

Exploratory Data Analysis

1 Personal Information

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Github link:

2 Data Context

This exploratory data analysis is conducted using R and RStudio. There are two main sets of data to be considered, linked to the two main sections of the project.

The first set of data represents the baseline or standard of care output runs produced by the existing patient pathway model. For the baseline there are 15 data output files (in csv format). Multiple files had been generated to account for model stochasticity, with the results of each simulation being based on a different set of random probabilities. The data files have a consistent structure and represent the population of individuals who move through the TB diagnostic patient pathway. In this, each column represents either a patient disease status or a point in the patient pathway that the individual may or may not have reached.

The second data set consists of TB burden estimates for Kenya produced by the World Health Organization as well as the accompanying data dictionary. The estimates cover a range of data variables and their estimated values between the years 2000 and 2022. Several key variables include estimates on TB incidence (new cases), notifications (diagnoses) and deaths. Estimates are also provided for different groups of individuals (for example HIV positive patients) and for different types of TB.

3 Data Description: Baseline TB model

4 Data Description: WHO Tuberculosis Data

4.1 Load and merge WHO TB burden data and data dictionary

```
#Path to directory
basePath="/Users/adenoooy/Library/CloudStorage/OneDrive-Personal/UVA/Thesis/MSc-Thesis/"

#Load data dictionary
datadict=read.csv(paste(basePath,"data/dynamic/TB_data_dictionary_2024-01-30.csv",sep=""))
colnames(datadict)
```

```
## [1] "variable_name" "dataset"      "code_list"    "definition"
```

```
print(datadict[1:3,])
```

```
##      variable_name dataset code_list
## 1 budget_cpp_dstb   Budget
## 2 budget_cpp_mdr    Budget
## 3 budget_cpp_tpt    Budget
##
## 1 Average cost of drugs budgeted per patient for drug-susceptible TB treatment, excluding buffer sto
## 2           Average cost of drugs budgeted per patient for MDR-TB treatment, excluding buffer sto
## 3           Average cost of drugs budgeted per patient for TB preventive treatment, excluding buffer sto
```

```
#Load TB data
```

```
tb_estimates=read_excel(paste(basePath,"data/dynamic/kenya_tb_burden.xlsx",sep=""))
colnames(tb_estimates)
```

```
## [1] "country"          "iso2"
## [3] "iso3"             "iso_numeric"
## [5] "g_whoregion"       "year"
## [7] "e_pop_num"         "e_inc_100k"
## [9] "e_inc_100k_lo"     "e_inc_100k_hi"
## [11] "e_inc_num"         "e_inc_num_lo"
## [13] "e_inc_num_hi"      "e_tbhiv_prct"
## [15] "e_tbhiv_prct_lo"   "e_tbhiv_prct_hi"
## [17] "e_inc_tbhiv_100k"  "e_inc_tbhiv_100k_lo"
## [19] "e_inc_tbhiv_100k_hi" "e_inc_tbhiv_num"
## [21] "e_inc_tbhiv_num_lo" "e_inc_tbhiv_num_hi"
## [23] "e_mort_exc_tbhiv_100k" "e_mort_exc_tbhiv_100k_lo"
## [25] "e_mort_exc_tbhiv_100k_hi" "e_mort_exc_tbhiv_num"
## [27] "e_mort_exc_tbhiv_num_lo" "e_mort_exc_tbhiv_num_hi"
## [29] "e_mort_tbhiv_100k" "e_mort_tbhiv_100k_lo"
## [31] "e_mort_tbhiv_100k_hi" "e_mort_tbhiv_num"
## [33] "e_mort_tbhiv_num_lo" "e_mort_tbhiv_num_hi"
## [35] "e_mort_100k"       "e_mort_100k_lo"
## [37] "e_mort_100k_hi"    "e_mort_num"
## [39] "e_mort_num_lo"     "e_mort_num_hi"
## [41] "cfr"               "cfr_lo"
## [43] "cfr_hi"            "cfr_pct"
## [45] "cfr_pct_lo"        "cfr_pct_hi"
## [47] "c_newinc_100k"     "c_cdr"
## [49] "c_cdr_lo"          "c_cdr_hi"
```

```
print(tb_estimates[1:3,])
```

```
## # A tibble: 3 x 50
##   country iso2 iso3 iso_numeric g_whor~1 year e_pop~2 e_inc~3 e_inc~4 e_inc~5
##   <chr>   <chr> <chr>      <dbl> <chr>   <dbl>   <dbl>   <dbl>   <dbl>
## 1 Kenya KE    KEN        404 AFR    2000  3.09e7   451    182    839
## 2 Kenya KE    KEN        404 AFR    2001  3.18e7   499    178    982
## 3 Kenya KE    KEN        404 AFR    2002  3.28e7   534    174   1090
## # ... with 40 more variables: e_inc_num <dbl>, e_inc_num_lo <dbl>,
## #   e_inc_num_hi <dbl>, e_tbhiv_prct <dbl>, e_tbhiv_prct_lo <dbl>,
## #   e_tbhiv_prct_hi <dbl>, e_inc_tbhiv_100k <dbl>, e_inc_tbhiv_100k_lo <dbl>,
```

```
## # e_inc_tbhiv_100k_hi <dbl>, e_inc_tbhiv_num <dbl>, e_inc_tbhiv_num_lo <dbl>,
## # e_inc_tbhiv_num_hi <dbl>, e_mort_exc_tbhiv_100k <dbl>,
## # e_mort_exc_tbhiv_100k_lo <dbl>, e_mort_exc_tbhiv_100k_hi <dbl>,
## # e_mort_exc_tbhiv_num <dbl>, e_mort_exc_tbhiv_num_lo <dbl>, ...
```

```
#Merge tb data with data dictionary
```

```
tbData=tb_estimates %>% gather("variable_name","value",7:50) %>% left_join(datadict)
```

```
## Joining, by = "variable_name"
```

```
#remove unnecessary regional columns, blank code_list column
```

```
tbData=subset(tbData, select = -c(iso2,iso3,iso_numeric,g_whoregion,code_list) )
print(tbData[1:5,])
```

```
## # A tibble: 5 x 6
```

	country	year	variable_name	value	dataset	definition
	<chr>	<dbl>	<chr>	<dbl>	<chr>	<chr>
## 1	Kenya	2000	e_pop_num	30851606	Estimates	Estimated total population num~
## 2	Kenya	2001	e_pop_num	31800343	Estimates	Estimated total population num~
## 3	Kenya	2002	e_pop_num	32779823	Estimates	Estimated total population num~
## 4	Kenya	2003	e_pop_num	33767122	Estimates	Estimated total population num~
## 5	Kenya	2004	e_pop_num	34791836	Estimates	Estimated total population num~

4.2 Exploring new incident infections (all infections and HIV)

Incident infections are the number of estimated people being infected with TB each year. The number of new infections is an estimate and is different from the number of reported cases or diagnoses - which is reliant on the identification, testing and treating of people with TB. This data represent a key element in the transmission model and it is important in understanding the past dynamics of TB in Kenya and provides an idea on the current trend.

HIV is an important factor to consider, given that Kenya has relatively high HIV/TB coinfection and because HIV impacts the likelihood of contracting TB, becoming infectious or of becoming severely ill.

```
#select relevant variables related to incidence
```

```
inc_data= tbData %>% filter(variable_name %in% c("e_inc_num","e_inc_num_lo","e_inc_num_hi")) %>% mutate(
hiv_inc=tbData %>% filter(variable_name%in% c("e_inc_tbhiv_num","e_inc_tbhiv_num_lo","e_inc_tbhiv_num_hi"))
hiv_perc_inc=tbData %>% filter(variable_name%in% c("e_tbhiv_prct","e_tbhiv_prct_lo","e_tbhiv_prct_hi")))
```

```
#label upper, lower and mean estimates
```

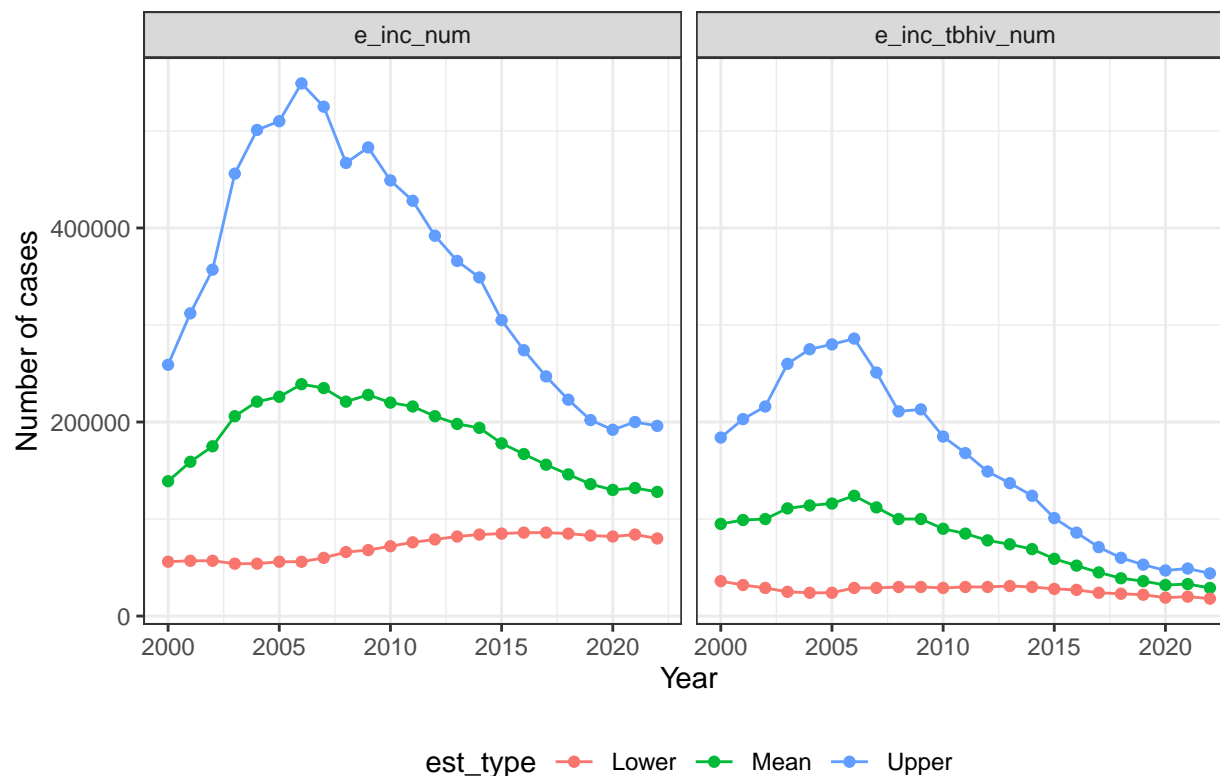
```
all_inc=rbind(inc_data,hiv_inc)
all_inc$est_type="Mean"
all_inc$est_type[grepl("_lo",all_inc$variable_name,fixed=TRUE)==TRUE]="Lower"
all_inc$est_type[grepl("_hi",all_inc$variable_name,fixed=TRUE)==TRUE]="Upper"
```

```
hiv_perc_inc$est_type="Mean"
hiv_perc_inc$est_type[grepl("_lo",hiv_perc_inc$variable_name,fixed=TRUE)==TRUE]="Lower"
hiv_perc_inc$est_type[grepl("_hi",hiv_perc_inc$variable_name,fixed=TRUE)==TRUE]="Upper"
```

```
#Incident cases (all and HIV)
```

```
ggplot(all_inc,aes(x=year,y=value,group=variable_name,color=est_type))+geom_point()+
  geom_line()+theme_bw()+xlab("Year")+ylab("Number of cases")+
  labs(title="Estimated number of new cases per year")+theme(legend.position = "bottom")+facet_wrap(.~v
```

Estimated number of new cases per year



#Percentage of new cases HIV positive

```
ggplot(hiv_perc_inc,aes(x=year,y=value,group=variable_name,color=est_type))+geom_point()+
  geom_line()+theme_bw()+xlab("Year")+ylab("Number of cases")+
  labs(title="Estimated number of new cases per year")+theme(legend.position = "bottom")+facet_wrap(.~v
```

Estimated number of new cases per year

