

International Nalaysis

```
##      dplyr      tidyr      ggplot2      plotly      knitr data.table
##      TRUE      TRUE      TRUE      TRUE      TRUE      TRUE
```

Landings data

- Estimate landings and revenue for each country form the Sea Around Us (<http://www.seaaroundus.org/>) database.

Notes

- Using reconstructed catch
- Filter out external countries (China, South Korea and “unknown”)
- Removed Chile’s reported landings in Peruvian waters

Questions to discuss with the team:

- Should we remove Recreational landings?
- Removed Chile’s reported landings in Peruvian waters?

```
#Data Path
Data_Path <- "./information/raw_databases/"
# File name
SAU_F_Name <- "SAU EEZ 604,152,76 v47-0.csv"
#read dataset
SAU <- fread(paste(Data_Path,SAU_F_Name,
                    sep = ""))

# Data exploration
# names(SAU)
# head(SAU)
# unique(SAU$fishing_sector) # "Industrial"    "Subsistence"    "Artisanal"    "Recreational"

# Countries excluded from the analysis
unique(SAU$fishing_entity) # There are 29 countries fishing in these waters, fill filter them
```

```
## [1] "Brazil"          "Chile"
## [3] "Peru"            "Venezuela"
## [5] "Sint Maarten"    "Cuba"
## [7] "Unknown Fishing Country" "Malta"
## [9] "Bahamas"         "USA"
## [11] "Japan"           "Argentina"
## [13] "Curacao"        "Russian Federation"
## [15] "Lithuania"       "South Africa"
## [17] "Taiwan"          "Estonia"
## [19] "Latvia"          "Ukraine"
## [21] "Korea (South)"   "Georgia"
## [23] "Uruguay"         "Spain"
## [25] "Mexico"          "Ecuador"
## [27] "Barbados"        "Panama"
## [29] "China"
```

```

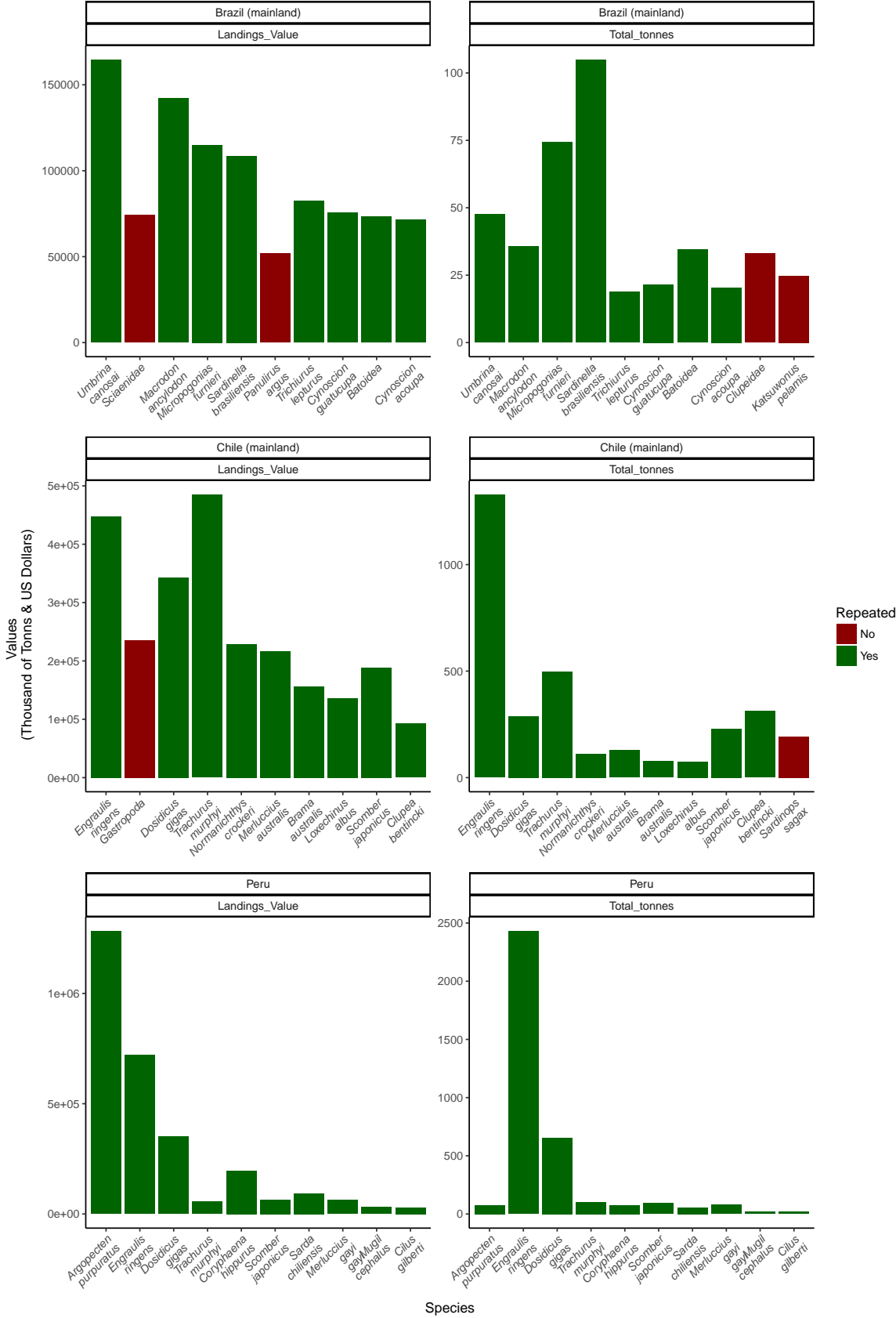
#Main fisheries

SAU$scientific_name <- gsub(" ", "\n",SAU$scientific_name)

# Plot <-
SAU %>%
  filter(scientific_name != "Marine\nfishes\nnot\nidentified") %>% #Removed but top in B and P
  group_by(area_name,
            year,
            scientific_name, # Trade data is too generic for species details
            fishing_entity#,
            # fishing_sector # #Trade data is too generic for sector details
  ) %>%
  summarise(
    Total_tonnes = sum(tonnes),
    Landings_Value = sum(landed_value)
  ) %>%
  filter(fishing_entity %in% Countries, # Remove China, Korea and unknown fishing country
         year == Year_an) %>% # Choose year to be analyzed
  rename(Reporter = fishing_entity) %>% # Change name of variable to match UN trade dataset
  filter(area_name != "Peru" | Reporter != "Chile") %>%
  tidyr::gather("Variable", "Value", 5:6) %>%
  group_by(area_name, Variable) %>%
  top_n(n = 10, wt = Value) %>% # Top 10, change if interactive plot
  # Determine what species are in both "Top" categories
  group_by(area_name, scientific_name) %>%
  mutate(Repeated = n()) %>%
  mutate(Repeated = ifelse(Repeated == 1, "No", "Yes")) %>%
  # Plot
  ggplot(.,
    aes(
      x= reorder(scientific_name, -Value),
      y = Value/1000, #thousands
      fill = Repeated
    )
  ) +
  geom_bar(stat = "identity") +
  theme_classic() +
  theme(axis.text.x = element_text(angle = 45,
                                     hjust = 1,
                                     face = "italic")
  ) +
  scale_fill_manual(values =c("darkred",
                              "darkgreen"),
                    name = "Repeated") +
  facet_wrap(~area_name + Variable,
             scales = "free",
             ncol = 2) +
  labs(title="Top 25 species by landings (left) and value (right)\n from the SAU dataset",
       x = "Species",
       y = "Values \n (Thousand of Tonns & US Dollars)"
  )

```

Top 25 species by landings (left) and value (right) from the SAU dataset



```

# For interactive plot
# Name the plot and then call it
# ggplotly(g_plot)

## Totals per Country

SAU_Clean <- SAU %>%
  group_by(area_name,
            year,
            # scientific_name, # Trade data is too generic for species details
            fishing_entity#,
            # fishing_sector # Trade data is too generic for sector details
            ) %>%
  summarise(
    Total_tonnes = sum(tonnes),
    Landings_Value = sum(landed_value)
  ) %>%
  filter(fishing_entity %in% Countries, # Remove China, Korea and unknown fishing country
         year == Year_an) %>% # Choose year to be analyzed
  rename(Reporter = fishing_entity) %>% # Change name of variable to match UN trade dataset
  filter(area_name != "Peru" | Reporter != "Chile") # Remove Chile report in Peruvian waters*

head(SAU_Clean)

## # A tibble: 3 x 5
## # Groups:   area_name, year [3]
##       area_name  year Reporter Total_tonnes Landings_Value
##       <chr> <int>   <chr>      <dbl>         <dbl>
## 1 Brazil (mainland) 2014   Brazil      879461.9      2298751150
## 2 Chile (mainland) 2014    Chile      3494985.2      2954129645
## 3 Peru 2014      Peru      3923988.5      3262441002

# names(SAU_Clean)
# head(SAU_Clean)
# unique(SAU_Clean$fishing_entity)

# Clenaed!

#Write csv
# write.csv(SAU_Clean,
#           "sau_clean_dataset.csv")

```

Trade data

- Estimate imports and exports of fish for each country from the UN come-trade database

Methods Notes:

- Data set does not contain weight information, just US dollars
- “Fish and crustaceans, molluscs and other aquatic invertebrates” is the only classification for all three countries
- Exports - Imports - Re-Imports

Questions to discuss with the team:

- What are “re-import”? Do we consider them? I’m assuming that they cost money to the country re-importing

```
#File name
UN_F_Name <- "UN_Comtrade_Statistics.csv"

UN <- fread(paste(Data_Path,UN_F_Name,
                  sep = ""))

# Data exploration

# head(UN)
# names(UN)
# unique(UN$Commodity) # [1] "Fish and crustaceans, molluscs and other aquatic invertebrates"
# unique(UN$Reporter) # [1] "Brazil" "Chile" "Peru"
# min(UN$Year) # 2013
# max(UN$Year) #2017

# max(UN$Gross_weight_kg)

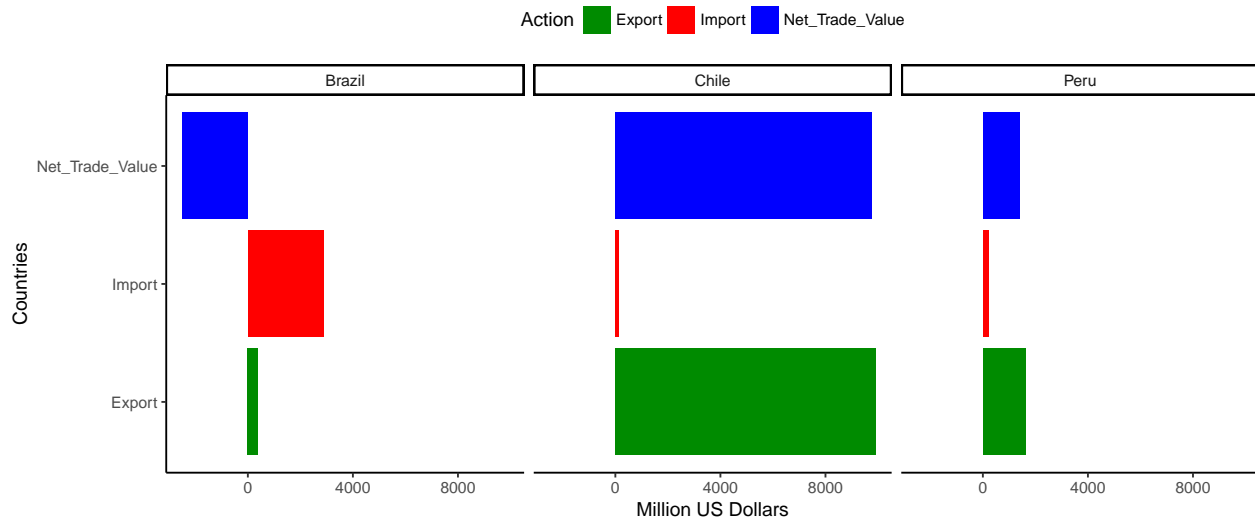
#### Trade Graph ####

# UN_Clean <-
UN %>%
  filter(Year == Year_an) %>%
  group_by(Reporter,
           Trade_Flow)
) %>%
  summarise(
    Trade_Value = sum(Trade_Value_US, na.rm =T)
  ) %>%
  spread(Trade_Flow,Trade_Value) %>%
  mutate(
    Net_Trade_Value = (Export-Import),
    Net_Re_Import_value = ifelse(!is.na(`Re-Import`), (Net_Trade_Value - `Re-Import`), Net_Trade_Value)
  ) %>%
  tidyr::gather("Action", "Value", 2:6) %>%
  filter(!Action %in% c("Re-Import", "Net_Export_Value", "Net_Re_Import_value")) %>%
  ggplot(.,
    aes(
      x = Action,
      y = Value/1000000,
      # colour = Action
      fill = Action
    )
  ) +
  # geom_point()
  geom_bar(stat = "identity") +
  facet_wrap(~Reporter) +
  scale_fill_manual(values =c("green4", # Export
                             "red", # Import
```

```

"blue" # Totals
)
) +
theme_classic() +
theme(legend.position = "top") +
coord_flip() +
labs(title="",
      x = "Countries",
      y = "Million US Dollars"
)

```



```
#### Trade Graph With Landings ####
```

```

# UN_Clean <-
UN %>%
  filter(Year == Year_an) %>%
  group_by(Reporter,
            Trade_Flow
  ) %>%
  summarise(
    Trade_Value = sum(Trade_Value_US, na.rm = T)
  ) %>%
  spread(Trade_Flow, Trade_Value) %>%
  mutate(
    Net_Trade_Value = (Export - Import),
    Net_Re_Import_value = ifelse(!is.na(`Re-Import`), (Net_Trade_Value - `Re-Import`), Net_Trade_Value)
  ) %>%
  left_join(SAU_Clean,
            by = "Reporter") %>%
  mutate(Fish_Industry_Total = Net_Trade_Value + Landings_Value) %>%
  select(1:3, 5, 10:11) %>%
  tidyr::gather("Action", "Value", 2:6) %>%
  ggplot(.,
    aes(
      x = Action,
      y = Value / 1000000,

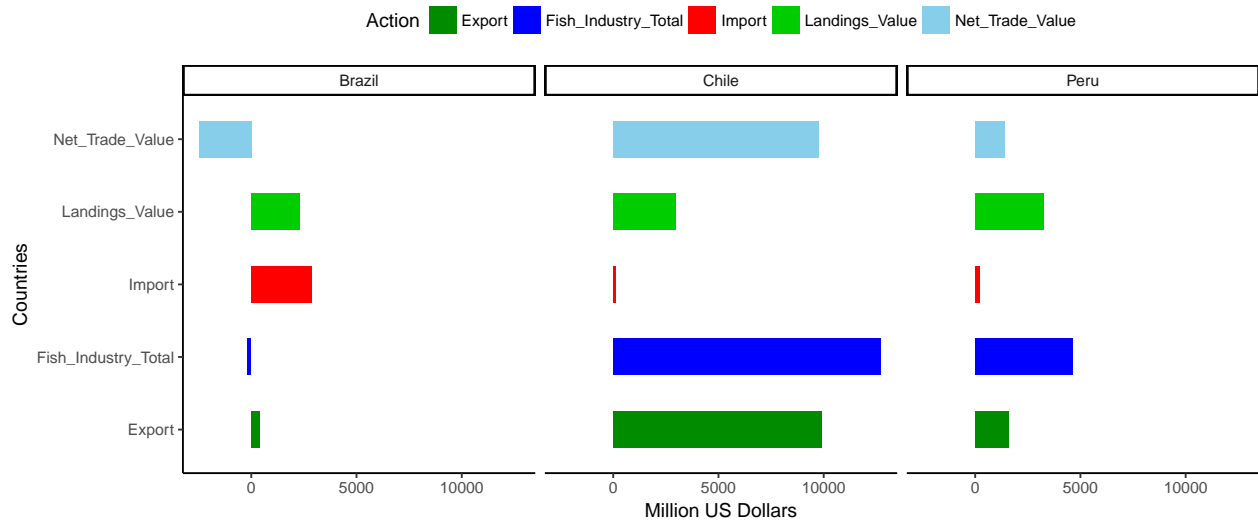
```

```

    # colour = Action
    fill = Action
  )
) +
# geom_point()
geom_bar(stat = "identity",
        width = 0.5 #Bar size
) +
facet_wrap(~Reporter) +
scale_fill_manual(values =c("green4",
                           "blue",
                           "red",
                           "green3",
                           "skyblue"

)
) +
theme_classic() +
theme(legend.position = "top") +
coord_flip() +
labs(title="",
     x = "Countries",
     y = "Million US Dollars"
)

```



Analysis

- Merge datasets and estimate total tons of fish and revenue from fishing activity

Methods Notes

- Total fishing value + Net exportation gain

Questions to discuss with the team

```

# Merge Datasets #####

SAU_Clean %>%
  left_join(UN_Clean,
            by = "Reporter") %>%
  mutate(
    Total_Fishing_Value = Total_value + Net_Re_Import_value
  ) %>%
  ggplot(.,
        aes(
          x = Reporter,
          y = Total_Fishing_Value,
          fill = Reporter
        )) +
  geom_bar(stat = "identity") +
  theme_classic()

```