

project1

February 17, 2021

```
[1]: !pip install wbdata

import pandas as pd
import wbdata
import numpy as np
import re
import seaborn as sns
import matplotlib.pyplot as plt
from matplotlib import animation as ani
import matplotlib.patches as mpatches
```

Requirement already satisfied: wbdata in /opt/conda/lib/python3.8/site-packages (0.3.0)
Requirement already satisfied: tabulate>=0.8.5 in /opt/conda/lib/python3.8/site-packages (from wbdata) (0.8.7)
Requirement already satisfied: requests>=2.0 in /opt/conda/lib/python3.8/site-packages (from wbdata) (2.25.1)
Requirement already satisfied: appdirs<2.0,>=1.4 in /opt/conda/lib/python3.8/site-packages (from wbdata) (1.4.4)
Requirement already satisfied: decorator>=4.0 in /opt/conda/lib/python3.8/site-packages (from wbdata) (4.4.2)
Requirement already satisfied: urllib3<1.27,>=1.21.1 in /opt/conda/lib/python3.8/site-packages (from requests>=2.0->wbdata) (1.25.7)
Requirement already satisfied: idna<3,>=2.5 in /opt/conda/lib/python3.8/site-packages (from requests>=2.0->wbdata) (2.8)
Requirement already satisfied: certifi>=2017.4.17 in /opt/conda/lib/python3.8/site-packages (from requests>=2.0->wbdata) (2019.11.28)
Requirement already satisfied: chardet<5,>=3.0.2 in /opt/conda/lib/python3.8/site-packages (from requests>=2.0->wbdata) (3.0.4)

0.1 Loading the Data

```
[2]: # Return list of all country/region codes:
wbdata.get_country()
```

```
[2]: id    name
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ABW	Aruba
AFG	Afghanistan
AFR	Africa
AGO	Angola
ALB	Albania
AND	Andorra
ARB	Arab World
ARE	United Arab Emirates
ARG	Argentina
ARM	Armenia
ASM	American Samoa
ATG	Antigua and Barbuda
AUS	Australia
AUT	Austria
AZE	Azerbaijan
BDI	Burundi
BEA	East Asia & Pacific (IBRD-only countries)
BEC	Europe & Central Asia (IBRD-only countries)
BEL	Belgium
BEN	Benin
BFA	Burkina Faso
BGD	Bangladesh
BGR	Bulgaria
BHI	IBRD countries classified as high income
BHR	Bahrain
BHS	Bahamas, The
BIH	Bosnia and Herzegovina
BLA	Latin America & the Caribbean (IBRD-only countries)
BLR	Belarus
BLZ	Belize
BMN	Middle East & North Africa (IBRD-only countries)
BMU	Bermuda
BOL	Bolivia
BRA	Brazil
BRB	Barbados
BRN	Brunei Darussalam
BSS	Sub-Saharan Africa (IBRD-only countries)
BTN	Bhutan
BWA	Botswana
CAA	Sub-Saharan Africa (IFC classification)
CAF	Central African Republic
CAN	Canada
CEA	East Asia and the Pacific (IFC classification)
CEB	Central Europe and the Baltics
CEU	Europe and Central Asia (IFC classification)
CHE	Switzerland
CHI	Channel Islands

CHL	Chile
CHN	China
CIV	Cote d'Ivoire
CLA	Latin America and the Caribbean (IFC classification)
CME	Middle East and North Africa (IFC classification)
CMR	Cameroon
COD	Congo, Dem. Rep.
COG	Congo, Rep.
COL	Colombia
COM	Comoros
CPV	Cabo Verde
CRI	Costa Rica
CSA	South Asia (IFC classification)
CSS	Caribbean small states
CUB	Cuba
CUW	Curacao
CYM	Cayman Islands
CYP	Cyprus
CZE	Czech Republic
DEA	East Asia & Pacific (IDA-eligible countries)
DEC	Europe & Central Asia (IDA-eligible countries)
DEU	Germany
DFS	IDA countries classified as Fragile Situations
DJI	Djibouti
DLA	Latin America & the Caribbean (IDA-eligible countries)
DMA	Dominica
DMN	Middle East & North Africa (IDA-eligible countries)
DNF	IDA countries not classified as Fragile Situations
DNK	Denmark
DNS	IDA countries in Sub-Saharan Africa not classified as fragile situations
DOM	Dominican Republic
DSA	South Asia (IDA-eligible countries)
DSF	IDA countries in Sub-Saharan Africa classified as fragile situations
DSS	Sub-Saharan Africa (IDA-eligible countries)
DZA	Algeria
EAP	East Asia & Pacific (excluding high income)
EAR	Early-demographic dividend
EAS	East Asia & Pacific
ECA	Europe & Central Asia (excluding high income)
ECS	Europe & Central Asia
ECU	Ecuador
EGY	Egypt, Arab Rep.
EMU	Euro area
ERI	Eritrea
ESP	Spain
EST	Estonia
ETH	Ethiopia

EUU	European Union
FCS	Fragile and conflict affected situations
FIN	Finland
FJI	Fiji
FRA	France
FRD	Faroe Islands
FSM	Micronesia, Fed. Sts.
FXS	IDA countries classified as fragile situations, excluding Sub-Saharan Africa
GAB	Gabon
GBR	United Kingdom
GEO	Georgia
GHA	Ghana
GIB	Gibraltar
GIN	Guinea
GMB	Gambia, The
GNB	Guinea-Bissau
GNQ	Equatorial Guinea
GRC	Greece
GRD	Grenada
GRL	Greenland
GTM	Guatemala
GUM	Guam
GUY	Guyana
HIC	High income
HKG	Hong Kong SAR, China
HND	Honduras
HPC	Heavily indebted poor countries (HIPC)
HRV	Croatia
HTI	Haiti
HUN	Hungary
IBB	IBRD, including blend
IBD	IBRD only
IBT	IDA & IBRD total
IDA	IDA total
IDB	IDA blend
IDN	Indonesia
IDX	IDA only
IMN	Isle of Man
IND	India
INX	Not classified
IRL	Ireland
IRN	Iran, Islamic Rep.
IRQ	Iraq
ISL	Iceland
ISR	Israel
ITA	Italy

JAM	Jamaica
JOR	Jordan
JPN	Japan
KAZ	Kazakhstan
KEN	Kenya
KGZ	Kyrgyz Republic
KHM	Cambodia
KIR	Kiribati
KNA	St. Kitts and Nevis
KOR	Korea, Rep.
KWT	Kuwait
LAC	Latin America & Caribbean (excluding high income)
LAO	Lao PDR
LBN	Lebanon
LBR	Liberia
LBY	Libya
LCA	St. Lucia
LCN	Latin America & Caribbean
LDC	Least developed countries: UN classification
LIC	Low income
LIE	Liechtenstein
LKA	Sri Lanka
LMC	Lower middle income
LMY	Low & middle income
LSO	Lesotho
LTE	Late-demographic dividend
LTU	Lithuania
LUX	Luxembourg
LVA	Latvia
MAC	Macao SAR, China
MAF	St. Martin (French part)
MAR	Morocco
MCO	Monaco
MDA	Moldova
MDE	Middle East (developing only)
MDG	Madagascar
MDV	Maldives
MEA	Middle East & North Africa
MEX	Mexico
MHL	Marshall Islands
MIC	Middle income
MKD	North Macedonia
MLI	Mali
MLT	Malta
MMR	Myanmar
MNA	Middle East & North Africa (excluding high income)
MNE	Montenegro

MNG	Mongolia
MNP	Northern Mariana Islands
MOZ	Mozambique
MRT	Mauritania
MUS	Mauritius
MWI	Malawi
MYS	Malaysia
NAC	North America
NAF	North Africa
NAM	Namibia
NCL	New Caledonia
NER	Niger
NGA	Nigeria
NIC	Nicaragua
NLD	Netherlands
NOR	Norway
NPL	Nepal
NRS	Non-resource rich Sub-Saharan Africa countries
NRU	Nauru
NXS	IDA countries not classified as fragile situations, excluding Sub-Saharan Africa
NZL	New Zealand
OED	OECD members
OMN	Oman
OSS	Other small states
PAK	Pakistan
PAN	Panama
PER	Peru
PHL	Philippines
PLW	Palau
PNG	Papua New Guinea
POL	Poland
PRE	Pre-demographic dividend
PRI	Puerto Rico
PRK	Korea, Dem. People's Rep.
PRT	Portugal
PRY	Paraguay
PSE	West Bank and Gaza
PSS	Pacific island small states
PST	Post-demographic dividend
PYF	French Polynesia
QAT	Qatar
ROU	Romania
RRS	Resource rich Sub-Saharan Africa countries
RUS	Russian Federation
RWA	Rwanda
SAS	South Asia

SAU	Saudi Arabia
SDN	Sudan
SEN	Senegal
SGP	Singapore
SLB	Solomon Islands
SLE	Sierra Leone
SLV	El Salvador
SMR	San Marino
SOM	Somalia
SRB	Serbia
SSA	Sub-Saharan Africa (excluding high income)
SSD	South Sudan
SSF	Sub-Saharan Africa
SST	Small states
STP	Sao Tome and Principe
SUR	Suriname
SVK	Slovak Republic
SVN	Slovenia
SWE	Sweden
SWZ	Eswatini
SXM	Sint Maarten (Dutch part)
SXZ	Sub-Saharan Africa excluding South Africa
SYC	Seychelles
SYR	Syrian Arab Republic
TCA	Turks and Caicos Islands
TCD	Chad
TEA	East Asia & Pacific (IDA & IBRD countries)
TEC	Europe & Central Asia (IDA & IBRD countries)
TGO	Togo
THA	Thailand
TJK	Tajikistan
TKM	Turkmenistan
TLA	Latin America & the Caribbean (IDA & IBRD countries)
TLS	Timor-Leste
TMN	Middle East & North Africa (IDA & IBRD countries)
TON	Tonga
TSA	South Asia (IDA & IBRD)
TSS	Sub-Saharan Africa (IDA & IBRD countries)
TTO	Trinidad and Tobago
TUN	Tunisia
TUR	Turkey
TUV	Tuvalu
TWN	Taiwan, China
TZA	Tanzania
UGA	Uganda
UKR	Ukraine
UMC	Upper middle income

URY	Uruguay
USA	United States
UZB	Uzbekistan
VCT	St. Vincent and the Grenadines
VEN	Venezuela, RB
VGB	British Virgin Islands
VIR	Virgin Islands (U.S.)
VNM	Vietnam
VUT	Vanuatu
WLD	World
WSM	Samoa
XXK	Kosovo
XZN	Sub-Saharan Africa excluding South Africa and Nigeria
YEM	Yemen, Rep.
ZAF	South Africa
ZMB	Zambia
ZWE	Zimbabwe

To see possible datasets we can access via the API, use `get_source()`

```
[3]: wldata.get_source()
```

```
[3]:  id  name
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     1  Doing Business
     2  World Development Indicators
     3  Worldwide Governance Indicators
     5  Subnational Malnutrition Database
     6  International Debt Statistics
    11  Africa Development Indicators
    12  Education Statistics
    13  Enterprise Surveys
    14  Gender Statistics
    15  Global Economic Monitor
    16  Health Nutrition and Population Statistics
    18  IDA Results Measurement System
    19  Millennium Development Goals
    20  Quarterly Public Sector Debt
    22  Quarterly External Debt Statistics SDDS
    23  Quarterly External Debt Statistics GDDS
    24  Poverty and Equity
    25  Jobs
    27  Global Economic Prospects
    28  Global Financial Inclusion
    29  The Atlas of Social Protection: Indicators of Resilience and Equity
    30  Exporter Dynamics Database - Indicators at Country-Year Level
    31  Country Policy and Institutional Assessment
```


32 Global Financial Development
 33 G20 Financial Inclusion Indicators
 34 Global Partnership for Education
 35 Sustainable Energy for All
 36 Statistical Capacity Indicators
 37 LAC Equity Lab
 38 Subnational Poverty
 39 Health Nutrition and Population Statistics by Wealth Quintile
 40 Population estimates and projections
 41 Country Partnership Strategy for India (FY2013 - 17)
 43 Adjusted Net Savings
 44 Readiness for Investment in Sustainable Energy
 45 Indonesia Database for Policy and Economic Research
 46 Sustainable Development Goals
 50 Subnational Population
 54 Joint External Debt Hub
 57 WDI Database Archives
 58 Universal Health Coverage
 59 Wealth Accounts
 60 Economic Fitness
 61 PPPs Regulatory Quality
 62 International Comparison Program (ICP) 2011
 63 Human Capital Index
 64 Worldwide Bureaucracy Indicators
 65 Health Equity and Financial Protection Indicators
 66 Logistics Performance Index
 67 PEFA 2011
 68 PEFA 2016
 69 Global Financial Inclusion and Consumer Protection Survey
 70 Economic Fitness 2
 71 International Comparison Program (ICP) 2005
 72 PEFA_Test
 73 Global Financial Inclusion and Consumer Protection Survey (Internal)
 75 Environment, Social and Governance (ESG) Data
 76 Remittance Prices Worldwide (Sending Countries)
 77 Remittance Prices Worldwide (Receiving Countries)
 78 ICP 2017
 79 PEFA_GRPFM
 80 Gender Disaggregated Labor Database (GDLD)
 81 International Debt Statistics: DSSI

```
[4]: SOURCE = 40 # "Population estimates and projections"
```

```

indicators_40 = wbdata.get_indicator(source=SOURCE)
indicators_40

```

[4]: id	name

SH.DTH.0509	Number of deaths ages 5-9 years
SH.DTH.1014	Number of deaths ages 10-14 years
SH.DTH.1519	Number of deaths ages 15-19 years
SH.DTH.2024	Number of deaths ages 20-24 years
SH.DTH.IMRT	Number of infant deaths
SH.DTH.IMRT.FE	Number of infant deaths, female
SH.DTH.IMRT.MA	Number of infant deaths, male
SH.DTH.MORT	Number of under-five deaths
SH.DTH.MORT.FE	Number of under-five deaths, female
SH.DTH.MORT.MA	Number of under-five deaths, male
SH.DTH.NMRT	Number of neonatal deaths
SH.DYN.0509	Probability of dying among children ages 5-9 years (per 1,000)
SH.DYN.1014	Probability of dying among adolescents ages 10-14 years (per 1,000)
SH.DYN.1519	Probability of dying among adolescents ages 15-19 years (per 1,000)
SH.DYN.2024	Probability of dying among youth ages 20-24 years (per 1,000)
SH.DYN.MORT	Mortality rate, under-5 (per 1,000 live births)
SH.DYN.MORT.FE	Mortality rate, under-5, female (per 1,000 live births)
SH.DYN.MORT.MA	Mortality rate, under-5, male (per 1,000 live births)
SH.DYN.NMRT	Mortality rate, neonatal (per 1,000 live births)
SM.POP.NETM	Net migration
SP.DYN.AMRT.FE	Mortality rate, adult, female (per 1,000 female adults)
SP.DYN.AMRT.MA	Mortality rate, adult, male (per 1,000 male adults)
SP.DYN.CBRT.IN	Birth rate, crude (per 1,000 people)
SP.DYN.CDRT.IN	Death rate, crude (per 1,000 people)
SP.DYN.IMRT.FE.IN	Mortality rate, infant, female (per 1,000 live births)
SP.DYN.IMRT.IN	Mortality rate, infant (per 1,000 live births)
SP.DYN.IMRT.MA.IN	Mortality rate, infant, male (per 1,000 live births)
SP.DYN.LE00.FE.IN	Life expectancy at birth, female (years)
SP.DYN.LE00.IN	Life expectancy at birth, total (years)
SP.DYN.LE00.MA.IN	Life expectancy at birth, male (years)
SP.DYN.TFRT.IN	Fertility rate, total (births per woman)
SP.POP.0004.FE	Population ages 00-04, female
SP.POP.0004.FE.5Y	Population ages 00-04, female (% of female population)
SP.POP.0004.MA	Population ages 00-04, male
SP.POP.0004.MA.5Y	Population ages 00-04, male (% of male population)
SP.POP.0014.FE.IN	Population ages 0-14, female
SP.POP.0014.FE.ZS	Population ages 0-14, female (% of female population)
SP.POP.0014.MA.IN	Population ages 0-14, male
SP.POP.0014.MA.ZS	Population ages 0-14, male (% of male population)
SP.POP.0014.TO	Population ages 0-14, total
SP.POP.0014.TO.ZS	Population ages 0-14 (% of total population)

SP.POP.0509.FE	Population ages 05-09, female
SP.POP.0509.FE.5Y	Population ages 05-09, female (% of female population)
SP.POP.0509.MA	Population ages 05-09, male
SP.POP.0509.MA.5Y	Population ages 05-09, male (% of male population)
SP.POP.1014.FE	Population ages 10-14, female
SP.POP.1014.FE.5Y	Population ages 10-14, female (% of female population)
SP.POP.1014.MA	Population ages 10-14, male
SP.POP.1014.MA.5Y	Population ages 10-14, male (% of male population)
SP.POP.1519.FE	Population ages 15-19, female
SP.POP.1519.FE.5Y	Population ages 15-19, female (% of female population)
SP.POP.1519.MA	Population ages 15-19, male
SP.POP.1519.MA.5Y	Population ages 15-19, male (% of male population)
SP.POP.1564.FE.IN	Population ages 15-64, female
SP.POP.1564.FE.ZS	Population ages 15-64, female (% of female population)
SP.POP.1564.MA.IN	Population ages 15-64, male
SP.POP.1564.MA.ZS	Population ages 15-64, male (% of male population)
SP.POP.1564.TO	Population ages 15-64, total
SP.POP.1564.TO.ZS	Population ages 15-64 (% of total population)
SP.POP.2024.FE	Population ages 20-24, female
SP.POP.2024.FE.5Y	Population ages 20-24, female (% of female population)
SP.POP.2024.MA	Population ages 20-24, male
SP.POP.2024.MA.5Y	Population ages 20-24, male (% of male population)
SP.POP.2529.FE	Population ages 25-29, female
SP.POP.2529.FE.5Y	Population ages 25-29, female (% of female population)
SP.POP.2529.MA	Population ages 25-29, male
SP.POP.2529.MA.5Y	Population ages 25-29, male (% of male population)
SP.POP.3034.FE	Population ages 30-34, female
SP.POP.3034.FE.5Y	Population ages 30-34, female (% of female population)
SP.POP.3034.MA	Population ages 30-34, male
SP.POP.3034.MA.5Y	Population ages 30-34, male (% of male population)
SP.POP.3539.FE	Population ages 35-39, female
SP.POP.3539.FE.5Y	Population ages 35-39, female (% of female population)
SP.POP.3539.MA	Population ages 35-39, male
SP.POP.3539.MA.5Y	Population ages 35-39, male (% of male population)
SP.POP.4044.FE	Population ages 40-44, female
SP.POP.4044.FE.5Y	Population ages 40-44, female (% of female population)
SP.POP.4044.MA	Population ages 40-44, male
SP.POP.4044.MA.5Y	Population ages 40-44, male (% of male population)
SP.POP.4549.FE	Population ages 45-49, female
SP.POP.4549.FE.5Y	Population ages 45-49, female (% of female population)
SP.POP.4549.MA	Population ages 45-49, male
SP.POP.4549.MA.5Y	Population ages 45-49, male (% of male population)
SP.POP.5054.FE	Population ages 50-54, female
SP.POP.5054.FE.5Y	Population ages 50-54, female (% of female population)
SP.POP.5054.MA	Population ages 50-54, male
SP.POP.5054.MA.5Y	Population ages 50-54, male (% of male population)
SP.POP.5559.FE	Population ages 55-59, female

SP.POP.5559.FE.5Y	Population ages 55-59, female (% of female population)
SP.POP.5559.MA	Population ages 55-59, male
SP.POP.5559.MA.5Y	Population ages 55-59, male (% of male population)
SP.POP.6064.FE	Population ages 60-64, female
SP.POP.6064.FE.5Y	Population ages 60-64, female (% of female population)
SP.POP.6064.MA	Population ages 60-64, male
SP.POP.6064.MA.5Y	Population ages 60-64, male (% of male population)
SP.POP.6569.FE	Population ages 65-69, female
SP.POP.6569.FE.5Y	Population ages 65-69, female (% of female population)
SP.POP.6569.MA	Population ages 65-69, male
SP.POP.6569.MA.5Y	Population ages 65-69, male (% of male population)
SP.POP.65UP.FE.IN	Population ages 65 and above, female
SP.POP.65UP.FE.ZS	Population ages 65 and above, female (% of female population)
SP.POP.65UP.MA.IN	Population ages 65 and above, male
SP.POP.65UP.MA.ZS	Population ages 65 and above, male (% of male population)
SP.POP.65UP.TO	Population ages 65 and above, total
SP.POP.65UP.TO.ZS	Population ages 65 and above (% of total population)
SP.POP.7074.FE	Population ages 70-74, female
SP.POP.7074.FE.5Y	Population ages 70-74, female (% of female population)
SP.POP.7074.MA	Population ages 70-74, male
SP.POP.7074.MA.5Y	Population ages 70-74, male (% of male population)
SP.POP.7579.FE	Population ages 75-79, female
SP.POP.7579.FE.5Y	Population ages 75-79, female (% of female population)
SP.POP.7579.MA	Population ages 75-79, male
SP.POP.7579.MA.5Y	Population ages 75-79, male (% of male population)
SP.POP.80UP.FE	Population ages 80 and above, female
SP.POP.80UP.FE.5Y	Population ages 80 and above, female (% of female population)
SP.POP.80UP.MA	Population ages 80 and above, male
SP.POP.80UP.MA.5Y	Population ages 80 and above, male (% of male population)
SP.POP.AG00.FE.IN	Age population, age 00, female, interpolated
SP.POP.AG00.MA.IN	Age population, age 00, male, interpolated
SP.POP.AG01.FE.IN	Age population, age 01, female, interpolated
SP.POP.AG01.MA.IN	Age population, age 01, male, interpolated
SP.POP.AG02.FE.IN	Age population, age 02, female, interpolated
SP.POP.AG02.MA.IN	Age population, age 02, male, interpolated
SP.POP.AG03.FE.IN	Age population, age 03, female, interpolated
SP.POP.AG03.MA.IN	Age population, age 03, male, interpolated
SP.POP.AG04.FE.IN	Age population, age 04, female, interpolated
SP.POP.AG04.MA.IN	Age population, age 04, male, interpolated
SP.POP.AG05.FE.IN	Age population, age 05, female, interpolated
SP.POP.AG05.MA.IN	Age population, age 05, male, interpolated
SP.POP.AG06.FE.IN	Age population, age 06, female, interpolated
SP.POP.AG06.MA.IN	Age population, age 06, male, interpolated
SP.POP.AG07.FE.IN	Age population, age 07, female, interpolated
SP.POP.AG07.MA.IN	Age population, age 07, male, interpolated
SP.POP.AG08.FE.IN	Age population, age 08, female, interpolated
SP.POP.AG08.MA.IN	Age population, age 08, male, interpolated

SP.POP.AG09.FE.IN	Age population, age 09, female, interpolated
SP.POP.AG09.MA.IN	Age population, age 09, male, interpolated
SP.POP.AG10.FE.IN	Age population, age 10, female, interpolated
SP.POP.AG10.MA.IN	Age population, age 10, male, interpolated
SP.POP.AG11.FE.IN	Age population, age 11, female, interpolated
SP.POP.AG11.MA.IN	Age population, age 11, male, interpolated
SP.POP.AG12.FE.IN	Age population, age 12, female, interpolated
SP.POP.AG12.MA.IN	Age population, age 12, male, interpolated
SP.POP.AG13.FE.IN	Age population, age 13, female, interpolated
SP.POP.AG13.MA.IN	Age population, age 13, male, interpolated
SP.POP.AG14.FE.IN	Age population, age 14, female, interpolated
SP.POP.AG14.MA.IN	Age population, age 14, male, interpolated
SP.POP.AG15.FE.IN	Age population, age 15, female, interpolated
SP.POP.AG15.MA.IN	Age population, age 15, male, interpolated
SP.POP.AG16.FE.IN	Age population, age 16, female, interpolated
SP.POP.AG16.MA.IN	Age population, age 16, male, interpolated
SP.POP.AG17.FE.IN	Age population, age 17, female, interpolated
SP.POP.AG17.MA.IN	Age population, age 17, male, interpolated
SP.POP.AG18.FE.IN	Age population, age 18, female, interpolated
SP.POP.AG18.MA.IN	Age population, age 18, male, interpolated
SP.POP.AG19.FE.IN	Age population, age 19, female, interpolated
SP.POP.AG19.MA.IN	Age population, age 19, male, interpolated
SP.POP.AG20.FE.IN	Age population, age 20, female, interpolated
SP.POP.AG20.MA.IN	Age population, age 20, male, interpolated
SP.POP.AG21.FE.IN	Age population, age 21, female, interpolated
SP.POP.AG21.MA.IN	Age population, age 21, male, interpolated
SP.POP.AG22.FE.IN	Age population, age 22, female, interpolated
SP.POP.AG22.MA.IN	Age population, age 22, male, interpolated
SP.POP.AG23.FE.IN	Age population, age 23, female, interpolated
SP.POP.AG23.MA.IN	Age population, age 23, male, interpolated
SP.POP.AG24.FE.IN	Age population, age 24, female, interpolated
SP.POP.AG24.MA.IN	Age population, age 24, male, interpolated
SP.POP.AG25.FE.IN	Age population, age 25, female, interpolated
SP.POP.AG25.MA.IN	Age population, age 25, male, interpolated
SP.POP.BRTH.MF	Sex ratio at birth (male births per female births)
SP.POP.DPND	Age dependency ratio (% of working-age population)
SP.POP.DPND.OL	Age dependency ratio, old (% of working-age population)
SP.POP.DPND.YG	Age dependency ratio, young (% of working-age population)
SP.POP.GROW	Population growth (annual %)
SP.POP.TOTL	Population, total
SP.POP.TOTL.FE.IN	Population, female
SP.POP.TOTL.FE.ZS	Population, female (% of total population)
SP.POP.TOTL.MA.IN	Population, male
SP.POP.TOTL.MA.ZS	Population, male (% of total population)
SP.RUR.TOTL	Rural population
SP.RUR.TOTL.ZG	Rural population growth (annual %)
SP.RUR.TOTL.ZS	Rural population (% of total population)

```

SP.URB.GROW      Urban population growth (annual %)
SP.URB.TOTL      Urban population
SP.URB.TOTL.IN.ZS Urban population (% of total population)

```

0.2 Population function

```

[5]: def population(sex, year, age, country):
      """Function that takes in a SEX ("Male", "Female"), ... , COUNTRY=, ..
      and returns a statistic for the given function arguments"""
      if sex == "Male":
          variable_labels = {"SP.POP." + str(age[0])+str(age[1]) + ".MA": sex}
      elif sex=="Female":
          variable_labels = {"SP.POP." + str(age[0])+str(age[1]) + ".FE": sex}
      pop_stats = wbdata.get_dataframe(variable_labels, country=country)
      pop_stats = pop_stats.filter(like=str(year), axis=0)
      return pop_stats[sex][0]

```

```

[6]: population("Female", 2010, (15,19), "CHN")

```

```

[6]: 45907253.0

```

```

[7]: def population_dataframes(indicators):
      """Returns a pandas DataFrame indexed by Region or Country and Year,
      with columns giving counts of people in different age-sex groups."""
      # Create a dictionary of all corresponding ID's and keys in SOURCE 40 from
      ↪WBDATA
      labels = {}
      for i in range(len(indicators)):
          col_id = indicators[i]['id']
          col_name = indicators[i]['name']
          labels[col_id] = col_name

      def find_labels(indicators):
          """Helper function that akes in an indicator object to filter through
          ↪variable
          keys and parse through necessary ID's to obtain relevant data. """
          # Filter through column ID strings to obtain relevant population data.
          r = re.compile("(SP.POP).[\d]{2}[A-Z0-9]{2}.[MAFE]{2}$")
          col_keys = list(filter(r.match, labels))

          # Add total population column at the end of COL_KEYS
          col_keys.append('SP.POP.TOTL')

          labels_filtered = {}
          for key, value in labels.items():
              if key in col_keys:

```

```

        labels_filtered[key] = value
    return labels_filtered

df_labels = find_labels(indicators)
return wbdata.get_dataframe(df_labels)

```

0.3 Population DataFrames

```

[8]: pop_df = population_dataframes(indicators_40)
     pop_df.head()

```

```

[8]:
      Population ages 00-04, female  Population ages 00-04, male  \
country    date
Afghanistan 1960                760938.0                780471.0
            1961                795378.0                808289.0
            1962                818678.0                831163.0
            1963                834934.0                851787.0
            1964                850992.0                873022.0

      Population ages 05-09, female  Population ages 05-09, male  \
country    date
Afghanistan 1960                583953.0                598721.0
            1961                594819.0                617341.0
            1962                611717.0                637537.0
            1963                632901.0                657620.0
            1964                654635.0                676153.0

      Population ages 10-14, female  Population ages 10-14, male  \
country    date
Afghanistan 1960                544194.0                523122.0
            1961                547964.0                528908.0
            1962                550019.0                538026.0
            1963                551891.0                550452.0
            1964                556047.0                565257.0

      Population ages 15-19, female  Population ages 15-19, male  \
country    date
Afghanistan 1960                447872.0                478075.0
            1961                460206.0                481616.0
            1962                477029.0                485476.0
            1963                495604.0                490025.0
            1964                511923.0                496099.0

      Population ages 20-24, female  Population ages 20-24, male  \
country    date
Afghanistan 1960                382542.0                415970.0
            1961                388240.0                421947.0

```

	1962		394896.0		429297.0
	1963		402961.0		437345.0
	1964		413203.0		445246.0
		... Population ages 60-64, male \			
country	date	...			
Afghanistan	1960	...	93053.0		
	1961	...	93359.0		
	1962	...	94140.0		
	1963	...	95186.0		
	1964	...	96187.0		
		Population ages 65-69, female	Population ages 65-69, male \		
country	date				
Afghanistan	1960	56624.0	67283.0		
	1961	57393.0	67330.0		
	1962	58326.0	67197.0		
	1963	59399.0	67052.0		
	1964	60569.0	67117.0		
		Population ages 70-74, female	Population ages 70-74, male \		
country	date				
Afghanistan	1960	34655.0	40748.0		
	1961	35363.0	41520.0		
	1962	36055.0	42186.0		
	1963	36757.0	42704.0		
	1964	37461.0	42987.0		
		Population ages 75-79, female	Population ages 75-79, male \		
country	date				
Afghanistan	1960	16990.0	19683.0		
	1961	17750.0	20500.0		
	1962	18364.0	21050.0		
	1963	18819.0	21331.0		
	1964	19098.0	21378.0		
		Population ages 80 and above, female \			
country	date				
Afghanistan	1960	7486.0			
	1961	8358.0			
	1962	9041.0			
	1963	9456.0			
	1964	9513.0			
		Population ages 80 and above, male	Population, total		
country	date				
Afghanistan	1960	8294.0	8996973.0		

1961	9283.0	9169410.0
1962	10013.0	9351441.0
1963	10376.0	9543205.0
1964	10269.0	9744781.0

[5 rows x 35 columns]

0.3.1 Cleaning the data

```
[9]: # Change "date" index to type INT.
pop_df.index = pop_df.index.set_levels(pop_df.index.levels[1].astype(int),
↳ level=1)

# Rename MultiIndex column "date" to "year"
pop_df = pop_df.rename_axis(index=['country', 'year'])

# Delete 2020 row for every country code
pop_df = pop_df[~pop_df.index.get_level_values('year').isin([2020])]

# Add population total breakdowns for male and female
cols_f = [col for col in pop_df.columns if 'female' in col]
cols_m = [col for col in pop_df.columns if col not in cols_f]
pop_df['Population, total female'] = pop_df.loc[:, cols_f].sum(axis=1)
pop_df['Population, total male'] = pop_df.loc[:, cols_m].sum(axis=1)
pop_df.head()
```

```
[9]: Population ages 00-04, female  Population ages 00-04, male  \
country    year
Afghanistan 1960          760938.0          780471.0
            1961          795378.0          808289.0
            1962          818678.0          831163.0
            1963          834934.0          851787.0
            1964          850992.0          873022.0
```

```
Population ages 05-09, female  Population ages 05-09, male  \
country    year
Afghanistan 1960          583953.0          598721.0
            1961          594819.0          617341.0
            1962          611717.0          637537.0
            1963          632901.0          657620.0
            1964          654635.0          676153.0
```

```
Population ages 10-14, female  Population ages 10-14, male  \
country    year
Afghanistan 1960          544194.0          523122.0
            1961          547964.0          528908.0
            1962          550019.0          538026.0
```

	1963	551891.0	550452.0
	1964	556047.0	565257.0
		Population ages 15-19, female	Population ages 15-19, male \
country	year		
Afghanistan	1960	447872.0	478075.0
	1961	460206.0	481616.0
	1962	477029.0	485476.0
	1963	495604.0	490025.0
	1964	511923.0	496099.0
		Population ages 20-24, female	Population ages 20-24, male \
country	year		
Afghanistan	1960	382542.0	415970.0
	1961	388240.0	421947.0
	1962	394896.0	429297.0
	1963	402961.0	437345.0
	1964	413203.0	445246.0
		... Population ages 65-69, male \	
country	year	...	
Afghanistan	1960	67283.0	
	1961	67330.0	
	1962	67197.0	
	1963	67052.0	
	1964	67117.0	
		Population ages 70-74, female	Population ages 70-74, male \
country	year		
Afghanistan	1960	34655.0	40748.0
	1961	35363.0	41520.0
	1962	36055.0	42186.0
	1963	36757.0	42704.0
	1964	37461.0	42987.0
		Population ages 75-79, female	Population ages 75-79, male \
country	year		
Afghanistan	1960	16990.0	19683.0
	1961	17750.0	20500.0
	1962	18364.0	21050.0
	1963	18819.0	21331.0
	1964	19098.0	21378.0
		Population ages 80 and above, female \	
country	year		
Afghanistan	1960	7486.0	
	1961	8358.0	

	1962	9041.0
	1963	9456.0
	1964	9513.0

country	year	Population ages 80 and above, male	Population, total \
Afghanistan	1960	8294.0	8996973.0
	1961	9283.0	9169410.0
	1962	10013.0	9351441.0
	1963	10376.0	9543205.0
	1964	10269.0	9744781.0

country	year	Population, total female	Population, total male
Afghanistan	1960	4347395.0	13646547.0
	1961	4439156.0	13899660.0
	1962	4535392.0	14167491.0
	1963	4636170.0	14450236.0
	1964	4741529.0	14748029.0

[5 rows x 37 columns]

0.4 Comparing Population Growth Rates of China and India

```
[10]: # Rename DATE to YEAR
def growth_df(region):
    df=wldata.get_dataframe({'SP.POP.GROW': 'Total Population growth (annual_
    ↳%)',
                            'SP.RUR.TOTL.ZG': 'Rural population growth (annual_
    ↳%)',
                            'SP.URB.GROW': 'Urban population growth (annual_
    ↳%)'}, country=region)
    df.index = df.index.rename('year')
    df.index = df.index.astype(int)
    df = df.drop([2020])
    return df
```

```
[11]: chn_gr = growth_df('CHN')
      chn_gr.head()
```

```
[11]:      Total Population growth (annual %)  Rural population growth (annual %) \
year
2019      0.357291      -2.513529
2018      0.455900      -2.420469
2017      0.559121      -2.310812
2016      0.541479      -2.275352
```

2015	0.508137	-2.242449
------	----------	-----------

Urban population growth (annual %)

year	
2019	2.292727
2018	2.491628
2017	2.693540
2016	2.744069
2015	2.769551

```
[12]: ind_gr = growth_df('IND')
      ind_gr.head()
```

```
[12]: Total Population growth (annual %) Rural population growth (annual %) \
      year
      2019      1.015106      0.342850
      2018      1.037323      0.387627
      2017      1.062597      0.435052
      2016      1.089880      0.485585
      2015      1.116633      0.533714
```

Urban population growth (annual %)

year	
2019	2.305597
2018	2.308965
2017	2.314448
2016	2.317931
2015	2.322891

```
[13]: wld_gr = growth_df('WLD')
      wld_gr.head()
```

```
[13]: Total Population growth (annual %) Rural population growth (annual %) \
      year
      2019      1.074675      0.072327
      2018      1.103609      0.102808
      2017      1.142040      0.138527
      2016      1.162578      0.162037
      2015      1.168120      0.176929
```

Urban population growth (annual %)

year	
2019	1.887461
2018	1.929861
2017	1.984537
2016	2.018914
2015	2.032497

```
[14]: def urb_rur_plot(df, region_name, color):
    fig = plt.figure(figsize=(11.75, 8.25))
    ax = plt.axes()

    plt.plot(df.index, df.iloc[:, 0], color=color)
    plt.plot(df.index, df.iloc[:, 1], '--', color=color)
    plt.plot(df.index, df.iloc[:, 2], '-.', color=color, linewidth=1)

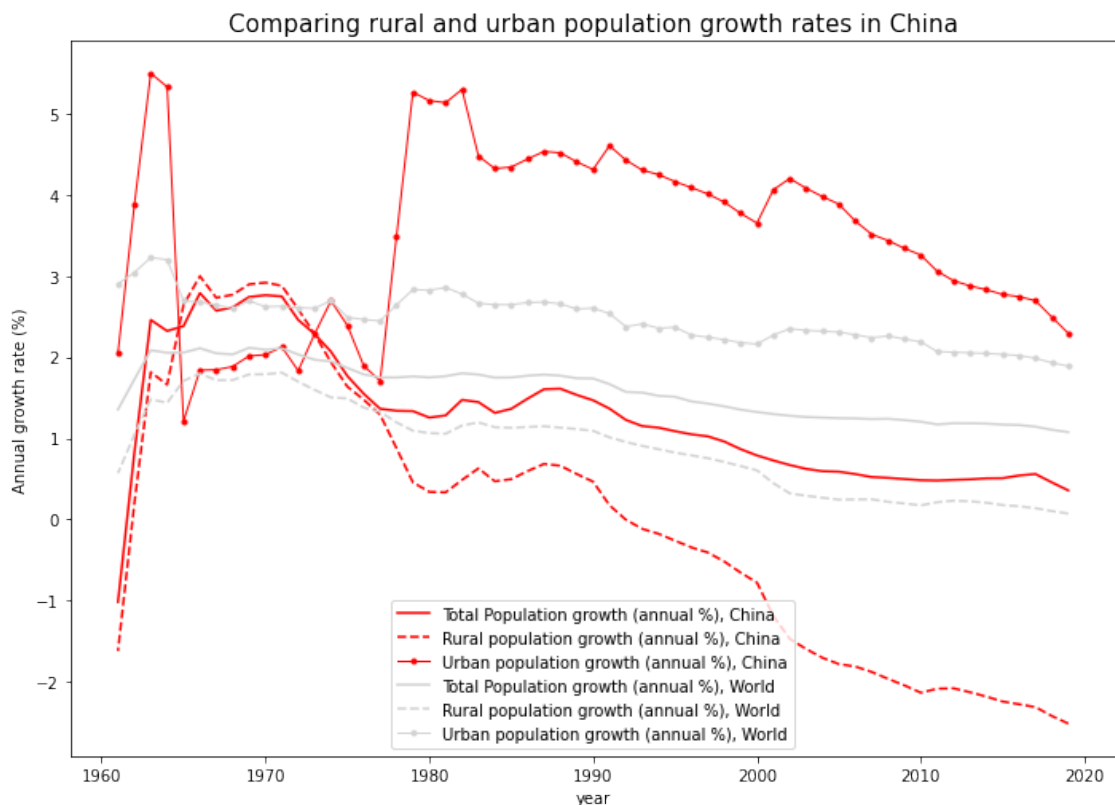
    plt.plot(wld_gr.index, wld_gr.iloc[:, 0], color='lightgray')
    plt.plot(wld_gr.index, wld_gr.iloc[:, 1], '--', color='lightgray')
    plt.plot(wld_gr.index, wld_gr.iloc[:, 2], '-.', color='lightgray',
    ↪linewidth=1);

    ax.legend((df.columns+", {}".format(region_name)).append(wld_gr.columns+", 
    ↪World"))

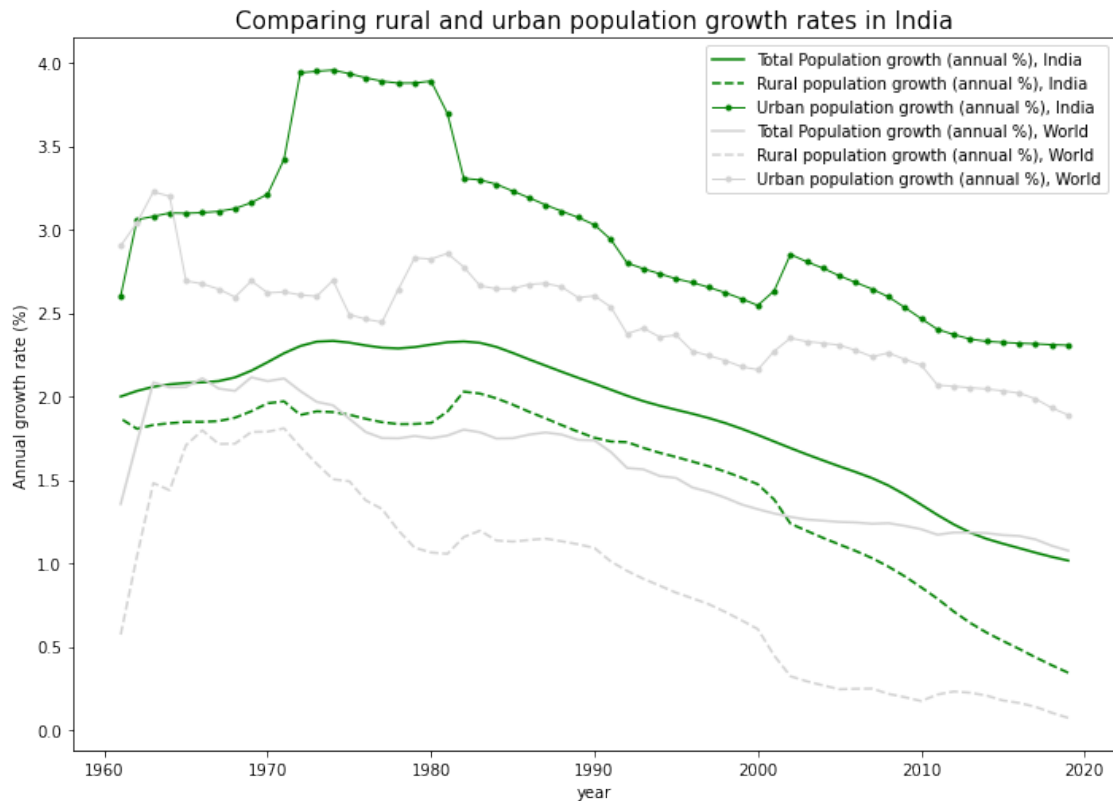
    plt.title('Comparing rural and urban population growth rates in {}'.
    ↪format(region_name), {'fontsize': 15})
    plt.xlabel('year')
    plt.ylabel('Annual growth rate (%)')

    plt.show();
```

```
[15]: urb_rur_plot(chn_gr, "China", 'red')
```



```
[16]: urb_rur_plot(ind_gr, "India", 'green')
```



0.5 Population Pyramids

For the sake of simplicity, visualizations will use world data only.

```
[17]: def generate_ppy_df(pop_df, country, year):
    """Takes in a population DataFrame and selects population
    data for a specific country/region in YEAR, then modifies the
    DataFrame in a way that can be visualized. """
    new_df = pop_df[pop_df.index.get_level_values('country')
    ↳isin([country])]
    country_df = new_df[new_df.index.get_level_values('year').isin([year])]
    ages = [col[16:21] for col in country_df.columns[:2]]
    ages = ages[:-3]
    ages.append('80+')
    f_pop = list(country_df.iloc[0])[:2][-2]
    m_pop = list(country_df.iloc[0])[1:2][-1]
    ppy_df = pd.DataFrame({'Age': ages,
```

```

        'Population, total female': f_pop,
        'Population, total male': m_pop})
    # Convert male population data to negative values for displaying
    → purposes.
    ppy_df['Population, total male'] = ppy_df['Population, total male']* -1
    return ppy_df

def population_pyramid(region, year):
    """Takes as input a pandas DataFrame with columns providing counts
    of people by age-sex groups, and constructs a population pyramid graph
    for visualizing the data."""
    df=generate_ppy_df(pop_df, region, year)
    fig = plt.figure(figsize=(11.75,8.25))
    ax = plt.axes()

    f_patch = mpatches.Patch(color='tomato', label='Female Population')
    m_patch = mpatches.Patch(color='dodgerblue', label='Male Population')
    plt.legend(handles=[m_patch, f_patch],
               loc='upper right')

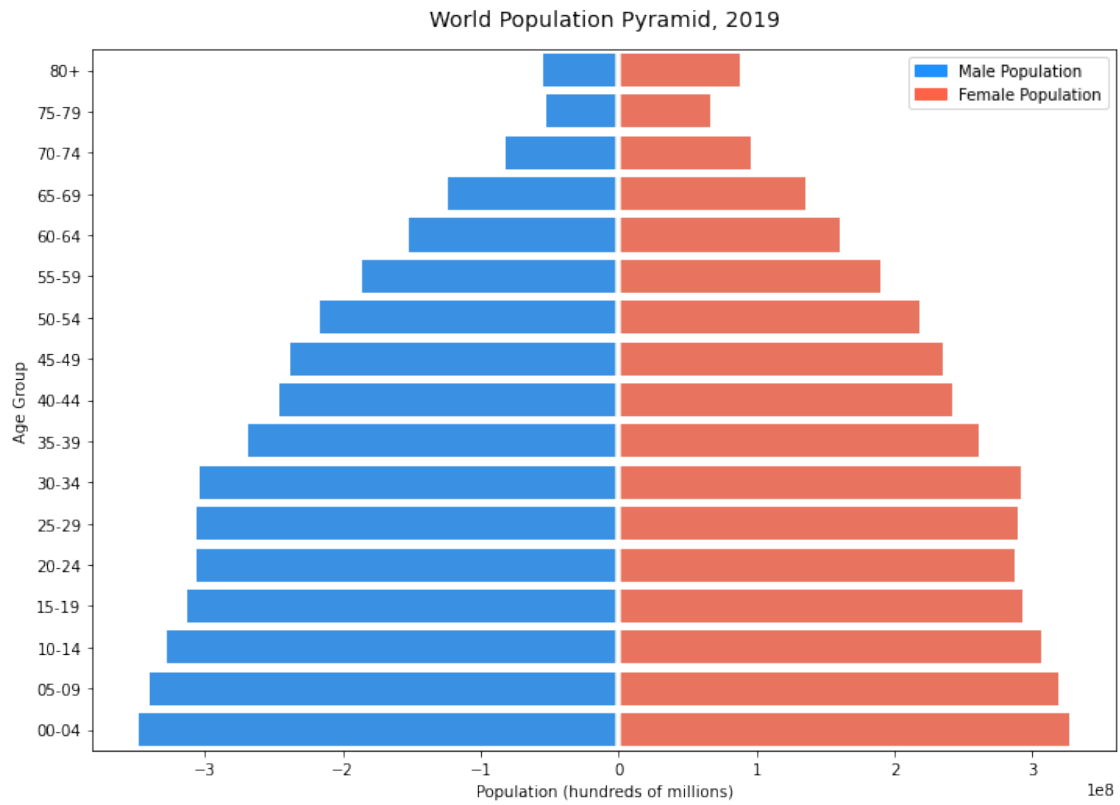
    ageAxis = df['Age'][:, :-1]
    fig = sns.barplot(x='Population, total female',
                     y='Age',
                     data=df,
                     order=ageAxis,
                     lw=0,
                     color='tomato')
    fig = sns.barplot(x='Population, total male',
                     y='Age',
                     data=df,
                     order=ageAxis,
                     lw=0,
                     color='dodgerblue')

    fig.axvline(lw=3.5, color='w')
    fig.set_title("{} Population Pyramid, {}".format(region, year),
    → fontdict={'fontsize':14, 'fontweight':3}, pad=15)
    fig.set(xlabel="Population (hundreds of millions)", ylabel="Age Group");

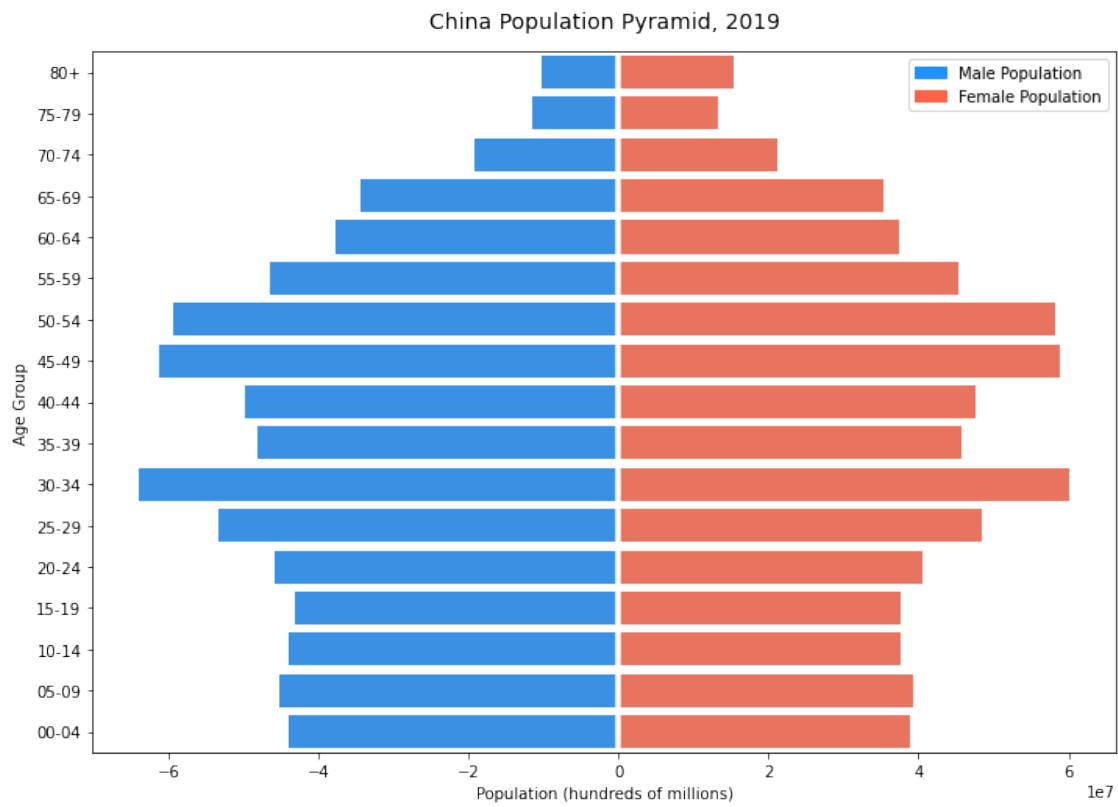
```

Reorganized DataFrame to make suitable for graphing.

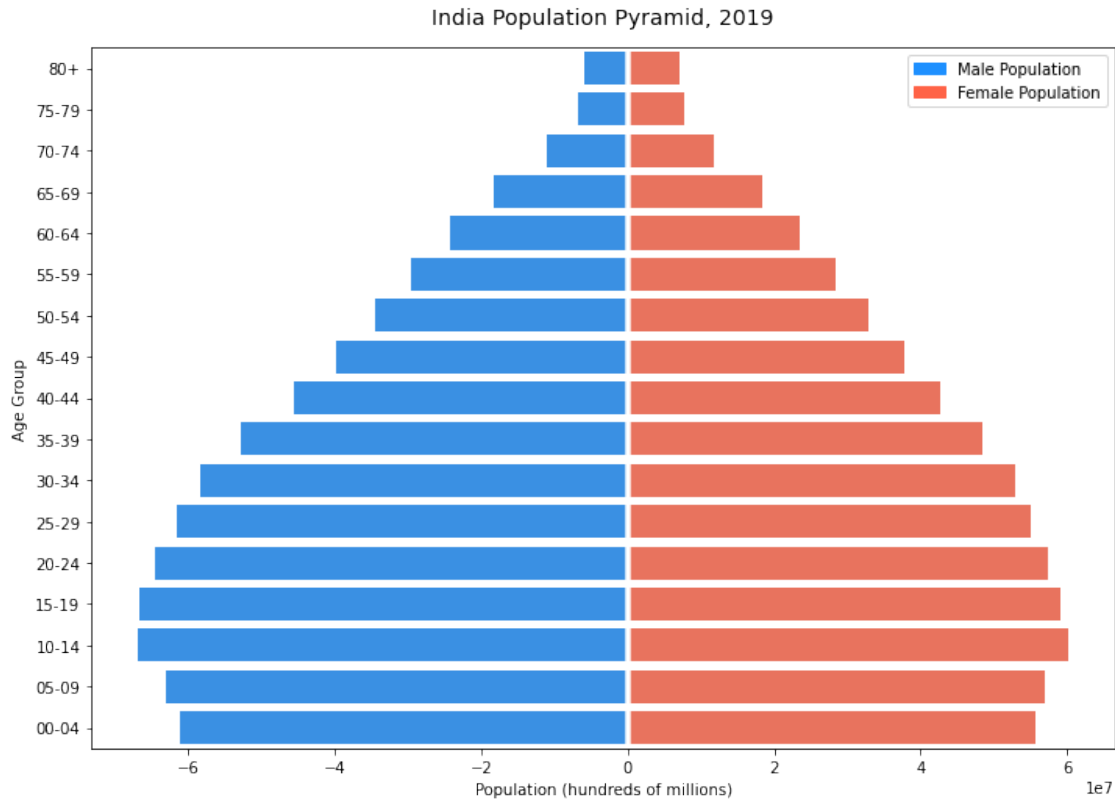
```
[18]: population_pyramid("World", 2019)
```



```
[19]: population_pyramid("China", 2019)
```

```
[20]: population_pyramid("India", 2019)
```



0.6 Animated Population Pyramid

```
[27]: %matplotlib notebook
from matplotlib import animation, rc
from IPython.display import HTML

Writer = animation.writers['ffmpeg']
writer = Writer(fps=60, metadata=dict(artist='Me'), bitrate=1800)

fig2 = plt.figure()
ax2 = plt.axes()
f_patch = mpatches.Patch(color='tomato', label='Female Population')
m_patch = mpatches.Patch(color='dodgerblue', label='Male Population')
plt.legend(handles=[m_patch, f_patch])

region = "World"          # Replace with desired region

def animate(i):
    df=generate_ppy_df(pop_df, region, 1960+i)
    ageAxis=df['Age'][:, :-1]
```

```

    p = sns.barplot(x='Population, total female', y='Age', data=df,
↳order=ageAxis, lw=0, color='tomato')
    p = sns.barplot(x='Population, total male', y='Age', data=df,
↳order=ageAxis, lw=0, color='dodgerblue')
    p.axvline(lw=3.5, color='w')
    p.set_title("{r} Population Pyramid, {y}".format(r=region, y=1960+i),
↳fontdict={'fontsize':14,'fontweight':3}, pad=15)
    plt.xlabel('Population (hundreds of millions)')
    plt.ylabel('Age Group')

anim=animation.
↳FuncAnimation(fig2,animate,blit=False,frames=60,interval=100,repeat=True)

```

<IPython.core.display.Javascript object>

<IPython.core.display.HTML object>

```

[22]: # anim.save('china_ppy.gif', writer=writer)      # Replace with appropriate
↳graph title

```

```

[ ]:

```