

blacklitterman

November 23, 2025

0.1 Black-Litterman Model Summary

0.1.1 Methodology

1. **Reverse Optimization:** Derived implied equilibrium returns (μ_e) from market cap weights using the formula $\mu_e = \Sigma w$
 - Calculated risk aversion coefficient (ρ) from market portfolio characteristics
 - These represent what the market “believes” about future returns
2. **Investor Views:** Defined subjective views about future performance
 - View 1: US Large Cap Growth outperforms Value by 3%
 - View 2: Emerging Markets absolute return of 8%
 - View 3: US Growth outperforms International by 4%
3. **Black-Litterman Blending:** Combined equilibrium returns with investor views
 - Views weighted by confidence levels
 - Higher confidence = greater influence on final returns
 - π parameter controls uncertainty in prior beliefs
4. **Portfolio Optimization:** Used BL expected returns to find optimal allocation
 - Maximized Sharpe ratio subject to long-only constraints
 - Compared to market cap weighted portfolio

0.1.2 Key Parameters

- **Risk-free rate:** 4%
- **(prior uncertainty):** 0.025
- **(risk aversion):** Derived from market portfolio
- **Historical data:** 5 years of daily returns

0.1.3 Interpretation

The resulting optimal weights reflect a balance between:
- Market equilibrium (what everyone collectively believes)
- Your subjective views (what you think will happen)
- Statistical uncertainty (confidence in predictions)

Adjust the views, confidence levels, and π parameter to reflect your market outlook.

1 Market Implied Equilibrium Weights

	Ticker	Market Cap	Weight (%)
0	R1KG	215000000000000	26.615087
1	R1KV	215000000000000	26.615087
2	R2K	220000000000000	2.723404
3	RMCG	325000000000000	4.023211
4	RMCV	325000000000000	4.023211
5	INT	250000000000000	29.032258
6	EM	600000000000000	6.967742

2 Historical Returns and Volatility

	Column	First Valid Date
0	EM	2003-04-14
1	INT	2003-04-14
2	R1KV	2003-04-14
3	R1KG	2003-04-14
4	R2K	2003-04-14
5	RMCG	2003-04-14
6	RMCV	2003-04-14

2.1 Returns

	Ticker	Total Compound Return (%)	Annualized Return (CAGR) (%)	\
0	EM	215.80	5.23	
1	INT	232.97	5.47	
2	R1KV	403.66	7.43	
3	R1KG	919.78	10.84	
4	R2K	344.15	6.83	
5	RMCG	610.11	9.07	
6	RMCV	498.08	8.25	

	Average Annual Return (%)	Min Annual Return (%)	\
0	8.80	-60.25	
1	7.49	-47.19	
2	8.99	-41.97	
3	12.21	-42.41	
4	9.47	-40.77	
5	10.99	-49.41	
6	10.02	-43.44	

	25th Percentile Annual Return (%)	Median Annual Return (%)	\
0	-5.22	10.11	
1	-1.38	10.53	
2	-2.47	13.15	

3	4.80	11.95
4	-0.39	11.81
5	4.57	10.87
6	-3.25	12.77
	75th Percentile Annual Return (%)	Max Annual Return (%)
0	26.34	63.34
1	20.35	46.46
2	16.30	32.80
3	29.68	39.94
4	18.00	49.42
5	23.97	40.33
6	21.87	41.21
	Years of Data	

2.2 Volatility

Volatility Statistics:

	Ticker	Total Volatility (%)	Average Rolling Volatility (%)	\
0	EM	27.22	24.63	
1	INT	20.73	18.94	
2	R1KV	19.04	17.06	
3	R1KG	19.56	18.04	
4	R2K	23.88	22.41	
5	RMCG	21.39	19.75	
6	RMCV	20.38	18.38	
	Period 1 (2003-04 to 2007-10)	Period 2 (2007-10 to 2012-04)	\	
0		22.94	44.40	
1		14.40	33.08	
2		12.03	29.28	
3		12.42	25.48	
4		17.90	33.88	
5		14.36	29.72	
6		12.24	30.69	
	Period 3 (2012-04 to 2016-10)	Period 4 (2016-10 to 2021-05)	\	
0		19.44	22.34	
1		15.99	18.04	
2		13.23	20.17	
3		13.38	21.33	
4		16.42	24.35	
5		14.48	21.72	
6		13.82	21.98	
	Period 5 (2021-05 to 2025-11)			
0		18.16		

1	16.30
2	15.13
3	21.82
4	22.83
5	22.73
6	17.68

2.3 Sharpe

Sharpe Ratio Statistics:

	Ticker	Total Sharpe Ratio	Period 1 (2003-04 to 2007-10)	\
0	EM	0.18	1.37	
1	INT	0.17	1.34	
2	R1KV	0.26	1.08	
3	R1KG	0.42	0.71	
4	R2K	0.23	0.83	
5	RMCG	0.33	1.03	
6	RMCV	0.30	1.32	
			Period 2 (2007-10 to 2012-04)	Period 3 (2012-04 to 2016-10) \
0		-0.16		-0.23
1		-0.34		0.08
2		-0.23		0.58
3		-0.08		0.58
4		-0.10		0.39
5		-0.08		0.44
6		-0.11		0.61
			Period 4 (2016-10 to 2021-05)	Period 5 (2021-05 to 2025-11)
0		0.28		-0.12
1		0.32		0.14
2		0.41		0.19
3		0.79		0.44
4		0.45		-0.06
5		0.65		0.09
6		0.37		0.05

3 Covariance & Correlation

Annual Expected Returns:

ticker	
EM	0.088022
INT	0.074877
R1KV	0.089854
R1KG	0.122106
R2K	0.094748

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RMCG    0.109859
RMCV    0.100194
dtype: float64
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Covariance Matrix:

ticker	EM	INT	R1KV	R1KG	R2K	RMCG	RMCV
ticker							
EM	0.074098	0.048820	0.041505	0.041111	0.049181	0.045529	0.043639
INT	0.048820	0.042964	0.034524	0.033313	0.039945	0.036465	0.036055
R1KV	0.041505	0.034524	0.036245	0.032419	0.040861	0.035966	0.037645
R1KG	0.041111	0.033313	0.032419	0.038256	0.039878	0.039483	0.034343
R2K	0.049181	0.039945	0.040861	0.039878	0.057016	0.046438	0.045363
RMCG	0.045529	0.036465	0.035966	0.039483	0.046438	0.045740	0.039478
RMCV	0.043639	0.036055	0.037645	0.034343	0.045363	0.039478	0.041552

Correlation Matrix:

ticker	EM	INT	R1KV	R1KG	R2K	RMCG	RMCV
ticker							
EM	1.000000	0.865255	0.800884	0.772152	0.756656	0.782059	0.786448
INT	0.865255	1.000000	0.874862	0.821689	0.807064	0.822576	0.853315
R1KV	0.800884	0.874862	1.000000	0.870616	0.898840	0.883310	0.970025
R1KG	0.772152	0.821689	0.870616	1.000000	0.853859	0.943870	0.861366
R2K	0.756656	0.807064	0.898840	0.853859	1.000000	0.909342	0.931965
RMCG	0.782059	0.822576	0.883310	0.943870	0.909342	1.000000	0.905554
RMCV	0.786448	0.853315	0.970025	0.861366	0.931965	0.905554	1.000000

Asset Correlation Analysis:

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Interpretation:

- Avg Correlation: Mean correlation with other assets (higher = more correlated)
- Mkt-Weighted Avg Corr: Correlation weighted by market cap (more relevant measure)
- Corr with Portfolio: Correlation with the market portfolio
- Diversification Ratio: Higher = better diversification benefit

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Ticker	Avg Correlation	Mkt-Weighted Avg Corr	Corr with Portfolio	\
0 EM	0.7939	0.8114	0.9334	
1 INT	0.8408	0.8420	0.9447	
2 R1KV	0.8831	0.8648	0.9247	
3 R1KG	0.8539	0.8505	0.9111	
4 R2K	0.8596	0.8377	0.8985	
5 RMCG	0.8745	0.8297	0.9349	

6	RMCV	0.8848	0.8577	0.9231
Diversification Ratio Standalone Volatility				
0		1.0714	0.2722	
1		1.0585	0.2073	
2		1.0814	0.1904	
3		1.0976	0.1956	
4		1.1130	0.2388	
5		1.0696	0.2139	
6		1.0833	0.2038	

4 Reverse Optimization

How = Σw Works

4.0.1 The Calculation Steps

1. $\Sigma @ w_{mkt} \rightarrow$ Multiplies the 7×7 covariance matrix by the 7×1 weight vector \rightarrow produces a 7×1 vector representing the **marginal contribution to portfolio risk** for each asset
 2. $\text{lambda_risk_aversion} \times \rightarrow$ Scales this by the risk aversion coefficient \rightarrow converts risk contributions into **expected return premiums**
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4.0.2 Equilibrium Intuition

In equilibrium, assets with higher contributions to portfolio risk must offer higher expected returns to compensate investors. The formula says:

- **Assets that are more volatile** (high variance) \rightarrow higher implied returns
- **Assets that correlate positively with the portfolio** \rightarrow higher implied returns
- **Assets with larger market weights** \rightarrow already reflect equilibrium pricing

Note on market weights: Assets with larger market cap weights don't necessarily need higher expected returns - their current weights already represent the market's equilibrium assessment of their risk/return profile. If an asset has a large market cap weight, there are two possible reasons:

1. It offers high expected returns (investors want to hold a lot of it)
2. It has low risk or diversification benefits (it's "safe" to hold a lot of it)

The market weight itself is the **outcome** of equilibrium pricing - it already incorporates all the information about risk and return that investors collectively believe.

4.0.3 Example

If emerging markets (EM) has high volatility and positive correlation with the rest of the portfolio, the formula will assign it a higher implied return to justify investors holding it at its current market weight.

Risk Aversion Coefficient (): 1.1939

Market Return: 9.3403%

Market Variance: 0.0447

Market Stdev: 0.2115

Implied Equilibrium Returns (from Reverse Optimization):

	Ticker	Market Weight (%)	Implied Return (%)
0	R1KG	26.615087	6.415429
1	R1KV	26.615087	4.944381
2	R2K	2.723404	4.445371
3	RMCG	4.023211	4.499714
4	RMCV	4.023211	5.417163
5	INT	29.032258	5.048777
6	EM	6.967742	4.751485

Individual Asset Sharpe Ratios (Implied Returns):

	Ticker	Implied Return (%)	Volatility (%)	Sharpe Ratio	Market Weight (%)
0	R1KG	6.4154	27.2209	0.0887	26.6151
4	RMCV	5.4172	23.8781	0.0593	4.0232
5	INT	5.0488	21.3869	0.0490	29.0323
1	R1KV	4.9444	20.7278	0.0456	26.6151
6	EM	4.7515	20.3844	0.0369	6.9677
3	RMCG	4.4997	19.5592	0.0255	4.0232
2	R2K	4.4454	19.0382	0.0234	2.7234

Highest Sharpe: R1KG = 0.0887

Lowest Sharpe: R2K = 0.0234

Sharpe Gap: 0.0653

Unconstrained mean-variance optimization allocates 100% to the highest Sharpe asset

when diversification benefits don't outweigh the risk-adjusted return advantage.

4.1 Scalar vs. Asset-Specific Risk Aversion

The choice between scalar vs. array risk aversion involves important tradeoffs:

4.1.1 Pro:

- Captures heterogeneous investor risk preferences across asset classes (e.g., institutions may be more risk-averse to emerging markets than domestic equities)
- Can reflect market segmentation where different investor bases dominate different assets
- Allows modeling of risk budgeting frameworks where risk tolerance varies by asset category

4.1.2 Cons:

- No clear market observable to estimate individual values - you'd need to make subjective assumptions
 - Loss of interpretability: a single asset-specific value has clear economic meaning (market price of risk), but asset-specific values are harder to justify
 - Violates the assumption of a representative investor with consistent preferences across all assets
 - In standard CAPM/equilibrium theory, all investors face the same efficient frontier, implying a single market-wide risk aversion
 - Can lead to arbitrage opportunities in theory (why would rational investors have different risk aversion for the same dollar of risk?)
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The Black-Litterman model already allows you to express asset-specific views through the P and Q matrices. If you believe certain assets deserve different treatment, you can express that through higher-confidence views rather than modifying the equilibrium framework.

The power of Black-Litterman is separating equilibrium (market consensus) from your subjective views. Using asset-specific risk aversion muddies this distinction and essentially bakes your views into the “neutral” equilibrium, which defeats the model’s elegance.

Weight Verification Test:

	Ticker	Original Market Weight	Calculated Weight	Difference	\
0	IWF	0.266151	0.266151	1.039022e-07	
1	IWD	0.266151	0.266150	-5.031852e-07	
2	IWM	0.027234	0.027240	5.628593e-06	
3	IWP	0.040232	0.040230	-1.928512e-06	
4	IWS	0.040232	0.040234	2.196299e-06	
5	EFA	0.290323	0.290325	2.041681e-06	
6	EEM	0.069677	0.069670	-7.538778e-06	

	Absolute Error
0	1.039022e-07
1	5.031852e-07
2	5.628593e-06
3	1.928512e-06
4	2.196299e-06
5	2.041681e-06
6	7.538778e-06

Maximum Absolute Error: 0.000008

Mean Absolute Error: 0.000003

Test PASSED (threshold: 0.001)

5 Investor Views

P Matrix (View Picks):

	R1KG	R1KV	R2K	RMCG	RMCV	INT	EM
0	1.0	-1.0	0.0	0.0	0.0	0.0	0.0
1	0.6	0.4	0.0	0.0	0.0	-1.0	0.0
2	0.6	0.4	-1.0	0.0	0.0	0.0	0.0
3	0.0	0.0	-1.0	0.6	0.4	0.0	0.0
4	0.0	0.0	0.0	0.0	0.0	-1.0	1.0

Q Vector (View Returns):

[0.05 0.04 0.05 0.025 0.03]

Omega Matrix (View Uncertainties):

[[0.02589538 0. 0. 0. 0.]
[0. 0.02522139 0. 0. 0.]
[0. 0. 0.02107162 0. 0.]
[0. 0. 0. 0.00891987 0.]
[0. 0. 0. 0. 0.01111388]]

6 Black-Litterman Expected Returns

Combining Equilibrium Returns with Investor Views

6.0.1 The Black-Litterman Formula

$$E[R] = [(\tau\Sigma)^{-1} + P'\Omega^{-1}P]^{-1}[(\tau\Sigma)^{-1}\pi + P'\Omega^{-1}Q]$$

Where: - τ (tau): Scalar representing uncertainty in the prior equilibrium (typically 0.025-0.05) - Σ (Sigma): Covariance matrix - P : Matrix picking out assets in each view - Ω (Omega): Diagonal matrix of view uncertainties - π : Implied equilibrium returns from reverse optimization - Q : Vector of expected returns according to investor views

6.0.2 The Parameter

τ = 0.025 controls how much you trust the market equilibrium returns ():

- **Smaller** (0.01-0.025): High confidence in equilibrium → views have **less** influence
- **Larger** (0.04-0.1): Low confidence in equilibrium → views have **more** influence

Mathematical effect: Larger τ → less precision in (Σ) ¹ → equilibrium estimates treated as more uncertain → more weight given to investor views (Q) relative to equilibrium ()

6.0.3 Interpretation

The posterior expected returns are a **precision-weighted average** of: 1. Market equilibrium beliefs () - what the market collectively thinks 2. Your subjective views (Q) - what you believe

will happen

The weighting is determined by both and the confidence levels in your views.

Black-Litterman Expected Returns:

	Ticker	Market Weight (%)	Implied Return (%)	BL Expected Return (%)	\
0	R1KG	26.615087	6.415429	6.867654	
1	R1KV	26.615087	4.944381	5.145896	
2	R2K	2.723404	4.445371	4.544507	
3	RMCG	4.023211	4.499714	4.585208	
4	RMCV	4.023211	5.417163	5.559484	
5	INT	29.032258	5.048777	5.113059	
6	EM	6.967742	4.751485	4.872743	

	Return Adjustment (%)
0	0.452225
1	0.201515
2	0.099136
3	0.085494
4	0.142321
5	0.064282
6	0.121257

6.0.4 Column Descriptions

- **Market Weight (%)**: Current market capitalization weight
 - Derived from free float-adjusted market caps; Reflects the “consensus” allocation across all investors
- **Implied Return (%)**: Equilibrium returns from reverse optimization
 - Calculated using $= \Sigma w_i \cdot R_i$; What the market “believes” returns should be to justify current weights; Represents the **prior** (before incorporating your views)
- **BL Expected Return (%)**: Black-Litterman posterior returns
 - Precision-weighted blend of implied returns (\hat{R}) and your views (Q); Represents the **posterior** (after incorporating your views)
- **Return Adjustment (%)**: BL Expected Return - Implied Return
 - Shows how much your views moved each asset’s expected return
 - Positive = your views increased expected return vs. equilibrium; Negative = your views decreased expected return vs. equilibrium; Magnitude depends on view confidence and parameter

7 Mean Variance Optimization with BL Expected Returns

Using Option 3: Market cap constraints ($\pm 10\%$) + minimum 2.5% allocation
(maximize Sharpe ratio)

Optimization successful: True

Market Portfolio (BL Returns):

Return: 5.55%

Volatility: 21.15%
 Sharpe Ratio: 0.0734

Optimal BL Portfolio:

Return: 5.90%
 Volatility: 22.42%
 Sharpe Ratio: 0.0848

Portfolio Allocation Comparison:

	Ticker	Market Weight (%)	BL Optimal Weight (%)	Weight Change (%)	\
0	R1KG	26.6151	41.6151	15.0000	
1	R1KV	26.6151	17.8294	-8.7856	
2	R2K	2.7234	2.5000	-0.2234	
3	RMCG	4.0232	2.5000	-1.5232	
4	RMCV	4.0232	19.0232	15.0000	
5	INT	29.0323	14.0323	-15.0000	
6	EM	6.9677	2.5000	-4.4677	
		Implied Return (%)	BL Expected Return (%)		
0		6.4154	6.8677		
1		4.9444	5.1459		
2		4.4454	4.5445		
3		4.4997	4.5852		
4		5.4172	5.5595		
5		5.0488	5.1131		
6		4.7515	4.8727		

Maximum Sharpe Ratio Portfolio (Unconstrained Long-Only):

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Return: 0.0687 (6.87%)
 Volatility: 0.2722 (27.22%)
 Sharpe Ratio: 0.1053

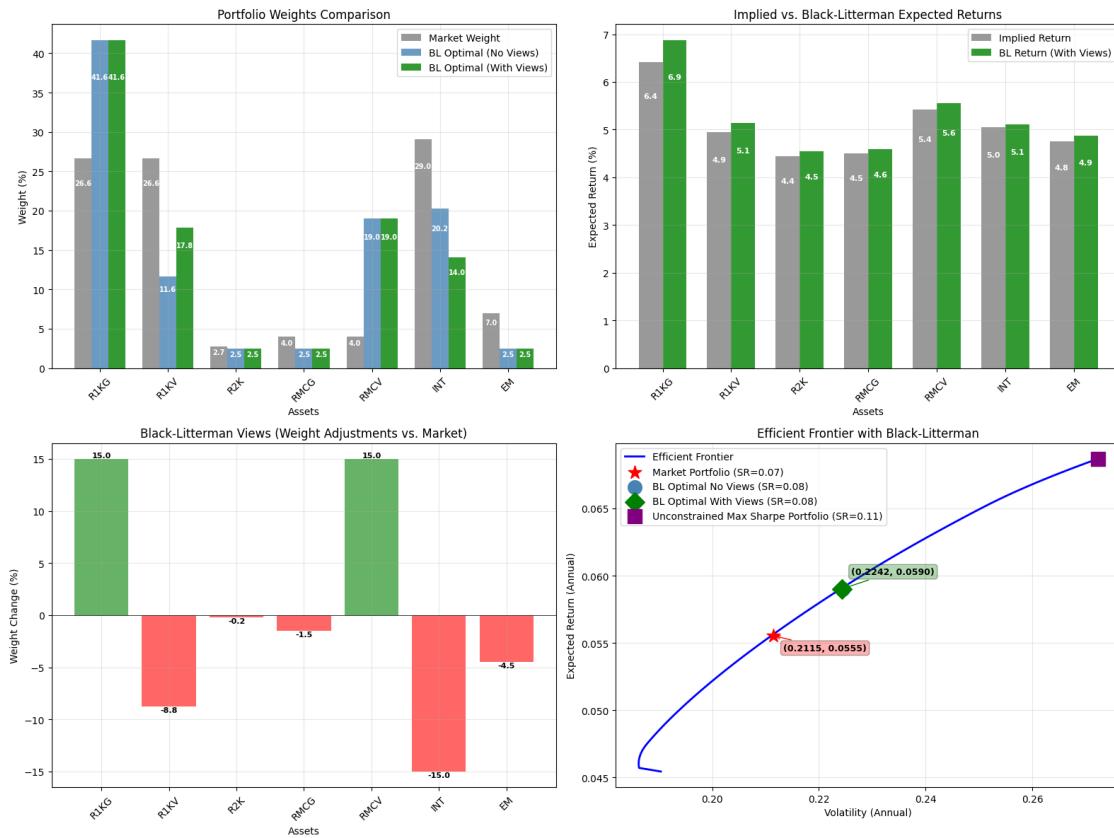
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	Ticker	Max Sharpe Weight (%)	Market Weight (%)	BL Optimal Weight (%)	\
0	R1KG	100.0	26.62	41.62	
1	R1KV	0.0	26.62	17.83	
2	R2K	0.0	2.72	2.50	
3	RMCG	0.0	4.02	2.50	
4	RMCV	0.0	4.02	19.02	
5	INT	0.0	29.03	14.03	
6	EM	0.0	6.97	2.50	
		Difference from Market (%)	Difference from BL Optimal (%)		
0		73.38	58.38		

1	-26.62	-17.83
2	-2.72	-2.50
3	-4.02	-2.50
4	-4.02	-19.02
5	-29.03	-14.03
6	-6.97	-2.50

8 BL Plot



9 Risk Analytics

Comprehensive risk analysis of the optimal Black-Litterman portfolio including:

- Risk Contribution:** Marginal and component risk by asset
- Tracking Error:** Deviation from market portfolio
- Downside Risk:** Semi-deviation and tail risk metrics
- Concentration Risk:** Portfolio diversification measures

Risk Contribution Analysis:

Portfolio Volatility: 0.2242

Tracking Error vs Market: 0.0256

Information Ratio: 0.1355

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	Ticker	Weight (%)	Marginal Risk	Component Risk	Risk Contribution (%)	\
0	R1KG	41.6151	0.2606	0.1085	48.3635	
1	R1KV	17.8294	0.1930	0.0344	15.3487	
2	R2K	2.5000	0.1735	0.0043	1.9342	
3	RMCG	2.5000	0.1730	0.0043	1.9289	
4	RMCV	19.0232	0.2145	0.0408	18.1976	
5	INT	14.0323	0.1943	0.0273	12.1592	
6	EM	2.5000	0.1855	0.0046	2.0680	

	Return Contribution (%)	Risk/Return Ratio	Active Weight (%)	\
0	48.4354	3.7948	15.0000	
1	15.5490	3.7514	-8.7856	
2	1.9254	3.8176	-0.2234	
3	1.9427	3.7735	-1.5232	
4	17.9235	3.8585	15.0000	
5	12.1594	3.8003	-15.0000	
6	2.0645	3.8067	-4.4677	

	TE Contribution (%)
0	107.4630
1	-25.0838
2	-0.5682
3	-2.5739
4	62.3371
5	-29.3480
6	-12.2262

Verification: Sum of Component Risks = 0.224247 (should equal portfolio vol = 0.224247)

Verification: Sum of Risk Contributions = 100.00% (should equal 100%)

Downside Risk Analysis:

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	Metric	BL Optimal Portfolio	Market Portfolio	Difference
0	Semi-Deviation (Annual)	0.1799	0.1721	0.0078
1	Sortino Ratio	0.1057	0.0903	0.0154
2	VaR 95% (Annual)	-0.3211	-0.3040	-0.0172
3	VaR 99% (Annual)	-0.6319	-0.6149	-0.0170
4	CVaR 95% (Annual)	-0.5390	-0.5123	-0.0267
5	CVaR 99% (Annual)	-0.9806	-0.9326	-0.0480
6	Maximum Drawdown	-0.6606	-0.6481	-0.0124
7	Downside Capture Ratio	1.0632	1.0000	0.0632

Interpretation:

- Lower semi-deviation = less downside volatility
- Higher Sortino ratio = better risk-adjusted returns considering only downside risk
- VaR/CVaR = potential losses at different confidence levels
- Lower maximum drawdown = smaller peak-to-trough decline
- Downside capture < 1.0 = portfolio declines less than market on down days

Concentration Risk Analysis:

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Metric	BL Optimal Portfolio	Market Portfolio	\
0 Herfindahl-Hirschman Index	0.2627	0.2348	
1 Effective Number of Assets	3.8063	4.2591	
2 Gini Coefficient	0.4735	0.4191	
3 Diversification Ratio	1.0721	1.0713	
4 Maximum Single Weight	0.4162	0.2903	
5 Top 3 Assets Concentration	0.7847	0.8226	
6 Number of Assets	7.0000	7.0000	

Difference
0 0.0279
1 -0.4528
2 0.0543
3 0.0008
4 0.1258
5 -0.0379
6 0.0000

Interpretation:

- Lower HHI = more diversified portfolio
- Higher effective N = more evenly distributed weights
- Lower Gini = more equal weight distribution
- Higher diversification ratio = better diversification benefits
- Lower top 3 concentration = less reliance on few positions

