

# Risk Budgeting Portfolios from Simulations

Rodrigo Targino<sup>1</sup>

School of Applied Mathematics (EMAp), Fundação Getulio  
Vargas (FGV)

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<sup>1</sup>Joint work with Silvana Pesenti (UoT) and Bernardo Costa (UFRJ)

# Introduction

- ▶ An investment concept used in industry, e.g., by pension funds, is **risk budgeting portfolios (RBP)**
  - ▶ **RBP**: diversified portfolios on the level of the **risk contribution** of each asset
  - ▶ **risk contribution**: the added risk of marginally increasing the portfolio's position in that asset
- ▶ Our contribution: to use **stochastic optimization** to construct RBPs based on a generic **risk measure** and **simulations** from the returns.

# The setup

- ▶ At time  $t = 0$ :
  - Assets prices:  $\mathbf{p} = (p_1, \dots, p_d)$
  - The investor holds:  $a_i$  shares of asset  $i$
  - Portfolio value:  $v(\mathbf{a}) = \mathbf{a}^\top \mathbf{p}$
- ▶ At time  $t = 1$ :
  - Assets prices:  $\mathbf{P} = (P_1, \dots, P_d)$
  - The investor holds:  $a_i$  shares of asset  $i$
  - Portfolio value:  $V(\mathbf{a}) = \mathbf{a}^\top \mathbf{P}$
- ▶ **Portfolio loss**:  $L(\mathbf{a}) = -[V(\mathbf{a}) - v(\mathbf{a})] = \sum_{i=1}^d a_i L_i$ 
  - $i$ -th asset loss:  $L_i = -(P_i - p_i)$
- ▶ **Initial endowment**:  $v_0$  (dollars) at time  $t = 0$ .

# Marginal Risk and Risk Contribution

- ▶ **Marginal risk** of asset  $i$  (with coherent risk measure  $\rho$ )

$$\mathcal{MR}_i(\mathbf{a}) = \frac{\partial}{\partial a_i} \rho(\mathbf{a}^\top \mathbf{L})$$

- ▶ **Risk contribution** of asset  $i$

$$\mathcal{RC}_i(\mathbf{a}) = a_i \mathcal{MR}_i(\mathbf{a}).$$

- ▶ For a portfolio  $\mathbf{a}$  and a risk measure  $\rho$ , it holds

$$\rho(L(\mathbf{a})) = \sum_{i=1}^d a_i \mathcal{MR}_i(\mathbf{a}) = \sum_{i=1}^d \mathcal{RC}_i(\mathbf{a}).$$

## The Risk Budgeting Portfolio

- ▶ The investor has a **risk appetite**  $B^\dagger$  split as  $B^\dagger = \sum_{i=1}^d B_i$
- ▶ For a **risk budget**  $\mathbf{B} = (B_1, \dots, B_d)$  the RB portfolio satisfies

$$B^\dagger = \rho(L(\mathbf{a})) \quad \text{and} \quad B_i \mathcal{RC}_j(\mathbf{a}) = B_j \mathcal{RC}_i(\mathbf{a})$$

### Proposition (RB portfolio)

*The RB portfolio's weights are the solution to*

$$\mathbf{a}^* := \operatorname{argmin}_{\mathbf{a}} \rho(L(\mathbf{a})), \text{ subject to } \sum_{i=1}^d B_i \log(a_i) \geq 0,$$

*where the  $a_i$  are implicitly constraint to be strictly positive, so that the log is well-defined.*

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- ▶ With **simulated** returns' scenarios, we can estimate

$$\hat{\rho}(L(\mathbf{a})) \approx \rho(L(\mathbf{a}))$$

and use the **Sample Average Approximation** method

## Proposition (**SAA**-RB portfolio)

The **SAA**-RB portfolio's weights are the solution to

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where the  $a_i$  are implicitly constraint to be strictly positive, so that the log is well-defined.

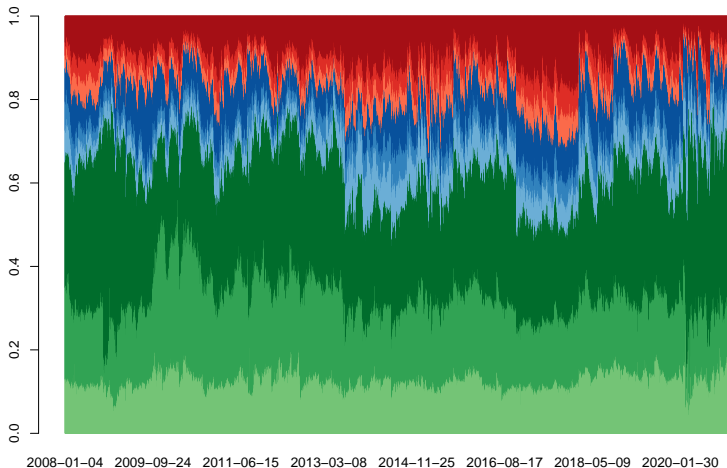
## Example: Multi-class RB portfolio

<b>Ticker</b>	<b>Name</b>	<b>Asset class</b>	<b>RB weight</b>
SPY	SPDR S&P 500 ETF	Equities	2
MDY	SPDR S&P MidCap 400 ETF	Equities	1
VB	Vanguard Small-Cap ETF	Equities	1
GLD	SPDR Gold Shares	Commodities	2
DBC	PowerShares DB Comm. Tracking ETF	Commodities	1
DBA	PowerShares DB Agriculture ETF	Commodities	1
IEF	iShares 7–10 year Treasury Bond	Bonds	2
AGG	iShares Core U.S. Aggregate Bond	Bonds	1
TIP	iShares TIPS Bond	Bonds	1





## Example: Multi-class RB portfolio



**Figure:** ■ SPY, ■ MDY, ■ VB, ■ GLD, ■ DBC, ■ DBA,  
■ IEF, ■ AGG, ■ TIP

## Example: Multi-class RB portfolio

	Vol.	Return	Sharpe	VaR <sub>5%</sub>	Drawdown
grp	5.72%	3.12%	<b>0.545</b>	-0.53%	-17.26%
gmv	5.96%	2.68%	0.450	-0.47%	-18.17%
ew	11.07%	<b>5.11%</b>	0.462	-1.00%	-30.58%
rp095	<b>5.06%</b>	2.71%	0.536	<b>-0.47%</b>	<b>-15.86%</b>

