

# Scenarios-based portfolio management: from theory to practice

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## **Hypothetical scenario**

March 2020:

- $\longrightarrow$  COVID-19 fear and equity markets have fallen by 40%
- $\longrightarrow$  But central banks are intervening in markets and providing liquidity and support
- $\longrightarrow$  A portfolio manager identifies two possible scenarios for the next 12 months:
- 1. Equities fall another 20% (30% likelihood)
- 2. Equities rally 50% (70% likelihood)



#### Incorporating scenarios into portfolio construction

How can insight into market scenarios be efficiently incorporated into portfolio management decisions?

- 1. We explore how academic research techniques address this problem
- 2. We consider the practical challenges regarding industry application



#### **Relevant literature**

- 1. Single-period mean-variance efficient portfolio construction
- $\longrightarrow$  Markowitz (1952), Tobin (1958), Sharpe (1963)
- 2. Multi-period utility maximization
  - a. Terminal wealth: Balduzzi and Lynch (1999), Johannes et al. (2014)
  - b. Lifetime income: Samuelson (1969), Merton (1969), Viceira and Campbell (2001)



#### **Relevant literature**

- 3. Regime-switching
- $\rightarrow$  Honda (2003), Graflund and Nilsson (2003), Guidolin and Timmermann (2007)
- 4. Parameter uncertainty
- $\rightarrow$  Barberis (2000), Hoevenaars et al (2014)
- 5. Robust portfolio construction
- $\rightarrow$  Peijnenburg (2011)



### Analysis

Explore the effectiveness of different approaches to acknowledging scenarios. Problem setting:

- $\rightarrow$  Seek to maximise utility of terminal wealth assuming CRRA preferences (Ar = 5)
- $\longrightarrow$ Institutional constraints: no leverage or net short positioning
- $\longrightarrow$ Two assets, each with normal i.i.d. returns

|        | 'Bad' Scenario |            | 'Good' Scenario |            |
|--------|----------------|------------|-----------------|------------|
|        | Exp. Return    | Volatility | Exp. Return     | Volatility |
| Cash   | 0%             | 0%         | 0%              | 0%         |
| Stocks | -20%           | 18%        | 40%             | 18%        |



# **Maximising utility**

Detail objective function

$$U(W_T) = \frac{W_T^{1 - A_R}}{(1 - A_R)}$$

Solution technique:

- Determine distribution to be sampled from
- Sample from this distribution to approximate the distribution of outcomes
- Consider different weights and compare which one maximises expected utility



## Alternative portfolio construction approaches

Four alternative portfolio construction approaches

|             | Approach 1                      | Approach 2                      | Approach 3   | Approach 4                        |
|-------------|---------------------------------|---------------------------------|--|-----------------------------------|
| Description | Use the most likely<br>scenario | Take a mean of the expectations | Account for<br>parameter<br>uncertainty in<br>expected returns | Sample from both<br>distributions |
| Mean        | 50%                             | 29%                             | 29% (p/u: 32.1%)   | N/A                               |
| Volatility  | 18%                             | 18%                             | 18%  | N/A                               |



### **Simulated distributions**





## **Sample summary statistics**

Comparison of sample summary statistics

|                      | Approach 1                      | Approach 2                      | Approach 3   | Approach 4                     |
|----------------------|---------------------------------|---------------------------------|--|--------------------------------|
| Description          | Use the most likely<br>scenario | Take a mean of the expectations | Account for<br>parameter<br>uncertainty in<br>expected returns | Sample from both distributions |
| Mean                 | 50%                             | 29%                             | 29% (p/u: 32.1%)   | N/A                            |
| Volatility           | 18%                             | 18%                             | 18%  | N/A                            |
| Mean (sampled)       | 50%                             | 29%                             | 29%  | 29%                            |
| Volatility (sampled) | 18%                             | 18%                             | 36.8%  | 36.8%                          |



## **Optimal allocations** \_\_\_\_

#### **Optimal allocations**

|                    | Approach 1                      | Approach 2                      | Approach 3   | Approach 4                        |
|--------------------|---------------------------------|---------------------------------|--|-----------------------------------|
| Description        | Use the most likely<br>scenario | Take a mean of the expectations | Account for<br>parameter<br>uncertainty in<br>expected returns | Sample from both<br>distributions |
| Optimal allocation | 100%                            | 100%                            | 42.5%  | 40%                               |



## **Efficiency assessment**

We use CEW (certainty equivalent wealth) to compare outcomes.

|             | Approach 1                      | Approach 2                      | Approach 3   | Approach 4                     |
|-------------|---------------------------------|---------------------------------|--|--------------------------------|
| Description | Use the most likely<br>scenario | Take a mean of the expectations | Account for<br>parameter<br>uncertainty in<br>expected returns | Sample from both distributions |
| CEW         | -80.8%                          | -80.8%                          | -0.04%   | N/A                            |



### **Discussion of results**

- $\rightarrow$  Ignoring information about expected returns is inefficient (Approach 1)
- →Ignoring information about the variability in outcomes is inefficient (Approach 1 and Approach 2)
- →In this case, using parameter uncertainty appears a reasonable proxy to sampling from both distributions
- →Sampling from both distributions is most efficient (if applying a sampling-based solution technique)



## In practice

- →It appears intuitive to make full use of scenario-based information when constructing portfolios
- $\longrightarrow$ In practice there are a range of challenges:
- 1. Large universe of assets
- 2. Potentially more than two scenarios
- 3. Correlation structure required (and possibly correlation scenarios)
- 4. Parametrisation challenges need to derive well-formed parameters for each asset / scenario combination
- 5. Computational challenges curse of dimensionality



## In practice

 $\longrightarrow$ Nonetheless, some interesting reflections:

 $\longrightarrow$ Use scenarios, where available, to sense-check parameter estimates

 $\longrightarrow$  Acknowledge that there can be great uncertainty in estimates of expected return



## Conclusion

- →The academic literature has considered techniques for accommodating scenario-based views on assets
- $\rightarrow$  Ignoring the uncertainty created by divergent scenarios can generate significant utility cost
- $\longrightarrow$ In practice there are many factors which make a pure scenarios-based approach difficult
- →But there is the possibility to integrate some of the academic techniques into industry practice

