

QF609

Risk Analysis



Time Horizon

1-Day



Reporting Currency

USD



Significance Level

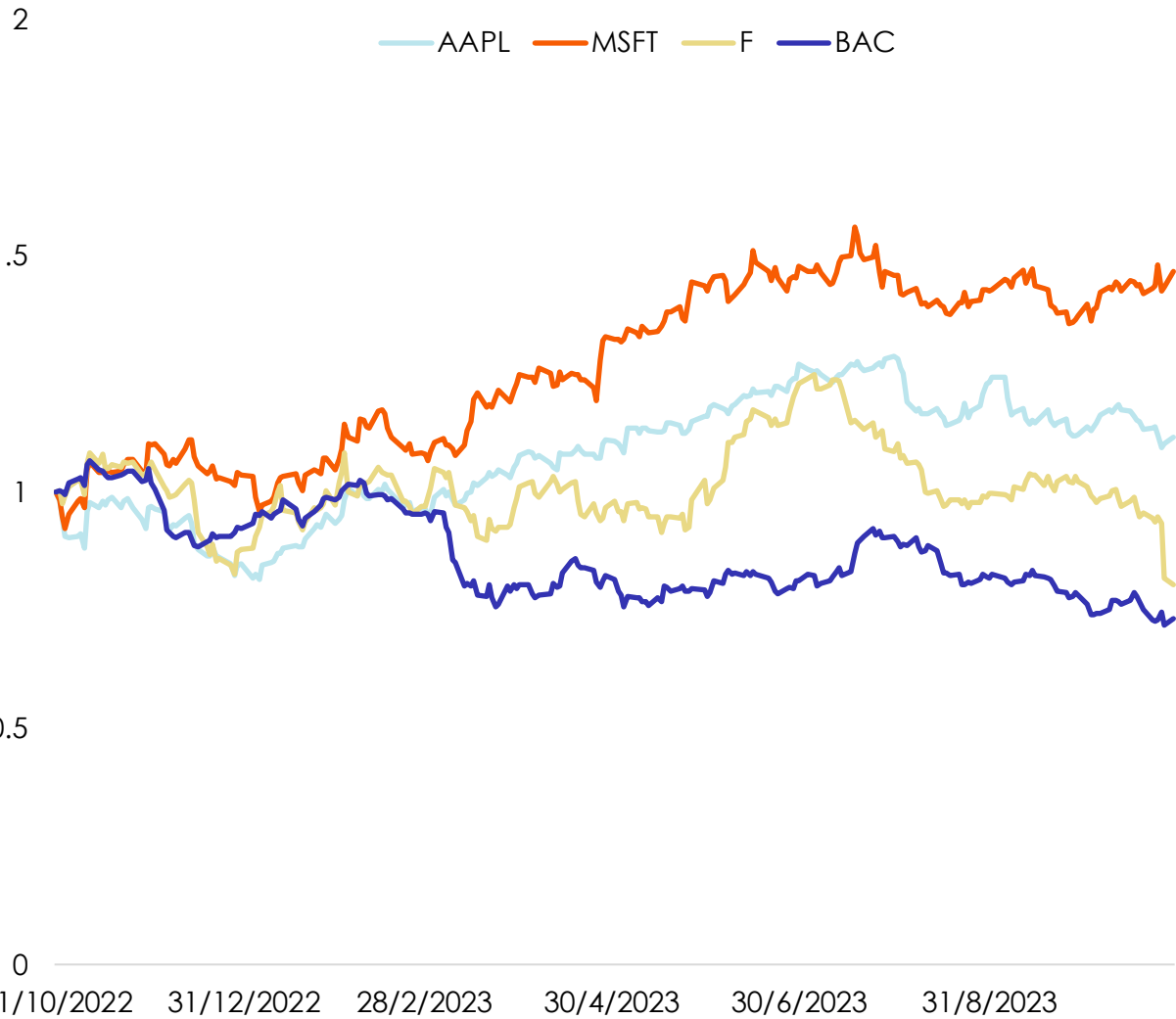
5%

(95% Confidence)



Stock Portfolio Overview

Historical Stocks Movements



Holdings @ 30th Oct 2023



Value: \$1,000,000
Shares: 5,887 @ \$169.85



Value: \$1,000,000
Shares: 2,970 @ \$336.63



Value: \$1,000,000
Shares: 103,950 @ \$12.74

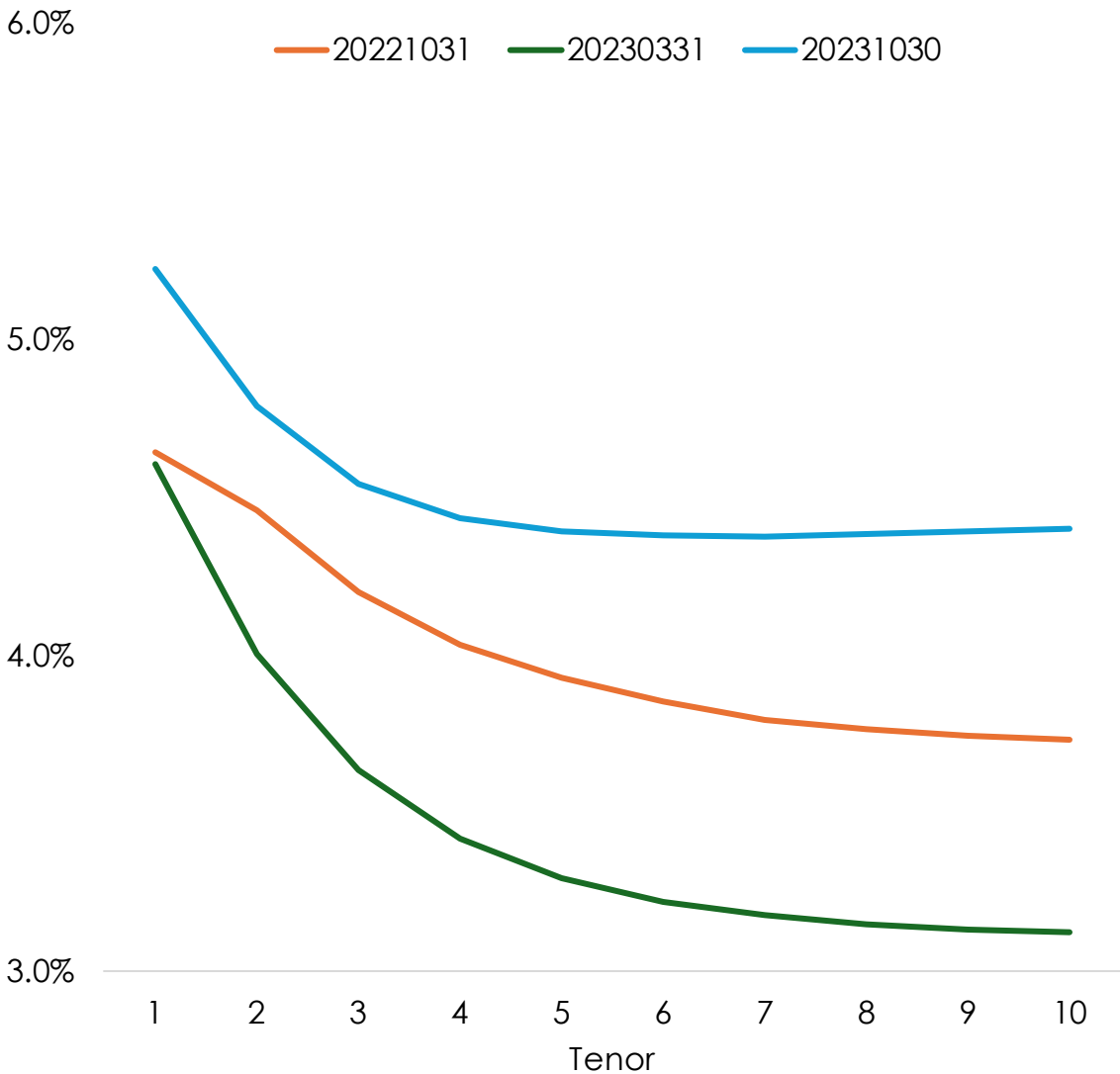


Value: \$1,000,000
Shares: 39,236 @ \$25.49



Swap Portfolio Overview

SOFR Zero Rate Curves



Payer Swap Valuation @ 30th Oct 2023



Start Date: 30th October 2023



Notional: \$100 Million, Fixed Rate: 4.2%



Float Leg Frequency: Annual
Float Leg = $1 - D(0, 10Y)$



Fixed Leg Frequency: Annual
Fixed Leg = $4.2\% * \sum_{i=1}^{10} D(0, i)$

Swap Value Today

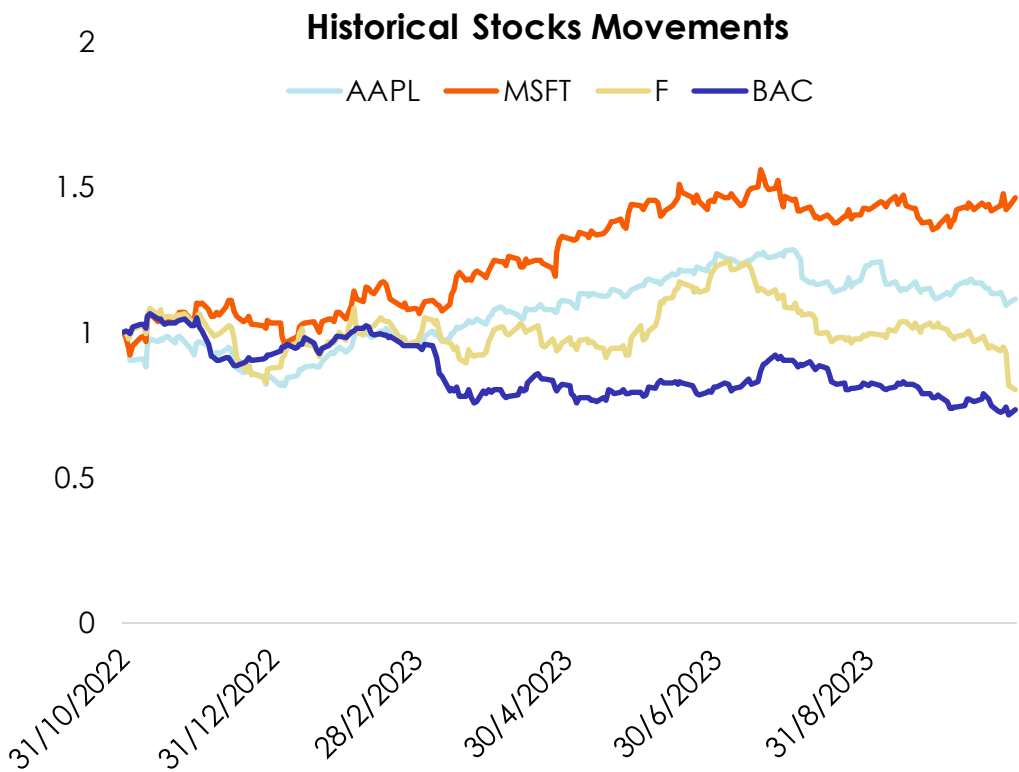
$Notional * (Float Leg - Fixed Leg)$

\$2,442,902



Stock

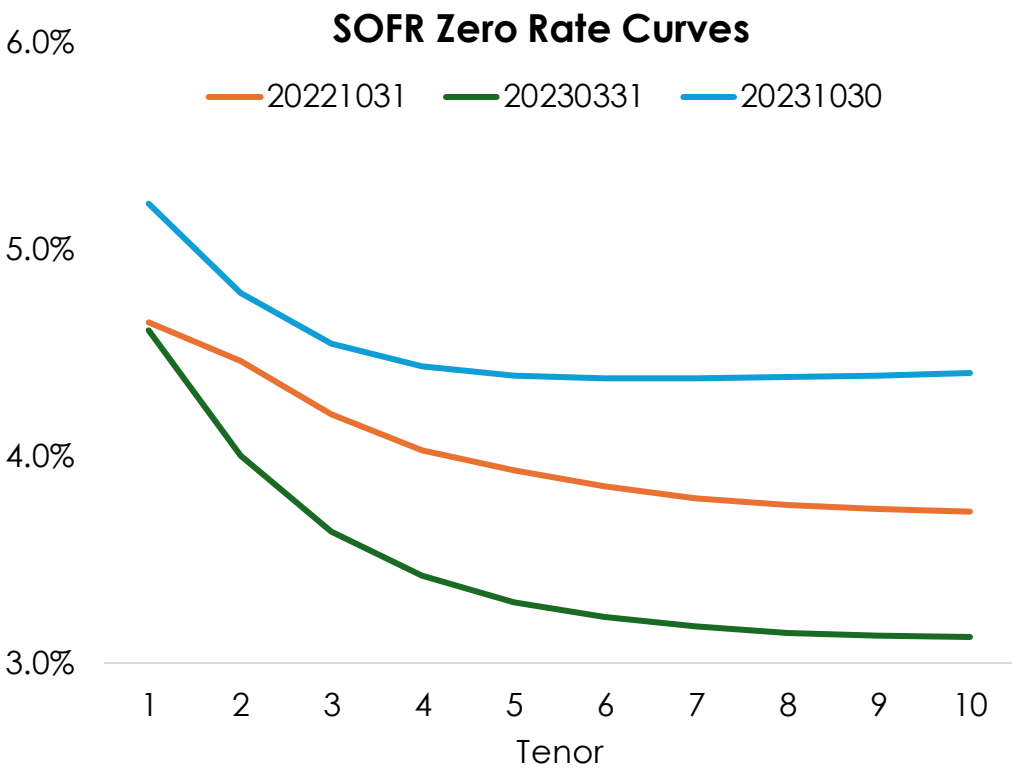
Risk Factor: Price



Daily % Change In Price

SOFR Rates

Risk Factor: 1-10y Zero Rates



Absolute Daily SOFR Rate Change



Stock

Date	Adj Close
14/6/2023	183.2264
15/6/2023	185.2783
16/6/2023	184.1926
19/6/2023	
20/6/2023	184.2823
21/6/2023	183.2364
22/6/2023	186.2644

*AAPL Data



19th June 2023

7th April 2023

SOFR Rates

Tenor	20231005	20231006	20231009	20231010	20231011
1D	0.052979	0.052939		0.052965	0.053028
1M	0.053099	0.053121		0.05309	0.053101
2M	0.053335	0.053462		0.053304	0.053256
3M	0.053532	0.053725		0.053488	0.053444
6M	0.053658	0.053969		0.053555	0.053602
9M	0.053223	0.053644		0.053028	0.053118
1Y	0.052471	0.05294		0.052191	0.052308
2Y	0.04805	0.048569		0.047529	0.047714
3Y	0.045262	0.045783		0.044683	0.044764



9th October 2023

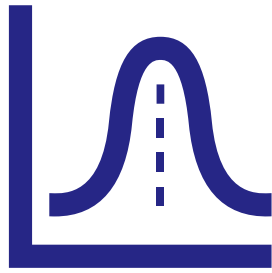
11th November 2023

Solution

Linear Interpolation

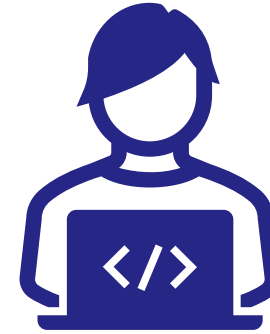
251 > 253 rows





Assumes a certain distribution
(typically normal)

μ & σ to estimate VaR



Simple & Fast Calculations

Parametric VaR (Stock)

Mean Returns

```
Stock_rets = Stocks.pct_change()
```

	Mean Return
AAPL	0.000568
MSFT	0.001692
F	-0.000619
BAC	-0.001105

Parametric mean
 $w * \mu = 536.21$

Covariance of Returns

```
Stock_rets = Stock_rets.cov()
```

	AAPL	MSFT	F	BAC
AAPL	0.000252	0.000183	0.000135	0.000075
MSFT	0.000183	0.000320	0.000134	0.000070
F	0.000135	0.000134	0.000508	0.000168
BAC	0.000075	0.000070	0.000168	0.000268

Parametric Standard Deviation
 $\sqrt{w * \Sigma * w^T} = 53,662.35$



Parametric VaR (Swap)

Computing KR01 (For Every Tenor)

Tenor	20231030		20231030		Discount		Tenor	KR01 / %
1Y	0.052245	+1bp →	0.052345	Compute Discount Factor Continuous compounding e^{-rt} →	0.949001	Change In Swap Value →	1Y	3,986,007
2Y	0.047904		0.047904		0.908638		2Y	7,631,797
3Y	0.045429		0.045429		0.872592		3Y	10,993,001
4Y	0.044345		0.044345		0.837461		4Y	14,066,554
5Y	0.043928		0.043928		0.802808		5Y	16,854,751
6Y	0.043794		0.043794		0.768923		6Y	19,371,056
7Y	0.043779		0.043779		0.736053		7Y	21,632,336
8Y	0.043828		0.043828		0.704249		8Y	23,653,228
9Y	0.043915		0.043915		0.673522		9Y	25,447,769
10Y	0.044023		0.044023		0.643888		10Y	670,595,263

Mean Absolute Change In Rates

Tenor	Mean
1Y	0.000023
2Y	0.000013
3Y	0.000014
4Y	0.000016
5Y	0.000019
6Y	0.000021
7Y	0.000023
8Y	0.000025
9Y	0.000026
10Y	0.000027

$$w * \mu = 20,927.25$$

Covariance Of Absolute Change In Rates

	1Y	2Y	...	9Y	10Y
1Y	0.0000006434	0.0000006910	...	0.0000003809	0.0000003612
2Y	0.0000006910	0.0000008367	...	0.0000005335	0.0000005078
...
9Y	0.0000003809	0.0000005335	...	0.0000005202	0.0000005079
10Y	0.0000003612	0.0000005078	...	0.0000005079	0.0000004972

$$\sqrt{w * \sum * w^T} = 53,662.35$$



Parametric VaR (Portfolio)

Mean Change In Risk Factors

W		μ	
1Y	3,986,007	1Y	0.000023
2Y	7,631,797	2Y	0.000013
3Y	10,993,001	3Y	0.000014
4Y	14,066,554	4Y	0.000016
5Y	16,854,751	5Y	0.000019
6Y	19,371,056	6Y	0.000021
7Y	21,632,336	7Y	0.000023
8Y	23,653,228	8Y	0.000025
9Y	25,447,769	9Y	0.000026
10Y	670,595,263	10Y	0.000027
AAPL	1,000,000	AAPL	0.000568
MSFT	1,000,000	MSFT	0.001692
F	1,000,000	F	-0.000619
BAC	1,000,000	BAC	-0.001105

Covariance Risk Factors

	1y	2y	...	9y	10y	AAPL	MSFT	F	BAC
1y	6.4e-07	6.8e-07	...	3.8e-07	3.6e-07	-2.7e-07	-1e-06	1.8e-06	4.5e-06
2y	6.8e-07	8.3e-07	...	5.3e-07	5e-07	-8.4e-07	-1.5e-06	1.4e-06	4.5e-06
...
9y	3.8e-07	5.3e-07	...	5.1e-07	5e-07	-1.4e-06	-1.2e-06	-7.7e-07	1.7e-06
10y	3.6e-07	5e-07	...	5e-07	4.9e-07	-1.4e-06	-1.1e-06	-9.2e-07	1.6e-06
AAPL	-2.7e-07	-8.4e-07	...	-1.4e-06	-1.4e-06	0.00025	0.00018	0.00013	7.5e-05
MSFT	-1e-06	-1.5e-06	...	-1.2e-06	-1.1e-06	0.00018	0.00032	0.00013	6.9e-05
F	1.8e-06	1.4e-06	...	-7.7e-07	-9.2e-07	0.00013	0.00013	0.0005	0.00017
BAC	4.5e-06	4.5e-06	...	1.7e-06	1.6e-06	7.5e-05	6.9e-05	0.00017	0.00027

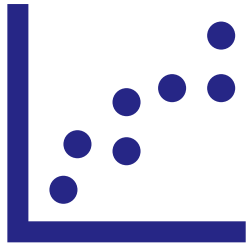
Parametric Mean: $w * \mu = 21,291.78$

Parametric SD: $\sqrt{w * \Sigma * w^T} = 576,604.50$

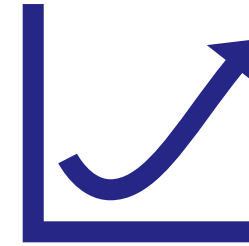
95% Parametric VaR

$|\mu + z * \sigma| = |21,291.78 + (-1.644) * 576,604.50| = \$927,138.23$





Simulate possible portfolio outcomes
using μ , σ^2 , & ρ



Does not require normal distribution

Ability to model complex, non-linear
relationships & path-dependent
options accurately

Sample Generation

```
np.random.default_rng().multivariate_normal(portfolio_mean,
                                             portfolio_cov, size = 2 ** 14 - 1, method = 'cholesky')
```

	1y	2y	3y	4y	5y	6y	7y	8y	9y	10y	AAPL	MSFT	F	BAC
0	-0.000122	0.000058	0.000197	0.000307	0.000348	0.000393	0.000442	0.000463	0.000465	0.000466	0.002634	0.023410	0.007983	0.029459
1	0.000453	0.000345	0.000203	0.000072	0.000058	0.000049	0.000015	-0.000004	-0.000009	-0.000012	0.011222	0.005479	0.001908	0.004710
2	-0.000696	-0.000712	-0.000590	-0.000531	-0.000554	-0.000581	-0.000586	-0.000574	-0.000559	-0.000561	0.004697	0.011977	0.006328	0.018192
3	0.002262	0.002291	0.002028	0.001801	0.001675	0.001612	0.001574	0.001554	0.001544	0.001535	0.012298	0.011177	-0.013796	-0.003296
4	-0.001105	-0.001044	-0.000730	-0.000555	-0.000374	-0.000294	-0.000293	-0.000301	-0.000300	-0.000290	0.041934	0.016775	0.034330	-0.011176
...
16378	0.000674	0.000660	0.000488	0.000427	0.000372	0.000315	0.000260	0.000204	0.000153	0.000114	-0.000379	-0.001437	0.011634	0.004586
16379	0.000783	0.000265	-0.000119	-0.000191	-0.000314	-0.000387	-0.000411	-0.000435	-0.000461	-0.000484	0.050214	0.040602	0.018115	0.013311
16380	0.000743	0.000816	0.000483	0.000206	-0.000041	-0.000197	-0.000283	-0.000354	-0.000414	-0.000453	0.029472	0.011619	0.004441	0.011247
16381	0.000489	0.000928	0.000804	0.000702	0.000579	0.000477	0.000410	0.000334	0.000263	0.000219	0.012512	-0.021983	0.000260	-0.016049
16382	0.000148	0.000434	0.000600	0.000557	0.000551	0.000597	0.000635	0.000631	0.000599	0.000565	-0.017340	-0.005116	0.013078	0.010021



Monte Carlo VaR Methodology (Stock)

Full Revaluation

Stock

Applying Risk Factor

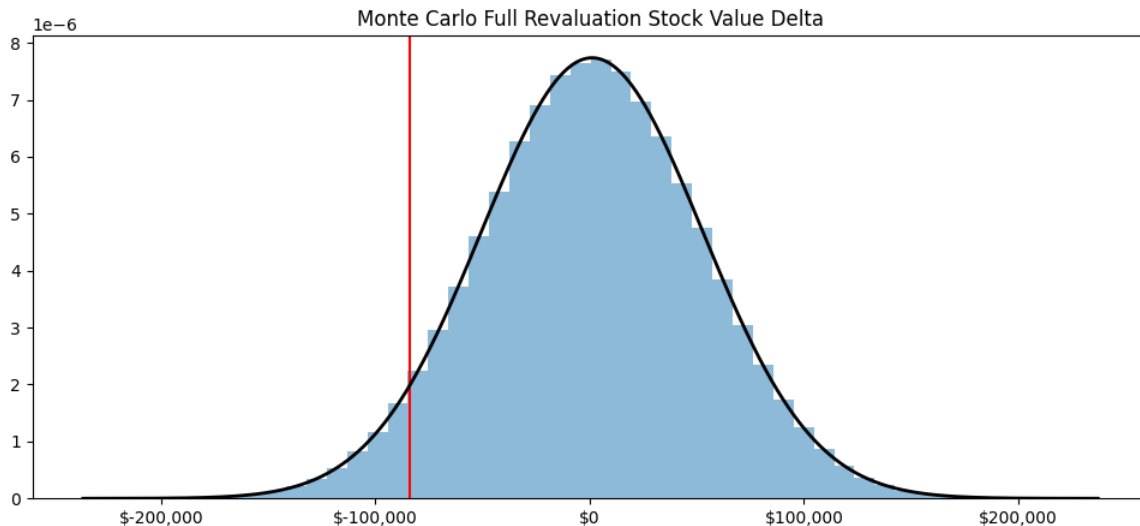
$$\text{Current Stock Price} * (1 + \text{Sim Return})$$

Computing New Stock Value

$$\text{Share Holding} * \text{New Stock Price}$$

Computing Stock PnL Change

$$\text{New Portfolio Value} - \$1,000,000$$

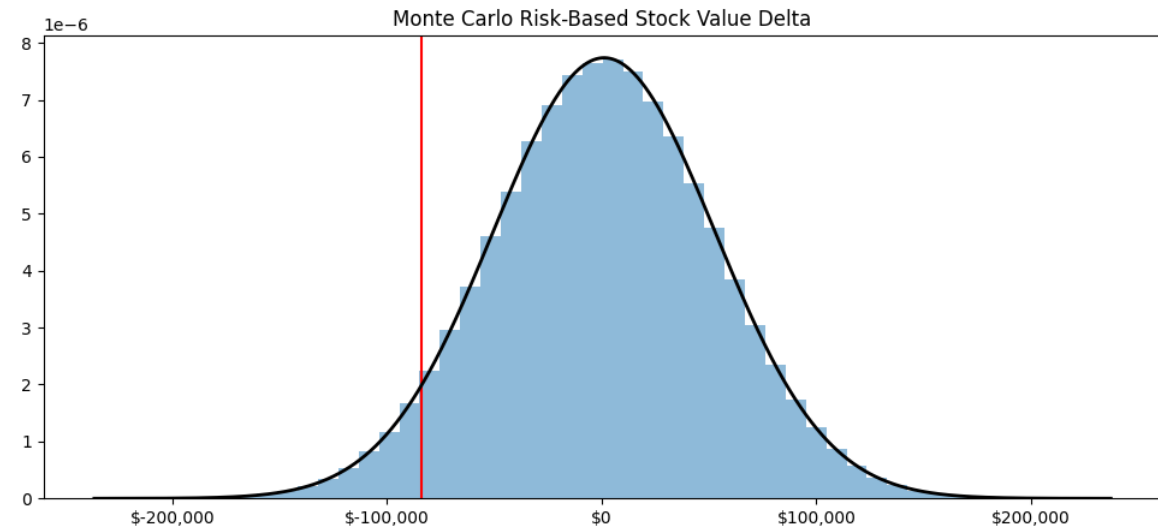


Risk-Based

Stock

Computing Stock PnL Change

$$\$1,000,000 * \text{Sim Return}$$



Monte Carlo VaR Methodology (Swap)

Full Revaluation

Swap

Applying Risk Factor

*Current Rates + **Sim** Rate Change*

Computing New Discount Factor

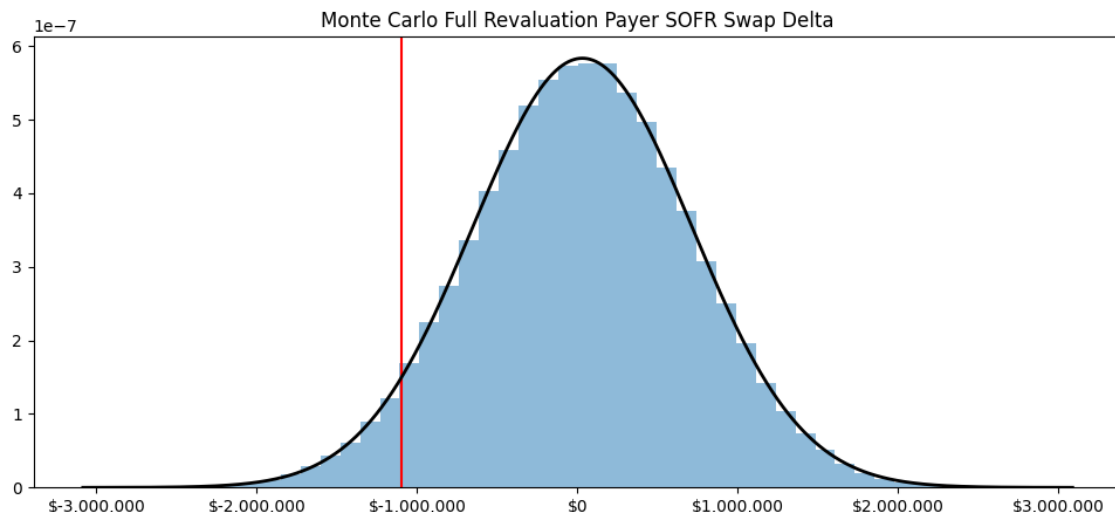
*$\text{Exp}(-\text{New Rate} * \text{Tenor})$*

Computing New Swap Value

*$\text{Notional} * (\text{Float Leg} - \text{Fixed Leg})$*

Computing Swap PnL Change

New Swap Value – \$2,442,902

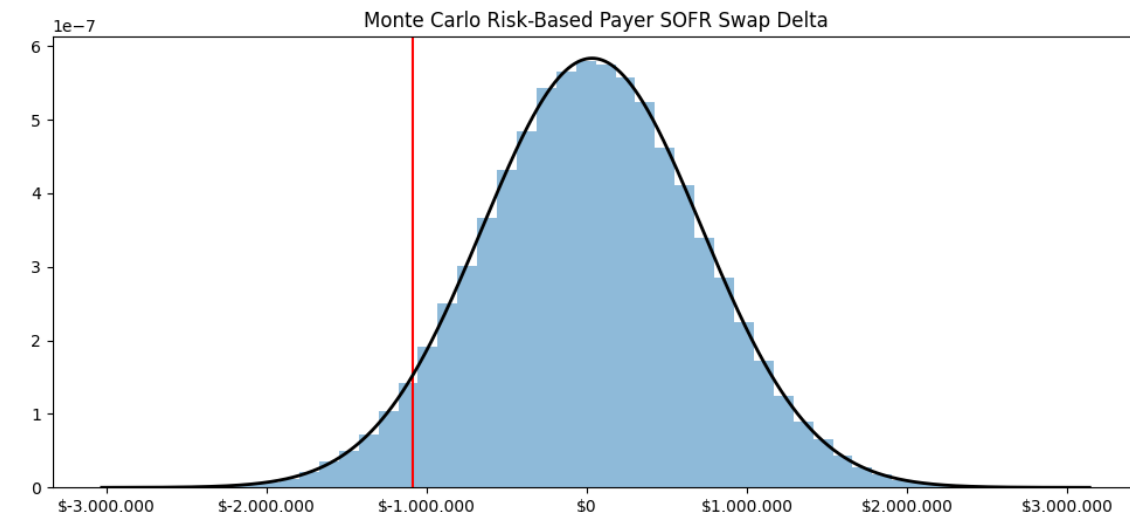


Risk-Based

Swap

Computing Swap PnL Change

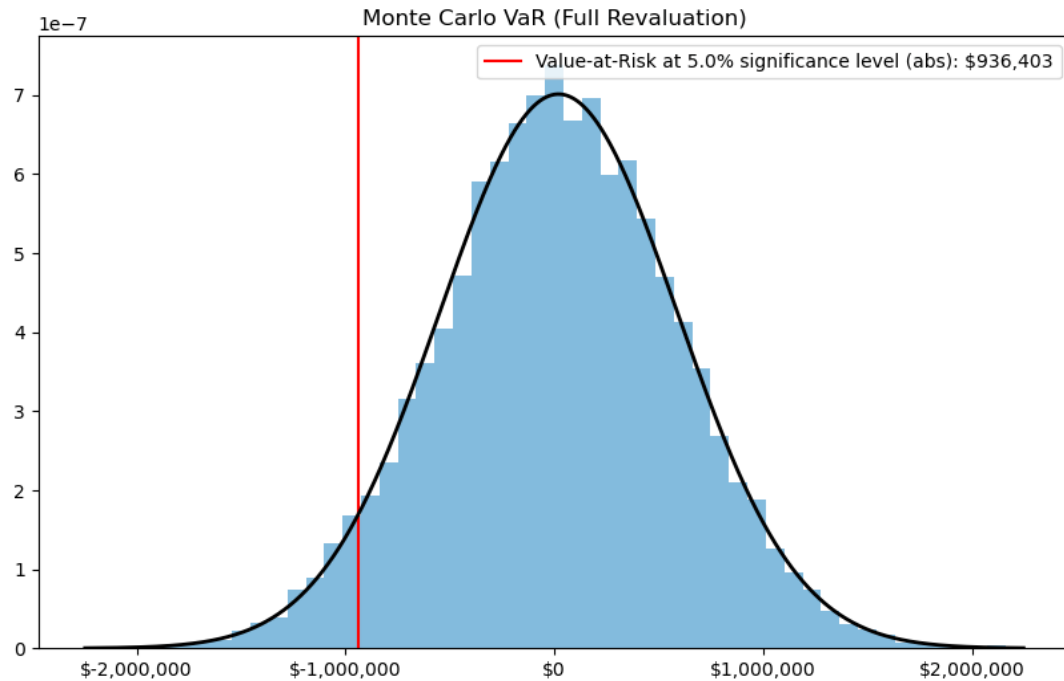
*$\text{KR01} * \text{Sim Rate Change}$*



Monte Carlo VaR (Portfolio)

Full Revaluation

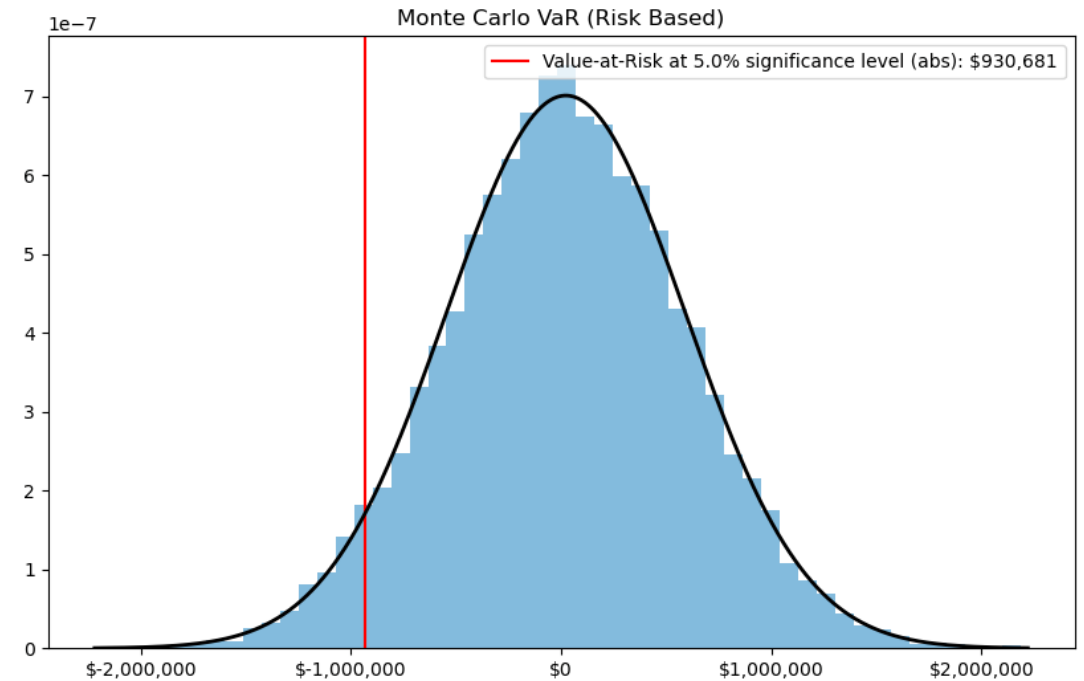
```
abs(np.percentile(MC_full_reval_portfolio, 5))
```



Full Revaluation 95% VaR
\$936,402.89

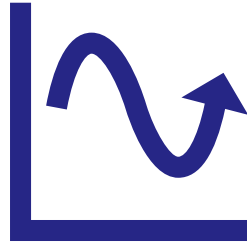
Risk-Based

```
abs(np.percentile(MC_Risk_Based_portfolio, 5))
```



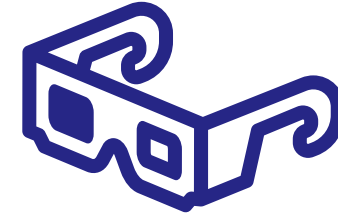
Risk-Based 95% VaR
\$930,680.76





Calculates VaR directly from **historical returns**, assuming historical movements are a good indicator of future risks

No assumptions about return distribution
straightforward & easy to understand



Backwards-looking, may not be as effective in capturing the risk of new or unprecedented market events

Less responsive to recent market conditions if the historical window is too long.

Historical VaR Methodology (Stock)

Full Revaluation

Stock

Applying Risk Factor

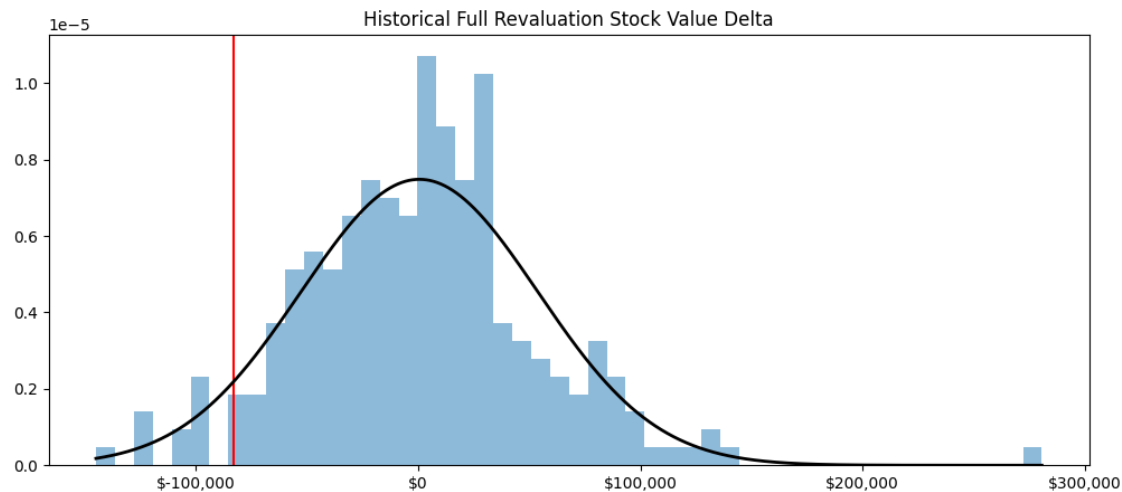
$$\text{Current Stock Price} * (1 + \text{Hist Return})$$

Computing New Portfolio Value

$$\text{Share Holding} * \text{New Stock Price}$$

Computing Stock PnL Change

$$\text{New Portfolio Value} - \$1,000,000$$

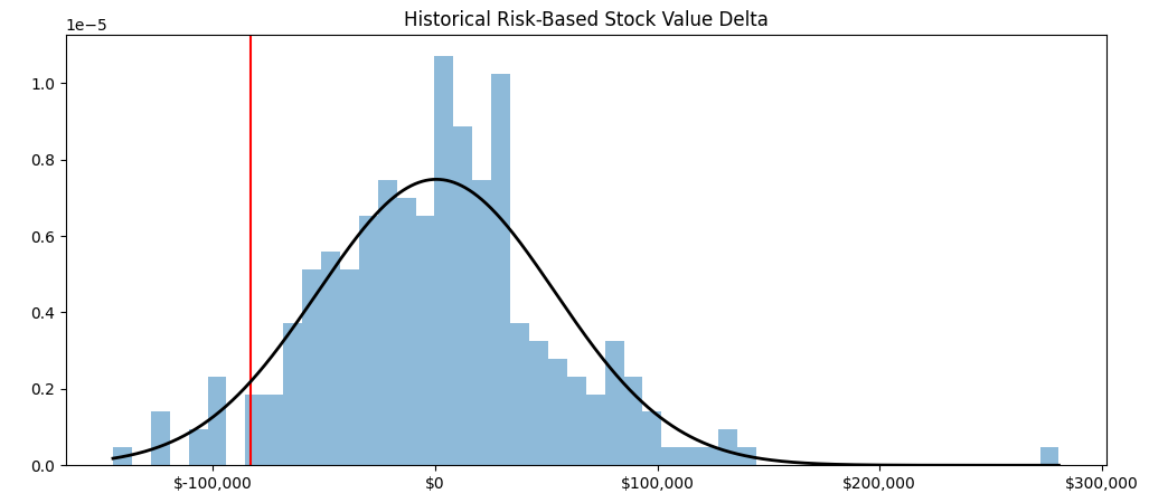


Risk-Based

Stock

Computing Stock PnL Change

$$\$1,000,000 * \text{Hist Return}$$



Historical VaR Methodology (Swap)

Full Revaluation

Swap

Applying Risk Factor

*Current Rates + **Hist** Rate Change*

Computing New Discount Factor

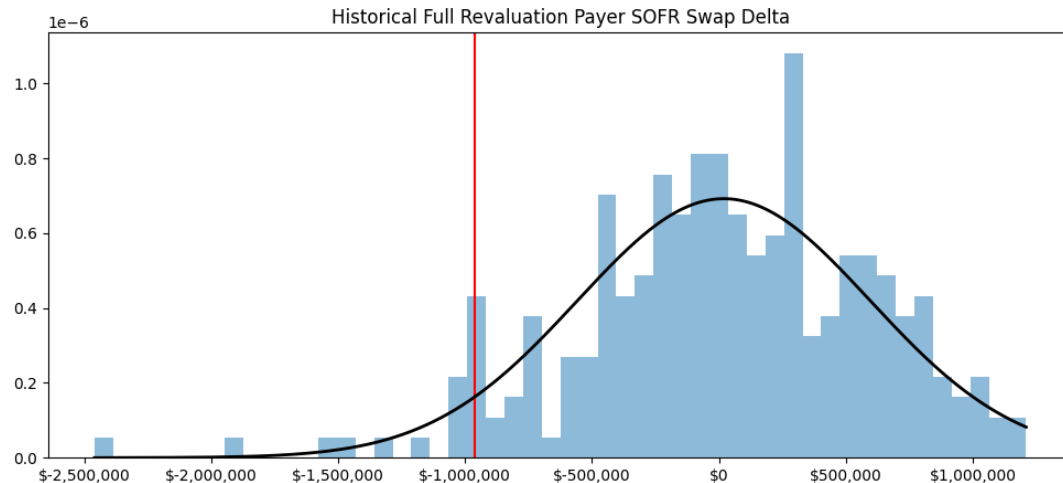
*$\text{Exp}(-\text{New Rate} * \text{Tenor})$*

Computing New Swap Value

*$\text{Notional} * (\text{Float Leg} - \text{Fixed Leg})$*

Computing Swap PnL Change

New Swap Value – \$2,442,902

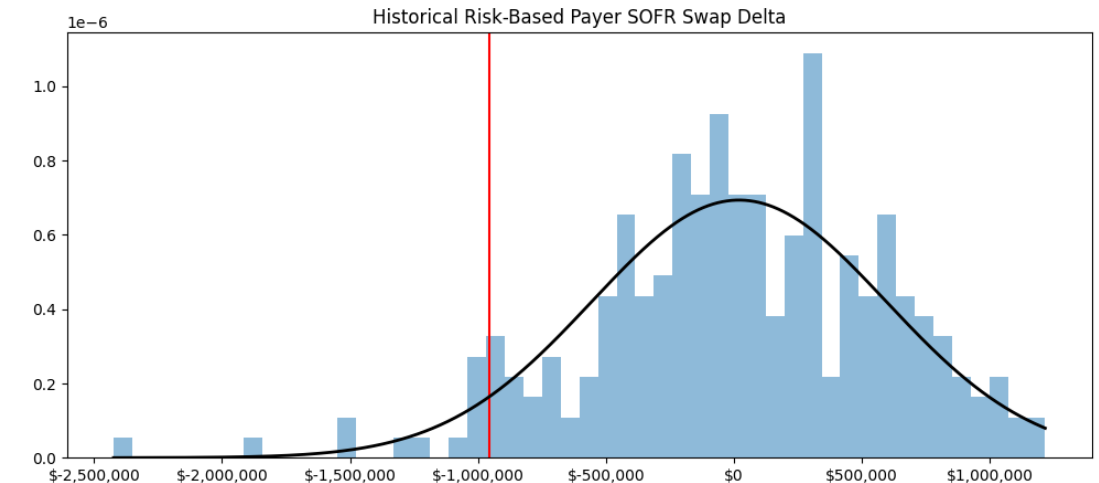


Risk-Based

Swap

Computing Swap PnL Change

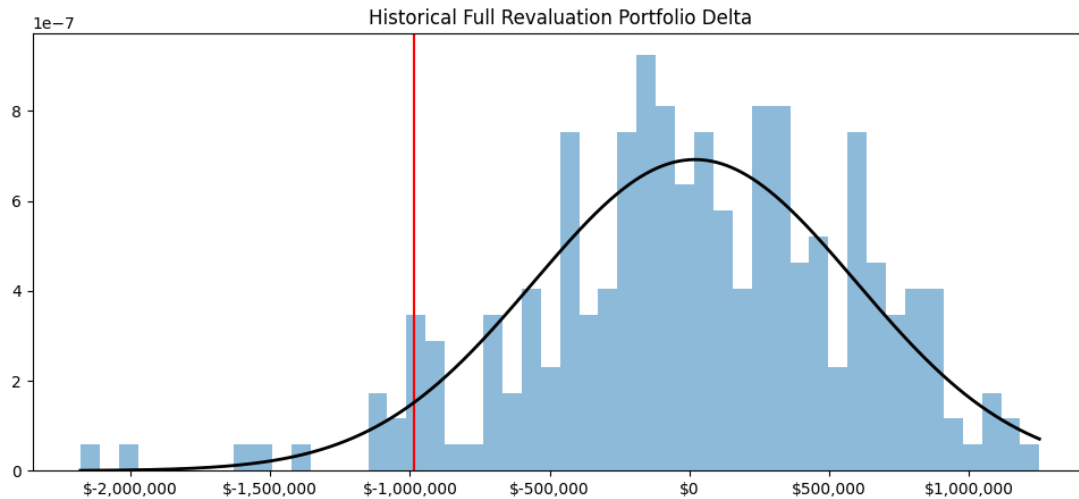
*$\text{KR01} * \text{Hist Rate Change}$*



Historical VaR (Portfolio)

Full Revaluation

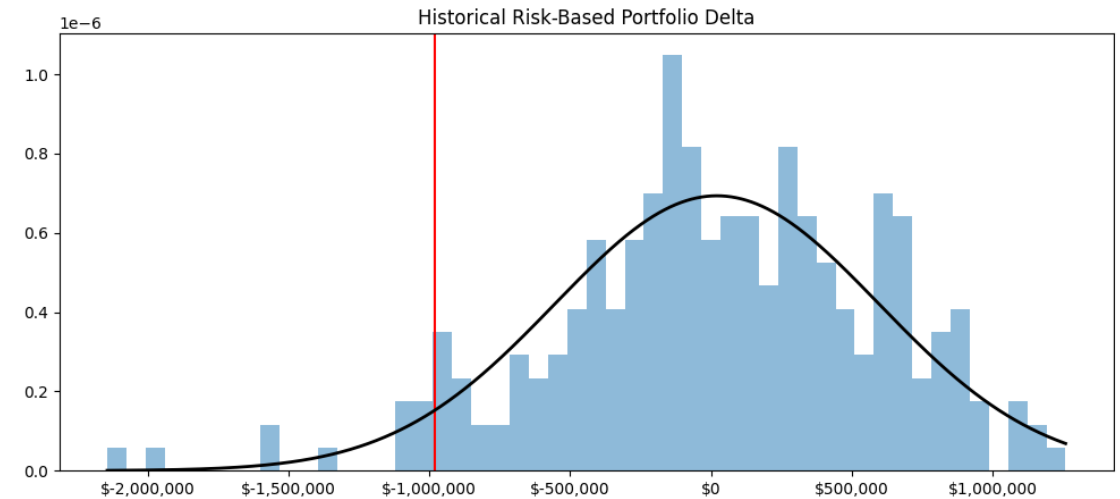
```
abs(np.percentile(Hist_full_reval_portfolio, 5))
```



Full Revaluation 95% VaR
\$984,781.39

Risk-Based

```
abs(np.percentile(Hist_Risk-Based_portfolio, 5))
```



Risk-Based 95% VaR
\$978,693.27

Parametric

- ✓ • Ease of application
- ✗ • Assumes risk factors normally distributed, market tend to have fatter tails (leptokurtosis)
- ✗ • Poor at handling non-linear exposures.

Monte Carlo

- ✓ • Accounts for both linear & non-linear risks.
- ✓ • Incorporate various scenarios, including extreme events not seen in the historical data.
- ✗ • Many assumptions on distribution, most computational effort
- ✗ • Results can vary significantly based on model assumptions
- ✗ • Requires detailed data on all portfolio components

Historical

- ✓ • No need to assume any particular distribution (Nonparametric)
- ✗ • Backwards-looking: Assumes past performance is representative of the future
- ✗ • Less responsive to recent market conditions.

Need to Weigh

Available computational resources
Acceptable level of approximation vs detail in risk estimation

1 Day 95% Value-at-Risk

Parametric	\$927,138.23	
	Full Revaluation	Risk-Based
Monte Carlo	\$936,402.89	\$930,680.76
Historical	\$984,781.39	\$978,693.27