



A REPORT ON THE VIABILITY OF A RISK PARITY
PORTFOLIO IN SOUTH AFRICA

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1 Introduction

Portfolio management has long been concerned with the trade-off between risk and return, with the Markowitz (1952) mean-variance framework serving as a foundational model in modern finance. This pioneering work introduced the concept of optimising portfolios by maximising expected return relative to risk, defined as variance. However, despite its theoretical significance, the mean-variance approach has faced substantial criticism over the years, particularly in its sensitivity to estimation errors and its tendency to concentrate risk in a few assets. These limitations became especially apparent during the 2008 global financial crisis, prompting a renewed interest in alternative portfolio construction methods that focus on risk diversification rather than just return optimisation.

One such alternative is the risk parity approach, which has gained traction among both academics and practitioners for its ability to equalise risk contributions across portfolio assets. Unlike traditional mean-variance optimisation, which often leads to concentrated risk exposure, risk parity ensures that each asset contributes equally to the total portfolio risk. This method has been particularly appealing to institutional investors, such as pension funds, seeking to construct more resilient portfolios. This essay investigates the application of the risk parity approach in a South African context, comparing its performance against two benchmark strategies: the Maximum Sharpe ratio portfolio and the equally weighted portfolio. By analysing these approaches, the study aims to assess the validity and practical effectiveness of risk parity in emerging markets like South Africa.

2 Literature Summary

The concept of risk-based portfolio construction has evolved significantly since the Markowitz (1952) seminal article. His model, while mathematically elegant, has been criticised for its reliance on precise estimates of expected returns and covariances, which are often subject to significant forecasting errors (Jagannathan & Ma, 2003). Moreover, empirical evidence suggests that mean-variance optimisation tends to concentrate capital in a few assets with high expected returns, inadvertently increasing the portfolio's vulnerability to shocks (DeMiguel, Garlappi & Uppal, 2009).

To address these shortcomings, risk-based allocation strategies have emerged as viable alternatives. One such approach is minimum variance portfolio optimisation, which seeks to minimise overall portfolio variance. However, this method still suffers from the issue of risk concentration, as assets with lower volatility tend to dominate the portfolio (Clarke, De Silva & Thorley, 2011). The risk parity approach, on the other hand, aims to distribute risk more evenly among assets, making it a more robust and resilient allocation strategy (Maillard, Roncalli & Teïletche, 2010). The first risk parity fund, Bridgewater Associates’ “All Weather” fund, was introduced in 1996 and gained prominence following the 2008 financial crisis due to its relative outperformance compared to traditional portfolio strategies (Asness, Frazzini & Pedersen, 2012).

Empirical studies on risk parity have demonstrated its effectiveness in enhancing portfolio stability. For instance, Qian (2005) highlights that risk parity portfolios tend to achieve higher risk-adjusted returns by avoiding concentrated risk exposure. Further research by Anderson, Bianchi & Goldberg (2012) supports the claim that risk parity portfolios exhibit lower draw-downs and greater resilience in periods of market turbulence compared to mean-variance optimised portfolios. Additionally, studies have examined the regional applicability of risk parity, with evidence suggesting its potential effectiveness in both developed and emerging markets (Lee, 2011).

Despite its theoretical appeal, the risk parity approach is not without its critics. Some portfolio managers argue that it inherently overweights low-volatility assets, such as bonds, leading to suboptimal returns in prolonged equity bull markets (Clarke, De Silva & Thorley, 2013). Others point out that risk parity relies heavily on historical volatility estimates, which may not fully capture the dynamic nature of financial markets (Roncalli, 2013). However, the strategy remains widely used by institutional investors seeking to mitigate downside risk while maintaining diversified exposure.

Given the increasing relevance of risk parity in global portfolio management, this study seeks to extend the literature by evaluating its application within South Africa’s financial landscape. By comparing the performance of a risk parity portfolio, a maximum Sharpe ratio portfolio, and an equally weighted portfolio, the analysis will provide insights into the robustness of risk parity in an emerging market setting. The findings will contribute to the ongoing debate on the efficacy of risk-based allocation strategies and their role in modern portfolio management.

3 Data

The data used for this report is a collection of local indices, to be used as proxies for different assets and sectors, as well as the rebalance dates for the portfolio during the period of interest. The following indices are used.

- **J203**: The JSE All Share Index (ALSI) as a proxy for equities.
- **ALBITR**: A composite bond index as a proxy for money market assets.
- **J311**: An index for the 25 largest industrial equities.
- **J212**: An index for the top 15 financial equities.
- **J310**: An index for the 10 largest resources equities, a proxy for commodities.
- **J253**: An index of companies in the Real Estate sector, a proxy for real estate assets.
- **STeFI**: Index for short-term fixed-interest assets, used as the risk-free rate.

This data consists of monthly returns from December 1999 until November 2024. However, some indices only have data starting from a later date, the latest being the start of 2002. For these indices the missing values were randomly drawn from the distribution of returns, based on the valid observations of that index.

4 Methodology

The risk parity approach to portfolio optimisation is done using the methods laid out in this section. This is an alternative method of diversification to something like an equal weighting portfolio, where all assets are allocated to equally in order to achieve diversification. Rather than the focus being on equal weights of assets, the focus shifts to the assets having equal contributions to the total risk of the portfolio.

From Euler's theorem, the risk of the portfolio $\sigma(w)$ is given by:

$$\sigma(w) = \sqrt{w^T \Sigma w}$$

which can be expanded to:

$$\sigma(w) = \sum_{i=1}^N w_i \frac{\partial \sigma}{\partial w_i} = \sum_{i=1}^N w_i \frac{(\Sigma w)_i}{\sqrt{w^T \Sigma w}}.$$

The risk contribution (RC) from the i th asset to the standard deviation $\sigma(w)$ is then defined as:

$$RC_i = w_i \frac{(\Sigma w)_i}{\sqrt{w^T \Sigma w}}$$

The risk parity approach includes a constraint into the optimisation process such that the RC's of all assets should be equal to one another.

$$RC_i = \frac{1}{N} \sigma(w)$$

This creates a portfolio where assets with lower standard deviations, less risk, receive higher weights. However, all assets still contribute to the final portfolio, as opposed to the popular Maximum Sharpe Ratio approach which tends to prefer high investment into specific asset classes and is more sensitive to small errors in parameter estimation (Qian, 2016). A Maximum Sharpe Ratio portfolio is also calculated for the same assets across the same time frame in order to be able to make comparisons to the risk parity portfolio and make conclusions. The Maximum Sharpe Ratio portfolio is based on the following functions. The Sharpe ratio is the difference in expected returns of the portfolio and the risk-free rate, divided by the standard deviation of the portfolio. The risk free rate in this report is taken as the rate of the Short-Term Fixed Interest (STeFI) index, which measures the performance of short-term fixed-interest instruments.

$$\max_w \frac{E[R_p] - R_f}{\sigma_p}$$

Where the expected returns ($E[R_p]$) and standard deviation (σ_p) is calculated as follows.

$$\text{where } E[R_p] = w^T \mu, \quad \sigma_p = \sqrt{w^T \Sigma w}$$

The sum of the weights of the assets must be equal to 1.

$$\text{subject to } \sum_{i=1}^N w_i = 1$$

A third approach is used for comparison, the Equal Weighting Portfolio (EWP) is one of the simplest methods of diversification, where all assets are given equal weights in order to get exposure to all asset classes. Comparing a Risk Parity Portfolio, a Maximum Sharpe Ratio Portfolio, and an Equal-Weight Portfolio provides valuable insights into the risk-return trade-offs and diversification benefits of risk parity investing. The Risk Parity Portfolio aims to allocate risk equally across assets, ensuring that no single asset class dominates the portfolio's overall risk. In contrast, the Maximum Sharpe Ratio Portfolio seeks to optimize the trade-off between expected return and risk, prioritizing assets that offer the highest risk-adjusted returns, which may lead to concentrated allocations in high-return, low-volatility assets. The Equal-Weight Portfolio, which assigns identical weights to all assets, serves as a baseline for comparison, offering simplicity and diversification without considering volatility or return expectations. By analysing these three approaches, one can assess whether risk parity delivers superior diversification benefits, how it compares in terms of return efficiency relative to the Maximum Sharpe portfolio, and whether its risk allocation provides advantages over a simple equal weighting approach. This comparison helps in understanding the validity of risk parity in achieving stable, well-balanced portfolio performance in the South African market.

5 Results

The first step of the investigation is to choose a method of risk parity calculation that can be used throughout this report. The first available methods is a naïve estimator, that assumes that assets are completely uncorrelated to each other for ease of computation, this is a very strong

assumption that is unlikely to hold in reality. The second method is the standard or vanilla method that uses non-linear multivariate root finders to find an optimal solution. The third is built on top of the vanilla method and includes the expected return as an additional objective in the optimisation problem, the parameter λ_μ controls the trade-off between minimising the risk parity function and maximising the expected returns.

Figure 1: Asset Risk Contribution and Weights for Different Portfolio Approaches

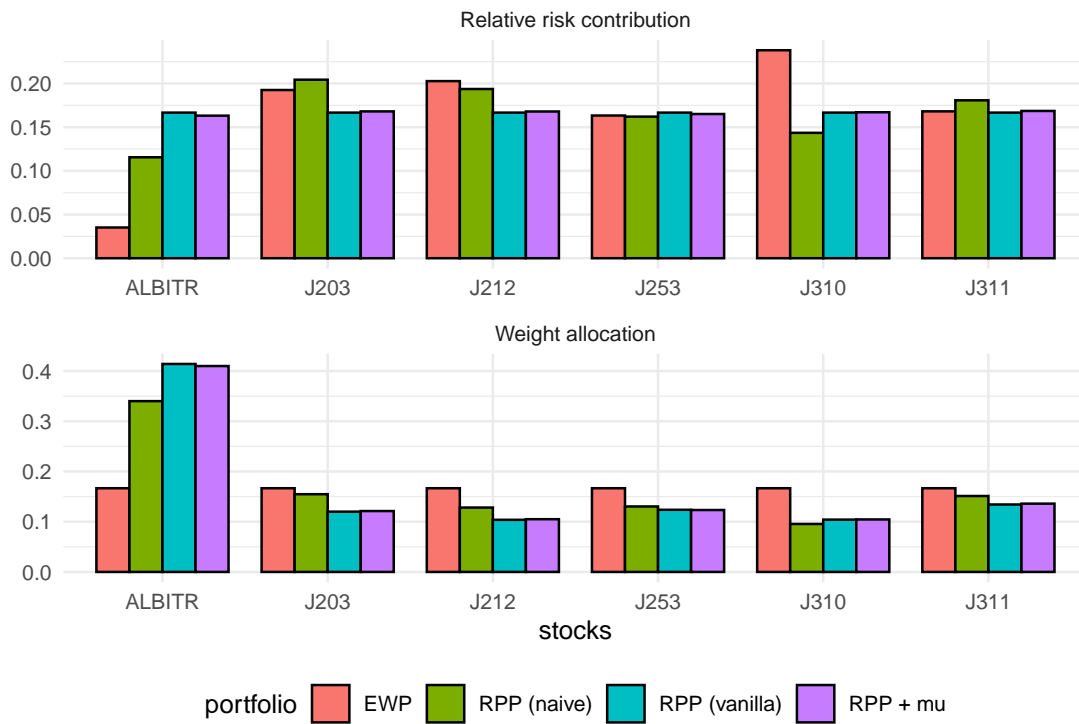


Figure 1 plots the relative risk contribution and weights of each asset for the different approach to the RPP, it also includes the equal weighting portfolio as a reference. It clear from the plot that the assumptions of the naïve estimator do not hold as the relative risks of the assets are not equal, the fundamental concept of RPP's. Whereas the equal weighting portfolio has equal weights in each asset, but each asset contributes a different level of risk, the RPP approach rather focuses on keeping the RC of each asset level and fluctuating the weights. There is value in a portfolio approach that can also take into account the expected returns of the portfolio, for this reason the RPP used for the remainder of the report is the RPP + mu estimation.

By altering the value of the trade-off parameter (λ_μ) I can plot the expected return versus how concentrated risk is in certain assets, for higher expected returns the portfolio must put more weight on asset with more risk. This plot is shown in *Figure 2*.

Figure 2: Expected Return vs Concentration of Risk

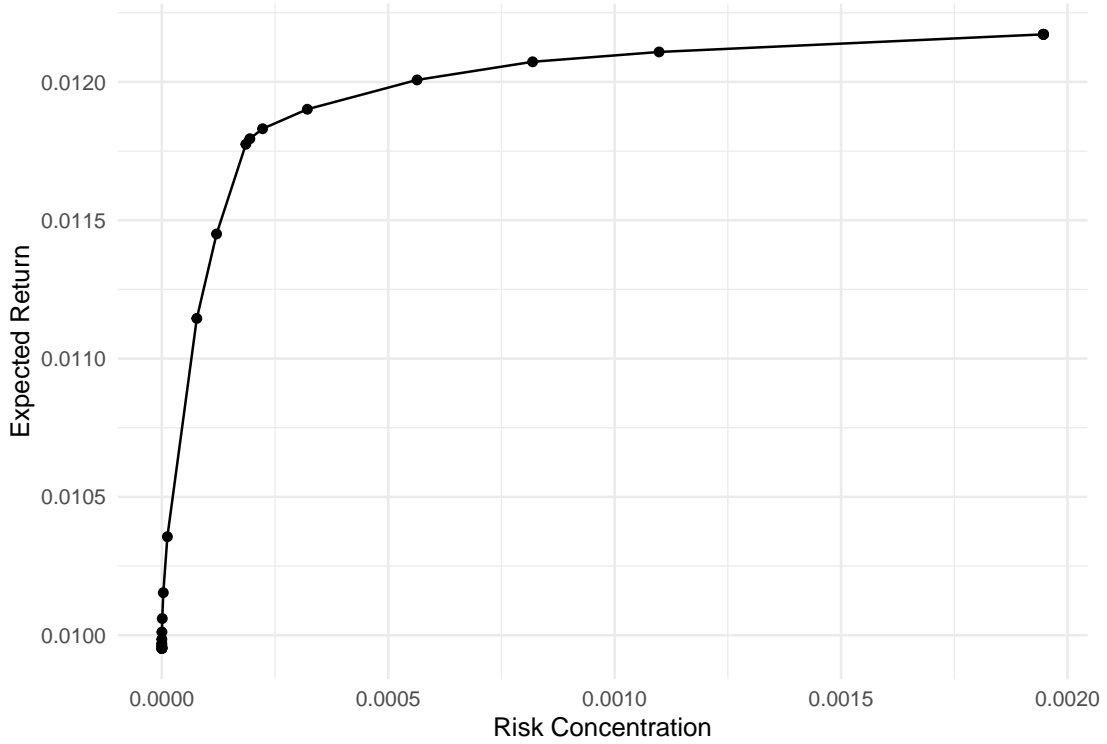
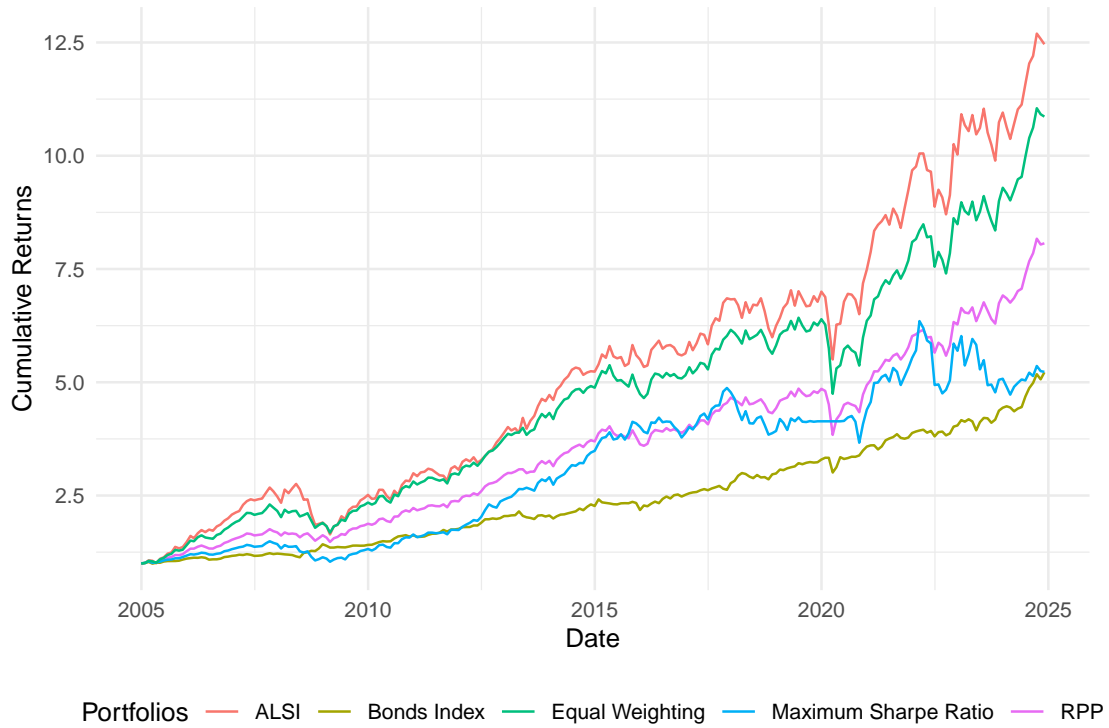


Figure 2 clearly shows diminishing returns in expected returns as risk becomes more concentrated, for the purposes of this report the value of λ_μ has been set to 0.00023 in order to maximise expected returns, without sacrificing the concentration of risk to a severe degree. Thus, for the remainder of this report, the RPP will include the expected returns term and have a λ_μ value of 0.00023.

I then plot the cumulative results of the RPP, equal weighting and Maximum Sharpe ratio portfolio, I also include the ALSI and Bond indices as references to provide the context of the assets with the most and least variation. This is shown in *Figure 3*.

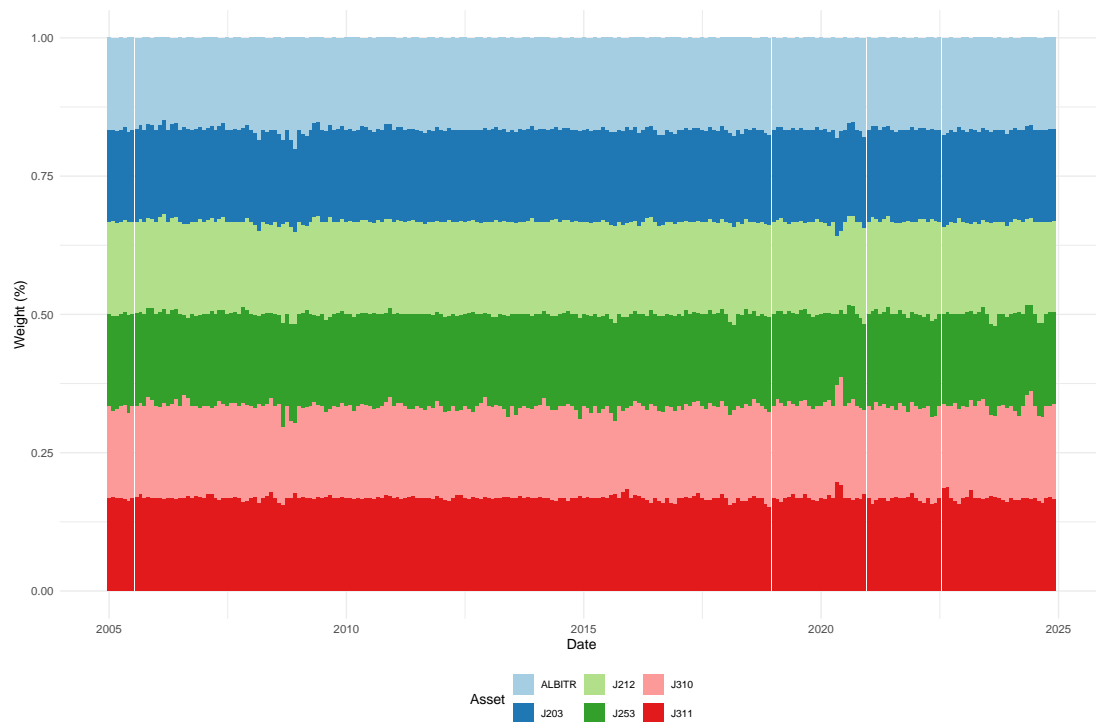
Figure 3: The Cumulative Returns of Portfolios and Certain Assets



From *Figure 3* it is clear that the Equal Weighting portfolio outperforms the other two, this is to be expected as this portfolio invests more heavily into assets that often carry more risk, rather than diversifying more along reduced risk lines as the Maximum Sharpe portfolio or RPP does. However, the RPP outperforms the Maximum Sharpe portfolio across the specified time period. The cumulative returns of all three portfolios are widely distributed between the cumulative returns of the ALSI and ALBITR Bond Index.

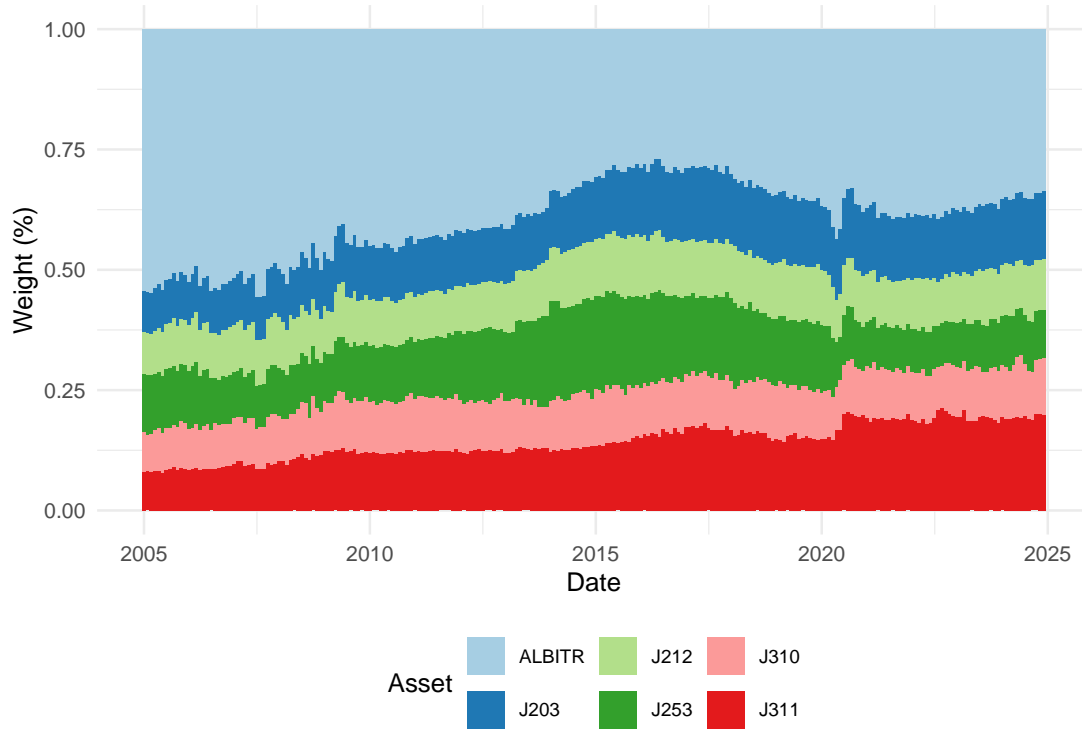
Next the I look at the historical weights for each portfolio to understand the changes in asset allocation throughout time. The optimal weights for each portfolio is calculated with a five year lookback period to calculate the optimal weights. The first figure, *Figure 4*, shows the historical weights of the equal weighting portfolio with the portfolio being rebalanced every quarter back to equal weights for every asset.

Figure 4: Portfolio Weights for Equal Weighting Portfolio Over Period



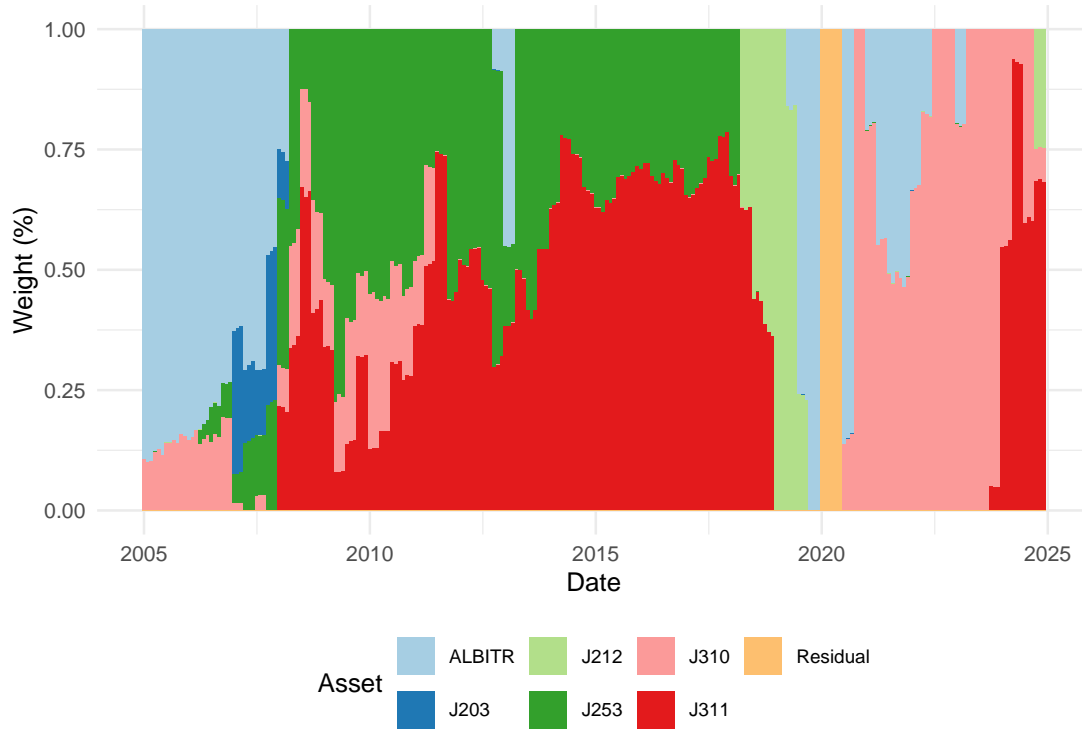
What is interesting in the above plot is the surge in the bonds weight in 2008 as other assets depreciated, the same can be seen for industrials during the COVID-19 pandemic in 2020. However, by construction, the weights of the portfolio remains relatively stable throughout the period. The next figure shows the change in portfolio weights for the RPP.

Figure 5: Portfolio Weights for Risk Parity Portfolio Over Period



In *Figure 5* we can see that the weights of each asset class fluctuates more than in the equal weighting portfolio, however each asset maintains a weight in the overall portfolio throughout the period. We see a consistent growth in the weights of real state from 2008 until 2017, while bonds slowly gets reduced weights over the same time period. The Covid period saw a quick spike in Bond weights before a spike in the industrials index directly after. The index for bonds maintains the largest weight throughout the period, this is expected as the bonds asset class is historically known for low risk, which in turn means a greater weight in order to have the same contribution of total risk as other assets. The following plot shows the weights of the Maximum Sharpe ratio portfolio.

Figure 6: Portfolio Weights for Risk Parity Portfolio Over Period



It is clear from looking at *Figure 6* that the weights of the Maximum Sharpe ratio portfolio changes considerably more drastically than the other two portfolios. This approach often prefers one or two assets rather than the constant 6 assets of the other portfolios. The drastic changes based on the calculated optimisation problem also open the portfolio to large consequences to small errors or misjudgments in calculation. During the period at the start of 2020 the historical mean returns above the risk-free rate were extremely close to zero, thus the portfolio suggested that full portfolio rather be allowed to appreciate at this risk free rate. This portfolio shows a preference towards a balance of industrial and real estate assets throughout the middle of the total period. However, more recently more preference has been given to the commodities assets.

Figure 7: Annualised 12 Month Rolling Returns



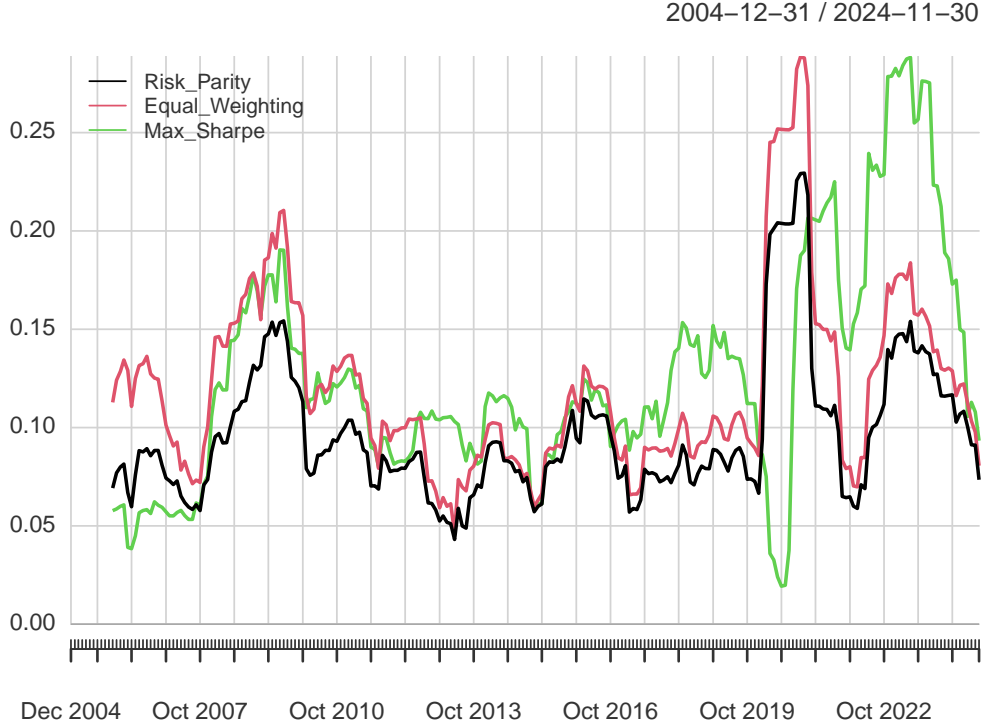
In *Figure 7* I plot the rolling one year returns for the portfolios, it is clear to see that the returns of the equal weighting portfolio and RPP are highly correlated, with the difference coming largely in the variance as the equal weighting portfolio tend to peak higher and trough lower. In contrast the Max Sharpe portfolio moves similarly in certain periods, such 2008 and 2021, but differently in other. This returns for the Max Sharpe portfolio is below the other portfolios at the start and end of the period, but above for stretches from 2010 until 2016.

Figure 8: Drawdowns for Each of the Portfolios



The largest drawdowns that can be seen in *Figure 8* is during the financial crisis in 2008 and during the COVID-19 period in 2020. However, the RPP and equal weighting portfolios recovered quite quickly, showing resilience. However, the Max Sharpe portfolio did not recover as quickly and spent longer periods before recovering. It is also important to note that the equal weighting has persistently lower drawdowns than the risk parity alternative.

Figure 9: Rolling 12 Month Standard Deviation



In *Figure 9* we can once again see the impact of the 2008 global financial crisis and Covid pandemic. It is important to note that the minimum standard deviation of the Max Sharpe portfolio in 2020 is due to the portfolio recommending the risk free rate, at zero deviation, making the standard deviation during this period to likely be biased towards zero for this portfolio. We again see that the standard deviation is consistently lower for the RPP than the equal weighting portfolio, in this plot we can also see that the RPP has consistently lower risk than the Max Sharpe portfolio.

Table 1: Performance Metrics for the Portfolios of Interest

	Ann_Return	Ann_Risk	Sharpe_Ratio	Max_Drawdown	E_Shortfall
Risk_Parity	0.1116045	0.1011426	0.3387643	0.2102686	-0.0759396
Equal_Weighting	0.1280526	0.1280177	0.3877621	0.2705263	-0.0893404
Max_Sharpe	0.0874733	0.1360629	0.0856339	0.3041064	-0.0953286

Table 1 provides different measures for the risk versus reward trade off in the South African context. The RPP demonstrates a balanced approach, yielding an annual return of 11.11% with a relatively low annualised risk of 10.10%. While its Sharpe ratio (0.335) is lower than that of the Equal Weighting portfolio (0.388), it significantly outperforms the Max Sharpe portfolio (0.090), which exhibits the highest risk (13.60%) but the lowest return (8.81%). This suggests that the Max Sharpe portfolio, despite its theoretical objective of maximizing risk-adjusted returns, is suboptimal in this context. This is likely due to the lack of additional constraints on the maximum weights of certain assets, causing the model to heavily weight a small number of assets, increasing risk.

A key advantage of risk parity is its lower maximum drawdown (21.03%) compared to Equal Weighting (27.05%) and Max Sharpe (30.41%). Similarly, its expected shortfall (-7.60%), which measures the average loss in the worst 5% of cases, is the smallest among the three. This indicates that risk parity minimises extreme downside risk, a critical consideration in the South African market, which is at times subject to high volatility and historical macroeconomic uncertainty. While the RPP does not perform the best in terms of returns, it performs the best in minimising risk, especially downside risk. The Max Sharpe portfolio's poor performance underscores that blindly optimizing for return-to-risk ratios may not always lead to practical or robust investment strategies without additional considerations or restrictions.

6 Conclusion

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