

Dimension Reduction in Stochastic Optimal Control

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1 Setting and Goals

- 2 Background, Literature and Motivations
- 3 Methodology
- 4 Preliminary Result

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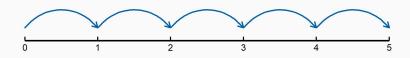
Monopoly



Monopoly

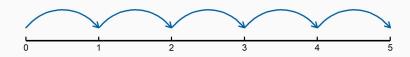
Elements in Monopoly:

- Randomness from the dice
- Moving in each round
- Decision to buy or not to buy the land
- Money to save
- Decision on whether to consume



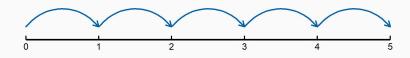
Elements in our setting:

- Randomness from the dice returns of risky assets
- Moving in each round forward in time
- Decision to buy or not how much to buy
- Money to save invest in risk-free asset
- Decision on whether how much to consume



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- Randomness from the dice returns of risky assets
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Difficult in high dimension!

Goals

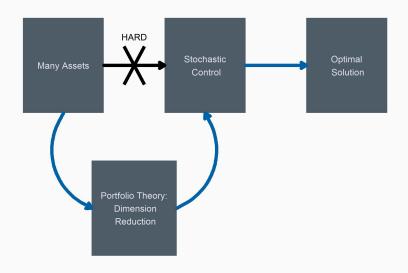
To develop an algorithm that achieves the optimal portfolio selection w.r.t. the objective utility function in an optimal control setting using dimension reduction, where the risky assets are projected into one risky portfolio using linear regression.

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Goals

- To develop an algorithm that achieves the optimal portfolio selection w.r.t. the objective utility function in an optimal control setting using dimension reduction, where the risky assets are projected into one risky portfolio using linear regression.
- To understand the convergence property of the algorithm.

The Big Picture



Significance

- Overcoming the curse of dimensionality.
- Relaxing the requirement on the computational power.
- Filling the gap to utilise the dynamics of portfolio selection in the stochastic control theory.

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Background: Stochastic Control

The classical stochastic optimal control problem has been limited to a small number of states.

$$V_t(\mathbf{x}_t) = \inf_{\{c_j\}_{j=t}^{T-1}} E\left[\sum_{j=t}^{T-1} U(\mathbf{X}_j, \mathbf{c}_j) + U(\mathbf{X}_T)\right]$$

where V_t is the value function at time t; U is the utility function; \mathbf{X}_t is a vector of states variable at time t; c_t is the control variable at time t.

Background: Stochastic Control

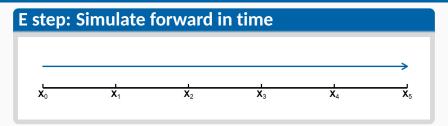
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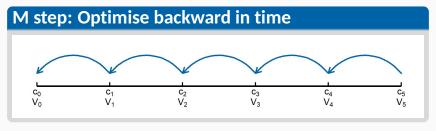
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Bellman Equation

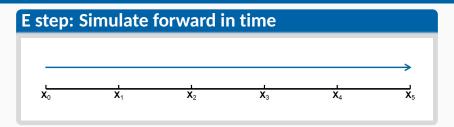
$$V_t(\mathbf{x}) = \inf_{c_t} \left(U(\mathbf{x}, c_t) + E[V_{t+1} | \mathcal{F}_t] \right)$$

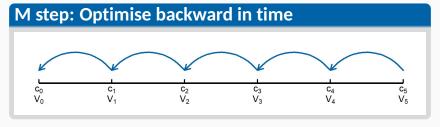
Literature: EM algorithm





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Neglecting the content specific structure

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Methodology: Portfolio Selection

The focus of portfolio selection has been on risk minimization.

$$\min_{\boldsymbol{\omega}^{\mathsf{T}} \mathbf{1} = 1} Var(\boldsymbol{\omega}^{\mathsf{T}} \mathbf{R})$$

where **R** is the return vector; ω is its portfolio allocation vector.

After simple algebra ...

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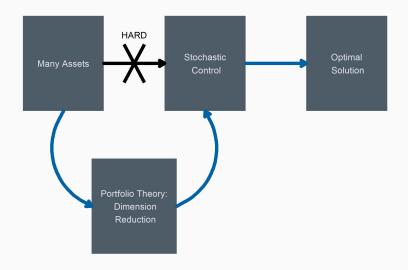
where **R** is the return vector; ω is its portfolio allocation vector.

After simple algebra ... OLS solvable!

$$R_t^r = \omega_t' \mathbf{R}_t$$

Dimension reduced.

The Big Picture

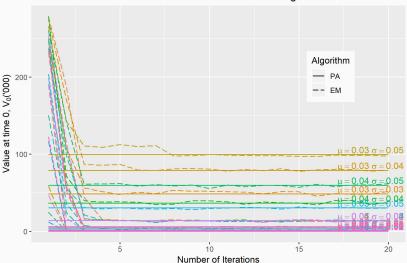


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Single asset without estimation error

The convergence paths of EM algorithm and the optimal value from PA with different means and standard deviations for a single asset



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- Exploring projection options
- Relaxing the independence assumption
- Applying restriction and constraint
- Considering the bound for optimal solution

R package stocon

The package is in development. You can install the development version

```
devtools::install_github(FinYang/stocon)
```

The documentation can be found at

https://pkg.yangzhuoranyang.com/stocon

References i