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## Post-Modern Portfolio Theory

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MPT

Modified MPT

PMPT

# *Post-Modern Portfolio Theory*

Cassy DeBacco

July 27, 2020

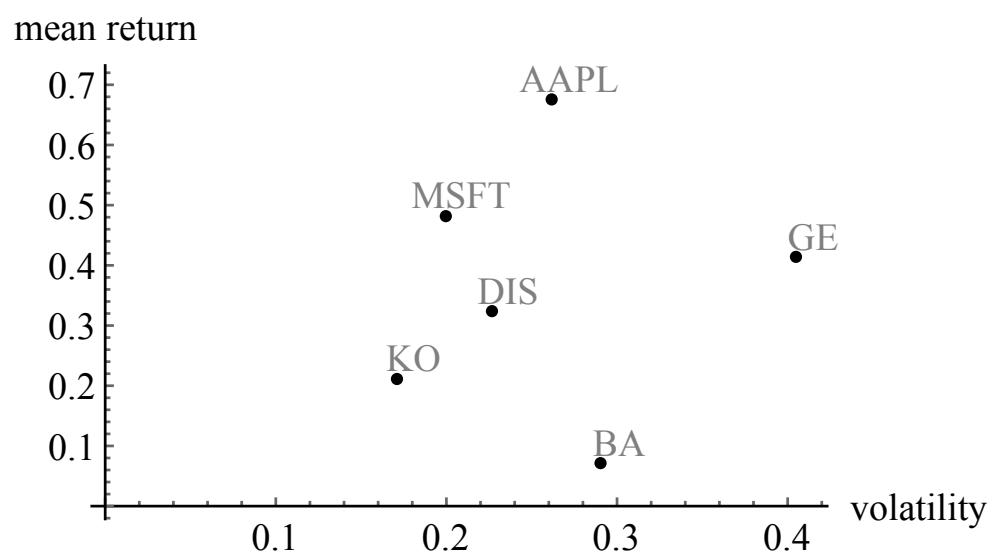


# How Should I Invest My Money?

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# Modern Portfolio Theory

The mean and variance for portfolio

$$\mathcal{P} = w_1 X_1 + \dots + w_n X_n$$

is  $\mu_{\mathcal{P}} = \boldsymbol{\mu}^T \mathbf{w}$  and  $\sigma_{\mathcal{P}}^2 = \mathbf{w}^T \boldsymbol{\Sigma} \mathbf{w}$ .

Modern Portfolio Theory (MPT) solves the quadratic program

$$\begin{aligned} & \text{Minimize } \sigma_{\mathcal{P}}^2 \\ & \text{such that } \mu_{\mathcal{P}} \geq \alpha \\ & \text{and } \mathbf{1}^T \mathbf{w} = 1. \end{aligned}$$



## Solution with Shorting Allowed

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If short sales are allowed, then the optimal weights for a minimum variance portfolio are

$$\mathbf{w} = \frac{1}{\mathbf{1}^T \boldsymbol{\Sigma}^{-1} \mathbf{1}} \boldsymbol{\Sigma}^{-1} \mathbf{1}.$$

The corresponding mean and variance are

$$\mu_{\mathcal{P}} = \frac{\boldsymbol{\mu}^T \boldsymbol{\Sigma}^{-1} \mathbf{1}}{\mathbf{1}^T \boldsymbol{\Sigma}^{-1} \mathbf{1}} \quad \text{and} \quad \sigma_{\mathcal{P}}^2 = \frac{1}{\mathbf{1}^T \boldsymbol{\Sigma}^{-1} \mathbf{1}}.$$

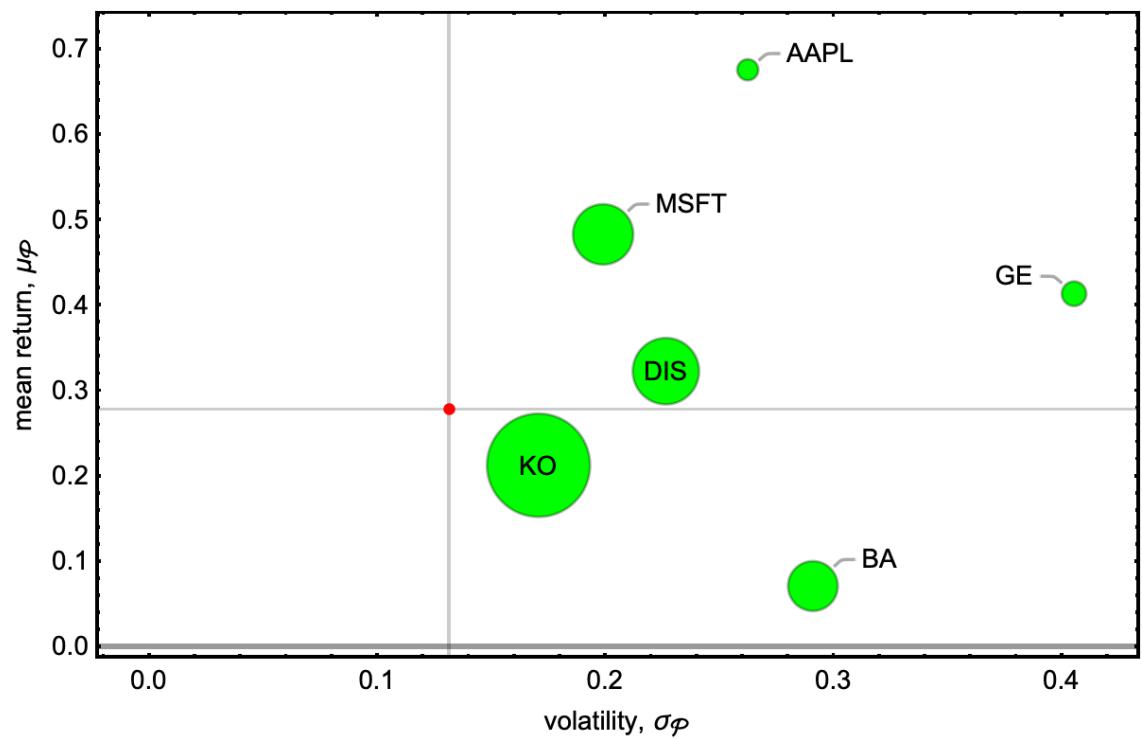


## Minimum Variance Portfolio

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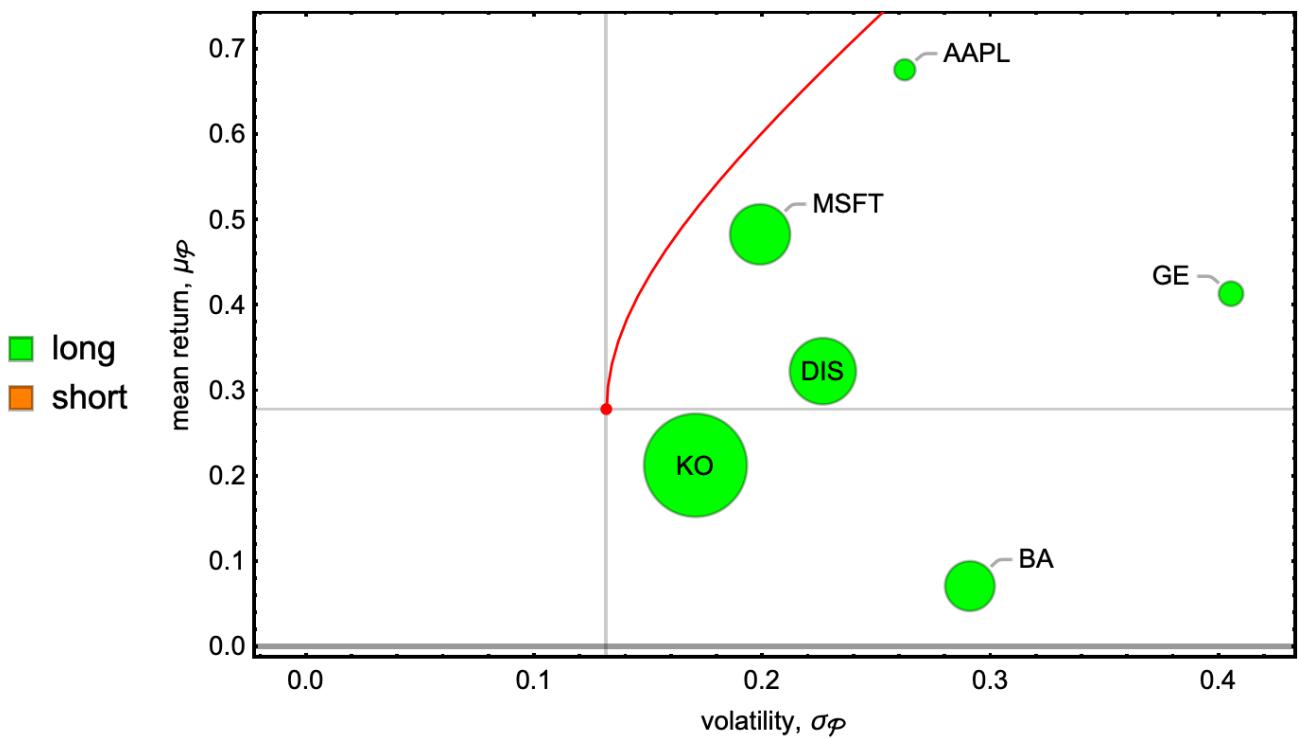
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## Efficient Frontier

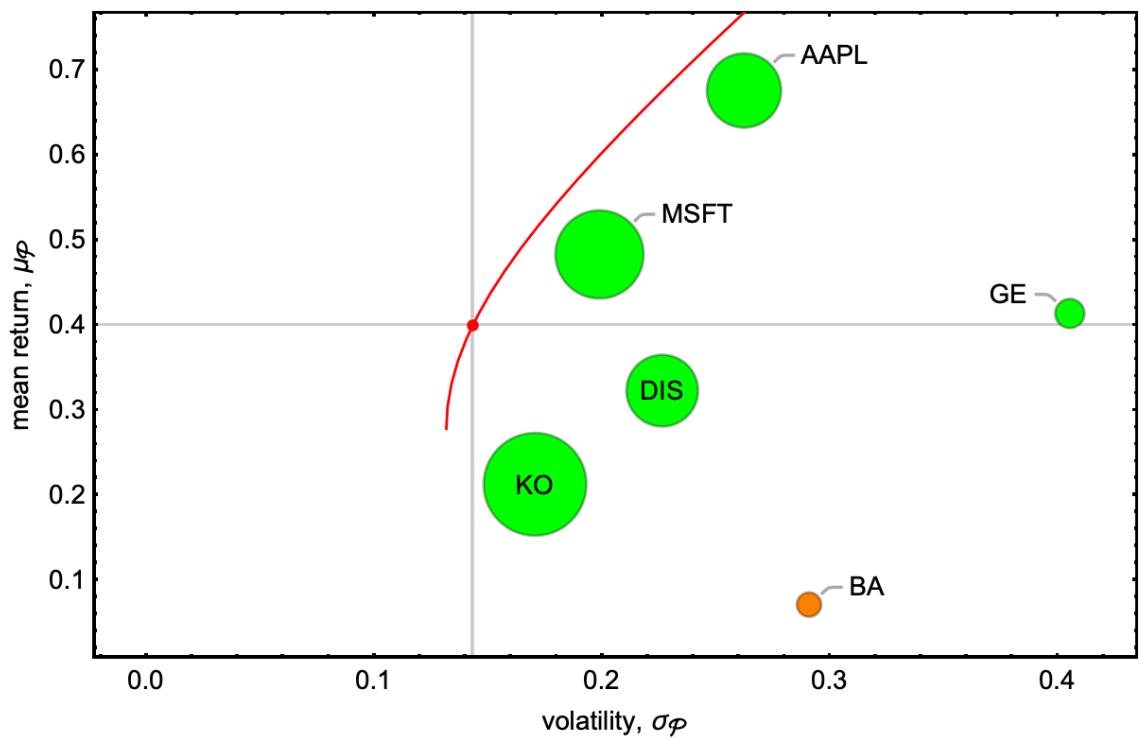
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## Efficient Frontier

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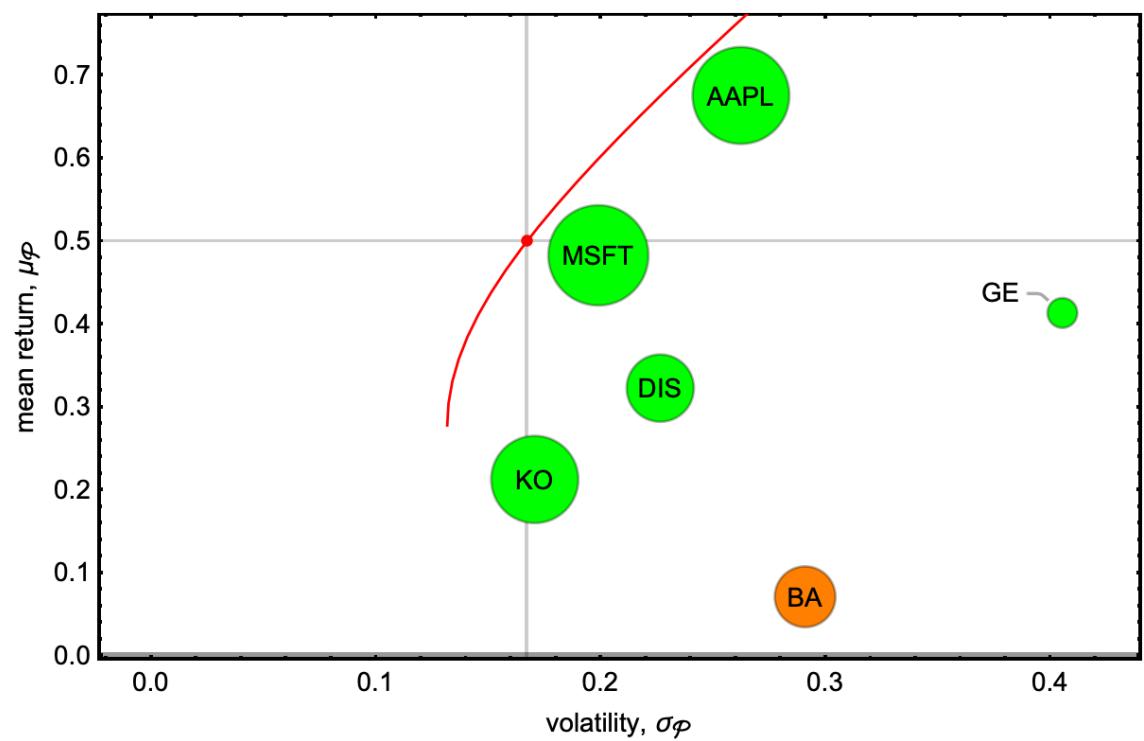


# Efficient Frontier

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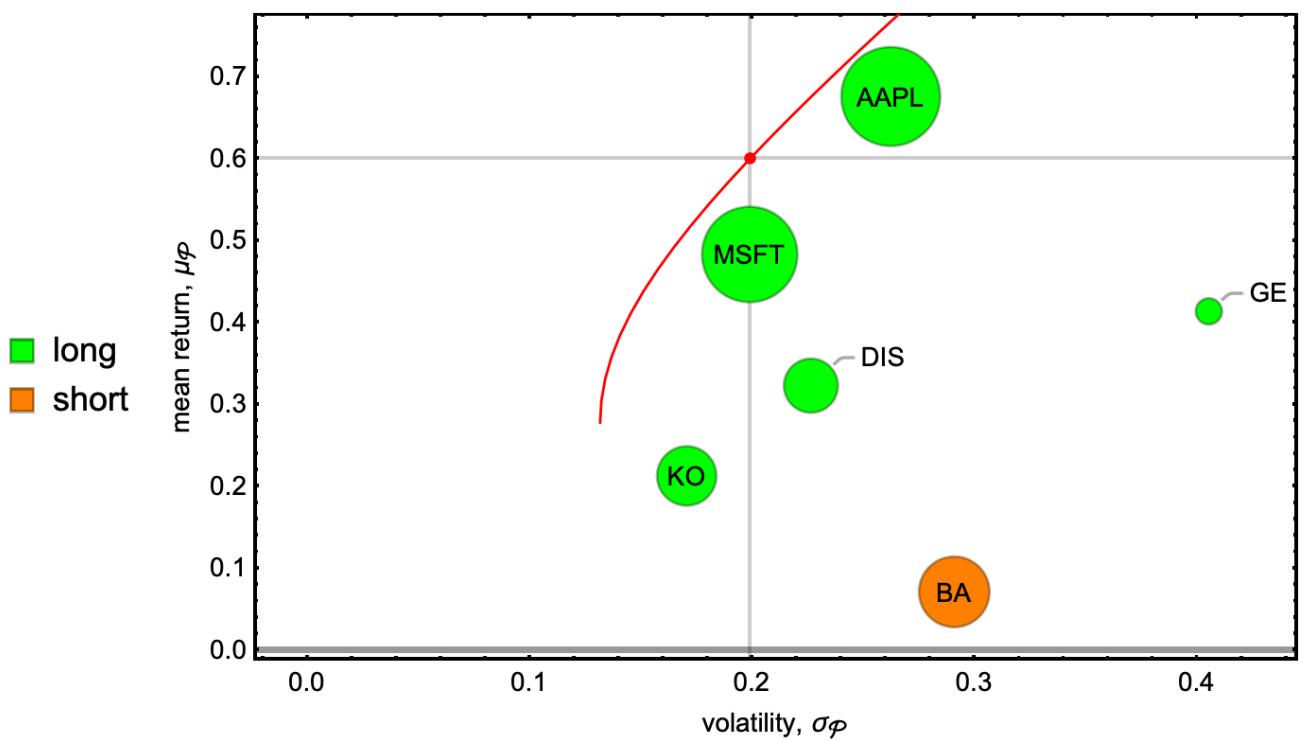


## Efficient Frontier

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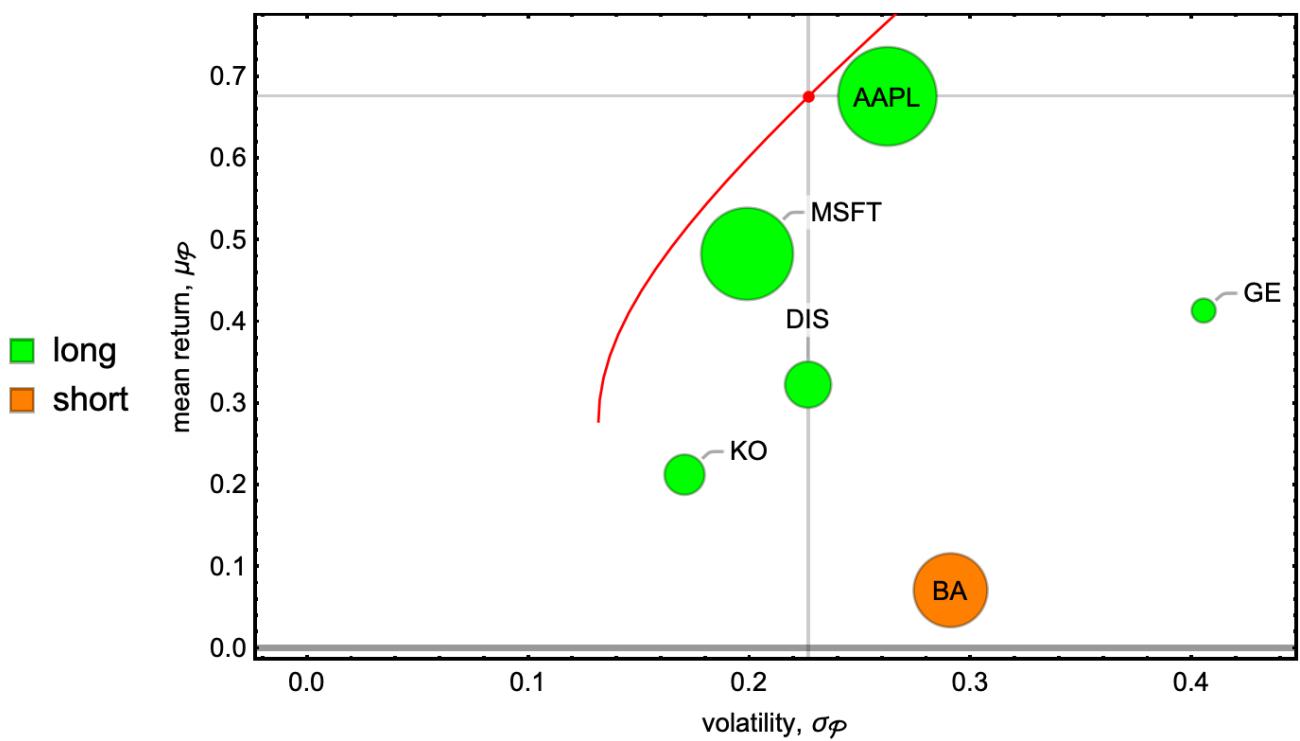
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## Efficient Frontier

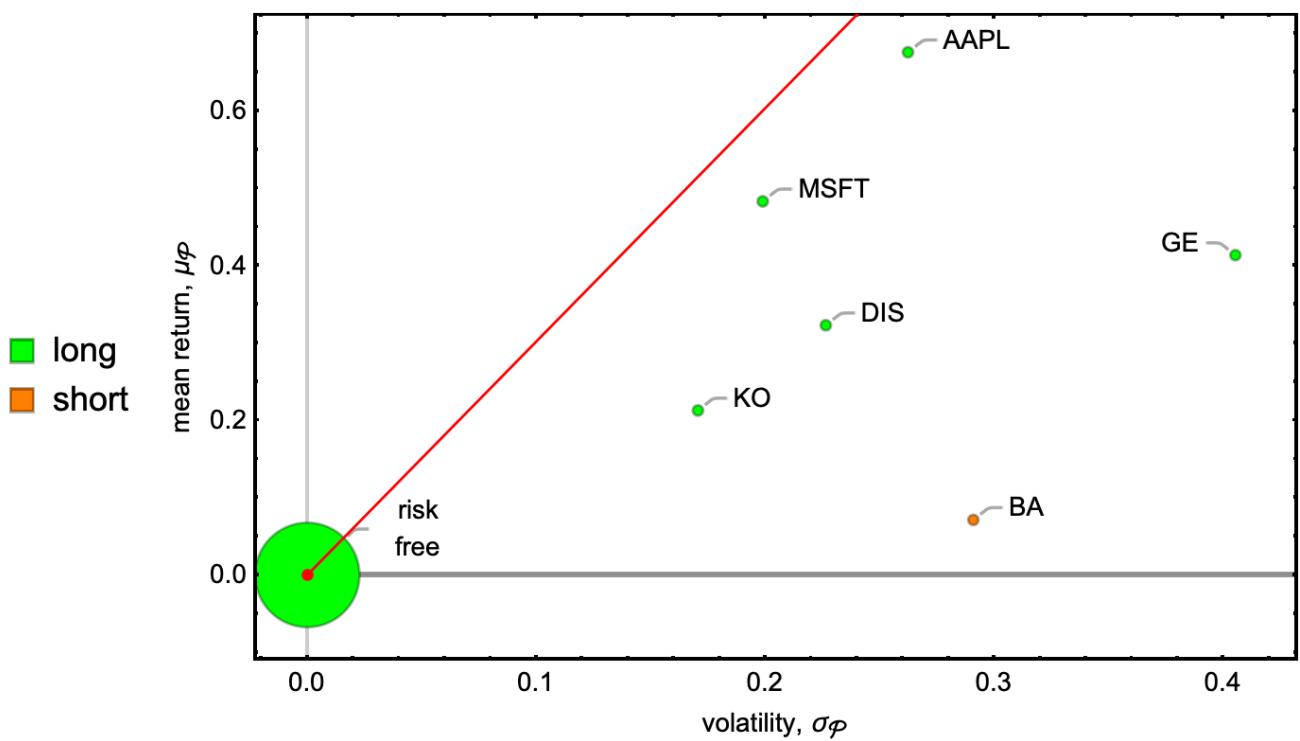
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## Efficient Frontier with a Risk-Free Asset

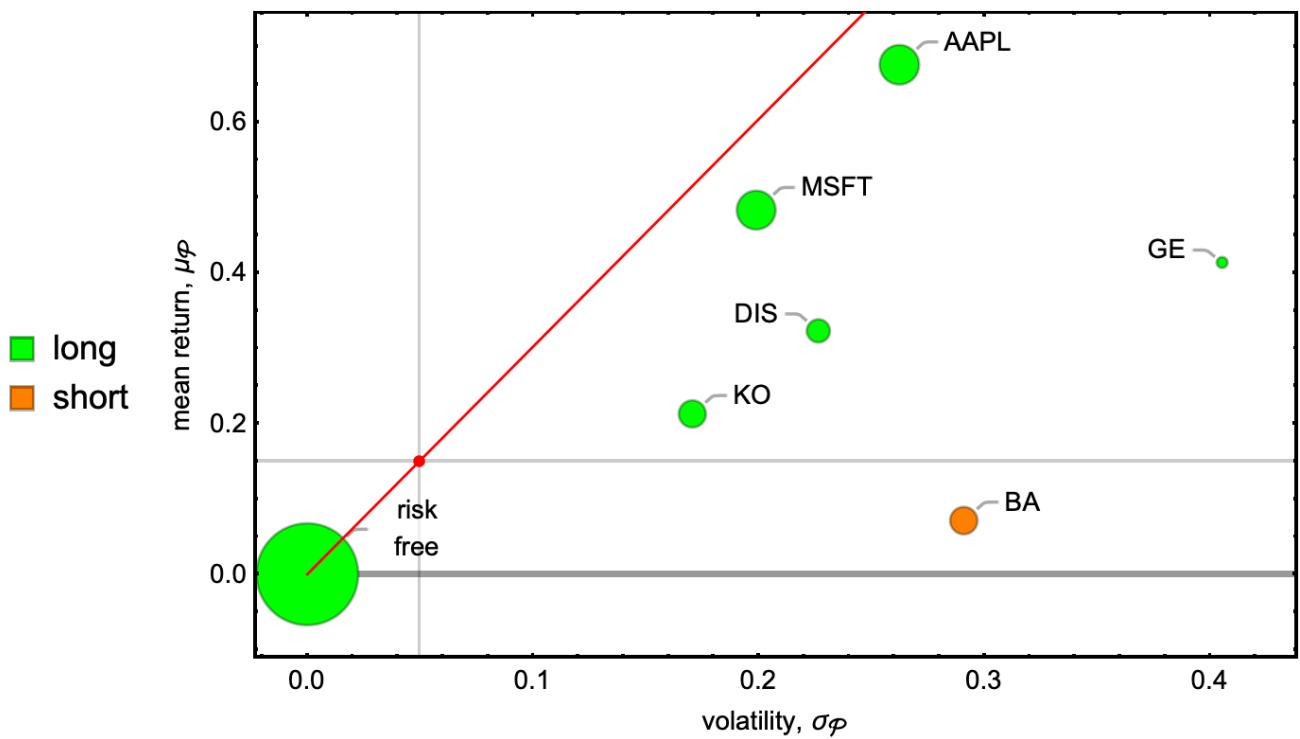
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## Efficient Frontier with a Risk-Free Asset

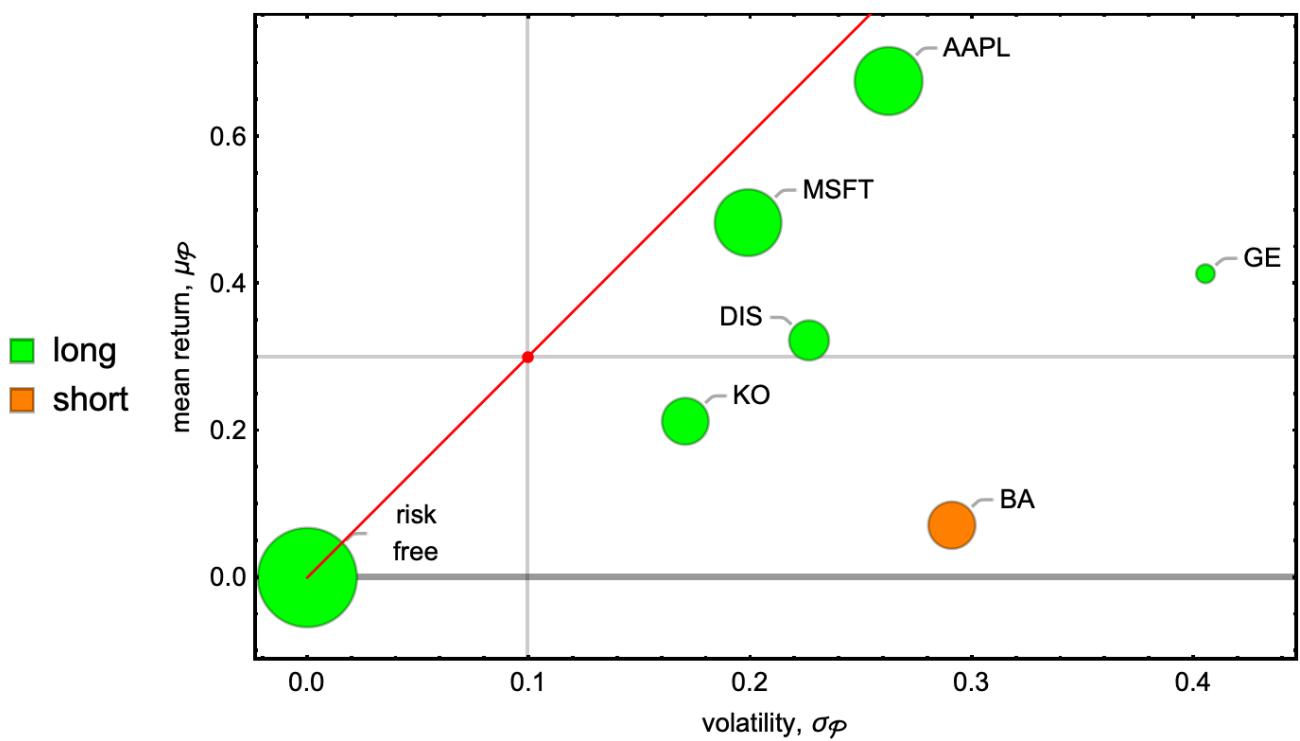
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## Efficient Frontier with a Risk-Free Asset

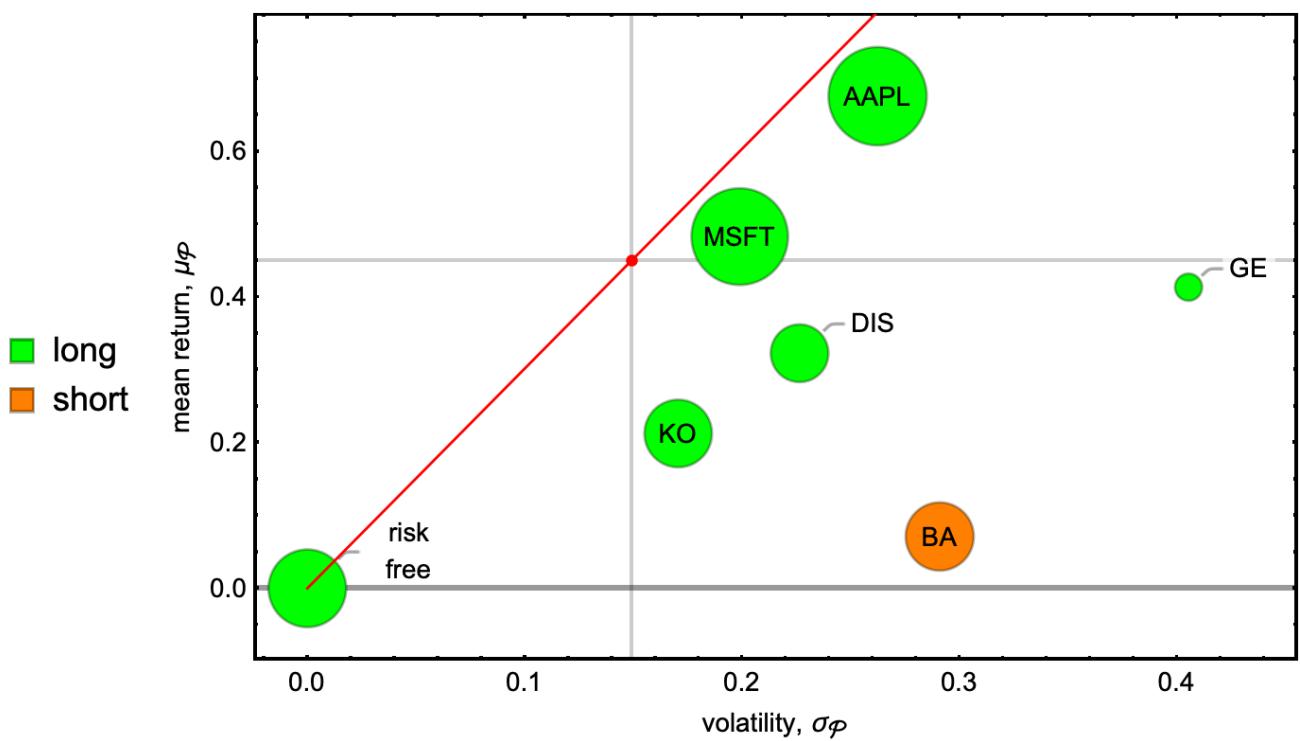
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## Efficient Frontier with a Risk-Free Asset

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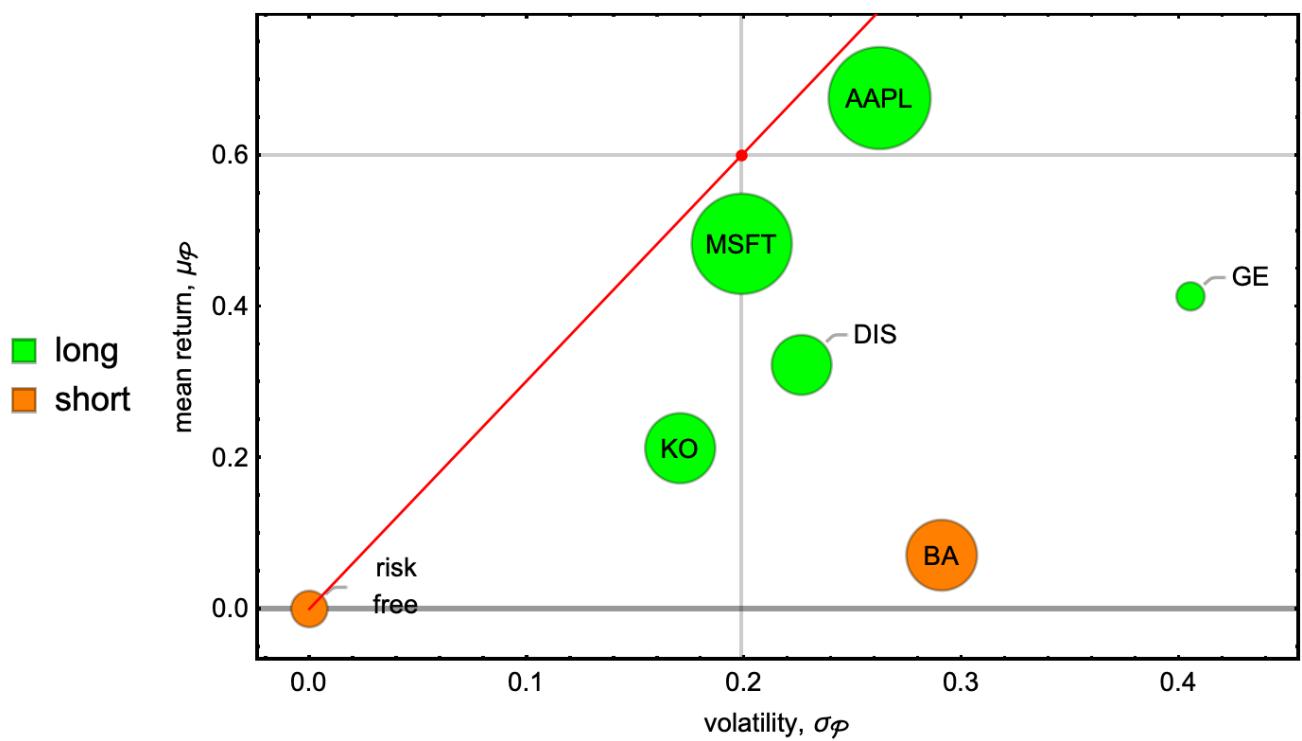


## Efficient Frontier with a Risk-Free Asset

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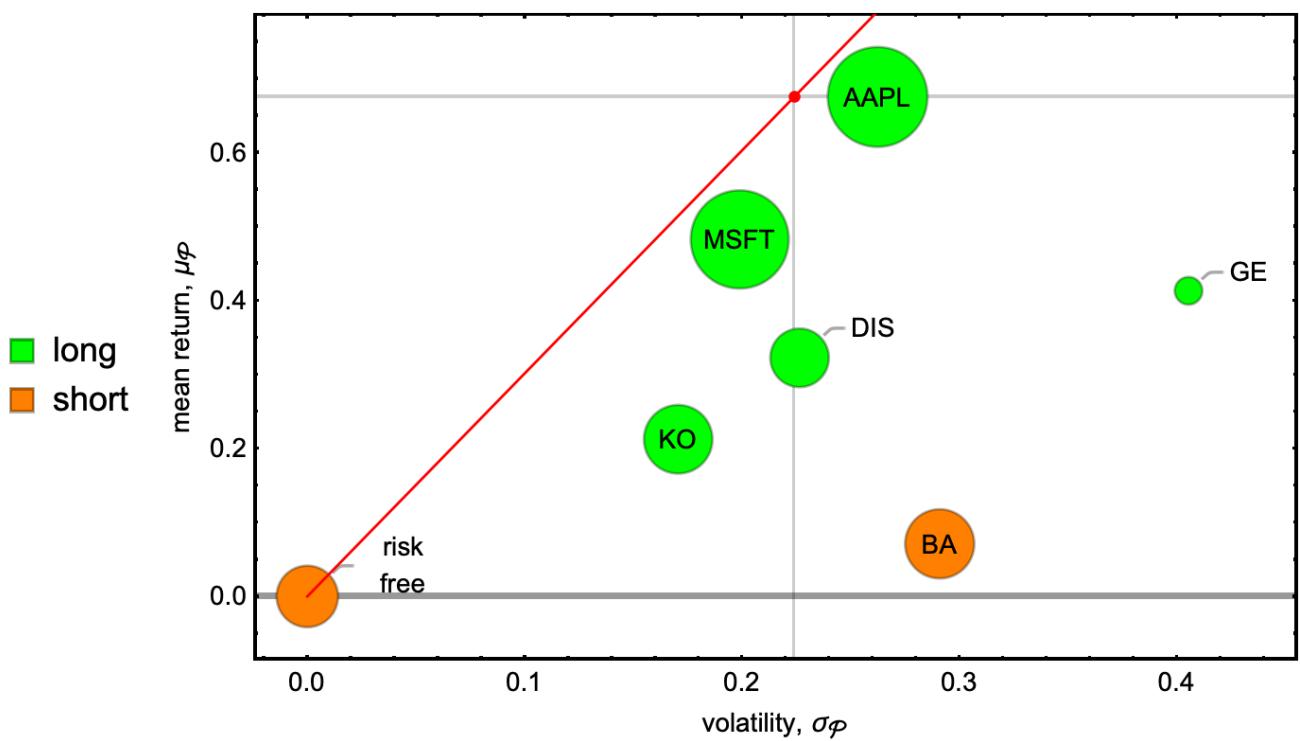
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## Efficient Frontier with a Risk-Free Asset

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## Some Shortcomings of MPT

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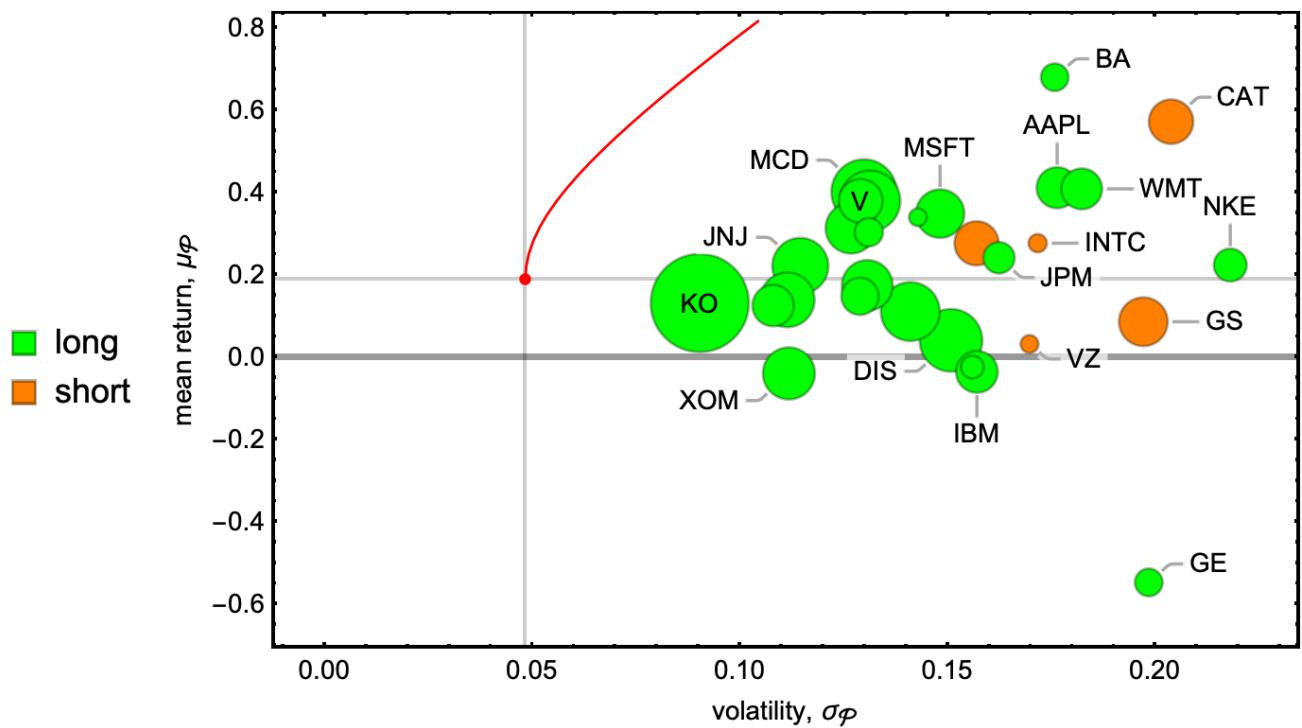
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- MPT can produce crazy portfolios
- Variance as a measure of risk is debatable
- Proceeds from short sales cannot be reinvested in practice



## MPT Can Produce Crazy Portfolios

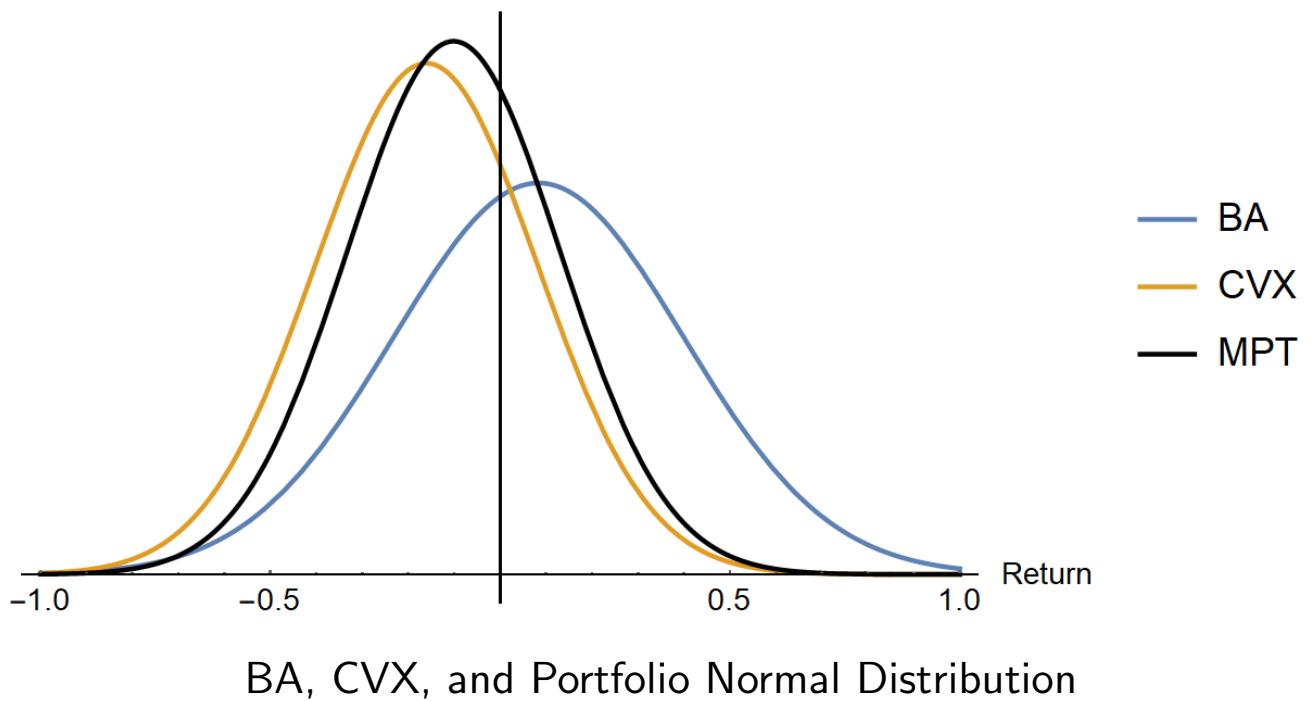
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Dow 30 returns for 2017



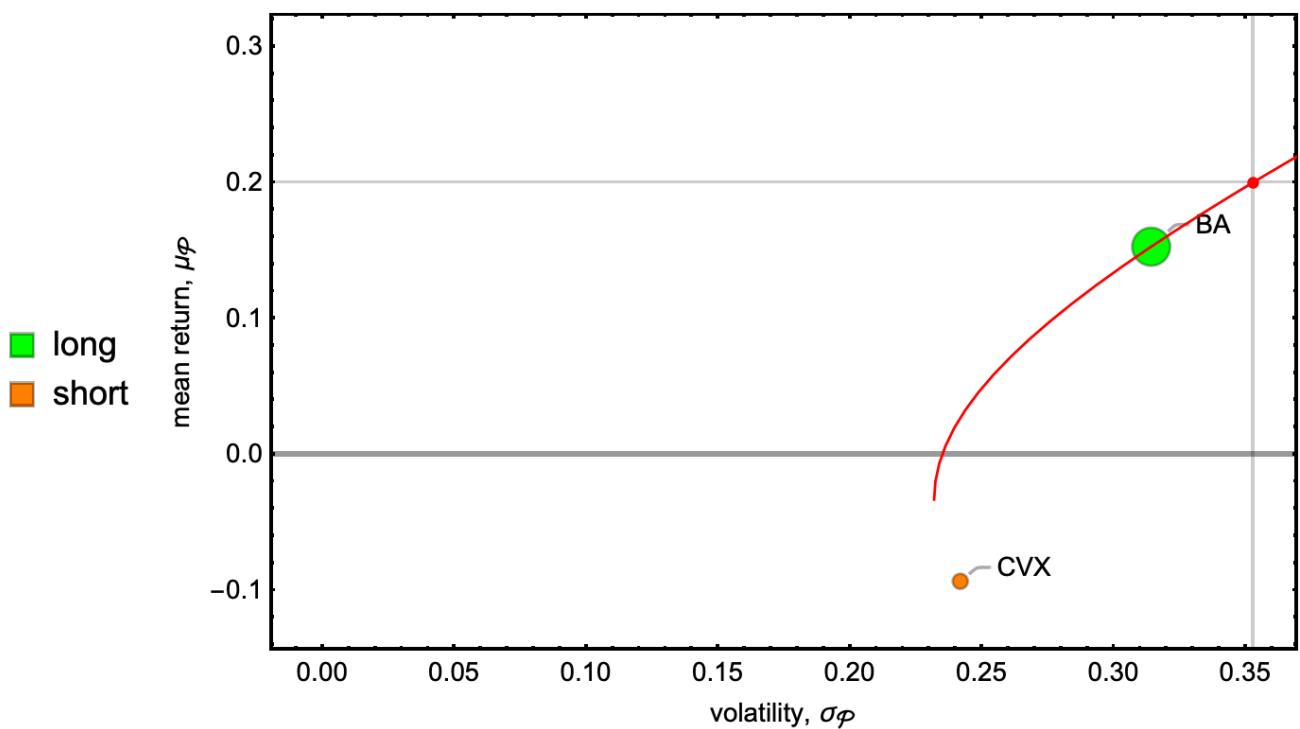
## Is Smaller Variance Really What We Want?





## Proceeds from the Short Sale are Reinvested

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Based on Data for BA and CVX in 2018

119% BA and -19% CVX



## Modified MPT

For  $n$  assets, modify MPT to minimize  $\sigma_{\mathcal{P}}^2$  such that

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$$\underbrace{\boldsymbol{\mu}^T \mathbf{w} + 1.5r \sum_{i=1}^n R(-w_i)}_{\mu_{\mathcal{P}}} \geq \alpha$$
$$\sum_{i=1}^n [w_i] = 1,$$

where

$$[x] = \begin{cases} x & x \geq 0 \\ -0.5x & x < 0 \end{cases} \quad \text{and} \quad R(x) = \begin{cases} x & x \geq 0 \\ 0 & x < 0 \end{cases}.$$



## Example Illustrating Constraints

Principal: \$1000

Spot Prices: Asset 1: \$100 Asset 2: \$100

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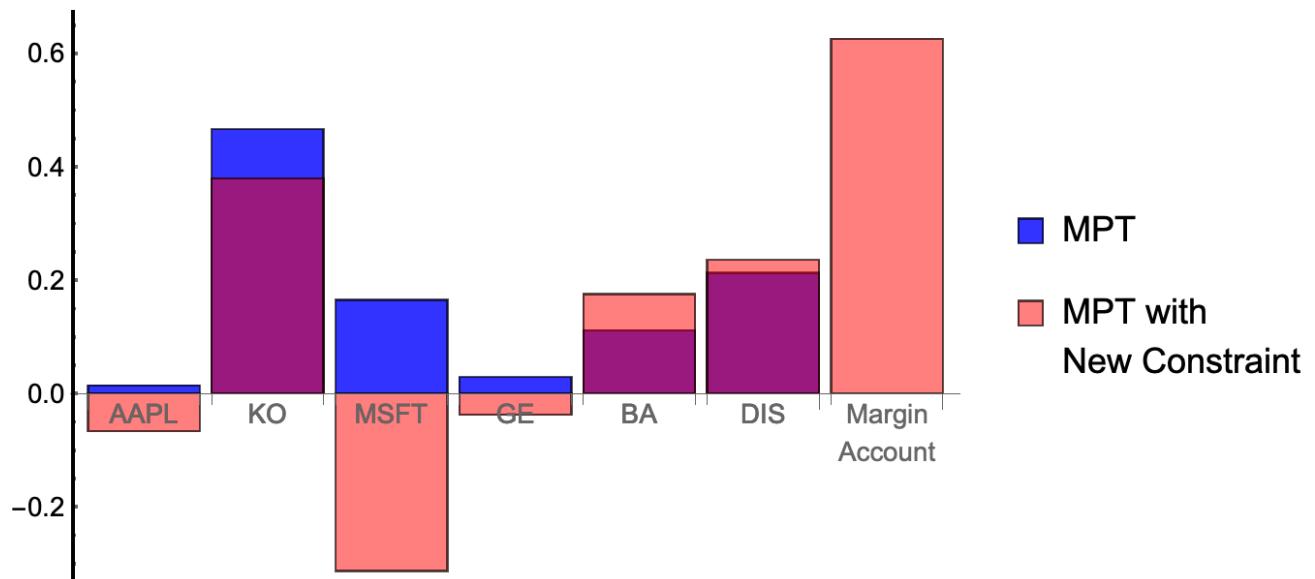
Long \$500 in asset 1 ( $w_1 = 0.5$ ) and short \$1000 of asset 2 ( $w_2 = -1$ ).

$$w_1 + 0.5(-w_2) = 0.5 + 0.5(+1) = 1.$$

Margin:  $\underbrace{\$0}_{\substack{\text{deposit} \\ \text{for asset 1}}} + \underbrace{\$500}_{\substack{\text{deposit} \\ \text{for asset 2}}} + \underbrace{\$1000}_{\substack{\text{proceeds from} \\ \text{short sale}}} = \$1500$

$$1.5 [R(-w_1) + R(-w_2)] = 1.5 [R(-0.5) + R(1)] = 1.5$$

## Example

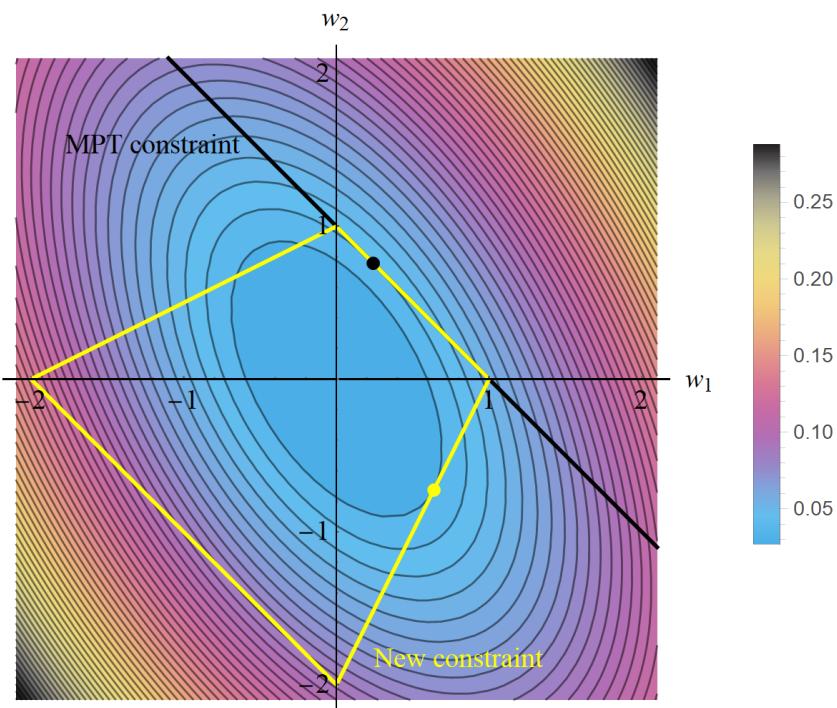


Weights for 2019 Portfolio with and without the new constraint



## Compare with MPT (2 Assets)

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Based on Data for BA and CVX in 2018.

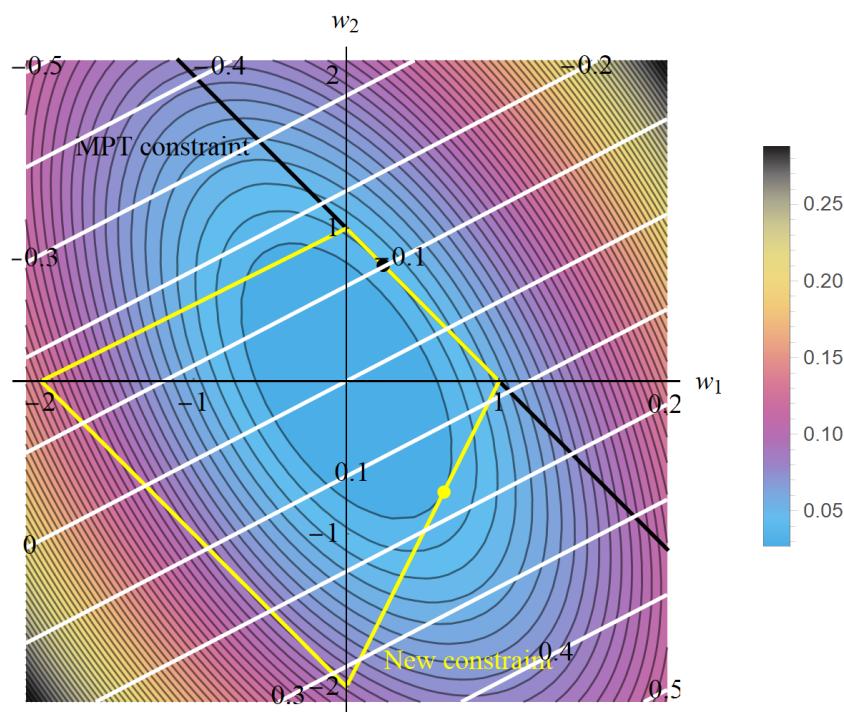


## Compare with MPT (2 Assets)

MPT

Modified MPT

PMPT



Based on Data for BA and CVX in 2018.



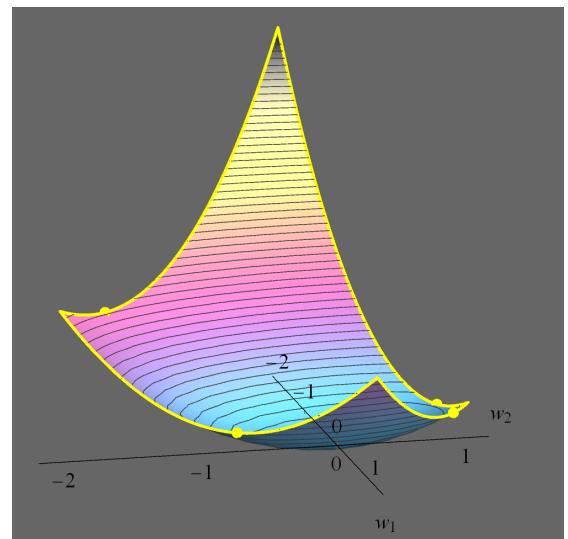
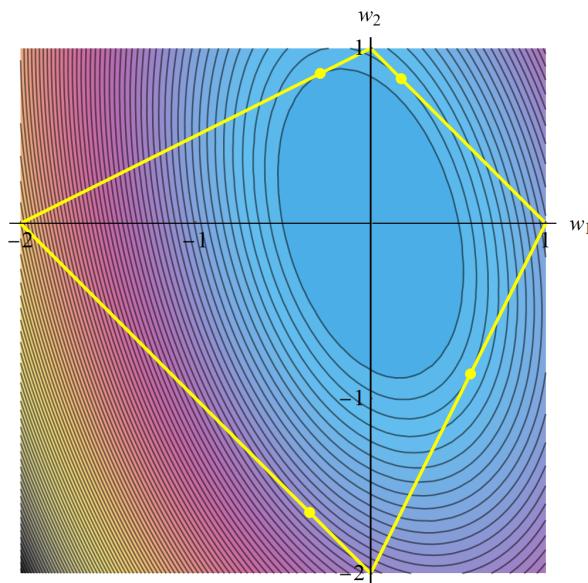
# Solution

Solutions must be found numerically, but the existence of many similar local minima can make the global minimum difficult to find.

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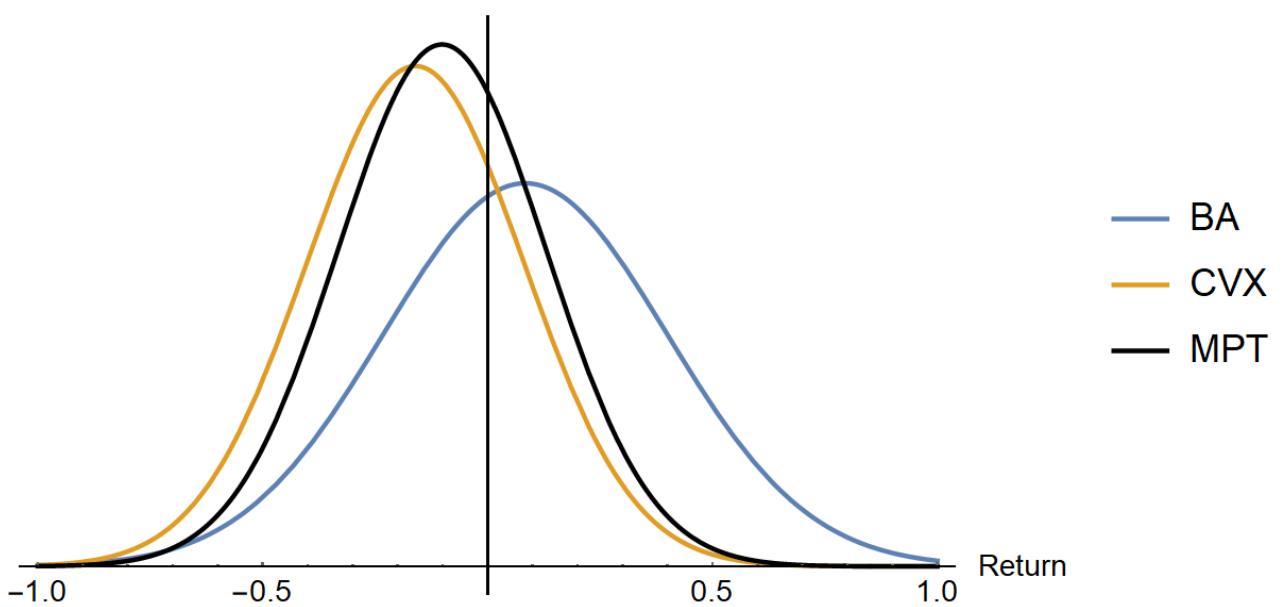
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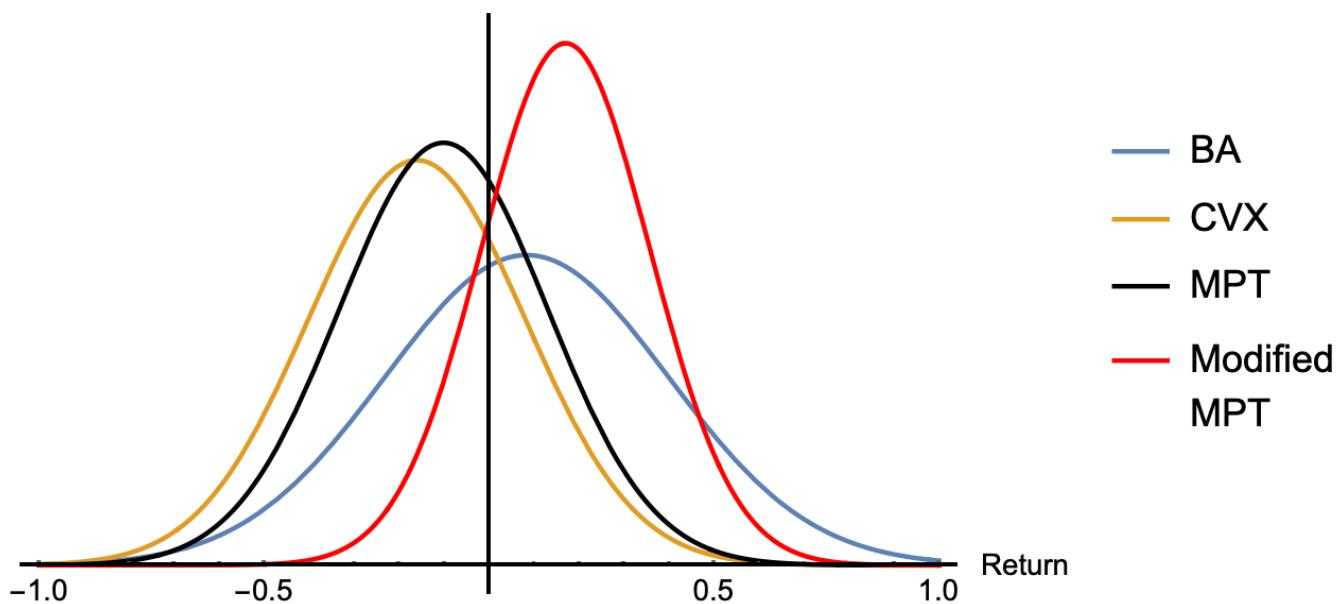
## Probability Density Function for Returns



Based on Data for BA and CVX in 2018.



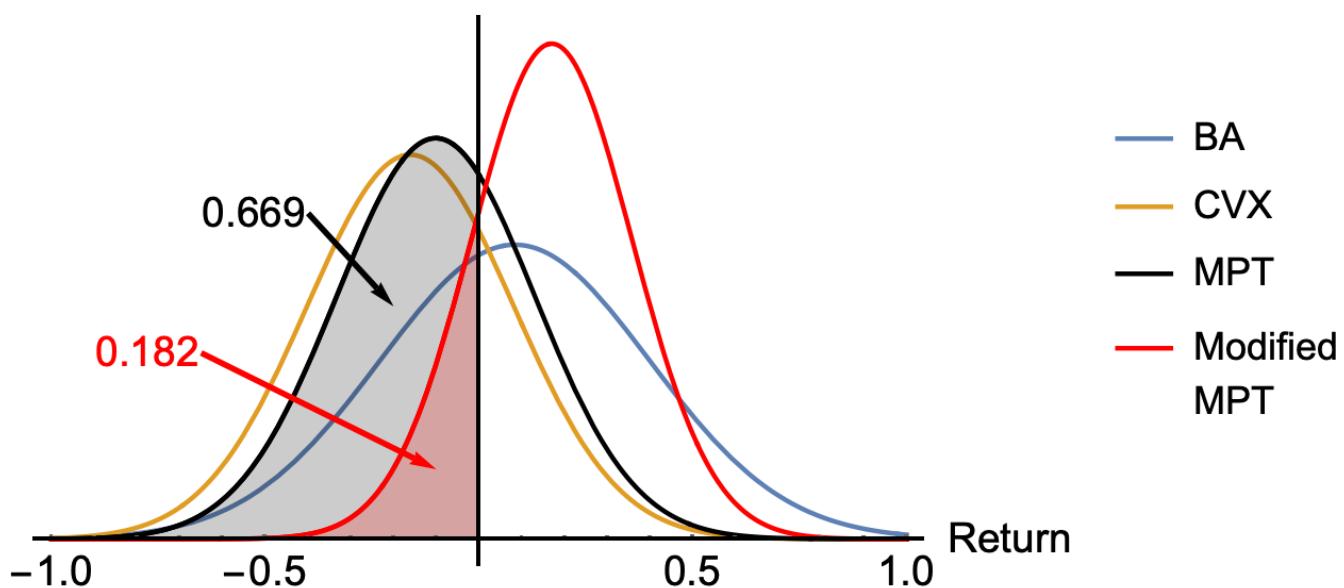
## Probability Density Function for Returns



Based on Data for BA and CVX in 2018.



## Value at Risk



Based on Data for BA and CVX in 2018.



## Value at Risk

If the asset returns are normally distributed, then

$$\begin{aligned}\text{Value at Risk} &= P(\text{portfolio return} < 0) \\ &= P\left(\frac{\text{return} - \mu_p}{\sigma_p} < \frac{-\mu_p}{\sigma_p}\right) \\ &= P\left(Z < -\frac{\mu_p}{\sigma_p}\right) \\ &= \Phi\left(-\frac{\mu_p}{\sigma_p}\right),\end{aligned}$$

where  $\Phi$  is the CDF for the standard normal distribution.

Note that minimizing the value at risk is equivalent to maximizing  $\frac{\mu_p}{\sigma_p}$  or minimizing  $\frac{\sigma_p}{\mu_p}$ , the coefficient of variation for the portfolio.

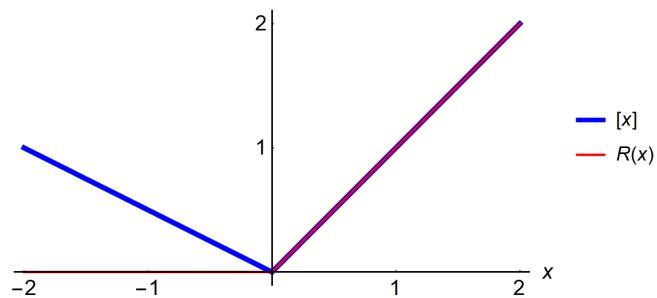


# Post-Modern Portfolio Theory (PMPT)

Minimize  $\frac{\sigma_{\mathcal{P}}}{\mu_{\mathcal{P}}} = \frac{\sqrt{\mathbf{w}^T \Sigma \mathbf{w}}}{\mu^T \mathbf{w} + 1.5r \sum_{i=1}^n R(-w_i)}$  such that

$\mu_{\mathcal{P}} \geq \alpha$  and  $\sum_{i=1}^n [w_i] = 1$ , where

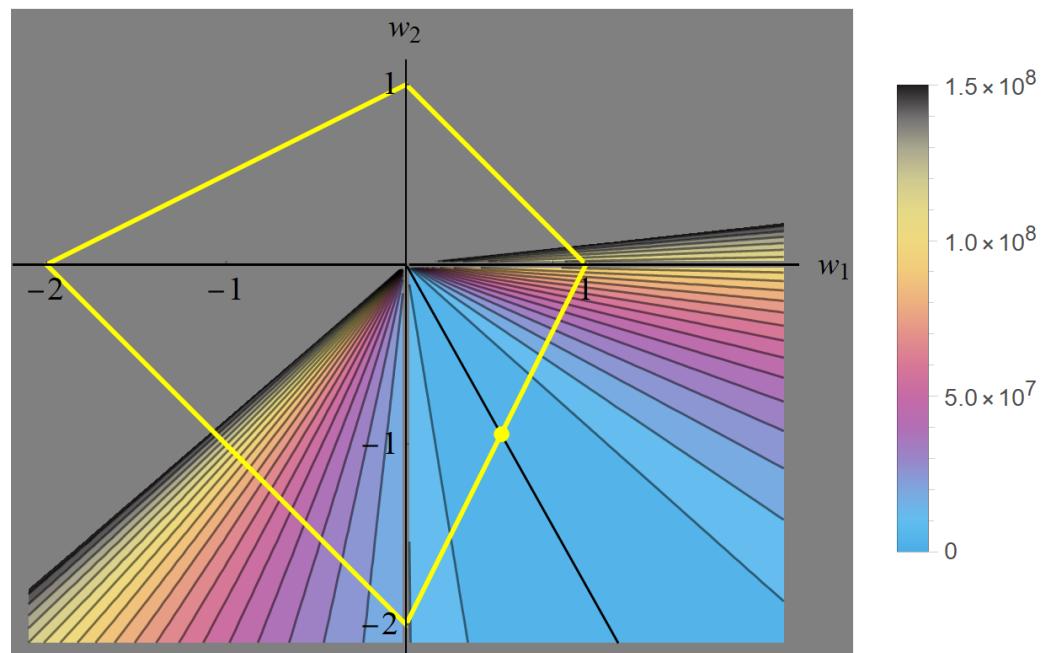
$$[x] = \begin{cases} x & x \geq 0 \\ -0.5x & x < 0 \end{cases} \quad \text{and} \quad R(x) = \begin{cases} x & x \geq 0 \\ 0 & x < 0 \end{cases}.$$





## Optimization Problem for PMPT

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Based on Data for BA and CVX in 2018.



## Almost Closed-Form Solution

$$\mathbf{w} = \Sigma^{-1} [\boldsymbol{\mu} - 1.5rH(-\mathbf{w})] t, \quad t > 0,$$

where

$$H(x) = \begin{cases} 1 & x \geq 0 \\ 0 & x < 0 \end{cases}$$

is the Heaviside step function and  $t$  is chosen so that

$$\sum_{i=1}^n [w_i] = 1.$$

If  $r > 0$ , then we can use fixed point iteration to quickly find the solution for any  $r$  using

$$\mathbf{w}_k = \Sigma^{-1} [\boldsymbol{\mu} - 1.5rH(-\mathbf{w}_{k-1})] t, \quad \mathbf{w}_0 = \Sigma^{-1} \boldsymbol{\mu} t_0.$$

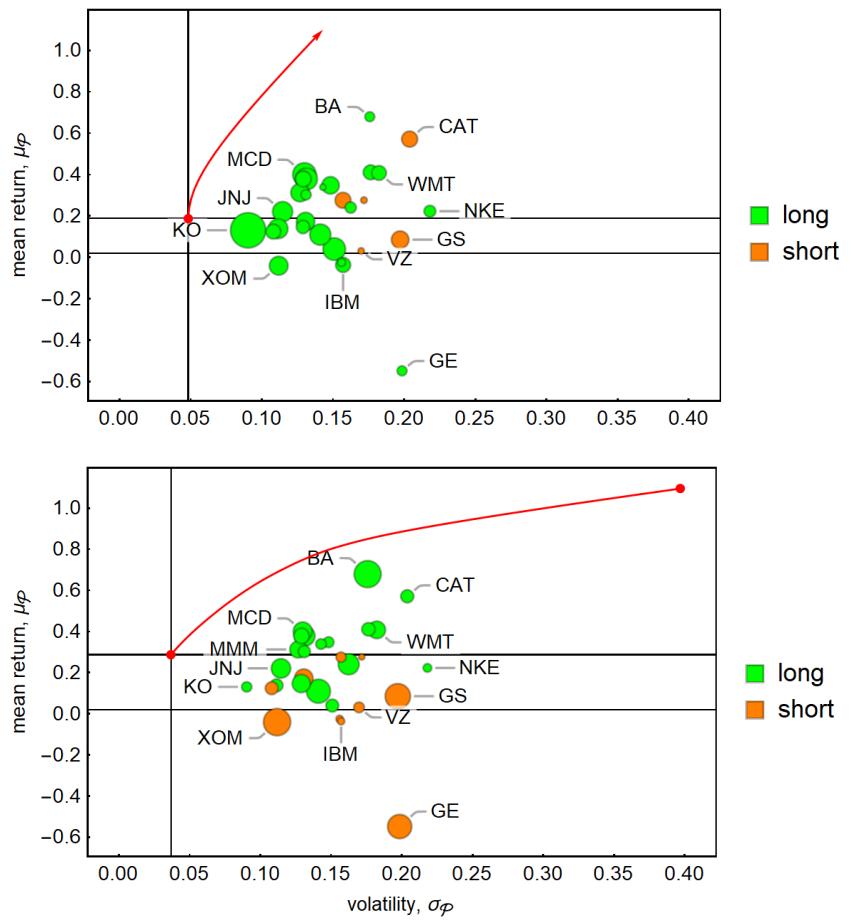


## MPT vs PMPT

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PMPT

MPT

PMPT





MPT

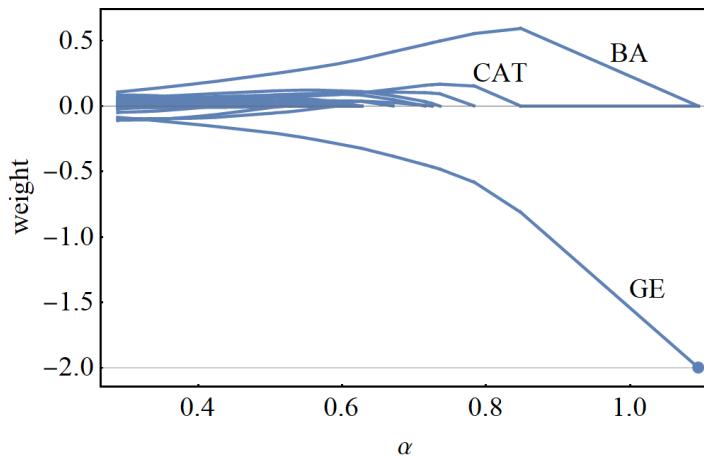
Modified MPT

PMPT

# Questions?



## PMPT Weights



Note that the weights are piecewise linear in  $\alpha$  and, except for GE, they achieve exact values of zero for larger values of  $\alpha$ .



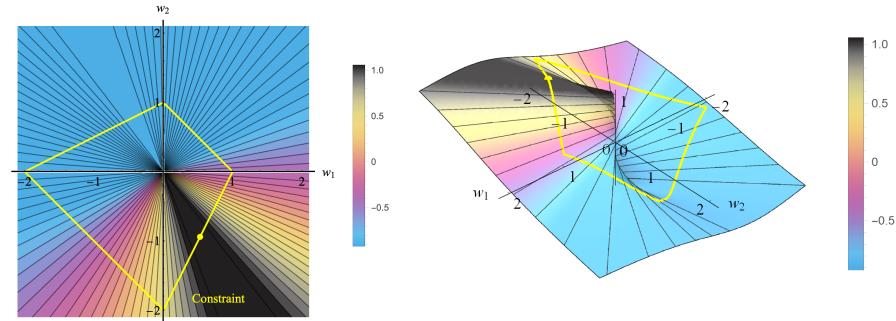
$$r = 0.02$$

It looks like the contours are still lines even with  $r > 0$ .  
Perhaps we can get a closed form solution here too?

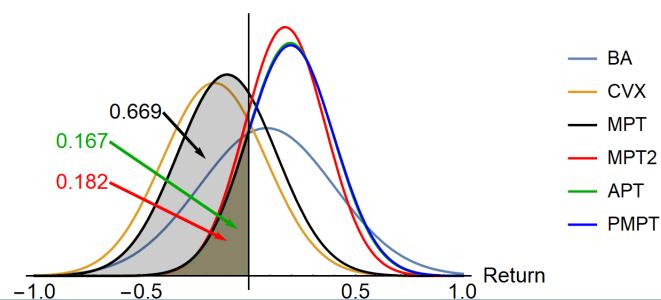
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Setting  $r = 0.02$  doesn't seem to do much to the portfolio in this case.





## Back Testing

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There may not be room for this in your presentation, but it's the kind of thing that a person from the audience may ask, so it might be good to prepare a slide on it.

Pictures comparing the results for all three models would be good to have.