

# Co-financing model code documentation

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# Setup

Loading packages.

```
library(tidyverse)
```

```
## -- Attaching packages -----  
  
## v ggplot2 3.2.0      v purrr  0.3.2  
## v tibble  2.1.3      v dplyr  0.8.3  
## v tidyr   0.8.3      v stringr 1.4.0  
## v readr   1.3.1      v forcats 0.4.0  
  
## -- Conflicts -----  
## x dplyr::filter() masks stats::filter()  
## x dplyr::lag()    masks stats::lag()
```

## Some useful functions (for later)

Set some global values.

```
countries_model <- c("Ghana", "India", "Kenya",  
                     "Nigeria", "Angola", "Senegal")
```

Capitalise first letter function (to be used later).

```
firstup <- function(x) {  
  substr(x, 1, 1) <- toupper(substr(x, 1, 1))  
  x  
}
```

Round up nicely for plotting (to be used later).

```
roundUp <- function(x, nice_val=c(1, 2, 3, 4, 5, 6, 7, 8, 9, 10)) {  
  if(length(x) != 1) stop("'x' can only take vectors of length 1")  
  10^floor(log10(x)) * nice_val[[which(x <= 10^floor(log10(x)) * nice_val)[[1]]]]  
}
```

## Donor-country (DC) model

Load the DC model function.

```
source("R/DCmodel.R")
```

### Examples using the DC model

```
# Input one vaccine and use the threshold input
```

```
DCmodel(cost = 10000,  
        benefit = 15,  
        intervention = "vaccine 1",  
        threshold = 500)
```

```
## # A tibble: 1 x 7
```

```
##   intervention cost benefit   cer threshold domestic donor  
##   <chr>         <dbl>  <dbl> <dbl>    <dbl>    <dbl> <dbl>  
## 1 vaccine 1    10000     15   667     500     7500  2500
```

```
# Input one vaccine and use gdp_per_capita and
```

```
#   gdp_threshold_multiple
```

```
DCmodel(cost = 1000,  
        benefit = 20,  
        intervention = "vaccine 1",  
        gdp_per_capita = 500,  
        gdp_threshold_multiple = 1.5) %>%
```

```
# Dropping gdp_threshold_multiple from output so see entire table
```

```
#   in RMarkdown document
```

```
select(-gdp_threshold_multiple)
```

```
## # A tibble: 1 x 8
```

```
##   intervention cost benefit   cer gdp_per_capita threshold domestic donor  
##   <chr>         <dbl>  <dbl> <dbl>    <dbl>    <dbl>    <dbl> <dbl>  
## 1 vaccine 1    1000     20   50      500      750    1000    0
```

```
# Input several vaccines
```

```
DCmodel(cost = c(1.5, 4, 0.8, 10) * 1000000,  
        benefit = c(2000, 2500, 1500, 1000),  
        intervention = c("v1", "v2", "v3", "v4"),  
        gdp_per_capita = 500,  
        gdp_threshold_multiple = 1.5) %>%
```

```
# Dropping gdp_threshold_multiple and gdp_per_capita from output so see entire
```

```
#   table in RMarkdown document
```

```
select(-gdp_threshold_multiple, -gdp_per_capita)
```

```
## # A tibble: 4 x 7
```

```
##   intervention cost benefit   cer threshold domestic donor  
##   <chr>         <dbl>  <dbl> <dbl>    <dbl>    <dbl>    <dbl>  
## 1 v1          1500000    2000   750     750  1500000     0  
## 2 v2          4000000    2500  1600     750  1875000 2125000  
## 3 v3           800000    1500   533     750   800000     0  
## 4 v4         10000000    1000 10000     750   750000 9250000
```

```
# Input one vaccine and several thresholds
DCmodel(cost = 10000,
        benefit = 15,
        intervention = "vaccine 1",
        gdp_per_capita = rep(500, 4), # rep repeats 500 four times,
        gdp_threshold_multiple = c(0.5, 1, 3, 4.5)) %>%
# Dropping gdp_threshold_multiple and gdp_per_capita from output so see entire
# table in RMarkdown document
select(-gdp_threshold_multiple, -gdp_per_capita)
```

```
## # A tibble: 4 x 7
##   intervention cost benefit cer threshold domestic donor
##   <chr>      <dbl>  <dbl> <dbl>    <dbl>    <dbl> <dbl>
## 1 vaccine 1  10000    15   667     250     3750  6250
## 2 vaccine 1  10000    15   667     500     7500  2500
## 3 vaccine 1  10000    15   667    1500    10000    0
## 4 vaccine 1  10000    15   667    2250    10000    0
```

```
# Input several vaccines and several thresholds
DCmodel(cost = c(1.5, 4, 0.8, 10) * 1000000,
        benefit = c(8000, 5000, 1500, 10000),
        intervention = c("v1", "v2", "v3", "v4"),
        gdp_per_capita = rep(500, 3),
        gdp_threshold_multiple = c(0.5, 1, 3)) %>%
# Dropping gdp_threshold_multiple and gdp_per_capita from output so see entire
# table in RMarkdown document
select(-gdp_threshold_multiple, -gdp_per_capita)
```

```
## # A tibble: 12 x 7
##   intervention cost benefit cer threshold domestic donor
##   <chr>      <dbl>  <dbl> <dbl>    <dbl>    <dbl> <dbl>
## 1 v1      1500000    8000   188     250  1500000    0
## 2 v2      4000000    5000    800     250  1250000  2750000
## 3 v3        800000    1500    533     250   375000  425000
## 4 v4     10000000   10000   1000     250  2500000  7500000
## 5 v1      1500000    8000   188     500  1500000    0
## 6 v2      4000000    5000    800     500  2500000  1500000
## 7 v3        800000    1500    533     500   750000   50000
## 8 v4     10000000   10000   1000     500  5000000  5000000
## 9 v1      1500000    8000   188    1500  1500000    0
## 10 v2     4000000    5000    800    1500  4000000    0
## 11 v3        800000    1500    533    1500   800000    0
## 12 v4     10000000   10000   1000    1500 10000000    0
```

# Data

## Load data

```
gdp_data <- read_csv("data/data_gdp_per_capita.csv")
```

```
## Parsed with column specification:
## cols(
##   Country = col_character(),
##   `2001` = col_double(),
##   `2002` = col_double(),
##   `2003` = col_double(),
##   `2004` = col_number(),
##   `2005` = col_number(),
##   `2006` = col_number(),
##   `2007` = col_number(),
##   `2008` = col_number(),
##   `2009` = col_number(),
##   `2010` = col_number(),
##   `2011` = col_number(),
##   `2012` = col_number(),
##   `2013` = col_number(),
##   `2014` = col_number(),
##   `2015` = col_number(),
##   `2016` = col_number(),
##   `2017` = col_number(),
##   `2018` = col_number()
## )
```

```
cea_data <- read_csv("data/data_cea.csv")
```

```
## Warning: Missing column names filled in: 'X15' [15], 'X16' [16],
## 'X17' [17], 'X18' [18], 'X19' [19], 'X20' [20], 'X21' [21], 'X22' [22],
## 'X23' [23], 'X24' [24]
```

```
## Parsed with column specification:
## cols(
##   .default = col_logical(),
##   Country = col_character(),
##   Vaccine = col_character(),
##   `Indicator (e.g. deaths averted)` = col_character(),
##   `Total cost for the period` = col_number(),
##   `Costs per year` = col_number(),
##   `Total benefit for the period` = col_number(),
##   `Benefits per year` = col_number(),
##   `Cost-effectiveness` = col_number(),
##   `Start year` = col_double(),
##   `End year` = col_double(),
##   `Year of dollar rate used` = col_double(),
##   Perspective = col_character(),
##   Assumptions = col_character(),

```

```
## `Author(s)` = col_character()
## )
```

```
## See spec(...) for full column specifications.
```

```
fund_data <- read_csv("data/data_vaccine_funding.csv")
```

```
## Warning: Missing column names filled in: 'X9' [9], 'X10' [10], 'X11' [11],
## 'X12' [12], 'X13' [13], 'X14' [14], 'X15' [15]
```

```
## Parsed with column specification:
## cols(
##   Country = col_character(),
##   Vaccine = col_character(),
##   Year = col_double(),
##   Total = col_double(),
##   `Domestic contribution ($)` = col_double(),
##   `Gavi contribution ($)` = col_double(),
##   `Domestic (%)` = col_double(),
##   `Gavi (%)` = col_double(),
##   X9 = col_logical(),
##   X10 = col_logical(),
##   X11 = col_logical(),
##   X12 = col_logical(),
##   X13 = col_logical(),
##   X14 = col_logical(),
##   X15 = col_logical()
## )
```

```
gdp_deflator <- read_csv("data/gdp_deflator.csv")
```

```
## Parsed with column specification:
## cols(
##   .default = col_double(),
##   `Country Name` = col_character(),
##   `Country Code` = col_character(),
##   `Indicator Name` = col_character(),
##   `Indicator Code` = col_character(),
##   `2019` = col_logical()
## )
## See spec(...) for full column specifications.
```

```
ex_rate <- read_csv("data/ex_rate.csv")
```

```
## Parsed with column specification:
## cols(
##   .default = col_double(),
##   `Country Name` = col_character(),
##   `Country Code` = col_character(),
##   `Indicator Name` = col_character(),
##   `Indicator Code` = col_character(),
```

```
## `2019` = col_logical()
## )
## See spec(...) for full column specifications.

population_data <- read_csv("data/population_data.csv")
```

```
## Parsed with column specification:
## cols(
##   `Country Name` = col_character(),
##   `2012` = col_double(),
##   `2013` = col_double(),
##   `2014` = col_double(),
##   `2015` = col_double(),
##   `2016` = col_double(),
##   `2017` = col_double(),
##   `2018` = col_double()
## )
```

## Clean data

### Population data

```
# Change to long format data
population_data <-
  population_data %>%
  rename(country = `Country Name`) %>%
  gather(key = year, value = population, `2012`:`2018`) %>%
  mutate(year = as.numeric(year)) %>%
  filter(country %in% countries_model)
```

### Gross domestic product deflator and exchange rate

Clean and combine Gross domestic product (GDP) deflator and exchange rate data.

```
# Clean and remove unnecessary columns
gdp_deflator <- gdp_deflator %>% rename(country = `Country Name`) %>%
  select(country, `2000`:`2018`) %>%
  filter(country %in% countries_model) %>%
  gather(key = year, value = deflator, `2000`:`2018`)

# Add 2018 year to get multiplier of def_2018/def_year
gdp_deflator <- left_join(gdp_deflator,
  gdp_deflator %>% filter(year == 2018) %>% select(-year) %>%
    rename(deflator2018 = deflator),
  by = "country") %>%
  mutate(def_multiplier = deflator2018/deflator) %>%
  select(country, year, def_multiplier)

# Clean exchange rate tibble
ex_rate <- ex_rate %>% rename(country = `Country Name`) %>%
```

```

select(country, `2000`:`2018`) %>%
filter(country %in% countries_model) %>%
gather(key = year, value = ex, `2000`:`2018`)

# Add exchange rate multiplier to get ex_2018/ex_year
# multiplying by both multipliers will give
# to local_currency = reported$$ * ex
# to 2018_local_currency = local_currency * def_2018/def
# to 2018_$ = 2018_local_currency / ex_2018
# i.e. (reported$$ * ex * def_2018/def) / ex_2018 =
# reported$$ * def_2018/def * ex/ex_2018
ex_rate <- left_join(ex_rate,
  ex_rate %>% filter(year == 2018) %>% select(-year) %>%
  rename(ex2018 = ex),
  by = "country") %>%
mutate(ex_multiplier = ex/ex2018) %>%
select(country, year, ex_multiplier)

# Create currency exchange deflating and exchanging
currency_conv <- left_join(gdp_deflator, ex_rate, by = c("country", "year")) %>%
mutate(currency_mult = def_multiplier * ex_multiplier) %>%
mutate(dollar_year = as.numeric(year)) %>%
select(country, dollar_year, currency_mult)

```

## Cost-effectiveness data and GDP per capita

Clean the cost-effectiveness analysis (CEA) data.

```

cea_data <- cea_data %>%
# Rename to more data-friendly
rename(country = Country, vaccine = Vaccine,
  indicator = `Indicator (e.g. deaths averted)`,
  cost_for_horizon = `Total cost for the period`,
  cost = `Costs per year`,
  benefit_for_horizon = `Total benefit for the period`,
  benefit = `Benefits per year`, cea = `Cost-effectiveness`,
  s_year = `Start year`, e_year = `End year`,
  dollar_year = `Year of dollar rate used`,
  perspective = Perspective, assumptions = Assumptions,
  authors = `Author(s)`) %>%
select(country:authors) %>%
# Consistent naming of indicators
mutate(indicator = firstup(indicator)) %>%
mutate(indicator = gsub("DALY ", "DALYs ", indicator)) %>%
mutate(indicator = gsub("Death ", "Deaths ", indicator)) %>%
mutate(indicator = gsub("Case ", "Cases ", indicator)) %>%
mutate(indicator = gsub("YLL", "YLLs averted", indicator)) %>%
# Some of the vaccine names differ between fund_data and cea_data
mutate(vaccine = ifelse(grepl("DTP-Hep", vaccine, ignore.case = TRUE),
  "DTP-hep B",
  vaccine)) %>%
mutate(vaccine = ifelse(grepl("Yellow Fever", vaccine),

```



```

        "Yellow Fever",
        vaccine)) %>%
mutate(vaccine = ifelse(grepl("Measles", vaccine),
        "Measles",
        vaccine)) %>%
mutate(vaccine = ifelse(grepl("Pneumococcal", vaccine),
        "PCV",
        vaccine)) %>%
mutate(vaccine = ifelse(grepl("Rota", vaccine),
        "Rotavirus",
        vaccine)) %>%
mutate(vaccine = ifelse(grepl("BCG", vaccine),
        "BCG",
        vaccine)) %>%
# Select only countries in the analysis
filter(country %in% countries_model)

```

Perform some checks on CEA data.

```

# Check diff between calculated and CEA reported in papers
# Small differences are likely due to rounding errors
cea_data %>% select(cost, benefit, cea) %>%
  mutate(cea2 = round(cost/benefit,2)) %>%
  mutate(diff = cea - cea2) %>%
  arrange(desc(abs(diff))) %>%
  filter(abs(diff) > 1e-09)

```

```

## # A tibble: 19 x 5
##       cost benefit    cea  cea2    diff
##       <dbl>   <dbl> <dbl> <dbl>   <dbl>
## 1  3689149     303 12196. 12175.  20.1
## 2  1409658     303  4660.  4652.   7.69
## 3   6409006     634 10102. 10109.  -6.37
## 4  11783120     888 13274. 13269.   4.48
## 5  18747118    1303 14392. 14388.   4.01
## 6   3721375     634  5866.  5870.  -3.7
## 7   1505569     311  4844.  4841.   3.11
## 8   7390364     888  8325.  8322.   2.81
## 9   1083447     311  3486.  3484.   2.24
## 10  379910     311  1222.  1222.   0.780
## 11 176198173  10601 16621. 16621.   0.470
## 12 176198173  10601 16621. 16621.   0.470
## 13  34223066   9193  3723.  3723.  -0.160
## 14   7766946   3053  2544.  2544.   0.160
## 15   7766946   3053  2544.  2544.   0.160
## 16  28770592   9193  3129.  3130.  -0.140
## 17  18747118   9228  2031.  2032.  -0.1000
## 18  3200000   12570   255.   255.   0.01
## 19   3721375  33380   111.   111.  -0.01000

```

Transform GDP per capita data to long format, convert to \$2018, and add cost & benefit columns.

```

# Max cost +, to create a sequence for plotting
max_x <- max(cea_data$cost) + 50000000
seq_x <- seq(0, max_x, by = 100000)

# Convert GDP data to 2018 $
gdp_data <- gdp_data %>%
  gather(key = "year", value = "gdp_pc", -Country) %>%
  rename(country = Country) %>%
  mutate(year = as.numeric(year)) %>%
  # Convert currency
  left_join(currency_conv %>% rename(year = dollar_year),
            by = c("country", "year")) %>%
  mutate(gdp_pc = gdp_pc * currency_mult) %>%
  select(-currency_mult)

# Add cost and benefit columns to gdp_data to
# draw the CEA threshold
gdp_data <- gdp_data %>%
  # Create rows for each analysis
  slice(rep(row_number(), length(seq_x))) %>%
  # Add a cost column, for plotting cost x-axis
  mutate(cost = unlist(sapply(seq_x, function(x) rep(x, nrow(gdp_data)), simplify = FALSE))) %>%
  # Add a number of benefit lines for plotting
  mutate(benefit_line = cost/gdp_pc,
         benefit_half_line = cost/gdp_pc/0.5,
         benefit_3_line = cost/gdp_pc/3,
         benefit_4_5_line = cost/gdp_pc/4.5)

# Convert to long data tibble
gdp_data <- gdp_data %>%
  gather(key = threshold, value = cea_line, benefit_line:benefit_4_5_line) %>%
  # Write clear headings for different thresholds
  mutate(threshold = ifelse(threshold == "benefit_line",
                           "1 x GDP\nper capita",
                           threshold)) %>%
  mutate(threshold = ifelse(threshold == "benefit_half_line",
                           "0.5 x GDP\nper capita",
                           threshold)) %>%
  mutate(threshold = ifelse(threshold == "benefit_3_line",
                           "3 x GDP\nper capita",
                           threshold)) %>%
  mutate(threshold = ifelse(threshold == "benefit_4_5_line",
                           "4.5 x GDP\nper capita",
                           threshold))

```

Convert the CEA data into 2018 US Dollars using WB deflator.

```

# Combine CEA data and currency conversion data
cea_data <- left_join(cea_data, currency_conv, by = c("country", "dollar_year")) %>%
  # Convert all costs
  mutate(cost_for_horizon = cost_for_horizon * currency_mult,
         cost = cost * currency_mult,
         cea = cea * currency_mult,
         benefit = ifelse(

```

```

grepl("Total healthcare cost averted|Out-of-pocket expenditure",
      indicator),
benefit * currency_mult,
benefit))

```

## Co-financing data

Clean co-financing data.

```

fund_data <-
fund_data %>%
# Rename columns so they are easier to work with.
rename(country = Country, vaccine = Vaccine, year = Year,
        total = Total,
        domestic = `Domestic contribution ($)` ,
        gavi = `Gavi contribution ($)` ,
        domestic_p = `Domestic (%)` , gavi_p = `Gavi (%)` ) %>%
# Remove some columns that are unnecessary
select(country:gavi_p) %>%
# Remove rows with countries that are not part of the analysis (there are some
# unnecessary rows in the data that explain the data)
filter(country %in% countries_model) %>%
# Convert numbers to numeric
mutate(year = as.numeric(year),
        total = as.numeric(gsub("\\\\", "", total)),
        domestic = as.numeric(gsub("\\\\", "", domestic)),
        gavi = as.numeric(gsub("\\\\", "", gavi)),
        domestic_p = as.numeric(domestic_p)/100,
        gavi_p = as.numeric(gavi_p)/100) %>%
# Convert currency using currency conversion table
left_join(currency_conv %>% rename(year = dollar_year),
          by = c("country", "year")) %>%
mutate(total = total * currency_mult,
        domestic = domestic * currency_mult,
        gavi = gavi * currency_mult) %>%
select(-currency_mult) %>%
# Some of the vaccine names differ between fund_data and cea_data;
# fix.
mutate(vaccine = ifelse(grepl("DTP-Hep", vaccine),
                        "DTP-hep B",
                        vaccine)) %>%
mutate(vaccine = ifelse(grepl("Yellow Fever", vaccine),
                        "Yellow Fever",
                        vaccine)) %>%
mutate(vaccine = ifelse(grepl("Measles", vaccine),
                        "Measles",
                        vaccine)) %>%
mutate(vaccine = ifelse(grepl("Pneumococcal", vaccine),
                        "PCV",
                        vaccine)) %>%
mutate(vaccine = ifelse(grepl("Rota", vaccine),
                        "Rotavirus",
                        vaccine))

```

## Combine CEA and funding data into a tibble

```
# Check what vaccines in CEA data are not in co-financing data
temp <- (cea_data %>% distinct(vaccine) %>% unlist) %in%
(fund_data %>% distinct(vaccine) %>% unlist)
(cea_data %>% distinct(vaccine) %>% unlist)[!temp]
```

```
## vaccine3
##      "BCG"
```

```
# Check what vaccines in co-financing data are not in CEA data
temp <- (fund_data %>% distinct(vaccine) %>% unlist) %in%
(cea_data %>% distinct(vaccine) %>% unlist)
(fund_data %>% distinct(vaccine) %>% unlist)[!temp]
```

```
##                vaccine6                vaccine7                vaccine8
##                "MenA" "MR Follow up campaign"                "MenA Routine"
##                vaccine9
##                "IPV"
```

```
# Combine CEA and co-financing data by country and vaccine.
# Include all of the vaccines we have in the CEA analysis data,
# but not ones we have only in the co-financing data.
```

```
data <- full_join(cea_data,
                  fund_data %>% rename(year_gavi = year),
                  by = c("country", "vaccine"))
```

```
# Add BCG vaccine, not supported by Gavi for plotting
```

```
BCG_data <- data %>%
  # Get BCG data; at this point only includes Levin et al 2007
  filter(vaccine == "BCG") %>%
  # Copy rows years to create years 2006-2018
  slice(rep(row_number(), length(2006:2018))) %>%
  arrange(indicator) %>%
  mutate(year_gavi = rep(2006:2018, 3)) %>%
  # Remove 2015 since do not have data
  filter(year_gavi != 2015)
```

```
# Add the BCG data
```

```
data <- data %>%
  # Remove the NA BCG data row
  filter(vaccine != "BCG") %>%
  # Add rest of BCG data
  rbind(BCG_data)
```

Clean combined data.

```
# Remove non-base-case
data <- data %>%
  # Include only DALY or YLL
  filter(grepl("DALY|YLL", indicator)) %>%
```

```

# Filter years that do not have any Gavi data or only for
# a few countries and based on future projections
# First add a year for BCG, so plot despite no Gavi
# funding.
mutate(year_gavi = ifelse(is.na(year_gavi) &
                          authors == "Levin et al. 2007" &
                          vaccine == "BCG",
                          2007,
                          year_gavi)) %>%
filter(year_gavi < 2019) %>%
# Only include base-case of Abbot et al. 2012
# and use earliest year of Gavi funding (since intervention is 2003-08).
filter(!(authors == "Abbott et al. 2012" &
            (assumptions != "$5 per dose"))) %>% # | year_gavi != 2012))) %>%
# Do not include the CEA with Gavi Subsidy accounted for
# and focus on health system perspective.
filter(!(authors == "Nonvignon et al. 2018" &
            (grepl("Ghana only", assumptions) |
             !grepl("Health system", perspective)))) %>%
# Use relevant Gavi data year for Krishnamoorthy et al. 2019
# based on paper intervention years
# filter(!(authors == "Krishnamoorthy et al. 2019" &
#             year_gavi != 2018)) %>%
# Use the analysis with the 90% intervention coverage and year 2018
# for Gavi funding based on paper intervention years
filter(!(authors == "Megiddo et al. 2018" &
            (grepl("Expanded coverage (90%)", assumptions)))) %>%
# Do not include the CEA with Gavi subsidy accounted for
filter(!(authors == "Debellut et al., 2019" &
            (assumptions == "Vaccine programme costs with Gavi subsidy"))) %>%
# Remove rows with no co-financing data
filter(!(is.na(total) & (!is.na(gavi) | !is.na(domestic)))) %>%
# Add a column with vaccine and paper
mutate(vaccine_cea_paper = paste0(vaccine, "\n", authors)) %>%
# Remove columns that do not need for figures
select(-cost_for_horizon, -benefit_for_horizon, -currency_mult)

```

## DC model with data

In the following code, we use the CEA study data for costs and benefits. We use the DC model to calculate the split between domestic and donor funding.

```

# Ochalek et al. 2018 proportion
ochalek_p <- 0.33

# Ghana data
ghana_data <-
  data %>%
  filter(country == "Ghana") %>%
  distinct(vaccine, cost, benefit)

# GDP per capita data

```

```

ghana_gdppc <-
  gdp_data %>%
  filter(country == "Ghana") %>%
  distinct(year, gdp_pc)

# Run DC model
DCmodel_out <- DCmodel(
  cost = ghana_data$cost,
  benefit = ghana_data$benefit,
  intervention = ghana_data$vaccine,
  gdp_per_capita = rep(ghana_gdppc$gdp_pc, 5),
  gdp_threshold_multiple = c(rep(ochalek_p, nrow(ghana_gdppc)),
                             rep(0.5, nrow(ghana_gdppc)),
                             rep(1, nrow(ghana_gdppc)),
                             rep(3, nrow(ghana_gdppc)),
                             rep(4.5, nrow(ghana_gdppc)))
)

# Print output
DCmodel_out %>%
  left_join(ghana_gdppc %>% rename(gdp_per_capita = gdp_pc),
            by = "gdp_per_capita") %>%
  # Remove rotavirus vaccines before 2012
  filter(!(intervention == "Rotavirus" & year < 2012)) %>%
  select(intervention, year, cer,
         gdp_threshold_multiple, threshold, domestic, donor) %>%
  print(n = nrow(DCmodel_out))

```

```

## # A tibble: 465 x 7
##   intervention year   cer gdp_threshold_mul~ threshold domestic donor
##   <chr>         <dbl> <dbl>         <dbl>         <dbl>         <dbl> <dbl>
## 1 Measles      2001  3397           0.33          388  5276800  4.09e7
## 2 DTP-hep B    2001  2665           0.33          388 13980804  8.21e7
## 3 Yellow Fever 2001   176           0.33          388  4719371.  0.
## 4 BCG          2001 154221         0.33          388    62080  2.46e7
## 5 Measles      2002  3397           0.33          396  5385600  4.08e7
## 6 DTP-hep B    2002  2665           0.33          396 14269068  8.18e7
## 7 Yellow Fever 2002   176           0.33          396  4719371.  0.
## 8 BCG          2002 154221         0.33          396    63360  2.46e7
## 9 Measles      2003  3397           0.33          407  5535200  4.07e7
## 10 DTP-hep B   2003  2665           0.33          407 14665431  8.14e7
## 11 Yellow Fever 2003   176           0.33          407  4719371.  0.
## 12 BCG         2003 154221         0.33          407    65120  2.46e7
## 13 Measles     2004  3397           0.33          419  5698400  4.05e7
## 14 DTP-hep B   2004  2665           0.33          419 15097827  8.09e7
## 15 Yellow Fever 2004   176           0.33          419  4719371.  0.
## 16 BCG         2004 154221         0.33          419    67040  2.46e7
## 17 Measles     2005  3397           0.33          433  5888800  4.03e7
## 18 DTP-hep B   2005  2665           0.33          433 15602289  8.04e7
## 19 Yellow Fever 2005   176           0.33          433  4719371.  0.
## 20 BCG         2005 154221         0.33          433    69280  2.46e7
## 21 Measles     2006  3397           0.33          449  6106400  4.01e7

```

##	22	DTP-hep B	2006	2665	0.33	449	16178817	7.99e7
##	23	Yellow Fever	2006	176	0.33	449	4719371.	0.
##	24	BCG	2006	154221	0.33	449	71840	2.46e7
##	25	Measles	2007	3397	0.33	456	6201600	4.00e7
##	26	DTP-hep B	2007	2665	0.33	456	16431048	7.96e7
##	27	Yellow Fever	2007	176	0.33	456	4719371.	0.
##	28	BCG	2007	154221	0.33	456	72960	2.46e7
##	29	Measles	2008	3397	0.33	485	6596000	3.96e7
##	30	DTP-hep B	2008	2665	0.33	485	17476005	7.86e7
##	31	Yellow Fever	2008	176	0.33	485	4719371.	0.
##	32	BCG	2008	154221	0.33	485	77600	2.46e7
##	33	Measles	2009	3397	0.33	496	6745600	3.94e7
##	34	DTP-hep B	2009	2665	0.33	496	17872368	7.82e7
##	35	Yellow Fever	2009	176	0.33	496	4719371.	0.
##	36	BCG	2009	154221	0.33	496	79360	2.46e7
##	37	Measles	2010	3397	0.33	522	7099200	3.91e7
##	38	DTP-hep B	2010	2665	0.33	522	18809226	7.72e7
##	39	Yellow Fever	2010	176	0.33	522	4719371.	0.
##	40	BCG	2010	154221	0.33	522	83520	2.46e7
##	41	Measles	2011	3397	0.33	581	7901600	3.83e7
##	42	DTP-hep B	2011	2665	0.33	581	20935173	7.51e7
##	43	Yellow Fever	2011	176	0.33	581	4719371.	0.
##	44	BCG	2011	154221	0.33	581	92960	2.46e7
##	45	Rotavirus	2012	113	0.33	620	2105497.	0.
##	46	Rotavirus	2012	387	0.33	620	4931933.	0.
##	47	Measles	2012	3397	0.33	620	8432000	3.78e7
##	48	DTP-hep B	2012	2665	0.33	620	22340460	7.37e7
##	49	Yellow Fever	2012	176	0.33	620	4719371.	0.
##	50	Rotavirus	2012	215	0.33	620	7183815.	0.
##	51	BCG	2012	154221	0.33	620	99200	2.46e7
##	52	Rotavirus	2013	113	0.33	650	2105497.	0.
##	53	Rotavirus	2013	387	0.33	650	4931933.	0.
##	54	Measles	2013	3397	0.33	650	8840000	3.74e7
##	55	DTP-hep B	2013	2665	0.33	650	23421450	7.26e7
##	56	Yellow Fever	2013	176	0.33	650	4719371.	0.
##	57	Rotavirus	2013	215	0.33	650	7183815.	0.
##	58	BCG	2013	154221	0.33	650	104000	2.46e7
##	59	Rotavirus	2014	113	0.33	654	2105497.	0.
##	60	Rotavirus	2014	387	0.33	654	4931933.	0.
##	61	Measles	2014	3397	0.33	654	8894400	3.73e7
##	62	DTP-hep B	2014	2665	0.33	654	23565582	7.25e7
##	63	Yellow Fever	2014	176	0.33	654	4719371.	0.
##	64	Rotavirus	2014	215	0.33	654	7183815.	0.
##	65	BCG	2014	154221	0.33	654	104640	2.46e7
##	66	Rotavirus	2015	113	0.33	653	2105497.	0.
##	67	Rotavirus	2015	387	0.33	653	4931933.	0.
##	68	Measles	2015	3397	0.33	653	8880800	3.73e7
##	69	DTP-hep B	2015	2665	0.33	653	23529549	7.25e7
##	70	Yellow Fever	2015	176	0.33	653	4719371.	0.
##	71	Rotavirus	2015	215	0.33	653	7183815.	0.
##	72	BCG	2015	154221	0.33	653	104480	2.46e7
##	73	Rotavirus	2016	113	0.33	661	2105497.	0.
##	74	Rotavirus	2016	387	0.33	661	4931933.	0.
##	75	Measles	2016	3397	0.33	661	8989600	3.72e7

##	76 DTP-hep B	2016	2665	0.33	661	23817813	7.22e7
##	77 Yellow Fever	2016	176	0.33	661	4719371.	0.
##	78 Rotavirus	2016	215	0.33	661	7183815.	0.
##	79 BCG	2016	154221	0.33	661	105760	2.46e7
##	80 Rotavirus	2017	113	0.33	699	2105497.	0.
##	81 Rotavirus	2017	387	0.33	699	4931933.	0.
##	82 Measles	2017	3397	0.33	699	9506400	3.67e7
##	83 DTP-hep B	2017	2665	0.33	699	25187067	7.09e7
##	84 Yellow Fever	2017	176	0.33	699	4719371.	0.
##	85 Rotavirus	2017	215	0.33	699	7183815.	0.
##	86 BCG	2017	154221	0.33	699	111840	2.46e7
##	87 Rotavirus	2018	113	0.33	727	2105497.	0.
##	88 Rotavirus	2018	387	0.33	727	4931933.	0.
##	89 Measles	2018	3397	0.33	727	9887200	3.63e7
##	90 DTP-hep B	2018	2665	0.33	727	26195991	6.98e7
##	91 Yellow Fever	2018	176	0.33	727	4719371.	0.
##	92 Rotavirus	2018	215	0.33	727	7183815.	0.
##	93 BCG	2018	154221	0.33	727	116320	2.46e7
##	94 Measles	2001	3397	0.5	589	8010400	3.82e7
##	95 DTP-hep B	2001	2665	0.5	589	21223437	7.48e7
##	96 Yellow Fever	2001	176	0.5	589	4719371.	0.
##	97 BCG	2001	154221	0.5	589	94240	2.46e7
##	98 Measles	2002	3397	0.5	600	8160000	3.80e7
##	99 DTP-hep B	2002	2665	0.5	600	21619800	7.44e7
##	100 Yellow Fever	2002	176	0.5	600	4719371.	0.
##	101 BCG	2002	154221	0.5	600	96000	2.46e7
##	102 Measles	2003	3397	0.5	616	8377600	3.78e7
##	103 DTP-hep B	2003	2665	0.5	616	22196328	7.38e7
##	104 Yellow Fever	2003	176	0.5	616	4719371.	0.
##	105 BCG	2003	154221	0.5	616	98560	2.46e7
##	106 Measles	2004	3397	0.5	635	8636000	3.76e7
##	107 DTP-hep B	2004	2665	0.5	635	22880955	7.32e7
##	108 Yellow Fever	2004	176	0.5	635	4719371.	0.
##	109 BCG	2004	154221	0.5	635	101600	2.46e7
##	110 Measles	2005	3397	0.5	655	8908000	3.73e7
##	111 DTP-hep B	2005	2665	0.5	655	23601615	7.24e7
##	112 Yellow Fever	2005	176	0.5	655	4719371.	0.
##	113 BCG	2005	154221	0.5	655	104800	2.46e7
##	114 Measles	2006	3397	0.5	680	9248000	3.69e7
##	115 DTP-hep B	2006	2665	0.5	680	24502440	7.15e7
##	116 Yellow Fever	2006	176	0.5	680	4719371.	0.
##	117 BCG	2006	154221	0.5	680	108800	2.46e7
##	118 Measles	2007	3397	0.5	691	9397600	3.68e7
##	119 DTP-hep B	2007	2665	0.5	691	24898803	7.11e7
##	120 Yellow Fever	2007	176	0.5	691	4719371.	0.
##	121 BCG	2007	154221	0.5	691	110560	2.46e7
##	122 Measles	2008	3397	0.5	735	9996000	3.62e7
##	123 DTP-hep B	2008	2665	0.5	735	26484255	6.96e7
##	124 Yellow Fever	2008	176	0.5	735	4719371.	0.
##	125 BCG	2008	154221	0.5	735	117600	2.46e7
##	126 Measles	2009	3397	0.5	752	10227200	3.60e7
##	127 DTP-hep B	2009	2665	0.5	752	27096816	6.89e7
##	128 Yellow Fever	2009	176	0.5	752	4719371.	0.
##	129 BCG	2009	154221	0.5	752	120320	2.46e7



## 130 Measles	2010	3397	0.5	791	10757600	3.54e7
## 131 DTP-hep B	2010	2665	0.5	791	28502103	6.75e7
## 132 Yellow Fever	2010	176	0.5	791	4719371.	0.
## 133 BCG	2010	154221	0.5	791	126560	2.45e7
## 134 Measles	2011	3397	0.5	880	11968000	3.42e7
## 135 DTP-hep B	2011	2665	0.5	880	31709040	6.43e7
## 136 Yellow Fever	2011	176	0.5	880	4719371.	0.
## 137 BCG	2011	154221	0.5	880	140800	2.45e7
## 138 Rotavirus	2012	113	0.5	940	2105497.	0.
## 139 Rotavirus	2012	387	0.5	940	4931933.	0.
## 140 Measles	2012	3397	0.5	940	12784000	3.34e7
## 141 DTP-hep B	2012	2665	0.5	940	33871020	6.22e7
## 142 Yellow Fever	2012	176	0.5	940	4719371.	0.
## 143 Rotavirus	2012	215	0.5	940	7183815.	0.
## 144 BCG	2012	154221	0.5	940	150400	2.45e7
## 145 Rotavirus	2013	113	0.5	985	2105497.	0.
## 146 Rotavirus	2013	387	0.5	985	4931933.	0.
## 147 Measles	2013	3397	0.5	985	13396000	3.28e7
## 148 DTP-hep B	2013	2665	0.5	985	35492505	6.06e7
## 149 Yellow Fever	2013	176	0.5	985	4719371.	0.
## 150 Rotavirus	2013	215	0.5	985	7183815.	0.
## 151 BCG	2013	154221	0.5	985	157600	2.45e7
## 152 Rotavirus	2014	113	0.5	991	2105497.	0.
## 153 Rotavirus	2014	387	0.5	991	4931933.	0.
## 154 Measles	2014	3397	0.5	991	13477600	3.27e7
## 155 DTP-hep B	2014	2665	0.5	991	35708703	6.03e7
## 156 Yellow Fever	2014	176	0.5	991	4719371.	0.
## 157 Rotavirus	2014	215	0.5	991	7183815.	0.
## 158 BCG	2014	154221	0.5	991	158560	2.45e7
## 159 Rotavirus	2015	113	0.5	990	2105497.	0.
## 160 Rotavirus	2015	387	0.5	990	4931933.	0.
## 161 Measles	2015	3397	0.5	990	13464000	3.27e7
## 162 DTP-hep B	2015	2665	0.5	990	35672670	6.04e7
## 163 Yellow Fever	2015	176	0.5	990	4719371.	0.
## 164 Rotavirus	2015	215	0.5	990	7183815.	0.
## 165 BCG	2015	154221	0.5	990	158400	2.45e7
## 166 Rotavirus	2016	113	0.5	1001	2105497.	0.
## 167 Rotavirus	2016	387	0.5	1001	4931933.	0.
## 168 Measles	2016	3397	0.5	1001	13613600	3.26e7
## 169 DTP-hep B	2016	2665	0.5	1001	36069033	6.00e7
## 170 Yellow Fever	2016	176	0.5	1001	4719371.	0.
## 171 Rotavirus	2016	215	0.5	1001	7183815.	0.
## 172 BCG	2016	154221	0.5	1001	160160	2.45e7
## 173 Rotavirus	2017	113	0.5	1059	2105497.	0.
## 174 Rotavirus	2017	387	0.5	1059	4931933.	0.
## 175 Measles	2017	3397	0.5	1059	14402400	3.18e7
## 176 DTP-hep B	2017	2665	0.5	1059	38158947	5.79e7
## 177 Yellow Fever	2017	176	0.5	1059	4719371.	0.
## 178 Rotavirus	2017	215	0.5	1059	7183815.	0.
## 179 BCG	2017	154221	0.5	1059	169440	2.45e7
## 180 Rotavirus	2018	113	0.5	1101	2105497.	0.
## 181 Rotavirus	2018	387	0.5	1101	4931933.	0.
## 182 Measles	2018	3397	0.5	1101	14973600	3.12e7
## 183 DTP-hep B	2018	2665	0.5	1101	39672333	5.64e7

## 184 Yellow Fever	2018	176	0.5	1101	4719371.	0.
## 185 Rotavirus	2018	215	0.5	1101	7183815.	0.
## 186 BCG	2018	154221	0.5	1101	176160	2.45e7
## 187 Measles	2001	3397	1	1177	16007200	3.02e7
## 188 DTP-hep B	2001	2665	1	1177	42410841	5.36e7
## 189 Yellow Fever	2001	176	1	1177	4719371.	0.
## 190 BCG	2001	154221	1	1177	188320	2.45e7
## 191 Measles	2002	3397	1	1200	16320000	2.99e7
## 192 DTP-hep B	2002	2665	1	1200	43239600	5.28e7
## 193 Yellow Fever	2002	176	1	1200	4719371.	0.
## 194 BCG	2002	154221	1	1200	192000	2.45e7
## 195 Measles	2003	3397	1	1232	16755200	2.94e7
## 196 DTP-hep B	2003	2665	1	1232	44392656	5.17e7
## 197 Yellow Fever	2003	176	1	1232	4719371.	0.
## 198 BCG	2003	154221	1	1232	197120	2.45e7
## 199 Measles	2004	3397	1	1269	17258400	2.89e7
## 200 DTP-hep B	2004	2665	1	1269	45725877	5.03e7
## 201 Yellow Fever	2004	176	1	1269	4719371.	0.
## 202 BCG	2004	154221	1	1269	203040	2.45e7
## 203 Measles	2005	3397	1	1311	17829600	2.84e7
## 204 DTP-hep B	2005	2665	1	1311	47239263	4.88e7
## 205 Yellow Fever	2005	176	1	1311	4719371.	0.
## 206 BCG	2005	154221	1	1311	209760	2.45e7
## 207 Measles	2006	3397	1	1359	18482400	2.77e7
## 208 DTP-hep B	2006	2665	1	1359	48968847	4.71e7
## 209 Yellow Fever	2006	176	1	1359	4719371.	0.
## 210 BCG	2006	154221	1	1359	217440	2.45e7
## 211 Measles	2007	3397	1	1382	18795200	2.74e7
## 212 DTP-hep B	2007	2665	1	1382	49797606	4.62e7
## 213 Yellow Fever	2007	176	1	1382	4719371.	0.
## 214 BCG	2007	154221	1	1382	221120	2.45e7
## 215 Measles	2008	3397	1	1470	19992000	2.62e7
## 216 DTP-hep B	2008	2665	1	1470	52968510	4.31e7
## 217 Yellow Fever	2008	176	1	1470	4719371.	0.
## 218 BCG	2008	154221	1	1470	235200	2.44e7
## 219 Measles	2009	3397	1	1503	20440800	2.58e7
## 220 DTP-hep B	2009	2665	1	1503	54157599	4.19e7
## 221 Yellow Fever	2009	176	1	1503	4719371.	0.
## 222 BCG	2009	154221	1	1503	240480	2.44e7
## 223 Measles	2010	3397	1	1582	21515200	2.47e7
## 224 DTP-hep B	2010	2665	1	1582	57004206	3.90e7
## 225 Yellow Fever	2010	176	1	1582	4719371.	0.
## 226 BCG	2010	154221	1	1582	253120	2.44e7
## 227 Measles	2011	3397	1	1761	23949600	2.22e7
## 228 DTP-hep B	2011	2665	1	1761	63454113	3.26e7
## 229 Yellow Fever	2011	176	1	1761	4719371.	0.
## 230 BCG	2011	154221	1	1761	281760	2.44e7
## 231 Rotavirus	2012	113	1	1880	2105497.	0.
## 232 Rotavirus	2012	387	1	1880	4931933.	0.
## 233 Measles	2012	3397	1	1880	25568000	2.06e7
## 234 DTP-hep B	2012	2665	1	1880	67742040	2.83e7
## 235 Yellow Fever	2012	176	1	1880	4719371.	0.
## 236 Rotavirus	2012	215	1	1880	7183815.	0.
## 237 BCG	2012	154221	1	1880	300800	2.44e7

## 238 Rotavirus	2013	113	1	1971	2105497.	0.
## 239 Rotavirus	2013	387	1	1971	4931933.	0.
## 240 Measles	2013	3397	1	1971	26805600	1.94e7
## 241 DTP-hep B	2013	2665	1	1971	71021043	2.50e7
## 242 Yellow Fever	2013	176	1	1971	4719371.	0.
## 243 Rotavirus	2013	215	1	1971	7183815.	0.
## 244 BCG	2013	154221	1	1971	315360	2.44e7
## 245 Rotavirus	2014	113	1	1982	2105497.	0.
## 246 Rotavirus	2014	387	1	1982	4931933.	0.
## 247 Measles	2014	3397	1	1982	26955200	1.92e7
## 248 DTP-hep B	2014	2665	1	1982	71417406	2.46e7
## 249 Yellow Fever	2014	176	1	1982	4719371.	0.
## 250 Rotavirus	2014	215	1	1982	7183815.	0.
## 251 BCG	2014	154221	1	1982	317120	2.44e7
## 252 Rotavirus	2015	113	1	1980	2105497.	0.
## 253 Rotavirus	2015	387	1	1980	4931933.	0.
## 254 Measles	2015	3397	1	1980	26928000	1.93e7
## 255 DTP-hep B	2015	2665	1	1980	71345340	2.47e7
## 256 Yellow Fever	2015	176	1	1980	4719371.	0.
## 257 Rotavirus	2015	215	1	1980	7183815.	0.
## 258 BCG	2015	154221	1	1980	316800	2.44e7
## 259 Rotavirus	2016	113	1	2002	2105497.	0.
## 260 Rotavirus	2016	387	1	2002	4931933.	0.
## 261 Measles	2016	3397	1	2002	27227200	1.90e7
## 262 DTP-hep B	2016	2665	1	2002	72138066	2.39e7
## 263 Yellow Fever	2016	176	1	2002	4719371.	0.
## 264 Rotavirus	2016	215	1	2002	7183815.	0.
## 265 BCG	2016	154221	1	2002	320320	2.44e7
## 266 Rotavirus	2017	113	1	2118	2105497.	0.
## 267 Rotavirus	2017	387	1	2118	4931933.	0.
## 268 Measles	2017	3397	1	2118	28804800	1.74e7
## 269 DTP-hep B	2017	2665	1	2118	76317894	1.97e7
## 270 Yellow Fever	2017	176	1	2118	4719371.	0.
## 271 Rotavirus	2017	215	1	2118	7183815.	0.
## 272 BCG	2017	154221	1	2118	338880	2.43e7
## 273 Rotavirus	2018	113	1	2202	2105497.	0.
## 274 Rotavirus	2018	387	1	2202	4931933.	0.
## 275 Measles	2018	3397	1	2202	29947200	1.62e7
## 276 DTP-hep B	2018	2665	1	2202	79344666	1.67e7
## 277 Yellow Fever	2018	176	1	2202	4719371.	0.
## 278 Rotavirus	2018	215	1	2202	7183815.	0.
## 279 BCG	2018	154221	1	2202	352320	2.43e7
## 280 Measles	2001	3397	3	3532	46193653.	0.
## 281 DTP-hep B	2001	2665	3	3532	96045852.	0.
## 282 Yellow Fever	2001	176	3	3532	4719371.	0.
## 283 BCG	2001	154221	3	3532	565120	2.41e7
## 284 Measles	2002	3397	3	3601	46193653.	0.
## 285 DTP-hep B	2002	2665	3	3601	96045852.	0.
## 286 Yellow Fever	2002	176	3	3601	4719371.	0.
## 287 BCG	2002	154221	3	3601	576160	2.41e7
## 288 Measles	2003	3397	3	3696	46193653.	0.
## 289 DTP-hep B	2003	2665	3	3696	96045852.	0.
## 290 Yellow Fever	2003	176	3	3696	4719371.	0.
## 291 BCG	2003	154221	3	3696	591360	2.41e7

## 292 Measles	2004	3397	3	3808	46193653.	0.
## 293 DTP-hep B	2004	2665	3	3808	96045852.	0.
## 294 Yellow Fever	2004	176	3	3808	4719371.	0.
## 295 BCG	2004	154221	3	3808	609280	2.41e7
## 296 Measles	2005	3397	3	3932	46193653.	0.
## 297 DTP-hep B	2005	2665	3	3932	96045852.	0.
## 298 Yellow Fever	2005	176	3	3932	4719371.	0.
## 299 BCG	2005	154221	3	3932	629120	2.40e7
## 300 Measles	2006	3397	3	4078	46193653.	0.
## 301 DTP-hep B	2006	2665	3	4078	96045852.	0.
## 302 Yellow Fever	2006	176	3	4078	4719371.	0.
## 303 BCG	2006	154221	3	4078	652480	2.40e7
## 304 Measles	2007	3397	3	4147	46193653.	0.
## 305 DTP-hep B	2007	2665	3	4147	96045852.	0.
## 306 Yellow Fever	2007	176	3	4147	4719371.	0.
## 307 BCG	2007	154221	3	4147	663520	2.40e7
## 308 Measles	2008	3397	3	4411	46193653.	0.
## 309 DTP-hep B	2008	2665	3	4411	96045852.	0.
## 310 Yellow Fever	2008	176	3	4411	4719371.	0.
## 311 BCG	2008	154221	3	4411	705760	2.40e7
## 312 Measles	2009	3397	3	4509	46193653.	0.
## 313 DTP-hep B	2009	2665	3	4509	96045852.	0.
## 314 Yellow Fever	2009	176	3	4509	4719371.	0.
## 315 BCG	2009	154221	3	4509	721440	2.40e7
## 316 Measles	2010	3397	3	4746	46193653.	0.
## 317 DTP-hep B	2010	2665	3	4746	96045852.	0.
## 318 Yellow Fever	2010	176	3	4746	4719371.	0.
## 319 BCG	2010	154221	3	4746	759360	2.39e7
## 320 Measles	2011	3397	3	5283	46193653.	0.
## 321 DTP-hep B	2011	2665	3	5283	96045852.	0.
## 322 Yellow Fever	2011	176	3	5283	4719371.	0.
## 323 BCG	2011	154221	3	5283	845280	2.38e7
## 324 Rotavirus	2012	113	3	5639	2105497.	0.
## 325 Rotavirus	2012	387	3	5639	4931933.	0.
## 326 Measles	2012	3397	3	5639	46193653.	0.
## 327 DTP-hep B	2012	2665	3	5639	96045852.	0.
## 328 Yellow Fever	2012	176	3	5639	4719371.	0.
## 329 Rotavirus	2012	215	3	5639	7183815.	0.
## 330 BCG	2012	154221	3	5639	902240	2.38e7
## 331 Rotavirus	2013	113	3	5912	2105497.	0.
## 332 Rotavirus	2013	387	3	5912	4931933.	0.
## 333 Measles	2013	3397	3	5912	46193653.	0.
## 334 DTP-hep B	2013	2665	3	5912	96045852.	0.
## 335 Yellow Fever	2013	176	3	5912	4719371.	0.
## 336 Rotavirus	2013	215	3	5912	7183815.	0.
## 337 BCG	2013	154221	3	5912	945920	2.37e7
## 338 Rotavirus	2014	113	3	5945	2105497.	0.
## 339 Rotavirus	2014	387	3	5945	4931933.	0.
## 340 Measles	2014	3397	3	5945	46193653.	0.
## 341 DTP-hep B	2014	2665	3	5945	96045852.	0.
## 342 Yellow Fever	2014	176	3	5945	4719371.	0.
## 343 Rotavirus	2014	215	3	5945	7183815.	0.
## 344 BCG	2014	154221	3	5945	951200	2.37e7
## 345 Rotavirus	2015	113	3	5939	2105497.	0.

## 346 Rotavirus	2015	387	3	5939	4931933.	0.
## 347 Measles	2015	3397	3	5939	46193653.	0.
## 348 DTP-hep B	2015	2665	3	5939	96045852.	0.
## 349 Yellow Fever	2015	176	3	5939	4719371.	0.
## 350 Rotavirus	2015	215	3	5939	7183815.	0.
## 351 BCG	2015	154221	3	5939	950240	2.37e7
## 352 Rotavirus	2016	113	3	6007	2105497.	0.
## 353 Rotavirus	2016	387	3	6007	4931933.	0.
## 354 Measles	2016	3397	3	6007	46193653.	0.
## 355 DTP-hep B	2016	2665	3	6007	96045852.	0.
## 356 Yellow Fever	2016	176	3	6007	4719371.	0.
## 357 Rotavirus	2016	215	3	6007	7183815.	0.
## 358 BCG	2016	154221	3	6007	961120	2.37e7
## 359 Rotavirus	2017	113	3	6353	2105497.	0.
## 360 Rotavirus	2017	387	3	6353	4931933.	0.
## 361 Measles	2017	3397	3	6353	46193653.	0.
## 362 DTP-hep B	2017	2665	3	6353	96045852.	0.
## 363 Yellow Fever	2017	176	3	6353	4719371.	0.
## 364 Rotavirus	2017	215	3	6353	7183815.	0.
## 365 BCG	2017	154221	3	6353	1016480	2.37e7
## 366 Rotavirus	2018	113	3	6607	2105497.	0.
## 367 Rotavirus	2018	387	3	6607	4931933.	0.
## 368 Measles	2018	3397	3	6607	46193653.	0.
## 369 DTP-hep B	2018	2665	3	6607	96045852.	0.
## 370 Yellow Fever	2018	176	3	6607	4719371.	0.
## 371 Rotavirus	2018	215	3	6607	7183815.	0.
## 372 BCG	2018	154221	3	6607	1057120	2.36e7
## 373 Measles	2001	3397	4.5	5297	46193653.	0.
## 374 DTP-hep B	2001	2665	4.5	5297	96045852.	0.
## 375 Yellow Fever	2001	176	4.5	5297	4719371.	0.
## 376 BCG	2001	154221	4.5	5297	847520	2.38e7
## 377 Measles	2002	3397	4.5	5402	46193653.	0.
## 378 DTP-hep B	2002	2665	4.5	5402	96045852.	0.
## 379 Yellow Fever	2002	176	4.5	5402	4719371.	0.
## 380 BCG	2002	154221	4.5	5402	864320	2.38e7
## 381 Measles	2003	3397	4.5	5545	46193653.	0.
## 382 DTP-hep B	2003	2665	4.5	5545	96045852.	0.
## 383 Yellow Fever	2003	176	4.5	5545	4719371.	0.
## 384 BCG	2003	154221	4.5	5545	887200	2.38e7
## 385 Measles	2004	3397	4.5	5712	46193653.	0.
## 386 DTP-hep B	2004	2665	4.5	5712	96045852.	0.
## 387 Yellow Fever	2004	176	4.5	5712	4719371.	0.
## 388 BCG	2004	154221	4.5	5712	913920	2.38e7
## 389 Measles	2005	3397	4.5	5898	46193653.	0.
## 390 DTP-hep B	2005	2665	4.5	5898	96045852.	0.
## 391 Yellow Fever	2005	176	4.5	5898	4719371.	0.
## 392 BCG	2005	154221	4.5	5898	943680	2.37e7
## 393 Measles	2006	3397	4.5	6117	46193653.	0.
## 394 DTP-hep B	2006	2665	4.5	6117	96045852.	0.
## 395 Yellow Fever	2006	176	4.5	6117	4719371.	0.
## 396 BCG	2006	154221	4.5	6117	978720	2.37e7
## 397 Measles	2007	3397	4.5	6221	46193653.	0.
## 398 DTP-hep B	2007	2665	4.5	6221	96045852.	0.
## 399 Yellow Fever	2007	176	4.5	6221	4719371.	0.

## 400 BCG	2007	154221	4.5	6221	995360	2.37e7
## 401 Measles	2008	3397	4.5	6617	46193653.	0.
## 402 DTP-hep B	2008	2665	4.5	6617	96045852.	0.
## 403 Yellow Fever	2008	176	4.5	6617	4719371.	0.
## 404 BCG	2008	154221	4.5	6617	1058720	2.36e7
## 405 Measles	2009	3397	4.5	6764	46193653.	0.
## 406 DTP-hep B	2009	2665	4.5	6764	96045852.	0.
## 407 Yellow Fever	2009	176	4.5	6764	4719371.	0.
## 408 BCG	2009	154221	4.5	6764	1082240	2.36e7
## 409 Measles	2010	3397	4.5	7119	46193653.	0.
## 410 DTP-hep B	2010	2665	4.5	7119	96045852.	0.
## 411 Yellow Fever	2010	176	4.5	7119	4719371.	0.
## 412 BCG	2010	154221	4.5	7119	1139040	2.35e7
## 413 Measles	2011	3397	4.5	7924	46193653.	0.
## 414 DTP-hep B	2011	2665	4.5	7924	96045852.	0.
## 415 Yellow Fever	2011	176	4.5	7924	4719371.	0.
## 416 BCG	2011	154221	4.5	7924	1267840	2.34e7
## 417 Rotavirus	2012	113	4.5	8458	2105497.	0.
## 418 Rotavirus	2012	387	4.5	8458	4931933.	0.
## 419 Measles	2012	3397	4.5	8458	46193653.	0.
## 420 DTP-hep B	2012	2665	4.5	8458	96045852.	0.
## 421 Yellow Fever	2012	176	4.5	8458	4719371.	0.
## 422 Rotavirus	2012	215	4.5	8458	7183815.	0.
## 423 BCG	2012	154221	4.5	8458	1353280	2.33e7
## 424 Rotavirus	2013	113	4.5	8867	2105497.	0.
## 425 Rotavirus	2013	387	4.5	8867	4931933.	0.
## 426 Measles	2013	3397	4.5	8867	46193653.	0.
## 427 DTP-hep B	2013	2665	4.5	8867	96045852.	0.
## 428 Yellow Fever	2013	176	4.5	8867	4719371.	0.
## 429 Rotavirus	2013	215	4.5	8867	7183815.	0.
## 430 BCG	2013	154221	4.5	8867	1418720	2.33e7
## 431 Rotavirus	2014	113	4.5	8918	2105497.	0.
## 432 Rotavirus	2014	387	4.5	8918	4931933.	0.
## 433 Measles	2014	3397	4.5	8918	46193653.	0.
## 434 DTP-hep B	2014	2665	4.5	8918	96045852.	0.
## 435 Yellow Fever	2014	176	4.5	8918	4719371.	0.
## 436 Rotavirus	2014	215	4.5	8918	7183815.	0.
## 437 BCG	2014	154221	4.5	8918	1426880	2.32e7
## 438 Rotavirus	2015	113	4.5	8908	2105497.	0.
## 439 Rotavirus	2015	387	4.5	8908	4931933.	0.
## 440 Measles	2015	3397	4.5	8908	46193653.	0.
## 441 DTP-hep B	2015	2665	4.5	8908	96045852.	0.
## 442 Yellow Fever	2015	176	4.5	8908	4719371.	0.
## 443 Rotavirus	2015	215	4.5	8908	7183815.	0.
## 444 BCG	2015	154221	4.5	8908	1425280	2.33e7
## 445 Rotavirus	2016	113	4.5	9010	2105497.	0.
## 446 Rotavirus	2016	387	4.5	9010	4931933.	0.
## 447 Measles	2016	3397	4.5	9010	46193653.	0.
## 448 DTP-hep B	2016	2665	4.5	9010	96045852.	0.
## 449 Yellow Fever	2016	176	4.5	9010	4719371.	0.
## 450 Rotavirus	2016	215	4.5	9010	7183815.	0.
## 451 BCG	2016	154221	4.5	9010	1441600	2.32e7
## 452 Rotavirus	2017	113	4.5	9530	2105497.	0.
## 453 Rotavirus	2017	387	4.5	9530	4931933.	0.

## 454 Measles	2017	3397	4.5	9530	46193653.	0.
## 455 DTP-hep B	2017	2665	4.5	9530	96045852.	0.
## 456 Yellow Fever	2017	176	4.5	9530	4719371.	0.
## 457 Rotavirus	2017	215	4.5	9530	7183815.	0.
## 458 BCG	2017	154221	4.5	9530	1524800	2.32e7
## 459 Rotavirus	2018	113	4.5	9910	2105497.	0.
## 460 Rotavirus	2018	387	4.5	9910	4931933.	0.
## 461 Measles	2018	3397	4.5	9910	46193653.	0.
## 462 DTP-hep B	2018	2665	4.5	9910	96045852.	0.
## 463 Yellow Fever	2018	176	4.5	9910	4719371.	0.
## 464 Rotavirus	2018	215	4.5	9910	7183815.	0.
## 465 BCG	2018	154221	4.5	9910	1585600	2.31e7

In the following code, we use the CEA study benefits but use data from Gavi reports on costs (this does not include many of the costs that are in the CEA studies). We use the DC model to calculate the split between domestic and donor funding.

```
# Ghana data
ghana_data <-
  data %>%
  filter(country == "Ghana") %>%
  distinct(vaccine, cost = total, benefit)

# GDP per capita data
ghana_gdppc <-
  gdp_data %>%
  filter(country == "Ghana") %>%
  distinct(year, gdp_pc)

# Run DC model
DCmodel_out <- DCmodel(
  cost = ghana_data$cost,
  benefit = ghana_data$benefit,
  intervention = ghana_data$vaccine,
  gdp_per_capita = rep(ghana_gdppc$gdp_pc, 5),
  gdp_threshold_multiple = c(rep(ochalek_p, nrow(ghana_gdppc)),
                             rep(0.5, nrow(ghana_gdppc)),
                             rep(1, nrow(ghana_gdppc)),
                             rep(3, nrow(ghana_gdppc)),
                             rep(4.5, nrow(ghana_gdppc)))
)

# Donor financing is 0 in all rows since the total amount
# is missing many costs included in CEAs. We only print
# out rows in which donor co-financing is higher than 0
# here.
DCmodel_out %>%
  left_join(ghana_gdppc %>% rename(gdp_per_capita = gdp_pc),
            by = "gdp_per_capita") %>%
  # Remove rotavirus vaccins before 2012
  filter(!(intervention == "Rotavirus" & year < 2012)) %>%
  select(intervention, year, cer,
         gdp_threshold_multiple, threshold, domestic, donor) %>%
  filter(donor > 0)
```

```
## # A tibble: 4 x 7
##   intervention year   cer gdp_threshold_multip~ threshold domestic donor
##   <chr>         <dbl> <dbl>                <dbl>      <dbl>      <dbl> <dbl>
## 1 DTP-hep B    2001   422                0.33      388 13980804 1.24e6
## 2 DTP-hep B    2002   422                0.33      396 14269068 9.47e5
## 3 DTP-hep B    2003   422                0.33      407 14665431 5.50e5
## 4 DTP-hep B    2004   422                0.33      419 15097827 1.18e5
```



# Figures

## All Ghana figure

```
# Add plotting themes.
library(ggthemes)

# Extract Ghana data
data_ghana <- data %>%
  rename(year = year_gavi) %>%
  filter(country == "Ghana" &
         grepl("DALY|YLL", indicator)) %>%
  select(vaccine_cea_paper, indicator, cost, benefit, cea, year, total,
         domestic, gavi, domestic_p, gavi_p, s_year, e_year, vaccine, perspective) %>%
  # Data for 2018 only on BCG which is not supported
  filter(year < 2018)

# Max axes
max_ghana_x <- data_ghana %>%
  summarise(max(cost)) %>% unlist %>%
  roundUp + 10000000
max_ghana_y <- data_ghana %>%
  summarise(max(benefit)) %>% unlist %>%
  roundUp

# Select relevant GDP data (depending on which years have data
# for Ghana)
gdp_data_ghana <- left_join(data_ghana %>% select(year),
                           gdp_data %>% filter(country == "Ghana"),
                           by = "year") %>%
  # Data for 2018 only on BCG which is not supported
  filter(year < 2018)

# Plot legend labels
legend_labels <- data_ghana %>%
  distinct(vaccine_cea_paper) %>%
  unlist %>% sort

# Plot Ghana
p1 <- ggplot() +
  # Facet by year and threshold
  facet_grid(year~threshold, scales = "free") +
  # Threshold line
  geom_line(data = gdp_data_ghana %>%
            filter(country == "Ghana") %>%
            filter(cost < max_ghana_x),
            mapping = aes(x = cost/1000000, y = cea_line/1000), linetype = 2) +
  # Costs and effectiveness point
  geom_point(data = data_ghana,
            mapping = aes(x = cost/1000000, y = benefit/1000,
                          color = vaccine_cea_paper,
```

```

        shape = vaccine_cea_paper),
      size = 2.5) +
  # Arrow based on the proportion of Gavi support relative to domestic
  geom_segment(data = data_ghana,
    aes(x = cost/1000000, xend = (cost - gavi)/1000000,
      y = benefit/1000, yend = benefit/1000),
    arrow = arrow(length = unit(0.1, "cm")))) +
  # x-axis name and $ formatting
  scale_x_continuous(name = "Cost (in millions USD 2018)",
    labels = scales::dollar) +
  # y-axis name and formatting
  scale_y_continuous(name = "DALYs averted (in thousands)",
    labels = scales::comma) +
  # Remove color/shape legend name
  # Display color legend in two rows and remove shape legend
  # Set colors and shapes
  guides(color = guide_legend(nrow = 2, byrow = TRUE)) +
  scale_color_manual(name = "",
    labels = legend_labels,
    values = gdocs_pal()(7)) +

  scale_shape_manual(name = "",
    labels = legend_labels,
    values = c(15, 16, 18, 17, 17, 17, 7)) +
  # Change theme of plot
  theme_fivethirtyeight() +
  theme(legend.position = "bottom",
    axis.title = element_text(),
    plot.caption = element_text(hjust = 0))

# Save figure
ggsave("figures/ghana_all.png", p1, dpi = 300,
  width = 8*2.5, height = 10*2.5, units = "cm")

```

## Warning: Removed 44 rows containing missing values (geom\_segment).

```

# Create and save figure with appendix caption
p1 +
  # Add a caption
  labs(caption = paste(
    "Figure A1. Gavi contribution to vaccines in Ghana.",
    "Columns differ by cost-effectiveness thresholds, represented by the dashed black line;",
    "Rows differ by the year of Gavi funding (years missing are due to lack of data);",
    "Points represent costs and benefits are extracted from cost-effectiveness (CE) studies;",
    "Arrows represent financing by Gavi (CE studies typically include costs not covered by Gavi",
    "such as distribution costs).",
    sep = "\n"))

```

## Warning: Removed 44 rows containing missing values (geom\_segment).

```

ggsave("figures/A1_ghana_all_appendix_caption.png", dpi = 300,
  width = 8*2.5, height = 10*2.5, units = "cm")

```

```
## Warning: Removed 44 rows containing missing values (geom_segment).
```

## All cross country figure 1 x GDP

```
# Get years that include non-Ghana countries
years_include <- data %>%
  filter(country != "Ghana") %>% distinct(year_gavi) %>% unlist

# All country data
data_all_countries <-
  data %>%
  rename(year = year_gavi) %>%
  filter(grepl("DALY|YLL", indicator)) %>%
  filter(year %in% years_include) %>%
  select(country, vaccine_cea_paper, indicator, cost, benefit, cea, year, total,
         domestic, gavi, domestic_p, gavi_p, s_year, e_year, vaccine, perspective) %>%
  # Data for 2018 only on BCG which is not supported
  # Also remove 2015 for which there is very limited data
  filter(year < 2018 & year != 2015) %>%
  # Add population data
  left_join(population_data, by = c("country", "year")) %>%
  # Convert values to per 100,000 population
  mutate(cost = cost/population*100000,
         benefit = benefit/population*100000,
         gavi = gavi/population*100000,
         domestic = domestic/population*100000)

# Max axes
max_all_countries_x <- data_all_countries %>%
  summarise(max(cost)) %>% unlist %>%
  roundUp + 10
max_all_countries_y <- data_all_countries %>%
  summarise(max(benefit)) %>% unlist %>%
  roundUp

# GDP data select
gdp_data_all_countries <-
  left_join(data_all_countries %>% select(year, country),
            gdp_data,
            by = c("year", "country")) %>%
  filter(year %in% years_include) %>%
  # Data for 2018 only on BCG which is not supported
  # Also remove 2015 for which there is very limited data
  filter(year < 2018 & year != 2015)

# Plot legend labels
legend_labels <- data_all_countries %>%
  distinct(vaccine_cea_paper) %>%
  unlist %>% sort

# Plot all countries using 1 X GDP per capita threshold
p2 <- ggplot() +
```

```

# Facet by country and year
facet_grid(year~country, scales = "free") +
# Threshold line
geom_line(data = gdp_data_all_countries %>%
  filter(threshold == "1 x GDP\nper capita") %>%
  filter(cost < max_all_countries_x),
  mapping = aes(x = cost/1000, y = cea_line), linetype = 2) +
# Costs and effectiveness point
geom_point(data = data_all_countries,
  mapping = aes(x = cost/1000, y = benefit,
    color = vaccine_cea_paper,
    shape = vaccine_cea_paper),
  size = 2.5) +
# Arrow based on the proportion of Gavi support relative to domestic
geom_segment(data = data_all_countries,
  aes(x = cost/1000, xend = (cost - gavi)/1000,
    y = benefit, yend = benefit),
  arrow = arrow(length = unit(0.1, "cm"))) +
# x-axis name and $ formatting
scale_x_continuous(name = "Cost per 100,000 people (in thousands USD 2018)",
  labels = scales::dollar) +
# y-axis name and formatting
scale_y_continuous(name = "DALYs averted per 100,000 people",
  labels = scales::comma) +
# Remove color/shape legend name
# Display color legend in two rows and remove shape legend
# Set colors and shapes
guides(color = guide_legend(nrow = 2, byrow = TRUE)) +
scale_color_manual(name = "",
  labels = legend_labels,
  values = gdocs_pal()(10)) +

scale_shape_manual(name = "",
  labels = legend_labels,
  values = c(15, 16, 18, 13, 13, 13, 17, 17, 17, 7)) +
# Display color legend in 4 columns
# Change theme of plot
theme_fivethirtyeight() +
theme(legend.position = "bottom",
  axis.title = element_text(),
  axis.text.x = element_text(angle = 45),
  plot.caption = element_text(hjust = 0))

# Save figure
ggsave("figures/all_country_compare_1GDP.png", p2, dpi = 300,
  width = 8*2.5, height = 10*2.5, units = "cm")

```

## Warning: Removed 5 rows containing missing values (geom\_segment).

```

p2 + labs(
  caption = paste(
    "Figure A2. Gavi contribution to vaccines in Ghana, Angola, India, Kenya, and Senegal.",
    "The dashed black line is the 1 x GDP cost-effectiveness threshold;"
  )

```

```

    "Points represent costs and benefits are extracted from cost-effectiveness (CE) studies;",
    "Arrows represent financing by Gavi (CE studies typically include costs not covered by Gavi",
    "such as distribution costs).",
    sep = "\n"))

```

```
## Warning: Removed 5 rows containing missing values (geom_segment).
```

```

# Create and save figure with appendix caption
ggsave("figures/A2_all_country_compare_1GDP.png", dpi = 300,
       width = 8*2.5, height = 10*2.5, units = "cm")

```

```
## Warning: Removed 5 rows containing missing values (geom_segment).
```

## All cross country figure 0.5 x GDP

```

# Plot all countries using 0.5 X GDP per capita threshold
p3 <- ggplot() +
  # Facet by country and year
  facet_grid(year~country, scales = "free") +
  # Threshold line
  geom_line(data = gdp_data_all_countries %>%
            filter(threshold == "0.5 x GDP\nper capita") %>%
            filter(cost < max_all_countries_x),
            mapping = aes(x = cost/1000, y = cea_line), linetype = 2) +
  # CCosts and effectiveness point
  geom_point(data = data_all_countries,
            mapping = aes(x = cost/1000, y = benefit,
                          color = vaccine_cea_paper,
                          shape = vaccine_cea_paper),
            size = 2.5) +
  # Arrow based on the proportion of Gavi support relative to domestic
  geom_segment(data = data_all_countries,
            aes(x = cost/1000, xend = (cost - gavi)/1000,
                y = benefit, yend = benefit),
            arrow = arrow(length = unit(0.1, "cm"))) +
  # x-axis name and $ formatting
  scale_x_continuous(name = "Cost per 100,000 people (in thousands USD 2018)",
                    labels = scales::dollar) +
  # y-axis name and formatting
  scale_y_continuous(name = "DALYs averted per 100,000 people",
                    labels = scales::comma) +
  # Remove color/shape legend name
  # Display color legend in four rows and remove shape legend
  # Set colors and shapes
  guides(color = guide_legend(nrow = 2, byrow = TRUE)) +
  scale_color_manual(name = "",
                    labels = legend_labels,
                    values = gdocs_pal()(10)) +

  scale_shape_manual(name = "",

```

```

        labels = legend_labels,
        values = c(15, 16, 18, 13, 13, 13, 17, 17, 17, 7)) +
# Display color legend in 4 columns
# Change theme of plot
theme_fivethirtyeight() +
theme(legend.position = "bottom",
      axis.title = element_text(),
      axis.text.x = element_text(angle = 45),
      plot.caption = element_text(hjust = 0))

# Save figure
ggsave("figures/all_country_compare_05GDP.png", p2, dpi = 300,
       width = 8*2.5, height = 10*2.5, units = "cm")

```

## Warning: Removed 5 rows containing missing values (geom\_segment).

```

p2 + labs(
  caption = paste(
    "Figure A2. Gavi contribution to vaccines in Ghana, Angola, India, Kenya, and Senegal.",
    "The dashed black line is the 0.5 x GDP cost-effectiveness threshold;",
    "Points represent costs and benefits are extracted from cost-effectiveness (CE) studies;",
    "Arrows represent financing by Gavi (CE studies typically include costs not covered by Gavi",
    "such as distribution costs).",
    sep = "\n"))

```

## Warning: Removed 5 rows containing missing values (geom\_segment).

```

# Create and save figure with appendix caption
ggsave("figures/A2_all_country_compare_05GDP.png", dpi = 300,
       width = 8*2.5, height = 10*2.5, units = "cm")

```

## Warning: Removed 5 rows containing missing values (geom\_segment).

## Ghana 1 x GDP 2012-2016 (most data for these years)

```

# Plot legend labels
legend_labels <- data_ghana %>%
  distinct(vaccine_cea_paper) %>%
  unlist %>% sort

# Plot Ghana
p4 <- ggplot() +
  # Facet by year and threshold
  facet_wrap(year~., scales = "free", ncol = 2) +
  # Threshold line
  geom_line(data = gdp_data_ghana %>%
    filter(country == "Ghana") %>%
    filter(cost < max_ghana_x) %>%
    filter(threshold == "1 x GDP\nper capita") %>%
    filter(year %in% 2012:2016),

```

```

    mapping = aes(x = cost/1000000, y = cea_line/1000), linetype = 2) +
# Costs and effectiveness point
geom_point(data = data_ghana %>%
  filter(year %in% 2012:2016),
  mapping = aes(x = cost/1000000, y = benefit/1000,
    color = vaccine_cea_paper,
    shape = vaccine_cea_paper),
  size = 4.5) +
# Arrow based on the proportion of Gavi support relative to domestic
geom_segment(data = data_ghana %>%
  filter(year %in% 2012:2016),
  aes(x = cost/1000000, xend = (cost - gavi)/1000000,
    y = benefit/1000, yend = benefit/1000),
  arrow = arrow(length = unit(0.2, "cm")),
  size = 1) +
# x-axis name and $ formatting
scale_x_continuous(name = "Cost (in millions USD 2018)",
  labels = scales::dollar) +
# y-axis name and formatting
scale_y_continuous(name = "DALYs averted (in thousands)",
  labels = scales::comma) +
# Remove color/shape legend name
# Display color legend in two rows and remove shape legend
# Set colors and shapes
guides(color = guide_legend(nrow = 2, byrow = TRUE)) +
scale_color_manual(name = "",
  labels = legend_labels,
  values = gdocs_pal()(7)) +
scale_shape_manual(name = "",
  labels = legend_labels,
  values = c(15, 16, 18, 17, 17, 17, 7)) +
# Change theme of plot
theme_fivethirtyeight() +
theme(legend.position = "bottom",
  axis.title = element_text(),
  plot.caption = element_text(hjust = 0),
  text = element_text(size = 16))

# Save figure
ggsave("figures/ghana_xlgdp.png", p4, dpi = 300,
  width = 8.5*2.5, height = 10*2.5, units = "cm")

```

## Warning: Removed 4 rows containing missing values (geom\_segment).

```

# Create and save figure with appendix caption
p4 +
# Add a caption
labs(caption = paste(
  "Figure 1. Gavi contribution to vaccines in Ghana.",
  "The dashed black line is the 0.5 x GDP-based cost-effectiveness threshold;",
  "Each plot represents a different year of Gavi co-financing;",
  "Points represents costs and benefits extracted from cost-effectiveness (CE) studies;",
  "Arrows represent financing by Gavi (CE studies typically include costs not covered by Gavi",

```

```
"such as distribution costs).",
  sep = "\n"))
```

```
## Warning: Removed 4 rows containing missing values (geom_segment).
```

```
ggsave("figures/A3_ghana_xlgdp_appendix_caption.png", dpi = 300,
  width = 8*2.5, height = 10*2.5, units = "cm")
```

```
## Warning: Removed 4 rows containing missing values (geom_segment).
```

## Country comparison 1 x GDP per capita

```
# Plot legend labels
legend_labels <- data_all_countries %>%
  distinct(vaccine_cea_paper) %>%
  unlist %>% sort

# Plot all countries using 1 X GDP per capita threshold
p5 <- ggplot() +
  # Facet by country and year
  facet_grid(year~country, scales = "free") +
  # Threshold line
  geom_line(data = gdp_data_all_countries %>%
    filter(threshold == "1 x GDP\nper capita") %>%
    filter(cost < max_all_countries_x) %>%
    filter(year %in% c(2014, 2017)),
    mapping = aes(x = cost/1000, y = cea_line), linetype = 2) +
  # Costs and effectiveness point
  geom_point(data = data_all_countries %>%
    filter(year %in% c(2014, 2017)),
    mapping = aes(x = cost/1000, y = benefit,
      color = vaccine_cea_paper,
      shape = vaccine_cea_paper),
    size = 4.5) +
  # Arrow based on proportion of Gavi support relative to domestic
  geom_segment(data = data_all_countries %>%
    filter(year %in% c(2014, 2017)),
    aes(x = cost/1000, xend = (cost - gavi)/1000,
      y = benefit, yend = benefit),
    arrow = arrow(length = unit(0.2, "cm")))) +
  # x-axis name and $ formatting
  scale_x_continuous(name = "Cost per 100,000 people (in thousands USD 2018)",
    labels = scales::dollar) +
  # y-axis name and formatting
  scale_y_continuous(name = "DALYs averted per 100,000 people",
    labels = scales::comma) +
  # Remove color/shape legend name
  # Display color legend in four rows and remove shape legend
  # Set colors and shapes
  guides(color = guide_legend(nrow = 2, byrow = TRUE)) +
```



```

scale_color_manual(name = "",
                   labels = legend_labels,
                   values = gdocs_pal()(10)) +
scale_shape_manual(name = "",
                   labels = legend_labels,
                   values = c(15, 16, 18, 13, 13, 13, 17, 17, 17, 7)) +
# Display color legend in 4 columns
# Change theme of plot
theme_fivethirtyeight() +
theme(legend.position = "bottom",
      axis.title = element_text(),
      axis.text.x = element_text(angle = 45),
      plot.caption = element_text(hjust = 0),
      text = element_text(size = 16))

# Save figure
ggsave("figures/country_compare_1GDP.png", p5, dpi = 300,
       width = 11*2.5, height = 9*2.5, units = "cm")

```

## Warning: Removed 2 rows containing missing values (geom\_segment).

```

p5 + labs(
  caption = paste(
    "Figure 2. Gavi contribution to vaccines in Ghana, Angola, India, Kenya, and Senegal.",
    "The dashed black line is the 1 x GDP cost-effectiveness threshold;",
    "Points represent costs and benefits are extracted from cost-effectiveness (CE) studies;",
    "Arrows represent financing by Gavi (CE studies typically include costs not covered by Gavi",
    "such as distribution costs).",
    sep = "\n"))

```

## Warning: Removed 2 rows containing missing values (geom\_segment).

```

# Create and save figure with appendix caption
ggsave("figures/A5_country_compare_1GDP.png", dpi = 300,
       width = 8*2.5, height = 10*2.5, units = "cm")

```

## Warning: Removed 2 rows containing missing values (geom\_segment).

## Gavi co-financing Ghana

```

financing_ghana_plot_data <-
  fund_data %>%
  filter(country == "Ghana" & !is.na(gavi)) %>%
  select(vaccine, year, domestic, gavi) %>%
  gather(key = source, value = amount, domestic:gavi) %>%
  mutate(source = gsub("gavi", "Gavi", source)) %>%
  mutate(source = gsub("domestic", "Domestic", source))

p6 <- ggplot(data = financing_ghana_plot_data,

```

```

    mapping = aes(x = factor(year), y = amount/1000000, fill = source)) +
  facet_wrap(vaccine~.) +
  geom_bar(stat = "identity", position = "dodge") +
  # x-axis name and $ formatting
  scale_x_discrete(name = "Year") +
  # y-axis name and formatting
  scale_y_continuous(name = "Funding in 2018 US Dollars",
    labels = scales::dollar_format(prefix = "$", suffix = "M")) +
  # Remove fill legend title
  # Display color legend in four rows and remove shape legend
  # Set colors and shapes
  scale_fill_discrete(name = "Source") +
  # Change theme of plot
  theme_fivethirtyeight() +
  theme(legend.position = "bottom",
    axis.title = element_text(),
    plot.caption = element_text(hjust = 0),
    axis.text.x = element_text(angle = 45))

# Save figure
ggsave("figures/ghana_gavi_cofinancing.png", p6, dpi = 300,
  width = 6*2.5, height = 6*2.5, units = "cm")

p6 + labs(
  caption = paste(
    "Figure 1. Gavi co-financing of vaccines in Ghana.",
    "Source of data: Gavi Alliance Annual Progress Reports.",
    "DTP-hep B: dephtheria, tetanus, pertusis, and hepatitis B vaccine.",
    "MenA: Meningococcal vaccine.",
    "PCV: Pneumococcal conjugate vaccine.",
    sep = "\n"))

# Create and save figure with appendix caption
ggsave("figures/A6_ghana_gavi_cofinancing.png.png", dpi = 300,
  width = 9*2.5, height = 6*2.5, units = "cm")

```

## Gavi co-financing All

```

financing_plot_data <-
  fund_data %>%
  filter(!is.na(gavi) & (year %in% 2012:2018)) %>%
  select(country, vaccine, year, domestic, gavi) %>%
  gather(key = source, value = amount, domestic:gavi) %>%
  mutate(source = gsub("gavi", "Gavi", source)) %>%
  mutate(source = gsub("domestic", "Domestic", source))

p7 <- ggplot(data = financing_plot_data,
  mapping = aes(x = factor(year), y = amount/1000000, fill = source)) +
  facet_wrap(vaccine~country, scales = "free") +
  geom_bar(stat = "identity", position = "dodge") +
  # x-axis name and $ formatting

```

```

scale_x_discrete(name = "Year") +
# y-axis name and formatting
scale_y_continuous(name = "Funding in 2018 US Dollars",
                    labels = scales::dollar_format(prefix = "$", suffix = "M")) +
# Remove fill legend title
# Display color legend in four rows and remove shape legend
# Set colors and shapes
scale_fill_discrete(name = "Source") +
# Change theme of plot
theme_fivethirtyeight() +
theme(legend.position = "bottom",
      axis.title = element_text(),
      plot.caption = element_text(hjust = 0),
      axis.text.x = element_text(angle = 45))

# Save figure
ggsave("figures/all_gavi_cofinancing.png", p7, dpi = 300,
       width = 10*2.5, height = 10*2.5, units = "cm")

p6 + labs(
  caption = paste(
    "Figure 1. Gavi co-financing of vaccines",
    "Source of data: Gavi Alliance Annual Progress Reports.",
    "DTP-hep B: dephtheria, tetanus, pertusis, and hepatitis B vaccine.",
    "MenA: Meningococcal vaccine.",
    "PCV: Pneumococcal conjugate vaccine.",
    sep = "\n"))

# Create and save figure with appendix caption
ggsave("figures/A7_all_gavi_cofinancing.png", dpi = 300,
       width = 8*2.5, height = 10*2.5, units = "cm")

```

## Reproduce figure from Morton et al 2018

```

threshold_data <- tibble(
  cost = 0:90,
  benefit = 0:90,
)

interventions_data <- tibble(
  cost = c(40, 80, 70, 90,
           20, 10, 40, 55, 15),
  benefit = c(10, 40, 50, 70,
              30, 40, 50, 80, 75),
  country_funded = c(rep(FALSE, 4),
                     rep(TRUE, 5))
)

library(grid)
library(pBrackets)
library(ggpubr)

```

```
## Warning: package 'ggpubr' was built under R version 3.6.3
```

```
## Loading required package: magrittr
```

```
##
```

```
## Attaching package: 'magrittr'
```

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

```
##
```

```
##      set_names
```

```
## The following object is masked from 'package:tidyr':
```

```
##
```

```
##      extract
```

```
bracketsGrob <- function(...){
  l <- list(...)
  e <- new.env()
  e$l <- l
  grid::recordGrob( {
    do.call(grid.brackets, l)
  }, e)
}

b1 <- bracketsGrob(0.79, 0.38, 0.41, 0.38, h=0.05, lwd=1, col="red")

ggplot() +
  geom_line(data = threshold_data,
            mapping = aes(x = cost, y = benefit),
            linetype = 2,
            size = 1) +
  geom_segment(data = interventions_data %>% filter(country_funded == FALSE),
              mapping = aes(x = cost, xend = benefit + 1, y = benefit, yend = benefit),
              arrow = arrow(length = unit(0.2, "cm")),
              size = 1) +
  annotation_custom(b1) +
  annotate("text", label = "Subsidy", x = 60, y = 30, color = "red", size = 6) +
  annotate("text", label = "Country funded\nprojects", x = 20, y = 60, hjust = 0,
          size = 6) +
  annotate("text", label = "Country rejected\nprojects", x = 60, y = 15, hjust = 0,
          size = 6) +
  annotate("text", label = "v[0]/c[0]", x = 95, y = 95, parse = TRUE, size = 6) +
  geom_point(data = interventions_data,
            mapping = aes(x = cost, y = benefit,
                          color = country_funded,
                          shape = country_funded),
            size = 4) +
  scale_color_manual(name = "", values = c("red", "darkgreen")) +
  scale_shape_manual(name = "", values = c("circle", "square")) +
  scale_x_continuous(name = "Costs", expand = c(0, 0), limits = c(0,100)) +
  scale_y_continuous(name = "Country Benefits", expand = c(0, 0), limits = c(0,100)) +
  theme_classic() +
  theme(legend.position = "none",
```

```
axis.line = element_line(arrow = arrow(length = unit(0.4, "cm"))),  
axis.text = element_blank(),  
axis.ticks = element_blank(),  
text = element_text(size = 16))
```

```
ggsave("figures/donor_funding_reproduce_morton18.png", dpi = 300,  
width = 7*2.5, height = 6*2.5, units = "cm")
```

## Tables

### Ghana Gavi co-financing table

```
table_out <- fund_data %>%
  filter(country == "Ghana") %>%
  filter(!is.na(gavi)) %>%
  mutate(domestic = paste(scales::dollar(signif(domestic, 3)),
    " (", scales::percent(domestic_p, accuracy = 1), "%)",
    sep = ""),
    gavi = paste(scales::dollar(signif(gavi, 3)),
    " (", scales::percent(gavi_p, accuracy = 1), "%)",
    sep = ""),
    total = scales::dollar(signif(total, 3))) %>%
  arrange(vaccine, year) %>%
  select(Vaccine = vaccine, Year = year,
    `Domestic support (%)` = domestic,
    `Gavi support (%)` = gavi, Total = total)
write_csv(table_out, "tables/csv_format/ghana_gavi_cofinancing.csv")
```

### Gavi CEA table

```
table_out <-
  cea_data %>%
  filter(grepl("DALY|YLL", indicator)) %>%
  # Only include base-case of Abbot et al. 2012
  # and use earliest year of Gavi funding (since intervention is 2003-08).
  filter(!(authors == "Abbott et al. 2012" &
    (assumptions != "$5 per dose"))) %>% # / year_gavi != 2012))) %>%
  # Do not include the CEA with Gavi Subsidy accounted for
  # and focus on health system perspective.
  filter(!(authors == "Nonvignon et al. 2018" &
    (grepl("Ghana only", assumptions) |
    !grepl("Health system", perspective)))) %>%
  # Use relevant Gavi data year for Krishnamoorthy et al. 2019
  # based on paper intervention years
  # filter(!(authors == "Krishnamoorthy et al. 2019" &
  #   year_gavi != 2018)) %>%
  # Use the analysis with the 90% intervention coverage and year 2018
  # for Gavi funding based on paper intervention years
  filter(!(authors == "Megidido et al. 2018" &
    (grepl("Expanded coverage (90%)", assumptions)))) %>%
  # Do not include the CEA with Gavi subsidy accounted for
  filter(!(authors == "Debellut et al., 2019" &
    (assumptions == "Vaccine programme costs with Gavi subsidy"))) %>%
  # Remove comma from authors column
  mutate(authors = gsub(",", "", authors)) %>%
  # Convert to dollars
  mutate(cost = scales::dollar(signif(cost, 3)),
    benefit = scales::comma(signif(benefit, 3)),
```

```

    cea = scales::dollar(signif(cea, 3)),
    year = as.numeric(gsub(".*\\. ", "", authors))) %>%
  arrange(country, vaccine, year) %>%
  select(Country = country, Vaccine = vaccine, Indicator = indicator,
         Costs = cost, Benefits = benefit, CER = cea, Perspective = perspective,
         Source = authors)
write_csv(table_out, "tables/csv_format/cea_table.csv")

```