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• Time is discrete and finite with time points n = 0, ..., N. We let:

$$t_0 < t_1 < \ldots < t_N,$$

represent the trading dates: $t_0 = 0$ today and $t_N = N$ the expiry date.

- Finite filtered probability space $(\Omega, \mathcal{F}, \{\mathcal{F}_n\}_{n=0}^N, \mathbb{P})$, i.e finite state space $\Omega = \{\omega_1, ..., \omega_M\}$ (paths).
- d risky assets, S^1, \ldots, S^d . If S^i_n denotes the price at time t_n of the i-th asset:

$$\begin{cases} S_0^i \in \mathbb{R}_+, \\ S_n^i = S_{n-1}^i \left(1 + \mu_n^i\right), \quad n = 1, \dots, N, \end{cases}$$

where μ_n^i is a real random variable that represents the rate of return of the *i*-th asset in the *n*-th period $[t_{n-1}, t_n]$.

• One risk-free asset (bank/money account) B where:

$$\begin{cases}
B_0 = 1, \\
B_n = B_{n-1}(1 + r_n^f), & n = 1, ..., N,
\end{cases}$$

where $r_n^f > -1$ denotes the risk-free rate in the *n*-th period $[t_{n-1}, t_n]$.

Model Set

Model Setup (2/2) (Pascucci & Runggaldier, 2012)

Utility function of class C¹:

$$u: I \to \mathbb{R}, \quad I =]a, \infty[$$
, where $a \le 0$,

strictly increasing, strictly concave.

Define (without consumption):

• Value of a trading strategy $(\alpha, \beta) = (\alpha_n^1, \dots, \alpha_n^d, \beta_n)_{n-1}$ at time n:

$$V_n^{(\alpha,\beta)} = \alpha_n S_n + \beta_n B_n,$$

where α_n^i (resp., β_n) represents the amount of asset S_i (resp., of bond) kept in the portfolio during the n-th period.

 $((\alpha_n, \beta_n)$ composition at $t_{n-1})$.

• The trading strategy (α, β) is self-financing if:

$$V_{n-1} = \alpha_n S_{n-1} + \beta_n B_{n-1}$$

• Proportions of the total value invested in asset *i* through recursive relation:

$$\pi_n^i = \frac{\alpha_n^i S_n^i}{V_{n-1}^i}: \quad V_n \underbrace{=}_{\text{prop. } 1.9 \text{ P&R}} V_{n-1} \left(1 + r_n + \sum_{i=1}^d \pi_n^i (\mu_n^i - r_n)\right)$$

The Dynamic Programming Method (Pascucci & Runggaldier, 2012)

Consider the stochastic process $(V_n)_{n=0,...,N}$ (think value of port.), and suppose it holds recursively that (V_n) depends on the choice of a control process and something stochastic i.e:

$$V_{k} = G_{k} \left(\underbrace{V_{k-1}, \mu_{k}}_{\text{stochastic}}; \underbrace{\eta_{k-1} \left(V_{k-1}\right)}_{\text{control}} \right), \quad k = 1, \dots, N, \tag{1}$$

where:

- μ_1, \ldots, μ_N : d-dim independent stochastic variables (risk factors that drive the dynamics of the asset prices),
- $\eta_0,...,\eta_N$: controls given as $\eta_k:\mathbb{R}\to\mathbb{R}^\ell$ (think strategy (stock fractions)),
- $G_k : \mathbb{R} \times \mathbb{R}^d \times \mathbb{R}^\ell \to \mathbb{R}$ for $k = 1, \dots, N$.

Denote by $(V_k^{n,v})_{k=n,\dots,N}$ the process with initial value $V_n^{n,v}=v$ and for k>n recursively by (1).

Finally, for functions $u_0, \ldots, u_N, u_n : \mathbb{R} \times \mathbb{R}^{\ell} \to \mathbb{R}$ (think utility), define:

$$U^{n,v}(\eta_n,\ldots,\eta_N)=\mathbb{E}\left[\sum_{k=n}^N u_k\left(\underbrace{V_k^{n,v},\eta_k\left(V_k^{n,v}\right)}_{\mathbb{CP}^n}\right)\right]. \tag{*}$$

The Bellman equation

Objective: Optimal expected utility of (\star) over the controls and optimal controls (when they exist) for which the supremum is attained, i.e:

$$\sup_{\eta_0,\ldots,\eta_N} U^{0,v}\left(\eta_0,\ldots,\eta_N\right) = \sup_{\eta_0,\ldots,\eta_N} \mathbb{E}\left[\sum_{k=0}^N u_k\left(V_k^{0,v},\eta_k\left(V_k^{0,v}\right)\right)\right],$$

"If the control is optimal over an entire sequence of periods, then it has to be optimal over each single period". Theorem 2.32 (Pascucci & Runggaldier, 2012, p. 90) DP solves optimization problem (\star) :

Theorem 2.32

For each n = 0, ..., N we have:

$$\sup_{\eta_n,\ldots,\eta_N} U^{n,\nu}(\eta_n,\ldots,\eta_N) = W_n(\nu),$$

where W_n is defined recursively by ([the Bellman equation]):

$$\begin{cases} W_{N}(v) &= \sup_{\xi \in \mathbb{R}^{\ell}} u_{N}(v, \xi), & \text{and, for } n = N, \dots, 1 \\ W_{n-1}(v) &= \sup_{\xi \in \mathbb{R}^{\ell}} \left(u_{n-1}(v, \xi) + \mathbb{E} \left[W_{n} \left(G_{n}(v, \mu_{n}; \xi) \right) \right] \right). \end{cases}$$

([ξ : Imagine we are at time $n-1 \to We$ then have to choose 2 quantities: consum. and inv.])

So: DP is a deterministic algorithm in which at every step we determine (by backwards recursion), the optimal value and the optimal controls by standard maximization procedure!

Example 2.4.2 - Terminal Utility in the Binomial Model: DP Method (1/3)

Objective: (See Extra Slides 2 for numerical example) Maximizing expected utility from terminal wealth where $u(x) = \log(x)$ in a standard binomial model with N periods and riskless interest rate r.

Firstly, dynamics of portolio value:

$$V_n = G_n(V_{n-1}, \mu_n; \pi_n) = \begin{cases} (\mathsf{Up}): & V_{n-1}(1+r+\pi_n(u-1-r)), & \text{if } \mu_n = u-1, \\ (\mathsf{Down}): & V_{n-1}(1+r+\pi_n(d-1-r)), & \text{if } \mu_n = d-1, \end{cases}$$
(2)

where the controls, π_n , denote the fraction of wealth held in the risky asset (*not a percentage*).

If $V_{n-1} > 0$, then $V_n > 0$ if d < 1 + r < u - absence of arbitrage holds, or equivalently:

$$\pi_n \in D =]a, b[$$
, where $a = -\frac{1+r}{n-1-r}$, $b = \frac{1+r}{1+r-d}$.

From theorem 2.32 and Example 2.33 - Maximization of expected utility of terminal wealth (Pascucci & Runggaldier, 2012, pp. 90-91) $\sup_{\pi_{n+1},...,\pi_N} \mathbb{E}\left[u(V_N^{v,n})\right] = W_n(v)$ where:

$$\begin{cases} W_{N}(v) = u(v) & \text{and, for } n = N, \dots, 1 \\ W_{n-1}(v) = \sup_{\bar{\pi}_{n} \in \mathbb{R}^{d}} \mathbb{E}\left[W_{n}\left(G_{n}(v, \mu_{n}; \bar{\pi}_{n})\right)\right]. \end{cases}$$

(W: Think indirect utility: represents the maximal expected utility that can be achieved with given initial wealth)

Example 2.4.2 - Terminal Utility in the Binomial Model: DP Method (2/3)

For initial wealth v > 0 and log-utility:

$$\sup_{\pi_{n+1},\ldots,\pi_N} \mathbb{E}\left(\log(V_N^{n,\nu})\right) = W_n(\nu),$$

where:

$$\begin{cases} W_N(v) &= \log(v) \quad \text{ and, for } n = N-1, \\ W_{N-1}(v) &= \max_{\bar{\pi}_N \in D} \mathbb{E}\left[\log(G_N(v, \mu_N; \bar{\pi}_N))\right] = \underbrace{\log(v) + \max_{\bar{\pi}_N \in D} f(\bar{\pi}_N),}_{\text{product property of log}} \end{cases}$$

where f from (2) (remember, W_{N-1} is the expectation of $V_n = \log(G_n)$:

$$f(\pi) = \underbrace{p}_{\text{UD}} \log \left(1 + r + \pi(u - 1 - r)\right) + \underbrace{\left(1 - p\right)}_{\text{down}} \log \left(1 + r + \pi(d - 1 - r)\right).$$

We find the global maximizer, $\bar{\pi}$, of f:

$$\begin{split} f'(\pi) &= p \frac{u - 1 - r}{1 + r + \pi(u - 1 - r)} + (1 - p) \frac{d - 1 - r}{1 + r + \pi(d - 1 - r)} = 0 \\ \Rightarrow \\ \bar{\pi} &= \frac{(1 + r)(pu + (1 - p)d - 1 - r)}{(u - 1 - r)(1 + r - d)} \in D, \quad \text{ for } p \in]0,1[\text{ and } d < 1 + r < u \\ \Rightarrow \max_{\bar{\pi}_N \in D} f(\bar{\pi}_N) = f(\bar{\pi}), \quad \text{as } \lim_{\pi \to s^+} f(\pi) = \lim_{\pi \to b^-} f(\pi) = -\infty. \end{split}$$

Example 2.4.2 - Terminal Utility in the Binomial Model: DP Method (3/3)

Lastly, substitute $\bar{\pi}$ into $f(\pi)$:

$$\max_{\bar{\pi}_N \in D} f(\pi) = f(\bar{\pi}) = p \log \left(\frac{p(u-d)}{1+r-d}\right) + (1-p) \log \left(\frac{(1-p)(u-d)}{u-1-r}\right) + \log \left(1+r\right).$$

Next step (N-2): The indirect utility function yields (remember, backwards recusion):

$$\begin{split} W_{N-2}(v) &= \underbrace{\max_{\bar{\pi}_{N-1}} E\left[\log(G_{N-1}(v,\mu_{N-1};\bar{\pi}_{N-1}))\right] + \max_{\bar{\pi}_{N-1} \in D} f(\pi)}_{\text{same problem but 2 steps/ticks}} \\ &= \log(v) + 2f(\bar{\pi}), \end{split}$$

i.e. recursively at generic step n:

$$W_{N-n}(v) = \underbrace{\log(v) + nf(\bar{\pi})}_{\text{same problem but } n \text{ steps/ticks}},$$

and specifically the optimal value of expected utility, starting from an initial v > 0 is:

$$W_0(v) = \log(v) + Nf(\bar{\pi}),$$

with a "constant" optimal strategy, $\pi_n^{\text{max}}(v) = \bar{\pi}$.

Implication and Discussion

Constant optimal strategy:

$$\bar{\pi} = \frac{(1+r)(pu+(1-p)d-1-r)}{(u-1-r)(1+r-d)}$$

Result: In any period the optimal strategy involves holding the same **ratio** in the risky asset, independent of stock price, wealth and time horizon ("myopia") and thus the solution for $n \le N-2$ is the same. Same ratio in wealth in every period does **not** imply, expressed in units of the assets kept in the portfolio, **remains constant**.

(same result as MG-method, requires DT!).

By proposition 2 (Blædel & Huge, 1999, p. 12): Can extend the analysis to any utility function with constant RRA.

Question: Should young investors (long time horizon) hold a higher proportion of stocks?

- Yes! Stock returns outperform bonds in the long run (out-performance probability).
- Yes! Future income/salaries is like a big bond investment ⇒ more should be invested in risky asset.
- No! Human capital is maybe not that risk-free unemployment (for months), lower salary after unemployment (stocks relatively uncorrelated between periods), i.e words of Rolf: "really nasty".
- · No! We have rigorous and robost results in dynamic opimization that says our ratio should be constant.

Conclusion: Inconclusive: it is not obvious whatsoever if you should have more in risky assets.

References



Blædel, N., & Huge, H. (1999). Multiperiod investment in discrete time.



Pascucci, A., & Runggaldier, W. J. (2012). Financial mathematics: Theory and problems for multi-period models. Springer Science & Business Media.

Extra Slides 1: Utility Functions

Special case of hyperbolic absolute risk aversion: Isoelastic function for utility / power utility function / CRRA utility function is given by:

$$u(c) = \begin{cases} \frac{c^{1-\gamma}}{1-\gamma} & \text{if } \gamma \neq 1, \gamma \geq 0, \\ \ln(c) & \text{if } \gamma = 1, \end{cases}$$

where c represents consumption and γ is the coefficient of relative risk aversion (CRRA).

Special cases:

- When $\gamma = 0$: This corresponds to risk neutrality, because utility is linear in c.
- When $\gamma = 1$: By virtue of l'Hôpital's rule, the limit of u(c) is $\ln c$ as γ goes to 1:

$$\lim_{\gamma \to 1} \frac{c^{1-\gamma} - 1}{1 - \gamma} = \ln(c),$$

which justifies the convention of using the limiting value $u(c) = \ln c$ when $\gamma = 1$. (Merton portfolio)/Growth optimal portfolio/growth rate maximizing investor \iff that has log-utility.

• When $\gamma \to \infty$: This is the case of infinite risk aversion.

CRRA: This utility function has the feature of constant relative risk aversion. Mathematically this means that:

$$\underbrace{\mathsf{RRA}}_{\mathsf{C}} = -\frac{cu''}{u'}$$

Extra Slide 2: Numerical Experiment (1/3)

PDF: A Numerical Experiment: The Gain from Dynamic Trading and R-code: DynamicStatic Model setup: not implausible numerical values:

- u = 1.29, d = 0.85, p = 0.5 this corresponds to a yearly expected return of 7% with a standard deviation of 22%.
- r = 0.04.
- T=30 it could be your pension savings,

which yields:

$$\hat{\pi} = 0.6568$$

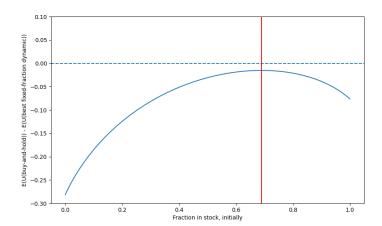
and

$$W_0(v) = \ln(v) + 30 \times 0.04861$$

The expected utility from a buy-and-hold strategy vs. Optimal dynamic trading is plotted on the next slide.

The figure shows $\mathbb{E}(U(\text{buy-and-hold})) - \mathbb{E}(U(\text{best fixed-fraction dynamic}))$ vs. Fraction in stock, initially.

- Full drawn is difference for some fixed fraction, initially.
- Dotted is when both strategies concide.
- Red vertical full drawn line is the optimal $\hat{\pi}_0 \approx 0.69$.
- Below zero everywhere → Dynamic trading can improve even the best buy-and-hold strategy.



Extra Slide 4: Numerical Experiment (3/3)

How large a fraction, say f, of his initial wealth would an investor constrained to buy-and-hold strategies be willing to give up in order to trade dynamicall. We can find the certainty equivalent (when are we indifferent) by solving the equation

$$W_0(\mathbf{v}\times(1-\mathbf{f}))=G(\mathbf{v};\hat{\pi}_0)$$

i.e how much of my wealth am I willing to give up to trade dynamically. The solution will be independent on v as the ln(v)-terms cancel out and is:

$$f = 0.0154 \approx 1.5\%,$$

of his initial wealth in order to be allowed to trade dynamically rather than using the best buy-and-hold strategy. On the one hand not a vanishing number, but on the other hand a gain that could easily be eaten up if there are transaction costs when trading dynamically; about 1/20th of a percent per year would seem to do the trick.