### Modeling and Simulation in Python

```
Chapter 8
```

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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 *
```

### Functions from the previous chapter

```
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
```

### Reading the data

```
def read_table2(filename = 'data/World_population_estimates.html'):
    tables = pd.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']
    return table2
#table2 = read_table2()
\#table2.\ to\_csv('data/World\_population\_estimates2.\ csv')
table2 = pd.read_csv('data/World_population_estimates2.csv')
table2.index = table2.Year
table2.head()
Year
census
prb
un
maddison
hyde
tanton
biraben
mj
thomlinson
durand
clark
Year
1950
1950
2557628654
2.516000e+09
2.525149e+09
2.544000e+09
2.527960e+09
2.400000e+09
2.527000e+09
2.500000e+09
2.400000e+09
```

NaN

2.486000e+09

1951

1951

2594939877

NaN

 $2.572851\mathrm{e}{+09}$ 

2.571663e+09

NaN

NaN

NaN

NaN

NaN

NaN

NaN

1952

1952

2636772306

NaN

2.619292e+09

2.617949e + 09

NaN

NaN

NaN

NaN

NaN

NaN

NaN

1953

1953

2682053389

NaN

2.665865e+09

2.665959e+09

NaN

NaN

```
NaN
NaN
NaN
NaN
NaN
1954
1954
2730228104
NaN
2.713172e+09
2.716927e+09
NaN
NaN
NaN
NaN
NaN
NaN
NaN
un = table2.un / 1e9
census = table2.census / 1e9
plot(census, ':', label='US Census')
plot(un, '--', label='UN DESA')
decorate(xlabel='Year',
```

### Running the quadratic model

Here's the update function for the quadratic growth model with parameters alpha and beta.

ylabel='World population (billion)',
title='Estimated world population')

```
def update_func_quad(pop, t, system):
    """Update population based on a quadratic model.

pop: current population in billions
    t: what year it is
    system: system object with model parameters
    """
    net_growth = system.alpha * pop + system.beta * pop**2
    return pop + net_growth
```

Extract the starting time and population.

## Estimated world population

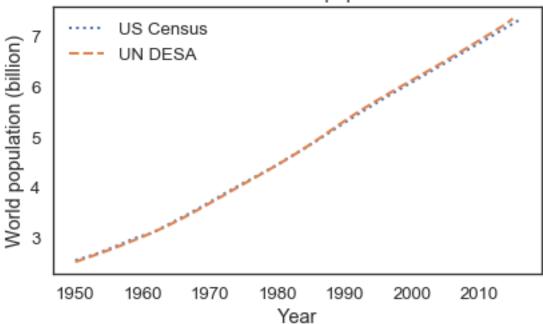


Figure 1: png

```
t_0 = get_first_label(census)
t_end = get_last_label(census)
p_0 = get_first_value(census)
```

### 2.557628654

Initialize the system object.

values

t\_0

1950.000000

 $t\_end$ 

2016.000000

 $p\_0$ 

2.557629

alpha

0.025000

beta

-0.001800

Run the model and plot results.

```
results = run_simulation(system, update_func_quad)
plot_results(census, un, results, 'Quadratic model')
```

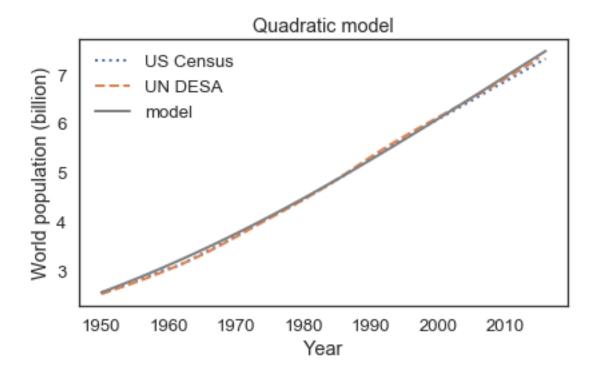


Figure 2: png

### Generating projections

To generate projections, all we have to do is change t\_end

```
system.t_end = 2250
results = run_simulation(system, update_func_quad)
plot_results(census, un, results, 'World population projection')
savefig('figs/chap08-fig01.pdf')
```

Saving figure to file figs/chap08-fig01.pdf

The population in the model converges on the equilibrium population, -alpha/beta

```
results[system.t_end]
```

13.856665141368708

# World population projection

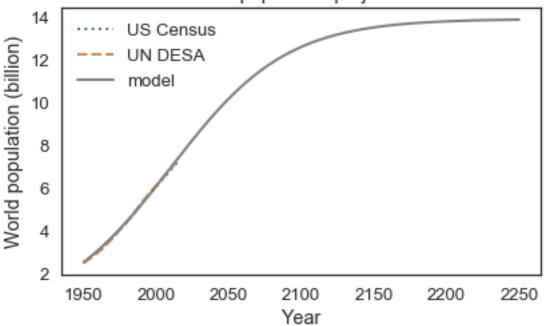


Figure 3: png

```
-system.alpha / system.beta
```

### 13.88888888888889

**Exercise:** What happens if we start with an initial population above the carrying capacity, like 20 billion? Run the model with initial populations between 1 and 20 billion, and plot the results on the same axes.

values

t\_0

1950.0000

 $t_{end}$ 

2016.0000

 $p_0$ 

```
1.0000
p_end
20.0000
alpha
0.0250
beta
-0.0018
```

```
results = run_simulation(system, update_func_quad)
plot_results(census, un, results, 'Quadratic model')
```

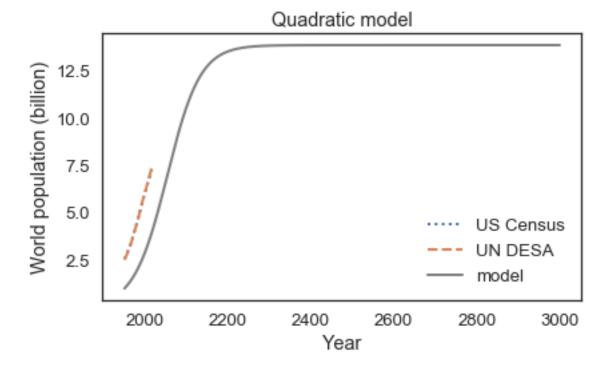


Figure 4: png

The ModSim Package assumes that the population growth has been limited due to a number of factors. The model is based on the assumption that population growth is limited by the access to resources. Meaning as the population appraches the carrying capacity, resources become less available.

### Comparing projections

We can compare the projection from our model with projections produced by people who know what they are doing.

```
def read_table3(filename = 'data/World_population_estimates.html'):
    tables = pd.read_html(filename, header=0, index_col=0, decimal='M')
    table3 = tables[3]
    table3.columns = ['census', 'prb', 'un']
    return table3
```

```
#table3 = read_table3()
\#table3.to\_csv('data/World\_population\_estimates3.csv')
table3 = pd.read_csv('data/World_population_estimates3.csv')
table3.index = table3.Year
table3.head()
Year
census
prb
un
Year
2016
2016
7.334772e+09
NaN
7.432663\mathrm{e}{+09}
2017
2017
7.412779e + 09
NaN
NaN
2018
2018
7.490428e+09
NaN
NaN
2019
2019
7.567403e+09
NaN
NaN
2020
2020
7.643402e+09
NaN
```

NaN is a special value that represents missing data, in this case because some agencies did not publish projections for some years.

7.758157e + 09

This function plots projections from the UN DESA and U.S. Census. It uses dropna to remove the NaN values from each series before plotting it.

```
def plot_projections(table):
    """Plot world population projections.

    table: DataFrame with columns 'un' and 'census'
    """
    census_proj = table.census / 1e9
    un_proj = table.un / 1e9

plot(census_proj.dropna(), ':', color='CO', label='US Census')
    plot(un_proj.dropna(), '--', color='C1', label='UN DESA')
```

Run the model until 2100, which is as far as the other projections go.

```
t_0 = get_first_label(census)
t_end = get_last_label(census)
p_0 = get_first_value(census)
```

#### 2.557628654

```
values
```

t 0

1950.000000

 $t_{end}$ 

2100.000000

 $p_0$ 

2.557629

alpha

0.025000

beta

-0.001800

```
results = run_simulation(system, update_func_quad)
plt.axvspan(1950, 2016, color='CO', alpha=0.05)
plot_results(census, un, results, 'World population projections')
plot_projections(table3)
savefig('figs/chap08-fig02.pdf')
```

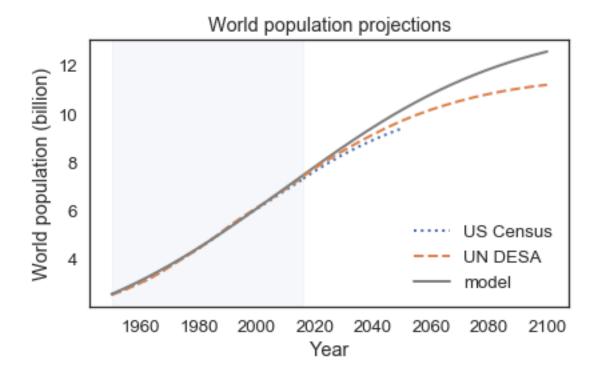


Figure 5: png

Saving figure to file figs/chap08-fig02.pdf

People who know what they are doing expect the growth rate to decline more sharply than our model projects.

### Exercises

**Exercise:** The net growth rate of world population has been declining for several decades. That observation suggests one more way to generate projections, by extrapolating observed changes in growth rate.

The modsim library provides a function, compute\_rel\_diff, that computes relative differences of the elements in a sequence.

Here's how we can use it to compute the relative differences in the census and un estimates:

```
alpha_census = compute_rel_diff(census)
plot(alpha_census, label='US Census')

alpha_un = compute_rel_diff(un)
plot(alpha_un, label='UN DESA')

decorate(xlabel='Year', label='Net growth rate')
```

Other than a bump around 1990, net growth rate has been declining roughly linearly since 1965. As an exercise, you can use this data to make a projection of world population until 2100.

1. Define a function, alpha\_func, that takes t as a parameter and returns an estimate of the net growth rate at time t, based on a linear function alpha = intercept + slope \* t. Choose values of slope and intercept to fit the observed net growth rates since 1965.

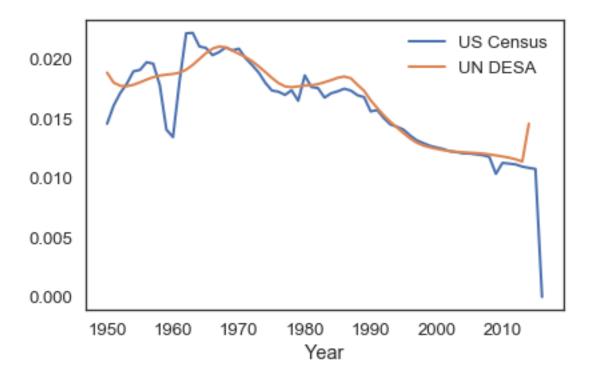


Figure 6: png

- 2. Call your function with a range of ts from 1960 to 2020 and plot the results.
- 3. Create a System object that includes alpha\_func as a system variable.
- 4. Define an update function that uses alpha\_func to compute the net growth rate at the given time t.
- 5. Test your update function with  $t_0 = 1960$  and  $p_0 = census[t_0]$ .
- 6. Run a simulation from 1960 to 2100 with your update function, and plot the results.
- 7. Compare your projections with those from the US Census and UN.

```
def alpha_func(t):
    intercept = 0.02
    slope = -0.00022
    y = 1970
    return intercept + slope * (t-y)
```

```
ts = linrange(1960, 2021)
```

```
array([1960, 1961, 1962, 1963, 1964, 1965, 1966, 1967, 1968, 1969, 1970, 1971, 1972, 1973, 1974, 1975, 1976, 1977, 1978, 1979, 1980, 1981, 1982, 1983, 1984, 1985, 1986, 1987, 1988, 1989, 1990, 1991, 1992, 1993, 1994, 1995, 1996, 1997, 1998, 1999, 2000, 2001, 2002, 2003, 2004, 2005, 2006, 2007, 2008, 2009, 2010, 2011, 2012, 2013, 2014, 2015, 2016, 2017, 2018, 2019, 2020], dtype=int32)
```

```
test_model = TimeSeries(alpha_func(ts), ts)
plot(test_model, color='gray', label='alpha_test_model')
plot(alpha_census)
plot(alpha_un)

decorate(xlabel='Year', ylabel='Net growth rate')
```

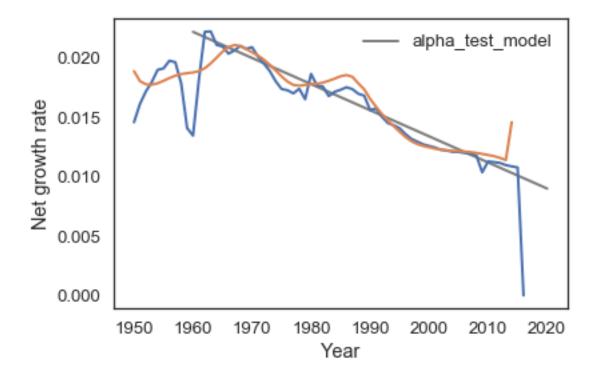


Figure 7: png

```
t_0 = 1960
p_0 = census[t_0]
```

### 3.043001508

values

t\_0

1960

 $t_{end}$ 

2100

 $p\_0$ 

```
3.043
```

alpha\_func

<function alpha\_func at 0x000001CD4601F5E8>

```
def update_func_alpha(pop, t, system):
    net_growth = system.alpha_func(t) * pop
    return pop + net_growth
```

```
update_func_alpha(p_0, t_0, system)
```

### 3.1105561414776

```
results = run_simulation(system, update_func_alpha);

plot_results(census, un, results, 'World population projections')
plot_projections(table3)
```



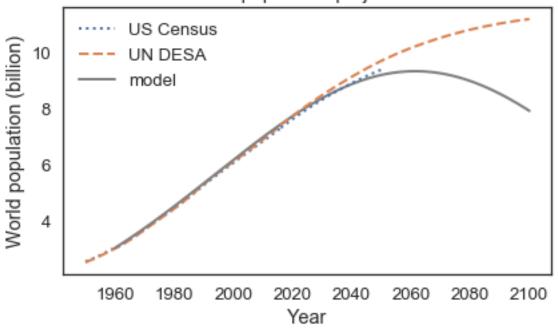


Figure 8: png

**Related viewing:** You might be interested in this video by Hans Rosling about the demographic changes we expect in this century.