

Index recruit Antarctic krill

Supplementary information HeatWaves Antarctic Paper

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```
rm(list = ls())
knitr::opts_chunk$set(echo = TRUE,
                      message = FALSE,
                      warning = FALSE,
                      fig.align = 'center',
                      dev = 'jpeg',
                      dpi = 300,
                      out.width='120%')
#XQuartz is a mess, put this in your onload to default to cairo instead
options(bitmapType = "cairo")
# (https://github.com/tidyverse/ggplot2/issues/2655)
# Lo mapas se hacen mas rapido
```

```
library(here)
#statistics
library(ggsignif)
library(lubridate)
library(easystats) # multiples unciones analiticas
library(readxl)
# vizualizacion
library(ggrepel)
library(ggpubr)
library(ggribes)
library(sf)
library(GGally)
library(tidyverse, quietly = TRUE)
library(knitr, quietly = TRUE)
library(kableExtra)
library(raster)
library(egg)
library(car) #Variance inflation Factor
library(ggthemes)
library(sjPlot)
library(CCAMLRGIS)
```

Contexto

Este análisis tiene como objetivo calcular un índice de reclutamiento (IR) del krill *Euphausia superba* a través de los datos monitoreados por la flota pesquera entre los años 1980 y 2020

Metodología

Los pasos para calcular el índice son los siguientes:

- Calcular el cuantíl 90% de la REF como proporción de individuos que ingresan a la pesquería.
- Con este valor del 90%, se calculan los individuos ajo esta talla provenientes de la pesquería. Luego se calcula la cantidad de inviduos bajo esa talla agrupados por distintas covariables.
- Se calcula la proporción respecto al total. Se identifica la distribución de los datos, luego se normalizan para que los datos tengan una distribución normal, y posteriormente, se estandarizan para llevarlos a un índice e tre -1 y 1, lo cual indica reclutamientos negativos y positivos respectivamente.
- Despliegue en diferentes plot y agrupaciones temporales y espaciales.

Load data

Data exploratory analysis

The object `ohbio2` come from data exploration analysis in data request CCAMLR data. This objetc have bio information from krill.

```
#carga objeto
meta <- get("METADATA")
c1 <- get("C1")
ohbio <- get("OBS_HAUL_BIOLOGY")
names(ohbio)

## [1] "c1_id" "obs_haul_id" "obs_logbook_id"
## [4] "haul_number" "taxon_code" "taxon_scientific_name"
## [7] "taxon_family" "maturity_stage" "sex_code"
## [10] "length_total_cm" "greenweight_kg"

dim(c1)

## [1] 360439 35

dim(ohbio)

## [1] 2107887 11

Join data set with master as c1 set. This join is trough obs_haul_id variable to get
geoposition variables

ohbio2 <- left_join(c1, ohbio, by="obs_haul_id")
dim(ohbio2)
```

```
## [1] 2443773      45
```

First glance. Test how many register have by year. In this case, `length_total_cm` by season `ccamlr`. Same exercise in date period `date_catchperiod_start` to separate dates.

```
ohbio3 <- ohbio2 %>%
  mutate(Year = year(date_catchperiod_start),
         Month = month(date_catchperiod_start),
         Day = day(date_catchperiod_start))
```

filter necessary data to further analysis

```
length481 <- ohbio3 %>%
  dplyr::select(7, 9, 11, 12, 14, 24, 25, 29, 42, 44, 46, 47, 43) %>%
  filter(asd_code=="481")
#save(length481, file = "length481.RData")
```

```
ohbio4 <- ohbio3 %>%
dplyr::select(7, 9, 11, 12, 14, 24, 25, 29, 42, 44, 46, 47)
names(ohbio4)
```

```
## [1] "vessel_nationality_code" "season_ccamlr"
## [3] "asd_code"                "trawl_technique"
## [5] "date_catchperiod_start" "latitude_set_end"
## [7] "longitude_set_end"      "gear_type"
## [9] "maturity_stage"         "length_total_cm"
## [11] "Year"                   "Month"
```

Maps works

First thing is get different raster layer to join krill data length according different porpoises.

```
# Cargo linea de costa
coast <- load_Coastline()
coast1 <- st_as_sf(coast)
coast2 = st_transform(coast1, "+proj=latlong +ellps=WGS84")

# con Statistical Areas con foco en 48.1
suba <- load_ASDs()
suba1 <- subset(suba[(3),])
suba1a <- st_as_sf(suba1)
suba1aa = st_transform(suba1a, "+proj=latlong +ellps=WGS84")

# Uso las agrupaciones de Strata
strata <- st_read("~/DOCAS/Mapas/Antarctic_SHPfiles/Strata.shp",
```

```

      quiet=T)
strata=st_transform(strata, "+proj=latlong +ellps=WGS84")
strata <- strata %>%
  dplyr::filter(ID != "Outer")

```

Strata maps

Show strata aggregation to join length data (Figure@ref(fig:maptest)).

```

# y testeo el mapa
ssmap <- ggplot(strata)+
  geom_sf(data = strata, aes(fill=strata$ID,
                             alpha=0.5))+
  geom_sf(data = subalaa,alpha=0.3,
          colour="red")+
  geom_sf(data = coast2, colour="black", fill=NA)+
  scale_fill_viridis_d(option = "F",
                       name="Stratum")+
  ylim(230000, 2220000)+
  xlim(-3095349 , -1858911)+
  coord_sf(crs = 6932)+
  scale_alpha(guide="none")+
  theme_bw()
ssmap

```

Grouping Length data into Strata

```

names(ohbio4)

## [1] "vessel_nationality_code" "season_ccamlr"
## [3] "asd_code"                "trawl_technique"
## [5] "date_catchperiod_start" "latitude_set_end"
## [7] "longitude_set_end"      "gear_type"
## [9] "maturity_stage"         "length_total_cm"
## [11] "Year"                   "Month"

ohbio5 <- ohbio4 %>%
  drop_na() %>%
  filter(asd_code==481) %>%
  dplyr::select(6, 7, 8, 9, 10, 11, 12)
ohbio6 <- st_as_sf(ohbio5 %>%
  drop_na(latitude_set_end),

```

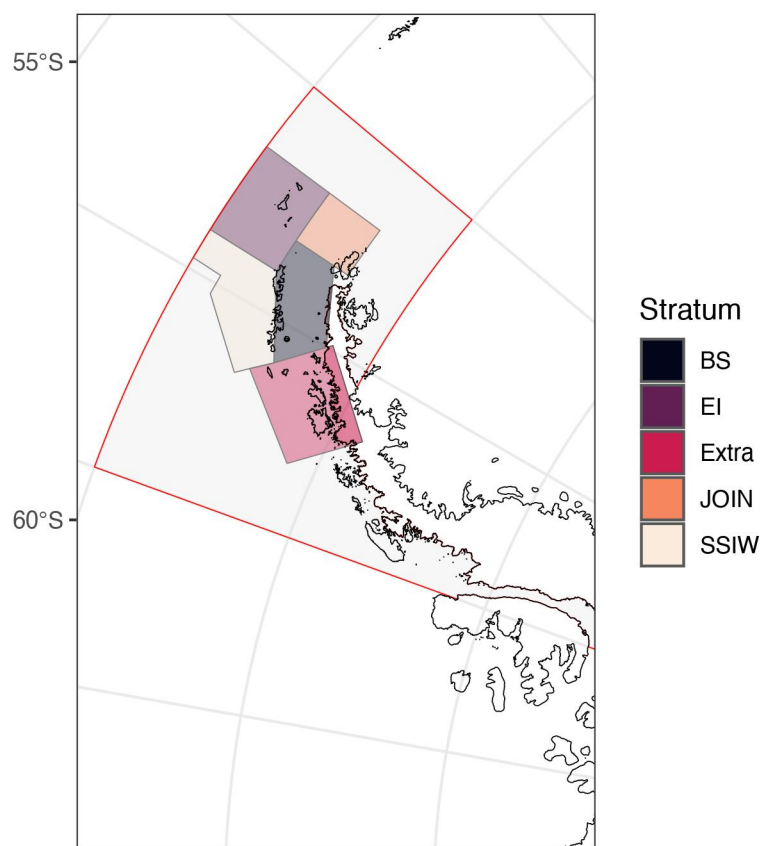


Figure 1: Strata Maps in 48.1

```

coords = c("longitude_set_end",
            "latitude_set_end"),
crs = "+proj=latlong +ellps=WGS84")

```

Comprobar si tengo datos duplicados

```

# comprobar si tengo datos duplicados
strata2 <- st_make_valid(strata)
ohbio7 <- st_make_valid(ohbio6)
krill.strata <- st_join(strata2, ohbio7)
#saveRDS(krill.strata, "KrillData.Rdata")
colSums(is.na(krill.strata))

```

```

##           ID           AreaKm2           Labx           Laby           TotArea
##           0              0             0             0             0
##   MarineArea      gear_type maturity_stage length_total_cm           Year
##           0              0             0             0             0
##           Month           geometry
##           0              0

```

```
dim(krill.strata)
```

```
## [1] 1050253      12
```

Exploración primaria

```

jz3 <- ggplot(krill.strata %>%
              drop_na(),
              aes(x=length_total_cm, y = as.factor(Month),
                  fill = factor(stat(quantile))))+
  stat_density_ridges(
    geom = "density_ridges_gradient",
    calc_ecdf = TRUE,
    quantiles = c(0.10, 0.90)) +
  scale_fill_manual(
    name = "Probability",
    values = c("#de2d26", "#fee0d2", "#de2d26"),
    labels = c("[0 - 0.10]",
               "[0.10 - 0.90]",
               "[0.90 - 1]"))+
  facet_grid(ID~Year) +
  geom_vline(xintercept = 3.6, color = "red")+
  scale_x_continuous(breaks = seq(from = 3,

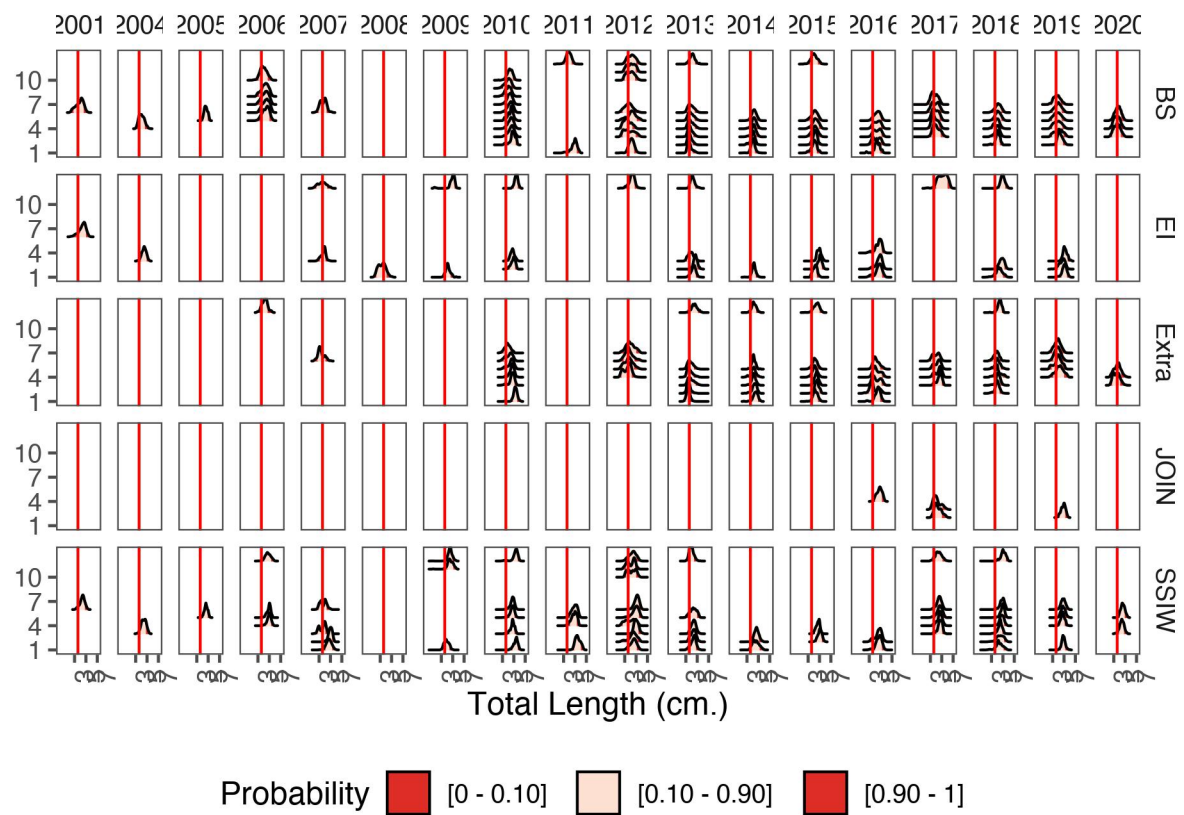
```



```

        to = 12,
        by = 2)))+
scale_y_discrete(breaks = seq(from = 1,
        to = 12,
        by = 3))+
#scale_fill_viridis_d(name="SubArea")+
theme_few()+
theme(axis.text.x = element_text(angle = 90, hjust = 1),
      legend.position = "bottom")+
#xlim(10,120)+
xlab("Total Length (cm.)")+
ylab("")
jz3

```



Ahora identifico los distintos cuantiles de los datos de pesquería y estaciones

```

# Calcular cuantiles por grupo
cuantiles_por_strata <- tapply(krill.strata$length_total_cm,
                              krill.strata$ID,

```

```
function(x) quantile(x,
                     c(0.10,
                       0.5,
                       0.90)))
```

Calculo el índice del reclutamiento de ([Maschette et al., 2020](#))

```
indice_reclutamiento <- krill.strata %>%
  filter(length_total_cm < 3.6 ) %>%
  group_by(Year, Month, ID) %>%
  summarize(PROP = n() / nrow(krill.strata)) %>%
  mutate(PROPLOG = log(PROP))
# Crear gráficos en facet_wrap de barras para representar el índice de reclutamiento
```

ahora estandarizo los datos entre -1 y 1.

```
a <- -1 # Límite inferior del rango objetivo
b <- 1  # Límite superior del rango objetivo

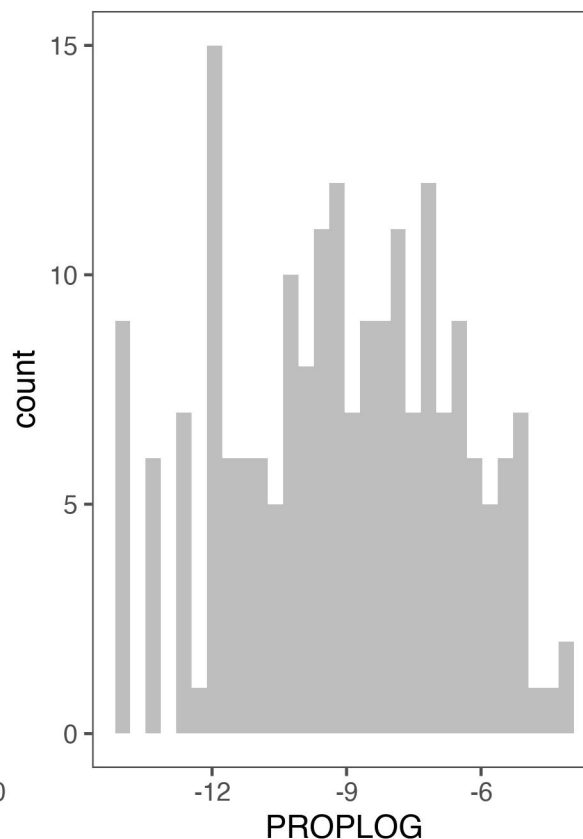
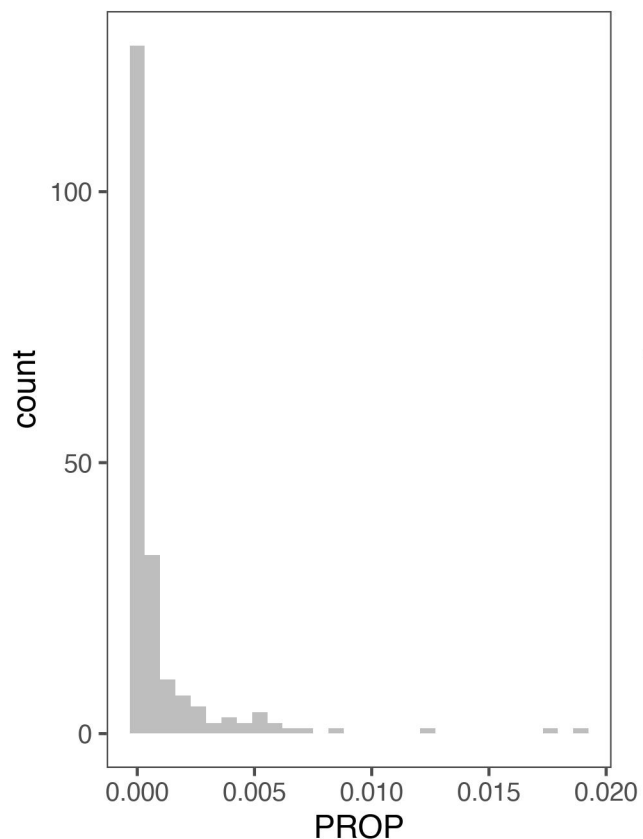
# Calcular el valor mínimo y máximo de tus datos
min_x <- min(indice_reclutamiento$PROPLOG)
max_x <- max(indice_reclutamiento$PROPLOG)

# Aplicar la fórmula de normalización
indice_reclutamiento$PROPLOG2 <- ((indice_reclutamiento$PROPLOG - min_x) / (max_x - min_x))
```

veo la distribucion de las variables

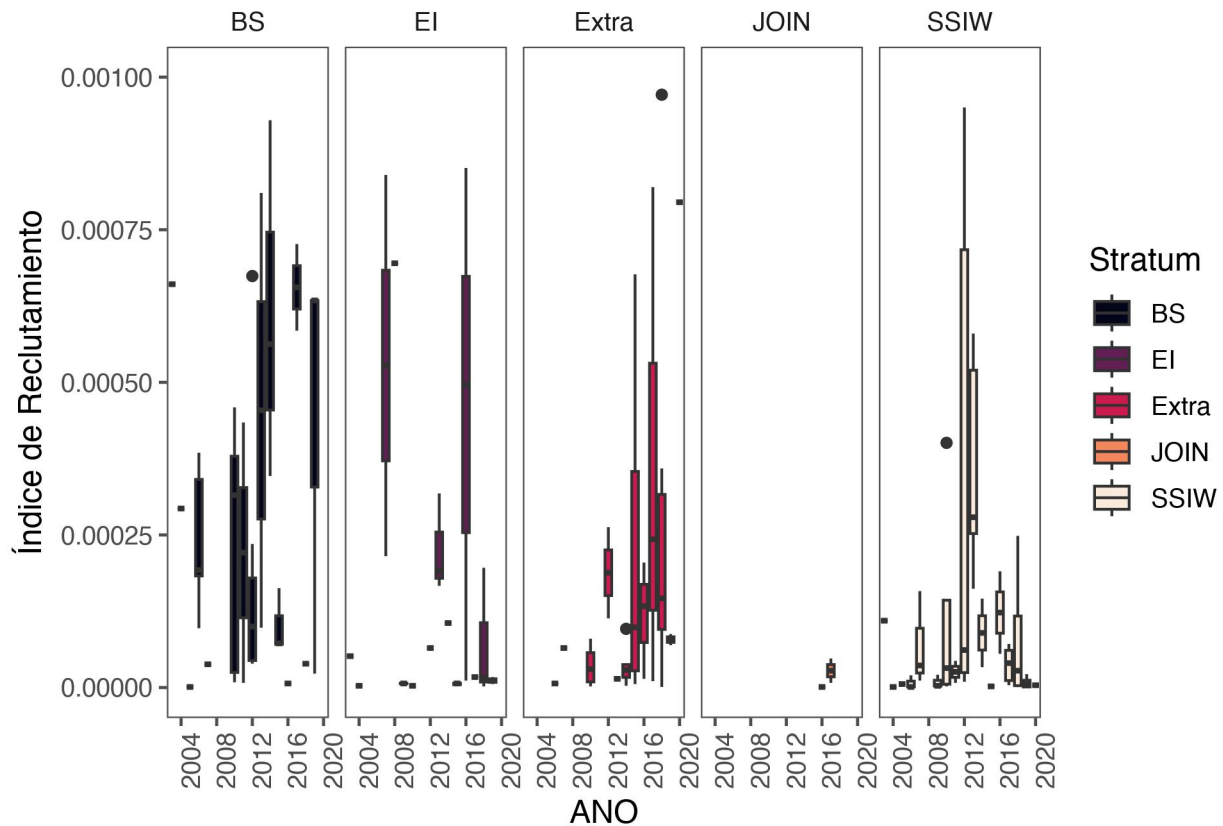
```
nor <- ggplot(indice_reclutamiento, aes(PROP)) +
  geom_histogram(fill="grey")+
  theme_few()
log <- ggplot(indice_reclutamiento, aes(PROPLOG)) +
  geom_histogram(fill="grey")+
  theme_few()

ggarrange(nor, log, ncol = 2)
```



```
indrec1 <- ggplot(indice_reclutamiento ,
  aes(x = factor(Year),
    y = PROP,
    fill=ID)) +
  geom_boxplot() +
  facet_wrap(ID~., ncol=5) +
  scale_fill_viridis_d(option = "F",
    name="Stratum")+
  scale_x_discrete(breaks = seq(from = 1996,
    to = 2022,
    by = 4))+

  theme_few()+
  theme(axis.text.x = element_text(angle = 90, hjust = 1))+
  labs(x = "ANO",
    y = "Índice de Reclutamiento")+
  ylim(0, 0.001)
indrec1
```



ahora como columnas por meses pero cambio los nombres

```
# Crear una nueva columna con los nombres de los meses
```

```
indice_reclutamiento2 <- indice_reclutamiento %>%
```

```
  mutate(Month = case_when(
```

```
    Month == 1 ~ "January",
```

```
    Month == 2 ~ "February",
```

```
    Month == 3 ~ "March",
```

```
    Month == 4 ~ "April",
```

```
    Month == 5 ~ "May",
```

```
    Month == 6 ~ "June",
```

```
    Month == 7 ~ "July",
```

```
    Month == 8 ~ "August",
```

```
    Month == 9 ~ "September",
```

```
    Month == 10 ~ "October",
```

```
    Month == 11 ~ "November",
```

```
    Month == 12 ~ "December")) %>%
```

```
  mutate(Month = factor(Month, levels = c("January",  
                                           "February",  
                                           "March",
```

```

"April",
"May",
"June",
"July",
"August",
"September",
"October",
"November",
"December"))))

```

#ahora trimestres

Definir una función para asignar trimestres

Definir reglas para asignar trimestres

```

indice_reclutamiento2<- indice_reclutamiento2 %>%
  mutate(quarter = case_when(
    Month %in% c("January", "February", "March") ~ "Q1",
    Month %in% c("April", "May", "June") ~ "Q2",
    Month %in% c("July", "August", "September") ~ "Q3",
    Month %in% c("October", "November", "December") ~ "Q4"))

```

```

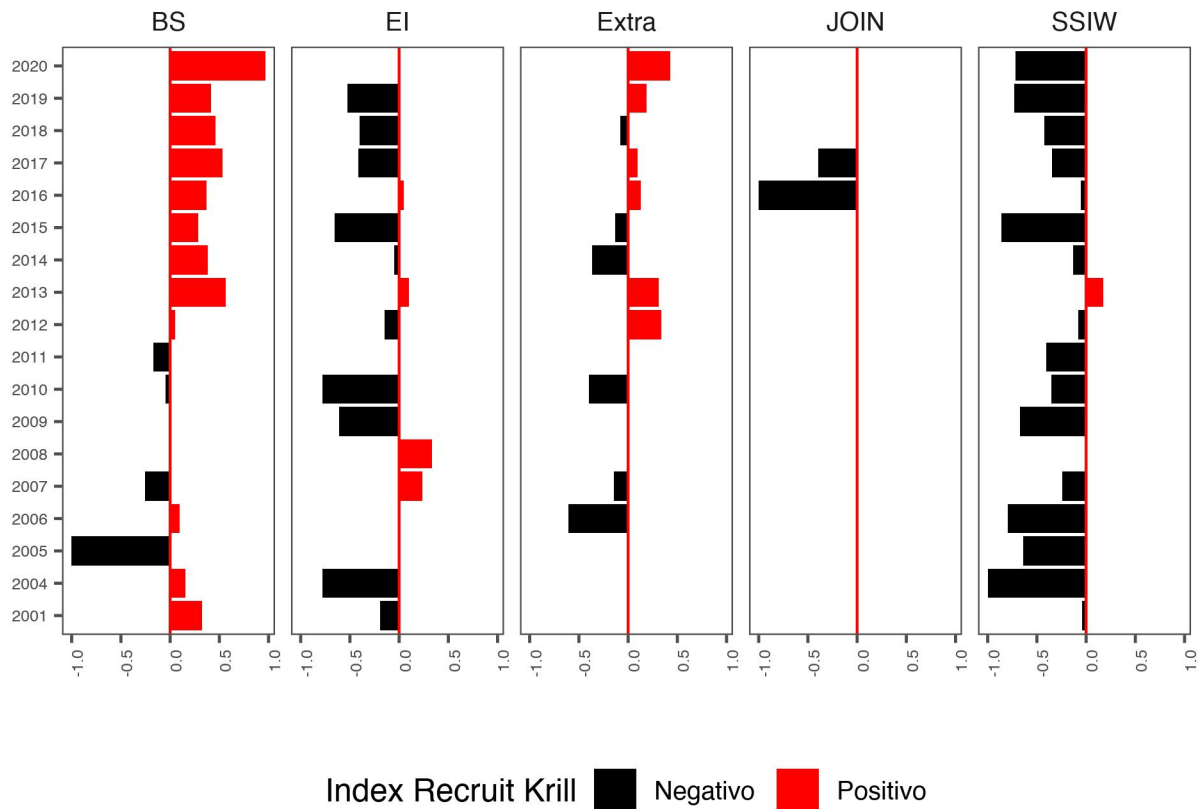
indseg3 <- ggplot(indice_reclutamiento2 %>%
  group_by(Year,ID) %>%
  summarise(PROLOG3=mean(PROLOG2)),
  aes(x = factor(Year),
      y = PROLOG3,
      fill=PROLOG3 > 0)) +
  geom_bar(stat = "identity") +
  scale_fill_manual(values = c("black", "red"),
                    labels = c("Negativo", "Positivo"),
                    name="Index Recruit Krill") +
  facet_wrap(~ID, ncol = 5) +
  geom_hline(yintercept = 0, color = "red")+
  scale_x_discrete(breaks = seq(from = 1996, to = 2022, by = 1))+
  theme_few()+
  theme(axis.text.x = element_text(angle = 90, hjust = 1),
        axis.text = element_text(size=6),
        legend.position = "bottom")+
  labs(x = "",

```

```

    y = "") +
  coord_flip()
indseg3

```



```

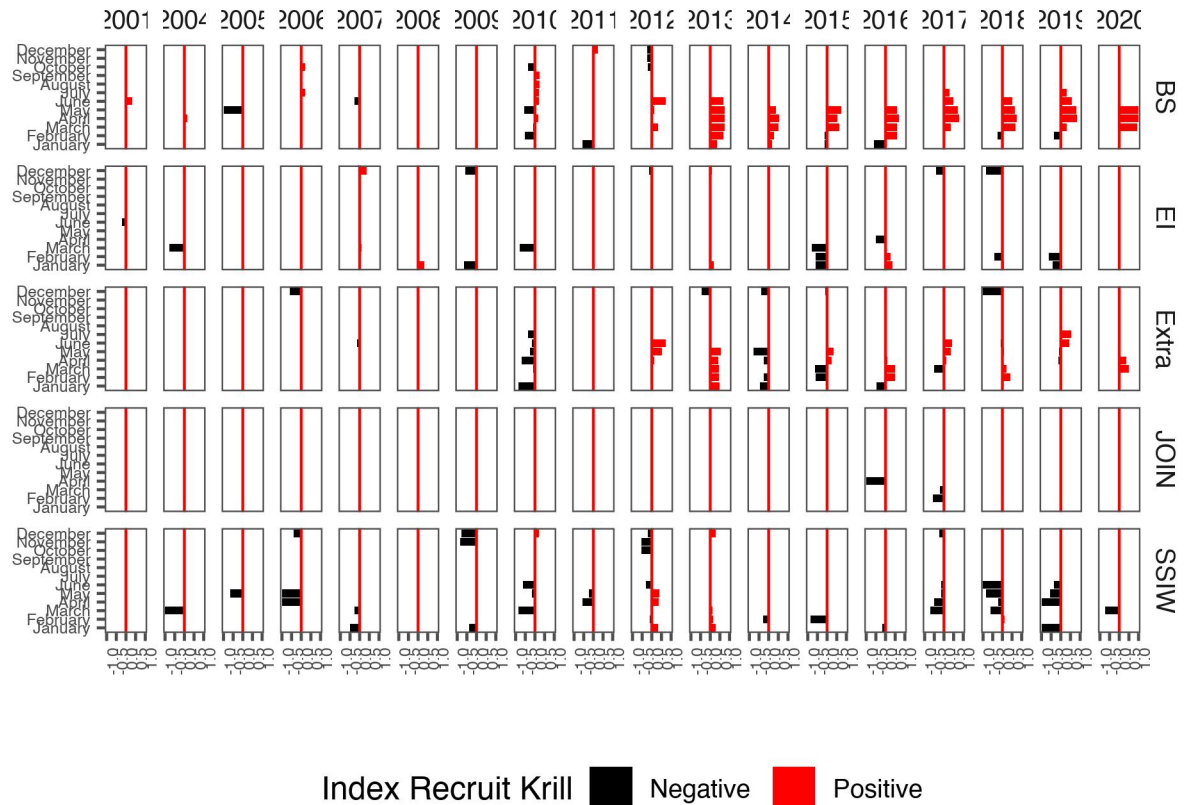
indrec4 <- ggplot(indice_reclutamiento2 %>%
  group_by(Year,ID,Month) %>%
  summarise(PROPLOG3=mean(PROPLOG2)),
  aes(x = Month,
    y = PROPLOG3,
    fill=PROPLOG3 > 0)) +
  geom_bar(stat = "identity") +
  scale_fill_manual(values = c("black", "red"),
    labels = c("Negative", "Positive"),
    name="Index Recruit Krill") +
  facet_grid(ID~Year) +
  geom_hline(yintercept = 0, color = "red")+
  #scale_x_discrete(breaks = seq(from = 1, to = 12, by = 2))+
  theme_few()+
  theme(axis.text.x = element_text(angle = 90, hjust = 1),

```

```

    axis.text = element_text(size=6),
    legend.position = "bottom")+
labs(x = "",
     y = "")+
coord_flip()
indrec4

```



Por trimestres

```

indrec5 <- ggplot(indice_reclutamiento2 %>%
  group_by(Year,ID,quarter) %>%
  summarise(PROPLOG3=mean(PROPLOG2)),
  aes(x = quarter,
      y = PROPLOG3,
      fill=PROPLOG3 > 0)) +
geom_bar(stat = "identity") +
scale_fill_manual(values = c("black", "red"),
                  labels = c("Negative", "Positive"),
                  name="Index Recruit Krill") +

```

```

facet_grid(ID~Year) +
geom_hline(yintercept = 0, color = "red")+
#scale_x_discrete(breaks = seq(from = 1, to = 12, by = 2))+
theme_few()+
theme(axis.text.x = element_text(angle = 90, hjust = 1),
      axis.text = element_text(size=6),
      legend.position = "bottom")+
labs(x = "",
     y = "")+
coord_flip()
indrec5

```

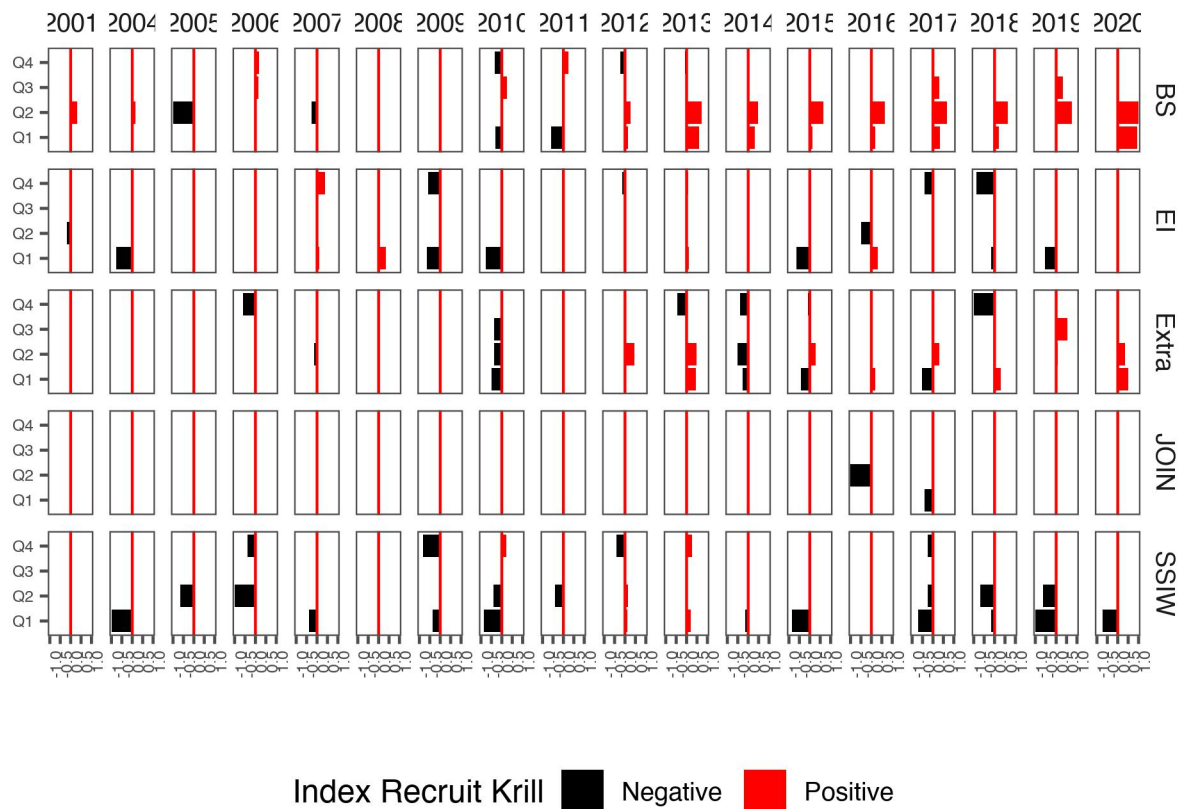


Grafico como Oscilación por Strata

```

recosc <- ggplot(indice_reclutamiento2 %>%
  group_by(Year,ID) %>%
  summarise(PROPLG3=mean(PROPLG2)),
  aes(x = Year, y = PROPLG3)) +
  geom_ribbon(aes(ymin = pmax(PROPLG3, 0),

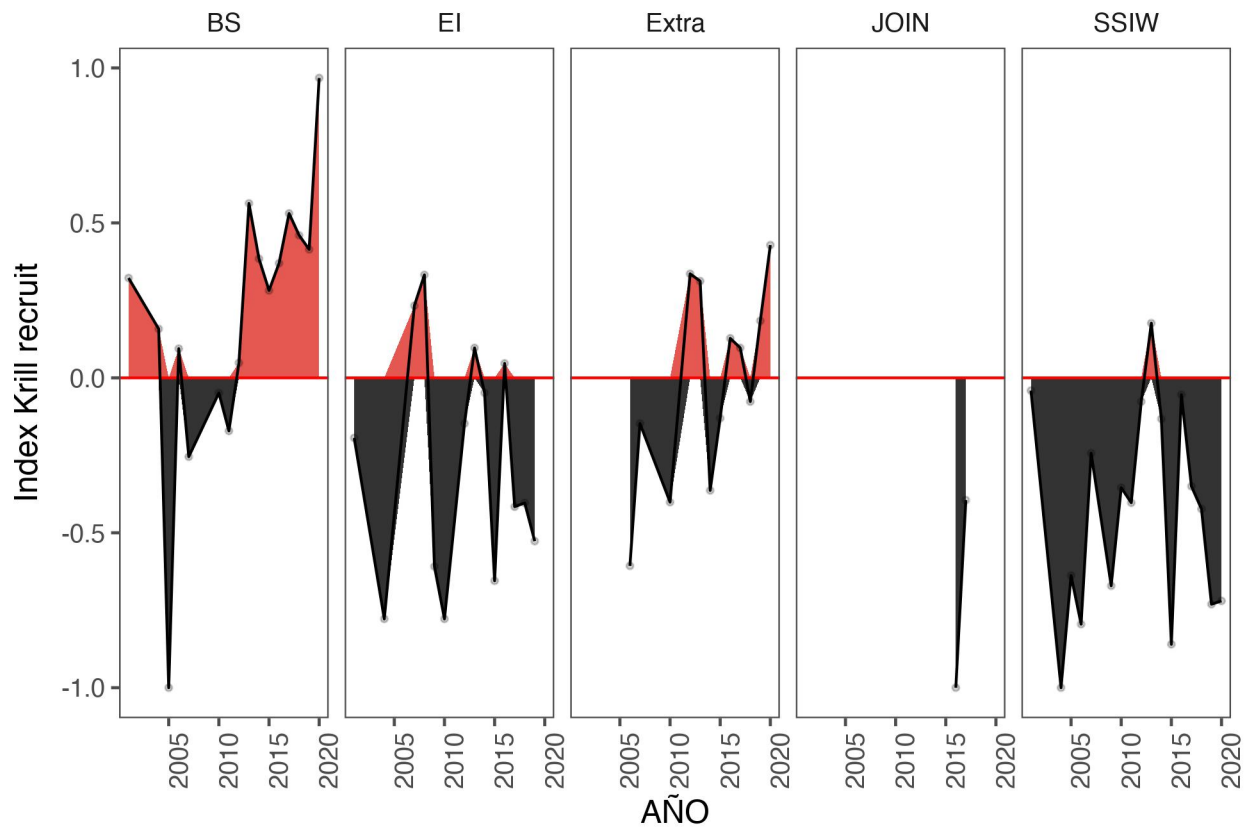
```



```

      ymax = 0),
      fill = "#de2d26",
      alpha = 0.8) + # Área por encima de la línea
geom_ribbon(aes(ymin = pmin(PROPLOG3, 0),
      ymax = 0),
      fill = "black",
      alpha = 0.8) + # Área por debajo de la línea
geom_hline(yintercept = 0, color = "red")+
geom_line(color = "black") + # Línea de anomalías
geom_point( alpha=0.2,
      size= 0.9)+
labs(x = "AÑO", y = "Index Krill recruit") +
  facet_wrap(~ID, ncol = 5)+
  theme_few()+
  theme(axis.text.x = element_text(angle = 90, hjust = 2))
recosc

```



References

Maschette, D., Wotherspoon, S., Pavez, C., Ziegler, P., Thanassekos, S., Reid, K., Kawaguchi, S., Welsford, D., & Constable, A. (2020). *Generalised R Yield Model (Grym)*. https://www.ccamlr.org/en/system/files/meeting%7B/_%7Ddocuments/with%7B/_%7Dcover/sc-39-bg-19.pdf