

Modeling and Simulation in Python

Chapter 6

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```
# Configure Jupyter so figures appear in the notebook
%matplotlib inline

# Configure Jupyter to display the assigned value after an assignment
%config InteractiveShell.ast_node_interactivity='last_expr_or_assign'

# import functions from the modsim.py module
from modsim import *

from pandas import read_html
```

Code from the previous chapter

```
filename = 'data/World_population_estimates.html'
tables = read_html(filename, header=0, index_col=0, decimal='M')
table2 = tables[2]
table2.columns = ['census', 'prb', 'un', 'maddison',
                  'hyde', 'tanton', 'biraben', 'mj',
                  'thomlinson', 'durand', 'clark']
```

```
un = table2.un / 1e9
un.head()
```

```
Year
1950    2.525149
1951    2.572851
1952    2.619292
1953    2.665865
1954    2.713172
Name: un, dtype: float64
```

```
census = table2.census / 1e9
census.head()
```

```
Year
1950    2.557629
1951    2.594940
1952    2.636772
1953    2.682053
1954    2.730228
Name: census, dtype: float64
```

```

t_0 = get_first_label(census)
t_end = get_last_label(census)
elapsed_time = t_end - t_0

p_0 = get_first_value(census)
p_end = get_last_value(census)
total_growth = p_end - p_0

annual_growth = total_growth / elapsed_time

```

0.07224800083333333

System objects

We can rewrite the code from the previous chapter using system objects.

```

system = System(t_0=t_0,
                t_end=t_end,
                p_0=p_0,
                annual_growth=annual_growth)

```

values

t_0

1950.000000

t_end

2016.000000

p_0

2.557629

annual_growth

0.072248

And we can encapsulate the code that runs the model in a function.

```

def run_simulation1(system):
    """Runs the constant growth model.

    system: System object

    returns: TimeSeries
    """
    results = TimeSeries()
    results[system.t_0] = system.p_0

    for t in linrange(system.t_0, system.t_end):
        results[t+1] = results[t] + system.annual_growth

    return results

```

We can also encapsulate the code that plots the results.

```
def plot_results(census, un, timeseries, title):
    """Plot the estimates and the model.

    census: TimeSeries of population estimates
    un: TimeSeries of population estimates
    timeseries: TimeSeries of simulation results
    title: string
    """
    plot(census, ':', label='US Census')
    plot(un, '--', label='UN DESA')
    plot(timeseries, color='gray', label='model')

    decorate(xlabel='Year',
             ylabel='World population (billion)',
             title=title)
```

Here's how we run it.

```
results = run_simulation1(system)
plot_results(census, un, results, 'Constant growth model')
```

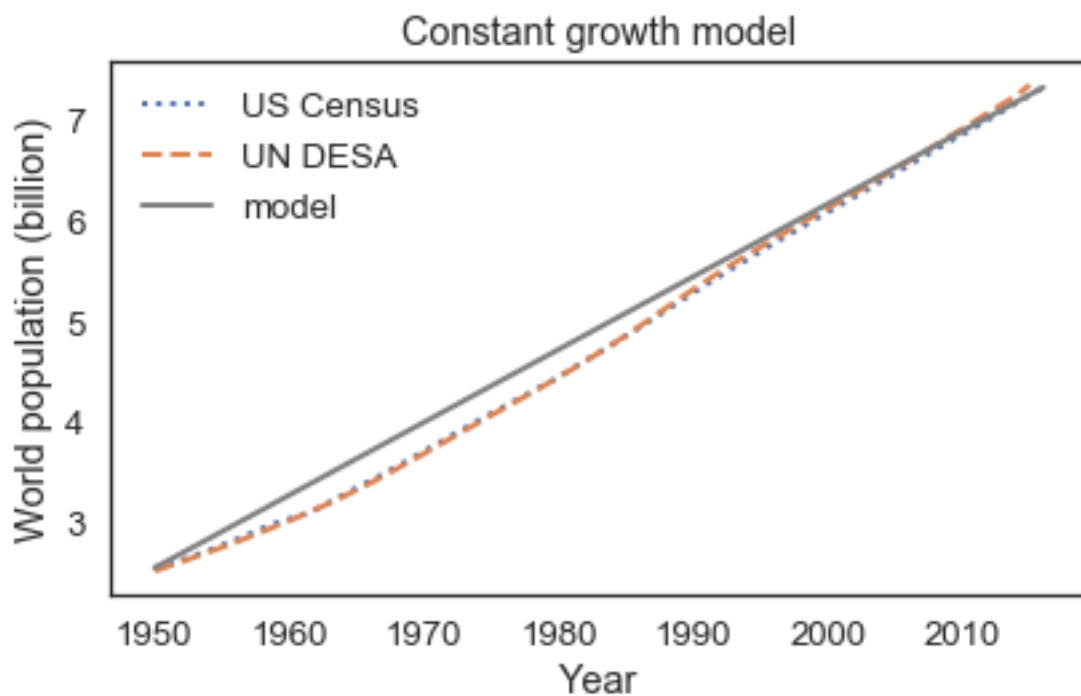


Figure 1: png

Proportional growth

Here's a more realistic model where the number of births and deaths is proportional to the current population.

```
def run_simulation2(system):
    """Run a model with proportional birth and death.

    system: System object

    returns: TimeSeries
    """
    results = TimeSeries()
    results[system.t_0] = system.p_0

    for t in linrange(system.t_0, system.t_end):
        births = system.birth_rate * results[t]
        deaths = system.death_rate * results[t]
        results[t+1] = results[t] + births - deaths

    return results
```

I picked a death rate that seemed reasonable and then adjusted the birth rate to fit the data.

```
system.death_rate = 0.01
system.birth_rate = 0.027
```

Here's what it looks like.

```
results = run_simulation2(system)
plot_results(census, un, results, 'Proportional model')
savefig('figs/chap06-fig01.pdf')
```

Saving figure to file figs/chap06-fig01.pdf

The model fits the data pretty well for the first 20 years, but not so well after that.

Factoring out the update function

run_simulation1 and run_simulation2 are nearly identical except the body of the loop. So we can factor that part out into a function.

```
def update_func1(pop, t, system):
    """Compute the population next year.

    pop: current population
    t: current year
    system: system object containing parameters of the model

    returns: population next year
    """
    births = system.birth_rate * pop
    deaths = system.death_rate * pop
    return pop + births - deaths
```

The name update_func refers to a function object.

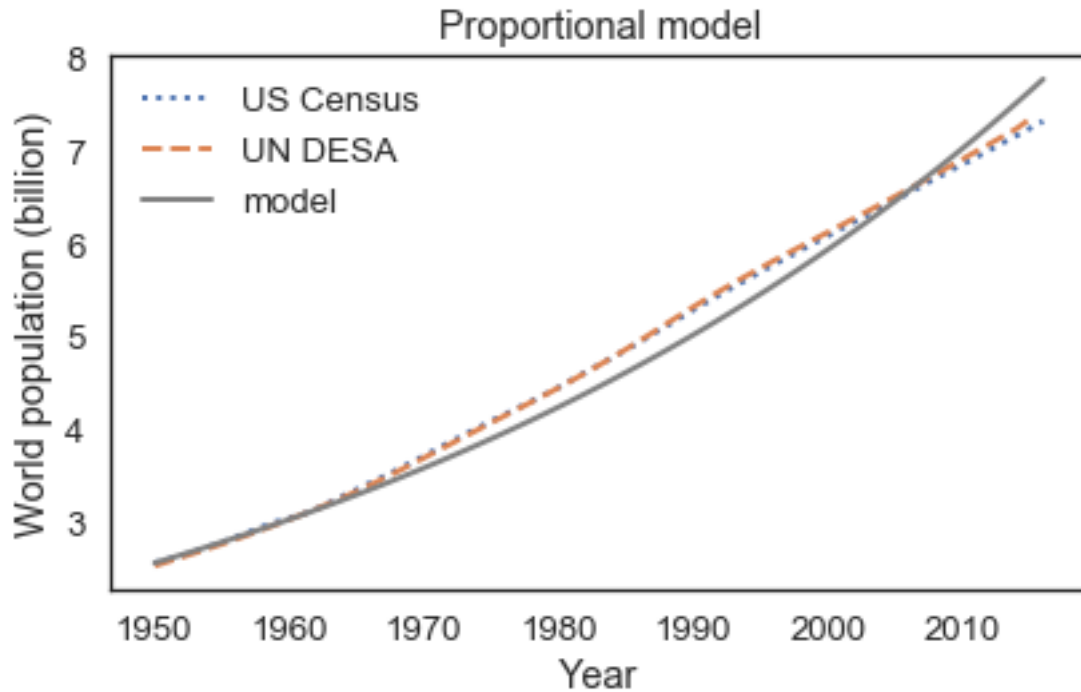


Figure 2: png

```
update_func1
```

```
<function __main__.update_func1(pop, t, system)>
```

Which we can confirm by checking its type.

```
type(update_func1)
```

```
function
```

`run_simulation` takes the update function as a parameter and calls it just like any other function.

```
def run_simulation(system, update_func):
    """Simulate the system using any update function.

    system: System object
    update_func: function that computes the population next year

    returns: TimeSeries
    """
    results = TimeSeries()
    results[system.t_0] = system.p_0

    for t in linrange(system.t_0, system.t_end):
        results[t+1] = update_func(results[t], t, system)

    return results
```

Here's how we use it.

```
t_0 = get_first_label(census)
t_end = get_last_label(census)
p_0 = census[t_0]

system = System(t_0=t_0,
                t_end=t_end,
                p_0=p_0,
                birth_rate=0.027,
                death_rate=0.01)
```

values

t_0

1950.000000

t_end

2016.000000

p_0

2.557629

birth_rate

0.027000

death_rate

0.010000

```
results = run_simulation(system, update_func1)
plot_results(census, un, results, 'Proportional model, factored')
```

Remember not to put parentheses after `update_func1`. What happens if you try?

Exercise: When you run `run_simulation`, it runs `update_func1` once for each year between `t_0` and `t_end`. To see that for yourself, add a print statement at the beginning of `update_func1` that prints the values of `t` and `pop`, then run `run_simulation` again.

Combining birth and death

Since births and deaths get added up, we don't have to compute them separately. We can combine the birth and death rates into a single net growth rate.

```
def update_func2(pop, t, system):
    """Compute the population next year.

    pop: current population
    t: current year
    system: system object containing parameters of the model

    returns: population next year
    """
    net_growth = system.alpha * pop
    return pop + net_growth
```

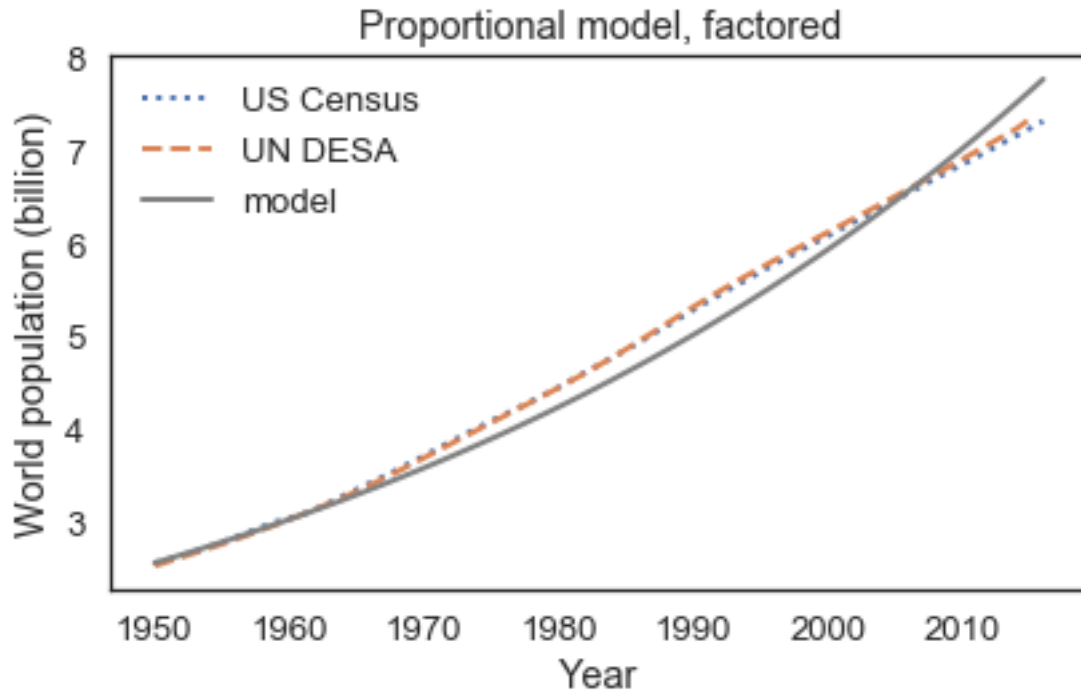


Figure 3: png

Here's how it works:

```
system.alpha = system.birth_rate - system.death_rate

results = run_simulation(system, update_func2)
plot_results(census, un, results, 'Proportional model, combined birth and death')
```

Exercises

Exercise: Maybe the reason the proportional model doesn't work very well is that the growth rate, α , is changing over time. So let's try a model with different growth rates before and after 1980 (as an arbitrary choice).

Write an update function that takes `pop`, `t`, and `system` as parameters. The system object, `system`, should contain two parameters: the growth rate before 1980, `alpha1`, and the growth rate after 1980, `alpha2`. It should use `t` to determine which growth rate to use. Note: Don't forget the `return` statement.

Test your function by calling it directly, then pass it to `run_simulation`. Plot the results. Adjust the parameters `alpha1` and `alpha2` to fit the data as well as you can.

```
t_0 = get_first_label(census)
t_end = get_last_label(census)
p_0 = census[t_0]
alpha1 = 0.027 - 0.01
alpha2 = 0.027 - 0.0075
```

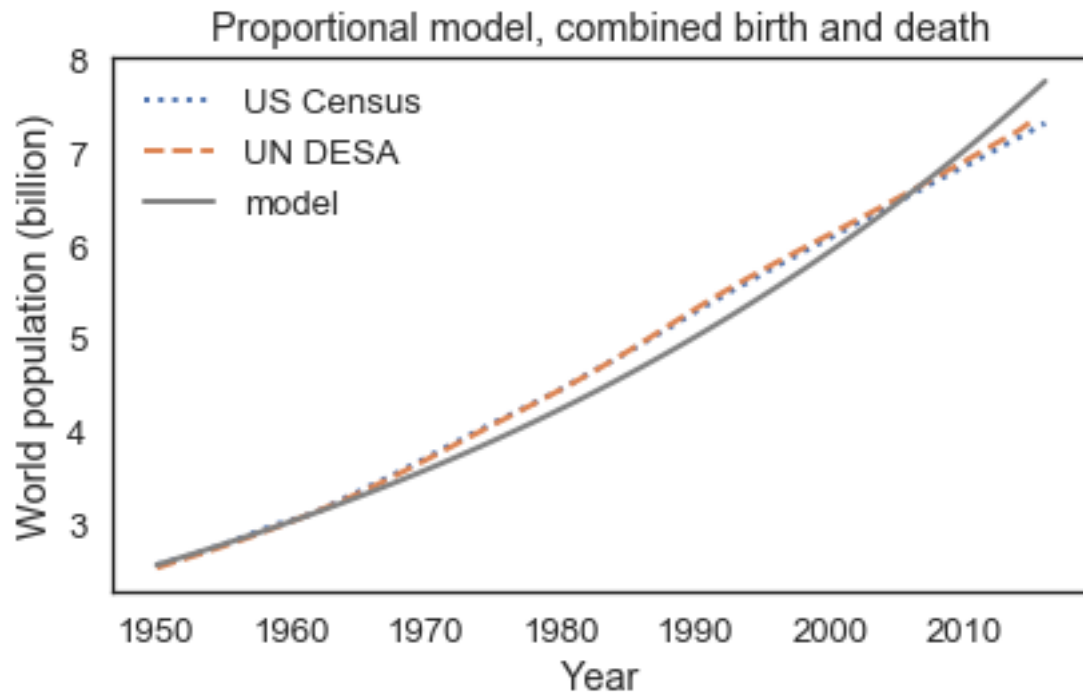


Figure 4: png

```
system = System(t_0=t_0,
                t_end=t_end,
                p_0=p_0,
                alpha1 = alpha1,
                alpha2 = alpha2)
```

values

t_0

1950.000000

t_end

2016.000000

p_0

2.557629

alpha1

0.017000

alpha2

0.019500

```
def update_func3(pop, t, system):
    if t <= 1979:
        net_growth = system.alpha1 * pop
```



```
if t >= 1980:  
    net_growth = system.alpha2 * pop  
    return pop + net_growth
```

```
results = run_simulation(system, update_func3)  
plot_results(census, un, results, 'Proportional model 2, combined birth and death')
```

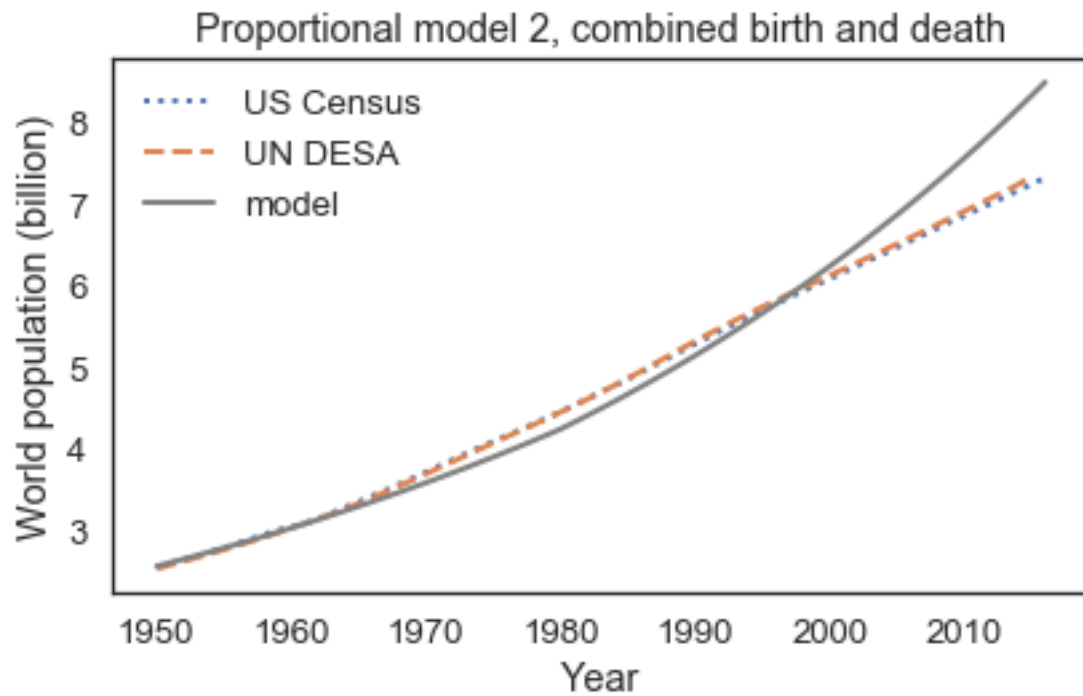


Figure 5: png