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Bachelor Thesis

Portfolio Choice: Optimal Investment - Consumption Decision

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Vienna, May 6, 2022

Affidavit

I declare in lieu of oath, that I wrote this thesis and performed the associated research myself, using only literature cited in this volume. If text passages from sources are used literally, they are marked as such.

I confirm that this work is original and has not been submitted elsewhere for any examination, nor is it currently under consideration for a thesis elsewhere.

Vienna, February, 2022

 
signature

Abstract

In this thesis I study a two period consumption and savings problem of an expected power-utility maximizing agent. In order to determine the optimum of the total expected utility, fundamentals of portfolio management and asset pricing are applied. I discuss the effects of a proportional change in returns, as well as a mean preserving spread on the optimal investment strategy. Also, the connection between the price of the investment and the central asset pricing formula is established. In order to simplify the analysis of the problem, I create a corresponding visualization using R-Shiny, which I also describe.

Table of Contents

1	Introduction	1
1.1	Risky investment	1
1.2	Risky and riskless investment	2
1.3	Execution	3
2	Theoretical Framework	4
2.1	Utility	4
2.2	Marginal utility	4
2.3	Expected utility	5
2.4	Risk aversion	5
2.5	Isoelastic utility function	5
3	Implementation	7
3.1	User interface	7
3.2	Server function	7
4	Model Analysis	11
4.1	Scenario 1	11
4.2	Scenario 2	18
5	Conclusion	25
	References	a
	List of Figures	b
	Appendix	c
	User Interface Object	c
	Server Function	h
	Call Function	y

1 Introduction

Uncertainty is a fundamental aspect of life. Every day, people are exposed to various risks. The economic and financial spheres in particular are characterized by uncertainty. When making investment or other financial decisions, there is usually no safe way to determine in advance what the value of the final outcome will be. Be it an investment decision on the stock exchange, or a simple purchase decision in the supermarket, no one can look into the future. So there is no other option than to rely on probabilities and approximate estimates.

One way to protect yourselves from unforeseen trends is portfolio diversification. On one hand, you may miss out on some of the potential gain from an asset with rising value, but on the other hand, you are better protected against losses from assets with falling value. The goal of a risk-averse investor is to maximize a portfolio's risk adjusted return, i.e., to optimize its risk/return tradeoff. In return, you give up part of the expected profit. Therefore, it often makes sense not to spend the entire capital on one asset, but to divide and spread it.

I take the concepts presented in this introductory section mostly from (Varian, 2010) and (Ingersoll, 1987). Some terms essential for understanding are defined and explained in the Theoretical Framework section.

1.1 Risky investment

In a first scenario, the only available investment is a risky asset. An investor, endowed with a certain capital K , must today ($t = 0$) decide how much to spend on an investment portfolio and how much to consume. In order to have consumption tomorrow ($t = 1$) the investor has to purchase an investment portfolio today such that the portfolio value tomorrow can be consumed. The amount of capital which is consumed today is referred to as c_0 , the amount available for consumption tomorrow is c_1 .

The portfolio's payoff is not known beforehand. It is assumed that there are N possible states, each is assigned a certain probability π_i with $i \in \{1, 2, \dots, N\}$ and a dividend per invested unit of capital, $d_{1,i}$. The probabilities must sum to 100% and each payoff must be larger than zero. One unit of investment has a price p_0 , the total amount of units purchased is denoted by x . Thus, the amount available for consumption tomorrow is $c_{1,i} \leq x \cdot d_{1,i}$. In the meantime, there is no other way to receive income at $t = 1$, so investing today is the only way to receive consumption tomorrow.

The investor derives utility from consumption. Utility from immediate consumption is assumed iso-elastic (i.e., with constant relative risk aversion γ):

$$u(c) = \begin{cases} \frac{c^{1-\gamma}-1}{1-\gamma} & \text{if } \gamma \neq 1, \\ \ln(c) & \text{if } \gamma = 1. \end{cases} \quad (1.1)$$

Since future consumption c_1 is uncertain, the investor seeks to maximize expected utility, which he/she determines in a time-separable way as

$$U = u(c_0) + \beta \cdot \sum_{i=1}^N \pi_i \cdot u(c_{1,i}) \quad (1.2)$$

where

$$\begin{aligned} c_0, c_{1,i} &> 0, \\ \beta &\leq 1, \\ \gamma &> 0. \end{aligned} \quad (1.3)$$

In order to express the investor's time preference the discount factor β is introduced. It can be chosen according to the weight the investor places on the utility received tomorrow compared to today.

The utility at $t = 1$ is stochastic. Expected utility can be calculated as the sum of the resulting utility in each state, each multiplied by the probability of its occurrence and the discount factor β .

Ultimately, the investor seeks to maximize the utility function (1.2) with the following constraints:

$$\begin{aligned} c_0 &= K - x \cdot p_0, \\ c_{1,i} &= x \cdot d_{1,i}, \\ x &\geq 0. \end{aligned} \quad (1.4)$$

1.2 Risky and riskless investment

In another scenario, it is assumed that the investor has an additional option to receive income tomorrow: a riskless investment.

One unit of the riskless investment has a price of b_0 and the quantity of riskless units purchased is denoted by y . The investor receives a fixed payment of 1 per unit in each state at $t = 1$. So total consumption tomorrow is the sum of a risky component $x \cdot d_{1,i}$ (the dividend) and a riskless component $y \cdot 1$ (savings in a riskless savings account).

The utility function to be maximized remains the same as in the first scenario, but there are new constraints:

$$\begin{aligned} c_0 &= K - x \cdot p_0 - y \cdot b_0, \\ c_{1,i} &= x \cdot d_{1,i} + y, \\ x &\geq 0, \\ y &\geq 0. \end{aligned} \quad (1.5)$$

The investor seeks to maximize the total expected utility by finding the optimal combination of x and y . This also leaves the possibility of not using one of the two investment options at all.

1.3 Execution

The problem was implemented and visualized as an R-Shiny project. This makes it very easy to identify the optimum and the parameters can be selected interactively.

2 Theoretical Framework

2.1 Utility

Although utility was originally introduced by philosophers and economists as a measure of happiness and overall pleasure, in the context of modern economics it is rather used as a tool to model worth or value. Overall this concept is based on the description of preferences and has more of a comparative character (Varian, 2010, p. 54).

A utility function $u(x, y)$ determines the utility which results from the consumption of a certain bundle of goods (x, y) as a numerical value such that a larger number indicates a higher preference. This helps to illustrate the order of preference, however, the numerical magnitude has no intrinsic meaning (Nicholson and Snyder, 2016, p. 90). Therefore, it is not possible to say that something gives twice as much utility as something else, as is attempted, for example, in the cardinal utility approach. The focus here is solely on ordering bundles based on the utility generated. To describe choice behavior, ordinal utility is sufficient. For example, if a monotonic transformation such as $f(u) = 2 \cdot u$ is applied to a particular utility function, the value of the utility obtained changes (doubles in this case), but the ordering of our preferences remains the same (Nicholson and Snyder, 2016, p. 56).

Nevertheless, it is possible and desirable to search for bundles which have the same utility. All bundles with the same constant utility $u(x_i, y_i) = k$ can be grouped into a set called the level set (Varian, 2010, p. 59). The graphical representation of the level set is called the indifference curve. The consumer is indifferent to all bundles on the indifference curve because they all lead to the same utility level.

For different utility levels different indifference curves are displayed. A utility function can be interpreted geometrically as a way of labeling the indifference curves (Varian, 2010, p. 57). A monotonic transformation would consequently mean a relabeling of the indifference curves.

2.2 Marginal utility

Given the utility that a consumer derives from consumption today and tomorrow, $u(c_0, c_1)$, the rate of change in utility as the consumption in one point of time is increased by a small margin Δc_0 , is called marginal utility MU_{c_0} .

If the additional value is infinitesimally small, one obtains the partial derivative of the utility function, which is equal to the slope of the utility function with respect to c_0 while c_1 remains fixed (Varian, 2010, p. 70):

$$MU_{c_0} = \lim_{\Delta c_0 \rightarrow 0} \frac{u(c_0 + \Delta c_0, c_1) - u(c_0, c_1)}{\Delta c_0} = \frac{\partial u(c_0, c_1)}{\partial c_0}. \quad (2.1)$$

The numerical magnitude also has no actual meaning, since it depends on the utility function in question. But it helps to identify where a small increase in immediate consumption would have a large impact on total utility.

2.3 Expected utility

If a consumer has to make a choice under uncertainty he/she might take into account the probabilities of different scenarios. The expected utility can then be expressed as a weighted sum of the utility functions in each state, where the weights are given by the probabilities (Varian, 2010, p. 223). A utility function of the form

$$U = \sum_i \pi_i u(c_i) \quad (2.2)$$

is called a von Neumann-Morgenstern utility function.

2.4 Risk aversion

Whether an agent is risk-averse or risk-seeking depends on the expected utility function. A risk-averse agent would rather choose a gamble with low uncertainty over a gamble with high uncertainty, even if the expected payoff of the relatively risky gamble is equal or slightly higher compared to the expected payoff of the relatively safe gamble (Arrow, 1996, p. 104).

Conversely, this means that a risk-averse agent is willing to trade risk for some amount of return (Pratt, 1964, p. 122) and the higher the risk aversion, the higher the amount the actor would be willing to pay for insurance.

One way to measure risk aversion is the concavity index or Arrow-Pratt measure of absolute risk aversion (Wakker, 2008, p. 1339)

$$A(c) = -\frac{u''(c)}{u'(c)}. \quad (2.3)$$

The Arrow-Pratt measure of relative risk aversion is

$$R(c) = -\frac{c \cdot u''(c)}{u'(c)}, \quad (2.4)$$

which, unlike the former, is dimensionless (Simon and Blume, 1994, p. 363). Relative risk aversion “is useful in analyzing risks expressed as a proportion of the gamble for example investment rates of return” (Ingersoll, 1987, p. 39).

2.5 Isoelastic utility function

To model the utility created by today’s consumption and expected utility of tomorrow’s consumption (the value of the investor’s portfolio at $t = 1$) the isoelastic utility function or power utility function (see equation 1.1) will be applied.

This function belongs to the class of hyperbolic absolute risk aversion functions, which are also called linear risk tolerance utility functions. This means that risk tolerance, which is the reciprocal of the absolute risk aversion, is a linear affine function in consumption (Ingersoll, 1987, p. 39):

$$T(c) = \frac{1}{A(c)} = \frac{c}{1-\gamma} + \frac{b}{a} = -\frac{u'(c)}{u''(c)}. \quad (2.5)$$

A special property of the power utility function is the constant relative risk aversion:

$$\begin{aligned} u'(c) &= c^{-\gamma} \\ u''(c) &= -\gamma \cdot c^{-\gamma-1} \end{aligned} \quad (2.6)$$

$$R(c) = -\frac{c \cdot u''(c)}{u'(c)} = \gamma. \quad (2.7)$$

The utility function $u(c)$ of a risk-averse agent is increasing, strictly concave and has positive absolute risk aversion (Poon, 2018, p. 2), provided that $c \geq 0$:

$$u'(c) > 0, \quad u''(c) < 0 \quad \text{and} \quad A(c) > 0. \quad (2.8)$$

Due to the negative second derivative and therefore strictly monotonically decreasing slope the first marginal unit of consumption has the largest impact $\lim_{c \rightarrow 0} u'(c) = \infty$. Thus, the so-called INADA-condition is satisfied. At the same time the effect of an additional unit of input converges to zero when an infinite amount of input is used $\lim_{c \rightarrow \infty} u'(c) = 0$ (Uzawa, 1971, p. 20). This is referred to as “saturation effect”.

With the isoelastic utility function an agent’s risk behaviour depends on the constant relative risk aversion γ . If γ is positive, then the actor is risk-averse, otherwise they would be risk seeking, $\gamma = 0$ corresponds to risk neutrality.

The case of $\gamma \rightarrow 1$ can be studied using L’Hospital’s rule. For this case, the utility function is equal to the logarithmic function.

In applying the utility function, it is assumed that the quantity consumed cannot be less than zero. In fact, as the function approaches zero the utility diverges to minus infinity, if the agent has a risk aversion larger or equal to 1. The zero-utility of a less risk-averse agent ($0 < \gamma < 1$) equals $u(0) = -\frac{1}{1-\gamma}$.

The limit approaching infinity does also depend on the value of relative risk aversion. If $\gamma \leq 1$, then the function diverges, but if $\gamma > 1$, then the value of the utility function converges to $\lim_{c \rightarrow \infty} u(c) = \frac{1}{\gamma-1}$.

3 Implementation

The problem visualisation was implemented as an application using the R-Shiny framework, which consists of three components: a user interface object, a server function and a call function. The code for each component can be found in the appendix.

Essentially the call function just specifies which user interface object and server function to call in the Shiny app.

3.1 User interface

The user interface object consists of three tabs. The default tab shows the description of the problem, similar to the introduction to this thesis.

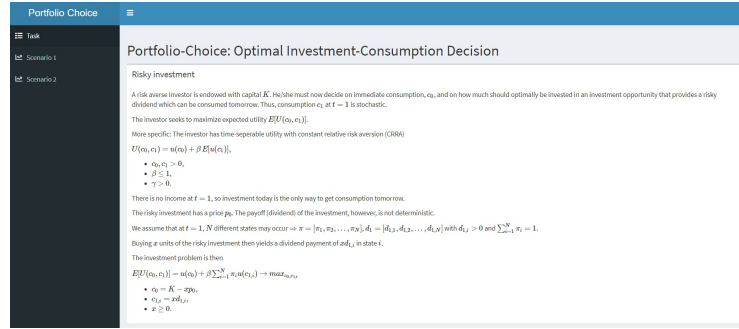


Figure 1: Problem description

The second tab is the visualization of the first scenario: The top diagram shows a plot of the utility generated at $t = 0$, the expected utility in $t = 1$ and the total expected utility.

Secondly, a diagram of indifference curves is displayed.

Finally, there is a plot of the total expected utility and the marginal utility.

All parameters can be set individually by the user. For this purpose, there are sliders on a panel on the left side. There are also two reset buttons, one for the function parameters and one for the investment parameters. At the bottom of the panel, the values of the resulting optimum are displayed.

The visualization of scenario 2 is located on the third tab, which is structured analogously.

3.2 Server function

The server function can be divided into two parts. Up to line 448 the functions for scenario 1 are defined, afterwards the functions for scenario 2. Both parts are once again similar in structure:

First, the function and the default values for the reset buttons are set. In the code, this is done from line 3 to line 79 for scenario 1 and from line 466 to 550 for scenario 2.

From line 80 to 130 and line 551 to 599, respectively, the function comes, which adjusts the probabilities so that the sum always adds up to 100%.

Between line 131 and 157 and line 600 and 629, respectively, the function parameters and investment specifications are imported and collected into lists for easier processing.

Then, in line 158-187 for scenario 1 and line 630-667 for scenario 2, respectively, the function follows, which maximizes the utility function. It is assumed that the constraints (1.4) and (1.5), respectively, are binding. In section 4.1 and section 4.2 it is shown that in the optimum this must be true.

For simplicity, the utility function is evaluated on a discrete set of points and the maximum can easily be determined. Since the utility function in scenario 1 is one-dimensional, evaluation over a vector is sufficient to find the optimal investment x^* . In Scenario 2, the utility function is evaluated over a matrix in order to find the best combination of x and y , which is denoted as (x^*, y^*) .

Between line 188 and 250 and line 668 and 771, respectively, the charts are generated: For Scenario 1, the first diagram simply consists of the three graphs, total utility, utility at $t = 0$ and expected utility at $t = 1$, plotted over the amount spent on the risky investment x_{p0} . The diagram for scenario 2 consists of two separate plots. First, the graphs are plotted over x_{p0} while y_{b0} remains fixed at y^*b_0 and then vice versa.

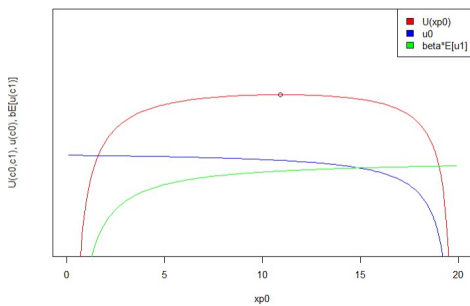


Figure 2: Utility, scen. 1

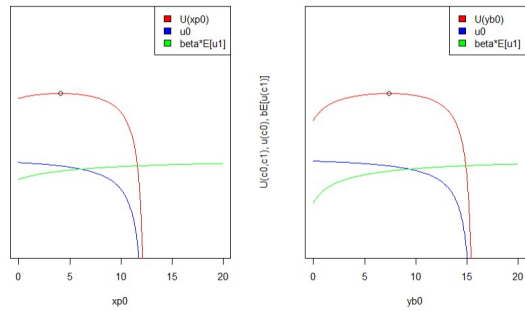


Figure 3: Utility, scen. 2

From line 251 to 340 and line 772 to 881, respectively, follows the generation of the plot of indifference curves. The abscissa corresponds to x_{p0} , i.e. the amount of capital spent on the risky investment. The ordinate refers to the amount consumed immediately, c_0 . For one indifference curve, the constant value is equal to the utility generated in the optimum, which is the maximum possible utility with the given parameters and constraints. Meanwhile, the share of riskless invested capital remains fixed at y^*b_0 . The other indifference curve can be chosen by the user by adjusting a

slider. From then on, things work a little differently for scenario 1 and scenario 2: In scenario 1, the user determines a certain percentage of the original capital. This proportion of capital is then added to (or subtracted from) the originally available capital. The constant value of the indifference curve then corresponds to the maximum utility possible with this newly available capital. In scenario 2, the user determines how much more or less (in percent) is spent on the riskless investment. The budget constraint remains binding. Thus, the amount spent on consumption and the risky investment also changes, and the resulting utility cannot be the optimum. The constant value of the new indifference curve corresponds to the maximum possible utility at the given risk-free expenditure. All points (xp_0, c_0) for which this holds are part of the indifference curve. The legend shows the resulting utilities for the two indifference curves and the amount spent on the riskless investment.

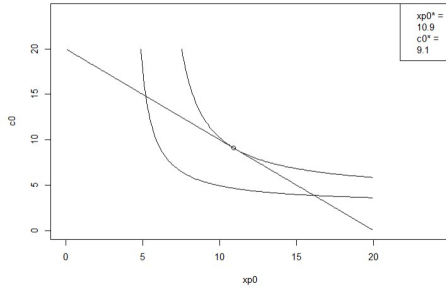


Figure 4: Indifference curves, scen. 1

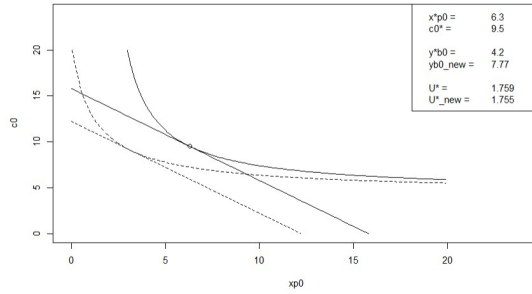


Figure 5: Indifference curves, scen. 2

The last diagram is generated between line 341 and 451 for scenario 1 and line 882 and 999. It shows in sub-diagrams the total expected utility plotted over c_0 , xp_0 and yb_0 , respectively, for scenario 2. The optimum is marked and the marginal utility is shown as a tangent through the optimum.

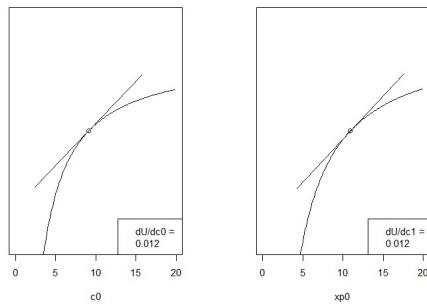


Figure 6: Marginal utilities, scen. 1

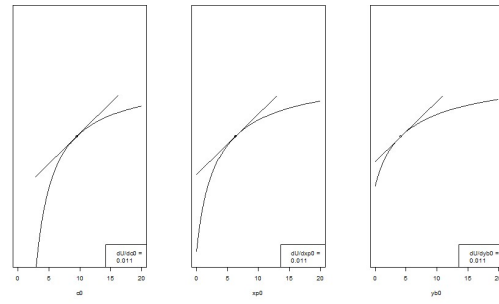


Figure 7: Marginal utilities, scen. 2

Finally (lines 452-465 and line 1000-1016, respectively), the function follows which gives the maximum expected utility, the amount consumed immediately and the

expenditure on the investments at the optimum. Small errors may occur, due to the discretization of the domain over which the optimum is evaluated.

4 Model Analysis

4.1 Scenario 1

The problem to be optimized can be expressed as

$$\begin{aligned}
 U(x) &= u(c_0) + \beta \cdot \sum_{i=1}^N \pi_i \cdot u(c_{1,i}) \quad \rightarrow \quad \max_x \\
 \text{subject to} \\
 c_0 &= K - x \cdot p_0, \\
 c_{1,i} &= x \cdot d_{1,i}, \\
 x &\geq 0.
 \end{aligned} \tag{4.1}$$

The Lagrangian associated to this optimization problem is

$$\begin{aligned}
 L(\lambda_i, x, \mu) &= u(c_0) + \beta \cdot \sum_{i=1}^N \pi_i \cdot u(c_{1,i}) + \lambda_0 \cdot (K - x \cdot p_0 - c_0) + \\
 &\quad + \sum_{i=1}^N \lambda_i \cdot (x \cdot d_{1,i} - c_{1,i}) + \mu \cdot x.
 \end{aligned} \tag{4.2}$$

The first order conditions of optimality are formulated as follows:

$$\frac{\partial L}{\partial \lambda_0} = K - x p_0 - c_0 \geq 0 \tag{4.3a} \qquad \frac{\partial L}{\partial \lambda_i} = x d_{1,i} - c_{1,i} \geq 0 \tag{4.4a}$$

$$\lambda_0 \geq 0 \tag{4.3b} \qquad \lambda_i \geq 0 \tag{4.4b}$$

$$\lambda_0 \cdot \frac{\partial L}{\partial \lambda_0} = 0 \tag{4.3c} \qquad \lambda_i \cdot \frac{\partial L}{\partial \lambda_i} = 0 \tag{4.4c}$$

$$\frac{\partial L}{\partial c_0} = u'(c_0) - \lambda_0 \leq 0 \tag{4.5a} \qquad \frac{\partial L}{\partial c_{1,i}} = \beta \pi_i u'(c_{1,i}) - \lambda_i \leq 0 \tag{4.6a}$$

$$c_0 \geq 0 \tag{4.5b} \qquad c_{1,i} \geq 0 \tag{4.6b}$$

$$c_0 \cdot \frac{\partial L}{\partial c_0} = 0 \tag{4.5c} \qquad c_{1,i} \cdot \frac{\partial L}{\partial c_{1,i}} = 0 \tag{4.6c}$$

$$\frac{\partial L}{\partial \mu} = x \geq 0 \tag{4.7a} \qquad \frac{\partial L}{\partial x} = -\lambda_0 p_0 + \sum_{i=1}^N \lambda_i d_{1,i} \leq 0 \tag{4.8a}$$

$$\mu \geq 0 \tag{4.7b} \qquad x \geq 0 \tag{4.8b}$$

$$\mu \cdot \frac{\partial L}{\partial \mu} = 0 \tag{4.7c} \qquad x \cdot \frac{\partial L}{\partial x} = 0 \tag{4.8c}$$

In section 2.5 it has already been stated that the INADA condition is satisfied. Hence, consumption must be larger than zero today c_0 and tomorrow $c_{1,i}$. Therefore, the investor will always choose to consume some share of his capital immediately and invest some in order to have consumption tomorrow.

From c_0 being strictly positive and the optimality condition (4.5) one can see that

$$\lambda_0 = u'(c_0), \quad (4.9)$$

which is always larger than zero according to equation (2.8). Therefore, the first constraint of the optimization problem (4.1) must be binding (see condition (4.3)), which has already been assumed.

The Lagrange-multiplier λ_0 indicates the marginal utility with respect to the available capital K : A small increase in capital ΔK causes a small gain in utility $\Delta u = \lambda_0 \cdot \Delta K$.

Similarly, it can be shown through the optimality condition (4.6) that

$$\lambda_i = \beta \pi_i u'(c_{1,i}). \quad (4.10)$$

The Lagrange-multiplier λ_i corresponds to the rate of change of the total utility with respect to $c_{1,i}$ and is again strictly positive, as long as the probability π_i is larger than zero. Otherwise this state can be neglected anyway.

From condition (4.4) follows that also the second boundary condition of the optimization problem (4.1) is binding. Therefore, also the number of units of the risky investment x purchased by the investor is strictly positive. The Lagrange-multiplier μ must be zero according to condition (4.7).

Furthermore, condition (4.8) with $x > 0$ leads to

$$-\lambda_0 p_0 + \sum_{i=1}^N \lambda_i d_{1,i} = 0. \quad (4.11)$$

Applying the equations (4.9) and (4.10) one ultimately gets

$$\begin{aligned} u'(c_0) p_0 &= \beta \sum_{i=1}^N \pi_i u'(c_{1,i}) d_{1,i}, \\ c_0 &= K - x p_0, \\ c_{1,i} &= x d_{1,i}, \quad \forall i. \end{aligned} \quad (4.12)$$

Thus, the optimal investment and consumption strategy is determined by the marginal utility of consumption. Today's consumption is a function of tomorrow's payoffs, state probabilities, the price and the discount factor (Dangl, 2021).

Considering the derivative of the utility function $u'(c) = c^{-\gamma}$, it is possible to transform equation (4.12) as follows:

$$\begin{aligned}
c_0^{-\gamma} p_0 &= \beta \sum_{i=1}^N \pi_i (c_{1,i})^{-\gamma} d_{1,i} \\
\Rightarrow c_0^{-\gamma} &= \beta \sum_{i=1}^N \pi_i (x d_{1,i})^{-\gamma} \frac{d_{1,i}}{p_0} \\
\Rightarrow \left(\frac{c_0}{x p_0} \right)^{-\gamma} &= \beta p_0^{\gamma-1} \sum_{i=1}^N \pi_i d_{1,i}^{1-\gamma} \\
\Rightarrow \frac{x p_0}{c_0} &= \beta^{\frac{1}{\gamma}} p_0^{1-\frac{1}{\gamma}} \left(\sum_{i=1}^N \pi_i d_{1,i}^{1-\gamma} \right)^{\frac{1}{\gamma}}
\end{aligned} \tag{4.13}$$

Hence, the ratio of capital spent on the investment and consumed immediately $\frac{x p_0}{c_0}$ is independent of the amount of initial capital, K , and does only depend on the discount factor, the price, the state probabilities and tomorrow's payoffs of the investment.

A smaller discount factor β causes the investor to invest less into tomorrow, due to the smaller weight placed on the utility gathered at $t = 1$.

The relation between the price p_0 and the spending ratio $\frac{x p_0}{c_0}$ depends on the constant relative risk aversion γ .

For $\gamma = 1$ the spending-proportion is independent of the price. A less risk averse agent ($0 < \gamma < 1$) responds to a higher price by spending less on the investment. An investor with higher relative risk aversion ($\gamma > 1$) responds to a higher price by investing more and consuming less.

The infinitely risk averse agent would always purchase the same number of investment units x , i.e., the spending-ratio is directly proportional to the price p_0 .

In order to model the influence of a dividend change, the dividends in each state are multiplied by a factor k :

$$d_{1,i} = k \cdot \tilde{d}_{1,i}$$

This results in a multiplication of the expected gross return $E(d_{1,i}/p_0)$ by k , as well as the standard deviation. Thus, the higher k , the higher the expected gross return. An increase in k affects the investment decision in two ways:

- (a) With higher return from investment, there is an incentive to consume less today in order to profit more from the higher return.
- (b) With higher return from investment, there is also an incentive to consume more today, because a lower investment is sufficient to have enough consumption tomorrow (Dangl, 2021).

After substitution in equation (4.13) the factor k can be separated and it can be seen that the spending ratio $\frac{xp_0}{c_0}$ is proportional to $k^{\frac{1}{\gamma}-1}$.

$$\begin{aligned} \frac{xp_0}{c_0} &= \beta^{\frac{1}{\gamma}} \cdot p_0^{1-\frac{1}{\gamma}} \cdot \left(\sum_{i=1}^N \pi_i \tilde{d}_{1,i}^{1-\gamma} \right)^{\frac{1}{\gamma}} \cdot k^{\frac{1}{\gamma}-1} \\ \Rightarrow \frac{xp_0}{c_0} &\sim k^{\frac{1}{\gamma}-1} \end{aligned} \quad (4.14)$$

When risk aversion is low ($0 < \gamma < 1$), an increase in dividends has the same effect as a price cut, because both are an increase in gross return: the investor will spend more on the investment and consume less immediately. Effect (a) is dominant.

Equally, when risk aversion is high ($\gamma > 1$), an increase in gross return causes the spending ratio to decrease and more is consumed immediately. Now effect (b) is dominating.

At $\gamma = 1$, the investor is called "myopic" (Dangl, 2021). In this case the investment decision is independent of the expected returns, as well as the price p_0 .

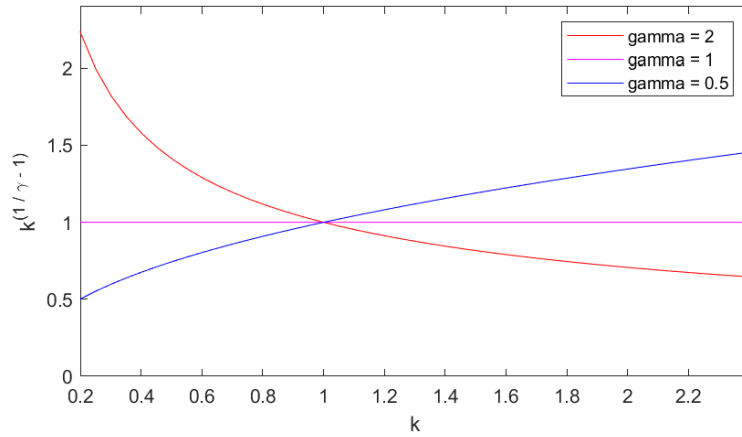


Figure 8: Effect of factor k on the spendings ratio

However, investment behavior is not only determined by the magnitude of the dividends, but also by their variance. I want to illustrate this effect, which is referred to as "mean preserving spread" (Michael Rothschild, 1970), with a simplified model with only two states at $t = 1$, $d_{1,1} \leq d_{1,2}$.

The probability of occurrence of the two states shall be equal $\pi_1 = \pi_2 = 0.5$. The dividends in each state are chosen such that the expected return $\sum_{i=1}^2 \pi_i d_{1,i}$ remains constant at 1. The abscissa is to describe the spread s of the dividends, modelled as $1 - \frac{d_{1,1}}{d_{1,2}}$:

$$d_{1,1} = \frac{2 - 2s}{2 - s}$$

$$d_{1,2} = \frac{2}{2 - s}$$

At $s = 0$ the dividends $d_{1,1}$ and $d_{1,2}$ are equal, there is no spread. As s approaches 1, $d_{1,1}$ converges towards 0, while the expected return must remain constant. Therefore, $d_{1,2}$ approaches 2. At $s = 1$ the spread is at its maximum. Meanwhile, β and p_0 are both fixed at a value of 1.

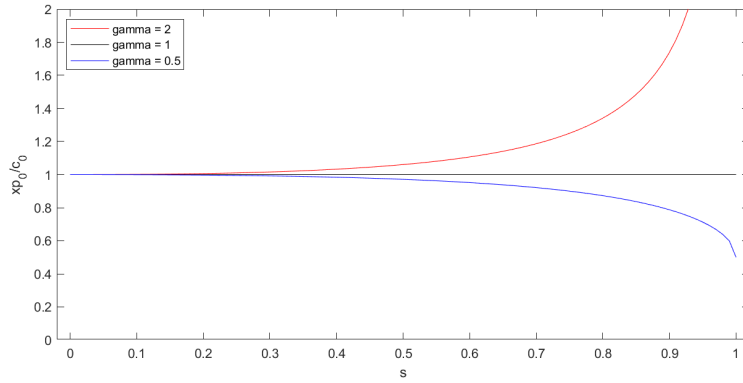


Figure 9: Effect of mean preserving spread on the spending ratio

This diagram shows that the influence of “mean preserving spread” depends on the relative risk aversion γ :

At $\gamma = 1$, the spending ratio once again remains unaffected by the dividends.

An investor with a strong risk aversion ($\gamma > 1$) reacts to an increase in dividend spread with a larger investment portion. This is due to the fact, that the highly risk averse investor fears to be left with little or even no consumption at $t = 1$. A larger investment should offset this risk.

Conversely, when risk aversion is lower than one, a large dividend spread results in more immediate consumption. The slightly risk averse investor does not consider it necessary to hedge against the risk of low returns because the minimum utility is limited (see page 6). More of the capital is spent on immediate consumption because its resulting utility is certain.

An alternative approach to determine the optimum is by substitution of c_0 with $K - xp_0$ in the original optimization problem (4.1):

$$f(x) = u(K - xp_0) + \beta \sum_{i=1}^N \pi_i u(c_{1,i}). \quad (4.15)$$

The model, which originally was a model in c_0 and uncertain c_1 , can now be described with a one-dimensional function in x . The optimum can be found through differentiation and solving for the maximum:

$$\frac{d}{dx}f(x) = -p_0 u'(K - xp_0) + \beta \sum_{i=1}^N \pi_i u'(c_{1,i}) d_{1,i} = 0 \quad (4.16)$$

This leads to the same result as obtained in equation (4.12). Since the objective function is globally concave in x and y and all constraints are linear in these investment decisions, 2nd-order optimality conditions are satisfied, i.e., the determined candidate (x^*, y^*) maximizes the investors utility (Dangl, 2022, p. 131).

It is noticeable that by taking the derivative with respect to xp_0 instead of x , one side of the equation yields exactly the derivative of the total utility function $U(c_0, c_1)$ from equation (4.1) with respect to c_0 :

$$\frac{\partial U}{\partial c_0} = u'(c_0) = MU_{c_0} \quad (4.17)$$

The other side of the equation equals the derivative of the total utility with respect to xp_0 :

$$\frac{\partial U}{\partial (xp_0)} = \beta \sum_{i=1}^N \pi_i u'(c_{1,i}) \frac{d_{1,i}}{p_0} = MU_{xp_0} \quad (4.18)$$

Thus, the marginal utilities with respect to c_0 and xp_0 must be equal, i.e., the increase in utility caused by an additional unit of capital must be the same, whether it is consumed immediately or invested.

Otherwise it would be possible to gain utility by simply transferring a small amount of capital from the investment option with lower marginal utility to the asset with higher marginal utility:

$$\text{if } MU_{c_0} > (<) MU_{xp_0} :$$

$$\begin{aligned} \Delta U &\cong -MU_{xp_0} \cdot \Delta K + MU_{c_0} \cdot \Delta K \\ &= \Delta K \cdot (MU_{c_0} - MU_{xp_0}) > (<) 0. \end{aligned}$$

This would be contrary to the fundamental properties of an optimum. Therefore, in the maximum point the marginal utilities with respect to c_0 and xp_0 must be equal.

Other interesting aspects of this problem can be determined by rearranging equation (4.12) as follows:

$$p_0 = \sum_{i=1}^N \pi_i \beta \frac{u'(c_{1,i})}{u'(c_0)} d_{1,i} \quad (4.19)$$

This equation is the "central asset pricing formula" (Cochrane, 2001, p. 6). The right side is called the expected discounted value of the asset's payoff. If the price does not match this function of payoffs, there exists a change in the investor's investment strategy such that the associated change in consumption would improve the investor's total expected utility. The term

$$m_i = \beta \frac{u'(c_{1,i})}{u'(c_0)} \quad (4.20)$$

is called the stochastic discount factor. It can be thought of as a weight placed on the different possible states, similar to the discount factor β , but state dependent, regarding differences in (marginal) utility.

For example, if there was a risk neutral investor ($\gamma = 0$), then the stochastic discount factor would be equal to β :

$$u'(c) = 1 \quad \Rightarrow \quad m_i = \beta.$$

Therefore, the investor's equilibrium price exactly matches the discounted expected dividend

$$p_0(\gamma = 0) = \beta \sum_{i=1}^N \pi_i d_{1,i}.$$

A risk averse agent has different stochastic discount factors m_i for different states. For a high level of consumption, tomorrow's marginal utility is low (due to saturation). Thus, a lower weight is placed on the states with higher dividends. States with a relatively low level of consumption have a disproportionately high weight (Dangl, 2021).

In other words, to a risk averse investor, the probability of a state i appears higher than it actually is if the dividend $d_{1,i}$ is lower compared to the others. Conversely, states with higher dividend receive lower weight.

4.2 Scenario 2

Compared to scenario 1, now the investor has a second option to invest his capital in order to receive consumption at $t=1$, a riskless investment. Therefore, there are additions in the constraints, the objective function of the optimization problem remains the same:

$$\begin{aligned}
 U(x, y) &= u(c_0) + \beta \cdot \sum_{i=1}^N \pi_i \cdot u(c_{1,i}) \quad \rightarrow \quad \max_{x,y} \\
 \text{subject to} \\
 c_0 &= K - x \cdot p_0 - y \cdot b_0 \\
 c_{1,i} &= x \cdot d_{1,i} + y \cdot 1 \\
 x &\geq 0 \\
 y &\geq 0
 \end{aligned} \tag{4.21}$$

The Lagrangian, which belongs to this optimization problem is

$$\begin{aligned}
 L(\lambda_i, x, \mu_x, y, \mu_y) &= u(c_0) + \beta \cdot \sum_{i=1}^N \pi_i \cdot u(c_{1,i}) + \lambda_0 \cdot (K - xp_0 - yb_0 - c_0) + \\
 &\quad + \sum_{i=1}^N \lambda_i \cdot (xd_{1,i} + y \cdot 1 - c_{1,i}) + \mu_x \cdot x + \mu_y \cdot y \quad (4.22)
 \end{aligned}$$

Next, the first order conditions of optimality are formulated:

$$\frac{\partial L}{\partial \lambda_0} = K - xp_0 - yb_0 - c_0 \geq 0 \quad (4.23a) \quad \frac{\partial L}{\partial \lambda_i} = xd_{1,i} + y - c_{1,i} \geq 0 \quad (4.24a)$$

$$\lambda_0 \geq 0 \quad (4.23b) \quad \lambda_i \geq 0 \quad (4.24b)$$

$$\lambda_0 \cdot \frac{\partial L}{\partial \lambda_0} = 0 \quad (4.23c) \quad \lambda_i \cdot \frac{\partial L}{\partial \lambda_i} = 0 \quad (4.24c)$$

$$\frac{\partial L}{\partial c_0} = u'(c_0) - \lambda_0 \leq 0 \quad (4.25a) \quad \frac{\partial L}{\partial c_{1,i}} = \beta \pi_i u'(c_{1,i}) - \lambda_i \leq 0 \quad (4.26a)$$

$$c_0 \geq 0 \quad (4.25b) \quad c_{1,i} \geq 0 \quad (4.26b)$$

$$c_0 \cdot \frac{\partial L}{\partial c_0} = 0 \quad (4.25c) \quad c_{1,i} \cdot \frac{\partial L}{\partial c_{1,i}} = 0 \quad (4.26c)$$

$$\frac{\partial L}{\partial \mu_x} = x \geq 0 \quad (4.27a) \quad \frac{\partial L}{\partial \mu_y} = y \geq 0 \quad (4.28a)$$

$$\mu_x \geq 0 \quad (4.27b) \quad \mu_y \geq 0 \quad (4.28b)$$

$$\mu_x \cdot \frac{\partial L}{\partial \mu_x} = 0 \quad (4.27c) \quad \mu_y \cdot \frac{\partial L}{\partial \mu_y} = 0 \quad (4.28c)$$

$$\frac{\partial L}{\partial x} = -\lambda_0 p_0 + \sum_{i=1}^N \lambda_i d_{1,i} \leq 0 \quad (4.29a) \quad \frac{\partial L}{\partial y} = -\lambda_0 b_0 + \sum_{i=1}^N \lambda_i \leq 0 \quad (4.30a)$$

$$x \geq 0 \quad (4.29b) \quad y \geq 0 \quad (4.30b)$$

$$x \cdot \frac{\partial L}{\partial x} = 0 \quad (4.29c) \quad y \cdot \frac{\partial L}{\partial y} = 0 \quad (4.30c)$$

Again, due to the satisfied INADA condition, c_0 and $c_{1,i}$ must be strictly positive. From condition (4.25) and $c_0 > 0$ follows that

$$\lambda_0 = u'(c_0), \quad (4.31)$$

which is strictly positive as well. λ_0 is the lagrange multiplier of the budget constraint, i.e., the marginal value of an additional unit of initial wealth. Equation (4.31) can be applied to condition (4.23) and it can be concluded that the first constraint of the optimization problem (4.21) is binding, as already assumed. Similarly, from $c_{1,i} > 0$ one can conclude by the condition (4.26) that

$$\lambda_i = \beta \pi_i u'(c_{1,i}) \quad \Rightarrow \quad \frac{\lambda_i}{\lambda_0} = m_i \cdot \pi_i. \quad (4.32)$$

Since λ_i is also strictly positive, the condition (4.24) can be applied, which leads to the conclusion that also the second boundary condition of the problem (4.21) must be binding.

It is, however, not immediately clear, whether the investor optimally builds a portfolio from both assets (the riskless and the risky asset) or whether she only uses one of those. The case of no investment is technically also an opportunity, but can be ruled out later. The four cases to be examined are the following:

- (I) $x = 0, y = 0$
- (II) $x > 0, y = 0$
- (III) $x = 0, y > 0$
- (IV) $x > 0, y > 0$

Case (I) can be ruled out, since

$$c_{1,i} = x \cdot d_{1,i} + y \cdot 1$$

is binding and $c_{1,i} > 0$.

In case (II) one has

$$c_{1,i} = x \cdot d_{1,i} \quad \text{and} \quad c_0 = K - x \cdot p_0.$$

From condition (4.27) can be concluded that $\mu_x = 0$. Condition (4.29) with equations (4.31) and (4.32) applied leads to

$$u'(c_0) = \beta \sum_{i=1}^N \pi_i u'(c_{1,i}) \frac{d_{1,i}}{p_0}. \quad (4.33)$$

Since $y = 0$, no new conclusions can be drawn from condition (4.28). Condition (4.30) shows that only if b_0 is large, i.e. the riskless rate is low, the investor is willing to hold only the risky asset.

Case (III) works similarly. Here one has

$$c_{1,i} = y \quad \text{and} \quad c_0 = K - y \cdot b_0.$$

$\mu_y = 0$, due to condition (4.28). Conditions (4.27) leads to no new conclusions. Condition (4.29) gives that only if p_0 is high compared to the dividends the investor will only hold the riskless asset.

Through application of equations (4.31) and (4.32), one finds that

$$u'(c_0) = \beta \sum_{i=1}^N \pi_i u'(c_{1,i}) \frac{1}{b_0}. \quad (4.34)$$

Case (IV) is the really interesting case where the prices of both assets are attractive in the sense that the investor optimally builds a portfolio from both.

$$c_{1,i} = x d_{1,i} + y \quad \text{and} \quad c_0 = K - x p_0 - y b_0.$$

Conditions (4.27) and (4.28) lead to $\mu_x = \mu_y = 0$. The evaluation of conditions (4.29) and (4.30) results in

$$\begin{aligned} u'(c_0) &= \beta \sum_{i=1}^N \pi_i u'(c_{1,i}) \frac{d_{1,i}}{p_0} = \\ &= \beta \sum_{i=1}^N \pi_i u'(c_{1,i}) \frac{1}{b_0}. \end{aligned} \quad (4.35)$$

Here, too, it is apparent that each part of this equation corresponds to the derivatives of the total expected utility function (4.21) with respect to c_0 , $x p_0$ and $y b_0$, which are the marginal utilities, respectively:

$$\begin{aligned}
\frac{\partial U}{\partial c_0} &= u'(c_0) = MU_{c_0}, \\
\frac{\partial U}{\partial(xp_0)} &= \beta \sum_{i=1}^N \pi_i u'(c_{1,i}) \frac{d_{1,i}}{p_0} = MU_{xp_0}, \\
\frac{\partial U}{\partial(yb_0)} &= \beta \sum_{i=1}^N \pi_i u'(c_{1,i}) \frac{1}{b_0} = MU_{yb_0}
\end{aligned}$$

$$\Rightarrow MU_{c_0} = MU_{xp_0} = MU_{yb_0}.$$

In case (IV) the marginal utilities must all be equal, such that the investor is indifferent how to invest an additional unit of capital ΔK .

In case (II), only $MU_{c_0} = MU_{xp_0}$ is known. However, MU_{yb_0} cannot be larger than MU_{c_0} and MU_{xp_0} , because then the investor could gain utility by reallocating invested capital from the risky investment to the riskless investment, but this would contradict the properties of an optimum, as already explained on page 12.

MU_{yb_0} can very well be smaller than MU_{c_0} and MU_{xp_0} , since there is no capital that could be shifted from the riskless investment to the risky investment and thus, no possible utility gain.

The same can be concluded from condition (4.30) after applying equations (4.31) and (4.32).

With no share of capital being invested risklessly, the investor is in the same situation as in scenario 1. Thus, the investment-consumption-ratio can again be described as

$$\frac{xp_0}{c_0} = \beta^{\frac{1}{\gamma}} \cdot p_0^{1-\frac{1}{\gamma}} \cdot \left(\sum_{i=1}^N \pi_i d_{1,i}^{1-\gamma} \right)^{\frac{1}{\gamma}}. \quad (4.36)$$

Case (III) works similarly. Here, MU_{c_0} and MU_{yb_0} are equal and MU_{xp_0} must be smaller or equal. This can also be concluded from condition (4.29).

The relation between the share of capital invested risklessly and consumed immediately is

$$\frac{yb_0}{c_0} = \beta^{\frac{1}{\gamma}} \cdot b_0^{1-\frac{1}{\gamma}}, \quad (4.37)$$

which is similar to the relation before: The spending ratio only depends on the discount factor β , the price b_0 and the relative risk aversion γ .

In order to know which case applies, one can use the observations about marginal utilities: For case (II) to apply ($x > 0$, $y = 0$), MU_{yb_0} cannot be larger than MU_{xp_0} .

$$\begin{aligned}
MU_{yb_0} &\leq MU_{xp_0} \\
\beta \sum_{i=1}^N \pi_i u'(c_{1,i}) \frac{1}{b_0} &\leq \beta \sum_{i=1}^N \pi_i u'(c_{1,i}) \frac{d_{1,i}}{p_0}
\end{aligned}$$

Also, consumption at $t = 1$ is only possible through the return of the risky investment, thus, $c_{1,i} = x \cdot d_{1,i}$. Therefore, one gets

$$\begin{aligned} \beta \sum_{i=1}^N \pi_i u'(x d_{1,i}) \frac{1}{b_0} &\leq \beta \sum_{i=1}^N \pi_i u'(x d_{1,i}) \frac{d_{1,i}}{p_0} \\ &\Leftrightarrow \\ \frac{p_0}{b_0} &\leq \frac{\sum_{i=1}^N \pi_i d_{1,i}^{1-\gamma}}{\sum_{i=1}^N \pi_i d_{1,i}^{-\gamma}}. \end{aligned} \tag{4.38}$$

From (4.29a), (4.30a) and (4.32) we see that

$$\begin{aligned} p_0 &= \sum_{i=1}^N \frac{\lambda_i \cdot d_{1,i}}{\lambda_0} = \sum_{i=1}^N \pi_i \cdot m_i \cdot d_{1,i} = E(m d_1), \\ b_0 &\geq \sum_{i=1}^N \frac{\lambda_i}{\lambda_0} \cdot 1 = E(m). \end{aligned}$$

Since only the risky asset is invested, the stochastic discount factor only prices the risky asset. But the asset pricing formula states that “the riskless asset is too expensive”.

(4.38) can be reformulated in terms of the stochastic discount factor m_i :

$$\frac{p_0}{b_0} \leq \frac{\sum_{i=1}^N \pi_i m_i d_{1,i}}{\sum_{i=1}^N \pi_i m_i} = \frac{E(m d_1)}{E(m)}.$$

Similarly, for case (III): Here, MU_{xp_0} cannot be larger than MU_{yb_0} . The investor gets consumption at $t = 1$ only through the return of the riskless investment, therefore, $c_{1,i} = y$.

Again, from (4.29a), (4.30a) and (4.32) we see that

$$\begin{aligned} p_0 &\geq E(m d_1), \\ b_0 &= E(m). \end{aligned}$$

Now only the riskless asset is priced, since the risky asset is not used. This means that “the risky asset is too expensive”. It follows that

$$\begin{aligned} \beta \sum_{i=1}^N \pi_i u'(y) \frac{1}{b_0} &\geq \beta \sum_{i=1}^N \pi_i u'(y) \frac{d_{1,i}}{p_0} \\ &\Leftrightarrow \\ \frac{p_0}{b_0} &\geq \frac{\sum_{i=1}^N \pi_i d_{1,i}}{\sum_{i=1}^N \pi_i} = \sum_{i=1}^N \pi_i d_{1,i} = E(d_1). \end{aligned} \tag{4.39}$$

Thus, it is possible to determine in advance which case applies, based on the prices, tomorrow's payoffs and state probabilities:

$$\begin{aligned}
& \text{case (II)} \quad \frac{p_0}{b_0} \leq \frac{\sum_{i=1}^N \pi_i d_{1,i}^{1-\gamma}}{\sum_{i=1}^N \pi_i d_{1,i}^{-\gamma}} \\
& \Rightarrow \quad x > 0, \quad y = 0 \\
& \text{case (III)} \quad \frac{p_0}{b_0} \geq \sum_{i=1}^N \pi_i d_{1,i} \\
& \Rightarrow \quad x = 0, \quad y > 0 \\
& \text{case (IV)} \quad \frac{\sum_{i=1}^N \pi_i d_{1,i}^{1-\gamma}}{\sum_{i=1}^N \pi_i d_{1,i}^{-\gamma}} < \frac{p_0}{b_0} < \sum_{i=1}^N \pi_i d_{1,i} \\
& \Rightarrow \quad x > 0, \quad y > 0
\end{aligned}$$

In order to find the optimal investment-consumption strategy the investor simply has to determine which case applies. If case (II) or (III) apply, he/she can easily determine the investment-consumption ratio through (4.36), or (4.37), respectively.

At $\gamma = 0$, the statements from equation (4.38) and (4.39) are equal, except for the inequality sign. Thus, the risk neutral investor would consider both investments only if the price ratio is equal to the expected return at $t = 1$. Otherwise, only one investment option will be considered. Then again a one-dimensional optimization problem has to be solved.

If case (IV) applies, equation (4.35) represents the optimality condition. By substitution of

$$c_{1,i} = x d_{1,i} + y \quad \text{and} \quad c_0 = K - x p_0 - y b_0$$

the problem to be solved consists of two equations, which are both two-dimensional. This can be solved by applying numerical methods or with the R-shiny implementation.

Prices are again consistent with the existence of a stochastic discount factor

$$\text{if } x > 0 : \quad p_0 = \sum_{i=1}^N \pi_i \beta \frac{u'(c_{1,i})}{u'(c_0)} d_{1,i} = E(m d_1) \quad (4.40)$$

$$\text{if } y > 0 : \quad b_0 = \sum_{i=1}^N \pi_i \beta \frac{u'(c_{1,i})}{u'(c_0)} \cdot 1 = E(m) \quad (4.41)$$

If multiple investors with different preferences participate in the market, their stochastic discount factors associated with the states of nature may be different. This may

be the case because investors have different risk aversions or different utility functions. Yet they agree on the prices of the financial instruments. This is the case when markets are incomplete ([Dangl, 2021](#)).

5 Conclusion

There exists an investment strategy that the investor would want to choose in order to maximize the total expected utility. Several observations are important to find this optimum:

- Consumption will be strictly positive both today and tomorrow, due to the property of the utility function that the first unit of consumption has infinite marginal value.
- The utility function is also strictly monotonously increasing, therefore the capital and the return on investment are always completely exhausted at the optimum.

- Scenario 2, where the investor has a risky and a riskless investment option, can be divided into four cases: The portfolio is either built from both investment options or from only one of the two. It is also an option to consume the whole capital immediately and not invest anything, however, this can be ruled out for an optimal investment-consumption strategy.

I derive simple conditions which include asset prices, the discount factor, state probabilities and state-contingent asset payoffs that determine uniquely which of the cases applies.

- The optimal investment-consumption strategy leads to equal marginal utilities for immediate consumption and for each investment option exerted, i.e., for an additional unit of capital there must be the same additional utility whether it is invested or immediately consumed.

The marginal utility with respect to an unused investment option is either equal to or smaller than the other marginal utilities.

- If only one investment option is used, the problem to be solved is one-dimensional. The solution can thus be found as a function of the discount factor, the prices, tomorrow's payoffs and state probabilities simply by rearranging the equation for the investment-consumption ratio.

If both investment options are used, the optimality condition consists of two equations, which are both two-dimensional. Numerical methods can be used to obtain a solution.

- With iso-elastic utility, the optimal investment-consumption ratio can be found as a function of the discount factor, the expected gross return and the parameter of relative risk aversion. It is independent of the available capital and is therefore independent of scale.

An increase in gross return leads a highly risk averse agent to invest less and consume more immediately. A slightly risk averse agent would react the other

way around.

A myopic investor makes his investment decision based solely on the discount factor and independently of the gross return.

- The impact of a mean preserving spread on the optimal investment decision depends on relative risk aversion:
Highly risk-averse investors fear low outcomes and try to compensate for them by investing more.
Slightly risk averse investors would consume more immediately, since they have a limited utility minimum.
- For those investments, which are part of the optimal portfolio, the price of the investment at the optimum must be equal to the expected discounted value of the asset's payoff, where discounting is done with the individual stochastic discount factor of the investor. If an asset is not in the optimal portfolio, the stochastic discount factor helps to determine a lower bound to the asset's price. With incomplete markets, the stochastic discount factors of the individual investors do not have to coincide, but they still agree on the prices of the financial instruments.

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List of Figures

1	Problem description	7
2	Utility, scen. 1	8
3	Utility, scen. 2	8
4	Indifference curves, scen. 1	9
5	Indifference curves, scen. 2	9
6	Marginal utilities, scen. 1	9
7	Marginal utilities, scen. 2	9
8	Effect of factor k on the spendings ratio	14
9	Effect of mean preserving spread on the spending ratio	15

Appendix

User Interface Object

```
1 OICD_UI <- dashboardPage(  
2   dashboardHeader(title = "Portfolio Choice"),  
3  
4   dashboardSidebar(  
5     sidebarMenu(  
6  
7       menuItem("Task", tabName = "Task",  
8         icon=icon("list")),  
9       menuItem("Scenario 1", tabName = "Risky_Visualisierung",  
10        icon=icon("chart-line")),  
11       menuItem("Scenario 2", tabName = "Riskless_Visualisierung",  
12        icon=icon("chart-line"))  
13     )  
14   ),  
15  
16   dashboardBody(  
17     withMathJax(),  
18     tags$head(  
19       tags$style(  
20         HTML(  
21           ".MathJax {  
22             font-size: 5pt !important;  
23           }"  
24         )  
25       )  
26     ),  
27     tabItems(  
28  
29 # Aufgabenstellung -----  
30     tabItem(tabName = "Task",  
31       h2("Portfolio-Choice: Optimal Investment-Consumption Decision"),  
32       fluidRow(width=12,  
33         box(width=12,title="Risky investment",  
34           p("A risk averse Investor is endowed with capital  
35             \\(K\\). He/she must now decide on immediate  
36             consumption, \\(c_0\\), and on how much should  
37             optimally be invested in an investment opportunity  
38             that provides a risky dividend which can be  
39             consumed tomorrow. Thus, consumption \\(c_1\\)  
40             at \\(t=1\\) is stochastic."),  
41           p("The investor seeks to maximize expected utility  
42             \\(E[U(c_0, c_1)]\\)."),  
43           p("More specific: The investor has time-seperable  
44             utility with constant relative risk aversion  
45             (CRRA)"),  
46           p("\\(U(c_0, c_1) = u(c_0) + \\beta \\,  
47             E[u(c_1)],\\)"),  
48           tags$ul(  
49             tags$li("\\(c_0, c_1 > 0\\),"),  
50             tags$li("\\(\\beta \\leq 1\\),"),  
51             tags$li("\\(\\gamma > 0\\).")  
52           ),  
53  
54           p("There is no income at \\(t=1\\), so investment  
55             today is the only way to get consumption  
56             tomorrow."),  
57           p("The risky investment has a price \\(p_0\\). The  
58             payoff (dividend) of the investment, however, is  
59             not deterministic."),  
60           p("We assume that at \\(t=1\\), \\(N\\) different
```

```

61         states may occur  $\rightarrow$ 
62          $\pi = [\pi_1, \pi_2, \dots, \pi_N]$ ,
63          $d_1 = [d_{1,1}, d_{1,2}, \dots, d_{1,N}]$ 
64         with  $d_{1,i} > 0$  and
65          $\sum_{i=1}^N \pi_i = 1$ .
66     p("Buying  $x$  units of the risky investment then
67       yields a dividend payment of  $x d_{1,i}$  in
68       state  $i$ ."),
69     p("The investment problem is then"),
70     p( $E[u(c_0, c_1)] = u(c_0) + \beta$ 
71        $\sum_{i=1}^N \pi_i u(c_{1,i}) \rightarrow$ 
72        $\max_{c_0, c_{1,i}}$ ),
73     tags$ul(
74       tags$li(" $c_0 = K - x p_0$ "),
75       tags$li(" $c_{1,i} = x d_{1,i}$ "),
76       tags$li(" $x \geq 0$ "),
77     )
78   ),
79
80   box(width=12, title="Risky and riskless Investment",
81     p("Now we assume that investors have in addition to
82       the risky investment also a riskless investment."),
83     p("Paying a price of  $b_0$  at  $t=0$ , the
84       investor receives a fixed payment of  $1$  in each
85       state at  $t=1$ ."),
86     p("The new optimization problem, which is to maximize
87       by deciding the magnitude of the risky investment,
88        $x$ , and the riskless investment,  $y$ , is
89       then"),
90     p( $E[u(c_0, c_1)] = u(c_0) + \beta$ 
91        $\sum_{i=1}^N \pi_i u(c_{1,i}) \rightarrow$ 
92        $\max_{c_0, c_{1,i}}$ ),
93     tags$ul(
94       tags$li(" $c_0 = K - x p_0 - y b_0$ "),
95       tags$li(" $c_{1,i} = x d_{1,i} + y \cdot 1$ "),
96       tags$li(" $x \geq 0$ "),
97       tags$li(" $y \geq 0$ "),
98     )
99   )
100 )
101 ),
102
103
104 # Risky Visualisierung -----
105   tabItem(tabName = "Risky_Visualisierung",
106     h2("Plot of the total expected utility:"),
107     fluidRow(
108       column(width=3,
109         box(width=NULL, title="Parameters",
110           sliderInput(inputId = "capital_x",
111             label="capital",
112             min = 1,
113             max = 100,
114             step = 1,
115             value = 20),
116           sliderInput(inputId = "price_x",
117             label="price",
118             min = 1,
119             max = 100,
120             step = 1,

```

```

121         value = 19),
122     sliderInput(inputId = "beta_x",
123         label= "discount factor",
124         min = 0,
125         max = 1,
126         step = 0.01,
127         value = 0.9),
128     sliderInput(inputId = "gamma_x",
129         label= "constant relative risk aversion",
130         min = 0.01,
131         max = 5,
132         step = 0.01,
133         value = 2),
134 ),
135
136 box(inputId = "Isoquant_delta", width=NULL,
137
138     sliderInput(inputId = "delta_x",
139         label= "isoquant delta",
140         min = -0.8,
141         max = 0.8,
142         step = 0.1,
143         value = 0)
144 ),
145
146 box(width=NULL,
147     actionButton(inputId= "setToDefault_Parameter_x",
148         label= "reset", width = '40%'),
149         align = "center",
150     ),
151
152 box(inputId = "Investment_box_x",
153     width=NULL, title= "Investment details",
154
155     sliderInput(inputId = "probability1_x",
156         label= "probability1",
157         min = 0,
158         max = 1,
159         step = 0.01,
160         value = 0.2),
161
162     sliderInput(inputId= "dividend1_x",
163         label="dividend1",
164         min = 1,
165         max = 100,
166         step= 1,
167         value = 5),
168
169     sliderInput(inputId = "probability2_x",
170         label= "probability2",
171         min = 0,
172         max = 1,
173         step = 0.01,
174         value = 0.5),
175
176     sliderInput(inputId= "dividend2_x",
177         label="dividend2",
178         min = 1,
179         max = 100,
180         step= 1,

```

```

181         value = 15),
182
183         sliderInput(inputId = "probability3_x",
184                     label= "probability3",
185                     min = 0,
186                     max = 1,
187                     step = 0.01,
188                     value = 0.3),
189
190         sliderInput(inputId= "dividend3_x",
191                     label="dividend3",
192                     min = 1,
193                     max = 100,
194                     step= 1,
195                     value = 25)
196     ),
197     box(width=NULL,
198         actionButton(inputId= "setDefault_investment_x",
199                     label= "reset", width = '40%'),
200                     align = "center",
201     ),
202     box(width=NULL,title="results",
203         htmlOutput("max_x"),
204         htmlOutput("maxZf_x")
205     )
206 ),
207 column(width=9,
208     box(width=NULL,
209         plotOutput(outputId = "PortfolioChoice_x",
210                     height="55vh"))
211     ),
212     column(width=9,
213         box(width=NULL,
214             plotOutput(outputId = "Isoquants_x",height="55vh"))
215     ),
216     column(width=9,
217         box(width=NULL,
218             plotOutput(outputId = "MarginalUtility_x",
219                     height="55vh"))
220     )
221 ),
222 ),
223
224
225 # Riskless Visualisierung -----
226
227 tabItem(tabName = "Riskless_Visualisierung",
228         h2("Plot of the total expected utility:"),
229         fluidRow(
230             column(width=3,
231
232                 box(width=NULL,title= "Parameters",
233                     sliderInput(inputId = "capital_xy",
234                                 label="capital",
235                                 min = 1,
236                                 max = 100,
237                                 step = 1,
238                                 value = 20),
239                     sliderInput(inputId = "price_p0_xy",
240                                 label= "price risky investment",

```

```

241         min = 1,
242         max = 100,
243         step = 1,
244         value = 19),
245     sliderInput(inputId = "price_b0_xy",
246               label= "price riskless investment",
247               min = 0.05,
248               max = 5,
249               step = 0.05,
250               value = 0.95),
251     sliderInput(inputId = "beta_xy",
252               label= "discount factor",
253               min = 0,
254               max = 1,
255               step = 0.01,
256               value = 0.9),
257     sliderInput(inputId = "gamma_xy",
258               label= "constant relative risk aversion",
259               min = 0.01,
260               max = 5,
261               step = 0.01,
262               value = 2),
263 ),
264
265 box(inputId = "Isoquant_delta_xy", width=NULL,
266
267       sliderInput(inputId = "delta_xy",
268                 label= "isoquant delta y",
269                 min = -1,
270                 max = 1,
271                 step = 0.05,
272                 value = 0)
273 ),
274
275 box(width=NULL,
276     actionButton(inputId = "setToDefault_Parameter_xy",
277                 label= "reset", width = '40%'),
278     align = "center",
279 ),
280
281 box(inputId = "Investment_box_xy", width=NULL,
282     title= "Investment details",
283
284     sliderInput(inputId = "probability1_xy",
285                 label= "probability1",
286                 min = 0,
287                 max = 1,
288                 step = 0.01,
289                 value = 0.2),
290
291     sliderInput(inputId= "dividend1_xy",
292                 label="dividend1",
293                 min = 1,
294                 max = 100,
295                 step= 1,
296                 value = 5),
297
298     sliderInput(inputId = "probability2_xy",
299                 label= "probability2",
300                 min = 0,

```

Server Function

```
1 OICD_server <- function(input,output,session) {
2
3 # set to default risky -----
4
5 observeEvent(input$setToDefault_Parameter_x,{
6   updateSliderInput(session, inputId= "capital_x",
7     label="capital",
8     min = 1,
9     max = 100,
10    step = 1,
11    value = 20)
12   updateSliderInput(session, inputId = "price_x",
13     label= "price",
14     min = 1,
15     max = 100,
16     step = 1,
17     value = 19)
18   updateSliderInput(session, inputId = "beta_x",
19     label= "discount factor",
20     min = 0,
21     max = 1,
22     step = 0.01,
23     value = 0.9)
24   updateSliderInput(session, inputId = "gamma_x",
25     label= "constant relative risk aversion",
26     min = 0.01,
27     max = 5,
28     step = 0.01,
29     value = 2)
30   updateSliderInput(session, inputId = "delta_x",
31     label= "isoquant delta",
32     min = -0.8,
33     max = 0.8,
34     step = 0.1,
35     value = 0)
36 })
37
38 observeEvent(input$setToDefault_investment_x,{
39   updateSliderInput(session, inputId = "probability1_x",
40     label= "probability1",
41     min = 0,
42     max = 1,
43     step = 0.01,
44     value = 0.2)
45   updateSliderInput(session, inputId= "dividend1_x",
46     label="dividend1",
47     min = 1,
48     max = 100,
49     step= 1,
50     value = 5)
51   updateSliderInput(session, inputId = "probability2_x",
52     label= "probability2",
53     min = 0,
54     max = 1,
55     step = 0.01,
56     value = 0.5)
57   updateSliderInput(session, inputId= "dividend2_x",
58     label="dividend2",
59     min = 1,
60     max = 100,
```

```

61         step= 1,
62         value = 15)
63
64     updateSliderInput(session, inputId = "probability3_x",
65                       label= "probability3",
66                       min = 0,
67                       max = 1,
68                       step = 0.01,
69                       value = 0.3)
70
71     updateSliderInput(session, inputId= "dividend3_x",
72                       label="dividend3",
73                       min = 1,
74                       max = 100,
75                       step= 1,
76                       value = 25)
77   })
78
79
80 # adjust probabilities risky -----
81
82   observeEvent(input$probability1_x,{
83     if(input$probability1_x + input$probability2_x + input$probability3_x > 1){
84       if(input$probability1_x + input$probability2_x > 1){
85         updateSliderInput(session, inputId = "probability3_x",
86                           value = 0)
87         updateSliderInput(session, inputId = "probability2_x",
88                           value = 1 - input$probability1_x)}
89       else if(input$probability1_x + input$probability2_x < 1){
90         updateSliderInput(session, inputId = "probability3_x",
91                           value = 1 - input$probability1_x - input$probability2_x)}}
92   else if (input$probability1_x + input$probability2_x + input$probability3_x <
93 1){
94     updateSliderInput(session, inputId = "probability3_x",
95                       value = 1 - input$probability1_x - input$probability2_x)}
96   })
97
98   observeEvent(input$probability2_x,{
99     if(input$probability1_x + input$probability2_x + input$probability3_x > 1){
100      if(input$probability1_x + input$probability2_x > 1){
101        updateSliderInput(session, inputId = "probability3_x",
102                          value = 0)
103        updateSliderInput(session, inputId = "probability1_x",
104                          value = 1 - input$probability2_x)}
105      else if(input$probability1_x + input$probability2_x < 1){
106        updateSliderInput(session, inputId = "probability3_x",
107                          value = 1 - input$probability1_x - input$probability2_x)}}
108   else if (input$probability1_x + input$probability2_x + input$probability3_x <
109 1){
110     updateSliderInput(session, inputId = "probability3_x",
111                       value = 1 - input$probability1_x - input$probability2_x)}
112   })
113
114   observeEvent(input$probability3_x,{
115     if(input$probability1_x + input$probability2_x + input$probability3_x > 1){
116       if(input$probability1_x + input$probability3_x > 1){
117         updateSliderInput(session, inputId = "probability2_x",
118                           value = 0)
119         updateSliderInput(session, inputId = "probability1_x",
120                           value = 1 - input$probability3_x)}

```

```

121     else if(input$probability1_x + input$probability3_x < 1){
122         updateSliderInput(session, inputId = "probability2_x",
123             value = 1 - input$probability1_x - input$probability3_x))}
124     else if(input$probability1_x + input$probability2_x + input$probability3_x <
125 1){
126         updateSliderInput(session, inputId = "probability2_x",
127             value = 1 - input$probability1_x - input$probability3_x))
128     })
129
130
131 # import parameters risky -----
132
133 parameters_x <- reactive({
134     par <- list()
135     par$capital <- input$capital_x
136     par$price <- input$price_x
137     par$beta <- input$beta_x
138     par$gamma <- input$gamma_x
139     par$delta <- input$delta_x
140     return(par)
141 })
142
143 probabilities_x <- reactive({
144     probs <- c(input$probability1_x, input$probability2_x, input$probability3_x)
145     return(probs)
146 })
147
148 dividends_x <- reactive({
149     divs <- c(input$dividend1_x, input$dividend2_x, input$dividend3_x)
150     return(divs)
151 })
152
153 x_axis_x <- reactive({
154     scale_x <- 200
155     return(seq(1/scale_x, (scale_x-1)/scale_x, 1/scale_x)))
156
157
158 # find maximal aggregated utility for given parameters risky -----
159
160 MaxExpTotUtility_x <- reactive({
161     max_x.l <- list()
162     par_x.l <- parameters_x()
163     pi.v <- probabilities_x()
164     dl.v <- dividends_x()
165     x.v <- par_x.l$capital/par_x.l$price*x_axis_x()
166     x.d <- c(min(x.v), max(x.v))
167
168     u.f <- function(c, gamma = par_x.l$gamma){
169         if(gamma == 1){return(log(c))}
170         else{return((c^(1-gamma)-1)/(1-gamma))}}
171
172     bE_ul.f <- function(x, p0 = par_x.l$price, beta = par_x.l$beta, pi = pi.v,
173         dl = dl.v){
174         return(beta*(pi[1]*u.f(x*dl[1])+pi[2]*u.f(x*dl[2])+pi[3]*u.f(x*dl[3])))}
175
176     U_x.f <- function(x, K=par_x.l$capital, p0 = par_x.l$price){
177         return(u.f(K - x * p0) + bE_ul.f(x))}
178
179     f.v <- U_x.f(x.v)
180     max_x.l$objective <- max(f.v, na.rm = TRUE)

```



```

181     max_x.l$maximum <- x.v[which.max(f.v)]
182
183     return(max_x.l)
184   })
185
186
187
188 # portfolio choice plot risky -----
189
190 output$PortfolioChoice_x <- renderPlot({
191   max_x.l <- MaxExpTotUtility_x()
192   par_x.l <- parameters_x()
193   pi.v <- probabilities_x()
194   dl.v <- dividends_x()
195
196   x.v <- par_x.l$capital*x_axis_x()
197
198
199   u.f <- function(c, gamma = par_x.l$gamma){
200     if(gamma == 1){return(log(c))}
201     else{return((c^(1-gamma)-1)/(1-gamma))}
202
203   bE_ul.f <- function(xp0, p0 = par_x.l$price, beta = par_x.l$beta, pi = pi.v,
204     dl = dl.v){
205     x <- xp0/p0
206     return(beta*(pi[1]*u.f(x*dl[1])+pi[2]*u.f(x*dl[2])+pi[3]*u.f(x*dl[3]))})
207
208   U_x.f <- function(xp0, K=par_x.l$capital, p0=par_x.l$price){
209     return(u.f(K - xp0) + bE_ul.f(xp0))}
210
211
212
213   U_x.v <- U_x.f(x.v)
214   u.v <- u.f(par_x.l$capital - x.v)
215   bE_ul.v <- mapply(bE_ul.f, x.v)
216
217
218
219
220   fmax <- max(u.f(par_x.l$capital - par_x.l$price*max_x.l$maximum),
221     bE_ul.f(par_x.l$price*max_x.l$maximum),
222     U_x.f(par_x.l$price*max_x.l$maximum))
223   fmin <- min(u.f(par_x.l$capital - par_x.l$price*max_x.l$maximum),
224     bE_ul.f(par_x.l$price*max_x.l$maximum),
225     U_x.f(par_x.l$price*max_x.l$maximum))
226
227
228   xmax <- max(x.v)
229   xmin <- min(x.v)
230   ymax <- fmax + (fmax - fmin)
231   ymin <- fmin - (fmax - fmin)
232
233
234   plot(NA,NA,xaxs="r",yaxs="r",xlim=c(xmin,xmax), ylim=c(ymin, ymax),
235     main="Portfolio Choice", xlab="xp0", ylab="U(c0,c1), u(c0), bE[u(c1)]",
236     yaxt='n')
237
238   legend("topright", c("U(xp0)", "u0", "beta*E[u1]"),
239     fill = c("red", "blue", "green"))
240

```

```

241   lines(x.v, U_x.v, col="red")
242   lines(x.v, u.v, col = "blue")
243   lines(x.v, bE_ul.v, col = "green")
244
245   points(par_x.l$price*max_x.l$maximum, max_x.l$objective)
246
247 })
248
249
250
251 # isoquant plot risky -----
252
253 output$Isoquants_x <- renderPlot({
254   max_x.l <- MaxExpTotUtility_x()
255   par_x.l <- parameters_x()
256   pi.v <- probabilities_x()
257   dl.v <- dividends_x()
258
259   x.v <- par_x.l$capital*x_axis_x()
260
261   maximum <- max_x.l$maximum
262   f_max <- max_x.l$objective
263
264   u.f <- function(c, gamma = par_x.l$gamma){
265     if(gamma == 1){return(log(c))}
266     else{return((c^(1-gamma)-1)/(1-gamma))}
267
268   bE_ul.f <- function(x, p0 = par_x.l$price, beta = par_x.l$beta, pi = pi.v,
269     dl = dl.v){
270     return(beta*(pi[1]*u.f(x/p0*d1[1])+pi[2]*u.f(x/p0*d1[2])+
271       pi[3]*u.f(x/p0*d1[3])))
272
273   U_cx.f <- function(c, xp0, K=par_x.l$capital, p0=par_x.l$price){
274     return(u.f(c) + bE_ul.f(xp0))}
275
276   KK <- function(xp0, K = par_x.l$capital){return(K - xp0)}
277
278   cons_x.f <- function(xp0, K=par_x.l$capital, p0=par_x.l$price,
279     beta=par_x.l$beta, gamma=par_x.l$gamma, f_const = max_x.l$objective){
280     if(gamma == 1){c0 <- exp(f_const - bE_ul.f(xp0))}
281     else{c0 <- ((f_const - bE_ul.f(xp0))*(1-gamma)+1)^(1/(1-gamma))}
282     return(c0)}
283
284   xmax <- 1.2*par_x.l$capital
285   xmin <- 0
286   ymax <- 1.2*par_x.l$capital
287   ymin <- 0
288   plot(NA,NA,xaxs="r",yaxs="r",xlim=c(xmin,xmax), ylim=c(ymin, ymax),
289     main="ISOQUANTS", xlab="xp0", ylab="c0")
290
291   lines(x.v, KK(x.v))
292
293
294
295   # Isoquant for U_max
296
297   f_c <- U_cx.f(c = (par_x.l$capital - max_x.l$maximum*par_x.l$price),
298     xp0 = max_x.l$maximum*par_x.l$price, par_x.l$capital)
299   x.v2 <- x.v
300   cons_x.v <- cons_x.f(x.v, f_const=f_c)

```

```

301 cons_x.v[1: which.max(cons_x.v)] <- NA
302 cons_x.v[cons_x.v > par_x.l$capital] <- NA
303
304 k <- which.min(abs(cons_x.v - par_x.l$capital))
305
306 cons_x.v[k - 1] <- par_x.l$capital
307 x.v2[k-1] <- x.v2[k] + (x.v2[k] - x.v2[k+1])/(cons_x.v[k]
308               - cons_x.v[k+1])*(cons_x.v[k-1] - cons_x.v[k])
309 lines(x.v2, cons_x.v)
310
311
312 # Isoquant for U(delta)
313
314 f_c <- U_cx.f(
315   c = (par_x.l$capital-max_x.l$maximum*par_x.l$price)*(1+par_x.l$delta),
316   xp0 = max_x.l$maximum*par_x.l$price*(1+par_x.l$delta))
317 x.v3 <- x.v
318 cons_x.v <- cons_x.f(x.v, f_const=f_c)
319 cons_x.v[1: which.max(cons_x.v)] <- NA
320 cons_x.v[cons_x.v > par_x.l$capital] <- NA
321 k <- which.min(abs(cons_x.v - par_x.l$capital))
322
323 cons_x.v[k - 1] <- par_x.l$capital
324 x.v3[k-1] <- x.v2[k] + (x.v2[k] - x.v2[k+1])/(cons_x.v[k]
325               - cons_x.v[k+1])*(cons_x.v[k-1] - cons_x.v[k])
326 lines(x.v3, cons_x.v)
327
328
329
330
331 points(max_x.l$maximum*par_x.l$price,
332        par_x.l$capital - max_x.l$maximum*par_x.l$price)
333 legend("topright", c(c("xp0* = ", round(max_x.l$maximum*par_x.l$price, 2)),
334                      c("c0* = ",
335                        round(par_x.l$capital - max_x.l$maximum*par_x.l$price, 2))))
336
337 })
338
339
340
341 # marginal utility plot risky -----
342
343 output$MarginalUtility_x <- renderPlot({
344   max_x.l <- MaxExpTotUtility_x()
345   par_x.l <- parameters_x()
346   pi.v <- probabilities_x()
347   d1.v <- dividends_x()
348
349   x.v <- par_x.l$capital*x_axis_x()
350
351
352   u.f <- function(c, gamma = par_x.l$gamma){
353     if(gamma == 1){return(log(c))}
354     else{return((c^(1-gamma)-1)/(1-gamma))}
355
356   bE_ul.f <- function(x, p0 = par_x.l$price, beta = par_x.l$beta, pi = pi.v,
357                      d1 = d1.v){
358     return(beta*(pi[1]*u.f(x/p0*d1[1])+pi[2]*u.f(x/p0*d1[2])+
359                  pi[3]*u.f(x/p0*d1[3])))
360

```

```

361 U_cx.f <- function(c, xp0, K=par_x.l$capital, p0=par_x.l$price){
362   return(u.f(c) + bE_ul.f(xp0))}
363
364 U_c.f <- function(c, xp0 = max_x.l$maximum*par_x.l$price, K=par_x.l$capital,
365   p0=par_x.l$price){
366   return(u.f(c) + bE_ul.f(xp0))}
367
368 U_x.f <- function(xp0, c = par_x.l$capital-max_x.l$maximum*par_x.l$price,
369   K=par_x.l$capital, p0=par_x.l$price){
370   return(u.f(c) + bE_ul.f(xp0))}
371
372
373 xmax <- par_x.l$capital
374 xmin <- 0
375
376 # fit y_axis limits to the plot
377 p <- max_x.l$maximum*par_x.l$price
378 q <- par_x.l$capital - p
379 x1 <- max(xmin, q - (xmax-xmin)/3, na.rm = TRUE)
380 x2 <- min(xmax, q + (xmax-xmin)/3, na.rm = TRUE)
381 slope <- (U_c.f(q*1.01)-U_c.f(q*0.99))/(q*1.01 - q*0.99)
382 U_c.f1 <- U_c.f(q) + (x1-q)*slope
383 U_c.f2 <- U_c.f(q) + (x2-q)*slope
384 x1 <- max(xmin, p - (xmax-xmin)/3, na.rm = TRUE)
385 x2 <- min(xmax, p + (xmax-xmin)/3, na.rm = TRUE)
386 slope <- (U_x.f(p*1.01)-U_x.f(p*0.99))/(p*1.01 - p*0.99)
387 U_x.f1 <- U_x.f(p) + (x1-p)*slope
388 U_x.f2 <- U_x.f(p) + (x2-p)*slope
389 yaxismax <- max(U_x.f2, U_c.f2)
390 yaxismin <- min(U_x.f1, U_c.f1)
391 yaxisdist <- yaxismax - yaxismin
392
393 ymax <- yaxismax + 0.5*yaxisdist
394 ymin <- yaxismin - 0.5*yaxisdist
395
396 par(mfrow = c(1, 2))
397 plot(NA,NA,xaxs="r",yaxs="r",xlim=c(xmin,xmax), ylim=c(ymin, ymax),
398   main="marginal utility c0", xlab="c0", ylab="", yaxt='n')
399
400 lines(x.v, U_c.f(x.v))
401
402
403 # tangent to immediate consumption part
404 tangent_U_c.f <- function(p){
405   x1 <- max(xmin, p - (xmax-xmin)/3, na.rm = TRUE)
406   x2 <- min(xmax, p + (xmax-xmin)/3, na.rm = TRUE)
407   slope <- (U_c.f(p*1.01)-U_c.f(p*0.99))/(p*1.01 - p*0.99)
408   U_c.f1 <- U_c.f(p) + (x1-p)*slope
409   U_c.f2 <- U_c.f(p) + (x2-p)*slope
410   lines(c(x1, x2), c(U_c.f1, U_c.f2))
411   points(p, U_c.f(p))
412   return(slope)}
413
414 s <- tangent_U_c.f(par_x.l$capital - max_x.l$maximum*par_x.l$price)
415 s <- round(s, abs(log10(s))+1)
416
417 legend("bottomright", c("dU/dc0 = ", s))
418
419
420

```

```

421
422
423
424   plot(NA,NA,xaxs="r",yaxs="r",xlim=c(xmin,xmax),
425         ylim=c(ymin, ymax), main="marginal utility beta*E[u(c1)]",
426         xlab="xp0", ylab="", yaxt='n')
427
428   lines(x.v, U_x.f(x.v))
429
430   # tangent to investment part
431   tangent_U_x.f <- function(p){
432     x1 <- max(xmin, p - (xmax-xmin)/3, na.rm = TRUE)
433     x2 <- min(xmax, p + (xmax-xmin)/3, na.rm = TRUE)
434     slope <- (U_x.f(p*1.01)-U_x.f(p*0.99))/(p*1.01 - p*0.99)
435     U_x.f1 <- U_x.f(p) + (x1-p)*slope
436     U_x.f2 <- U_x.f(p) + (x2-p)*slope
437     lines(c(x1, x2), c(U_x.f1, U_x.f2))
438     points(p, U_x.f(p))
439     return(slope)}
440
441   t <- tangent_U_x.f(max_x.l$maximum*par_x.l$price)
442   t <- round(t, abs(log10(t))+1)
443
444   legend("bottomright", c("dU/dc1 =", t))
445
446
447
448   })
449
450
451
452 # results output risky -----
453
454   output$maxZf_x <- renderText({
455     max_x.l <- MaxExpTotUtility_x()
456     paste("max. exp. Utility: ", round(max_x.l$objective, 3))
457   })
458
459   output$max_x <- renderText({
460     max_x.l <- MaxExpTotUtility_x()
461     paste("x*: ", round(max_x.l$maximum, 3))
462   })
463
464
465
466 # set to default riskless -----
467 -
468
469   observeEvent(input$setToDefault_Parameter_xy,{
470     updateSliderInput(session, inputId= "capital_xy",
471                       label="capital",
472                       min = 1,
473                       max = 100,
474                       step = 1,
475                       value = 20)
476     updateSliderInput(session, inputId = "price_p0_xy",
477                       label= "price risky investment",
478                       min = 1,
479                       max = 100,
480                       step = 1,

```

```

481         value = 19)
482     updateSliderInput(session, inputId = "price_b0_xy",
483         label= "price riskless investment",
484         min = 0.05,
485         max = 5,
486         step = 0.05,
487         value = 0.95)
488     updateSliderInput(session, inputId = "beta_xy",
489         label= "discount factor",
490         min = 0,
491         max = 1,
492         step = 0.01,
493         value = 0.9)
494     updateSliderInput(session, inputId = "gamma_xy",
495         label= "constant relative risk aversion",
496         min = 0.01,
497         max = 5,
498         step = 0.01,
499         value = 2)
500     updateSliderInput(session, inputId = "delta_xy",
501         label= "isoquant delta y",
502         min = -1,
503         max = 1,
504         step = 0.05,
505         value = 0)
506 })
507
508 observeEvent(input$setToDefault_investment_xy,{
509     updateSliderInput(session, inputId = "probability1_xy",
510         label= "probability1",
511         min = 0,
512         max = 1,
513         step = 0.01,
514         value = 0.2)
515     updateSliderInput(session, inputId= "dividend1_xy",
516         label="dividend1",
517         min = 1,
518         max = 100,
519         step= 1,
520         value = 5)
521     updateSliderInput(session, inputId = "probability2_xy",
522         label= "probability2",
523         min = 0,
524         max = 1,
525         step = 0.01,
526         value = 0.5)
527     updateSliderInput(session, inputId= "dividend2_xy",
528         label="dividend2",
529         min = 1,
530         max = 100,
531         step= 1,
532         value = 15)
533
534     updateSliderInput(session, inputId = "probability3_xy",
535         label= "probability3",
536         min = 0,
537         max = 1,
538         step = 0.01,
539         value = 0.3)
540

```

```

541     updateSliderInput(session, inputId= "dividend3_xy",
542                       label="dividend3",
543                       min = 1,
544                       max = 100,
545                       step= 1,
546                       value = 25)
547   })
548
549
550
551 # adjust probabilities riskless -----
552
553   observeEvent(input$probability1_xy,{
554     if(input$probability1_xy+input$probability2_xy+input$probability3_xy > 1){
555       if(input$probability1_xy + input$probability2_xy > 1){
556         updateSliderInput(session, inputId = "probability3_xy", value = 0)
557         updateSliderInput(session, inputId = "probability2_xy",
558                           value = 1 - input$probability1_xy)
559       } else if(input$probability1_xy + input$probability2_xy < 1){
560         updateSliderInput(session, inputId = "probability3_xy",
561                           value = 1 - input$probability1_xy - input$probability2_xy)}
562     } else if(input$probability1_xy+input$probability2_xy+input$probability3_xy <
563 1){
564       updateSliderInput(session, inputId = "probability3_xy",
565                         value = 1 - input$probability1_xy - input$probability2_xy)
566     }
567   })
568
569   observeEvent(input$probability2_xy,{
570     if(input$probability1_xy+input$probability2_xy+input$probability3_xy > 1){
571       if(input$probability1_xy + input$probability2_xy > 1){
572         updateSliderInput(session, inputId = "probability3_xy", value = 0)
573         updateSliderInput(session, inputId = "probability1_xy",
574                           value = 1 - input$probability2_xy)
575       } else if(input$probability1_xy + input$probability2_xy < 1){
576         updateSliderInput(session, inputId = "probability3_xy",
577                           value = 1 - input$probability1_xy - input$probability2_xy)}
578     } else if(input$probability1_xy+input$probability2_xy+input$probability3_xy <
579 1){
580       updateSliderInput(session, inputId = "probability3_xy",
581                         value = 1 - input$probability1_xy - input$probability2_xy)
582     }
583   })
584
585   observeEvent(input$probability3_xy,{
586     if(input$probability1_xy+input$probability2_xy+input$probability3_xy > 1){
587       if(input$probability1_xy + input$probability3_xy > 1){
588         updateSliderInput(session, inputId = "probability2_xy", value = 0)
589         updateSliderInput(session, inputId = "probability1_xy",
590                           value = 1 - input$probability3_xy)
591       } else if(input$probability1_xy + input$probability3_xy < 1){
592         updateSliderInput(session, inputId = "probability2_xy",
593                           value = 1 - input$probability1_xy - input$probability3_xy)}
594     } else if(input$probability1_xy+input$probability2_xy+input$probability3_xy <
595 1){
596       updateSliderInput(session, inputId = "probability2_xy",
597                         value = 1 - input$probability1_xy - input$probability3_xy)
598     }
599   })
600 # import parameters riskless -----

```

```

601
602 parameters_xy <- reactive({
603   par <- list()
604   par$capital <- input$capital_xy
605   par$price <- input$price_p0_xy
606   par$brice <- input$price_b0_xy
607   par$beta <- input$beta_xy
608   par$gamma <- input$gamma_xy
609   par$delta <- input$delta_xy
610   return(par)
611 })
612
613 probabilities_xy <- reactive({
614   probs <- c(input$probability1_xy, input$probability2_xy,
615             input$probability3_xy)
616   return(probs)
617 })
618
619 dividends_xy <- reactive({
620   divs <- c(input$dividend1_xy, input$dividend2_xy, input$dividend3_xy)
621   return(divs)
622 })
623
624 x_axis_xy <- reactive({
625   scale_xy <- 200
626   return(seq(0/scale_xy, (scale_xy - 1)/scale_xy, 1/scale_xy)))
627
628
629
630 # find maximal aggregated utility for given parameters riskless -----
631
632 MaxExpTotUtility_xy <- reactive({
633   max_xy.l <- list()
634   par_xy.l <- parameters_xy()
635   pi.v <- probabilities_xy()
636   dl.v <- dividends_xy()
637   v.v <- par_xy.l$capital*x_axis_xy()
638   v.d <- c(min(v.v), max(v.v))
639
640   u.f <- function(c, gamma = par_xy.l$gamma){
641     if(gamma == 1){return(log(c))}
642     else{return((c^(1-gamma)-1)/(1-gamma))}
643
644   bE_ul.f <- function(xp0, yb0, p0=par_xy.l$price, b0=par_xy.l$brice, pi=pi.v,
645                     dl=dl.v, beta=par_xy.l$beta){
646     return(beta*(pi[1]*u.f(xp0/p0*d1[1]+yb0/b0))+
647            (pi[2]*u.f(xp0/p0*d1[2]+yb0/b0))+
648            (pi[3]*u.f(xp0/p0*d1[3]+yb0/b0)))
649
650   U_xp0yb0.f <- function(xp0, yb0, K=par_xy.l$capital, p0=par_xy.l$price,
651                        b0=par_xy.l$brice){
652     mapply(function(xp, yb){
653       if(K - xp - yb < 0){return(NA)}
654       else{return(u.f(K - xp - yb) + bE_ul.f(xp, yb))}}, xp = xp0, yb = yb0)
655
656   f.m <- outer(v.v, v.v, U_xp0yb0.f)
657
658   max_xy.l$objective <- max(f.m, na.rm = TRUE)
659   max_xy.l$maximum_x <- v.v[which.max(f.m)%length(v.v)]
660   max_xy.l$maximum_y <- v.v[which.max(f.m)%length(v.v)+1]

```



```

661
662
663     return(max_xy.l)
664   })
665
666
667
668 # portfolio choice plot riskless -----
669
670 output$PortfolioChoice_xy <- renderPlot({
671   max_xy.l <- MaxExpTotUtility_xy()
672   par_xy.l <- parameters_xy()
673   pi.v <- probabilities_xy()
674   dl.v <- dividends_xy()
675
676   v.v <- par_xy.l$capital*x_axis_xy()
677   v.d <- c(min(v.v), max(v.v))
678
679   u.f <- function(c, gamma = par_xy.l$gamma){
680     if(gamma == 1){return(log(c))}
681     else{return((c^(1-gamma)-1)/(1-gamma))}
682
683   bE_ul.f <- function(xp0, yb0, p0=par_xy.l$price, b0=par_xy.l$brice, pi=pi.v,
684                       dl=dl.v, beta=par_xy.l$beta){
685     return(beta*(pi[1]*u.f(xp0/p0*d1[1]+yb0/b0))+
686            (pi[2]*u.f(xp0/p0*d1[2]+yb0/b0))+
687            (pi[3]*u.f(xp0/p0*d1[3]+yb0/b0)))
688
689   U_xp0yb0.f <- function(xp0, yb0, K=par_xy.l$capital, p0=par_xy.l$price,
690                          b0=par_xy.l$brice){
691     mapply(function(xp, yb){
692       if(K - xp - yb < 0){return(NA)}
693       else{return(u.f(K - xp - yb) + bE_ul.f(xp, yb))}, xp = xp0, yb = yb0)
694
695   par(mfrow = c(1, 2))
696
697
698
699   U1.v <- U_xp0yb0.f(v.v, max_xy.l$maximum_y)
700
701   if(max_xy.l$maximum_y > 0){
702     U1.v[which.min(U1.v):length(U1.v)] <- NA
703
704   ul.v <- u.f(par_xy.l$capital - v.v - max_xy.l$maximum_y)
705   ul.v[which.min(ul.v):length(ul.v)] <- NA
706
707   bE_ul.v <- mapply(bE_ul.f, v.v, max_xy.l$maximum_y)
708
709
710   fmax <- max(u.f(par_xy.l$capital - max_xy.l$maximum_x - max_xy.l$maximum_y),
711              bE_ul.f(max_xy.l$maximum_x, max_xy.l$maximum_y),
712              U_xp0yb0.f(max_xy.l$maximum_x, max_xy.l$maximum_y))
713   fmin <- min(u.f(par_xy.l$capital - max_xy.l$maximum_x - max_xy.l$maximum_y),
714              bE_ul.f(max_xy.l$maximum_x, max_xy.l$maximum_y),
715              U_xp0yb0.f(max_xy.l$maximum_x, max_xy.l$maximum_y))
716
717   xmax <- max(v.v)
718   xmin <- min(v.v)
719   ymax <- fmax + (fmax - fmin)
720   ymin <- fmin - (fmax - fmin)

```

```

721
722     plot(NA,NA,xaxs="r",yaxs="r",xlim=c(xmin,xmax), ylim=c(ymin, ymax),
723           main=NA, xlab="xp0", ylab=NA, yaxt='n')
724
725     legend("topright", c("U(xp0)", "u0", "beta*E[u1]"),
726           fill = c("red", "blue", "green"))
727
728     lines(v.v, U1.v, col="red")
729     lines(v.v, u1.v,
730           col = "blue")
731     lines(v.v, bE_u1.v, col = "green")
732
733     points(v.v[which.max(U1.v)], max(U1.v, na.rm = TRUE))
734
735
736
737
738
739     U2.v <- U_xp0yb0.f(max_xy.l$maximum_x, v.v)
740
741     if(max_xy.l$maximum_x > 0){
742       U2.v[which.min(U2.v):length(U2.v)] <- NA}
743
744     u2.v <- u.f(par_xy.l$capital - v.v - max_xy.l$maximum_x)
745     u2.v[which.min(u2.v):length(u2.v)] <- NA
746
747     bE_u1.v <- mapply(bE_u1.f, max_xy.l$maximum_x, v.v)
748
749     plot(NA,NA,xaxs="r",yaxs="r",xlim=c(xmin,xmax), ylim=c(ymin, ymax),
750           main=NA, xlab="yb0", ylab="U(c0,c1), u(c0), bE[u(c1)]", yaxt='n')
751
752     legend("topright", c("U(yb0)", "u0", "beta*E[u1]"),
753           fill = c("red", "blue", "green"))
754
755     lines(v.v, U2.v, col="red")
756     lines(v.v, u2.v,
757           col = "blue")
758     lines(v.v, bE_u1.v, col = "green")
759
760     points(v.v[which.max(U2.v)], max(U2.v, na.rm = TRUE))
761
762
763
764     title("PORTFOLIO CHOICE", line = -2, outer=TRUE)
765
766
767
768   })
769
770
771
772 # isoquant plot riskless -----
773
774   output$Isoquants_xy <- renderPlot({
775     max_xy.l <- MaxExpTotUtility_xy()
776     par_xy.l <- parameters_xy()
777     pi.v <- probabilities_xy()
778     dl.v <- dividends_xy()
779
780     v.v <- par_xy.l$capital*x_axis_xy()

```

```

781 v.d <- c(min(v.v), max(v.v))
782
783 u.f <- function(c, gamma = par_xy.l$gamma){
784   if(gamma == 1){return(log(c))}
785   else{return((c^(1-gamma)-1)/(1-gamma))}}
786
787 bE_ul.f <- function(xp0, yb0, p0=par_xy.l$price, b0=par_xy.l$brice, pi=pi.v,
788   d1=d1.v, beta=par_xy.l$beta){
789   return(beta*(pi[1]*u.f(xp0/p0*d1[1]+yb0/b0))+
790     (pi[2]*u.f(xp0/p0*d1[2]+yb0/b0))+
791     (pi[3]*u.f(xp0/p0*d1[3]+yb0/b0))))}
792
793 U_cxp0.f <- function(xp0, yb0=max_xy.l$maximum_y, K=par_xy.l$capital,
794   p0=par_xy.l$price, b0=par_xy.l$brice){
795   mapply(function(xp, yb){
796     if(K - xp - yb < 0){return(NA)}
797     else{return(u.f(K - xp - yb) + bE_ul.f(xp, yb))}},
798     xp = xp0, yb = yb0)}
799
800 KK <- function(xp0, K=par_xy.l$capital, yb0=max_xy.l$maximum_y){
801   return(K - xp0 - yb0)}
802
803 cons_f.f <- function(xp0, yb0=max_xy.l$maximum_y, K=par_xy.l$capital,
804   p0=par_xy.l$price, b0=par_xy.l$brice,
805   beta=par_xy.l$beta, gamma = par_xy.l$gamma,
806   f_const=max_xy.l$objective){
807   if(gamma == 1){
808     c0 <- exp(f_const - bE_ul.f(xp0, yb0))
809   } else{c0 <- ((f_const - bE_ul.f(xp0, yb0))*(1-gamma)+1)^(1/(1-gamma))
810   }
811   return(c0)}
812
813
814 xmax <- 1.2*par_xy.l$capital
815 xmin <- 0
816 ymax <- 1.2*par_xy.l$capital
817 ymin <- 0
818
819 plot(NA, NA, xaxs="r", yaxs="r", xlim=c(xmin, xmax), ylim=c(ymin,ymax),
820   main="ISOQUANTS", xlab="xp0", ylab="c0")
821
822 vK.v <- head(v.v, which.min(abs(KK(v.v))))
823 lines(vK.v, KK(vK.v))
824
825 v.v2 <- v.v
826 cons_f.v <- cons_f.f(v.v2)
827 cons_f.v[1: which.max(cons_f.v)] <- NA
828 cons_f.v[cons_f.v > par_xy.l$capital] <- NA
829
830 k <- which.min(abs(cons_f.v - par_xy.l$capital))
831 cons_f.v[k - 1] <- par_xy.l$capital
832 v.v2[k-1] <- v.v2[k] + (v.v2[k] - v.v2[k+1])/(cons_f.v[k]-cons_f.v[k+1])*
833   (cons_f.v[k-1]-cons_f.v[k])
834
835 lines(v.v2, cons_f.v)
836
837
838
839
840

```

```

841
842 delta_yb0 <- max_xy.l$maximum_y * par_xy.l$delta
843 yb0_new <- max_xy.l$maximum_y * (1 + par_xy.l$delta)
844
845 KK.v <- KK(v.v, K= par_xy.l$capital, yb0 = yb0_new)
846 KK.v[KK.v < 0] <- NA
847 lines(v.v, KK.v, lty = "dashed")
848
849 U_c0xp0.v <- U_c0xp0.f(v.v, yb0 = yb0_new)
850 f_new <- max(U_c0xp0.v, na.rm = TRUE)
851 xp0_new <- which.max(U_c0xp0.v)
852
853 v.v3 <- v.v
854 cons_f2.v <- cons_f.f(v.v, yb0=yb0_new, f_const=f_new)
855
856 cons_f2.v[1: which.max(cons_f2.v)] <- NA
857 cons_f2.v[cons_f2.v > par_xy.l$capital] <- NA
858
859 n <- which.min(abs(cons_f2.v - par_xy.l$capital))
860 cons_f2.v[n - 1] <- par_xy.l$capital
861 v.v3[n-1] <- v.v3[n] + (v.v3[n] - v.v3[n+1])/(cons_f2.v[n]-cons_f2.v[n+1])*
862   (cons_f2.v[n-1]-cons_f2.v[n])
863
864 lines(v.v3, cons_f2.v, lty = "dashed")
865
866 points(max_xy.l$maximum_x,
867        par_xy.l$capital - max_xy.l$maximum_x - max_xy.l$maximum_y)
868
869 legend("topright", c(c("x*p0 =", "c0* =", "", "y*b0 =", "yb0_new =", "",
870   "U* =", "U*_new ="),
871   c(round(max_xy.l$maximum_x, log10(max_xy.l$maximum_x)+3),
872     round(par_xy.l$capital - max_xy.l$maximum_x -max_xy.l$maximum_y,
873   log10(par_xy.l$capital - max_xy.l$maximum_x -max_xy.l$maximum_y)+3),
874   "", round(max_xy.l$maximum_y, log10(max_xy.l$maximum_y)+3),
875   round(yb0_new, log10(abs(yb0_new))+3), "",
876   round(max_xy.l$objective, log10(max_xy.l$objective)+3),
877   round(f_new, log10(f_new)+3))),
878   ncol = 2)
879 })
880
881
882 # marginal utility plot riskless -----
883
884 output$MarginalUtility_xy <- renderPlot({
885   max_xy.l <- MaxExpTotUtility_xy()
886   par_xy.l <- parameters_xy()
887   pi.v <- probabilities_xy()
888   d1.v <- dividends_xy()
889
890   v.v <- par_xy.l$capital*x_axis_xy()
891   v.d <- c(min(v.v), max(v.v))
892
893   u.f <- function(c, gamma = par_xy.l$gamma){
894     if(gamma == 1){return(log(c))}
895     else{return((c^(1-gamma)-1)/(1-gamma))}
896
897   bE_ul.f <- function(xp0, yb0, p0=par_xy.l$price, b0=par_xy.l$brice, pi=pi.v,
898     d1=d1.v, beta=par_xy.l$beta){
899     return(beta*(pi[1]*u.f(xp0/p0*d1[1]+yb0/b0))+
900       (pi[2]*u.f(xp0/p0*d1[2]+yb0/b0))+

```

```

901         (pi[3]*u.f(xp0/p0*d1[3]+yb0/b0)))}
902
903     U_c0.f <- function(c0, xp0=max_xy.l$maximum_x, yb0=max_xy.l$maximum_y,
904                       K=par_xy.l$capital, p0=par_xy.l$price, b0=par_xy.l$brice)
905       {return(u.f(c0) + bE_u1.f(xp0, yb0))}
906
907     U_xp0.f <- function(xp0, yb0=max_xy.l$maximum_y, K=par_xy.l$capital,
908                        p0=par_xy.l$price, b0=par_xy.l$brice)
909       {return(u.f(K - max_xy.l$maximum_x - yb0) + bE_u1.f(xp0, yb0))}
910
911     U_yb0.f <- function(yb0, xp0=max_xy.l$maximum_x, K=par_xy.l$capital,
912                        p0=par_xy.l$price, b0=par_xy.l$brice)
913       {return(u.f(K - max_xy.l$maximum_y - xp0) + bE_u1.f(xp0, yb0))}
914
915     xmax <- par_xy.l$capital
916     xmin <- 0
917
918     # fit y_axis limits to the plot
919     p <- par_xy.l$capital - max_xy.l$maximum_x - max_xy.l$maximum_y
920
921     p1 <- max(xmin, p - (xmax-xmin)/3, na.rm = TRUE)
922     p2 <- min(xmax, p + (xmax-xmin)/3, na.rm = TRUE)
923     slope <- (U_c0.f(p+0.01*xmax)-U_c0.f(p-0.01*xmax))/(0.02*xmax)
924     yaxismax <- U_c0.f(p) + (p2-p)*slope
925     yaxismin <- U_c0.f(p) + (p1-p)*slope
926     yaxisdist <- yaxismax - yaxismin
927     ymax <- yaxismax + yaxisdist
928     ymin <- yaxismin - yaxisdist
929
930     par(mfrow = c(1, 3))
931     plot(NA,NA,xaxs="r",yaxs="r",xlim=c(xmin, xmax), ylim=c(ymin, ymax),
932          main="marginal utility c0", xlab="c0", ylab="", yaxt='n')
933
934     lines(v.v, U_c0.f(v.v))
935
936     # tangent to immediate consumption part
937     tangent_U_c0.f <- function(p){
938       x1 <- max(xmin, p - (xmax-xmin)/3, na.rm = TRUE)
939       x2 <- min(xmax, p + (xmax-xmin)/3, na.rm = TRUE)
940       slope <- (U_c0.f(p+0.01*xmax)-U_c0.f(p-0.01*xmax))/(0.02*xmax)
941       U_c0.f1 <- U_c0.f(p) + (x1-p)*slope
942       U_c0.f2 <- U_c0.f(p) + (x2-p)*slope
943       lines(c(x1, x2), c(U_c0.f1, U_c0.f2))
944       points(p, U_c0.f(p))
945       return(slope)}
946
947     s <- tangent_U_c0.f(par_xy.l$capital-max_xy.l$maximum_x-max_xy.l$maximum_y)
948     s <- round(s, abs(log10(s))+1)
949
950     legend("bottomright", c("dU/dc0 = ", s))
951
952
953
954     plot(NA,NA,xaxs="r",yaxs="r",xlim=c(xmin, xmax), ylim=c(ymin, ymax),
955          main="marginal utility beta*E[u(xp0)]", xlab="xp0", ylab="", yaxt='n')
956
957     lines(v.v, U_xp0.f(v.v))
958
959     # tangent to risky investment part
960     tangent_U_xp0.f <- function(p){

```

```

961     x1 <- max(xmin, p - (xmax-xmin)/3, na.rm = TRUE)
962     x2 <- min(xmax, p + (xmax-xmin)/3, na.rm = TRUE)
963     slope <- (U_xp0.f(p+0.01*xmax)-U_xp0.f(p-0.01*xmax))/(0.02*xmax)
964     U_xp0.f1 <- U_xp0.f(p) + (x1-p)*slope
965     U_xp0.f2 <- U_xp0.f(p) + (x2-p)*slope
966     lines(c(x1, x2), c(U_xp0.f1, U_xp0.f2))
967     points(p, U_xp0.f(p))
968     return(slope)}
969
970     t <- tangent_U_xp0.f(max_xy.l$maximum_x)
971     t <- round(t, abs(log10(t))+1)
972
973     legend("bottomright", c("dU/dxp0 = ", t))
974
975
976     plot(NA,NA,xaxs="r",yaxs="r",xlim=c(xmin, xmax), ylim=c(ymin, ymax),
977          main="marginal utility beta*E[u(yb0)]", xlab="yb0", ylab="", yaxt='n')
978
979     lines(v.v, U_yb0.f(v.v))
980
981     # tangent to riskless investment part
982     tangent_U_yb0.f <- function(p){
983         x1 <- max(xmin, p - (xmax-xmin)/3, na.rm = TRUE)
984         x2 <- min(xmax, p + (xmax-xmin)/3, na.rm = TRUE)
985         slope <- (U_yb0.f(p+0.01*xmax)-U_yb0.f(p-0.01*xmax))/(0.02*xmax)
986         U_yb0.f1 <- U_yb0.f(p) + (x1-p)*slope
987         U_yb0.f2 <- U_yb0.f(p) + (x2-p)*slope
988         lines(c(x1, x2), c(U_yb0.f1, U_yb0.f2))
989         points(p, U_yb0.f(p))
990         return(slope)}
991
992     u <- tangent_U_yb0.f(max_xy.l$maximum_y)
993     u <- round(u, abs(log10(u))+1)
994
995     legend("bottomright", c("dU/dyb0 = ", u))
996 })
997
998
999
1000 # results output riskless -----
1001
1002 output$maxZf_xy <- renderText({
1003     max_xy.l <- MaxExpTotUtility_xy()
1004     paste("max. exp. Utility: ", round(max_xy.l$objective, 3))
1005 })
1006
1007 output$max_xy_x <- renderText({
1008     max_xy.l <- MaxExpTotUtility_xy()
1009     paste("maximum in x*p0: ", round(max_xy.l$maximum_x, 3))
1010 })
1011
1012 output$max_xy_y <- renderText({
1013     max_xy.l <- MaxExpTotUtility_xy()
1014     paste("maximum in y*b0: ", round(max_xy.l$maximum_y, 3))
1015 })
1016 }

```

Call Function

```
1  # Import packages -----  
2  library(shiny)  
3  library(shinydashboard)  
4  
5  # Import UI and server -----  
6  source('OICD_UI.R')  
7  source('OICD_server.R')  
8  
9  # Run application -----  
10 shinyApp(ui = OICD_UI, server = OICD_server)
```