# Making figures for Fisk- och skaldjursöversikt i hav och sötvatten - resursöversikt in R

# Max Lindmark 28 april 2019

## Contents

1	Introduction 1		
	1.1	Why R?	
	1.2	Basic prerequisitis	
	1.3	What do these scripts do?	
	1.4	How do I use this for my species?	
<b>2</b>	Pre	paration	
	2.1	Load libraries	
	2.2	Load example data and clean it up!	
	2.3	Define ggplot theme	
3	Let'	et's start making figures!	
	3.1	Fig. 1. Single series	
		Fig. 2. Multiple series	
		Fig. 3. Multiple series, two y-axes	
	3.4	Fig. 4. Landings by country	

# 1 Introduction

First, this is a first draft and a work in progress so any feedback on this document or the figures are highly appreciated! This document has been put together by Max Lindmark (K-lab), with code and input from Massimiliano Cardinale massimiliano.cardinale@slu.se (H-lab) and Martin Ogonowski martin.ogonowski@slu.se (Sö-lab). We are also the R-contact-persons for RoM work, so please send any questions to your corresponding lab-contact. And if it's a general R-question, please do also email the R-mailing list Aqua r-users aqua-r-users@slu.se.

These code snippets are aimed to be as generic as possible. However, because almost all non-assessed species have different data, there is no guarantee that the code will work straight from pdf with your species. Do give it a try if you feel adventurous, but if you get stuck don't hesitate contacting us!

# 1.1 Why R?

The main argument for using code to create RoM figures is to standardize them across species and to do so while limiting repetitive work (you only need to write a script once!).

#### 1.2 Basic prerequisitis

It is an advantage if you know some basic R. I strongly recommend using R-studio and working in a so called R-studio project (but it is not needed to reproduce this code). To create a project, open R-studio, click

File/New Project/New Directory and specify where you want to save it. Open the "your\_project\_name". Rproj and click File/New Script and save that in your project folder. And that's it! The best thing is that now all your search paths are relative and not absolute. If you want to read in data, put the data inside the project folder and you don't have to specify the full search path ("C:/R/RoM/data-file.csv" or whatever), it's enough you give the name of the file only (more on that below!). Another benefit with a relative search path is that I don't have to worry about setting the working directory and changing the directory each time I try to rerun anyone elses script.

Lastly, this is an R Markdown document. This means R-code is text with grey background. You can copy these chunks of code to a new R-script in your R-studio project and run it from there.

# 1.3 What do these scripts do?

These scripts give an example of how you can use R to create standardized figures for RoM, and save them in a predefined, ready to be used without further editing.

I have chosen data for freshwater pike (*Esox lucious*) and Baltic turbot (*Scophthalmus maximus*) to illustrate how you can use this code, because it has all potential data for a RoM species (recreational, multiple areas, error bars, landings-by-country etc.). These data have been uploaded on github.

# 1.4 How do I use this for my species?

Most basic RoM figures that all species have in some form can be grouped into different categories. Each category will have its own script. Once you identify the best category for your species you should not have to modify the actual plotting code that much, as a minimum you only have to:

(1) Pre-define your variables (set y-axis names), (2) Prepare your data in the right format (see examples below and do the same for your species), (3) If a multi-series plot: put hash tags in front of stuff you don't need! E.g. if you don't have points to plot, make sure to put a hash tag before the geom\_point()-function, which plots points. But do not worry! Instructions for how to do that will be in the described below!

The categories are:

- Fig. 1. Single series
- Fig. 2. Multiple series
- Fig. 3. Multiple series and y-axes
- Fig. 4. Landings-by-country

Once you know which plot-type you want, make sure you follow how the data should be structured in your csv-file, and load the theme (see "Define gaplot theme").

# 2 Preparation

#### 2.1 Load libraries

If you haven't done it already, you can now open a new script (File->New Script) and copy paste code chunks into that. The first thing to do is to install a few packages (this might take a moment!):

```
"RCurl",
    "tidyr",
    "dplyr",
    "scales",
    "png",
    "knitr")

# Install packages if not installed
if (length(setdiff(pkgs, rownames(installed.packages()))) > 0) {
    install.packages(setdiff(pkgs, rownames(installed.packages())))
}

# Load all packages
invisible(lapply(pkgs, function(x) require(x, character.only = T, quietly = T)))
```

# 2.2 Load example data and clean it up!

Now let's read in the example data (freshwater pike):

```
# Go to https://github.com/maxlindmark/ROM to view the data in the browser
dat <- read.csv(
  text = getURL("https://raw.githubusercontent.com/maxlindmark/ROM/master/pike.csv"),
  sep = ";")</pre>
```

For your own species, the code would typically look like this if you put the data inside your R Project folder:

```
#dat <- read.csv("pike.csv", sep = ";")
```

Inspect the data. It is important that you use the same column names as I do here. "Year" is Year, "error\_plus" and "error\_minus" are min and max of the confidence interval of the response variable - if you have any, otherwise just type NA. "Group" is our grouping variable, typically different areas and response\_variable could be biomass, cpue or abundance. You will specify the unit before plotting, so that everyone can use the same base code, irrespective of which data the species have.

```
head(dat)
```

```
##
                                             Group Response_variable
     Year error_plus error_minus
## 1 1997
                  NA
                              NA Stora sj<f6>arna
## 2 1998
                  NA
                              NA Stora sj<f6>arna
                                                                  114
## 3 1999
                              NA Stora sj<f6>arna
                  NA
                                                                  149
## 4 2000
                              NA Stora sj<f6>arna
                                                                  145
                  NA
## 5 2001
                  NA
                              NA Stora sj<f6>arna
                                                                  121
## 6 2002
                  NA
                              NA Stora sj<f6>arna
                                                                  145
# list(dat) # This function shows all the data,
# for you to see the levels in the Group column
```

Importing data in the right format is often non-trivial. Here I show you how you can rename the levels of the column Group (i.e. the lake-names in this case). We do this because the beauty of *ggplot* is that all important information in the plot will be inherited from the data.

```
"Vättern")
head(dat)
##
     Year error_plus error_minus
                                           Group Response_variable
## 1 1997
                   NA
                               NA Stora sjöarna
                                                                 115
## 2 1998
                               NA Stora sjöarna
                   NΑ
                                                                 114
## 3 1999
                   NA
                               NA Stora sjöarna
                                                                 149
## 4 2000
                   NA
                               NA Stora sjöarna
                                                                 145
## 5 2001
                   NA
                               NA Stora sjöarna
                                                                 121
## 6 2002
                               NA Stora sjöarna
                                                                 145
                   NΑ
tail(dat)
##
       Year error_plus error_minus
                                            Group Response_variable
## 121 2012
                     NA
                                 NA Fritidsfiske
## 122 2013
                     NA
                                 NA Fritidsfiske
                                                                   NA
## 123 2014
                    223
                                  77 Fritidsfiske
                                                                  150
## 124 2015
                                  82 Fritidsfiske
                                                                  133
                    184
```

There is one last thing you might need to do with the data before proceeding with plotting, and that is to change the order of the levels in the data. When plotting, ggplot sets the levels of the data in alphabetical order, but here we want a specific order: the first "level" should correspond to the total and the last to any special level, such as the recreational data. This is how you would do that, just replace the levels with your own names.

NA

NA

NA Fritidsfiske

NA Fritidsfiske

Now the data look much better! You might not need to do these modification on your own data. But you need to make sure it is structured in the same way, that is: 1 row = 1 observation (not multiple columns for different areas)

# 2.3 Define *ggplot* theme

## 125 2016

## 126 2017

NA

NA

Now that you have loaded and cleaned up your or the example data, we can move on with general plotting settings. First the color palette:

```
pal <- c("#56B4E9", "#009E73", "#F0E442", "#0072B2", "#E69F00", "#D55E00")
```

Second, we define the theme we will use for all plots. This is made to match as closely as possible to the RoM style. This applies to all figures styles.

```
theme_rom <- function(base_size = 12, base_family = "") {
  theme_bw(base_size = 12, base_family = "") +
    theme(
        axis.text = element_text(size = 8),
        axis.title = element_text(size = 8),
        axis.ticks.length = unit(0.05, "cm"),
        axis.line = element_line(colour = "black",</pre>
```

```
size = 0.3),
      text = element_text(family = "sans"),
      panel.grid.major = element_blank(),
      panel.grid.minor = element_blank(),
      panel.border = element_blank(),
      plot.title = element_text(hjust = 0.5,
                                margin = margin(b = -3),
                                 size = 9.6,
                                face = "bold"),
      legend.position = "bottom",
      legend.text = element_text(size = 8),
      legend.background = element_rect(fill = "transparent"),
      legend.key = element rect(fill = "transparent"),
      legend.box.margin = margin(-25, -25, -25, -25),
      aspect.ratio = 1,
      plot.margin = unit(c(5.5, 5.5, 20, 5.5),
                         "points")
      )
}
# And also replace points with commas in the plot
options(OutDec= ",")
```

You can also source the theme-function directly from github. That way you'll always have the latest version!

```
u<-c("https://raw.githubusercontent.com/maxlindmark/scaling/master/R/raincloud_plot.R")
script <- getURL(u, ssl.verifypeer = FALSE)
eval(parse(text = script))</pre>
```

# 3 Let's start making figures!

# 3.1 Fig. 1. Single series

For illustration purposes, we will now pretend our example pike data set only is a single series, by filtering it to only contain the area "Stora Sjöarna". If you have a "Fig.1 situation"", your data might look something like this, with one column for year, one for area and one for tonnes.

```
dat1 <- dat %>%
   select(Year, Response_variable, Group) %>%
   filter(Group == "Stora sjöarna")
head(dat1)
```

```
## Year Response_variable Group
## 1 1997 115 Stora sjöarna
## 2 1998 114 Stora sjöarna
## 3 1999 149 Stora sjöarna
## 4 2000 145 Stora sjöarna
## 5 2001 121 Stora sjöarna
## 6 2002 145 Stora sjöarna
```

We now need to specify the **y-axis title**. For the pike data we can set them as:

```
y_axis <- c("Landningar (ton)")</pre>
```

Now go ahead and create the plot:

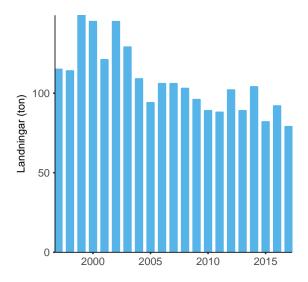


Figure 1: Gädda in the "Great Lakes" - example of a single-series plot using R

Save the file (to your working directory)

```
ggsave("Fig_1.tiff", plot = p1, dpi = 300, width = 8, height = 8, units = "cm")
```

## 3.2 Fig. 2. Multiple series

We now need to specify y-axis title, and in addition we need to define how many levels we have! For the pike data we can set them as:

Note that we in this example use multiple data series, and that one of them is recreational fisheries that in turn has error bars. These need to be in columns, so that we have a high (error\_plus) and

a low (error\_minuss) column. Note also that they are NA when the Group is not equal to recreational fisheries

```
head(dat)
                                           Group Response_variable
##
     Year error_plus error_minus
## 1 1997
                  NA
                               NA Stora sjöarna
## 2 1998
                  NA
                               NA Stora sjöarna
                                                                114
## 3 1999
                  NA
                               NA Stora sjöarna
                                                                149
                               NA Stora sjöarna
## 4 2000
                  NA
                                                                145
## 5 2001
                               NA Stora sjöarna
                                                                121
                  NΑ
## 6 2002
                  NA
                               NA Stora sjöarna
                                                                145
tail(dat)
```

```
##
       Year error_plus error_minus
                                           Group Response variable
## 121 2012
                    NA
                                 NA Fritidsfiske
## 122 2013
                    NA
                                 NA Fritidsfiske
                                                                  NA
## 123 2014
                    223
                                 77 Fritidsfiske
                                                                 150
## 124 2015
                    184
                                 82 Fritidsfiske
                                                                 133
## 125 2016
                    NA
                                 NA Fritidsfiske
                                                                  NA
## 126 2017
                    NA
                                 NA Fritidsfiske
                                                                  NA
```

And with that set, we can make the second figure, with multiple levels. The key to have a single legend is to have a layer that has all Groups. In this case it's the line. But we don't want all series to have lines (not the point or bar series). We can overcome this by setting those series to have full transparancy. We control that in the scale\_alpha\_manual()-function. 0 means transparant and 1 is normal. The override.aes part in guides() makes sure this transparancy is not inherited in the legend!

```
p2 <- ggplot(dat, aes(Year, Response_variable, color = Group)) +</pre>
  geom_bar(data = subset(dat, Group == main_series),
           aes(x = Year, y = Response_variable),
           stat = "identity", color = pal[1], fill = pal[1],
           width = 0.6) +
  geom_line(data = dat, aes(Year, Response_variable, color = Group, alpha = Group),
            size = 1) +
  geom_point(data = subset(dat, Group == special_series), # here we set our special series
             aes(Year, Response_variable, fill = Group),
             size = 2, color = pal[max(n lev)]) +
  # above the number of level enters (max(n_lev))
  geom_errorbar(data = dat, aes(x = Year, ymin = error_minus, ymax = error_plus,
                                color = Group),
                show.legend = FALSE, width = 1) +
  scale_alpha_manual(values = c(0, rep(1, (n_lev-2)), 0)) +
  scale_color_manual(values = pal[seq(1, n_lev)]) +
  # above we set the line between rec fisheries transparent
  labs(x = "", y = y_axis) +
  guides(fill = FALSE,
         alpha = FALSE,
         color = guide_legend(nrow = 3,
                              title = "",
                              override.aes = list(size = 1.3,
                                                   color = pal[seq(1, n_lev)]),
                              keywidth = 0.3.
                              keyheight = 0.1,
                              default.unit = "inch")) +
```

```
scale_x_continuous(expand = c(0, 0), breaks = scales::pretty_breaks(n = 6)) +
scale_y_continuous(expand = c(0, 0), breaks = scales::pretty_breaks(n = 5)) +
theme_rom()
```

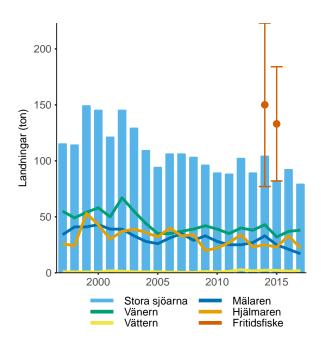


Figure 2: Gädda in the Great Lakes - example of a multiple-series plot using R

Now save the file (to your working directory)

```
ggsave("Fig_2.tiff", plot = p2, dpi = 300, width = 8, height = 8, units = "cm")
```

Let's say that you have 3 different areas and no special series and no error bars. Then you simply need to hash tag the code the plots those features. We subset the full data to select only the total and lake Mälaren as an example:

```
dat2 <- dat %>%
  select(Year, Response_variable, Group) %>%
  filter(Group %in% c("Stora sjöarna", "Mälaren"))
```

We now need to update the number of levels in the data:

```
n_lev <- length(unique(dat2$Group))</pre>
```

Repeat the general Fig. 2 plot, but remove the points and error bars using hashtags. (see also changes to scale alpha manual()!)

```
p2a <- ggplot(dat2, aes(Year, Response_variable, color = Group)) +
  geom_bar(data = subset(dat2, Group == main_series),</pre>
```

```
aes(x = Year, y = Response_variable),
          stat = "identity", color = pal[1], fill = pal[1],
          width = 0.6) +
 geom_line(data = dat2, aes(Year, Response_variable, color = Group, alpha = Group), # remove alpha = G
           size = 1) +
#geom_point(data = subset(dat, Group == special_series), # here we define our special series
            aes(Year, Response\_variable, fill = Group), size = 2, color = pal[max(n\_lev)]) +
\#geom\_errorbar(data = dat, aes(x = Year, ymin = error\_minus, ymax = error\_plus,
                               color = Group),
              show.legend = FALSE, width = 1) +
 scale_alpha_manual(values = c(0, 1)) +
 scale_color_manual(values = pal[seq(1, n_lev)]) +
 labs(x = "", y = y_axis) +
 guides(fill = FALSE,
        alpha = FALSE,
        color = guide_legend(nrow = 3,
                              title = "",
                              override.aes = list(size = 1.3,
                                                  color = pal[seq(1, n_lev)]),
                              keywidth = 0.3,
                              keyheight = 0.1,
                              default.unit = "inch")) +
 scale_x_continuous(expand = c(0, 0), breaks = scales::pretty_breaks(n = 6)) +
 scale_y_continuous(expand = c(0, 0), breaks = scales::pretty_breaks(n = 5)) +
 theme rom()
```

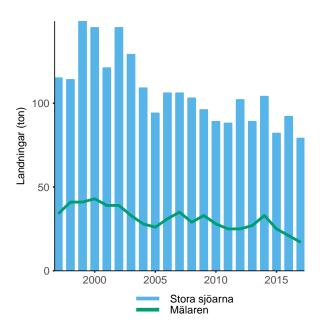


Figure 3: Gädda in the Great Lakes - example of a multiple-series plot using R

# 3.3 Fig. 3. Multiple series, two y-axes

In some cases, you may be forced to plot different series on two different axes, e.g. when you have things in different units (landings and and an index perhaps). Using the pike-data set, we will invent such a scenario treating the "Fritidsfiske" as if on a different unit.

```
## Increase the value of Fritidsfiske:
dat3 <- dat
dat3$Response_variable <- ifelse(dat3$Group == "Fritidsfiske",
                                   dat3$Response_variable * 100,
                                   dat3$Response_variable)
## And the errorbars...
dat3$error_minus <- dat3$error_minus * 100</pre>
dat3$error_plus <- dat3$error_plus * 100</pre>
# Define plot settings again
y_axis <- c("Landningar (ton)")</pre>
y_axis2 <- c("Fångst per ansträngning (antal/timme)") # 2nd axis title
main_series <- c("Stora sjöarna")</pre>
                                      # This is the sum of all other series
n_lev <- length(unique(dat3$Group)) # Number of unique lakes</pre>
special_series <- c("Fritidsfiske") # This series must be drawn with points, so we need to
                                      # separate it from the rest.
```

And with that set, we can make the second figure, with multiple levels. Hashtag geom\_point and geom\_errorbar as these are no longer on the same scale.

```
p3 <- ggplot(dat3, aes(Year, Response_variable, color = Group)) +
  geom_bar(data = subset(dat3, Group == main_series),
           aes(x = Year, y = Response_variable), stat = "identity",
           color = pal[1], fill = pal[1],
           width = 0.6) +
  geom_line(data = dat3, aes(Year, Response_variable, color = Group, alpha = Group),
            size = 1) +
#geom_point(data = subset(dat, Group == special_series), # here we set our special series
            aes(Year, Response\_variable, fill = Group), size = 2, color = pal[max(n\_lev)]) +
\#qeom\_errorbar(data = dat, aes(x = Year, ymin = error\_minus, ymax = error\_plus,
                               color = Group),
               show.legend = FALSE, width = 1) +
  scale_alpha_manual(values = c(0, rep(1, (n_lev-2)), 0)) +
  scale_color_manual(values = pal[seq(1, n_lev)]) +
  labs(x = "", y = y_axis) +
  guides(fill = FALSE,
         alpha = FALSE,
         color = guide_legend(nrow = 3,
                              title = "".
                              override.aes = list(size = 1.3,
                                                  color = pal[seq(1, n_lev)]),
                              keywidth = 0.3,
                              keyheight = 0.1,
                              default.unit = "inch")) +
  scale_x_continuous(expand = c(0, 0), breaks = scales::pretty_breaks(n = 6)) +
  scale_y_continuous(expand = c(0, 0),
                     limits = c(0, max_yax), breaks = scales::pretty_breaks(n = 5)) +
  theme_rom()
```

Now the trick is to first transform the variable that is to be plotted on the second y-axis to a similar scale, and then reset the scale to the original values::

## Scale for 'y' is already present. Adding another scale for 'y', which ## will replace the existing scale.

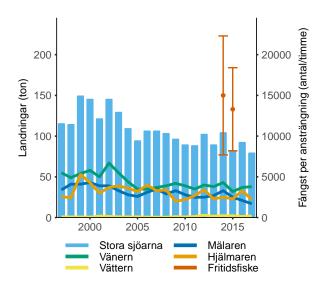


Figure 4: Gädda in the Great Lakes with secondary y-axes- example of a multiple-series plot using R

Now save the file (to your working directory)

```
ggsave("Fig_3.tiff", plot = p3f, dpi = 300, width = 8, height = 8, units = "cm")
tail(dat3)
```

This is not a very straightforward plot to make. First you need to manually find a value that scales the different units so they become close (see above when making p3f). When scaling back you need to verify all the data-points are within the plot range (which now is manual! By default it's based on your data but we need to bypass that). Also, in order to inheret the legend, we need at least one geom with all data we want in the legend. This is geom\_line in the above case. In order to not get a gigantic primary y-axis, we set it manually and simple plot the series that is on a different scale outside the plot window. So, many things to keep in mind when/if adapting this code to your data!

# 3.4 Fig. 4. Landings by country

For this example, we will use Baltic turbot data:

```
u<-c("https://raw.githubusercontent.com/maxlindmark/ROM/master/Landingsbycountry_Baltic_Turbot.csv")
tur <- read.csv(
  text = getURL(u),
  sep = ";")</pre>
```

```
head(tur)
     Year Danmark Estland Finland Lettland Litauen Polen Sverige Tyskland
## 1 1965
            0,001
                    0,001
                            0,001
                                      0,001
                                              0,001 0,001
                                                            0,001
## 2 1966
           37,000
                    0,001
                            0,001
                                      0,001
                                              0,001 0,001
                                                            0,001
                                                                         58
## 3 1967
           34,000
                    0,001
                            0,001
                                      0.001
                                              0,001 0,001
                                                            0.001
                                                                         17
                                                                         70
## 4 1968
           32,000
                    0,001
                            0,001
                                      0,001
                                              0,001 0,001
                                                            0,001
## 5 1969
           26,000
                    0,001
                            0,001
                                      0,001
                                              0,001 0,001
                                                            0,001
                                                                         61
## 6 1970 24,000
                    0,001
                            0,001
                                      0,001
                                              0,001 0,001
                                                            2,000
                                                                         45
     Ryssland Andra.1.e4.nder
                                 FishStock
                                                          Swedishname
## 1
        0.001
                        0,001 tur.27.22-32 Piggvar i <d6>stersj<f6>n
## 2
        0,001
                        0,001 tur.27.22-32 Piggvar i <d6>stersj<f6>n
## 3
        0,001
                        0,001 tur.27.22-32 Piggvar i <d6>stersj<f6>n
## 4
        0,001
                        0,001 tur.27.22-32 Piggvar i <d6>stersj<f6>n
## 5
        0,001
                        0,001 tur.27.22-32 Piggvar i <d6>stersj<f6>n
## 6
                        0,001 tur.27.22-32 Piggvar i <d6>stersj<f6>n
        0,001
# Rename column
tur <- tur %>% rename("Andra länder" = "Andra.l.e4.nder")
# Make data long (1 row, 1 observation)
tur_l <- tur %>% gather(land, landningar, 2:11, na.rm = TRUE)
# Show total tonnes by country
tur_1 %>%
  group_by(land) %>%
  summarize(tot_land = sum(landningar)) %>%
  arrange(desc(tot_land))
## # A tibble: 10 x 2
##
      land
                   tot_land
                      <dbl>
##
      <chr>
## 1 Danmark
                   7136.
## 2 Polen
                   5287.
## 3 Tyskland
                   3295
## 4 Sverige
                   2224.
## 5 Ryssland
                    649.
## 6 Litauen
                    466.
## 7 Lettland
                    365.
## 8 Andra länder
                      0.053
## 9 Estland
                      0.053
## 10 Finland
                      0.053
# We will only plot the top 5 countries and the total. Other countries will
# be regrouped to "Andra länder"
unique(tur_l$land)
    [1] "Danmark"
##
                       "Estland"
                                       "Finland"
                                                      "Lettland"
##
    [5] "Litauen"
                       "Polen"
                                                      "Tyskland"
                                       "Sverige"
   [9] "Ryssland"
                       "Andra länder"
tur_l$land <- ifelse(tur_l$land %in% c("Litauen",
                                        "Lettland",
                                        "Estland",
                                        "Finland"),
```

```
"Andra länder",
tur_l$land)

unique(tur_l$land)

## [1] "Danmark" "Andra länder" "Polen" "Sverige"
## [5] "Tyskland" "Ryssland"

We now need to specify the y-axis title. For the turbot landings data we can set them as:
```

```
y_axis <- c("Landningar (1000 ton)")</pre>
```

Now go ahead and create the plot:

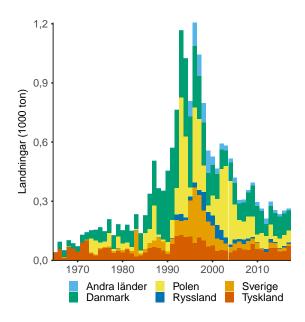


Figure 5: Landings of turbot in the Baltic Sea - example of a landings-by-country plot using R

Save the file (to your working directory)

ggsave("Fig\_4.tiff", plot = p4, dpi = 300, width = 8, height = 8, units = "cm")