Lab 2

Introduction to Python Programming for Economics & Finance

Richard Foltyn *University of Glasgow*

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1 Portfolio choice problem

Consider a portfolio choice problem where an investor chooses the fraction α to invest in a risky asset in order to maximise expected utility,

$$\max_{\alpha \in [0,1]} E_t \left[u(W_{t+1}) \right]$$

Assume that the investor consumes all of next-period's wealth W_{t+1} which is given by

$$W_{t+1} = R_{t+1}\alpha W_t + R_f(1-\alpha)W_t$$

where W_t is the initial investable wealth in period t, R_{t+1} is the gross return on the risky investment and R_f is the risk-free return on the fraction of the portfolio which is invested a risk-free asset (e.g., a bank deposit). The utility function $u(\bullet)$ has a constant relative risk aversion (CRRA) form and is given by

$$u(W) = \begin{cases} \frac{W^{1-\gamma}}{1-\gamma} & \text{if } \gamma \neq 1\\ \log(W) & \text{if } \gamma = 1 \end{cases}$$

where γ is a parameter governing the investor's risk aversion (higher values of γ imply that the investor is more averse to taking risk).

For simplicity, let the gross risk-free return be $R_f = 1$. Finally, assume that the risky return can take on two realisations, high and low, with equal probability,

$$R_{t+1} = \begin{cases} 1 + \mu + \epsilon & \text{with probability } \frac{1}{2} \\ 1 + \mu - \epsilon & \text{with probability } \frac{1}{2} \end{cases}$$

where $\mu > 0$ is the risk premium and $\epsilon > 0$ parametrises the volatility of risky returns.

2 Expected utility function

Write a Python function that takes as arguments the risky share α , the initial wealth W_t , and the parameters μ , ϵ and γ , and returns the expected utility associated with the given values. Your function signature should look like this:

```
[1]: def expected_util(alpha, W, mu, epsilon, gamma):
    # Compute the associated expected utility
    # eu = ...
    return eu
```

Make sure that your function works correctly for both $\gamma = 1$ and $\gamma \neq 1$. Moreover, the function should allow for the arguments α and W to be passed as both scalar values as well as NumPy arrays!

3 Plot expected utility against the risky share

Use the function you wrote above to plot expected utility for 100 values of α on the unit interval [0,1]. Assume the following values for the remaining parameters:

Label both axes and add a legend to your plot.

4 Optimal risky share

In the above section, you plotted expected utility for all possible risky shares. As a next step, you are asked to compute the *optimal* risky share for a given set of parameters and initial wealth W_t .

You can find the optimal α numerically using one of the minimiser function included in SciPy's optimization package. Specifically, for this task you should use the function minimize_scalar() with the argument method='bounded' since the risky share is a scalar variable which is bounded between [0,1]. Use the expected_util() function from the previous question as the minimiser's objective.

The following code illustrates how to set up the minimisation for a specific set of parameters:

```
[5]: from scipy.optimize import minimize_scalar
                        # Initial wealth
     W = 1.0
     gamma = 2.0
                        # Relative risk aversion
     mu = 0.04
                         # Risk premium
     epsilon = 0.2
                         # Standard deviation of risky return
     # Compute optimal risky share for given parameters
     result = minimize_scalar(
         # SciPy minimiser passes a single argument (alpha), which we plug into
         # expected_util() together with any other parameters.
         lambda alpha: - expected_util(alpha, W, mu, epsilon, gamma),
         method='bounded',
         bounds=[0, 1]
     )
```

```
# Print optimal risky share
print(f'Optimal risky share: {result.x}')
```

Optimal risky share: 0.5155169286613541

The following comments should help you understand what the above code does:

- SciPy's routines all perform *minimisation*, whereas we need to *maximise* expected utility. We get around this problem by using a lambda expression and returning *negative* expected utility.
- The expected_util() function you wrote requires several arguments, SciPy's minimize_scalar(), however, expects a function that takes only a single argument. We can again get around this using the lambda expression as shown above.
- The maximisation result is returned as an OptimizeResult object. The only thing you need to know about this object is that the optimal value is stored in the attribute x and can be retrieved as illustrated above.

4.1 Optimal risky share by wealth

Consider a set of initial wealth levels W_t uniformly spread over the interval [1, 10],

```
[6]: W = np.linspace(1.0, 10.0, 100)
```

Write a loop that computes the optimal risky share for each of these wealth levels, using the same values for the remaining parameters as above:

```
[7]: gamma = 2.0  # Relative risk aversion
mu = 0.04  # Risk premium
epsilon = 0.2  # Standard deviation of risky return
```

Plot the optimal risky share against initial wealth. Set the plot range to the interval [0,1.1] using plt.ylim((0.0, 1.1)) to clearly see the results. How does the optimal risky share depend on initial wealth?

4.2 Optimal risky share by relative risk aversion (RRA)

Now consider a set of RRA parameters γ uniformly spread over the interval [1,5],

```
[10]: gammas = np.linspace(1.0, 5.0, 100)
```

Write a loop that computes the optimal risky share for each γ , using the following values for initial wealth and the remaining parameters:

```
[11]: W = 1.0 # Initial wealth
mu = 0.04 # Risk premium
epsilon = 0.2 # Standard deviation of risky return
```

Plot the optimal risky share as a function of these RRA values. How does the optimal risky share depend on risk aversion γ ?