

Week 1: Intro to demographic concepts

SOC6708 ADA

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Read in the data

All these data come from the [UN World Population Prospects 2024](#).

Packages:

```
library(tidyverse)
library(here)
library(readxl)
library(janitor)
```

Population data:

```
d_male <- read_xlsx("../data/WPP2024_POP_F01_2_POPULATION_SINGLE AGE MALE.xlsx", skip = 16)
d_male$sex <- "Male"
d_male <- d_male |> drop_na(Year)
d_female <- read_xlsx("../data/WPP2024_POP_F01_3_POPULATION_SINGLE AGE FEMALE.xlsx", skip = 16)
d_female$sex <- "Female"
d_female <- d_female |> drop_na(Year)

d <- rbind(d_male, d_female)
rm(d_male, d_female)

d <- d |>
  clean_names() |>
  select(region_subregion_country_or_area, iso3_alpha_code, year, x0:sex) |>
  rename(region = region_subregion_country_or_area) |>
  mutate(across(x0:x100, as.numeric))

head(d)
```

```

# A tibble: 6 x 105
  region iso3_alpha_code year     x0     x1     x2     x3     x4     x5     x6
  <chr>  <lgl>          <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>
1 World  NA              1950 41603. 36997. 34245. 32359. 30007. 28065. 27335.
2 World  NA              1951 42615. 38588. 35577. 33456. 31876. 29664. 27829.
3 World  NA              1952 43984. 39615. 37182. 34798. 32980. 31545. 29427.
4 World  NA              1953 45183. 41001. 38243. 36421. 34326. 32646. 31320.
5 World  NA              1954 45934. 42199. 39641. 37489. 35955. 33999. 32422.
6 World  NA              1955 47059. 42972. 40848. 38889. 37023. 35636. 33784.
# i 95 more variables: x7 <dbl>, x8 <dbl>, x9 <dbl>, x10 <dbl>, x11 <dbl>,
#   x12 <dbl>, x13 <dbl>, x14 <dbl>, x15 <dbl>, x16 <dbl>, x17 <dbl>,
#   x18 <dbl>, x19 <dbl>, x20 <dbl>, x21 <dbl>, x22 <dbl>, x23 <dbl>,
#   x24 <dbl>, x25 <dbl>, x26 <dbl>, x27 <dbl>, x28 <dbl>, x29 <dbl>,
#   x30 <dbl>, x31 <dbl>, x32 <dbl>, x33 <dbl>, x34 <dbl>, x35 <dbl>,
#   x36 <dbl>, x37 <dbl>, x38 <dbl>, x39 <dbl>, x40 <dbl>, x41 <dbl>,
#   x42 <dbl>, x43 <dbl>, x44 <dbl>, x45 <dbl>, x46 <dbl>, x47 <dbl>, ...

```

Mortality data:

```

d_male <- read_xlsx(here("data/WPP2024_MORT_F01_2_DEATHS_SINGLE_AGE_MALE.xlsx"), skip = 16)
d_male$sex <- "Male"
d_male <- d_male |> drop_na(Year)
d_female <- read_xlsx(here("data/WPP2024_MORT_F01_3_DEATHS_SINGLE_AGE_FEMALE.xlsx"), skip = 16)
d_female$sex <- "Female"
d_female <- d_female |> drop_na(Year)

dm <- rbind(d_male, d_female)
rm(d_male, d_female)

```

```

dm <- dm |>
  clean_names() |>
  select(region_subregion_country_or_area, iso3_alpha_code, year, x0:sex) |>
  rename(region = region_subregion_country_or_area) |>
  mutate(across(x0:x100, as.numeric()))

```

```
head(dm)
```

```

# A tibble: 6 x 105
  region iso3_alpha_code year     x0     x1     x2     x3     x4     x5     x6     x7
  <chr>  <lgl>          <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>
1 World  NA              1950 6779. 1815. 967. 588. 380. 255. 183. 142.
2 World  NA              1951 6780. 1796. 968. 582. 382. 257. 181. 138.

```

```

3 World NA 1952 6867. 1743. 943. 578. 375. 252. 176. 132.
4 World NA 1953 6831. 1726. 933. 572. 379. 264. 191. 142.
5 World NA 1954 6841. 1706. 928. 568. 375. 252. 180. 136.
6 World NA 1955 6815. 1663. 917. 566. 373. 256. 181. 137.
# i 94 more variables: x8 <dbl>, x9 <dbl>, x10 <dbl>, x11 <dbl>, x12 <dbl>,
# x13 <dbl>, x14 <dbl>, x15 <dbl>, x16 <dbl>, x17 <dbl>, x18 <dbl>,
# x19 <dbl>, x20 <dbl>, x21 <dbl>, x22 <dbl>, x23 <dbl>, x24 <dbl>,
# x25 <dbl>, x26 <dbl>, x27 <dbl>, x28 <dbl>, x29 <dbl>, x30 <dbl>,
# x31 <dbl>, x32 <dbl>, x33 <dbl>, x34 <dbl>, x35 <dbl>, x36 <dbl>,
# x37 <dbl>, x38 <dbl>, x39 <dbl>, x40 <dbl>, x41 <dbl>, x42 <dbl>,
# x43 <dbl>, x44 <dbl>, x45 <dbl>, x46 <dbl>, x47 <dbl>, x48 <dbl>, ...

```

Crude Rates

Calculate the crude death rates for Kenya and Canada in 2023

```

# get total populations across all age and sex
total_pops <- d  |>
  filter(region=="Kenya" | region=="Canada") |>
  pivot_longer(x0:x100, names_to = "age", values_to = "pop") |>
  mutate(age = as.numeric(str_remove(age, "x")))) |>
  group_by(region, year) |>
  summarize(total_pop = sum(pop))

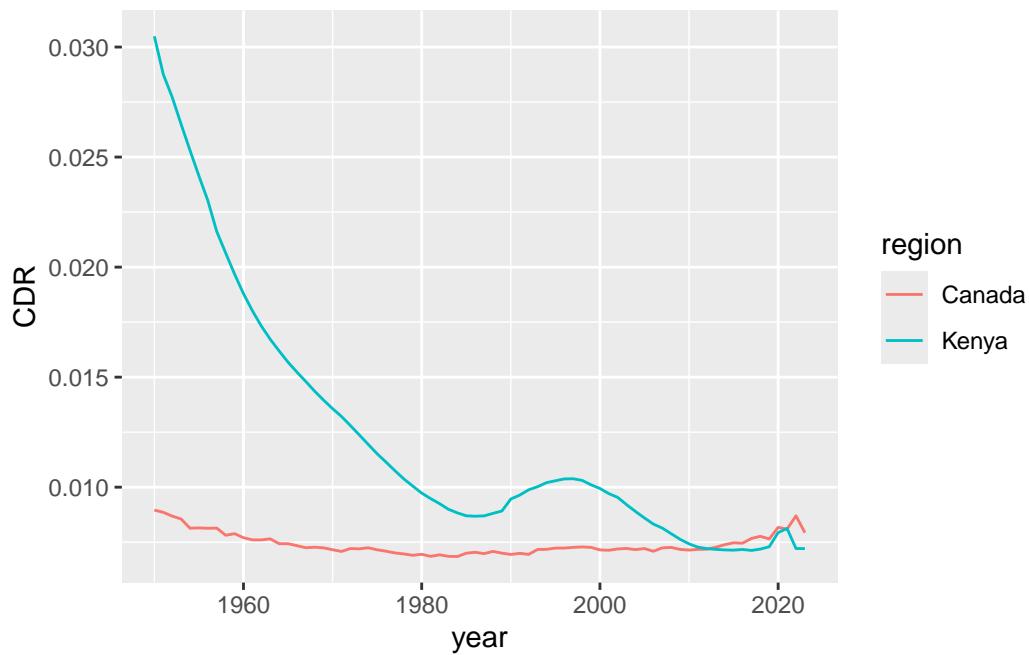
# get total deaths

total_deaths <- dm  |>
  filter(region=="Kenya" | region=="Canada") |>
  pivot_longer(x0:x100, names_to = "age", values_to = "deaths") |>
  mutate(age = as.numeric(str_remove(age, "x")))) |>
  group_by(region, year) |>
  summarize(total_deaths = sum(deaths))

# join these

total_pops |>
  left_join(total_deaths) |>
  mutate(CDR = total_deaths/total_pop) |>
  ggplot(aes(year, CDR, color = region)) +
  geom_line()

```



```
total_pops |>
  left_join(total_deaths) |>
  mutate(CDR = total_deaths/total_pop) |>
  filter(year==2023)
```

```
# A tibble: 2 x 5
# Groups:   region [2]
  region  year total_pop total_deaths      CDR
  <chr>   <dbl>     <dbl>        <dbl>    <dbl>
1 Canada   2023     39299.       312.  0.00793
2 Kenya    2023     55339.       399.  0.00721
```

Population Pyramids

Population age structures of Kenya and Canada in 2019:

```
d_long <- d |>
  pivot_longer(x0:x100, names_to = "age", values_to = "pop") |>
  mutate(age = as.numeric(str_remove(age, "x")))

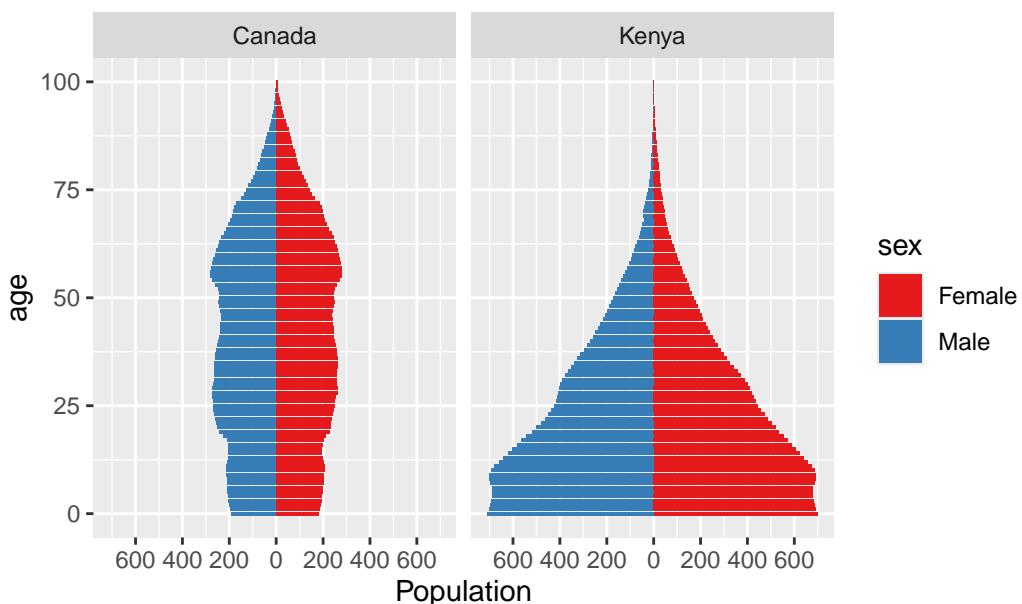
d_long|>
```

```

filter(region=="Kenya" | region=="Canada", year == 2019) |>
  mutate(population = ifelse(sex == "Male", -pop, pop)) |>
  ggplot(aes(x = age, y = population, fill = sex)) +
  facet_wrap(~region) +
  geom_bar(stat = "identity") +
  ggttitle("Population in each age group") +
  ylab("Population") +
  coord_flip() +
  scale_y_continuous(breaks = seq(-600, 600, 200),
                     labels = c(seq(600, 0, -200), seq(200, 600, 200))) +
  scale_fill_brewer(palette = "Set1")

```

Population in each age group



```
ggsave(here("plots", "CAN_KEN_pyramid.pdf"))
```

Change in age structures over time:

```

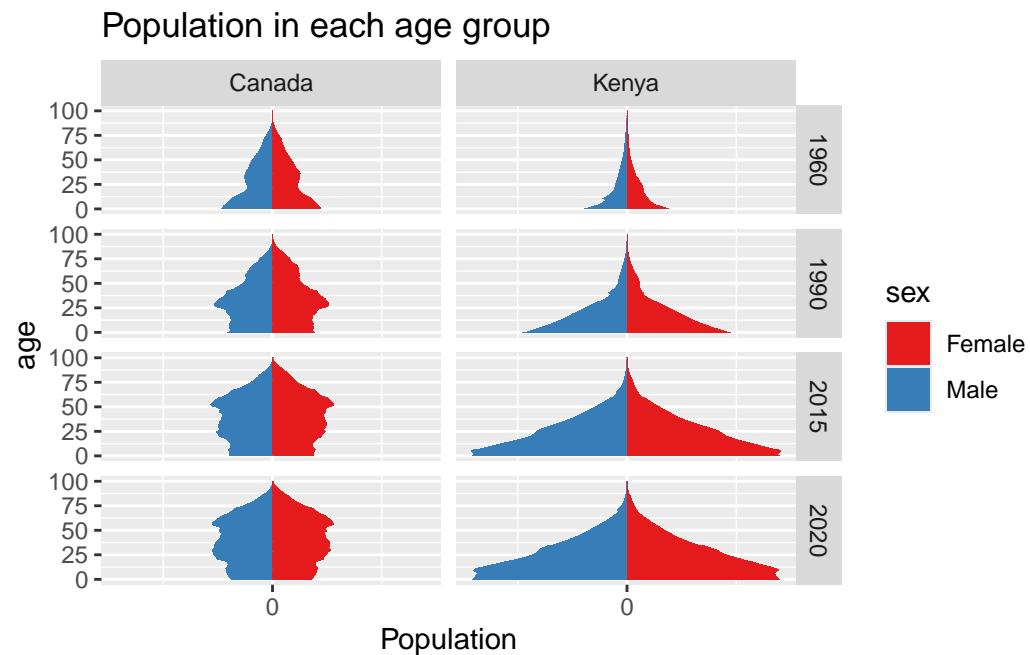
d_long |>
  filter(region == "Kenya" | region == "Canada", year %in% c(1960, 1990, 2015, 2020)) |>
  mutate(population = ifelse(sex == "Male", -pop, pop)) %>%
  ggplot(aes(x = age, y = population, fill = sex)) +
  facet_grid(year ~ region) +
  geom_bar(stat = "identity") +

```

```

ggtitle("Population in each age group")+
  ylab("Population")+
  coord_flip() +
  scale_y_continuous(breaks = seq(-4000, 4000, 1000),
                     labels = c(seq(4000, 0, -1000), seq(1000, 4000, 1000))) +
  scale_fill_brewer(palette = "Set1")

```



```
ggsave(here("plots", "CAN_KEN_pyramid_time.pdf"))
```

Age-specific rates

Create and plot age-specific mortality rates (across both sexes)

```

pops <- d_long

dm_long <- dm |>
  pivot_longer(x0:x100, names_to = "age", values_to = "deaths") |>
  mutate(age = as.numeric(str_remove(age, "x")))

# join these two tibbles and calculate rates

```

```

asmr <- d_long |>
  left_join(dm_long) |>
  mutate(mx = deaths/pop)

head(asmr)

# A tibble: 6 x 8
  region iso3_alpha_code year sex     age   pop deaths      mx
  <chr>    <lgl>        <dbl> <chr> <dbl> <dbl> <dbl>    <dbl>
1 World    NA            1950 Male     0 41603. 6779. 0.163
2 World    NA            1950 Male     1 36997. 1815. 0.0491
3 World    NA            1950 Male     2 34245. 967. 0.0282
4 World    NA            1950 Male     3 32359. 588. 0.0182
5 World    NA            1950 Male     4 30007. 380. 0.0127
6 World    NA            1950 Male     5 28065. 255. 0.00909

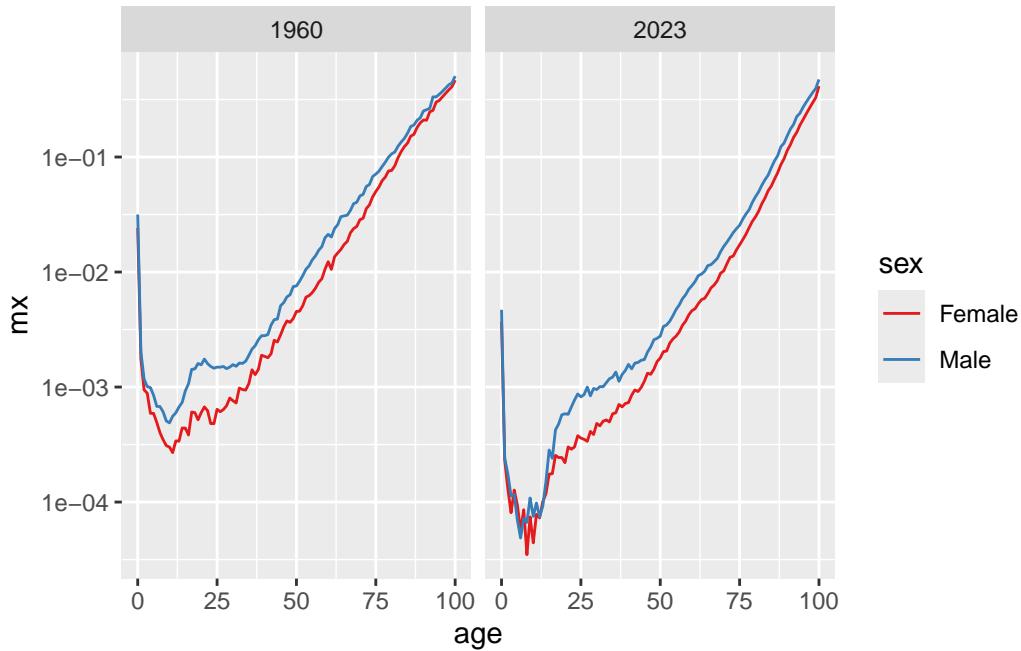
```

Mortality rates for Canada

```

asmr |>
  filter(region == "Canada", year %in% c(1960, 2023)) |>
  ggplot(aes(age, mx, color = sex)) +
  geom_line()+
  scale_y_log10()+
  facet_wrap(~year)+
  labs("Age- and sex-specific mortality rates for Canada")+
  scale_color_brewer(palette = "Set1")

```



```
ggsave(here("plots", "CAN_mortality.pdf"), width = 6, height = 4)
```

Age-standardized rates

What would Kenya mortality look like in 2023 with Canada's age structure?

```
kenya_2023 <- asmr |>
  filter(year==2023, region=="Kenya") |>
  rename(kpop = pop, kdeath = deaths, kmx = mx) |>
  select(-region)

canada_2023 <- asmr |>
  ungroup() |>
  filter(year==2023, region=="Canada") |>
  rename(cpop = pop, cdeath = deaths, cmx = mx) |>
  select(-region, -age, -year, -sex, -iso3_alpha_code)

kc_2023 <- bind_cols(kenya_2023, canada_2023)

# now calculate age-standardized rates using canada's population

kc_2023 |>
```

```

    mutate(std_deaths_kenya = cpop*kmx,
           std_deaths_canada = cpop*cmx) |>
summarise(std_rate_kenya = sum(std_deaths_kenya)/sum(cpop),
           std_rate_canada = sum(std_deaths_canada)/sum(cpop))

```

```

# A tibble: 1 x 2
  std_rate_kenya std_rate_canada
  <dbl>          <dbl>
1     0.0223        0.00793

```

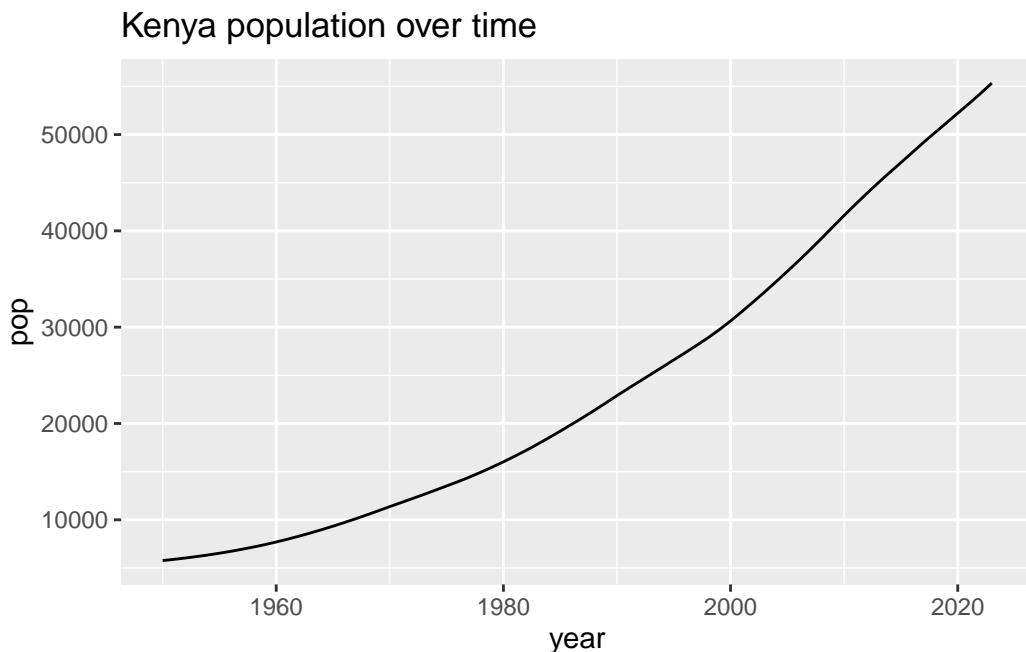
Population Growth

Plot Kenya total population over time

```

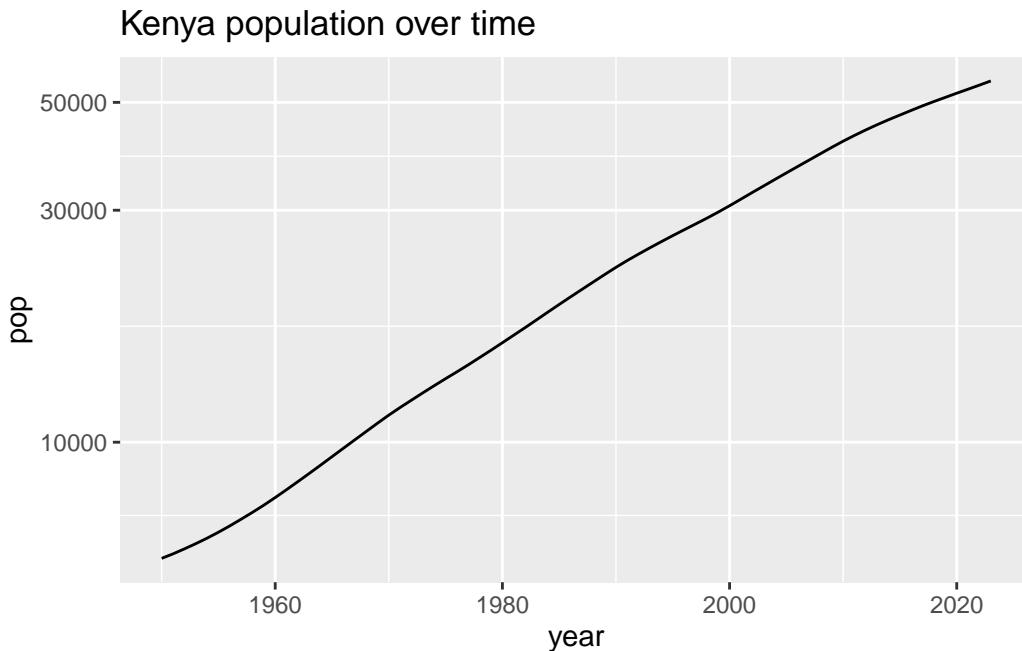
d_long |>
  filter(region=="Kenya") |>
  group_by(year) |>
  summarize(pop = sum(pop)) |>
  ggplot(aes(year, pop)) +
  geom_line() +
  ggtitle("Kenya population over time")

```



What does the log look like?

```
d_long |>
  filter(region=="Kenya") |>
  group_by(year) |>
  summarize(pop = sum(pop)) |>
  ggplot(aes(year, pop)) +
  geom_line() +
  ggtitle("Kenya population over time")+
  scale_y_log10()
```



```
ggsave(here("plots", "KEN_pop.pdf"))
```

Pretty straight! Let's calculate the growth rate from 1950 to 2023. This is just the slope of the logged graph. It's about 3% per year.

```
d_long |>
  filter(region=="Kenya") |>
  group_by(year) |>
  summarize(pop = sum(pop)) |>
  mutate(log_pop = log(pop))|>
  summarise(growth_rate = (log_pop[year==2023] - log_pop[year==1950])/(2023-1950))
```

```
# A tibble: 1 x 1
  growth_rate
  <dbl>
1      0.0310
```

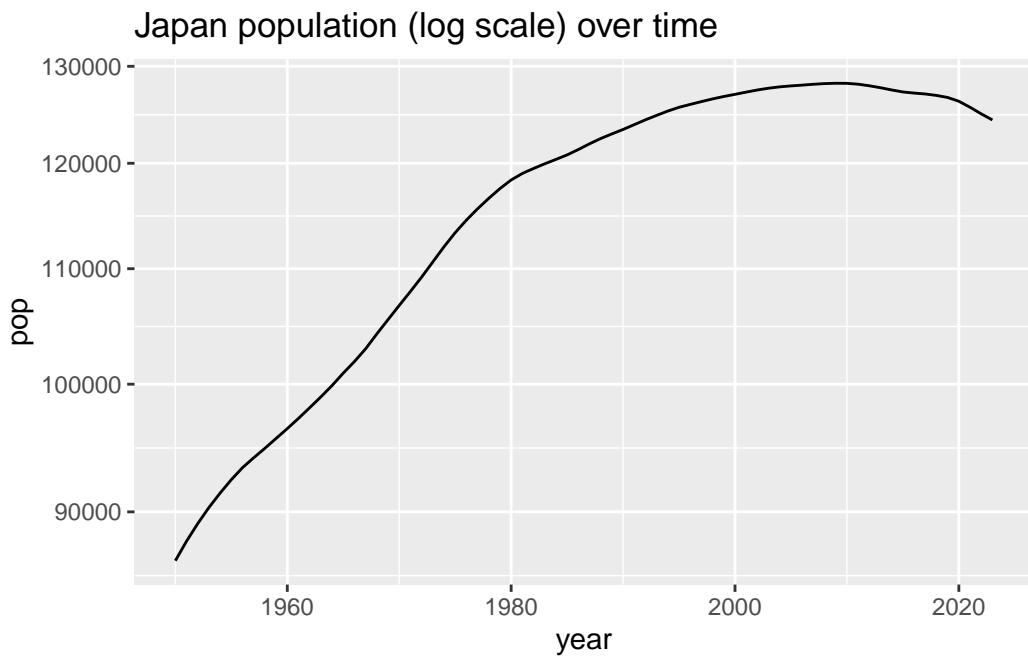
What about Canada? About half that.

```
d_long |>
  filter(region=="Canada") |>
  group_by(year) |>
  summarize(pop = sum(pop)) |>
  mutate(log_pop = log(pop)) |>
  summarise(growth_rate = (log_pop[year==2023] - log_pop[year==1950])/(2023-1950))
```

```
# A tibble: 1 x 1
  growth_rate
  <dbl>
1      0.0144
```

Some countries have stagnated:

```
d_long |>
  filter(region=="Japan") |>
  group_by(year) |>
  summarize(pop = sum(pop)) |>
  ggplot(aes(year, pop)) + geom_line()+
  scale_y_log10()+
  ggtitle("Japan population (log scale) over time")
```



Questions

1. Pick two countries. Plot their population pyramids in 1960, 1990, and 2020.
2. Based on the shape and change in population pyramids in 1, do you think the population growth rate in each of your chosen countries is positive or negative in recent years?
3. Confirm your guesses in 2 by plotting population over time in both countries.
4. Taking the US population in 2000 as the standard population, calculate the standardized mortality rates for US in 2023 and Australia in 2023. How much higher was the death rate in US compared to Australia?