"    "sycamore" "beech"
[67] "ash"      "elm"      "maple"    "sycamore" "beech"    "ash"
[73] "elm"      "maple"    "sycamore" "beech"    "ash"      "elm"
[79] "maple"    "sycamore" "beech"    "ash"      "elm"      "maple"
[85] "sycamore" "beech"    "ash"      "elm"      "maple"    "sycamore"
[91] "beech"    "ash"      "elm"      "maple"    "sycamore" "beech"
[97] "ash"      "elm"      "maple"    "sycamore" "beech"    "ash"
[103] "elm"      "maple"    "sycamore" "beech"    "ash"      "elm"
[109] "maple"    "sycamore" "beech"    "ash"      "elm"      "maple"
[115] "sycamore" "beech"    "ash"      "elm"      "maple"    "sycamore"
[121] "beech"    "ash"      "elm"      "maple"    "sycamore" "beech"
[127] "ash"      "elm"      "maple"    "sycamore" "beech"    "ash"
[133] "elm"      "maple"    "sycamore" "beech"    "ash"      "elm"
[139] "maple"    "sycamore" "beech"    "ash"      "elm"      "maple"
[145] "sycamore" "beech"    "ash"      "elm"      "maple"    "sycamore"
[151] "beech"    "ash"      "elm"      "maple"    "sycamore" "beech"
[157] "ash"      "elm"      "maple"    "sycamore" "beech"    "ash"
[163] "elm"      "maple"    "sycamore" "beech"    "ash"      "elm"
[169] "maple"    "sycamore" "beech"    "ash"      "elm"      "maple"
[175] "sycamore" "beech"    "ash"      "elm"      "maple"    "sycamore"
[181] "beech"    "ash"      "elm"      "maple"    "sycamore" "beech"
[187] "ash"      "elm"      "maple"    "sycamore" "beech"    "ash"
[193] "elm"      "maple"    "sycamore" "beech"    "ash"      "elm"
[199] "maple"    "sycamore"

```

```

dbh <- runif(200, 5, 40)
dbh

```

```

[1] 10.044094 32.757767 8.925053 13.807362 23.551105 19.234134 29.784401
[8] 19.914805 8.917980 35.235702 24.112720 30.786224 23.873515 13.694458
[15] 26.677193 24.903134 6.582594 35.465086 15.766389 36.379711 28.268448
[22] 31.397243 18.517488 28.823234 28.858379 8.060898 34.694842 21.227169

```

```

[29] 20.789003  5.464636 18.221045 23.572341 16.128950 39.221586 34.064109
[36] 22.823254 16.107606 31.412299 14.301667 13.495726 35.563361 20.350248
[43]  6.479554 34.554516  6.153258 38.462263 25.040311 33.003961 39.740898
[50] 14.037295 25.109435 31.948507 21.998141 13.225884 32.093805 29.552895
[57] 21.772292 12.623135 37.886395 26.589330 22.457700  5.915094 11.906434
[64]  6.783141 14.684739 38.473724  7.001651  5.881051 32.913674 10.432873
[71] 30.676020 28.439690  5.183506 31.533047 14.069628 36.106217 13.369217
[78] 15.573841 20.503041 27.824678 35.685208 30.168768 10.182187 23.196161
[85] 39.480306 19.605731  9.248206 26.402371  5.412345 18.879057 37.590255
[92] 20.982300 33.570737 39.147252 18.013256 29.459367 16.397009 24.289191
[99] 26.379787 27.030762 23.856859 16.227416 38.557839 20.207973 18.276160
[106] 16.346004 14.641058 12.194734 33.289332 22.184364 12.231942 19.598883
[113] 28.886417  5.301439 32.198566 25.840199 31.700178 38.854138 17.195720
[120] 13.535357 30.451491 30.208220 10.320936 14.031847 32.070404  7.557059
[127]  9.693760 33.213005 18.586024 33.356090 27.621806 19.878239  6.328824
[134] 17.467950 28.705810 38.572616 21.921200  5.882015 17.224443 11.144815
[141] 30.721523 31.089054 10.981030  7.999222 23.121387 21.696210 25.456123
[148] 31.890229 32.570559 15.022879 13.601822 16.012817 26.485338 33.294767
[155]  8.768550 33.398899 21.827025 33.347179 32.690425 21.520964  7.534485
[162]  9.019292 28.141417 25.774356 24.811242 27.714324 14.011637 36.114202
[169] 34.667347 36.334699  9.025190 35.700573 38.051683  6.168635 26.200303
[176] 25.830076 35.270737 13.999843  8.021445 12.125467 28.337453 29.331106
[183] 33.047502 11.701738 32.420077 26.045249 29.456374 19.806934 26.052081
[190] 31.710512  7.031836 17.751345 28.109164 13.525772  7.660293 14.977394
[197] 12.344184 35.594130 10.804181  6.559926

```

```

age <- as.integer(runif(200, 20, 120))
age

```

```

[1] 74 67 117 118 102 72 27 35 85 103 107 23 113 86 75 29 115 25
[19] 63 110 40 99 99 72 73 34 88 103 67 111 40 65 78 71 112 117
[37] 59 99 108 33 100 100 74 66 45 89 70 68 22 104 38 66 79 44
[55] 85 116 92 39 76 69 117 47 82 74 89 39 24 28 35 80 80 58
[73] 44 103 72 24 113 45 65 28 26 23 81 38 92 56 58 48 61 104
[91] 54 93 41 103 114 106 80 52 111 109 36 37 93 112 35 29 62 21
[109] 51 95 24 54 109 28 64 87 72 62 31 66 76 31 92 35 113 56
[127] 75 114 55 54 55 60 89 42 97 27 105 26 95 59 75 51 92 118
[145] 94 107 72 44 88 103 60 87 56 65 44 67 62 68 109 50 51 52
[163] 24 105 84 43 34 77 72 43 105 74 57 54 108 92 81 109 89 51
[181] 45 67 30 38 91 86 102 24 72 113 95 107 39 29 87 69 55 106
[199] 40 47

```

```
df1 <- data.frame(species, dbh, age)
df1
```

species	dbh	age
beech	10.044094	74
ash	32.757767	67
elm	8.925053	117
maple	13.807362	118
sycamore	23.551105	102
beech	19.234133	72
ash	29.784401	27
elm	19.914805	35
maple	8.917980	85
sycamore	35.235702	103
beech	24.112720	107
ash	30.786224	23
elm	23.873515	113
maple	13.694458	86
sycamore	26.677193	75
beech	24.903134	29
ash	6.582594	115
elm	35.465086	25
maple	15.766389	63
sycamore	36.379711	110
beech	28.268448	40
ash	31.397243	99
elm	18.517488	99
maple	28.823234	72
sycamore	28.858379	73
beech	8.060898	34
ash	34.694842	88
elm	21.227169	103
maple	20.789003	67
sycamore	5.464636	111
beech	18.221045	40
ash	23.572341	65
elm	16.128950	78
maple	39.221586	71
sycamore	34.064109	112
beech	22.823254	117
ash	16.107606	59

species	dbh	age
elm	31.412299	99
maple	14.301667	108
sycamore	13.495726	33
beech	35.563361	100
ash	20.350248	100
elm	6.479554	74
maple	34.554516	66
sycamore	6.153258	45
beech	38.462263	89
ash	25.040311	70
elm	33.003961	68
maple	39.740898	22
sycamore	14.037295	104
beech	25.109435	38
ash	31.948508	66
elm	21.998141	79
maple	13.225884	44
sycamore	32.093805	85
beech	29.552895	116
ash	21.772292	92
elm	12.623135	39
maple	37.886395	76
sycamore	26.589330	69
beech	22.457700	117
ash	5.915094	47
elm	11.906434	82
maple	6.783141	74
sycamore	14.684739	89
beech	38.473724	39
ash	7.001651	24
elm	5.881051	28
maple	32.913674	35
sycamore	10.432873	80
beech	30.676020	80
ash	28.439690	58
elm	5.183506	44
maple	31.533047	103
sycamore	14.069628	72
beech	36.106217	24
ash	13.369217	113
elm	15.573841	45

species	dbh	age
maple	20.503041	65
sycamore	27.824678	28
beech	35.685208	26
ash	30.168768	23
elm	10.182187	81
maple	23.196161	38
sycamore	39.480306	92
beech	19.605731	56
ash	9.248206	58
elm	26.402371	48
maple	5.412345	61
sycamore	18.879057	104
beech	37.590255	54
ash	20.982300	93
elm	33.570737	41
maple	39.147252	103
sycamore	18.013256	114
beech	29.459367	106
ash	16.397009	80
elm	24.289191	52
maple	26.379787	111
sycamore	27.030762	109
beech	23.856859	36
ash	16.227416	37
elm	38.557839	93
maple	20.207973	112
sycamore	18.276160	35
beech	16.346004	29
ash	14.641058	62
elm	12.194734	21
maple	33.289333	51
sycamore	22.184364	95
beech	12.231942	24
ash	19.598883	54
elm	28.886417	109
maple	5.301439	28
sycamore	32.198566	64
beech	25.840199	87
ash	31.700178	72
elm	38.854138	62
maple	17.195720	31

species	dbh	age
sycamore	13.535357	66
beech	30.451491	76
ash	30.208220	31
elm	10.320936	92
maple	14.031847	35
sycamore	32.070404	113
beech	7.557059	56
ash	9.693760	75
elm	33.213005	114
maple	18.586024	55
sycamore	33.356090	54
beech	27.621806	55
ash	19.878239	60
elm	6.328824	89
maple	17.467950	42
sycamore	28.705810	97
beech	38.572616	27
ash	21.921200	105
elm	5.882015	26
maple	17.224443	95
sycamore	11.144815	59
beech	30.721523	75
ash	31.089054	51
elm	10.981030	92
maple	7.999222	118
sycamore	23.121387	94
beech	21.696210	107
ash	25.456123	72
elm	31.890229	44
maple	32.570559	88
sycamore	15.022879	103
beech	13.601822	60
ash	16.012817	87
elm	26.485338	56
maple	33.294767	65
sycamore	8.768550	44
beech	33.398899	67
ash	21.827025	62
elm	33.347179	68
maple	32.690425	109
sycamore	21.520964	50

species	dbh	age
beech	7.534485	51
ash	9.019292	52
elm	28.141417	24
maple	25.774356	105
sycamore	24.811242	84
beech	27.714324	43
ash	14.011637	34
elm	36.114202	77
maple	34.667347	72
sycamore	36.334699	43
beech	9.025190	105
ash	35.700573	74
elm	38.051683	57
maple	6.168635	54
sycamore	26.200303	108
beech	25.830076	92
ash	35.270737	81
elm	13.999843	109
maple	8.021445	89
sycamore	12.125467	51
beech	28.337453	45
ash	29.331106	67
elm	33.047503	30
maple	11.701738	38
sycamore	32.420077	91
beech	26.045249	86
ash	29.456374	102
elm	19.806934	24
maple	26.052081	72
sycamore	31.710511	113
beech	7.031836	95
ash	17.751345	107
elm	28.109164	39
maple	13.525772	29
sycamore	7.660293	87
beech	14.977394	69
ash	12.344184	55
elm	35.594130	106
maple	10.804180	40
sycamore	6.559926	47

To access values, we can use the same approaches as for matrices:

```
df1[1:12, 1:2]
```

species	dbh
beech	10.044094
ash	32.757767
elm	8.925053
maple	13.807362
sycamore	23.551105
beech	19.234133
ash	29.784401
elm	19.914805
maple	8.917980
sycamore	35.235702
beech	24.112720
ash	30.786224

but can also access and filter the columns directly using their names like this:

```
df1$species
```

```
[1] "beech"  "ash"    "elm"    "maple"  "sycamore" "beech"
[7] "ash"    "elm"    "maple"  "sycamore" "beech"  "ash"
[13] "elm"    "maple"  "sycamore" "beech"  "ash"    "elm"
[19] "maple"  "sycamore" "beech"  "ash"    "elm"    "maple"
[25] "sycamore" "beech"  "ash"    "elm"    "maple"  "sycamore"
[31] "beech"  "ash"    "elm"    "maple"  "sycamore" "beech"
[37] "ash"    "elm"    "maple"  "sycamore" "beech"  "ash"
[43] "elm"    "maple"  "sycamore" "beech"  "ash"    "elm"
[49] "maple"  "sycamore" "beech"  "ash"    "elm"    "maple"
[55] "sycamore" "beech"  "ash"    "elm"    "maple"  "sycamore"
[61] "beech"  "ash"    "elm"    "maple"  "sycamore" "beech"
[67] "ash"    "elm"    "maple"  "sycamore" "beech"  "ash"
[73] "elm"    "maple"  "sycamore" "beech"  "ash"    "elm"
[79] "maple"  "sycamore" "beech"  "ash"    "elm"    "maple"
[85] "sycamore" "beech"  "ash"    "elm"    "maple"  "sycamore"
[91] "beech"  "ash"    "elm"    "maple"  "sycamore" "beech"
[97] "ash"    "elm"    "maple"  "sycamore" "beech"  "ash"
[103] "elm"    "maple"  "sycamore" "beech"  "ash"    "elm"
```

```

[109] "maple"      "sycamore" "beech"     "ash"       "elm"       "maple"
[115] "sycamore"   "beech"     "ash"       "elm"       "maple"     "sycamore"
[121] "beech"     "ash"       "elm"       "maple"     "sycamore"   "beech"
[127] "ash"       "elm"       "maple"     "sycamore"   "beech"     "ash"
[133] "elm"       "maple"     "sycamore"   "beech"     "ash"       "elm"
[139] "maple"     "sycamore"   "beech"     "ash"       "elm"       "maple"
[145] "sycamore"   "beech"     "ash"       "elm"       "maple"     "sycamore"
[151] "beech"     "ash"       "elm"       "maple"     "sycamore"   "beech"
[157] "ash"       "elm"       "maple"     "sycamore"   "beech"     "ash"
[163] "elm"       "maple"     "sycamore"   "beech"     "ash"       "elm"
[169] "maple"     "sycamore"   "beech"     "ash"       "elm"       "maple"
[175] "sycamore"   "beech"     "ash"       "elm"       "maple"     "sycamore"
[181] "beech"     "ash"       "elm"       "maple"     "sycamore"   "beech"
[187] "ash"       "elm"       "maple"     "sycamore"   "beech"     "ash"
[193] "elm"       "maple"     "sycamore"   "beech"     "ash"       "elm"
[199] "maple"     "sycamore"

```

```
df1[df1$dbh > 15, ]
```

	species	dbh	age
2	ash	32.75777	67
5	sycamore	23.55111	102
6	beech	19.23413	72
7	ash	29.78440	27
8	elm	19.91480	35
10	sycamore	35.23570	103
11	beech	24.11272	107
12	ash	30.78622	23
13	elm	23.87352	113
15	sycamore	26.67719	75
16	beech	24.90313	29
18	elm	35.46509	25
19	maple	15.76639	63
20	sycamore	36.37971	110
21	beech	28.26845	40
22	ash	31.39724	99
23	elm	18.51749	99
24	maple	28.82323	72
25	sycamore	28.85838	73
27	ash	34.69484	88
28	elm	21.22717	103

	species	dbh	age
29	maple	20.78900	67
31	beech	18.22105	40
32	ash	23.57234	65
33	elm	16.12895	78
34	maple	39.22159	71
35	sycamore	34.06411	112
36	beech	22.82325	117
37	ash	16.10761	59
38	elm	31.41230	99
41	beech	35.56336	100
42	ash	20.35025	100
44	maple	34.55452	66
46	beech	38.46226	89
47	ash	25.04031	70
48	elm	33.00396	68
49	maple	39.74090	22
51	beech	25.10943	38
52	ash	31.94851	66
53	elm	21.99814	79
55	sycamore	32.09380	85
56	beech	29.55290	116
57	ash	21.77229	92
59	maple	37.88640	76
60	sycamore	26.58933	69
61	beech	22.45770	117
66	beech	38.47372	39
69	maple	32.91367	35
71	beech	30.67602	80
72	ash	28.43969	58
74	maple	31.53305	103
76	beech	36.10622	24
78	elm	15.57384	45
79	maple	20.50304	65
80	sycamore	27.82468	28
81	beech	35.68521	26
82	ash	30.16877	23
84	maple	23.19616	38
85	sycamore	39.48031	92
86	beech	19.60573	56
88	elm	26.40237	48
90	sycamore	18.87906	104

	species	dbh	age
91	beech	37.59025	54
92	ash	20.98230	93
93	elm	33.57074	41
94	maple	39.14725	103
95	sycamore	18.01326	114
96	beech	29.45937	106
97	ash	16.39701	80
98	elm	24.28919	52
99	maple	26.37979	111
100	sycamore	27.03076	109
101	beech	23.85686	36
102	ash	16.22742	37
103	elm	38.55784	93
104	maple	20.20797	112
105	sycamore	18.27616	35
106	beech	16.34600	29
109	maple	33.28933	51
110	sycamore	22.18436	95
112	ash	19.59888	54
113	elm	28.88642	109
115	sycamore	32.19857	64
116	beech	25.84020	87
117	ash	31.70018	72
118	elm	38.85414	62
119	maple	17.19572	31
121	beech	30.45149	76
122	ash	30.20822	31
125	sycamore	32.07040	113
128	elm	33.21300	114
129	maple	18.58602	55
130	sycamore	33.35609	54
131	beech	27.62181	55
132	ash	19.87824	60
134	maple	17.46795	42
135	sycamore	28.70581	97
136	beech	38.57262	27
137	ash	21.92120	105
139	maple	17.22444	95
141	beech	30.72152	75
142	ash	31.08905	51
145	sycamore	23.12139	94

	species	dbh	age
146	beech	21.69621	107
147	ash	25.45612	72
148	elm	31.89023	44
149	maple	32.57056	88
150	sycamore	15.02288	103
152	ash	16.01282	87
153	elm	26.48534	56
154	maple	33.29477	65
156	beech	33.39890	67
157	ash	21.82703	62
158	elm	33.34718	68
159	maple	32.69043	109
160	sycamore	21.52096	50
163	elm	28.14142	24
164	maple	25.77436	105
165	sycamore	24.81124	84
166	beech	27.71432	43
168	elm	36.11420	77
169	maple	34.66735	72
170	sycamore	36.33470	43
172	ash	35.70057	74
173	elm	38.05168	57
175	sycamore	26.20030	108
176	beech	25.83008	92
177	ash	35.27074	81
181	beech	28.33745	45
182	ash	29.33111	67
183	elm	33.04750	30
185	sycamore	32.42008	91
186	beech	26.04525	86
187	ash	29.45637	102
188	elm	19.80693	24
189	maple	26.05208	72
190	sycamore	31.71051	113
192	ash	17.75135	107
193	elm	28.10916	39
198	elm	35.59413	106

2.3.1 Exercise

- a) Using `df1`, select only entries corresponding to ash and maple with an age over 50 and a diameter less than 30.
- b) Add a new column to the dataframe called “year”. Generate data for this column so that there are 10 different years and the same number of entries for each tree species per year.

Tip

Use the function `rep` for this exercise

3 Data, packages, and some more functions

3.1 Setting up your working environment

Usually when working in **R**, you want to look at your own data, and not generate it from random distributions. To read in a data file, we first need to tell **R** where the working directory is located.

```
setwd("/home/of22haqi/Documents/TEACHING/MEE-WS25-26/")
```

The path will look different on your machine of course.

Now that `R` knows where to find it, we are ready to read in a data file.

```
# The table contains headers, and the fields are separated by commas
be <- read.table("data/butterfly_ecology.csv", header = TRUE, sep = ",")

# Let's have a glimpse at the data
head(be)
```

[illegible]


```
mean(be$range.size, na.rm = TRUE)
```

```
[1] 261.4116
```

`mean` is the function, `be$range.size` is the object (1 vector from the dataframe we just read into R) and `na.rm = TRUE` is the option to remove NAs from the vector before calculating the mean.

In some cases, you may want to do things to your data that cannot be addressed by a single function. In this case, you may have to perform a number of different operations on the dataset. If you are likely to use the same set of operations in the future, it may be advisable to use your own functions.

A very simple example. Let's assume the mean function didn't exist and we would need to write our own.

```
mean2 <- function(x){  
  x <- na.omit(x)  
  sum(x) / length(x)  
}  
  
mean2(be$range.size)
```

```
[1] 261.4116
```

We define `mean2` as a function that requires an object (here called `x` as an input). Looking into the function, we can see that it first removes the NAs from the object and next calculates the sum of `x` divided by the number of elements of `x` (this is how the mean is defined). Testing it, we can see that it gives the same result as the native mean function.

3.3 Loops

In many cases, we need to apply a function to a number of elements. In this case, loops come in handy. In the simple examples below, the structure of a for loop is illustrated.

```
for(i in 1:10) # how often is the loop repeated  
{  
  print(i)    # what is to be done each iteration  
}
```

```
[1] 1
[1] 2
[1] 3
[1] 4
[1] 5
[1] 6
[1] 7
[1] 8
[1] 9
[1] 10
```

```
j<-0
for(i in 1:5)
{
  j<-i+j
  print(j)
}
```

```
[1] 1
[1] 3
[1] 6
[1] 10
[1] 15
```

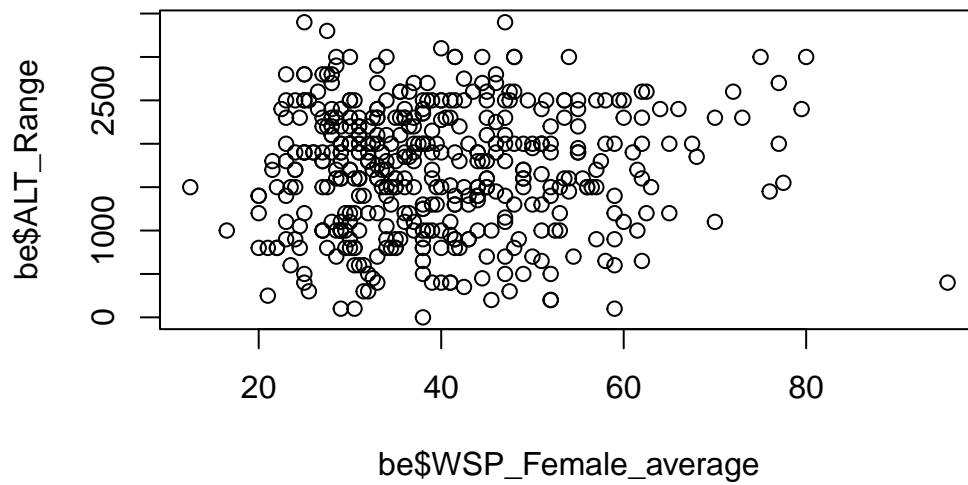
Observe and try to explain what happens in each iteration to the variables used in these examples.

3.4 Plots

For many use cases `ggplot2` is the best approach of plotting, and we will get to know this package later. However, for very simple and quick plots, base R plotting functions are sufficient and superior to other options because of simplicity and speed.

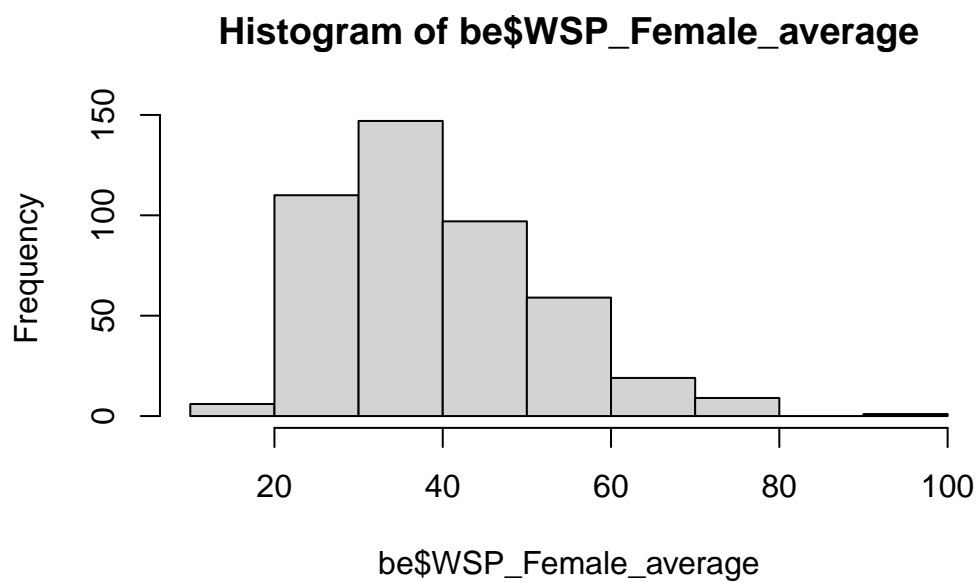
Scatter plots can be created by just naming the variables to be plotted against each other.

```
plot(be$WSP_Female_average, be$ALT_Range)
```



Histograms showing frequency distributions are also very easily generated

```
hist(be$WSP_Female_average)
```



3.5 Exercises

- a) Using a loop, plot histograms for the columns “WSP_Female_average”, “Alt_Range”, “Alt_min”, and “range.size”.
- b) Write a function that creates these plots with only the dataframe as argument.

4 Tidyverse

4.1 What is the tidyverse?

- A collection of R packages for data science
- All packages share a “philosophy” about design and data structure
- All packages are highly compatible and functions complement each other

We will only be looking at a couple of functions from a 2 packages (`dplyr` & `ggplot2`). All functions are about **data manipulation and visualisation** and are especially well suited for exploring very large data sets.

You can install all tidyverse packages by running

```
install.packages("tidyverse", dependencies = TRUE)
```

The following package(s) will be installed:

- tidyverse [2.0.0]

These packages will be installed into "~/work/MEE-WS25-26/MEE-WS25-26/renv/library/R-4.2/x86_64-pc-linux-gnu".

```
# Installing packages -----  
- Installing tidyverse ... OK [linked from cache]  
Successfully installed 1 package in 6.2 milliseconds.
```

Let's refresh what we learned earlier this week:

1. What different types of data structures are used in R?
2. Which of these do you think is most likely to be used in the `tidyverse`?

(Remember, you can just add the answers into this document for future reference!)

4.2 Our data set for today

We will be looking at a data set of ecological traits of european butterflies. Download the table and read it into R.

```
# The table contains headers, and the fields are separated by commas
be <- read.table("data/butterfly_ecology.csv", header = TRUE, sep = ",")

# Let's have a glimpse at the data
head(be)
```

[illegible]

Each of the rows contains data for 1 European species, and the columns contain the following information:

[illegible]

[illegible]

		WSP_Fe-																			
con-		GEN_Av- male_av-										FM_Av-									
fam- serv-		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_Fe-										ELT_sin-									
con- OWS		WSP_F																			

[illegible]

[illegible]

The filtering criterion can be specified using the methods you are already familiar with (e.g., `>`, `>=`, `!=`, `%in%`).

Notice that the variable names can be used directly here, so instead of using `be$OWS_egg`, `filter()` lets you use `OWS_egg` directly. All **tidyverse** functions work like that. Let's look at more complex filtering:

```
# combine 2 filters with boolean "AND" ...
filter(be, OWS_egg == 1 , ALT_Min > 500)
```


species	range	size	con- fam- serv- com- OWS	OWS page	OWS page	GEN	Av- male_av- er-WSP_Fe-	WSP_Fe-	ELT_sin-	FM_Av- er-	AFB_AFB- an- ey- i- AFB_min-	AFB_min- power																		
													WSP_Fe-	ELT_sin-	FM_Av- er-	AFB_AFB- an- ey- i- AFB_min-	AFB_min- power													
EreNy34pha5	biai-eri-	phylae	1	1	0	0	0.50	0.50	0.034	4	0.50	0.00	1	0	0	NAN	NAN	1200	5000	1	0	0	0	0	0	0	0			
EreNy2031pha5	biai-eu-	ryadae	1	1	0	0	0.70	0.51	0.00	0.544	4	0.00	0.96	0	1	0	0	1	0	0	600	1900	0.02	1	0	0	0	0	0	
EreNy89pha5	biai-manto	dae	1	1	0	0	0.50	0.50	0.039	10	0.30	0.60	1	0	0	1	0	0	900	1600	0.00	1	0	0	0	0	0	0		
PaPa484	naspil-	siusophoe-	4	4	1	1	0	0	1.00	0.01	0.00	0.065	30	0.15	0.81	1	0	0	1	0	0	1600	1200	0.00	1	0	0	0	0	0
PolKy1631	omacidae	tus_da-	4	4	1	1	0	0	1.00	0.01	0.00	0.032	4	0.44	0.71	0	0	0	1	0	0	600	1800	0.00	1	0	0	0	0	1
SatKyilmNANA	erenidae		0	0	0	0	1.00	0.01	0.00	0.029	4	0.50	0.77	0	0	1	0	1	0	0	600	1900	0.53	1	0	0	0	0	0	0

```
# ... or boolean "OR"
filter(be, OWS_egg == 1 | AFB_honeydew == 1)
```

species	range	size	con- fam- serv- com- OWS	OWS page	OWS page	GEN	Av- male_av- er-WSP_Fe-	WSP_Fe-	ELT_sin-	FM_Av- er-	AFB_AFB- an- ey- i- AFB_min-	AFB_min- power																	
ApNy5281pha5	atura_ilia	dae	0	1	0	0	1.50	0.02	0.01	0.068	0	0.20	0.70	0	1	1	1	0	0	50	1850	0.53	1	0	0	1	1	1	1
ApNy5711pha5	atura_iris	dae	0	1	0	0	1.00	0.01	0.00	0.070	0.00	0.30	0.60	0	1	0	1	0	0	0	2300	0.51	1	0	0	1	1	1	1
ApNy471pha5	atura_metis	dae	0	1	0	0	1.50	0.02	0.01	0.062	0	1.00	0.00	0	1	1	1	0	0	0	650	1.00	1	0	1	1	1	1	1

[illegible]

[illegible]

species	con- fam- ser- con- size	OWS	OWS	SV	sup- size	GEN	Av- male	WSP_Fe- er	WSP_Fe- er	ELT_sin- er	FM_Av- er	AFB_Av- ey-	AFB_i-	AFB_an- i-	AFB_min- i-	AFB_max- i-	AFB_h- i-	AFB_l- i-	AFB_s- i-	AFB_p- i-	AFB_r- i-	AFB_m- i-	AFB_n- i-	AFB_o- i-	AFB_u- i-	AFB_d- i-	AFB_c- i-	AFB_e- i-	AFB_f- i-	AFB_g- i-	AFB_h- i-	AFB_i- i-	AFB_j- i-	AFB_k- i-	AFB_l- i-	AFB_m- i-	AFB_n- i-	AFB_o- i-	AFB_p- i-	AFB_q- i-	AFB_r- i-	AFB_s- i-	AFB_t- i-	AFB_u- i-	AFB_v- i-	AFB_w- i-	AFB_x- i-	AFB_y- i-	AFB_z- i-	AFB_0- i-	AFB_1- i-	AFB_2- i-	AFB_3- i-	AFB_4- i-	AFB_5- i-	AFB_6- i-	AFB_7- i-	AFB_8- i-	AFB_9- i-	AFB_10- i-	AFB_11- i-	AFB_12- i-	AFB_13- i-	AFB_14- i-	AFB_15- i-	AFB_16- i-	AFB_17- i-	AFB_18- i-	AFB_19- i-	AFB_20- i-	AFB_21- i-	AFB_22- i-	AFB_23- i-	AFB_24- i-	AFB_25- i-	AFB_26- i-	AFB_27- i-	AFB_28- i-	AFB_29- i-	AFB_30- i-	AFB_31- i-	AFB_32- i-	AFB_33- i-	AFB_34- i-	AFB_35- i-	AFB_36- i-	AFB_37- i-	AFB_38- i-	AFB_39- i-	AFB_40- i-	AFB_41- i-	AFB_42- i-	AFB_43- i-	AFB_44- i-	AFB_45- i-	AFB_46- i-	AFB_47- i-	AFB_48- i-	AFB_49- i-	AFB_50- i-	AFB_51- i-	AFB_52- i-	AFB_53- i-	AFB_54- i-	AFB_55- i-	AFB_56- i-	AFB_57- i-	AFB_58- i-	AFB_59- i-	AFB_60- i-	AFB_61- i-	AFB_62- i-	AFB_63- i-	AFB_64- i-	AFB_65- i-	AFB_66- i-	AFB_67- i-	AFB_68- i-	AFB_69- i-	AFB_70- i-	AFB_71- i-	AFB_72- i-	AFB_73- i-	AFB_74- i-	AFB_75- i-	AFB_76- i-	AFB_77- i-	AFB_78- i-	AFB_79- i-	AFB_80- i-	AFB_81- i-	AFB_82- i-	AFB_83- i-	AFB_84- i-	AFB_85- i-	AFB_86- i-	AFB_87- i-	AFB_88- i-	AFB_89- i-	AFB_90- i-	AFB_91- i-	AFB_92- i-	AFB_93- i-	AFB_94- i-	AFB_95- i-	AFB_96- i-	AFB_97- i-	AFB_98- i-	AFB_99- i-	AFB_100- i-	AFB_101- i-	AFB_102- i-	AFB_103- i-	AFB_104- i-	AFB_105- i-	AFB_106- i-	AFB_107- i-	AFB_108- i-	AFB_109- i-	AFB_110- i-	AFB_111- i-	AFB_112- i-	AFB_113- i-	AFB_114- i-	AFB_115- i-	AFB_116- i-	AFB_117- i-	AFB_118- i-	AFB_119- i-	AFB_120- i-	AFB_121- i-	AFB_122- i-	AFB_123- i-	AFB_124- i-	AFB_125- i-	AFB_126- i-	AFB_127- i-	AFB_128- i-	AFB_129- i-	AFB_130- i-	AFB_131- i-	AFB_132- i-	AFB_133- i-	AFB_134- i-	AFB_135- i-	AFB_136- i-	AFB_137- i-	AFB_138- i-	AFB_139- i-	AFB_140- i-	AFB_141- i-	AFB_142- i-	AFB_143- i-	AFB_144- i-	AFB_145- i-	AFB_146- i-	AFB_147- i-	AFB_148- i-	AFB_149- i-	AFB_150- i-	AFB_151- i-	AFB_152- i-	AFB_153- i-	AFB_154- i-	AFB_155- i-	AFB_156- i-	AFB_157- i-	AFB_158- i-	AFB_159- i-	AFB_160- i-	AFB_161- i-	AFB_162- i-	AFB_163- i-	AFB_164- i-	AFB_165- i-	AFB_166- i-	AFB_167- i-	AFB_168- i-	AFB_169- i-	AFB_170- i-	AFB_171- i-	AFB_172- i-	AFB_173- i-	AFB_174- i-	AFB_175- i-	AFB_176- i-	AFB_177- i-	AFB_178- i-	AFB_179- i-	AFB_180- i-	AFB_181- i-	AFB_182- i-	AFB_183- i-	AFB_184- i-	AFB_185- i-	AFB_186- i-	AFB_187- i-	AFB_188- i-	AFB_189- i-	AFB_190- i-	AFB_191- i-	AFB_192- i-	AFB_193- i-	AFB_194- i-	AFB_195- i-	AFB_196- i-	AFB_197- i-	AFB_198- i-	AFB_199- i-	AFB_200- i-	AFB_201- i-	AFB_202- i-	AFB_203- i-	AFB_204- i-	AFB_205- i-	AFB_206- i-	AFB_207- i-	AFB_208- i-	AFB_209- i-	AFB_210- i-	AFB_211- i-	AFB_212- i-	AFB_213- i-	AFB_214- i-	AFB_215- i-	AFB_216- i-	AFB_217- i-	AFB_218- i-	AFB_219- i-	AFB_220- i-	AFB_221- i-	AFB_2
---------	--------------------------	-----	-----	----	-----------	-----	----------	------------	------------	-------------	-----------	-------------	--------	------------	-------------	-------------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------------	-------

[illegible]

[illegible]

[illegible]

NOTE

`filter()` (and many other `tidyverse` functions) return a data frame. In the `tidyverse`, these are called `tibble()` and behave slightly different to regular data frames. For our purposes however, these differences are not important.

4.4 The pipe %>% for combining commands

The filtering using `filter()` is very useful, but you can see that the commands can become very long when you have many filters. Also, trying out many different filters to see what they do with the data can be cumbersome. This where `%>%` comes in really handy.

The “pipe” `%>%` (keyboard shortcut: `Ctrl+Shift+M`) simply passes the result of one function to the next function. For the next function, one does not have to specify the data frame. Let’s see an example.

```
# this is how we filtered our data frame earlier
filter(be, OWS_egg == 1)
```

[illegible]

[illegible]

		WSP_Fe-																					
con-		GEN_Av- male_av-																FM_Av-		AFB_Av-		AFB_an-	
fam- serv- con-	OWS_Sup-	er-WSP_Fe-																ELT_sin-		er-		ey- i- AFB_min-	
spec- range- size- OWS- WSP- er- WSP_Fe-	ELT- sin- er- ey- i- AFB_min-	AFB_max- AFB_min- A																					

[illegible]

[illegible]

	con- fam- spec	ser- rang	con- size	OWS var	SVS cov	WSP page	GEN age	Av- can	WSP H1	Fe- H2	Av- H3	male_av- er-WSP	Fe- H4	ELT H5	sin- H6	er- H7	FM_Av- H8	AFB_Av- ey- H9	AFB_an- i- H10	AFB_min- power H11											
Favosites niusiaceae cus	Ly775	5	1	0	0	0	1.00	0.01	0.00	0.03	2.6	0.40	80	0	0	1	1	0	0	0	200	500	2	0	0	0	1	0	1	1	
Gelidium troperidaceae	He655	5	1	1	1	1	3.00	0.03	0.00	0.03	2.0	0.28	90	1	0	0	1	0	0	0	190	900	0	1	0	0	0	0	0	1	
Goniophlebium uledae	He655	5	1	1	1	1	3.00	0.03	0.00	0.06	1.53	0.57	70	0	1	1	1	0	0	300	1700	0	4	1	0	0	0	0	0	0	
He655	807	5	1	1	0	0	1.00	0.01	0.00	0.03	1.6	0.33	31	1	0	0	1	0	0	0	220	305	3	1	0	0	0	0	0	0	
Is-Nyctiphanes so-i- ria	Ly765	5	1	1	1	1	2.00	0.03	0.02	0.04	2.5	0.40	81	1	0	0	1	0	0	0	275	500	6	1	0	0	0	0	0	1	
Ladosia boriidae	Ly825	5	1	0	0	0	1.00	0.01	0.00	0.02	7.6	0.70	71	0	0	1	1	0	0	100	1500	0	0	1	0	0	0	0	0	0	
Ladosia descabridae	Ly865	5	1	1	1	0	3.00	0.03	0.00	0.03	3.6	0.05	80	1	0	0	1	0	0	0	270	900	0	1	0	0	0	0	0	0	
Lychnis caenidae	Ly685	5	4	1	1	0	0	1.00	0.01	0.00	0.30	5	0.28	91	1	0	0	1	0	0	0	250	405	1	1	0	0	0	0	0	0
Lychnis dracachidae	Ly405	5	1	0	0	0	1.00	0.01	0.00	0.03	9.6	1.00	00	1	0	0	NAN	NAN	500	1000	0	300	0	1	0	0	0	0	0	1	
Lychnis dracachidae	Ly575	5	4	1	0	0	0	1.50	0.02	0.01	0.03	3.6	1.00	00	1	0	0	1	0	0	100	240	300	0	1	0	0	0	0	1	1
Lychnis dracachidae	Ly505	5	1	0	0	0	0	2.00	0.02	0.00	0.03	4.0	1.00	00	1	0	0	1	0	0	100	900	7.00	1	0	0	0	0	0	0	1
Muscivora peri- idae	He885	5	1	0	0	0	1.00	0.01	0.00	0.03	4.9	0.57	71	1	0	0	1	0	0	0	200	400	2	1	0	0	0	0	0	0	1

[illegible]

[illegible]

[illegible]

```
# Multiple filters are connected by pipes
be %>%
  filter(OWS_egg == 1) %>%
  filter(LEV_ground_layer == 1) %>%
  filter(AFB_honeydew == 1)
```

```
# As always in R, assign the result to a new variable using "<-"
be_filtered <- be %>%
  filter(OWS_egg == 1) %>%
  filter(LEV_ground_layer == 1) %>%
  filter(AFB_honeydew == 1)
```

4.5 Sort by column with arrange()

This doesn't change the dataframe itself, it simply orders the columns (similar to the sort function in Excel):

```
# sort by age (ascending) and weight (descending)
be %>%
  arrange(conserv.eu, -range.size)
```

	con-	GEN_Av-	WSP_Fe-	FM_Av-	AFB_Av-	AFB_an-	AFB_min-	AFB_max-	AFB_power
species	range	size	OWS	OWS	OWS	OWS	OWS	OWS	OWS
Tomias	3	n6	NANANAN	1.00	0.01	0.03	2.5	1.00	NANANANANANANANAN
geliidae									
CoPie	1	2	1	0	1	0	0	2.50	0.03
l- dae									
ias_myr-									
mi-									
done									
Phy	5	2	2	0	1	0	0	1.00	0.01
garidae									
ion									
LepPie	4	2	0	0	1	0	2.00	0.02	0.04
tidae									
sei									
BolNympha	2	0	1	0	0	1.00	0.01	0.03	2.5
ria_i-im-									
probae									
TuLy	3	2	2	0	0	1	0	1.50	0.02
ranidae									
ica									
Agly	2	2	2	0	1	0	0	1.00	0.01
adesidae									
lichi									
Polly	2	2	2	0	1	0	0	1.00	0.01
omidae									
tus_humedasae									
GolPie	2	0	0	0	0	3.00	0.03	0.06	1.53
sis dae									

[illegible]

[illegible]

species	range	size	con-	fam-	ser-	con-	OVS	WSp	sup-	GEN_Av-	male_av-	er-WSP_Fe-	WSP_Fe-	ELT_sin-	er-	FM_Av-	AFB_Av-	AFB_an-	AFB_min-	power
Lys-5751	4	1	0	0	0	1.50	0.02	0.1	0.033	0	1.00	0.01	0	0	0	1	0	0	0	1
dracacoidae																				
Li-Ny-5261	4	0	1	0	0	1.00	0.01	0.0	0.077	0.51	0.57	0.70	0	0	1	1	0	0	0	1
men-i-																				
tis-dep-																				
uli																				
Ly-Ly-5195	4	0	1	0	0	1.00	0.01	0.0	0.034	0	0.57	0.71	1	0	0	1	0	0	0	0
caeracidae																				
ci-																				
phron																				
Thy-Ho-496	4	0	1	0	0	1.50	0.02	0.1	0.024	0	0.10	0.80	1	0	0	0	1	0	0	0
cuspoiteon																				
idae																				
Me-Ny-4671	4	0	1	0	0	1.50	0.02	0.1	0.035	0.9	0.10	0.26	1	0	0	0	0	1	0	1
taei-di-																				
amidae																				
Hip-Ny-3661	4	0	1	0	0	1.00	0.01	0.0	0.045	0	0.10	0.23	1	0	0	1	0	0	0	1
parhia_statil-																				
i- dae																				
nus																				
Pa-Pa-354	4	1	1	0	0	1.00	0.01	0.0	0.080	0.00	0.20	0.41	0	0	0	1	1	0	0	0
naspil-																				
siusonpollo																				
idae																				
Ph-Hy-354	4	0	1	0	0	1.00	0.01	0.0	0.036	0	0.57	0.71	1	0	0	1	1	0	0	0
gariscidae																				
con																				
Cha-Ny-3531	4	0	1	0	0	1.00	0.01	0.0	0.055	0.06	0.33	0.31	1	0	0	1	1	0	0	1
arai-bri-																				
seidae																				
Pol-Ly-330	4	0	1	0	0	1.50	0.02	0.1	0.032	0	0.20	0.41	1	0	0	1	0	0	0	0
omacidae																				
tus-do-																				
ry-																				
las																				

[illegible]

[illegible]

[illegible]

		WSP_Fe-																	
con-	GEN_Av-	male_av-															FM_Av-	AFB_Av-	AFB_an-
fam- ser-	con-	OVS-	sup-	er-WSP_Fe-	ELT_sin-												er-	ey- i-	AFB_min-
species	range	size	OWS	pag	ag	GEN	CHN	IND	NEP	PHI	THA	THA	THA	THA	THA	THA	THA	THA	THA
Hip-Ny-NA1	4	0	1	0	0	1.00	01.00	052.0	0.28	NAN	NAN	NAN	NAN	NAN	NAN	NAN	2002.00	NAN	NAN
parhia_sbor-																			
dondae																			
Man-Ny-NA1	4	0	1	0	0	1.00	01.00	044.5	0.28	91	1	0	0	1	1	0	0	0	0
iola_hali-																			
cardae																			
nas-																			
sus																			
Pol-Ly-NA1	4	NAN	NAN	NAN	1.00	01.00	030.5	1.00	NAN	NAN	NAN	NAN	NAN	NAN	NAN	NAN	2002.00	NAN	NAN
omnae																			
tus_nepho-																			
hip-																			
ta-																			
menos																			
Pol-Ly-1476	5	0	1	0	0	2.50	04.03	028.51	0.10	41	1	0	0	1	0	0	0	0	1
omnae																			
tus_icarus																			
Pie-Pie-1476	5	0	0	1	0	2.00	03.02	040.06	0.10	41	1	0	0	1	0	0	0	0	0
dae																			
Van-Ny-1371	5	0	0	0	1	2.00	03.02	054.06	0.08	30	1	0	0	1	0	0	0	0	0
dui-																			
dae																			
Co-Ny-1371	5	0	0	0	1	2.00	03.02	028.00	0.09	60	1	0	0	1	0	0	0	0	0
philus																			
dae																			
Pie-Pie-1363	5	0	0	1	0	2.00	03.02	044.51	0.10	81	1	0	0	0	1	0	0	0	1
paedae																			
Pie-Pie-1353	5	0	0	1	0	2.50	03.01	057.02	0.10	51	1	0	0	0	1	1	0	0	0
si- dae																			
cae																			
Van-Ny-1371	5	0	0	0	1	2.00	03.02	059.53	0.30	60	1	0	0	1	0	0	0	0	0
lanta																			
dae																			
Ly-Ly-1340	5	0	1	0	0	2.00	03.02	026.51	0.33	31	0	0	0	1	0	0	0	0	1
caenidae																			
Man-Ny-1289	5	0	1	0	0	1.00	01.00	041.9	0.28	91	1	0	0	0	1	0	0	0	0
iola_ju-																			
rtidae																			

		WSP_Fe-																	FM_Av-		AFB_AFB_an-																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																															
con-		GEN_Av-																	er-		ey- i- AFB_min-																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																															
fam-	serv-	con-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-	OWS_Sap-

[illegible]

[illegible]

species	range	size	con-	fam-	ser-	con-	OVS	WSP	sup-	GEN_Av-	male_av-	er-WSP_Fe-	WSP_Fe-	ELT_sin-	er-	FM_Av-	AFB_Av-	AFB_an-	AFB_min-	power
FavLy-775	5	1	0	0	0	1.00	01.00	032.0	0.40	80	0	0	1	1	0	0	0	1	0	1
niusacridae																				
cus																				
PleLy-774	5	1	0	0	0	1.50	02.01	024.5	0.04	10	1	0	0	1	0	0	1	0	0	0
be-caenidae																				
jus_idas																				
Ly-Ly-756	5	0	1	0	0	1.00	01.00	027.0	0.40	81	0	0	0	1	0	0	1	0	0	0
caenidae																				
gau-																				
reae																				
IphPa-755	5	0	0	1	0	2.00	03.02	077.06	0.10	51	1	0	0	1	0	0	0	1	0	0
i- pil-																				
cliden-po-																				
dalidae																				
ius																				
CuLy-753	5	0	1	0	0	1.50	02.01	023.0	0.10	50	1	0	0	1	0	0	0	0	0	1
pidcaenidae																				
imus																				
CoenLy-761	5	0	1	0	0	1.25	01.50	537.0	0.16	90	1	0	0	1	1	0	0	1	0	0
ca-i-																				
niadae																				
AriLy-700	5	0	1	0	0	2.00	03.02	025.0	0.10	21	1	0	0	1	0	0	0	0	0	1
ciaacridae																				
BreLy-699	5	1	1	0	0	1.00	01.00	037.8	0.20	41	1	0	0	1	1	0	0	1	0	0
this_ino																				
dae																				
CoPic-682	5	0	1	1	0	2.50	03.01	045.00	0.20	41	1	0	0	1	0	0	0	0	0	0
l- dae																				
ias_hyale																				
Ly-Ly-673	5	0	1	0	0	2.00	03.02	030.0	0.13	61	1	0	0	1	0	0	0	0	0	0
caenidae																				
MeLy-671	5	0	1	0	0	1.50	02.01	035.9	0.28	91	1	0	0	0	0	1	0	0	0	0
taen_cinxia																				
dae																				
TheLy-628	5	1	0	0	0	1.00	01.00	035.00	0.12	90	0	1	0	1	0	0	0	1	0	0
clacanthidae																				
tu-																				
lae																				

Species	con- fam- serv- con- OWS	size	var	page	agg	Vag	Cach	Hag	Mag	WSP_Fe- male_av- er-WSP_Fe-	RHS	ELE	VEN	MTC	MTC	ELT_sin- er-	FM_Av- er-	AFB_Av- ey-	AFB_i- AFB_min-	AFB_an- AFB_max-	AFB_all- power						
																						WSP_Fe- male_av- er-WSP_Fe-	ELT_sin- er-	FM_Av- er-	AFB_Av- ey-	AFB_i- AFB_min-	AFB_an- AFB_max-
SatLyri555_w- alb-caenidae	1	0	0	0	1.00	0.01	0.00	0.30	0.0	0.06	50	0	1	1	1	0	0	0	1700	000	1	0	0	1	0	0	0
PolLy544_5 om-caenidae	5	0	1	0	0	1.50	0.02	0.01	0.31	0.0	0.20	41	1	0	0	1	0	0	10200	000	1	0	0	0	0	0	1
tus_aman- dus																											
EuNy538_5 phidryas_aur- rinidae	5	0	1	0	0	1.00	0.01	0.00	0.36	0.53	0.10	21	1	0	0	0	0	1	2600	053	1	0	0	0	0	0	0
ApNy528_5 atura_ilia dae	5	0	1	0	0	1.50	0.02	0.01	0.68	0.8	0.26	70	0	1	1	1	0	0	50185	053	1	0	0	1	1	1	1
LiNy526_5 mei- tis-caemilla	5	0	1	0	0	1.00	0.01	0.00	0.56	0.8	0.28	90	0	1	1	1	0	0	0	1500	002	1	1	1	1	1	1
MeNy525_5 taei-phoebe dae	5	0	1	0	0	1.50	0.02	0.01	0.44	0.51	0.16	40	1	0	0	0	0	1	2700	055	1	0	0	0	0	0	1
PyNy515_5 ro-i- niadacthonus	5	0	1	0	0	1.00	0.01	0.00	0.36	0.0	0.08	11	1	0	0	1	0	0	2300	053	1	0	1	0	0	0	0
HaRe495_5 cina- dinidae	5	0	0	1	0	1.00	0.01	0.00	0.32	0.3	0.57	71	1	0	0	0	1	0	0	1700	000	1	0	0	0	0	0
CoPie494_5 l- ias_al- facarien- sis	5	0	1	0	0	2.50	0.03	0.01	0.43	0.3	0.16	71	1	0	0	1	0	0	2600	051	1	0	0	0	0	0	0
SatLyri485_5 caenidae	5	0	0	0	0	1.00	0.01	0.00	0.30	0.0	0.16	70	0	1	0	1	1	0	2000	051	1	0	1	1	0	0	0
AgLy480_5 adesacomp- tilete	5	0	1	0	0	1.00	0.01	0.00	0.25	0.8	0.20	81	1	0	0	1	0	0	2800	000	1	0	1	0	0	1	0
BolNy478_5 ria i-aquilonar- is dae	5	0	1	0	0	1.00	0.01	0.00	0.36	0.8	0.28	91	1	0	0	1	0	0	10190	002	1	0	1	0	0	0	0

Species	Family	con- serva- tion status	size	OVS	SVL	sup- ra- umbil-	GEN	Av- age	WSP_Fe- male	av- er- WSP_Fe-	ELT	sin- er-	FM_Av- er-	AFB_Av- er-	AFB_i-	AFB_min- power	WSP_Fe-												
																	1	2	3	4	5	6	7	8	9	10	11	12	13
CuLy475	caeciliidae	5	0	1	0	0	2.00	0.02	0.00	0.29	0.07	20	1	0	0	1	0	0	0	1600	000	1	0	0	0	0	0	0	
PyHe476	gustavidae	5	0	1	0	0	1.00	0.01	0.00	0.28	0.35	40	1	0	0	1	0	0	0	2300	053	1	0	0	0	0	0	0	
EuLy441	mecaenidae	5	0	1	0	0	1.00	0.01	0.00	0.28	0.20	81	1	0	0	1	0	0	0	2400	000	1	0	0	0	0	0	0	
ArLy435	ciaacnidae	5	0	1	0	0	1.00	0.01	0.00	0.27	0.05	61	1	0	0	1	0	0	0	2200	000	1	0	0	0	0	0	0	
CoPie117	l- dae	5	1	1	0	0	1.00	0.01	0.00	0.47	0.00	70	1	1	0	0	1	0	0	0	2500	000	1	0	1	0	0	0	0
BrNy416	ias_palaeno	5	0	1	0	0	1.00	0.01	0.00	0.73	0.04	0.20	41	1	0	0	0	1	0	0	2300	053	1	0	0	0	1	1	1
LiNy410	meini- tis_dae	5	0	1	0	0	2.00	0.03	0.02	0.50	0.05	77	0	1	0	1	0	0	0	1950	004	1	0	1	1	1	1	1	
SatLy408	caenidae	5	0	0	0	0	1.00	0.01	0.00	0.31	0.13	40	0	1	0	1	0	0	0	1000	003	00	1	0	1	1	0	0	0
LyLy383	caenidae	5	0	1	0	0	2.00	0.03	0.02	0.37	0.04	0.81	1	0	0	1	1	0	0	1000	051	1	0	1	1	0	0	0	
SpHe380	par- to-idae	5	0	1	0	0	2.00	0.03	0.02	0.24	0.10	00	1	1	0	0	1	0	0	0	1700	004	1	0	0	0	0	0	0

species	range	size	con-	fam-	serv-	OVS	WSP	sup-	GEN_Av-	male_av-	WSP_Fe-										ELT_sin-	er-	FM_Av-				AFB_Av-				AFB_min-	power
											er	WSP_Fe-	ELT	sin	er	FM_Av-	AFB_Av-	AFB_min-	power													
Zerthia	125	5	0	0	1	0	1.25	0.15	0.54	5.0	0.57	771	1	0	0	1	1	0	0	1500	004	1	0	0	0	0	0	0	0	0	0	0
thiapi-																																
min-																																
idae																																
Chara	125	5	0	1	0	0	1.50	0.02	0.10	79.5	0.10	20	0	0	1	1	1	0	0	2400	053	0	0	0	0	0	0	1	1	1	1	1
sius-																																
dae																																
Pieris	125	5	0	0	1	0	2.00	0.03	0.02	0.42	0.40	81	1	0	0	1	0	0	0	2200	055	1	0	0	0	0	0	0	0	1	1	1
gan-																																
Erebia	125	5																														
biai-em-																																
bladae																																
Pieris	125	5	0	0	1	0	1.50	0.02	0.10	38.5	0.20	41	0	0	0	1	1	0	0	6000	000	1	0	0	0	0	0	0	0	0	0	0
onidae																																
Bolita	125	5	0	1	0	0	1.00	0.01	0.00	39.6	0.35	41	1	0	0	1	0	0	0	2500	000	1	0	0	0	0	0	0	0	0	0	0
ria i-na-																																
pacidae																																
Melipotis	125	5	0	1	0	0	1.00	0.01	0.00	49.6	0.13	61	1	0	0	1	0	0	0	1600	051	1	0	0	0	0	0	0	0	0	0	0
nari-																																
giada-																																
ach-																																
esis																																
Melipotis	125	5	0	1	0	0	1.00	0.01	0.00	55.00	0.12	01	1	0	0	1	0	0	0	5000	000	1	0	0	0	0	0	0	0	0	0	0
nari-																																
giada-																																
us-																																
siae																																
Melipotis	125	5	0	1	0	0	1.50	0.02	0.10	39.04	0.12	01	1	0	0	0	1	1	0	2000	053	1	0	0	0	0	0	0	0	1	1	1
taei deione																																
dae																																
Coenonympha	125	5	0	1	0	0	1.00	0.01	0.00	31.6	0.16	90	1	0	0	1	0	0	0	1000	000	1	0	0	0	0	0	0	1	0	0	0
i-																																
dae																																
Satyrus	105	5	1	0	0	0	1.00	0.01	0.00	32.9	0.70	70	0	1	1	1	0	0	0	1000	000	1	0	0	0	0	0	0	0	0	0	0
culicaenidae																																
Pyrausta	105	5	0	1	0	0	2.50	0.03	0.10	25.6	0.57	771	1	0	0	1	0	0	0	2800	000	1	0	0	0	0	0	0	0	1	1	1
guspon-																																
pondidae																																

	WSP_Fe-												FM_Av-												AFB_AFB-an-																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																											
	con-				GEN_Av-				male_av-								er-WSP_Fe-				ELT_sin-				er-				ey-				i- AFB_min-																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																			
	fam-	serv-	con-	OWS	OWS	Sap-	OWS	CEN	CEN	M	RHS	EVEN	MENTAL	HEAT	HAUT	THIGH	AID	DEEP	AFB_h-	AFB_l-	power																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																															

[illegible]

[illegible]

[illegible]

	WSP_Fe-													FM_Av-													AFB_AFB-an-																										
	con-	GEN_Av- male_av-													ELT_sin-													er-													ey- i- AFB_min-												
	fam-	ser-	con-	OVS	SV	sup-	er-WSP_Fe-													er-													AFB_min-																				
species	range	size	OVS	SV	sup-	GEN	Av	male	av	WSP	Fe	ELT	sin	er	ey	i	AFB	min	power																																		
AriLy434	4	5	0	1	0	0	2.00	0.03	0.02	0.28	8.0	0.28	90	1	0	0	1	0	0	30	2100	000	1	0	0	0	0	0	0																								
cia	canidae																																																				
teros																																																					
EreNy425	5	0	1	0	0	0	1.00	0.01	0.00	0.39	0.00	0.77	NAN	NAN	NAN	NAN	NAN	NAN	NAN	45	2150	002	1	0	0	0	0	0	0																								
biai	ot-																																																				
ton	dae																																																				
HipNy425	5	NAN	NAN	NAN	NAN	1.00	0.01	0.00	0.53	5	1.00	001	1	0	0	1	0	0	0	250	002	1	0	0	0	0	0	0	0																								
par	hia_vol-																																																				
gendae																																																					
sis																																																					
MeNy415	5	0	1	0	0	0	1.00	0.01	0.00	0.34	8	0.28	91	1	0	0	0	1	1	120	003	000	1	0	0	0	0	0	0																								
taei	_varia																																																				
dae																																																					
LyLy405	5	5	1	0	0	0	1.00	0.01	0.00	0.39	6	1.00	00	1	0	0	NAN	NAN	NAN	50	1003	000	1	0	0	0	0	0	1																								
drac	hidae																																																				
bi-																																																					
cans																																																					
CoNy395	5	0	1	0	0	0	1.00	0.01	0.00	0.33	0	0.35	40	1	0	0	1	0	0	75	1750	51	1	0	0	0	0	0	0																								
sis	i-																																																				
dae																																																					
EreNy385	5	0	1	0	0	0	1.00	0.01	0.00	0.47	6	0.35	41	0	0	0	NAN	NAN	NAN	110	003	000	1	0	0	0	0	0	0																								
biai	mon-																																																				
tandae																																																					
BoNy375	5	0	1	0	0	0	1.00	0.01	0.00	0.36	8	1.00	00	1	0	0	1	0	0	75	1850	02	1	0	0	0	0	0	0																								
ria	i-graeca																																																				
dae																																																					
CaLy375	5	0	0	1	0	0	1.00	0.01	0.00	0.35	3	0.25	NAN	NAN	NAN	NAN	1	0	0	100	003	000	1	0	0	0	0	0	0																								
loph	hidae																																																				
LyLy375	5	0	1	0	0	0	2.00	0.02	0.00	0.29	0	1.00	NAN	NAN	NAN	NAN	NAN	NAN	NAN	160	002	1	0	0	0	0	0	0	0																								
caer	enidae																																																				
tomana																																																					
AnPi365	5	5	0	0	1	0	1.00	0.01	0.00	0.33	0	0.48	NAN	NAN	NAN	NAN	1	0	0	0	230	004	1	0	0	0	0	0	0																								
tho	charis_gruneri																																																				
EreNy365	5	0	1	0	0	0	0.50	0.50	0.50	0.45	5	0.57	NAN	NAN	NAN	NAN	NAN	NAN	NAN	300	200	2.00	NAN	NAN	NAN	NAN	NAN	NAN																									
biai	-disa																																																				
dae																																																					

[illegible]

[illegible]

	con-	GEN_Av-	WSP_Fe-	FM_Av-	AFB_AFB-an-																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																			
fam- serv- con- OWS_Sup-	er-WSP_Fe-	ELT_sin-	er-	ey- i- AFB_min-																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																				
species	range	size	env	page	age	Ca	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co	Co

[illegible]

[illegible]

[illegible]

		WSP_Fe-																							
con-		GEN_Av- male_av-																FM_Av-		AFB_Av-		AFB_an-			
fam- sercon-		er-WSP_Fe-																ELT_sin-		er-		ey- i-		AFB_min-	
speciesrangexizeOVS- pag- OVS- C- H- C- H- M- B- S- E- A- T- E- M- H- H- H- A- T- E- H- H- M- A- B- B- A- E- B- h- s- a- f-																									
Tan-4	5 5	NANANANA	3.00	03.00	021.0	0.35	40	0	1	0	1	0	0	0	250	6.00	NANANANANANA								
cusca- midphar- tus																									
HipN-2pha-	5 0 1 0 0	1.00	01.00	051.06	0.28	91	1	0	0	1	0	0	0	1000	02	1 0 0 0 0 1 0									
par- hia_pel- lu-dae cida																									
Man-2pha-	5 0 1 0 0	1.00	01.00	041.9	0.28	91	1	0	0	1	1	0	0	800	5.00	1 0 0 0 0 0 0									
iola- chia dae																									
HipN-2pha-	5 NANANANA	1.00	01.00	051.06	0.28	91	NANANANANANANANA	250	652.02	NANANANANANANA															
par- hia_chris- tend- eri																									
Psd-1	5 5 0 1 0 0	1.00	01.00	022.0	1.00	00	1	0	0	NANANA	700	800	4.00	1 0 0 0 0 0 0											
dop- plaid- idae barba- giae																									
Cal-H-NA	5 5 NANANANA	3.00	03.00	030.8	1.00	NANANANANANANANA	0	250	000	NANANANANANANA															
chaperi- o- idae dus_tripoli- nus																									
CyLy-NA	5 5 1 1 1 1	3.00	03.00	027.5	0.28	90	1	0	0	0	1	0	200	300	ANA	1 0 0 0 0 0 0									
cly- rius- idae bianus																									
EreN-NA	5 0 1 0 0	1.00	01.00	038.8	1.00	NANANANANANANANA	150	000	200	1 0 0 0 0 0 0															
biai-ron- doudae																									
EuPienA	5 5 0 0 1 0	2.50	03.01	034.0	0.03	61	1	0	0	1	1	0	0	300	051	1 0 0 0 0 0 0									
chla- echar- lonia																									
EuPienA	5 5 0 0 1 0	2.00	02.00	040.8	0.28	91	NANANANANA	1 0 0	170	00	4.00	1 0 0 0 0 0 0													
chla- e-ev- ersi																									
EuPienA	5 5 NANANANA	2.00	02.00	040.8	1.00	NANANANANANANANA	50	190	000	NANANANANANANA															
chla- e-granca- narien- sis																									

		WSP_Fe-																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																					
--	--	---------	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--

[illegible]

		WSP_Fe-																	
		con-	GEN_Av-		male_av-		WSP_Fe-		ELT_sin-		FM_Av-		AFB_Av-		AFB_an-				
species	range	con-	OWS	OWS	OWS	OWS	OWS	OWS	OWS	OWS	OWS	OWS	OWS	OWS	OWS	OWS	OWS	OWS	
Pyris	He-29	NAN	0	1	0	0	1.00	0.01	0.00	0.28	3	0.5	NAN	NAN	NAN	NAN	NAN	NAN	
guspe	frulquieri																		
idae																			
KrLy	16 4	NAN	0	1	0	0	1.00	0.01	0.00	0.29	3	1.00	0	0	1	0	0	0	
ta-caenidae																			
nia_py-																			
laon																			
Dan	15	NAN	0	0	0	1	3.00	0.03	0.00	0.95	53	0.50	0	0	1	0	0	0	
ip-i-																			
pusdae																			
Van	12	NAN																	
ginien-																			
sis dae																			
Coen	11	NAN																	
en-i-																			
talidae																			
Col	10	NAN	0	0	1	0	3.00	0.03	0.00	0.33	6	0.70	0	1	0	0	1	0	
tis dev-																			
agore																			
Hip	9	NAN																	
parhia_																			
blachieri																			
dae																			
KrLy	6 4	NAN	0	1	0	0	1.00	0.01	0.00	0.31	6	0.70	1	1	0	0	NAN	NAN	
ta-caenidae																			
nia trappi																			
Ly-Ly	6	NAN																	
caenidae																			
caenidae																			
Pse	6	NAN	0	1	0	0	1.00	0.01	0.00	0.50	6	1.00	0	0	0	0	0	0	
dod-																			
hazdae																			
ara_mer-																			
curius																			
Pro	5	NAN	0	1	0	0	1.00	0.01	0.00	0.46	6	1.00	0	1	1	0	15	14	
tere-																			
biadphegea																			
Pse	5	NAN	0	1	0	0	1.00	0.01	0.00	0.50	6	NAN	NAN	NAN	NAN	NAN	NAN	NAN	
dod-																			
hazdae																			
ara williamsi																			

[illegible]

[illegible]

[illegible]

```
# choose which columns to keep
be %>%
  select(species, range.size, conserv.eu, FM Average, WSP_Female average)
```

species	range.size	conserv.eu	FM_Average	WSP_Female_average
Aglais_io	1136	5	7.5	53.5
Aglais_urticae	1271	5	7.5	48.0
Agriades_aquilo	21	5	3.0	23.0
Agriades_dardanus	2	4	4.0	25.0
Agriades_glandon	45	5	2.0	23.0
Agriades_optilete	480	5	3.0	25.0
Agriades_orbitulus	63	5	3.0	26.0
Agriades_pyrenaicus	5	5	NA	25.0
Agriades_zullichii	2	2	3.0	23.5
Anthocharis_cardamines	1245	5	4.0	40.5
Anthocharis_damone	19	5	3.0	39.5
Anthocharis_euphenoides	131	5	3.5	37.0
Anthocharis_gruneri	36	5	4.0	33.0
Apatura_ilia	528	5	4.5	68.0
Apatura_iris	571	5	2.5	70.0
Apatura_metis	47	5	4.0	62.0
Aphantopus_hyperantus	1002	5	3.0	38.5
Aporia_crataegi	894	5	3.5	62.5
Araschnia_levana	626	5	5.0	33.5
Archon_apollinus	13	5	3.5	57.0
Arethusana_arethusa	229	5	3.5	41.0
Argynnis_laodice	100	4	2.0	55.0
Argynnis_pandora	320	5	5.0	72.0
Argynnis_paphia	1024	5	4.0	62.0
Aricia_agestis	700	5	7.0	25.0
Aricia_anteros	43	5	5.0	28.0
Aricia_artaxerxes	435	5	4.0	27.5
Aricia_cramera	138	5	9.5	25.0
Aricia_montensis	32	5	4.0	27.5
Aricia_morronensis	15	5	4.0	28.0
Aricia_nicias	78	5	2.0	25.5
Azanus_jesous	NA	NA	NA	21.5
Azanus_ubaldus	NA	NA	NA	12.5
Boloria_aquilonaris	478	5	2.0	36.0
Boloria_chariclea	22	4	1.5	38.0
Boloria_dia	554	5	5.5	33.0
Boloria_eunomia	311	5	3.0	36.5
Boloria_euphrosyne	963	5	4.0	36.0
Boloria_freija	144	5	2.5	34.5
Boloria_frigga	145	5	2.0	42.5

species	range.size	conserv.eu	FM_Average	WSP_Female_average
Boloria_graeca	37	5	4.0	36.0
Boloria_improba	6	2	2.0	32.5
Boloria_napaea	112	5	3.0	39.0
Boloria_pales	101	5	3.0	36.0
Boloria_polaris	18	3	2.5	41.5
Boloria_selene	998	5	3.5	36.5
Boloria_thore	79	5	2.0	44.0
Boloria_titania	141	5	3.0	39.5
Borbo_borbonica	1	NA	4.5	29.0
Brenthis_daphne	323	5	3.5	47.0
Brenthis_hecate	190	5	3.0	40.0
Brenthis_ino	699	5	2.5	37.0
Brintesia_circe	416	5	4.5	73.0
Cacyreus_marshalli	33	NA	8.0	23.5
Callophrys_avis	37	5	3.0	35.5
Callophrys_rubi	1192	5	6.0	23.0
Carcharodus_alceae	615	5	8.0	30.0
Carcharodus_baeticus	56	5	4.0	30.0
Carcharodus_floccifera	277	5	4.5	30.0
Carcharodus_lavatherae	146	4	4.0	31.0
Carcharodus_orientalis	54	5	4.5	29.0
Carcharodus_stauderi	2	NA	8.0	30.0
Carcharodus_tripolinus	NA	5	7.0	30.0
Carterocephalus_palaemon	585	5	2.5	27.0
Carterocephalus_silvicola	222	5	2.5	25.0
Catopsilia_florella	NA	NA	11.0	50.0
Celastrina_argiolus	1190	5	7.0	27.0
Charaxes_jasius	127	5	6.5	79.5
Chazara_briseis	353	4	4.0	55.0
Chazara_prieuri	8	5	2.0	60.0
Coenonympha_arcania	736	5	3.5	37.0
Coenonympha_corinna	17	5	4.5	29.0
Coenonympha_dorus	107	5	3.0	31.0
Coenonympha_gardetta	61	5	4.0	31.0
Coenonympha_glycerion	574	5	3.0	28.0
Coenonympha_hero	173	3	3.5	30.0
Coenonympha_leander	27	5	3.5	33.0
Coenonympha_oedippus	52	5	3.0	38.0
Coenonympha_orientalis	11	NA	3.0	30.5
Coenonympha_pamphilus	1371	5	7.0	28.0

species	range.size	conserv.eu	FM_Average	WSP_Female_average
Coenonympha_rhodopenis	39	5	2.5	33.0
Coenonympha_thyrsis	9	5	5.0	32.0
Coenonympha_tullia	553	5	2.0	31.5
Colias_alfacariensis	494	5	6.5	43.5
Colias_aurorina	14	5	5.0	55.0
Colias_caucasica	13	5	3.0	52.0
Colias_chrysotheme	36	3	7.0	44.0
Colias_crocea	1028	5	7.0	48.0
Colias_erate	138	5	4.5	49.0
Colias_hecla	24	4	3.0	43.0
Colias_hyale	682	5	6.5	45.0
Colias_myrmidone	94	1	5.5	47.0
Colias_palaeno	417	5	3.0	47.0
Colias_phicomone	83	4	3.5	45.0
Colias_tyche	6	5	3.0	45.0
Colotis_evagore	10	NA	9.0	33.0
Cupido_alcetas	146	5	7.0	29.0
Cupido_argiades	478	5	6.0	29.0
Cupido_decoloratus	67	5	5.0	25.0
Cupido_lorquini	17	5	3.0	25.0
Cupido_minimus	753	5	6.0	23.0
Cupido_osiris	131	5	6.0	27.0
Cyaniris_semiargus	930	5	6.0	29.0
Cyclyrius_webbianus	NA	5	NA	27.5
Danaus_chrysippus	38	NA	10.5	70.0
Danaus_plexippus	15	NA	11.0	95.5
Erebia_aethiopella	10	5	2.0	38.0
Erebia_aethiops	326	5	3.5	47.0
Erebia_albergana	65	5	3.0	43.0
Erebia_calcarius	9	5	3.0	38.0
Erebia_cassioides	65	5	3.0	37.0
Erebia_christi	1	3	2.5	38.0
Erebia_claudina	8	4	3.0	35.0
Erebia_disa	36	5	2.0	45.5
Erebia_embla	123	5	2.0	47.5
Erebia_epiphron	137	5	2.5	38.0
Erebia_epistygne	24	4	3.0	47.0
Erebia_eriphyle	34	5	3.0	34.0
Erebia_euryale	203	5	3.0	44.0
Erebia_flavofasciata	5	4	3.0	35.0

species	range.size	conserv.eu	FM_Aver- age	WSP_Female_aver- age
Erebia_gorge	92	5	3.0	37.0
Erebia_gorgone	13	5	2.0	41.0
Erebia_hispania	17	5	2.5	38.0
Erebia_lefebvrei	12	5	3.0	44.0
Erebia_ligea	564	5	3.0	41.0
Erebia_manto	89	5	3.0	39.0
Erebia_medusa	361	5	3.5	38.0
Erebia_melampus	56	5	3.0	33.0
Erebia_melas	31	5	3.5	45.0
Erebia_meolans	139	5	4.0	46.0
Erebia_mnestra	22	5	2.0	36.0
Erebia_montana	38	5	3.0	47.0
Erebia_neoridas	48	5	2.5	41.0
Erebia_nivalis	14	5	1.5	32.0
Erebia_oeme	89	5	3.0	42.0
Erebia_orientalis	5	5	2.5	30.5
Erebia_ottomana	42	5	3.0	39.0
Erebia_palarica	5	5	3.0	58.0
Erebia_pandrose	194	5	2.5	40.0
Erebia_pharte	59	5	2.0	36.0
Erebia_pluto	51	5	2.5	45.0
Erebia_polaris	16	5	1.5	39.0
Erebia_pronoe	81	5	3.5	46.0
Erebia_rhodopensis	10	5	1.5	36.5
Erebia_rondoui	NA	5	2.0	38.0
Erebia_scipio	9	5	2.0	48.0
Erebia_sthenno	10	5	2.5	42.0
Erebia_stiria	17	5	3.0	49.0
Erebia_styx	25	5	3.0	51.0
Erebia_sudetica	13	3	2.5	28.5
Erebia_triarius	72	5	4.0	47.0
Erebia_tyndarus	25	5	3.0	35.0
Erebia_zapateri	6	5	3.5	38.0
Erynnis_marloyi	31	5	6.0	31.0
Erynnis_tages	842	5	4.0	30.0
Euchloe_ausonia	352	5	5.0	44.0
Euchloe_bazae	5	3	4.0	34.0
Euchloe_belemia	63	5	5.0	40.0
Euchloe_charltonia	NA	5	8.5	34.0
Euchloe_crameri	197	5	3.0	44.0

species	range.size	conserv.eu	FM_Average	WSP_Female_average
Euchloe_eversi	NA	5	4.0	40.0
Euchloe_grancanariensis	NA	5	7.0	40.0
Euchloe_hesperidum	NA	5	6.0	40.0
Euchloe_insularis	16	5	5.0	34.5
Euchloe_penia	14	5	5.0	34.0
Euchloe_simplonia	32	5	4.0	43.0
Euchloe_tagis	56	5	3.0	35.0
Eumedonia_eumedon	441	5	4.0	28.0
Euphydryas_aurinia	538	5	3.5	36.5
Euphydryas_cynthia	58	5	3.0	37.0
Euphydryas_desfontainii	41	4	3.0	46.0
Euphydryas_iduna	24	4	2.0	41.0
Euphydryas_intermedia	28	5	3.0	40.0
Euphydryas_materna	210	5	2.5	41.5
Fabriciana_adippe	888	5	3.5	53.5
Fabriciana_elisa	8	5	3.0	49.0
Fabriciana_niobe	646	4	2.5	49.5
Favonius_quercus	777	5	5.0	32.0
Freyeria_trochylus	19	5	8.0	16.5
Gegenes_nostrodamus	65	5	9.0	32.0
Gegenes_pumilio	76	5	9.0	32.0
Glaucopsyche_alexis	609	5	5.0	29.0
Glaucopsyche_melanops	103	5	2.0	27.0
Glaucopsyche_paphos	NA	5	NA	29.0
Gonepteryx_cleobule	NA	3	4.0	61.5
Gonepteryx_cleopatra	303	5	7.5	59.0
Gonepteryx_farinosa	50	5	11.0	60.0
Gonepteryx_maderensis	NA	2	9.0	61.5
Gonepteryx_rhamni	1177	5	8.0	58.0
Hamearis_lucina	497	5	6.0	32.5
Hesperia_comma	807	5	3.5	31.0
Heteropterus_morpheus	270	5	3.0	33.5
Hipparchia_aristaeus	13	5	4.5	52.0
Hipparchia_autonoe	NA	NA	2.0	54.0
Hipparchia_azorina	NA	5	5.0	41.0
Hipparchia_bacchus	NA	3	2.0	59.0
Hipparchia_blachieri	9	NA	4.5	52.0
Hipparchia_christenseni	1	5	2.0	51.0
Hipparchia_cretica	9	5	5.5	56.0
Hipparchia_cypriensis	NA	5	9.0	56.5

species	range.size	conserv.eu	FM_Average	WSP_Female_average
Hipparchia_fagi	309	4	4.5	66.0
Hipparchia_fatua	48	5	6.5	57.5
Hipparchia_fidia	92	5	3.5	59.0
Hipparchia_gomera	NA	5	5.0	59.0
Hipparchia_hermione	237	4	2.0	61.0
Hipparchia_leighebi	1	4	4.0	52.0
Hipparchia_maderensis	NA	5	3.0	44.0
Hipparchia_mersina	3	NA	4.5	51.0
Hipparchia_miguelensis	NA	5	4.0	41.0
Hipparchia_neapolitana	NA	5	3.0	NA
Hipparchia_neomiris	9	5	3.0	48.0
Hipparchia_pellucida	2	5	4.0	51.0
Hipparchia_sbordonii	NA	4	2.0	52.0
Hipparchia_semele	814	5	3.5	51.0
Hipparchia_senthes	56	5	7.0	52.0
Hipparchia_statilinus	360	4	4.0	45.0
Hipparchia_syriaca	54	5	5.5	65.0
Hipparchia_tamadabae	NA	5	6.0	55.5
Hipparchia_tilosi	NA	3	3.0	59.0
Hipparchia_volgensis	42	5	4.0	53.5
Hipparchia_wyssii	NA	5	3.0	59.0
Hyponephele_lupina	143	5	4.5	45.0
Hyponephele_lycaon	365	5	3.5	38.0
Iolana_debilitata	NA	NA	5.0	39.0
Iolana_iolas	102	4	4.0	39.0
Iphiclides_feisthamelii	NA	NA	7.5	77.0
Iphiclides_podalirius	755	5	6.0	77.0
Issoria_lathonia	976	5	7.0	42.5
Kirinia_climene	9	5	2.5	47.0
Kirinia_roxelana	86	5	6.5	60.0
Kretania_eurypilus	2	NA	2.0	29.5
Kretania_hesperica	8	5	4.0	30.0
Kretania_psylorita	5	5	2.0	24.5
Kretania_pylaon	16	NA	2.0	29.5
Kretania_sephirus	NA	5	NA	31.0
Kretania_trappi	6	NA	3.0	31.0
Laeosopis_roboris	82	5	3.0	27.0
Lampides_boeticus	361	5	9.0	33.0
Lasiommata_maera	888	5	5.5	47.5
Lasiommata_megera	1053	5	6.5	42.5

species	range.size	conserv.eu	FM_Aver- age	WSP_Female_aver- age
Lasiommata_paramegaera	15	5	6.0	42.5
Lasiommata_petropolitana	352	5	3.5	38.0
Leptidea_duponcheli	67	5	5.0	38.0
Leptidea_juvernica	NA	NA	3.0	42.0
Leptidea_morsei	49	2	5.0	41.5
Leptidea_reali	NA	5	6.0	38.0
Leptidea_sinapis	1089	5	6.0	38.0
Leptotes_pirithous	287	5	8.0	25.0
Libythea_celtis	181	5	6.5	44.5
Limenitis_camilla	526	5	3.0	56.0
Limenitis_populi	529	4	2.5	77.5
Limenitis_reducta	410	5	5.0	50.0
Lopinga_achine	275	3	2.5	44.0
Luthrodes_galba	NA	NA	NA	20.0
Lycaena_alciphron	519	4	4.0	34.0
Lycaena_candens	NA	5	3.5	34.5
Lycaena_dispar	383	5	6.5	37.0
Lycaena_helle	148	5	6.0	24.0
Lycaena_hippothoe	689	4	4.5	30.5
Lycaena_ottomana	37	5	7.0	29.0
Lycaena_phlaeas	1340	5	9.5	26.5
Lycaena_thersamon	199	5	9.0	30.0
Lycaena_thetis	6	NA	2.0	27.5
Lycaena_tityrus	673	5	7.0	30.0
Lycaena_virgaureae	756	5	5.0	27.0
Lysandra_albicans	40	5	3.0	39.0
Lysandra_bellargus	601	5	7.0	31.0
Lysandra_coridon	575	4	3.0	33.0
Lysandra_corydonius	NA	NA	3.0	33.0
Lysandra_hispana	50	5	7.0	34.0
Maniola_chia	2	5	5.0	41.5
Maniola_cypricola	NA	5	4.0	NA
Maniola_halicarnassus	NA	4	3.0	44.5
Maniola_jurtina	1289	5	5.0	41.5
Maniola_megala	2	NA	3.5	41.5
Maniola_nurag	5	5	3.5	38.0
Maniola_telmessia	7	5	7.5	41.5
Melanargia_arge	33	5	2.0	53.0
Melanargia_galathea	802	5	3.5	47.5
Melanargia_ines	79	5	4.0	48.0

species	range.size	conserv.eu	FM_Aver- age	WSP_Female_aver- age
Melanargia_lachesis	111	5	2.5	49.0
Melanargia_larissa	73	5	4.0	55.0
Melanargia_occitanica	85	5	2.5	51.0
Melanargia_pherusa	4	5	2.5	48.0
Melanargia_russiae	111	5	3.0	55.0
Melitaea_aetherie	26	5	2.5	44.0
Melitaea_arduinna	14	5	3.5	44.0
Melitaea_asteria	15	5	2.0	29.0
Melitaea_athalia	1011	5	3.5	35.5
Melitaea_aurelia	268	5	2.0	30.0
Melitaea_britomartis	99	4	3.0	33.0
Melitaea_celadussa	NA	NA	3.0	35.5
Melitaea_cinxia	673	5	4.0	35.5
Melitaea_deione	108	5	4.5	39.0
Melitaea_diamina	461	4	3.5	35.5
Melitaea_didyma	621	5	5.5	37.0
Melitaea_ornata	NA	NA	4.0	41.5
Melitaea_parthenoides	172	5	3.0	33.0
Melitaea_phoebe	521	5	4.5	44.5
Melitaea_trivia	260	4	5.0	33.0
Melitaea_varia	41	5	3.0	34.0
Minois_dryas	48	5	3.0	62.0
Muschampia_cribellum	7	4	3.0	29.0
Muschampia_proto	88	5	4.0	34.5
Muschampia_tessellum	22	5	4.5	33.5
Neptis_rivularis	150	5	3.5	52.0
Neptis_sappho	104	5	4.5	54.0
Nymphalis_antiope	910	5	5.5	67.5
Nymphalis_polychloros	845	3	6.5	62.0
Nymphalis_vauualbum	72	5	3.5	63.0
Nymphalis_xanthomelas	113	4	4.5	62.0
Ochlodes_sylvanus	1058	5	4.0	30.0
Oeneis_bore	19	5	1.5	43.0
Oeneis_glacialis	46	5	3.0	53.0
Oeneis_jutta	166	5	2.5	53.0
Oeneis_norna	54	4	2.0	47.0
Papilio_alexanor	49	5	4.0	64.0
Papilio_hospiton	14	5	3.0	76.0
Papilio_machaon	1224	5	6.5	75.0
Pararge_aegeria	1178	5	7.0	40.0

species	range.size	conserv.eu	FM_Aver- age	WSP_Female_aver- age
Pararge_xiphia	NA	2	12.0	45.5
Pararge_xiphioides	NA	5	12.0	37.5
Parnassius_apollo	354	4	4.5	80.0
Parnassius_mnemosyne	371	5	3.0	58.0
Parnassius_phoebus	48	4	3.0	65.0
Pelopidas_thrax	1	NA	10.0	39.5
Phengaris_alcon	354	4	3.0	36.0
Phengaris_arion	522	2	4.0	36.0
Phengaris_nausithous	188	4	3.0	35.0
Phengaris_teleius	252	3	3.0	34.0
Pieris_balcana	NA	5	7.0	38.5
Pieris_brassicae	1353	5	7.0	57.0
Pieris_bryoniae	114	5	4.0	38.5
Pieris_cheiranthi	NA	2	12.0	62.5
Pieris_ergane	125	5	7.5	42.0
Pieris_krueperi	51	5	8.0	46.0
Pieris_mannii	203	5	7.5	43.0
Pieris_napi	1405	5	5.5	40.0
Pieris_rapae	1363	5	9.5	44.5
Plebejidea_loewii	3	NA	3.5	28.5
Plebejus_argus	991	5	4.0	22.5
Plebejus_argyrognomon	290	5	4.0	31.0
Plebejus_bellieri	10	5	3.0	23.0
Plebejus_idas	774	5	5.0	24.5
Polygonia_c-album	1098	5	6.5	46.0
Polygonia_egea	158	5	8.5	45.0
Polyommatus_admetus	57	5	3.0	34.0
Polyommatus_amandus	544	5	3.0	31.0
Polyommatus_aroaniensis	13	5	3.0	30.5
Polyommatus_celina	NA	NA	11.0	27.0
Polyommatus_damocles	NA	NA	2.5	31.5
Polyommatus_damon	163	4	2.0	32.0
Polyommatus_daphnis	305	5	3.0	35.0
Polyommatus_dolus	24	5	2.0	35.0
Polyommatus_dorylas	330	4	4.0	32.0
Polyommatus_eros	88	4	3.0	31.0
Polyommatus_escheri	167	5	3.0	28.0
Polyommatus_fabressei	11	5	3.0	30.5
Polyommatus_fulgens	27	5	2.0	32.0
Polyommatus_golgus	3	3	2.0	28.0

species	range.size	conserv.eu	FM_Average	WSP_Female_average
Polyommatus_humedasae	2	2	2.0	30.5
Polyommatus_icarus	1476	5	7.0	28.5
Polyommatus_iphigenia	2	NA	2.5	30.0
Polyommatus_nephohip- tenos	NA	4	2.0	30.5
Polyommatus_nivescens	26	4	4.0	33.0
Polyommatus_orphicus	9	3	NA	33.0
Polyommatus_ripartii	59	4	4.0	31.0
Polyommatus_thersites	351	5	7.0	29.0
Polyommatus_violetae	3	3	2.0	31.0
Pontia_callidice	60	5	4.0	47.0
Pontia_chloridice	11	5	6.0	38.5
Pontia_daplidice	921	5	5.5	41.5
Pontia_edusa	NA	5	5.5	41.5
Proterebia_phegea	5	NA	2.5	46.0
Pseudochazara_amymone	NA	3	2.5	53.0
Pseudochazara_anthelea	45	5	4.0	51.0
Pseudochazara_cingovskii	2	NA	2.0	52.0
Pseudochazara_euxina	NA	NA	3.0	54.5
Pseudochazara_geyeri	6	5	2.0	49.0
Pseudochazara_graeca	19	5	3.0	51.0
Pseudochazara_mercurius	6	NA	2.0	50.0
Pseudochazara_orestes	1	3	3.0	53.5
Pseudochazara_williamsi	5	NA	3.0	50.0
Pseudophilotes_abencerragus	27	5	2.0	20.0
Pseudophilotes_barbagiae	1	5	4.0	22.0
Pseudophilotes_baton	222	5	5.0	24.0
Pseudophilotes_bavius	20	5	3.0	27.0
Pseudophilotes_panoptes	41	4	6.0	21.5
Pseudophilotes_vicrama	212	4	5.0	23.0
Pyrgus_alveus	476	5	3.5	28.5
Pyrgus_andromedae	83	5	3.0	28.5
Pyrgus_armoricanus	311	5	5.5	26.0
Pyrgus_cacaliae	55	5	2.5	28.5
Pyrgus_carlinae	30	5	2.0	29.0
Pyrgus_carthami	264	5	5.0	32.0
Pyrgus centaureae	162	5	2.5	29.5
Pyrgus_cinarae	14	5	4.0	31.0
Pyrgus_cirsii	94	3	3.0	27.0
Pyrgus_foulquieri	29	NA	3.0	28.5

species	range.size	conserv.eu	FM_Aver- age	WSP_Female_aver- age
Pyrgus_malvae	959	5	5.0	23.0
Pyrgus_malvoides	NA	5	3.5	23.0
Pyrgus_onopordi	105	5	7.0	25.0
Pyrgus_serratulae	318	4	3.0	28.0
Pyrgus_sidae	77	5	4.0	35.0
Pyrgus_warrenensis	29	5	2.0	24.5
Pyronia_bathseba	104	5	4.5	37.0
Pyronia_cecilia	219	5	3.5	31.0
Pyronia_tithonus	515	5	3.5	36.0
Satyrium_acaciae	349	5	3.0	30.0
Satyrium_esculi	106	5	4.0	32.0
Satyrium_ilicis	586	5	4.0	34.0
Satyrium_ledereri	1	NA	3.5	29.0
Satyrium_pruni	408	5	3.0	31.5
Satyrium_spini	483	5	3.5	30.0
Satyrium_w-album	553	5	3.0	30.0
Satyrus_actaea	73	5	3.0	55.0
Satyrus_ferula	144	5	4.0	55.0
Scolitantides_orion	260	4	5.0	27.5
Speyeria_aglaja	1132	5	3.5	51.5
Spialia_orbifer	129	5	5.5	25.0
Spialia_phlomidis	25	5	4.0	27.0
Spialia_rosae	NA	NA	NA	20.0
Spialia_sertorius	380	5	7.0	24.0
Spialia_therapne	7	5	5.0	24.0
Tarucus_balkanicus	19	5	7.0	20.0
Tarucus_theophrastus	4	5	6.0	21.0
Thecla_betulae	628	5	4.5	35.0
Thymelicus_acteon	496	4	4.5	24.0
Thymelicus_christi	NA	5	9.0	24.0
Thymelicus_hyrax	10	5	3.0	32.5
Thymelicus_lineola	926	5	3.5	25.5
Thymelicus_sylvestris	832	5	4.5	26.5
Tomares_ballus	51	5	4.0	29.0
Tomares_callimachus	NA	NA	3.0	28.0
Tomares_nogelii	2	0	3.5	32.5
Turanana_taygetica	3	2	4.0	21.0
Vanessa_atalanta	1343	5	8.5	59.5
Vanessa_cardui	1373	5	8.0	54.0
Vanessa_virginiensis	12	NA	10.0	45.0

species	range.size	conserv.eu	FM_Aver- age	WSP_Female_aver- age
Vanessa_vulcania	NA	5	12.0	57.0
Ypthima_asterope	2	NA	7.5	30.0
Zegris_eupheme	36	4	4.0	48.5
Zerynthia_cassandra	44	NA	4.0	49.0
Zerynthia_cerisy	61	4	5.0	57.0
Zerynthia_cretica	7	5	4.5	52.5
Zerynthia_polyxena	215	5	4.0	49.0
Zerynthia_rumina	128	5	5.0	45.0
Zizeeria_karsandra	NA	NA	NA	22.0
Zizeeria_knysna	39	NA	9.0	22.0

[illegible]

species	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32	33	34	35	36	37	38	39	40	41	42	43	44	45	46	47	48	49	50	51	52	53	54	55	56	57	58	59	60	61	62	63	64	65	66	67	68	69	70	71	72	73	74	75	76	77	78	79	80	81	82	83	84	85	86	87	88	89	90	91	92	93	94	95	96	97	98	99	100	101	102	103	104	105	106	107	108	109	110	111	112	113	114	115	116	117	118	119	120	121	122	123	124	125	126	127	128	129	130	131	132	133	134	135	136	137	138	139	140	141	142	143	144	145	146	147	148	149	150	151	152	153	154	155	156	157	158	159	160	161	162	163	164	165	166	167	168	169	170	171	172	173	174	175	176	177	178	179	180	181	182	183	184	185	186	187	188	189	190	191	192	193	194	195	196	197	198	199	200	201	202	203	204	205	206	207	208	209	210	211	212	213	214	215	216	217	218	219	220	221	222	223	224	225	226	227	228	229	230	231	232	233	234	235	236	237	238	239	240	241	242	243	244	245	246	247	248	249	250	251	252	253	254	255	256	257	258	259	260	261	262	263	264	265	266	267	268	269	270	271	272	273	274	275	276	277	278	279	280	281	282	283	284	285	286	287	288	289	290	291	292	293	294	295	296	297	298	299	300	301	302	303	304	305	306	307	308	309	310	311	312	313	314	315	316	317	318	319	320	321	322	323	324	325	326	327	328	329	330	331	332	333	334	335	336	337	338	339	340	341	342	343	344	345	346	347	348	349	350	351	352	353	354	355	356	357	358	359	360	361	362	363	364	365	366	367	368	369	370	371	372	373	374	375	376	377	378	379	380
---------	---	---	---	---	---	---	---	---	---	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----

species	range	size	con- serva- tion	GEN_Av- er- age	WSP_Fe- male_av- er- WSP_Fe- range	WSP_Fe- range										ELT_sin- er-		FM_Av- er-		AFB_Av- ey- i-		AFB_an- AFB_min- AFB_max	
						GEN_Av- er- age	WSP_Fe- male_av- er- WSP_Fe- range	WSP_Fe- male_av- er- WSP_Fe- range	WSP_Fe- male_av- er- WSP_Fe- range	WSP_Fe- male_av- er- WSP_Fe- range	WSP_Fe- male_av- er- WSP_Fe- range	WSP_Fe- male_av- er- WSP_Fe- range	WSP_Fe- male_av- er- WSP_Fe- range	WSP_Fe- male_av- er- WSP_Fe- range	WSP_Fe- male_av- er- WSP_Fe- range	WSP_Fe- male_av- er- WSP_Fe- range	WSP_Fe- male_av- er- WSP_Fe- range	WSP_Fe- male_av- er- WSP_Fe- range	WSP_Fe- male_av- er- WSP_Fe- range	WSP_Fe- male_av- er- WSP_Fe- range	WSP_Fe- male_av- er- WSP_Fe- range	WSP_Fe- male_av- er- WSP_Fe- range	WSP_Fe- male_av- er- WSP_Fe- range
Aph-Ny-1005	15	1.00	01	00	038.57	0.063	1	1	0	0	1	1	0	0	2000	0.00	1	0	1	0	0	0	0
to- i-																							
pus_dhy-																							
per-																							
an-																							
tus																							
Apo-Pie-8945	5	1.00	01	00	062.51	0.082	0	1	1	0	0	0	1	0	2600	0.51	1	1	0	1	0	0	1
ria_cataegi																							
Aras-Ny-6205	15	2.00	03	02	033.51	0.707	0	1	0	0	0	1	1	0	1700	0.02	1	0	1	0	0	0	0
nia_ilev-																							
ana dae																							
Ar- Pa-134	5	1.00	01	00	057.0	0.507	0	1	0	0	0	1	0	0	900	0.53	1	0	0	0	0	0	0
chompilapolli-																							
nus ion-																							
idae																							
Are-Ny-2295	15	1.00	01	00	041.0	0.383	1	1	0	0	0	1	0	0	2300	0.51	1	0	0	0	0	0	0
sana_arethusa																							
dae																							
Arg-Ny-1005	4	1.00	01	00	055.0	0.707	N	A	N	A	N	A	N	A	N	A	N	A	N	A	N	A	N
yn- i-																							
nis_daodice																							
Arg-Ny-3205	15	1.50	02	01	072.06	0.517	1	1	0	0	1	0	0	0	2600	0.04	1	0	0	0	0	1	1
yn- i-																							
nis_dan-																							
dora																							
Arg-Ny-1025	15	1.00	01	00	062.0	0.268	0	1	0	0	1	0	0	0	2300	0.02	1	0	1	1	0	0	1
yn- i-																							
nis_dae-																							
phia																							
Ari-Ly-7005	5	2.00	03	02	025.0	0.102	1	1	0	0	1	0	0	0	1900	0.00	1	0	0	0	0	1	1
cia_cacitidae																							
Ari-Ly-434	5	2.00	03	02	028.0	0.289	0	1	0	0	1	0	0	300	0.10	0.00	1	0	0	0	0	0	0
cia_cacitidae																							
teros																							
Ari-Ly-4355	5	1.00	01	00	027.57	0.066	1	1	0	0	1	0	0	0	2200	0.00	1	0	0	0	0	0	0
cia_cacitidae																							
tax-																							
erxes																							

[illegible]

[illegible]

species	range	size	con- serva- tion	GEN_Av- er- age	WSP_Fe- male_av- er- age	WSP_Fe- range	HIS_EVE	KEY_EVE	FAT_EVE	ELT_sin- gular	ELT_sin- gular	ELT_sin- gular	ELT_sin- gular	FM_Av- er- age	FM_Av- er- age	AFB_Av- er- age	AFB_Av- er- age	AFB_an- i- AFB_min- AFB_max	AFB_min- AFB_max
Brenthys thisi-ino- dae	695	5	1.00	0.01	0.00	0.37	0.8	0.20	4	1	1	0	0	1	1	0	0	200	251
Brinthis te- i- sia_darce	416	5	1.00	0.01	0.00	0.73	0.4	0.20	4	1	1	0	0	0	1	0	0	230	453
Ca-Ly-33 cyrenidae shalli	NANA	3.00	0.03	0.00	0.23	0.5	1.00	NANA	NANA	NANA	1	0	0	0	150	0.00	1	0	0
Cal-Ly-37 lophyrus idais	5	5	1.00	0.01	0.00	0.35	0.3	0.25	NANA	NANA	NANA	1	0	0	100	0.00	1	0	0
Cal-Ly-119 lophyrus idabi	2	5	1.25	0.01	0.50	0.52	0.6	0.00	3	1	1	1	0	1	0	0	0	230	602
CarHes-615 charperi- o- idae dus_al- ceae	5	5	2.50	0.03	0.01	0.30	0.8	0.10	7	0	1	0	0	1	0	0	0	300	804
CarHes-56 charperi- o- idae dus_baeti- cus	5	5	2.50	0.03	0.01	0.30	0.4	0.40	8	1	1	0	0	1	0	0	500	1104	
CarHes-277 charperi- o- idae dus_floc- cifera	4	5	1.50	0.02	0.01	0.30	0.4	0.10	4	0	1	0	0	1	0	0	0	200	451
CarHes-146 charperi- o- idae dus_lavatherae	4	4	1.00	0.01	0.00	0.31	0.6	0.40	8	1	1	0	0	1	0	0	0	160	402
CarHes-54 charperi- o- idae dus_ori- en- talis	5	5	2.50	0.03	0.01	0.29	0.2	0.50	7	NANA	NANA	NANA	NANA	NANA	200	0.51	NANA	NANA	NANA

species	range	size	GEN	Av- con-	Av- con-	WSP_Fe- er-	WSP_Fe- er-	ELT_sin-	FM_Av-	FM_Av-	AFB_Av-	AFB_an-	AFB_min-
range	size	GEN	Av- con-	Av- con-	WSP_Fe- er-	WSP_Fe- er-	ELT_sin-	FM_Av-	FM_Av-	AFB_Av-	AFB_an-	AFB_min-	
CarHe2	NANA	3.03	0.03	0.00	0.30	0.1	0.28	NANANANANANANANA	700	1700	100	NANANANANANANA	
charperi-													
o- idae													
dus_staud-													
eri													
CarHeNA5	5	3.03	0.03	0.00	0.30	0.8	1.00	NANANANANANANANA	0	2500	100	NANANANANANANA	
charperi-													
o- idae													
dus_tripoli-													
nus													
CarHe5855	5	1.00	0.01	0.00	0.27	0.6	0.10	6 0 1 0 0 0 1 0 0 0 0	1800	251	1 0 0 0 0 0 0 0		
ro- peri-													
cephidus_													
palae-													
mon													
CarHe2225	5	1.00	0.01	0.00	0.25	0.6	0.20	0 0 1 0 0 0 1 0 0 0 0	500	251	1 0 0 0 0 0 0 0		
ro- peri-													
cephidus_													
sil-													
vi-													
cola													
CatHeNANA	3.03	0.03	0.00	0.50	0.10	NA	0 0 0 1 1 1 1 0 0	2000	1.2	NANANANANANANA			
siliada-													
rella													
CelaLy-1190	5	1.71	0.02	0.51	0.52	0.10	0.00	0 0 0 1 1 1 0 0 0 0	1900	100	1 0 1 1 0 0 0 0		
t- caenidae													
rina_ar-													
gi-													
o-													
lus													
ChaNy-1871	5	1.50	0.02	0.01	0.79	0.5	0.10	2 0 0 0 1 1 1 0 0	2400	53	0 0 0 0 0 0 1 1		
siusi-													
dae													
ChaNy-3531	4	1.00	0.01	0.00	0.55	0.26	0.30	3 1 1 0 0 1 1 0 0	2500	104	1 0 0 0 0 0 1 1		
ara_i-bri-													
seis dae													
ChaNy-5151	5	1.00	0.01	0.00	0.60	0.12	1.00	NANANANANANANANA	900	1100	1 0 0 0 0 0 0 0		
ara_i-prieuri													
dae													

species	range	size	GEN	Av- GEN	male_av- er- WSP_Fe-	WSP_Fe-										FM_Av-				AFB_Av-				AFB_an-			
						WSP_Fe-	ELT_sin-	er-	er-	er-	er-	er-	er-	er-	er-	er-	er-	er-	er-	er-	er-	er-	er-	er-	er-	er-	
Coenonympha arcania dae	5200	115	21.01	50.53	7.6	0.169	0	1	0	0	1	1	0	0	230	0.53	1	0	1	0	0	0	0	0	0	0	
Coenonympha idae	5200	115	21.01	50.53	7.6	0.57	N	A	N	A	N	A	N	A	1	0	0	0	200	0.51	1	0	0	0	0	0	0
Coenonympha idae	5200	115	21.01	50.53	7.6	0.169	0	1	0	0	1	0	0	100	160	0.00	1	0	0	0	0	0	1	0	0	0	
Coenonympha idae	5200	115	21.01	50.53	7.6	0.57	N	A	N	A	N	A	N	A	1	1	0	800	220	0.00	1	0	0	0	0	0	0
Coenonympha e-riondae	5200	115	21.01	50.53	7.6	0.268	0	1	0	0	1	0	0	200	220	0.00	1	0	0	0	0	0	1	0	0	0	
Coenonympha idae	5200	115	21.01	50.53	7.6	0.164	0	1	0	0	1	0	0	0	800	0.53	1	0	1	0	0	0	0	0	0	0	
Coenonympha ander dae	5200	115	21.01	50.53	7.6	0.500	0	1	0	0	1	0	0	0	210	0.53	1	0	0	0	0	0	0	0	0	0	
Coenonympha pus i-dae	5200	115	21.01	50.53	7.6	0.268	0	1	0	0	1	0	0	0	500	0.00	1	0	1	1	0	0	0	0	0	0	
Coenonympha entalidae	5200	115	21.01	50.53	7.6	0.500	N	A	N	A	N	A	N	A	N	A	800	120	0.00	N	A	N	A	N	A	N	A
Coenonympha philis dae	5200	115	21.01	50.53	7.6	0.096	0	1	0	0	1	0	0	0	270	0.06	1	0	0	0	0	0	0	0	0	0	
Coenonympha sis i-dae	5200	115	21.01	50.53	7.6	0.364	0	1	0	0	1	0	0	750	175	0.51	1	0	0	0	0	0	0	0	0	0	
Coenonympha sis i-dae	5200	115	21.01	50.53	7.6	0.289	0	1	0	0	N	A	N	A	0	180	0.04	1	0	0	0	0	0	0	0	0	

|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|

species	l-rang	p-size	con- fam- 1	ser- com- 2	GEN_Av- 1	GEN_Av- 2	GEN_Av- 3	GEN_Av- 4	GEN_Av- 5	WSP_Fe- er- WSP_Fe- 1	WSP_Fe- er- WSP_Fe- 2	WSP_Fe- er- WSP_Fe- 3	WSP_Fe- er- WSP_Fe- 4	WSP_Fe- er- WSP_Fe- 5	ELT_sin- er- 1	ELT_sin- er- 2	ELT_sin- er- 3	ELT_sin- er- 4	ELT_sin- er- 5	FM_Av- er- 1	FM_Av- er- 2	FM_Av- er- 3	FM_Av- er- 4	FM_Av- er- 5	AFB_an- ey- i- AFB_min- 1	AFB_an- ey- i- AFB_min- 2	AFB_an- ey- i- AFB_min- 3	AFB_an- ey- i- AFB_min- 4	AFB_an- ey- i- AFB_min- 5	AFB_an- ey- i- AFB_min- 6	AFB_an- ey- i- AFB_min- 7	AFB_an- ey- i- AFB_min- 8	AFB_an- ey- i- AFB_min- 9	AFB_an- ey- i- AFB_min- 10	AFB_an- ey- i- AFB_min- 11	AFB_an- ey- i- AFB_min- 12	AFB_an- ey- i- AFB_min- 13	AFB_an- ey- i- AFB_min- 14	AFB_an- ey- i- AFB_min- 15	AFB_an- ey- i- AFB_min- 16	AFB_an- ey- i- AFB_min- 17	AFB_an- ey- i- AFB_min- 18	AFB_an- ey- i- AFB_min- 19	AFB_an- ey- i- AFB_min- 20	AFB_an- ey- i- AFB_min- 21	AFB_an- ey- i- AFB_min- 22	AFB_an- ey- i- AFB_min- 23	AFB_an- ey- i- AFB_min- 24	AFB_an- ey- i- AFB_min- 25	AFB_an- ey- i- AFB_min- 26	AFB_an- ey- i- AFB_min- 27	AFB_an- ey- i- AFB_min- 28	AFB_an- ey- i- AFB_min- 29	AFB_an- ey- i- AFB_min- 30	AFB_an- ey- i- AFB_min- 31	AFB_an- ey- i- AFB_min- 32	AFB_an- ey- i- AFB_min- 33	AFB_an- ey- i- AFB_min- 34	AFB_an- ey- i- AFB_min- 35	AFB_an- ey- i- AFB_min- 36	AFB_an- ey- i- AFB_min- 37	AFB_an- ey- i- AFB_min- 38	AFB_an- ey- i- AFB_min- 39	AFB_an- ey- i- AFB_min- 40	AFB_an- ey- i- AFB_min- 41	AFB_an- ey- i- AFB_min- 42	AFB_an- ey- i- AFB_min- 43	AFB_an- ey- i- AFB_min- 44	AFB_an- ey- i- AFB_min- 45	AFB_an- ey- i- AFB_min- 46	AFB_an- ey- i- AFB_min- 47	AFB_an- ey- i- AFB_min- 48	AFB_an- ey- i- AFB_min- 49	AFB_an- ey- i- AFB_min- 50	AFB_an- ey- i- AFB_min- 51	AFB_an- ey- i- AFB_min- 52	AFB_an- ey- i- AFB_min- 53	AFB_an- ey- i- AFB_min- 54	AFB_an- ey- i- AFB_min- 55	AFB_an- ey- i- AFB_min- 56	AFB_an- ey- i- AFB_min- 57	AFB_an- ey- i- AFB_min- 58	AFB_an- ey- i- AFB_min- 59	AFB_an- ey- i- AFB_min- 60	AFB_an- ey- i- AFB_min- 61	AFB_an- ey- i- AFB_min- 62	AFB_an- ey- i- AFB_min- 63	AFB_an- ey- i- AFB_min- 64	AFB_an- ey- i- AFB_min- 65	AFB_an- ey- i- AFB_min- 66	AFB_an- ey- i- AFB_min- 67	AFB_an- ey- i- AFB_min- 68	AFB_an- ey- i- AFB_min- 69	AFB_an- ey- i- AFB_min- 70	AFB_an- ey- i- AFB_min- 71	AFB_an- ey- i- AFB_min- 72	AFB_an- ey- i- AFB_min- 73	AFB_an- ey- i- AFB_min- 74	AFB_an- ey- i- AFB_min- 75	AFB_an- ey- i- AFB_min- 76	AFB_an- ey- i- AFB_min- 77	AFB_an- ey- i- AFB_min- 78	AFB_an- ey- i- AFB_min- 79	AFB_an- ey- i- AFB_min- 80	AFB_an- ey- i- AFB_min- 81	AFB_an- ey- i- AFB_min- 82	AFB_an- ey- i- AFB_min- 83	AFB_an- ey- i- AFB_min- 84	AFB_an- ey- i- AFB_min- 85	AFB_an- ey- i- AFB_min- 86	AFB_an- ey- i- AFB_min- 87	AFB_an- ey- i- AFB_min- 88	AFB_an- ey- i- AFB_min- 89	AFB_an- ey- i- AFB_min- 90	AFB_an- ey- i- AFB_min- 91	AFB_an- ey- i- AFB_min- 92	AFB_an- ey- i- AFB_min- 93	AFB_an- ey- i- AFB_min- 94	AFB_an- ey- i- AFB_min- 95	AFB_an- ey- i- AFB_min- 96	AFB_an- ey- i- AFB_min- 97	AFB_an- ey- i- AFB_min- 98	AFB_an- ey- i- AFB_min- 99	AFB_an- ey- i- AFB_min- 100	AFB_an- ey- i- AFB_min- 101	AFB_an- ey- i- AFB_min- 102	AFB_an- ey- i- AFB_min- 103	AFB_an- ey- i- AFB_min- 104	AFB_an- ey- i- AFB_min- 105	AFB_an- ey- i- AFB_min- 106	AFB_an- ey- i- AFB_min- 107	AFB_an- ey- i- AFB_min- 108	AFB_an- ey- i- AFB_min- 109	AFB_an- ey- i- AFB_min- 110	AFB_an- ey- i- AFB_min- 111	AFB_an- ey- i- AFB_min- 112	AFB_an- ey- i- AFB_min- 113	AFB_an- ey- i- AFB_min- 114	AFB_an- ey- i- AFB_min- 115	AFB_an- ey- i- AFB_min- 116	AFB_an- ey- i- AFB_min- 117	AFB_an- ey- i- AFB_min- 118	AFB_an- ey- i- AFB_min- 119	AFB_an- ey- i- AFB_min- 120	AFB_an- ey- i- AFB_min- 121	AFB_an- ey- i- AFB_min- 122	AFB_an- ey- i- AFB_min- 123	AFB_an- ey- i- AFB_min- 124	AFB_an- ey- i- AFB_min- 125	AFB_an- ey- i- AFB_min- 126	AFB_an- ey- i- AFB_min- 127	AFB_an- ey- i- AFB_min- 128	AFB_an- ey- i- AFB_min- 129	AFB_an- ey- i- AFB_min- 130	AFB_an- ey- i- AFB_min- 131	AFB_an- ey- i- AFB_min- 132	AFB_an- ey- i- AFB_min- 133	AFB_an- ey- i- AFB_min- 134	AFB_an- ey- i- AFB_min- 135	AFB_an- ey- i- AFB_min- 136	AFB_an- ey- i- AFB_min- 137	AFB_an- ey- i- AFB_min- 138	AFB_an- ey- i- AFB_min- 139	AFB_an- ey- i- AFB_min- 140	AFB_an- ey- i- AFB_min- 141	AFB_an- ey- i- AFB_min- 142	AFB_an- ey- i- AFB_min- 143	AFB_an- ey- i- AFB_min- 144	AFB_an- ey- i- AFB_min- 145	AFB_an- ey- i- AFB_min- 146	AFB_an- ey- i- AFB_min- 147	AFB_an- ey- i- AFB_min- 148	AFB_an- ey- i- AFB_min- 149	AFB_an- ey- i- AFB_min- 150	AFB_an- ey- i- AFB_min- 151	AFB_an- ey- i- AFB_min- 152	AFB_an- ey- i- AFB_min- 153	AFB_an- ey- i- AFB_min- 154	AFB_an- ey- i- AFB_min- 155	AFB_an- ey- i- AFB_min- 156	AFB_an- ey- i- AFB_min- 157	AFB_an- ey- i- AFB_min- 158	AFB_an- ey- i- AFB_min- 159	AFB_an- ey- i- AFB_min- 160	AFB_an- ey- i- AFB_min- 161	AFB_an- ey- i- AFB_min- 162	AFB_an- ey- i- AFB_min- 163	AFB_an- ey- i- AFB_min- 164	AFB_an- ey- i- AFB_min- 165	AFB_an- ey- i- AFB_min- 166	AFB_an- ey- i- AFB_min- 167	AFB_an- ey- i- AFB_min- 168	AFB_an- ey- i- AFB_min- 169	AFB_an- ey- i- AFB_min- 170	AFB_an- ey- i- AFB_min- 171	AFB_an- ey- i- AFB_min- 172	AFB_an- ey- i- AFB_min- 173	AFB_an- ey- i- AFB_min- 174	AFB_an- ey- i- AFB_min- 175	AFB_an- ey- i- AFB_min- 176	AFB_an- ey- i- AFB_min- 177	AFB_an- ey- i- AFB_min- 178	AFB_an- ey- i- AFB_min- 179	AFB_an- ey- i- AFB_min- 180	AFB_an- ey- i- AFB_min- 181	AFB_an- ey- i- AFB_min- 182	AFB_an- ey- i- AFB_min- 183	AFB_an- ey- i- AFB_min- 184	AFB_an- ey- i- AFB_min- 185	AFB_an- ey- i- AFB_min- 186	AFB_an- ey- i- AFB_min- 187	AFB_an- ey- i- AFB_min- 188	AFB_an- ey- i- AFB_min- 189	AFB_an- ey- i- AFB_min- 190	AFB_an- ey- i- AFB_min- 191	AFB_an- ey- i- AFB_min- 192	AFB_an- ey- i- AFB_min- 193	AFB_an- ey- i- AFB_min- 194	AFB_an- ey- i- AFB_min- 195	AFB_an- ey- i- AFB_min- 196	AFB_an- ey- i- AFB_min- 197	AFB_an- ey- i- AFB_min- 198	AFB_an- ey- i- AFB_min- 199	AFB_an- ey- i- AFB_min- 200	AFB_an- ey- i- AFB_min- 201	AFB_an- ey- i- AFB_min- 202	AFB_an- ey- i- AFB_min- 203	AFB_an- ey- i- AFB_min- 204	AFB_an- ey- i- AFB_min- 205	AFB_an- ey- i- AFB_min- 206	AFB_an- ey- i- AFB_min- 207	AFB_an- ey- i- AFB_min- 208	AFB_an- ey- i- AFB_min- 209	AFB_an- ey- i- AFB_min- 210	AFB_an- ey- i- AFB_min- 211	AFB_an- ey- i- AFB_min- 212	AFB_an- ey- i- AFB_min- 213	AFB_an- ey- i- AFB_min- 214	AFB_an- ey- i- AFB_min- 215	AFB_an- ey- i- AFB_min- 216	AFB_an- ey- i- AFB_min- 217	AFB_an- ey- i- AFB_min- 218	AFB_an- ey- i- AFB_min- 219	AFB_an- ey- i- AFB_min- 220	AFB_an- ey- i- AFB_min- 221	AFB_an- ey- i- AFB_min- 222	AFB_an- ey- i- AFB_min- 223	AFB_an- ey- i- AFB_min- 224	AFB_an- ey- i- AFB_min- 225	AFB_an- ey- i- AFB_min- 226	AFB_an- ey- i- AFB_min- 227	AFB_an- ey- i- AFB_min- 228	AFB_an- ey- i- AFB_min- 229	AFB_an- ey- i- AFB_min- 230	AFB_an- ey- i- AFB_min- 231	AFB_an- ey- i- AFB_min- 232	AFB_an- ey- i- AFB_min- 233	AFB_an- ey- i- AFB_min- 234	AFB_an- ey- i- AFB_min- 235	AFB_an- ey- i- AFB_min- 236	AFB_an- ey- i- AFB_min- 237	AFB_an- ey- i- AFB_min- 238	AFB_an- ey- i- AFB_min- 239	AFB_an- ey- i- AFB_min- 240	AFB_an- ey- i- AFB_min- 241	AFB_an- ey- i- AFB_min- 242	AFB_an- ey- i- AFB_min- 243	AFB_an- ey- i- AFB_min- 244	AFB_an- ey- i- AFB_min- 245	AFB_an- ey- i- AFB_min- 246	AFB_an- ey- i- AFB_min- 247	AFB_an- ey- i- AFB_min- 248	AFB_an- ey- i- AFB_min- 249	AFB_an- ey- i- AFB_min- 250	AFB_an- ey- i- AFB_min- 251	AFB_an- ey- i- AFB_min- 252	AFB_an- ey- i- AFB_min- 253	AFB_an- ey- i- AFB_min- 254	AFB_an- ey- i- AFB_min- 255	AFB_an- ey- i- AFB_min- 256	AFB_an- ey- i- AFB_min- 257	AFB_an- ey- i- AFB_min- 258	AFB_an- ey- i- AFB_min- 259	AFB_an- ey- i- AFB_min- 260	AFB_an- ey- i- AFB_min- 261	AFB_an- ey- i- AFB_min- 262	AFB_an- ey- i- AFB_min- 263	AFB_an- ey- i- AFB_min- 264	AFB_an- ey- i- AFB_min- 265	AFB_an- ey- i- AFB_min- 266	AFB_an- ey- i- AFB_min- 267	AFB_an- ey- i- AFB_min- 268	AFB_an- ey- i- AFB_min- 269	AFB_an- ey- i- AFB_min- 270	AFB_an- ey- i- AFB_min- 271	AFB_an- ey- i- AFB_min- 272	AFB_an- ey- i- AFB_min- 273	AFB_an- ey- i- AFB_min- 274	AFB_an- ey- i- AFB_min- 275	AFB_an- ey- i- AFB_min- 276	AFB_an- ey- i- AFB_min- 277	AFB_an- ey- i- AFB_min- 278	AFB_an- ey- i- AFB_min- 279	AFB_an- ey- i- AFB_min- 280	AFB_an- ey- i- AFB_min- 281	AFB_an- ey- i- AFB_min- 282	AFB_an- ey- i- AFB_min- 283	AFB_an- ey- i- AFB_min- 284	AFB_an- ey- i- AFB_min- 285	AFB_an- ey- i- AFB_min- 286	AFB_an- ey- i- AFB_min- 287	AFB_an- ey- i- AFB_min- 288	AFB_an- ey- i- AFB_min- 289	AFB_an- ey- i- AFB_min- 290	AFB_an- ey- i- AFB_min- 291	AFB_an- ey- i- AFB_min- 292	AFB_an- ey- i- AFB_min- 293	AFB_an- ey- i- AFB_min- 294	AFB_an- ey- i- AFB_min- 295	AFB_an- ey- i- AFB_min- 296	AFB_an- ey- i- AFB_min- 297	AFB_an- ey- i- AFB_min- 298	AFB_an- ey- i- AFB_min- 299	AFB_an- ey- i- AFB_min- 300	AFB_an- ey- i- AFB_min- 301	AFB_an- ey- i- AFB_min- 302	AFB_an- ey- i- AFB_min- 303	AFB_an- ey- i- AFB_min- 304	AFB_an- ey- i- AFB_min- 305	AFB_an- ey- i- AFB_min- 306	AFB_an- ey- i- AFB_min- 307	AFB_an- ey- i- AFB_min- 308	AFB_an- ey- i- AFB_min- 309	AFB_an- ey- i- AFB_min- 310	AFB_an- ey- i- AFB_min- 311	AFB_an- ey- i- AFB_min- 312	AFB_an- ey- i- AFB_min- 313	AFB_an- ey- i- AFB_min- 314	AFB_an- ey- i- AFB_min- 315	AFB_an- ey- i- AFB_min- 316	AFB_an- ey- i- AFB_min- 317	AFB_an- ey- i- AFB_min- 318	AFB_an- ey- i- AFB_min- 319	AFB_an- ey- i- AFB_min- 320	AFB_an- ey- i- AFB_min- 321	AFB_an- ey- i- AFB_min- 322	AFB_an- ey- i- AFB_min- 323	AFB_an- ey- i- AFB_min- 324	AFB_an- ey- i- AFB_min- 325	AFB_an- ey- i- AFB_min- 326	AFB_an- ey- i- AFB_min- 327	AFB_an- ey- i- AFB_min- 328	AFB_an- ey- i- AFB_min- 329	AFB_an- ey- i- AFB_min- 330	AFB_an- ey- i- AFB_min- 331	AFB_an- ey- i- AFB_min- 332	AFB_an- ey- i- AFB_min- 333	AFB_an- ey- i- AFB_min- 334	AFB_an- ey- i- AFB_min- 335	AFB_an- ey- i- AFB_min- 336	AFB_an- ey- i- AFB_min- 337	AFB_an- ey- i- AFB_min- 338	AFB_an- ey- i- AFB_min- 339	AFB_an- ey- i- AFB_min- 340	AFB_an- ey- i- AFB_min- 341	AFB_an- ey- i- AFB_min- 342	AFB_an-<
---------	--------	--------	-------------------	-------------------	--------------	--------------	--------------	--------------	--------------	--------------------------------	--------------------------------	--------------------------------	--------------------------------	--------------------------------	----------------------	----------------------	----------------------	----------------------	----------------------	--------------------	--------------------	--------------------	--------------------	--------------------	---------------------------------------	---------------------------------------	---------------------------------------	---------------------------------------	---------------------------------------	---------------------------------------	---------------------------------------	---------------------------------------	---------------------------------------	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	----------

species	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32	33	34	35	36	37	38	39	40	41	42	43	44	45	46	47	48	49	50	51	52	53	54	55	56	57	58	59	60	61	62	63	64	65	66	67	68	69	70	71	72	73	74	75	76	77	78	79	80	81	82	83	84	85	86	87	88	89	90	91	92	93	94	95	96	97	98	99	100	101	102	103	104	105	106	107	108	109	110	111	112	113	114	115	116	117	118	119	120	121	122	123	124	125	126	127	128	129	130	131	132	133	134	135	136	137	138	139	140	141	142	143	144	145	146	147	148	149	150	151	152	153	154	155	156	157	158	159	160	161	162	163	164	165	166	167	168	169	170	171	172	173	174	175	176	177	178	179	180	181	182	183	184	185	186	187	188	189	190	191	192	193	194	195	196	197	198	199	200	201	202	203	204	205	206	207	208	209	210	211	212	213	214	215	216	217	218	219	220	221	222	223	224	225	226	227	228	229	230	231	232	233	234	235	236	237	238	239	240	241	242	243	244	245	246	247	248	249	250	251	252	253	254	255	256	257	258	259	260	261	262	263	264	265	266	267	268	269	270	271	272	273	274	275	276	277	278	279	280	281	282	283	284	285	286	287	288	289	290	291	292	293	294	295	296	297	298	299	300	301	302	303	304	305	306	307	308	309	310	311	312	313	314	315	316	317	318	319	320	321	322	323	324	325	326	327	328	329	330	331	332	333	334	335	336	337	338	339	340	341	342	343	344	345	346	347	348	349	350	351	352	353	354	355	356	357	358	359	360	361	362	363	364	365	366	367	368	369	370	371	372	373	374	375	376	377	378	379	380
---------	---	---	---	---	---	---	---	---	---	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----

[illegible]

species	range	size	con- serva- tion	WSP_Fe-										FM_Av-					AFB_Av-					AFB_an-				
				GEN_Av-	male_av-	er- WSP_Fe-	ELT_sin-	er-	ey-	i-	AFB_min-	AFB_max-	AFB_min-	AFB_max-	AFB_min-	AFB_max-	AFB_min-	AFB_max-	AFB_min-	AFB_max-	AFB_min-	AFB_max-	AFB_min-	AFB_max-	AFB_min-	AFB_max-	AFB_min-	AFB_max-
EreNympha- bia_imanto- dae	89	5	0.50	0.50	0.039	0.10	0.30	6	0	1	0	0	1	0	0	900	1600	1	0	0	0	0	0	0	0	0	0	0
EreNympha- bia_imedusa- dae	36	5	0.75	0.51	0.00	0.53	8	0	0	2	4	1	1	0	0	1	0	0	200	2350	1	51	1	0	0	0	0	0
EreNympha- bia_imelam- pus dae	56	5	1.00	0.01	0.00	0.33	0	0	2	8	NAN	NAN	NAN	NAN	NAN	NAN	NAN	800	1600	1	0	0	0	0	0	0	0	
EreNympha- bia_imelas- dae	31	5	1.00	0.01	0.00	0.45	0	0	1	0	NAN	NAN	NAN	NAN	NAN	NAN	NAN	200	2600	1	51	1	0	0	0	0	0	
EreNympha- bia_imel- olandae	139	5	1.00	0.01	0.00	0.46	0	0	2	0	0	1	1	0	0	1	0	0	150	2250	1	0	0	0	0	0	0	0
EreNympha- bia_imnes- tra dae	22	5	0.75	0.51	0.00	0.36	0	0	5	0	0	1	0	0	0	NAN	NAN	1500	1000	1	0	0	0	0	0	0	0	
EreNympha- bia_imon- tanadae	38	5	1.00	0.01	0.00	0.47	0	0	3	0	4	1	0	0	0	NAN	NAN	1100	1400	1	0	0	0	0	0	0	0	
EreNympha- bia_in- ori- das	48	5	1.00	0.01	0.00	0.41	0	0	2	8	9	0	1	0	0	1	0	0	500	1100	1	51	1	0	0	0	0	0
EreNympha- bia_ini- validae	14	5	0.50	0.50	0.00	0.32	0	0	1	0	NAN	NAN	NAN	NAN	NAN	NAN	NAN	1	0	0	2100	100	1	51	1	0	0	0
EreNympha- bia_ioeme- dae	89	5	0.75	0.51	0.00	0.54	2	0	0	1	0	0	1	0	0	800	1800	1	0	0	1	0	0	0	0	0	0	
EreNympha- bia_iori- en- talis	5	5	1.00	0.01	0.00	0.30	0	3	0	2	8	NAN	NAN	NAN	NAN	NAN	NAN	1800	100	2	51	1	0	0	0	0	0	
EreNympha- bia_iot- tomidae	42	5	1.00	0.01	0.00	0.39	0	0	7	0	NAN	NAN	NAN	NAN	NAN	NAN	NAN	450	2150	1	0	2	1	0	0	0	0	

species	1s	2s	3s	4s	5s	6s	7s	8s	9s	10s	11s	12s	13s	14s	15s	16s	17s	18s	19s	20s	21s	22s	23s	24s	25s	26s	27s	28s	29s	30s	31s	32s	33s	34s	35s	36s	37s	38s	39s	40s	41s	42s	43s	44s	45s	46s	47s	48s	49s	50s	51s	52s	53s	54s	55s	56s	57s	58s	59s	60s	61s	62s	63s	64s	65s	66s	67s	68s	69s	70s	71s	72s	73s	74s	75s	76s	77s	78s	79s	80s	81s	82s	83s	84s	85s	86s	87s	88s	89s	90s	91s	92s	93s	94s	95s	96s	97s	98s	99s	100s	101s	102s	103s	104s	105s	106s	107s	108s	109s	110s	111s	112s	113s	114s	115s	116s	117s	118s	119s	120s	121s	122s	123s	124s	125s	126s	127s	128s	129s	130s	131s	132s	133s	134s	135s	136s	137s	138s	139s	140s	141s	142s	143s	144s	145s	146s	147s	148s	149s	150s	151s	152s	153s	154s	155s	156s	157s	158s	159s	160s	161s	162s	163s	164s	165s	166s	167s	168s	169s	170s	171s	172s	173s	174s	175s	176s	177s	178s	179s	180s	181s	182s	183s	184s	185s	186s	187s	188s	189s	190s	191s	192s	193s	194s	195s	196s	197s	198s	199s	200s	201s	202s	203s	204s	205s	206s	207s	208s	209s	210s	211s	212s	213s	214s	215s	216s	217s	218s	219s	220s	221s	222s	223s	224s	225s	226s	227s	228s	229s	230s	231s	232s	233s	234s	235s	236s	237s	238s	239s	240s	241s	242s	243s	244s	245s	246s	247s	248s	249s	250s	251s	252s	253s	254s	255s	256s	257s	258s	259s	260s	261s	262s	263s	264s	265s	266s	267s	268s	269s	270s	271s	272s	273s	274s	275s	276s	277s	278s	279s	280s	281s	282s	283s	284s	285s	286s	287s	288s	289s	290s	291s	292s	293s	294s	295s	296s	297s	298s	299s	300s	301s	302s	303s	304s	305s	306s	307s	308s	309s	310s	311s	312s	313s	314s	315s	316s	317s	318s	319s	320s	321s	322s	323s	324s	325s	326s	327s	328s	329s	330s	331s	332s	333s	334s	335s	336s	337s	338s	339s	340s	341s	342s	343s	344s	345s	346s	347s	348s	34
---------	----	----	----	----	----	----	----	----	----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	----

species	isrange	fam- ser-	con- com-	GEN_Av- size	GEN_Av- age	male_av- er- WSP_Fe-	WSP_Fe- range	ELT_sin-	FM_Av- er-	AFB_Av- ey-	AFB_an- i-	AFB_min- power	WSP_Fe-														
													ELT_sin-	FM_Av- er-	AFB_Av- ey-	AFB_an- i-	AFB_min- power	ELT_sin-	FM_Av- er-	AFB_Av- ey-	AFB_an- i-	AFB_min- power	ELT_sin-	FM_Av- er-	AFB_Av- ey-	AFB_an- i-	AFB_min- power
Ere-Nympha- bia isude- t- dae ica	13	Bal	3	1.00	0.01	0.00	0.28	5	1.00	NANANANANANA	1	0	0	600	160	2.51	1	0	0	0	0	0	0	0	0	0	0
Ere-Nympha- bia itri- ar- dae ius	70	Bal	5	1.00	0.01	0.00	0.47	6	0.28	9	1	1	0	0	NANANA	400	210	4.00	1	0	0	0	0	0	0	1	1
Ere-Nympha- bia ityn- dar- dae rude	25	Bal	5	1.00	0.01	0.00	0.35	2	0.50	NANANANANANANANANA	1	20	150	0.02	1	0	0	0	0	0	0	0	0	0	0	0	
Ere-Nympha- bia iza- p- dae a- teri	6	Bal	5	1.00	0.01	0.00	0.38	4	0.28	9	NANANANANANANANANA	100	65	3.51	1	0	0	0	0	0	0	0	0	0	0	0	
Erythra- nis par- loyiidae	31	5	5	2.00	0.03	0.02	0.31	6	1.00	NANANANANANANANANA	0	210	6.04	1	0	0	0	0	0	0	0	0	0	0	0	0	
Erythra- nis par- loides idae	84	5	5	1.50	0.02	0.01	0.30	4	0.10	9	0	1	0	0	0	1	0	0	0	0	0	0	0	0	0	0	
Eu-Pie- chloda- so- nia	35	5	5	2.00	0.03	0.02	0.44	8	0.10	6	0	1	0	0	0	1	0	0	0	0	0	0	0	0	0	0	
Eu-Pie- chloda- bazae	51	3	3	1.50	0.02	0.01	0.34	4	0.50	0	1	0	0	0	0	1	1	0	0	800	4.00	1	0	0	0	0	0
Eu-Pie- chloda- belemia	63	5	5	2.00	0.02	0.00	0.40	8	0.30	4	1	0	0	0	0	1	0	0	0	150	0.00	1	0	0	0	0	0
Eu-Pie- chloda- achar- lonia	45	5	5	2.50	0.03	0.01	0.34	4	0.00	6	1	1	0	0	0	1	1	0	0	300	0.51	1	0	0	0	0	0
Eu-Pie- chloda- crameri	19	5	5	1.50	0.02	0.01	0.44	8	0.10	4	0	1	0	0	0	1	0	0	NANA	3.00	1	0	0	0	0	0	0
Eu-Pie- chloda- ev- ersi	45	5	5	2.00	0.02	0.00	0.40	8	0.28	9	NANANANANANA	1	0	0	170	800	4.00	1	0	0	0	0	0	0	0	0	

species	range	size	con- serv	GEN_Av- er	WSP_Fe- male_av- er	WSP_Fe- range	ELT_sin- er	FM_Av- er	AFB_Av- er	AFB_an- i	AFB_min- AFB_max	
Eu-Pier- chloa- granca- narien- sis	NA5	5	2.02	0.02	0.04	0.8	1.0	NANANANANANANANA	50	190	0.00	NANANANANANANA
Eu-Pier- chloa- ches- peridum	NA5	5	2.02	0.02	0.04	0.8	0.28	NANANANANANANANA	0	406	6.00	1 0 0 0 0 0 0
Eu-Pier- chloa- ain- su- laris	16 5	5	1.75	0.52	0.05	34.3	0.26	0 0 1 0 0 0 1 0 0 0	130	6.00	1 0 0 0 0 0 0	
Eu-Pier- chloa- ape- nia	14 5	5	1.50	0.02	0.01	0.34	0.4	0.8 1 1 0 0 0 1 0 0 0	210	6.00	1 0 0 0 0 0 0	
Eu-Pier- chloa- sim- plo- nia	32 5	5	0.75	0.51	0.05	43.6	0.36	3 1 1 0 0 0 1 1 0 100	140	6.00	1 0 0 0 0 0 0	
Eu-Pier- chloa- tagis	56 5	5	1.00	0.01	0.00	0.35	0.10	0.57 1 1 1 0 1 0 0 0	160	6.00	1 0 0 0 0 0 0	
Eu-Ly- me-caenidae do- nia_eu- me- don	44 15	5	1.00	0.01	0.00	0.28	0.8	0.20 8 1 1 0 0 0 1 0 0 0	240	6.00	1 0 0 0 0 0 0	
Eu-Ny- phydryas_au- rinidae	53 15	5	1.00	0.01	0.00	0.36	0.10	0.2 1 1 0 0 0 0 0 1 0	260	6.53	1 0 0 0 0 0 0	
Eu-Ny- phydryas_cyn- thiidae	58 15	5	0.75	0.51	0.05	37.0	0.28	9 1 1 0 0 0 0 1 1 900	220	6.00	1 0 0 0 0 0 0	
Eu-Ny- phydryas_des- fontainei	41 14	4	1.00	0.01	0.00	0.46	0.12	1.00 0 1 0 0 0 0 0 1 0	280	6.00	1 0 0 0 0 0 0	
Eu-Ny- phydryas_iduna dae	41 14	4	1.00	0.01	0.00	0.41	0.8	0.30 6 0 1 0 0 0 0 1 0	300	102.00	NANANANANANANA	

	con- fam-	GEN_serv-	Av-comer-	WSP_male-	Fe-av-	WSP_Fe-	ELT_sin-	FM_Av-	AFB_ey-	AFB_i-	AFB_an-	AFB_min-		
species	rangs	sizaga	GENE	NW_R	HW_R	Range	Vingc	HTA	HLI	SALH	M_BAA	FBB	CABD	slap
Eu-Ny-285al-5 phydryas_in- ter-dae	0.75	0.51	0.05	0.54	0.0	0.57	7 1 0 1 0 0 1 1	14000	0.00	1 0 0 0 0 0 0	0			
me- dia														
Eu-Ny-210al-5 phydryas_matura dae	0.75	0.51	0.05	0.54	1.9	0.06	0 0 1 1 1 0 0 1	10000	2.51	1 0 0 1 0 0 0	0			
FabN-885al-5 ciana_adippe dae	1.00	0.01	0.00	0.53	3.9	0.50	0 0 1 0 0 0 1 0 0 0	2300	0.53	1 0 1 0 0 0 0	1			
FabN-815al-5 ciana_elisa dae	1.00	0.01	0.00	0.49	0.6	1.00	0 0 1 0 0 0 1 0 0 0	400160	0.00	1 0 0 0 0 0 0	0			
FabN-645al-4 ciana_niobe dae	1.00	0.01	0.00	0.49	3.9	0.40	8 1 1 0 0 0 1 0 0 0	2500	0.53	1 0 0 0 0 0 0	0			
FavLy-775 niuscaquid cus	5	1.00	0.01	0.00	0.32	0.6	0.40	8 0 0 0 0 1 1 0 0 0	2000	0.02	0 0 0 0 1 0 1	1		
FreyLy-19 e-caenidae ria_trochy- lus	5 5	3.03	0.03	0.00	0.16	3.3	0.70	7ANANANANANANANA	0	1000	0.00	1 0 0 0 0 0 0	0	
GegHes-65_hos tro-peri- daniae	5	3.03	0.03	0.00	0.32	0.4	0.28	9 0 1 0 0 0 1 0 0 0	1900	0.00	1 0 0 0 0 0 0	1		
GegHes-76_punil peri- idae	5	3.03	0.03	0.00	0.32	0.4	0.36	4 0 1 0 0 0 1 0 0 0	1800	0.00	1 0 0 0 0 0 0	0		
GlaLy-605 copysyenidae che_alaxis	5 5	1.00	0.01	0.00	0.29	0.8	0.06	5 0 1 0 0 0 1 0 0 0	1900	0.00	1 0 0 0 0 0 1	1		
GlaLy-1035 copysyenidae che_melanops	5 5	1.00	0.01	0.00	0.27	0.0	0.22	4 0 1 0 0 0 1 0 0 0	100100	0.00	1 0 0 0 0 0 0	1		

[illegible]

species	range	size	con- serv	GEN_Av- er- GEN	WSP_Fe- male_av- er- WSP_Fe- range	ELT_sin- er- ELT	FM_Av- er- FM	AFB_Av- er- AFB	AFB_an- i- AFB	AFB_min- power
HipNymphal5	1.00	0.01	0.00	0.052	0.1	0.28	0	0	0	0
pardhia_blachieri	1.00	0.01	0.00	0.052	0.1	0.28	0	0	0	0
pardhia_chris- tensdae	1.00	0.01	0.00	0.051	0.16	0.28	0	0	0	0
HipNymphal5	1.00	0.01	0.00	0.056	0.8	0.28	0	0	0	0
pardhia_cret- ica dae	1.00	0.01	0.00	0.056	5.3	0.408	1	1	0	0
HipNymphal5	1.00	0.01	0.00	0.056	5.3	0.408	1	1	0	0
pardhia_cy- priedae	1.00	0.01	0.00	0.066	0.8	0.250	1	1	0	0
pardhia_fagi	1.00	0.01	0.00	0.066	0.8	0.250	1	1	0	0
HipNymphal5	1.00	0.01	0.00	0.057	5.5	0.354	1	1	0	0
pardhia_fatua	1.00	0.01	0.00	0.057	5.5	0.354	1	1	0	0
HipNymphal5	1.00	0.01	0.00	0.059	0.6	0.144	0	1	0	0
pardhia_fidia	1.00	0.01	0.00	0.059	0.6	0.144	0	1	0	0
HipNymphal5	1.00	0.01	0.00	0.059	0.6	0.28	0	0	0	0
pardhia_gomera	1.00	0.01	0.00	0.059	0.6	0.28	0	0	0	0
HipNymphal4	1.00	0.01	0.00	0.061	0.0	0.408	1	1	0	0
pardhia_hermione	1.00	0.01	0.00	0.061	0.0	0.408	1	1	0	0
HipNymphal4	1.00	0.01	0.00	0.052	0.1	0.28	0	0	0	0
pardhia_leighebi	1.00	0.01	0.00	0.052	0.1	0.28	0	0	0	0
HipNymphal5	1.00	0.01	0.00	0.044	0.8	NA	0	0	1	0
pardhia_maderen- sis dae	1.00	0.01	0.00	0.044	0.8	NA	0	0	1	0
HipNymphal5	1.00	0.01	0.00	0.051	0.16	0.28	0	0	0	0
pardhia_mersina	1.00	0.01	0.00	0.051	0.16	0.28	0	0	0	0

[illegible]

[illegible]

[illegible]

species	fam-	con-	GEN_Av-	male_av-	WSP_Fe-										FM_Av-			AFB_Av-			AFB_an-						
					er- WSP_Fe-	ELT_sin-	er-	ey-	i-	AFB_min-	range	size	GEN	AV	RA	SL	VE	VE	VE	FA	FA	FA	FA				
La-Nymphal5	3.03	0.03	0.00	0.42	5	0.35	4	0	1	0	0	0	1	0	0	0	1500	0.00	1	0	0	0	0	0			
sioni-																											
matdaaparamegaera																											
La-Nymphal5	1.50	0.02	0.01	0.38	6	0.26	8	0	1	0	0	0	1	0	0	0	1001	250	1	0	0	0	0	0	1		
sioni-																											
matdaapetropoli-																											
tana																											
LepPie67	5	5	2.00	0.02	0.00	0.38	8	0.40	8	1	1	0	0	0	1	0	0	0	2400	0.00	1	0	0	0	0	0	0
tidedaapuloncheli																											
LepPieNANA	1.00	0.01	0.00	0.42	0	NA	0	0	1	0	0	0	1	0	0	0	NANA	3.00	1	0	1	0	0	0	0		
tidedaagu-																											
ver-																											
nica																											
LepPie49	4	2	2.00	0.02	0.00	0.41	5	0.50	7	0	1	0	0	0	1	0	0	0	1400	0.00	1	0	0	0	0	0	1
tidedaamor-																											
sei																											
LepPieNA	5	5	2.00	0.03	0.02	0.38	0	1.00	0	0	1	0	0	0	1	0	0	0	2000	0.04	0	0	1	1	1	1	1
tidedare-																											
ali																											
LepPie1089	5	5	2.00	0.03	0.02	0.38	0	0.10	4	0	1	0	0	0	1	0	0	0	2000	0.04	1	0	0	0	0	0	1
tidedacinapis																											
LepLy-2875	5	5	3.03	0.03	0.00	0.25	0	0.00	5	0	1	0	0	0	1	0	0	0	1900	0.00	1	0	1	0	0	0	0
totesapiidae																											
hous																											
LibyNymphal5	1.25	0.01	0.50	0.54	4.5	0.70	7	0	0	0	1	1	0	0	0	0	1800	0.53	1	0	0	0	0	0	0	1	
i-																											
dae																											
Li-Nymphal5	1.00	0.01	0.00	0.56	8	0.28	9	0	0	0	1	1	1	0	0	0	0	1500	0.02	1	1	1	1	1	1	1	
memi-																											
tis_damilla																											
Li-Nymphal4	1.00	0.01	0.00	0.77	5.11	0.50	7	0	0	0	1	1	0	0	0	0	1001	552	0	0	0	0	1	1	0		
memi-																											
tis_dap-																											
uli																											
Li-Nymphal5	2.00	0.03	0.02	0.50	8	0.50	7	0	0	0	1	0	1	0	0	0	0	1950	0.04	1	0	1	1	1	1	1	
memi-																											
tis_dae																											
ducta																											

[illegible]

species	range	size	con- serv	GEN_Av- con- serv	WSP_Fe-		ELT_sin- er-	FM_Av- er-	AFB_Av- ey-	AFB_an- i-	AFB_min- AFB_max														
					male_av- er- WSP_Fe-	WSP_Fe- range																			
Lysdra_cad- bi- cans	Ly-40	5	5	1.00	0.01	0.00	0.39	0.6	1.000	0	1	0	0	NANANA	500	100	0	1	0	0	0	0	1		
Lysdra_cad- largus	Ly-60	15	5	2.00	0.03	0.02	0.31	0.6	0.183	1	0	0	0	1	0	0	100	190	0	0	0	0	0	1	1
Lysdra_cad- don	Ly-57	4	4	1.50	0.02	0.01	0.33	0.6	1.000	1	0	0	0	1	0	0	100	240	0	0	0	0	0	1	1
Lysdra_cad- do- nius	Ly-NA	5	NA	1.00	0.01	0.00	0.33	0.6	NA	0	1	1	0	0	1	0	0	0	290	0	0	0	0	0	0
Lysdra_cad- pana	Ly-50	5	5	2.00	0.02	0.00	0.34	0.4	1.000	0	1	0	0	1	0	0	100	90	0	0	0	0	0	0	1
ManNympha- iola i-chia dae	ManNympha- 5	1	5	1.00	0.01	0.00	0.41	0.5	0.289	1	1	0	0	1	1	0	0	800	5.00	1	0	0	0	0	0
ManNympha- iola i-cypri- coladae	ManNympha- 5	1	5	1.00	0.01	0.00	0	NANA	0.289	1	1	0	0	1	1	0	NANA	4.00	1	0	1	0	0	0	0
ManNympha- iola i-hali- car-dae	ManNympha- 4	1	4	1.00	0.01	0.00	0.44	0.5	0.289	1	1	0	0	1	1	0	0	450	3.04	1	0	0	0	0	0
ManNympha- iola i-ju- rtinalae	ManNympha- 5	1	5	1.00	0.01	0.00	0.41	0.5	0.289	1	1	0	0	0	1	0	0	250	0.06	1	0	1	0	0	0
ManNympha- iola i-megala dae	ManNympha- NA	1	NA	1.00	0.01	0.00	0.41	0.5	0.289	1	1	0	0	1	1	0	0	900	3.53	1	0	1	0	0	0
ManNympha- iola i-nurag dae	ManNympha- 5	1	5	1.00	0.01	0.00	0.38	0.4	1.000	1	1	0	0	1	1	0	400	0	0	0	0	0	0	0	0

species	isrange	fam- Nympha-	con- servat-	GEN_size	Av- con-	WSP_Fe-		male_av- er- WSP_Fe-	ELT_sin-	FM_Av- er-	AFB_Av-				AFB_an- ey- i- AFB_min-												
						GEN	WSP_Fe-				ELT	sin	FM	AFB		AFB											
Maniola telmessia dae	7	5	1.00	0.01	0.00	0.41	5	0.36	4	1	1	0	0	1	1	0	0	130	7	55	1	0	0	0	0	0	0
Melanargia naricia dae	33	5	1.00	0.01	0.00	0.53	6	0.28	N	A	N	A	N	A	N	A	N	A	300	120	2	0	0	1	0	0	0
Melanargia naricia dae	80	5	1.00	0.01	0.00	0.47	5	0.10	7	1	1	0	0	1	1	0	0	250	0	53	1	0	0	0	0	0	0
Melanargia naricia dae	79	5	1.00	0.01	0.00	0.48	4	0.50	0	0	1	0	0	N	A	N	A	0	260	0	0	1	0	0	0	0	0
Melanargia naricia dae	11	5	1.00	0.01	0.00	0.49	8	0.16	6	1	1	0	0	1	0	0	0	160	2	51	1	0	0	0	0	0	0
Melanargia naricia dae	73	5	1.00	0.01	0.00	0.55	10	0.70	7	1	1	0	0	1	0	0	0	240	0	02	1	0	0	0	0	0	0
Melanargia naricia dae	85	5	1.00	0.01	0.00	0.51	10	0.16	4	1	1	0	0	1	0	0	0	200	2	51	1	0	0	0	0	0	0
Melanargia naricia dae	26	5	1.00	0.01	0.00	0.48	4	0.28	N	A	N	A	N	A	N	A	N	A	600	800	2	51	1	0	0	0	0
Melanargia naricia dae	11	5	1.00	0.01	0.00	0.55	10	0.10	2	0	1	1	0	0	1	0	0	500	190	0	0	1	0	0	0	0	0
Melanargia naricia dae	26	5	1.50	0.02	0.01	0.44	4	0.40	N	A	N	A	N	A	N	A	N	A	0	1	1	0	180	2	51	1	0

	con- fam- serv-com- GEN_Av- size GEN_Fe- NYmin- Range WSP_Fe- er- WSP_Fe- range WSP_Fe
--	--

	con- fam- species	GEN_Av- serv-comer- range	WSP_Fe- male_av- er- WSP_Fe- range	ELT_sin- er- range	FM_Av- er- range	AFB_AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i- AFB_min- range	AFB_an- ey- i-<
--	-------------------------	---------------------------------	--	--------------------------	------------------------	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	-----------------------

[illegible]

species	family	range	con- serva-	GEN- size	Av- range	male_av- er- WSP_Fe-	WSP_Fe-										FM_Av-				AFB_Av-				AFB_min- power
							WSP_Fe- range	WSP_Fe- min	WSP_Fe- max	WSP_Fe- range	WSP_Fe- min	WSP_Fe- max	WSP_Fe- range	WSP_Fe- min	WSP_Fe- max	WSP_Fe- range	WSP_Fe- min	WSP_Fe- max	WSP_Fe- range	WSP_Fe- min	WSP_Fe- max	WSP_Fe- range	WSP_Fe- min	WSP_Fe- max	
ParN-11	Nyctali- dae	3.00	3.00	0.04	5.3	0.10	2	1	1	0	0	1	0	0	0	100	12.0	0	0	1	1	1	0	0	
ParN-15	Nyctali- dae	3.00	3.00	0.03	7.5	0.10	2	1	1	0	0	1	0	0	0	200	12.0	0	0	1	1	1	0	0	
ParPa-354	nas-pil- siusion- apollo idae	4	1.00	0.01	0.08	0.30	0.20	4	1	0	0	0	1	1	0	0	300	4.51	1	0	0	0	0	0	0
ParPa-374	nas-pil- siusion- memosyne idae	5	1.00	0.01	0.05	8.12	0.35	4	1	0	0	0	1	0	0	0	250	3.00	1	0	0	0	0	0	0
ParPa-48	nas-pil- siusion- phoe- busidae	4	1.00	0.01	0.06	5.30	0.16	8	1	1	0	0	1	0	0	160	12.0	1	0	0	0	0	0	0	
PelH-1	i- peri- dasidae	NANA	3.00	3.00	0.03	9.5	0.20	8	0	1	0	0	1	0	0	500	1500.4	1	0	0	0	0	0	0	
PheLy-354	garisac- ulidae con	4	1.00	0.01	0.03	6.04	0.57	7	1	1	0	0	1	1	0	100	120.0	1	0	0	0	0	0	0	
PheLy-522	garisac- uridae ion	2	1.00	0.01	0.03	6.6	0.50	0	1	1	0	0	1	0	0	0	240	4.00	1	0	0	0	0	0	0
PheLy-1884	garisac- mid- sit- hous	4	1.00	0.01	0.03	5.2	1.00	0	1	1	0	0	1	0	0	100	150.0	1	0	0	0	0	0	0	
PheLy-2523	garisac- oldus	3	1.00	0.01	0.03	4.04	1.00	0	1	1	0	0	1	0	0	100	150.0	1	0	0	0	0	0	0	
PierSi-1353	canalae	5	2.50	0.03	0.01	0.38	5.13	0.28	NANA	NANA	NANA	NANA	NANA	NANA	NANA	100	7.00	1	0	0	0	0	0	0	
PierSi-1353	si- dae cae	5	2.50	0.03	0.01	0.57	0.12	0.10	5	1	1	0	0	0	1	1	0	250	7.06	1	0	1	0	0	0

[illegible]

species	range	size	con-	GEN_Av-	WSP_Fe-	male_av-	er- WSP_Fe-	ELT_sin-	er-	FM_Av-	AFB_Av-	AFB_an-	AFB_min-
PolyNympha- go- i- nia_daga	155	5	2.00	0.03	0.02	0.45	0.2	0.57	7	0	1	0	0
PolyLy-57	5	5	1.00	0.01	0.00	0.34	0.8	0.70	7	1	1	0	0
ommaenidae tus_ad- me- tus													
PolyLy-54	5	5	1.50	0.02	0.01	0.31	0.6	0.20	4	1	1	0	0
ommaenidae tus_aman- dus													
PolyLy-13	5	5	1.00	0.01	0.00	0.30	0.5	1.00	0	0	1	0	0
ommaenidae tus_aroanien- sis													
PolyLy-NANANA	4	0	0.04	0.00	0.27	0.12	1.00	NANANANANANANANA	0	280	1.0	1	0
ommaenidae tus_celina													
PolyLy-NANANA	1	0	0.01	0.00	0.31	0.5	NANANANANANANANANA	170	30	2.51	NANANANANANANANA		
ommaenidae tus_damo- cles													
PolyLy-16	4	4	1.00	0.01	0.00	0.32	0.1	0.40	7	1	0	0	0
ommaenidae tus_da- mon													
PolyLy-30	5	5	1.00	0.01	0.00	0.35	0.6	0.10	9	1	1	0	0
ommaenidae tus_daph- nis													
PolyLy-24	5	5	1.00	0.01	0.00	0.35	0.6	1.00	NANANANANA	1	0	0	0
ommaenidae tus_do- lus													

|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|--|

species	range	size	con- serva- tion	GEN_Av- er- age	WSP_Fe- male_av- er- WSP_Fe- range	WSP_Fe- range	ELT_sin- er- age	FM_Av- er- age	AFB_Av- er- age	AFB_an- i- AFB_min- AFB_max																		
PolyLy-NA4	4	1.00	0.01	0.00	0.03	0.5	1.00	NANANANANANANANA	1200	2.00	NANANANANANANANA																	
ommaenidae																												
tus_nepho-																												
hip-																												
ta-																												
menos																												
PolyLy-26	4	4	1.00	0.01	0.00	0.33	0.6	1.00	0	1	1	0	0	1	0	0	1	0	0	0	0	0	1					
ommaenidae																												
tus_nivescens																												
PolyLy-9	3	3	1.00	0.01	0.00	0.33	0.6	0.5	7	NANANANANANANANA	400	160	NANANANANANANANANA															
ommaenidae																												
tus_or-																												
phi-																												
cus																												
PolyLy-59	5	4	1.00	0.01	0.00	0.31	0.6	0.4	7	1	1	0	0	1	0	0	100	190	1.00	1	0	0	0	0	0	0		
ommaenidae																												
tus_ri-																												
par-																												
tii																												
PolyLy-35	15	5	2.00	0.03	0.02	0.29	0.6	0.5	0	1	1	0	0	1	0	0	0	220	70	1.00	1	0	0	0	0	0	0	
ommaenidae																												
tus_ther-																												
sites																												
PolyLy-3	3	3	1.00	0.01	0.00	0.31	0.6	1.00	0	1	0	0	NANANA	1200	2.00	1	0	0	1	0	0	1	0	0	1			
ommaenidae																												
tus_vi-																												
o-																												
le-																												
tae																												
PonPier-60	5	5	1.50	0.02	0.01	0.47	0.0	0.1	02	1	0	0	0	1	0	0	0	340	40	1.00	1	0	0	0	0	0	0	
tia_dad-																												
lidice																												
PonPier-11	5	5	2.50	0.03	0.01	0.38	0.3	0.2	06	0	1	0	0	1	0	0	0	250	0	04	NANANANANANANANA							
tia_dad-																												
ridice																												
PonPier-21	5	5	3.00	0.03	0.00	0.41	0.5	0.0	06	1	0	1	0	0	1	1	0	0	300	0	53	1	0	0	0	0	0	0
tia_dap-																												
lidice																												

[illegible]

	con- fam- serv- con- GEN_Av- male_av- er- WSP_Fe- ELT_sin- FM_Av- AFB_Av- AFB_an- AFB_min-
species	range size GEN_Av- male_av- er- WSP_Fe- ELT_sin- FM_Av- AFB_Av- AFB_an- AFB_min-
Pseudocampoceros	1 1.00 0.01 0.05 0.04 1.00 NANANANANANANANANA14000.00 1 0 0 0 0 0 0
docihazdae	
ara_mercurius	
Pseudocampoceros	3 1.00 0.01 0.05 0.33 0.28 NANANANANANANANANA30000.02 1 0 0 0 0 0 0
docihazdae	
ara_orestes	
Pseudocampoceros	1 1.00 0.01 0.05 0.04 NANANANANANANANANANA20000.00 1 0 0 0 0 0 1
docihazdae	
ara_williamsi	
Pseudocampoceros	5 1.50 0.02 0.01 0.20 0.04 0.36 4 0 1 0 0 NANANA10000.00 1 0 0 0 0 0 1
dophalotus	
abencer-ragus	
Pseudocampoceros	5 1.00 0.01 0.05 0.22 0.02 1.00 0 1 0 0 NANANA70000.00 1 0 0 0 0 0 0
dophalotus	
adabargiae	
Pseudocampoceros	5 1.50 0.02 0.01 0.24 0.04 0.10 4 0 1 0 0 1 0 0 10000.00 1 0 0 0 0 0 0
dophalotus	
adabaton	
Pseudocampoceros	5 1.00 0.01 0.05 0.27 0.06 0.50 7 0 1 0 0 1 0 0 50000.00 1 0 1 0 0 0 0
dophalotus	
adavivius	
Pseudocampoceros	4 2.00 0.02 0.00 0.21 0.51 0.50 NANANANANANANANANA20000.00 1 0 0 0 0 0 1
dophalotus	
apanoptes	
Pseudocampoceros	4 1.50 0.02 0.01 0.23 0.06 0.06 9 0 1 0 0 NANANA0 20000.00 1 0 0 0 0 0 0
dophalotus	
aevicrama	
Pseudocampoceros	5 1.00 0.01 0.05 0.28 0.57 0.36 4 0 1 0 0 1 0 0 23000.53 1 0 0 0 0 0 0
gus_pallidus	
idae	
Pseudocampoceros	5 1.00 0.01 0.05 0.28 0.57 0.28 NANANANANANANANANA16000.02 1 0 0 0 0 0 0
gus_pandronidae	

species	family	range	con- serva-	GEN_ size	Av- range	male_av- er	WSP_Fe- er	WSP_Fe- Rang										ELT_sin- er	FM_Av- er	AFB_ ey-	AFB_ i-	AFB_ an- i-	AFB_min- power				
								WSP_Fe- Rang	WSP_Fe- Rang	WSP_Fe- Rang	WSP_Fe- Rang	WSP_Fe- Rang	WSP_Fe- Rang	WSP_Fe- Rang	WSP_Fe- Rang	WSP_Fe- Rang	WSP_Fe- Rang							WSP_Fe- Rang	WSP_Fe- Rang		
PyrHes3115	gus_pari- moridae	5	5	1.50	0.02	0.01	0.26	0.4	0.18	3	0	1	1	0	1	0	0	0	1900	0.55	1	0	0	0	0	0	1
PyrHes555	gus_pai- caliadae	5	5	1.00	0.01	0.00	0.28	0.5	0.57	NA	NA	NA	NA	NA	NA	NA	NA	1800	0.2	1	0	0	0	0	0	0	
PyrHes305	gus_pai- li-idae	5	5	1.00	0.01	0.00	0.29	0.6	0.50	NA	NA	NA	NA	NA	NA	NA	NA	1500	0.2	1	0	0	0	0	0	0	
PyrHes2645	gus_pai- thami- idae	5	5	1.00	0.01	0.00	0.32	0.4	0.20	8	1	1	0	0	1	1	0	0	1800	0.00	1	0	0	0	0	0	1
PyrHes1625	gus_pai- tau-idae	5	5	1.00	0.01	0.00	0.29	0.5	1.00	NA	NA	NA	NA	NA	NA	NA	NA	0	1000	0.2	1	0	0	0	0	0	0
PyrHes145	gus_pai- arae- idae	5	5	1.00	0.01	0.00	0.31	0.2	1.00	NA	NA	NA	NA	NA	NA	NA	NA	0	2000	0.02	1	0	0	0	0	0	0
PyrHes943	gus_pai- sii-idae	3	3	1.00	0.01	0.00	0.27	0.2	0.40	8	1	1	0	0	1	0	0	0	1000	0.02	1	0	0	0	0	0	1
PyrHes29	gus_pai- quieri- idae	NA	NA	1.00	0.01	0.00	0.28	0.3	0.50	NA	NA	NA	NA	NA	NA	NA	NA	1000	0.03	0.2	1	0	0	0	0	0	0
PyrHes9595	gus_pai- vae-idae	5	5	1.50	0.02	0.01	0.23	0.6	0.10	2	0	1	0	0	1	0	0	0	2500	0.04	1	0	0	0	0	0	0
PyrHesNA5	gus_pai- voldes- idae	5	5	1.50	0.02	0.01	0.23	0.6	NA	0	1	1	0	0	1	0	0	0	1800	0.53	1	0	0	0	0	0	1
PyrHes1055	gus_pai- pordidae	5	5	2.50	0.03	0.01	0.25	0.6	0.50	7	1	1	0	0	1	0	0	0	2800	0.00	1	0	0	0	0	0	1

[illegible]

[illegible]

species	fam-range	con-serv-size	GEN_Av-con-size	GEN_Min-con-size	GEN_Max-con-size	WSP_Fe-er- WSP_Fe-range	HIS_Min-range	HIS_Max-range	EVE_Min-range	EVE_Max-range	FAT_Min-range	FAT_Max-range	ELT_sin-er-ELT_sin-range	FM_Av-er-FM_Av-range	FM_Min-range	FM_Max-range	AFB_Min-range	AFB_Max-range	AFB_an-ey-i- AFB_min- AFB_max-power						
Thecla caenidae	Ly-6285	5	1.00	0.01	0.00	0.035	0.10	0.10	0	0	1	0	1	0	0	0	1600	0.51	1	0	0	1	0	0	0
tu-lae	Thy-Hell-4964	4	1.50	0.02	0.01	0.024	0.0	0.10	0	1	0	0	0	1	0	0	2500	0.53	1	0	0	0	0	0	0
cus pericleonidae	Thy-Hell-NA5	5	2.50	0.03	0.01	0.024	0.0	NA	0	1	0	0	NANANA	1000	0.00	NANANANANANANA									
cus perchristidae	Thy-Hell-10	5	1.00	0.01	0.00	0.032	0.5	0.10	NA	NANANANANANANANANA	0	2000	0.00	1	0	0	0	0	0	0	0	0	0	0	
cus phyraxidae	Thy-Hell-4265	5	1.00	0.01	0.00	0.025	0.7	0.00	0	1	0	0	0	1	1	0	2500	0.53	1	0	0	0	0	0	1
cus phirine-idae	Thy-Hell-8325	5	1.00	0.01	0.00	0.026	0.5	0.10	1	0	0	0	0	1	0	0	2600	0.53	1	0	0	0	0	0	1
cus perylvestrisidae	Tom-Ly-51	5	1.00	0.01	0.00	0.029	0.2	0.10	0	1	0	0	1	0	0	0	2000	1.00	0.02	1	0	0	0	0	0
lus caenidae	Tom-Ly-NA5	5	1.00	0.01	0.00	0.028	0.0	NA	1	1	0	0	1	0	0	0	2300	0.00	NANANANANANANANA						
li-caenidae	Tom-Ly-2	3	1.00	0.01	0.00	0.032	0.5	1.00	NA	NANANANANANANANANA	0	2300	0.51	NANANANANANANANA											
gelicaenidae	Tu-Ly-3	2	1.50	0.02	0.01	0.021	0.2	0.50	NA	NANANANANANANANANA	1500	0.00	1	0	0	0	0	0	0	0	0	0	0	1	
ranacae nithygetica	Van-Ly-13	5	2.00	0.03	0.02	0.059	0.3	0.30	0	1	0	0	1	0	0	0	2500	0.57	1	0	1	1	1	1	1
lanta dae	Van-Ly-13	5	2.00	0.03	0.02	0.054	0.6	0.00	0	1	0	0	1	0	0	0	3000	0.08	1	0	1	1	0	0	0
Van-Ly-13 dui i-dae																									

species	range	con- serva- tion	GEN_Av- size	GEN_Male- size	WSP_Fe- er- WSP_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe- range	HIS_Fe
---------	-------	------------------------	-----------------	-------------------	------------------------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	------------------	--------

```
# contains is another useful command to select columns.
```

```
be %>%
  select(species, contains("LEV")) %>%
  drop_na()
```

species	LEV_buried_layer	LEV_ground_layer	LEV_field_layer	LEV_shrub_layer	LEV_canopy_layer
Aglais_io	0	0	1	0	0
Aglais_urticae	0	0	1	0	0
Agriades_aquilo	0	0	1	0	0
Agriades_glandon	0	0	1	0	0
Agriades_optilete	0	1	1	0	0
Agriades_pyrenaicus	0	1	0	0	0
Agriades_zullichii	1	1	1	0	0
Anthocharis_cardamines	0	0	1	0	0
Anthocharis_euphenoides	0	0	1	0	0
Apatura_ilia	0	0	0	1	1
Apatura_iris	0	0	0	1	0
Apatura_metis	0	0	0	1	1
Aphantopus_hyperantus	0	1	1	0	0
Aporia_crataegi	0	0	1	1	0
Araschnia_levana	0	0	1	0	0
Archon_apollinus	0	0	1	0	0
Arethusana_arethusa	0	1	1	0	0
Argynnis_pandora	1	1	1	0	0
Argynnis_paphia	0	0	1	0	0
Aricia_agestis	0	1	1	0	0
Aricia_anteros	0	0	1	0	0
Aricia_artaxerxes	0	1	1	0	0
Aricia_cramera	0	1	1	0	0
Aricia_morroneis	0	0	1	0	0
Aricia_nicias	0	0	1	0	0
Azanus_ubaldus	0	0	0	1	0
Boloria_aquilonaris	0	1	1	0	0
Boloria_dia	0	0	1	0	0
Boloria_eunomia	0	0	1	0	0
Boloria_euphrosyne	0	0	1	0	0
Boloria_freija	0	0	0	1	0
Boloria_frigga	0	0	0	1	0

species	LEV_buried_layer	LEV_ground_layer	LEV_field_layer	LEV_shrub_layer	LEV_canopy_layer
Boloria_graeca	0	0	1	0	0
Boloria_improba	0	0	1	0	0
Boloria_napaea	0	1	1	0	0
Boloria_pales	0	0	1	0	0
Boloria_selene	0	0	1	0	0
Boloria_thore	0	0	1	0	0
Boloria_titania	0	0	1	0	0
Borbo_borbonica	0	1	1	0	0
Brenthis_daphne	0	0	0	1	0
Brenthis_hecate	0	0	1	0	0
Brenthis_ino	0	1	1	0	0
Brintesia_circe	0	1	1	0	0
Callophrys_rubi	0	1	1	1	0
Carcharodus_alceae	0	0	1	0	0
Carcharodus_baeticus	0	1	1	0	0
Carcharodus_floccifera	0	0	1	0	0
Carcharodus_lavatherae	0	1	1	0	0
Carterocephalus_palaemon	0	0	1	0	0
Carterocephalus_silvicola	0	0	1	0	0
Catopsilia_florella	0	0	0	1	1
Celastrina_argiolus	0	0	0	1	1
Charaxes_jasius	0	0	0	0	1
Chazara_briseis	0	1	1	0	0
Coenonympha_arcania	0	0	1	0	0
Coenonympha_dorus	0	0	1	0	0
Coenonympha_glycerion	0	0	1	0	0
Coenonympha_hero	0	0	1	0	0
Coenonympha_leander	0	0	1	0	0
Coenonympha_oedippus	0	0	1	0	0
Coenonympha_pamphilus	0	0	1	0	0

species	LEV_buried_layer	LEV_ground_layer	LEV_field_layer	LEV_shrub_layer	LEV_canopy_layer
Coenonympha_rhodopen- sis	0	0	1	0	0
Coenonympha_thyr- sis	0	0	1	0	0
Coenonympha_tullia	0	0	1	0	0
Colias_alfacariensis	0	1	1	0	0
Colias_aurorina	0	0	1	0	0
Colias_caucasica	0	0	1	0	0
Colias_chrysotheme	0	1	1	0	0
Colias_crocea	0	0	1	0	0
Colias_hecla	0	1	0	0	0
Colias_hyale	0	1	1	0	0
Colias_myrmidone	0	0	1	0	0
Colias_palaeno	0	1	1	0	0
Colias_phicomone	0	1	0	0	0
Colias_tyche	0	1	0	0	0
Colotis_evagore	0	0	1	0	0
Cupido_alcetas	0	0	1	0	0
Cupido_argiades	0	0	1	0	0
Cupido_decoloratus	0	0	1	0	0
Cupido_lorquinii	0	0	1	0	0
Cupido_minimus	0	0	1	0	0
Cupido_osiris	0	0	1	0	0
Cyaniris_semiargus	0	0	1	0	0
Cyclotrius_webbianus	0	0	1	0	0
Danaus_chrysippus	0	0	1	0	0
Danaus_plexippus	0	0	1	0	0
Erebia_aethiops	0	1	1	0	0
Erebia_cassioides	0	1	1	0	0
Erebia_claudina	0	0	1	0	0
Erebia_embla	0	0	1	0	0
Erebia_epiphron	0	0	1	0	0
Erebia_eriphyle	0	0	1	0	0
Erebia_euryale	0	0	1	0	0
Erebia_gorge	0	0	1	0	0
Erebia_ligea	0	1	1	0	0
Erebia_manto	0	0	1	0	0
Erebia_medusa	0	1	1	0	0
Erebia_meolans	0	1	1	0	0
Erebia_mnestra	0	0	1	0	0
Erebia_montana	0	1	0	0	0

species	LEV_buried_layer	LEV_ground_layer	LEV_field_layer	LEV_shrub_layer	LEV_canopy_layer
Erebia_neoridas	0	0	1	0	0
Erebia_oeme	0	1	1	0	0
Erebia_palarica	0	1	1	0	0
Erebia_pandrose	0	1	1	0	0
Erebia_pluto	0	1	1	0	0
Erebia_polaris	0	0	1	0	0
Erebia_triarius	0	1	1	0	0
Erynnis_tages	0	0	1	0	0
Euchloe_ausonia	0	0	1	0	0
Euchloe_bazae	0	1	0	0	0
Euchloe_belemia	0	1	0	0	0
Euchloe_charlonia	0	1	1	0	0
Euchloe_crameri	0	0	1	0	0
Euchloe_insularis	0	0	1	0	0
Euchloe_penia	0	1	1	0	0
Euchloe_simplonia	0	1	1	0	0
Euchloe_tagis	0	1	1	1	0
Eumedonia_eumedon	0	1	1	0	0
Euphydryas_aurinia	0	1	1	0	0
Euphydryas_cynthia	0	1	1	0	0
Euphydryas_desfontainii	0	0	1	0	0
Euphydryas_iduna	0	0	1	0	0
Euphydryas_intermedia	0	1	0	1	0
Euphydryas_materna	0	0	1	1	1
Fabriciana_adippe	0	0	1	0	0
Fabriciana_elisa	0	0	1	0	0
Fabriciana_niobe	0	1	1	0	0
Favonius_quercus	0	0	0	0	1
Gegenes_nostrodamus	0	0	1	0	0
Gegenes_pumilio	0	0	1	0	0
Glaucopsyche_alexis	0	0	1	0	0
Glaucopsyche_melanops	0	0	1	0	0
Glaucopsyche_paphos	0	0	1	0	0
Gonepteryx_cleobule	0	0	0	1	1

species	LEV_buried_layer	LEV_ground_layer	LEV_field_layer	LEV_shrub_layer	LEV_canopy_layer
Gonepteryx_cleopatra	0	0	0	1	0
Gonepteryx_farinosa	0	0	0	1	0
Gonepteryx_rhamni	0	0	0	1	0
Hamearis_lucina	0	1	1	0	0
Hesperia_comma	0	1	1	0	0
Heteropterus_morphheus	0	0	1	0	0
Hipparchia_azorina	0	0	1	0	0
Hipparchia_cypriensis	0	1	1	0	0
Hipparchia_fagi	0	1	1	0	0
Hipparchia_fatua	0	1	1	0	0
Hipparchia_fidia	0	0	1	0	0
Hipparchia_hermione	0	1	1	0	0
Hipparchia_maderensis	0	0	1	0	0
Hipparchia_pellucida	0	1	1	0	0
Hipparchia_semele	0	1	1	0	0
Hipparchia_statilinus	0	1	1	0	0
Hipparchia_syriaca	0	1	1	0	0
Hipparchia_volgensis	0	1	1	0	0
Hipparchia_wyssii	0	1	1	0	0
Hyponephele_lupina	0	1	1	0	0
Hyponephele_lycaon	0	0	1	0	0
Iolana_iolas	0	1	1	0	0
Iphiclides_feisthamelii	0	0	0	1	0
Iphiclides_podalirius	0	1	1	0	0
Issoria_lathonia	0	1	1	0	0
Kirinia_climene	0	1	1	0	0
Kretania_eurypilus	0	0	1	0	0
Kretania_hesperica	0	0	1	0	0
Kretania_psylorita	0	0	1	0	0
Kretania_pylaon	0	0	1	0	0
Kretania_trappi	0	1	1	0	0
Laeosopis_roboris	0	1	0	0	1
Lampides_boeticus	0	0	1	0	0
Lasiommata_maera	0	0	1	0	0
Lasiommata_megera	0	0	1	0	0

species	LEV_buried_layer	LEV_ground_layer	LEV_field_layer	LEV_shrub_layer	LEV_canopy_layer
Lasiom-	0	0	1	0	0
mata_paramegaera					
Lasiom-	0	0	1	0	0
mata_petropolitana					
Leptidea_duponcheli	0	1	1	0	0
Leptidea_juvernica	0	0	1	0	0
Leptidea_morsei	0	0	1	0	0
Leptidea_reali	0	0	1	0	0
Leptidea_sinapis	0	0	1	0	0
Leptotes_pirithous	0	0	1	0	0
Libythea_celtis	0	0	0	0	1
Limenitis_camilla	0	0	0	1	1
Limenitis_populi	0	0	0	0	1
Limenitis_reducta	0	0	0	1	0
Lopinga_achine	0	1	1	0	0
Lycaena_alciphron	0	1	1	0	0
Lycaena_candens	0	1	0	0	0
Lycaena_dispar	0	1	1	0	0
Lycaena_helle	0	0	1	0	0
Lycaena_hippothoe	0	1	1	0	0
Lycaena_phlaeas	0	1	0	0	0
Lycaena_tityrus	0	1	1	0	0
Lycaena_virgaureae	0	1	0	0	0
Lysandra_albicans	0	0	1	0	0
Lysandra_bellargus	0	1	0	0	0
Lysandra_coridon	0	1	0	0	0
Lysandra_corydonius	0	1	1	0	0
Lysandra_hispana	0	0	1	0	0
Maniola_chia	0	1	1	0	0
Maniola_cypricola	0	1	1	0	0
Maniola_halicarnas-	0	1	1	0	0
sus					
Maniola_jurtina	0	1	1	0	0
Maniola_megala	0	1	1	0	0
Maniola_nurag	0	1	1	0	0
Maniola_telmessia	0	1	1	0	0
Melanargia_galathea	0	1	1	0	0
Melanargia_ines	0	0	1	0	0
Melanargia_lachesis	0	1	1	0	0
Melanargia_larissa	0	1	1	0	0

species	LEV_buried_layer	LEV_ground_layer	LEV_field_layer	LEV_shrub_layer	LEV_canopy_layer
Melanargia_occitanica	0	1	1	0	0
Melanargia_russiae	0	1	1	0	0
Melitaea_arduinna	0	1	0	0	0
Melitaea_asteria	0	1	1	0	0
Melitaea_athalia	0	1	1	0	0
Melitaea_aurelia	0	0	1	0	0
Melitaea_britomartis	0	0	1	0	0
Melitaea_cinxia	0	1	1	0	0
Melitaea_deione	0	1	1	0	0
Melitaea_diamina	0	1	1	0	0
Melitaea_didyma	0	0	1	0	0
Melitaea_ornata	0	0	1	0	0
Melitaea_parthenoides	0	0	1	0	0
Melitaea_phoebe	0	0	1	0	0
Melitaea_trivia	0	0	1	0	0
Melitaea_varia	0	1	1	0	0
Minois_dryas	0	1	1	0	0
Muschampia_proto	0	1	1	0	0
Nymphalis_antiope	0	0	0	1	1
Nymphalis_poly-chloros	0	0	0	1	1
Nymphalis_vaulsleyana	0	0	0	0	1
Nymphalis_xanthomelas	0	0	0	0	1
Ochlodes_sylvanus	0	0	1	0	0
Oeneis_bore	0	0	1	0	0
Oeneis_jutta	0	0	1	0	0
Oeneis_norna	0	0	1	0	0
Papilio_alexanor	0	0	1	0	0
Papilio_hospiton	0	1	1	0	0
Papilio_machaon	0	0	1	0	0
Pararge_aegeria	0	0	1	0	0
Pararge_xiphia	0	1	1	0	0
Pararge_xiphioides	0	1	1	0	0
Parnassius_apollo	0	1	0	0	0
Parnassius_mnemosyne	0	1	0	0	0
Parnassius_phoebus	0	1	1	0	0

species	LEV_buried_layer	LEV_ground_layer	LEV_field_layer	LEV_shrub_layer	LEV_canopy_layer
Pelopidas_thrax	0	0	1	0	0
Phengaris_alcon	1	1	1	0	0
Phengaris_arion	1	1	1	0	0
Phengaris_nausithous	1	1	1	0	0
Phengaris_teleius	1	1	1	0	0
Pieris_brassicae	0	1	1	0	0
Pieris_bryoniae	0	1	0	0	0
Pieris_cheiranthi	0	1	1	1	0
Pieris_ergane	0	1	1	0	0
Pieris_krueperi	0	0	1	0	0
Pieris_mannii	0	1	1	0	0
Pieris_napi	0	1	1	0	0
Pieris_rapae	0	1	1	0	0
Plebejus_argus	0	1	1	0	0
Plebejus_argyrogonomon	0	0	1	0	0
Plebejus_idas	0	0	1	0	0
Polygonia_c-album	0	0	1	1	1
Polygonia_egea	0	0	1	0	0
Polyommatus_admetus	1	1	1	0	0
Polyommatus_aman-dus	0	1	1	0	0
Polyommatus_aroaniensis	0	0	1	0	0
Polyommatus_damon	0	1	0	0	0
Polyommatus_daph-nis	0	1	1	0	0
Polyommatus_dory-las	0	1	1	0	0
Polyommatus_eros	0	0	1	0	0
Polyommatus_escheri	0	1	1	0	0
Polyommatus_fabres-sei	0	0	1	0	0
Polyommatus_ful-gens	0	0	1	0	0
Polyommatus_golgus	1	1	1	0	0
Polyommatus_icarus	0	1	1	0	0

species	LEV_buried_layer	LEV_ground_layer	LEV_field_layer	LEV_shrub_layer	LEV_canopy_layer
Polyommatus_nivescens	0	1	1	0	0
Polyommatus_ripar-tii	0	1	1	0	0
Polyommatus_ther-sites	0	1	1	0	0
Polyommatus_viole-tae	0	0	1	0	0
Pontia_callidice	0	1	0	0	0
Pontia_chloridice	0	0	1	0	0
Pontia_daplidice	0	0	1	0	0
Pontia_edusa	0	1	0	0	0
Proterebia_phegea	0	1	1	0	0
Pseudochazara_an-thelea	1	1	1	0	0
Pseu-dophilotes_abencer-ragus	0	0	1	0	0
Pseu-dophilotes_barbagiae	0	0	1	0	0
Pseudophilotes_ba-ton	0	0	1	0	0
Pseudophilotes_bav-ius	0	0	1	0	0
Pseudophilotes_vi-crama	0	0	1	0	0
Pyrgus_alveus	0	0	1	0	0
Pyrgus_armoricanus	0	0	1	1	0
Pyrgus_carthami	0	1	1	0	0
Pyrgus_cirsii	0	1	1	0	0
Pyrgus_malvae	0	0	1	0	0
Pyrgus_malvoides	0	1	1	0	0
Pyrgus_onopordi	0	1	1	0	0
Pyrgus_serratulae	0	1	1	0	0
Pyronia_bathseba	0	1	1	0	0
Pyronia_cecilia	0	0	1	0	0
Pyronia_tithonus	0	1	1	0	0
Satyrium_acaciae	0	1	0	1	0
Satyrium_esculi	0	0	0	1	1
Satyrium_ilicis	0	0	0	1	0
Satyrium_ledereri	0	0	0	1	0

species	LEV_buried_layer	LEV_ground_layer	LEV_field_layer	LEV_shrub_layer	LEV_canopy_layer
Satyrium_pruni	0	0	0	1	0
Satyrium_spini	0	0	0	1	0
Satyrium_w-album	0	0	0	1	1
Satyrus_actaea	0	1	1	0	0
Satyrus_ferula	0	1	1	0	0
Scolitantides_orion	0	0	1	0	0
Speyeria_aglaja	0	1	1	0	0
Spialia_sertorius	0	1	1	0	0
Spialia_therapne	0	1	1	0	0
Tarucus_balkanicus	0	0	1	0	0
Tarucus_theophrastus	0	0	0	1	0
Thecla_betulae	0	0	0	1	0
Thymelicus_acteon	0	0	1	0	0
Thymelicus_christi	0	0	1	0	0
Thymelicus_lineola	0	0	1	0	0
Thymelicus_sylvestris	0	0	1	0	0
Tomares_ballus	0	0	1	0	0
Tomares_callimachus	1	1	1	0	0
Vanessa_atalanta	0	0	1	0	0
Vanessa_cardui	0	0	1	0	0
Vanessa_virginiensis	0	0	1	0	0
Vanessa_vulcania	0	0	1	0	0
Ypthima_asterope	0	0	1	0	0
Zegris_eupheme	0	0	1	0	0
Zerynthia_cassandra	0	0	1	0	0
Zerynthia_cerisy	0	0	1	0	0
Zerynthia_cretica	0	0	1	0	0
Zerynthia_polyxena	0	0	1	0	0
Zerynthia_rumina	0	1	1	0	0

Functions like `contains` can be powerful for filtering and selecting. `starts_with` and `ends_with` work just the same way and are equally useful.

4.7 Create new variables with `mutate()`

This is a very powerful and flexible function that uses existing variables to create novel ones. Let's look at a simple example

```
# Create a new variable summarizing all the overwintering stages that are not adults
be %>%
  mutate(OWS_juvenile = 1-OWS_adult) %>%
  select(OWS_juvenile, OWS_adult) %>%
  drop_na()
```

OWS_juvenile	OWS_adult
0	1
0	1
0	1
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
1	0
0	1
1	0
1	0
1	0
0	1
1	0
1	0

[illegible]

[illegible]

[illegible]

[illegible]

OWS_juvenile	OWS_adult
1	0
1	0
1	0
1	0
1	0
1	0
1	0
0	1
0	1

```
# Determine how different the protection levels between EU and Europe and extract the species
be %>%
  mutate(protect_diff = abs(conserv.europe - conserv.eu)) %>%
  filter(protect_diff > 2)
```

[illegible]

4.8 Exercise

- From our dataset, filter out all butterflies with average wingspans larger than 60mm and smaller than 30mm. Only keep the species that have a conservation classification on the EU level. Only keep the species names and all variables associated with adult feeding, and store this in a new data frame. How many rows and columns does the new data frame have?
- Re-calculate the generation range from the provided minima and maxima. Check if your calculations match the original range values given in the data.

4.9 group_by() and summarise() as powerful data exploration tools

Although `dplyr` has a simpler syntax overall, everything we have looked at so far could have been done fairly easily with base R functions: data frame filtering, sorting, and adding and removing columns. One of the strengths of `dplyr` is explorative data analysis, and this is where `group_by()` and `summarize()` are really helpful. We'll only look at very simple examples today.

When browsing through the complete data table, it is very hard to recognize any patterns. Let's assume we wanted to compare the average wing span of butterflies with that overwinter as adults vs all other butterflies:

```
# Are butterflies that overwinter as adults larger than other species?
be %>%
  drop_na() %>%
  group_by(OWS_adult) %>%
  summarise(mean_wsp = mean(WSP_Female_average))
```

OWS_adult	mean_wsp
0	39.42489
1	51.85294

After choosing which variable to group by (here: `OWS_adult`), `summarise()` then calculates a function for each group. In our simple example, there are 2 groups: 0 (not overwintering as adult) and 1 (overwintering as adult); and the function to be calculated is the mean of the female wing span. This is a very flexible set of functions, because you can group by multiple groups and also use `summarise()` with many different functions (e.g., `mean()`, `sum()`, `min()`, `max()`, `median()` – just to name a few). Let's look at a more complex example:

```
# Let's add another group. How large is the standard deviation? How large is each group?
be %>%
  drop_na() %>%
  group_by(OWS_adult, LEV_ground_layer) %>%
  summarise(mean_wsp = mean(WSP_Female_average),
            sd = sd(WSP_Female_average),
            group_size=n())
```

``summarise()`` has grouped output by `'OWS_adult'`. You can override using the ``.groups`` argument.

OWS_adult	LEV_ground_layer	mean_wsp	sd	group_size
0	0	38.12174	12.312380	115
0	1	40.69492	12.181504	118
1	0	54.26667	9.460192	15
1	1	33.75000	12.374369	2

4.10 More exercises

Using `dplyr` functions, determine

- If butterflies overwintering as pupae have higher level of legal protection
- If butterflies occurring at higher altitudes on average have a higher level of protection
- If feeding on honeydew is more common in larger butterflies.
- How many butterfly species are there per family?

For a–c also determine how many species belong to each group.

5 The ggplot2 package

5.1 Very (!) brief introduction

ggplot2 is a graphing library, i.e., a tool to make graphs in R. Compared with base graphs and other graphics packages, it comes with a number of advantages:

- Beautiful!
- Highly customizable (which is not always necessary though)
- Easiest way to create very complex plots
- Tightly integrated into the **tidyverse**

Compared with other packages the major drawbacks would be that it comes with a steep(ish) learning curve and is probably less intuitive for beginners. The reason is that **ggplot2** doesn't have fixed commands for scatterplots, boxplots, barplots, etc, but rather creates the plot in layers. The most important elements (or layers) of a plot in **ggplot2** are:

- *Data*: as we are still in the **tidyverse**, this is always a data frame
- *Aesthetics*: i.e., **what** you want to plot. Often, this will correspond to variables (columns) in your dataframe
- *Geometric objects*: i.e., **how** you want to plot the data. This can be points, bars, boxplots, lines, etc..
- *Facets*: more about this later
- *Additional (optional) adjustments*: this includes themes that specify the overall design

5.2 Building up the plot

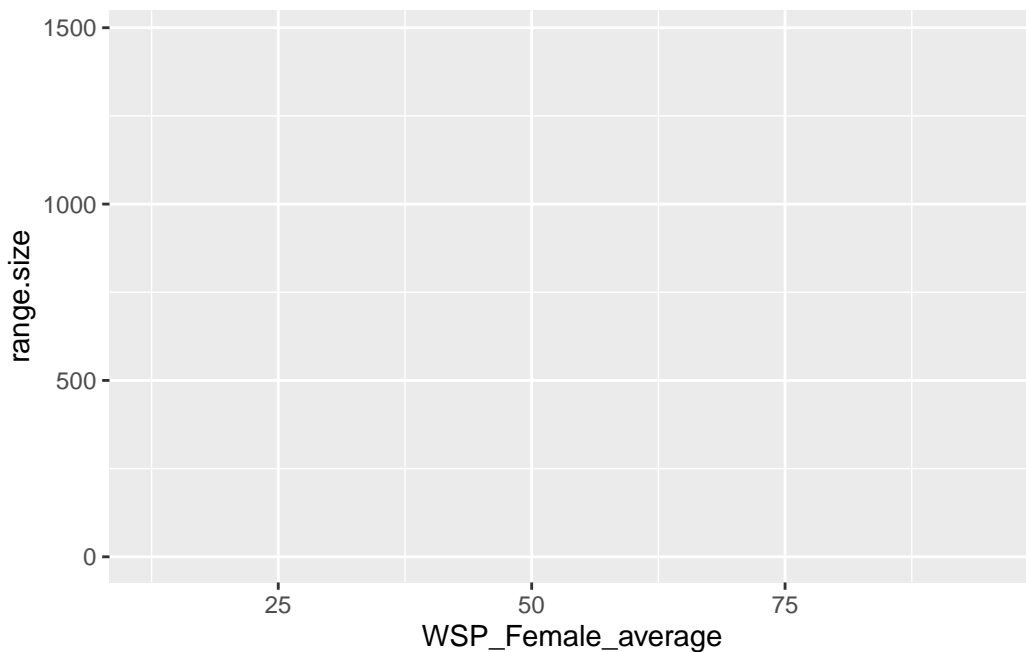
We will start off with a very simple scatterplot and gradually increase the complexity to illustrate **ggplot2** functionality.

```
# data and packages
library(tidyverse, quietly = TRUE)
```

```
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr      1.1.4      v readr      2.1.6
v forcats    1.0.1      v stringr    1.6.0
v ggplot2     4.0.1      v tibble     3.3.1
v lubridate   1.9.4      v tidyr      1.3.2
v purrr       1.2.1
-- Conflicts ----- tidyverse_conflicts() --
x dplyr::filter() masks stats::filter()
x dplyr::lag()     masks stats::lag()
i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become
```

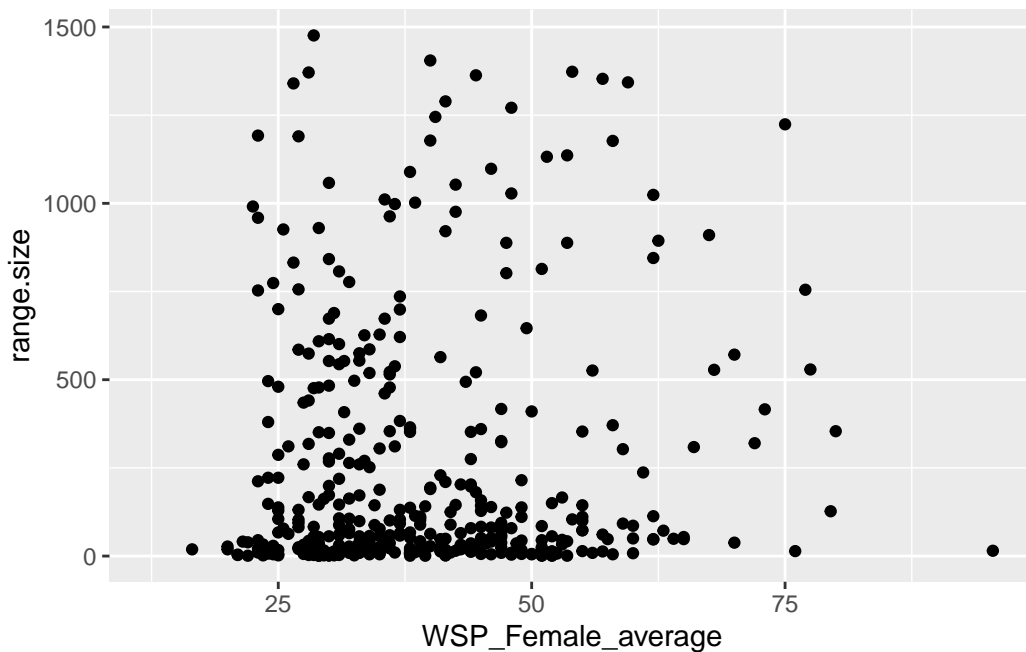
```
be <- read.table("data/butterfly_ecology.csv", header = TRUE, sep = ",")

# simple plot
be %>%                                # DATA
  ggplot(aes(x = WSP_Female_average, # AESTHETICS
             y = range.size))
```



In the above example, the data is the data frame that we have been using the whole time. Notice how we can simply pipe it to `ggplot2`. `aes` specifies our aesthetics, i.e., **what** we want to plot. What is missing?

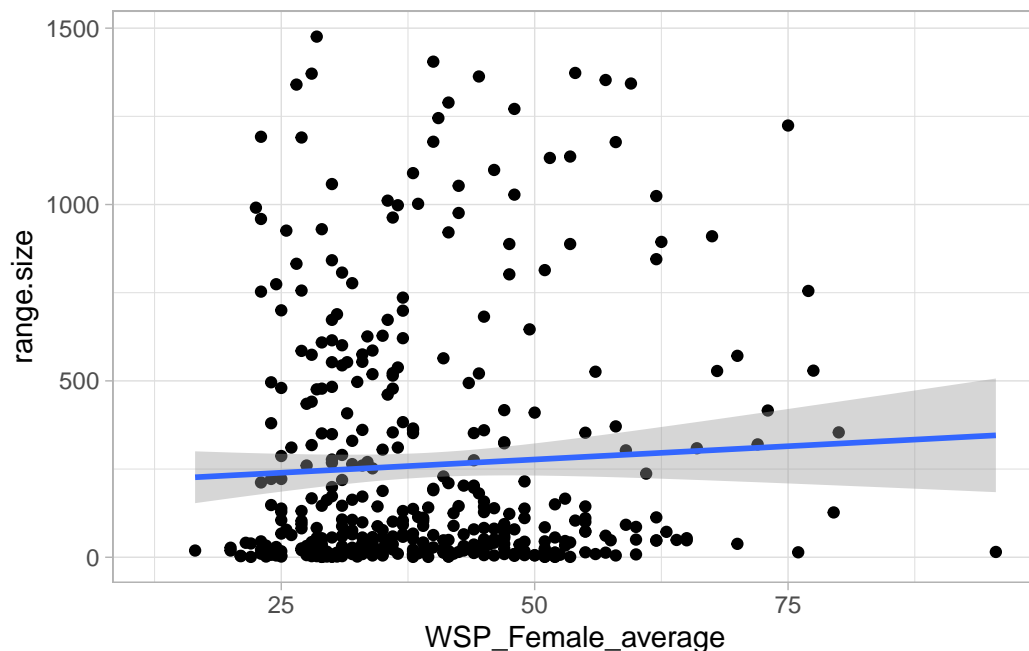

```
# simple scatter plot
be %>%                                # DATA
  ggplot(aes(x = WSP_Female_average, # Aesthetics
             y = range.size)) +
  geom_point()                        # Geometric object
```



The geometric object, i.e., **how** we want to plot our aesthetics. Notice that elements in `ggplot2` are added with the `+` symbol (this is specific to `ggplot2`). We can add more geometric objects that will use the same aesthetics:

```
# lets add another geom (a regression line), and also change the theme
be %>%                                # DATA
  ggplot(aes(x = WSP_Female_average, # Aesthetics
             y = range.size)) +
  geom_point() +                      # Geometric object
  geom_smooth(method = "lm") +       # Another geometric object
  theme_light()                      # Let's also change the theme
```

``geom_smooth()`` using formula = `'y ~ x'`



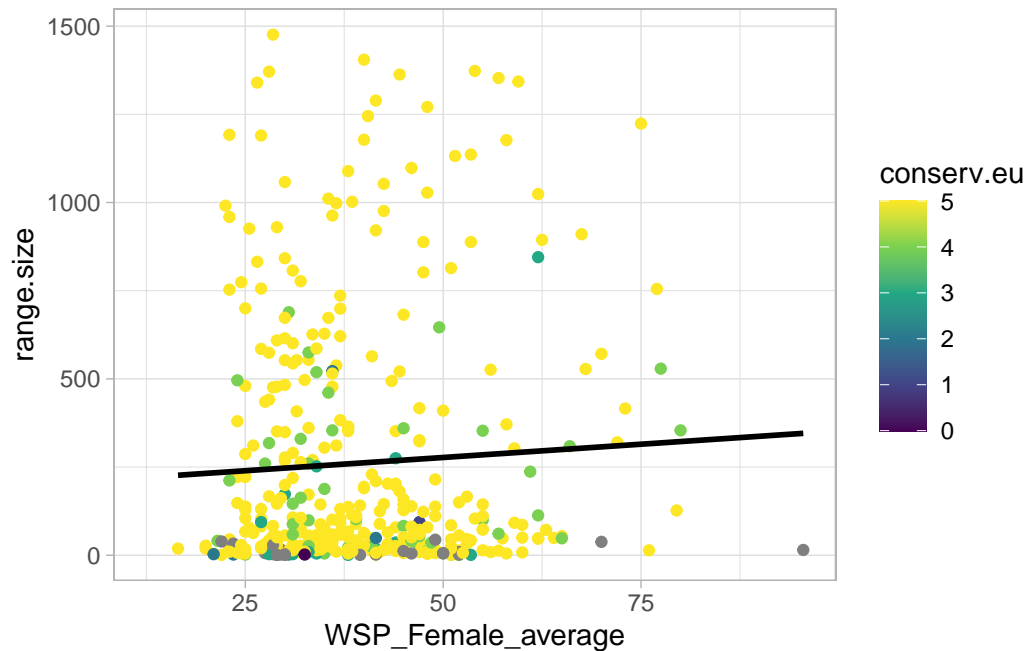
Changing the theme changes many layout options. For different applications, different themes might be appropriate. There are many additional themes available through packages such as [ggthemr](#) or [ggthemes](#).

A bit more on aesthetics: have you noticed that you only specify x and y once, and all geoms know **what** you want to plot. You can also specify additional aesthetics for each geom.

```
# Add additional aesthetics, here: we want to plot the sonsevation status. How? With colour

be %>%                                # DATA
  ggplot(aes(x = WSP_Female_average,    # Aesthetics
             y = range.size)) +
  geom_point(aes(color = conserv.eu)) + # aesthetics specific to the points only
  geom_smooth(method = "lm",
             color = "black",
             se = FALSE) +
  theme_light() +
  scale_color_viridis_c()              # let's use some nicer colors
```

```
`geom_smooth()` using formula = 'y ~ x'
```



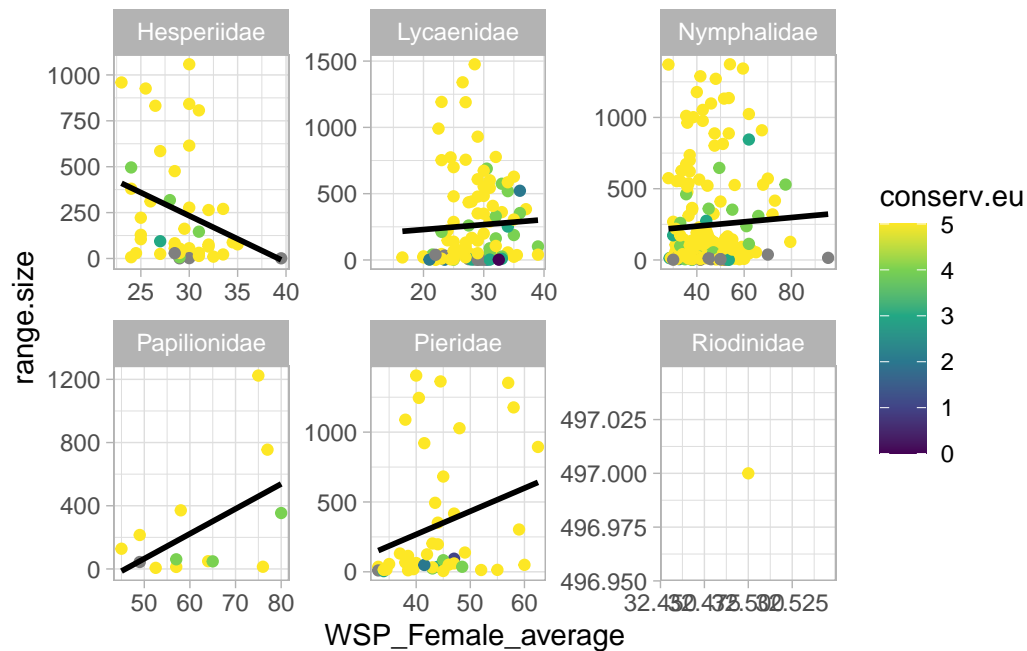
Aesthetics can be added through colors or shapes

5.3 Faceting

So far, we have cramped as much information as possible into the plot. This was useful to illustrate the functionality of `ggplot2`, but did not create very readable plots. Often, faceting is a better solution. The implementation in `ggplot2` is very straightforward.

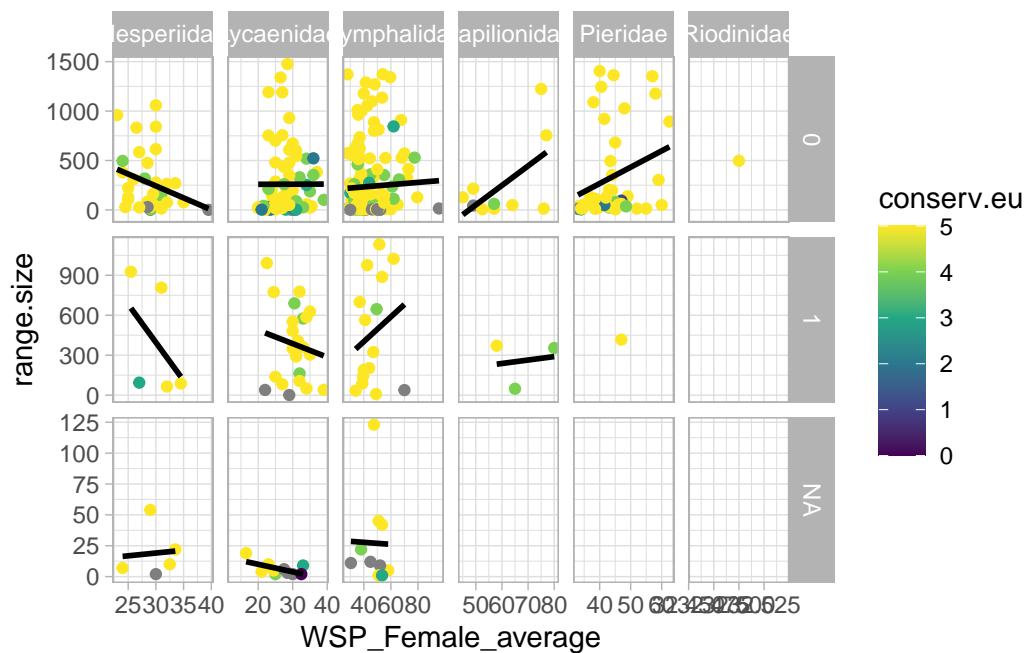
```
# Same example as before, faceted over family
be %>%
  ggplot(aes(x = WSP_Female_average,
             y = range.size)) +
  geom_point(aes(color = conserv.eu)) +
  geom_smooth(method = "lm",
             color = "black",
             se = FALSE) +
  theme_light() +
  scale_color_viridis_c() +
  facet_wrap(~family, scales = "free")
```

``geom_smooth()`` using formula = 'y ~ x'



```
# And now, faceting over 2 variables
be %>%
  ggplot(aes(x = WSP_Female_average,
             y = range.size)) +
  geom_point(aes(color = conserv.eu)) +
  geom_smooth(method = "lm",
             color = "black",
             se = FALSE) +
  theme_light() +
  scale_color_viridis_c() +
  facet_grid(OWS_egg ~ family, scales = "free") # faceting
```

`geom_smooth()` using formula = 'y ~ x'

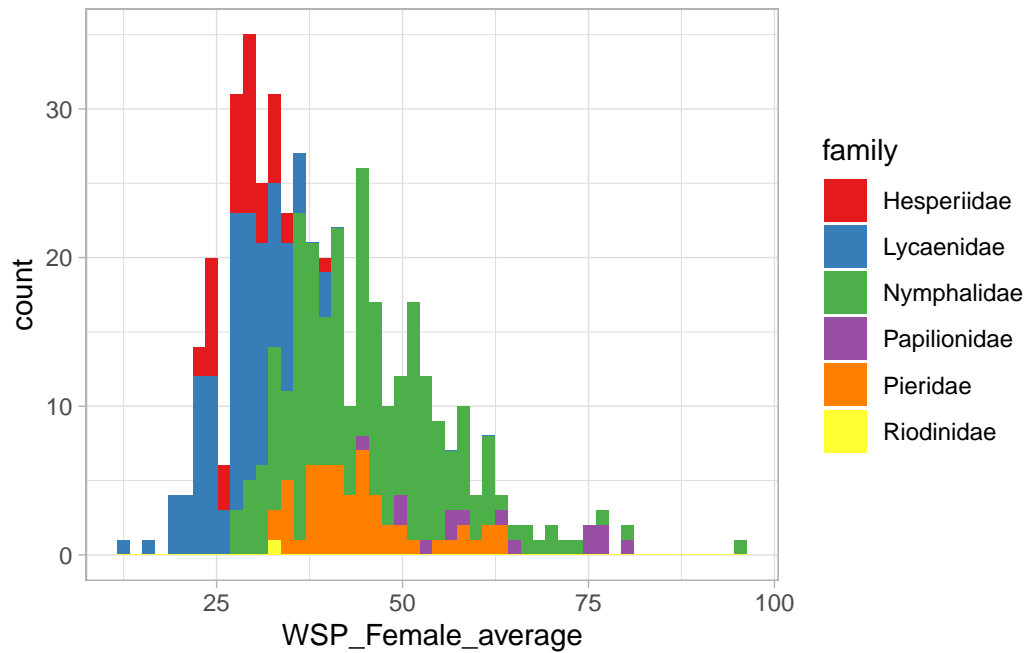


The above plot would need a little ‘cleaning up’. How would you do that?

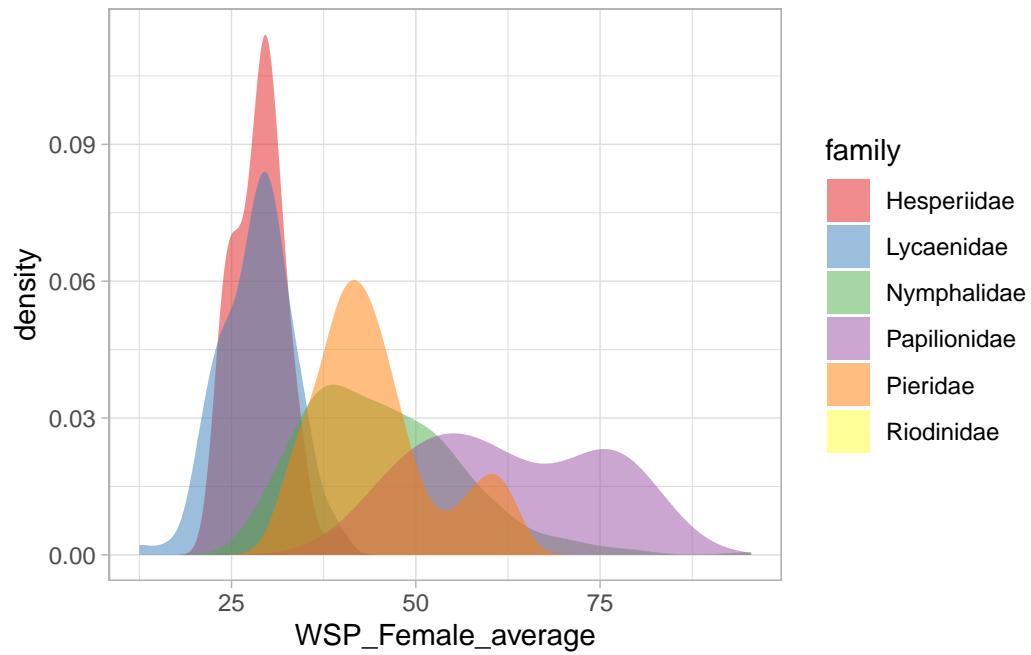
5.4 Some common plot types

We have looked at scatterplots, now let’s look at a number of other commonly used plots.

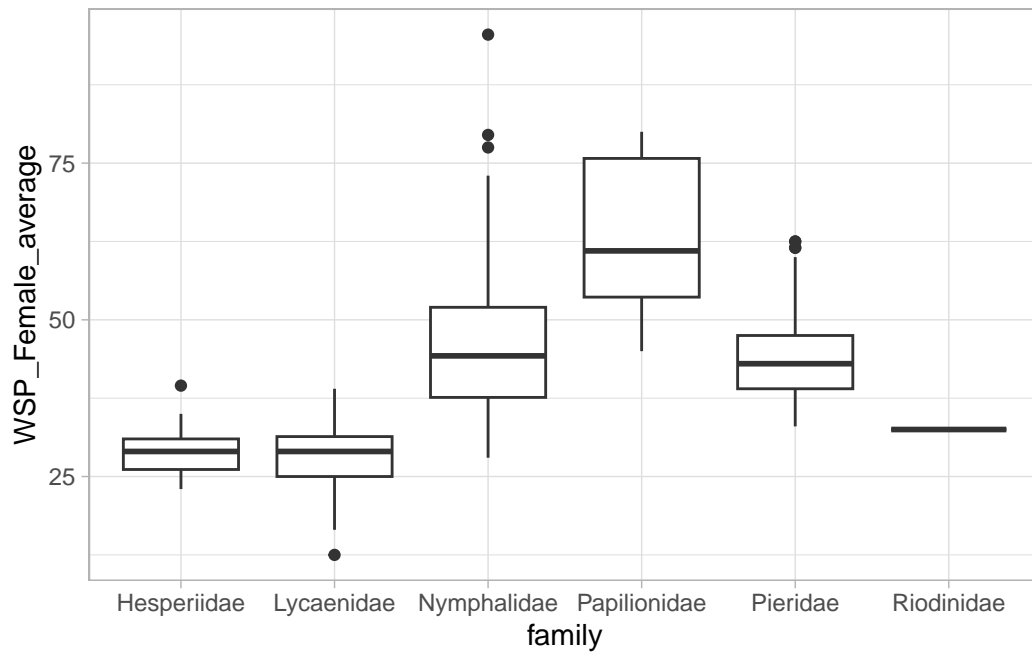
```
# histograms
be %>%
  ggplot(aes(x = WSP_Female_average, fill = family)) +
  geom_histogram(bins = 50) +
  theme_light() +
  scale_fill_brewer(palette = "Set1")
```



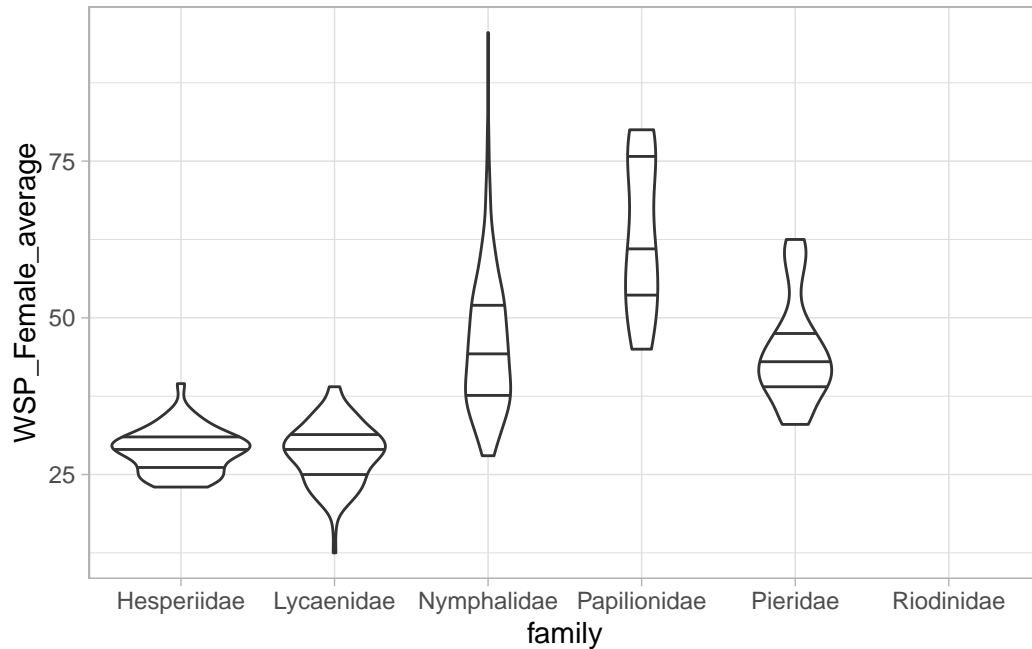
```
# better alternative are often density plots
be %>%
  ggplot(aes(x = WSP_Female_average, fill = family)) +
  geom_density(alpha = 0.5, colour = NA) +
  theme_light() +
  scale_fill_brewer(palette = "Set1")
```



```
# boxplots  
be %>%  
  ggplot(aes(y = WSP_Female_average, x = family)) +  
  geom_boxplot() +  
  theme_light()
```



```
# better alternative are violin plots
be %>%
  ggplot(aes(y = WSP_Female_average, x = family)) +
  geom_violin(draw_quantiles = c(0.25, 0.5, 0.75)) +
  theme_light()
```

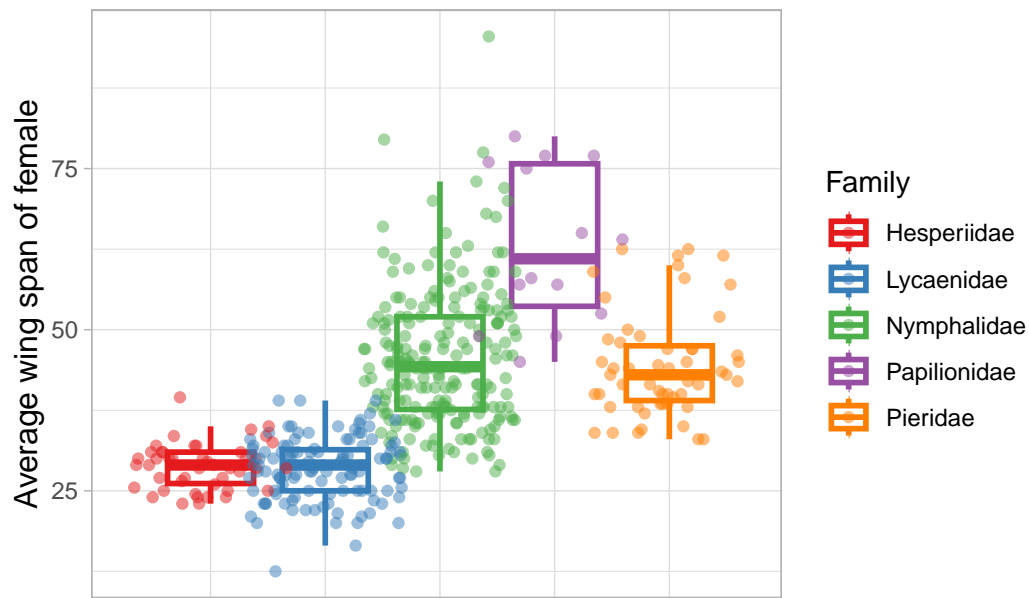



5.5 Fine tuning plots

We now know how to plot some common chart types but most of these don't look publishable yet. Lets return to one of our first examples and see how to polish it a little.

```
# A publication ready plot
be %>%
  filter(family != "Riodinidae") %>%
  ggplot(aes(y = WSP_Female_average, x = family, color = family)) +
  geom_boxplot(lwd = 1, outlier.shape = NA) +
  geom_point(position = position_jitterdodge(jitter.width = 2), alpha = 0.5) +
  theme_light() +
  labs(title = "Wing span across European butterfly families",
       y = "Average wing span of female")+
  theme(plot.title = element_text(face = "bold"),
        axis.title.y = element_text(size = 12),
        axis.title.x = element_blank(),
        axis.text.x = element_blank(),
        strip.text = element_text(size = 11)) +
  scale_color_brewer(palette = "Set1", name = "Family")
```

Wing span across European butterfly families



5.6 Exercise

Using `ggplot2`, explore how range size differs between butterfly families and plot check if butterflies with smaller ranges have higher protection status. Try to find appropriate plot types for this, use the help pages to find plot types that were not introduced to you yet. Explore other variables that may explain some trends in the data. Find a theme that you like! Remember, start with the **data**, add **aesthetics (what do you want to plot)**, and then think about **geometric objects (how do you want to plot the data)**. How can colour help in your visualisations? Does faceting make sense?

Part II

UNIX

6 Introduction to Linux Environment and Command Line.

6.1 The Unix / Linux environment

Unix and Linux (a variant of Unix) are operating systems (like Windows or macOS). They belong to a “family” of operating systems that share a common ancestor, have been around since 1969 and it’s not likely to disappear any time soon.

- Commonly used among the scientific and technical community (e.g. servers and scientific clusters).
- macOS is Unix-based system.
- Most supercomputers are powered by Unix-like operating systems.

6.2 Why learn Unix command line?

- Is the foundation of scientific computing (e.g. bioinformatics and data analysis)
- Powerful for working on large datasets and files
- Helps automate repetitive tasks (e.g. imagine you need to need to rename or modify 1,000 files?)
- Enables use of higher-powered computers elsewhere (clusters and servers/cloud-computing)

6.3 Some terminology

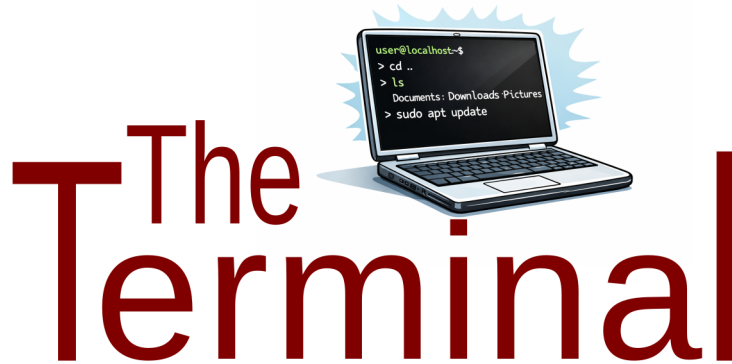


Figure 6.1: image1.png

As a user you can “communicate” with your Linux system either by a **Graphical User Interface (GUI)** or by typing instructions (commands) using a **Command Line Interface (CLI)**. At first it might look quite complex and confusing but once you understand the concept and the basics then its quite simple and intuitive!

Command Line: is the written instructions we type.

Terminal: also known as terminal emulator is the text-based environment (software) capable of taking input and providing output.

Shell: a program that interprets command-line input and executes commands. There are different shells available e.g bash, zsh, sh, csh etc. Each of them offering unique features and functionalities. Some are more basic and some are more fancy but all serve the same purpose, to interpret the commands provided by the user and output the results. The most commonly used is the **bash** shell.

6.4 Some important rules

1. **Be aware of the case!** The command line is case sensitive so be careful when typing. For example, typing **Echo** is not the same as **echo**, nor are the directory names “**/Results**” and “**/results**”.
2. **Spaces are having special use!** The command line uses spaces as separators between arguments. Using spaces in filenames or directory names will certainly cause problems sooner or later. Avoid using names that contain spaces, but rather it’s better to use

dashes (-) or underscores (_). e.g., “results_2026.txt” is preferred over “results 2026.txt”.

3. Apart from spaces there are several other characters (special characters) that can be used to perform special operations. See some examples below.

Character	Description
/	Directory separator, used to separate a string of directory names. Example: /usr/src/linux
\	Escape — (backslash) prevents the next character from being interpreted as a special character. This works outside of quoting, inside double quotes, and generally ignored in single quotes.
.	Current directory. Can also “hide” files when it is the first character in a filename.
..	Parent directory
~	The tilde is a representation of the current user’s home directory.
*	Represents 0 or more characters in a filename, or by itself, all files in a directory.
?	Represents a single character in a filename.
\$	Expansion — introduces various types of expansion: parameter expansion (e.g. \$var or \${var}), command substitution (e.g. \$(command)), or arithmetic expansion (e.g. \$((expression))). More on expansions later.
[]	Can be used to represent a range of values, e.g. [0-9], [A-Z], etc. Example: hello[0-2].txt represents the names hello0.txt, hello1.txt, and hello2.txt
	Pipe — send the output from one command to the input of another command. This is a method of chaining commands together. Example: echo “Hello beautiful.” grep -o beautiful.
>	Redirect output of a command into a new file. If the file already exists, over-write it. Example: ls > myfiles.txt
»	Redirect and appends the output of a command onto the end of an existing file.
<	Redirect a file as input to a program.

Character	Description
;	Command separator. Allows you to execute multiple commands on a single line. Example: cd /var/log ; ls -l
&&	Command separator as above, but only runs the second command if the first one finished without errors.
&	Background – when used at the end of a command, run the command in the background (do not wait for it to complete).
#	Comment — the # character begins a commentary that extends to the end of the line. Comments are notes of explanation and are not processed by the shell.

6.5 Accessing your Terminal

You are using a **Kubuntu** Linux system. In Kubuntu, the terminal emulator is called **Konsole**. You can start it in any of the following ways:

1. Press **Ctrl + Alt + T** to open **Konsole** instantly.
2. Open the **Krunner** by pressing **Alt + Space**, type **Konsole**, then press **Enter**.
3. Open the **Application Launcher** → **System** → **Konsole**.

Now you are ready to start typing your first commands!

7 Part 1. Your first Unix commands - Navigating the Unix File-System Structure

What will be cover:

1. The Unix file-system structure and how to navigate
2. Some general rules
3. Relative vs absolute path
4. Running some basic commands and general syntax

Unix systems, like most operating systems, store file locations in a hierarchical structure. In the UNIX file-system each file and directory has its own “address”, and that address is called a “**path**”.

There are two special locations in all Unix-based systems That you should be familiar. The “**root**” location is where the address system of the computer starts. The “**home**” location is where the current user’s location starts.

By default every time you open a new terminal you start in your own “**home**” directory(containing files and directories that only you can modify). The path of home directory is usually represented by the “~” character.

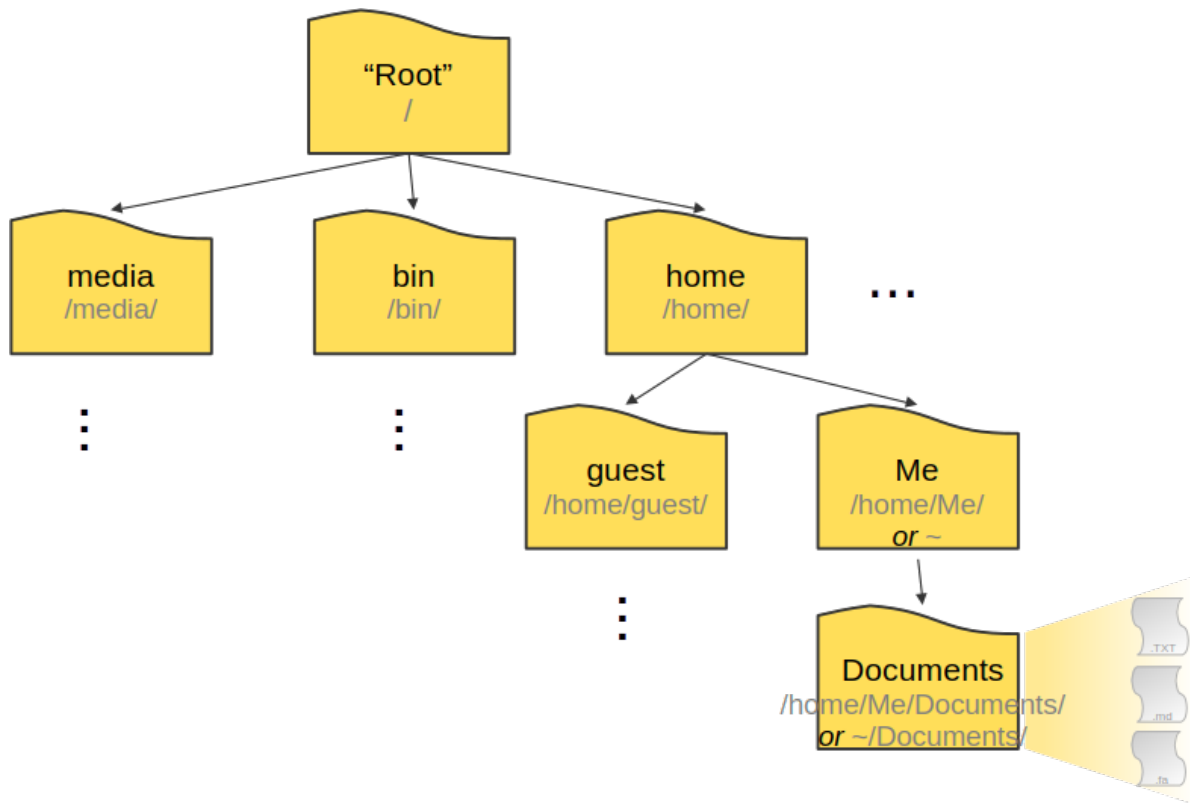


Figure 7.1: image2.png

Basic commands we will use

Command	Description
<code>pwd</code>	print working directory
<code>ls</code>	list items in directory
<code>cd</code>	change directory
<code>cd ..</code>	go one level up
<code>ls -l</code>	long format listing
<code>ls -lh</code>	human readable sizes

7.1 Finding out where you are

The `pwd` command in Linux is short for *print working directory*. It's only function is to print the absolute path of the current directory. It's handy when you're not exactly sure what directory you're in. So make it a good habit to get used to running the `pwd` command a lot.

```
# Example. Try running pwd. What do you see?
pwd
# What do you see if you run PWD instead?
PWD
# Try now to run this "echo $PWD". The command echo just prints the parameters we give it. Try
echo $PWD
```

7.2 The command `ls`

`ls` is short for *list*, and is used to list the files and sub-directories in your present working directory or some other directory if you specify one.

```
# Examples
# List the files and directories in your current directory
ls
# or
ls .
# ls can accept several options. Try running the following commands and observe how their output
ls -l
# or
ls -lh
# You can use ls to list the contents of any directory. Try the following.
ls -l /etc
```

Note

I. The anatomy of a command (or command syntax)

Each command is usually composed of three parts:

1. **The command** itself
2. **The options:** These are optional parameters that can be used to customise the behavior of a command. (e.g. on the previous examples `ls -l` shows the list of files in a long format)
3. **The arguments:** specify the target of the command. (e.g. on the previous example `ls -l /etc` you instructed the command to list the contents of the `/etc` directory)

II. Getting help

Most Unix commands can accept several parameters. How do we know which ones to use and why? Luckily, most Unix commands have built-in help documentation that we can access by providing `-help` as the only argument.

Try for example: `ls --help`

Another way to access the documentation for a command is by using the `man` command and providing the command's name as an argument. For example:

```
man ls
```

7.3 Relative vs absolute path and getting help

There are two ways to specify the path (the file's address on the computer):

- An **absolute path** starts from a fixed location, either the root directory (/) or the home directory (~). Note: A “full path” usually refers to an absolute path that starts from the root (/).
- A **relative path** starts from your current directory.

When working at the command line, it's important to always be aware of your current location in the system. One of the most common mistakes is trying to operate on a file that isn't where you think it is. To avoid this, it's good practice to use absolute paths, which clearly specify a file's exact location regardless of where you are.

7.4 Moving around

One of the most commonly used commands in Linux is the ***change directory*** command, or `cd`. It allows you to change your working directory from the current location to another directory you want to navigate to. The `cd` command takes a positional argument: the path (address) of the directory you want to move into. This path can be either **absolute** or **relative**. Let's try moving from our current directory to a directory present in your home directory called *Documents*.

```
# The relative way
cd Documents
# The absolute way (~ stands for /home/<user>/). Can you see the change in your command prompt?
cd ~/Documents
# But how do we go back "up" to the parent directory? We can use the ".." special characters
cd ..
# When you need to navigate back to the previous working directory from the current working directory
cd -
# Note: running the cd command without any arguments will always bring you back to your home directory
cd
```

7.5 Exercise

Practice moving around the filesystem with `cd` and listing directory contents with `ls`, including navigating by relative and absolute paths or using special characters `..` . Use `pwd` frequently to see your current working directory. Practice navigating home with `cd`.

8 Part 2. Working with Files and Directories

What we will cover:

1. Working with simple text files
2. Working with directories
3. Intro to a command-line text editor

8.1 Creating new directories and files

You can create a new directory using the **mkdir** command. It takes as a parameter a relative or absolute path to the directory to create

```
# Let's try some examples. From your home directory navigate to Documents/ and create 2 new directories
pwd
cd Documents
mkdir projects
cd projects
mkdir projects1
pwd
# Try to create a new directory with the same name e.g. projects1. What do you see?
```

Exercise

In the previous example we created the two directories in two separate steps. Can we do it in one step instead? (check the help page for **mkdir** command).

Now that we saw how to create a new directory let's see how we can create a new file. There are several ways to do this in Linux command line but we will start with the basic one. The command **touch** will create a new, empty file.

```
# Create an Empty File within the new directory project1.
touch notes.txt
# can you do it if you are outside the project1 directory?
```

8.2 Move or Rename a File or Directory

The **mv** command serves for both moving and renaming files and directories. Works by specifying a “source_path” and a “destination_path”, where “source_path” is the path (absolute or relative) of the file/directory to move or rename, and “destination_path” is the new name or location to give it.

Copying files and directories is similar operation, except that the original file or directory is not removed! We will use the **cp** command for this.

NOTE: copying, moving or renaming files at the command line will overwrite files if they have the same name!!

Let’s try them both.

```
# move the notes.txt file one directory up assuming your current directory is project1
mv notes.txt ../notes.txt
# rename the notes.txt to README.txt
mv ../notes.txt ../README.txt
# copy the README.txt file to project1 directory as README1.txt
cp ../README.txt ../README1.txt
```

Exercise

Try to make a copy of the complete project1 directory on the same parent location with the name project2. What do you see?

Note

Wildcards: friend and foe at the command prompt.

As we saw earlier some characters have special use in Unix command line. For example:

- the asterisk character “*” represents 0 or more characters in a filename, or by itself, all files in a directory.
- the square brackets [...] can be used define a range of values e.g. [0-9], [A-Z], etc.
- the question mark “?” can represent any single character.

These characters can be used as “Wildcards” to perform operations on multiple files at the same time. Lets see some examples.

```
# Create a directory with the name wildcard_practice and, inside it, create some empty files
mkdir wildcard_practice
cd wildcard_practice
touch file1.txt file2.txt file10.txt
touch image1.png image2.png imageA.png
touch data_2022.csv data_2023.csv data_2024.csv
# List all the .txt files
ls *.txt
# create a directory called data and move the .csv files into data/
mkdir data
mv *.csv data/
```

Exercise List only the .csv files with years 2022 or 2023

8.3 Delete (remove) files and directories

WARNING!!

Using the `rm` command permanently deletes files and directories without moving them to a trash or recycle bin. This action cannot be undone! Whenever you use the `rm` command, ALWAYS double-check your syntax.

```
# Lets remove some unwanted files and directories
rm ~/Documents/projects/README.txt
# the command rmdir can remove an empty directory, so it is safer!
mkdir test_dir
rmdir test_dir
# The dangerous way. Use rm recursively "-r" to remove a non-empty directory. BE CAUTIOUS! YOU CAN LOSE DATA!
rm -r ~/Documents/projects/project1
# check the help page for rm command. Is there a safer way to do this?
```

Exercise

1. Create a directory `data`.
2. Inside `data`, create three empty files: `sample1.txt`, `sample2.txt`, `sample3.txt`
3. Copy `sample1.txt` into a new directory called `backup`.
4. Rename `sample2.txt` to `s2.txt`.
5. Remove `sample3.txt`.

8.4 Viewing and inspecting Files

Sometimes we want to quickly look at files, either to inspect their structure or to get some basic information about their contents. These are some commands we can use to do so.

Command	Description
cat	print entire file
less	scroll through file
head	show first n lines
tail	show last n lines
wc	count (w ord c ount)

```
# Lets use the file "butterfly_ecology.csv" you previously used with R.
# You can either navigate to the folder you saved the file copy it to a new directory.
# You can use the "cat" command for viewing the contents of a file
cat butterfly_ecology.csv
# Another command for viewing the contents of a file is "less". This command allows to scroll
less butterfly_ecology.csv
# However, most of the time we only need to see part of a file rather than the entire file.
head butterfly_ecology.csv
tail butterfly_ecology.csv
# or you can specify the number of lines to print (by default will print 10)
head -n 2 butterfly_ecology.csv
tail -n 3 butterfly_ecology.csv
# The command wc (word count) is useful for counting how many lines, words, and characters there are
wc butterfly_ecology.csv
# Can you find an option for wc that will let us get only the number of lines of the file?
```

Note

The **cat** command can also be used to create new files or to concatenate existing files. Lets try:

```
cat > file1
Stefanos
# Exit by pressing Ctrl + D
cat > file2
some random text
# Exit by pressing Ctrl + D
cat file1 file2 > file3
```


8.5 A text editor for the terminal

A simple text editor available on most systems is nano. To run it, simply specify a file name to edit. If the file doesn't exist already, it will be created after saving.

```
nano NOTES.txt
```

Part III

Phylogenetics

Part IV

Microbiome

References