

# Jam

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```
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.4      v readr      2.1.5
## v forcats    1.0.0      v stringr   1.5.1
## v ggplot2    3.5.0      v tibble    3.2.1
## v lubridate  1.9.3      v tidyr     1.3.1
## v purrr      1.0.2
## -- 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 errors
```

```
library(corrplot)
```

```
## corrplot 0.95 loaded
```

```
library(ggplot2)
library(dbplyr)
```

```
##
## Attaching package: 'dbplyr'
##
## The following objects are masked from 'package:dplyr':
##
##      ident, sql
```

```
library(colorspace)
library(tidyr)
library(astsa)
library(maps)
```

```
##
## Attaching package: 'maps'
##
## The following object is masked from 'package:astsa':
##
##      unemp
##
```

```
## The following object is masked from 'package:purrr':
##
##      map
```

```
library(sf)
```

```
## Linking to GEOS 3.11.2, GDAL 3.8.2, PROJ 9.3.1; sf_use_s2() is TRUE
```

```
# data upload
age_consent <- read.csv("data/Age-of-Sexual-Consent.csv")
slavery_index <- read.csv("data/Global_Slavery_Index_2023.csv")
marriage <- read.csv("data/Percentage-of-Females-Married-by-15-years-old.csv")
human_trafficking <- read.csv("data/human-trafficking-victims_new.csv")
inequality <- read.csv("data/Gender_Inequality_Index.csv")
```

```
nrow(age_consent) #201
```

```
## [1] 201
```

```
nrow(slavery_index) # 180
```

```
## [1] 180
```

```
nrow(marriage) #202
```

```
## [1] 202
```

```
nrow(human_trafficking) # 364
```

```
## [1] 364
```

```
head(human_trafficking) # 84
```

```
##      Entity Code Year Detected.victims.of.human.trafficking
## 1 Burkina Faso BFA 2007                                289
## 2   Colombia COL 2007                                    6
## 3   Germany DEU 2007                                    7
## 4     Mali MLI 2007                                     5
## 5 Netherlands NLD 2007                                 16
## 6   Norway NOR 2007                                    19
```

```
unique(human_trafficking$Entity)
```

```
## [1] "Burkina Faso"      "Colombia"          "Germany"
## [4] "Mali"              "Netherlands"       "Norway"
## [7] "Philippines"       "Romania"           "Serbia"
## [10] "Tajikistan"        "Togo"              "Belarus"
## [13] "Ghana"             "Mexico"            "Qatar"
```

```
## [16] "Armenia"           "Bolivia"           "Nepal"
## [19] "Paraguay"          "Peru"               "United Kingdom"
## [22] "Argentina"         "Egypt"              "Ireland"
## [25] "Myanmar"           "Nigeria"            "United States"
## [28] "Bosnia and Herzegovina" "Bulgaria"           "Congo"
## [31] "Dominican Republic" "Moldova"             "Venezuela"
## [34] "Canada"            "Czechia"             "Ecuador"
## [37] "Honduras"          "Italy"               "Jamaica"
## [40] "Mozambique"         "Thailand"             "Uzbekistan"
## [43] "Guinea"             "Uganda"              "Guatemala"
## [46] "Hungary"            "Laos"                "Austria"
## [49] "Denmark"            "France"              "Lithuania"
## [52] "Mauritius"          "Niger"               "Turkey"
## [55] "Ukraine"            "Algeria"             "Cote d'Ivoire"
## [58] "Croatia"            "Finland"             "Greece"
## [61] "Slovakia"          "Tunisia"             "Angola"
## [64] "Benin"              "Cameroon"            "India"
## [67] "Kazakhstan"         "Malaysia"            "Morocco"
## [70] "Russia"             "Brazil"              "El Salvador"
## [73] "Estonia"            "Indonesia"           "Namibia"
## [76] "South Africa"       "Spain"               "Sudan"
## [79] "Costa Rica"         "Iraq"                "Lebanon"
## [82] "Portugal"           "North Macedonia"     "Sierra Leone"
```

```
head(age_consent)
```

```
##      Country Age.Of.Consent
## 1  Afghanistan Must be married
## 2    Albania           14
## 3    Algeria           16
## 4 American Samoa           16
## 5    Andorra           16
## 6    Angola            12
```

```
head(slavery_index)
```

```
##      Country      Population      Region
## 1  Afghanistan  38,928,000  Asia and the Pacific
## 2    Albania    2,878,000  Europe and Central Asia
## 3    Algeria    43,851,000      Africa
## 4    Angola     32,866,000      Africa
## 5 Antigua and Barbuda    98,000  Americas
## 6    Argentina   45,196,000  Americas
## Estimated.prevalence.of.modern.slavery.per.1.000.population
## 1                                     13.0
## 2                                     11.8
## 3                                     1.9
## 4                                     4.1
## 5                                     NA
## 6                                     4.2
## Estimated.number.of.people.in.modern.slavery
## 1                                     505,000
## 2                                     34,000
```

```
## 3 84,000
## 4 136,000
## 5
## 6 189,000
```

```
slavery_index <- slavery_index |>
  rename(Per_1000 = Estimated.prevalence.of.modern.slavery.per.1.000.population) |>
  rename(Num_modern_slavery = Estimated.number.of.people.in.modern.slavery)

head(marriage)
```

```
## Countries Percentage.of.Females.Married.by.15.years.old
## 1 Afghanistan 10
## 2 Albania 1
## 3 Algeria 0
## 4 Andorra -
## 5 Angola 8
## 6 Anguilla -
```

```
marriage <- marriage |>
  rename(Percent_married_15 = Percentage.of.Females.Married.by.15.years.old) |>
  rename(Country = Countries)

head(human_trafficking)
```

```
## Entity Code Year Detected.victims.of.human.trafficking
## 1 Burkina Faso BFA 2007 289
## 2 Colombia COL 2007 6
## 3 Germany DEU 2007 7
## 4 Mali MLI 2007 5
## 5 Netherlands NLD 2007 16
## 6 Norway NOR 2007 19
```

```
human_trafficking <- human_trafficking |>
  rename(Detected_Number = Detected.victims.of.human.trafficking) |>
  rename(Country = Entity)
```

```
#slavery_index |> group_by(Region)
```

```
slavery_index <- left_join(slavery_index, age_consent, by = "Country")

slavery_index <- left_join(slavery_index, marriage, by = "Country")

#slavery_index <- slavery_index |> mutate(across(everything(), ~replace_na(.x, NA)))

head(slavery_index)
```

```
## Country Population Region Per_1000
## 1 Afghanistan 38,928,000 Asia and the Pacific 13.0
## 2 Albania 2,878,000 Europe and Central Asia 11.8
## 3 Algeria 43,851,000 Africa 1.9
```

```
## 4          Angola    32,866,000          Africa    4.1
## 5 Antigua and Barbuda    98,000          Americas    NA
## 6          Argentina    45,196,000          Americas    4.2
##   Num_modern_slavery Age.Of.Consent Percent_married_15
## 1          505,000 Must be married          10
## 2           34,000          14          1
## 3           84,000          16          0
## 4          136,000          12          8
## 5           16          16          -
## 6          189,000          18          2
```

```
# slavery_index <- slavery_index |>
#   mutate(Percent_married_15 = recode(Percent_married_15, - = NA))

#slavery_index[5, "Percent_married_15"]
```

```
# Column Prep
```

```
slavery_index$Num_modern_slavery <- replace(slavery_index$Num_modern_slavery,
                                             slavery_index$Num_modern_slavery=="", NA)

slavery_index$Num_modern_slavery <- as.numeric(gsub(",", "", slavery_index$Num_modern_slavery))

slavery_index$Population <- as.double(gsub(",", "", slavery_index$Population))
```

```
## Warning: NAs introduced by coercion
```

```
slavery_index$Percent_married_15 <- as.double(replace(slavery_index$Percent_married_15, slavery_index$P
```

```
slavery_index |> group_by(Region)
```

```
## # A tibble: 180 x 7
## # Groups:   Region [5]
##   Country      Population Region Per_1000 Num_modern_slavery Age.Of.Consent
##   <chr>          <dbl> <chr>    <dbl>          <dbl> <chr>
## 1 Afghanistan    38928000 Asia ~      13          505000 Must be marri~
## 2 Albania         2878000 Europ~    11.8          34000 14
## 3 Algeria         43851000 Africa     1.9          84000 16
## 4 Angola          32866000 Africa     4.1         136000 12
## 5 Antigua and Bar~    98000 Ameri~    NA           NA 16
## 6 Argentina       45196000 Ameri~    4.2         189000 18
## 7 Armenia         2963000 Europ~    8.9          26000 16
## 8 Australia       25500000 Asia ~     1.6          41000 16
## 9 Austria          9006000 Europ~    1.9          17000 14
## 10 Azerbaijan     10139000 Europ~   10.6         107000 16
## # i 170 more rows
## # i 1 more variable: Percent_married_15 <dbl>
```

```
unique(slavery_index$Region)
```

```
## [1] "Asia and the Pacific"    "Europe and Central Asia"
## [3] "Africa"                  "Americas"
## [5] "Arab States"
```

```
Asia_Pacific_mean <- slavery_index |> filter(Region == "Asia and the Pacific") |>
  summarize(across(Per_1000, mean, na.rm=T))
```

```
## Warning: There was 1 warning in 'summarize()'.
## i In argument: 'across(Per_1000, mean, na.rm = T)'.
## Caused by warning:
## ! The '...' argument of 'across()' is deprecated as of dplyr 1.1.0.
## Supply arguments directly to '.fns' through an anonymous function instead.
##
## # Previously
## across(a:b, mean, na.rm = TRUE)
##
## # Now
## across(a:b, \(x) mean(x, na.rm = TRUE))
```

```
Americas_mean <- slavery_index |> filter(Region == "Americas") |> summarize(across(Per_1000, mean, na.rm=T))
```

```
Euro_Central_Asia_mean <- slavery_index |>
  filter(Region == "Europe and Central Asia") |>
  summarize(across(Per_1000, mean, na.rm=T))
```

```
Africa_mean <- slavery_index |>
  filter(Region == "Africa") |>
  summarize(across(Per_1000, mean, na.rm=T))
```

```
Arab_States_mean <- slavery_index |>
  filter(Region == "Arab States") |>
  summarize(across(Per_1000, mean, na.rm=T))
```

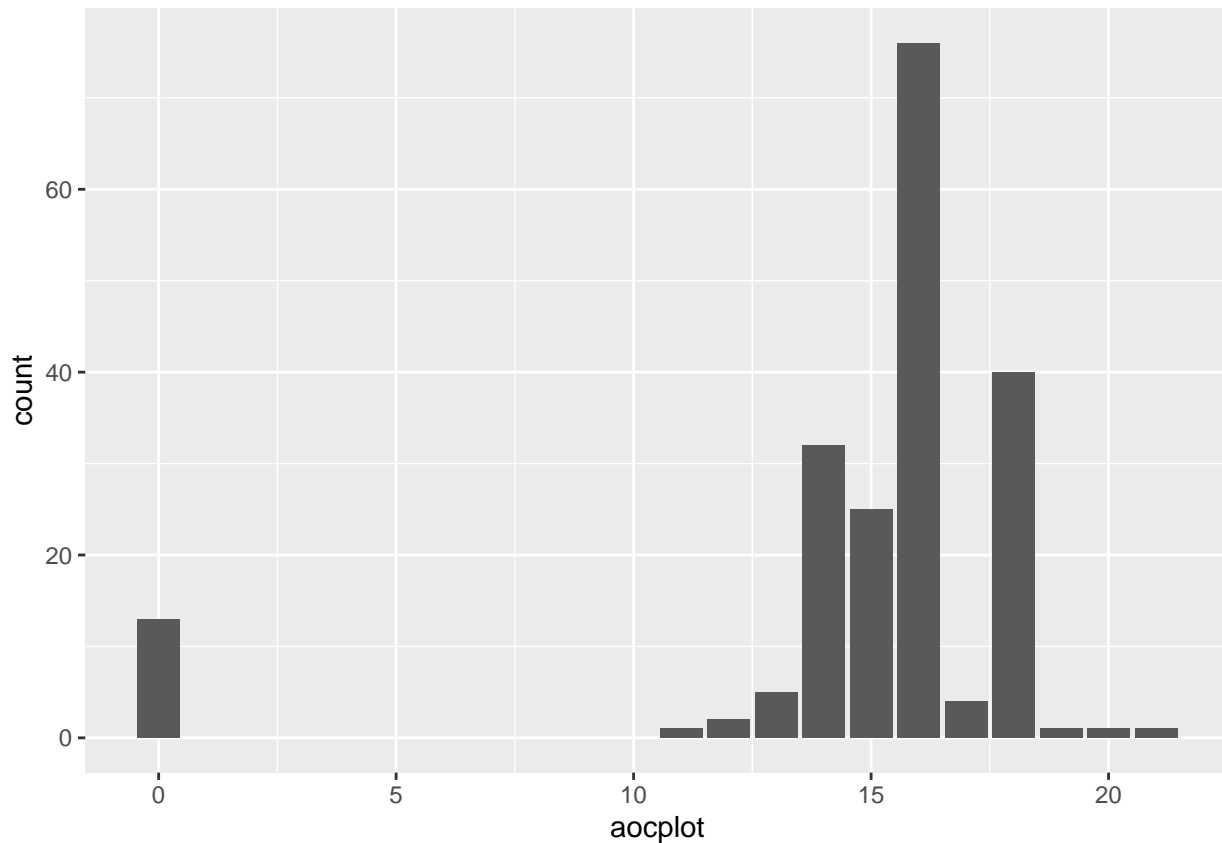
```
Regions <- data.frame(
  Region = c("Asia and the Pacific",
             "Americas", "Europe and Central Asia",
             "Africa", "Arab States"),
  Mean_Per_1000 = c(Asia_Pacific_mean[1,1],
                    Americas_mean[1,1], Euro_Central_Asia_mean[1,1],
                    Africa_mean[1,1], Arab_States_mean[1,1])
)
```

```
regions_graph <- Regions |> mutate(Region = as.factor(Region))|>
  mutate(Region = fct_reorder(Region, desc(Mean_Per_1000))) |> ggplot(aes(x = Region, y = Mean_Per_1000)) +
  geom_col(fill = "cornflowerblue", col = "royalblue4") +
  ylab("Average Modern Slavery per 1000 People")
```

```
aocplot <- as.double(replace(age_consent$Age.Of.Consent, age_consent$Age.Of.Consent=="Must be married",
```

```
df1 <- data.frame(x = aocplot, y = 1:201)
```

```
ggplot(df1, aes(x=aocplot)) +
  geom_bar()
```



```
country_max <- slavery_index[which.max(slavery_index$Per_1000), ]
country_min <- slavery_index[which.min(slavery_index$Per_1000), ]

slavery_index |> arrange(desc(Per_1000))
```

##	Country	Population	Region
## 1	North Korea	25779000	Asia and the Pacific
## 2	Eritrea	3546000	Africa
## 3	Mauritania	4650000	Africa
## 4	Saudi Arabia	34814000	Arab States
## 5	Türkiye	84339000	Europe and Central Asia
## 6	Tajikistan	9538000	Europe and Central Asia
## 7	United Arab Emirates	9890000	Arab States
## 8	Afghanistan	38928000	Asia and the Pacific
## 9	Kuwait	4271000	Arab States
## 10	Russia	145934000	Europe and Central Asia
## 11	Ukraine	43734000	Europe and Central Asia
## 12	North Macedonia	2083000	Europe and Central Asia
## 13	Myanmar	54410000	Asia and the Pacific
## 14	Turkmenistan	6031000	Europe and Central Asia
## 15	Albania	2878000	Europe and Central Asia
## 16	Belarus	9449000	Europe and Central Asia
## 17	Kazakhstan	18777000	Europe and Central Asia
## 18	Azerbaijan	10139000	Europe and Central Asia
## 19	Pakistan	220892000	Asia and the Pacific
## 20	Papua New Guinea	8947000	Asia and the Pacific

## 21	South Sudan	11194000	Africa
## 22	Bosnia and Herzegovina	3281000	Europe and Central Asia
## 23	Jordan	10203000	Arab States
## 24	Moldova	4034000	Europe and Central Asia
## 25	Venezuela	28436000	Americas
## 26	Armenia	2963000	Europe and Central Asia
## 27	Kyrgyzstan	6524000	Europe and Central Asia
## 28	Syria	17501000	Arab States
## 29	Bulgaria	6948000	Europe and Central Asia
## 30	Haiti	11403000	Americas
## 31	El Salvador	6486000	Americas
## 32	Cyprus	1207000	Europe and Central Asia
## 33	India	1380004000	Asia and the Pacific
## 34	Kosovo	NA	Europe and Central Asia
## 35	Republic of the Congo	5518000	Africa
## 36	Colombia	50883000	Americas
## 37	Equatorial Guinea	1403000	Africa
## 38	Georgia	3989000	Europe and Central Asia
## 39	Guatemala	17916000	Americas
## 40	Nigeria	206140000	Africa
## 41	Philippines	109581000	Asia and the Pacific
## 42	Slovakia	5460000	Europe and Central Asia
## 43	Ecuador	17643000	Americas
## 44	Gabon	2226000	Africa
## 45	Lebanon	6825000	Arab States
## 46	Burundi	11891000	Africa
## 47	Romania	19238000	Europe and Central Asia
## 48	Uzbekistan	33469000	Europe and Central Asia
## 49	Côte d'Ivoire	26378000	Africa
## 50	Jamaica	2961000	Americas
## 51	Nicaragua	6625000	Americas
## 52	Bolivia	11673000	Americas
## 53	Bangladesh	164689000	Asia and the Pacific
## 54	Djibouti	988000	Africa
## 55	Iran	83993000	Asia and the Pacific
## 56	Peru	32972000	Americas
## 57	Honduras	9905000	Americas
## 58	Serbia	8737000	Europe and Central Asia
## 59	Libya	6871000	Africa
## 60	Qatar	2881000	Arab States
## 61	Bahrain	1702000	Arab States
## 62	Indonesia	273524000	Asia and the Pacific
## 63	Dominican Republic	10848000	Americas
## 64	Hungary	9660000	Europe and Central Asia
## 65	Mexico	128933000	Americas
## 66	Gambia	2417000	Africa
## 67	Oman	5107000	Arab States
## 68	Sri Lanka	21413000	Asia and the Pacific
## 69	Greece	10423000	Europe and Central Asia
## 70	Paraguay	7133000	Americas
## 71	Ethiopia	114964000	Africa
## 72	Malaysia	32366000	Asia and the Pacific
## 73	Somalia	15893000	Africa
## 74	Lithuania	2722000	Europe and Central Asia



## 75	Timor-Leste	1318000	Asia and the Pacific
## 76	Yemen	29826000	Arab States
## 77	Chad	16426000	Africa
## 78	Cameroon	26546000	Africa
## 79	Thailand	69800000	Asia and the Pacific
## 80	Iraq	40223000	Arab States
## 81	Poland	37847000	Europe and Central Asia
## 82	Cuba	11327000	Americas
## 83	Central African Republic	4830000	Africa
## 84	Croatia	4105000	Europe and Central Asia
## 85	Lao PDR	7276000	Asia and the Pacific
## 86	Mali	20251000	Africa
## 87	Zambia	18384000	Africa
## 88	Brazil	212559000	Americas
## 89	Cambodia	16719000	Asia and the Pacific
## 90	Kenya	53771000	Africa
## 91	Zimbabwe	14863000	Africa
## 92	Malawi	19130000	Africa
## 93	Panama	4315000	Americas
## 94	Trinidad and Tobago	1399000	Americas
## 95	Madagascar	27691000	Africa
## 96	Niger	24207000	Africa
## 97	Democratic Republic of the Congo	89561000	Africa
## 98	Guinea-Bissau	1968000	Africa
## 99	Slovenia	2079000	Europe and Central Asia
## 100	Egypt	102334000	Africa
## 101	Rwanda	12952000	Africa
## 102	Argentina	45196000	Americas
## 103	Czechia	10709000	Europe and Central Asia
## 104	Guyana	787000	Americas
## 105	Uganda	45741000	Africa
## 106	Angola	32866000	Africa
## 107	Estonia	1327000	Europe and Central Asia
## 108	Viet Nam	97339000	Asia and the Pacific
## 109	China	1439324000	Asia and the Pacific
## 110	Guinea	13133000	Africa
## 111	Mongolia	3278000	Asia and the Pacific
## 112	Sudan	43849000	Africa
## 113	Israel	8656000	Europe and Central Asia
## 114	Portugal	10197000	Europe and Central Asia
## 115	Burkina Faso	20903000	Africa
## 116	Eswatini	1160000	Africa
## 117	South Korea	51269000	Asia and the Pacific
## 118	Latvia	1886000	Europe and Central Asia
## 119	Sierra Leone	7977000	Africa
## 120	Italy	60462000	Europe and Central Asia
## 121	Nepal	29137000	Asia and the Pacific
## 122	Togo	8279000	Africa
## 123	United States of America	331003000	Americas
## 124	Chile	19116000	Americas
## 125	Costa Rica	5094000	Americas
## 126	Liberia	5058000	Africa
## 127	Benin	12123000	Africa
## 128	Mozambique	31255000	Africa

## 129	Ghana	31073000	Africa
## 130	Senegal	16744000	Africa
## 131	Tanzania	59734000	Africa
## 132	Hong Kong	7497000	Asia and the Pacific
## 133	South Africa	59309000	Africa
## 134	Namibia	2541000	Africa
## 135	Morocco	36911000	Africa
## 136	Spain	46755000	Europe and Central Asia
## 137	Tunisia	11819000	Africa
## 138	France	65274000	Europe and Central Asia
## 139	Singapore	5850000	Asia and the Pacific
## 140	Algeria	43851000	Africa
## 141	Austria	9006000	Europe and Central Asia
## 142	Uruguay	3474000	Americas
## 143	Botswana	2352000	Africa
## 144	Canada	37742000	Americas
## 145	United Kingdom	67886000	Europe and Central Asia
## 146	Taiwan	23817000	Asia and the Pacific
## 147	Australia	25500000	Asia and the Pacific
## 148	Lesotho	2142000	Africa
## 149	New Zealand	4822000	Asia and the Pacific
## 150	Mauritius	1272000	Africa
## 151	Finland	5541000	Europe and Central Asia
## 152	Ireland	4938000	Europe and Central Asia
## 153	Japan	126476000	Asia and the Pacific
## 154	Belgium	11590000	Europe and Central Asia
## 155	Denmark	5792000	Europe and Central Asia
## 156	Germany	83784000	Europe and Central Asia
## 157	Netherlands	17135000	Europe and Central Asia
## 158	Sweden	10099000	Europe and Central Asia
## 159	Norway	5421000	Europe and Central Asia
## 160	Switzerland	8655000	Europe and Central Asia
## 161	Antigua and Barbuda	98000	Americas
## 162	Bahamas	393000	Americas
## 163	Barbados	287000	Americas
## 164	Belize	398000	Americas
## 165	Brunei Darussalam	437000	Asia and the Pacific
## 166	Cape Verde	NA	Africa
## 167	Fiji	896000	Asia and the Pacific
## 168	Iceland	341000	Europe and Central Asia
## 169	Liechtenstein	NA	Europe and Central Asia
## 170	Luxembourg	626000	Europe and Central Asia
## 171	Maldives	541000	Asia and the Pacific
## 172	Malta	442000	Europe and Central Asia
## 173	Montenegro	628000	Europe and Central Asia
## 174	Palau	NA	Asia and the Pacific
## 175	Saint Lucia	184000	Americas
## 176	Saint Vincent and the Grenadines	111000	Americas
## 177	Seychelles	98000	Africa
## 178	Solomon Islands	687000	Asia and the Pacific
## 179	Suriname	587000	Americas
## 180	Vanuatu	307000	Asia and the Pacific
##	Per_1000 Num_modern_slavery	Age.Of.Consent	Percent_married_15
## 1	104.6	2696000	15 NA

## 2	90.3	320000	18	13.0
## 3	32.0	149000	16	16.0
## 4	21.3	740000 Must be married		NA
## 5	15.6	1320000 <NA>		2.0
## 6	14.0	133000	16	0.0
## 7	13.4	132000 Must be married		NA
## 8	13.0	505000 Must be married		10.0
## 9	13.0	55000 Must be married		NA
## 10	13.0	1899000	16	NA
## 11	12.8	559000	16	0.0
## 12	12.6	26000 <NA>		0.0
## 13	12.1	657000	14	2.0
## 14	11.9	72000	16	0.0
## 15	11.8	34000	14	1.0
## 16	11.3	107000	16	0.0
## 17	11.1	208000	16	0.0
## 18	10.6	107000	16	2.0
## 19	10.6	2349000 Must be married		4.0
## 20	10.3	93000	16	8.0
## 21	10.3	115000	18	9.0
## 22	10.1	33000 <NA>		0.0
## 23	10.0	102000	16	2.0
## 24	9.5	38000	16	NA
## 25	9.5	270000	16	NA
## 26	8.9	26000	16	0.0
## 27	8.7	57000	16	0.0
## 28	8.7	153000	15	NA
## 29	8.5	59000	14	NA
## 30	8.2	94000	18	2.0
## 31	8.1	52000	18	4.0
## 32	8.0	10000	17	NA
## 33	8.0	11050000	18	5.0
## 34	8.0	14000 <NA>		NA
## 35	8.0	44000	18	NA
## 36	7.8	397000	14	5.0
## 37	7.8	11000	18	9.0
## 38	7.8	31000	16	0.0
## 39	7.8	140000	18	6.0
## 40	7.8	1611000	11	12.0
## 41	7.8	859000	12	2.0
## 42	7.7	42000 <NA>		NA
## 43	7.6	135000	14	4.0
## 44	7.6	17000	18	3.0
## 45	7.6	52000	18	1.0
## 46	7.5	89000	18	3.0
## 47	7.5	145000	15	0.5
## 48	7.4	249000	16	0.0
## 49	7.3	193000 <NA>		7.0
## 50	7.3	22000	16	1.0
## 51	7.3	49000	18	10.0
## 52	7.2	83000	14	NA
## 53	7.1	1162000	14	16.0
## 54	7.1	7000	18	1.0
## 55	7.1	597000 Must be married		NA

## 56	7.1	234000	14	2.0
## 57	7.0	69000	15	9.0
## 58	7.0	61000	14	1.0
## 59	6.8	47000	Must be married	NA
## 60	6.8	20000	Must be married	0.0
## 61	6.7	11000	21	NA
## 62	6.7	1833000	16	2.0
## 63	6.6	72000	18	9.0
## 64	6.6	63000	14	NA
## 65	6.6	850000	17	4.0
## 66	6.5	16000	18	6.0
## 67	6.5	33000	Must be married	1.0
## 68	6.5	139000	16	1.0
## 69	6.4	66000	15	NA
## 70	6.4	46000	14	4.0
## 71	6.3	727000	18	14.0
## 72	6.3	202000	16	NA
## 73	6.2	98000	18	8.0
## 74	6.1	17000	16	0.0
## 75	6.1	8000	<NA>	3.0
## 76	6.0	180000	Must be married	7.0
## 77	5.9	97000	14	24.0
## 78	5.8	155000	16	11.0
## 79	5.7	401000	15	6.0
## 80	5.5	221000	18	7.0
## 81	5.5	209000	15	NA
## 82	5.4	61000	16	5.0
## 83	5.2	25000	<NA>	26.0
## 84	5.2	22000	15	NA
## 85	5.2	38000	<NA>	NA
## 86	5.2	106000	18	16.0
## 87	5.1	94000	16	5.0
## 88	5.0	1053000	14	6.0
## 89	5.0	83000	15	2.0
## 90	5.0	269000	18	2.0
## 91	5.0	74000	16	5.0
## 92	4.9	93000	14	8.0
## 93	4.7	20000	18	7.0
## 94	4.7	7000	16	1.0
## 95	4.6	127000	14	13.0
## 96	4.6	112000	13	28.0
## 97	4.5	407000	14	8.0
## 98	4.5	9000	<NA>	8.0
## 99	4.4	9000	15	NA
## 100	4.3	442000	18	2.0
## 101	4.3	55000	18	0.0
## 102	4.2	189000	18	2.0
## 103	4.2	45000	<NA>	NA
## 104	4.2	3000	<NA>	6.0
## 105	4.2	190000	18	7.0
## 106	4.1	136000	12	8.0
## 107	4.1	5000	14	NA
## 108	4.1	396000	<NA>	1.0
## 109	4.0	5771000	14	0.0

## 110	4.0	53000	15	17.0
## 111	4.0	13000	16	1.0
## 112	4.0	174000	Must be married	12.0
## 113	3.8	33000	16	NA
## 114	3.8	39000	14	NA
## 115	3.7	77000	13	9.0
## 116	3.6	4000	<NA>	0.0
## 117	3.5	180000	20	NA
## 118	3.4	6000	16	NA
## 119	3.4	27000	18	9.0
## 120	3.3	197000	14	NA
## 121	3.3	97000	16	6.0
## 122	3.3	28000	16	6.0
## 123	3.3	1091000	<NA>	NA
## 124	3.2	61000	18	NA
## 125	3.2	16000	15	2.0
## 126	3.1	16000	18	6.0
## 127	3.0	37000	18	6.0
## 128	3.0	93000	<NA>	17.0
## 129	2.9	91000	16	3.0
## 130	2.9	49000	16	9.0
## 131	2.9	171000	18	NA
## 132	2.8	21000	<NA>	NA
## 133	2.7	158000	16	1.0
## 134	2.4	6000	16	2.0
## 135	2.3	85000	<NA>	1.0
## 136	2.3	108000	16	NA
## 137	2.3	27000	18	0.0
## 138	2.1	135000	15	NA
## 139	2.1	12000	16	0.0
## 140	1.9	84000	16	0.0
## 141	1.9	17000	14	NA
## 142	1.9	7000	15	1.0
## 143	1.8	4000	16	NA
## 144	1.8	69000	16	NA
## 145	1.8	122000	16	NA
## 146	1.7	40000	16	NA
## 147	1.6	41000	16	NA
## 148	1.6	4000	16	1.0
## 149	1.6	8000	16	NA
## 150	1.5	2000	14	NA
## 151	1.4	8000	16	NA
## 152	1.1	5000	17	NA
## 153	1.1	144000	13	NA
## 154	1.0	11000	16	0.0
## 155	0.6	4000	15	NA
## 156	0.6	47000	14	NA
## 157	0.6	10000	16	NA
## 158	0.6	6000	15	NA
## 159	0.5	3000	16	0.0
## 160	0.5	4000	16	NA
## 161	NA	NA	16	NA
## 162	NA	NA	16	NA
## 163	NA	NA	16	8.0

## 164	NA	NA	16	6.0
## 165	NA	NA	<NA>	NA
## 166	NA	NA	14	NA
## 167	NA	NA	16	0.0
## 168	NA	NA	15	NA
## 169	NA	NA	14	NA
## 170	NA	NA	16	NA
## 171	NA	NA Must be married		0.0
## 172	NA	NA	18	NA
## 173	NA	NA	14	2.0
## 174	NA	NA	16	NA
## 175	NA	NA	16	4.0
## 176	NA	NA	15	NA
## 177	NA	NA	18	NA
## 178	NA	NA	15	6.0
## 179	NA	NA	16	9.0
## 180	NA	NA	16	3.0

```
slavery_index |> filter(Region == "Asia and the Pacific") |> arrange(desc(Per_1000))
```

##	Country	Population	Region	Per_1000
## 1	North Korea	25779000	Asia and the Pacific	104.6
## 2	Afghanistan	38928000	Asia and the Pacific	13.0
## 3	Myanmar	54410000	Asia and the Pacific	12.1
## 4	Pakistan	220892000	Asia and the Pacific	10.6
## 5	Papua New Guinea	8947000	Asia and the Pacific	10.3
## 6	India	1380004000	Asia and the Pacific	8.0
## 7	Philippines	109581000	Asia and the Pacific	7.8
## 8	Bangladesh	164689000	Asia and the Pacific	7.1
## 9	Iran	83993000	Asia and the Pacific	7.1
## 10	Indonesia	273524000	Asia and the Pacific	6.7
## 11	Sri Lanka	21413000	Asia and the Pacific	6.5
## 12	Malaysia	32366000	Asia and the Pacific	6.3
## 13	Timor-Leste	1318000	Asia and the Pacific	6.1
## 14	Thailand	69800000	Asia and the Pacific	5.7
## 15	Lao PDR	7276000	Asia and the Pacific	5.2
## 16	Cambodia	16719000	Asia and the Pacific	5.0
## 17	Viet Nam	97339000	Asia and the Pacific	4.1
## 18	China	1439324000	Asia and the Pacific	4.0
## 19	Mongolia	3278000	Asia and the Pacific	4.0
## 20	South Korea	51269000	Asia and the Pacific	3.5
## 21	Nepal	29137000	Asia and the Pacific	3.3
## 22	Hong Kong	7497000	Asia and the Pacific	2.8
## 23	Singapore	5850000	Asia and the Pacific	2.1
## 24	Taiwan	23817000	Asia and the Pacific	1.7
## 25	Australia	25500000	Asia and the Pacific	1.6
## 26	New Zealand	4822000	Asia and the Pacific	1.6
## 27	Japan	126476000	Asia and the Pacific	1.1
## 28	Brunei Darussalam	437000	Asia and the Pacific	NA
## 29	Fiji	896000	Asia and the Pacific	NA
## 30	Maldives	541000	Asia and the Pacific	NA
## 31	Palau	NA	Asia and the Pacific	NA
## 32	Solomon Islands	687000	Asia and the Pacific	NA
## 33	Vanuatu	307000	Asia and the Pacific	NA

	Num_modern_slavery	Age.Of.Consent	Percent_married_15
## 1	2696000	15	NA
## 2	505000	Must be married	10
## 3	657000	14	2
## 4	2349000	Must be married	4
## 5	93000	16	8
## 6	11050000	18	5
## 7	859000	12	2
## 8	1162000	14	16
## 9	597000	Must be married	NA
## 10	1833000	16	2
## 11	139000	16	1
## 12	202000	16	NA
## 13	8000	<NA>	3
## 14	401000	15	6
## 15	38000	<NA>	NA
## 16	83000	15	2
## 17	396000	<NA>	1
## 18	5771000	14	0
## 19	13000	16	1
## 20	180000	20	NA
## 21	97000	16	6
## 22	21000	<NA>	NA
## 23	12000	16	0
## 24	40000	16	NA
## 25	41000	16	NA
## 26	8000	16	NA
## 27	144000	13	NA
## 28	NA	<NA>	NA
## 29	NA	16	0
## 30	NA	Must be married	0
## 31	NA	16	NA
## 32	NA	15	6
## 33	NA	16	3

```
# population of top 20
# area for top 20
```

```
top_20 <- slavery_index |> arrange(desc(Per_1000)) |> head(n=20)
bottom_20 <- slavery_index |> arrange(Per_1000) |> head(n=20)
```

```
highest_pop <- slavery_index |> arrange(desc(Population))
#cutoff <- quantile(slavery_index$Population, probs = 0.992, na.rm = T)
# outliers <- slavery_index %>% filter(Population > cutoff)
```

```
lowest_pop <- slavery_index |> arrange(Population) |> filter(!is.na(Per_1000))
```

```
# biggest countries - don't use
```

```
biggest_pop_graph <- head(highest_pop, n=10) |> mutate(Country = as.factor(Country)) |>
  mutate(Country = fct_reorder(Country, Population, .desc = T)) |> ggplot(aes(x = Country, y = Per_1000))
  geom_col(fill = "cornflowerblue", col = "royalblue4")
```

```
# lowest pop - don't use
```

```

lowest_pop_graph <- head(lowest_pop, n=10) |> mutate(Country = as.factor(Country)) |>
  mutate(Country = fct_reorder(Country, Population)) |>
  ggplot(aes(x = Country, y = Per_1000)) +
  geom_col(fill = "cornflowerblue", col = "royalblue4")

# top 20
top_20_graph <- top_20 |> mutate(Country = as.factor(Country)) |>
  mutate(Country = fct_reorder(Country, Per_1000)) |>
  ggplot(aes(x = Country, y = Per_1000, fill = Region)) +
  geom_col() +
  coord_flip() +
  ylab("Modern Slavery per 1000 People") +
  labs(title = "Top 20 Countries")

# bottom 20
bottom_20_graph <- bottom_20 |> mutate(Country = as.factor(Country)) |>
  mutate(Country = fct_reorder(Country, Per_1000)) |>
  ggplot(aes(x = Country, y = Per_1000, fill = Region)) +
  geom_col() +
  coord_flip() +
  ylab("Modern Slavery per 1000 People") +
  labs(title = "Bottom 20 Countries")

# population
population_graph <- slavery_index |> ggplot(aes(x=Population, y=Per_1000)) +
  geom_point() +
  geom_point(col = "darkslategray") +
  ylab("Modern Slavery per 1000 People")

population_without_outliers_graph <- slavery_index |> filter(!(Country == "China" |
  Country == "India" | Country == "Eritrea" |
  Country == "North Korea")) |>
  ggplot(aes(x=Population, y=Per_1000)) +
  geom_point(col = "darkslategray") +
  ylab("Modern Slavery per 1000 People")

# Age of Consent
age_consent_graph <- slavery_index |> ggplot(aes(x=Age.Of.Consent, y=Per_1000, col = Region)) +
  geom_point() +
  xlab("Age of Consent") +
  ylab("Modern Slavery per 1000 People")

# Percent married before 15
percent_married_15_graph <- slavery_index |> ggplot(aes(x=Percent_married_15, y=Per_1000)) +
  geom_col(fill = "darkseagreen") +
  xlab("Percent of Females Married before age 15") +
  ylab("Modern Slavery per 1000 People")

# all countries w/ values w/o country labels
all_countries_per_1000 <- slavery_index |> filter(!is.na(Per_1000)) |> mutate(Country = as.factor(Country)) |>
  mutate(Country = fct_reorder(Country, Per_1000)) |>
  ggplot(aes(x = Country, y = Per_1000, fill = Region)) +

```



```
geom_col() +
coord_flip() +
theme(
  axis.ticks.y = element_blank(),
  axis.text.y = element_blank()
) +
ylab("Modern Slavery per 1000 People")
```

```
# correlation
```

```
x_mx <- slavery_index[, !(names(slavery_index) %in% c("Country", "Region"))] |>
  filter(!(Age.Of.Consent == "Must be married")) |> apply(2, as.double)
corr_mx <- cor(x_mx, use = "pairwise.complete.obs")

sorted_corr <- sort(corr_mx[, "Per_1000"])
```

```
slavery_index |> filter(Country == "United Kingdom")
```

```
##           Country Population           Region Per_1000 Num_modern_slavery
## 1 United Kingdom  67886000 Europe and Central Asia      1.8          122000
##   Age.Of.Consent Percent_married_15
## 1              16              NA
```

forced labour, forced marriage, debt bondage, human trafficking, and child exploitation

```
climate_change <- read.csv("data/climate_change_indicators.csv")

head(climate_change)
```

```
##   ObjectId          Country ISO2 ISO3
## 1         1 Afghanistan, Islamic Rep. of AF AFG
## 2         2           Albania AL ALB
## 3         3           Algeria DZ DZA
## 4         4 American Samoa AS ASM
## 5         5 Andorra, Principality of AD AND
## 6         6           Angola AO AGO
##
##                                     Indicator
## 1 Temperature change with respect to a baseline climatology, corresponding to the period 1951-1980
## 2 Temperature change with respect to a baseline climatology, corresponding to the period 1951-1980
## 3 Temperature change with respect to a baseline climatology, corresponding to the period 1951-1980
## 4 Temperature change with respect to a baseline climatology, corresponding to the period 1951-1980
## 5 Temperature change with respect to a baseline climatology, corresponding to the period 1951-1980
## 6 Temperature change with respect to a baseline climatology, corresponding to the period 1951-1980
##           Unit
## 1 Degree Celsius
## 2 Degree Celsius
## 3 Degree Celsius
## 4 Degree Celsius
## 5 Degree Celsius
## 6 Degree Celsius
##
## 1 Food and Agriculture Organization of the United Nations (FAO). 2022. FAOSTAT Climate Change, Climate
```

##	2	Food and Agriculture Organization of the United Nations (FAO). 2022. FAOSTAT Climate Change, Climate Indicators, Surface Temperature Change
##	3	Food and Agriculture Organization of the United Nations (FAO). 2022. FAOSTAT Climate Change, Climate Indicators, Surface Temperature Change
##	4	Food and Agriculture Organization of the United Nations (FAO). 2022. FAOSTAT Climate Change, Climate Indicators, Surface Temperature Change
##	5	Food and Agriculture Organization of the United Nations (FAO). 2022. FAOSTAT Climate Change, Climate Indicators, Surface Temperature Change
##	6	Food and Agriculture Organization of the United Nations (FAO). 2022. FAOSTAT Climate Change, Climate Indicators, Surface Temperature Change
##		CTS_Code CTS_Name
##	1	ECCS Surface Temperature Change
##	2	ECCS Surface Temperature Change
##	3	ECCS Surface Temperature Change
##	4	ECCS Surface Temperature Change
##	5	ECCS Surface Temperature Change
##	6	ECCS Surface Temperature Change
##		CTS_Full_Descriptor
##	1	Environment, Climate Change, Climate Indicators, Surface Temperature Change
##	2	Environment, Climate Change, Climate Indicators, Surface Temperature Change
##	3	Environment, Climate Change, Climate Indicators, Surface Temperature Change
##	4	Environment, Climate Change, Climate Indicators, Surface Temperature Change
##	5	Environment, Climate Change, Climate Indicators, Surface Temperature Change
##	6	Environment, Climate Change, Climate Indicators, Surface Temperature Change
##		F1961 F1962 F1963 F1964 F1965 F1966 F1967 F1968 F1969 F1970 F1971
##	1	-0.113 -0.164 0.847 -0.764 -0.244 0.226 -0.371 -0.423 -0.539 0.813 0.619
##	2	0.627 0.326 0.075 -0.166 -0.388 0.559 -0.074 0.081 -0.013 -0.106 -0.195
##	3	0.164 0.114 0.077 0.250 -0.100 0.433 -0.026 -0.067 0.291 0.116 -0.385
##	4	0.079 -0.042 0.169 -0.140 -0.562 0.181 -0.368 -0.187 0.132 -0.047 -0.477
##	5	0.736 0.112 -0.752 0.308 -0.490 0.415 0.637 0.018 -0.137 0.121 -0.326
##	6	0.041 -0.152 -0.190 -0.229 -0.196 0.175 -0.081 -0.193 0.188 0.248 -0.097
##		F1972 F1973 F1974 F1975 F1976 F1977 F1978 F1979 F1980 F1981 F1982
##	1	-1.124 0.232 -0.489 -0.445 -0.286 0.513 0.129 0.361 0.600 0.483 -0.346
##	2	-0.069 -0.288 -0.139 -0.211 -0.683 0.545 -0.814 0.203 -0.414 -0.351 0.173
##	3	-0.348 -0.015 -0.503 -0.539 -0.782 0.504 0.012 0.654 0.232 0.215 0.399
##	4	-0.067 0.330 -0.308 -0.118 -0.177 0.156 0.092 0.341 0.350 0.179 0.280
##	5	-0.499 0.025 -0.371 0.246 -0.045 -0.093 -0.163 0.058 -0.188 0.178 1.044
##	6	-0.035 0.475 -0.158 -0.029 -0.313 0.272 0.037 0.291 0.279 -0.071 0.164
##		F1983 F1984 F1985 F1986 F1987 F1988 F1989 F1990 F1991 F1992 F1993
##	1	0.164 0.145 0.283 -0.141 0.391 0.919 -0.205 0.730 -0.168 -0.294 0.220
##	2	-0.128 -0.270 -0.103 0.569 -0.106 0.370 -0.066 0.795 -0.269 0.106 0.076
##	3	0.560 -0.004 0.508 0.296 0.975 1.304 0.386 1.266 0.031 -0.312 0.552
##	4	0.313 0.277 0.256 0.394 0.354 0.509 0.143 0.497 0.641 0.344 -0.069
##	5	0.859 -0.157 0.059 0.387 0.397 0.883 1.162 1.736 0.231 0.386 0.174
##	6	0.487 0.631 0.694 0.176 0.689 0.572 -0.055 0.687 0.341 0.466 0.256
##		F1994 F1995 F1996 F1997 F1998 F1999 F2000 F2001 F2002 F2003 F2004 F2005
##	1	0.430 0.359 -0.116 0.471 0.675 1.198 0.993 1.311 1.365 0.587 1.373 0.401
##	2	1.330 -0.172 -0.038 0.075 0.795 0.670 1.065 1.532 0.492 0.970 0.444 0.189
##	3	0.732 0.595 0.846 1.059 1.109 1.476 0.820 1.856 1.258 1.585 0.988 1.264
##	4	0.189 0.755 0.784 NA NA 0.242 0.626 0.904 1.152 0.716 0.191 0.801
##	5	1.508 1.279 0.570 1.788 1.018 1.055 1.050 1.480 0.835 1.949 0.936 0.851
##	6	0.212 0.753 0.370 0.107 1.064 0.417 0.169 0.295 0.735 0.889 0.414 1.021
##		F2006 F2007 F2008 F2009 F2010 F2011 F2012 F2013 F2014 F2015 F2016 F2017 F2018
##	1	1.720 0.675 0.704 0.895 1.613 1.397 0.223 1.281 0.456 1.093 1.555 1.540 1.544
##	2	0.345 1.316 0.978 0.910 1.191 1.055 1.487 1.333 1.198 1.569 1.464 1.121 2.028
##	3	1.395 1.220 1.185 0.945 2.265 1.398 1.147 1.192 1.690 1.121 1.757 1.512 1.210
##	4	0.403 1.032 0.670 NA 1.311 0.854 0.924 1.257 1.170 1.009 1.539 1.435 1.189
##	5	1.485 1.024 0.946 1.413 0.471 1.677 1.265 0.831 1.946 1.690 1.990 1.925 1.919
##	6	0.561 0.885 0.501 0.708 1.194 0.880 0.552 1.044 0.828 1.331 1.609 0.870 1.395

##	CTS_Code	CTS_Name
## 1	ECCS Surface Temperature Change	
## 2	ECCS Surface Temperature Change	
## 3	ECCS Surface Temperature Change	
## 4	ECCS Surface Temperature Change	
## 5	ECCS Surface Temperature Change	
## 6	ECCS Surface Temperature Change	

```
## CTS_Full_Descriptor
```

## 1	Environment, Climate Change, Climate Indicators, Surface Temperature Change
## 2	Environment, Climate Change, Climate Indicators, Surface Temperature Change
## 3	Environment, Climate Change, Climate Indicators, Surface Temperature Change
## 4	Environment, Climate Change, Climate Indicators, Surface Temperature Change
## 5	Environment, Climate Change, Climate Indicators, Surface Temperature Change
## 6	Environment, Climate Change, Climate Indicators, Surface Temperature Change

##	F1961	F1962	F1963	F1964	F1965	F1966	F1967	F1968	F1969	F1970	F1971
----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

```
## 1 -0.113 -0.164  0.847 -0.764 -0.244  0.226 -0.371 -0.423 -0.539  0.813  0.619
```

```
## 2  0.627  0.326  0.075 -0.166 -0.388  0.559 -0.074  0.081 -0.013 -0.106 -0.195
```

```
## 3  0.164  0.114  0.077  0.250 -0.100  0.433 -0.026 -0.067  0.291  0.116 -0.385
```

```
## 4  0.079 -0.042  0.169 -0.140 -0.562 0.181 -0.368 -0.187  0.132 -0.047 -0.477
```

```
## 5  0.736  0.112 -0.752  0.308 -0.490  0.415  0.637  0.018 -0.137  0.121 -0.326
```

```
## 6  0.041 -0.152 -0.190 -0.229 -0.196 0.175 -0.081 -0.193  0.188  0.248 -0.097
```

##	F1972	F1973	F1974	F1975	F1976	F1977	F1978	F1979	F1980	F1981	F1982
----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

```
## 1 -1.124  0.232 -0.489 -0.445 -0.286  0.513  0.129 0.361  0.600  0.483 -0.346
```

```
## 2 -0.069 -0.288 -0.139 -0.211 -0.683  0.545 -0.814 0.203 -0.414 -0.351  0.173
```

```
## 3 -0.348 -0.015 -0.503 -0.539 -0.782  0.504  0.012 0.654  0.232  0.215  0.399
```

```
## 4 -0.067  0.330 -0.308 -0.118 -0.177  0.156  0.092 0.341  0.350  0.179  0.280
```

```
## 5 -0.499  0.025 -0.371  0.246 -0.045 -0.093 -0.163 0.058 -0.188  0.178  1.044
```

```
## 6 -0.035  0.475 -0.158 -0.029 -0.313  0.272  0.037 0.291  0.279 -0.071  0.164
```

##	F1983	F1984	F1985	F1986	F1987	F1988	F1989	F1990	F1991	F1992	F1993
----	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------	-------

```
## 1  0.164  0.145  0.283 -0.141  0.391  0.919 -0.205  0.730 -0.168 -0.294  0.220
```

```
## 2 -0.128 -0.270 -0.103  0.569 -0.106 0.370 -0.066 0.795 -0.269  0.106  0.076
```

```
## 3  0.560 -0.004  0.508  0.296  0.975 1.304  0.386 1.266  0.031 -0.312  0.552
```

```
## 4  0.313  0.277  0.256  0.394  0.354 0.509  0.143 0.497  0.641  0.344 -0.069
```

```
## 5  0.859 -0.157  0.059  0.387  0.397 0.883  1.162 1.736  0.231  0.386  0.174
```

```
## 6  0.487  0.631  0.694  0.176  0.689 0.572 -0.055 0.687  0.341  0.466  0.256
```

```
##      F1994   F1995   F1996 F1997 F1998 F1999 F2000 F2001 F2002 F2003 F2004 F2005
```

```
## 1 0.430 0.359 -0.116 0.471 0.675 1.198 0.993 1.311 1.365 0.587 1.373 0.401
```

```
## 2 1.330 -0.172 -0.038 0.075 0.795 0.670 1.065 1.532 0.492 0.970 0.444 0.189
```

```
## 3 0.732 0.595 0.846 1.059 1.109 1.476 0.820 1.856 1.258 1.585 0.988 1.264
```

```
## 4 0.189 0.755 0.784 NA NA 0.242 0.626 0.904 1.152 0.716 0.191 0.801
```

```
## 5 1.508 1.279 0.570 1.788 1.018 1.055 1.050 1.480 0.835 1.949 0.936 0.851
```

```
## 6 0.212  0.753  0.370 0.107 1.064 0.417 0.169 0.295 0.735 0.889 0.414 1.021
```

```
##      F2006 F2007 F2008 F2009 F2010 F2011 F2012 F2013 F2014 F2015 F2016 F2017 F20
```

```
## 1 1.720 0.675 0.704 0.895 1.613 1.397 0.223 1.281 0.456 1.093 1.555 1.540 1.5
```

```
## 2 0.345 1.316 0.978 0.910 1.191 1.055 1.487 1.333 1.198 1.569 1.464 1.121 2.0
```

```
## 3 1.395 1.220 1.185 0.945 2.265 1.398 1.147 1.192 1.690 1.121 1.757 1.512 1.2
```

```
## 4 0.403 1.032 0.670    NA 1.311 0.854 0.924 1.257 1.170 1.009 1.539 1.435 1.1
```

```
## 5 1.485 1.024 0.946 1.413 0.471 1.677 1.265 0.831 1.946 1.690 1.990 1.925 1.9
```

```
## 6 0.561 0.885 0.501 0.708 1.194 0.880 0.552 1.044 0.828 1.331 1.609 0.870 1.3
```

```
##   F2019 F2020 F2021 F2022
## 1 0.910 0.498 1.327 2.012
## 2 1.675 1.498 1.536 1.518
## 3 1.115 1.926 2.330 1.688
## 4 1.539 1.430 1.268 1.256
## 5 1.964 2.562 1.533 3.243
## 6 1.752 1.162 1.553 1.212
```

```
# Temperature change with respect to a baseline climatology, corresponding to the period 1951-1980
# celcius
```

```
climate_change <- climate_change |> select(!c(Source, Unit,
                                                Indicator, IS02, CTS_Code, CTS_Name, CTS_Full_Descriptor))

climate_2022 <- climate_change |> select(c(Country, IS03, F2022))

nrow(climate_2022)
```

```
## [1] 225
```

```
climate_2022 |> arrange(desc(F2022))
```

```
##           Country IS03 F2022
## 1   Andorra, Principality of AND 3.243
## 2           France FRA 2.929
## 3       Luxembourg LUX 2.926
## 4         Belgium BEL 2.807
## 5       Switzerland CHE 2.798
## 6           Monaco MCO 2.737
## 7 Kazakhstan, Rep. of KAZ 2.712
## 8       Kyrgyz Rep. KGZ 2.619
## 9  Uzbekistan, Rep. of UZB 2.607
## 10  Netherlands, The NLD 2.601
## 11           Germany DEU 2.596
## 12  Liechtenstein LIE 2.534
## 13  Turkmenistan TKM 2.528
## 14         Austria AUT 2.498
## 15           Spain ESP 2.489
## 16  Slovenia, Rep. of SVN 2.435
## 17  New Caledonia NCL 2.421
## 18       Czech Rep. CZE 2.394
## 19  Iran, Islamic Rep. of IRN 2.370
## 20         Portugal PRT 2.318
## 21         Morocco MAR 2.303
## 22         Tunisia TUN 2.291
## 23         Kuwait KWT 2.275
## 24  Russian Federation RUS 2.261
## 25         Gibraltar GIB 2.228
## 26         Denmark DNK 2.219
## 27  Croatia, Rep. of HRV 2.209
## 28         Mongolia MNG 2.197
## 29           Iraq IRQ 2.147
## 30  Moldova, Rep. of MDA 2.143
```

## 31	Hungary	HUN	2.142
## 32	Italy	ITA	2.135
## 33	San Marino, Rep. of	SMR	2.115
## 34	Sweden	SWE	2.110
## 35	Bosnia and Herzegovina	BIH	2.091
## 36	Azerbaijan, Rep. of	AZE	2.080
## 37	Tajikistan, Rep. of	TJK	2.074
## 38	Finland	FIN	2.062
## 39	Poland, Rep. of	POL	2.056
## 40	Holy See	VAT	2.028
## 41	Slovak Rep.	SVK	2.018
## 42	Bahrain, Kingdom of	BHR	2.017
## 43	Afghanistan, Islamic Rep. of	AFG	2.012
## 44	Western Sahara	ESH	1.970
## 45	Estonia, Rep. of	EST	1.961
## 46	Lithuania	LTU	1.960
## 47	Malta	MLT	1.955
## 48	Latvia	LVA	1.953
## 49	Romania	ROU	1.949
## 50	Saint Pierre and Miquelon	SPM	1.944
## 51	Serbia, Rep. of	SRB	1.938
## 52	Ukraine	UKR	1.931
## 53	Belarus, Rep. of	BLR	1.922
## 54	Norway	NOR	1.918
## 55	United Kingdom	GBR	1.911
## 56	China, P.R.: Mainland	CHN	1.906
## 57	Montenegro	MNE	1.865
## 58	Qatar	QAT	1.855
## 59	Cabo Verde	CPV	1.853
## 60	Isle of Man	IMN	1.840
## 61	Bulgaria	BGR	1.831
## 62	Ireland	IRL	1.764
## 63	Saudi Arabia	SAU	1.745
## 64	Mauritania, Islamic Rep. of	MRT	1.711
## 65	Armenia, Rep. of	ARM	1.707
## 66	Algeria	DZA	1.688
## 67	Saint Helena	SHN	1.686
## 68	Palau, Rep. of	PLW	1.672
## 69	Korea, Rep. of	KOR	1.659
## 70	Korea, Dem. People's Rep. of	PRK	1.610
## 71	Liberia	LBR	1.556
## 72	Djibouti	DJI	1.547
## 73	Syrian Arab Rep.	SYR	1.535
## 74	Bhutan	BTN	1.523
## 75	Albania	ALB	1.518
## 76	Georgia	GEO	1.510
## 77	United Arab Emirates	ARE	1.482
## 78	Bahamas, The	BHS	1.480
## 79	Congo, Dem. Rep. of the	COD	1.480
## 80	Vanuatu	VUT	1.479
## 81	Ethiopia, The Federal Dem. Rep. of	ETH	1.455
## 82	Somalia	SOM	1.449
## 83	Brunei Darussalam	BRN	1.446
## 84	Cuba	CUB	1.437

## 85	Myanmar	MMR	1.436
## 86	China, P.R.: Hong Kong	HKG	1.432
## 87	China, P.R.: Macao	MAC	1.432
## 88	Uganda	UGA	1.416
## 89	Greenland	GRL	1.413
## 90	Philippines	PHL	1.405
## 91	Senegal	SEN	1.405
## 92	North Macedonia, Republic of	MKD	1.404
## 93	Faroe Islands	FRO	1.396
## 94	World	WLD	1.394
## 95	Pakistan	PAK	1.389
## 96	Taiwan Province of China	TWN	1.383
## 97	Lesotho, Kingdom of	LSO	1.372
## 98	Eswatini, Kingdom of	SWZ	1.371
## 99	Malaysia	MYS	1.367
## 100	Jordan	JOR	1.358
## 101	Fiji, Rep. of	FJI	1.346
## 102	Japan	JPN	1.335
## 103	Sierra Leone	SLE	1.328
## 104	Gambia, The	GMB	1.323
## 105	São Tomé and Príncipe, Dem. Rep. of	STP	1.321
## 106	New Zealand	NZL	1.319
## 107	Timor-Leste, Dem. Rep. of	TLS	1.315
## 108	Singapore	SGP	1.311
## 109	Suriname	SUR	1.296
## 110	Guinea-Bissau	GNB	1.295
## 111	Lebanon	LBN	1.290
## 112	Kenya	KEN	1.280
## 113	Canada	CAN	1.268
## 114	American Samoa	ASM	1.256
## 115	Mexico	MEX	1.249
## 116	Equatorial Guinea, Rep. of	GNQ	1.247
## 117	Congo, Rep. of	COG	1.245
## 118	Cyprus	CYP	1.237
## 119	Norfolk Island	NFK	1.225
## 120	Papua New Guinea	PNG	1.221
## 121	United States	USA	1.217
## 122	Bangladesh	BGD	1.216
## 123	Angola	AGO	1.212
## 124	Iceland	ISL	1.201
## 125	Guinea	GIN	1.195
## 126	Samoa	WSM	1.156
## 127	Jamaica	JAM	1.140
## 128	Nepal	NPL	1.132
## 129	Lao People's Dem. Rep.	LAO	1.112
## 130	Haiti	HTI	1.095
## 131	South Africa	ZAF	1.094
## 132	Indonesia	IDN	1.085
## 133	Israel	ISR	1.079
## 134	West Bank and Gaza	PSE	1.074
## 135	Guyana	GUY	1.066
## 136	Oman	OMN	1.055
## 137	Cameroon	CMR	1.044
## 138	Greece	GRC	1.040

## 139	Solomon Islands	SLB	1.038
## 140	Vietnam	VNM	1.033
## 141	Central African Rep.	CAF	1.032
## 142	Belize	BLZ	1.031
## 143	Thailand	THA	1.014
## 144	Sri Lanka	LKA	1.004
## 145	Ghana	GHA	0.996
## 146	Gabon	GAB	0.994
## 147	South Sudan, Rep. of	SSD	0.983
## 148	Dominican Rep.	DOM	0.982
## 149	Togo	TGO	0.975
## 150	Cayman Islands	CYM	0.974
## 151	Wallis and Futuna Islands	WLF	0.951
## 152	Cambodia	KHM	0.948
## 153	Turks and Caicos Islands	TCA	0.944
## 154	Brazil	BRA	0.926
## 155	Guatemala	GTM	0.916
## 156	Tanzania, United Rep. of	TZA	0.911
## 157	United States Virgin Islands	VIR	0.894
## 158	Mali	MLI	0.887
## 159	Benin	BEN	0.884
## 160	Peru	PER	0.878
## 161	British Virgin Islands	VGB	0.875
## 162	Colombia	COL	0.874
## 163	Seychelles	SYC	0.872
## 164	Madagascar, Rep. of	MDG	0.860
## 165	Puerto Rico	PRI	0.848
## 166	Tonga	TON	0.846
## 167	Nicaragua	NIC	0.842
## 168	Anguilla	AIA	0.839
## 169	Mozambique, Rep. of	MOZ	0.834
## 170	Malawi	MWI	0.816
## 171	Burkina Faso	BFA	0.802
## 172	Nigeria	NGA	0.791
## 173	India	IND	0.790
## 174	Maldives	MDV	0.784
## 175	Guadeloupe	GLP	0.773
## 176	Sudan	SDN	0.773
## 177	Antigua and Barbuda	ATG	0.770
## 178	Montserrat	MSR	0.770
## 179	St. Kitts and Nevis	KNA	0.770
## 180	Australia	AUS	0.754
## 181	Chad	TCD	0.738
## 182	Ecuador	ECU	0.736
## 183	Panama	PAN	0.728
## 184	Honduras	HND	0.723
## 185	Trinidad and Tobago	TTO	0.714
## 186	Grenada	GRD	0.698
## 187	St. Vincent and the Grenadines	VCT	0.698
## 188	St. Lucia	LCA	0.689
## 189	Dominica	DMA	0.688
## 190	Zambia	ZMB	0.686
## 191	Martinique	MTQ	0.684
## 192	Mayotte	MYT	0.678

```
## 193 Tuvalu TUV 0.663
## 194 El Salvador SLV 0.658
## 195 Egypt, Arab Rep. of EGY 0.655
## 196 Paraguay PRY 0.649
## 197 Bolivia BOL 0.644
## 198 Argentina ARG 0.643
## 199 Micronesia, Federated States of FSM 0.627
## 200 Mauritius MUS 0.595
## 201 Libya LBY 0.559
## 202 Venezuela, Rep. Bolivariana de VEN 0.533
## 203 Eritrea, The State of ERI 0.530
## 204 Comoros, Union of the COM 0.521
## 205 Cook Islands COK 0.479
## 206 Niger NER 0.428
## 207 Chile CHL 0.407
## 208 Uruguay URY 0.382
## 209 French Polynesia PYF 0.359
## 210 Namibia NAM 0.239
## 211 Marshall Islands, Rep. of the MHL 0.007
## 212 Zimbabwe ZWE -0.490
## 213 Botswana BWA -1.305
## 214 Aruba, Kingdom of the Netherlands ABW NA
## 215 Barbados BRB NA
## 216 Burundi BDI NA
## 217 Costa Rica CRI NA
## 218 Falkland Islands (Malvinas) FLK NA
## 219 Kiribati KIR NA
## 220 Nauru, Rep. of NRU NA
## 221 Niue NIU NA
## 222 Pitcairn Islands PCN NA
## 223 Rwanda RWA NA
## 224 Tokelau TKL NA
## 225 Yemen, Rep. of YEM NA
```

```
climate_10_20 <- climate_change |> select(Country, IS03, F2010:F2020) |> rowwise() |>
  mutate(mean_10_20 = mean(c_across(F2010:F2020)))

climate_10_20 |> arrange(desc(mean_10_20))
```

```
## # A tibble: 225 x 14
```

```
## # Rowwise:
```

```
## Country IS03 F2010 F2011 F2012 F2013 F2014 F2015 F2016 F2017 F2018 F2019
## <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 Russian Fe~ RUS 0.88 1.70 1.93 1.17 1.75 2.25 2.46 2.13 2.04 2.18
## 2 Belarus, R~ BLR 1.46 1.02 1.56 1.21 2.18 2.25 2.20 1.59 2.34 2.69
## 3 Estonia, R~ EST 0.554 1.30 1.47 1.18 2.45 2.38 1.94 1.50 2.41 2.22
## 4 Ukraine UKR 1.65 0.581 1.96 1.52 1.72 2.08 2.11 1.42 2.24 2.37
## 5 Latvia LVA 0.704 1.15 1.41 1.08 2.33 2.17 1.95 1.50 2.36 2.46
## 6 Slovenia, ~ SVN 0.769 1.35 1.93 1.23 2.43 2.12 2.05 1.72 2.41 2.34
## 7 Austria AUT 0.656 1.37 1.78 1.10 2.41 2.17 2.10 1.74 2.52 2.37
## 8 Bahrain, K~ BHR 2.29 1.43 1.38 1.40 1.69 2.24 1.60 2.08 2.25 2.11
## 9 Lithuania LTU 0.766 0.957 1.37 0.967 2.24 2.10 1.94 1.54 2.43 2.64
## 10 Finland FIN 0.242 1.35 1.54 1.42 2.59 2.61 2.15 1.49 2.16 1.45
## # i 215 more rows
```

```
## # i 2 more variables: F2020 <dbl>, mean_10_20 <dbl>
```

```
climate_10_20 <- climate_10_20 |> rowwise() |>  
  mutate(sum_10_20 = sum(c_across(F2010:F2020)))
```

```
climate_10_20 |> arrange(desc(sum_10_20))
```

```
## # A tibble: 225 x 15
```

```
## # Rowwise:
```

```
##   Country      IS03 F2010 F2011 F2012 F2013 F2014 F2015 F2016 F2017 F2018 F2019  
##   <chr>        <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>  
## 1 Russian Fe~ RUS  0.88  1.70  1.93  1.17  1.75  2.25  2.46  2.13  2.04  2.18  
## 2 Belarus, R~ BLR  1.46  1.02  1.56  1.21  2.18  2.25  2.20  1.59  2.34  2.69  
## 3 Estonia, R~ EST  0.554 1.30  1.47  1.18  2.45  2.38  1.94  1.50  2.41  2.22  
## 4 Ukraine     UKR  1.65  0.581 1.96  1.52  1.72  2.08  2.11  1.42  2.24  2.37  
## 5 Latvia      LVA  0.704 1.15  1.41  1.08  2.33  2.17  1.95  1.50  2.36  2.46  
## 6 Slovenia, ~ SVN  0.769 1.35  1.93  1.23  2.43  2.12  2.05  1.72  2.41  2.34  
## 7 Austria     AUT  0.656 1.37  1.78  1.10  2.41  2.17  2.10  1.74  2.52  2.37  
## 8 Bahrain, K~ BHR  2.29  1.43  1.38  1.40  1.69  2.24  1.60  2.08  2.25  2.11  
## 9 Lithuania   LTU  0.766 0.957 1.37  0.967 2.24  2.10  1.94  1.54  2.43  2.64  
## 10 Finland    FIN  0.242 1.35  1.54  1.42  2.59  2.61  2.15  1.49  2.16  1.45
```

```
## # i 215 more rows
```

```
## # i 3 more variables: F2020 <dbl>, mean_10_20 <dbl>, sum_10_20 <dbl>
```

```
pivot <- pivot_wider(human_trafficking, names_from = Year, values_from = Detected_Number)
```

```
first <- c()
```

```
last <- c()
```

```
for(i in 1:nrow(pivot)) {  
  pivmin <- pivot[i,]  
  pivmin2 <- as.vector(pivmin[-c(1,2)])  
  first[i] <- pivot[i, 2 + which(!is.na(pivmin2))[1]]  
  last[i] <- tail(pivmin2[!is.na(pivmin2)], 1)  
}
```

```
last <- as.numeric(last)
```

```
first <- as.numeric(first)
```

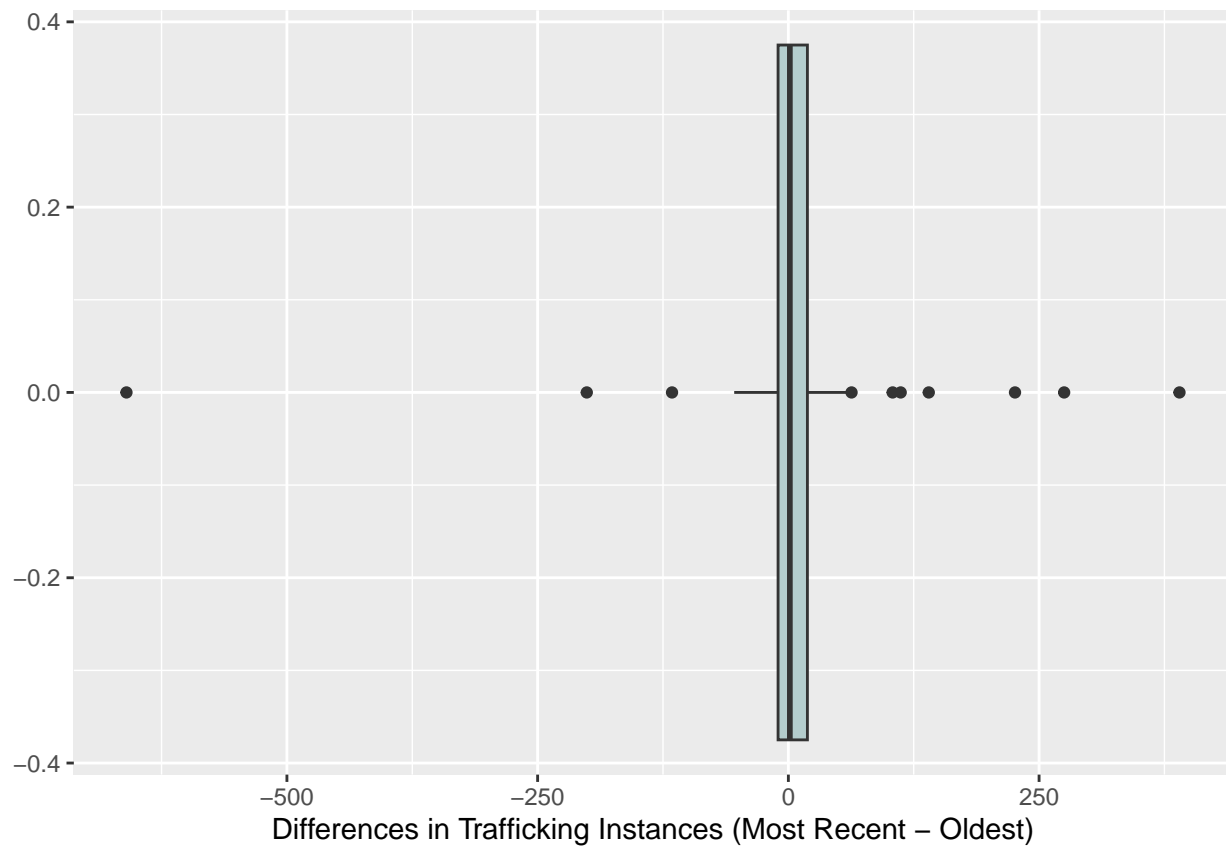
```
diffs <- last - first
```

```
diffs.noz <- if(length(which(diffs==0)!=0)) diffs[-which(diffs==0)]
```

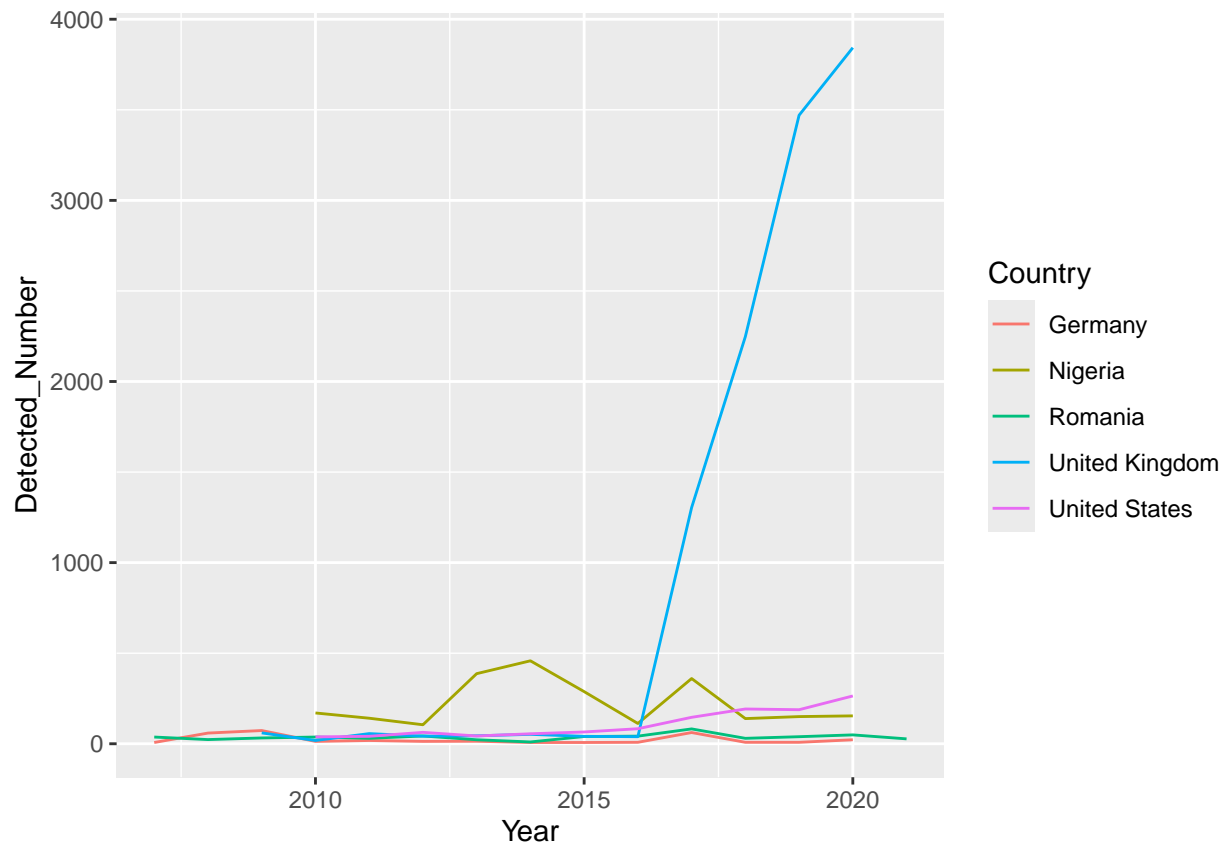
```
diffs.df <- data.frame(diffs.noz[-16])
```

```
ggplot(diffs.df, aes(x = diffs.noz[-16])) +  
  geom_boxplot(fill = "lightcyan3") +  
  xlab("Differences in Trafficking Instances (Most Recent - Oldest)")
```





```
human_trafficking %>%
  filter(Code %in% c("USA", "GBR", "DEU", "ROU", "NGA")) %>%
  drop_na() %>%
  ggplot() +
    geom_line(aes(x=Year, y = Detected_Number, col = Country,))
```



```
corr <- left_join(slavery_index, inequality, by="Country")

temp <- data.frame(year = rep(c("1990", "2000", "2010", "2020"), each = nrow(inequality)),
  values = c(inequality$Gender.Inequality.Index..1990., inequality$Gender.Inequality.I

ggplot(temp, aes(x = year, y = values, fill = year)) +
  geom_boxplot()
```

