# DigHum100Project

### June 14, 2024

# Requirement already satisfied: seaborn in /srv/conda/lib/python3.11/site-packages (0.13.2) Requirement already satisfied: numpy!=1.24.0,>=1.20 in /srv/conda/lib/python3.11/site-packages (from seaborn) (1.26.4) Requirement already satisfied: pandas>=1.2 in /srv/conda/lib/python3.11/site-packages (from seaborn) (2.2.2)

Requirement already satisfied: matplotlib!=3.6.1,>=3.4 in /srv/conda/lib/python3.11/site-packages (from seaborn) (3.7.3)

Requirement already satisfied: contourpy>=1.0.1 in

[1]: pip install --upgrade seaborn

/srv/conda/lib/python3.11/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn)
(1.2.1)

Requirement already satisfied: cycler>=0.10 in /srv/conda/lib/python3.11/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (0.12.1)

Requirement already satisfied: fonttools>=4.22.0 in

/srv/conda/lib/python3.11/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn)
(4.53.0)

Requirement already satisfied: kiwisolver>=1.0.1 in

/srv/conda/lib/python3.11/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn)
(1.4.5)

Requirement already satisfied: packaging>=20.0 in

/srv/conda/lib/python3.11/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn)
(24.0)

Requirement already satisfied: pillow>=6.2.0 in /srv/conda/lib/python3.11/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (10.3.0)

Requirement already satisfied: pyparsing>=2.3.1 in

/srv/conda/lib/python3.11/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn)
(3.1.2)

Requirement already satisfied: python-dateutil>=2.7 in

/srv/conda/lib/python3.11/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn)
(2.9.0)

Requirement already satisfied: pytz>=2020.1 in /srv/conda/lib/python3.11/site-packages (from pandas>=1.2->seaborn) (2024.1)

Requirement already satisfied: tzdata>=2022.7 in /srv/conda/lib/python3.11/site-packages (from pandas>=1.2->seaborn) (2024.1)

Requirement already satisfied: six>=1.5 in /srv/conda/lib/python3.11/site-packages (from python-dateutil>=2.7->matplotlib!=3.6.1,>=3.4->seaborn) (1.16.0)

Note: you may need to restart the kernel to use updated packages.

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

def isolate_vals(dataset):
    new = dataset[['Location', 'Value', 'Period']]
    value_str = dataset['Indicator'].values[0]
    new = new.rename(columns = {'Location' : 'income_group', 'Value' :___
    value_str, 'Period' : 'year'})
    return new
def extract_percentage(dataset, column_name):
    percentage_str = dataset[column_name].str[:4]
    dataset[column_name] = percentage_str.str.strip('['))
    return dataset
```

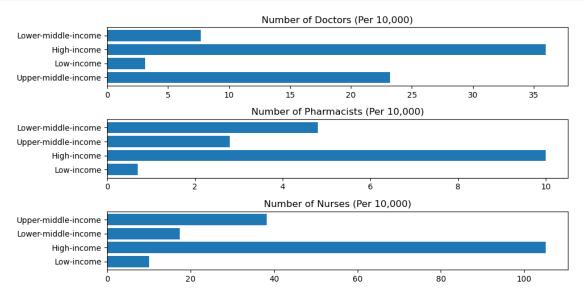
### 0.1 Staff availability

```
[3]: income_group count year
0 Upper-middle-income 23.2 2022
1 Low-income 3.1 2022
2 High-income 36.0 2022
3 Lower-middle-income 7.7 2022
```

```
[4]: nurses = pd.read_csv('dighum nurse dataset.csv')
nurses_cleaned = isolate_vals(nurses)
nurses_cleaned = nurses_cleaned.rename(columns = {'Nursing and midwifery_
personnel (per 10,000)' : 'count'})
nurses_cleaned
```

```
[5]: income_group count year
0 Low-income 0.7 2022
1 High-income 10.0 2022
2 Upper-middle-income 2.8 2022
3 Lower-middle-income 4.8 2022
```

```
[6]: df_list = [cleaned_doc, pharma_cleaned, nurses_cleaned]
  fig, ax = plt.subplots(3, 1, figsize = (10, 5))
  for i in range(len(df_list)):
        ax[i].barh(y = 'income_group', width = 'count', data = df_list[i])
        ax[0].set_title('Number of Doctors (Per 10,000)')
        ax[1].set_title('Number of Pharmacists (Per 10,000)')
        ax[2].set_title('Number of Nurses (Per 10,000)')
        plt.tight_layout();
```



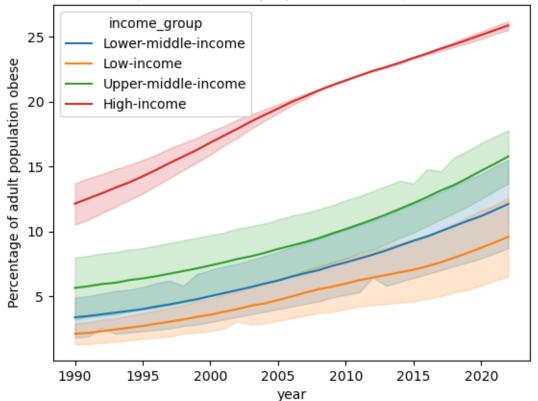
### 0.2 Nutrition related conditions

```
[7]: obesity = pd.read_csv('dighum obesity dataset.csv')
obesity
cleaned_obesity = isolate_vals(obesity)
cleaned_obesity = extract_percentage(cleaned_obesity, 'Prevalence of obesity_
among adults, BMI ≥ 30 (age-standardized estimate) (%)')
```

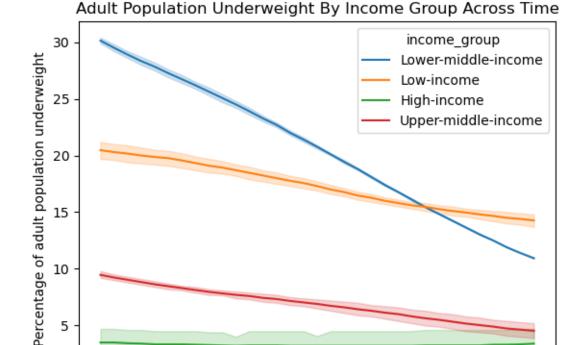
```
cleaned_obesity['Prevalence of obesity among adults, BMI ≥ 30_\( \) \( \) (age-standardized estimate) (%)'] = cleaned_obesity['Prevalence of obesity_\( \) \( \) \( \) among adults, BMI ≥ 30 (age-standardized estimate) (%)']. \( \) \( \) astype(float) \( \) grouped = cleaned_obesity.drop(columns = ['year']).groupby(['income_group']). \( \) \( \) \( \) mean() \( \) grouped
```

```
[8]: lp = sns.lineplot(data = cleaned_obesity, x = 'year', y = 'Prevalence of obesity among adults, BMI ≥ 30 (age-standardized estimate) (%)', hue = 'income_group')
lp.set_ylabel('Percentage of adult population obese')
lp.set_title('Adult Population Obesity By Income Group Across Time');
```

## Adult Population Obesity By Income Group Across Time



```
[9]: under = pd.read_csv('dighum underweight dataset.csv')
     under_cleaned = isolate_vals(under)
     under_cleaned = extract_percentage(under_cleaned, 'Prevalence of underweight_
      →among adults, BMI < 18 (age-standardized estimate) (%)')
     under_cleaned['Prevalence of underweight among adults, BMI < 18__
      among adults, BMI < 18 (age-standardized estimate) (%)'].astype(float)</pre>
     grouped = under_cleaned.drop(columns = ['year']).groupby(['income_group']).
      →mean()
     grouped
[9]:
                         Prevalence of underweight among adults, BMI < 18 (age-
     standardized estimate) (%)
     income_group
     High-income
                                                                3.273737
     Low-income
                                                               17.298990
     Lower-middle-income
                                                               20.567677
     Upper-middle-income
                                                                6.864646
[11]: | lp = sns.lineplot(data = under_cleaned, x = 'year', y = 'Prevalence of_u
      ounderweight among adults, BMI < 18 (age-standardized estimate) (%)',
                 hue = 'income_group')
     lp.set_ylabel('Percentage of adult population underweight')
     lp.set_title('Adult Population Underweight By Income Group Across Time');
```



2005

year

2010

Stunting prevalence among children under 5 years of age

2015

2020

4.651515 33.336364

44.078788

40.148485

```
[17]: stunted = pd.read csv('dighum stunted dataset.csv')
      stunted_cleaned = isolate_vals(stunted)
      stunted_cleaned = extract_percentage(stunted_cleaned, 'Stunting prevalence_
       →among children under 5 years of age (% height-for-age <-2 SD), model-based
       ⇔estimates')
      stunted_cleaned['Stunting prevalence among children under 5 years of age (%__
       ⇔height-for-age <-2 SD), model-based estimates'] = stunted cleaned['Stunting__</pre>
       oprevalence among children under 5 years of age (% height-for-age <-2 SD), □

¬model-based estimates'].astype(float)
      grouped = stunted_cleaned.drop(columns = ['year']).groupby(['income_group']).
       →mean()
      grouped
```

5

[17]:

income\_group High-income

Low-income

Low- and middle-income

Lower-middle-income

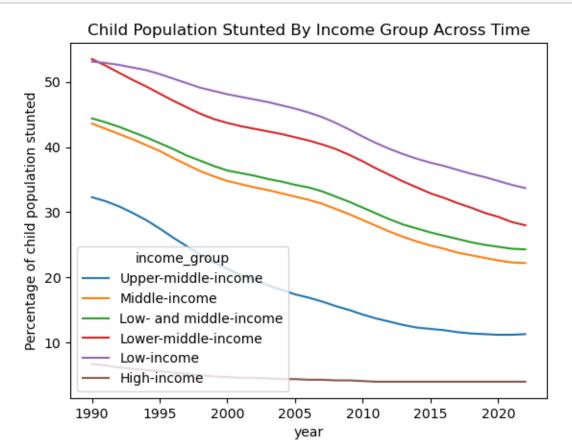
1990

1995

(% height-for-age <-2 SD), model-based estimates

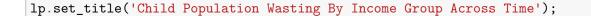
2000

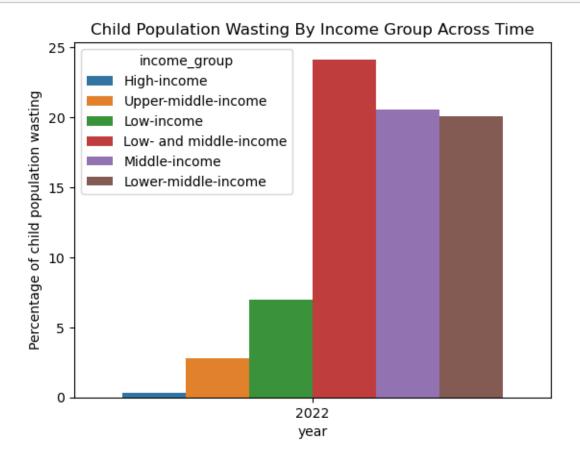
```
Middle-income 31.627273
Upper-middle-income 18.648485
```



```
wasting = pd.read_csv('dighum wasting dataset.csv')
wasting_cleaned = isolate_vals(wasting)
wasting_cleaned = extract_percentage(wasting_cleaned, 'Wasted prevalence among_
children under 5 years of age (%), model-based estimates')
wasting_cleaned['Wasted prevalence among children under 5 years of age (%),__
model-based estimates'] = wasting_cleaned['Wasted prevalence among children__
under 5 years of age (%), model-based estimates'].astype(float)
wasting_cleaned
```

```
[21]:
                     income_group \
      0
                     High-income
             Upper-middle-income
      1
      2
                       Low-income
          Low- and middle-income
      3
      4
                   Middle-income
      5
             Lower-middle-income
                      High-income
      6
      7
             Upper-middle-income
             Lower-middle-income
      8
      9
                   Middle-income
      10
          Low- and middle-income
                       Low-income
      11
          Wasted prevalence among children under 5 years of age (%), model-based
      estimates \
      0
                                                           0.4
                                                           2.1
      1
      2
                                                           6.7
      3
                                                           6.9
      4
                                                           7.0
      5
                                                           9.6
      6
                                                           0.2
      7
                                                           3.5
      8
                                                          30.5
      9
                                                          34.1
                                                          41.4
      10
      11
                                                           7.3
          year
          2022
      0
          2022
      1
      2
          2022
      3
          2022
      4
          2022
      5
          2022
      6
          2022
      7
          2022
          2022
      8
          2022
      9
         2022
      10
          2022
      11
[32]: | lp = sns.barplot(data = wasting_cleaned, x = 'year', y = 'Wasted prevalence_
       →among children under 5 years of age (%), model-based estimates',
                       errorbar = None, hue = 'income_group')
      lp.set_ylabel('Percentage of child population wasting')
```

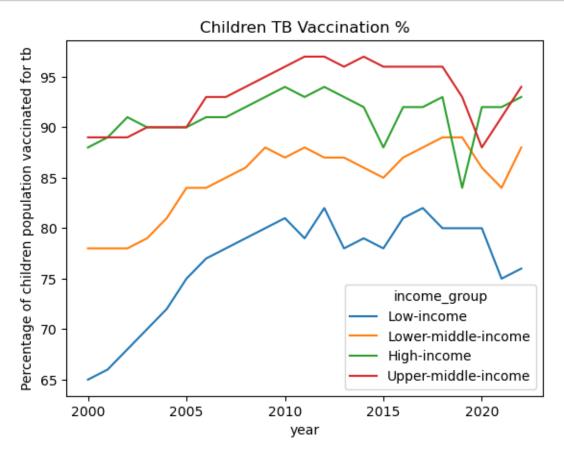




### 0.3 Vaccine Available Diseases Coverage

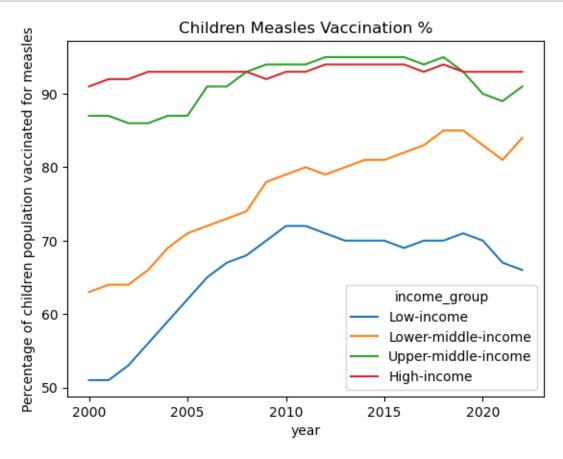
```
[33]: tb = pd.read_csv('dighum tb dataset.csv')
      cleaned_tb = isolate_vals(tb)
      cleaned_tb
      grouped = cleaned_tb.drop(columns = ['year']).groupby(['income_group']).mean()
      grouped
[33]:
                           BCG immunization coverage among 1-year-olds (%)
      income_group
      High-income
                                                                  91.173913
      Low-income
                                                                  76.565217
      Lower-middle-income
                                                                  84.869565
      Upper-middle-income
                                                                  93.260870
[34]: | lp = sns.lineplot(data = cleaned_tb, x = 'year', y = 'BCG immunization coverage_
       →among 1-year-olds (%)',
```

```
hue = 'income_group')
lp.set_ylabel('Percentage of children population vaccinated for tb')
lp.set_title('Children TB Vaccination %');
```

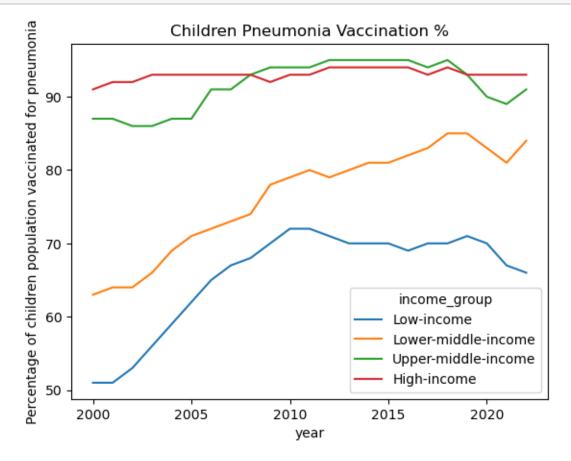


```
[35]: Measles-containing-vaccine first-dose (MCV1) immunization coverage among 1-year-olds (%) income_group
High-income 93.043478
Low-income 65.652174
Lower-middle-income 76.391304
Upper-middle-income 91.478261
```

```
[37]: lp = sns.lineplot(data = measles_cleaned, x = 'year', y = \( \) 'Measles-containing-vaccine first-dose (MCV1) immunization coverage among \( \) \( \) '1-year-olds (%)', \( \) hue = 'income_group') lp.set_ylabel('Percentage of children population vaccinated for measles') lp.set_title('Children Measles Vaccination %');
```



```
[38]: Measles-containing-vaccine first-dose (MCV1) immunization coverage among 1-year-olds (%) income_group
High-income 93.043478
Low-income 65.652174
Lower-middle-income 76.391304
```



[]: