

Sparse Portfolio Design

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Background

- Portfolio Optimization: Optimizing the allocation of funds w in a portfolio to maximize return and minimize risk
- Mean-Variance Model: $\min_w \underbrace{w^T K w}_{\text{Risk}} - \beta \underbrace{u^T w}_{\text{Return}} \text{ s.t. } \underbrace{w^T \mathbf{1}}_{\text{Budget}} = 1$
- K is the covariance matrix and u is the average daily return of the assets, and β is risk tolerance
- MV model results in dense portfolio which is more expensive and complex
- K considers both overperformance and underperformance risky

Objective

Create a optimization which:

- Considers underperformance more risky
- Produces a long-only sparse portfolio
- Enforces reallocation limits

Methodology

Downside deviation diagonal matrix D :

- Each value on the diagonal is $\sqrt{LPM_2(0)}$ of the returns of an asset
- Penalizes stocks with large or frequent underperformance

$$\text{Modified } \hat{K} = K_{\text{denoised}} + \gamma D \times \frac{\text{mean}(\text{diag}(K_{\text{denoised}}))}{\text{mean}(\text{diag}(D))}$$

- γ to control weighting of D

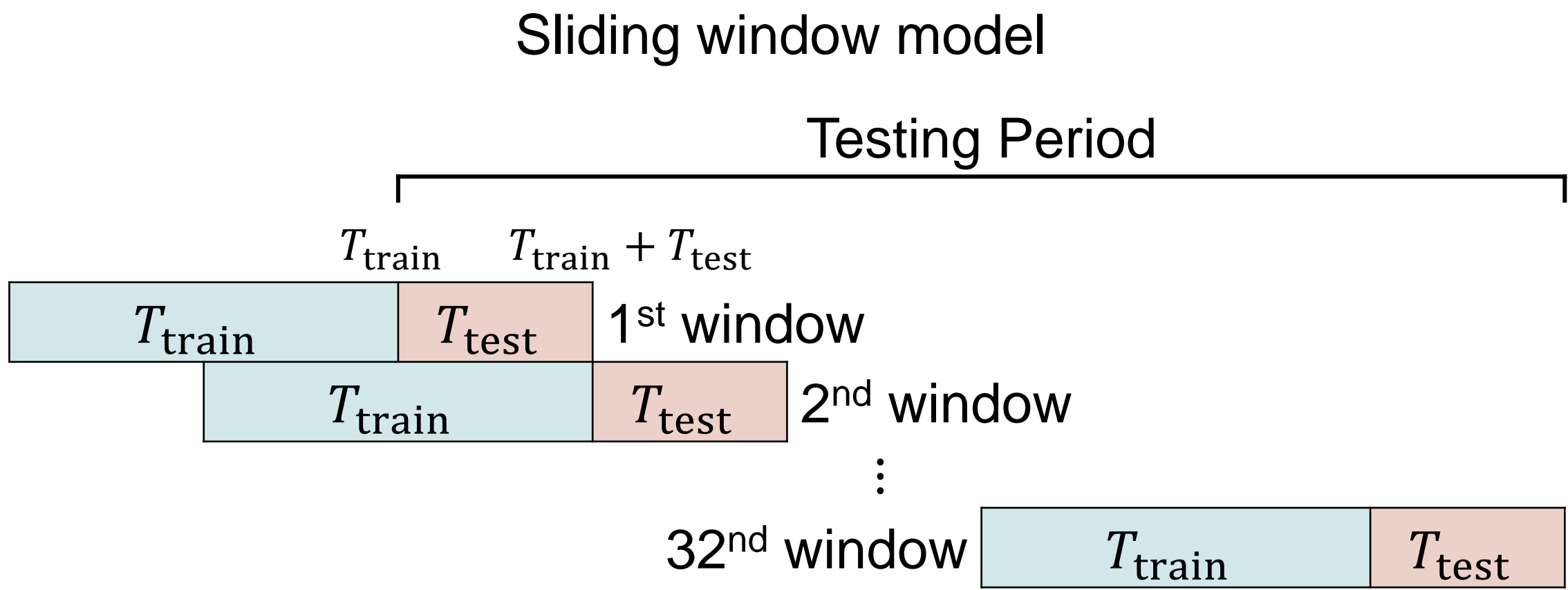
$$\text{Formulation: } \min_w f(w) = \underbrace{w^T \hat{K} w}_{\text{Risk}} + \beta \underbrace{u^T w}_{\text{Return}} + \alpha \underbrace{(w^T \mathbf{1} - 1)^2}_{\text{Budget}}$$

Projection for sparsity, weight and reallocation constraints

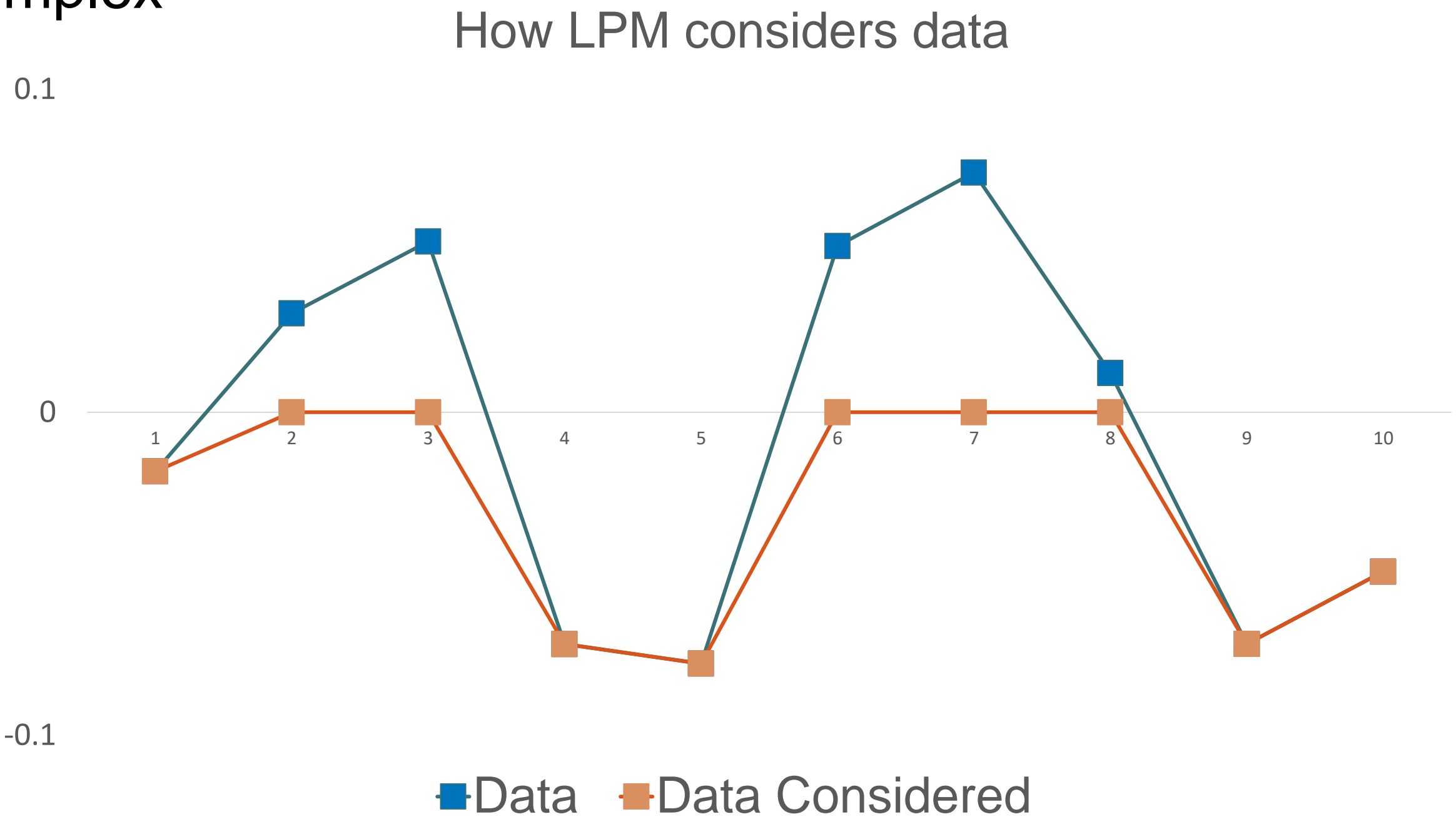
- First project values with non-zero lower bounds
- Project remaining until desired sparsity
- Done at each step of gradient descent

Results

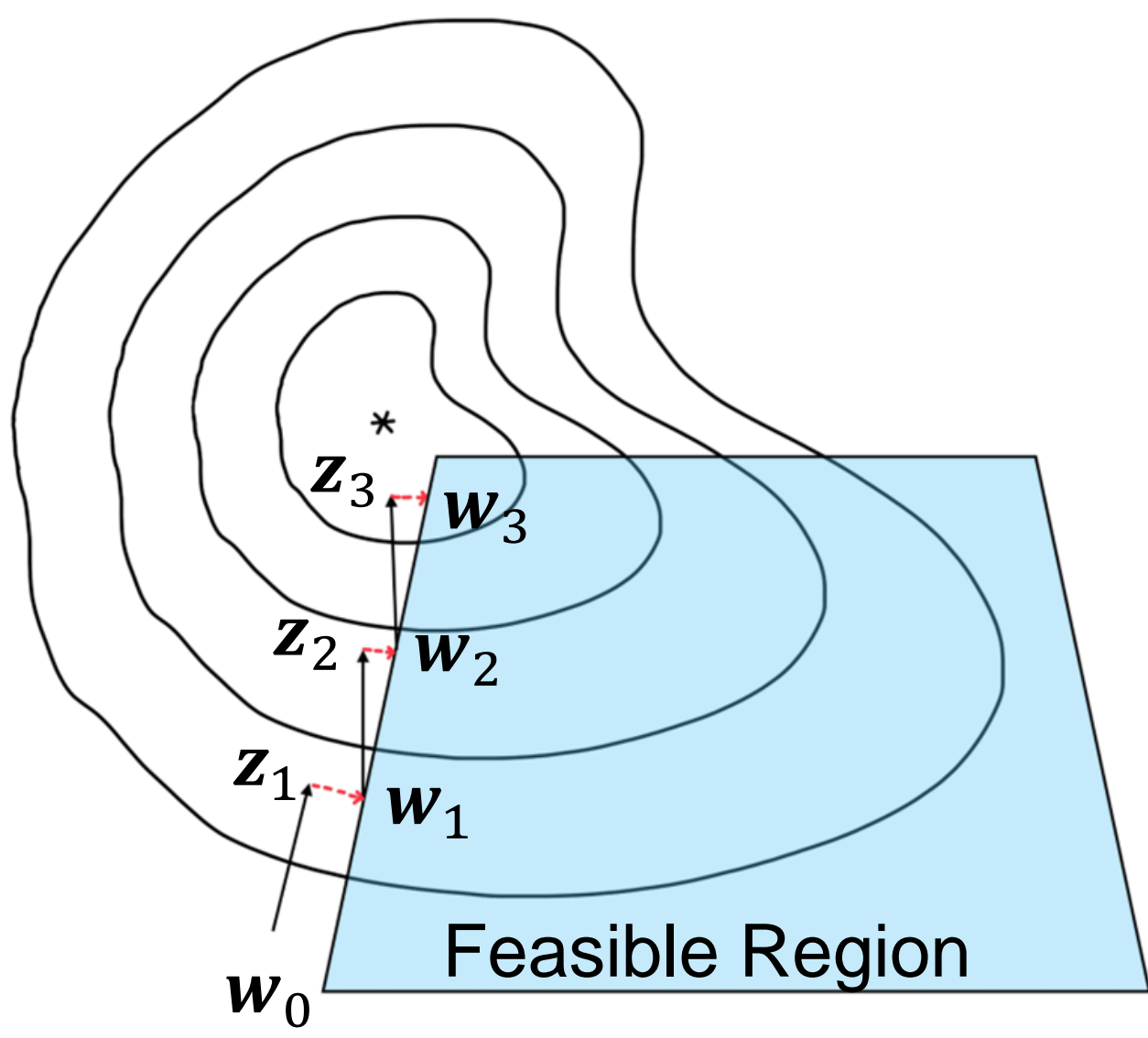
- Tested using sliding window model



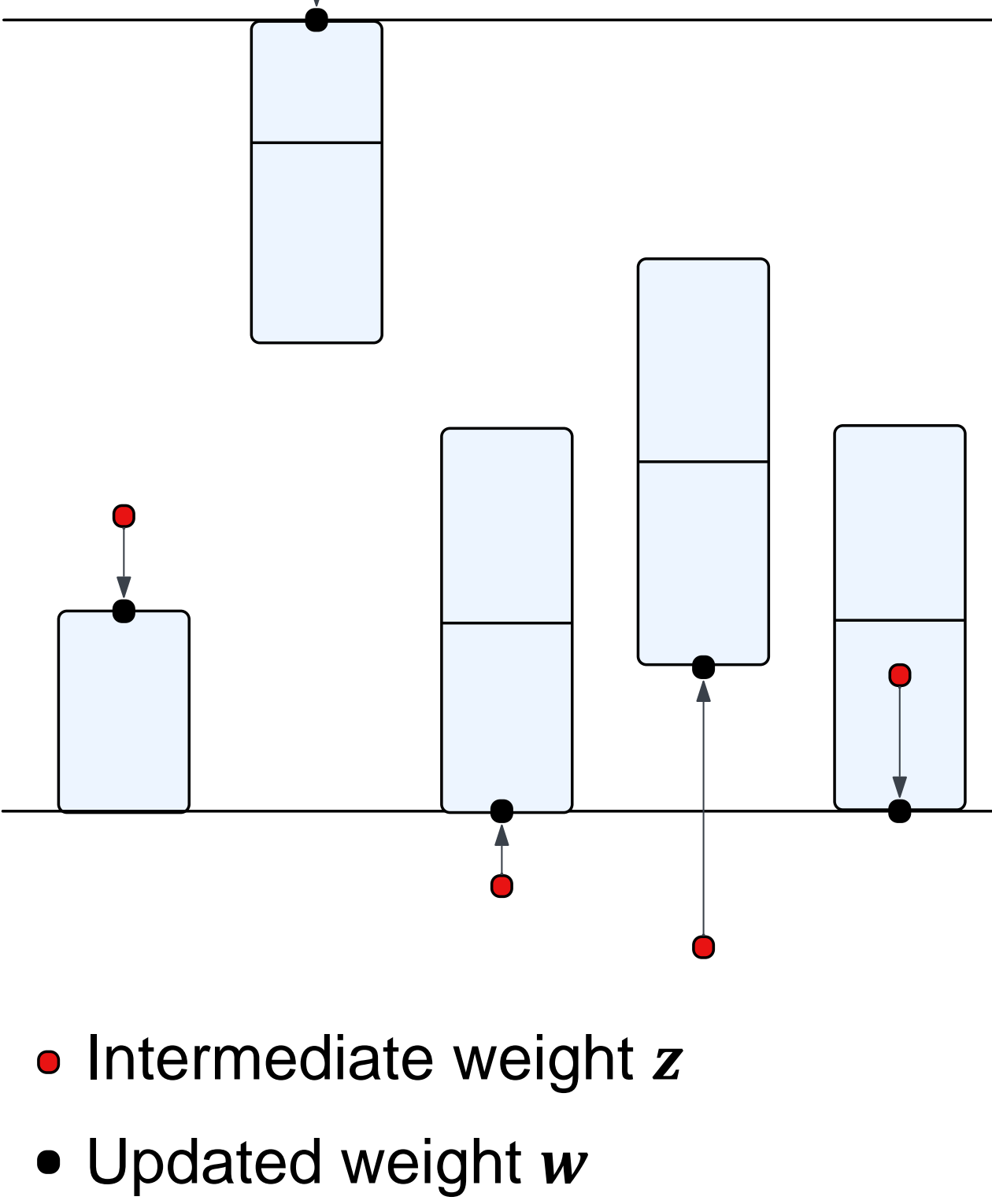
Index	Assets	T_{train}	T_{test}	Total days	Start Date	End Date
S&P 500	472	504	63	2520	3/2/2015	3/2/2025
Russell 1000	807					



Projected Gradient Descent



Projection operation with sparsity 3



	Algorithm	Bounded Reallocation	Note
■	PGDBDR (Mine)	Yes	
■	PGDB		Mean-Variance
■	PGDDR (Mine)	Yes	
■	PGD		Mean-Variance
■	Fixed Target	Yes	Minimizes returns below target

