Irrigation's real impact on global water and food security $_{\rm R~code}$

Arnald Puy

Contents

1	Retrieve all corpus	6
	1.1 Abstract corpus	6
	1.2 Full text corpus	6
	1.3 Policy corpus	8
2	Split datasets for analysis	g
3	Network analysis	10
	3.1 Network metrics	12
	3.2 Network plots	14
4	Aquastat data	25
5	Session information	27

```
sensobol::load_packages(c("openxlsx", "data.table", "tidyverse", "bibliometrix",
                        "igraph", "ggraph", "cowplot", "tidygraph", "benchmarkme",
                        "parallel", "wesanderson", "scales"))
# 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),
         strip.background = element_rect(fill = "white"),
         legend.margin = margin(0.5, 0.1, 0.1, 0.1),
         legend.box.margin = margin(0.2, -4, -7, -7),
         plot.margin = margin(3, 4, 0, 4),
         legend.text = element_text(size = 8),
         axis.title = element_text(size = 10),
         legend.key.width = unit(0.4, "cm"),
         legend.key.height = unit(0.4, "cm"),
         legend.title = element_text(size = 9))
}
```

```
database <- c("wos", "scopus", "dimensions")</pre>
topic <- c("water", "food")</pre>
# Create all possible combinations
combinations <- expand.grid(database = database, topic = topic)</pre>
# Combine the vectors with an underscore
file.name <- paste(combinations$database, "dt", combinations$topic, sep = "_")</pre>
# Loop to create the file names -----
for (i in 1:length(file.name)) {
 database.type <- str_extract(file.name, "^(wos|scopus|dimensions)")</pre>
 if(isTRUE(database.type[i] == "wos")) {
   file.name[i] <- paste(file.name[i], "bib", sep = ".")</pre>
 } else {
   file.name[i] <- paste(file.name[i], "csv", sep = ".")</pre>
 }
}
# vector with new column names ------
new_colnames <- c("doi", "authors", "year", "title", "journal", "abstract", "database")</pre>
to_lower <- c("authors", "title", "journal", "abstract")</pre>
# Loop to read in the datasets -----
out <- list()</pre>
for (i in 1:length(file.name)) {
 database.type <- str_extract(file.name[i], "^(wos|scopus|dimensions)")</pre>
 if(isTRUE(database.type == "wos")) {
   out[[i]] <- convert2df(file = file.name[i],</pre>
            dbsource = "wos",
```

```
format = "bibtex") %>%
      data.table() %>%
      .[, .(DI, AU, PY, TI, SO, AB)] %>%
      .[, database:= "wos"]
  } else if (isTRUE(database.type == "dimensions")) {
    out[[i]] <- fread(file.name[i], skip = 1) %>%
      .[, .(DOI, Authors, PubYear, Title, `Source title`, Abstract)] %>%
      .[, database:= "dimensions"]
  } else if(isTRUE(database.type == "scopus")) {
    out[[i]] <- fread(file.name[i]) %>%
      .[, .(DOI, Authors, Year, Title, `Source title`, Abstract)] %>%
      .[, database:= "scopus"]
  }
  setnames(out[[i]], colnames(out[[i]]), new_colnames) %>%
    .[, (to_lower):= lapply(.SD, tolower), .SDcols = (to_lower)] %>%
    .[, abstract:= sub("references.*", "", abstract)]
}
##
## Converting your wos collection into a bibliographic dataframe
##
##
## Warning:
## In your file, some mandatory metadata are missing. Bibliometrix functions may not work prop
##
## Please, take a look at the vignettes:
## - 'Data Importing and Converting' (https://www.bibliometrix.org/vignettes/Data-Importing-and-
## - 'A brief introduction to bibliometrix' (https://www.bibliometrix.org/vignettes/Introduction
##
##
## Missing fields: C1 CR
## Done!
##
## Converting your wos collection into a bibliographic dataframe
##
##
## Warning:
## In your file, some mandatory metadata are missing. Bibliometrix functions may not work prop
## Please, take a look at the vignettes:
```

```
## - 'Data Importing and Converting' (https://www.bibliometrix.org/vignettes/Data-Importing-and
## - 'A brief introduction to bibliometrix' (https://www.bibliometrix.org/vignettes/Introduction
##
##
## Missing fields: C1 CR
## Done!
names(out) <- combinations$topic</pre>
# Arrange -----
dt <- rbindlist(out, idcol = "topic")</pre>
tmp <- split(dt, list(dt$topic, dt$database))</pre>
cols_to_merge_by <- c("doi", "year", "title", "journal", "abstract")</pre>
dt.water <- merge(merge(tmp$water.dimensions, tmp$water.scopus, by = cols_to_merge_by,
           all = TRUE), tmp$water.wos, by = cols_to_merge_by,
     all = TRUE)
dt.food <- merge(merge(tmp$food.dimensions, tmp$food.scopus, by = cols_to_merge_by,
           all = TRUE), tmp$food.wos, by = cols_to_merge_by,
     all = TRUE)
# Filer out duplicated studies by doi ------
tmp.list <- list(dt.water, dt.food)</pre>
duplicated.dois <- final.dt <- list()</pre>
for (i in 1:length(tmp.list)) {
 duplicated.dois[[i]] <- duplicated(tmp.list[[i]]$doi, incomparables = NA, na.rm = TRUE)</pre>
 final.dt[[i]] <- tmp.list[[i]][!duplicated.dois[[i]]][, location.belief.system:= "abstract"]</pre>
}
names(final.dt) <- topic</pre>
# Check if there is any duplicated doi ------
any(duplicated(final.dt$food$doi, na.rm = TRUE, incomparables = NA))
## [1] FALSE
# Export to xlsx -----
```

1 Retrieve all corpus

1.1 Abstract corpus

```
final.dt.water.screened <- data.table(read.xlsx("final.dt.water_screened.xlsx"))</pre>
final.dt.food.screened <- data.table(read.xlsx("final.dt.food_screened.xlsx"))</pre>
screened.dt <- list(final.dt.water.screened, final.dt.food.screened)</pre>
names(screened.dt) <- c("water", "food")</pre>
lapply(screened.dt, function(x) x[, .N, screening])
## $water
##
      screening
##
         <char> <int>
              F
## 1:
                  168
              Т
## 2:
                  163
##
## $food
      screening
         <char> <int>
##
## 1:
              F
                  465
## 2:
              Т
                   39
# Export for close-reading only the references that do include
# the belief system in the abstract -----
for (i in names(screened.dt)) {
  screened.dt[[i]][screening == "T"] %>%
    unique(., by = "title") %>%
    .[, .(doi, title, year)] %>%
    write.xlsx(., paste("abstract.corpus", i, "xlsx", sep = "."))
}
```

1.2 Full text corpus

```
colnames.full.text <- c("doi", "year", "title", "journal", "topic")</pre>
keywords <- c("water", "irrigat")</pre>
# Function to load and preprocess data -----
load_and_preprocess_data <- function(file_path, topic) {</pre>
 fread(file_path, skip = 1)[, topic := topic]
}
# Load and preprocess water data -----
dimensions.full.text.water <- load_and_preprocess_data("dimensions_dt_full_text_water_2022_202
dimensions.full.text.food <- rbind(</pre>
  load_and_preprocess_data("dimensions_dt_full_text_food_2022.csv", "food"),
 load_and_preprocess_data("dimensions dt_full_text_food_2023.csv", "food")
# Combine water and food data ------
result <- rbind(dimensions.full.text.water, dimensions.full.text.food) %>%
  .[, .(DOI, PubYear, Title, `Source title`, topic)] %>%
 setnames(., colnames(.), colnames.full.text) %>%
  # remove the references that are included in the dataset that
  # collects mentions in the abstracts
  .[!.$doi %in% final.dt.water.screened$doi]
# Create a logical condition for pattern matching using grepl ------
pattern_condition <- sapply(keywords, function(keyword)</pre>
  grepl(keyword, result$title, ignore.case = TRUE))
full.text.dt <- result[rowSums(pattern condition) > 0]
# Sample just 50% for the analysis -----
# Create function
random_sample <- function(input_dt) {</pre>
  set.seed(123) # Set a seed for reproducibility
 sampled_dt <- input_dt[sample(.N, .N * 0.5), ]</pre>
 return(sampled_dt)
}
# sample
full.text.sampled <- full.text.dt[, random_sample(.SD), topic]</pre>
full.text.sampled[, .N, topic]
```

```
## topic N
## <char> <int>
## 1: water 4607
## 2: food 8108
# Export -----

for (i in c("water", "food")) {
  full.text.sampled[topic == i] %>%
    write.xlsx(paste("full.text.corpus", i, "xlsx", sep = "."))
}
```

1.3 Policy corpus

```
dt.policy.water <- load_and_preprocess_data("dimensions_dt_policy.csv", "water")</pre>
dt.policy.food <- load_and_preprocess_data("dimensions_dt_policy_food.csv", "food")
dimensions.full.text.policy <- rbind(dt.policy.food, dt.policy.water) %>%
  .[, .(`Policy document ID`, PubYear, Title, `Publishing Organization`,
       `Sustainable Development Goals`, `Source Linkout`, topic)]
dimensions.full.text.policy[, .N, topic]
##
      topic
##
     <char> <int>
## 1:
       food 10573
## 2: water 3455
# Create a logical condition for pattern matching using grepl
pattern_condition_policy <- sapply(keywords, function(keyword)</pre>
 grep1(keyword, dimensions.full.text.policy$Title, ignore.case = TRUE))
# Combine conditions with OR using rowSums
matching.rows.policy <- dimensions.full.text.policy[rowSums(pattern_condition_policy) > 0]
matching.rows.policy[, .N, topic]
##
      topic
               N
##
     <char> <int>
## 1:
       food
              750
## 2: water
              450
# Export -----
for (i in c("water", "food")) {
```

```
matching.rows.policy[topic == i] %>%
    write.xlsx(paste("policy.corpus", i, "xlsx", sep = "."))
}
```

2 Split datasets for analysis

```
# Function to split dataset in n chunks -----
split_dt_fun <- function(dt, num_parts) {</pre>
  split_dt <- list()</pre>
  # Calculate the number of rows in each part
 rows_per_part <- nrow(dt) %/% num_parts</pre>
  # Split the data.table into roughly equal parts
 for (i in 1:num_parts) {
    start_row <- (i - 1) * rows_per_part + 1
    end_row <- i * rows_per_part</pre>
   if (i == num_parts) {
     end_row <- nrow(dt)</pre>
    split_dt[[i]] <- dt[start_row:end_row, ]</pre>
 return(split_dt)
}
# Create the datasets for close reading -----
times.nanxin <- 4
times.arnald <- 2
nanxin <- paste(rep("nanxin", times.nanxin), 1:times.nanxin, sep = "")</pre>
arnald <- paste(rep("arnald", times.arnald), 1:times.arnald, sep = "")
names_surveyors <- c(arnald, nanxin, "seth", paste("student", 1:4, sep = ""))</pre>
n.surveyors <- length(names_surveyors)</pre>
full.text.corpus.water <- read.xlsx("full.text.corpus.water.xlsx")</pre>
survey.dt.split <- split_dt_fun(dt = full.text.corpus.water, num_parts = n.surveyors)</pre>
```

3 Network analysis

```
tmp <- list()</pre>
names.files <- c("WORK", "NETWORK")</pre>
topics <- c("water")</pre>
corpus <- c("abstract.corpus", "policy.corpus", "full.text.corpus")</pre>
cols_of_interest <- c("title", "author", "claim", "citation")</pre>
# Paste all possible combinations of names -----
combs <- expand.grid(corpus = corpus, topics = topics, approach = names.files)</pre>
all.files <- paste(paste(combs$corpus, combs$topics, sep = "."), combs$approach, sep = "
                "xlsx", sep = ".")
tmp <- list()</pre>
for (i in 1:length(all.files)) {
 tmp[[i]] <- data.table(read.xlsx(all.files[i]))</pre>
 if (!str_detect(all.files[i], "NETWORK")) {
   tmp[[i]][, title:= tolower(title)]
 } else {
   tmp[[i]][, (cols_of_interest):= lapply(.SD, tolower), .SDcols = (cols_of_interest)]
 }
names(tmp) <- all.files</pre>
```

```
dataset.networks <- all.files[str detect(all.files, "NETWORK")]</pre>
network.dt <- tmp[dataset.networks] %>%
 rbindlist() %>%
 .[, policy:= grepl("^policy", doi)]
network.dt[, author:= ifelse(policy == TRUE, doi, author)]
# Remove the year from mentions to FAO Aquastat ------
pattern <- "\\b(?:19|20)\\d{2}\\b" # Matches years between 1900 and 2099
for (col in c("citation", "author")) {
 matches <- grepl("^fao aquastat\\s+\\d+$", network.dt[[col]], ignore.case = TRUE)</pre>
 network.dt[matches, (col) := gsub("\\d+", "", network.dt[[col]][matches], perl = TRUE)]
 network.dt[, (col) := trimws(network.dt[[col]])]
}
# Rename columns -----
setnames(network.dt, c("author", "citation"), c("from", "to"))
# Create copy ------
network.dt.claim <- copy(network.dt)</pre>
# Convert all to lower caps ------
network.dt <- network.dt[, .(from, to, document.type, nature.claim)]</pre>
cols_to_change <- colnames(network.dt)</pre>
network.dt[, (cols_to_change):= lapply(.SD, trimws), .SDcols = (cols_to_change)]
# Check proportio of studies making the claim but without providing any citation and
# not being primary papers (!modelling) ------
no.citation.total <- nrow(network.dt.claim[nature.claim == "no citation" &
                                  !document.type == "modelling"])
no.citation.total / nrow(network.dt.claim)
## [1] 0.229572
# Count document type by nature of claim -----
a <- network.dt[, .N, .(nature.claim, document.type)] %>%
na.omit() %>%
```

```
ggplot(., aes(reorder(nature.claim, N), N)) +
  coord flip() +
  geom bar(stat = "identity") +
  facet_wrap(~document.type) +
  scale y continuous(breaks = breaks pretty(n = 2)) +
  labs(x = "", y = "Counts") +
  theme_AP()
# Count how many documents make the claim and cite / do not cite,
# by document.type -----
b <- network.dt[, .(without.citation = sum(is.na(to)),</pre>
               with.citation = .N - sum(is.na(to))), document.type] %>%
 melt(., measure.vars = c("without.citation", "with.citation")) %>%
  ggplot(., aes(document.type, value)) +
  geom_bar(stat = "identity") +
  scale_y_continuous(breaks = breaks_pretty(n = 2)) +
  scale_x_discrete(guide = guide_axis(n.dodge = 2)) +
  labs(x = "", y = "Counts") +
 facet_wrap(~variable) +
  theme_AP()
plot_grid(a, b, ncol = 2, rel_widths = c(0.63, 0.37), labels = "auto")
                                                      b
\mathbf{a}
                                                            without.citation
                                                                          with.citation
                  other
                               policy
                                            review
  citation backup
                                                      Counts
     no citation
                                                        50
      modelling
       no claim
                                                           other
                                                                  review other
                                 50
                   50
                                              50
                                                               policy
                                                                            policy
```

3.1 Network metrics

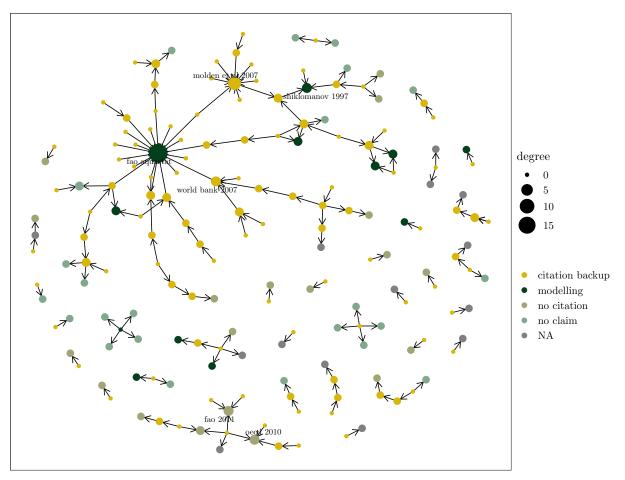
Counts

```
# Calculate network metrics
edge_density(citation_graph)
## [1] 0.00477964
# Modularity:
# - c.1: Strong community structure, where nodes within groups are highly connected.
# - c. -1: Opposite of community structure, where nodes between groups are more connected.
# - c. 0: Indicates absence of community structure or anti-community structure in the network.
wtc <- cluster_walktrap(citation_graph)</pre>
modularity(wtc)
## [1] 0.8629828
network_metrics <- data.table(node = V(citation_graph)$name,</pre>
                               # Degree of a node: The number of connections or
                               # edges linked to that node.
                               # It represents how well-connected or central a
                               # node is within the graph.
                               degree = degree(citation_graph, mode = "in"),
                               # Betweenness centrality of a node: Measures the
                               # extent to which a node lies on the shortest
                               # paths between all pairs of other nodes in the graph.
                               # Nodes with high betweenness centrality act as
                               # bridges or intermediaries, facilitating
                               # communication and information flow between other nodes.
                               betweenness = betweenness(citation_graph),
                               # Closeness centrality of a node: Measures how
                               # close a node is to all other nodes in the graph,
                               # taking into account the length of the shortest paths.
                               # Nodes with high closeness centrality are able to
                               # efficiently communicate or interact with other
                               # nodes in the graph.
                               closeness = closeness(citation graph),
                               pagerank = page_rank(citation_graph)$vector
# Define the max number of rows
max.number <- 3</pre>
degree.nodes <- network_metrics[order(-degree)][1:max.number]</pre>
betweenness.nodes <- network_metrics[order(-betweenness)][1:max.number]</pre>
pagerank.nodes <- network_metrics[order(-closeness)][1:max.number]</pre>
```

```
degree.nodes
##
                 node degree betweenness closeness
                                                   pagerank
##
                       <num>
                                  <num>
               <char>
                                           <num>
                                                      <num>
                                             NaN 0.07754896
## 1:
          fao aquastat
                          19
## 2: molden et al 2007
                          7
                                     16 0.3333333 0.02246264
       world bank 2007
                          3
                                      8 1.0000000 0.01889437
betweenness.nodes
##
                  node degree betweenness closeness
                                                   pagerank
##
                <char> <num>
                                   <num>
                                            <niim>
                                                       <niim>
## 1: molden et al 2007
                          7
                                     16 0.3333333 0.02246264
## 2: oki and kanae 2006
                           2
                                     12 1.0000000 0.02369278
## 3:
        world bank 2007
                           3
                                      8 1.0000000 0.01889437
pagerank.nodes
##
                     node degree betweenness closeness
                                                       pagerank
##
                   <char> <num>
                                      <num>
                                               <num>
                                                          <num>
## 1: sharma and irmak 2012
                                                  1 0.003099364
                                         0
## 2:
          world bank 2007
                              3
                                         8
                                                  1 0.018894373
## 3:
       brajovic et al 2015
                              0
                                         Ω
                                                   1 0.003099364
3.2 Network plots
# Retrieve a vector with the node names ------
graph <- tidygraph::as_tbl_graph(network.dt.complete, directed = TRUE)</pre>
vec.names <- graph %>%
 activate(nodes) %>%
 pull() %>%
 data.table(name = .)
# Merge with info from the network.dt -----
vec.nature.claim <- merge(merge(vec.names, unique(network.dt[, .(from, nature.claim)]),</pre>
                             by.x = "name", by.y = "from", all.x = TRUE),
                        unique(network.dt[, .(from, document.type)]),
                        by.x = "name", by.y = "from", all.x = TRUE)
# Merge with the correct order -----
order_indices <- match(vec.names$name, vec.nature.claim$name)
final.vec.nature.claim <- vec.nature.claim[order_indices, ] %>%
  .[, nature.claim]
```

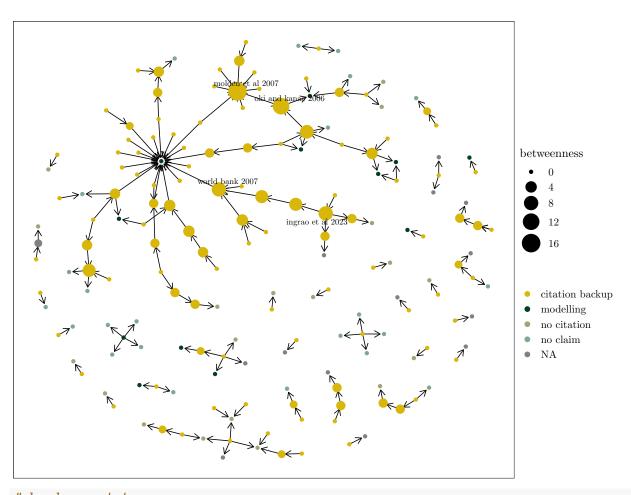
final.vec.document.type <- vec.nature.claim[order_indices,] %>%

```
.[, document.type]
# Attach to the graph -----
graph <- graph %>%
 activate(nodes) %>%
 mutate(nature.claim = final.vec.nature.claim,
        document.type = final.vec.document.type,
        degree = network_metrics$degree,
        betweenness = network_metrics$betweenness,
        pagerank = network_metrics$pagerank)
seed <- 123
# by nature of claim -------
set.seed(seed)
# Label the nodes with highest degree ------
ggraph(graph, layout = "igraph", algorithm = "nicely") +
 geom_edge_link(arrow = arrow(length = unit(1.8, 'mm')),
               end_cap = circle(1, "mm")) +
 geom_node_point(aes(color = nature.claim, size = degree)) +
 geom_node_text(aes(label = ifelse(degree >= min(degree.nodes$degree), name, NA)),
               repel = TRUE, size = 2.2) +
 labs(x = "", y = "") +
 scale_color_manual(name = "",
                  values = wes_palette(name = "Cavalcanti1", 4)) +
 theme_AP() +
 theme(axis.text.x = element_blank(),
       axis.ticks.x = element_blank(),
       axis.text.y = element blank(),
       axis.ticks.y = element_blank(),
       legend.position = "right")
## Warning: Using the `size` aesthetic in this geom was deprecated in ggplot2 3.4.0.
## i Please use `linewidth` in the `default_aes` field and elsewhere instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
## Warning: Removed 174 rows containing missing values (`geom_text_repel()`).
```



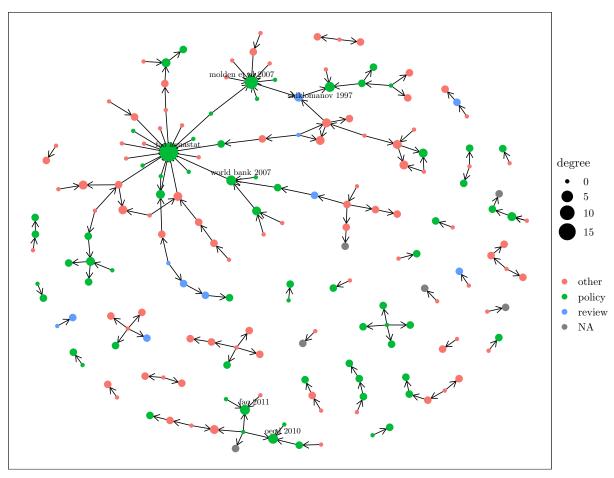
```
set.seed(seed)
# Label the nodes with highest betweenness ---
ggraph(graph, layout = "igraph", algorithm = "nicely") +
  geom_edge_link(arrow = arrow(length = unit(1.8, 'mm')),
                 end_cap = circle(1, "mm")) +
  geom_node_point(aes(color = nature.claim, size = betweenness)) +
  geom_node_text(aes(label = ifelse(betweenness >= min(betweenness.nodes$betweenness), name, N.
                 repel = TRUE, size = 2.2) +
  labs(x = "", y = "") +
  scale_color_manual(name = "",
                     values = wes_palette(name = "Cavalcanti1", 4)) +
  theme_AP() +
  theme(axis.text.x = element_blank(),
        axis.ticks.x = element_blank(),
        axis.text.y = element_blank(),
        axis.ticks.y = element_blank(),
        legend.position = "right")
```

Warning: Removed 176 rows containing missing values (`geom_text_repel()`).



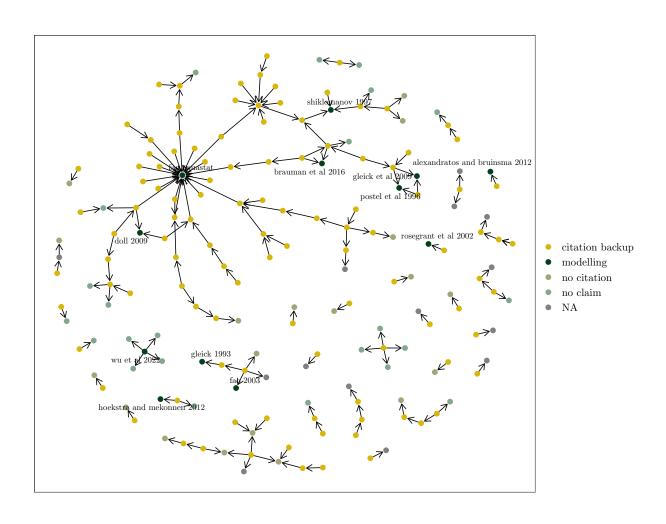
```
# by document.type----
set.seed(seed)
ggraph(graph, layout = "igraph", algorithm = "nicely") +
  geom_edge_link(arrow = arrow(length = unit(1.8, 'mm')),
                 end_cap = circle(1, "mm")) +
  geom_node_point(aes(color = document.type, size = degree)) +
 geom_node_text(aes(label = ifelse(degree >= min(degree.nodes$degree), name, NA)),
                 repel = TRUE, size = 2.2) +
 labs(x = "", y = "") +
  scale_color_discrete(name = "") +
 theme_AP() +
  theme(axis.text.x = element_blank(),
        axis.ticks.x = element_blank(),
        axis.text.y = element_blank(),
        axis.ticks.y = element_blank(),
        legend.position = "right")
```

Warning: Removed 174 rows containing missing values (`geom_text_repel()`).



```
# Label nodes that are modelling exercises -----
set.seed(seed)
ggraph(graph, layout = "igraph", algorithm = "nicely") +
  geom_edge_link(arrow = arrow(length = unit(1.8, 'mm')),
                 end_cap = circle(1, "mm")) +
 geom_node_point(aes(color = nature.claim)) +
 geom_node_text(aes(label = ifelse(nature.claim == "modelling", name, NA)),
                repel = TRUE, size = 2.2) +
 labs(x = "", y = "") +
  scale_color_manual(name = "",
                     values = wes_palette(name = "Cavalcanti1", 4)) +
 theme_AP() +
  theme(axis.text.x = element_blank(),
        axis.ticks.x = element_blank(),
        axis.text.y = element_blank(),
        axis.ticks.y = element_blank(),
        legend.position = "right")
```

Warning: Removed 168 rows containing missing values (`geom_text_repel()`).

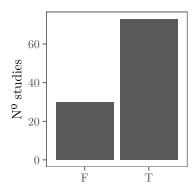


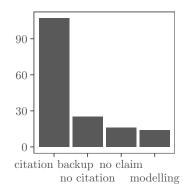
```
tmp <- list()</pre>
names.files <- c("WORK", "NETWORK")</pre>
cols_of_interest <- c("title", "author", "claim", "citation")</pre>
files.abstract.water <- paste(paste("abstract.corpus.water2_", names.files, sep = ""), "xlsx",
for (i in 1:length(files.abstract.water)) {
    tmp[[i]] <- data.table(read.xlsx(files.abstract.water[i]))</pre>
    if (i == 1) {
        tmp[[i]][, title:= tolower(title)]
    } else {
        tmp[[i]][, (cols_of_interest):= lapply(.SD, tolower), .SDcols = (cols_of_interest)]
names(tmp) <- names.files</pre>
abstract.water.dt <- merge(tmp[[1]][claim.in.text == "F"], tmp[[2]], by = c("doi", "title"), a
abstract.water.dt[, claim.in.text:= ifelse(is.na(claim.in.text), "TRUE", "FALSE")]
abstract.water.dt[, c(cols_of_interest, "nature.claim"):= lapply(.SD, trimws), .SDcols = c(cols_of_interest, "nature.claim"):= lapply(.SD, trimws), .SDcols_of_interest, "nature.claim"
abstract.water.dt[, year:= ifelse(is.na(year), as.numeric(gsub("\\D", "", abstract.water.dt$au
a <- tmp$WORK[claim.in.text %in% c("T", "F")] %>%
    .[, .N, claim.in.text] %>%
    ggplot(., aes(claim.in.text, N)) +
    geom_bar(stat = "identity") +
    labs(x = "", y = "N^{\circ} studies") +
    theme AP()
b <- tmp$NETWORK %>%
    .[complete.cases(.$nature.claim), ] %>%
    .[, nature.claim:= trimws(nature.claim)] %>%
    .[, .N, nature.claim] %>%
    ggplot(., aes(reorder(nature.claim, -N), N)) +
    scale_x_discrete(guide = guide_axis(n.dodge = 2)) +
```

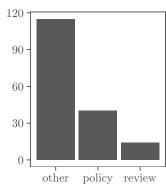
```
geom_bar(stat = "identity") +
labs(x = "", y = "") +
theme_AP()

c <- tmp$NETWORK %>%
    .[, .N, document.type] %>%
    ggplot(., aes(reorder(document.type, -N), N)) +
    geom_bar(stat = "identity") +
    labs(x = "", y = "") +
    theme_AP()

plot_grid(a, b, c, ncol = 3)
```








```
## [1] 0.005838807
```

```
network_metrics <- data.table(node = V(citation_graph)$name,

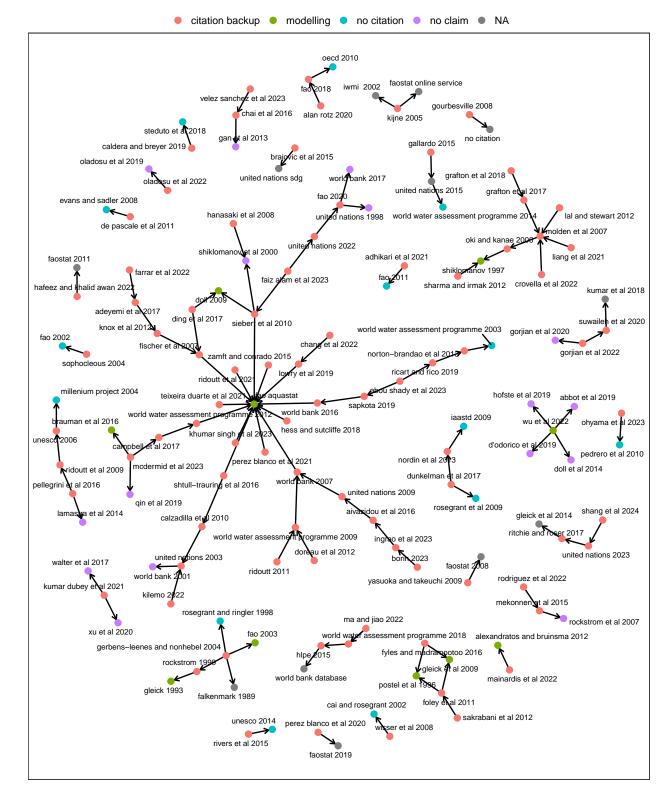
# Degree of a node: The number of connections or
# edges linked to that node.</pre>
```

```
# It represents how well-connected or central a
                               # node is within the graph.
                               degree = degree(citation_graph, mode = "in"),
                               # Betweenness centrality of a node: Measures the
                               # extent to which a node lies on the shortest
                               # paths between all pairs of other nodes in the graph.
                               # Nodes with high betweenness centrality act as
                               # bridges or intermediaries, facilitating
                               # communication and information flow between other nodes.
                               betweenness = betweenness(citation_graph),
                               # Closeness centrality of a node: Measures how
                               # close a node is to all other nodes in the graph,
                               # taking into account the length of the shortest paths.
                               # Nodes with high closeness centrality are able to
                               # efficiently communicate or interact with other
                               # nodes in the graph.
                               closeness = closeness(citation_graph),
                               pagerank = page_rank(citation_graph)$vector
network_metrics[order(-degree)][1:5]
##
                                        node degree betweenness closeness
##
                                      <char>
                                              <n11m>
                                                           <n11m>
                                                                     <niim>
## 1:
                                                                       NaN
                                fao aquastat
                                                 13
                                                              10 0.3333333
## 2:
                          molden et al 2007
                                                  4
## 3: world water assessment programme 2009
                                                  2
                                                              4 0.3333333
                                                  2
                                                              7 1.0000000
                             world bank 2007
## 5: world water assessment programme 2018
                                                  2
                                                               2 0.3333333
##
         pagerank
##
            <num>
## 1: 0.079718516
## 2: 0.020085501
## 3: 0.010586794
## 4: 0.023540448
## 5: 0.007253915
network metrics[order(-betweenness)][1:5]
##
                      node degree betweenness closeness
                                                            pagerank
##
                             <num>
                    <char>
                                         <num>
                                                    <num>
                                                               <num>
## 1:
         molden et al 2007
                                            10 0.3333333 0.02008550
## 2:
           world bank 2007
                                 2
                                             7 1.0000000 0.02354045
                                             6 0.1666667 0.01008686
## 3: aivazidou et al 2016
                                 1
## 4: united nations 2009
                                 1
                                             6 0.3333333 0.01249487
## 5:
        oki and kanae 2006
                                             6 1.0000000 0.02099371
```

```
network metrics[order(-closeness)][1:5]
##
                        node degree betweenness closeness
                                                        pagerank
                       <char> <num>
##
                                       <num>
                                                <num>
                                                          <num>
## 1:
         sharma and irmak 2012
                                           0
                                                   1 0.003921035
                                0
               world bank 2007
                                 2
                                           7
                                                   1 0.023540448
## 2:
## 3:
           brajovic et al 2015
                                 0
                                           0
                                                   1 0.003921035
## 4:
             rivers et al 2015
                                                   1 0.003921035
                                 0
## 5: hafeez and khalid awan 2022
                                 0
                                           0
                                                   1 0.003921035
# Retrieve a vector with the node names -----
graph <- tidygraph::as_tbl_graph(network.dt.complete, directed = TRUE)</pre>
vec.names <- graph %>%
 activate(nodes) %>%
 pull() %>%
 data.table(name = .)
# Merge with info from the network.dt -----
vec.nature.claim <- merge(vec.names, unique(network.dt[, .(from, nature.claim)]),</pre>
     by.x = "name", by.y = "from", all.x = TRUE)
# Merge with the correct order ------
order_indices <- match(vec.names$name, vec.nature.claim$name)
final.vec.nature.claim <- vec.nature.claim[order_indices, ] %>%
 .[, nature.claim]
# Attach to the graph ------
graph <- graph %>%
 activate(nodes) %>%
 mutate(nature.claim = final.vec.nature.claim)
# Plot network ------
ggraph(graph, layout = "igraph", algorithm = "nicely") +
 geom_edge_link(arrow = arrow(length = unit(1.5, 'mm')),
              end_cap = circle(1, "mm")) +
 geom_node_point(size = 2, aes(color = nature.claim)) +
 geom_node_text(aes(label = name), repel = TRUE, size = 2.1) +
 labs(x = "", y = "") +
 scale_color_discrete(name = "") +
 theme AP() +
 theme(axis.text.x = element_blank(),
```

```
axis.ticks.x = element_blank(),
    axis.text.y = element_blank(),
    axis.ticks.y = element_blank(),
    legend.position = "top")

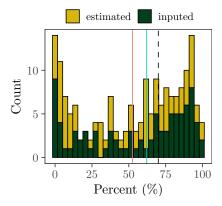
## Warning: Using the `size` aesthetic in this geom was deprecated in ggplot2 3.4.0.
## i Please use `linewidth` in the `default_aes` field and elsewhere instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
```



4 Aquastat data

```
aquastat <- data.table(read.xlsx("aquastat_dt.xlsx"))</pre>
aquastat.dt <- aquastat[Variable == "Agricultural water withdrawal as % of total water withdrawal water withdrawal as % of total water withdrawal water water withdrawal water water
     .[, .(Area, Year, Value, Symbol)] %>%
     .[Year == 2020] \%
     .[, Symbol:= ifelse(Symbol == "E", "estimated", "inputed")]
# Count proportion of estimated and imputed values
# E: Estimate either calculated as sum or identify (yield) from
# official values or from an AQUASTAT estimation.
# I: Imputed (carry forward, vertical imputation, linear interpolation)
# Calculate mean and median -----
mean.and.median <- aquastat.dt[, .(mean = mean(Value), median = median(Value))] %>%
    melt(., measure.vars = colnames(.))
# Plot histogram -----
a <- ggplot(aquastat.dt, aes(Value)) +
     geom_histogram(color = "black", aes(fill = Symbol)) +
    geom_vline(data = mean.and.median, aes(xintercept = value, color = variable),
                              show.legend = FALSE) +
    geom vline(xintercept = 70, lty = 2) +
     scale_fill_manual(name = "",
                                               values = wes_palette(name = "Cavalcanti1", 5)) +
    theme AP() +
    labs(x = "Percent (\\\\)", y = "Count") +
    theme(legend.position = "top")
```





5 Session information

[46] yaml_2.3.7

```
sessionInfo()
## R version 4.3.2 (2023-10-31)
## Platform: aarch64-apple-darwin20 (64-bit)
## Running under: macOS Sonoma 14.2.1
##
## Matrix products: default
          /Library/Frameworks/R.framework/Versions/4.3-arm64/Resources/lib/libRblas.0.dylib
## BLAS:
## LAPACK: /Library/Frameworks/R.framework/Versions/4.3-arm64/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
## time zone: Europe/London
## tzcode source: internal
## attached base packages:
## [1] parallel stats
                          graphics grDevices utils
                                                       datasets methods
## [8] base
##
## other attached packages:
   [1] wesanderson_0.3.6 benchmarkme_1.0.8 tidygraph_1.3.0
                                                               cowplot_1.1.1
## [5] ggraph_2.1.0
                          igraph_1.6.0
                                             bibliometrix_4.0.1 lubridate_1.9.2
## [9] forcats_1.0.0
                          stringr_1.5.1
                                            dplyr_1.1.4
                                                               purrr_1.0.2
## [13] readr_2.1.4
                          tidyr_1.3.0
                                            tibble_3.2.1
                                                               ggplot2_3.4.4
## [17] tidyverse_2.0.0
                          data.table_1.14.99 openxlsx_4.2.5.2
##
## loaded via a namespace (and not attached):
     [1] Rdpack_2.6
##
                               gridExtra_2.3
                                                     readxl_1.4.2
     [4] rlang_1.1.3
                               magrittr_2.0.3
                                                     tidytext_0.4.1
##
     [7] compiler_4.3.2
                                                     crayon_1.5.2
##
                               vctrs_0.6.5
   [10] pkgconfig_2.0.3
                               fastmap_1.1.1
                                                     ellipsis_0.3.2
##
   [13] labeling_0.4.3
                               utf8_1.2.4
                                                     promises_1.2.0.1
## [16] rmarkdown_2.21
                               tzdb_0.3.0
                                                     tinytex_0.45
## [19] bit_4.0.5
                               xfun_0.39
                                                      jsonlite_1.8.4
## [22] flashClust_1.01-2
                                                     SnowballC_0.7.1
                               highr_0.10
## [25] later_1.3.0
                               tweenr_2.0.2
                                                     cluster_2.1.4
## [28] R6_2.5.1
                               stringi_1.8.3
                                                     RColorBrewer_1.1-3
## [31] cellranger_1.1.0
                               estimability_1.4.1
                                                     iterators_1.0.14
## [34] Rcpp_1.0.12
                               knitr_1.42
                                                     filehash_2.4-5
## [37] httpuv_1.6.9
                               rentrez_1.2.3
                                                     Matrix_1.6-1.1
## [40] timechange_0.2.0
                                                     viridis_0.6.4
                               tidyselect_1.2.0
## [43] rstudioapi_0.15.0
                               stringdist_0.9.10
                                                     pubmedR_0.0.3
```

codetools_0.2-19

doParallel_1.0.17

```
[49] lattice_0.21-9
                                                       shiny_1.7.4
##
                                plyr_1.8.8
   [52] withr_3.0.0
##
                                benchmarkmeData_1.0.4
                                                       coda_0.19-4
##
    [55] evaluate_0.20
                                polyclip_1.10-6
                                                       zip_2.3.0
    [58] pillar_1.9.0
                                janeaustenr_1.0.0
                                                       foreach_1.5.2
##
                                plotly_4.10.1
    [61] DT 0.27
                                                       generics 0.1.3
##
    [64] vroom_1.6.1
                                hms_1.1.3
                                                       munsell_0.5.0
##
##
    [67] scales 1.3.0
                                sensobol_1.1.4
                                                       xtable 1.8-4
##
    [70] leaps_3.1
                                glue_1.7.0
                                                       tikzDevice_0.12.4
## [73] emmeans_1.8.5
                                scatterplot3d_0.3-43
                                                       lazyeval_0.2.2
##
    [76] tools_4.3.2
                                tokenizers_0.3.0
                                                       mvtnorm_1.1-3
    [79] graphlayouts_1.0.2
                                XML_3.99-0.14
                                                       grid_4.3.2
##
    [82] rbibutils_2.2.16
                                rscopus_0.6.6
                                                       colorspace_2.1-0
##
                                                       bibliometrixData_0.3.0
    [85] dimensionsR_0.0.3
                                ggforce_0.4.1
##
    [88] cli_3.6.2
##
                                fansi_1.0.6
                                                       viridisLite_0.4.2
## [91] gtable_0.3.4
                                digest_0.6.34
                                                       ggrepel_0.9.5
## [94] FactoMineR_2.8
                                htmlwidgets_1.6.2
                                                       farver_2.1.1
## [97] htmltools_0.5.5
                                factoextra_1.0.7
                                                       lifecycle_1.0.4
                                multcompView_0.1-9
## [100] httr_1.4.5
                                                       mime_0.12
## [103] bit64_4.0.5
                                MASS_7.3-60
## Return the machine CPU
               "); print(get_cpu()$model_name)
cat("Machine:
## Machine:
## [1] "Apple M1 Max"
## Return number of true cores
cat("Num cores: "); print(detectCores(logical = FALSE))
## Num cores:
## [1] 10
## Return number of threads
cat("Num threads: "); print(detectCores(logical = FALSE))
## Num threads:
## [1] 10
```