# Code of Fallacies in the Global Water Crisis statistics $_{\rm R~code}$

Arnald Puy

## Contents

```
sensobol::load_packages(c("sensobol", "tidyverse", "data.table", "cowplot",
                        "scales", "rvest", "janitor", "fitdistrplus", "wesanderson"))
theme_AP <- function() {</pre>
 theme_bw() +
   theme(panel.grid.major = element_blank(),
         panel.grid.minor = element_blank(),
         legend.background = element_rect(fill = "transparent",
                                        color = NA),
         legend.margin=margin(0, 0, 0, 0),
         legend.box.margin=margin(-5,-5,-5),
         legend.key = element_rect(fill = "transparent",
                                 color = NA),
         strip.background = element_rect(fill = "white"),
         axis.title = element_text(size = 9),
         legend.text = element_text(size = 9),
         legend.title = element_text(size = 9),
         legend.key.width = unit(0.3, "cm"),
         legend.key.height = unit(0.3, "cm"))
}
## ----calculations, warning=FALSE-----
# Values used in the paper
precipitation_estimate <- 120000</pre>
precipitation_min <- precipitation_estimate - (precipitation_estimate * 0.1)</pre>
precipitation_max <- precipitation_estimate + (precipitation_estimate * 0.1)</pre>
land_runoff_estimate <- 46000</pre>
land_runoff_min <- land_runoff_estimate - (land_runoff_estimate * 0.1)</pre>
land_runoff_max <- land_runoff_estimate + (land_runoff_estimate * 0.1)</pre>
# Read the HTML content of the website
webpage <- read_html("https://en.wikipedia.org/wiki/List_of_countries_by_food_energy_intake#ci</pre>
# Select the table using CSS selector
table_node <- html_nodes(webpage, "table")</pre>
# Extract the table content
table_content <- data.table(html_table(table_node, header = TRUE)[[1]]) %>%
 row_to_names(row_number = 1)
```

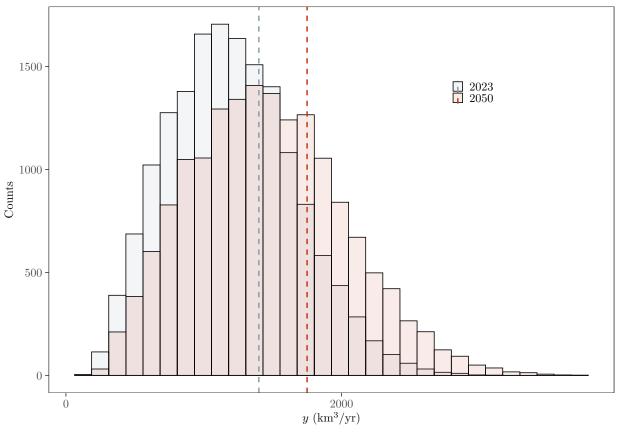
```
# Arrange and clean columns ----
old_colnames <- colnames(table_content)</pre>
new_colnames <- c("rank", "country", "kcal", "year")</pre>
setnames(table_content, old_colnames, new_colnames)
table_content[, kcal:= as.numeric(gsub(",", "", kcal))]
# Check best distribution
fg <- fitdist(table_content$kcal, "gamma")</pre>
fln <- fitdist(table_content$kcal, "lnorm")</pre>
fw <- fitdist(table content$kcal, "weibull")</pre>
# Plot goodness of fit
par(mfrow = c(2, 2))
plot.legend <- c("gamma", "lognormal", "weibull")</pre>
denscomp(list(fg, fln, fw), legendtext = plot.legend)
qqcomp(list(fg, fln, fw), legendtext = plot.legend)
cdfcomp(list(fg, fln, fw), legendtext = plot.legend)
ppcomp(list(fg, fln, fw), legendtext = plot.legend)
       Histogram and theoretical densities
                                                                        Q-Q plot
    8e-04
                                                  Empirical quantiles
                                     gamma
                                     lognormal
Density
                                                       3000
    4e-04
                                     <del>wei</del>bull
                                                                                       gamma
                                                                                     \circ lognormal
                                                       2000
    00+00
                                                                                     • weibull
           2000
                   2500
                           3000
                                    3500
                                                                       2500
                                                                           3000
                                                            1500
                                                                 2000
                                                                                  3500
                                                                                       4000 4500
                         data
                                                                     Theoretical quantiles
                                                                         P-P plot
         Empirical and theoretical CDFs
                                                  Empirical probabilities
                                                       0.4
                                     gamma
                                                                                     o gamma
                                     lognormal
                                                                                     o lognormal
                                     weibull
                                                                                     weibull
    0.0
           2000
                   2500
                            3000
                                    3500
                                                                  0.2
                                                                        0.4
                                                           0.0
                                                                               0.6
                                                                                      0.8
                                                                                             1.0
                         data
                                                                    Theoretical probabilities
```

```
# Opt for truncated weibull ------
shape <- fw$estimate[[1]]
scale <- fw$estimate[[2]]
minimum <- min(table_content$kcal)
maximum <- max(table_content$kcal)
weibull_dist <- sapply(c(minimum, maximum), function(x)</pre>
```

```
pweibull(x, shape = shape, scale = scale))
N < - 2^13
params <- c("precipitation", "et_crops", "et_vegetation", "global_consumption",</pre>
           "planetary_boundary", "W_g", "W_i", "F_i", "F_u", "$k$", "F_b",
           "$F_m$", "$F_{m_w}$", "$F_{v_w}$")
mat <- sobol_matrices(N = N, params = params)</pre>
# Uncertain parameters in Figure 1 -----
mat[, "precipitation"] <- qunif(mat[, "precipitation"], precipitation_min, precipitation_max)</pre>
mat[, "et_crops"] <- qunif(mat[, "et_crops"], 5200, 5800)</pre>
mat[, "et_vegetation"] <- qunif(mat[, "et_vegetation"], 68200, 68800)</pre>
mat[, "global_consumption"] <- qunif(mat[, "global_consumption"], 3391, 5349)</pre>
mat[, "planetary_boundary"] <- qunif(mat[, "planetary_boundary"], 4000, 6000)</pre>
# Uncertain parameters for water exceedance in 2023 ------
# Estimates groundwater consumption
mat[, "W_g"] <- qunif(mat[, "W_g"], 84, 304)</pre>
# estimates irrigation water consumption
mat[, "W_i"] <- qunif(mat[, "W_i"], 1083, 1550)</pre>
# fraction of irrigation over total water consumption
mat[, "F_i"] <- qunif(mat[, "F_i"], 0.57, 0.71)
# fraction of unsustainable irrigation
mat[, "F_u"] <- qunif(mat[, "F_u"], 0.1, 0.34)
# Uncertain parameters for water exceedance in 2025 -----
mat[, "$k$"] <- qunif(mat[, "$k$"], weibull_dist[[1]], weibull_dist[[2]])</pre>
# kilocalories
mat[, "$k$"] <- qweibull(mat[, "$k$"], shape, scale)</pre>
# Fraction of blue water
mat[, "F_b"] <- qunif(mat[, "F_b"], 0.13, 0.15)
# Fraction of diet based on meat
mat[, "$F_m$"] <- qunif(mat[, "$F_m$"], 0.01, 0.35)
# Cubic meters needed to produce 1000 kcal of meat
mat[, "$F_{m_w}$"] \leftarrow qunif(mat[, "$F_{m_w}$"], 1.08, 3.8)
```

```
# Cubic meters needed to produce 1000 kcal of vegetables
mat[, "$F_{v_w}$"] <- qunif(mat[, "$F_{v_w}$"], 0.16, 1.25)
Fv <- 1 - mat[, "$F m$"] # Fraction of diet based on vegetables
mat <- cbind(mat, F v)</pre>
fun_exceedance_2023 <- function(mat) mat[, "W_g"] + mat[, "F_i"] * mat[, "W_i"] * mat[, "F_u"]</pre>
projection_fun <- function(mat, P) {</pre>
 W \leftarrow 365 * (mat[, "$k$"] * mat[, "$F_m$"] * mat[, "$F_{m_w}$"] +
             mat[, "$k$"] * mat[, "F_v"] * mat[, "$F_{v_w}$"]) / 1000
 y <- P * mat[, "F_b"] * W
 out <- list(W, y)</pre>
 names(out) <- c("\vec{W}", "\vec{V}")</pre>
 return(out)
}
land_runoff <- mat[, "precipitation"] - mat[, "et_vegetation"] - mat[, "et_crops"]</pre>
y_2023 <- fun_exceedance_2023(mat)</pre>
exceedance_2023 <- -1 * (mat[, "planetary_boundary"] - mat[, "global_consumption"])</pre>
population \leftarrow c(8, 9.7)
y <- lapply(population, function(P) projection_fun(mat = mat, P = P))
tmp <- lapply(y, function(x) data.table(do.call(cbind, x)))</pre>
names(tmp) \leftarrow c(2023, 2050)
dt.projections <- rbindlist(tmp, idcol = "year")</pre>
dt.projections.ua <- dt.projections[, .SD[1:(2 * N)], year]
# Stats -----
melt(dt.projections.ua, measure.vars = c("W", "y")) %>%
```

```
.[, .(min = min(value),
        max = max(value),
        median = median(value)), .(year, variable)]
# Plots -----
hist.w <- dt.projections.ua[year == 2023] %>%
  ggplot(., aes(W)) +
  geom_histogram(fill = "grey", color = "black") +
  theme_AP() +
  geom_vline(xintercept = 1300, lty = 2, linewidth = 1) +
  labs(x = "$W$ (km$^3$/yr)", y = "Counts") +
  scale_x_continuous(breaks = pretty_breaks(n = 3))
dt.year \leftarrow data.table(year = c(2023, 2050), value = c(1400, 1750))
dt.year[, year:= as.factor(year)]
selected_wesanderson <- "Royal1"</pre>
hist.y <- dt.projections.ua %>%
  ggplot(., aes(y, fill = year)) +
  geom histogram(colour = "black", alpha = 0.1, position="identity") +
  labs(x = "$y$ (km$^3$/yr)", y = "Counts") +
  theme_AP() +
  scale_x_continuous(breaks = pretty_breaks(n = 2)) +
  geom_vline(data = dt.year, aes(xintercept = value, color = year, group = year),
             linetype = 2, linewidth = 1) +
  scale_fill_manual(values = wes_palette(name = selected_wesanderson, 2),
                    name = "") +
  scale_color_manual(values = wes_palette(name = selected_wesanderson, 2),
                    name = "") +
  theme(legend.position = c(0.75, 0.8))
hist.y
```



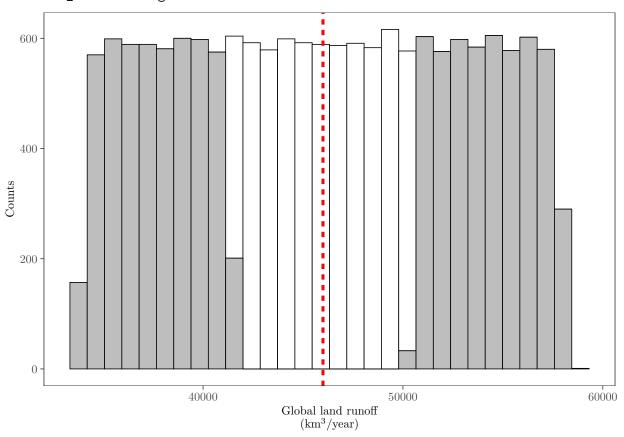
```
# Sensitivity analysis -
ind <- dt.projections[year == 2023] %>%
  .[, sobol_indices(Y = y, params = params, N = N, boot = TRUE, R = 10^3,
         first = "jansen", total = "jansen")]
plot.ind <- ind$results[parameters %in% params[10:14]] %>%
  .[!parameters == "F_b"] %>%
 ggplot(., aes(parameters, original, fill = sensitivity)) +
  geom_bar(stat = "identity", position = position_dodge(0.6), color = "black") +
  scale_y_continuous(breaks = pretty_breaks(n = 3)) +
 labs(x = "", y = "Sobol' index") +
 geom_errorbar(aes(ymin = low.ci, ymax = high.ci),
               position = position_dodge(0.6)) +
  scale_fill_discrete(name = "",
                             labels = c(expression(S[italic(i)]),
                                       expression(T[italic(i)]))) +
  theme AP() +
 theme(legend.position = c(0.8, 0.85))
food.fraction <- fread("daily-per-capita-caloric-supply.csv")</pre>
old_colnames <- colnames(food.fraction)</pre>
```

```
new_colnames <- c("entity", "code", "year", "kcal")</pre>
setnames(food.fraction, old_colnames, new_colnames)
plot.caloric <- food.fraction[year == 2018] %>%
  ggplot(., aes(kcal)) +
  geom histogram() +
  scale_x_continuous(breaks = pretty_breaks(n = 3)) +
  labs(x = "Kcal", y = "Counts") +
  theme AP()
plot_grid(plot.caloric, hist.w, hist.y, plot.ind, ncol = 2, labels = "auto")
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
\mathbf{a}_{15}
                             1000
Counts
                           Counts
                              500
      2000
              3000
                       4000
                                      1000
                                             2000
                                      W \, (\mathrm{km^3/yr})
             Kcal
                           \mathbf{d}
  1500
                                              S_i
T_i
                             0.4
                 \begin{array}{c} \boxed{\phantom{0}} \quad 2023 \\ \boxed{\phantom{0}} \quad 2050 \end{array}
                           Sobol' index
Counts
  1000
                             0.2
   500
                             0.0
                                      \dot{F_{v_w}}
              2000
           y \, (\mathrm{km^3/yr})
tmp <- split(dt.projections, dt.projections$year) %>%
  lapply(., function(x) x[, year:= NULL])
years \leftarrow c(2023, 2050)
out <- list()</pre>
for(i in 1:length(tmp)) {
  out[[i]] <- setnames(tmp[[i]], colnames(tmp[[i]]),</pre>
                        paste(colnames(tmp[[i]]), years[[i]], sep = "."))
```

```
}
dt <- do.call(cbind, out) %>%
 cbind(land_runoff, exceedance_2023, .) %>%
  .[1:(2 * N)] \%
  .[, outside:= ifelse(land_runoff < land_runoff_min |
                      land_runoff > land_runoff_max, "Yes", "No")] %>%
 .[, accessible_water_runoff:= land_runoff - 7800 - 20400] %>%
  .[, outside_runoff:= ifelse(accessible_water_runoff < 12500 |
                             accessible_water_runoff > 18500, "Yes", "No")] %>%
 .[, water_deficit:= ifelse(exceedance_2023 > 0, "Yes", "No")]
dt <- dt[, additional_water_2050:= y.2050 - y.2023] %>%
  .[, exceedance_2050:= exceedance_2023 + additional_water_2050] %>%
 .[, exceedance_by_2050:= ifelse(exceedance_2050 > 0, "Yes", "No")]
cols <- c("land runoff", "accessible water runoff", "exceedance 2023")
summary_fun = function(x) list(min = min(x), max = max(x))
dt[, lapply(.SD, summary_fun), .SDcols = (cols)]
##
     land_runoff accessible_water_runoff exceedance_2023
## 1:
        33441.89
                              5241.895
                                            -2593.442
## 2:
        58552.83
                              30352.83
                                             1333.539
tmp <- melt(dt, measure.vars = c("outside", "outside_runoff", "water_deficit")) %>%
  .[, .N, .(variable, value)]
tmp[, total:= (2^13 * 2)] \%
 .[, prop:= N / total] %>%
 print()
##
          variable value
                            N total
                                        prop
## 1:
                    No 6275 16384 0.3829956
           outside
## 2:
           outside
                   Yes 10109 16384 0.6170044
## 3: outside_runoff
                    No 4089 16384 0.2495728
## 4: outside_runoff
                   Yes 12295 16384 0.7504272
## 5: water_deficit
                    No 12571 16384 0.7672729
## 6: water_deficit
                    Yes 3813 16384 0.2327271
plot_land_runoff <- ggplot(dt, aes(land_runoff, fill = outside)) +</pre>
 geom_histogram(colour = "black") +
 scale_fill_manual(values = c("white", "grey")) +
 theme AP() +
```

```
geom_vline(xintercept = land_runoff_estimate, color = "red", lty = 2, size = 2) +
labs(x = "Global land runoff \n (km\$^3\$/year)", y = "Counts") +
theme(legend.position = "none")
plot_land_runoff
```

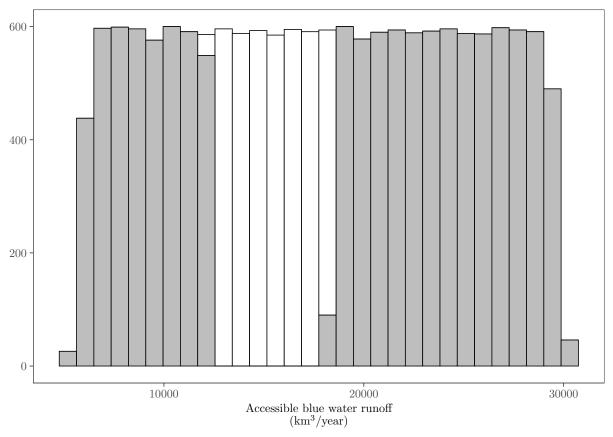
## `stat\_bin()` using `bins = 30`. Pick better value with `binwidth`.

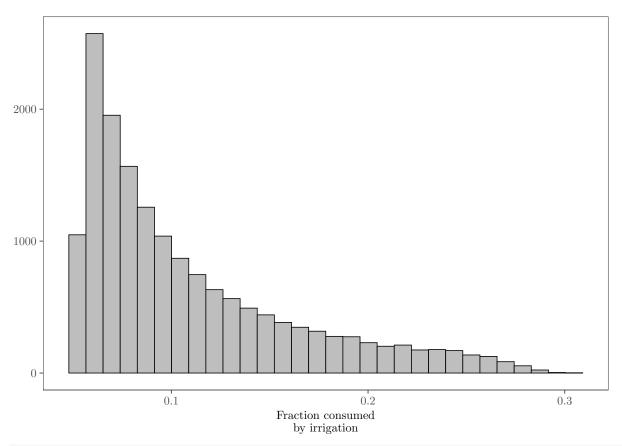


### 

```
plot_accessible_runoff <- ggplot(dt, aes(accessible_water_runoff, fill = outside_runoff)) +
    geom_histogram(colour = "black") +
    scale_fill_manual(values = c("white", "grey")) +
    theme_AP() +
    labs(x = "Accessible blue water runoff \n (km$^3$/year)", y = "") +
    theme(legend.position = "none")

plot_accessible_runoff</pre>
```





#### 

```
da <- data.table(readxl::read_xls("/Users/arnaldpuy/Documents/papers/fallacies_water_crisis/codd
da <- da[, fraction.gw:= `ICU_GW (m3 yr-1)` / `ICU (m3 yr-1)`] %>%
    .[, fraction.irr:= `ICU_SW (m3 yr-1)` / `ICU (m3 yr-1)`]

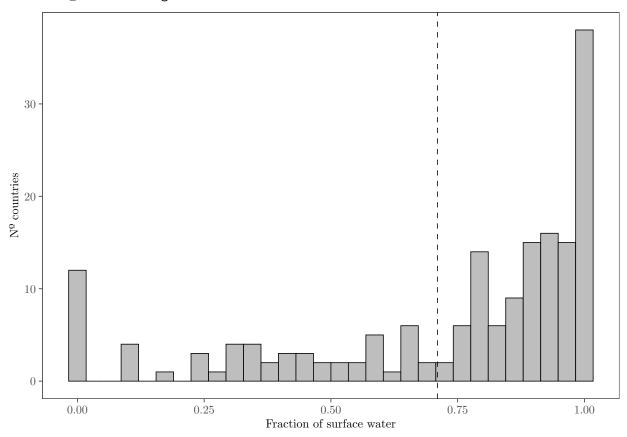
da[fraction.irr < 0.05] %>%
    .[, .(COUNTRY, fraction.irr, fraction.gw)]
```

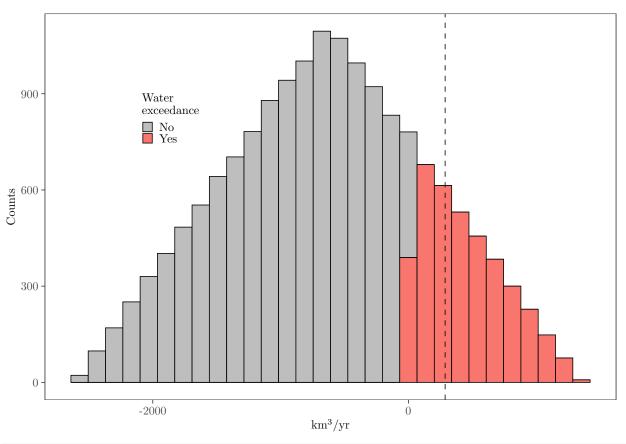
COUNTRY fraction.irr fraction.gw ## ## 1: Bahrain 0.000000000 0.9031133 2: Denmark 0.000000000 1.0000000 ## ## 3: Djibouti 0.000000000 1.000000 0.000000000 ## 4: Kuwait 0.6100000 Libyan Arab Jamahiriya 0.005554530 0.9888909 ## 5: Malta ## 6: 0.005586855 0.9944131 Montenegro 0.003619909 0.9963801 ## 7: 8: Occupied Palestinian Territory ## 0.000000000 1.000000 ## 9: Oman 0.000000000 1.000000 ## 10: Qatar 0.000000000 0.9339782 ## 11: Saudi Arabia 0.000000000 0.9700133 United Arab Emirates 0.000000000 ## 12: 1.0000000

```
fraction.surface <- ggplot(da, aes(fraction.irr)) +
  geom_histogram(color = "black", fill = "grey") +
  theme_AP() +
  geom_vline(xintercept = 0.71, lty = 2) +
  labs(x = "Fraction of surface water", y = "Nº countries")

fraction.surface</pre>
```

## `stat\_bin()` using `bins = 30`. Pick better value with `binwidth`.





## 

