Global irrigation water demands biased by unreliable irrigation efficiencies

R code

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1 Preliminary steps

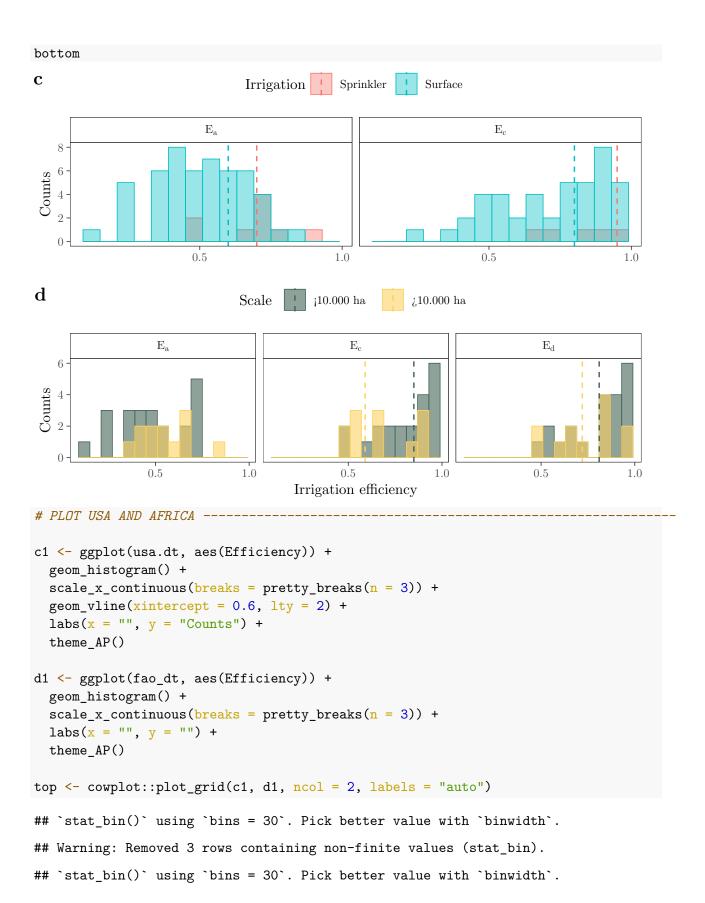
```
# Function to read in all required packages in one go:
loadPackages <- function(x) {</pre>
  for(i in x) {
    if(!require(i, character.only = TRUE)) {
      install.packages(i, dependencies = TRUE)
      library(i, character.only = TRUE)
   }
 }
}
# Load the packages
loadPackages(c("data.table", "tidyverse", "sensobol", "wesanderson",
               "cowplot", "parallel", "foreach", "doParallel",
               "countrycode", "ggridges", "scales", "overlapping",
               "sp", "rworldmap", "ncdf4", "benchmarkme"))
# Create custom theme
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.key = element_rect(fill = "transparent",
                                     color = NA),
          legend.position = "top",
          strip.background = element_rect(fill = "white"),
          plot.margin = margin(t = 0, r = 0.3, b = 0, l = 0.3, unit ="cm"))
}
# Set checkpoint
dir.create(".checkpoint")
library("checkpoint")
checkpoint("2021-08-02",
           R.version = 4.0.3,
           checkpointLocation = getwd())
```

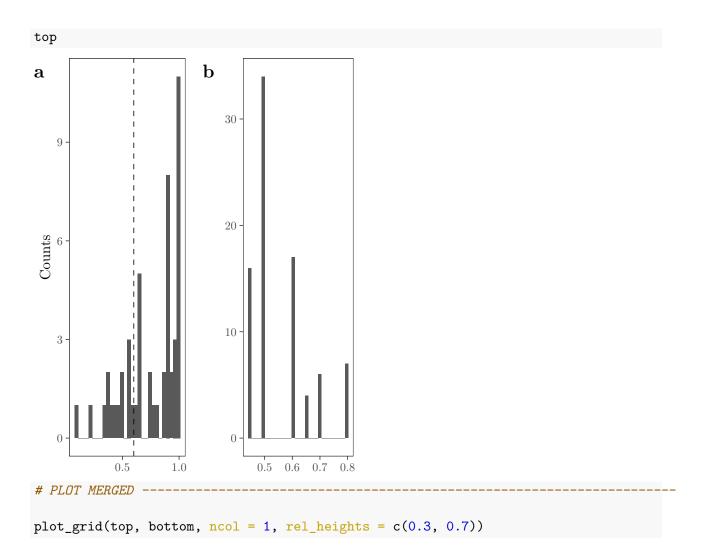
2 Read in data

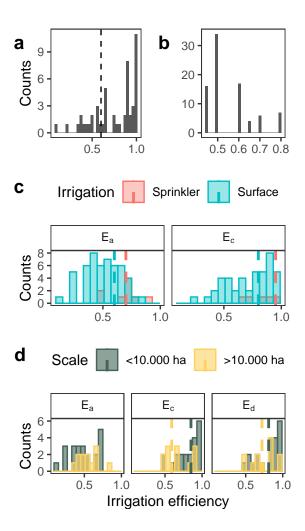
```
# READ IN DATA -----
# Rohwer data
rohwer <- fread("rohwer_data_all.csv")</pre>
rohwer[rohwer == ""] <- NA</pre>
rohwer <- rohwer[, Large_fraction:= Large_fraction / 100]</pre>
# Jager data
jager <- fread("jager_data.csv")</pre>
jager.list <- split(jager, jager$Country)</pre>
# Bos data
bos <- fread("bos_data.csv")</pre>
bos <- bos[, Scale := ifelse(Irrigated_area < 10000, "<10.000 ha", ">10.000 ha")]
# Solley data (USA)
usa.dt <- fread("usa_efficiency.csv")</pre>
usa.dt <- usa.dt[, Efficiency:= consumptive.use / total.withdrawal]
# FAO 1997 data (Irrigation potential in Africa)
fao_dt <- fread("fao_1997.csv")</pre>
fao_dt <- fao_dt[, Efficiency:= Efficiency / 100]</pre>
# Create data set with E_a values as defined by Rohwer
bos.rohwer.ea <- data.table("Irrigation" = c("Surface", "Sprinkler"),</pre>
                              "Value" = c(0.6, 0.7),
                              "variable" = "E[a]")
# Create data set with E_c values as defined by Rohwer
bos.rohwer.ec <- data.table("Irrigation" = c("Surface", "Sprinkler"),</pre>
                              "Value" = c(0.8, 0.95),
                              "variable" = "E[c]")
bos.rohwer.all <- rbind(bos.rohwer.ec, bos.rohwer.ea)</pre>
# As a function of scale
bos.rohwer.mf.ec <- data.table("Scale" = c("<10.000 ha", ">10.000 ha"),
                                 "Value" = c(0.85, 0.59),
                                 "variable" = "E[c]")
bos.rohwer.mf.ed \leftarrow data.table("Scale" = c("<10.000 ha", ">10.000 ha"),
                                 "Value" = c(0.81, 0.72),
                                 "variable" = "E[d]")
bos.rohwer.mf.all <- rbind(bos.rohwer.mf.ec, bos.rohwer.mf.ed)</pre>
```

```
bos2 <- copy(bos)</pre>
bos2 <- setnames(bos2, c("E_a", "E c", "E_d"), c("E[a]", "E[c]", "E[d]"))
# Field and conveyance efficiency -----
a <- bos2 %>%
 melt(., measure.vars = c("E[a]", "E[c]")) %>%
  ggplot(., aes(value, fill = Irrigation, color = Irrigation)) +
  geom_histogram(position = "identity", alpha = 0.4, bins = 15) +
 facet_wrap(~variable, labeller = label_parsed) +
  scale_x_continuous(breaks = pretty_breaks(n = 3)) +
  geom_vline(data = bos.rohwer.all, aes(xintercept = Value,
                                       color = Irrigation,
                                        group = variable),
            lty = 2,
            size = 1) +
  labs(x = "", y = "Counts") +
  theme_AP()
# As a function of scale -----
b <- melt(bos2, measure.vars = c("E[c]", "E[a]", "E[d]")) %>%
 na.omit() %>%
  ggplot(., aes(value, fill = Scale, color = Scale)) +
 geom histogram(bins = 15, position = "identity", alpha = 0.6) +
 labs(x = "Irrigation efficiency", y = "Counts") +
 facet_wrap(~ variable, labeller = label_parsed) +
  geom_vline(data = bos.rohwer.mf.all, aes(xintercept = Value,
                                         color = Scale,
                                         group = variable),
            lty = 2,
             size = 1) +
  scale_x_continuous(breaks = pretty_breaks(n = 3)) +
  scale_color_manual(values = wes_palette(2, name = "Chevalier1"),
                    name = "Scale",
                    labels = c("<10.000 ha", ">10.000 ha")) +
  scale_fill_manual(values = wes_palette(2, name = "Chevalier1"),
                    name = "Scale",
                    labels = c("<10.000 ha", ">10.000 ha")) +
  theme_AP()
bottom <- plot_grid(a, b, ncol = 1, labels = c("c", "d"))
```

Warning: Removed 74 rows containing non-finite values (stat_bin).







3 The model

3.1 Function to create sample matrix

```
sample_matrix_fun <- function(IFT) {
  params <- params_fun(IFT = IFT)
  mat <- sensobol::sobol_matrices(N = N, params = params)
  out <- list(params, mat)
  names(out) <- c("parameters", "matrix")
  return(out)
}</pre>
```

3.2 Define distributions

```
# DEFINE TRUNCATED DISTRIBUTIONS -
# EA SURFACE -----
Ea.surface <- bos[Irrigation == "Surface"][, .(min = min(E_a, na.rm = TRUE),</pre>
                                                 max = max(E_a, na.rm = TRUE))]
shape <- 3.502469
scale <- 0.5444373
minimum <- Ea.surface$min
maximum <- Ea.surface$max</pre>
weibull_dist <- sapply(c(minimum, maximum), function(x)</pre>
  pweibull(x, shape = shape, scale = scale))
# EC SURFACE -----
Ec.surface <- bos[Irrigation == "Surface"][, .(min = min(E_c, na.rm = TRUE),</pre>
                                                 \max = \max(E_c, na.rm = TRUE))]
shape1 <- 5.759496
shape2 <- 1.403552
minimum.beta <- Ec.surface$min
maximum.beta <- Ec.surface$max</pre>
beta_dist <- sapply(c(minimum.beta, maximum.beta), function(x)</pre>
  pbeta(x, shape1 = shape1, shape2 = shape2))
# EA SPRINKLER -----
Ea.sprinkler <- bos[Irrigation == "Sprinkler"][, .(min = min(E_a, na.rm = TRUE),</pre>
                                                 max = max(E_a, na.rm = TRUE))]
shape.spr <- 6.9913711
scale.spr <- 0.7451178
minimum.spr <- Ea.sprinkler$min
maximum.spr <- Ea.sprinkler$max</pre>
weibull_dist_spr <- sapply(c(minimum.spr, maximum.spr), function(x)</pre>
  pweibull(x, shape = shape.spr, scale = scale.spr))
# MANAGEMENT FACTOR (m) ----
```

```
shape1.m < -5.759496
shape2.m < -1.403552
minimum.m < - 0.65
maximum.m <- 1</pre>
beta_dist.m <- sapply(c(minimum.m, maximum.m), function(x)</pre>
 pbeta(x, shape1 = shape1.m, shape2 = shape2.m))
# FUNCTION TO TRANSFORM TO APPROPRIATE DISTRIBUTIONS -----
distributions_fun <- list(</pre>
  # SURFACE IRRIGATION
  # -----
  "Ea_surf" = function(x) {
   out <- qunif(x, weibull_dist[[1]], weibull_dist[[2]])</pre>
   out <- qweibull(out, shape, scale)</pre>
 },
  "Ec_surf" = function(x) {
   out <- qunif(x, beta_dist[[1]], beta_dist[[2]])</pre>
   out <- qbeta(out, shape1, shape2)</pre>
 },
  # SPRINKLER IRRIGATION
  # -----
  "Ea_sprinkler" = function(x) {
   out <- qunif(x, weibull_dist_spr[[1]], weibull_dist_spr[[2]])</pre>
   out <- qweibull(out, shape.spr, scale.spr)</pre>
 },
  "Ec_sprinkler" = function(x) qunif(x, 0.64, 0.96),
  # MICRO (DRIP) IRRIGATION
  "Ea_micro" = function(x) out <- qunif(x, 0.75, 0.95),
  "Ec_micro" = function(x) out <- qunif(x, 0.9, 0.95),
  # PROPORTION LARGE
  # -----
  "Proportion_large" = function(x) x,
```

3.3 Uncertainty in the proportion of large-scale irrigated areas

```
# DEFINE THE UNCERTAINTY IN THE LARGE FRACTION AT THE COUNTRY LEVEL -----
eff10 <- fread("efficiency 10.csv")</pre>
eff30 <- fread("efficiency_30.csv")</pre>
eff100 <- fread("efficiency_100.csv")</pre>
# CHECK WHICH COUNTRIES FROM ROHWER ET AL. ARE MISSING IN THE
# LARGE-SCALE IRRIGATED AREA DATASETS ------
countryDiff <- setdiff(rohwer$Country, eff100$Country)</pre>
countryMissing <- data.table(Country = rep(countryDiff, each = 12),</pre>
                           X1 = rep(1:4, each = 3),
                           X2 = rep(1:3, times = 4),
                           Proportion large = 0)
# ARRANGE DATASETS -------
largescale.dt <- rbind(eff10, eff30, eff100) %>%
 melt(., measure.vars = 3:6, variable.name = "X1",
      value.name = "Proportion_large") %>%
 .[, Code:= NULL] %>%
 setcolorder(., c("Country", "X1", "X2", "Proportion_large")) %>%
```

3.4 Function to create sample matrix and transfrom to appropriate distributions

3.5 Define the model

```
# FULL MODEL ----
full_model <- function(IFT, Country, sample.size, R) {
  country.differences <- setdiff(rohwer$Country, jager$Country)
  tmp <- full_sample_matrix(IFT = IFT, Country = Country)
  mat <- tmp$matrix

if(IFT == "Surface" | IFT == "Mixed" | IFT == "Jager") {
   X1 <- mat[, "X1"]
   X2 <- mat[, "X2"]
   index <- paste(Country, X1, X2, sep = ".")</pre>
```

```
Proportion_large <- triggers.dt[index][, Proportion_large]</pre>
}
if(IFT == "Surface") {
  Mf <- mat[, "m"] - mat[, "r_L"] * Proportion_large</pre>
  y <- mat[, "Ea_surf"] * mat[, "Ec_surf"] * Mf</pre>
} else if(IFT == "Sprinkler") {
  Mf <- mat[, "m"]</pre>
  y <- mat[, "Ea_sprinkler"] * mat[, "Ec_sprinkler"] * Mf
} else if(IFT == "Mixed") {
  Mf.surf <- mat[, "m"] - mat[, "r_L"] * Proportion_large</pre>
  y.surf <- mat[, "Ea_surf"] * mat[, "Ec_surf"] * Mf.surf</pre>
  Mf.sprink <- mat[, "m"]</pre>
  y.sprink <- mat[, "Ea_sprinkler"] * mat[, "Ec_sprinkler"] * Mf.sprink</pre>
  y \leftarrow 0.5 * y.surf + 0.5 * y.sprink
} else if(IFT == "Micro") {
  Mf <- mat[, "m"]</pre>
  y <- mat[, "Ea_micro"] * mat[, "Ec_micro"] * Mf
} else if(IFT == "Jager") {
  if(Country %in% country.differences == TRUE) {
    next
  }
  Mf.surf <- mat[, "m"] - mat[, "r_L"] * Proportion_large</pre>
  y.surf <- mat[, "Ea_surf"] * mat[, "Ec_surf"] * Mf.surf</pre>
  Mf.spr <- mat[, "m"]</pre>
  y.spr <- mat[, "Ea_sprinkler"] * mat[, "Ec_sprinkler"] * Mf.spr</pre>
  Mf.micro <- mat[, "m"]</pre>
  y.micro <- mat[, "Ea_micro"] * mat[, "Ec_micro"] * Mf.micro</pre>
  y <- jager.list[[Country]]$Surface_fraction * y.surf +</pre>
    jager.list[[Country]]$Sprinkler_fraction * y.spr +
    jager.list[[Country]]$Drip_fraction * y.micro
```

3.6 Define settings

```
# DEFINE SETTINGS -----
N <- 2^14
R <- 10^2
list_continents <- list(c("Africa", "Asia"), c("Americas", "Europe"))</pre>
```

3.7 Run the model in parallel

3.8 Extract model output

```
# EXTRACT MODEL OUTPUT -----
names(y) <- c("Rohwer et al. 2007", "Jägermeyr et al. 2015")

output <- tmp <- list()
for(i in names(y)) {
  output[[i]] <- lapply(y[[i]], function(x) x[["output"]][1:(2 * N)])

if(i == "Rohwer et al. 2007") {</pre>
```

```
names(output[[i]]) <- rohwer$Country</pre>
  } else if(i == "Jägermeyr et al. 2015") {
    names(output[[i]]) <- new.rohwer$Country</pre>
  tmp[[i]] <- lapply(output[[i]], data.table) %>%
    rbindlist(., idcol = "Country")
  if(i == "Rohwer et al. 2007") {
    tmp[[i]] <- merge(tmp[[i]], rohwer[, .(Country, IFT)], all.x = TRUE) %>%
      .[, IFT:= factor(IFT, levels = c("Surface", "Sprinkler", "Micro", "Mixed"))]
  } else if(i == "Jägermeyr et al. 2015") {
    tmp[[i]] <- tmp[[i]][, IFT:= "Jager"]</pre>
  }
 tmp[[i]] <- tmp[[i]][, Continent:= countrycode(tmp[[i]][, Country],</pre>
                                                   origin = "country.name",
                                                   destination = "continent")]
}
## Warning in countrycode_convert(sourcevar = sourcevar, origin = origin, destination = dest,
## Warning in countrycode_convert(sourcevar = sourcevar, origin = origin, destination = dest,
uncertainty.dt <- rbindlist(tmp, idcol = "Approach")</pre>
uncertainty.dt <- uncertainty.dt[, Study:= ifelse(IFT == "Jager",
                                                    "Jägermeyr et al. approach",
                                                    "Rohwer et al. approach")]
# FXPORT UNCERTAINTY IN TRRIGATION FFFICIENCY -----
fwrite(uncertainty.dt, "uncertainty.dt.csv")
```

4 Uncertainty analysis

4.1 Coefficient of variation

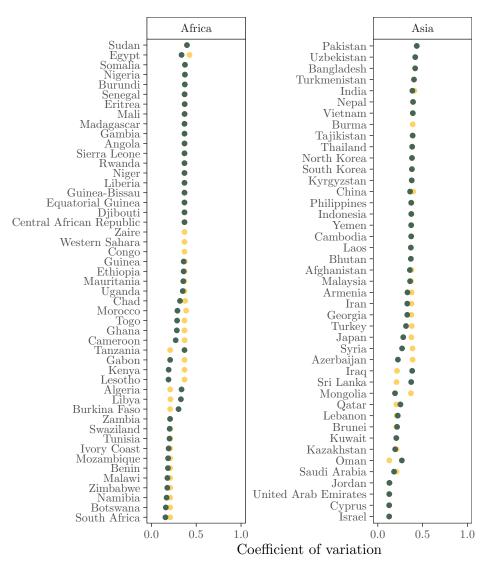
```
dd <- list()</pre>
for (i in 1:length(list_continents)) {
  dd[[i]] <- ggplot(cv.dt[Continent %in% list_continents[[i]]],</pre>
                    aes(reorder(Country, cv), cv, color = Approach)) +
    geom point() +
    scale_color_manual(values = wes_palette("Chevalier1"),
                       labels = c("Jägermeyr et al. approach",
                                   "Rohwer et al. approach")) +
    labs(y = "Coefficient of variation",
         x = "") +
    facet_wrap(~Continent, scales = "free_y") +
    scale_y_continuous(limits = c(0, 1),
                       breaks = pretty_breaks(n = 3)) +
    coord_flip() +
    theme_AP() +
    guides(color = guide_legend(nrow = 2, byrow = TRUE))
}
dd
```

[[1]]

• Jägermeyr et al. approach

Approach

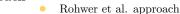
Rohwer et al. approach

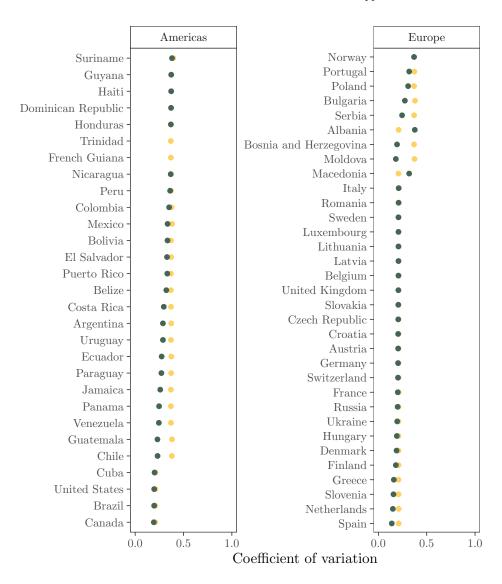


##

[[2]]

Jägermeyr et al. approach Approach





Ranges 4.2

```
# COMPUTE RANGES
calc <- uncertainty.dt[, .(min = min(V1), max = max(V1)), .(Continent, Country)] %>%
  .[, .(range = max - min), .(Continent, Country)] %>%
  .[order(range)]
print(calc, n = Inf)
```

Continent Country range ## 1: Asia Cyprus 0.4639176 2: United Arab Emirates 0.4854683 ## Asia

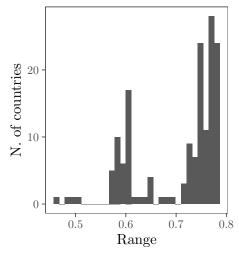
	_			
##	3:	Asia		0.4944147
##	4:	Asia		0.5042251
##	5:		Africa Tunisia	
##		6: Asia Saudi Arabia		
##	7:	Europe		0.5721363
##	8:	Africa		0.5726990
##	9:	Americas	United States	
##	10:	Europe	Netherlands	
##	11:	Americas		0.5818560
##	12:	Africa	Mozambique	
##	13:	Americas		0.5855978
##	14:	Africa		0.5856878
##	15:	Africa		0.5856878
##	16:	Africa		0.5856878
##	17:	Asia		0.5864389
##	18:	Europe		0.5869800
##	19:	Africa	South Africa	
##	20:	Africa		0.5890368
##	21:	Asia	Kazakhstan	
##	22:	Africa	Ivory Coast	
##	23:	Asia		0.5922976
##	24:	Americas		0.5944970
##	25:	Europe	-	0.5984780
##	26:	Asia		0.5998517
##	27:	Europe		0.6014893
##	28:	Europe	_	0.6014893
##	29:	<na></na>	· ·	0.6014893
##	30:	Europe		0.6014893
##	31:	Europe		0.6014893
##	32:	Europe		0.6014893
##	33:	Europe	Greece	0.6014893
##	34:	Europe	Hungary	0.6014893
##	35:	Europe	Latvia	0.6014893
##	36:	Europe		0.6014893
##	37:	Europe	Luxembourg	
##	38:	Africa		0.6014893
##	39:	Europe	Slovakia	0.6014893
##	40:	Europe		0.6014893
##	41:	Europe	Switzerland	0.6014893
##	42:	Africa	Botswana	0.6071124
##	43:	Europe	Ukraine	0.6122925
##	44:	Asia	Qatar	0.6222104
##	45:	Europe	Czech Republic	0.6394465
##	46:	Europe	Croatia	0.6449088
##	47:	Europe	Russia	0.6468832
##	48:	Europe	United Kingdom	0.6527024
##	49:	Europe	Slovenia	0.6529393
##	50:	Europe	Romania	0.6658015

##	51:	Africa	Burkina Faso	0 6010500
##	52:	Africa		0.6956438
##	53:	Asia	Azerbaijan	
##	54:	Asia		
	55:	Africa	-	0.7165713
##			_	0.7176049
##	56:	Asia	_	0.7217608
##	57: 58:	Africa		0.7219235
##		Oceania	New Zealand	
##	59:	Asia	•	0.7249771
##	60:	Asia		0.7257428
##	61:	Oceania	Australia	
##	62:	Africa		0.7287327
##	63:	Americas		0.7295769
##	64:	Asia		0.7310074
##	65:	Asia		0.7320971
##	66:	Americas		0.7341855
##	67:	Asia		0.7368912
##	68:	Asia	_	0.7370818
##	69:	Europe	•	0.7393857
##	70:	Asia		0.7407613
##	71:	Asia	Afghanistan	
##	72:	Asia	Philippines	
##	73:	Americas	_	0.7460277
##	74:	Asia	Turkmenistan	0.7470244
##	75:	Asia	Vietnam	0.7474112
##	76:	Asia	Pakistan	0.7475370
##	77:	Americas	Ecuador	0.7475424
##	78:	Europe	Macedonia	0.7477843
##	79:	Asia	North Korea	0.7482417
##	80:	Americas	Peru	0.7485998
##	81:	Asia	Uzbekistan	0.7486592
##	82:	Europe	Moldova	0.7488747
##	83:	Africa	Chad	0.7498686
##	84:	Americas	Paraguay	0.7498686
##	85:	Americas	Uruguay	0.7498686
##	86:	Africa	Sudan	0.7498884
##	87:	Asia	Malaysia	0.7503642
##	88:	Europe	Portugal	0.7505188
##	89:	Asia	Thailand	0.7505896
##	90:	Americas	Bolivia	0.7506488
##	91:	Asia	Bangladesh	0.7508831
##	92:	Americas	Venezuela	0.7512531
##	93:	Asia	Sri Lanka	0.7513038
##	94:	Asia	Nepal	0.7520751
##	95:	Africa	-	0.7527058
##	96:	Asia	Kyrgyzstan	
##	97:	Asia		0.7566068
##	98:	Americas		0.7568347
			,	

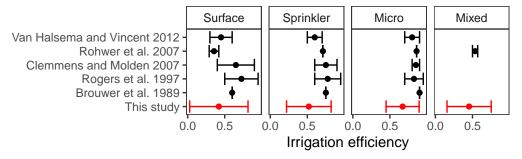
```
##
    99:
                                  Tajikistan 0.7569981
             Asia
## 100:
             Asia
                                     Georgia 0.7580011
## 101:
                                 South Korea 0.7584425
             Asia
## 102:
                                    Cambodia 0.7615212
             Asia
## 103:
           Africa
                                    Ethiopia 0.7628903
                                     Albania 0.7632624
## 104:
           Europe
## 105:
             Asia
                                       Yemen 0.7644014
## 106:
           Africa
                                    Tanzania 0.7646283
## 107:
           Africa
                                     Nigeria 0.7656979
## 108:
         Americas
                                       Haiti 0.7658826
## 109:
                                        Oman 0.7711095
             Asia
## 110:
         Americas
                                      Belize 0.7728079
## 111:
                     Bosnia and Herzegovina 0.7728079
           Europe
## 112:
           Africa
                                       Congo 0.7728079
## 113:
         Americas
                              French Guiana 0.7728079
## 114:
                                       Gabon 0.7728079
           Africa
## 115:
           Africa
                                       Ghana 0.7728079
## 116:
                                   Guatemala 0.7728079
         Americas
## 117:
           Africa
                                      Guinea 0.7728079
## 118:
                                     Jamaica 0.7728079
         Americas
## 119:
           Africa
                                     Lesotho 0.7728079
## 120:
           Africa
                                  Mauritania 0.7728079
## 121:
         Americas
                                      Panama 0.7728079
## 122:
                           Papua New Guinea 0.7728079
          Oceania
## 123:
                                      Poland 0.7728079
           Europe
## 124:
                                 Puerto Rico 0.7728079
         Americas
## 125:
           Africa
                                        Togo 0.7728079
## 126:
         Americas
                                    Trinidad 0.7728079
## 127:
           Africa
                                      Uganda 0.7728079
## 128:
           Africa
                             Western Sahara 0.7728079
## 129:
                                       Zaire 0.7728079
           Africa
## 130:
                                      Serbia 0.7733555
           Europe
## 131:
         Americas
                                 El Salvador 0.7743265
## 132:
           Africa
                                       Kenya 0.7743758
## 133:
         Americas
                                  Costa Rica 0.7745177
## 134:
             Asia
                                    Mongolia 0.7749708
## 135:
         Americas
                                    Colombia 0.7779004
## 136:
           Africa
                                     Senegal 0.7793409
## 137:
         Americas
                                   Nicaragua 0.7803721
## 138:
           Africa
                                  Madagascar 0.7806578
## 139:
                                        Mali 0.7808877
           Africa
## 140:
           Africa
                                      Angola 0.7814033
## 141:
                                      Bhutan 0.7814033
             Asia
## 142:
           Africa
                                     Burundi 0.7814033
## 143:
           Africa Central African Republic 0.7814033
## 144:
           Africa
                                    Djibouti 0.7814033
## 145:
         Americas
                         Dominican Republic 0.7814033
## 146:
                          Equatorial Guinea 0.7814033
           Africa
```

```
## 147:
           Africa
                                    Eritrea 0.7814033
## 148:
                                     Gambia 0.7814033
           Africa
                              Guinea-Bissau 0.7814033
## 149:
           Africa
## 150:
         Americas
                                   Honduras 0.7814033
## 151:
             Asia
                                       Laos 0.7814033
## 152:
           Africa
                                    Liberia 0.7814033
## 153:
           Africa
                                      Niger 0.7814033
## 154:
           Europe
                                     Norway 0.7814033
## 155:
           Africa
                                     Rwanda 0.7814033
## 156:
                               Sierra Leone 0.7814033
           Africa
## 157:
                                    Somalia 0.7814033
           Africa
## 158:
         Americas
                                   Suriname 0.7830037
        Continent
##
                                    Country
                                                 range
ggplot(calc, aes(range)) +
  geom_histogram() +
  labs(x = "Range", y = "N. of countries") +
  theme_AP()
```

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.



```
"Van Halsema and Vincent 2012"))] %>%
na.omit() %>%
ggplot(., aes(mean.value, Study, color = ifelse(Study == "This study", "red", "black"))) +
geom_point() +
scale_x_continuous(breaks = pretty_breaks(n = 3)) +
geom_errorbar(aes(xmin = lower, xmax = higher)) +
scale_color_identity() +
facet_wrap(~IFT, ncol = 4) +
labs(x = "Irrigation efficiency", y = "") +
theme_AP()
```



4.3 Overlap between irrigation efficiencies

```
# CHECK OVERLAP -
dd <- uncertainty.dt[!Continent == "Oceania"][Study == "Rohwer et al. approach"] %>%
  split(., .$Continent, drop = TRUE)
overlap.dt <- lapply(dd, function(x) split(x, x$IFT, drop = TRUE)) %>%
  lapply(., function(x) lapply(x, function(y) y[, V1])) %>%
  lapply(., function(x) overlap(x)$0V)
overlap.dt
## $Africa
## Surface-Sprinkler
                         Surface-Mixed
                                         Sprinkler-Mixed
           0.3329115
##
                             0.5084454
                                               0.5255184
##
## $Americas
## Surface-Mixed
##
       0.5084078
##
## $Asia
## Surface-Micro Surface-Mixed
                                 Micro-Mixed
      0.05314604
                    0.42996793
                                  0.07870342
##
##
## $Europe
## Surface-Sprinkler
                         Surface-Mixed
                                         Sprinkler-Mixed
##
          0.3372350
                             0.5362841
                                               0.4917646
```

```
ff <- uncertainty.dt[!Continent == "Oceania"] %>%
  .[Country %in% intersect(rohwer[, Country], jager[, Country])] %>%
  split(., .$Country, drop = TRUE) %>%
  lapply(., function(x) split(x, x$Approach, drop = TRUE)) %>%
  lapply(., function(x) lapply(x, function(y) y[, V1])) %>%
  lapply(., function(x) overlap(x)$0V) %>%
  lapply(., data.table) %>%
 rbindlist(., idcol = "Country") %>%
  .[, Continent:= countrycode(.[, Country],
                             origin = "country.name",
                             destination = "continent")]
ff[, .(median = median(V1)), Continent]
##
      Continent
                   median
## 1:
           Asia 0.9198489
## 2:
        Europe 0.7806813
## 3:
         Africa 0.9335912
## 4: Americas 0.8630331
list_continents <- list(c("Africa", "Asia"), c("Americas", "Europe"))</pre>
# PLOT OVERLAP -----
dd <- list()</pre>
for(i in 1:length(list_continents)) {
  dd[[i]] <- ff[Continent %in% list_continents[[i]]] %>%
    ggplot(., aes(reorder(Country, V1), V1)) +
    geom_point() +
    scale_color_discrete(name = "GM") +
    labs(y = "Overlap", x = "") +
    facet_wrap(~Continent, scales = "free_y") +
    coord flip() +
    theme_AP()
}
dd
```

[[1]]



[[2]]

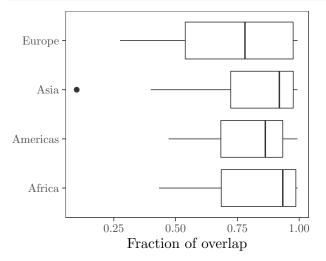


PLOT OVERLAP AS HISTOGRAMS AND BOXPLOTS

```
ggplot(ff, aes(V1)) +
  geom_histogram() +
  facet_wrap(~Continent) +
  scale_x_continuous(breaks = pretty_breaks(n = 3)) +
  theme_AP() +
  labs(x = "Fraction of overlap", y = "Nº of countries")
```

```
Africa
                                               Americas
    15
    10
N° of countries
     5
                    Asia
                                                 Europe
    15
    10
     5
     0
                   0.5
                                  1.0
                                                  0.5
                        Fraction of overlap
```

```
ggplot(ff, aes(Continent, V1)) +
  geom_boxplot() +
  coord_flip() +
  theme_AP() +
  labs(y = "Fraction of overlap", x = "")
```



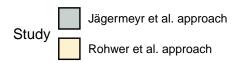
```
# CHECK CORRESPONDENCE BETWEEN SHARES OF IFT AND PREDOMINANT TECHNOLOGY -----
```

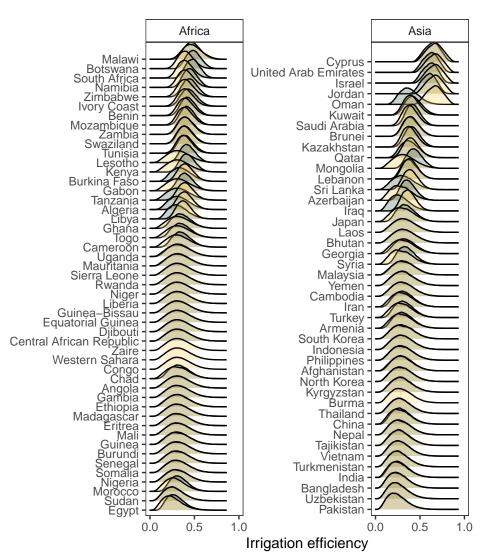
```
# Retrieve countries where overlap is <0.3
merge(jager, rohwer, by = c("Country")) %>%
    .[Country %in% ff[V1 < 0.3][, Country]] %>%
    .[, .(Country, Surface_fraction, Sprinkler_fraction, Drip_fraction, IFT)]
```

```
##
       Country Surface_fraction Sprinkler_fraction Drip_fraction
                                                                         IFT
          {\tt Oman}
                                                 0.113
## 1:
                            0.793
                                                                0.094 Micro
## 2: Slovenia
                            0.000
                                                 0.693
                                                                0.307 Mixed
## 3:
         Spain
                            0.297
                                                 0.226
                                                                0.478 Mixed
```

```
# PLOT UNCERTAINTY -----
gg <- list()
```

```
for (i in 1:length(list_continents)) {
  gg[[i]] <- ggplot(uncertainty.dt[Continent %in% list_continents[[i]]],</pre>
                    aes(x = V1, y = fct_reorder(Country, V1), fill = Study)) +
    geom_density_ridges(scale = 2, alpha = 0.3) +
    labs(x = "Irrigation efficiency", y = "") +
    facet_wrap(~Continent, scales = "free") +
    scale_x_continuous(breaks = pretty_breaks(n = 3),
                       limits = c(0, 1)) +
    scale_fill_manual(values = wes_palette("Chevalier1")) +
    theme_AP() +
    theme(legend.position = "top") +
    guides(fill = guide_legend(nrow = 2, byrow = TRUE))
}
# MERGE PLOTS -----
gg
## [[1]]
## Picking joint bandwidth of 0.0121
## Picking joint bandwidth of 0.0118
```

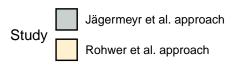


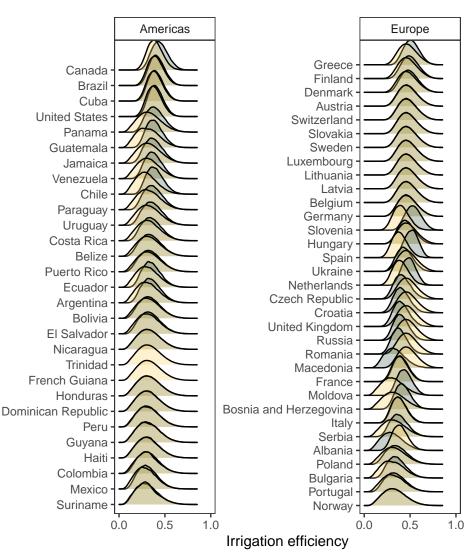


[[2]]

Picking joint bandwidth of 0.0123

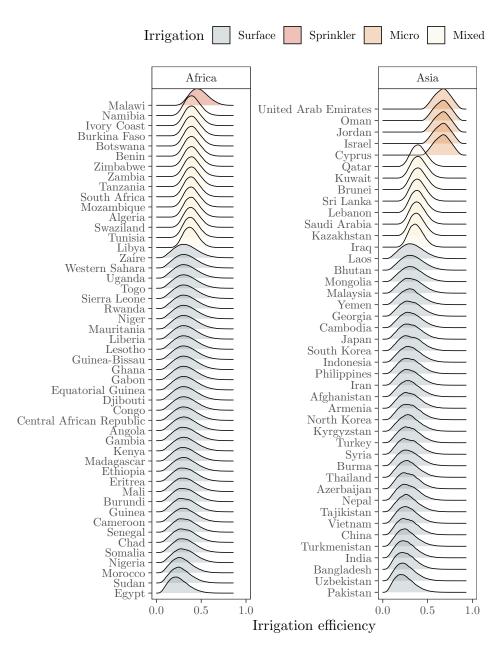
Picking joint bandwidth of 0.0109





PLOT UNCERTAINTY IN EACH IRRIGATION TECHNOLOGY -----

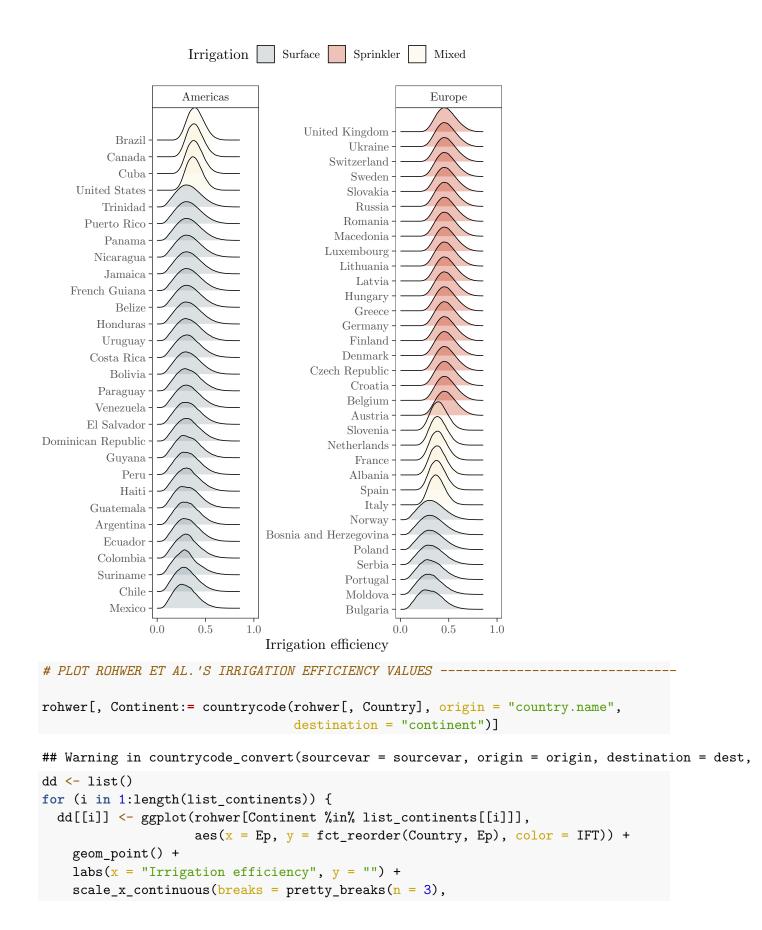
```
## [[1]]
## Picking joint bandwidth of 0.0123
## Picking joint bandwidth of 0.0119
```



[[2]]

Picking joint bandwidth of 0.0129

Picking joint bandwidth of 0.0113

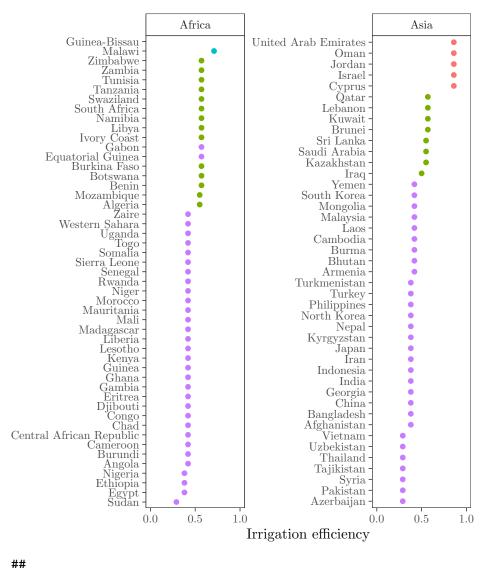


```
limits = c(0, 1)) +
facet_wrap(~Continent, scales = "free") +
scale_color_discrete(name = "Irrigation") +
theme_AP()
}
dd
```

[[1]]

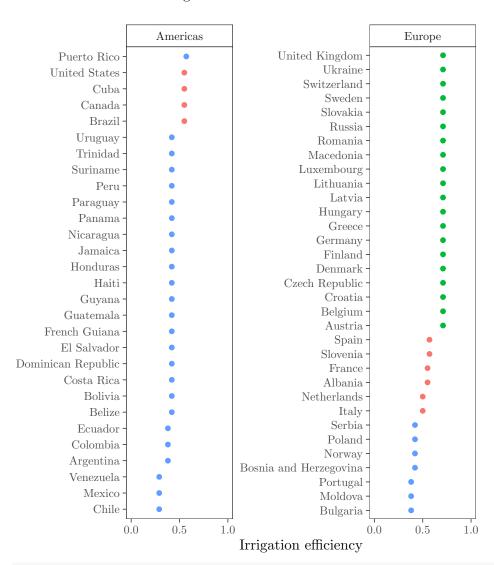
Warning: Removed 1 rows containing missing values (geom_point).



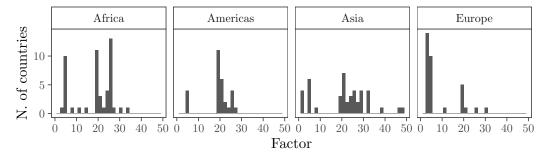


[[2]]





`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.



```
# Number of countries whose irrigation water withdrawals fluctuate a factor of x
# due to uncertainty in irrigation efficiency
factor_unc %>%
    .[, factor:= floor(max / min)] %>%
    .[, .(number.countries = .N), factor] %>%
    .[order(factor)] %>%
    print()
```

##		factor	number.countries
##	1:	2	4
##	2:	3	15
##	3:	4	28
##	4:	5	2
##	5:	7	1
##	6:	8	1
##	7:	11	2
##	8:	13	1
##	9:	19	29
##	10:	20	16
##	11:	21	4
##	12:	22	3
##	13:	23	3
##	14:	24	5
##	15:	25	20
##	16:	26	3
##	17:	27	3
##	18:	28	3
##	19:	29	1
##	20:	30	1
##	21:	31	3
##	22:	32	2
##	23:	33	1
##	24:	39	1
##	25:	47	1
##	26:	48	1
##		${\tt factor}$	${\tt number.countries}$

4.4 Retrieve data from ISIMIP

```
# FUNCTIONS TO EXTRACT DATA FROM .NC FILES -
coords2country = function(points) {
  countriesSP <- rworldmap::getMap(resolution = 'low')</pre>
 pointsSP = sp::SpatialPoints(points, proj4string=CRS(proj4string(countriesSP)))
  indices = sp::over(pointsSP, countriesSP)
  indices$ADMIN
}
# Function to load and extract data from .nc files from ISIMIP
open_nc_files <- function(file, dname, selected.years, vec) {
 ncin <- nc_open(file)</pre>
  # get longitude, latitude, time
 lon <- ncvar_get(ncin, "lon")</pre>
 lat <- ncvar_get(ncin, "lat")</pre>
  # Get variable
 tmp array <- ncvar get(ncin, dname)</pre>
 m <- lapply(selected.years, function(x) vec[[x]])</pre>
  out <- lapply(m, function(x) {
    tmp_slice <- lapply(x, function(y) tmp_array[, , y])</pre>
    # create dataframe -- reshape data
    # matrix (nlon*nlat rows by 2 cols) of lons and lats
    lonlat <- as.matrix(expand.grid(lon,lat))</pre>
    # vector of `tmp` values
    tmp_vec <- lapply(tmp_slice, function(x) as.vector(x))</pre>
    # create dataframe and add names
    tmp_df01 <- lapply(tmp_vec, function(x) data.frame(cbind(lonlat, x)))</pre>
    names(tmp_df01) <- x
    da <- lapply(tmp_df01, data.table) %>%
      rbindlist(., idcol = "month") %>%
      na.omit()
    # Convert coordinates to country
    Country <- coords2country(da[1:nrow(da), 2:3])
    df <- cbind(Country, da)</pre>
    setDT(df)
    out <- na.omit(df)[, .(Water.Withdrawn = sum(x)), Country]</pre>
    out[, Water.Withdrawn:= Water.Withdrawn * 10000]
    out[, Continent:= countrycode(out[, Country],
                                    origin = "country.name",
                                    destination = "continent")] %>%
      .[, Code:= countrycode(out[, Country],
                              origin = "country.name",
                              destination = "un")] %>%
      .[, Country:= countrycode(out[, Code],
                                 origin = "un",
```

```
destination = "country.name")] %>%
      .[!Continent == "Oceania"]
    setcolorder(out, c("Country", "Continent", "Code", "Water.Withdrawn"))
 })
 return(out)
# READ IN NC FILES -----
# Define settings
vecs <- 1:((2010 - 1970) * 12)
vec <- split(vecs, ceiling(seq_along(vecs) / 12))</pre>
names(vec) <- 1971:2010
selected.years <- "2010"
dname <- "pirrww"</pre>
files <- list("h08_wfdei nobc_hist_varsoc_co2_pirrww_global_monthly_1971_2010.nc",
              "pcr-globwb_wfdei_nobc_hist_varsoc_co2_pirrww_global_monthly_1971_2010.nc",
              "lpjml_wfdei_nobc_hist_varsoc_co2_pirrww_global_monthly_1971_2010.nc",
              "watergap2_wfdei_nobc_hist_varsoc_co2_pirrww_global_monthly_1971_2010.nc")
names.isimip <- c("HO8", "PCR-GLOBWB", "LPJmL", "WaterGap")</pre>
isimip.dt <- mclapply(files, function(x)</pre>
  open nc files(file = x, dname = dname, selected.years = selected.years, vec = vec),
mc.cores = detectCores() * 0.75)
# EXTRACT CORRECTIVE COEFFICIENTS FOR IRRIGATION EFFICIENCY FOR LPJML ---------
ncin <- nc_open("irrigation_project_efficiencies.nc")</pre>
lon <- ncvar_get(ncin, "lon")</pre>
lat <- ncvar_get(ncin, "lat")</pre>
tmp_array <- ncvar_get(ncin)</pre>
lonlat <- as.matrix(expand.grid(lon,lat))</pre>
da <- na.omit(cbind(lonlat, as.vector(tmp_array))) %>%
  data.frame() %>%
  na.omit()
Country <- coords2country(da[1:nrow(da), 1:2])</pre>
lpjml_efficiencies <- cbind(Country, da) %>%
 na.omit() %>%
  data.table() %>%
  [, (Ep = mean(V3)), Country]
# ARRANGE NC FILES ---
names(isimip.dt) <- names.isimip</pre>
isimip.dt <- lapply(isimip.dt, function(x) rbindlist(x)) %>%
```

```
rbindlist(., idcol = "Model") %>%
 na.omit() %>%
  # To correct for duplicate country in Cyprus
  .[, .(Water.Withdrawn = mean(Water.Withdrawn)), .(Model, Country, Continent, Code)]
lpjml_harmonized <- merge(isimip.dt[Model == "LPJmL"], lpjml_efficiencies, all.x = TRUE) %>%
  .[, Water.Withdrawn:= Water.Withdrawn * Ep] %>%
  .[, Ep:= NULL]
isimip.dt <- rbind(isimip.dt[!Model == "LPJmL"], lpjml_harmonized)</pre>
fwrite(isimip.dt, "isimip.dt")
# MERGE UNCERTAINTY IN EP WITH ISIMIP DATA -----
efficiency.dt <- copy(uncertainty.dt) %>%
  setnames(., "V1", "Ep")
ghm.dt <- dcast(isimip.dt, Country + Continent + Code ~ Model, value.var = "Water.Withdrawn")</pre>
full.dt <- merge(efficiency.dt, ghm.dt, by = c("Country", "Continent"), all.x = TRUE) %>%
  .[, (names.isimip):= lapply(.SD, function(x) x / Ep), .SDcols = names.isimip]
tmp.dt <- melt(full.dt, measure.vars = names.isimip, variable.name = "Model",</pre>
               value.name = "IWW_corrected")
ghm.large <- melt(ghm.dt, measure.vars = names.isimip, variable.name = "Model",</pre>
     value.name = "IWW")
gm.uncertainty <- tmp.dt[, .(min = min(IWW_corrected), max = max(IWW_corrected)),</pre>
                          .(Country, Continent, Model)]
gm.dt <- merge(ghm.large, gm.uncertainty)</pre>
```

4.5 Retrieve data from ISIMIP (climate change in 2050)

```
# READ IN FILES ON CLIMATE CHANGE UNCERTAINTY (2050) -------

files <- list(
    "watergap2_miroc5_ewembi_rcp85_2005soc_co2_pirrww_global_monthly_2006_2099.nc",
    "watergap2_miroc5_ewembi_rcp60_2005soc_co2_pirrww_global_monthly_2006_2099.nc",
    "watergap2_miroc5_ewembi_rcp45_2005soc_co2_pirrww_global_monthly_2006_2099.nc",
    "watergap2_miroc5_ewembi_rcp26_2005soc_co2_pirrww_global_monthly_2006_2099.nc",
    "lpjml_miroc5_ewembi_rcp85_2005soc_co2_pirrww_global_monthly_2006_2099.nc",
    "lpjml_miroc5_ewembi_rcp60_2005soc_co2_pirrww_global_monthly_2006_2099.nc",
    "pcr-globwb_miroc5_ewembi_rcp26_2005soc_co2_pirrww_global_monthly_2006_2099.nc",
    "pcr-globwb_miroc5_ewembi_rcp60_2005soc_co2_pirrww_global_monthly_2006_2099.nc",
    "h08_miroc5_ewembi_rcp85_2005soc_co2_pirrww_global_monthly_2006_2099.nc",
    "h08_miroc5_ewembi_rcp60_2005soc_co2_pirrww_global_monthly_2006_2099.nc",
    "h08_miroc5_ewembi_rcp26_2005soc_co2_pirrww_global_monthly_2006_2099.nc",
    "h08_miroc5_ewembi_rcp26_2005soc_co2_pirrww_gl
```

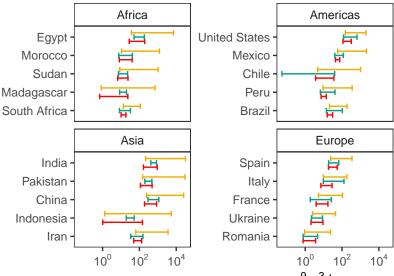
```
vecs <- 1:((2099 - 2005) * 12)
vec <- split(vecs, ceiling(seq_along(vecs) / 12))</pre>
names(vec) <- 2006:2099
dname <- "pirrww"</pre>
selected.years <- as.character(seq(2030, 2050, 10))</pre>
# Read in datasets
isimip.climate <- mclapply(</pre>
 files, function(x)
    open_nc_files(file = x, dname = dname, selected.years = selected.years, vec = vec),
 mc.cores = detectCores() * 0.75
)
# ARRANGE DATASETS ---
ghms <- c(rep("WaterGap", times = 4),</pre>
          rep("LPJmL", times = 3),
          rep("PCR-GLOBWB", times = 2),
          rep("H08", times = 3))
climate_scenario <- c(85, 60, 45, 26, 85, 60, 26, 60, 26, 85, 60, 26)
names.isimip <- paste(ghms, climate_scenario, sep = "/")</pre>
# Name the slots
names(isimip.climate) <- names.isimip</pre>
for(i in names(isimip.climate)) {
 names(isimip.climate[[i]]) <- selected.years</pre>
}
# Arrange data
isimip.climate.dt <- lapply(isimip.climate, function(x) rbindlist(x, idcol = "Year")) %%
 rbindlist(., idcol = "Model") %>%
  .[!Continent == "Oceania"] %>%
  separate(., "Model", c("Model", "Climate scenario"), "/") %>%
 na.omit() %>%
  .[Year == 2050]
# Export
fwrite(isimip.climate.dt, "isimip.climate.dt.csv")
# PLOT RANGES OF STRUCTURAL UNCERTAINTY AND RANGES OF
# STRUCTURAL UNCERTAINTY + UNCERTAINTY IN IRRIGATION EFFICIENCY +
# UNCERTAINTY IN CLIMATE CHANGE -----
countries_list <- c("Egypt", "Sudan", "South Africa", "Morocco", "Madagascar",</pre>
                    "United States", "Mexico", "Brazil", "Chile", "Peru",
```

```
"India", "China", "Pakistan", "Iran", "Indonesia",
                    "Italy", "Spain", "France", "Ukraine", "Romania")
range.gm <- gm.dt %>%
  .[, .(min = min(IWW, na.rm = TRUE), max = max(IWW, na.rm = TRUE)),
    .(Country, Continent)] %>%
  .[, Approach:= "GM"]
range.study <- gm.dt %>%
  .[, .(min = min(min, na.rm = TRUE), max = max(max, na.rm = TRUE)),
    .(Country, Continent)] %>%
  .[, Approach:= "GM + uncertainty in irrigation efficiency"]
range.climate <- isimip.climate.dt %>%
  .[, .(min = min(Water.Withdrawn), max = max(Water.Withdrawn)),
    .(Country, Continent)] %>%
  .[, Approach:= "GM + uncertainty in climate change"]
all.uncertainties <- rbind(range.gm, range.study, range.climate) %>%
  [, mean:= (min + max) / 2]
# Substitute O by NA -----
all.uncertainties[all.uncertainties == 0] <- NA
all.uncertainties %>%
  .[Country %in% countries_list] %>%
 ggplot(., aes(reorder(Country, mean), mean, color = Approach)) +
  geom_errorbar(aes(ymin = min,
                    ymax = max),
                position = position_dodge(0.7)) +
  scale_y_log10(breaks = trans_breaks("log10", function(x) 10 ^ (2 * x)),
                labels = trans_format("log10", math_format(10 ^ .x))) +
  scale_color_manual(name = "GMs", values = wes_palette("Darjeeling1")) +
  labs(y = expression(paste("Irrigation water withdrawal ", " ", "(", 10^9, m^3/year, "", ")")
       x = "") +
  facet_wrap(~Continent, scales = "free_y") +
  coord_flip() +
  theme AP() +
  guides(color = guide_legend(nrow = 3, byrow = TRUE))
```

<u> —</u> GМ

GMs — GM + uncertainty in climate change

— GM + uncertainty in irrigation efficiency



FXPORT ----

Irrigation water withdrawal (10⁹m³/year)

```
fwrite(all.uncertainties, "all.uncertainties.csv")
# PLOT RANGES OF STRUCTURAL UNCERTAINTY AND RANGES OF
# STRUCTURAL UNCERTAINTY + UNCERTAINTY IN IRRIGATION EFFICIENCY (COMPLETE)
vec1 <- all.uncertainties[Approach == "GM", Country]</pre>
vec2 <- all.uncertainties[Approach == "GM + uncertainty in climate change", Country]</pre>
vec3 <- all.uncertainties[Approach == "GM + uncertainty in irrigation efficiency", Country]</pre>
common_countries <- Reduce(intersect, list(vec1, vec2, vec3))</pre>
dd <- list()
for (i in 1:length(list_continents)) {
  dd[[i]] <- all.uncertainties %>%
    .[Country %in% common_countries] %>%
    na.omit() %>%
    .[Continent %in% list_continents[[i]]] %>%
    ggplot(., aes(reorder(Country, mean), mean, color = Approach)) +
    geom_errorbar(aes(ymin = min,
                      ymax = max),
                  position = position_dodge(0.7)) +
    scale_y_log10(breaks = trans_breaks("log10", function(x) 10 ^ (2 * x)),
                  labels = trans_format("log10", math_format(10 ^ .x))) +
    scale_color_manual(name = "GM", values = wes_palette("Darjeeling1")) +
    labs(y = expression(paste("Irrigation water withdrawal ", " ", "(", 10^9, m^3/year, "", ")
         x = "") +
```

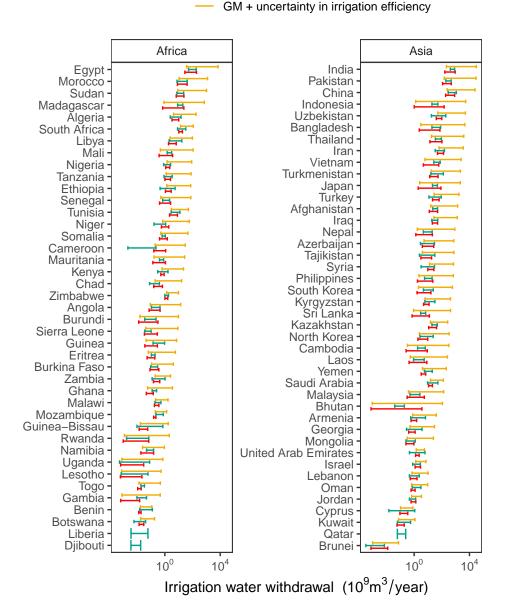
```
facet_wrap(~Continent, scales = "free_y") +
    coord_flip() +
    theme_AP() +
    guides(color = guide_legend(nrow = 3, byrow = TRUE))
}

dd

## [[1]]

— GM

GM — GM + uncertainty in climate change
```



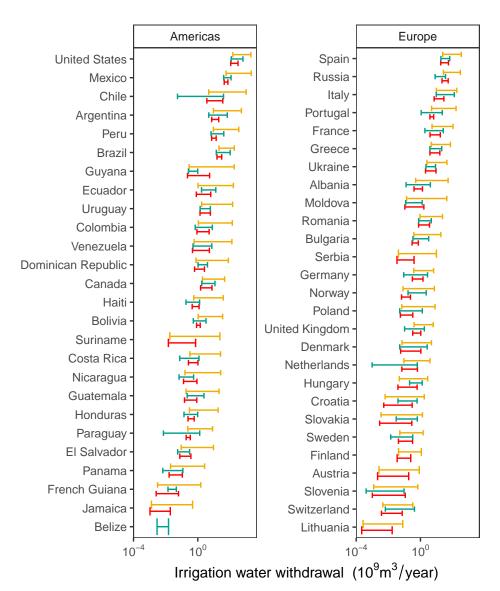
##

```
## [[2]]
```

<u> —</u> GМ

GM — GM + uncertainty in climate change

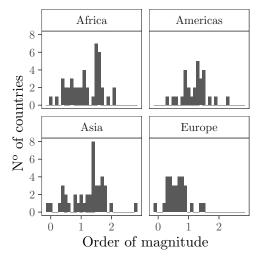
GM + uncertainty in irrigation efficiency



```
dd <- melt(all.uncertainties, measure.vars = c("min", "max")) %>%
  dcast(., Country + Continent + variable ~ Approach, value.var = "value") %>%
  .[variable == "max"] %>%
  na.omit() %>%
  # Calculate differencs in orders of magnitude
```

```
.[, order.magnitude:= log10(`GM + uncertainty in irrigation efficiency`) -
      log10(`GM + uncertainty in climate change`)] %>%
  .[order(-order.magnitude)]
ggplot(dd, aes(order.magnitude)) +
 geom_histogram() +
 facet_wrap(~Continent) +
 labs(x = "Order of magnitude", y = "N^{\circ} of countries") +
 theme_AP()
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## Warning in (function (texString, cex = 1, face = 1, engine =
## getOption("tikzDefaultEngine"), : Attempting to calculate the width of a Unicode
## stringusing the pdftex engine. This may fail! See the Unicodesection of ?
## tikzDevice for more information.
## Warning in (function (texString, cex = 1, face = 1, engine =
## getOption("tikzDefaultEngine"), : Attempting to calculate the width of a Unicode
## stringusing the pdftex engine. This may fail! See the Unicodesection of ?
## tikzDevice for more information.
## Warning in (function (texString, cex = 1, face = 1, engine =
## getOption("tikzDefaultEngine"), : Attempting to calculate the width of a Unicode
## stringusing the pdftex engine. This may fail! See the Unicodesection of ?
## tikzDevice for more information.
## Warning in (function (texString, cex = 1, face = 1, engine =
## getOption("tikzDefaultEngine"), : Attempting to calculate the width of a Unicode
## stringusing the pdftex engine. This may fail! See the Unicodesection of ?
## tikzDevice for more information.
## Warning in (function (texString, cex = 1, face = 1, engine =
## getOption("tikzDefaultEngine"), : Attempting to calculate the width of a Unicode
## stringusing the pdftex engine. This may fail! See the Unicodesection of ?
## tikzDevice for more information.
## Warning in (function (texString, cex = 1, face = 1, engine =
## getOption("tikzDefaultEngine"), : Attempting to calculate the width of a Unicode
## stringusing the pdftex engine. This may fail! See the Unicodesection of ?
## tikzDevice for more information.
## Warning in (function (texString, cex = 1, face = 1, engine =
## getOption("tikzDefaultEngine"), : Attempting to calculate the width of a Unicode
## stringusing the pdftex engine. This may fail! See the Unicodesection of ?
## tikzDevice for more information.
## Warning in (function (texString, cex = 1, face = 1, engine =
```

getOption("tikzDefaultEngine"), : Attempting to calculate the width of a Unicode
stringusing the pdftex engine. This may fail! See the Unicodesection of ?
tikzDevice for more information.



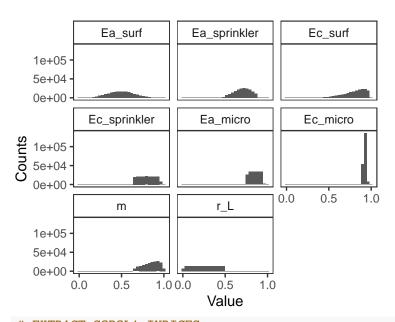
5 Sensitivity analysis

5.1 Probability distributions

```
# SAMPLE MATRIX DISTRIBUTIONS ----
mat <- data.table(full_sample_matrix(IFT = "Jager", Country = "Spain")$matrix)

melt(mat[, 1:8], measure.vars = colnames(mat)[-c(9,10)]) %>%
    ggplot(., aes(value)) +
    geom_histogram() +
    labs(x = "Value", y = "Counts") +
    scale_x_continuous(breaks = pretty_breaks(n = 3)) +
    facet_wrap(~variable, labeller = labeller(type = label_parsed)) +
    theme_AP()
```

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

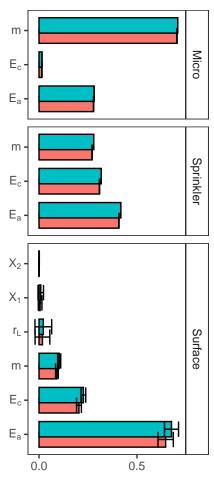


out[[i]] <- ind[Country %in% tmp.ift[[i]][, Country]]</pre>

5.2 Sobol' indices

}

for(i in names(tmp.ift)) {

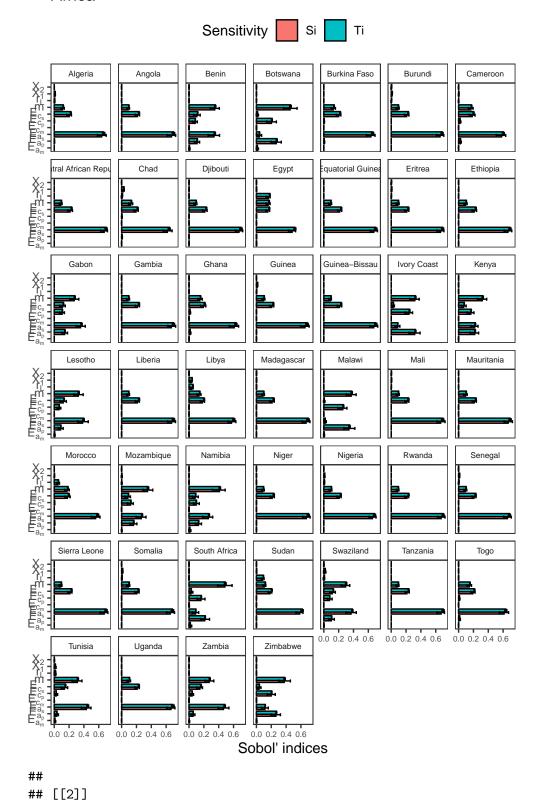


EXTRACT SOBOL' INDICES FOR JAGER ---jager.tmp <- lapply(y[["Jägermeyr et al. 2015"]], function(x) x\$indices\$results)
names(jager.tmp) <- new.rohwer\$Country</pre>

```
jager.ind <- rbindlist(jager.tmp, idcol = "Country") %>%
  .[, Continent:= countrycode(.[, Country],
                             origin = "country.name",
                             destination = "continent")] %>%
  .[, parameters:= ifelse(parameters == "Ea_surf", "E[a[s]]",
                         ifelse(parameters == "Ec_surf", "E[c[s]]",
                                ifelse(parameters == "Ea_sprinkler", "E[a[p]]",
                                       ifelse(parameters == "Ec_sprinkler", "E[c[p]]",
                                              ifelse(parameters == "Ea_micro", "E[a[m]]",
                                                      ifelse(parameters == "Ec_micro", "E[c[m]]
                                                             ifelse(parameters == "r_L", "r[L]"
                                                                    ifelse(parameters == "X1",
                                                                           ifelse(parameters ==
## Warning in countrycode_convert(sourcevar = sourcevar, origin = origin, destination = dest,
Continent_vector <- c("Africa", "Americas", "Asia", "Europe")</pre>
lapply(Continent_vector, function(x)
  ggplot(jager.ind[Continent == x], aes(parameters, original, fill = sensitivity), color = bla
    geom_bar(stat = "identity", position = position_dodge(0.6), color = "black") +
    geom_errorbar(aes(ymin = low.ci, ymax = high.ci),
                  position = position_dodge(0.6)) +
    scale_fill_discrete(name = "Sensitivity", labels = c("Si", "Ti")) +
    labs(x = "", y = "Sobol' indices") +
    scale_x_discrete(labels = ggplot2:::parse_safe) +
    coord_flip() +
    scale_y_continuous(breaks = pretty_breaks(n = 3)) +
    facet_wrap(~Country) +
    theme_AP() +
    theme(strip.text.x = element_text(size = 6),
          axis.text.x = element_text(size = 6)) +
    ggtitle(x)
```

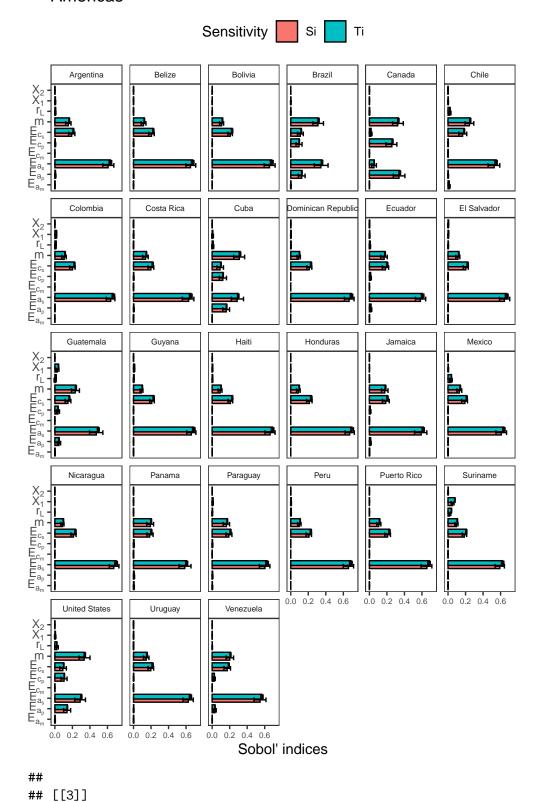
[[1]]

Africa



49

Americas

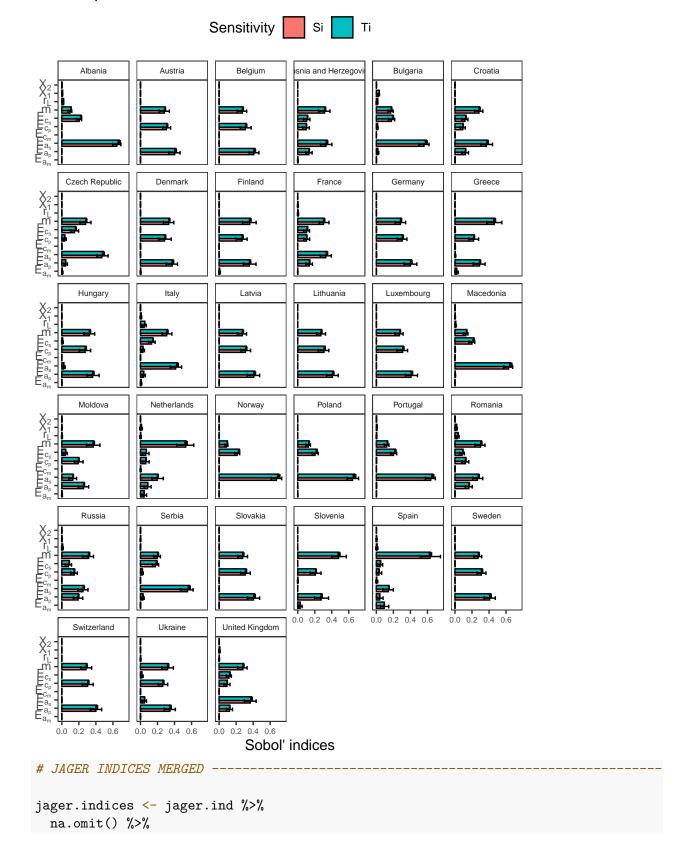


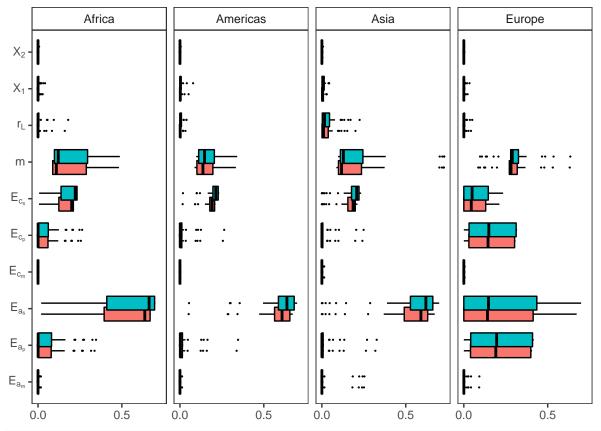
Asia



[[4]]

Europe





```
y = 6, x = 0.55, vjust = 5.5, size = 10))
                           Sensitivity
   m
                                  Africa
                                            Americas
                  Micro
                        b
                                                           Asia
                                                                     Europe
a
   Ec
                            X_2 -
   E_a
                            X_1 -
                  Sprinkler
   m
                             r<sub>L</sub> -
                            m -
                                                                    =: | : ::
                               7
                                           ::#
                                                       =:#
   X_2
                            E<sub>c,</sub>
                               :::
                                           : :
                                                       == ::
   X_1
                           \mathsf{E}_{c_m}
                  Surface
   r_{\text{\tiny L}}
                                               :: -₩
                                                       m·
                           \mathsf{E}_{\mathsf{a}_{\mathsf{p}}}
                                           t: =
                                               :
                                                       | :: ::
   Ec
                           \mathsf{E}_{a_m}
                                          0.0
                                                      0.0
                                                                         0.5
     0.0
             0.5
                              0.0
                                     0.5
                                                 0.5
                                                             0.5
                                                                  0.0
                                     Sobol' indices
# SESSION INFORMATION ----
sessionInfo()
## R version 4.0.3 (2020-10-10)
## Platform: x86 64-apple-darwin17.0 (64-bit)
## Running under: macOS Big Sur 10.16
##
## Matrix products: default
            /Library/Frameworks/R.framework/Versions/4.0/Resources/lib/libRblas.dylib
## LAPACK: /Library/Frameworks/R.framework/Versions/4.0/Resources/lib/libRlapack.dylib
##
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/c/en_US.UTF-8/en_US.UTF-8
##
## attached base packages:
## [1] parallel stats
                               graphics grDevices utils
                                                                  datasets methods
## [8] base
##
## other attached packages:
   [1] checkpoint_1.0.0 benchmarkme_1.0.7 ncdf4_1.17
                                                                        rworldmap_1.3-6
##
    [5] sp_1.4-5
                              overlapping_1.6
                                                   testthat_3.0.4
                                                                        scales_1.1.1
##
   [9] ggridges_0.5.3
                              countrycode_1.3.0 doParallel_1.0.16 iterators_1.0.13
                              cowplot_1.1.1
## [13] foreach 1.5.1
                                                   wesanderson_0.3.6 sensobol_1.0.3
## [17] forcats_0.5.1
                              stringr_1.4.0
                                                   dplyr_1.0.7
                                                                        purrr_0.3.4
```

loaded via a namespace (and not attached):

tidyr_1.1.3

data.table_1.14.0

[21] readr_2.0.1

##

[25] tidyverse_1.3.1

tibble_3.1.3

ggplot2_3.3.5

```
## [1] fs_1.5.0
                              lubridate_1.7.10
                                                     httr_1.4.2
## [4] tools_4.0.3
                              backports_1.2.1
                                                     utf8_1.2.2
## [7] R6_2.5.0
                              DBI_1.1.1
                                                     colorspace_2.0-2
## [10] withr_2.4.2
                              tidyselect_1.1.1
                                                     gridExtra_2.3
## [13] compiler 4.0.3
                              cli 3.0.1
                                                     rvest 1.0.1
## [16] xml2_1.3.2
                              labeling_0.4.2
                                                     digest_0.6.27
## [19] foreign 0.8-81
                              rmarkdown 2.10
                                                     benchmarkmeData 1.0.4
## [22] pkgconfig_2.0.3
                              htmltools_0.5.1.1
                                                     highr_0.9
## [25] dbplyr_2.1.1
                              maps_3.3.0
                                                     rlang_0.4.11
## [28] readxl_1.3.1
                              rstudioapi_0.13
                                                     farver_2.1.0
## [31] generics_0.1.0
                                                     jsonlite_1.7.2
                              tikzDevice_0.12.3.1
## [34] magrittr_2.0.1
                              dotCall64_1.0-1
                                                     Matrix_1.3-4
## [37] Rcpp_1.0.7
                              munsell_0.5.0
                                                     fansi_0.5.0
## [40] viridis_0.6.1
                              lifecycle_1.0.0
                                                     stringi_1.7.3
## [43] yaml_2.2.1
                              plyr_1.8.6
                                                     grid_4.0.3
## [46] maptools_1.1-1
                              crayon_1.4.1
                                                     lattice_0.20-44
## [49] haven_2.4.3
                              hms_1.1.0
                                                     knitr_1.33
## [52] pillar_1.6.2
                              randtoolbox_1.30.1
                                                     codetools_0.2-18
## [55] reprex_2.0.1
                              glue_1.4.2
                                                     evaluate_0.14
## [58] modelr 0.1.8
                              vctrs 0.3.8
                                                     spam 2.7-0
## [61] tzdb 0.1.2
                                                     cellranger_1.1.0
                              Rdpack 2.1.2
## [64] gtable 0.3.0
                              assertthat 0.2.1
                                                     xfun 0.25
## [67] rbibutils_2.2.3
                              rngWELL_0.10-6
                                                     broom_0.7.9
                                                     fields_12.5
## [70] filehash_2.4-2
                              viridisLite_0.4.0
## [73] ellipsis_0.3.2
## Return the machine CPU
                  "); print(get cpu()$model name)
cat("Machine:
## Machine:
## [1] "Intel(R) Core(TM) i9-9900K CPU @ 3.60GHz"
## Return number of true cores
cat("Num cores: "); print(detectCores(logical = FALSE))
## Num cores:
## [1] 8
## Return number of threads
cat("Num threads: "); print(detectCores(logical = FALSE))
## Num threads:
## [1] 8
```