

Statistical Validity of the Fama French Carhart Four Model

by

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I dedicate this thesis to my immediate family members: my parents, my brothers David and Alvin, and my sister-in-law Lily. Thank you for all of their love and support throughout my life. In particular, I would like to thank my parents for always believing in me, their continuous love, and their support in all my life decisions. Without whom I could not have made it here. My bothers David and Alvin have never left my side and are very special.

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Abstract

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We examine whether OLS is the right method to apply on the Fama French Carhart Four Model using the observed data of the four-factor and the monthly returns of each of the 10 portfolios formed monthly on momentum, where the 10 portfolios are constructed monthly using NYSE prior (2-12) return in ascending order with decile breakpoints. Next, we use Cook's distance and covariance ratio to measure the influence of an observation on betas and t-statistics respectively. Finally, we compare two mean-variance optimal portfolios, in which the portfolios are both comprised of the 10 portfolios formed on momentum but with two different weighted methods.

Chapter 1

Introduction

1.1 The Fama French Carhart Four Model

The Fama French Carhart Four Model [6] is a 4-factor model using Fama and French's 3-factor model plus an additional factor capturing Jegadeesh and Titman's one-year momentum anomaly. The Fama French Carhart Four Model is formulated as

$$r_{i,t} = \alpha_i + \beta_{i,RMRF}RMRF_t + \beta_{i,SMB}SMB_t + \beta_{i,HML}HML_t + \beta_{i,MOM}MOM_t + \epsilon_{i,t} \quad (1.1)$$

where $i = 1, \dots, 10$ indexes the 10 portfolios formed monthly on momentum factor, t indexes for month, r is the portfolio return in excess of the risk-free rate, RMRF, SMB, and HML are the Fama-French market, size, and value factors, MOM is the momentum factor, ϵ is the unobservable random error, and the various β s are the corresponding factor loadings. The paper focus on the empirical study of the equation (1.1) using the data from Kenneth R. French's Data Library.¹

1.2 Data

We are using the following three sets of data from Kenneth R. French's Data Library [11].

- Fama/French Factors
- Momentum Factor (MOM)
- 10 Portfolios Formed Monthly on Momentum

¹The Kenneth R. French's Data Library http://mba.tuck.dartmouth.edu/pages/faculty/ken.french/data_library.html contains data for U.S. research returns, historical benchmark returns, U.S. research breakpoints, U.S. book equity, international research returns, and developed market factors and returns.

1.2.1 Fama/French Factors

This dataset includes the monthly data for the three Fama/French Factors RMRF, SMB, and HML. It also includes the risk-free rate factor R_f so that we can get the market rate factor R_m by adding the risk-free factor to RMRF factor.

SMB (Small Minus Big) [11] is the average return on the three small portfolios minus the average return on the three big portfolios, which is equal to

$$\frac{1}{3}(Small\ Value + Small\ Neutral + Small\ Growth) - \frac{1}{3}(Big\ Value + Big\ Neutral + Big\ Growth)$$

HML (High Minus Low) [11] is the average return on the two value portfolios minus the average return on the two growth portfolios, which is equal to

$$\frac{1}{2}(Small\ Value + Big\ Value) - \frac{1}{2}(Small\ Growth + Big\ Growth)$$

RMRF ($R_m - R_f$) [11] is the excess return on the market, value-weight return of all CRSP firms incorporated in the US and listed on the NYSE, AMEX, or NASDAQ that have a CRSP share code of 10 or 11 at the beginning of month t , good shares and price data at the beginning of t , and good return data for t minus the one-month Treasury bill rate.

These Fama/French factors [11] are constructed using the 6 value-weight portfolios formed on size and book-to-market. The portfolios, which are constructed at the end of each June, are the intersections of 2 portfolios formed on size (market equity, ME) and 3 portfolios formed on the ratio of book equity to market equity (BE/ME). The size breakpoint for year t is the median NYSE market equity at the end of June of year t . BE/ME for June of year t is the book equity for the last fiscal year end in $t - 1$ divided by ME for December of $t - 1$. The BE/ME breakpoints are the 30th and 70th NYSE percentiles.

1.2.2 Momentum Factor (MOM)

This dataset includes the monthly data for the Momentum (MOM) factor [11], MOM is the average return on the two high prior return portfolios minus the average return on the two low prior return portfolios, which is equal to

$$\frac{1}{2}(Small\ High + Big\ High) - \frac{1}{2}(Small\ Low + Big\ Low)$$

1.2.3 10 Portfolios Formed Monthly on Momentum

This dataset includes the monthly data for return of 10 portfolios formed monthly on momentum.

The portfolios [11] constructed each month include NYSE, AMEX, and NASDAQ stocks with prior return data. To be included in a portfolio for month t (formed at the end of month $t - 1$), a stock must have a price for the end of month $t - 13$ and a good return for $t - 2$.

Stocks are sorted at the beginning of the first trading day of each month from January 1925 to August 2013 into decile portfolios based on their previous calendar year's return. We have two sets of data; one is average equal weighted (AEW) and the other one is average value weighted (AVW). The portfolios are weighted monthly so the weights are readjusted whenever a stock disappears. Stocks with the lowest NYSE prior (2-12) return comprise decile 1 and stocks with the highest comprise decile 10 [11].

1.3 Programming Language and Reproducibility

We use the open-source programming language R for the computation and data analysis that are necessary for our research. All the figures and other output appeared in this thesis are reproducible using the R codes in appendix G together with the three sets of data from Kenneth R. French's Data Library. We will also use GitHub to facilitate reproducible computational results and share our research. All our datasets and programming files, together with this thesis, can be found here: <https://github.com/henrykmwong>.

1.4 Multiple Linear Regression

In this chapter, we set up a multiple linear regression model for the equation (1.1) and identify the influential observations for each of the portfolios. The model [10] is:

$$Y = X\beta + \epsilon \quad (1.2)$$

On the left, Y is an $N \times 1$ vector of observable random variables. The Y vector is called the *dependent* or *response* variable. On the right, the first term is X , an $N \times p$ matrix of observable random variables, one column for each data variable. The X matrix is called the *design matrix* and the data variables are called the *explanatory* or *independent* variables. We assume that $N > p$ and the design matrix has full rank. The second term on the right is β , a $p \times 1$ vector of unknown coefficients or parameters to be estimated. The final term on the right is ϵ , an $N \times 1$ vector of unobservable random errors.

In the application of the Fama French Carhart Four Model on our data, N is the number of monthly observations, which is equal to 1,039 and p is the number of coefficient variables, which is equal to 5 (intercept plus coefficients for the four factors).

We will write Y_i for the i th observation of Y , X_i for the i th row of X , and ϵ_i for the i th component of ϵ . The matrix equation (1.2) is equivalent to a set of N ordinary equations, one for each observation. For the i th observation, the equation is

$$Y_i = X_i \beta + \epsilon_i \quad (1.3)$$

To estimate the β in equation (1.2), we use the method of ordinary least squares (OLS) with the following assumptions [10].

1. The ϵ_i are independent and identically distributed with mean zero and constant variance (or homoscedasticity) σ^2 .
2. If X is random, ϵ is independent of X .

Using the method of OLS, we have:

$$\hat{\beta} = (X^T X)^{-1} X^T Y \quad (1.4)$$

$$\hat{Y} = X \hat{\beta} \quad (1.5)$$

where

$$P = X(X^T X)^{-1} X^T \quad (1.6)$$

$$e = Y - \hat{Y} = (I - P)Y \quad (1.7)$$

$$\hat{\sigma}^2 = \frac{e^T e}{N - p} \quad (1.8)$$

$$\begin{aligned} \Psi &= cov(\hat{\beta}|X) \\ &= \sigma^2 (X^T X)^{-1} \end{aligned} \quad (1.9)$$

$$cov(\hat{Y}|X) = \sigma^2 P \quad (1.10)$$

and

$$cov(e|X) = \sigma^2 (I - P) \quad (1.11)$$

We denote the vector of estimated coefficients and the vector of fitted values by $\hat{\beta}$ and \hat{Y} . The $N \times N$ matrix P in equation (1.6) is called the *hat matrix*. It is also called the *projection matrix* because it projects Y into the column space of X . The vector of the (least squares) residuals is denoted by e in equation (1.7).

The following equation will be used in this chapter for identifying influential observations.

$$t_i = \frac{e_i}{\hat{\sigma} \sqrt{1 - p_i}} \quad (1.12)$$

where e_i is the i th component of the vector of residuals e and p_i is the i th diagonal element of P .

1.5 Regression Diagnostic

The OLS assumptions are difficult to check because ϵ is not observable [10]. One way to work around the difficulty is to use the OLS residuals.

We can use the Lilliefors test [12] to test the null hypothesis that the residuals come from the Gaussian distribution. The 2nd and 3rd columns of table 1.1 show the p-value of the Lilliefors test for each portfolio from each weighting method. We reject all the null that the residuals are Gaussian distributed with 5% α level and therefore the OLS estimates are not equivalent to the maximum-likelihood estimates. However, it does not violate the first OLS assumption as long as the residuals are independent and identically distributed with mean zero and constant variance.

Table 1.1: Lilliefors Test for Normality with 0.05 Alpha Level

portfolio	pvalAEW	pvalAVW	nullAEW	nullAVW
1	8.79e-40	1.65e-17	Reject	Reject
2	3.55e-28	8.10e-07	Reject	Reject
3	3.81e-24	2.31e-13	Reject	Reject
4	1.52e-15	2.17e-18	Reject	Reject
5	3.87e-22	1.03e-10	Reject	Reject
6	1.55e-20	2.57e-14	Reject	Reject
7	9.17e-20	1.92e-11	Reject	Reject
8	1.52e-24	8.64e-10	Reject	Reject
9	1.83e-19	2.79e-10	Reject	Reject
10	6.37e-15	9.20e-09	Reject	Reject

As shown in appendix A, each of the scatterplots of residuals versus the observed values of the independent variables does not show any pattern, therefore the second OLS assumption seem to be valid. For illustration, figure 1.1 shows the residual plots for portfolio 2 with average equal weighted (AEW) method. In addition, as shown in table 1.2, the mean of residuals for each portfolio and weighting method is equal to zero

with accuracy of 18 decimal places.

Table 1.2: Mean of Residuals

portfolio	meanresidAEW	meanresidAVW
1	1.03e-19	-2.62e-19
2	1.19e-18	2.00e-18
3	-3.45e-19	-1.01e-18
4	8.90e-19	9.34e-20
5	1.53e-19	1.27e-19
6	2.61e-20	2.03e-19
7	-2.36e-19	4.82e-19
8	5.49e-19	-2.78e-19
9	6.30e-19	-1.83e-20
10	1.32e-19	1.09e-19

However, based on the times series of the residuals in appendix A, the residuals seem not to be independent and identically distributed because the residuals have larger fluctuations before 1940 and after 1980. The time series plots also suggest that the homoscedasticity (constant variance) assumption is violated. We can use the non-constant variance score test [5] to test the null hypothesis that the residuals are distributed with constant variance. The 2nd and 3rd columns of table 1.3 shows the p-value of the non-constant variance score test for each portfolio and weighting method. We reject all the null that the residuals are distributed with constant variance with 5% α level except for portfolios 7 and 10 with average value weighted method.

Our failure of meeting some of the OLS assumptions suggests that our dataset contains influential observations. Chapter 2 will discuss the identification of the influential observations. Even though we do not meet all the OLS assumptions, we can still apply the method of OLS on the multiple linear regression model for the equation (1.1), but we need to exercise care in interpreting the results. In particular, we can use Newey-West t-statistic to handle the autocorrelation and heteroscedasticity problem, which will be discussed next.

1.6 Heteroskedasticity and Autocorrelation Consistent Covariance

When random errors exhibit autocorrelation and/or heteroskedasticity, $cov(\epsilon|X)$ will not be equal to $\sigma^2 I$, therefore equation (1.9) no longer holds. In other words, $\hat{\sigma}^2(X^T X)^{-1}$,

Table 1.3: Non-constant Variance Score Test with 0.05 Alpha Level

portfolio	pvalAEW	pvalAVW	nullAEW	nullAVW
1	5.27e-94	8.65e-16	Reject	Reject
2	1.35e-155	2.86e-24	Reject	Reject
3	4.28e-57	1.62e-06	Reject	Reject
4	6.37e-107	3.84e-65	Reject	Reject
5	2.05e-192	1.55e-42	Reject	Reject
6	2.41e-81	1.28e-03	Reject	Reject
7	5.47e-160	4.88e-01	Reject	Accept
8	6.09e-256	6.89e-29	Reject	Reject
9	1.70e-109	8.00e-20	Reject	Reject
10	1.55e-11	2.52e-01	Reject	Accept

denoted by $\hat{\Psi}$, will be a biased estimator for $cov(\hat{\beta}|X)$. Instead, we have

$$E(\epsilon|X) = 0_{N \times 1} \quad (1.13)$$

$$cov(\epsilon|X) = \Omega \quad (1.14)$$

where Ω is a $N \times N$ matrix. The formula for $cov(\hat{\beta}|X)$ becomes:

$$\begin{aligned} cov(\hat{\beta}|X) &= (X^T X)^{-1} X^T \Omega X (X^T X)^{-1} \\ &= (\frac{1}{N} X^T X)^{-1} \frac{1}{N} \Phi (\frac{1}{N} X^T X)^{-1} \end{aligned} \quad (1.15)$$

where $\Phi = N^{-1} X^T \Omega X$.

To tackle the biased problem, Newey and West [14] have suggested to use the heteroskedasticity and autocorrelation consistent (HAC) covariance estimator $\hat{\Psi}_{HAC}$, which can be computed by substituting $\hat{\Phi}$ into equation (1.15) with

$$\hat{\Phi} = \frac{1}{N} \sum_{i,j=1}^N \omega_{|i-j|} \hat{V}_i \hat{V}_j^T \quad (1.16)$$

where $\omega = (\omega_0, \dots, \omega_{N-1})^T$ is a vector of weight with lag $l = |i - j|$. Furthermore, Newey and West [14] have proved that $\hat{\Psi}_{HAC}$ is positive semi-definite. Thus, in the presence of heteroskedasticity and autocorrelation, to test the null hypothesis that $\beta_k = 0$ against the alternative $\beta_k \neq 0$ with $k = 1, \dots, p$, we should look at the Newey-West

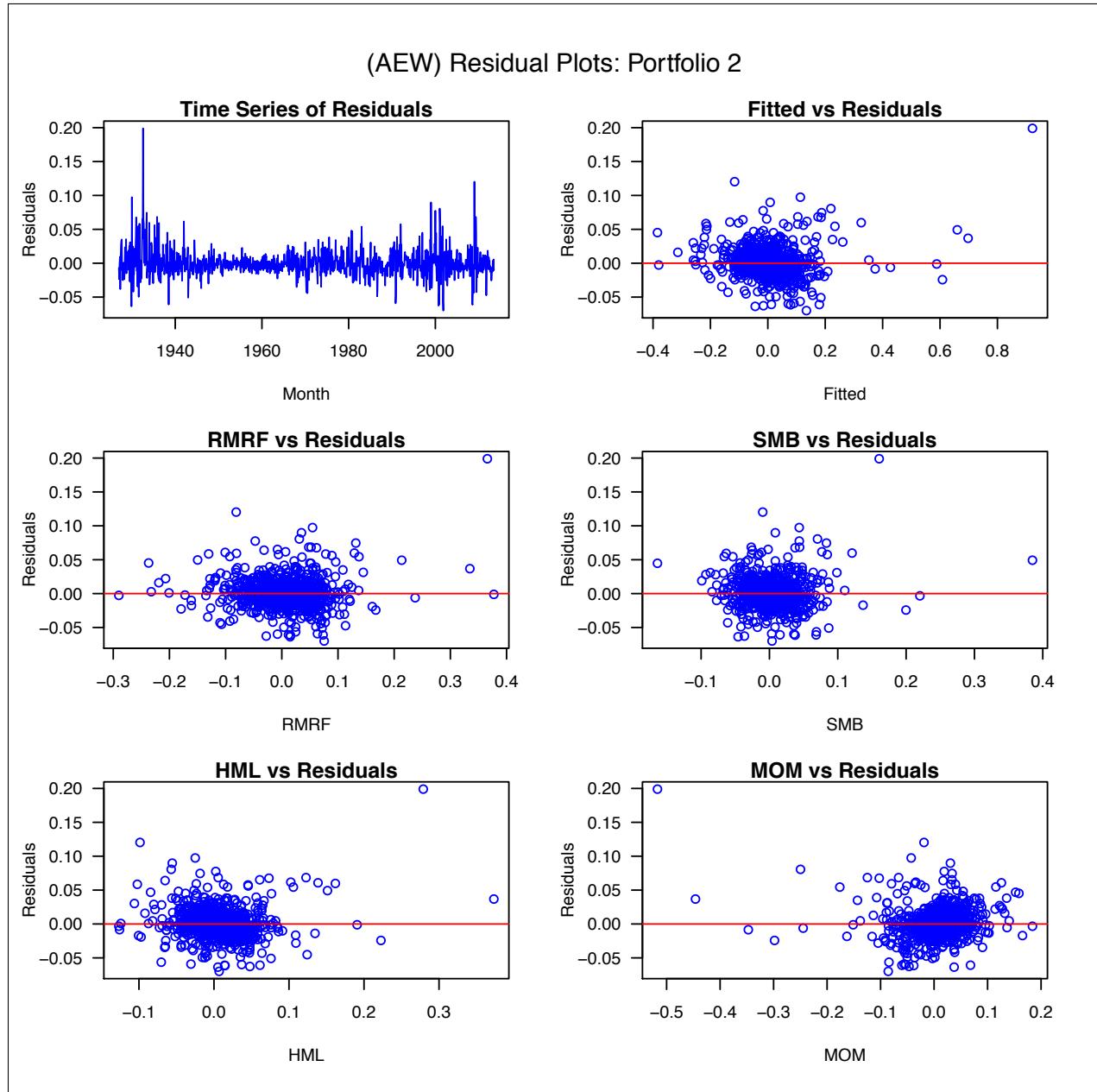


Figure 1.1: Residual Plots (e.g. AEW Portfolio 2)

t-statistic:

$$t_{NW} = \frac{\hat{\beta}_k}{\hat{SE}_{NW}(\beta_k)} \quad (1.17)$$

where $\hat{SE}_{NW}(\beta_k)$ is the kk th element of $\hat{\Psi}_{HAC}$, in addition to the ordinary t-statistic.

Chapter 2

Identification of Influential Observation

2.1 Influence Measure

Belsley, Kuh, and Welsch [3] have given an appropriate definition of influential observation.

An influential observation is one which, either individually or together with several other observations, has a demonstrably larger impact on the calculated values of various estimates (coefficients, standard errors, t-values, etc.) than is the case for most of other observations. One obvious means for examining such an impact is to delete each row, one at a time, and note the resultant effect on the various calculated values. Rows whose deletion produces relatively large changes in the calculated values are deemed influential.

There are numerous influence measures and there is a great deal of redundancy among the measures [7]. After experiencing with several data sets, Chatterjee and Hadi [7] have concluded that only a few influence measures are sufficient for identifying influential observations. Influence measures can be classified into the following five groups [7]:

1. Measures based on residuals
2. Measures based on the prediction matrix
3. Measures based on the volume of confidence ellipsoids
4. Measures based on influence function
5. Measures based on partial influence

In particular, measures based on the volume of confidence ellipsoids measure the impact of influential observations on the t-statistic of regression coefficient; whereas the

measures based on influence function measure the impact of influential observations on the estimated regression coefficient. For instance, Anderson, Bianchi, and Goldberg [1] use covariance ratio to examine the impact of the influential observations on the Newey-West t-statistic of the FVIX betas. In this thesis, we will use *Cook's Distance* and *Covariance Ratio*, in which Cook's Distance is the measure based on influence function and Covariance Ratio is the measure based on the volume of confidence ellipsoid.

2.1.1 Cook's Distance

Cook's distance was proposed by Cook [9] in 1977.

$$C_i = \frac{t_i}{p} \frac{p_i}{1 - p_i} \quad (2.1)$$

where t_i is defined in equation (1.12) and p_i is the i th diagonal element of P as defined in equation (1.6).

We denote C_i as the Cook's distance for the i th observation. Chatterjee and Hadi [7] interpret C_i as:

The scaled Euclidean distance between the two vectors of fitted values when the fitting is done by including or excluding the i th observation.

Cook [8] suggests that the i th observation can be considered as influential if C_i is greater than 1.

2.1.2 Covariance Ratio

Covariance Ratio was proposed by Belsley, Kuh, and Welsch [3] in 1980.

$$CVR_i = \left(\frac{N - p - t_i^2}{N - p - 1} \right)^p (1 - p_i) \quad (2.2)$$

We denote CVR_i as the covariance ratio for the i th observation. The data point with $|CVR_i - 1| > 3p/N$ is considered as influential observation.

2.2 Influential Observation

With the equations (2.1) and (2.2), along with the influential point criteria, we can easily find out the influential points for each of the ten portfolios and weighting methods based on the influence measures of Cook's distance and covariance ratio. Appendix B includes the plots of Cook's distance and covariance ratios for each of the ten portfolios and weighting methods. The influential observations for the corresponding influence

measures are highlighted with solid red. Appendix D shows the information of the influential observations identified by the two influence measures for each portfolio and weighting methods.

The plots in appendix B show that that covariance ratio identifies more influential observations than Cook's distance. Figure 2.1 shows an example for portfolio 2. We then combine our identification results from the ten portfolios. Out of 1,039 total monthly observations, we find that for the average equal weighted method, Cook's distance and covariance ratio identify 5 and 187 influential observations respectively. For the average value weighted method, Cook's distance and covariance ratio identify 4 and 232 influential observations respectively. Appendix E shows the corresponding months of the influential observations.

2.2.1 Influential Observations Identified by Cook's Distance

Based on the combined identification results in appendix E, the average equal weighted method has one additional influential observation **1939-09** and the rest of four influential observations are the same as the ones for the average value weighted method. The influential observations identified by Cook's distance are as follows.

- Average Equal Weighted: 1932-07, 1932-08, 1933-04, 1933-05, and 1939-09
- Average Value Weighted: 1932-07, 1932-08, 1933-04, and 1933-05

The influential observations are noticeable from the plots of momentum factor MOM against portfolio returns in appendix F. The influential observations have the largest difference between the portfolio return and the MOM factor, in which the corresponding portfolio return are always the peaks and the corresponding MOM factor are always the troughs. For instance, as shown in figure 2.1 for portfolio 2 with average equal weighted method, the five influential observations are noticeable by inspecting the graphical distance between the MOM factor and the portfolio return.

It was not a coincidence that the influential observations happened in those five months. In fact, they were linked to the major historical events [4].

- **July - August 1932:** On July 8th, 1932, the Dow reached its lowest level of the 20th century and did not return to pre-1929 levels until the 23rd November, 1954.
- **April - May 1933:** FDR took office in March 1933. In the first 100 days many acts and agencies were introduced which were to form the basis of the New Deal. The acts and agencies included but not limited to the Economy Act, Emergency Banking Act, National Industrial Recovery Act, Agricultural Adjustment Act, and FDIC. The New Deal made huge impact on economy in years to come.

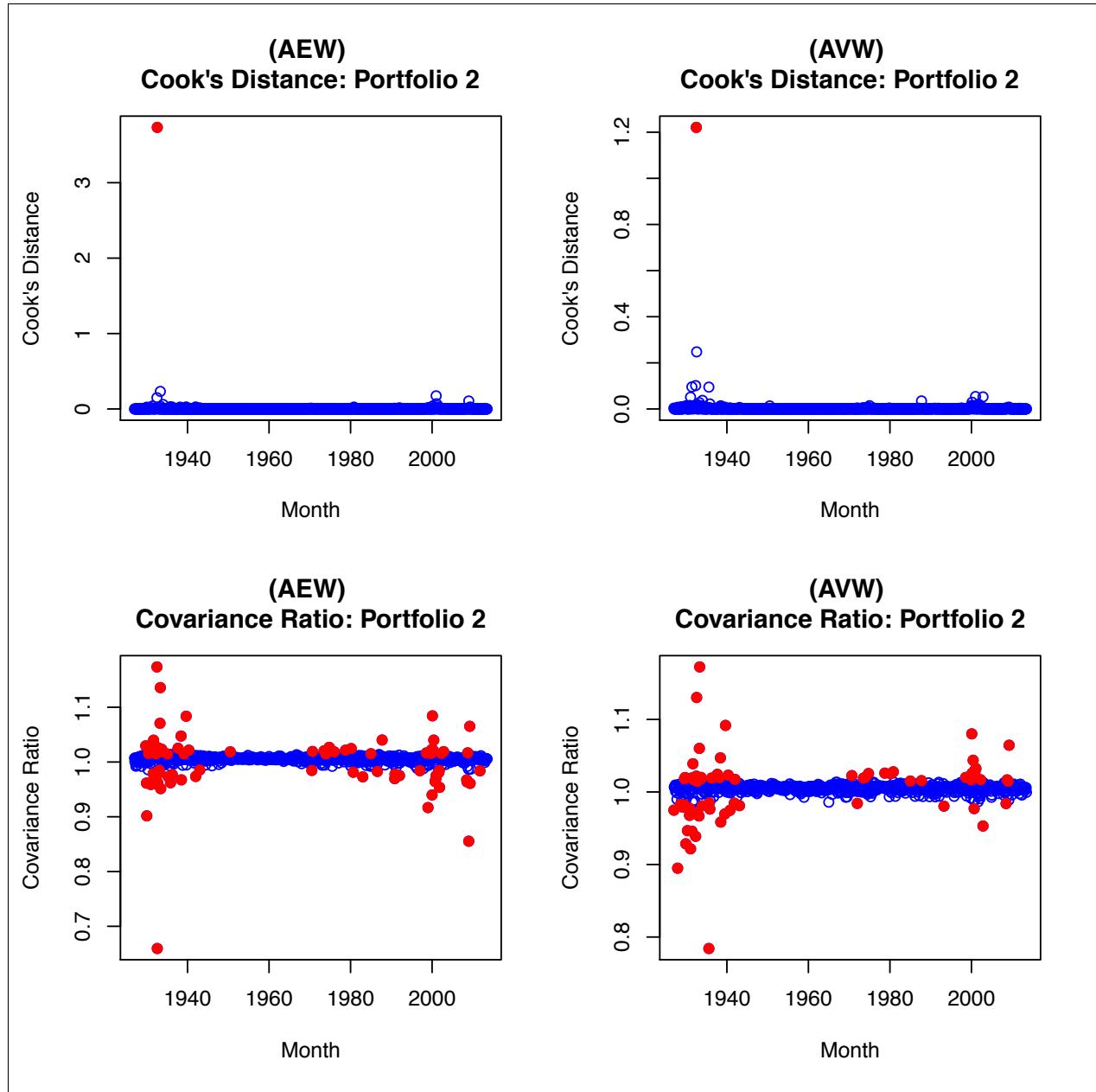


Figure 2.1: Plot of Cook's Distance and Covariance Ratio (e.g. Portfolio 2)

- **September 1939:** The start of World War II.

In addition, momentum was noticeable on Dow Jones before and after July 1932. There was tendency for falling asset prices to keep falling from 1929 to July 1932 and tendency for rising asset prices to keep rising from July 1932 to 1937.

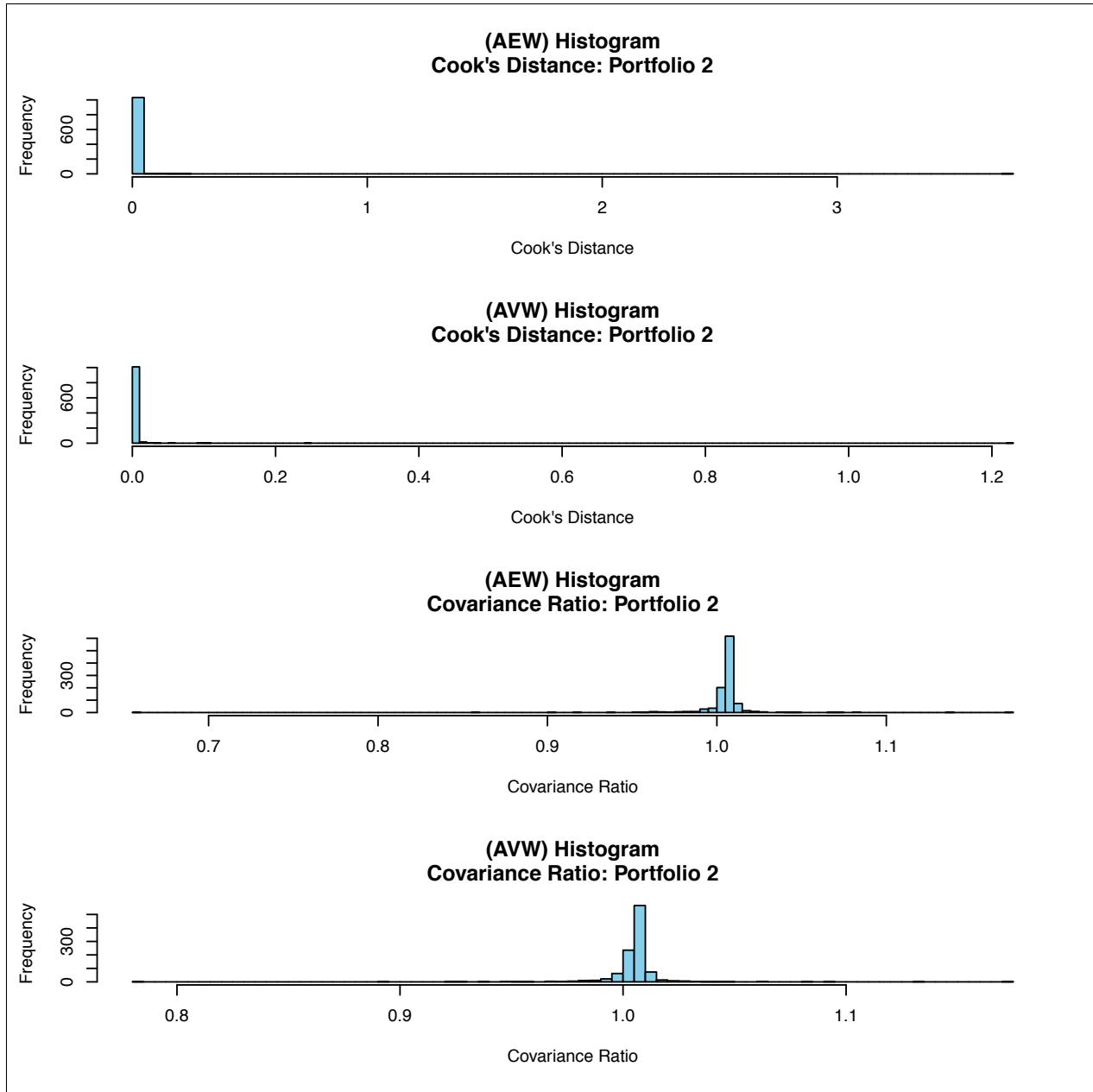


Figure 2.2: Histogram of Cook's Distance and Covariance Ratio (e.g. Portfolio 2)

Furthermore, momentum was noticeable on FT 30 but not Dow Jones before and after September 1939. There was tendency for falling asset prices to keep falling from 1936 to September 1939 and tendency for rising asset prices to keep rising from September 1939 to 1946.

In the next chapter, we will compare the the β estimates and the t-statistics between the full dataset and the same dataset with the set of influential observations excluded.

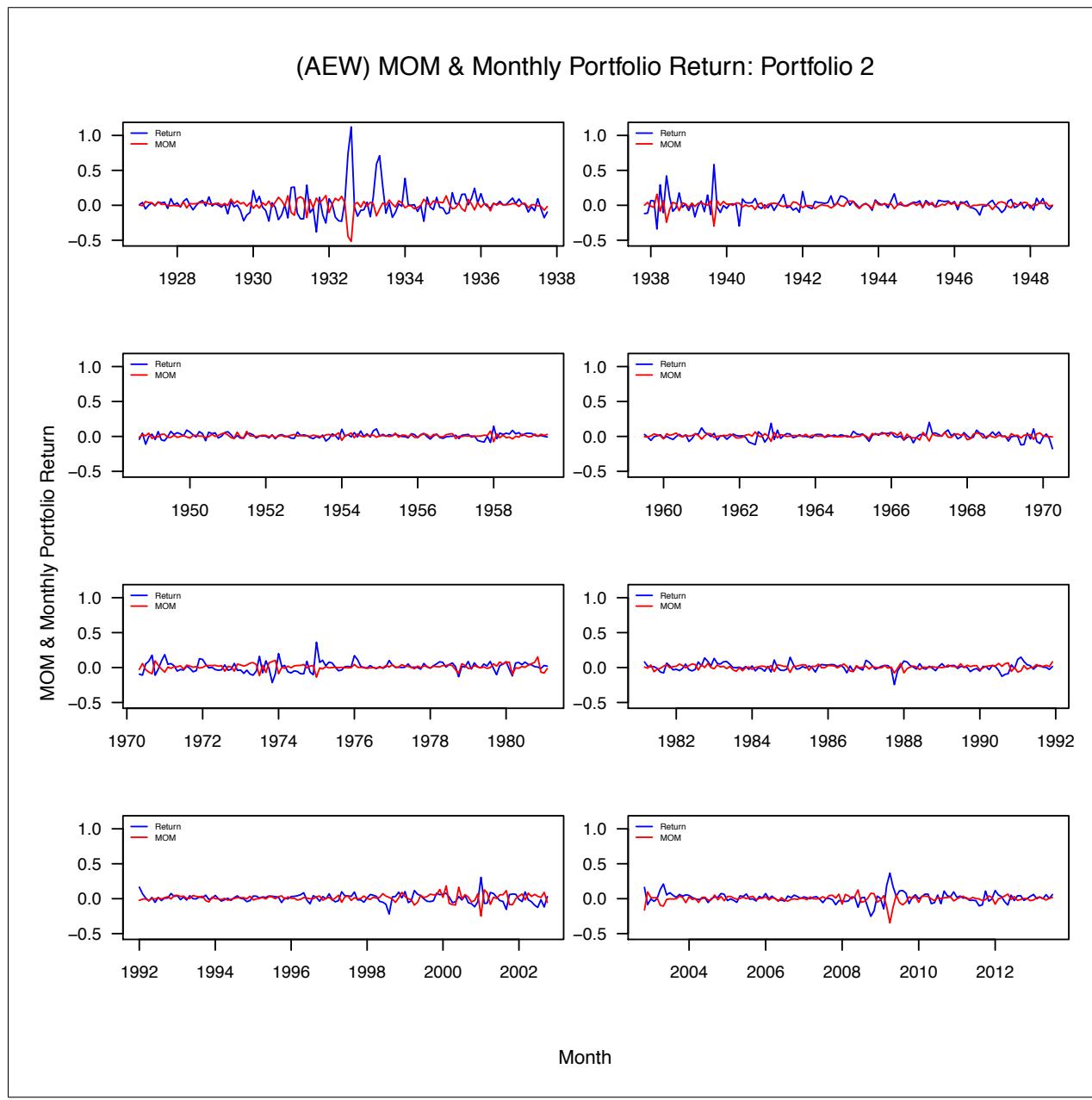


Figure 2.3: Plot of MOM vs Monthly Portfolio Return (e.g. AEW Portfolio 2)

Chapter 3

Influence of Influential Observation

In this chapter, we use the influence measures of Cook's distance and covariance ratio to measure the influence of an observation on the β estimates and the t-statistics respectively. We also use both influence measures to examine the impact on the OLS assumptions. To do so, we compare the results of the multiple linear regression on the full dataset and the same dataset with the set of influential observations excluded. **Panel A** refers to the regression result on the full dataset, **Panel B** refers to the regression result on the full dataset with the influential points identified by Cook's distance excluded, and **Panel C** refers to the regression result on the full dataset with the influential points identified by covariance ratio excluded.

3.1 Influence on Beta Estimates

Table 3.1 summarizes the key findings on each of the β estimates for both weighting methods. Figures 3.1 and 3.2 show the plots of each β estimate against the portfolio rankings for the average equal weighted method and the average value weighted method respectively. In sum, the β_{RMRF} and β_{SMB} estimates are positive and convex on the portfolio rankings for both panels, whereas the β_{HML} estimates are positive and concave on the portfolio rankings for both panels. As we should have expected, the β_{MOM} estimates are monotonically increasing on the portfolio rankings because the portfolios are constructed monthly using NYSE prior (2-12) return decile breakpoints in which stocks with the lowest NYSE prior (2-12) return comprise decile 1 (or portfolio 1) and stocks with the highest comprise decile 10 (or portfolio 10). Note that the β_{MOM} estimates are positive for the average equal weighted method, but for the average value weighted method, the β_{MOM} estimates are positive for portfolio ranking less than 7 and negative otherwise.

By excluding the influential observations identified by Cook's distance, the β_{RMRF} and β_{SMB} estimates are more convex on the portfolio rankings for the average equal weighted method. In addition, β_{HML} estimates are more concave on the portfolio rank-

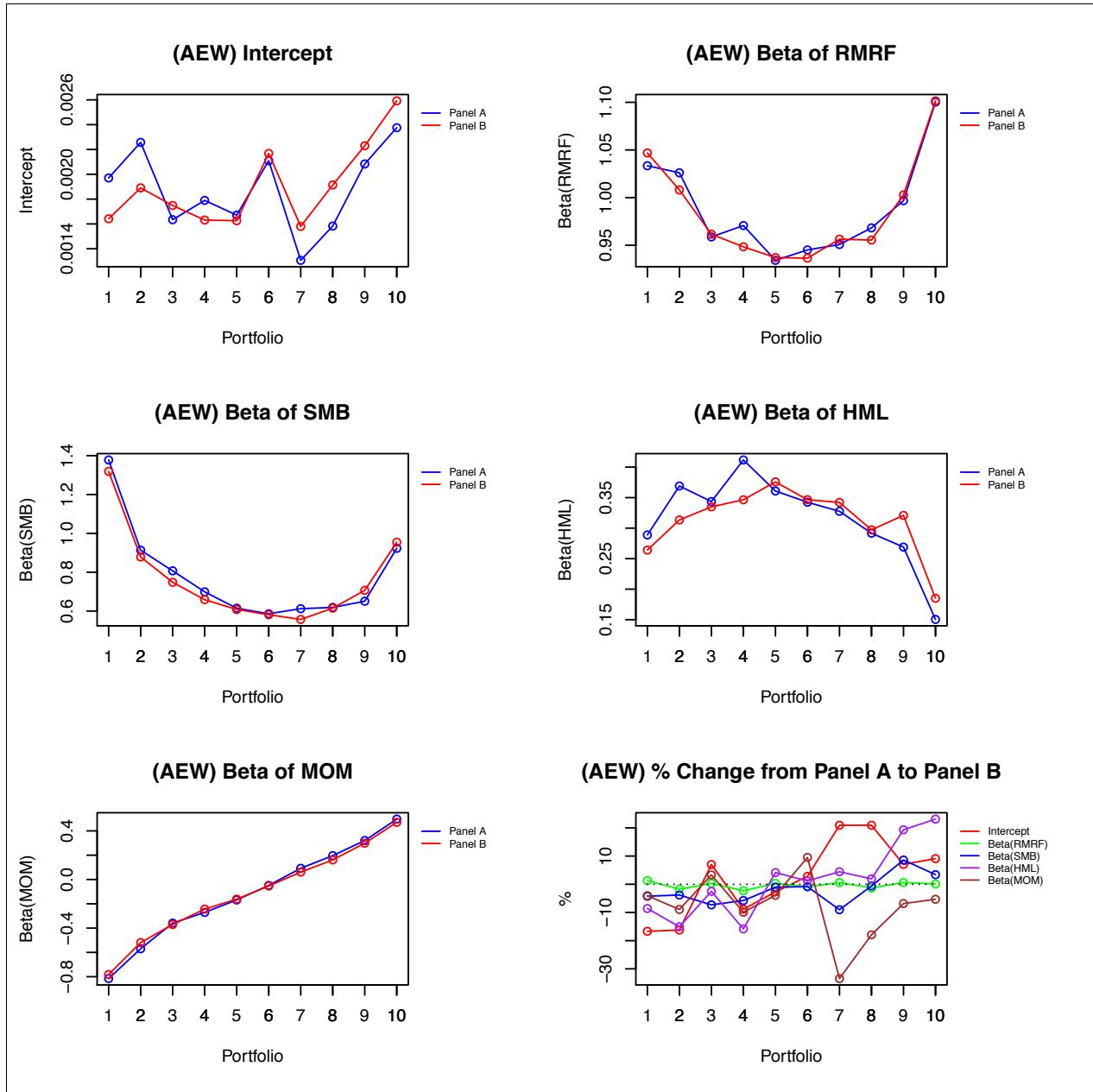
Table 3.1: $\hat{\beta}$ vs Portfolio Ranking

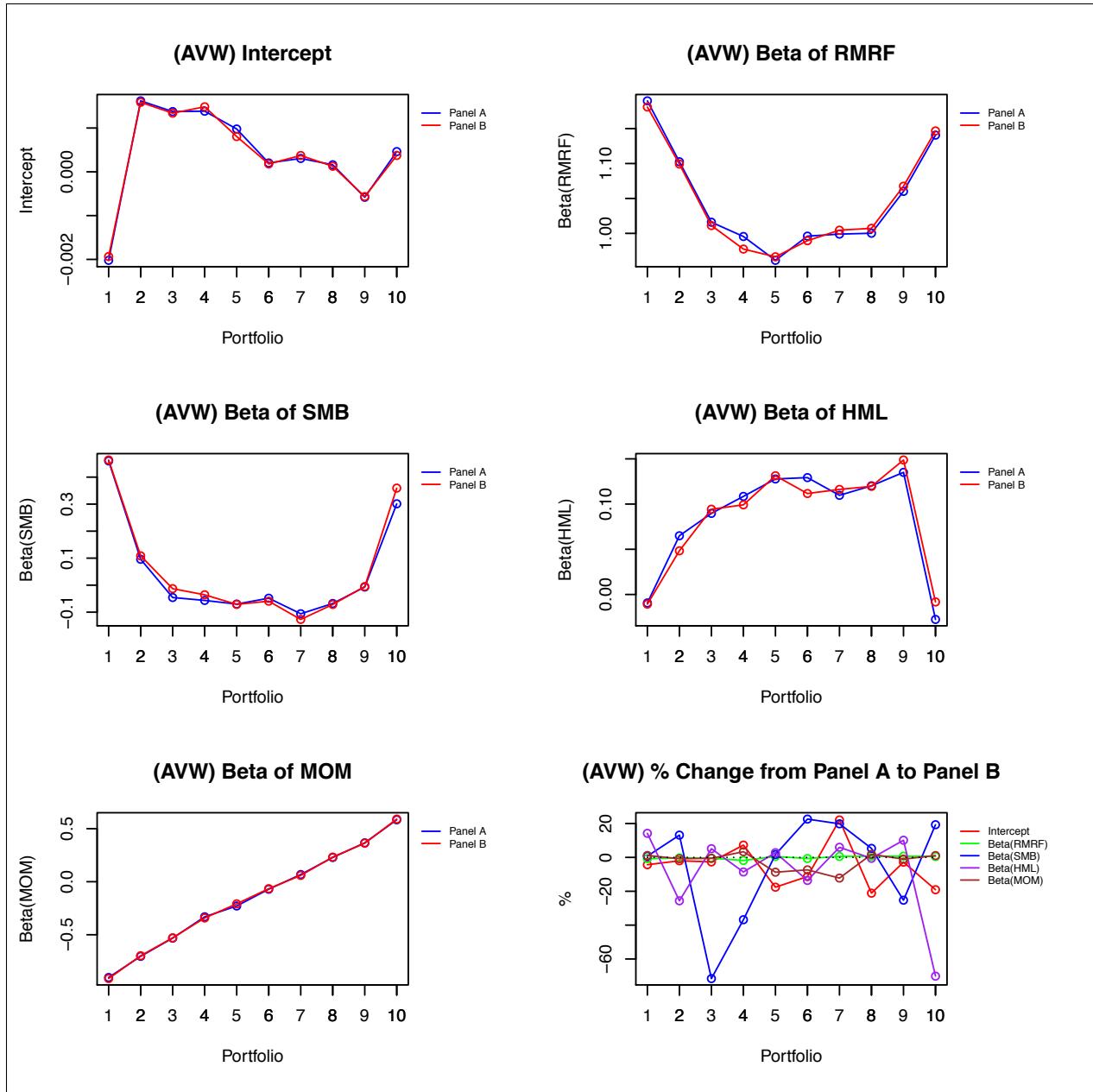
$\hat{\beta}$	AEW	AVW
β_{RMRF}	Positive; Convex on portfolio ranking	Positive; Convex on portfolio ranking
β_{SMB}	Positive; Convex on portfolio ranking	Positive; Convex on portfolio ranking
β_{HML}	Positive; Concave on portfolio ranking	Positive; Concave on portfolio ranking
β_{MOM}	Positive; Monotonically increasing on portfolio ranking	Positive for portfolio ranking less than 7 and negative otherwise; Monotonically increasing on portfolio ranking

Table 3.2: $\hat{\beta}$ vs Portfolio Ranking: Change from Panel A to Panel B

$\hat{\beta}$	AEW	AVW
β_{RMRF}	More convex	Insignificant change
β_{SMB}	More convex	Insignificant change
β_{HML}	More concave	Insignificant change
β_{MOM}	Insignificant change	Insignificant change

ings for the average equal weighted method. For the rest of the β estimates, the graphical relationships and the percentage changes are insignificant. Table 3.2 summarizes the comparisons and the bottom right plots in figures 3.1 and 3.2 show the influence in terms of percentage changes.

Figure 3.1: (AEW) Plots of β

Figure 3.2: (AVW) Plots of β

3.2 Influence on t-statistics

In this section, we want to examine the influence on the t-statistic and also the Newey-West t-statistic. As we have discussed in section 1.6, we use the Newey-West t-statistic as defined in equation (1.17) to overcome autocorrelation and heteroskedasticity in the random errors in the Fama French Carhart Four Model (1.1). Tables 3.3 and 3.4 summarize the key findings on both versions of t-statistic of the β estimates for the average equal weighted method and the average value weighted method respectively. Figures 3.4, 3.5, 3.6, and 3.7 show the plots of both versions of t-statistic against the portfolio rankings for β_{RMRF} , β_{SMB} , β_{HML} , and β_{MOM} respectively. I also include the same kind of plots for the intercept in figure 3.3.

For the average equal weighted method, both versions of t-statistic are positive and highly significant for all the β estimates except for the β_{MOM} estimates in portfolios 7-10. As for the average value weighted method, both versions of t-statistic are positive for all the β estimates except for the β_{SMB} estimates in portfolios 3-9 and the β_{MOM} estimates in portfolios 7-10. The β_{SMB} estimates are highly significant except for portfolios 3 and 9, whereas the β_{HML} estimates are highly significant except for portfolios 1 and 10. In addition, β_{MOM} estimates are monotonically increasing on portfolio ranking starting from portfolios 2 or 3 across the board.

By excluding the influential observations identified by covariance ratio, the sign of both t-statistic, the level of statistical significance, and the graphical relationship between the t-statistic and the porfolio rankings remain the same at large. However, when we look into the percentage change as shown in figure 3.8, the percentage changes from panel A to panel C are much higher for the Newey-West t-statistic than the ones for t-statistic across the board.

Table 3.3: (AEW) t-statistic and Newey-West t-statistic for $\hat{\beta}$

$\hat{\beta}$	t-statistic	Newey-West t-statistic
β_{RMRF}	Positive; Highly significant; Concave on portfolio ranking	Positive; Highly significant; Multi-modal on portfolio ranking
β_{SMB}	Positive; Highly significant; Multi-modal on portfolio ranking	Positive; Highly significant; Multi-modal on portfolio ranking
β_{HML}	Positive; Highly significant; Near concave on portfolio ranking	Positive; Highly significant; Near concave on portfolio ranking
β_{MOM}	Positive for portfolio ranking less than 7 and negative otherwise; Highly significant; Monotonically increasing on portfolio ranking starting from portfolios 2 or 3	Positive for portfolio ranking less than 7 and negative otherwise; Highly significant; Monotonically increasing on portfolio ranking starting from portfolios 2 or 3

Table 3.4: (AVW) t-statistic and Newey-West t-statistic for $\hat{\beta}$

$\hat{\beta}$	t-statistic	Newey-West t-statistic
β_{RMRF}	Positive; Highly significant; Multi-modal on portfolio ranking	Positive; Highly significant; Multi-modal on portfolio ranking
β_{SMB}	Positive for portfolios 1, 2, 10, and negative otherwise; Highly significant except for portfolios 3 and 9; Convex on portfolio ranking	Positive for portfolios 1, 2, 10, and negative otherwise; Highly significant except for portfolios 3 and 9; Convex on portfolio ranking
β_{HML}	Positive; Highly significant except for portfolios 1 and 10; Near concave on portfolio ranking	Positive; Highly significant except for portfolios 1 and 10; Near concave on portfolio ranking
β_{MOM}	Positive for portfolio ranking less than 7 and negative otherwise; Highly significant; Monotonically increasing on portfolio ranking starting from portfolios 2 or 3	Positive for portfolio ranking less than 7 and negative otherwise; Highly significant; Monotonically increasing on portfolio ranking starting from portfolios 2 or 3

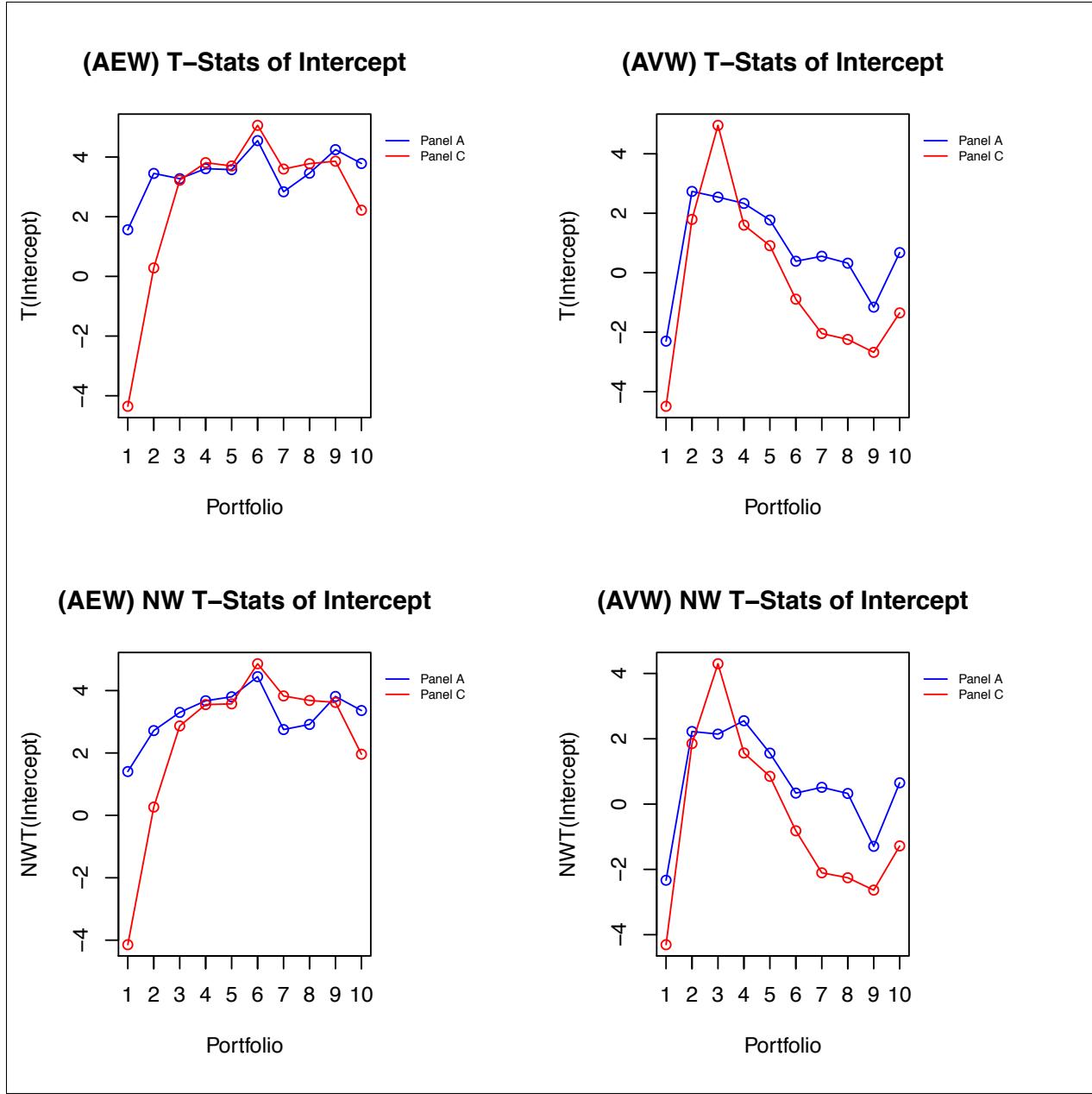
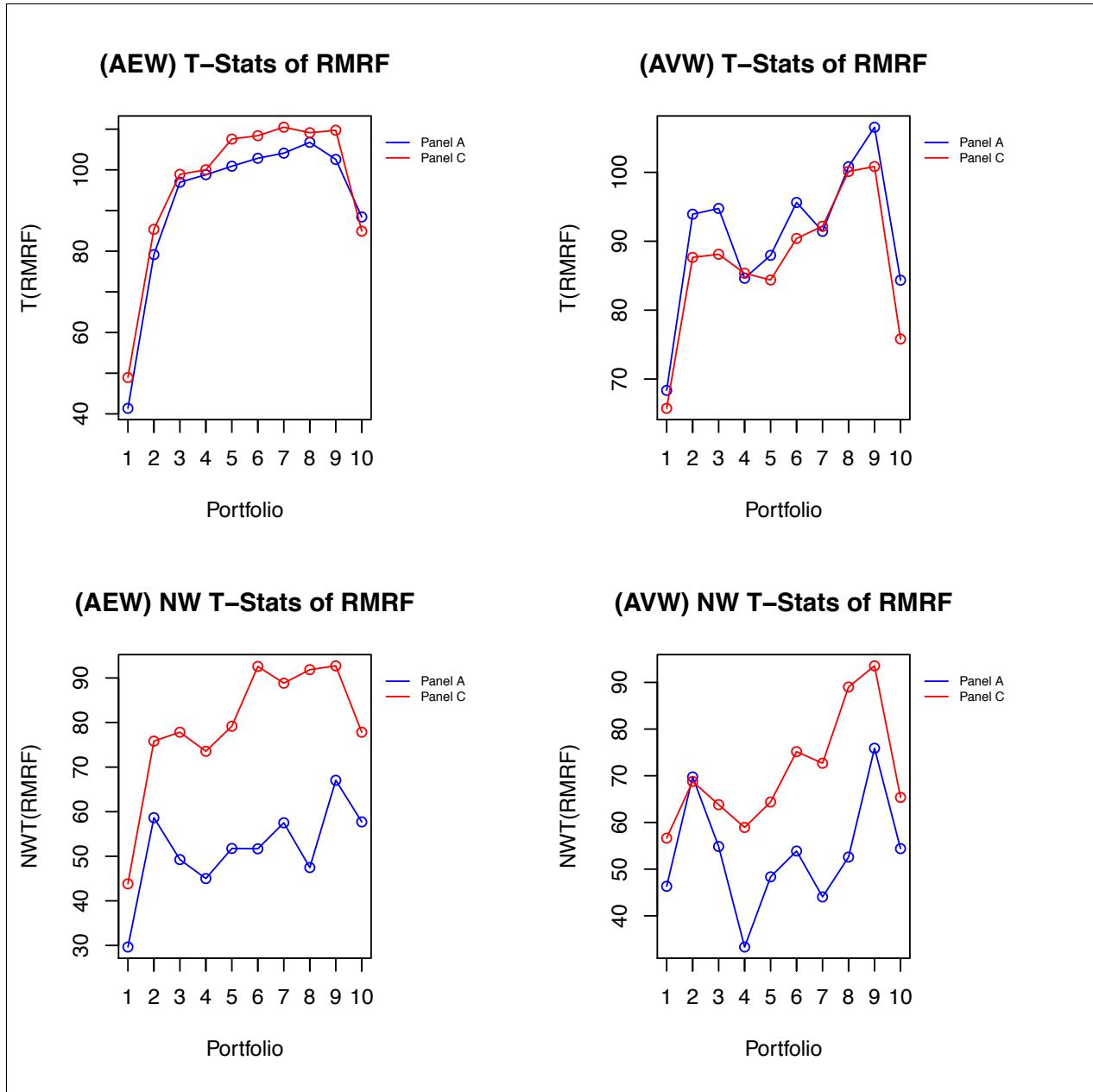
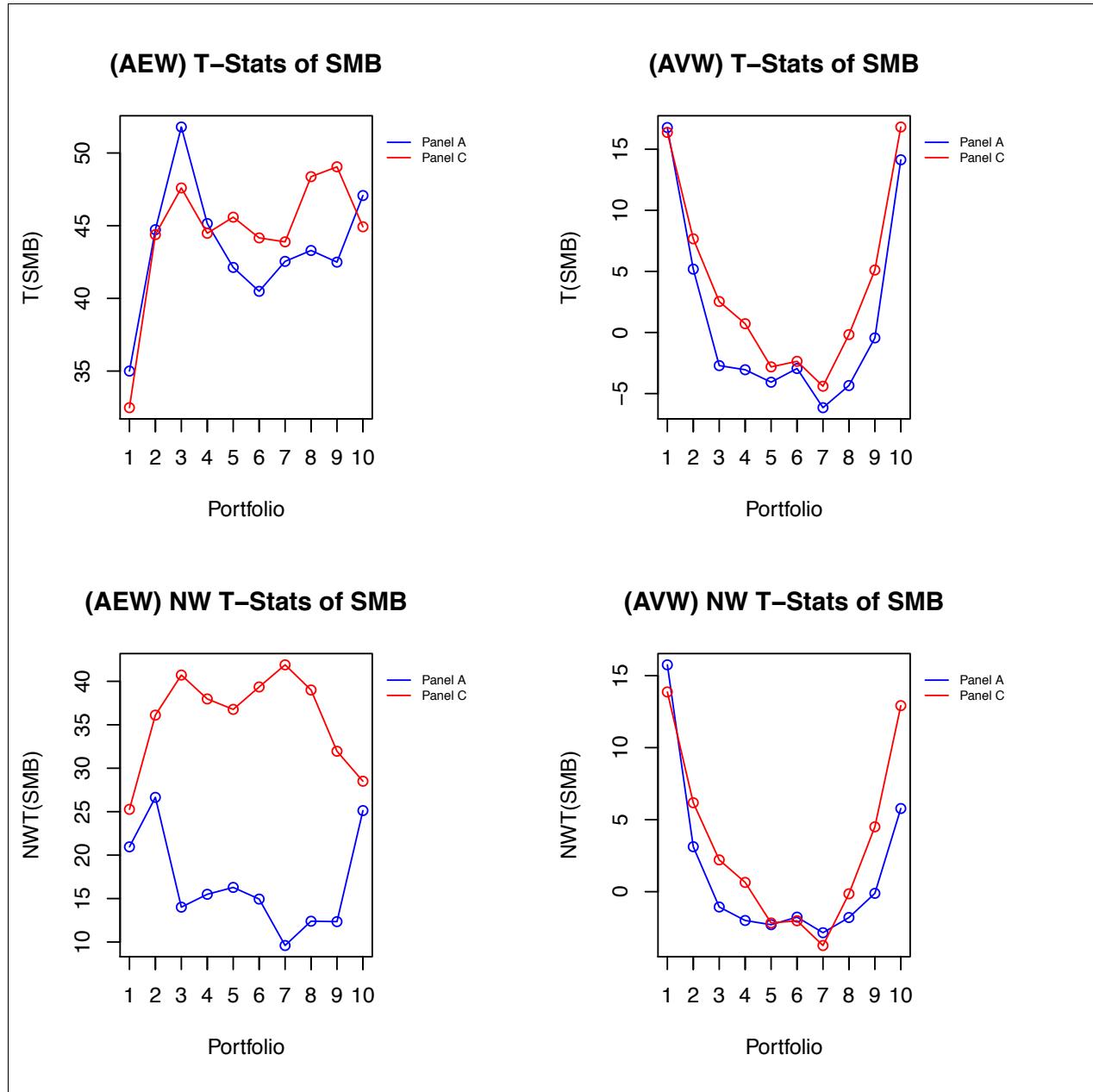
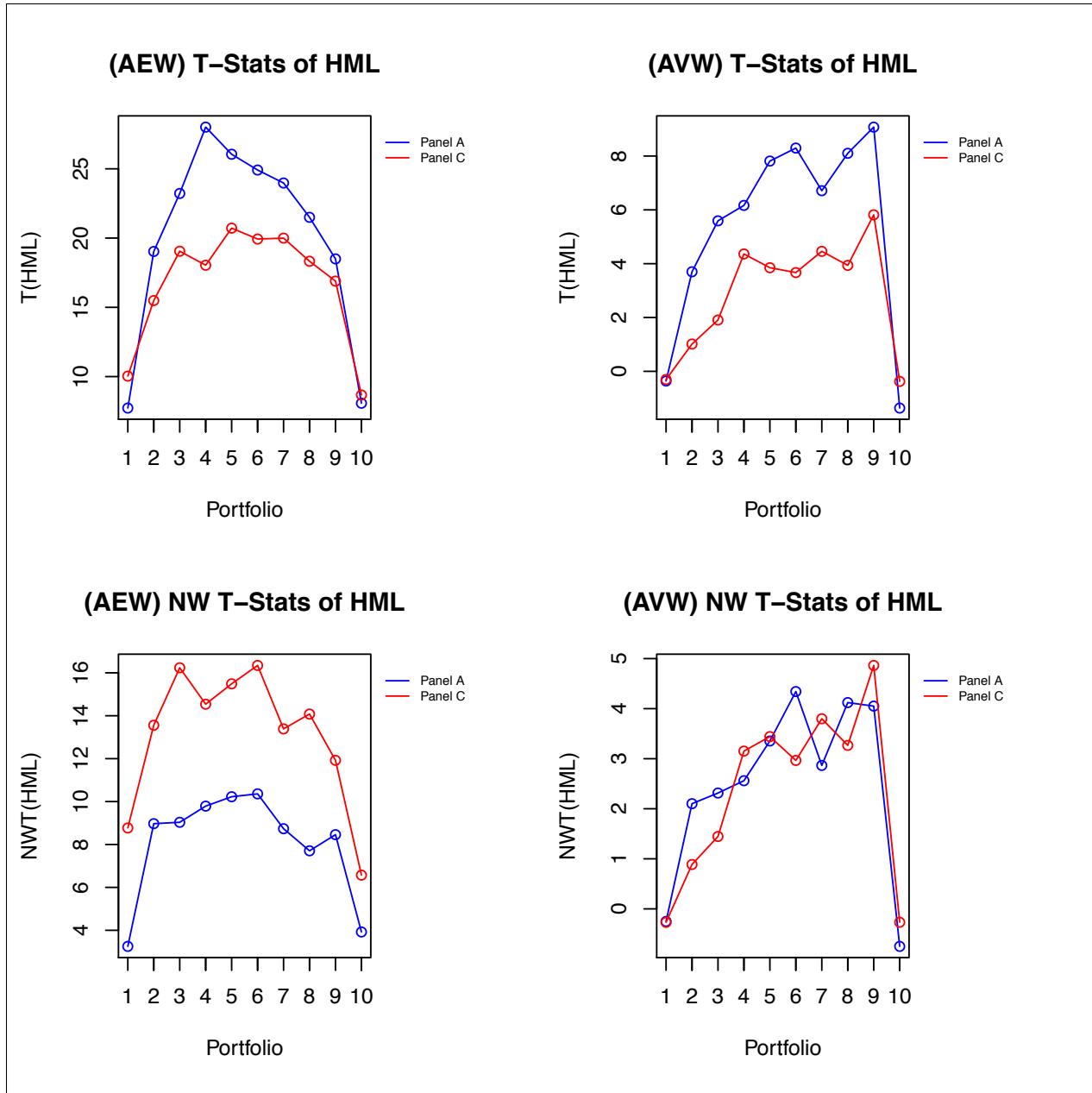
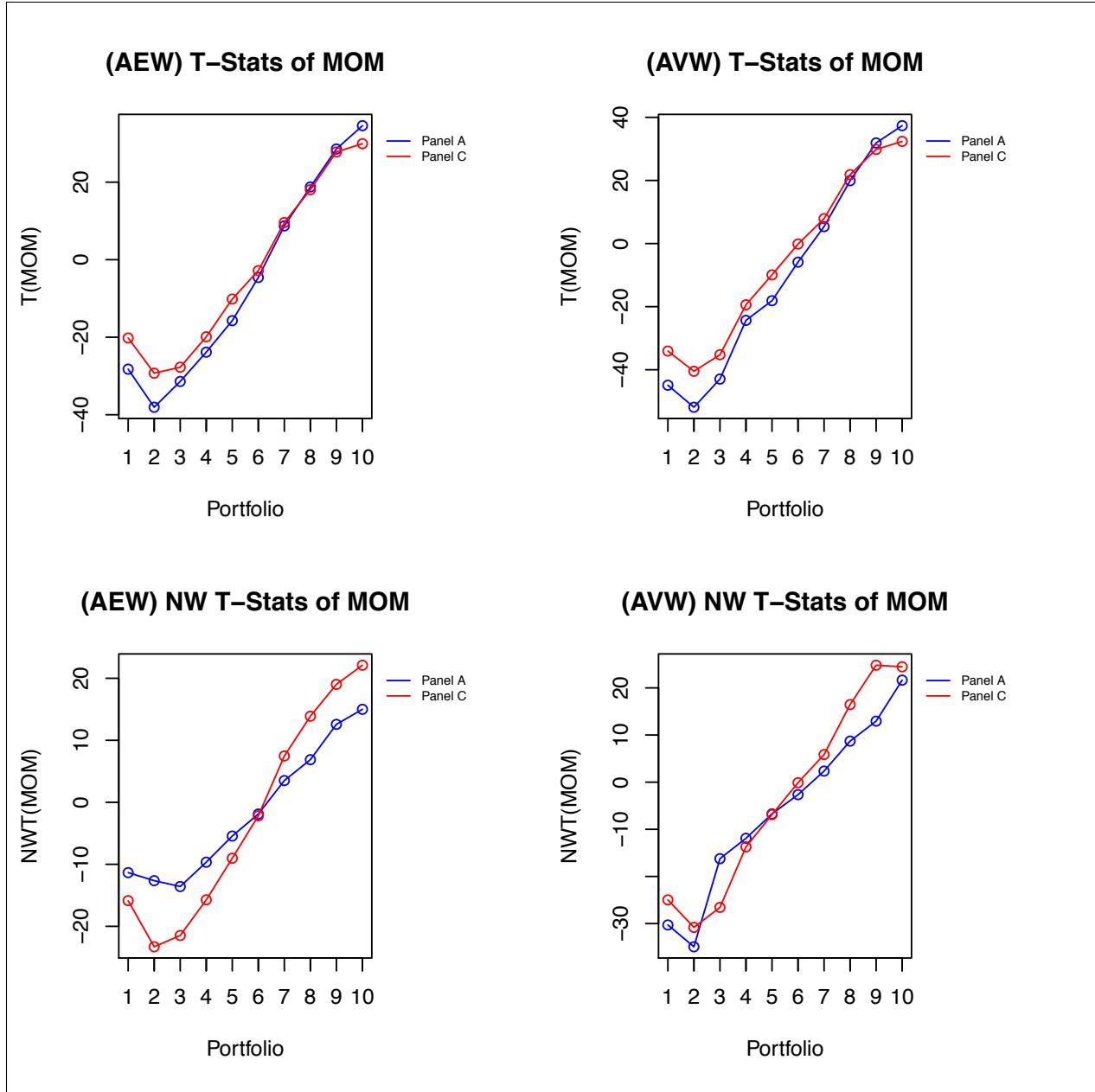


Figure 3.3: Plots of Intercept t-statistic

**Figure 3.4:** Plots of β_{RMRF} t-statistic

**Figure 3.5:** Plots of β_{SMB} t-statistic

**Figure 3.6:** Plots of β_{HML} t-statistic

**Figure 3.7:** Plots of β_{MOM} t-statistic

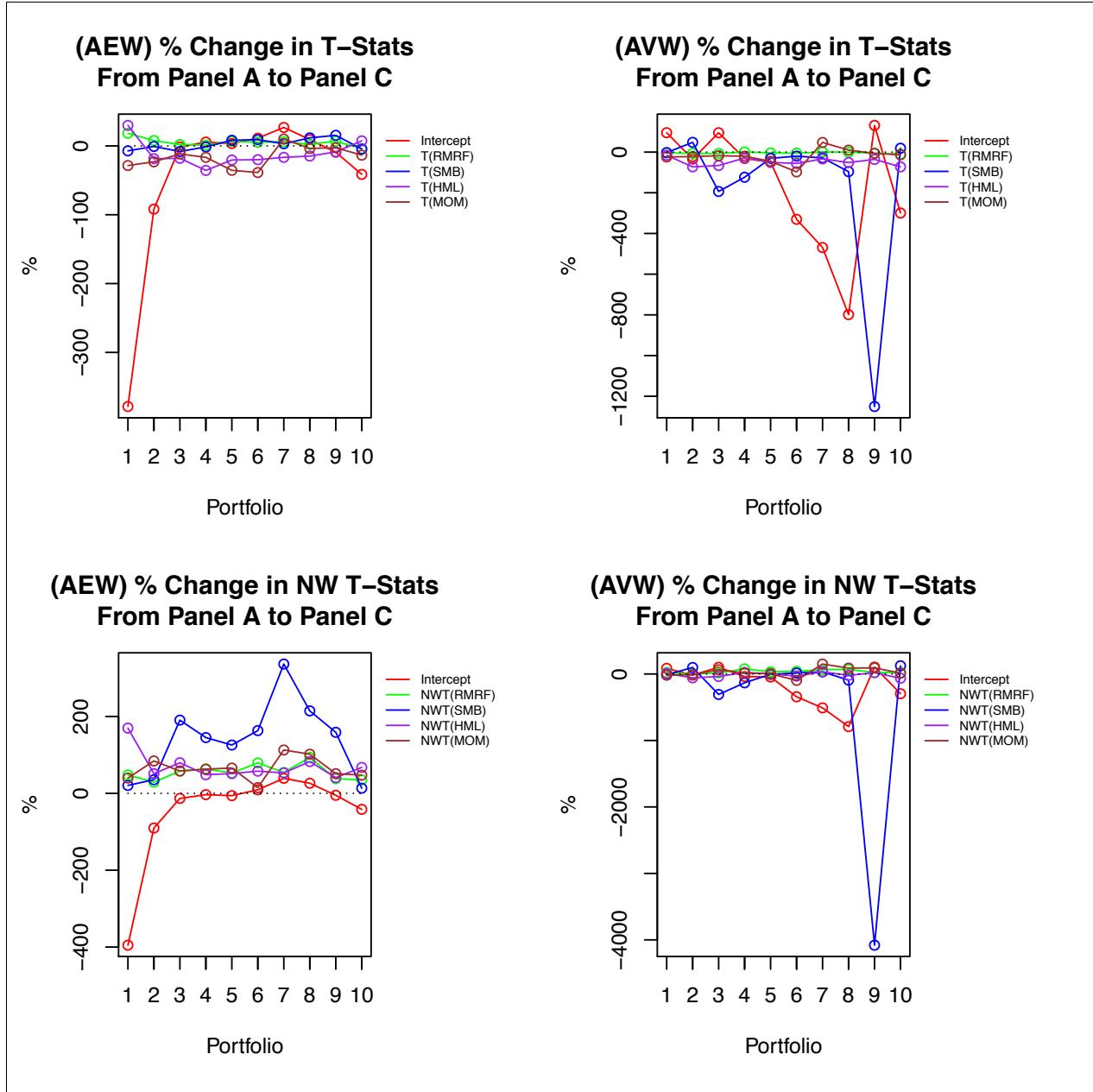


Figure 3.8: Plots of Percentage Change in t-statistic

3.3 Influence on the OLS Assumptions

In this section, we want to examine the impact of the influential observations on the OLS assumptions, in particular, normality and variance of residuals.

3.3.1 Lilliefors Test for Normality

We apply the Lilliefors test on panels B and C to test the null hypothesis that the residuals come from the Gaussian distribution. Tables 3.5 and 3.6 show the test results for panels B and C respectively. The Lilliefors test results for panel B remain the same as the ones for panel A. However, for panel C, the normality assumption seem to be valid for 3 and 8 out of 10 portfolios for the average equal weighted and average value weighted methods respectively.

Table 3.5: (Panel B) Lilliefors Test for Normality with 0.05 Alpha Level

portfolio	pvalAEW	pvalAVW	nullAEW	nullAVW
1	2.31e-40	2.59e-16	Reject	Reject
2	2.02e-26	4.58e-06	Reject	Reject
3	3.26e-22	3.28e-14	Reject	Reject
4	2.03e-15	2.70e-14	Reject	Reject
5	4.99e-18	8.70e-09	Reject	Reject
6	4.91e-17	3.46e-13	Reject	Reject
7	9.28e-11	2.30e-10	Reject	Reject
8	2.08e-08	2.61e-09	Reject	Reject
9	3.19e-17	3.47e-09	Reject	Reject
10	1.58e-13	1.79e-07	Reject	Reject

3.3.2 Non-constant Variance Score Test

We apply the non-constant variance score test on panels B and C to test the null hypothesis that the residuals are distributed with constant variance. Tables 3.7 and 3.8 show the test results for panels B and C respectively. For panel B, there is 1 additional portfolio seem to be distributed with constant variance for each weighting methods. Whereas for panel C, there are 7 and 5 additional portfolios seem to be distributed with constant variance for the average equal weighted and average value weighted methods respectively.

Table 3.6: (Panel C) Lilliefors Test for Normality with 0.05 Alpha Level

portfolio	pvalAEW	pvalAVW	nullAEW	nullAVW
1	3.83e-08	2.69e-02	Reject	Reject
2	6.46e-04	9.19e-01	Reject	Accept
3	3.48e-01	1.36e-02	Accept	Reject
4	1.18e-01	5.69e-02	Accept	Accept
5	1.99e-02	7.89e-01	Reject	Accept
6	2.53e-03	4.12e-01	Reject	Accept
7	8.22e-02	6.60e-01	Accept	Accept
8	4.53e-02	8.85e-01	Reject	Accept
9	3.34e-03	1.68e-01	Reject	Accept
10	5.10e-04	1.49e-02	Reject	Reject

Table 3.7: (Panel B) Non-constant Variance Score Test with 0.05 Alpha Level

portfolio	pvalAEW	pvalAVW	nullAEW	nullAVW
1	1.88e-55	3.77e-12	Reject	Reject
2	1.96e-19	5.33e-06	Reject	Reject
3	2.09e-10	2.07e-04	Reject	Reject
4	3.81e-16	1.58e-01	Reject	Accept
5	1.86e-17	4.63e-04	Reject	Reject
6	1.36e-14	1.34e-03	Reject	Reject
7	6.80e-06	8.59e-01	Reject	Accept
8	7.53e-01	1.35e-11	Accept	Reject
9	2.50e-19	1.49e-18	Reject	Reject
10	2.09e-06	4.90e-06	Reject	Reject

Table 3.8: (Panel C) Non-constant Variance Score Test with 0.05 Alpha Level

portfolio	pvalAEW	pvalAVW	nullAEW	nullAVW
1	1.15e-06	2.84e-03	Reject	Reject
2	2.08e-02	6.25e-01	Reject	Accept
3	3.45e-01	4.11e-02	Accept	Reject
4	2.73e-01	5.20e-01	Accept	Accept
5	5.26e-02	1.48e-01	Accept	Accept
6	6.97e-01	8.59e-01	Accept	Accept
7	6.70e-01	2.82e-02	Accept	Reject
8	1.97e-01	5.92e-01	Accept	Accept
9	1.95e-01	4.64e-01	Accept	Accept
10	6.85e-02	2.85e-01	Accept	Accept

Chapter 4

Portfolio Optimization

As a continuation from the last chapter, we want to examine the impact of the influence observations on the expected monthly return and the standard deviation of the monthly return by comparing their results among panels A, B, and C. Then, with the two different weighting methods for consideration, our logical next step is to compare their investment performances by looking at their mean-variance optimal portfolios.

4.1 Mean and SD of Monthly Portfolio Returns

Figure 4.1 shows the plots of the expected monthly portfolio returns, their standard deviations, and the percentage change from panel A to panels B and C against the portfolio rankings for both weighting methods. In sum, for both weighting methods,

- The expected monthly portfolio return increases as the portfolio ranking increases.
- The standard deviations of monthly portfolio returns are convex on the portfolio ranking.
- The percentage changes for the expected monthly portfolio returns and their standard deviations are negative from panel A to panels B and C, except for the expected monthly portfolio returns for portfolios 9 and 10 with the average value weighted method in panel C.
- The absolute percentage changes from panel A to panels B and C for the two estimates (mean and SD) are diminishing on the portfolio rankings at large.

To go one step further, we want to look into the relationship between the expected monthly portfolio returns and their standard deviations for each of the weighting methods and panels. As shown in the top two plots in figure 4.2, for both weighting methods, the risk and return curve shifts to the left when we change from panel A to panel B. Similarly, for both weighting methods, the risk and return curve shifts to the left when

we change from either panel A or panel B to panel C. These results are what we have expected because the influential observations have abnormal high monthly portfolio returns and therefore their removal would reduce the risk and return. The bottom two plots show that the risk and return curve shifts to the left when we change from the average equal weighted method to the average value weighted method in all panels. The explanation will be given in the latter section *Optimal Portfolio Weights*.

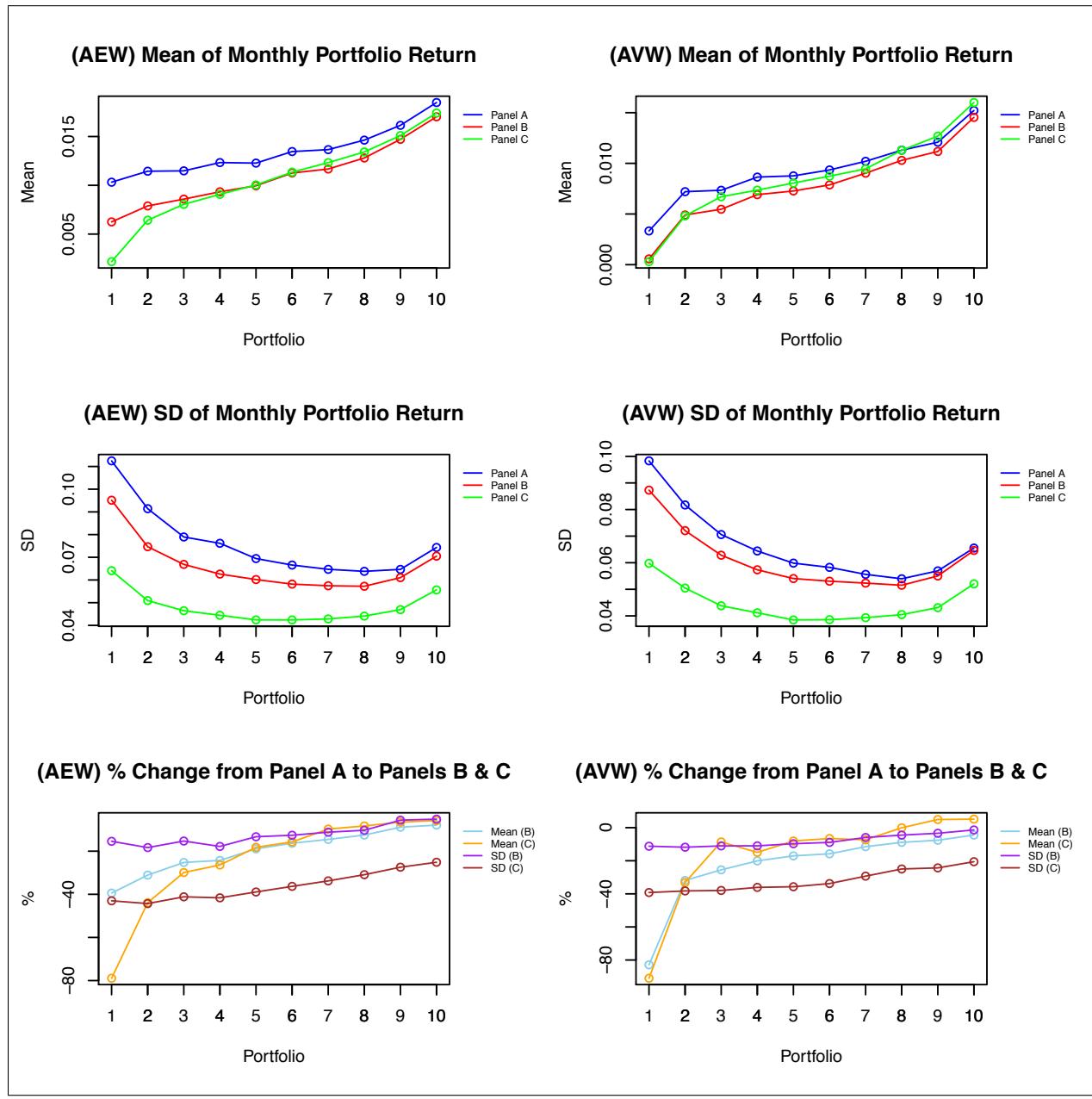


Figure 4.1: Plots of Mean and SD of Monthly Portfolio Returns

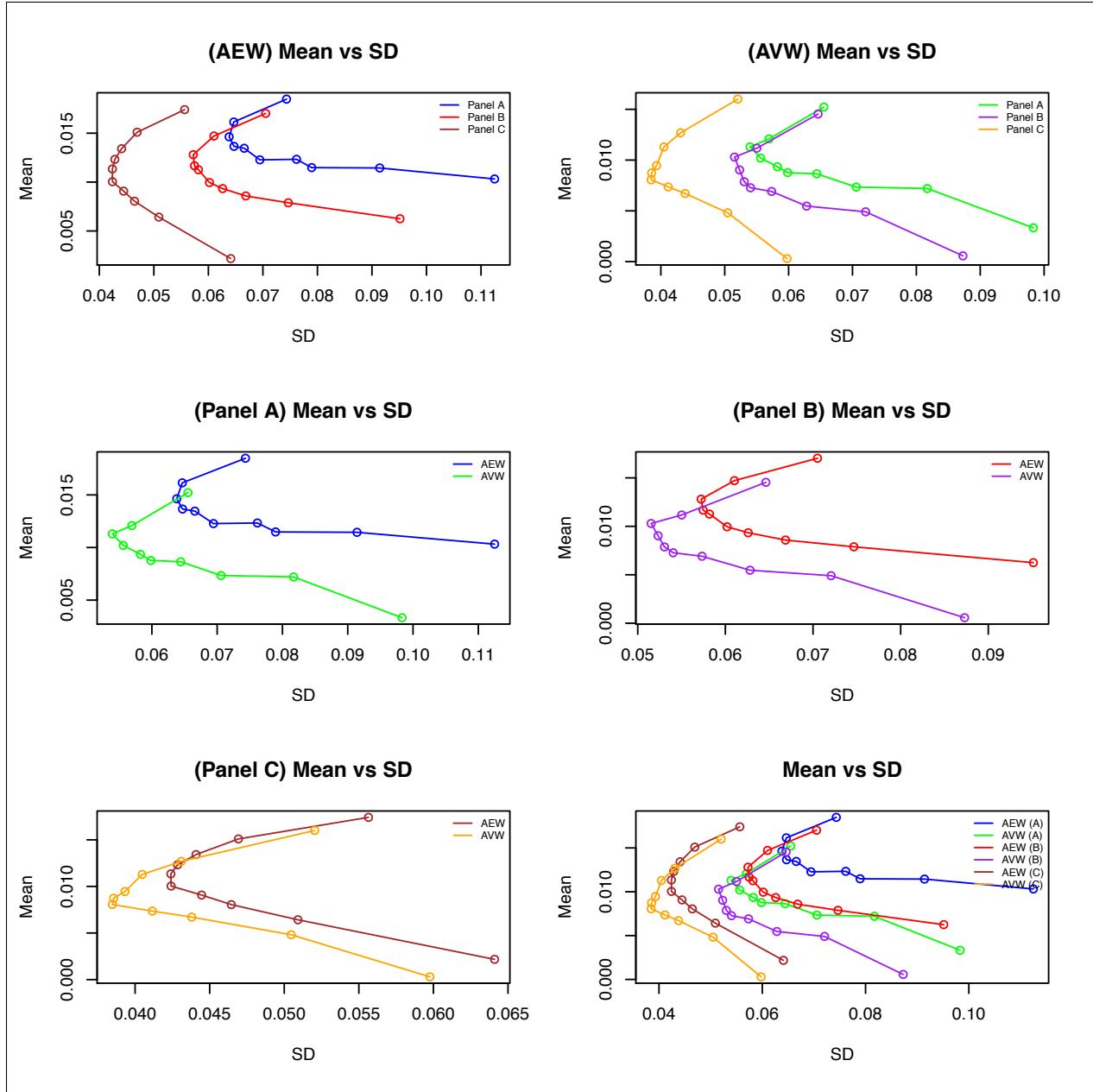


Figure 4.2: Plots of Mean versus SD of Monthly Portfolio Returns

4.2 Mean-Variance Optimal Portfolio

In this section, we want to compare the investment performances of two mean-variance optimal portfolios, in which one portfolio is comprised of the ten momentum portfolios formed with the average equal weighted method and the other portfolio is comprised of the ten momentum portfolios formed with the average value weighted method.

To find a mean-variance optimal portfolio with ten assets [13], i.e. the ten momentum portfolios, first, we compute the mean rates of return $\bar{r}_1, \bar{r}_2, \dots, \bar{r}_{10}$ and the covariances $\sigma_{ij}, i, j = 1, 2, \dots, 10$ for the ten momentum portfolios, using all the months in our dataset. Our portfolio is defined by a set of ten weights $w_i, i = 1, 2, \dots, 10$, that sum to 1 (Negative weight corresponds to short selling). Then, we fix the mean value at some arbitrary value \bar{r} and find the feasible portfolio of minimum variance that has this mean. The mathematical problem of solving for the mean-variance optimal portfolio as known as the Markowitz problem is equivalent to the following problem:

$$\text{minimize } \frac{1}{2} \sum_{i,j=1}^{10} w_i w_j \sigma_{ij}$$

$$\text{subject to } \sum_{i=1}^{10} w_i \bar{r}_i = \bar{r}$$

$$\sum_{i=1}^{10} w_i = 1$$

To find the solutions for the Markowitz problem, we form the *Lagrangian* L with the *Lagrange multipliers* λ and μ .

$$L = \frac{1}{2} \sum_{i,j=1}^{10} w_i w_j \sigma_{ij} - \lambda \left(\sum_{i=1}^{10} w_i \bar{r}_i - \bar{r} \right) - \mu \left(\sum_{i=1}^{10} w_i - 1 \right) \quad (4.1)$$

After setting the first derivatives of the *Lagrangian* L with respect to λ , μ , and the 10 portfolio weights w_i ($i = 1, 2, \dots, 10$) equal to zero, we obtain

$$\sum_{j=1}^{10} \sigma_{ij} w_j - \lambda \bar{r}_i - \mu = 0 \text{ for } i = 1, 2, \dots, 10 \quad (4.2)$$

$$\sum_{i=1}^{10} w_i \bar{r}_i = \bar{r} \quad (4.3)$$

$$\sum_{i=1}^{10} w_i = 1 \quad (4.4)$$

With equations (4.2), (4.3), and (4.4), we can generate the mean-variance optimal set of portfolios as shown in figure 4.3. The top two plots compares the mean-variance optimal sets with selling short allowed and going long only, in which the plot on the top refers to the average equal weighted method and the plot in the middle refers to the average value weighted method. Both plots show that the mean-variance optimal set shifts to the left when selling short is allowed, which is what we have expected because the feasible solution set for going long only is the subset of the feasible solution set for selling short allowed.

Absent risk-free borrowing, the efficient frontier is the upper arc of the parabola (the mean-variance optimal set) because for any portfolio on the lower arc you can always find a portfolio on the upper arc with the same standard deviation and higher expected return. Once a risk-free rate is introduced, the efficient frontier is a line that is tangent to the upper arc of the parabola. [13].

With the assumption of selling short allowed, the plot at the bottom compares the mean-variance optimal sets for both weighting method. The plot shows that the mean-variance optimal set shifts to the left when we change from the average equal weighted method to the average value weighted method, but with the intersection near the expected monthly return of 0.018 and its standard deviation of 0.060. Therefore, if our risk appetite (or standard deviation) is lower than 0.060, we should hold the optimal portfolio formed with the average value weighted method, otherwise, we should hold the optimal portfolio formed with the average equal weighted method.

The result that the mean-variance optimal set shifts to the left when we change from the average equal weighted method to the average value weighted method can be explained by Banz's finding [2] that smaller firms have had higher risk adjusted returns, on average, than larger firms. The average equal weighted method put much more weight on the smaller firms than the larger firms because we have numerous of smaller firms in the market. Therefore, the lowest level of risk that is achievable with the average equal weighted method is higher than the one that is achievable with the average value weighted method. However, with the effect of diversification and selling short allowed, we can achieve higher expected monthly return with risk or standard deviation greater than 0.060 using the average equal weighted method.

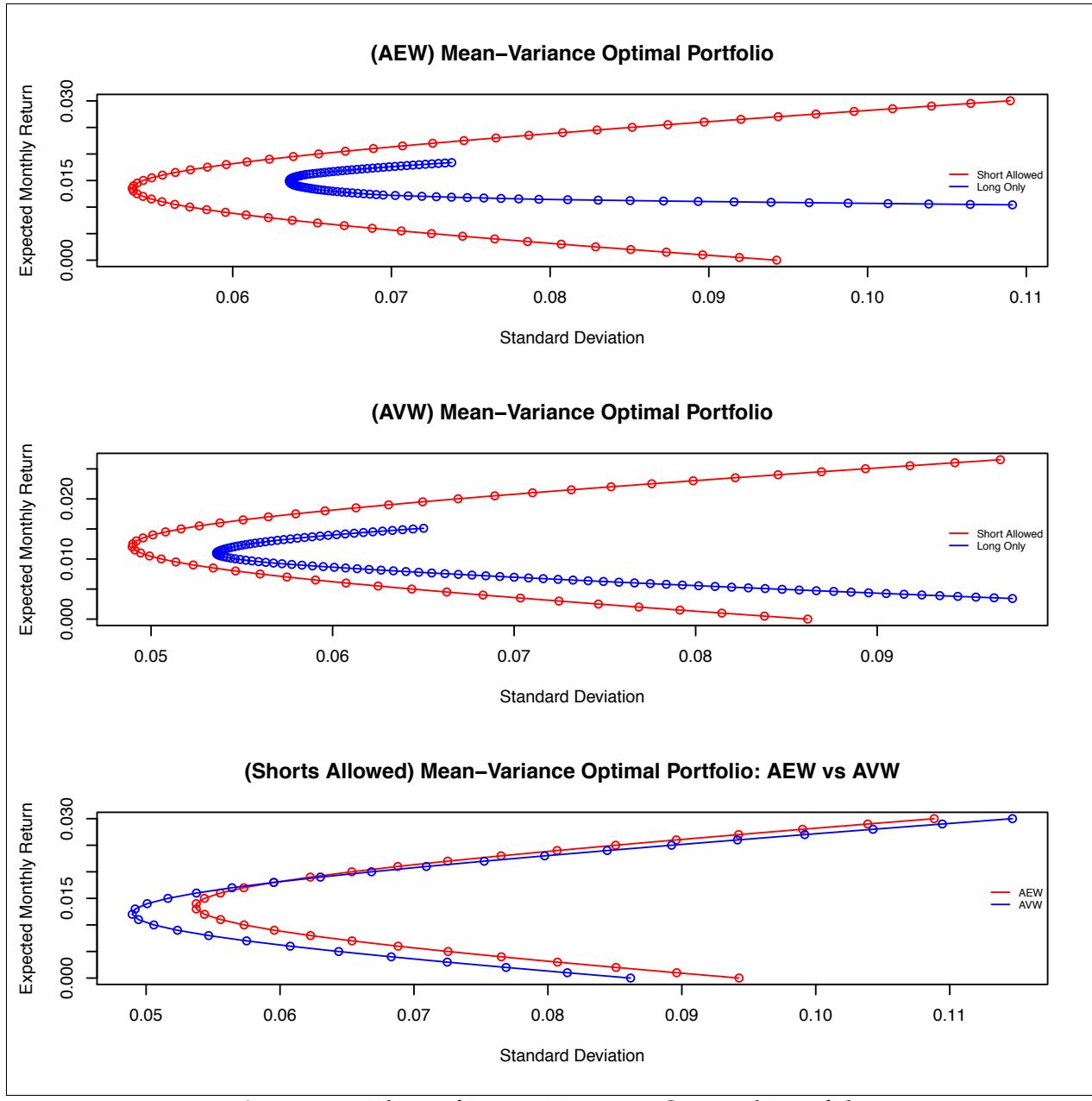


Figure 4.3: Plots of Mean–Variance Optimal Portfolio

4.3 Optimal Portfolio Weights

Tables 4.1 and 4.2 contain the solutions for the mean-variance optimal portfolio for the average equal weighted method and the average value weighted method respectively. It is easier to see the relationships among the minimum portfolio set and the corresponding optimal portfolio weights by looking at their correlations. The numerical results for the average equal weighted method and the average value weighted method are shown in tables 4.3 and 4.4 respectively; whereas the graphic results are shown in figures 4.4 and 4.5 respectively.

We find that for the average equal weighted method, the optimal weights for portfolios 2, 4, 6, 9, and 10 are positively correlated with the expected month return and risk; whereas for the average value weighted method, the optimal weights for portfolios 2, 4, and 10 are positively correlated with the expected month return and risk. Therefore, the average equal weighted method includes two additional portfolio (9 and 10) for the positive correlation.

Table 4.1: (AEW) Mean-Variance Optimal Portfolio: Solution

Mean	SD	W1	W2	W3	W4	W5	W6	W7	W8	W9	W10
0.000	0.094	0.40	-2.06	1.59	-1.39	3.06	-0.80	1.63	1.38	-0.45	-2.36
0.001	0.090	0.37	-1.97	1.47	-1.30	2.90	-0.71	1.54	1.32	-0.38	-2.23
0.002	0.085	0.33	-1.88	1.35	-1.21	2.74	-0.62	1.44	1.25	-0.31	-2.10
0.003	0.081	0.30	-1.79	1.24	-1.12	2.58	-0.53	1.35	1.19	-0.24	-1.97
0.004	0.077	0.26	-1.69	1.12	-1.04	2.42	-0.44	1.25	1.12	-0.17	-1.84
0.005	0.073	0.23	-1.60	1.00	-0.95	2.26	-0.34	1.16	1.05	-0.10	-1.71
0.006	0.069	0.20	-1.51	0.88	-0.86	2.10	-0.25	1.06	0.99	-0.03	-1.58
0.007	0.065	0.16	-1.41	0.76	-0.78	1.94	-0.16	0.97	0.92	0.04	-1.44
0.008	0.062	0.13	-1.32	0.64	-0.69	1.78	-0.07	0.87	0.86	0.11	-1.31
0.009	0.060	0.09	-1.23	0.53	-0.60	1.62	0.02	0.78	0.79	0.19	-1.18
0.010	0.057	0.06	-1.14	0.41	-0.51	1.46	0.12	0.68	0.72	0.26	-1.05
0.011	0.056	0.02	-1.04	0.29	-0.43	1.30	0.21	0.59	0.66	0.33	-0.92
0.012	0.054	-0.01	-0.95	0.17	-0.34	1.14	0.30	0.49	0.59	0.40	-0.79
0.013	0.054	-0.04	-0.86	0.05	-0.25	0.98	0.39	0.40	0.53	0.47	-0.66
0.014	0.054	-0.08	-0.77	-0.06	-0.16	0.82	0.49	0.30	0.46	0.54	-0.53
0.015	0.054	-0.11	-0.67	-0.18	-0.08	0.66	0.58	0.21	0.39	0.61	-0.39
0.016	0.056	-0.15	-0.58	-0.30	0.01	0.50	0.67	0.11	0.33	0.68	-0.26
0.017	0.057	-0.18	-0.49	-0.42	0.10	0.33	0.76	0.01	0.26	0.75	-0.13
0.018	0.060	-0.22	-0.39	-0.54	0.18	0.17	0.85	-0.08	0.19	0.82	-0.00
0.019	0.062	-0.25	-0.30	-0.65	0.27	0.01	0.95	-0.18	0.13	0.89	0.13
0.020	0.065	-0.29	-0.21	-0.77	0.36	-0.15	1.04	-0.27	0.06	0.96	0.26
0.021	0.069	-0.32	-0.12	-0.89	0.45	-0.31	1.13	-0.37	-0.00	1.03	0.39
0.022	0.073	-0.35	-0.02	-1.01	0.53	-0.47	1.22	-0.46	-0.07	1.11	0.52
0.023	0.076	-0.39	0.07	-1.13	0.62	-0.63	1.31	-0.56	-0.14	1.18	0.66
0.024	0.081	-0.42	0.16	-1.25	0.71	-0.79	1.41	-0.65	-0.20	1.25	0.79
0.025	0.085	-0.46	0.26	-1.36	0.79	-0.95	1.50	-0.75	-0.27	1.32	0.92
0.026	0.090	-0.49	0.35	-1.48	0.88	-1.11	1.59	-0.84	-0.33	1.39	1.05
0.027	0.094	-0.53	0.44	-1.60	0.97	-1.27	1.68	-0.94	-0.40	1.46	1.18
0.028	0.099	-0.56	0.53	-1.72	1.06	-1.43	1.77	-1.03	-0.47	1.53	1.31
0.029	0.104	-0.59	0.63	-1.84	1.14	-1.59	1.87	-1.13	-0.53	1.60	1.44
0.030	0.109	-0.63	0.72	-1.95	1.23	-1.75	1.96	-1.22	-0.60	1.67	1.57

Table 4.2: (AVW) Mean-Variance Optimal Portfolio: Solution

Mean	SD	W1	W2	W3	W4	W5	W6	W7	W8	W9	W10
0.000	0.086	0.82	-1.29	0.30	-0.33	0.72	0.89	0.55	0.66	0.13	-1.45
0.001	0.081	0.72	-1.20	0.28	-0.28	0.70	0.82	0.52	0.64	0.12	-1.33
0.002	0.077	0.63	-1.12	0.26	-0.24	0.69	0.76	0.49	0.63	0.12	-1.22
0.003	0.072	0.54	-1.04	0.24	-0.19	0.67	0.70	0.46	0.61	0.11	-1.10
0.004	0.068	0.45	-0.95	0.22	-0.14	0.66	0.64	0.43	0.59	0.10	-0.99
0.005	0.064	0.35	-0.87	0.20	-0.10	0.64	0.58	0.40	0.58	0.09	-0.87
0.006	0.061	0.26	-0.79	0.18	-0.05	0.63	0.52	0.37	0.56	0.09	-0.76
0.007	0.058	0.17	-0.70	0.15	-0.01	0.61	0.46	0.34	0.54	0.08	-0.64
0.008	0.055	0.07	-0.62	0.13	0.04	0.59	0.39	0.31	0.52	0.07	-0.52
0.009	0.052	-0.02	-0.53	0.11	0.08	0.58	0.33	0.28	0.51	0.06	-0.41
0.010	0.051	-0.11	-0.45	0.09	0.13	0.56	0.27	0.25	0.49	0.05	-0.29
0.011	0.049	-0.20	-0.37	0.07	0.17	0.55	0.21	0.22	0.47	0.05	-0.18
0.012	0.049	-0.30	-0.28	0.05	0.22	0.53	0.15	0.19	0.46	0.04	-0.06
0.013	0.049	-0.39	-0.20	0.03	0.26	0.52	0.09	0.16	0.44	0.03	0.06
0.014	0.050	-0.48	-0.12	0.01	0.31	0.50	0.03	0.13	0.42	0.02	0.17
0.015	0.052	-0.58	-0.03	-0.02	0.35	0.49	-0.04	0.10	0.41	0.02	0.29
0.016	0.054	-0.67	0.05	-0.04	0.40	0.47	-0.10	0.07	0.39	0.01	0.40
0.017	0.056	-0.76	0.13	-0.06	0.45	0.46	-0.16	0.04	0.37	0.00	0.52
0.018	0.060	-0.85	0.22	-0.08	0.49	0.44	-0.22	0.01	0.36	-0.01	0.64
0.019	0.063	-0.95	0.30	-0.10	0.54	0.43	-0.28	-0.02	0.34	-0.01	0.75
0.020	0.067	-1.04	0.39	-0.12	0.58	0.41	-0.34	-0.05	0.32	-0.02	0.87
0.021	0.071	-1.13	0.47	-0.14	0.63	0.40	-0.40	-0.08	0.31	-0.03	0.98
0.022	0.075	-1.23	0.55	-0.16	0.67	0.38	-0.47	-0.11	0.29	-0.04	1.10
0.023	0.080	-1.32	0.64	-0.19	0.72	0.37	-0.53	-0.14	0.27	-0.04	1.21
0.024	0.084	-1.41	0.72	-0.21	0.76	0.35	-0.59	-0.17	0.26	-0.05	1.33
0.025	0.089	-1.50	0.80	-0.23	0.81	0.34	-0.65	-0.19	0.24	-0.06	1.45
0.026	0.094	-1.60	0.89	-0.25	0.85	0.32	-0.71	-0.22	0.22	-0.07	1.56
0.027	0.099	-1.69	0.97	-0.27	0.90	0.31	-0.77	-0.25	0.21	-0.08	1.68
0.028	0.104	-1.78	1.05	-0.29	0.95	0.29	-0.83	-0.28	0.19	-0.08	1.79
0.029	0.109	-1.88	1.14	-0.31	0.99	0.28	-0.90	-0.31	0.17	-0.09	1.91
0.030	0.115	-1.97	1.22	-0.33	1.04	0.26	-0.96	-0.34	0.16	-0.10	2.03

Table 4.3: (AEW) Mean-Variance Optimal Portfolio: Correlation

	Mean	SD	W1	W2	W3	W4	W5	W6	W7	W8	W9	W10
Mean	1.0	0.3	-1.0	1.0	-1.0	1.0	-1.0	1.0	-1.0	-1.0	1.0	1.0
SD	0.3	1.0	-0.3	0.3	-0.3	0.3	-0.3	0.3	-0.3	-0.3	0.3	0.3
W1	-1.0	-0.3	1.0	-1.0	1.0	-1.0	1.0	-1.0	1.0	1.0	-1.0	-1.0
W2	1.0	0.3	-1.0	1.0	-1.0	1.0	-1.0	1.0	-1.0	-1.0	1.0	1.0
W3	-1.0	-0.3	1.0	-1.0	1.0	-1.0	1.0	-1.0	1.0	1.0	-1.0	-1.0
W4	1.0	0.3	-1.0	1.0	-1.0	1.0	-1.0	1.0	-1.0	-1.0	1.0	1.0
W5	-1.0	-0.3	1.0	-1.0	1.0	-1.0	1.0	-1.0	1.0	1.0	-1.0	-1.0
W6	1.0	0.3	-1.0	1.0	-1.0	1.0	-1.0	1.0	-1.0	-1.0	1.0	1.0
W7	-1.0	-0.3	1.0	-1.0	1.0	-1.0	1.0	-1.0	1.0	1.0	-1.0	-1.0
W8	-1.0	-0.3	1.0	-1.0	1.0	-1.0	1.0	-1.0	1.0	1.0	-1.0	-1.0
W9	1.0	0.3	-1.0	1.0	-1.0	1.0	-1.0	1.0	-1.0	-1.0	1.0	1.0
W10	1.0	0.3	-1.0	1.0	-1.0	1.0	-1.0	1.0	-1.0	-1.0	1.0	1.0

Table 4.4: (AVW) Mean-Variance Optimal Portfolio: Correlation

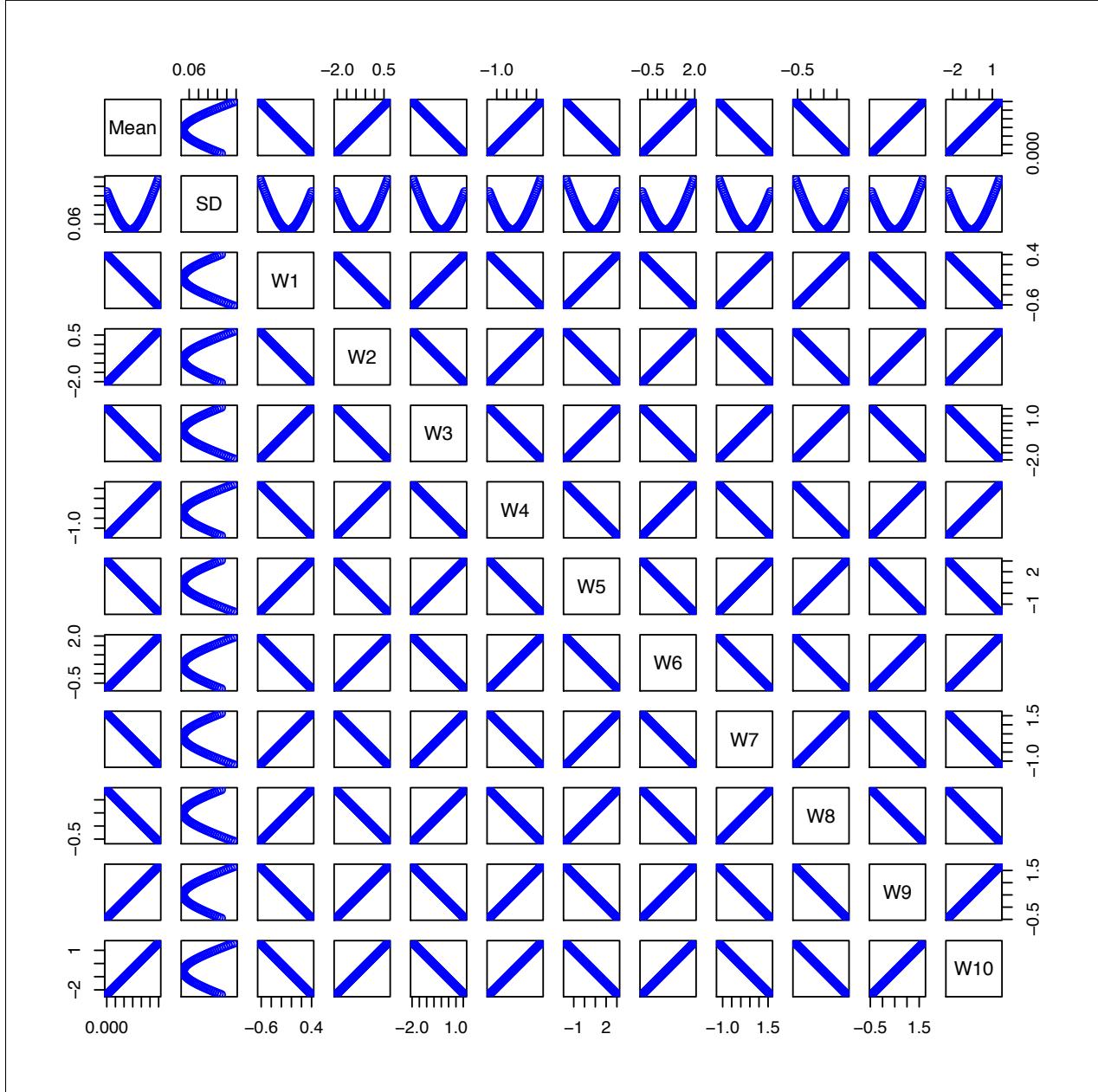


Figure 4.4: (AEW) Scatterplot of the Portfolio's Mean, SD, and Weight

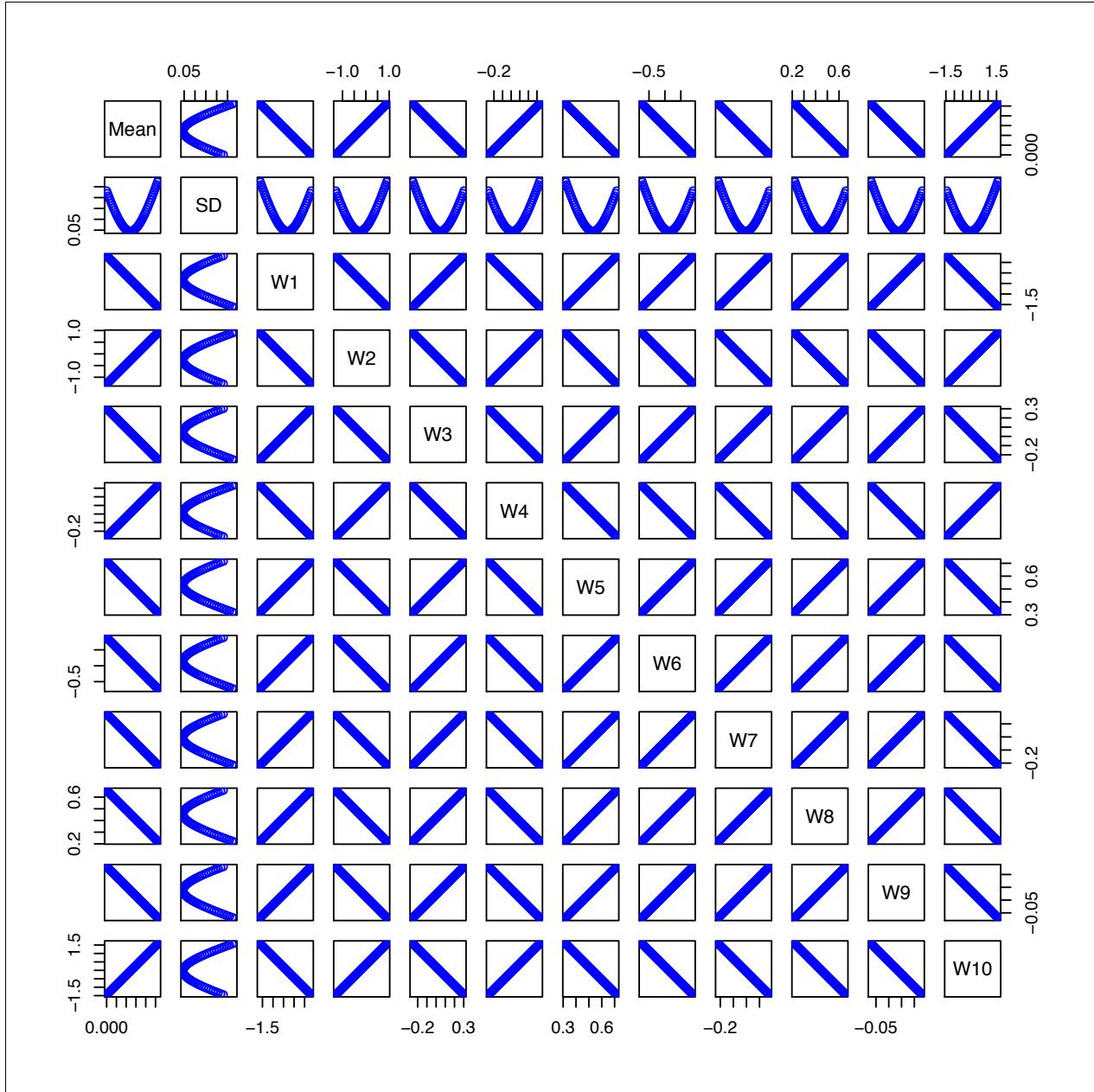


Figure 4.5: (AVW) Scatterplot of the Portfolio's Mean, SD, and Weight

Chapter 5

Conclusion

Based on the results of Lilliefors test for normality, with 5% α level, the random errors are not Gaussian distributed, therefore the OLS estimates are not equivalent to the maximum-likelihood estimates. For OLS, the random errors do not have to be Gaussian distributed.

Through our regression diagnostic, the random errors appear to be distributed with mean zero. In addition, the random errors appear to be independent of the observed data of the Fama French Carhart four factors. However, based on the times series of residuals, the residuals seem not to be independent and identically distributed because they have larger fluctuations before 1940 and after 1980. The time series plots also suggest that the homoscedasticity (constant variance) assumption is violated, which are confirmed by the results of the non-constant variance score test with 5% alpha level. Our failure of meeting some of the OLS assumptions suggests that our dataset contains influential observations.

Covariance ratio identifies more influential observations than Cook's distance. Out of 1,039 monthly observations, for the average equal weighted method, Cook's distance and covariance ratio identify 5 and 187 influential observations respectively. For the average value weighted method, Cook's distance and covariance ratio identify 4 and 232 influential observations respectively.

We use Cook's distance to examine influence on betas estimates. After excluding the influential observations identified by Cook's distance from the original data set, for the average equal weighted method, the β_{RMRF} and β_{SMB} estimates are more convex on portfolio rankings and β_{HML} estimates are more concave on portfolio rankings. We have also found that all the influential observations identified by Cook's Distance are connected with major historical events.

We use covariance ratio to examine influence on t-statistics. After excluding the in-

fluential observations identified by covariance ratio from the original data set, there are substantial percentage changes for both the Newey-West t-statistics and the standard t-statistics. The Newey-West t-statistics appear to be more sensitive to the elimination of influential observations than the ordinary t-statistics.

We apply the Lilliefors test on panels B and C to test the null hypothesis that the residuals come from the Gaussian distribution. For panel B, the results of Lilliefors test for normality remain the same as the ones for panel A. However, for panel C, the normality assumption seem to be valid for 3 and 8 out of 10 portfolios for the average equal weighted and average value weighted methods respectively.

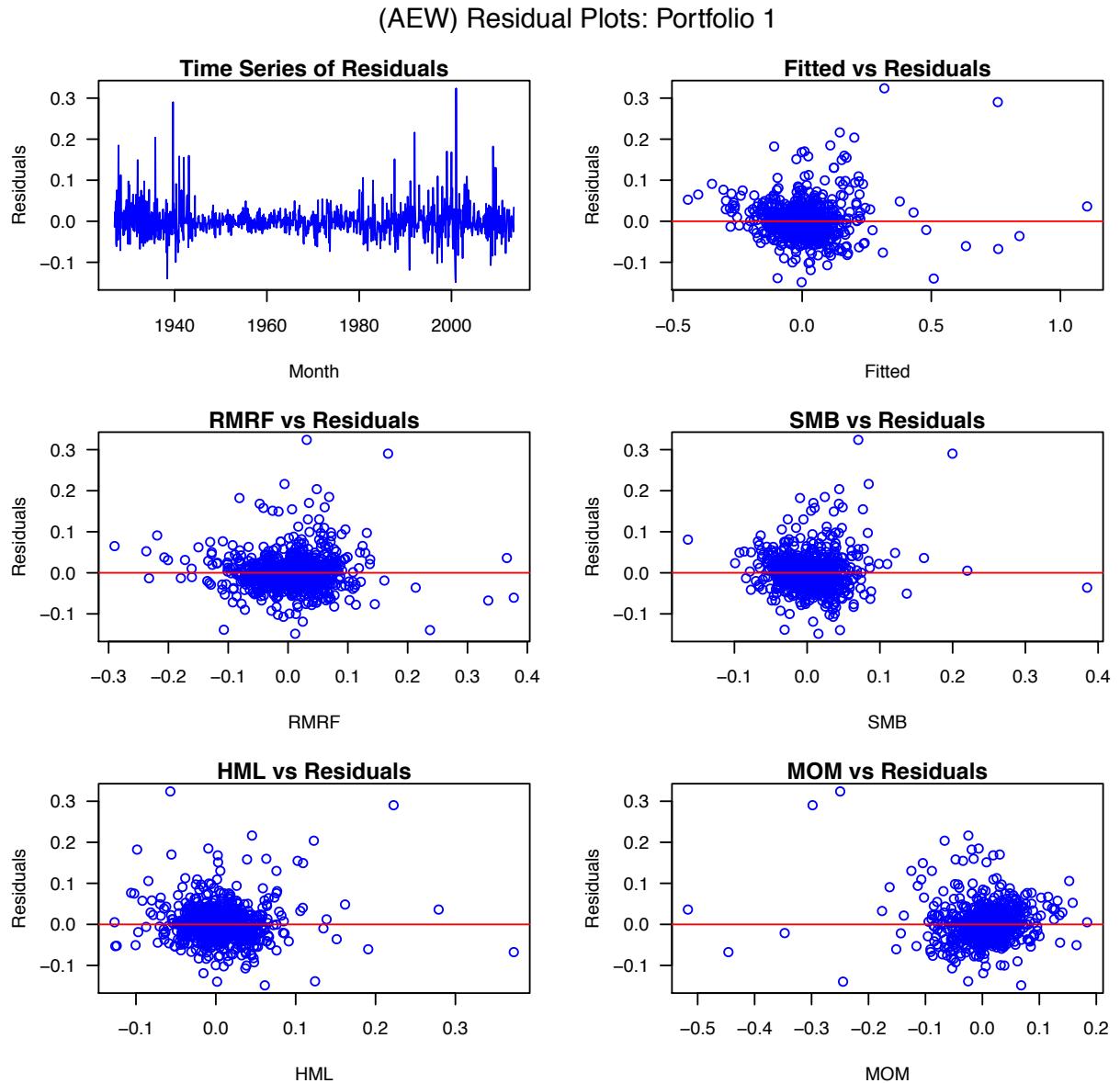
We also apply the non-constant variance score test on panels B and C to test the null hypothesis that the residuals are distributed with constant variance. For panel B, there is 1 additional portfolio seem to be distributed with constant variance for each weighting methods. Whereas for panel C, there are 7 and 5 additional portfolios seem to be distributed with constant variance for the average equal weighted and average value weighted methods respectively.

By comparing the two mean-variance optimal portfolios, in which the portfolios are both comprised of the 10 portfolios formed on momentum but with two different weighted methods, we have found that the mean-variance optimal set shifts to the left when we change from the average equal weighted method to the average value weighted method, but with the intersection near the expected monthly return of 0.018 and its standard deviation of 0.060. To increase the expected monthly return and therefore the associated risk of an efficient portfolio, the optimal portfolio weights for portfolios 2, 4, and 10 have to be increased for both weighted methods, in addition to increasing the optimal portfolio weights for portfolios 6 and 9 for average equal weighted method.

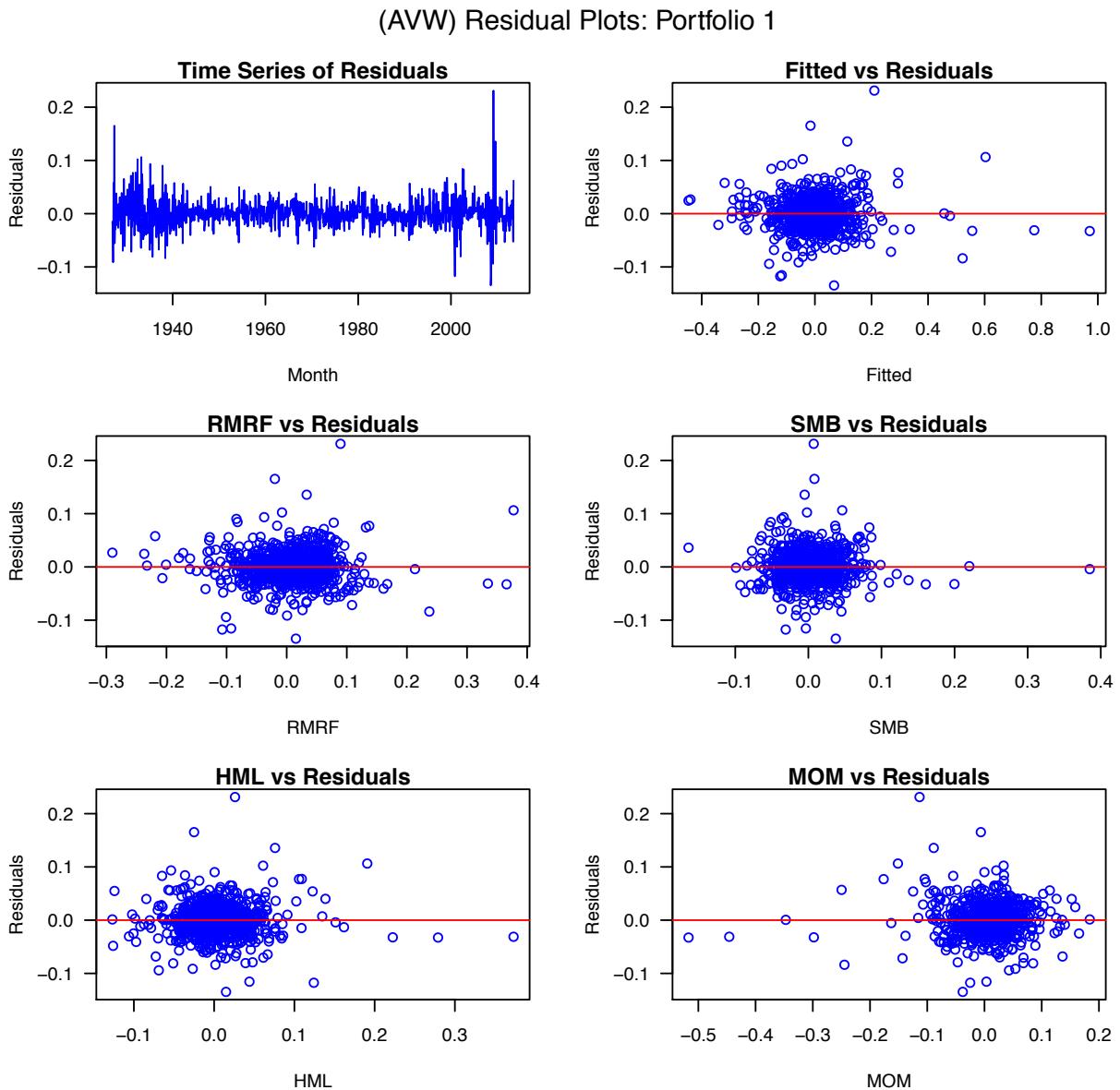
Appendix A

Residual Plots

A.1 (AEW) Portfolio 1

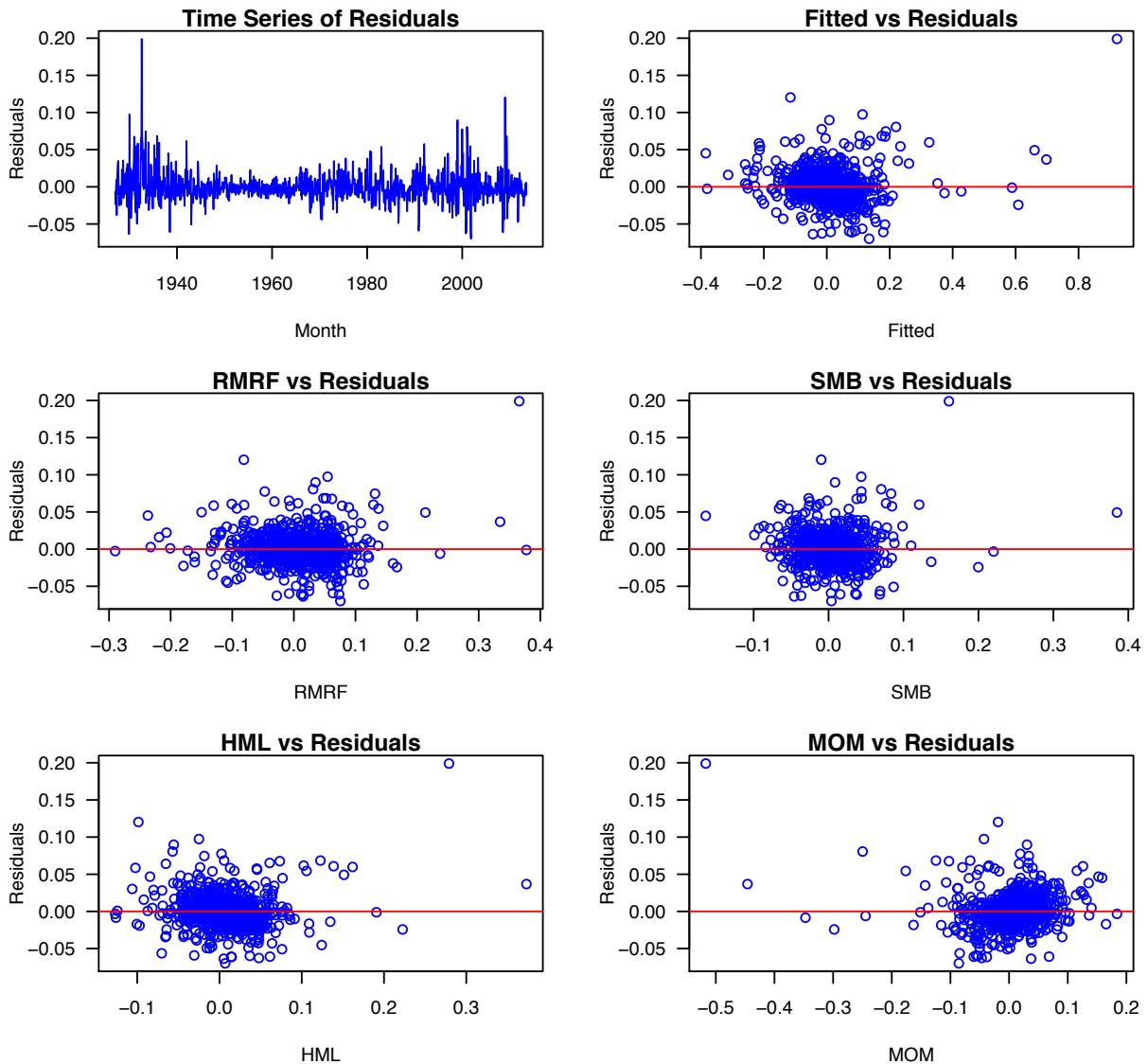


A.2 (AVW) Portfolio 1

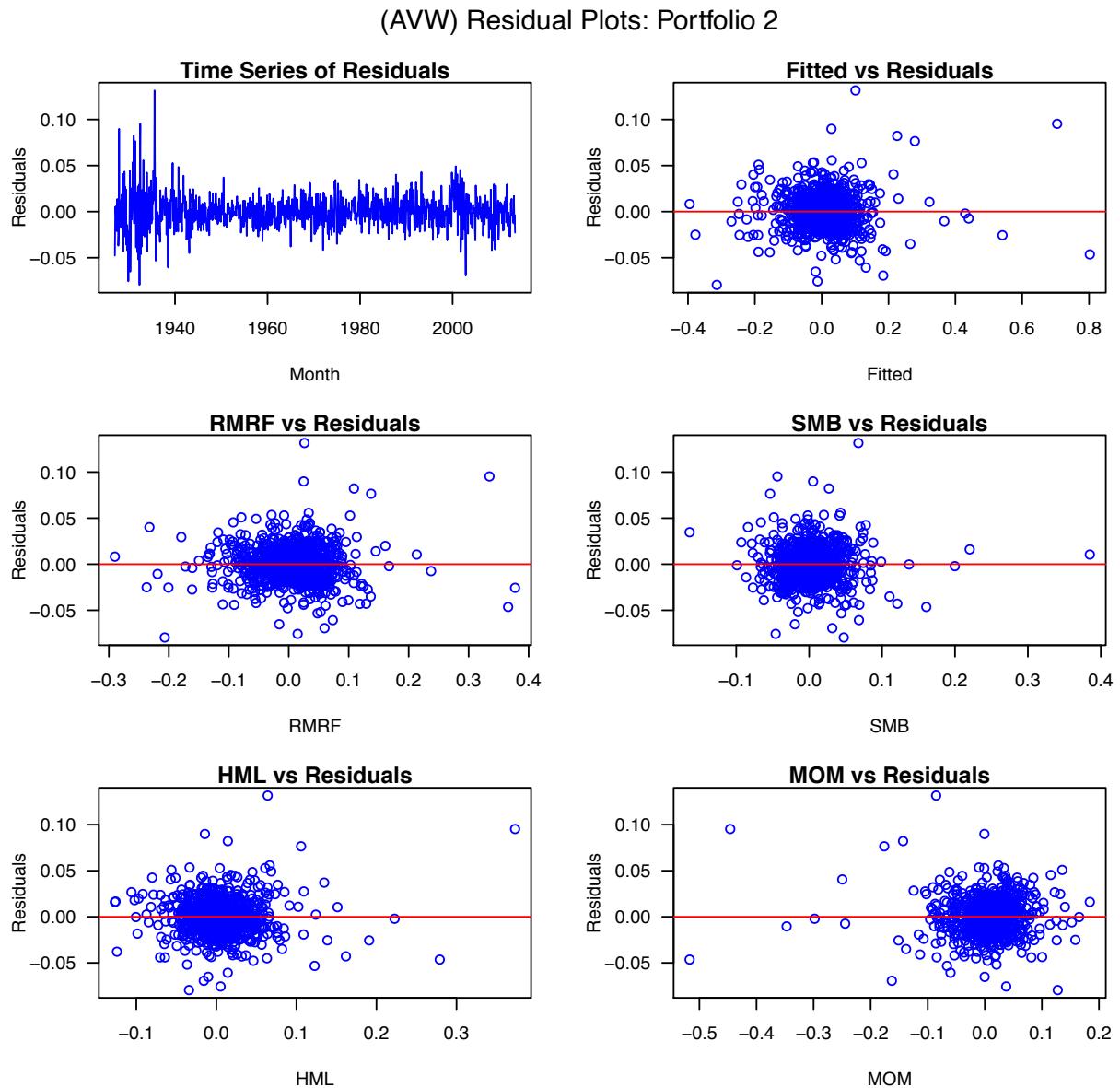


A.3 (AEW) Portfolio 2

(AEW) Residual Plots: Portfolio 2

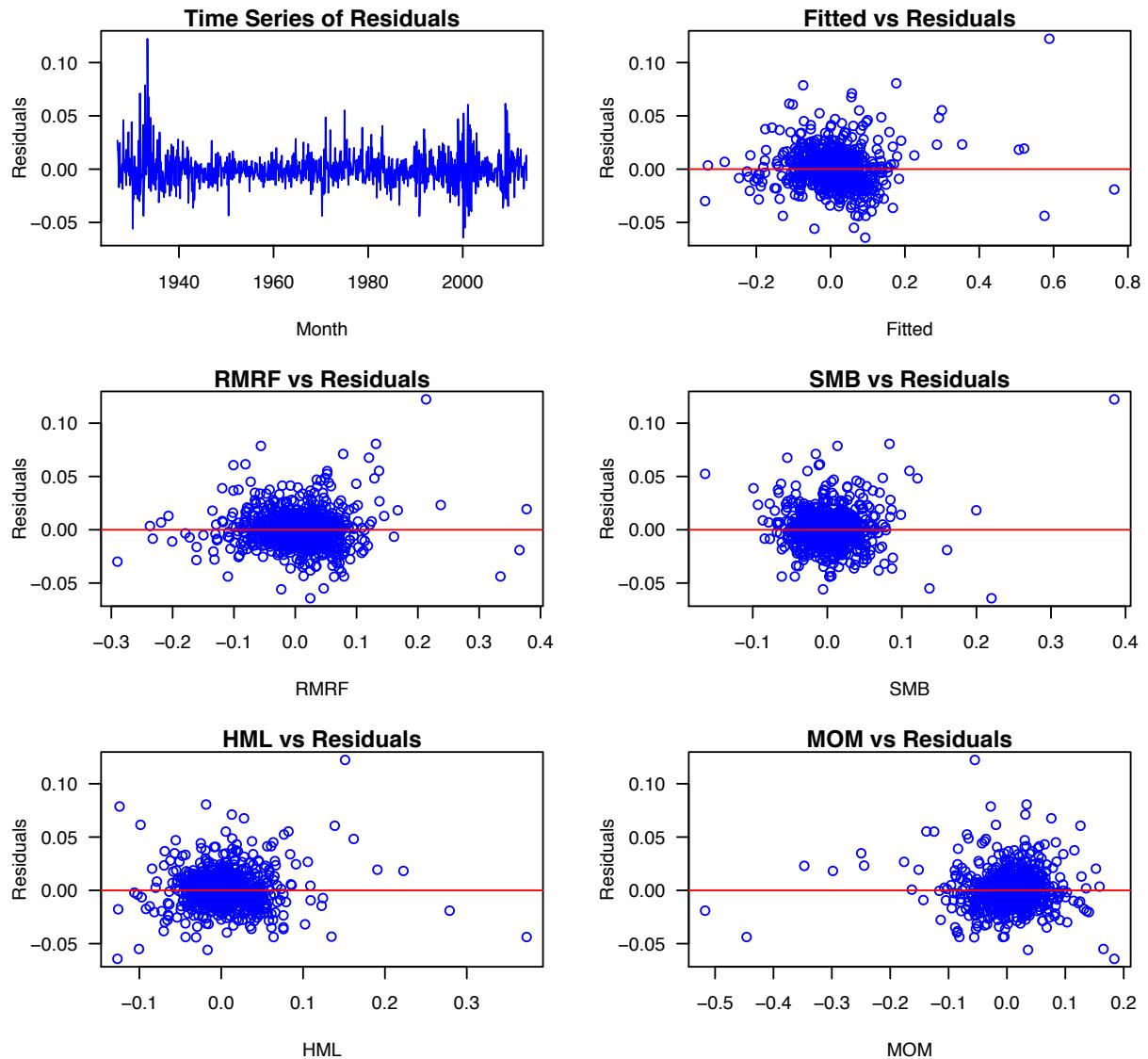


A.4 (AVW) Portfolio 2



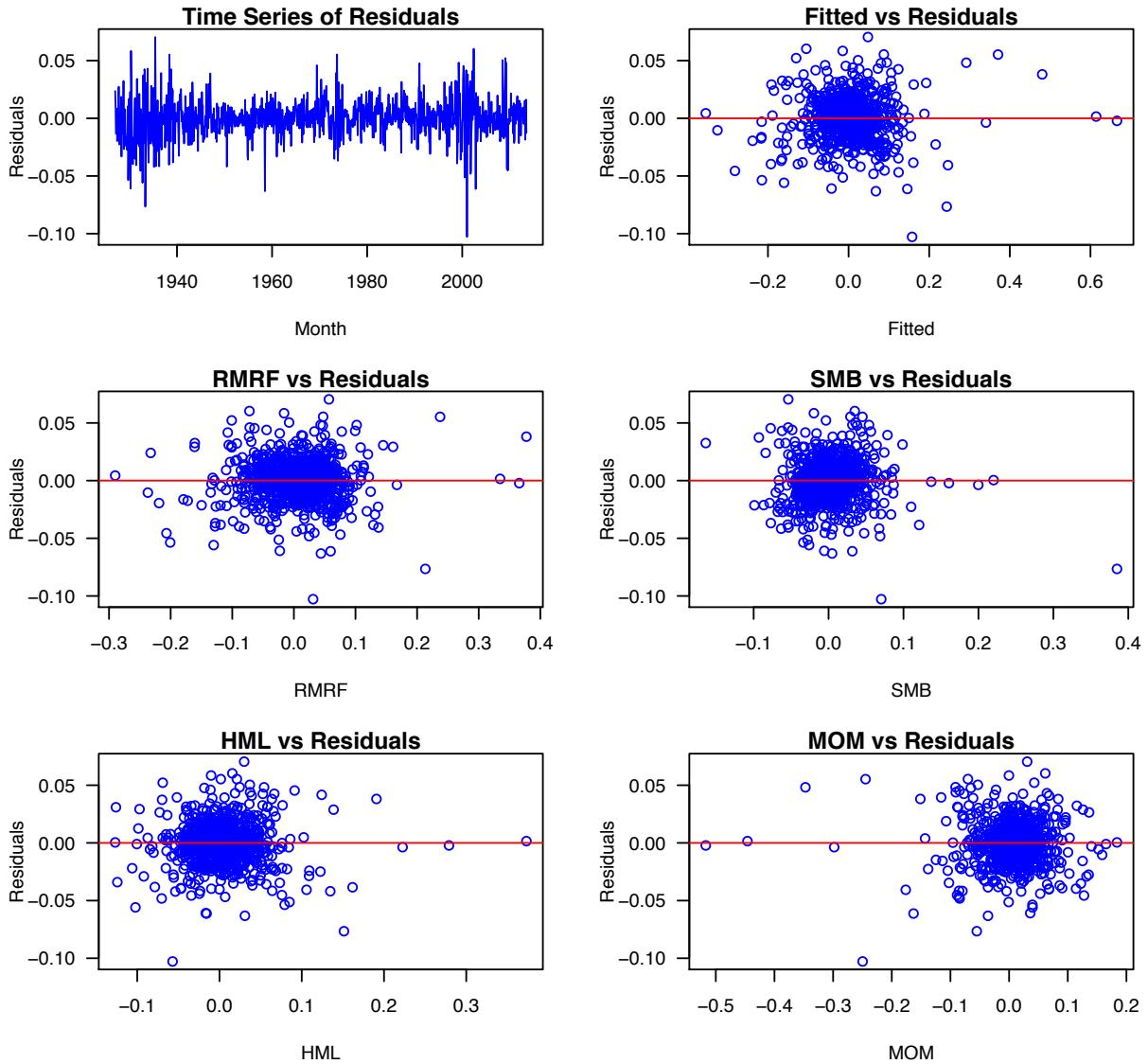
A.5 (AEW) Portfolio 3

(AEW) Residual Plots: Portfolio 3

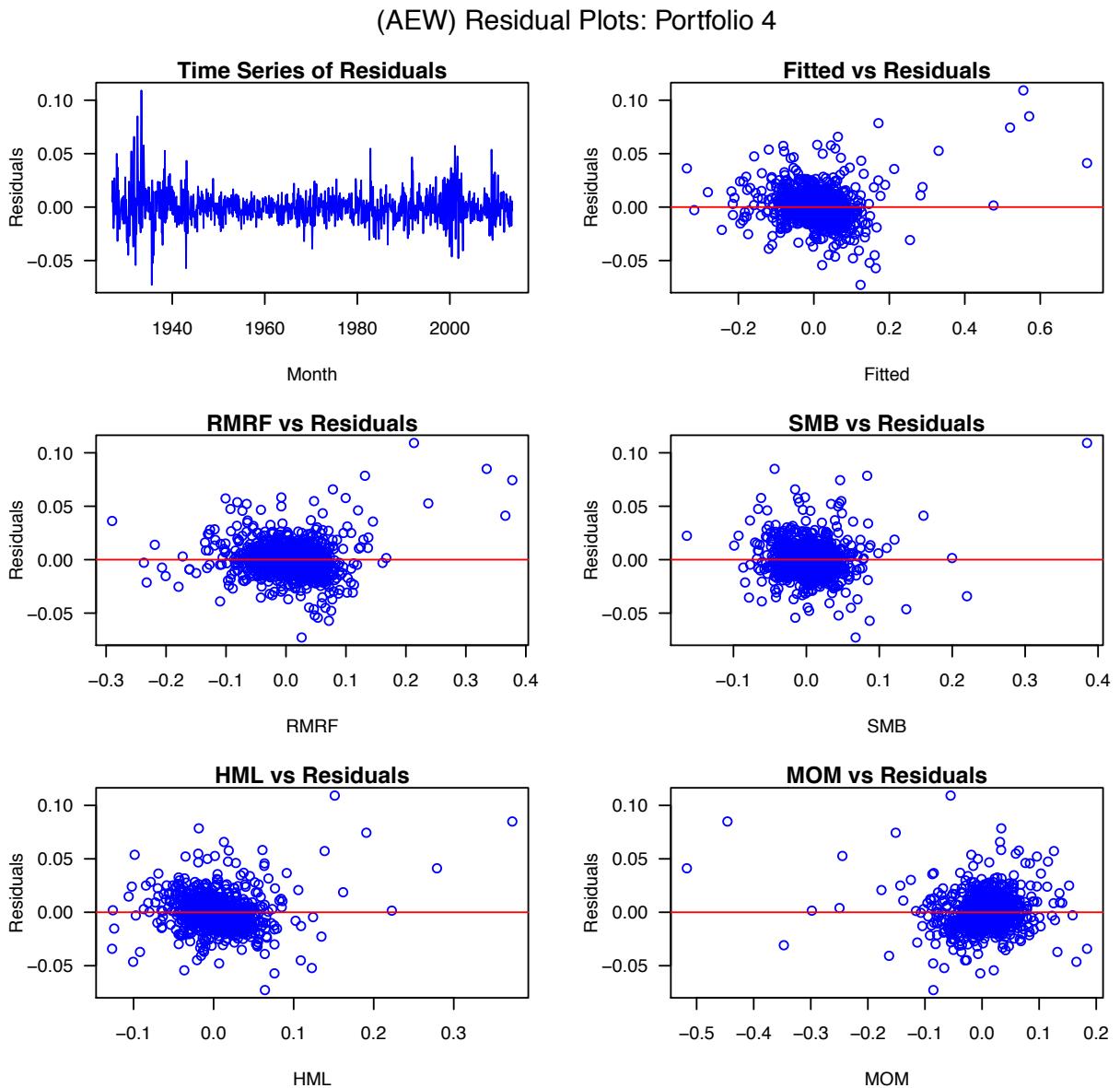


A.6 (AVW) Portfolio 3

(AVW) Residual Plots: Portfolio 3

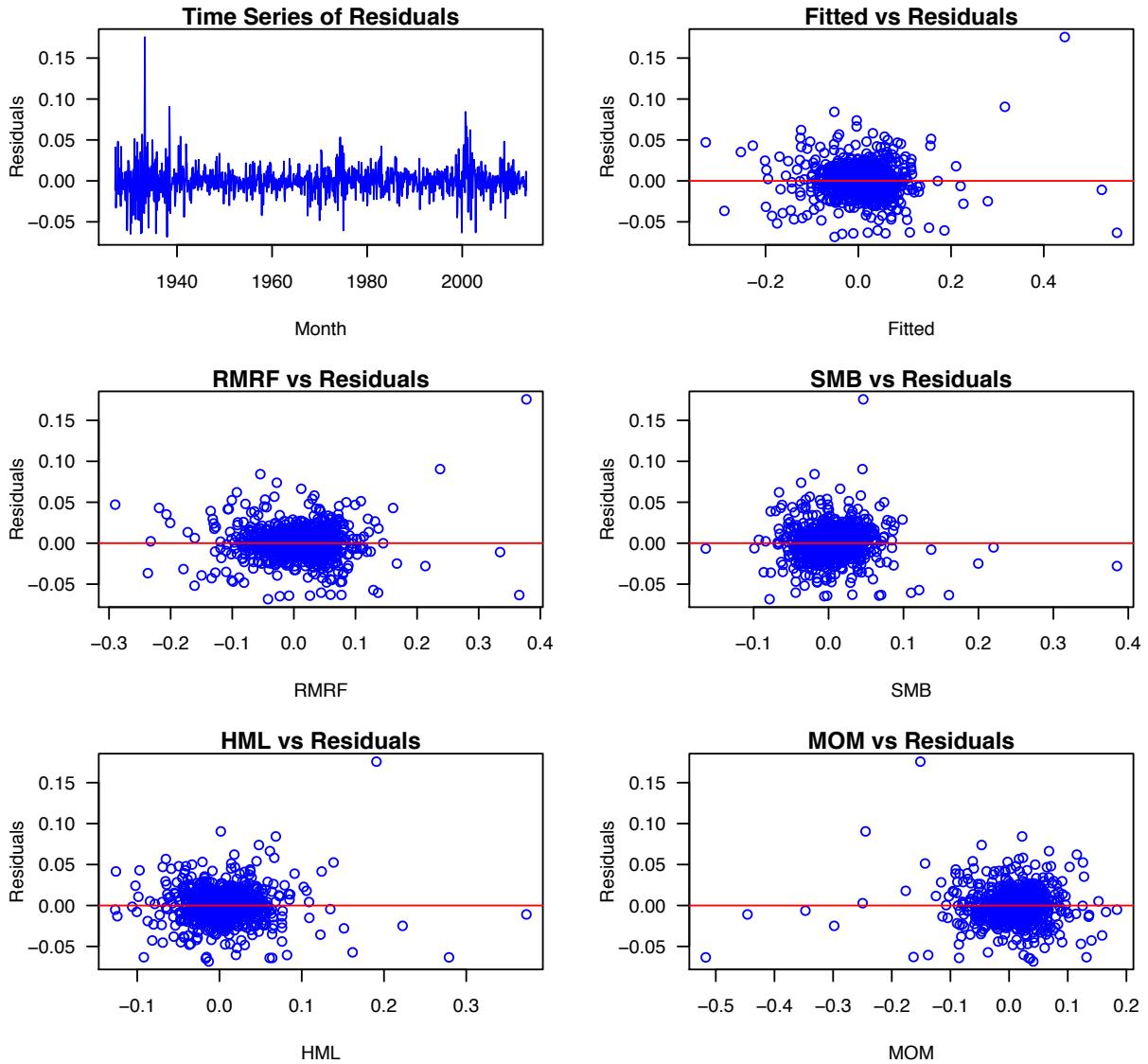


A.7 (AEW) Portfolio 4



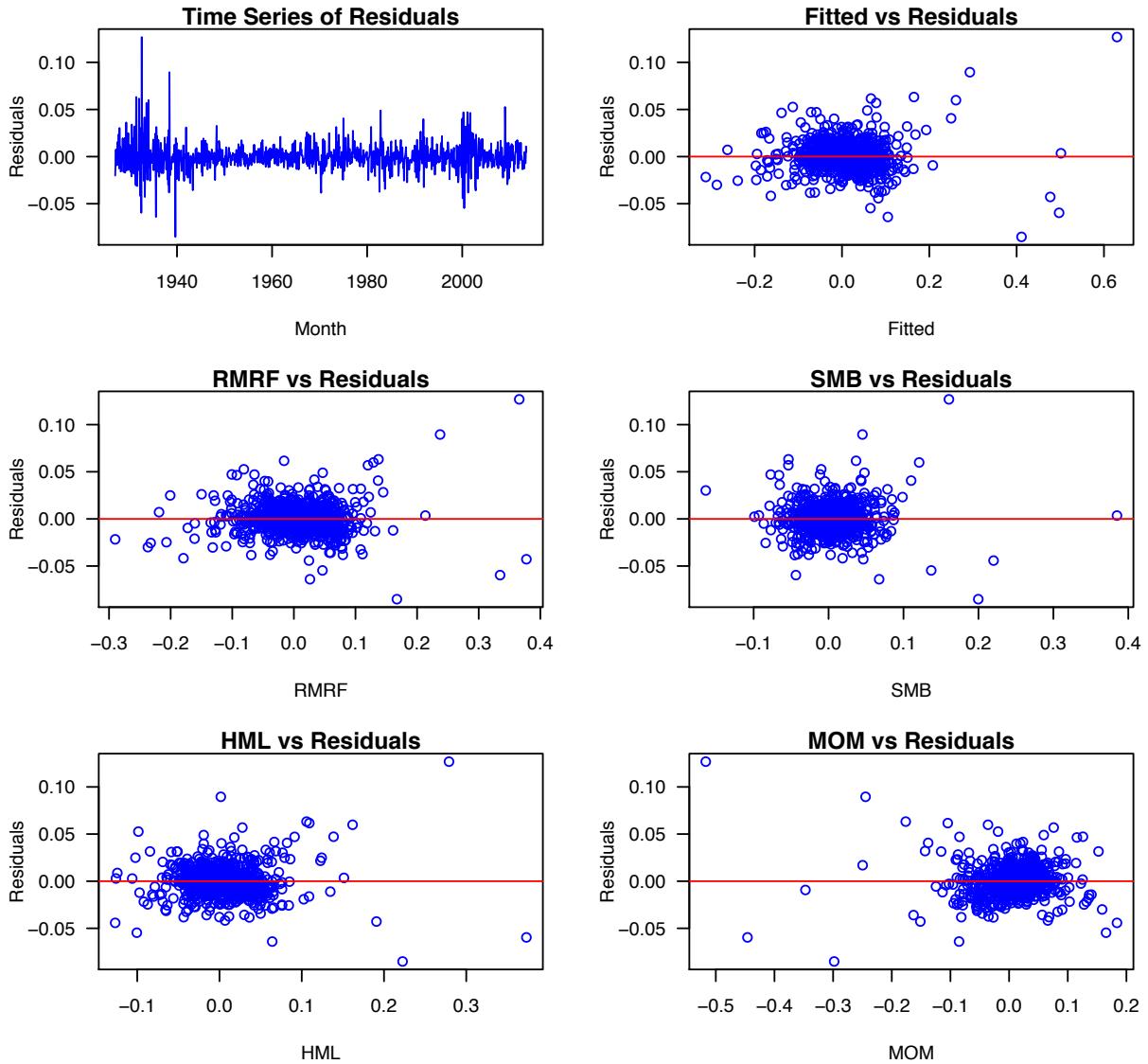
A.8 (AVW) Portfolio 4

(AVW) Residual Plots: Portfolio 4



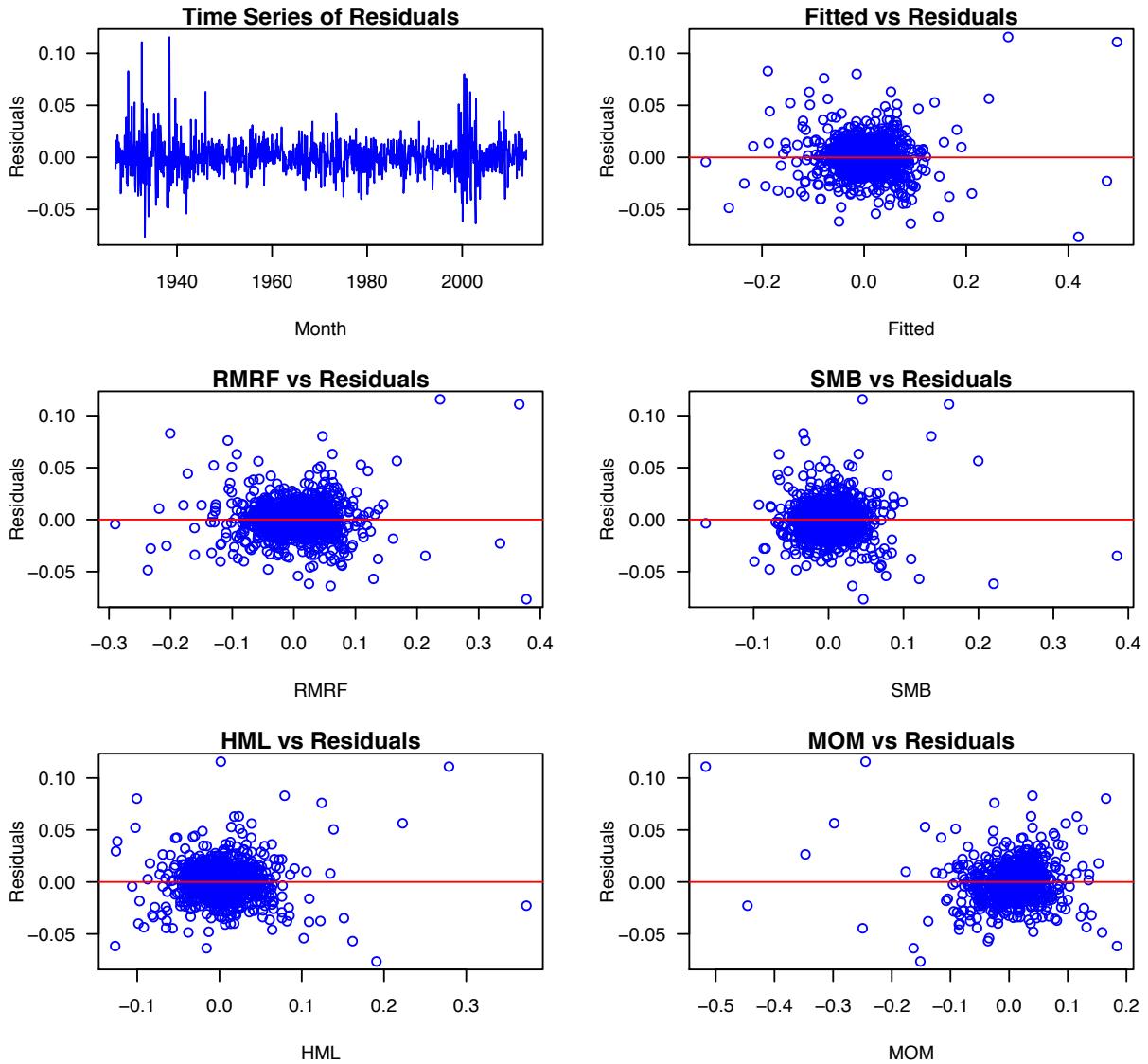
A.9 (AEW) Portfolio 5

(AEW) Residual Plots: Portfolio 5



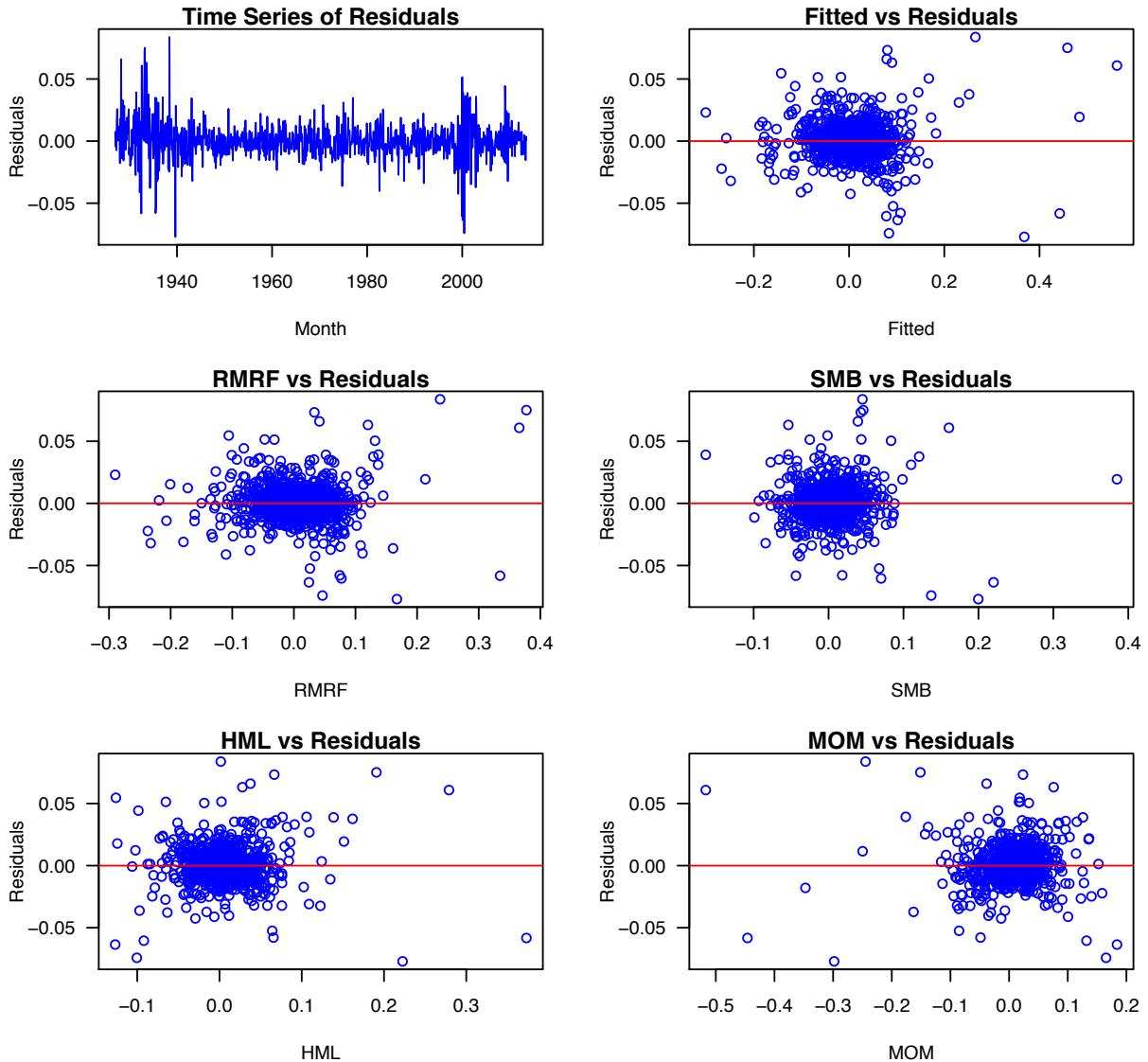
A.10 (AVW) Portfolio 5

(AVW) Residual Plots: Portfolio 5



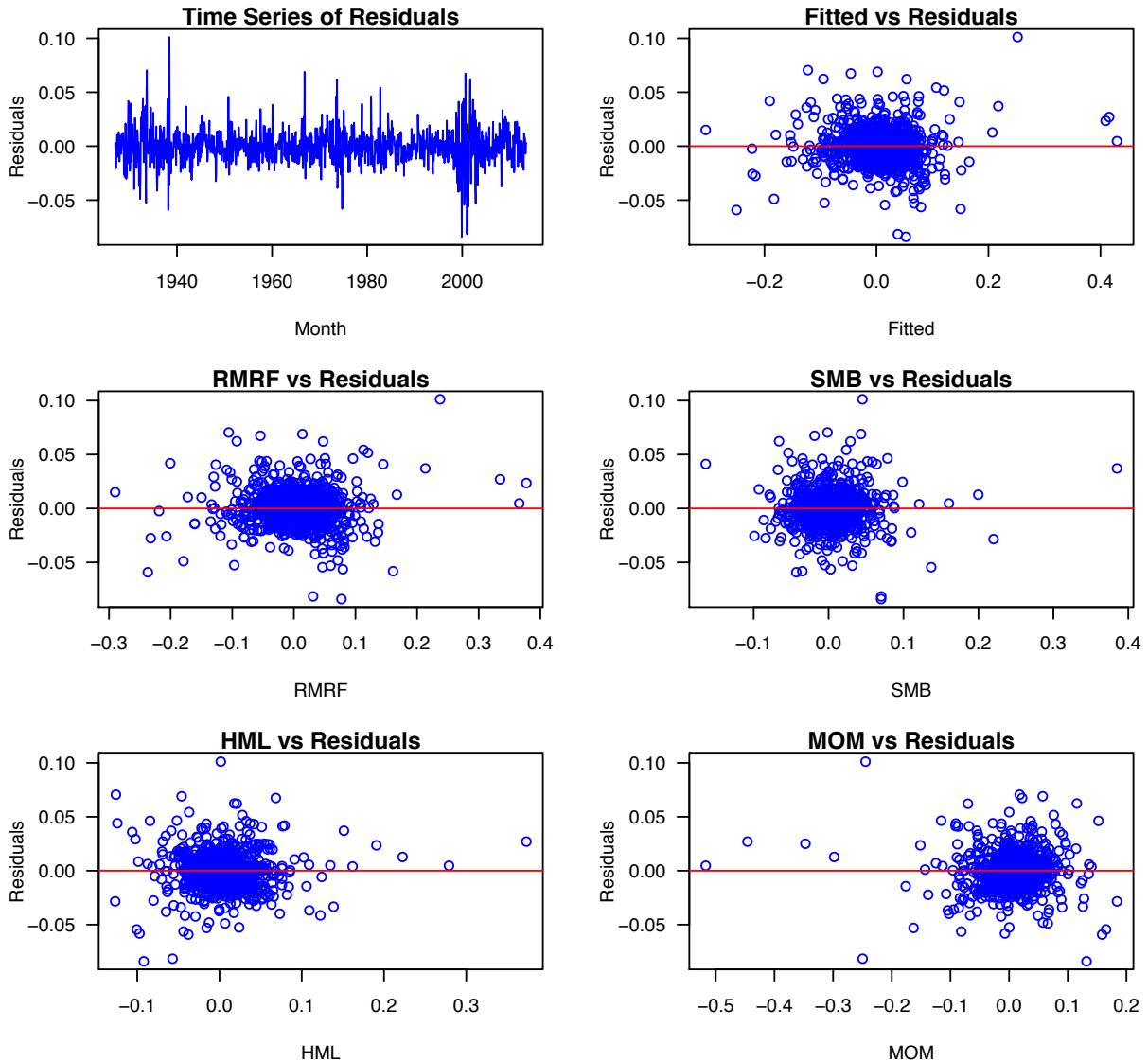
A.11 (AEW) Portfolio 6

(AEW) Residual Plots: Portfolio 6



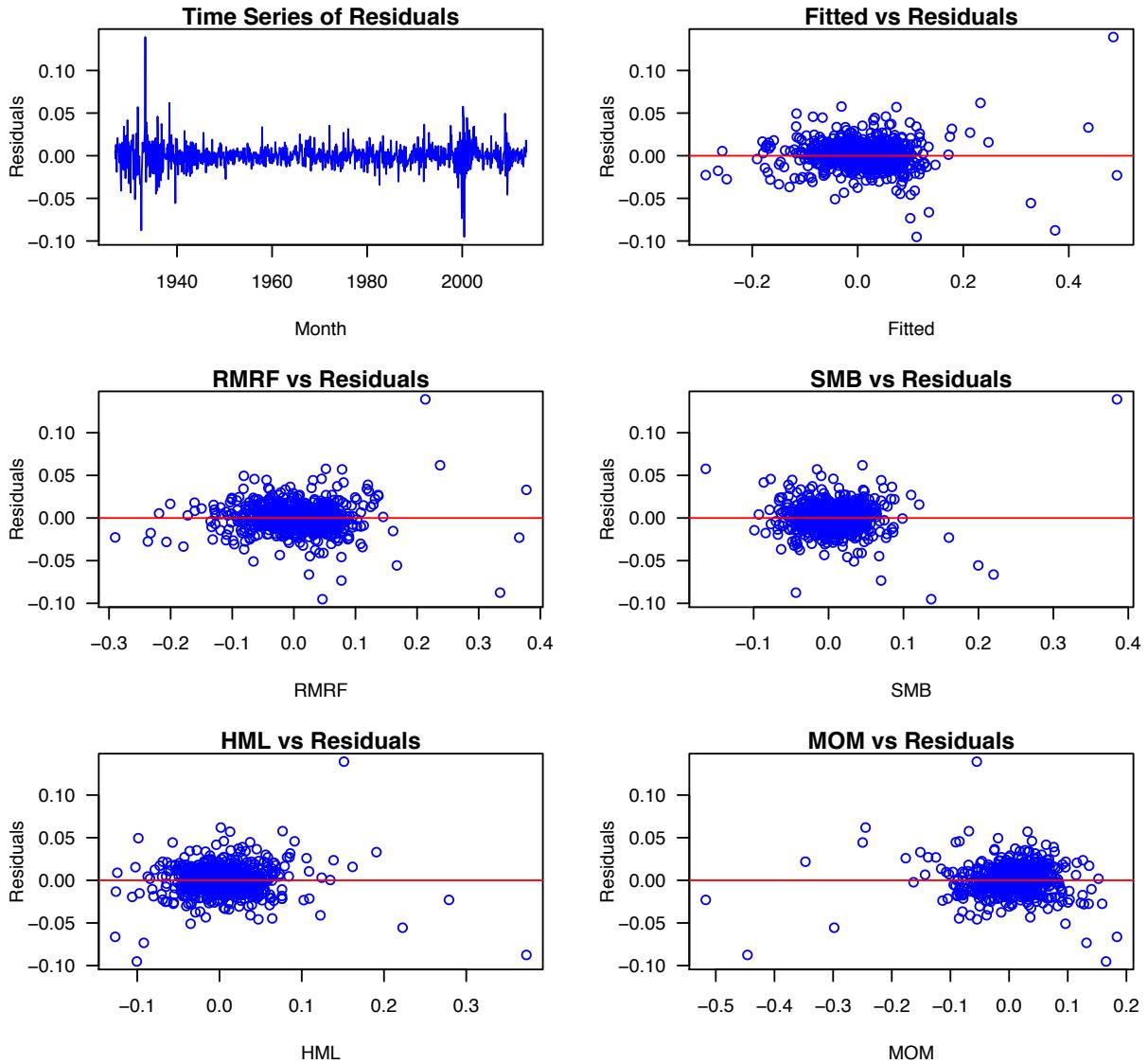
A.12 (AVW) Portfolio 6

(AVW) Residual Plots: Portfolio 6



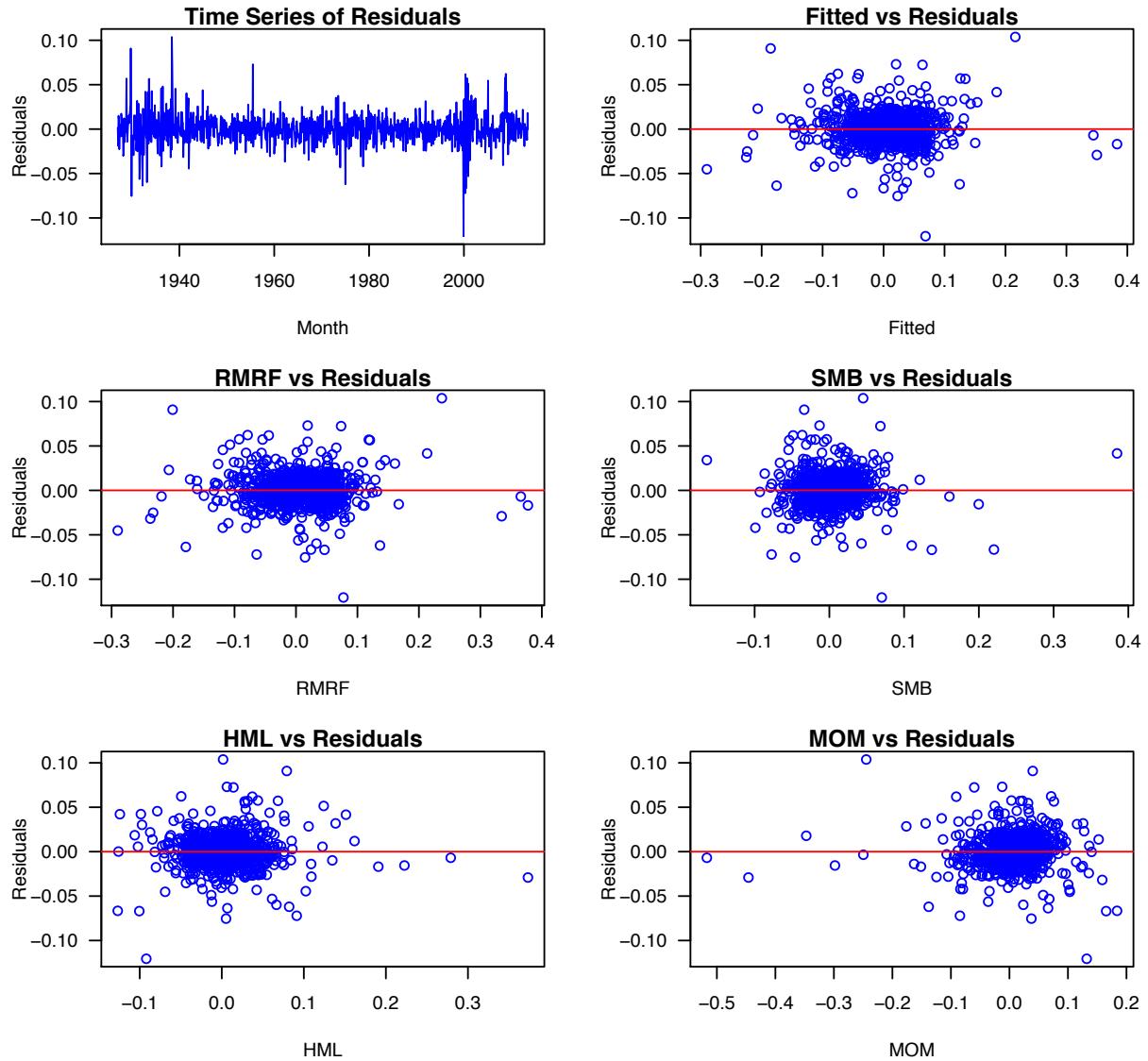
A.13 (AEW) Portfolio 7

(AEW) Residual Plots: Portfolio 7



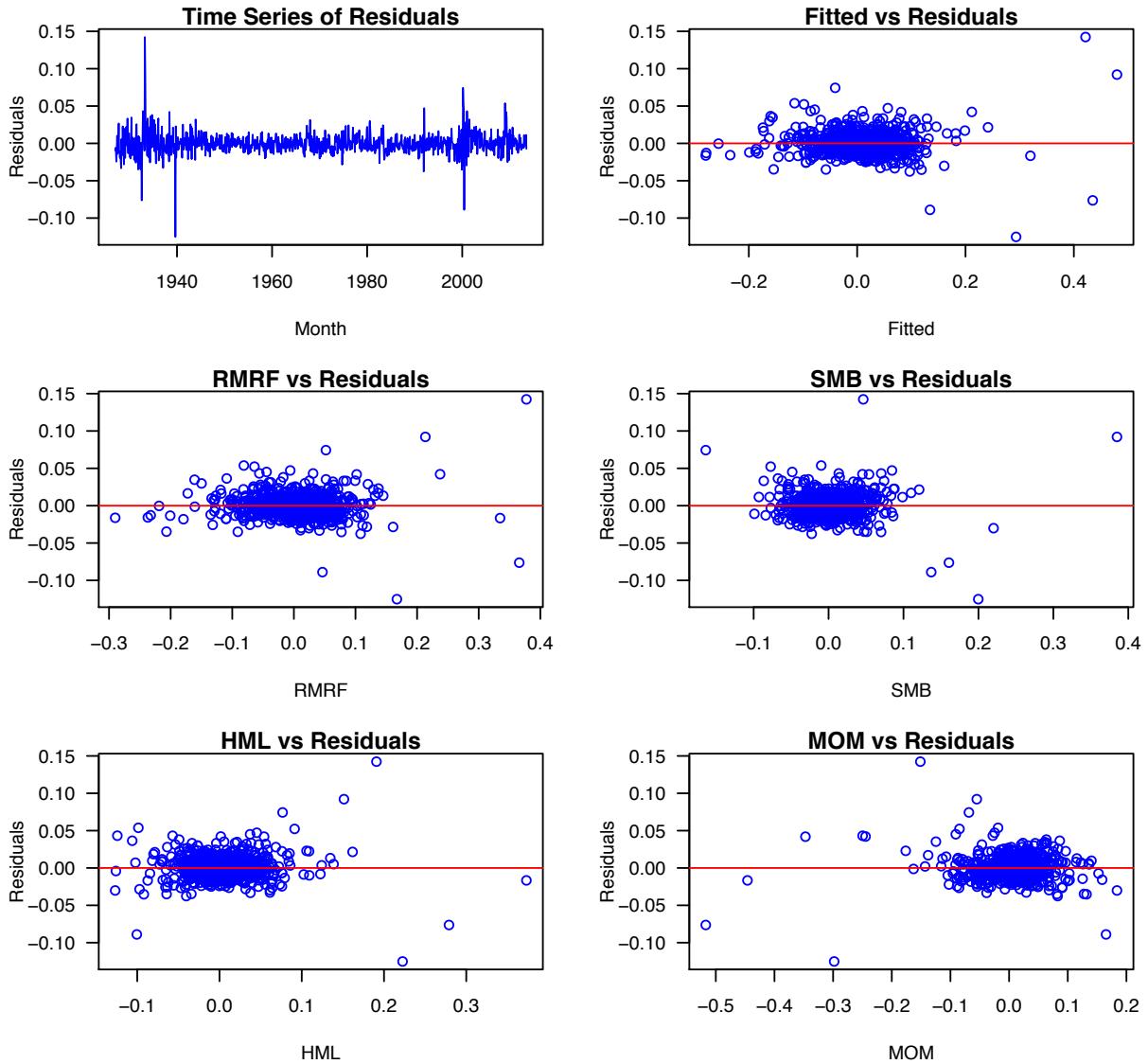
A.14 (AVW) Portfolio 7

(AVW) Residual Plots: Portfolio 7

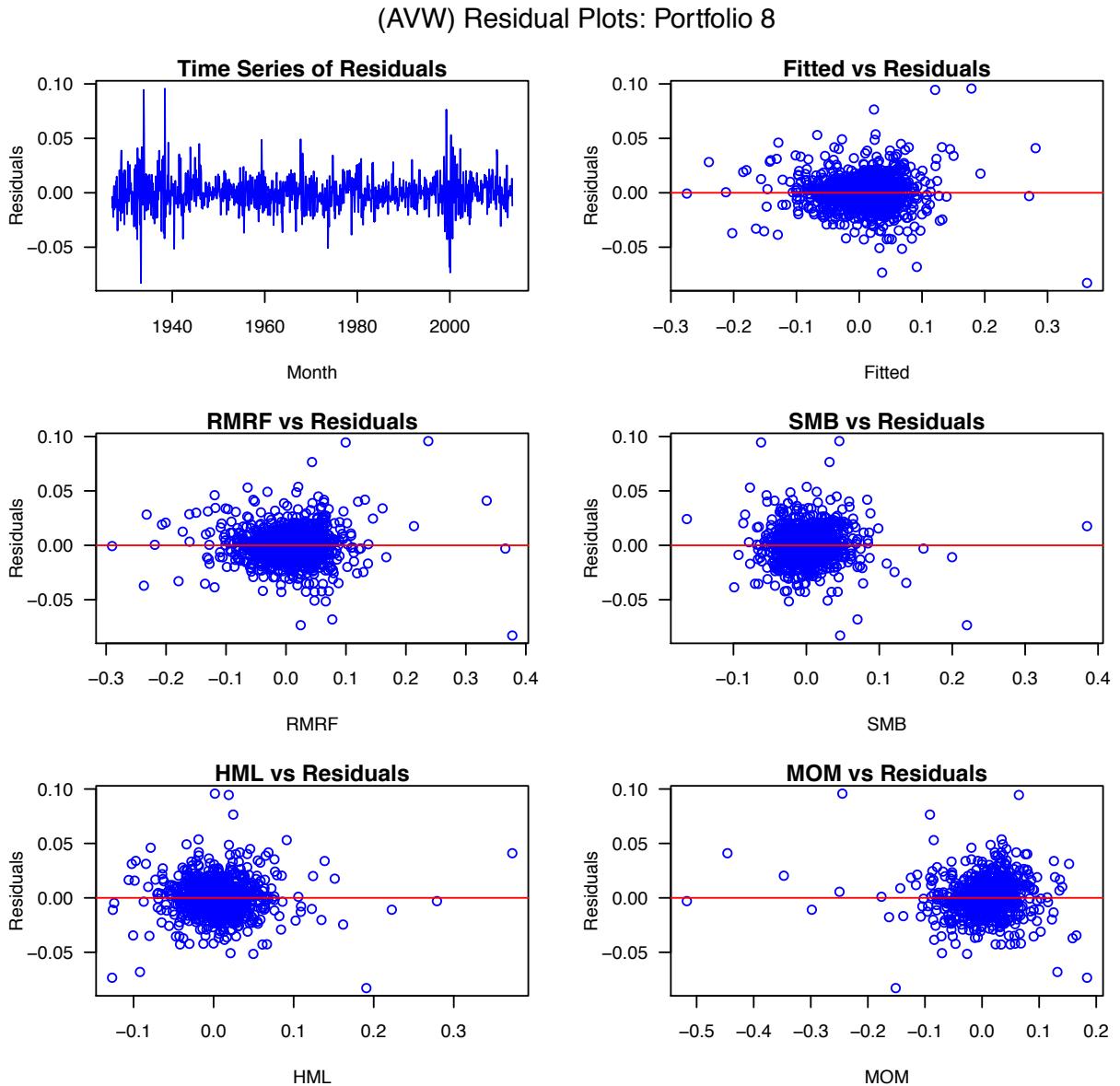


A.15 (AEW) Portfolio 8

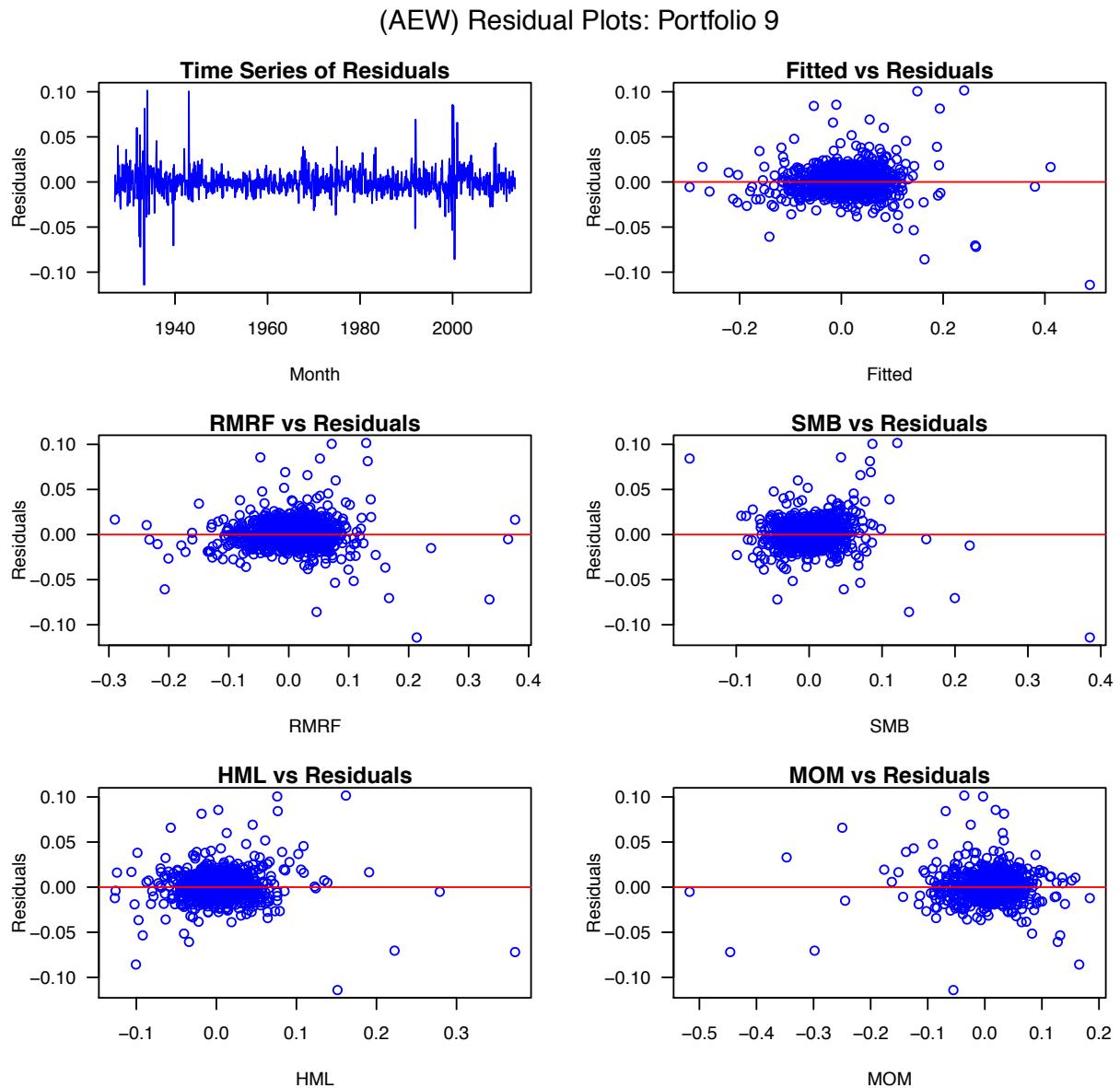
(AEW) Residual Plots: Portfolio 8



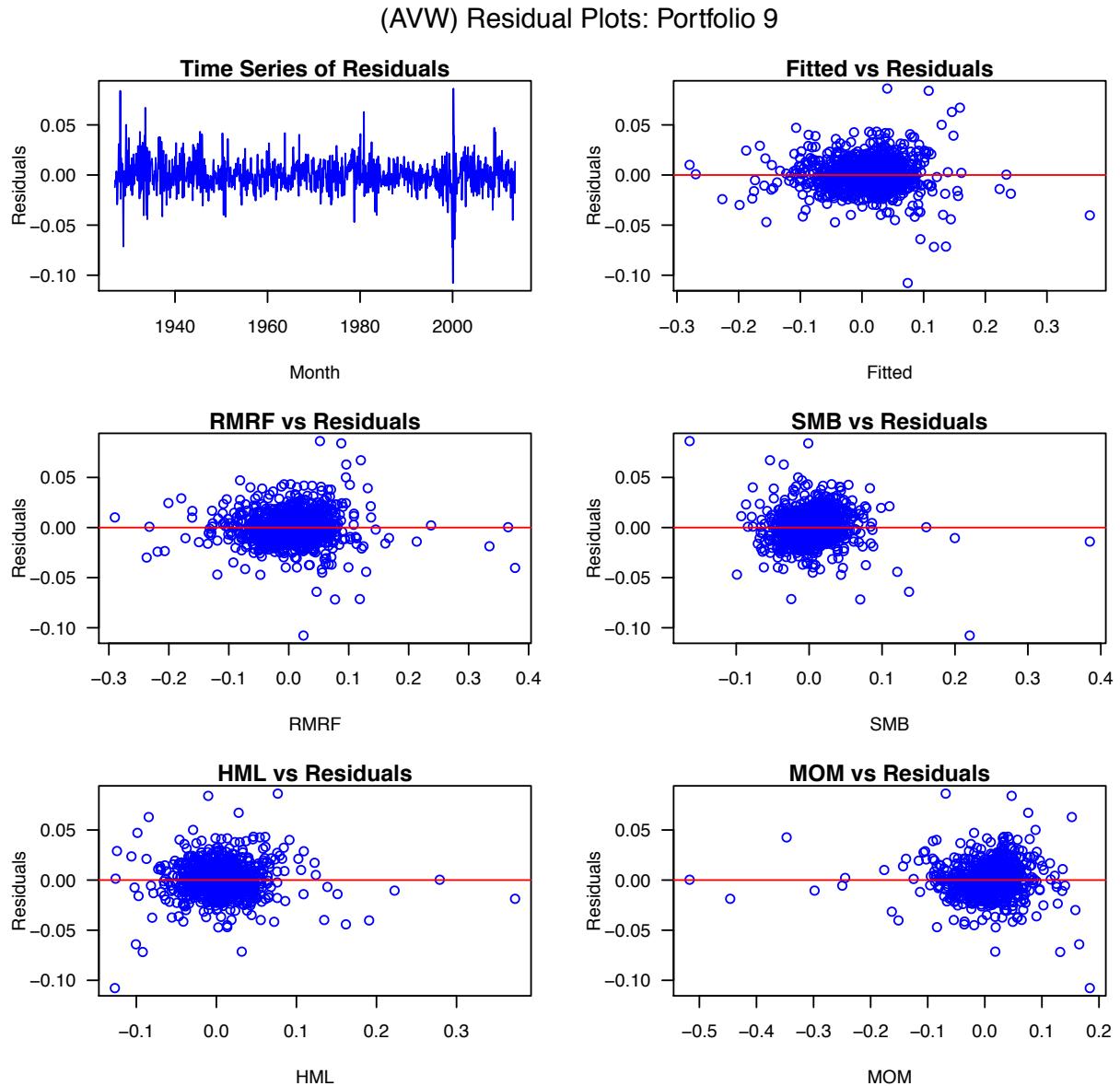
A.16 (AVW) Portfolio 8



A.17 (AEW) Portfolio 9

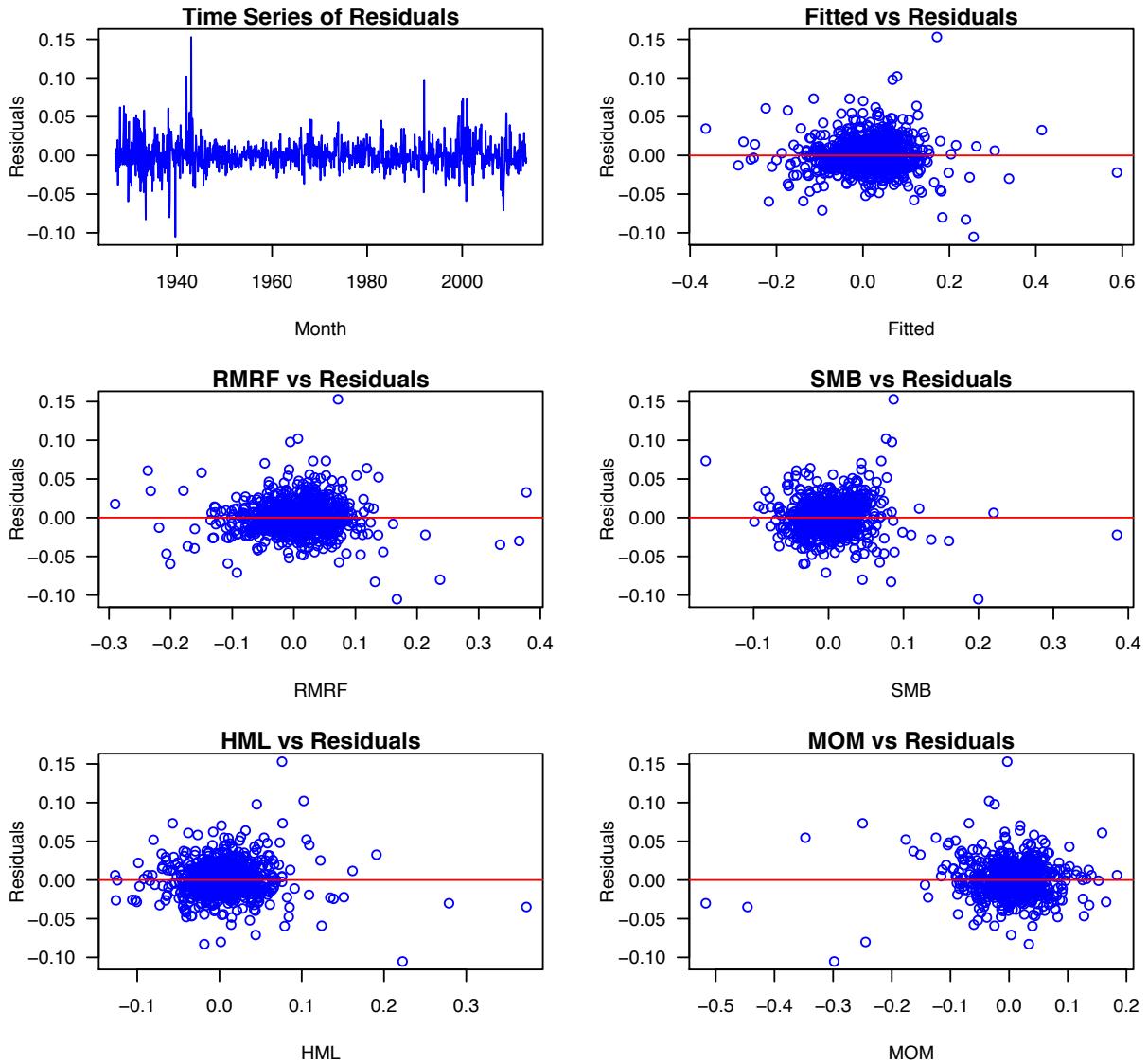


A.18 (AVW) Portfolio 9

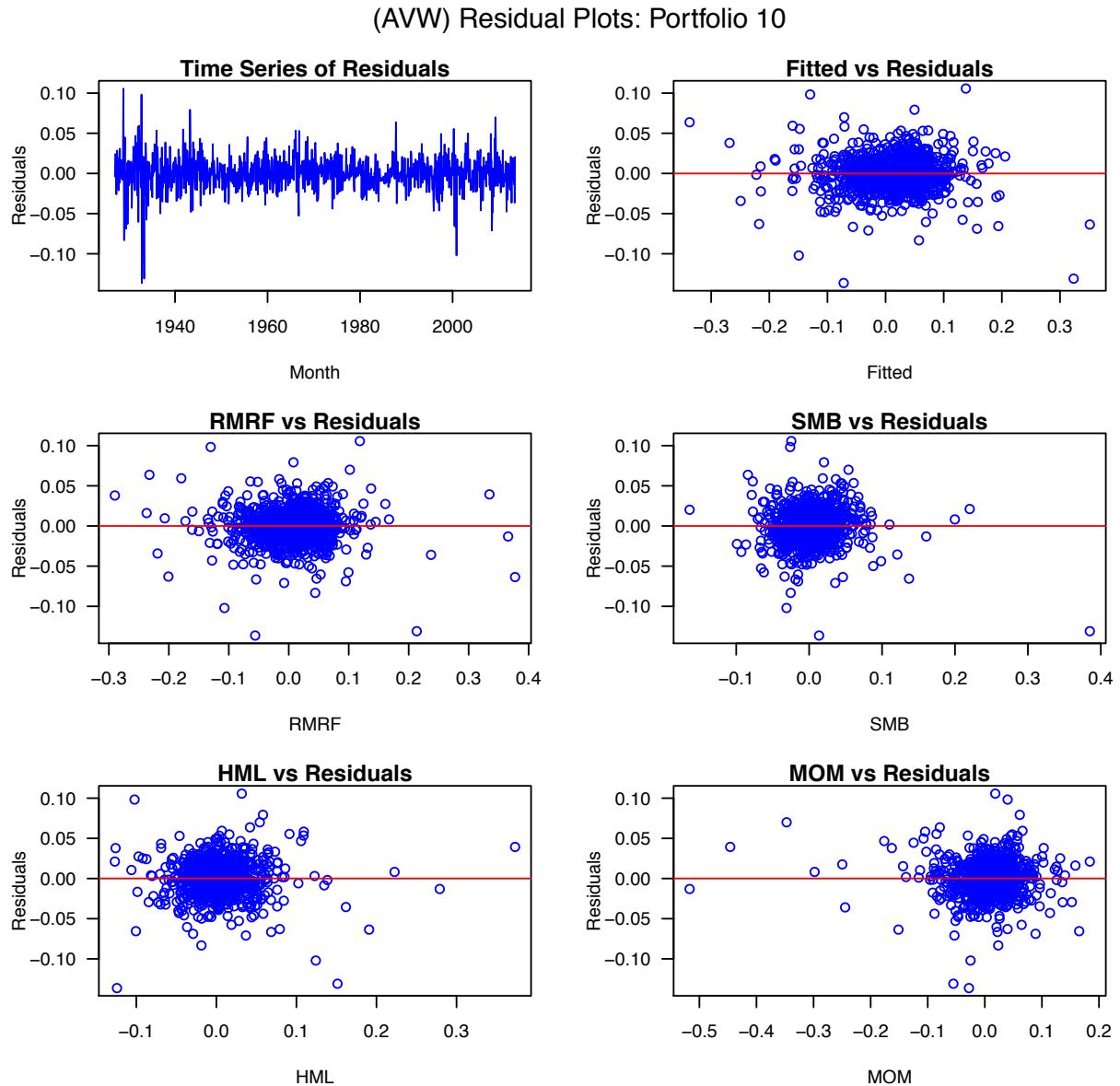


A.19 (AEW) Portfolio 10

(AEW) Residual Plots: Portfolio 10



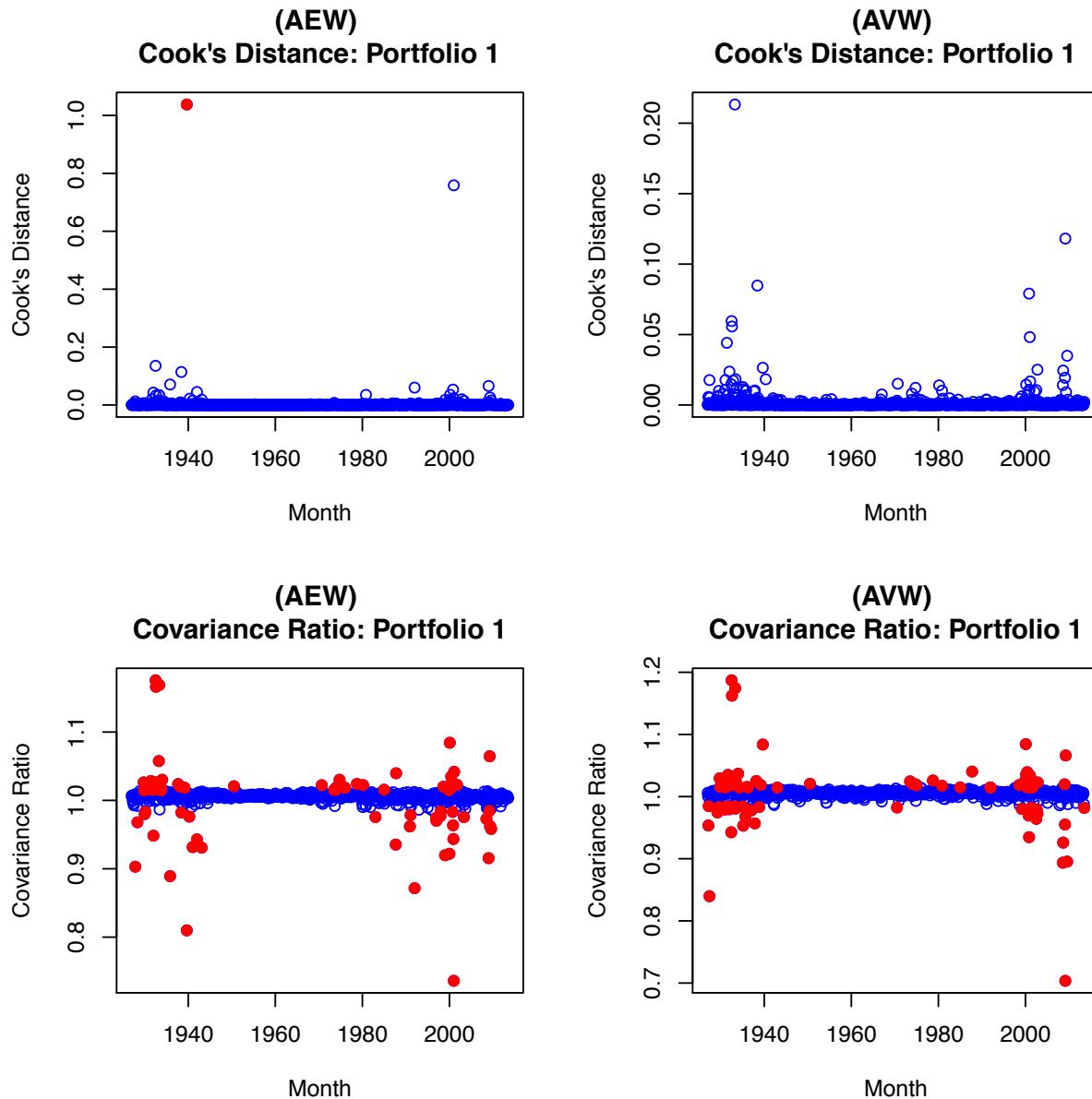
A.20 (AVW) Portfolio 10



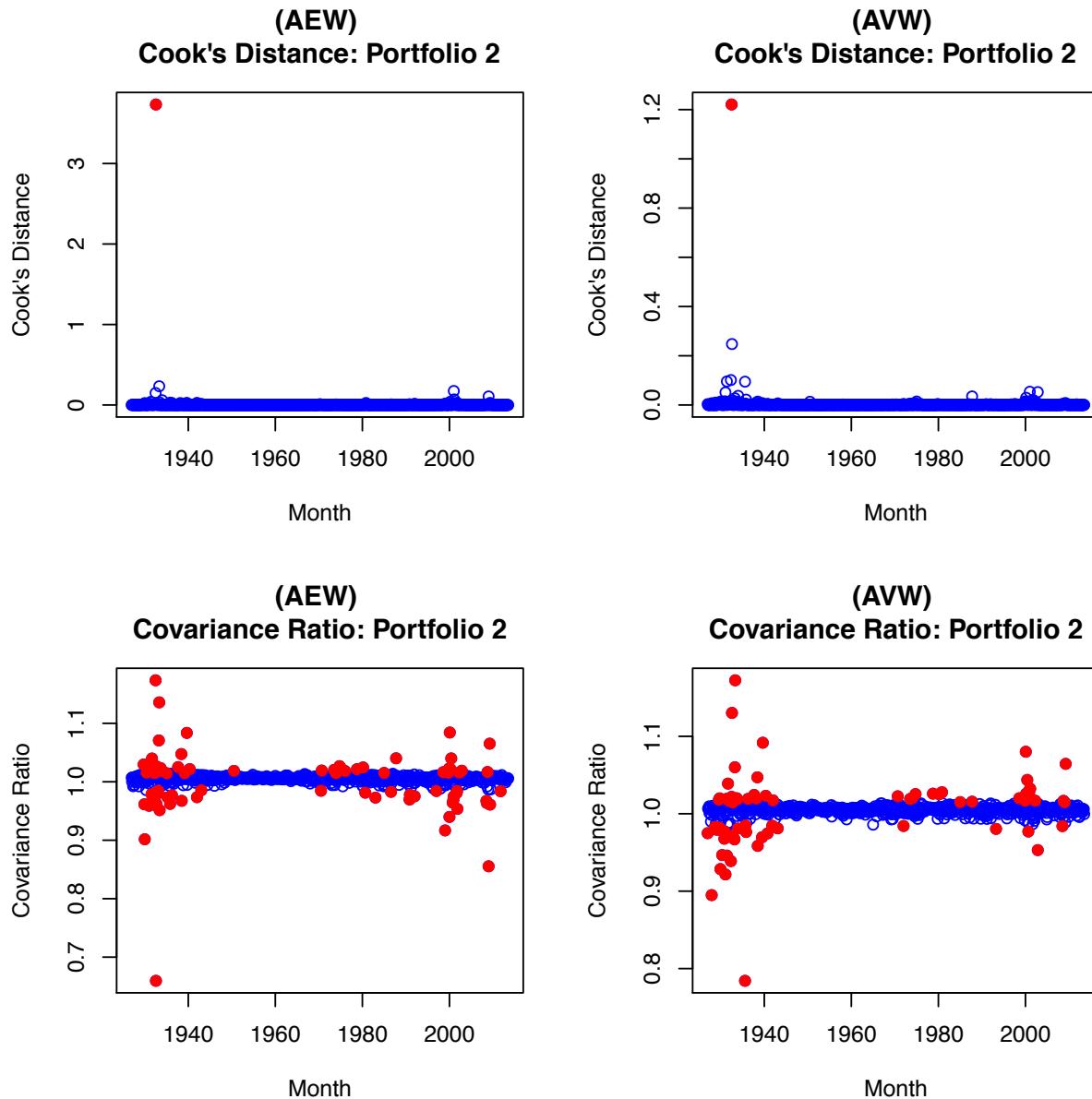
Appendix B

Plot of Cook's Distance and Covariance ratio

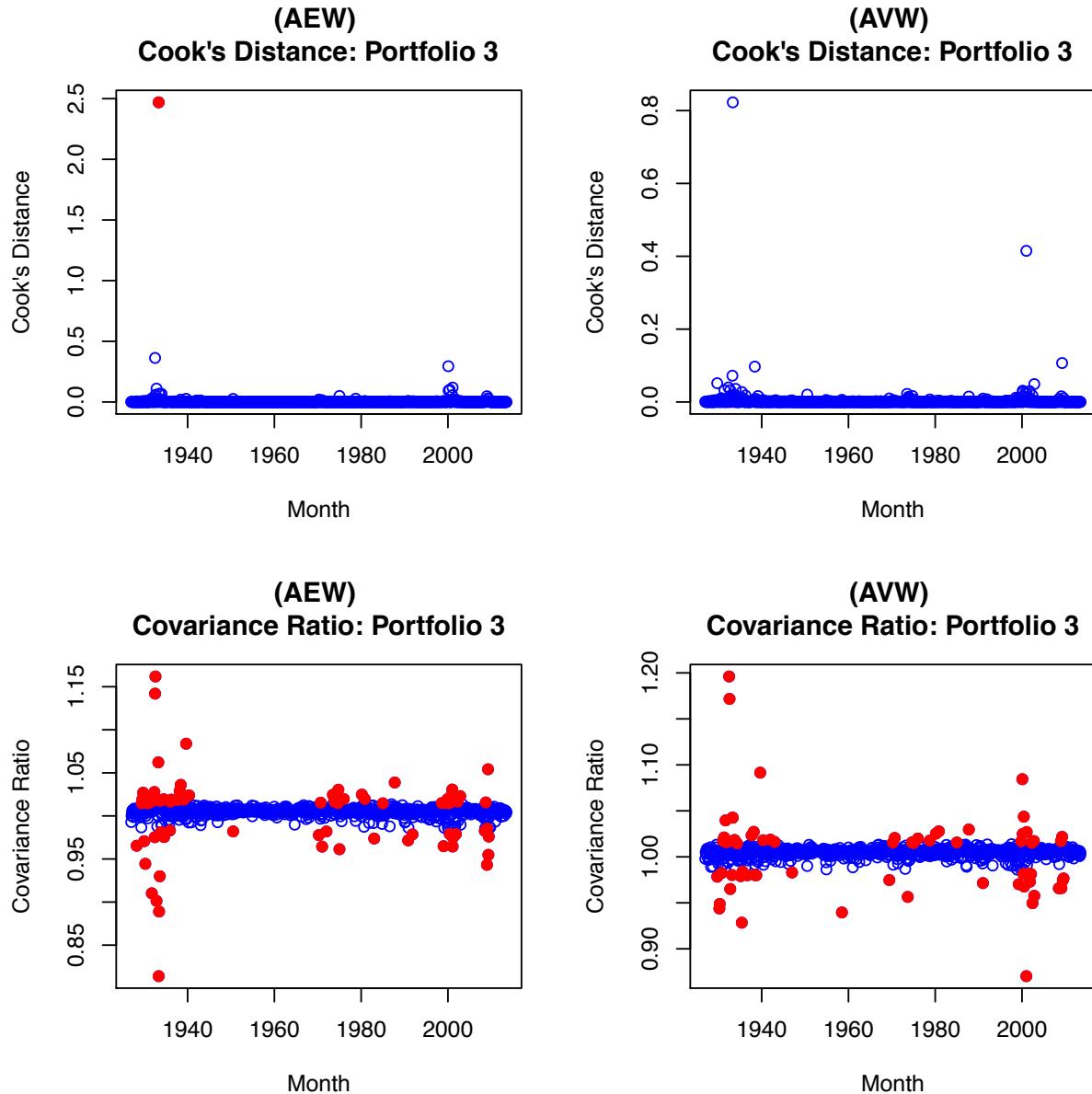
B.1 Portfolio 1



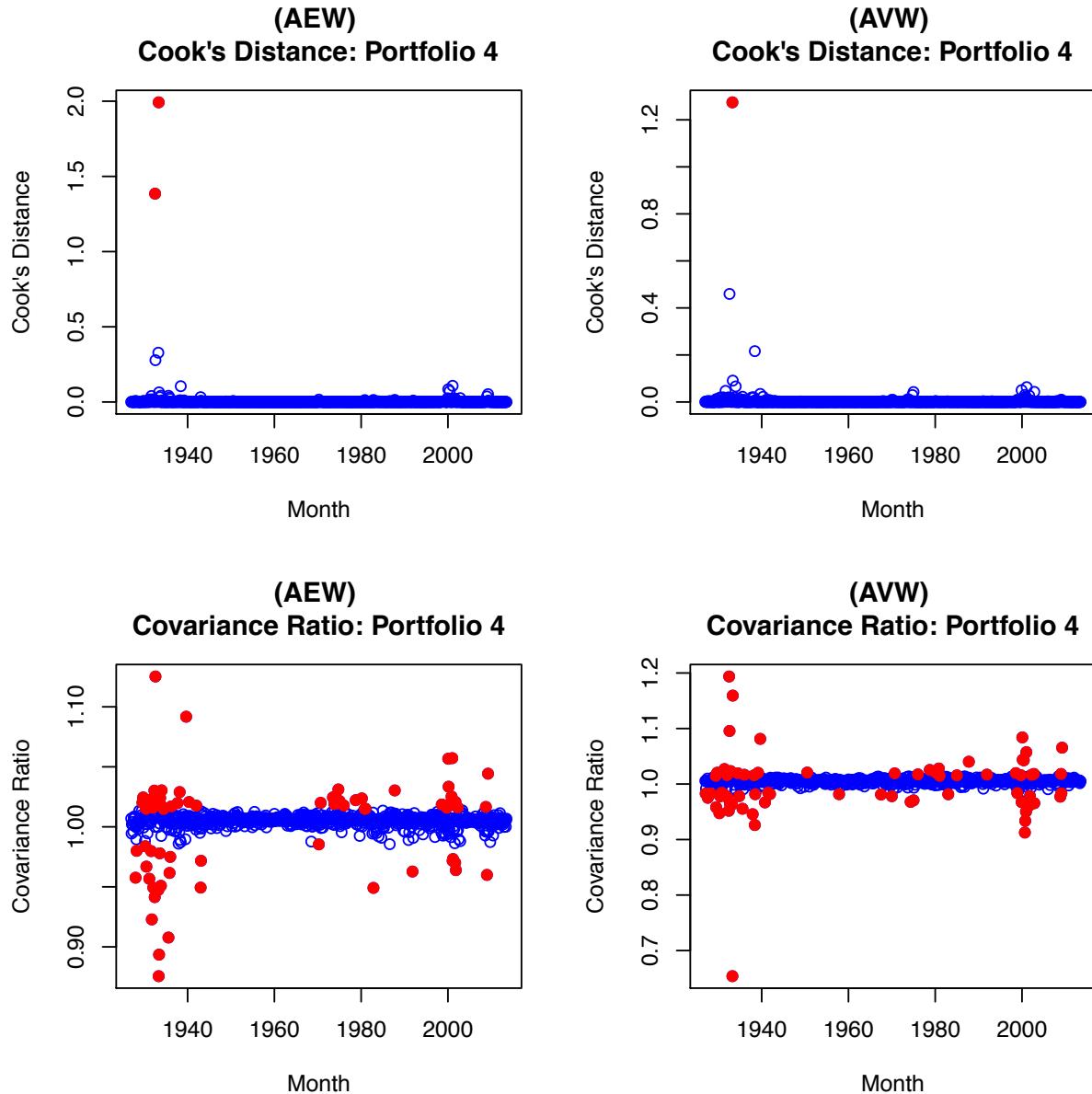
B.2 Portfolio 2



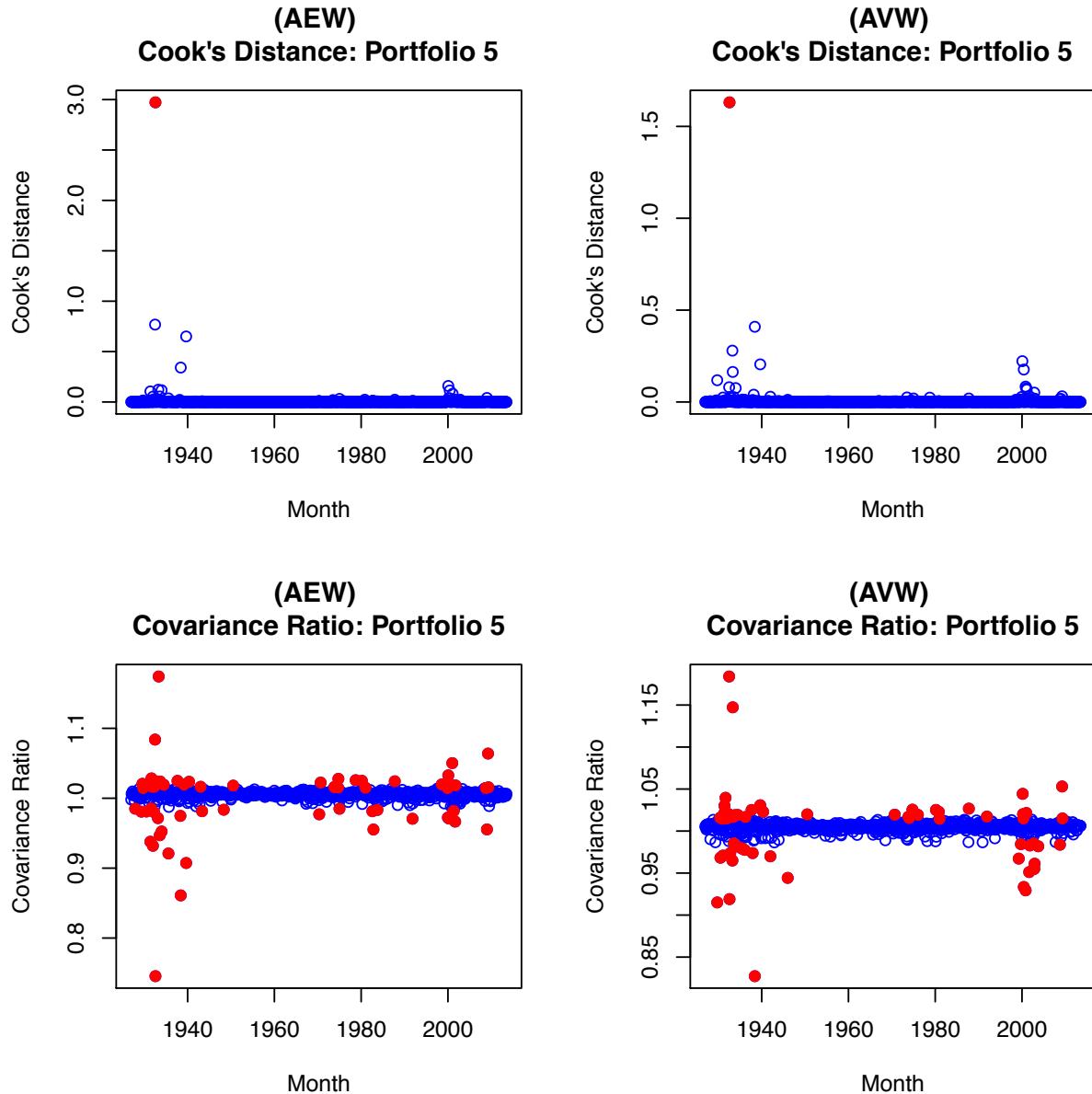
B.3 Portfolio 3



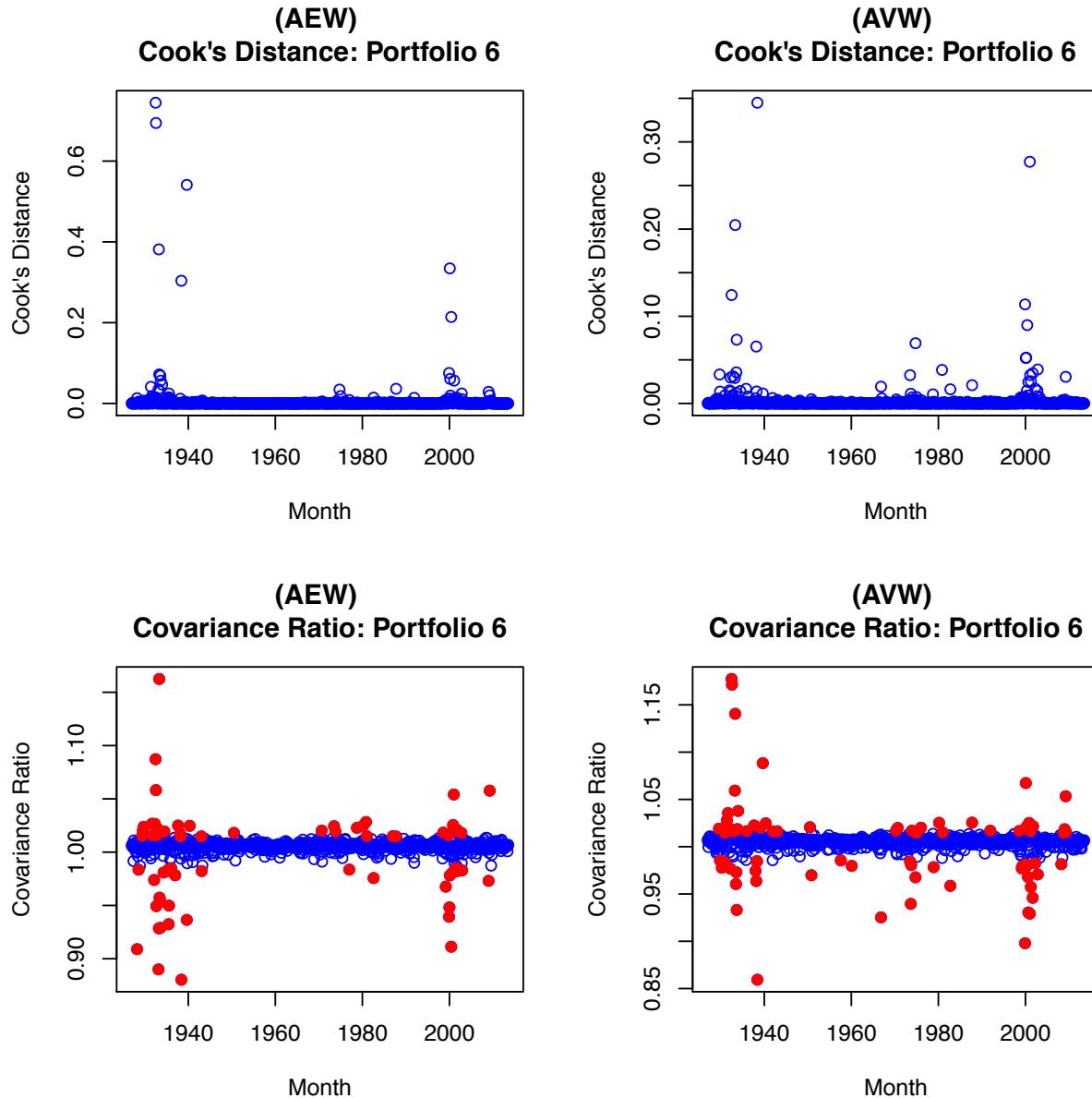
B.4 Portfolio 4



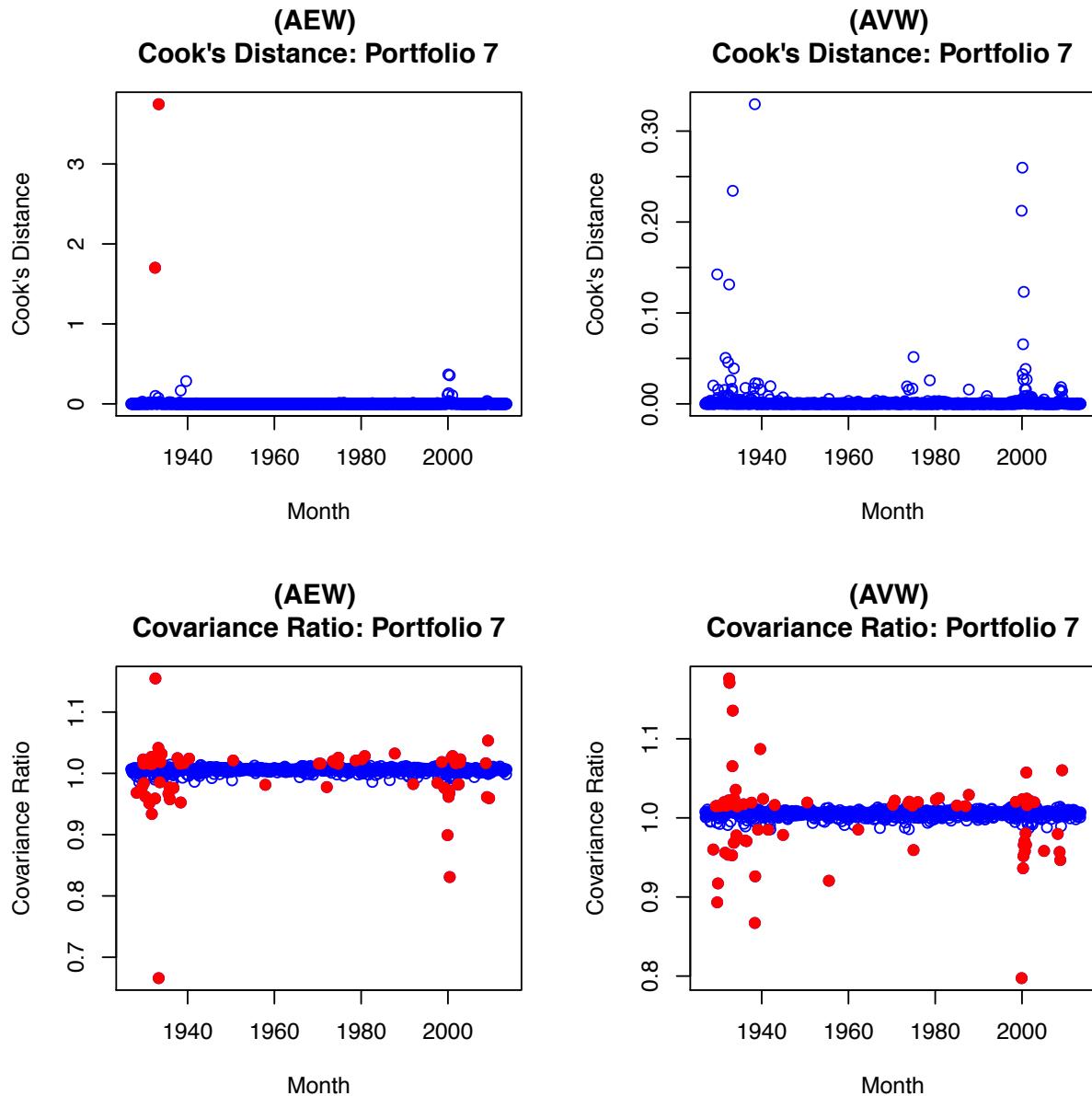
B.5 Portfolio 5



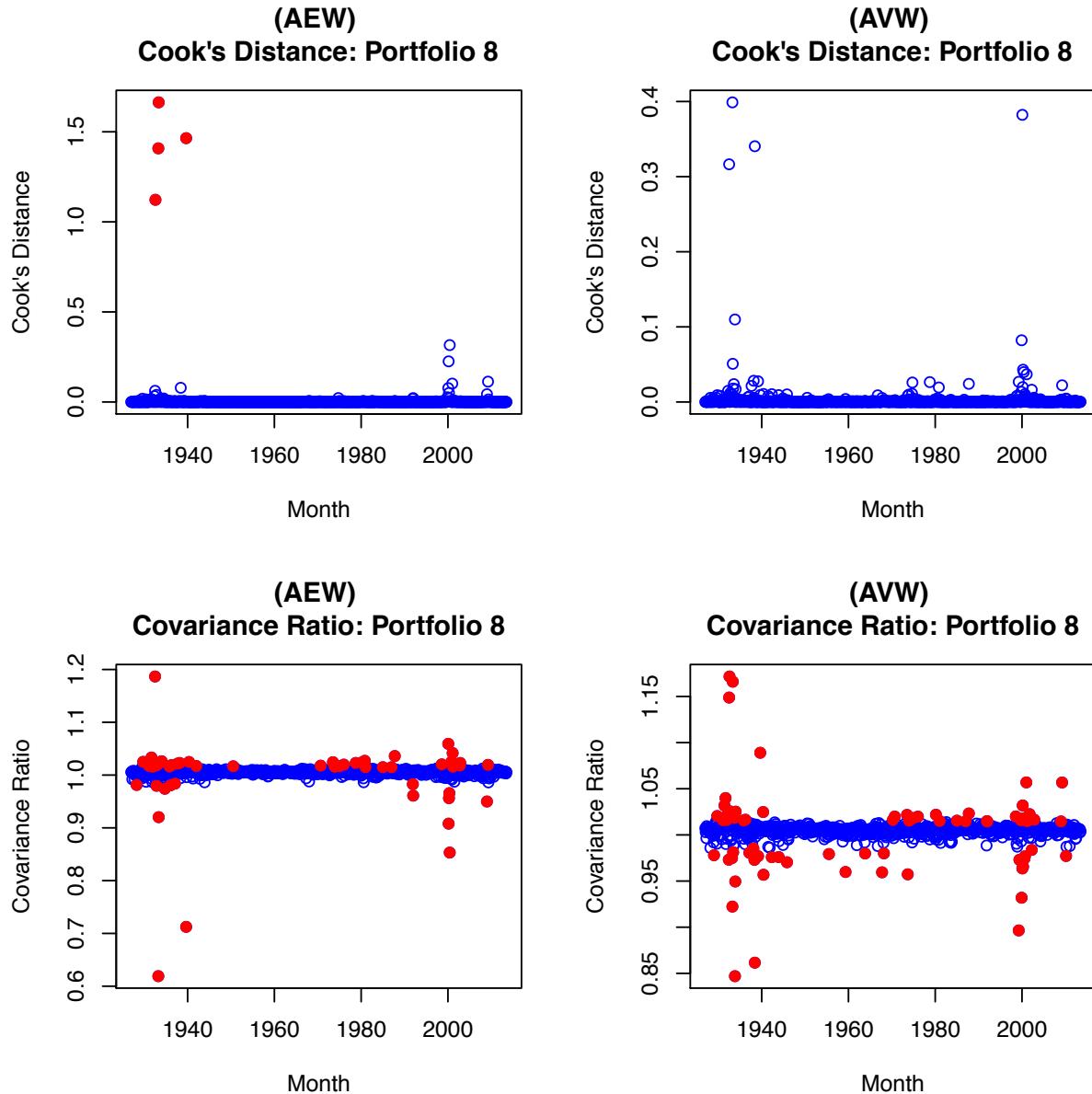
B.6 Portfolio 6



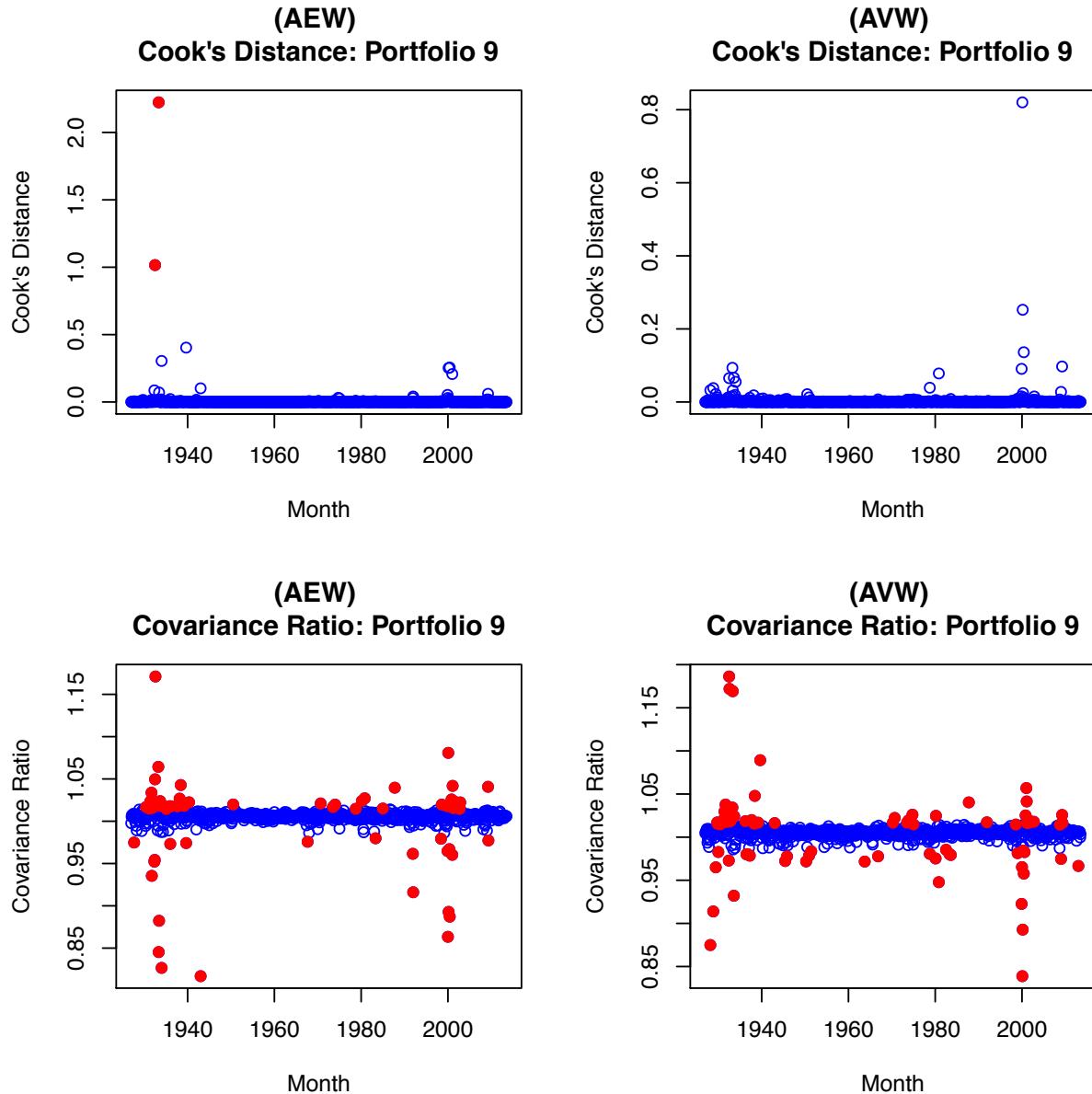
B.7 Portfolio 7



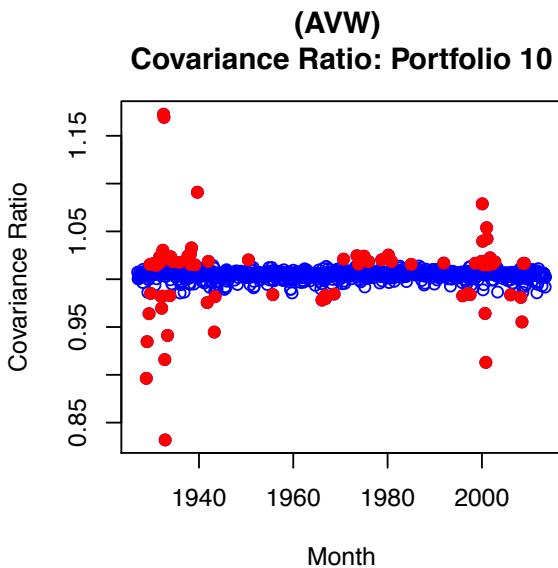
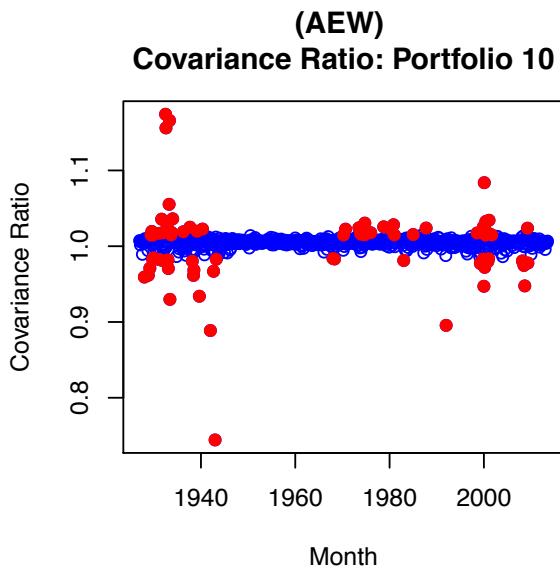
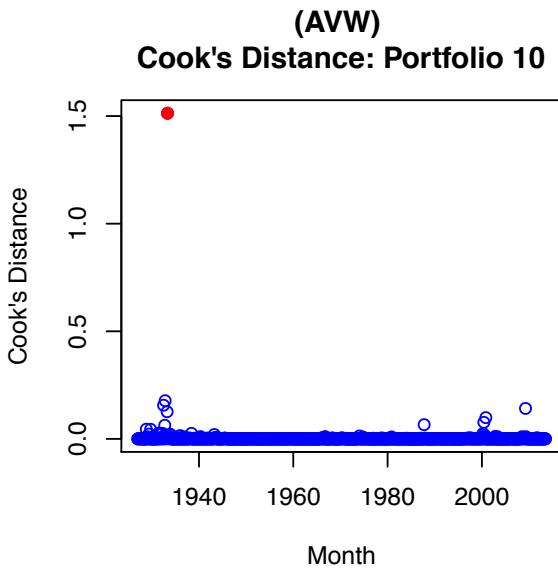
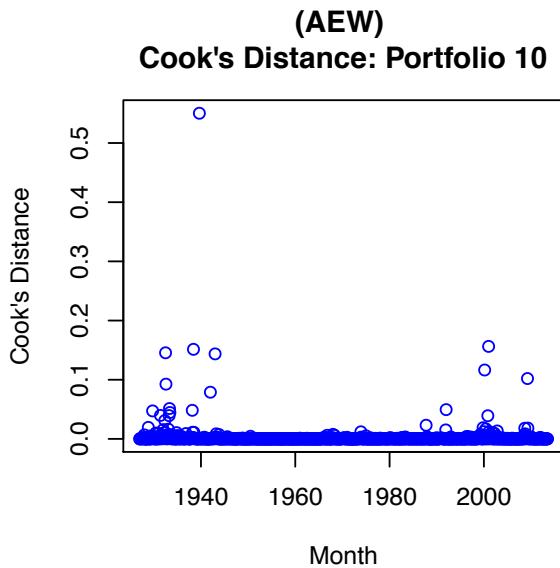
B.8 Portfolio 8



B.9 Portfolio 9



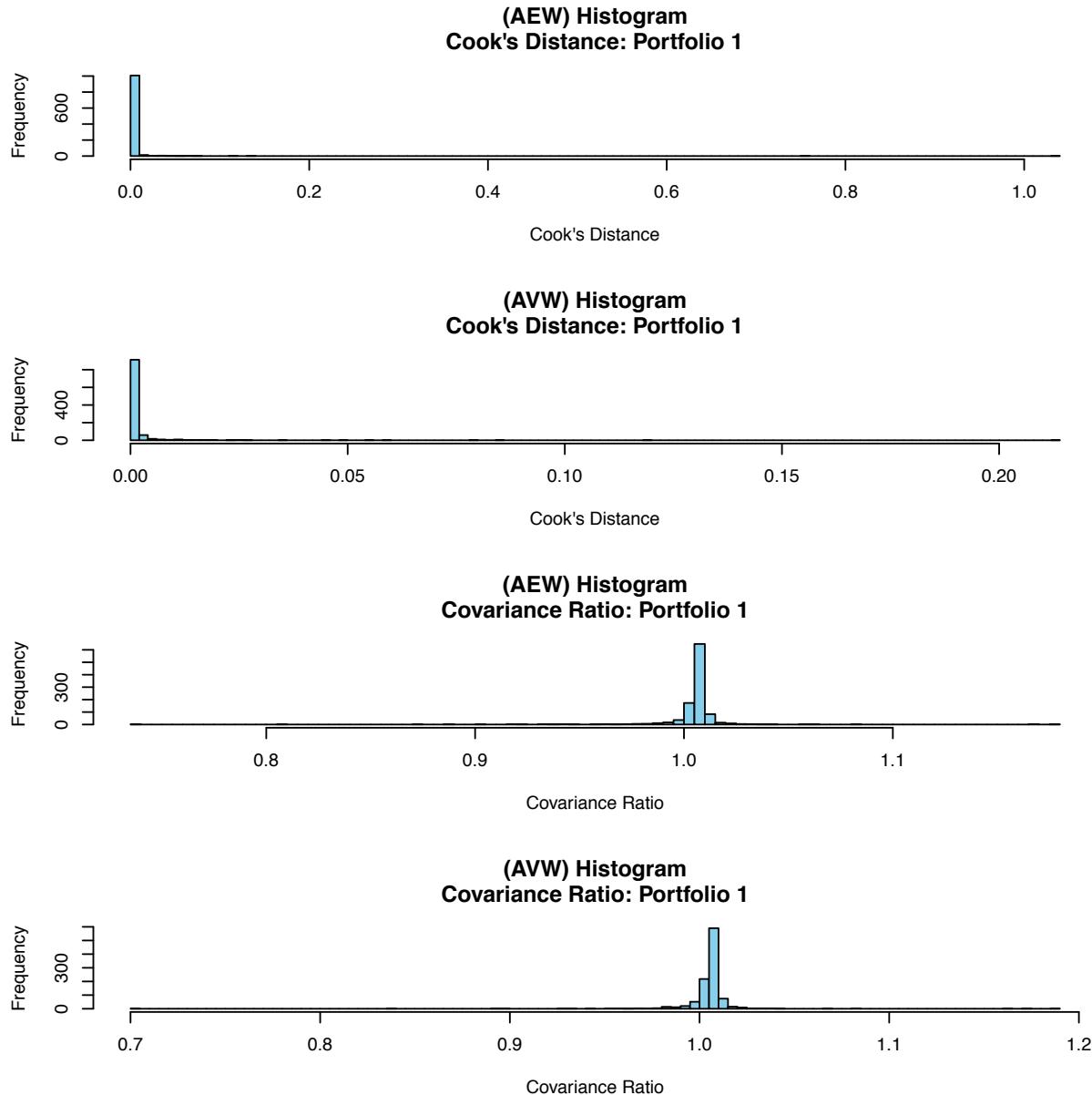
B.10 Portfolio 10



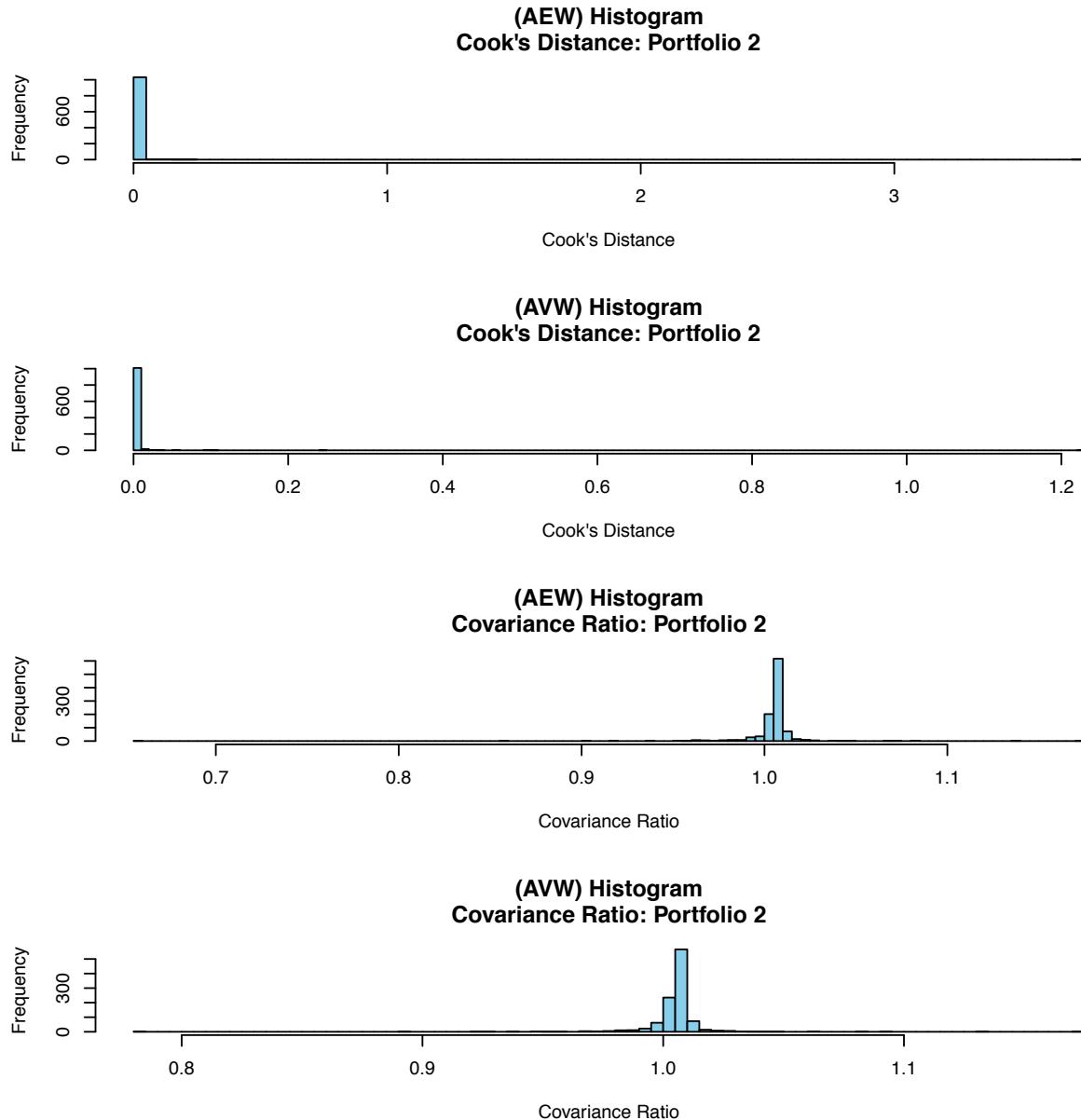
Appendix C

Histogram of Cook's Distance and Covariance ratio

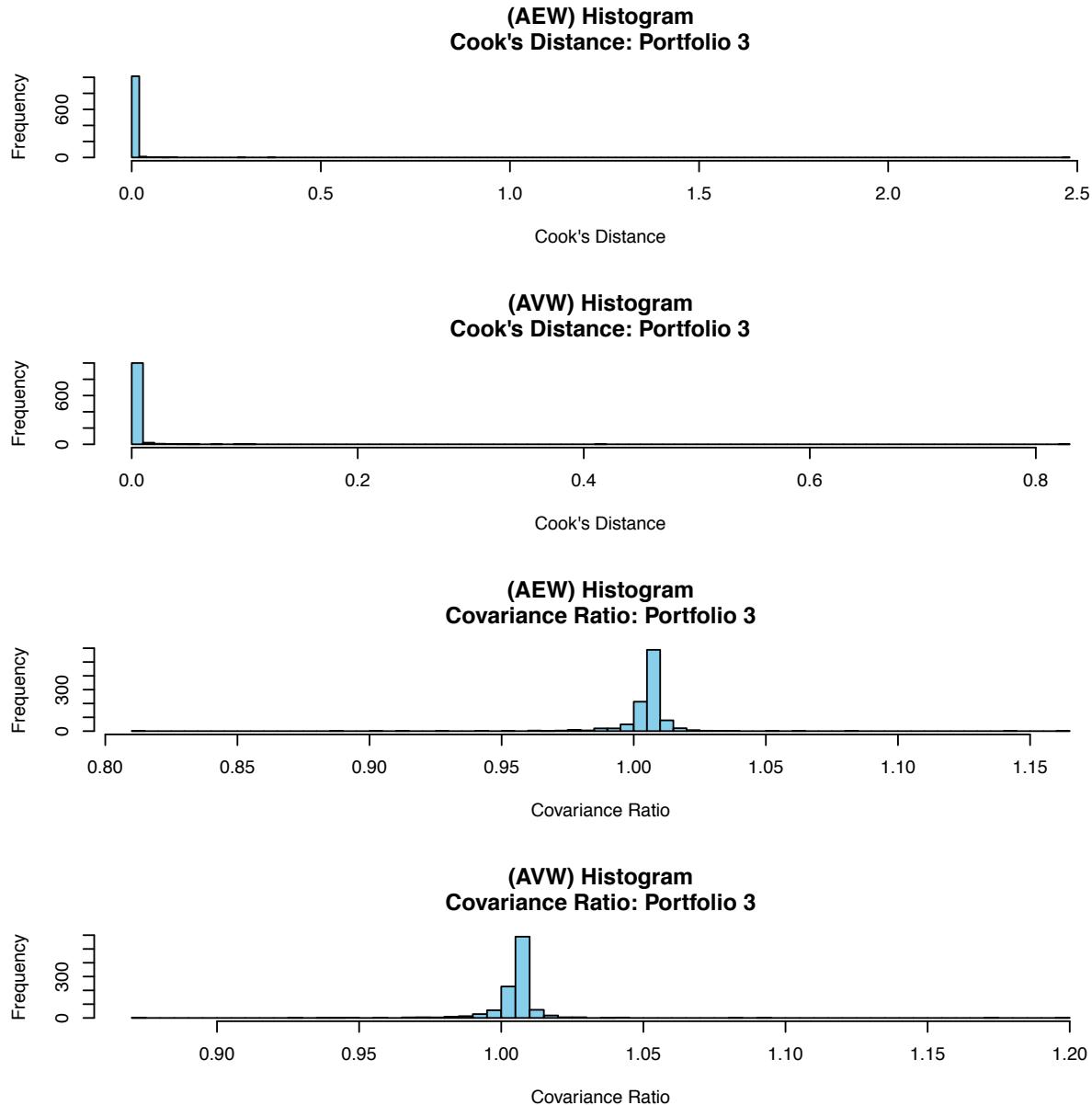
C.1 Portfolio 1



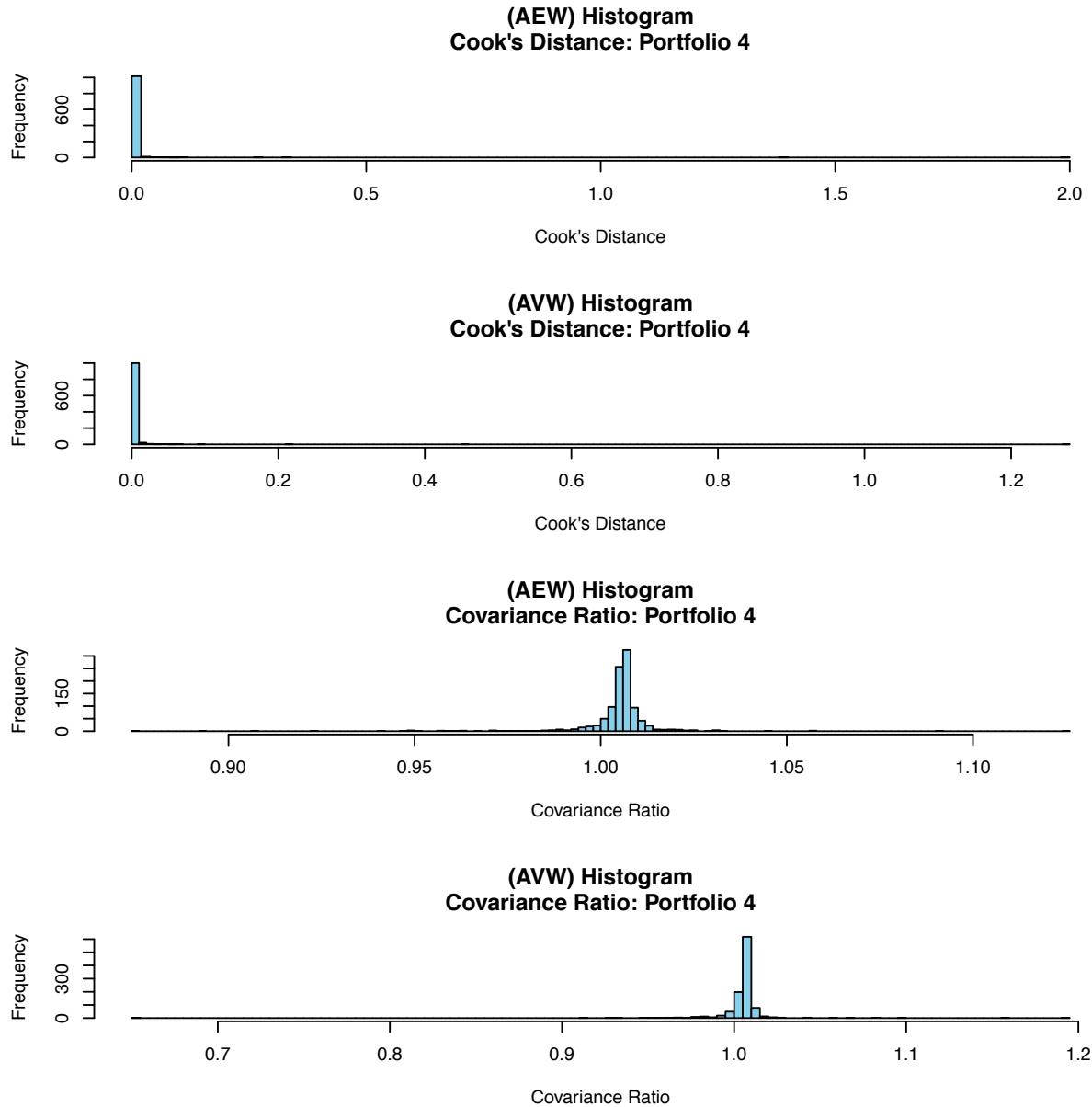
C.2 Portfolio 2



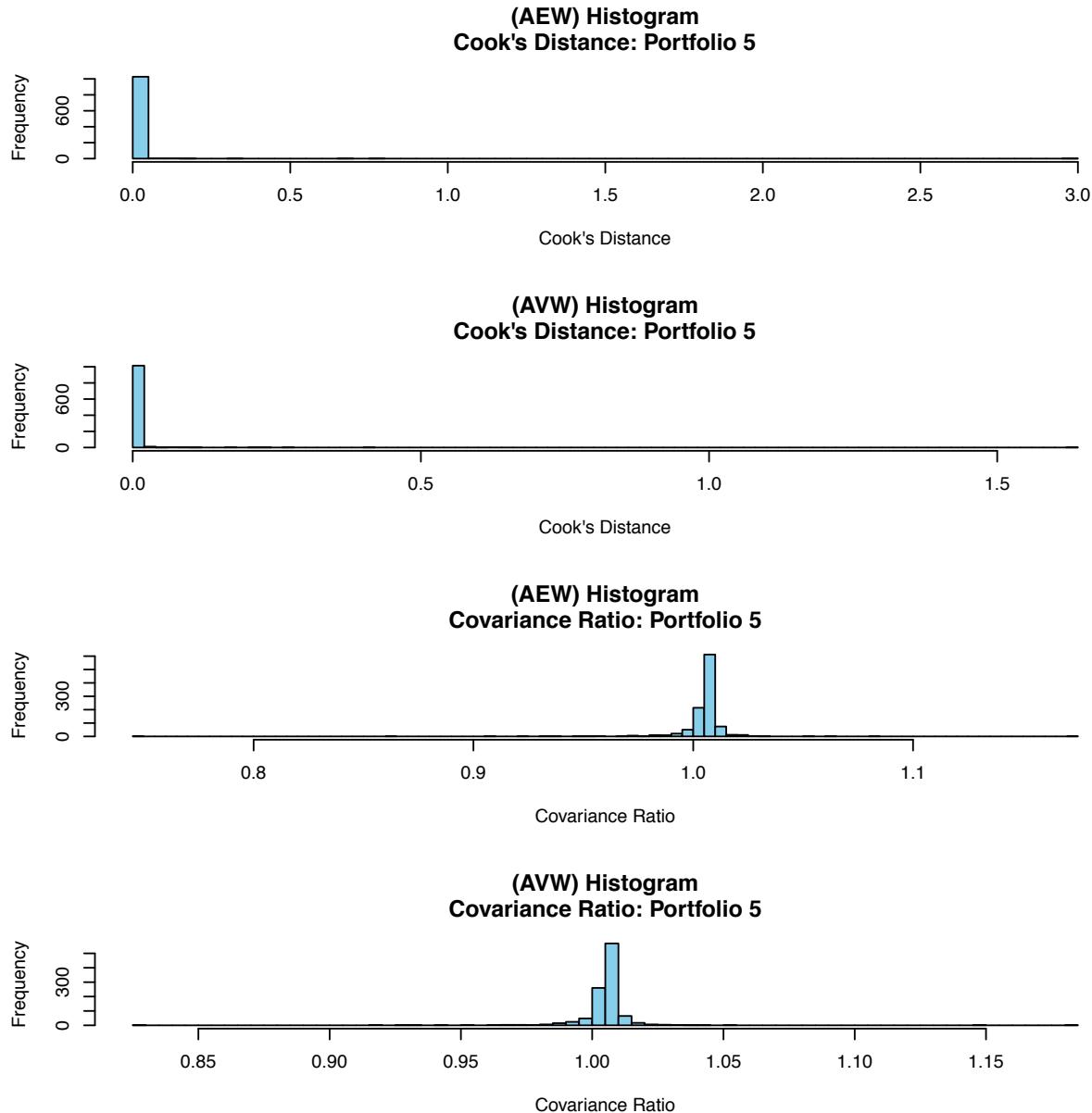
C.3 Portfolio 3



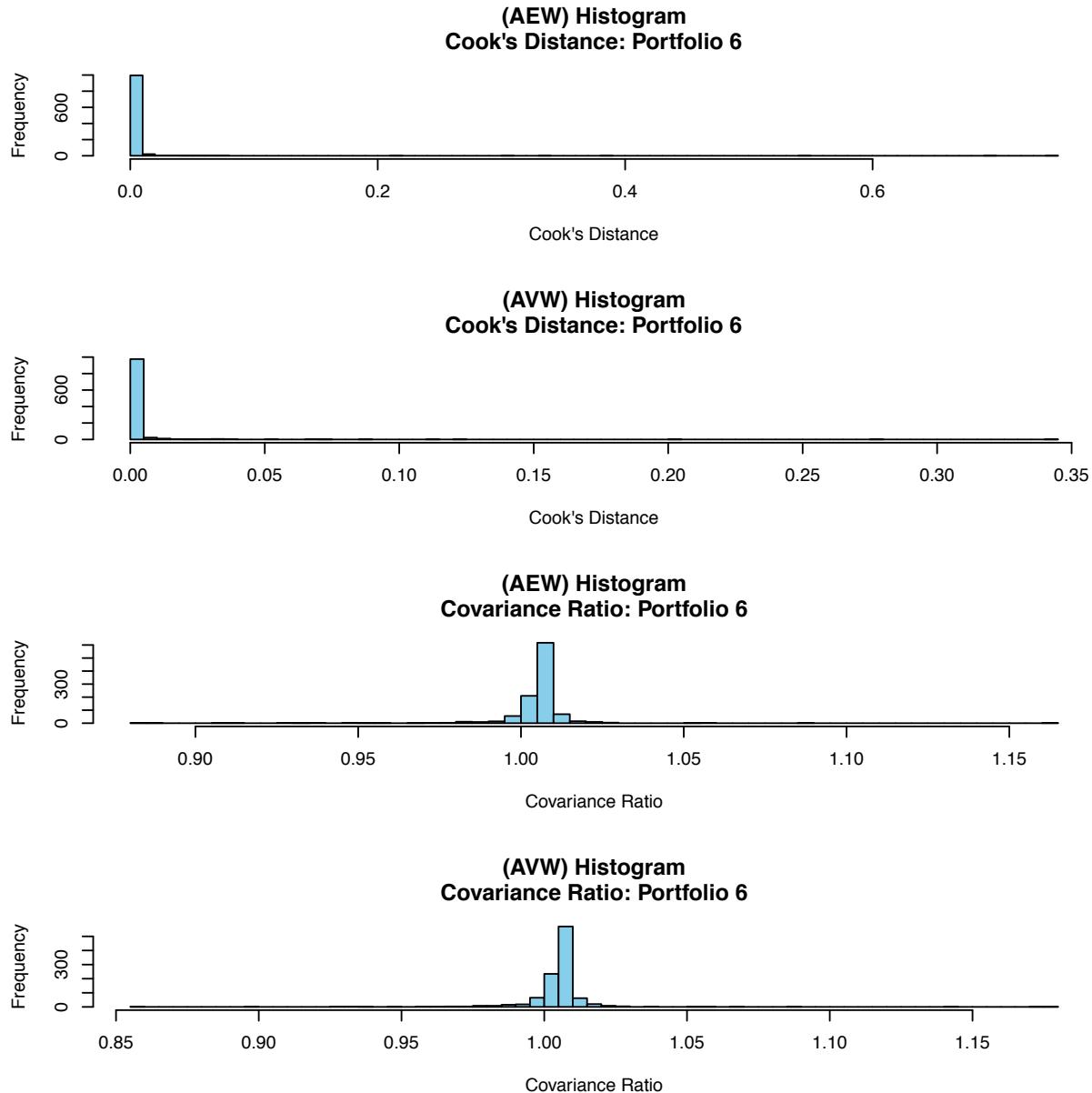
C.4 Portfolio 4



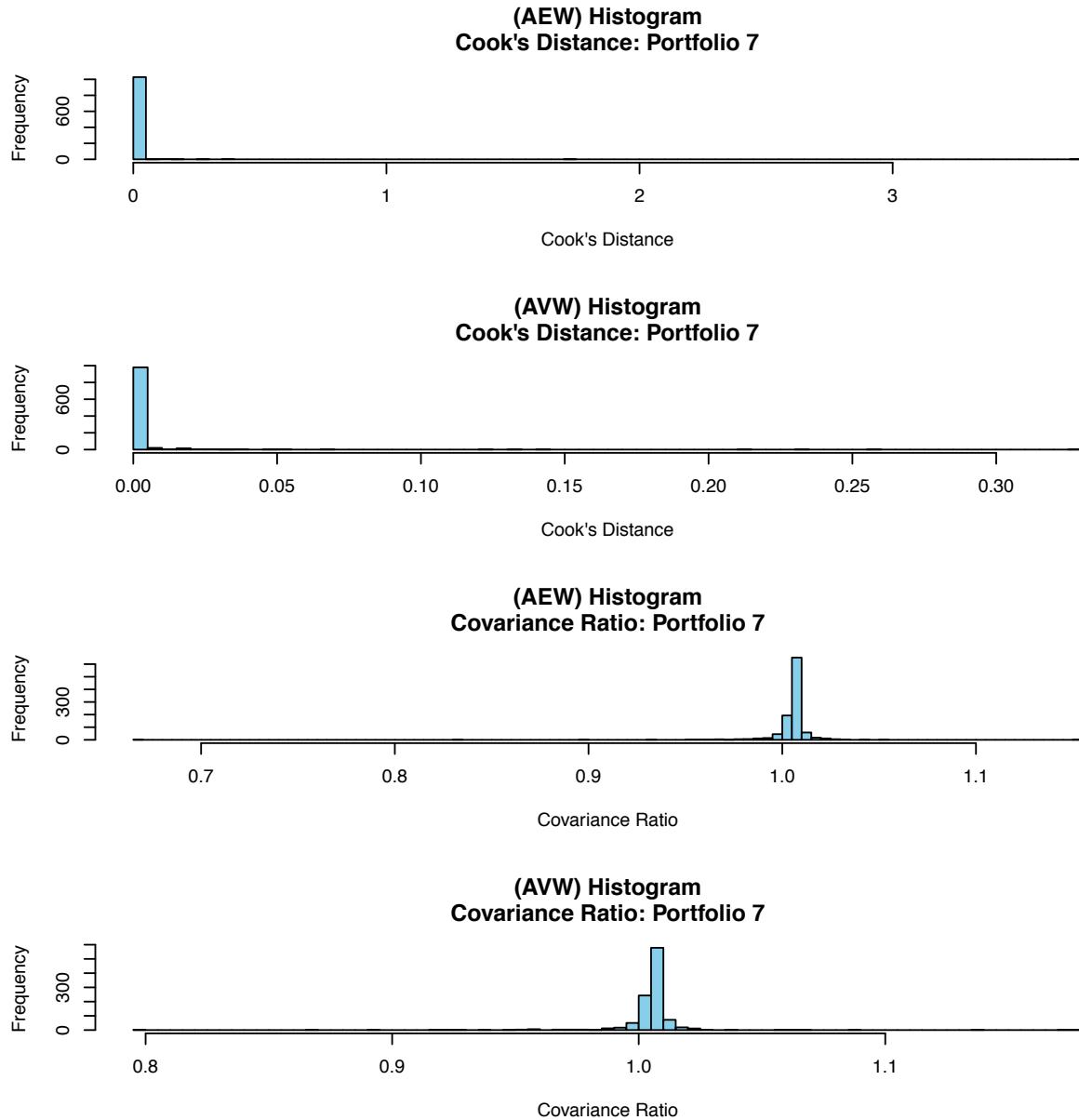
C.5 Portfolio 5



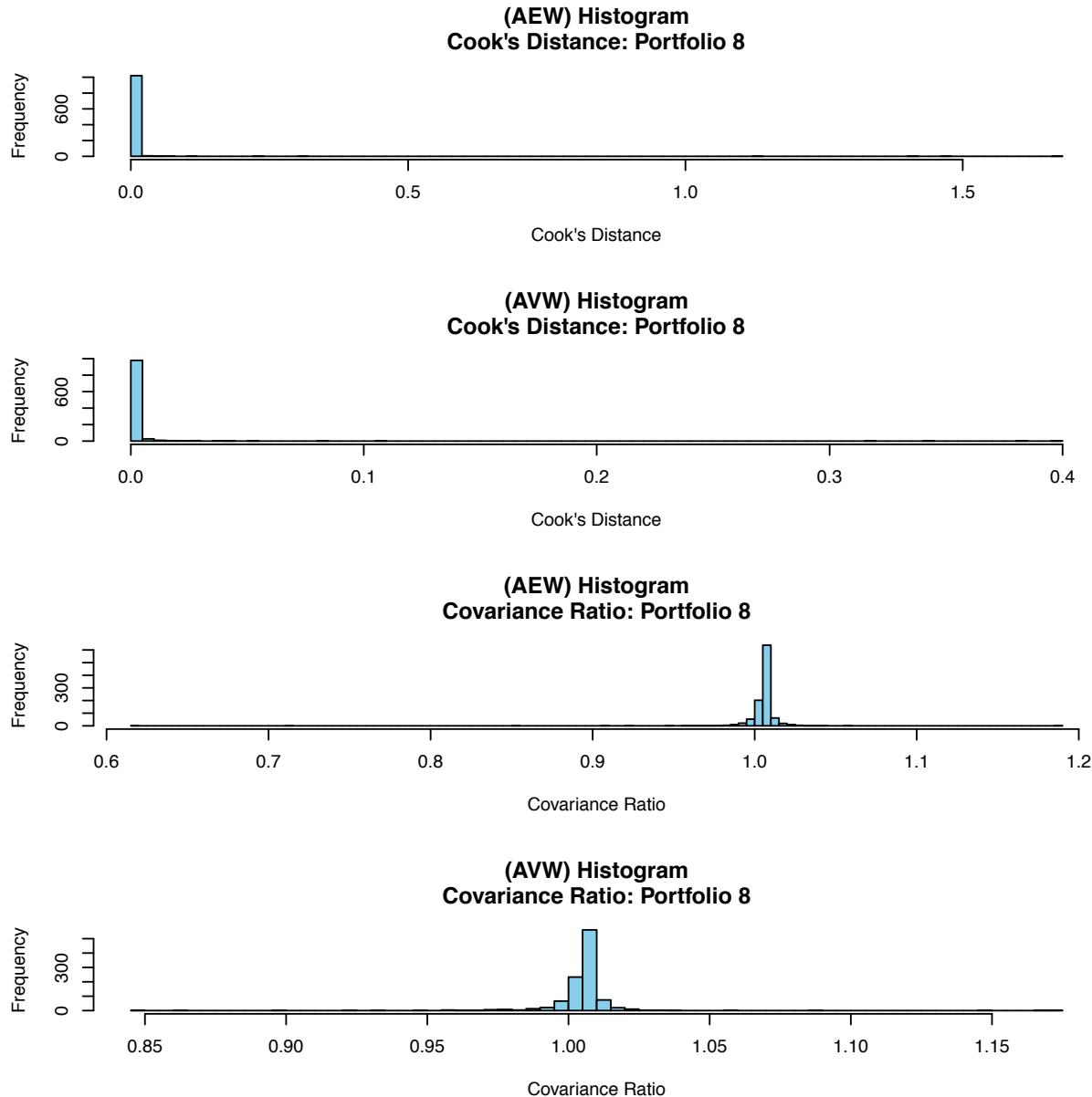
C.6 Portfolio 6



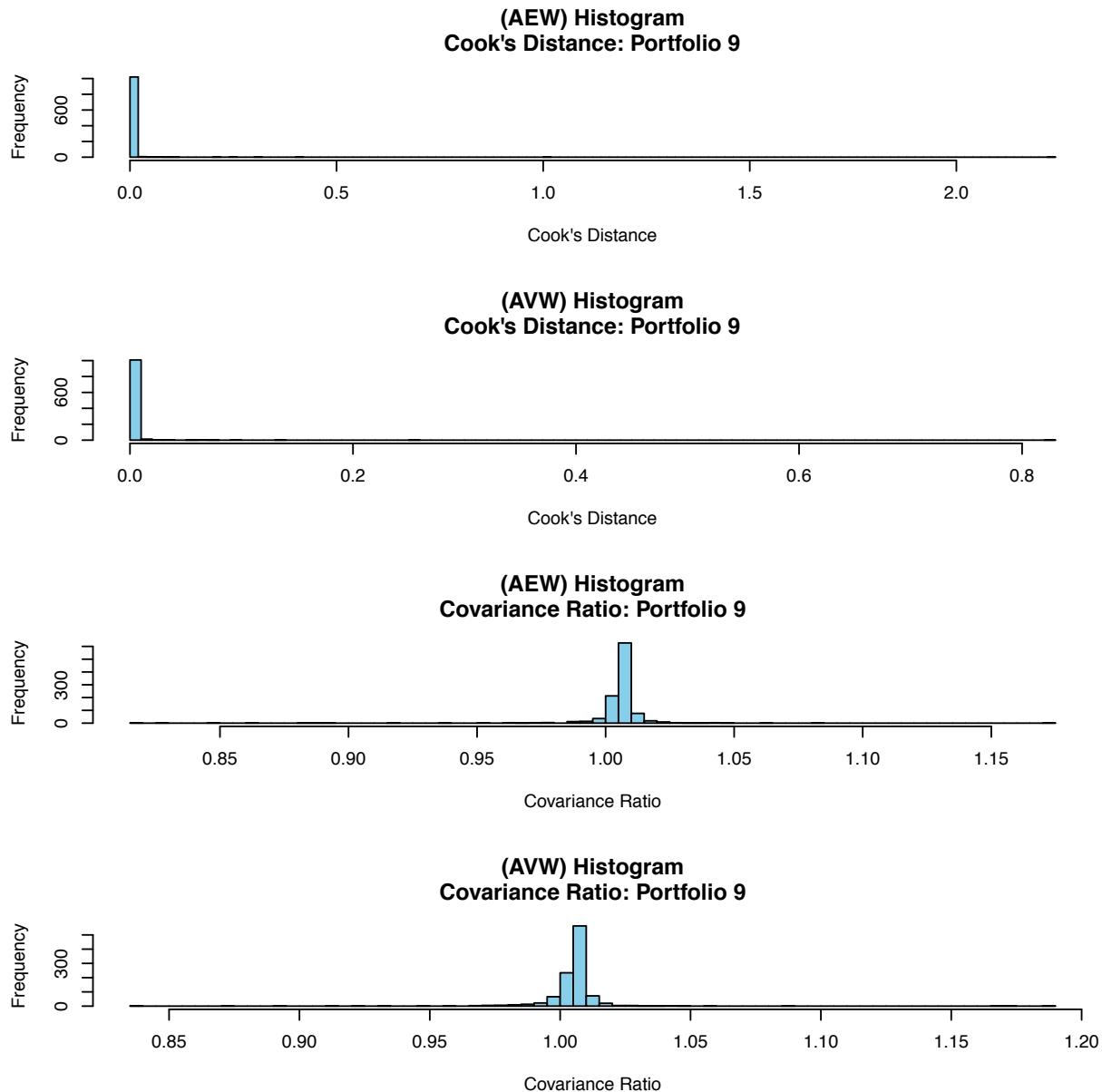
C.7 Portfolio 7



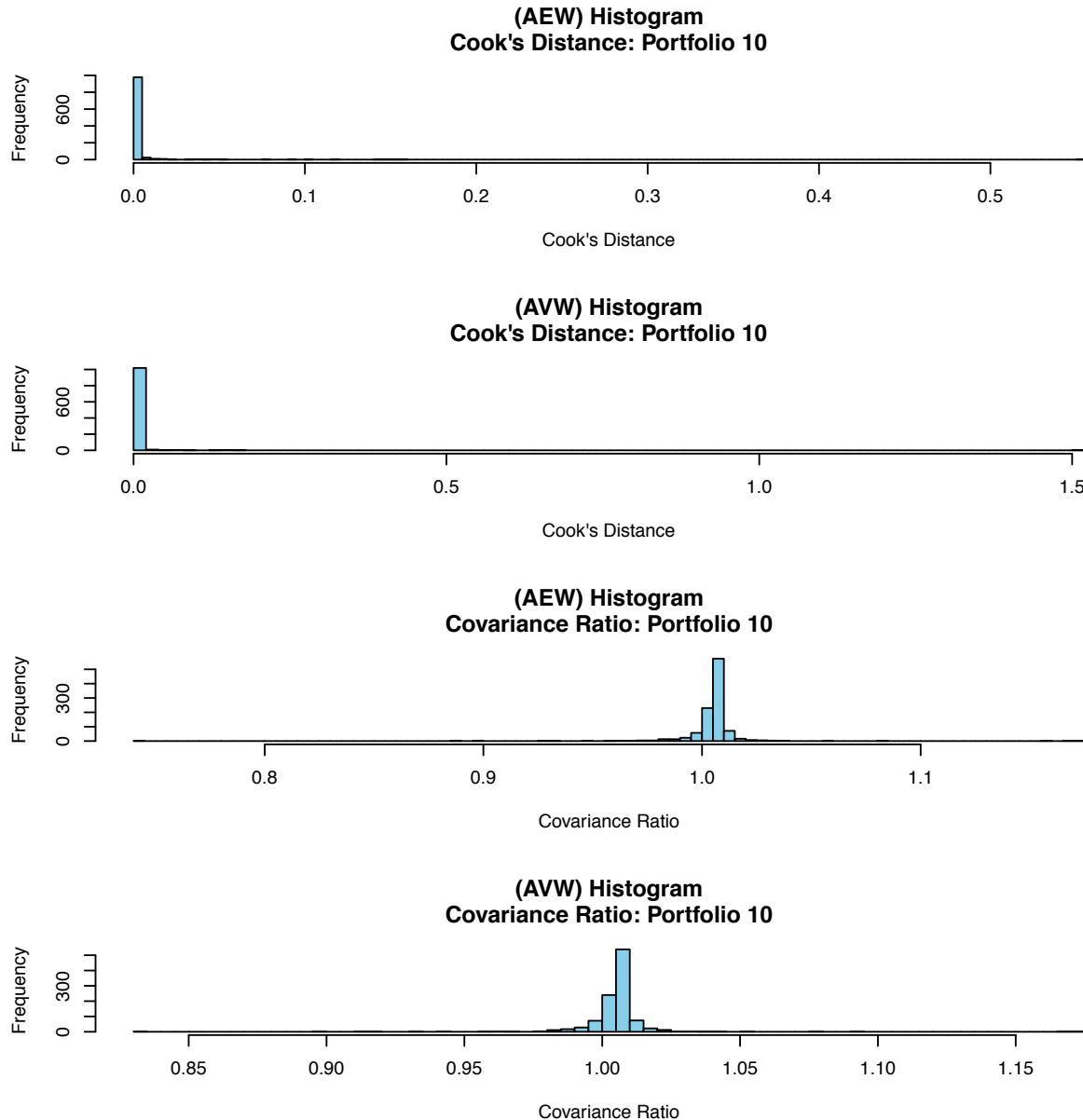
C.8 Portfolio 8



C.9 Portfolio 9



C.10 Portfolio 10



Appendix D

Influence Measures of Influential Observations

D.1 (AEW) Portfolio 1

```

## $`Cook's Distance`
##      Month   Ci
## 153 1939-09-01 1.038
##
## $`Covariance Ratio`
##      Month CVRi abs.CVRi_1 calpt.crit
## 889 2001-01-01 0.7361    0.26388  0.01444
## 153 1939-09-01 0.8099    0.19009  0.01444
## 67  1932-07-01 1.1757    0.17570  0.01444
## 77  1933-05-01 1.1689    0.16888  0.01444
## 68  1932-08-01 1.1662    0.16624  0.01444
## 781 1992-01-01 0.8716    0.12843  0.01444
## 107 1935-11-01 0.8892    0.11077  0.01444
## 11  1927-11-01 0.9030    0.09699  0.01444
## 985 2009-01-01 0.9155    0.08446  0.01444
## 878 2000-02-01 1.0844    0.08435  0.01444
## 865 1999-01-01 0.9198    0.08020  0.01444
## 877 2000-01-01 0.9219    0.07810  0.01444
## 194 1943-02-01 0.9309    0.06913  0.01444
## 169 1941-01-01 0.9318    0.06823  0.01444
## 729 1987-09-01 0.9354    0.06460  0.01444
## 988 2009-04-01 1.0646    0.06457  0.01444
## 76  1933-04-01 1.0576    0.05756  0.01444
## 181 1942-01-01 0.9430    0.05696  0.01444
## 888 2000-12-01 0.9436    0.05640  0.01444
## 61  1932-01-01 0.9485    0.05154  0.01444
## 890 2001-02-01 1.0419    0.04188  0.01444
## 992 2009-08-01 0.9583    0.04167  0.01444
## 730 1987-10-01 1.0397    0.03972  0.01444
## 768 1990-12-01 0.9620    0.03801  0.01444
## 989 2009-05-01 0.9622    0.03776  0.01444
## 887 2000-11-01 0.9636    0.03645  0.01444
## 882 2000-06-01 1.0349    0.03491  0.01444
## 17  1928-05-01 0.9680    0.03198  0.01444
## 574 1974-10-01 1.0302    0.03017  0.01444
## 85  1934-01-01 1.0302    0.03016  0.01444
## 841 1997-01-01 0.9702    0.02980  0.01444
## 54  1931-06-01 1.0283    0.02829  0.01444
## 979 2008-07-01 0.9730    0.02702  0.01444

```

## 65	1932-05-01	1.0266	0.02664	0.01444
## 34	1929-10-01	1.0264	0.02637	0.01444
## 840	1996-12-01	0.9742	0.02577	0.01444
## 57	1931-09-01	1.0256	0.02561	0.01444
## 917	2003-05-01	0.9756	0.02441	0.01444
## 673	1983-01-01	0.9757	0.02431	0.01444
## 622	1978-10-01	1.0241	0.02413	0.01444
## 129	1937-09-01	1.0239	0.02390	0.01444
## 160	1940-04-01	0.9763	0.02371	0.01444
## 852	1997-12-01	0.9770	0.02302	0.01444
## 898	2001-10-01	1.0227	0.02266	0.01444
## 577	1975-01-01	1.0226	0.02262	0.01444
## 639	1980-03-01	1.0225	0.02248	0.01444
## 525	1970-09-01	1.0222	0.02222	0.01444
## 879	2000-03-01	1.0222	0.02221	0.01444
## 770	1991-02-01	0.9786	0.02136	0.01444
## 283	1950-07-01	1.0207	0.02069	0.01444
## 64	1932-04-01	1.0206	0.02061	0.01444
## 135	1938-03-01	1.0199	0.01994	0.01444
## 860	1998-08-01	1.0199	0.01992	0.01444
## 37	1930-01-01	0.9801	0.01992	0.01444
## 53	1931-05-01	1.0191	0.01915	0.01444
## 565	1974-01-01	1.0189	0.01886	0.01444
## 147	1939-03-01	1.0186	0.01860	0.01444
## 82	1933-10-01	1.0185	0.01854	0.01444
## 589	1976-01-01	1.0185	0.01852	0.01444
## 138	1938-06-01	0.9823	0.01774	0.01444
## 880	2000-04-01	1.0174	0.01740	0.01444
## 71	1932-11-01	1.0169	0.01688	0.01444
## 886	2000-10-01	0.9832	0.01684	0.01444
## 39	1930-03-01	0.9833	0.01670	0.01444
## 42	1930-06-01	1.0165	0.01651	0.01444
## 853	1998-01-01	0.9839	0.01612	0.01444
## 50	1931-02-01	1.0160	0.01598	0.01444
## 697	1985-01-01	1.0156	0.01562	0.01444
## 564	1973-12-01	1.0155	0.01552	0.01444
## 60	1931-12-01	1.0155	0.01546	0.01444
## 559	1973-07-01	1.0154	0.01535	0.01444
## 987	2009-03-01	0.9848	0.01521	0.01444
## 876	1999-12-01	1.0149	0.01491	0.01444
## 81	1933-09-01	1.0149	0.01488	0.01444

```
## 83 1933-11-01 1.0147    0.01469    0.01444  
## 35 1929-11-01 1.0145    0.01454    0.01444
```

D.2 (AVW) Portfolio 1

```

## $`Cook's Distance`
## [1] "None"
##
## $`Covariance Ratio`
##           Month    CVRi abs.CVRi_1 calpt.crit
## 987 2009-03-01 0.7036   0.29640   0.01444
## 67   1932-07-01 1.1871   0.18711   0.01444
## 77   1933-05-01 1.1744   0.17439   0.01444
## 68   1932-08-01 1.1624   0.16239   0.01444
## 6    1927-06-01 0.8398   0.16023   0.01444
## 980  2008-08-01 0.8936   0.10639   0.01444
## 992  2009-08-01 0.8954   0.10461   0.01444
## 878  2000-02-01 1.0844   0.08444   0.01444
## 153  1939-09-01 1.0838   0.08378   0.01444
## 981  2008-09-01 0.9260   0.07400   0.01444
## 988  2009-04-01 1.0662   0.06617   0.01444
## 887  2000-11-01 0.9347   0.06533   0.01444
## 66   1932-06-01 0.9426   0.05742   0.01444
## 99   1935-03-01 0.9532   0.04681   0.01444
## 3    1927-03-01 0.9536   0.04644   0.01444
## 986  2009-02-01 0.9551   0.04493   0.01444
## 131  1937-11-01 0.9569   0.04312   0.01444
## 730  1987-10-01 1.0403   0.04028   0.01444
## 882  2000-06-01 1.0393   0.03929   0.01444
## 85   1934-01-01 1.0368   0.03684   0.01444
## 907  2002-07-01 0.9642   0.03579   0.01444
## 879  2000-03-01 1.0353   0.03531   0.01444
## 57   1931-09-01 1.0350   0.03496   0.01444
## 889  2001-01-01 1.0343   0.03430   0.01444
## 105  1935-09-01 0.9677   0.03233   0.01444
## 902  2002-02-01 0.9678   0.03219   0.01444
## 890  2001-02-01 1.0311   0.03114   0.01444
## 886  2000-10-01 0.9694   0.03057   0.01444
## 34   1929-10-01 1.0294   0.02944   0.01444
## 65   1932-05-01 1.0283   0.02826   0.01444
## 910  2002-10-01 0.9725   0.02745   0.01444
## 622  1978-10-01 1.0259   0.02594   0.01444
## 28   1929-04-01 0.9745   0.02545   0.01444
## 135  1938-03-01 1.0250   0.02500   0.01444

```

## 876	1999-12-01	1.0249	0.02489	0.01444
## 559	1973-07-01	1.0245	0.02445	0.01444
## 911	2002-11-01	1.0228	0.02281	0.01444
## 121	1937-01-01	0.9777	0.02232	0.01444
## 880	2000-04-01	1.0223	0.02230	0.01444
## 80	1933-08-01	1.0220	0.02203	0.01444
## 49	1931-01-01	0.9793	0.02074	0.01444
## 283	1950-07-01	1.0207	0.02066	0.01444
## 908	2002-08-01	0.9794	0.02061	0.01444
## 565	1974-01-01	1.0203	0.02034	0.01444
## 61	1932-01-01	0.9797	0.02033	0.01444
## 898	2001-10-01	1.0203	0.02030	0.01444
## 867	1999-03-01	0.9801	0.01992	0.01444
## 574	1974-10-01	1.0199	0.01989	0.01444
## 985	2009-01-01	1.0197	0.01974	0.01444
## 888	2000-12-01	0.9803	0.01966	0.01444
## 78	1933-06-01	0.9807	0.01932	0.01444
## 64	1932-04-01	1.0193	0.01925	0.01444
## 147	1939-03-01	1.0189	0.01886	0.01444
## 70	1932-10-01	1.0186	0.01864	0.01444
## 860	1998-08-01	1.0186	0.01860	0.01444
## 577	1975-01-01	1.0182	0.01825	0.01444
## 82	1933-10-01	1.0182	0.01823	0.01444
## 1039	2013-07-01	0.9818	0.01816	0.01444
## 875	1999-11-01	1.0181	0.01809	0.01444
## 31	1929-07-01	0.9820	0.01800	0.01444
## 893	2001-05-01	0.9822	0.01779	0.01444
## 60	1931-12-01	1.0177	0.01772	0.01444
## 523	1970-07-01	0.9825	0.01752	0.01444
## 647	1980-11-01	1.0175	0.01751	0.01444
## 143	1938-11-01	0.9832	0.01682	0.01444
## 33	1929-09-01	0.9832	0.01675	0.01444
## 42	1930-06-01	1.0167	0.01673	0.01444
## 50	1931-02-01	0.9835	0.01650	0.01444
## 62	1932-02-01	0.9840	0.01597	0.01444
## 98	1935-02-01	0.9842	0.01583	0.01444
## 53	1931-05-01	1.0158	0.01575	0.01444
## 35	1929-11-01	1.0157	0.01567	0.01444
## 75	1933-03-01	0.9844	0.01562	0.01444
## 109	1936-01-01	1.0155	0.01553	0.01444
## 4	1927-04-01	0.9847	0.01534	0.01444

```
## 881 2000-05-01 1.0150 0.01500 0.01444
## 897 2001-09-01 1.0148 0.01485 0.01444
## 697 1985-01-01 1.0148 0.01477 0.01444
## 48   1930-12-01 1.0147 0.01470 0.01444
## 780  1991-12-01 1.0147 0.01468 0.01444
## 112  1936-04-01 1.0147 0.01468 0.01444
## 89   1934-05-01 1.0146 0.01461 0.01444
## 193  1943-01-01 1.0146 0.01459 0.01444
## 891  2001-03-01 1.0146 0.01458 0.01444
## 781  1992-01-01 1.0145 0.01453 0.01444
```

D.3 (AEW) Portfolio 2

```

## $`Cook's Distance`
##      Month   Ci
## 68 1932-08-01 3.733
##
## $`Covariance Ratio`
##      Month   CVRi abs.CVRi_1 calpt.crit
## 68 1932-08-01 0.6595    0.34053   0.01444
## 67 1932-07-01 1.1736    0.17364   0.01444
## 985 2009-01-01 0.8554    0.14460   0.01444
## 77 1933-05-01 1.1359    0.13586   0.01444
## 37 1930-01-01 0.9017    0.09832   0.01444
## 878 2000-02-01 1.0843    0.08430   0.01444
## 153 1939-09-01 1.0836    0.08363   0.01444
## 865 1999-01-01 0.9168    0.08318   0.01444
## 76 1933-04-01 1.0708    0.07076   0.01444
## 988 2009-04-01 1.0652    0.06522   0.01444
## 877 2000-01-01 0.9399    0.06012   0.01444
## 78 1933-06-01 0.9515    0.04853   0.01444
## 138 1938-06-01 1.0475    0.04745   0.01444
## 899 2001-11-01 0.9537    0.04632   0.01444
## 49 1931-01-01 0.9590    0.04103   0.01444
## 730 1987-10-01 1.0402    0.04022   0.01444
## 882 2000-06-01 1.0400    0.03999   0.01444
## 57 1931-09-01 1.0399    0.03987   0.01444
## 66 1932-06-01 0.9603    0.03974   0.01444
## 989 2009-05-01 0.9610    0.03900   0.01444
## 36 1929-12-01 0.9615    0.03848   0.01444
## 107 1935-11-01 0.9622    0.03778   0.01444
## 69 1932-09-01 0.9623    0.03769   0.01444
## 980 2008-08-01 0.9638    0.03620   0.01444
## 886 2000-10-01 0.9643    0.03573   0.01444
## 979 2008-07-01 0.9674    0.03262   0.01444
## 139 1938-07-01 0.9675    0.03254   0.01444
## 888 2000-12-01 0.9693    0.03073   0.01444
## 58 1931-10-01 0.9696    0.03039   0.01444
## 767 1990-11-01 0.9697    0.03032   0.01444
## 34 1929-10-01 1.0295    0.02955   0.01444
## 673 1983-01-01 0.9729    0.02709   0.01444
## 574 1974-10-01 1.0268    0.02680   0.01444

```

## 100	1935-04-01	0.9734	0.02663	0.01444
## 181	1942-01-01	0.9736	0.02637	0.01444
## 71	1932-11-01	1.0259	0.02585	0.01444
## 65	1932-05-01	1.0252	0.02523	0.01444
## 129	1937-09-01	1.0251	0.02513	0.01444
## 889	2001-01-01	0.9752	0.02480	0.01444
## 781	1992-01-01	0.9752	0.02479	0.01444
## 639	1980-03-01	1.0246	0.02457	0.01444
## 577	1975-01-01	1.0239	0.02387	0.01444
## 81	1933-09-01	1.0233	0.02326	0.01444
## 112	1936-04-01	0.9770	0.02295	0.01444
## 876	1999-12-01	1.0228	0.02278	0.01444
## 768	1990-12-01	0.9779	0.02212	0.01444
## 880	2000-04-01	1.0218	0.02176	0.01444
## 161	1940-05-01	1.0217	0.02166	0.01444
## 70	1932-10-01	0.9784	0.02163	0.01444
## 80	1933-08-01	1.0216	0.02156	0.01444
## 622	1978-10-01	1.0215	0.02153	0.01444
## 56	1931-08-01	0.9789	0.02108	0.01444
## 559	1973-07-01	1.0204	0.02039	0.01444
## 83	1933-11-01	1.0194	0.01945	0.01444
## 525	1970-09-01	1.0192	0.01924	0.01444
## 879	2000-03-01	1.0192	0.01922	0.01444
## 897	2001-09-01	0.9810	0.01897	0.01444
## 911	2002-11-01	1.0189	0.01895	0.01444
## 82	1933-10-01	1.0187	0.01866	0.01444
## 645	1980-09-01	0.9813	0.01865	0.01444
## 283	1950-07-01	1.0186	0.01865	0.01444
## 589	1976-01-01	1.0186	0.01855	0.01444
## 60	1931-12-01	1.0180	0.01797	0.01444
## 563	1973-11-01	1.0175	0.01751	0.01444
## 716	1986-08-01	0.9828	0.01720	0.01444
## 982	2008-10-01	1.0167	0.01669	0.01444
## 860	1998-08-01	1.0165	0.01648	0.01444
## 1018	2011-10-01	0.9839	0.01610	0.01444
## 841	1997-01-01	0.9842	0.01579	0.01444
## 904	2002-04-01	1.0158	0.01579	0.01444
## 42	1930-06-01	1.0156	0.01562	0.01444
## 898	2001-10-01	0.9844	0.01556	0.01444
## 874	1999-10-01	1.0155	0.01545	0.01444
## 523	1970-07-01	0.9847	0.01532	0.01444

```
## 74 1933-02-01 0.9848 0.01523 0.01444
## 697 1985-01-01 1.0152 0.01518 0.01444
## 64 1932-04-01 1.0149 0.01491 0.01444
## 564 1973-12-01 1.0149 0.01487 0.01444
## 147 1939-03-01 1.0147 0.01475 0.01444
## 98 1935-02-01 1.0146 0.01459 0.01444
## 193 1943-01-01 0.9855 0.01445 0.01444
```

D.4 (AVW) Portfolio 2

```

## $`Cook's Distance`
##      Month   Ci
## 67 1932-07-01 1.221
##
## $`Covariance Ratio`
##      Month CVRi abs.CVRi_1 calpt.crit
## 104 1935-08-01 0.7842    0.21580  0.01444
## 77 1933-05-01 1.1724    0.17238  0.01444
## 68 1932-08-01 1.1302    0.13025  0.01444
## 12 1927-12-01 0.8949    0.10510  0.01444
## 153 1939-09-01 1.0917    0.09174  0.01444
## 878 2000-02-01 1.0801    0.08011  0.01444
## 50 1931-02-01 0.9216    0.07838  0.01444
## 36 1929-12-01 0.9286    0.07145  0.01444
## 988 2009-04-01 1.0644    0.06443  0.01444
## 65 1932-05-01 0.9389    0.06113  0.01444
## 76 1933-04-01 1.0601    0.06008  0.01444
## 54 1931-06-01 0.9459    0.05406  0.01444
## 41 1930-05-01 0.9467    0.05334  0.01444
## 138 1938-06-01 1.0471    0.04706  0.01444
## 911 2002-11-01 0.9531    0.04691  0.01444
## 882 2000-06-01 1.0437    0.04370  0.01444
## 139 1938-07-01 0.9585    0.04150  0.01444
## 57 1931-09-01 1.0389    0.03892  0.01444
## 75 1933-03-01 0.9669    0.03307  0.01444
## 890 2001-02-01 1.0323    0.03233  0.01444
## 47 1930-11-01 0.9678    0.03221  0.01444
## 889 2001-01-01 1.0316    0.03160  0.01444
## 151 1939-07-01 0.9697    0.03033  0.01444
## 66 1932-06-01 0.9699    0.03007  0.01444
## 62 1932-02-01 0.9705    0.02952  0.01444
## 647 1980-11-01 1.0278    0.02778  0.01444
## 622 1978-10-01 1.0259    0.02594  0.01444
## 879 2000-03-01 1.0259    0.02592  0.01444
## 887 2000-11-01 1.0256    0.02563  0.01444
## 574 1974-10-01 1.0254    0.02539  0.01444
## 166 1940-10-01 0.9746    0.02538  0.01444
## 639 1980-03-01 1.0251    0.02515  0.01444
## 1 1927-01-01 0.9749    0.02512  0.01444

```

## 129	1937-09-01	1.0243	0.02431	0.01444
## 107	1935-11-01	0.9764	0.02356	0.01444
## 161	1940-05-01	1.0231	0.02314	0.01444
## 885	2000-09-01	0.9771	0.02293	0.01444
## 48	1930-12-01	0.9772	0.02279	0.01444
## 525	1970-09-01	1.0226	0.02257	0.01444
## 40	1930-04-01	0.9780	0.02196	0.01444
## 67	1932-07-01	1.0218	0.02181	0.01444
## 880	2000-04-01	1.0208	0.02079	0.01444
## 26	1929-02-01	0.9794	0.02062	0.01444
## 565	1974-01-01	1.0203	0.02032	0.01444
## 860	1998-08-01	1.0200	0.02003	0.01444
## 81	1933-09-01	1.0200	0.02001	0.01444
## 34	1929-10-01	1.0199	0.01986	0.01444
## 135	1938-03-01	1.0197	0.01968	0.01444
## 796	1993-04-01	0.9805	0.01952	0.01444
## 101	1935-05-01	0.9806	0.01942	0.01444
## 53	1931-05-01	1.0192	0.01923	0.01444
## 112	1936-04-01	1.0192	0.01917	0.01444
## 84	1933-12-01	0.9809	0.01912	0.01444
## 559	1973-07-01	1.0190	0.01903	0.01444
## 194	1943-02-01	0.9811	0.01885	0.01444
## 82	1933-10-01	1.0186	0.01858	0.01444
## 32	1929-08-01	1.0182	0.01823	0.01444
## 33	1929-09-01	0.9823	0.01772	0.01444
## 181	1942-01-01	1.0173	0.01730	0.01444
## 21	1928-09-01	0.9830	0.01697	0.01444
## 904	2002-04-01	1.0168	0.01683	0.01444
## 876	1999-12-01	1.0167	0.01672	0.01444
## 982	2008-10-01	1.0167	0.01666	0.01444
## 60	1931-12-01	1.0165	0.01654	0.01444
## 979	2008-07-01	0.9840	0.01596	0.01444
## 541	1972-01-01	0.9842	0.01578	0.01444
## 730	1987-10-01	1.0156	0.01561	0.01444
## 179	1941-11-01	0.9846	0.01540	0.01444
## 105	1935-09-01	0.9849	0.01513	0.01444
## 697	1985-01-01	1.0151	0.01508	0.01444
## 985	2009-01-01	1.0149	0.01485	0.01444
## 70	1932-10-01	1.0146	0.01459	0.01444

D.5 (AEW) Portfolio 3

```

## $`Cook's Distance`
##      Month   Ci
## 77 1933-05-01 2.469
##
## $`Covariance Ratio`
##      Month CVRi abs.CVRi_1 calpt.crit
## 77 1933-05-01 0.8140    0.18600  0.01444
## 68 1932-08-01 1.1619    0.16186  0.01444
## 67 1932-07-01 1.1420    0.14198  0.01444
## 78 1933-06-01 0.8890    0.11097  0.01444
## 71 1932-11-01 0.9014    0.09863  0.01444
## 58 1931-10-01 0.9102    0.08984  0.01444
## 153 1939-09-01 1.0839    0.08386  0.01444
## 80 1933-08-01 0.9299    0.07006  0.01444
## 76 1933-04-01 1.0622    0.06219  0.01444
## 985 2009-01-01 0.9432    0.05678  0.01444
## 40 1930-04-01 0.9444    0.05562  0.01444
## 988 2009-04-01 1.0542    0.05416  0.01444
## 989 2009-05-01 0.9548    0.04517  0.01444
## 730 1987-10-01 1.0388    0.03880  0.01444
## 577 1975-01-01 0.9613    0.03867  0.01444
## 138 1938-06-01 1.0361    0.03608  0.01444
## 529 1971-01-01 0.9644    0.03562  0.01444
## 890 2001-02-01 0.9645    0.03549  0.01444
## 865 1999-01-01 0.9649    0.03511  0.01444
## 16 1928-04-01 0.9654    0.03460  0.01444
## 889 2001-01-01 1.0306    0.03063  0.01444
## 574 1974-10-01 1.0305    0.03045  0.01444
## 37 1930-01-01 0.9707    0.02933  0.01444
## 135 1938-03-01 1.0289    0.02886  0.01444
## 767 1990-11-01 0.9716    0.02840  0.01444
## 65 1932-05-01 1.0277    0.02774  0.01444
## 34 1929-10-01 1.0270    0.02697  0.01444
## 673 1983-01-01 0.9738    0.02623  0.01444
## 891 2001-03-01 0.9750    0.02501  0.01444
## 639 1980-03-01 1.0248    0.02476  0.01444
## 66 1932-06-01 0.9753    0.02466  0.01444
## 887 2000-11-01 1.0246    0.02459  0.01444
## 559 1973-07-01 1.0246    0.02457  0.01444

```

## 92	1934-08-01	0.9755	0.02449	0.01444
## 990	2009-06-01	0.9761	0.02391	0.01444
## 161	1940-05-01	1.0238	0.02384	0.01444
## 892	2001-04-01	0.9762	0.02377	0.01444
## 911	2002-11-01	1.0230	0.02301	0.01444
## 520	1970-04-01	0.9777	0.02227	0.01444
## 895	2001-07-01	0.9778	0.02222	0.01444
## 779	1991-11-01	0.9785	0.02146	0.01444
## 899	2001-11-01	0.9789	0.02106	0.01444
## 882	2000-06-01	0.9790	0.02096	0.01444
## 64	1932-04-01	1.0209	0.02092	0.01444
## 57	1931-09-01	1.0206	0.02062	0.01444
## 32	1929-08-01	1.0196	0.01963	0.01444
## 589	1976-01-01	1.0196	0.01960	0.01444
## 647	1980-11-01	1.0196	0.01956	0.01444
## 876	1999-12-01	1.0194	0.01943	0.01444
## 91	1934-07-01	1.0194	0.01936	0.01444
## 70	1932-10-01	1.0191	0.01907	0.01444
## 112	1936-04-01	1.0189	0.01893	0.01444
## 147	1939-03-01	1.0189	0.01891	0.01444
## 83	1933-11-01	0.9812	0.01880	0.01444
## 129	1937-09-01	1.0185	0.01846	0.01444
## 541	1972-01-01	0.9819	0.01806	0.01444
## 875	1999-11-01	1.0180	0.01797	0.01444
## 283	1950-07-01	0.9821	0.01794	0.01444
## 563	1973-11-01	1.0179	0.01789	0.01444
## 61	1932-01-01	1.0176	0.01756	0.01444
## 81	1933-09-01	1.0175	0.01749	0.01444
## 979	2008-07-01	0.9827	0.01726	0.01444
## 564	1973-12-01	1.0169	0.01695	0.01444
## 108	1935-12-01	0.9831	0.01692	0.01444
## 54	1931-06-01	1.0166	0.01664	0.01444
## 109	1936-01-01	1.0166	0.01662	0.01444
## 904	2002-04-01	1.0166	0.01658	0.01444
## 565	1974-01-01	1.0163	0.01629	0.01444
## 50	1931-02-01	1.0158	0.01580	0.01444
## 982	2008-10-01	1.0157	0.01565	0.01444
## 525	1970-09-01	1.0154	0.01537	0.01444
## 874	1999-10-01	1.0149	0.01491	0.01444
## 573	1974-09-01	1.0149	0.01490	0.01444
## 881	2000-05-01	1.0149	0.01488	0.01444

```
## 860 1998-08-01 1.0148    0.01484   0.01444
## 30   1929-06-01 1.0147    0.01471   0.01444
## 986  2009-02-01 0.9854    0.01465   0.01444
## 48   1930-12-01 1.0146    0.01463   0.01444
## 697  1985-01-01 1.0146    0.01457   0.01444
```

D.6 (AVW) Portfolio 3

```

## $`Cook's Distance`
## [1] "None"
##
## $`Covariance Ratio`
##           Month    CVRi abs.CVRi_1 calpt.crit
## 67 1932-07-01 1.1961     0.19608   0.01444
## 68 1932-08-01 1.1718     0.17175   0.01444
## 889 2001-01-01 0.8703     0.12972   0.01444
## 153 1939-09-01 1.0915     0.09154   0.01444
## 878 2000-02-01 1.0844     0.08445   0.01444
## 102 1935-06-01 0.9284     0.07155   0.01444
## 379 1958-07-01 0.9396     0.06043   0.01444
## 40 1930-04-01 0.9438     0.05619   0.01444
## 41 1930-05-01 0.9488     0.05124   0.01444
## 906 2002-06-01 0.9496     0.05038   0.01444
## 882 2000-06-01 1.0437     0.04368   0.01444
## 561 1973-09-01 0.9565     0.04354   0.01444
## 76 1933-04-01 1.0426     0.04258   0.01444
## 77 1933-05-01 1.0426     0.04258   0.01444
## 911 2002-11-01 0.9576     0.04239   0.01444
## 57 1931-09-01 1.0396     0.03960   0.01444
## 70 1932-10-01 0.9650     0.03495   0.01444
## 979 2008-07-01 0.9658     0.03415   0.01444
## 986 2009-02-01 0.9660     0.03404   0.01444
## 883 2000-07-01 0.9677     0.03230   0.01444
## 868 1999-04-01 0.9702     0.02976   0.01444
## 730 1987-10-01 1.0297     0.02969   0.01444
## 769 1991-01-01 0.9715     0.02850   0.01444
## 647 1980-11-01 1.0278     0.02777   0.01444
## 135 1938-03-01 1.0272     0.02717   0.01444
## 890 2001-02-01 1.0270     0.02701   0.01444
## 899 2001-11-01 0.9730     0.02696   0.01444
## 990 2009-06-01 0.9746     0.02540   0.01444
## 639 1980-03-01 1.0253     0.02527   0.01444
## 510 1969-06-01 0.9748     0.02524   0.01444
## 879 2000-03-01 1.0249     0.02487   0.01444
## 992 2009-08-01 0.9765     0.02347   0.01444
## 129 1937-09-01 1.0233     0.02333   0.01444
## 988 2009-04-01 1.0217     0.02174   0.01444

```

## 34	1929-10-01	0.9784	0.02158	0.01444
## 53	1931-05-01	1.0210	0.02105	0.01444
## 99	1935-03-01	0.9790	0.02100	0.01444
## 525	1970-09-01	1.0206	0.02055	0.01444
## 117	1936-09-01	0.9800	0.02000	0.01444
## 142	1938-10-01	0.9800	0.02000	0.01444
## 589	1976-01-01	1.0199	0.01989	0.01444
## 75	1933-03-01	0.9802	0.01977	0.01444
## 137	1938-05-01	0.9803	0.01969	0.01444
## 902	2002-02-01	0.9813	0.01870	0.01444
## 181	1942-01-01	1.0186	0.01864	0.01444
## 881	2000-05-01	0.9815	0.01854	0.01444
## 82	1933-10-01	1.0183	0.01834	0.01444
## 898	2001-10-01	0.9819	0.01810	0.01444
## 60	1931-12-01	1.0181	0.01807	0.01444
## 161	1940-05-01	1.0180	0.01803	0.01444
## 622	1978-10-01	1.0177	0.01773	0.01444
## 46	1930-10-01	0.9824	0.01760	0.01444
## 50	1931-02-01	1.0173	0.01727	0.01444
## 910	2002-10-01	1.0172	0.01720	0.01444
## 985	2009-01-01	1.0170	0.01702	0.01444
## 241	1947-01-01	0.9830	0.01702	0.01444
## 875	1999-11-01	1.0168	0.01683	0.01444
## 64	1932-04-01	1.0164	0.01645	0.01444
## 193	1943-01-01	1.0162	0.01618	0.01444
## 104	1935-08-01	0.9840	0.01595	0.01444
## 59	1931-11-01	1.0159	0.01589	0.01444
## 574	1974-10-01	1.0158	0.01584	0.01444
## 520	1970-04-01	1.0156	0.01561	0.01444
## 697	1985-01-01	1.0156	0.01558	0.01444
## 573	1974-09-01	1.0156	0.01556	0.01444
## 904	2002-04-01	1.0152	0.01525	0.01444
## 577	1975-01-01	1.0149	0.01493	0.01444
## 89	1934-05-01	1.0146	0.01456	0.01444

D.7 (AEW) Portfolio 4

```

## $`Cook's Distance`
##      Month   Ci
## 77 1933-05-01 1.992
## 67 1932-07-01 1.385
##
## $`Covariance Ratio`
##      Month CVRi abs.CVRi_1 calpt.crit
## 68 1932-08-01 1.1252    0.12525  0.01444
## 77 1933-05-01 0.8756    0.12439  0.01444
## 78 1933-06-01 0.8935    0.10645  0.01444
## 104 1935-08-01 0.9078    0.09221  0.01444
## 153 1939-09-01 1.0918    0.09176  0.01444
## 58 1931-10-01 0.9229    0.07713  0.01444
## 66 1932-06-01 0.9415    0.05854  0.01444
## 889 2001-01-01 1.0572    0.05723  0.01444
## 878 2000-02-01 1.0567    0.05674  0.01444
## 76 1933-04-01 0.9475    0.05249  0.01444
## 671 1982-11-01 0.9492    0.05084  0.01444
## 62 1932-02-01 0.9493    0.05066  0.01444
## 193 1943-01-01 0.9494    0.05059  0.01444
## 83 1933-11-01 0.9509    0.04911  0.01444
## 988 2009-04-01 1.0443    0.04431  0.01444
## 51 1931-03-01 0.9567    0.04326  0.01444
## 13 1928-01-01 0.9576    0.04236  0.01444
## 985 2009-01-01 0.9600    0.04004  0.01444
## 107 1935-11-01 0.9616    0.03841  0.01444
## 779 1991-11-01 0.9628    0.03724  0.01444
## 899 2001-11-01 0.9639    0.03605  0.01444
## 879 2000-03-01 1.0335    0.03346  0.01444
## 43 1930-07-01 0.9669    0.03313  0.01444
## 574 1974-10-01 1.0312    0.03116  0.01444
## 85 1934-01-01 1.0304    0.03038  0.01444
## 730 1987-10-01 1.0303    0.03033  0.01444
## 65 1932-05-01 1.0302    0.03015  0.01444
## 897 2001-09-01 0.9701    0.02991  0.01444
## 135 1938-03-01 1.0289    0.02895  0.01444
## 890 2001-02-01 0.9717    0.02826  0.01444
## 194 1943-02-01 0.9717    0.02826  0.01444
## 891 2001-03-01 0.9732    0.02681  0.01444

```

```

## 887 2000-11-01 1.0253    0.02526   0.01444
## 109 1936-01-01 0.9750    0.02504   0.01444
## 559 1973-07-01 1.0245    0.02454   0.01444
## 34  1929-10-01 1.0245    0.02450   0.01444
## 81  1933-09-01 1.0240    0.02405   0.01444
## 639 1980-03-01 1.0237    0.02370   0.01444
## 54  1931-06-01 1.0225    0.02252   0.01444
## 622 1978-10-01 1.0222    0.02221   0.01444
## 80  1933-08-01 0.9780    0.02203   0.01444
## 577 1975-01-01 1.0216    0.02158   0.01444
## 71  1932-11-01 1.0209    0.02089   0.01444
## 898 2001-10-01 1.0207    0.02069   0.01444
## 161 1940-05-01 1.0207    0.02068   0.01444
## 56  1931-08-01 0.9798    0.02020   0.01444
## 32  1929-08-01 1.0201    0.02009   0.01444
## 525 1970-09-01 1.0200    0.02001   0.01444
## 16  1928-04-01 0.9800    0.01996   0.01444
## 53  1931-05-01 1.0198    0.01983   0.01444
## 129 1937-09-01 1.0197    0.01972   0.01444
## 565 1974-01-01 1.0197    0.01967   0.01444
## 563 1973-11-01 1.0190    0.01896   0.01444
## 860 1998-08-01 1.0186    0.01860   0.01444
## 82  1933-10-01 1.0185    0.01849   0.01444
## 875 1999-11-01 1.0181    0.01809   0.01444
## 589 1976-01-01 1.0178    0.01778   0.01444
## 181 1942-01-01 1.0177    0.01770   0.01444
## 112 1936-04-01 1.0170    0.01705   0.01444
## 60  1931-12-01 1.0166    0.01662   0.01444
## 982 2008-10-01 1.0166    0.01656   0.01444
## 904 2002-04-01 1.0162    0.01618   0.01444
## 40  1930-04-01 0.9838    0.01617   0.01444
## 61  1932-01-01 1.0159    0.01588   0.01444
## 874 1999-10-01 1.0158    0.01584   0.01444
## 649 1981-01-01 1.0151    0.01511   0.01444
## 42  1930-06-01 1.0151    0.01505   0.01444
## 647 1980-11-01 1.0149    0.01494   0.01444
## 91  1934-07-01 1.0149    0.01486   0.01444
## 520 1970-04-01 0.9854    0.01460   0.01444

```

D.8 (AVW) Portfolio 4

```

## $`Cook's Distance`
##      Month   Ci
## 76 1933-04-01 1.274
##
## $`Covariance Ratio`
##      Month   CVRi abs.CVRi_1 calpt.crit
## 76 1933-04-01 0.6539    0.34609   0.01444
## 67 1932-07-01 1.1937    0.19372   0.01444
## 77 1933-05-01 1.1593    0.15933   0.01444
## 68 1932-08-01 1.0956    0.09559   0.01444
## 885 2000-09-01 0.9126    0.08737   0.01444
## 878 2000-02-01 1.0840    0.08401   0.01444
## 153 1939-09-01 1.0814    0.08138   0.01444
## 138 1938-06-01 0.9262    0.07380   0.01444
## 886 2000-10-01 0.9337    0.06634   0.01444
## 988 2009-04-01 1.0655    0.06552   0.01444
## 889 2001-01-01 1.0575    0.05746   0.01444
## 132 1937-12-01 0.9457    0.05430   0.01444
## 40 1930-04-01 0.9475    0.05252   0.01444
## 888 2000-12-01 0.9502    0.04980   0.01444
## 66 1932-06-01 0.9519    0.04814   0.01444
## 104 1935-08-01 0.9555    0.04450   0.01444
## 879 2000-03-01 1.0439    0.04385   0.01444
## 882 2000-06-01 1.0428    0.04275   0.01444
## 31 1929-07-01 0.9581    0.04188   0.01444
## 730 1987-10-01 1.0402    0.04024   0.01444
## 897 2001-09-01 0.9612    0.03882   0.01444
## 75 1933-03-01 0.9633    0.03674   0.01444
## 69 1932-09-01 0.9646    0.03537   0.01444
## 911 2002-11-01 0.9655    0.03455   0.01444
## 166 1940-10-01 0.9669    0.03311   0.01444
## 876 1999-12-01 0.9674    0.03255   0.01444
## 569 1974-05-01 0.9679    0.03213   0.01444
## 577 1975-01-01 0.9707    0.02931   0.01444
## 647 1980-11-01 1.0279    0.02792   0.01444
## 54 1931-06-01 1.0270    0.02699   0.01444
## 622 1978-10-01 1.0254    0.02538   0.01444
## 8 1927-08-01 0.9752    0.02476   0.01444
## 62 1932-02-01 0.9762    0.02381   0.01444

```

## 71	1932-11-01	1.0233	0.02330	0.01444
## 983	2008-11-01	0.9773	0.02272	0.01444
## 42	1930-06-01	0.9779	0.02210	0.01444
## 58	1931-10-01	0.9780	0.02196	0.01444
## 72	1932-12-01	1.0219	0.02193	0.01444
## 517	1970-01-01	0.9781	0.02189	0.01444
## 95	1934-11-01	0.9782	0.02177	0.01444
## 899	2001-11-01	0.9784	0.02162	0.01444
## 283	1950-07-01	1.0207	0.02071	0.01444
## 639	1980-03-01	1.0207	0.02068	0.01444
## 50	1931-02-01	0.9794	0.02061	0.01444
## 34	1929-10-01	1.0204	0.02040	0.01444
## 147	1939-03-01	1.0203	0.02028	0.01444
## 860	1998-08-01	1.0197	0.01969	0.01444
## 91	1934-07-01	1.0194	0.01941	0.01444
## 525	1970-09-01	1.0192	0.01923	0.01444
## 487	1967-07-01	0.9812	0.01881	0.01444
## 673	1983-01-01	0.9815	0.01846	0.01444
## 52	1931-04-01	0.9817	0.01828	0.01444
## 139	1938-07-01	0.9817	0.01828	0.01444
## 371	1957-11-01	0.9818	0.01824	0.01444
## 985	2009-01-01	1.0181	0.01812	0.01444
## 910	2002-10-01	1.0181	0.01810	0.01444
## 180	1941-12-01	0.9820	0.01802	0.01444
## 589	1976-01-01	1.0174	0.01737	0.01444
## 2	1927-02-01	0.9827	0.01728	0.01444
## 16	1928-04-01	0.9829	0.01705	0.01444
## 780	1991-12-01	1.0170	0.01696	0.01444
## 986	2009-02-01	0.9832	0.01676	0.01444
## 109	1936-01-01	1.0167	0.01675	0.01444
## 904	2002-04-01	1.0167	0.01666	0.01444
## 864	1998-12-01	0.9837	0.01625	0.01444
## 61	1932-01-01	1.0160	0.01601	0.01444
## 874	1999-10-01	1.0158	0.01584	0.01444
## 46	1930-10-01	0.9843	0.01567	0.01444
## 136	1938-04-01	1.0157	0.01567	0.01444
## 697	1985-01-01	1.0156	0.01564	0.01444
## 175	1941-07-01	0.9845	0.01545	0.01444
## 649	1981-01-01	1.0148	0.01482	0.01444
## 30	1929-06-01	1.0146	0.01462	0.01444

D.9 (AEW) Portfolio 5

```

## $`Cook's Distance`
##      Month   Ci
## 68 1932-08-01 2.973
##
## $`Covariance Ratio`
##      Month CVRi abs.CVRi_1 calpt.crit
## 68 1932-08-01 0.7455    0.25451  0.01444
## 77 1933-05-01 1.1741    0.17415  0.01444
## 138 1938-06-01 0.8610    0.13901  0.01444
## 153 1939-09-01 0.9074    0.09264  0.01444
## 67 1932-07-01 1.0839    0.08393  0.01444
## 104 1935-08-01 0.9210    0.07904  0.01444
## 61 1932-01-01 0.9320    0.06799  0.01444
## 988 2009-04-01 1.0639    0.06391  0.01444
## 54 1931-06-01 0.9378    0.06218  0.01444
## 80 1933-08-01 0.9471    0.05288  0.01444
## 889 2001-01-01 1.0502    0.05024  0.01444
## 85 1934-01-01 0.9524    0.04758  0.01444
## 671 1982-11-01 0.9553    0.04473  0.01444
## 985 2009-01-01 0.9554    0.04459  0.01444
## 897 2001-09-01 0.9666    0.03339  0.01444
## 878 2000-02-01 1.0327    0.03275  0.01444
## 779 1991-11-01 0.9707    0.02933  0.01444
## 880 2000-04-01 0.9710    0.02905  0.01444
## 882 2000-06-01 0.9712    0.02883  0.01444
## 75 1933-03-01 0.9715    0.02847  0.01444
## 57 1931-09-01 1.0283    0.02835  0.01444
## 574 1974-10-01 1.0277    0.02774  0.01444
## 877 2000-01-01 0.9724    0.02765  0.01444
## 622 1978-10-01 1.0258    0.02583  0.01444
## 137 1938-05-01 0.9748    0.02524  0.01444
## 639 1980-03-01 1.0250    0.02497  0.01444
## 129 1937-09-01 1.0249    0.02485  0.01444
## 730 1987-10-01 1.0242    0.02421  0.01444
## 71 1932-11-01 1.0240    0.02402  0.01444
## 81 1933-09-01 1.0239    0.02392  0.01444
## 161 1940-05-01 1.0236    0.02361  0.01444
## 76 1933-04-01 1.0234    0.02345  0.01444
## 521 1970-05-01 0.9770    0.02296  0.01444

```

## 525	1970-09-01	1.0224	0.02239	0.01444
## 879	2000-03-01	1.0219	0.02192	0.01444
## 72	1932-12-01	1.0219	0.02185	0.01444
## 32	1929-08-01	1.0207	0.02065	0.01444
## 64	1932-04-01	0.9802	0.01979	0.01444
## 860	1998-08-01	1.0197	0.01974	0.01444
## 147	1939-03-01	1.0195	0.01949	0.01444
## 91	1934-07-01	1.0193	0.01926	0.01444
## 29	1929-05-01	0.9808	0.01920	0.01444
## 43	1930-07-01	0.9810	0.01896	0.01444
## 892	2001-04-01	0.9814	0.01861	0.01444
## 668	1982-08-01	0.9814	0.01858	0.01444
## 197	1943-05-01	0.9814	0.01855	0.01444
## 898	2001-10-01	1.0184	0.01835	0.01444
## 893	2001-05-01	0.9818	0.01819	0.01444
## 283	1950-07-01	1.0181	0.01807	0.01444
## 682	1983-10-01	0.9833	0.01672	0.01444
## 65	1932-05-01	1.0165	0.01648	0.01444
## 257	1948-05-01	0.9836	0.01642	0.01444
## 193	1943-01-01	1.0164	0.01638	0.01444
## 59	1931-11-01	1.0163	0.01634	0.01444
## 563	1973-11-01	1.0160	0.01605	0.01444
## 564	1973-12-01	1.0160	0.01601	0.01444
## 989	2009-05-01	1.0156	0.01563	0.01444
## 573	1974-09-01	1.0154	0.01536	0.01444
## 577	1975-01-01	0.9850	0.01502	0.01444
## 649	1981-01-01	1.0150	0.01499	0.01444
## 12	1927-12-01	0.9851	0.01491	0.01444
## 56	1931-08-01	0.9853	0.01467	0.01444
## 34	1929-10-01	1.0146	0.01458	0.01444
## 982	2008-10-01	1.0146	0.01455	0.01444
## 876	1999-12-01	1.0145	0.01446	0.01444

D.10 (AVW) Portfolio 5

```

## $`Cook's Distance`
##      Month   Ci
## 68 1932-08-01 1.631
##
## $`Covariance Ratio`
##      Month CVRi abs.CVRi_1 calpt.crit
## 67 1932-07-01 1.1840    0.18396  0.01444
## 138 1938-06-01 0.8273    0.17269  0.01444
## 77 1933-05-01 1.1473    0.14728  0.01444
## 34 1929-10-01 0.9151    0.08494  0.01444
## 68 1932-08-01 0.9190    0.08100  0.01444
## 887 2000-11-01 0.9294    0.07057  0.01444
## 882 2000-06-01 0.9335    0.06655  0.01444
## 229 1946-01-01 0.9443    0.05575  0.01444
## 988 2009-04-01 1.0531    0.05308  0.01444
## 897 2001-09-01 0.9512    0.04879  0.01444
## 911 2002-11-01 0.9550    0.04503  0.01444
## 879 2000-03-01 1.0443    0.04431  0.01444
## 57 1931-09-01 1.0396    0.03962  0.01444
## 912 2002-12-01 0.9611    0.03886  0.01444
## 76 1933-04-01 0.9649    0.03510  0.01444
## 868 1999-04-01 0.9672    0.03279  0.01444
## 43 1930-07-01 0.9683    0.03171  0.01444
## 153 1939-09-01 1.0307    0.03072  0.01444
## 54 1931-06-01 1.0301    0.03014  0.01444
## 181 1942-01-01 0.9699    0.03006  0.01444
## 50 1931-02-01 0.9708    0.02918  0.01444
## 730 1987-10-01 1.0267    0.02670  0.01444
## 70 1932-10-01 0.9737    0.02630  0.01444
## 132 1937-12-01 0.9738    0.02620  0.01444
## 574 1974-10-01 1.0255    0.02549  0.01444
## 639 1980-03-01 1.0252    0.02516  0.01444
## 129 1937-09-01 1.0251    0.02513  0.01444
## 161 1940-05-01 1.0229    0.02288  0.01444
## 647 1980-11-01 1.0228    0.02280  0.01444
## 110 1936-02-01 0.9776    0.02238  0.01444
## 889 2001-01-01 1.0215    0.02151  0.01444
## 104 1935-08-01 0.9788    0.02119  0.01444
## 53 1931-05-01 1.0210    0.02096  0.01444

```

## 65	1932-05-01	1.0203	0.02030	0.01444
## 880	2000-04-01	1.0201	0.02013	0.01444
## 99	1935-03-01	0.9801	0.01990	0.01444
## 283	1950-07-01	1.0199	0.01987	0.01444
## 525	1970-09-01	1.0195	0.01954	0.01444
## 589	1976-01-01	1.0193	0.01926	0.01444
## 91	1934-07-01	1.0191	0.01913	0.01444
## 83	1933-11-01	1.0189	0.01891	0.01444
## 72	1932-12-01	1.0187	0.01867	0.01444
## 82	1933-10-01	1.0186	0.01855	0.01444
## 85	1934-01-01	0.9818	0.01817	0.01444
## 922	2003-10-01	0.9819	0.01809	0.01444
## 64	1932-04-01	1.0179	0.01787	0.01444
## 780	1991-12-01	1.0173	0.01728	0.01444
## 899	2001-11-01	0.9828	0.01716	0.01444
## 563	1973-11-01	1.0170	0.01698	0.01444
## 564	1973-12-01	1.0169	0.01688	0.01444
## 112	1936-04-01	1.0168	0.01676	0.01444
## 78	1933-06-01	1.0166	0.01665	0.01444
## 982	2008-10-01	0.9838	0.01622	0.01444
## 42	1930-06-01	1.0157	0.01575	0.01444
## 874	1999-10-01	0.9844	0.01558	0.01444
## 565	1974-01-01	1.0155	0.01554	0.01444
## 989	2009-05-01	1.0150	0.01502	0.01444
## 881	2000-05-01	1.0150	0.01501	0.01444
## 61	1932-01-01	1.0150	0.01497	0.01444
## 910	2002-10-01	0.9852	0.01482	0.01444
## 48	1930-12-01	1.0146	0.01465	0.01444
## 80	1933-08-01	0.9854	0.01459	0.01444
## 649	1981-01-01	1.0146	0.01458	0.01444

D.11 (AEW) Portfolio 6

```

## $`Cook's Distance`
## [1] "None"
##
## $`Covariance Ratio`
##           Month    CVRi abs.CVRi_1 calpt.crit
## 77  1933-05-01 1.1625    0.16250   0.01444
## 138 1938-06-01 0.8803    0.11971   0.01444
## 75  1933-03-01 0.8900    0.11004   0.01444
## 16   1928-04-01 0.9090    0.09103   0.01444
## 882 2000-06-01 0.9112    0.08875   0.01444
## 67   1932-07-01 1.0872    0.08723   0.01444
## 76   1933-04-01 0.9285    0.07153   0.01444
## 80   1933-08-01 0.9289    0.07105   0.01444
## 103  1935-07-01 0.9324    0.06760   0.01444
## 153  1939-09-01 0.9365    0.06354   0.01444
## 876  1999-12-01 0.9394    0.06061   0.01444
## 68   1932-08-01 1.0582    0.05818   0.01444
## 988  2009-04-01 1.0577    0.05769   0.01444
## 889  2001-01-01 1.0541    0.05411   0.01444
## 877  2000-01-01 0.9482    0.05182   0.01444
## 69   1932-09-01 0.9494    0.05064   0.01444
## 104  1935-08-01 0.9498    0.05016   0.01444
## 81   1933-09-01 0.9533    0.04667   0.01444
## 78   1933-06-01 0.9573    0.04272   0.01444
## 867  1999-03-01 0.9676    0.03237   0.01444
## 647  1980-11-01 1.0283    0.02828   0.01444
## 985  2009-01-01 0.9731    0.02692   0.01444
## 57   1931-09-01 1.0267    0.02667   0.01444
## 65   1932-05-01 1.0265    0.02650   0.01444
## 63   1932-03-01 0.9739    0.02611   0.01444
## 887  2000-11-01 1.0254    0.02542   0.01444
## 129  1937-09-01 1.0249    0.02493   0.01444
## 639  1980-03-01 1.0249    0.02490   0.01444
## 161  1940-05-01 1.0247    0.02469   0.01444
## 559  1973-07-01 1.0243    0.02434   0.01444
## 668  1982-08-01 0.9757    0.02431   0.01444
## 34   1929-10-01 1.0237    0.02373   0.01444
## 622  1978-10-01 1.0228    0.02280   0.01444
## 72   1932-12-01 1.0221    0.02208   0.01444

```

## 878	2000-02-01	0.9782	0.02182	0.01444
## 121	1937-01-01	0.9783	0.02171	0.01444
## 898	2001-10-01	1.0209	0.02086	0.01444
## 32	1929-08-01	1.0203	0.02033	0.01444
## 525	1970-09-01	1.0202	0.02021	0.01444
## 91	1934-07-01	1.0194	0.01944	0.01444
## 885	2000-09-01	0.9806	0.01936	0.01444
## 563	1973-11-01	1.0193	0.01932	0.01444
## 83	1933-11-01	1.0193	0.01932	0.01444
## 89	1934-05-01	0.9808	0.01915	0.01444
## 82	1933-10-01	1.0186	0.01863	0.01444
## 860	1998-08-01	1.0183	0.01833	0.01444
## 71	1932-11-01	1.0181	0.01809	0.01444
## 283	1950-07-01	1.0181	0.01806	0.01444
## 910	2002-10-01	1.0180	0.01802	0.01444
## 899	2001-11-01	0.9822	0.01777	0.01444
## 194	1943-02-01	0.9823	0.01774	0.01444
## 105	1935-09-01	0.9825	0.01752	0.01444
## 912	2002-12-01	0.9829	0.01710	0.01444
## 135	1938-03-01	1.0170	0.01700	0.01444
## 875	1999-11-01	1.0166	0.01665	0.01444
## 20	1928-08-01	0.9835	0.01648	0.01444
## 102	1935-06-01	0.9835	0.01647	0.01444
## 601	1977-01-01	0.9836	0.01638	0.01444
## 874	1999-10-01	1.0158	0.01585	0.01444
## 70	1932-10-01	1.0157	0.01573	0.01444
## 59	1931-11-01	1.0157	0.01571	0.01444
## 30	1929-06-01	1.0154	0.01540	0.01444
## 896	2001-08-01	0.9849	0.01508	0.01444
## 110	1936-02-01	0.9850	0.01502	0.01444
## 649	1981-01-01	1.0150	0.01501	0.01444
## 895	2001-07-01	0.9851	0.01494	0.01444
## 721	1987-01-01	1.0149	0.01488	0.01444
## 193	1943-01-01	1.0148	0.01478	0.01444
## 136	1938-04-01	1.0148	0.01475	0.01444
## 730	1987-10-01	1.0147	0.01468	0.01444

D.12 (AVW) Portfolio 6

```

## $`Cook's Distance`
## [1] "None"
##
## $`Covariance Ratio`
##           Month    CVRi abs.CVRi_1 calpt.crit
## 67 1932-07-01 1.1774     0.17736   0.01444
## 68 1932-08-01 1.1713     0.17133   0.01444
## 138 1938-06-01 0.8593     0.14066   0.01444
## 77 1933-05-01 1.1405     0.14053   0.01444
## 876 1999-12-01 0.8979     0.10210   0.01444
## 153 1939-09-01 1.0883     0.08833   0.01444
## 479 1966-11-01 0.9253     0.07475   0.01444
## 889 2001-01-01 0.9292     0.07075   0.01444
## 885 2000-09-01 0.9304     0.06958   0.01444
## 878 2000-02-01 1.0673     0.06731   0.01444
## 81 1933-09-01 0.9331     0.06688   0.01444
## 561 1973-09-01 0.9395     0.06045   0.01444
## 76 1933-04-01 1.0593     0.05930   0.01444
## 897 2001-09-01 0.9460     0.05403   0.01444
## 988 2009-04-01 1.0533     0.05335   0.01444
## 892 2001-04-01 0.9573     0.04269   0.01444
## 670 1982-10-01 0.9586     0.04139   0.01444
## 79 1933-07-01 0.9605     0.03951   0.01444
## 85 1934-01-01 1.0378     0.03778   0.01444
## 135 1938-03-01 0.9636     0.03641   0.01444
## 57 1931-09-01 1.0355     0.03553   0.01444
## 574 1974-10-01 0.9677     0.03234   0.01444
## 884 2000-08-01 0.9680     0.03198   0.01444
## 287 1950-11-01 0.9698     0.03025   0.01444
## 911 2002-11-01 0.9707     0.02928   0.01444
## 54 1931-06-01 1.0278     0.02784   0.01444
## 80 1933-08-01 0.9730     0.02699   0.01444
## 133 1938-01-01 0.9745     0.02548   0.01444
## 730 1987-10-01 1.0255     0.02546   0.01444
## 639 1980-03-01 1.0253     0.02527   0.01444
## 887 2000-11-01 1.0251     0.02508   0.01444
## 161 1940-05-01 1.0247     0.02472   0.01444
## 64 1932-04-01 0.9767     0.02328   0.01444
## 867 1999-03-01 0.9771     0.02292   0.01444

```

## 129	1937-09-01	1.0224	0.02241	0.01444
## 40	1930-04-01	0.9777	0.02227	0.01444
## 870	1999-06-01	0.9782	0.02184	0.01444
## 624	1978-12-01	0.9783	0.02165	0.01444
## 898	2001-10-01	1.0215	0.02155	0.01444
## 53	1931-05-01	1.0211	0.02115	0.01444
## 880	2000-04-01	1.0210	0.02095	0.01444
## 868	1999-04-01	0.9791	0.02092	0.01444
## 890	2001-02-01	1.0206	0.02064	0.01444
## 283	1950-07-01	1.0206	0.02055	0.01444
## 398	1960-02-01	0.9797	0.02027	0.01444
## 881	2000-05-01	0.9798	0.02021	0.01444
## 589	1976-01-01	1.0201	0.02009	0.01444
## 525	1970-09-01	1.0198	0.01985	0.01444
## 147	1939-03-01	1.0198	0.01984	0.01444
## 562	1973-10-01	0.9805	0.01945	0.01444
## 32	1929-08-01	1.0190	0.01904	0.01444
## 83	1933-11-01	1.0186	0.01862	0.01444
## 877	2000-01-01	0.9814	0.01860	0.01444
## 65	1932-05-01	1.0185	0.01851	0.01444
## 976	2008-04-01	0.9815	0.01848	0.01444
## 985	2009-01-01	1.0185	0.01847	0.01444
## 60	1931-12-01	1.0180	0.01797	0.01444
## 82	1933-10-01	1.0180	0.01795	0.01444
## 50	1931-02-01	1.0175	0.01752	0.01444
## 904	2002-04-01	0.9825	0.01747	0.01444
## 564	1973-12-01	1.0170	0.01701	0.01444
## 780	1991-12-01	1.0169	0.01692	0.01444
## 520	1970-04-01	1.0169	0.01688	0.01444
## 860	1998-08-01	1.0166	0.01660	0.01444
## 49	1931-01-01	0.9834	0.01658	0.01444
## 109	1936-01-01	1.0165	0.01645	0.01444
## 193	1943-01-01	1.0163	0.01633	0.01444
## 72	1932-12-01	1.0162	0.01625	0.01444
## 181	1942-01-01	1.0162	0.01622	0.01444
## 891	2001-03-01	1.0162	0.01621	0.01444
## 559	1973-07-01	0.9843	0.01574	0.01444
## 137	1938-05-01	0.9845	0.01551	0.01444
## 989	2009-05-01	1.0155	0.01547	0.01444
## 136	1938-04-01	0.9847	0.01534	0.01444
## 649	1981-01-01	1.0151	0.01511	0.01444

```
## 577 1975-01-01 1.0147    0.01469   0.01444
## 982 2008-10-01 1.0147    0.01468   0.01444
## 35  1929-11-01 0.9854    0.01462   0.01444
## 48  1930-12-01 1.0145    0.01452   0.01444
## 368 1957-08-01 0.9855    0.01449   0.01444
```

D.13 (AEW) Portfolio 7

```

## $`Cook's Distance`
##      Month   Ci
## 77 1933-05-01 3.747
## 67 1932-07-01 1.702
##
## $`Covariance Ratio`
##      Month CVRi abs.CVRi_1 calpt.crit
## 77 1933-05-01 0.6659    0.33412  0.01444
## 882 2000-06-01 0.8308    0.16916  0.01444
## 68 1932-08-01 1.1549    0.15489  0.01444
## 876 1999-12-01 0.8992    0.10082  0.01444
## 58 1931-10-01 0.9338    0.06623  0.01444
## 988 2009-04-01 1.0535    0.05347  0.01444
## 51 1931-03-01 0.9511    0.04885  0.01444
## 138 1938-06-01 0.9524    0.04755  0.01444
## 108 1935-12-01 0.9577    0.04230  0.01444
## 76 1933-04-01 1.0416    0.04159  0.01444
## 67 1932-07-01 0.9589    0.04105  0.01444
## 991 2009-07-01 0.9595    0.04053  0.01444
## 985 2009-01-01 0.9613    0.03872  0.01444
## 879 2000-03-01 0.9617    0.03828  0.01444
## 40 1930-04-01 0.9622    0.03776  0.01444
## 104 1935-08-01 0.9665    0.03350  0.01444
## 881 2000-05-01 0.9674    0.03257  0.01444
## 730 1987-10-01 1.0324    0.03241  0.01444
## 878 2000-02-01 0.9680    0.03205  0.01444
## 85 1934-01-01 1.0318    0.03179  0.01444
## 16 1928-04-01 0.9683    0.03166  0.01444
## 890 2001-02-01 1.0282    0.02824  0.01444
## 647 1980-11-01 1.0282    0.02823  0.01444
## 880 2000-04-01 0.9724    0.02760  0.01444
## 57 1931-09-01 1.0268    0.02685  0.01444
## 574 1974-10-01 1.0255    0.02553  0.01444
## 887 2000-11-01 1.0255    0.02547  0.01444
## 129 1937-09-01 1.0249    0.02485  0.01444
## 161 1940-05-01 1.0241    0.02410  0.01444
## 71 1932-11-01 1.0239    0.02391  0.01444
## 118 1936-10-01 0.9762    0.02382  0.01444
## 867 1999-03-01 0.9764    0.02360  0.01444

```

##	639	1980-03-01	1.0229	0.02290	0.01444
##	911	2002-11-01	1.0229	0.02289	0.01444
##	107	1935-11-01	0.9773	0.02272	0.01444
##	34	1929-10-01	1.0227	0.02270	0.01444
##	542	1972-02-01	0.9776	0.02237	0.01444
##	72	1932-12-01	1.0218	0.02175	0.01444
##	283	1950-07-01	1.0210	0.02100	0.01444
##	622	1978-10-01	1.0209	0.02089	0.01444
##	32	1929-08-01	0.9792	0.02081	0.01444
##	565	1974-01-01	1.0204	0.02042	0.01444
##	559	1973-07-01	1.0200	0.02002	0.01444
##	81	1933-09-01	1.0198	0.01978	0.01444
##	372	1957-12-01	0.9813	0.01868	0.01444
##	860	1998-08-01	1.0185	0.01851	0.01444
##	905	2002-05-01	0.9816	0.01837	0.01444
##	82	1933-10-01	1.0182	0.01819	0.01444
##	563	1973-11-01	1.0182	0.01817	0.01444
##	910	2002-10-01	1.0176	0.01765	0.01444
##	907	2002-07-01	0.9824	0.01760	0.01444
##	781	1992-01-01	0.9826	0.01741	0.01444
##	147	1939-03-01	1.0165	0.01655	0.01444
##	982	2008-10-01	1.0165	0.01651	0.01444
##	50	1931-02-01	1.0165	0.01649	0.01444
##	898	2001-10-01	1.0164	0.01643	0.01444
##	59	1931-11-01	1.0164	0.01640	0.01444
##	36	1929-12-01	0.9841	0.01595	0.01444
##	525	1970-09-01	1.0159	0.01586	0.01444
##	520	1970-04-01	1.0158	0.01582	0.01444
##	136	1938-04-01	1.0156	0.01562	0.01444
##	60	1931-12-01	1.0156	0.01557	0.01444
##	848	1997-08-01	0.9844	0.01557	0.01444
##	35	1929-11-01	1.0156	0.01556	0.01444
##	573	1974-09-01	1.0154	0.01543	0.01444
##	54	1931-06-01	1.0150	0.01504	0.01444
##	80	1933-08-01	0.9855	0.01449	0.01444

D.14 (AVW) Portfolio 7

```

## $`Cook's Distance`
## [1] "None"
##
## $`Covariance Ratio`
##           Month    CVRi abs.CVRi_1 calpt.crit
## 876 1999-12-01 0.7974     0.20262   0.01444
## 67  1932-07-01 1.1763     0.17632   0.01444
## 68  1932-08-01 1.1708     0.17079   0.01444
## 77  1933-05-01 1.1357     0.13566   0.01444
## 138 1938-06-01 0.8671     0.13288   0.01444
## 34  1929-10-01 0.8932     0.10675   0.01444
## 153 1939-09-01 1.0871     0.08706   0.01444
## 36  1929-12-01 0.9171     0.08286   0.01444
## 343 1955-07-01 0.9205     0.07948   0.01444
## 139 1938-07-01 0.9260     0.07395   0.01444
## 76  1933-04-01 1.0654     0.06540   0.01444
## 880 2000-04-01 0.9363     0.06368   0.01444
## 988 2009-04-01 1.0602     0.06019   0.01444
## 889 2001-01-01 1.0574     0.05737   0.01444
## 983 2008-11-01 0.9467     0.05331   0.01444
## 881 2000-05-01 0.9516     0.04837   0.01444
## 75  1933-03-01 0.9527     0.04734   0.01444
## 64  1932-04-01 0.9534     0.04656   0.01444
## 56  1931-08-01 0.9560     0.04404   0.01444
## 981 2008-09-01 0.9569     0.04308   0.01444
## 885 2000-09-01 0.9579     0.04210   0.01444
## 938 2005-02-01 0.9582     0.04179   0.01444
## 577 1975-01-01 0.9593     0.04070   0.01444
## 23  1928-11-01 0.9600     0.04002   0.01444
## 85  1934-01-01 1.0356     0.03555   0.01444
## 882 2000-06-01 0.9657     0.03425   0.01444
## 888 2000-12-01 0.9660     0.03401   0.01444
## 80  1933-08-01 0.9685     0.03146   0.01444
## 884 2000-08-01 0.9706     0.02942   0.01444
## 115 1936-07-01 0.9706     0.02941   0.01444
## 730 1987-10-01 1.0291     0.02911   0.01444
## 111 1936-03-01 0.9714     0.02856   0.01444
## 647 1980-11-01 1.0250     0.02503   0.01444
## 890 2001-02-01 1.0244     0.02437   0.01444

```

```

## 81 1933-09-01 1.0241 0.02412 0.01444
## 161 1940-05-01 1.0241 0.02406 0.01444
## 879 2000-03-01 1.0238 0.02377 0.01444
## 639 1980-03-01 1.0230 0.02297 0.01444
## 72 1932-12-01 1.0221 0.02214 0.01444
## 87 1934-03-01 0.9779 0.02211 0.01444
## 65 1932-05-01 1.0221 0.02209 0.01444
## 898 2001-10-01 1.0219 0.02189 0.01444
## 525 1970-09-01 1.0218 0.02179 0.01444
## 216 1944-12-01 0.9784 0.02156 0.01444
## 565 1974-01-01 1.0204 0.02041 0.01444
## 976 2008-04-01 0.9797 0.02026 0.01444
## 860 1998-08-01 1.0202 0.02023 0.01444
## 589 1976-01-01 1.0199 0.01994 0.01444
## 53 1931-05-01 1.0199 0.01990 0.01444
## 911 2002-11-01 1.0197 0.01972 0.01444
## 887 2000-11-01 0.9806 0.01941 0.01444
## 283 1950-07-01 1.0194 0.01936 0.01444
## 129 1937-09-01 1.0193 0.01925 0.01444
## 70 1932-10-01 1.0189 0.01887 0.01444
## 563 1973-11-01 1.0185 0.01849 0.01444
## 60 1931-12-01 1.0182 0.01822 0.01444
## 54 1931-06-01 1.0177 0.01766 0.01444
## 107 1935-11-01 1.0172 0.01723 0.01444
## 78 1933-06-01 1.0170 0.01703 0.01444
## 83 1933-11-01 1.0167 0.01674 0.01444
## 520 1970-04-01 1.0165 0.01645 0.01444
## 193 1943-01-01 1.0162 0.01624 0.01444
## 891 2001-03-01 1.0161 0.01608 0.01444
## 697 1985-01-01 1.0156 0.01558 0.01444
## 574 1974-10-01 1.0155 0.01547 0.01444
## 59 1931-11-01 1.0153 0.01531 0.01444
## 32 1929-08-01 1.0151 0.01514 0.01444
## 175 1941-07-01 0.9851 0.01493 0.01444
## 42 1930-06-01 1.0149 0.01486 0.01444
## 147 1939-03-01 0.9852 0.01480 0.01444
## 721 1987-01-01 1.0148 0.01478 0.01444
## 424 1962-04-01 0.9853 0.01473 0.01444
## 30 1929-06-01 1.0146 0.01463 0.01444
## 89 1934-05-01 1.0145 0.01449 0.01444

```

D.15 (AEW) Portfolio 8

```

## $`Cook's Distance`
##           Month   Ci
## 77 1933-05-01 1.663
## 153 1939-09-01 1.465
## 76 1933-04-01 1.408
## 68 1932-08-01 1.122
##
## $`Covariance Ratio`
##           Month CVRi abs.CVRi_1 calpt.crit
## 76 1933-04-01 0.6191    0.38093   0.01444
## 153 1939-09-01 0.7125    0.28753   0.01444
## 67 1932-07-01 1.1867    0.18672   0.01444
## 882 2000-06-01 0.8532    0.14679   0.01444
## 879 2000-03-01 0.9080    0.09201   0.01444
## 77 1933-05-01 0.9202    0.07977   0.01444
## 878 2000-02-01 1.0592    0.05919   0.01444
## 985 2009-01-01 0.9499    0.05008   0.01444
## 880 2000-04-01 0.9563    0.04367   0.01444
## 890 2001-02-01 1.0417    0.04171   0.01444
## 781 1992-01-01 0.9611    0.03893   0.01444
## 730 1987-10-01 1.0361    0.03605   0.01444
## 881 2000-05-01 0.9658    0.03420   0.01444
## 57 1931-09-01 1.0332    0.03318   0.01444
## 647 1980-11-01 1.0270    0.02697   0.01444
## 85 1934-01-01 1.0262    0.02621   0.01444
## 94 1934-10-01 0.9741    0.02590   0.01444
## 887 2000-11-01 1.0254    0.02539   0.01444
## 34 1929-10-01 1.0250    0.02502   0.01444
## 161 1940-05-01 1.0248    0.02483   0.01444
## 559 1973-07-01 1.0244    0.02443   0.01444
## 639 1980-03-01 1.0239    0.02387   0.01444
## 81 1933-09-01 1.0237    0.02370   0.01444
## 911 2002-11-01 1.0230    0.02297   0.01444
## 622 1978-10-01 1.0230    0.02297   0.01444
## 135 1938-03-01 1.0229    0.02287   0.01444
## 129 1937-09-01 1.0216    0.02158   0.01444
## 898 2001-10-01 1.0208    0.02085   0.01444
## 71 1932-11-01 0.9797    0.02030   0.01444
## 860 1998-08-01 1.0202    0.02023   0.01444

```

## 589	1976-01-01	1.0198	0.01982	0.01444
## 110	1936-02-01	0.9807	0.01929	0.01444
## 988	2009-04-01	1.0193	0.01928	0.01444
## 563	1973-11-01	1.0192	0.01918	0.01444
## 112	1936-04-01	1.0188	0.01882	0.01444
## 72	1932-12-01	1.0188	0.01881	0.01444
## 16	1928-04-01	0.9813	0.01866	0.01444
## 54	1931-06-01	1.0185	0.01849	0.01444
## 70	1932-10-01	1.0182	0.01818	0.01444
## 910	2002-10-01	1.0180	0.01804	0.01444
## 525	1970-09-01	1.0178	0.01783	0.01444
## 50	1931-02-01	1.0174	0.01743	0.01444
## 780	1991-12-01	0.9828	0.01720	0.01444
## 181	1942-01-01	1.0172	0.01718	0.01444
## 565	1974-01-01	1.0172	0.01717	0.01444
## 61	1932-01-01	1.0170	0.01697	0.01444
## 577	1975-01-01	1.0168	0.01685	0.01444
## 283	1950-07-01	1.0167	0.01667	0.01444
## 122	1937-02-01	0.9838	0.01619	0.01444
## 107	1935-11-01	1.0161	0.01610	0.01444
## 60	1931-12-01	1.0159	0.01588	0.01444
## 891	2001-03-01	1.0159	0.01587	0.01444
## 564	1973-12-01	1.0157	0.01570	0.01444
## 697	1985-01-01	1.0150	0.01495	0.01444
## 649	1981-01-01	1.0149	0.01493	0.01444
## 721	1987-01-01	1.0148	0.01483	0.01444

D.16 (AVW) Portfolio 8

```

## $`Cook's Distance`
## [1] "None"
##
## $`Covariance Ratio`
##           Month    CVRi abs.CVRi_1 calpt.crit
## 68 1932-08-01 1.1716    0.17162   0.01444
## 77 1933-05-01 1.1660    0.16602   0.01444
## 83 1933-11-01 0.8470    0.15302   0.01444
## 67 1932-07-01 1.1488    0.14883   0.01444
## 138 1938-06-01 0.8616    0.13841   0.01444
## 868 1999-04-01 0.8965    0.10350   0.01444
## 153 1939-09-01 1.0890    0.08899   0.01444
## 76 1933-04-01 0.9224    0.07764   0.01444
## 876 1999-12-01 0.9320    0.06804   0.01444
## 889 2001-01-01 1.0569    0.05689   0.01444
## 988 2009-04-01 1.0568    0.05680   0.01444
## 84 1933-12-01 0.9495    0.05046   0.01444
## 162 1940-06-01 0.9568    0.04315   0.01444
## 561 1973-09-01 0.9573    0.04268   0.01444
## 490 1967-10-01 0.9595    0.04053   0.01444
## 389 1959-05-01 0.9598    0.04018   0.01444
## 57 1931-09-01 1.0399    0.03995   0.01444
## 878 2000-02-01 0.9637    0.03631   0.01444
## 880 2000-04-01 0.9655    0.03451   0.01444
## 879 2000-03-01 1.0319    0.03190   0.01444
## 54 1931-06-01 1.0318    0.03182   0.01444
## 227 1945-11-01 0.9704    0.02961   0.01444
## 137 1938-05-01 0.9730    0.02699   0.01444
## 870 1999-06-01 0.9731    0.02693   0.01444
## 66 1932-06-01 0.9732    0.02684   0.01444
## 884 2000-08-01 0.9745    0.02553   0.01444
## 71 1932-11-01 1.0254    0.02538   0.01444
## 85 1934-01-01 1.0252    0.02520   0.01444
## 75 1933-03-01 0.9749    0.02509   0.01444
## 161 1940-05-01 1.0248    0.02483   0.01444
## 185 1942-05-01 0.9759    0.02409   0.01444
## 204 1943-12-01 0.9760    0.02399   0.01444
## 65 1932-05-01 1.0236    0.02364   0.01444
## 885 2000-09-01 0.9768    0.02321   0.01444

```

## 730	1987-10-01	1.0232	0.02321	0.01444
## 147	1939-03-01	0.9770	0.02304	0.01444
## 999	2010-03-01	0.9772	0.02281	0.01444
## 898	2001-10-01	1.0225	0.02249	0.01444
## 25	1929-01-01	0.9780	0.02198	0.01444
## 639	1980-03-01	1.0219	0.02189	0.01444
## 559	1973-07-01	1.0217	0.02169	0.01444
## 81	1933-09-01	1.0216	0.02158	0.01444
## 343	1955-07-01	0.9793	0.02072	0.01444
## 72	1932-12-01	1.0206	0.02056	0.01444
## 34	1929-10-01	1.0204	0.02044	0.01444
## 443	1963-11-01	0.9799	0.02010	0.01444
## 860	1998-08-01	1.0200	0.02004	0.01444
## 495	1968-03-01	0.9800	0.01997	0.01444
## 525	1970-09-01	1.0199	0.01988	0.01444
## 589	1976-01-01	1.0198	0.01983	0.01444
## 123	1937-03-01	0.9807	0.01934	0.01444
## 78	1933-06-01	0.9813	0.01873	0.01444
## 53	1931-05-01	1.0186	0.01858	0.01444
## 82	1933-10-01	1.0184	0.01841	0.01444
## 577	1975-01-01	1.0183	0.01826	0.01444
## 61	1932-01-01	1.0181	0.01812	0.01444
## 882	2000-06-01	1.0180	0.01800	0.01444
## 890	2001-02-01	1.0177	0.01773	0.01444
## 887	2000-11-01	1.0173	0.01730	0.01444
## 112	1936-04-01	1.0165	0.01645	0.01444
## 911	2002-11-01	1.0165	0.01645	0.01444
## 897	2001-09-01	1.0163	0.01631	0.01444
## 520	1970-04-01	1.0162	0.01624	0.01444
## 904	2002-04-01	0.9838	0.01622	0.01444
## 60	1931-12-01	1.0161	0.01612	0.01444
## 50	1931-02-01	1.0159	0.01592	0.01444
## 874	1999-10-01	1.0159	0.01585	0.01444
## 107	1935-11-01	1.0156	0.01561	0.01444
## 697	1985-01-01	1.0154	0.01543	0.01444
## 565	1974-01-01	1.0154	0.01542	0.01444
## 649	1981-01-01	1.0151	0.01508	0.01444
## 780	1991-12-01	1.0149	0.01493	0.01444
## 891	2001-03-01	1.0149	0.01491	0.01444
## 721	1987-01-01	1.0147	0.01472	0.01444
## 134	1938-02-01	0.9853	0.01471	0.01444

```
## 985 2009-01-01 1.0146      0.01460     0.01444
## 781 1992-01-01 1.0145      0.01454     0.01444
```

D.17 (AEW) Portfolio 9

```

## $`Cook's Distance`
##      Month   Ci
## 77 1933-05-01 2.223
## 67 1932-07-01 1.016
##
## $`Covariance Ratio`
##      Month CVRi abs.CVRi_1 calpt.crit
## 193 1943-01-01 0.8168    0.18316  0.01444
## 85  1934-01-01 0.8266    0.17338  0.01444
## 68  1932-08-01 1.1711    0.17111  0.01444
## 77  1933-05-01 0.8453    0.15468  0.01444
## 877 2000-01-01 0.8634    0.13665  0.01444
## 78  1933-06-01 0.8823    0.11769  0.01444
## 882 2000-06-01 0.8872    0.11281  0.01444
## 879 2000-03-01 0.8928    0.10716  0.01444
## 781 1992-01-01 0.9161    0.08394  0.01444
## 878 2000-02-01 1.0808    0.08083  0.01444
## 58  1931-10-01 0.9354    0.06456  0.01444
## 76  1933-04-01 1.0643    0.06429  0.01444
## 67  1932-07-01 1.0496    0.04958  0.01444
## 65  1932-05-01 0.9523    0.04774  0.01444
## 66  1932-06-01 0.9543    0.04572  0.01444
## 138 1938-06-01 1.0428    0.04280  0.01444
## 890 2001-02-01 1.0418    0.04182  0.01444
## 988 2009-04-01 1.0407    0.04070  0.01444
## 889 2001-01-01 0.9603    0.03969  0.01444
## 730 1987-10-01 1.0396    0.03962  0.01444
## 780 1991-12-01 0.9615    0.03850  0.01444
## 876 1999-12-01 0.9648    0.03522  0.01444
## 57  1931-09-01 1.0337    0.03373  0.01444
## 881 2000-05-01 0.9670    0.03303  0.01444
## 647 1980-11-01 1.0271    0.02714  0.01444
## 109 1936-01-01 0.9732    0.02682  0.01444
## 135 1938-03-01 1.0267    0.02671  0.01444
## 153 1939-09-01 0.9742    0.02580  0.01444
## 887 2000-11-01 1.0257    0.02568  0.01444
## 9   1927-09-01 0.9748    0.02518  0.01444
## 489 1967-09-01 0.9757    0.02427  0.01444
## 639 1980-03-01 1.0239    0.02394  0.01444

```

## 81	1933-09-01	1.0238	0.02377	0.01444
## 54	1931-06-01	1.0235	0.02351	0.01444
## 989	2009-05-01	0.9773	0.02274	0.01444
## 898	2001-10-01	1.0227	0.02269	0.01444
## 161	1940-05-01	1.0224	0.02239	0.01444
## 911	2002-11-01	1.0223	0.02227	0.01444
## 80	1933-08-01	1.0215	0.02149	0.01444
## 525	1970-09-01	1.0210	0.02098	0.01444
## 858	1998-06-01	0.9793	0.02071	0.01444
## 71	1932-11-01	1.0202	0.02019	0.01444
## 677	1983-05-01	0.9798	0.02017	0.01444
## 283	1950-07-01	1.0198	0.01977	0.01444
## 860	1998-08-01	1.0196	0.01965	0.01444
## 565	1974-01-01	1.0196	0.01962	0.01444
## 147	1939-03-01	1.0181	0.01806	0.01444
## 64	1932-04-01	1.0180	0.01802	0.01444
## 82	1933-10-01	1.0180	0.01801	0.01444
## 107	1935-11-01	1.0177	0.01775	0.01444
## 112	1936-04-01	1.0177	0.01768	0.01444
## 875	1999-11-01	1.0177	0.01768	0.01444
## 129	1937-09-01	1.0175	0.01753	0.01444
## 559	1973-07-01	1.0168	0.01678	0.01444
## 42	1930-06-01	1.0168	0.01676	0.01444
## 904	2002-04-01	1.0166	0.01660	0.01444
## 891	2001-03-01	1.0161	0.01612	0.01444
## 59	1931-11-01	1.0156	0.01560	0.01444
## 910	2002-10-01	1.0155	0.01555	0.01444
## 892	2001-04-01	1.0154	0.01538	0.01444
## 50	1931-02-01	1.0150	0.01503	0.01444
## 697	1985-01-01	1.0150	0.01499	0.01444
## 622	1978-10-01	1.0147	0.01468	0.01444
## 98	1935-02-01	1.0145	0.01454	0.01444
## 909	2002-09-01	1.0145	0.01453	0.01444

D.18 (AVW) Portfolio 9

```

## $`Cook's Distance`
## [1] "None"
##
## $`Covariance Ratio`
##           Month    CVRi abs.CVRi_1 calpt.crit
## 67 1932-07-01 1.1862     0.18625   0.01444
## 68 1932-08-01 1.1719     0.17186   0.01444
## 77 1933-05-01 1.1692     0.16919   0.01444
## 878 2000-02-01 0.8389     0.16115   0.01444
## 15 1928-03-01 0.8749     0.12507   0.01444
## 879 2000-03-01 0.8928     0.10720   0.01444
## 153 1939-09-01 1.0892     0.08921   0.01444
## 23 1928-11-01 0.9139     0.08610   0.01444
## 876 1999-12-01 0.9226     0.07739   0.01444
## 80 1933-08-01 0.9321     0.06791   0.01444
## 889 2001-01-01 1.0569     0.05690   0.01444
## 647 1980-11-01 0.9478     0.05220   0.01444
## 138 1938-06-01 1.0478     0.04783   0.01444
## 882 2000-06-01 0.9577     0.04225   0.01444
## 890 2001-02-01 1.0414     0.04138   0.01444
## 730 1987-10-01 1.0403     0.04031   0.01444
## 57 1931-09-01 1.0378     0.03776   0.01444
## 30 1929-06-01 0.9650     0.03496   0.01444
## 877 2000-01-01 0.9653     0.03470   0.01444
## 76 1933-04-01 1.0343     0.03434   0.01444
## 1033 2013-01-01 0.9666     0.03340   0.01444
## 54 1931-06-01 1.0297     0.02969   0.01444
## 442 1963-10-01 0.9717     0.02829   0.01444
## 280 1950-04-01 0.9718     0.02816   0.01444
## 222 1945-06-01 0.9724     0.02765   0.01444
## 66 1932-06-01 0.9728     0.02715   0.01444
## 574 1974-10-01 1.0260     0.02602   0.01444
## 988 2009-04-01 1.0257     0.02575   0.01444
## 985 2009-01-01 0.9748     0.02524   0.01444
## 887 2000-11-01 1.0252     0.02517   0.01444
## 639 1980-03-01 1.0247     0.02471   0.01444
## 638 1980-02-01 0.9753     0.02471   0.01444
## 81 1933-09-01 1.0241     0.02408   0.01444
## 525 1970-09-01 1.0223     0.02230   0.01444

```

## 227	1945-11-01	0.9777	0.02229	0.01444
## 479	1966-11-01	0.9778	0.02221	0.01444
## 124	1937-04-01	0.9786	0.02143	0.01444
## 288	1950-12-01	0.9787	0.02132	0.01444
## 53	1931-05-01	1.0211	0.02112	0.01444
## 680	1983-08-01	0.9793	0.02066	0.01444
## 115	1936-07-01	0.9801	0.01989	0.01444
## 129	1937-09-01	1.0198	0.01981	0.01444
## 65	1932-05-01	1.0197	0.01975	0.01444
## 118	1936-10-01	0.9803	0.01968	0.01444
## 72	1932-12-01	1.0195	0.01951	0.01444
## 676	1983-04-01	0.9806	0.01940	0.01444
## 563	1973-11-01	1.0194	0.01939	0.01444
## 622	1978-10-01	0.9806	0.01936	0.01444
## 898	2001-10-01	1.0189	0.01892	0.01444
## 865	1999-01-01	0.9815	0.01853	0.01444
## 112	1936-04-01	1.0185	0.01849	0.01444
## 70	1932-10-01	1.0183	0.01827	0.01444
## 910	2002-10-01	1.0178	0.01780	0.01444
## 60	1931-12-01	1.0176	0.01758	0.01444
## 884	2000-08-01	0.9827	0.01734	0.01444
## 780	1991-12-01	1.0172	0.01719	0.01444
## 37	1930-01-01	0.9828	0.01719	0.01444
## 34	1929-10-01	1.0171	0.01706	0.01444
## 147	1939-03-01	1.0168	0.01683	0.01444
## 520	1970-04-01	1.0168	0.01680	0.01444
## 559	1973-07-01	1.0168	0.01678	0.01444
## 897	2001-09-01	1.0166	0.01663	0.01444
## 294	1951-06-01	0.9836	0.01637	0.01444
## 989	2009-05-01	1.0163	0.01635	0.01444
## 193	1943-01-01	1.0163	0.01630	0.01444
## 891	2001-03-01	1.0163	0.01628	0.01444
## 881	2000-05-01	0.9844	0.01561	0.01444
## 136	1938-04-01	1.0156	0.01558	0.01444
## 35	1929-11-01	1.0154	0.01541	0.01444
## 42	1930-06-01	1.0148	0.01481	0.01444
## 577	1975-01-01	1.0148	0.01476	0.01444
## 860	1998-08-01	1.0146	0.01461	0.01444
## 667	1982-07-01	0.9854	0.01458	0.01444
## 982	2008-10-01	1.0144	0.01444	0.01444

D.19 (AEW) Portfolio 10

```

## $`Cook's Distance`
## [1] "None"
##
## $`Covariance Ratio`
##           Month    CVRi abs.CVRi_1 calpt.crit
## 193 1943-01-01 0.7444     0.25565   0.01444
## 67  1932-07-01 1.1742     0.17419   0.01444
## 77  1933-05-01 1.1659     0.16594   0.01444
## 68  1932-08-01 1.1562     0.15617   0.01444
## 181 1942-01-01 0.8887     0.11126   0.01444
## 781 1992-01-01 0.8956     0.10445   0.01444
## 878 2000-02-01 1.0839     0.08388   0.01444
## 78  1933-06-01 0.9298     0.07015   0.01444
## 153 1939-09-01 0.9340     0.06602   0.01444
## 76  1933-04-01 1.0553     0.05532   0.01444
## 877 2000-01-01 0.9471     0.05287   0.01444
## 981 2008-09-01 0.9476     0.05238   0.01444
## 13  1928-01-01 0.9593     0.04071   0.01444
## 138 1938-06-01 0.9615     0.03852   0.01444
## 23  1928-11-01 0.9617     0.03833   0.01444
## 85  1934-01-01 1.0362     0.03619   0.01444
## 57  1931-09-01 1.0357     0.03572   0.01444
## 890 2001-02-01 1.0343     0.03426   0.01444
## 189 1942-09-01 0.9669     0.03306   0.01444
## 882 2000-06-01 1.0326     0.03263   0.01444
## 139 1938-07-01 0.9686     0.03139   0.01444
## 574 1974-10-01 1.0305     0.03048   0.01444
## 74  1933-02-01 0.9707     0.02935   0.01444
## 27  1929-03-01 0.9710     0.02898   0.01444
## 647 1980-11-01 1.0283     0.02831   0.01444
## 879 2000-03-01 0.9721     0.02786   0.01444
## 71  1932-11-01 1.0259     0.02586   0.01444
## 876 1999-12-01 1.0258     0.02582   0.01444
## 622 1978-10-01 1.0256     0.02559   0.01444
## 979 2008-07-01 0.9746     0.02539   0.01444
## 129 1937-09-01 1.0251     0.02510   0.01444
## 559 1973-07-01 1.0243     0.02425   0.01444
## 730 1987-10-01 1.0240     0.02395   0.01444
## 988 2009-04-01 1.0238     0.02378   0.01444

```

## 868	1999-04-01	0.9773	0.02269	0.01444
## 161	1940-05-01	1.0226	0.02262	0.01444
## 639	1980-03-01	1.0226	0.02255	0.01444
## 525	1970-09-01	1.0225	0.02251	0.01444
## 980	2008-08-01	0.9778	0.02219	0.01444
## 989	2009-05-01	0.9778	0.02215	0.01444
## 880	2000-04-01	1.0214	0.02141	0.01444
## 887	2000-11-01	0.9798	0.02021	0.01444
## 147	1939-03-01	1.0198	0.01984	0.01444
## 135	1938-03-01	0.9803	0.01970	0.01444
## 32	1929-08-01	1.0195	0.01950	0.01444
## 975	2008-03-01	0.9806	0.01940	0.01444
## 72	1932-12-01	1.0193	0.01930	0.01444
## 112	1936-04-01	1.0191	0.01906	0.01444
## 673	1983-01-01	0.9812	0.01879	0.01444
## 58	1931-10-01	0.9813	0.01866	0.01444
## 82	1933-10-01	1.0182	0.01817	0.01444
## 73	1933-01-01	0.9821	0.01788	0.01444
## 589	1976-01-01	1.0178	0.01780	0.01444
## 60	1931-12-01	1.0177	0.01772	0.01444
## 860	1998-08-01	1.0175	0.01746	0.01444
## 577	1975-01-01	1.0174	0.01745	0.01444
## 34	1929-10-01	0.9826	0.01745	0.01444
## 49	1931-01-01	0.9827	0.01732	0.01444
## 196	1943-04-01	0.9829	0.01714	0.01444
## 50	1931-02-01	1.0170	0.01701	0.01444
## 563	1973-11-01	1.0170	0.01699	0.01444
## 875	1999-11-01	0.9832	0.01683	0.01444
## 497	1968-05-01	0.9832	0.01682	0.01444
## 80	1933-08-01	1.0167	0.01669	0.01444
## 889	2001-01-01	0.9836	0.01642	0.01444
## 891	2001-03-01	1.0163	0.01631	0.01444
## 493	1968-01-01	0.9838	0.01618	0.01444
## 565	1974-01-01	1.0157	0.01573	0.01444
## 36	1929-12-01	0.9846	0.01543	0.01444
## 697	1985-01-01	1.0154	0.01541	0.01444
## 573	1974-09-01	1.0151	0.01511	0.01444
## 649	1981-01-01	1.0150	0.01497	0.01444
## 81	1933-09-01	1.0149	0.01494	0.01444
## 35	1929-11-01	1.0148	0.01485	0.01444
## 897	2001-09-01	1.0148	0.01485	0.01444

```
## 520 1970-04-01 1.0148    0.01483    0.01444
## 30  1929-06-01 1.0147    0.01473    0.01444
## 881 2000-05-01 1.0147    0.01470    0.01444
```

D.20 (AVW) Portfolio 10

```

## $`Cook's Distance`
##      Month   Ci
## 77 1933-05-01 1.513
##
## $`Covariance Ratio`
##      Month   CVRi abs.CVRi_1 calpt.crit
## 67 1932-07-01 1.1725    0.17255  0.01444
## 68 1932-08-01 1.1693    0.16935  0.01444
## 71 1932-11-01 0.8319    0.16808  0.01444
## 23 1928-11-01 0.8963    0.10366  0.01444
## 153 1939-09-01 1.0910    0.09096  0.01444
## 887 2000-11-01 0.9130    0.08696  0.01444
## 70 1932-10-01 0.9160    0.08404  0.01444
## 878 2000-02-01 1.0788    0.07883  0.01444
## 25 1929-01-01 0.9347    0.06535  0.01444
## 77 1933-05-01 0.9412    0.05880  0.01444
## 196 1943-04-01 0.9447    0.05531  0.01444
## 889 2001-01-01 1.0539    0.05387  0.01444
## 979 2008-07-01 0.9553    0.04469  0.01444
## 890 2001-02-01 1.0423    0.04234  0.01444
## 879 2000-03-01 1.0397    0.03975  0.01444
## 30 1929-06-01 0.9640    0.03605  0.01444
## 885 2000-09-01 0.9642    0.03584  0.01444
## 138 1938-06-01 1.0328    0.03277  0.01444
## 62 1932-02-01 0.9697    0.03032  0.01444
## 65 1932-05-01 1.0303    0.03030  0.01444
## 135 1938-03-01 1.0262    0.02618  0.01444
## 639 1980-03-01 1.0252    0.02524  0.01444
## 559 1973-07-01 1.0246    0.02456  0.01444
## 178 1941-10-01 0.9756    0.02441  0.01444
## 129 1937-09-01 1.0244    0.02437  0.01444
## 577 1975-01-01 1.0241    0.02410  0.01444
## 85 1934-01-01 1.0235    0.02352  0.01444
## 57 1931-09-01 1.0233    0.02333  0.01444
## 574 1974-10-01 1.0228    0.02280  0.01444
## 898 2001-10-01 1.0224    0.02239  0.01444
## 76 1933-04-01 1.0219    0.02186  0.01444
## 470 1966-02-01 0.9782    0.02182  0.01444
## 80 1933-08-01 1.0212    0.02124  0.01444

```

## 525	1970-09-01	1.0209	0.02095	0.01444
## 622	1978-10-01	1.0203	0.02030	0.01444
## 479	1966-11-01	0.9797	0.02025	0.01444
## 283	1950-07-01	1.0202	0.02019	0.01444
## 976	2008-04-01	0.9810	0.01900	0.01444
## 876	1999-12-01	1.0187	0.01871	0.01444
## 181	1942-01-01	1.0187	0.01866	0.01444
## 647	1980-11-01	1.0185	0.01854	0.01444
## 589	1976-01-01	1.0184	0.01837	0.01444
## 198	1943-06-01	0.9817	0.01828	0.01444
## 91	1934-07-01	1.0182	0.01821	0.01444
## 910	2002-10-01	1.0182	0.01815	0.01444
## 60	1931-12-01	1.0180	0.01805	0.01444
## 61	1932-01-01	0.9820	0.01805	0.01444
## 875	1999-11-01	1.0180	0.01798	0.01444
## 64	1932-04-01	0.9822	0.01775	0.01444
## 107	1935-11-01	1.0177	0.01765	0.01444
## 828	1995-12-01	0.9827	0.01729	0.01444
## 83	1933-11-01	0.9828	0.01715	0.01444
## 478	1966-10-01	0.9830	0.01697	0.01444
## 780	1991-12-01	1.0169	0.01685	0.01444
## 985	2009-01-01	1.0167	0.01671	0.01444
## 860	1998-08-01	1.0167	0.01668	0.01444
## 42	1930-06-01	1.0166	0.01660	0.01444
## 891	2001-03-01	1.0165	0.01648	0.01444
## 982	2008-10-01	1.0163	0.01634	0.01444
## 950	2006-02-01	0.9837	0.01631	0.01444
## 53	1931-05-01	1.0162	0.01620	0.01444
## 345	1955-09-01	0.9839	0.01612	0.01444
## 847	1997-07-01	0.9839	0.01610	0.01444
## 563	1973-11-01	1.0158	0.01583	0.01444
## 904	2002-04-01	1.0158	0.01583	0.01444
## 501	1968-09-01	0.9845	0.01554	0.01444
## 697	1985-01-01	1.0155	0.01548	0.01444
## 136	1938-04-01	1.0154	0.01537	0.01444
## 32	1929-08-01	1.0153	0.01527	0.01444
## 35	1929-11-01	1.0152	0.01525	0.01444
## 881	2000-05-01	1.0150	0.01499	0.01444
## 147	1939-03-01	1.0150	0.01497	0.01444
## 50	1931-02-01	1.0149	0.01493	0.01444
## 34	1929-10-01	0.9851	0.01486	0.01444

```
## 892 2001-04-01 1.0148    0.01482    0.01444
## 842 1997-02-01 0.9853    0.01472    0.01444
## 48   1930-12-01 1.0145    0.01447    0.01444
```

Appendix E

Excluded Months By Panel and Weighting Method

```

## $` (AEW) Panel B's Excluded Months` 
## [1] "1932-07-01" "1932-08-01" "1933-04-01" "1933-05-01" "1939-09-01"
## 
## $` (AEW) Panel B's Number of Excluded Months` 
## [1] 5
## 
## $` (AVW) Panel B's Excluded Months` 
## [1] "1932-07-01" "1932-08-01" "1933-04-01" "1933-05-01"
## 
## $` (AVW) Panel B's Number of Excluded Months` 
## [1] 4
## 
## $` (AEW) Panel C's Excluded Months` 
##      [,1]      [,2]      [,3]      [,4]      [,5]
## [1,] "1927-09-01" "1927-11-01" "1927-12-01" "1928-01-01" "1928-04-01"
## [2,] "1928-05-01" "1928-08-01" "1928-11-01" "1929-03-01" "1929-05-01"
## [3,] "1929-06-01" "1929-08-01" "1929-10-01" "1929-11-01" "1929-12-01"
## [4,] "1930-01-01" "1930-03-01" "1930-04-01" "1930-06-01" "1930-07-01"
## [5,] "1930-12-01" "1931-01-01" "1931-02-01" "1931-03-01" "1931-05-01"
## [6,] "1931-06-01" "1931-08-01" "1931-09-01" "1931-10-01" "1931-11-01"
## [7,] "1931-12-01" "1932-01-01" "1932-02-01" "1932-03-01" "1932-04-01"
## [8,] "1932-05-01" "1932-06-01" "1932-07-01" "1932-08-01" "1932-09-01"
## [9,] "1932-10-01" "1932-11-01" "1932-12-01" "1933-01-01" "1933-02-01"
## [10,] "1933-03-01" "1933-04-01" "1933-05-01" "1933-06-01" "1933-08-01"
## [11,] "1933-09-01" "1933-10-01" "1933-11-01" "1934-01-01" "1934-05-01"
## [12,] "1934-07-01" "1934-08-01" "1934-10-01" "1935-02-01" "1935-04-01"

```

```

## [13,] "1935-06-01" "1935-07-01" "1935-08-01" "1935-09-01" "1935-11-01"
## [14,] "1935-12-01" "1936-01-01" "1936-02-01" "1936-04-01" "1936-10-01"
## [15,] "1937-01-01" "1937-02-01" "1937-09-01" "1938-03-01" "1938-04-01"
## [16,] "1938-05-01" "1938-06-01" "1938-07-01" "1939-03-01" "1939-09-01"
## [17,] "1940-04-01" "1940-05-01" "1941-01-01" "1942-01-01" "1942-09-01"
## [18,] "1943-01-01" "1943-02-01" "1943-04-01" "1943-05-01" "1948-05-01"
## [19,] "1950-07-01" "1957-12-01" "1967-09-01" "1968-01-01" "1968-05-01"
## [20,] "1970-04-01" "1970-05-01" "1970-07-01" "1970-09-01" "1971-01-01"
## [21,] "1972-01-01" "1972-02-01" "1973-07-01" "1973-11-01" "1973-12-01"
## [22,] "1974-01-01" "1974-09-01" "1974-10-01" "1975-01-01" "1976-01-01"
## [23,] "1977-01-01" "1978-10-01" "1980-03-01" "1980-09-01" "1980-11-01"
## [24,] "1981-01-01" "1982-08-01" "1982-11-01" "1983-01-01" "1983-05-01"
## [25,] "1983-10-01" "1985-01-01" "1986-08-01" "1987-01-01" "1987-09-01"
## [26,] "1987-10-01" "1990-11-01" "1990-12-01" "1991-02-01" "1991-11-01"
## [27,] "1991-12-01" "1992-01-01" "1996-12-01" "1997-01-01" "1997-08-01"
## [28,] "1997-12-01" "1998-01-01" "1998-06-01" "1998-08-01" "1999-01-01"
## [29,] "1999-03-01" "1999-04-01" "1999-10-01" "1999-11-01" "1999-12-01"
## [30,] "2000-01-01" "2000-02-01" "2000-03-01" "2000-04-01" "2000-05-01"
## [31,] "2000-06-01" "2000-09-01" "2000-10-01" "2000-11-01" "2000-12-01"
## [32,] "2001-01-01" "2001-02-01" "2001-03-01" "2001-04-01" "2001-05-01"
## [33,] "2001-07-01" "2001-08-01" "2001-09-01" "2001-10-01" "2001-11-01"
## [34,] "2002-04-01" "2002-05-01" "2002-07-01" "2002-09-01" "2002-10-01"
## [35,] "2002-11-01" "2002-12-01" "2003-05-01" "2008-03-01" "2008-07-01"
## [36,] "2008-08-01" "2008-09-01" "2008-10-01" "2009-01-01" "2009-02-01"
## [37,] "2009-03-01" "2009-04-01" "2009-05-01" "2009-06-01" "2009-07-01"
## [38,] "2009-08-01" "2011-10-01" ""           ""           ""
## 
## $` (AEW) Panel C's Number of Excluded Months`
## [1] 187
## 
## $` (AVW) Panel C's Excluded Months` 
## [,1]      [,2]      [,3]      [,4]      [,5]
## [1,] "1927-01-01" "1927-02-01" "1927-03-01" "1927-04-01" "1927-06-01"
## [2,] "1927-08-01" "1927-12-01" "1928-03-01" "1928-04-01" "1928-09-01"
## [3,] "1928-11-01" "1929-01-01" "1929-02-01" "1929-04-01" "1929-06-01"
## [4,] "1929-07-01" "1929-08-01" "1929-09-01" "1929-10-01" "1929-11-01"
## [5,] "1929-12-01" "1930-01-01" "1930-04-01" "1930-05-01" "1930-06-01"
## [6,] "1930-07-01" "1930-10-01" "1930-11-01" "1930-12-01" "1931-01-01"
## [7,] "1931-02-01" "1931-04-01" "1931-05-01" "1931-06-01" "1931-08-01"
## [8,] "1931-09-01" "1931-10-01" "1931-11-01" "1931-12-01" "1932-01-01"
## [9,] "1932-02-01" "1932-04-01" "1932-05-01" "1932-06-01" "1932-07-01"

```

```

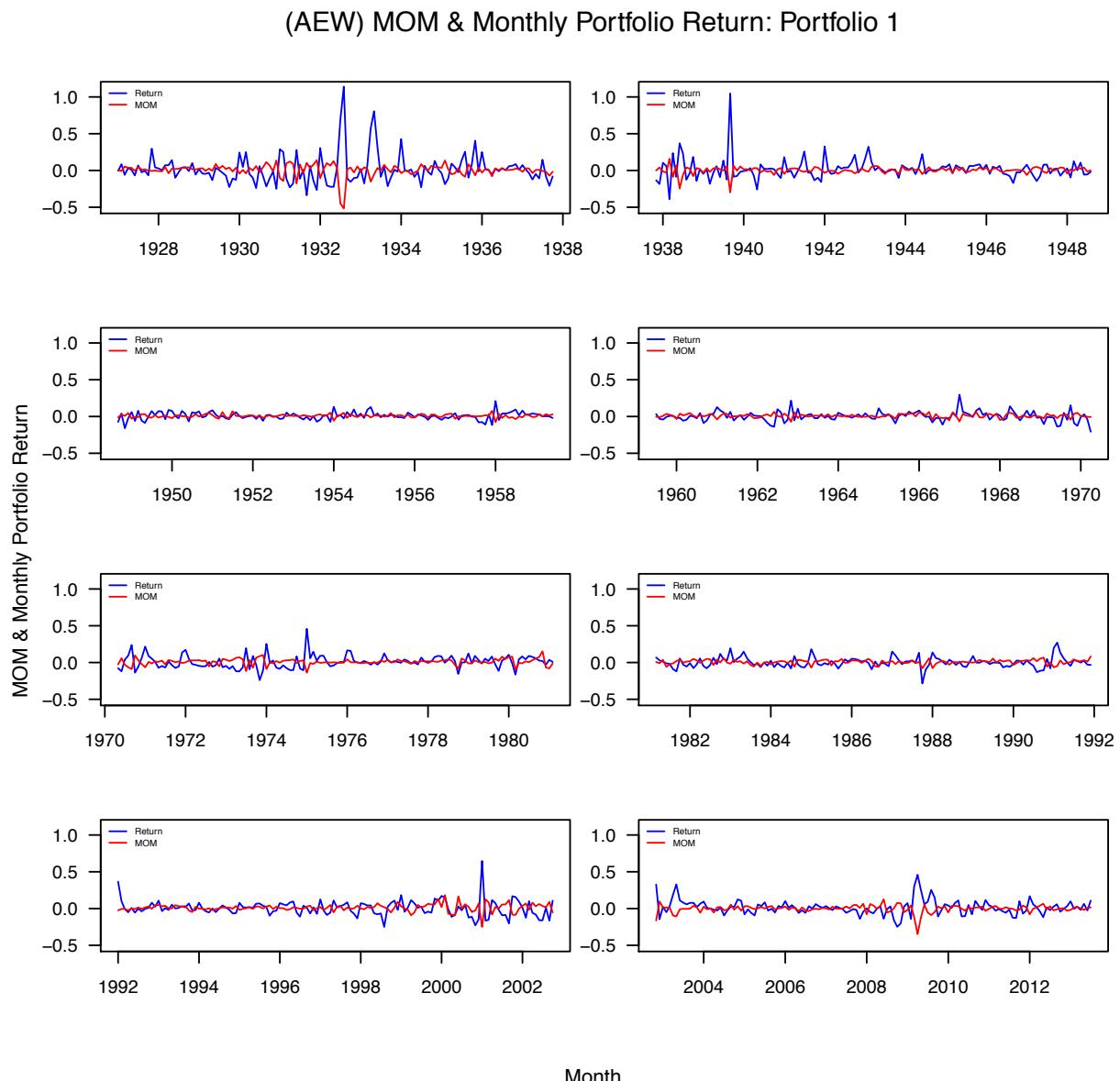
## [10,] "1932-08-01" "1932-09-01" "1932-10-01" "1932-11-01" "1932-12-01"
## [11,] "1933-03-01" "1933-04-01" "1933-05-01" "1933-06-01" "1933-07-01"
## [12,] "1933-08-01" "1933-09-01" "1933-10-01" "1933-11-01" "1933-12-01"
## [13,] "1934-01-01" "1934-03-01" "1934-05-01" "1934-07-01" "1934-11-01"
## [14,] "1935-02-01" "1935-03-01" "1935-05-01" "1935-06-01" "1935-08-01"
## [15,] "1935-09-01" "1935-11-01" "1936-01-01" "1936-02-01" "1936-03-01"
## [16,] "1936-04-01" "1936-07-01" "1936-09-01" "1936-10-01" "1937-01-01"
## [17,] "1937-03-01" "1937-04-01" "1937-09-01" "1937-11-01" "1937-12-01"
## [18,] "1938-01-01" "1938-02-01" "1938-03-01" "1938-04-01" "1938-05-01"
## [19,] "1938-06-01" "1938-07-01" "1938-10-01" "1938-11-01" "1939-03-01"
## [20,] "1939-07-01" "1939-09-01" "1940-05-01" "1940-06-01" "1940-10-01"
## [21,] "1941-07-01" "1941-10-01" "1941-11-01" "1941-12-01" "1942-01-01"
## [22,] "1942-05-01" "1943-01-01" "1943-02-01" "1943-04-01" "1943-06-01"
## [23,] "1943-12-01" "1944-12-01" "1945-06-01" "1945-11-01" "1946-01-01"
## [24,] "1947-01-01" "1950-04-01" "1950-07-01" "1950-11-01" "1950-12-01"
## [25,] "1951-06-01" "1955-07-01" "1955-09-01" "1957-08-01" "1957-11-01"
## [26,] "1958-07-01" "1959-05-01" "1960-02-01" "1962-04-01" "1963-10-01"
## [27,] "1963-11-01" "1966-02-01" "1966-10-01" "1966-11-01" "1967-07-01"
## [28,] "1967-10-01" "1968-03-01" "1968-09-01" "1969-06-01" "1970-01-01"
## [29,] "1970-04-01" "1970-07-01" "1970-09-01" "1972-01-01" "1973-07-01"
## [30,] "1973-09-01" "1973-10-01" "1973-11-01" "1973-12-01" "1974-01-01"
## [31,] "1974-05-01" "1974-09-01" "1974-10-01" "1975-01-01" "1976-01-01"
## [32,] "1978-10-01" "1978-12-01" "1980-02-01" "1980-03-01" "1980-11-01"
## [33,] "1981-01-01" "1982-07-01" "1982-10-01" "1983-01-01" "1983-04-01"
## [34,] "1983-08-01" "1985-01-01" "1987-01-01" "1987-10-01" "1991-01-01"
## [35,] "1991-12-01" "1992-01-01" "1993-04-01" "1995-12-01" "1997-02-01"
## [36,] "1997-07-01" "1998-08-01" "1998-12-01" "1999-01-01" "1999-03-01"
## [37,] "1999-04-01" "1999-06-01" "1999-10-01" "1999-11-01" "1999-12-01"
## [38,] "2000-01-01" "2000-02-01" "2000-03-01" "2000-04-01" "2000-05-01"
## [39,] "2000-06-01" "2000-07-01" "2000-08-01" "2000-09-01" "2000-10-01"
## [40,] "2000-11-01" "2000-12-01" "2001-01-01" "2001-02-01" "2001-03-01"
## [41,] "2001-04-01" "2001-05-01" "2001-09-01" "2001-10-01" "2001-11-01"
## [42,] "2002-02-01" "2002-04-01" "2002-06-01" "2002-07-01" "2002-08-01"
## [43,] "2002-10-01" "2002-11-01" "2002-12-01" "2003-10-01" "2005-02-01"
## [44,] "2006-02-01" "2008-04-01" "2008-07-01" "2008-08-01" "2008-09-01"
## [45,] "2008-10-01" "2008-11-01" "2009-01-01" "2009-02-01" "2009-03-01"
## [46,] "2009-04-01" "2009-05-01" "2009-06-01" "2009-08-01" "2010-03-01"
## [47,] "2013-01-01" "2013-07-01" ""          ""          ""
## 
## $`AVW) Panel C's Number of Excluded Months`
## [1] 232

```

Appendix F

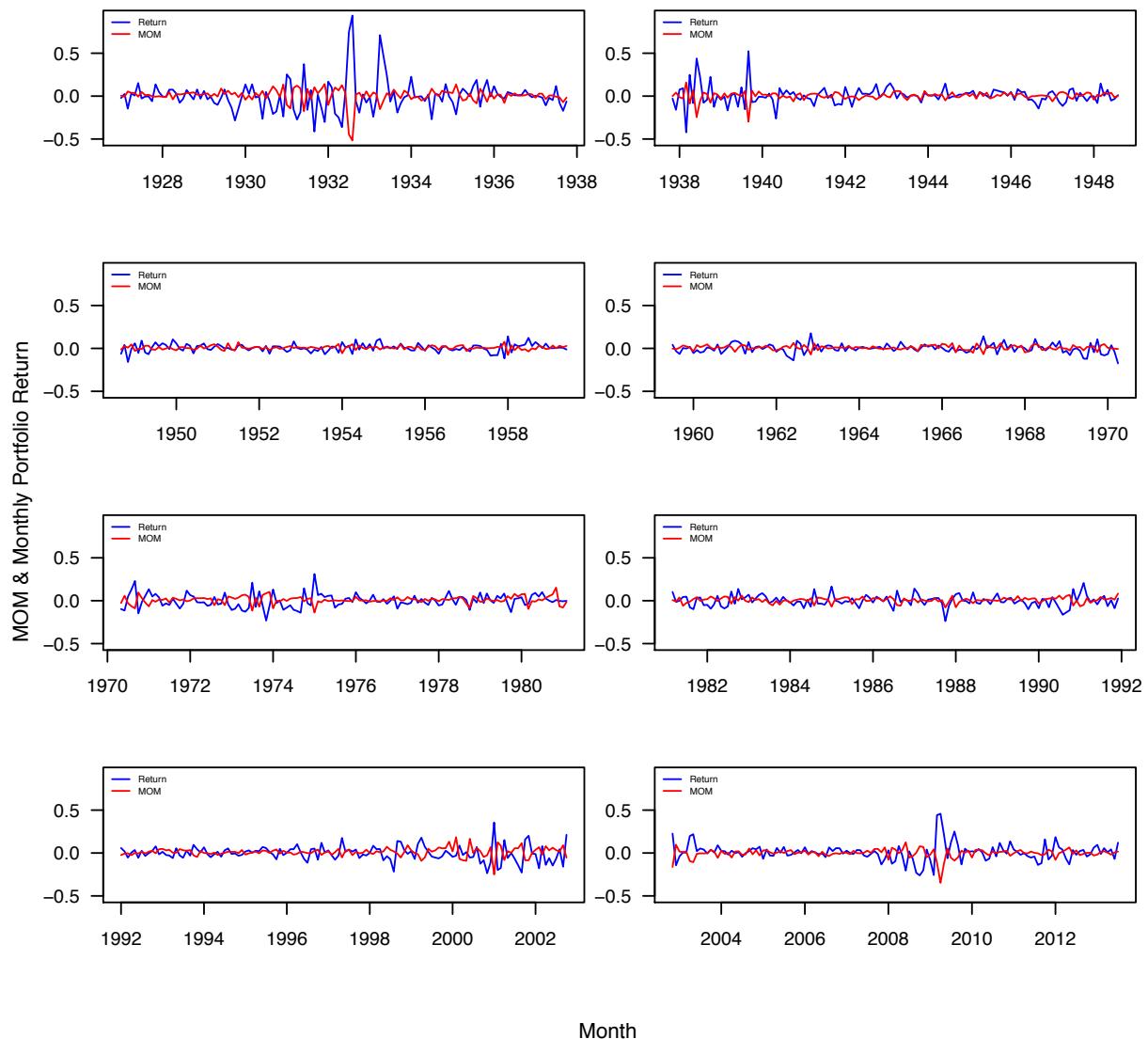
Plot of M_{OM} vs Monthly Portfolio Return

F.1 (AEW) Portfolio 1



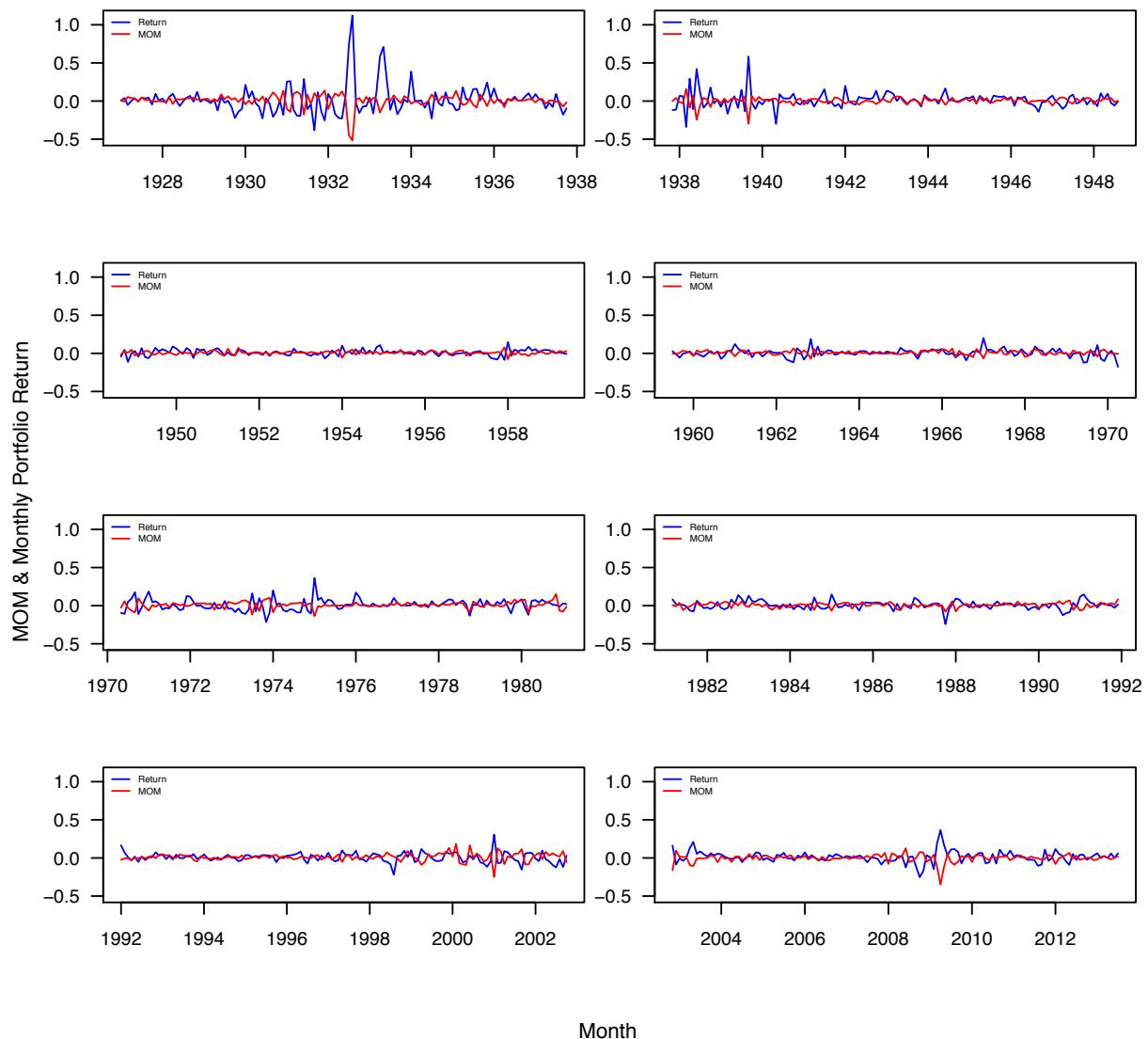
F.2 (AVW) Portfolio 1

(AVW) MOM & Monthly Portfolio Return: Portfolio 1



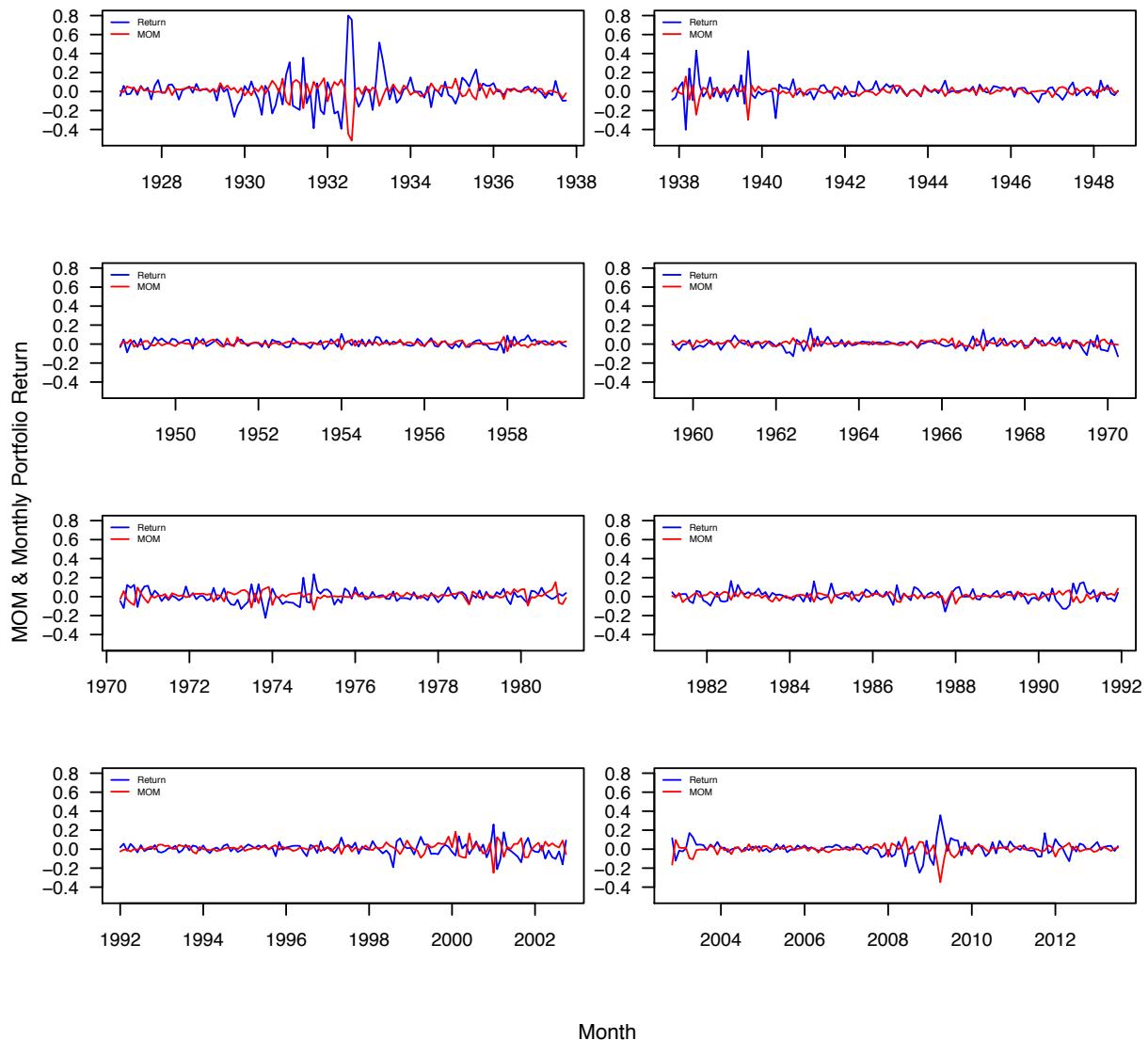
F.3 (AEW) Portfolio 2

(AEW) MOM & Monthly Portfolio Return: Portfolio 2



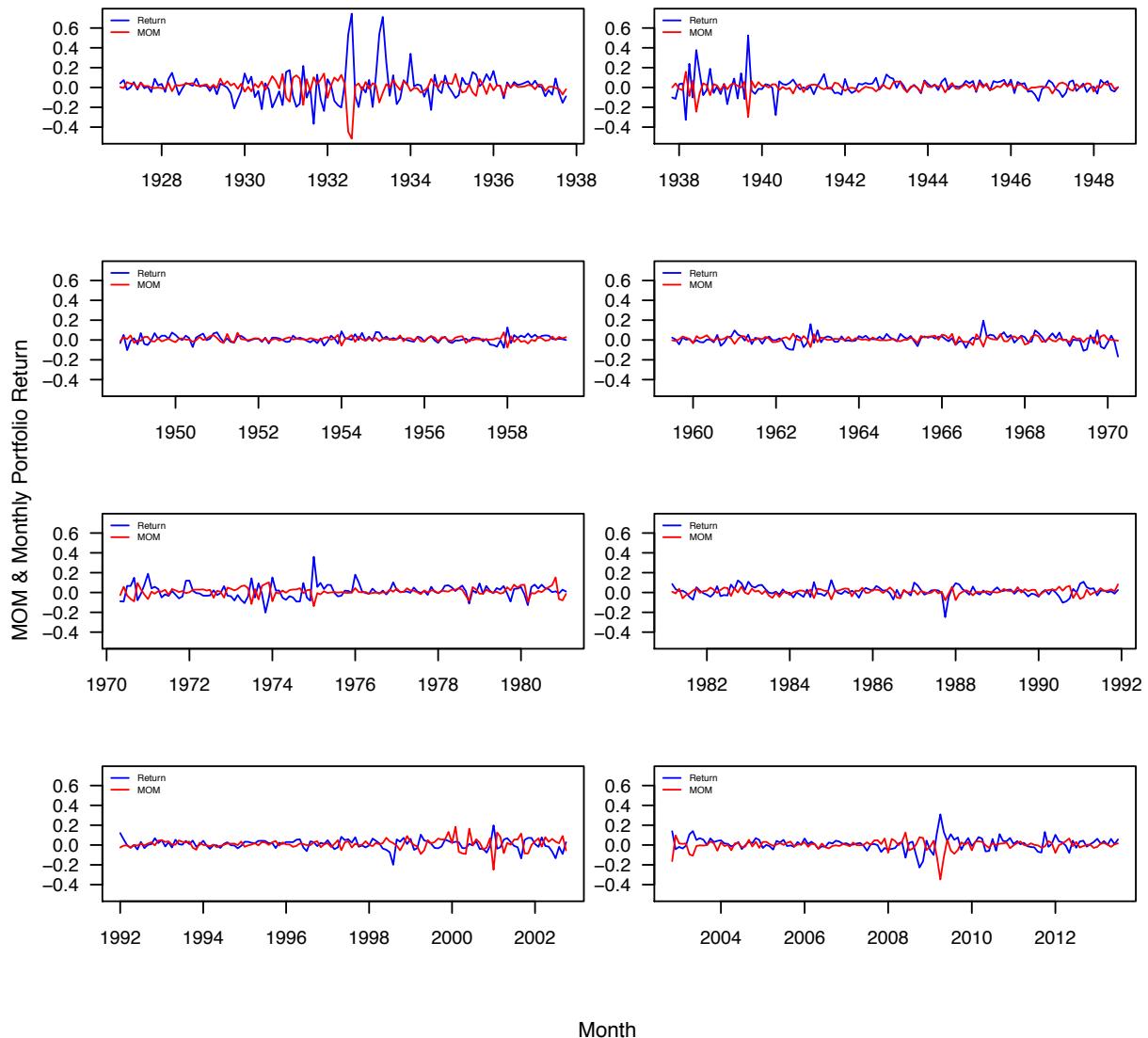
F.4 (AVW) Portfolio 2

(AVW) MOM & Monthly Portfolio Return: Portfolio 2



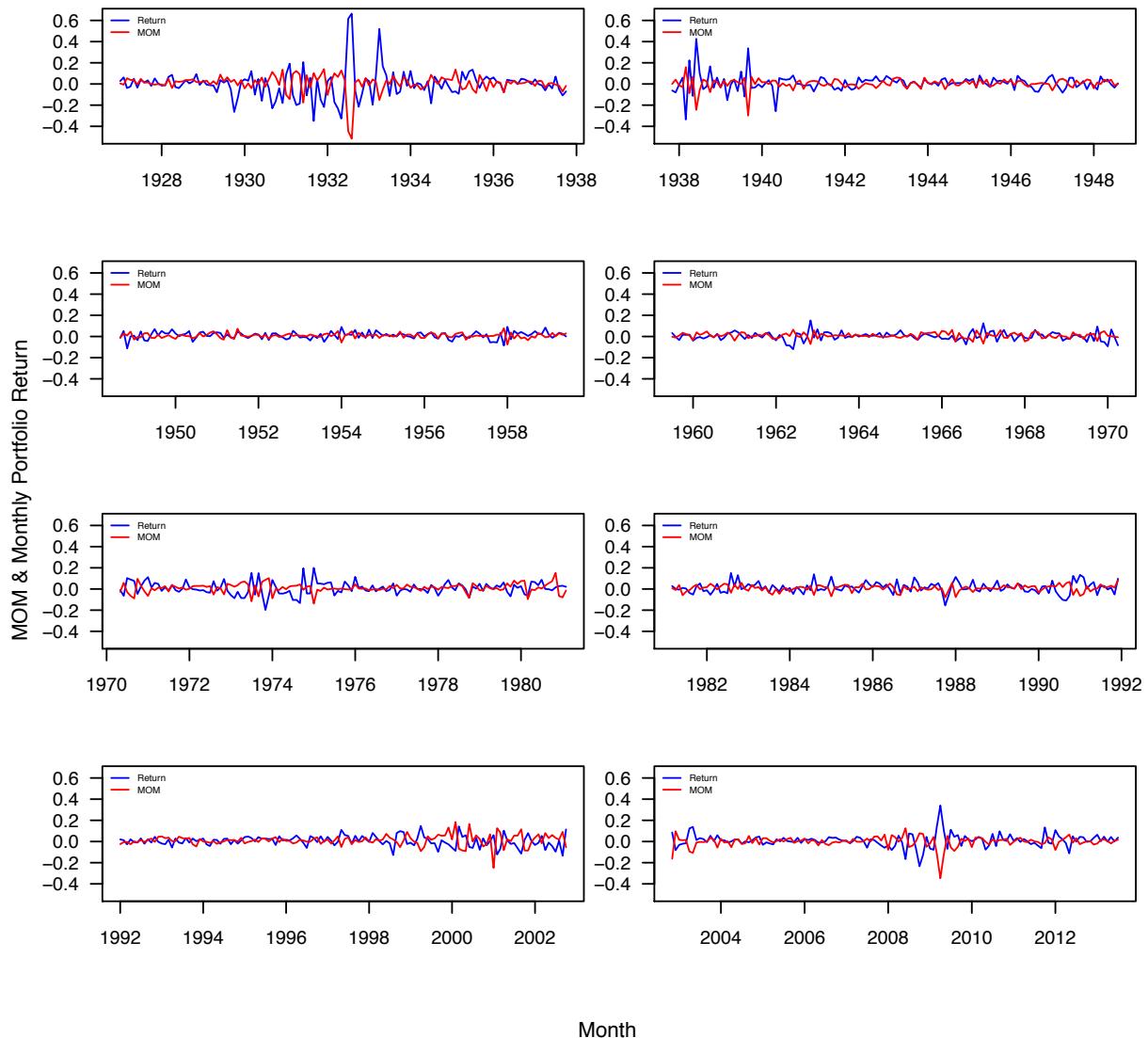
F.5 (AEW) Portfolio 3

(AEW) MOM & Monthly Portfolio Return: Portfolio 3



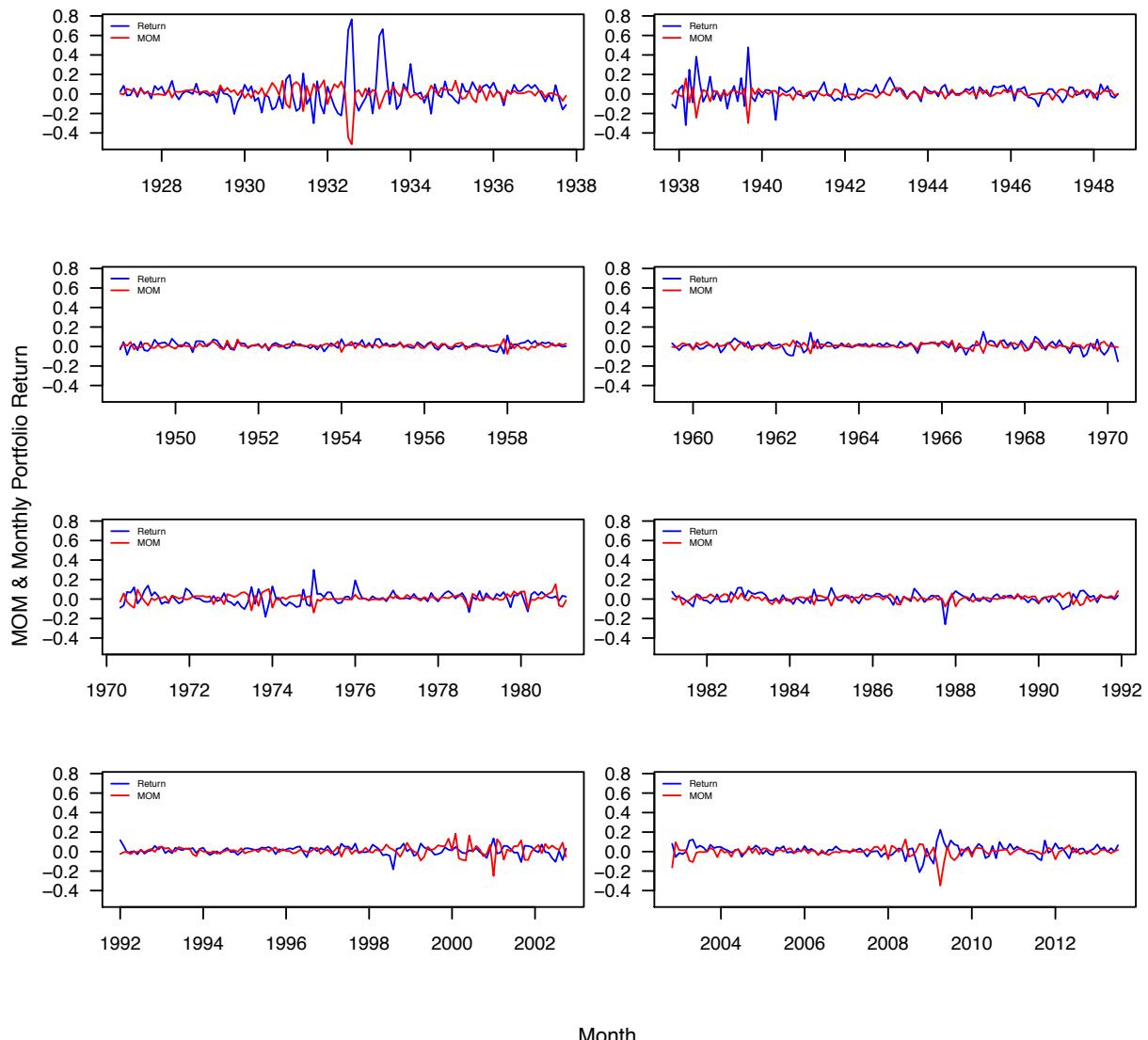
F.6 (AVW) Portfolio 3

(AVW) MOM & Monthly Portfolio Return: Portfolio 3



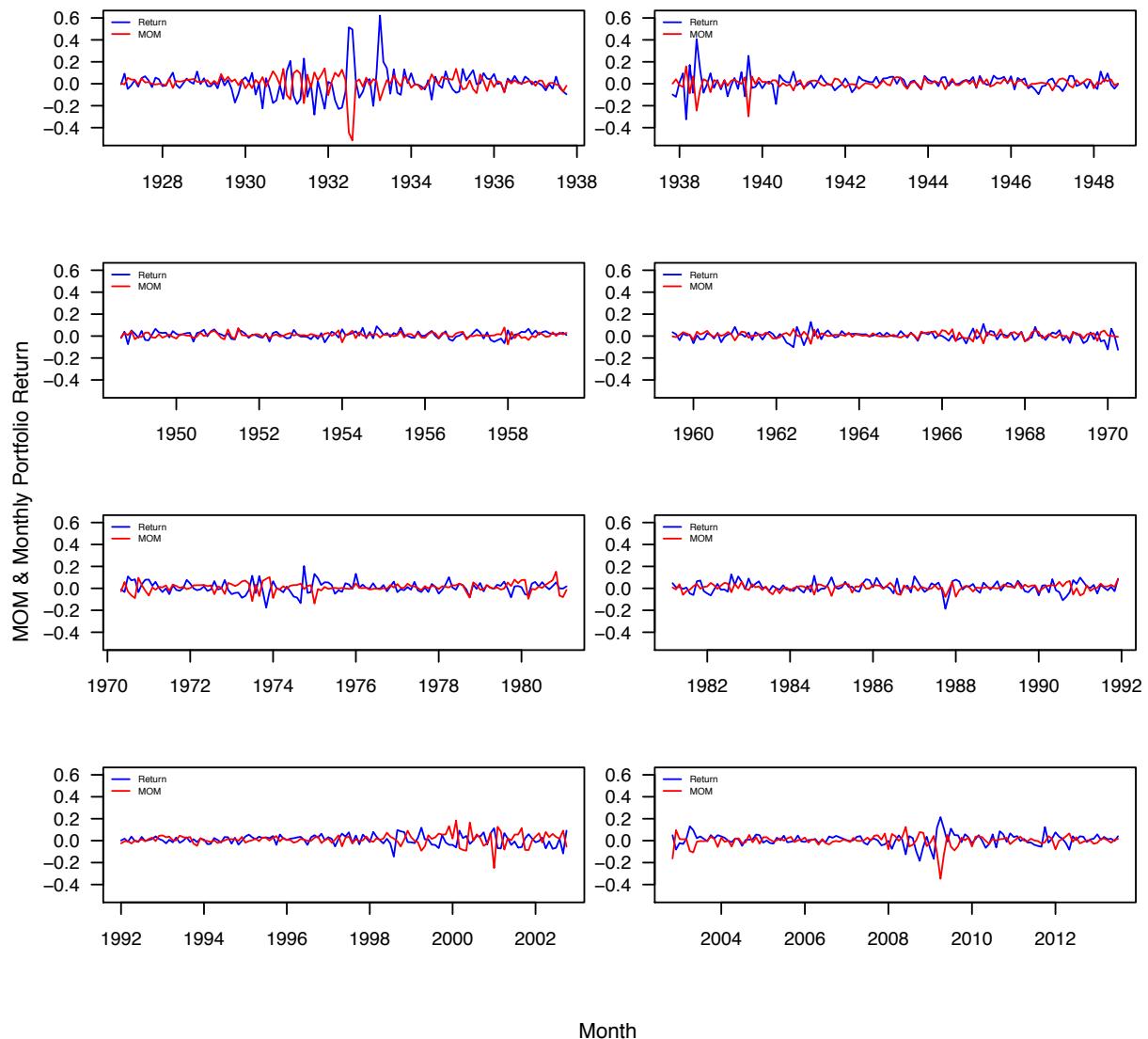
F.7 (AEW) Portfolio 4

(AEW) MOM & Monthly Portfolio Return: Portfolio 4



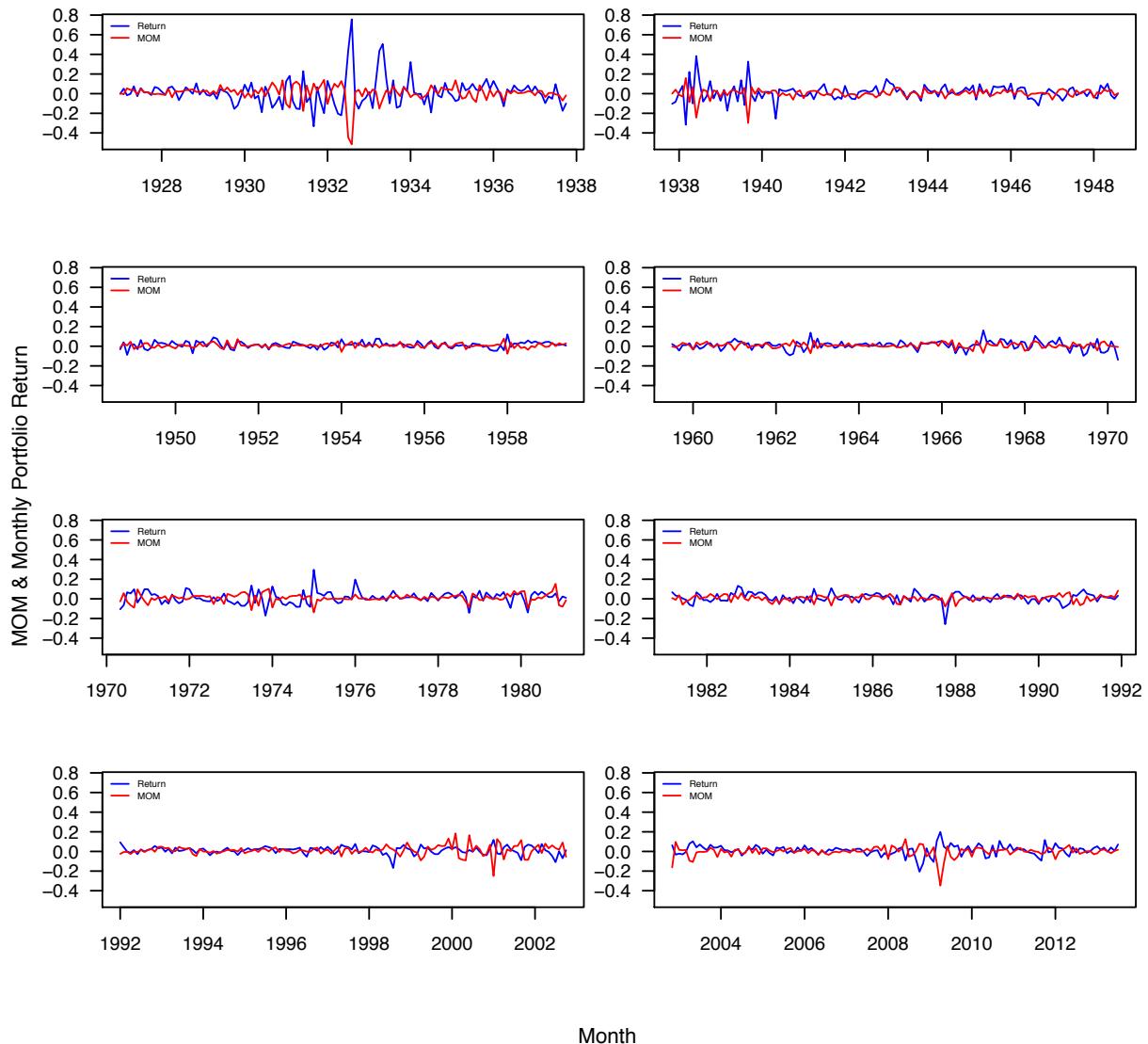
F.8 (AVW) Portfolio 4

(AVW) MOM & Monthly Portfolio Return: Portfolio 4



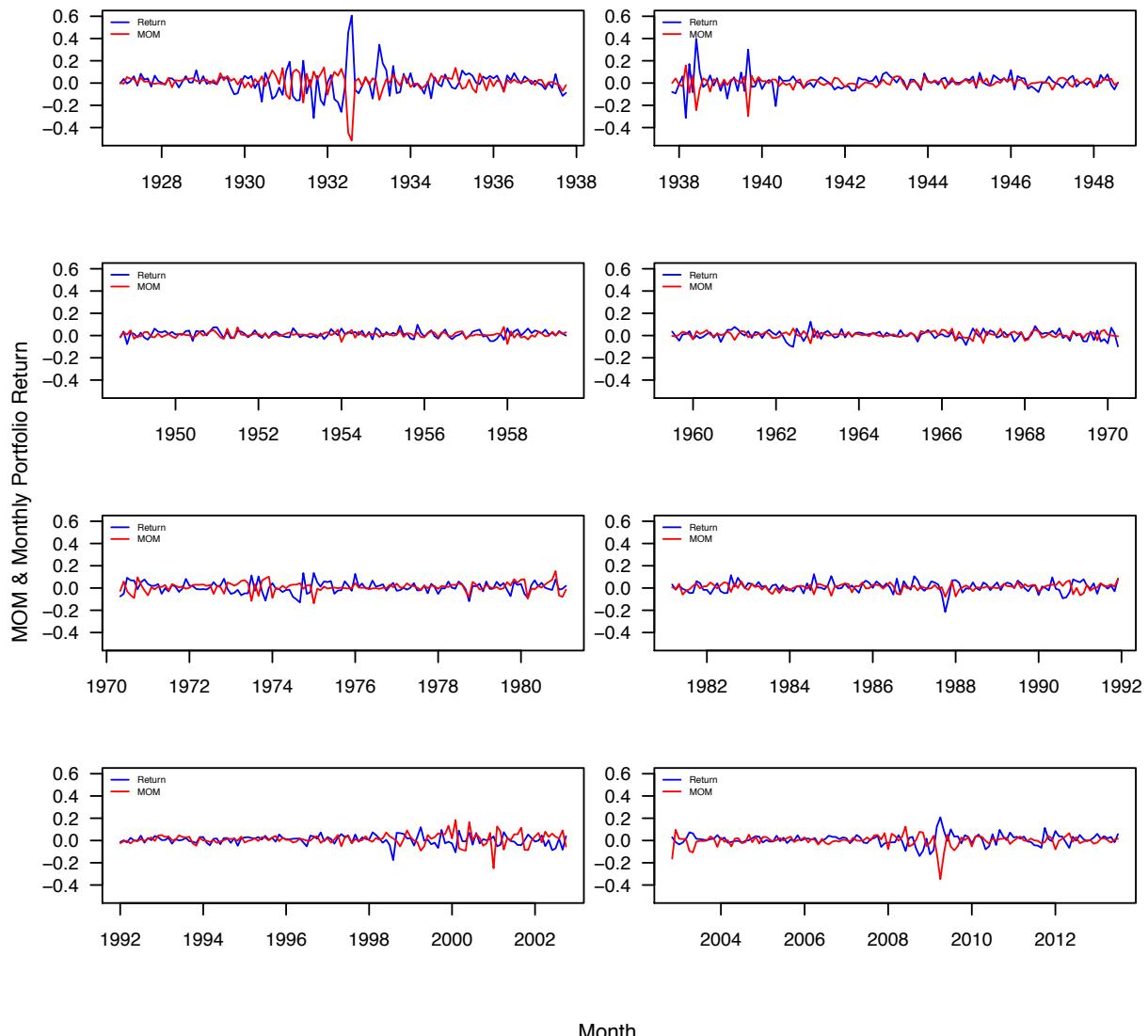
F.9 (AEW) Portfolio 5

(AEW) MOM & Monthly Portfolio Return: Portfolio 5



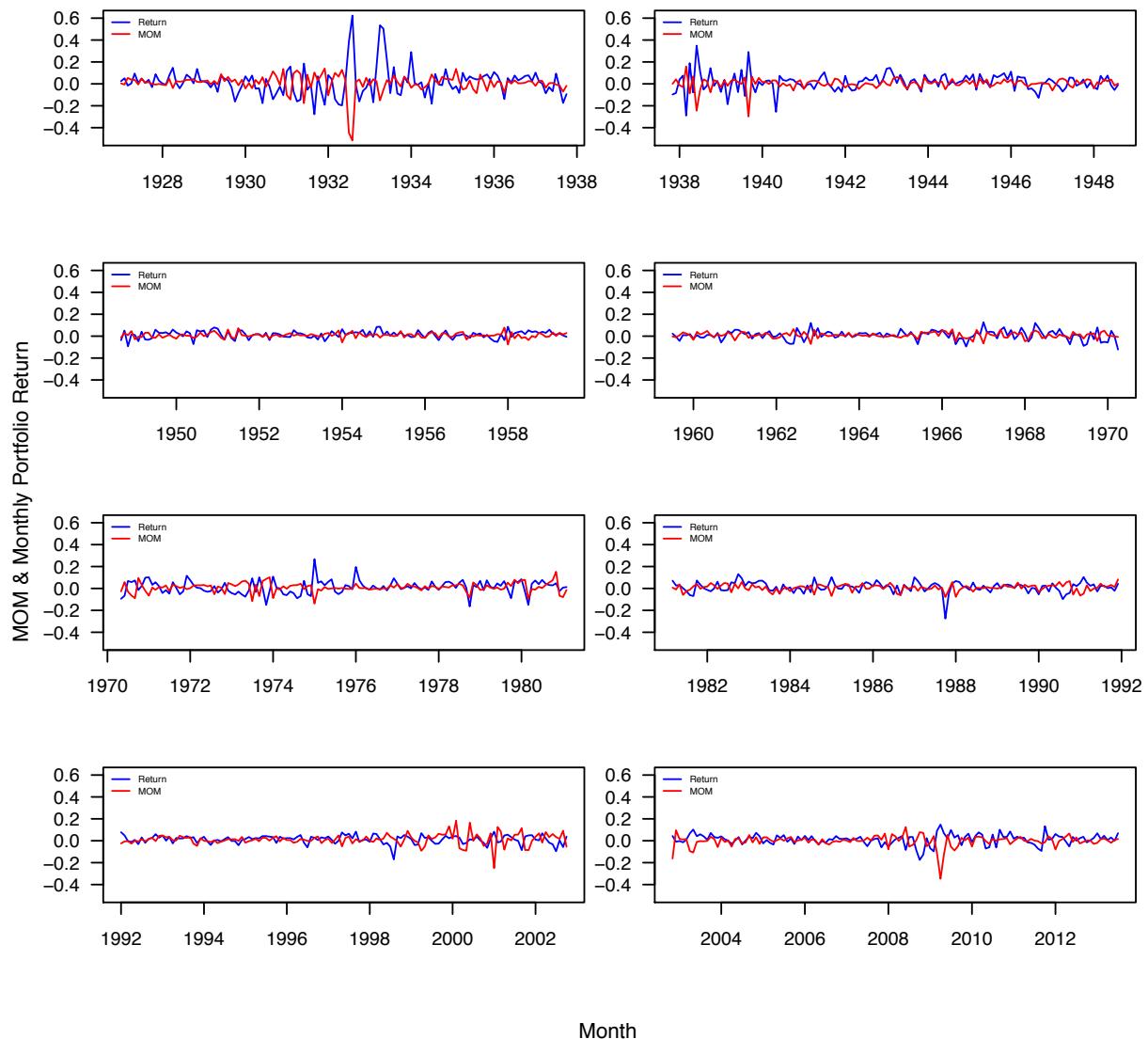
F.10 (AVW) Portfolio 5

(AVW) MOM & Monthly Portfolio Return: Portfolio 5



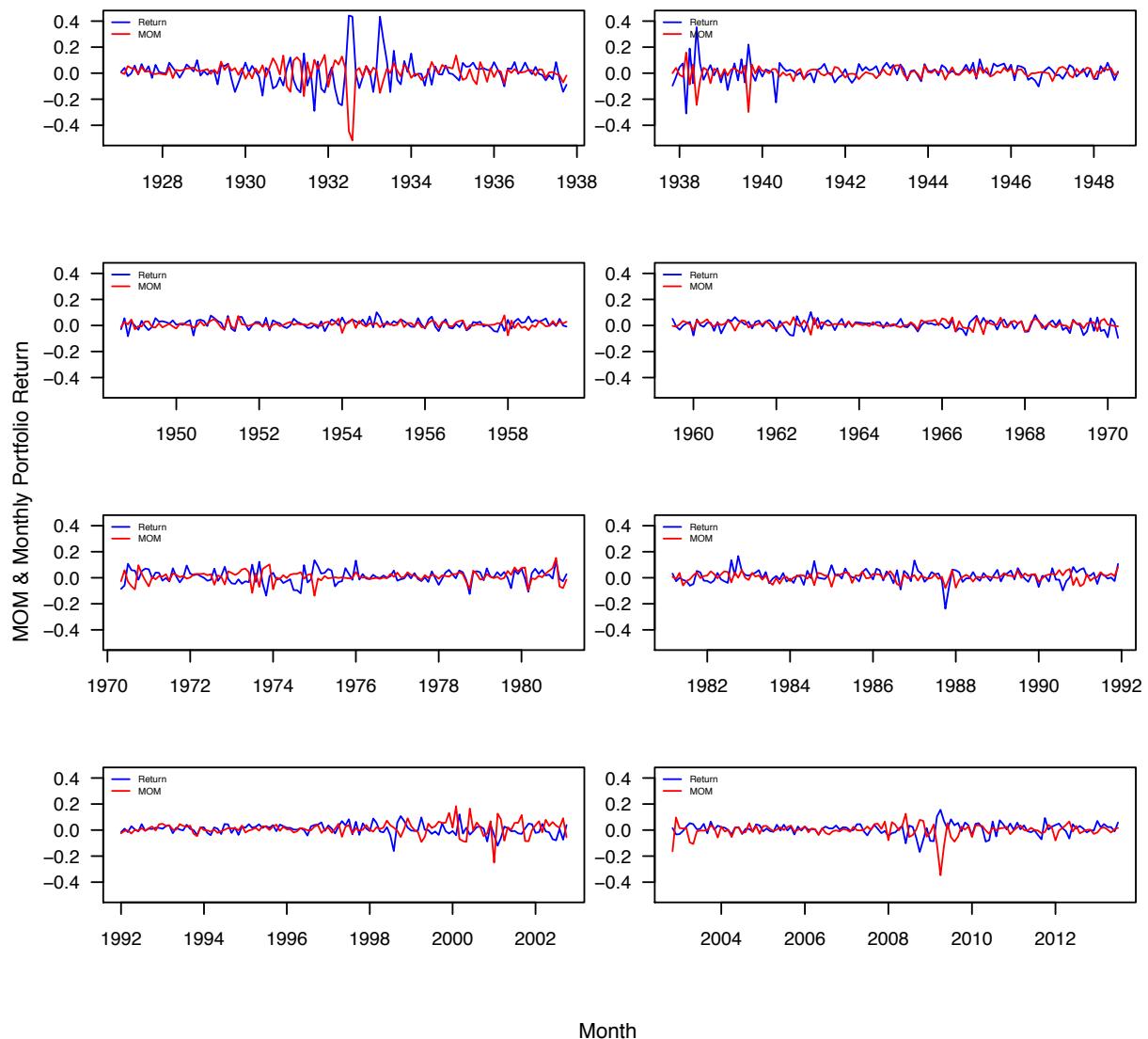
F.11 (AEW) Portfolio 6

(AEW) MOM & Monthly Portfolio Return: Portfolio 6



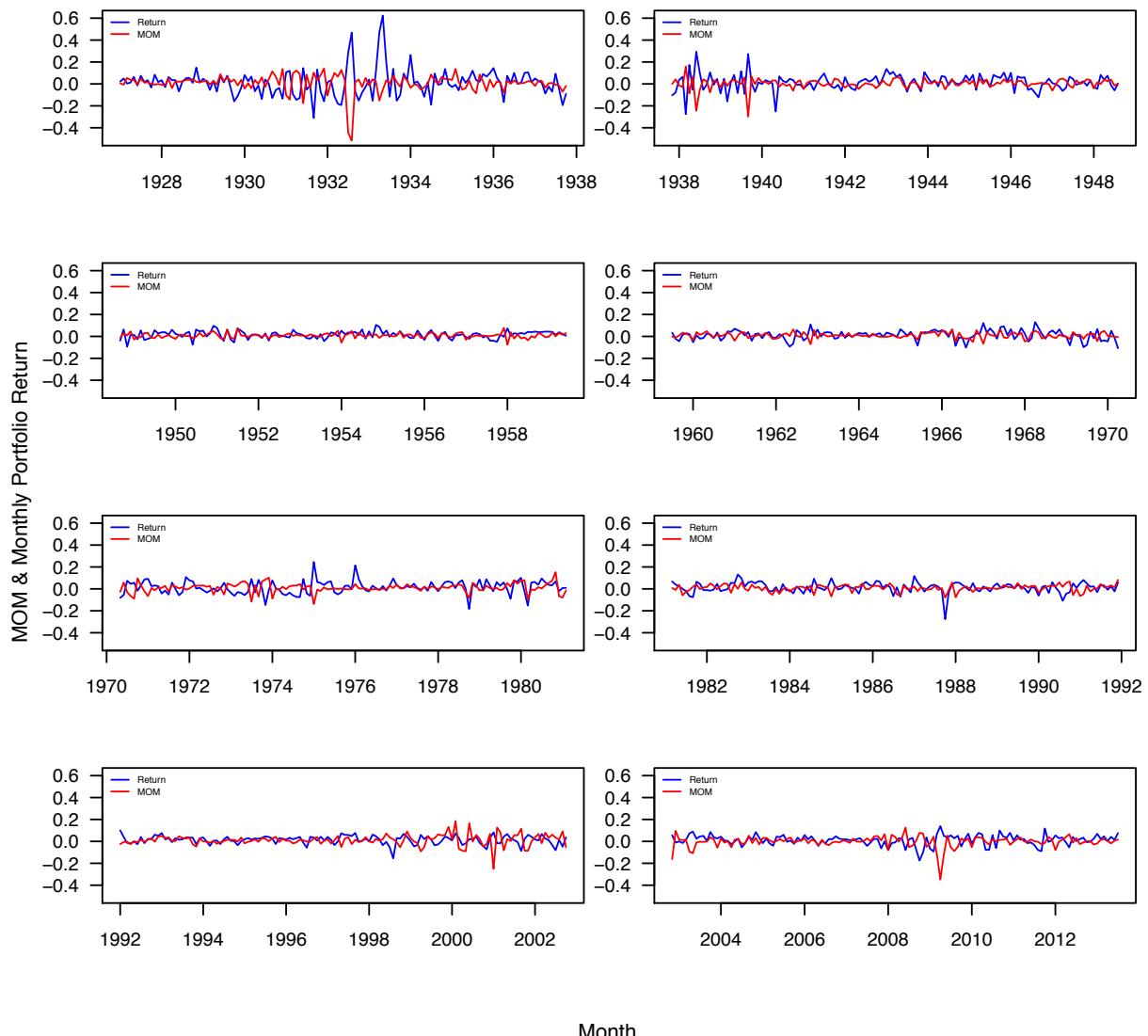
F.12 (AVW) Portfolio 6

(AVW) MOM & Monthly Portfolio Return: Portfolio 6



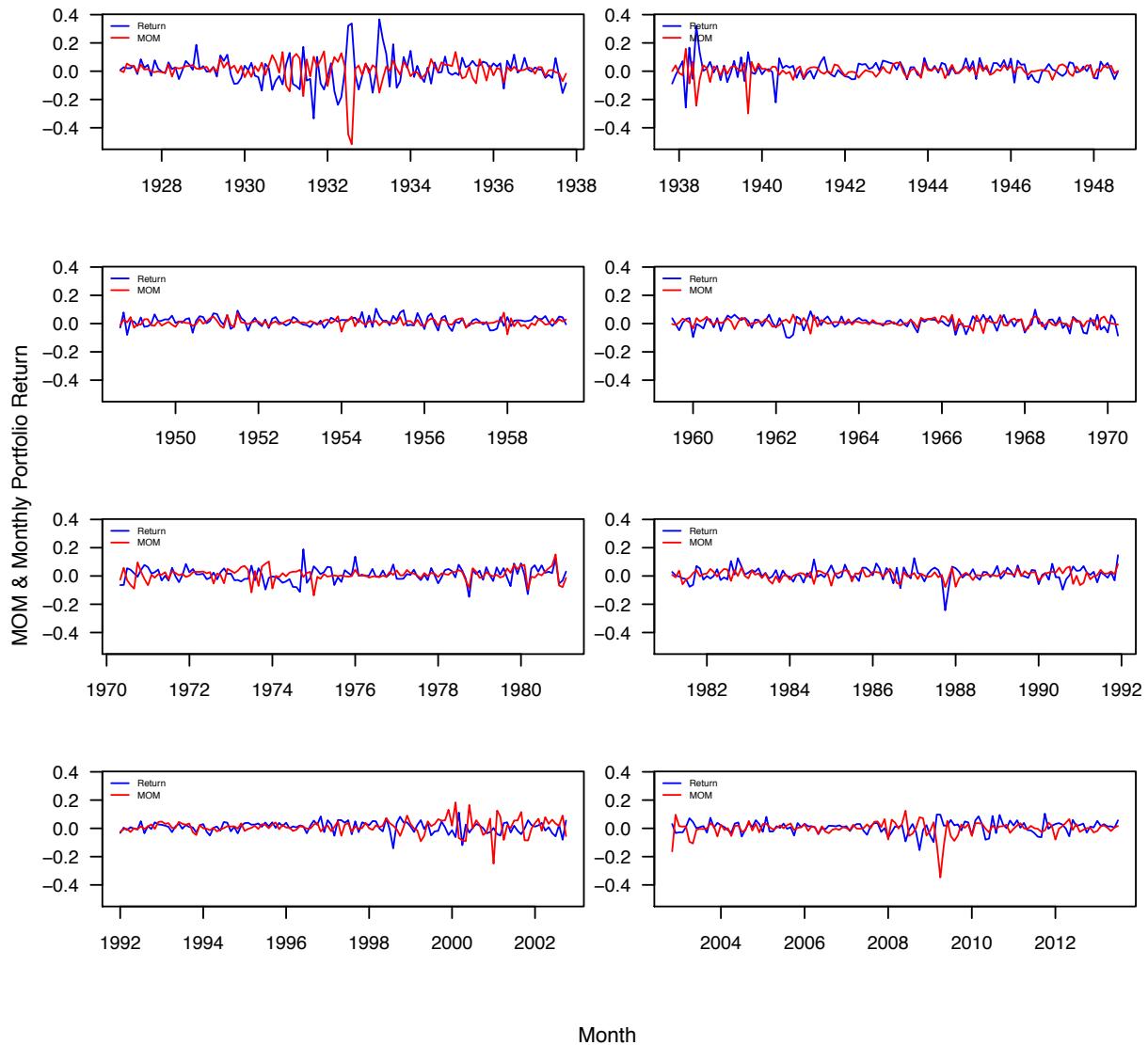
F.13 (AEW) Portfolio 7

(AEW) MOM & Monthly Portfolio Return: Portfolio 7



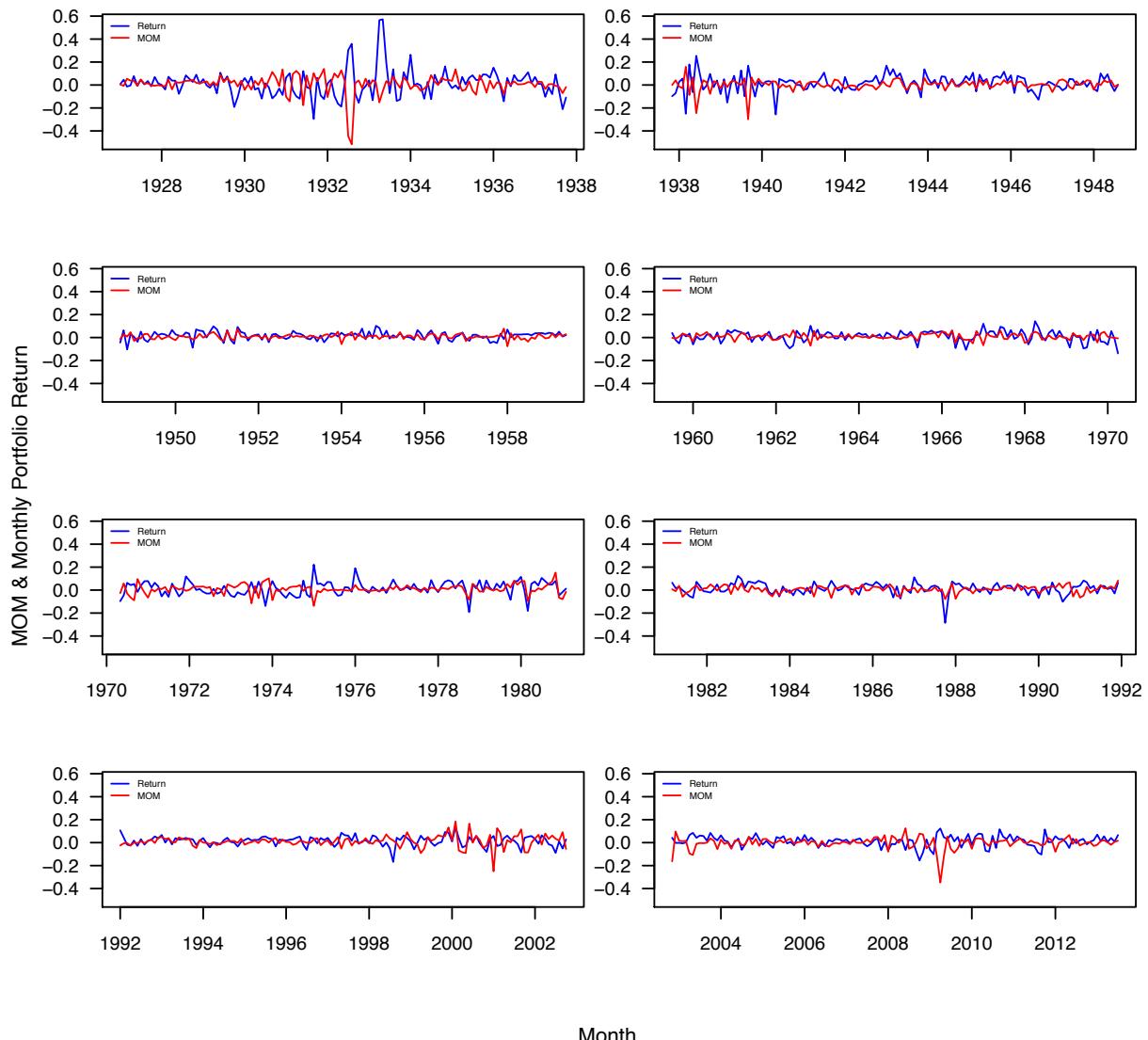
F.14 (AVW) Portfolio 7

(AVW) MOM & Monthly Portfolio Return: Portfolio 7

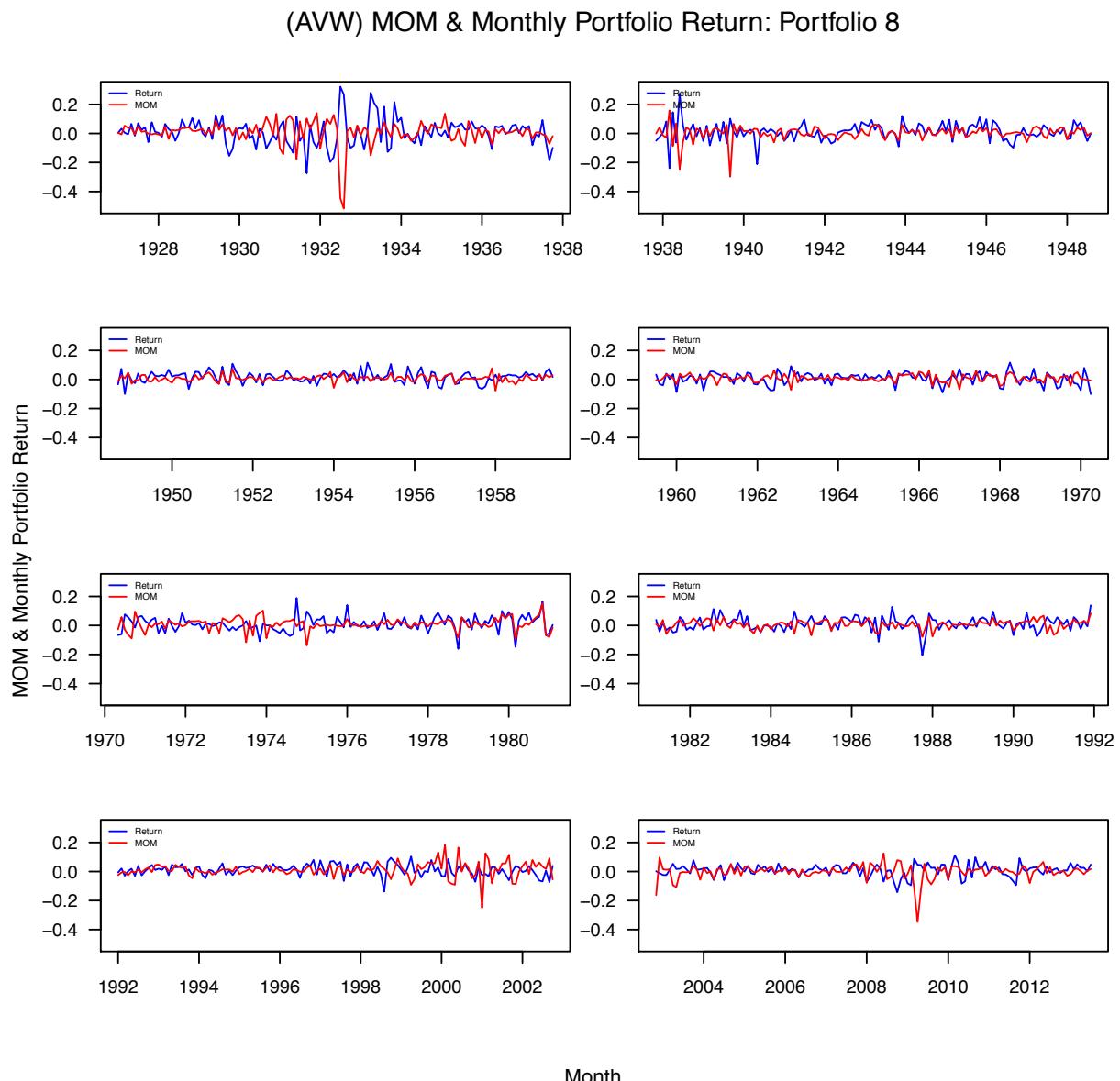


F.15 (AEW) Portfolio 8

(AEW) MOM & Monthly Portfolio Return: Portfolio 8

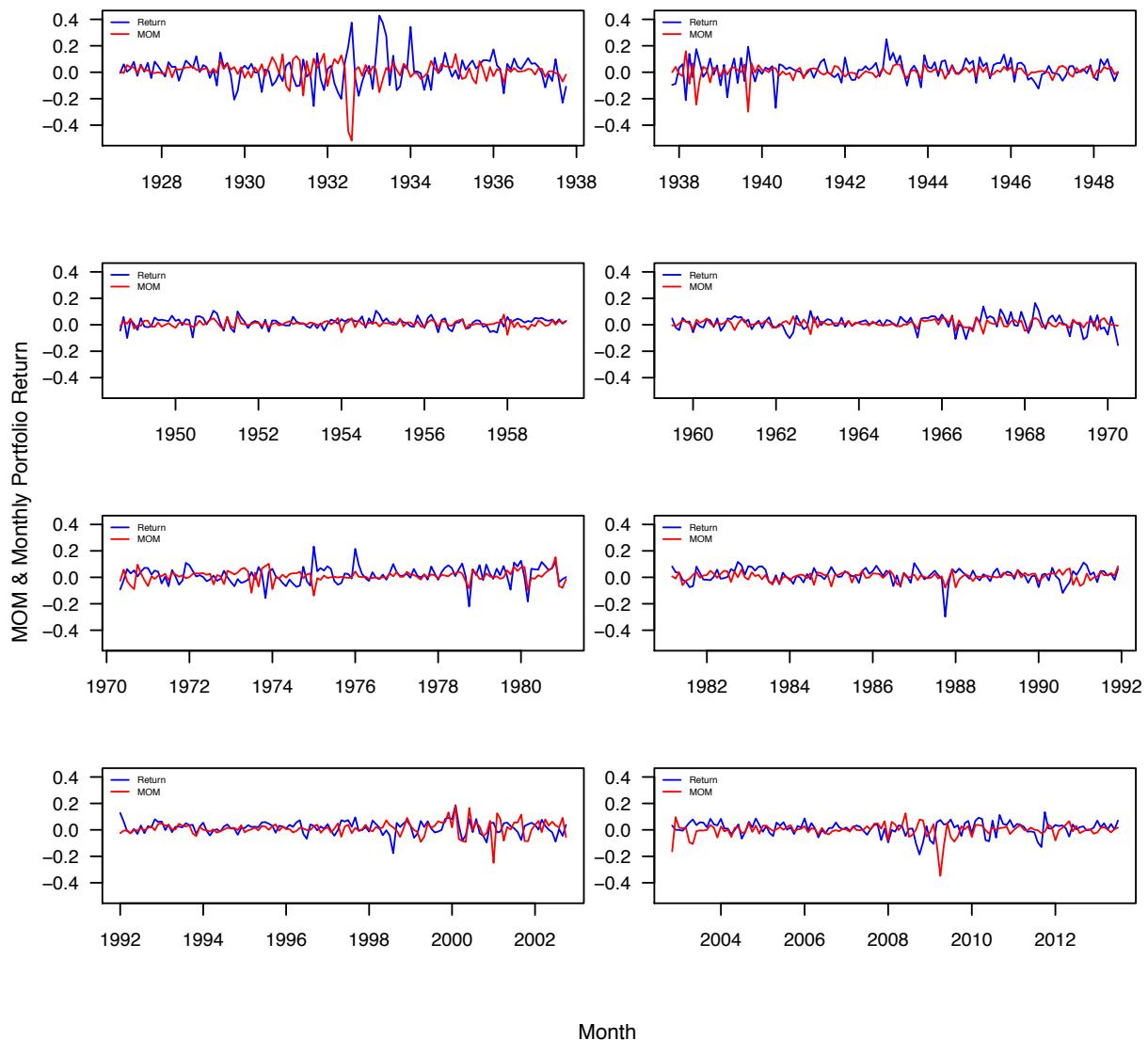


F.16 (AVW) Portfolio 8

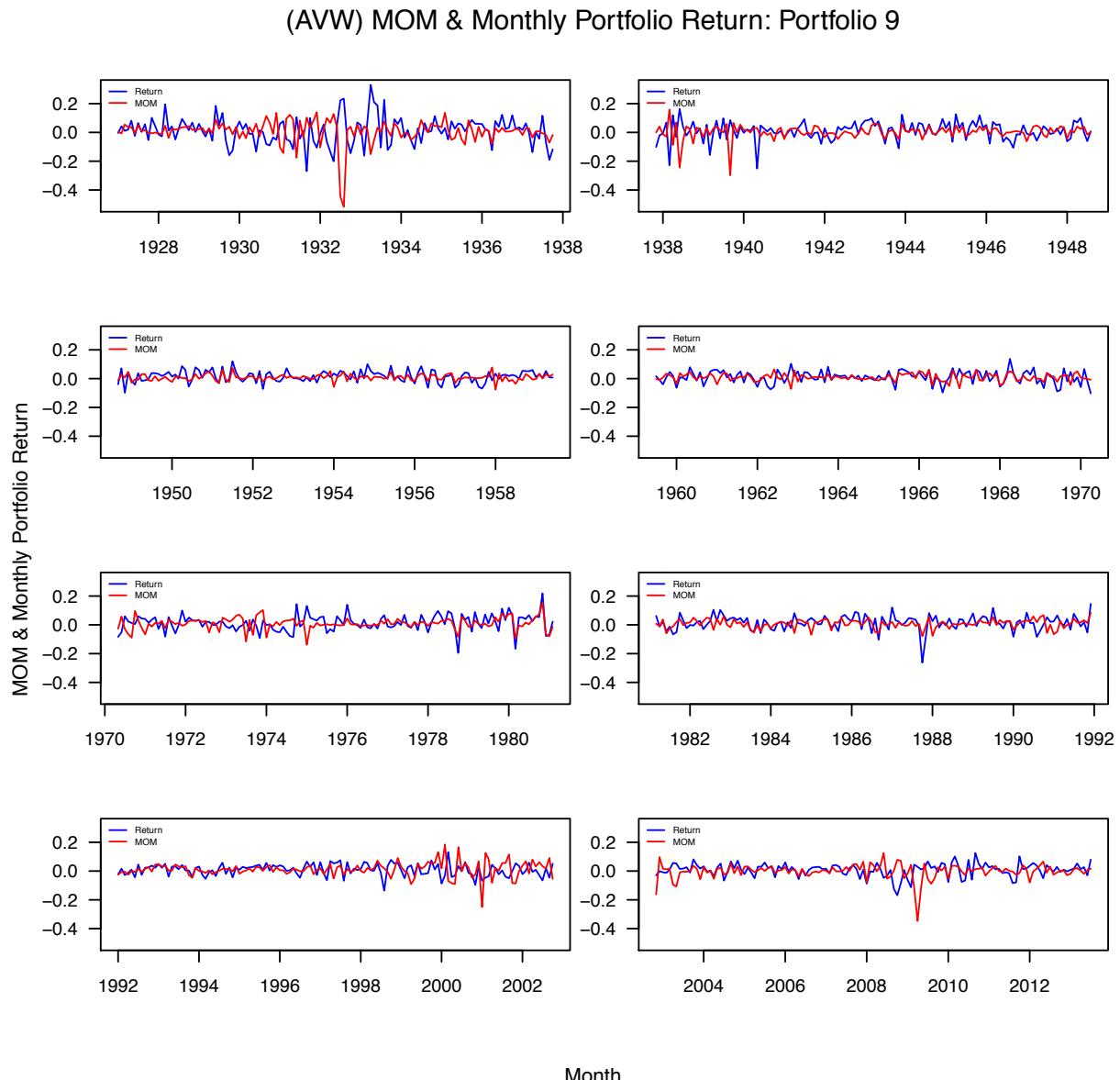


F.17 (AEW) Portfolio 9

(AEW) MOM & Monthly Portfolio Return: Portfolio 9

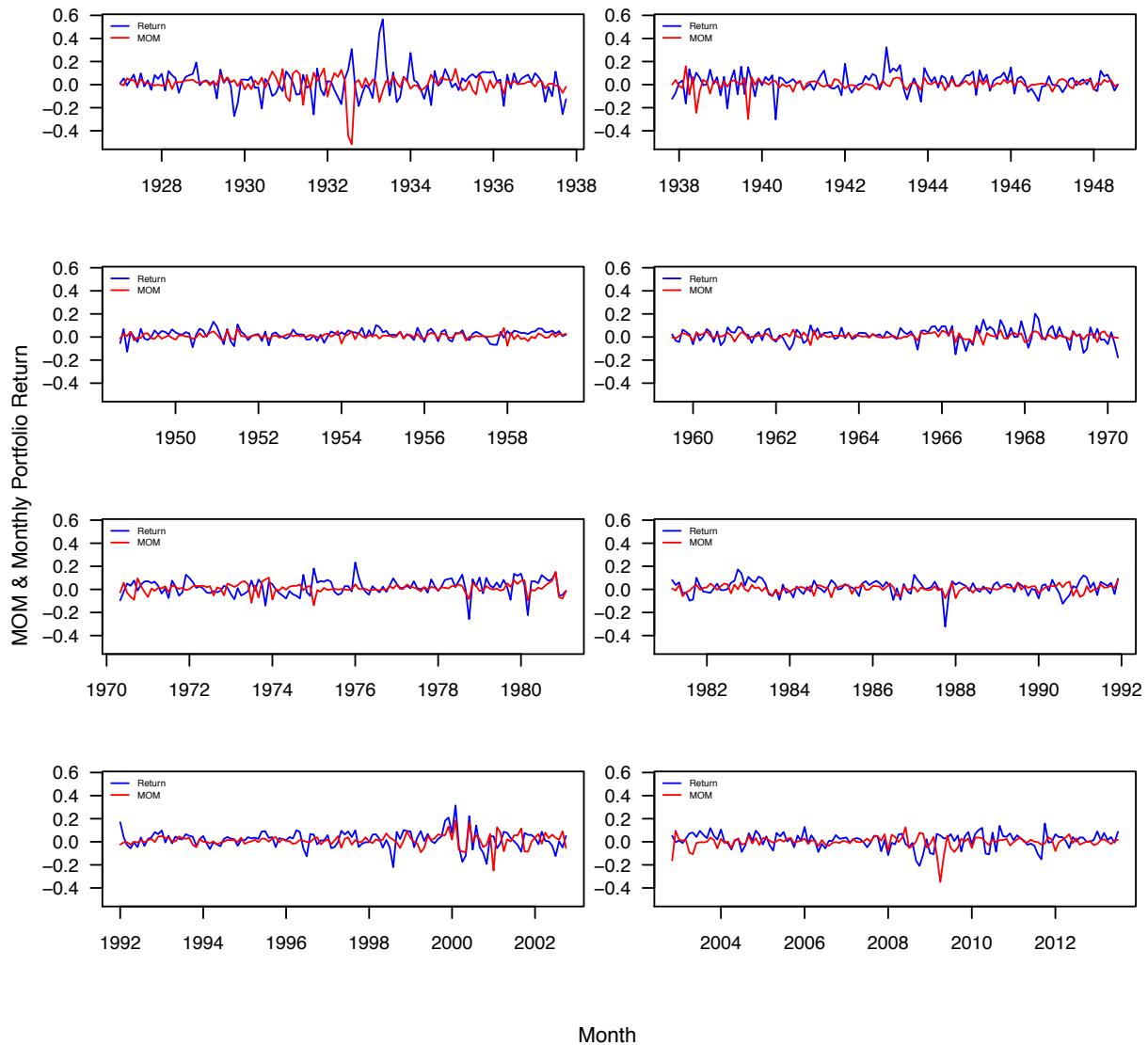


F.18 (AVW) Portfolio 9

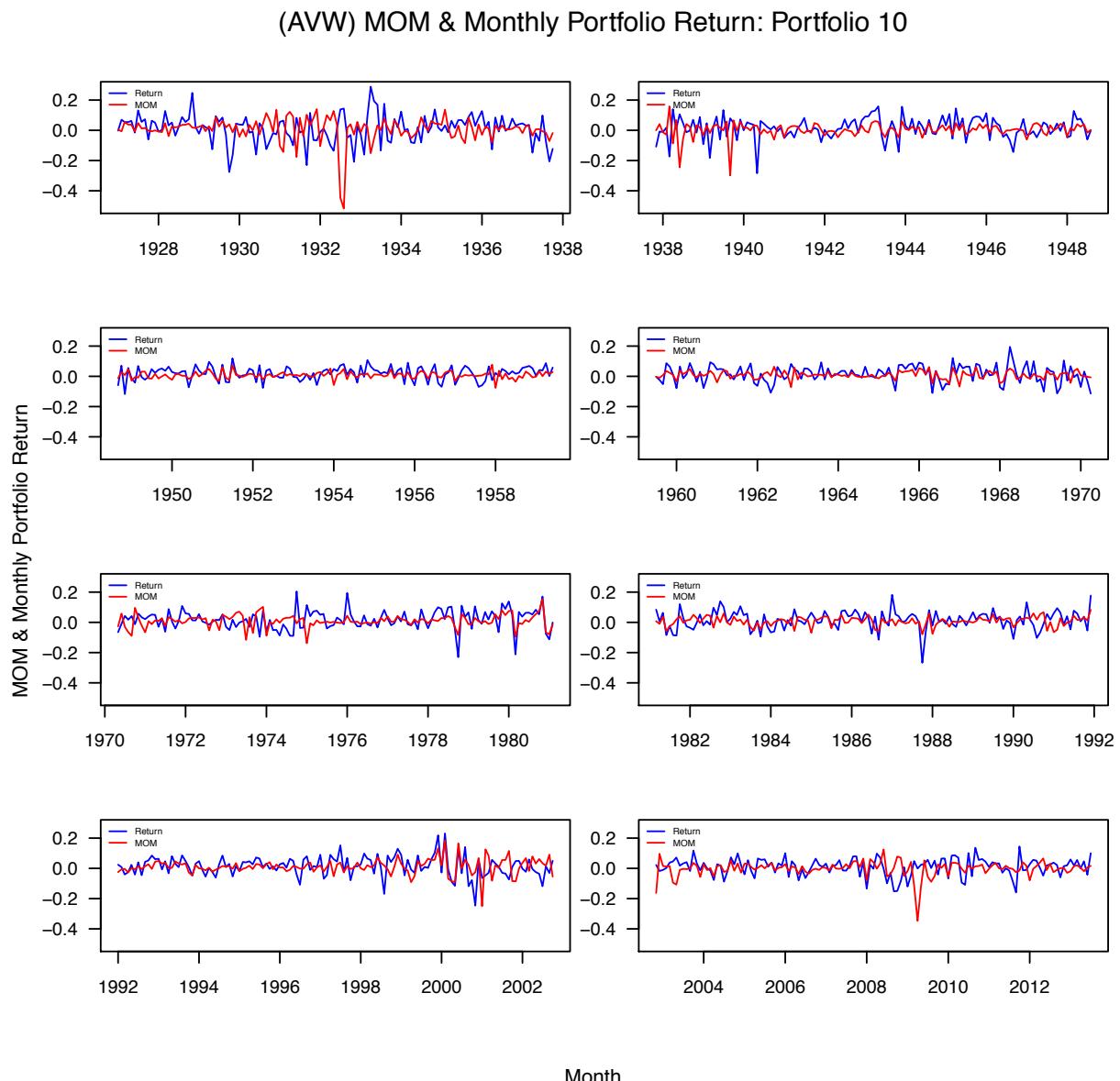


F.19 (AEW) Portfolio 10

(AEW) MOM & Monthly Portfolio Return: Portfolio 10



F.20 (AVW) Portfolio 10



Appendix G

R Codes

```
#####
### The Beginning of R codes for functions and variables
#####
### Import the data #####
#####
library(sandwich)
library(lmtest)
library(tseries)
library(xtable)
library(nortest)
library(car)
### Import the data
portfolios.e <- read.table("10_Portfolios_AveEquWeight.txt", header=T,
quote="\n")
portfolios.v <- read.table("10_Portfolios_AveValWeight.txt", header=T,
quote="\n")
researchfactor <- read.table("F-F_Research_Data_Factors.txt", header=T,
quote="\n")
momfactor <- read.table("F-F_Momentum_Factor.TXT", header=T,
quote="\n")
### Create function to combine the data
dataFun <- function(portfolios, researchfactor, momfactor){
  ### Combine the data
  data <- cbind(portfolios, researchfactor[,2:4], momfactor[,2],
researchfactor[,5])
  ### Rename the variables
  names(data)[2:11] <- paste("Portfolio", 1:10, sep="")
  names(data)[12] <- "RMRF"
```

```

names(data)[15:16] <- c("MOM", "RF")
### Convert the data in percentage
data[,2:16] <- data[,2:16] / 100
### Compute the excess portfolio returns
data[,2:11] <- data[,2:11] - data[,16]
### Convert the month from numeric to date format
month.temp <- as.character(data[,1])
data[,1] <- paste(substring(month.temp, 1, 4),
substring(month.temp, 5, 6), "01", sep="-")
data[,1] <- as.Date(data[,1])
return (data)
}

### Get the average equal weighted (AEW) data with excess return
data.e <- dataFun(portfolios.e, researchfactor, momfactor)
### Get the average value weighted (AVW) data with excess return
data.v <- dataFun(portfolios.v, researchfactor, momfactor)
### Data with real return
pdata.e <- data.e
pdata.v <- data.v
pdata.e[,2:11] <- pdata.e[,2:11] + pdata.e[,16]
pdata.v[,2:11] <- pdata.v[,2:11] + pdata.v[,16]
### List of data with excess return
data.l <- list(data.e, data.v)
### List of data with real return
pdata.l <- list(pdata.e, pdata.v)
#####
### Create functions and variables for Chapter 1
#####
### Create function to generate residual plots
residplotFun <- function(datalist, datalistnum, portfolio){
  data <- datalist[[datalistnum]]
  i <- portfolio + 1
  fit <- lm(data[,i] ~ RMRF + SMB + HML + MOM, data = data)
  titlename <- paste(c("(AEW)", "(AVW)")[datalistnum],
                     "Residual Plots: Portfolio", portfolio, sep = " ")
  par(mfrow=c(3,2), oma=c(0,0,4,0), mar=c(5.1,4.1,1.1,2.1), las=1, cex=0.7)
  plot(data[, "Month"], residuals(fit),
        main = "Time Series of Residuals",
        xlab = "Month", ylab = "Residuals",
        type = "l", col="blue")
  plot(fitted(fit), residuals(fit),
        main = "Normality Test: Portfolio", portfolio,
        xlab = "Fitted Values", ylab = "Residuals",
        type = "p", pch=19)
}

```

```

    main = "Fitted vs Residuals",
    xlab = "Fitted", ylab = "Residuals", col = "blue")
abline (h=0, col = "red")
plot(data[, "RMRF"], residuals(fit),
      main = "RMRF vs Residuals",
      xlab = "RMRF", ylab = "Residuals", col = "blue")
abline (h=0, col = "red")
plot(data[, "SMB"], residuals(fit),
      main = "SMB vs Residuals",
      xlab = "SMB", ylab = "Residuals", col = "blue")
abline (h=0, col = "red")
plot(data[, "HML"], residuals(fit),
      main = "HML vs Residuals",
      xlab = "HML", ylab = "Residuals", col = "blue")
abline (h=0, col = "red")
plot(data[, "MOM"], residuals(fit),
      main = "MOM vs Residuals",
      xlab = "MOM", ylab = "Residuals", col = "blue")
abline (h=0, col = "red")
mtext(titlename, 3, 1, outer=TRUE, cex=1)
# Restore default clipping rect
par(mar = c(5, 4, 4, 2) + 0.1)
par(mfrow=c(1,1))
}

#####
### Create function to compute p-value of Lilliefors test for normality
normtestpvalFun <- function(datalist){
  portfolio <- 1:10
  pvalAEW <- c(NA, 10)
  pvalAVW <- c(NA, 10)
  nullAEW <- c(NA, 10)
  nullAVW <- c(NA, 10)
  for(i in 1:10){
    fit1 <- lm(datalist[[1]][,i+1] ~ RMRF + SMB + HML + MOM, data=datalist[[1]])
    resid1 <- residuals(fit1)
    pvalAEW[i] <- lillie.test(resid1)$p.value
    if(pvalAEW[i] < 0.05){
      nullAEW[i] <- "Reject"
    } else {nullAEW[i] <- "Accept"}
    fit2 <- lm(datalist[[2]][,i+1] ~ RMRF + SMB + HML + MOM, data=datalist[[2]])
    resid2 <- residuals(fit2)
  }
}

```

```

pvalAVW[i] <- lillie.test(resid2)$p.value
if(pvalAVW[i] < 0.05){
  nullAVW[i] <- "Reject"
} else {nullAVW[i] <- "Accept"}
}
return(data.frame(portfolio, pvalAEW, pvalAVW, nullAEW, nullAVW))
}

### Compute the p-value
normtestpval.A <- normtestpvalFun(data.l)
#####
### Create function to compute mean of residuals
meanresidFun <- function(datalist){
  portfolio <- 1:10
  meanresidAEW <- c(NA, 10)
  meanresidAVW <- c(NA, 10)
  for(i in 1:10){
    fit1 <- lm(datalist[[1]][,i+1] ~ RMRF + SMB + HML + MOM, data=datalist[[1]])
    meanresidAEW[i] <- mean(residuals(fit1))
    fit2 <- lm(datalist[[2]][,i+1] ~ RMRF + SMB + HML + MOM, data=datalist[[2]])
    meanresidAVW[i] <- mean(residuals(fit2))
  }
  return(data.frame(portfolio, meanresidAEW, meanresidAVW))
}

### Compute the mean of residuals
meanresid.A <- meanresidFun(data.l)
#####
### Compute the p-values of non-constant variance score test
datalist <- data.l
portfolio <- 1:10
pvalAEW <- c(NA, 10)
pvalAVW <- c(NA, 10)
nullAEW <- c(NA, 10)
nullAVW <- c(NA, 10)
for(i in 1:10){
  fit1 <- lm(datalist[[1]][,i+1] ~ RMRF + SMB + HML + MOM,
             data=datalist[[1]])
  pvalAEW[i] <- ncvTest(fit1)$p
  if(pvalAEW[i] < 0.05){
    nullAEW[i] <- "Reject"
  } else {nullAEW[i] <- "Accept"}
  fit2 <- lm(datalist[[2]][,i+1] ~ RMRF + SMB + HML + MOM,

```

```

        data=datalist[[2]])
pvalAVW[i] <- ncvTest(fit2)$p
if(pvalAVW[i] < 0.05){
  nullAVW[i] <- "Reject"
} else {nullAVW[i] <- "Accept"}
}
ncvtestpval.A <- data.frame(portfolio, pvalAEW, pvalAVW, nullAEW, nullAVW)
rm(datalist, portfolio, pvalAEW, pvalAVW, nullAEW, nullAVW, i, fit1, fit2)
#####
#### Create functions and variables for Chapter 2
#####
#### Create function to plot Cook's distance
CiPlotFun <- function(datalist, datalistnum, portfolio, toplot = TRUE){
  data <- datalist[[datalistnum]]
  colnum <- portfolio + 1
  N <- nrow(data)
  X <- as.matrix(data[,12:15])
  X <- cbind(as.matrix(rep(1,N)), X)
  p <- ncol(X)
  Y <- as.matrix(data[,colnum, drop=FALSE])
  betahat <- solve(t(X) %*% X) %*% t(X) %*% Y
  Yhat <- X %*% betahat
  P <- X %*% solve(t(X) %*% X) %*% t(X)
  resid <- Y - Yhat
  sigma2hat <- (t(resid) %*% resid) / (N - p)
  ti <- as.matrix(rep(1,N))
  Ci <- as.matrix(rep(1,N))
  for(j in 1:N){
    ti[j] <- resid[j] / (sqrt(sigma2hat) * sqrt(1 - P[j,j]))
    Ci[j] <- ( (ti[j])^2 ) / p * ( P[j,j] / (1 - P[j,j]) )
  }
  calpt.find <- Ci > 1
  Month <- as.matrix(data[,1])
  Cidata <- cbind(Month, Ci)
  colnames(Cidata) <- c("Month", "Ci")
  Cidata <- as.data.frame(Cidata)
  Cidata[,1] <- as.Date(Cidata[,1])
  calpt.count <- sum(calpt.find)
  Cidata.sort <- Cidata[order(-Ci),]
  dataname <- c("(AEW)", "(AVW)") [datalistnum]
  if(toplot == TRUE){

```

```

plot(Cidata[, "Month"], Cidata[, "Ci"],
  main = paste(dataname, "\nCook's Distance: Portfolio", portfolio,
  sep = " "),
  xlab = "Month", ylab = "Cook's Distance", col="blue")
points(Cidata[, "Month"] [calpt.find], Cidata[, "Ci"] [calpt.find],
  col="red", pch=19)
} else {
  hist(Cidata[, "Ci"], breaks = 100,
  col = "skyblue", xlab = "Cook's Distance",
  main = paste(dataname, "Histogram\nCook's Distance: Portfolio",
  portfolio, sep = " "))
}
#####
#### Create function to plot covariance ratio
CVRiPlotFun <- function(datalist, datalistnum, portfolio, toplot = TRUE){
  data <- datalist[[datalistnum]]
  colnum <- portfolio + 1
  N <- nrow(data)
  X <- as.matrix(data[,12:15])
  X <- cbind(as.matrix(rep(1,N)), X)
  p <- ncol(X)
  Y <- as.matrix(data[,colnum, drop=FALSE])
  betahat <- solve(t(X) %*% X) %*% t(X) %*% Y
  Yhat <- X %*% betahat
  P <- X %*% solve(t(X) %*% X) %*% t(X)
  resid <- Y - Yhat
  sigma2hat <- (t(resid) %*% resid) / (N - p)
  sigma2hat.i <- as.matrix(rep(1,N))
  for(j in 1:N){
    sigma2hat.i[j] <- (N - p)*sigma2hat/(N - p -1) -
      resid[j]^2/((N - p - 1)*(1 - P[j,j]))
  }
  CVRi <- as.matrix(rep(1,N))
  for(j in 1:N){
    CVRi[j] <- ((sigma2hat.i[j]/sigma2hat)^p) / (1 - P[j,j])
  }
  calpt.find <- abs(CVRi - 1 ) > 3*p/N
  abs.CVRi_1 <- abs(CVRi - 1 )
  calpt.crit <- 3*p/N
  Month <- as.matrix(data[,1])
}

```

```

CVRidata <- data.frame(Month, CVRi, abs.CVRi_1, calpt.crit)
CVRidata[,1] <- as.Date(CVRidata[,1])
calpt.count <- sum(calpt.find)
CVRidata.sort <- CVRidata[order(-abs.CVRi_1),]
dataname <- c("(AEW)", "(AVW)")[datalistnum]
if(toplot == TRUE){
  plot(CVRidata$Month, CVRidata$CVRi,
    main = paste(dataname, "\nCovariance Ratio: Portfolio", portfolio,
    sep = " "),
    xlab = "Month", ylab = "Covariance Ratio", col="blue")
  points(CVRidata$Month[calpt.find], CVRidata$CVRi[calpt.find], col="red",
    pch=19)
} else {
  hist(CVRidata$CVRi, breaks = 100,
    col = "skyblue", xlab = "Covariance Ratio",
    main = paste(dataname, "Histogram\nCovariance Ratio: Portfolio",
    portfolio, sep = " "))
}
#####
#### Create function to combine plots of Cook's distance and covariance ratio
toplotFun <- function(datalist, portfolio, toplot = TRUE){
  if(toplot == TRUE){
    par(mfcol=c(2,2))
    for(i in 1:2){
      CiPlotFun(datalist, i, portfolio)
      CVRiPlotFun(datalist, i, portfolio)
    }
    par(mfrow=c(1,1))
  } else {
    par(mfcol=c(4,1))
    for(i in 1:2){
      CiPlotFun(datalist, i, portfolio, FALSE)
    }
    for(i in 1:2){
      CVRiPlotFun(datalist, i, portfolio, FALSE)
    }
    par(mfrow=c(1,1))
  }
}
#####

```

```

#### Create function to plot MOM vs monthly portfolio returns
momreturnplotFun <- function(datalist, datalistnum, portfolio){
  data <- datalist[[datalistnum]]
  i <- c(1:7)
  i <- 130 * i
  start <- c(1, i + 1)
  end <- c(i, nrow(data))
  j <- portfolio + 1
  temp <- data[,j] + data[,16]
  ymin <- min(min(temp), min(data[,15]))
  ymax <- max(max(temp), max(data[,15]))
  titlename <- paste(c("(AEW)", "(AVW)")[datalistnum],
  "MOM & Monthly Portfolio Return: Portfolio", portfolio,
  sep = " ")
  par(mfrow=c(4,2), oma=c(3,3,4,0), mar=c(4,2,1,1), las=1, cex=0.7)
  for(k in 1:8){
    plot(data[start[k]:end[k],1], temp[start[k]:end[k]],
      xlab = "", ylab = "", ylim=c(ymin,ymax), type="l", col = "blue")
    lines(data[start[k]:end[k],1], data[start[k]:end[k],15], type="l",
      col = "red")
    legend("topleft", c("Return", "MOM"), cex=0.5, col=c("blue", "red"),
      lty=1, bty="n")
  }
  mtext("Month", 1, 1, outer=TRUE, cex = 0.75)
  mtext("MOM & Monthly Portfolio Return", 2, 1, outer=TRUE, las=0,
  cex = 0.75)
  mtext(titlename, 3, 1, outer=TRUE, cex=1)
  # Restore default clipping rect
  par(mar = c(5, 4, 4, 2) + 0.1)
  par(mfrow=c(1,1))
}
#####
#### Create function to compute the Cook's distance for Influential Points
CiCalptFun <- function(data, colnum){
  N <- nrow(data)
  X <- as.matrix(data[,12:15])
  X <- cbind(as.matrix(rep(1,N)), X)
  p <- ncol(X)
  Y <- as.matrix(data[,colnum, drop=FALSE])
  betahat <- solve(t(X) %*% X) %*% t(X) %*% Y
  Yhat <- X %*% betahat
}

```

```

P <- X %*% solve(t(X) %*% X) %*% t(X)
resid <- Y - Yhat
sigma2hat <- (t(resid) %*% resid) / (N - p)
ti <- as.matrix(rep(1,N))
Ci <- as.matrix(rep(1,N))
for(j in 1:N){
  ti[j] <- resid[j] / (sqrt(sigma2hat) * sqrt(1 - P[j,j]))
  Ci[j] <- ((ti[j])^2) / p * (P[j,j] / (1 - P[j,j]))
}
calpt.find <- Ci > 1
Month <- as.matrix(data[,1])
Cidata <- cbind(Month, Ci)
colnames(Cidata) <- c("Month", "Ci")
Cidata <- as.data.frame(Cidata)
Cidata[,1] <- as.Date(Cidata[,1])
calpt.count <- sum(calpt.find)
Cidata.sort <- Cidata[order(-Ci),]
if(calpt.count == 0){
  return(NULL)
} else {
  return(Cidata.sort[1:calpt.count,])
}
}

#####
### Create function to compute the covariance ratios for Influential Points
CVRiCalptFun <- function(data, colnum){
  N <- nrow(data)
  X <- as.matrix(data[,12:15])
  X <- cbind(as.matrix(rep(1,N)), X)
  p <- ncol(X)
  Y <- as.matrix(data[,colnum, drop=FALSE])
  betahat <- solve(t(X) %*% X) %*% t(X) %*% Y
  Yhat <- X %*% betahat
  P <- X %*% solve(t(X) %*% X) %*% t(X)
  resid <- Y - Yhat
  sigma2hat <- (t(resid) %*% resid) / (N - p)
  sigma2hat.i <- as.matrix(rep(1,N))
  for(j in 1:N){
    sigma2hat.i[j] <- (N - p)*sigma2hat/(N - p - 1) -
      resid[j]^2/((N - p - 1)*(1 - P[j,j]))
  }
}

```

```

CVRi <- as.matrix(rep(1,N))
for(j in 1:N){
  CVRi[j] <- ((sigma2hat.i[j]/sigma2hat)^p) / (1 - P[j,j])
}
calpt.find <- abs(CVRi - 1) > 3*p/N
abs.CVRi_1 <- abs(CVRi - 1)
calpt.crit <- 3*p/N
Month <- as.matrix(data[,1])
CVRidata <- data.frame(Month, CVRi, abs.CVRi_1, calpt.crit)
CVRidata[,1] <- as.Date(CVRidata[,1])
calpt.count <- sum(calpt.find)
CVRidata.sort <- CVRidata[order(-abs.CVRi_1),]
if(calpt.count == 0){
  return(NULL)
} else {
  return(CVRidata.sort[1:calpt.count,])
}
#####
### Create function to compute influence measures of influential observations
exclusioninfoFun <- function(data, portfolio){
  Ci <- as.list(c(1:10))
  CVARi <- as.list(c(1:10))
  for(i in 1:10){
    if(is.null(CiCalptFun(data, i+1)) == TRUE){
      Ci[[i]] <- "None"
    } else {
      Ci[[i]] <- CiCalptFun(data, i+1)
    }
    if(is.null(CVRiCalptFun(data, i+1)) == TRUE){
      CVARi[[i]] <- "None"
    } else {
      CVARi[[i]] <- CVRiCalptFun(data, i+1)
    }
  }
  exclusion.info <- list("Cook's Distance" = Ci[[portfolio]],
    "Covariance Ratio" = CVARi[[portfolio]])
  return(exclusion.info)
}
#####
### Create functions and variables for Chapter 3

```

```

#####
#### Get the excluded months
exclusion.Be <- c()
exclusion.Bv <- c()
exclusion.Ce <- c()
exclusion.Cv <- c()
for(i in 1:10){
  exclusion.Be <- c(exclusion.Be,
  CiCalptFun(data.e, i+1)[,1])
  exclusion.Bv <- c(exclusion.Bv,
  CiCalptFun(data.v, i+1)[,1])
  exclusion.Ce <- c(exclusion.Ce,
  CVRiCalptFun(data.e, i+1)[,1])
  exclusion.Cv <- c(exclusion.Cv,
  CVRiCalptFun(data.v, i+1)[,1])
}
exclusion.Be <- as.Date(sort(unique(exclusion.Be)))
exclusion.Bv <- as.Date(sort(unique(exclusion.Bv)))
exclusion.Ce <- as.Date(sort(unique(exclusion.Ce)))
exclusion.Cv <- as.Date(sort(unique(exclusion.Cv)))
#####
#### Get the subset lists
subset.Ae <- as.list(c(1:10))
subset.Av <- as.list(c(1:10))
subset.Be <- as.list(c(1:10))
subset.Bv <- as.list(c(1:10))
subset.Ce <- as.list(c(1:10))
subset.Cv <- as.list(c(1:10))
for(i in 1:10){
  subset.Ae[[i]] <- rep(TRUE, nrow(data.e))
  subset.Av[[i]] <- rep(TRUE, nrow(data.v))
  subset.Be[[i]] <- !(data.e[,1] %in% exclusion.Be)
  subset.Bv[[i]] <- !(data.v[,1] %in% exclusion.Bv)
  subset.Ce[[i]] <- !(data.e[,1] %in% exclusion.Ce)
  subset.Cv[[i]] <- !(data.v[,1] %in% exclusion.Cv)
}
#####
## Panel B data
data.Be <- data.e[subset.Be[[1]],] ### AEW excess return
data.Bv <- data.v[subset.Bv[[1]],] ### AVW excess return
pdata.Be <- pdata.e[subset.Be[[1]],] ### AEW real return
pdata.Bv <- pdata.v[subset.Bv[[1]],] ### AVW real return
#####
## List of data

```

```

data.B1 <- list(data.Be, data.Bv)
pdata.B1 <- list(pdata.Be, pdata.Bv)
### Panel C data
data.Ce <- data.e[subset.Ce[[1]],] ### AEW excess return
data.Cv <- data.v[subset.Cv[[1]],] ### AVW excess return
pdata.Ce <- pdata.e[subset.Ce[[1]],] ### AEW real return
pdata.Cv <- pdata.v[subset.Cv[[1]],] ### AVW real return
### List of data
data.Cl <- list(data.Ce, data.Cv)
pdata.Cl <- list(pdata.Ce, pdata.Cv)
#####
### Create function to compute the statistics
olsFun <- function(data, portfolio){
  varindex <- portfolio + 1
  fit <- lm(data[,varindex] ~ RMRF + SMB + HML + MOM, data=data)
  varnames <- c("Intercept", "RMRF", "SMB", "HML", "MOM")
  beta <- matrix(NA, 1, 5)
  beta[1,] <- fit$coefficients
  colnames(beta) <- varnames
  beta.t <- matrix(NA, 1, 5)
  beta.t[1,] <- coeftest(fit)[,"t value"]
  colnames(beta.t) <- varnames
  beta.NWt <- matrix(NA, 1, 5)
  beta.NWt[1,] <- coeftest(fit, NeweyWest(fit))[, "t value"]
  colnames(beta.NWt) <- varnames
  return(list("Beta" = beta, "Beta t-statistic" = beta.t,
             "Beta Newey-West t-statistic" = beta.NWt))
}
#####
### Create function to generate the panel summary
panelsummaryFun <- function(data, subset){
  panel.sum <- olsFun(data[subset[[1]], ], 1)
  for(i in 2:10){
    stats.temp <- olsFun(data[subset[[i]], ], i)
    for(j in 1:3){
      panel.sum[[j]] <- rbind(panel.sum[[j]], stats.temp[[j]])
    }
  }
  Portfolio <- 1:10
  for(j in 1:3){
    panel.sum[[j]] <- cbind(Portfolio, panel.sum[[j]])
  }
}

```

```

        }
      return(panel.sum)
}

#####
### Generate summary for each panel
### AEW
panelAe.summary <- panelsummaryFun(data.e, subset.Ae)
panelBe.summary <- panelsummaryFun(data.e, subset.Be)
panelCe.summary <- panelsummaryFun(data.e, subset.Ce)
### AVW
panelAv.summary <- panelsummaryFun(data.v, subset.Av)
panelBv.summary <- panelsummaryFun(data.v, subset.Bv)
panelCv.summary <- panelsummaryFun(data.v, subset.Cv)
#####
### (AEW) Beta
### Panel A
beta.Ae <- panelAe.summary[[1]]
### Panel B
beta.Be <- panelBe.summary[[1]]
### Percentage change from panel A to panel B
beta.ABe <- ((panelBe.summary[[1]] - panelAe.summary[[1]]) /
  panelAe.summary[[1]]) * 100
beta.ABe[,1] <- 1:10
#####
### (AVW) Beta
### Panel A
beta.Av <- panelAv.summary[[1]]
### Panel B
beta.Bv <- panelBv.summary[[1]]
### Percentage change from panel A to panel B
beta.ABv <- ((panelBv.summary[[1]] - panelAv.summary[[1]]) /
  panelAv.summary[[1]]) * 100
beta.ABv[,1] <- 1:10
#####
### (AEW) Beta t-statistic
### Panel A
beta_t.Ae <- panelAe.summary[[2]]
### Panel C
beta_t.Ce <- panelCe.summary[[2]]
### Percentage change from panel A to panel C
beta_t.ACe <- ((panelCe.summary[[2]] - panelAe.summary[[2]]) /

```

```

panelAe.summary[[2]]) * 100
beta_t.ACe[,1] <- 1:10
#####
### (AEW) Beta Newey-West t-statistic
### Panel A
beta_nwt.Ae <- panelAe.summary[[3]]
### Panel C
beta_nwt.Ce <- panelCe.summary[[3]]
### Percentage change from panel A to panel C
beta_nwt.ACe <- ((panelCe.summary[[3]] - panelAe.summary[[3]]) /
panelAe.summary[[3]]) * 100
beta_nwt.ACe[,1] <- 1:10
#####
### (AVW) Beta t-statistic
### Panel A
beta_t.Av <- panelAv.summary[[2]]
### Panel C
beta_t.Cv <- panelCv.summary[[2]]
### Percentage change from panel A to panel C
beta_t.ACv <- ((panelCv.summary[[2]] - panelAv.summary[[2]]) /
panelAv.summary[[2]]) * 100
beta_t.ACv[,1] <- 1:10
#####
### (AVW) Beta Newey-West t-statistic
### Panel A
beta_nwt.Av <- panelAv.summary[[3]]
### Panel C
beta_nwt.Cv <- panelCv.summary[[3]]
### Percentage change from panel A to panel C
beta_nwt.ACv <- ((panelCv.summary[[3]] - panelAv.summary[[3]]) /
panelAv.summary[[3]]) * 100
beta_nwt.ACv[,1] <- 1:10
#####
### Examine the OLS assumptions for panels B and C
### Panel B
### Compute the p-Values of Lilliefors test for normality
normtestpval.B <- normtestpvalFun(data.B1)
### Compute the p-values of non-constant variance score test
datalist <- data.B1
portfolio <- 1:10
pvalAEW <- c(NA, 10)

```

```

pvalAVW <- c(NA, 10)
nullAEW <- c(NA, 10)
nullAVW <- c(NA, 10)
for(i in 1:10){
  fit1 <- lm(datalist[[1]][,i+1] ~ RMRF + SMB + HML + MOM,
             data=datalist[[1]])
  pvalAEW[i] <- ncvTest(fit1)$p
  if(pvalAEW[i] < 0.05){
    nullAEW[i] <- "Reject"
  } else {nullAEW[i] <- "Accept"}
  fit2 <- lm(datalist[[2]][,i+1] ~ RMRF + SMB + HML + MOM,
             data=datalist[[2]])
  pvalAVW[i] <- ncvTest(fit2)$p
  if(pvalAVW[i] < 0.05){
    nullAVW[i] <- "Reject"
  } else {nullAVW[i] <- "Accept"}
}
ncvtestpval.B <- data.frame(portfolio, pvalAEW, pvalAVW, nullAEW, nullAVW)
rm(datalist, portfolio, pvalAEW, pvalAVW, nullAEW, nullAVW, i, fit1, fit2)
### Panel C
### Compute the p-Values of Lilliefors test for normality
normtestpval.C <- normtestpvalFun(data.C1)
### Compute the p-values of non-constant variance score test
datalist <- data.C1
portfolio <- 1:10
pvalAEW <- c(NA, 10)
pvalAVW <- c(NA, 10)
nullAEW <- c(NA, 10)
nullAVW <- c(NA, 10)
for(i in 1:10){
  fit1 <- lm(datalist[[1]][,i+1] ~ RMRF + SMB + HML + MOM,
             data=datalist[[1]])
  pvalAEW[i] <- ncvTest(fit1)$p
  if(pvalAEW[i] < 0.05){
    nullAEW[i] <- "Reject"
  } else {nullAEW[i] <- "Accept"}
  fit2 <- lm(datalist[[2]][,i+1] ~ RMRF + SMB + HML + MOM,
             data=datalist[[2]])
  pvalAVW[i] <- ncvTest(fit2)$p
  if(pvalAVW[i] < 0.05){
    nullAVW[i] <- "Reject"
  }
}

```

```

} else {nullAVW[i] <- "Accept"}
}
ncvtestpval.C <- data.frame(portfolio, pvalAEW, pvalAVW, nullAEW, nullAVW)
rm(datalist, portfolio, pvalAEW, pvalAVW, nullAEW, nullAVW, i, fit1, fit2)
#####
#### Create functions and variables for Chapter 4
#####
#### Create function to compute mean and sd of monthly portfolio return
statFun <- function(pdata, subset){
  statsummary <- matrix(NA, 10, 3)
  rownames(statsummary) <- rep("", 10)
  colnames(statsummary) <- c("Portfolio", "Mean", "SD")
  statsummary[,1] <- 1:10
  for(i in 1:10){
    statsummary[i,2] <- mean(pdata[subset[[i]]], i+1)
    statsummary[i,3] <- sd(pdata[subset[[i]]], i+1)
  }
  return(statsummary)
}
#####
#### (AEW) Mean and sd of return
#####
### Panel A
meansd.Ae <- statFun(pdata.e, subset.Ae)
### Panel B
meansd.Be <- statFun(pdata.e, subset.Be)
### Panel C
meansd.Ce <- statFun(pdata.e, subset.Ce)
#####
### Percentage change from panel A to panel B
meansd.ABe <- ((meansd.Be - meansd.Ae) / meansd.Ae) * 100
meansd.ABe[,1] <- 1:10
#####
### Percentage change from panel A to panel C
meansd.ACe <- ((meansd.Ce - meansd.Ae) / meansd.Ae) * 100
meansd.ACe[,1] <- 1:10
#####
#### (AVW) Mean and sd of return
#####
### Panel A
meansd.Av <- statFun(pdata.v, subset.Av)
### Panel B
meansd.Bv <- statFun(pdata.v, subset.Bv)
### Panel C
meansd.Cv <- statFun(pdata.v, subset.Cv)

```

```

#### Percentage change from panel A to panel B
meansd.ABv <- ((meansd.Bv - meansd.Av) / meansd.Av) * 100
meansd.ABv[,1] <- 1:10
#### Percentage change from panel A to panel C
meansd.ACv <- ((meansd.Cv - meansd.Av) / meansd.Av) * 100
meansd.ACv[,1] <- 1:10
#####
##### Create function to compute numbers for mean-variance optimal portfolio
effpfFun <- function(pdata, mu.min, mu.max, nobs, toshorts = TRUE){
  pfoptFun <- function(mu, pdata, toshorts){
    pfmu <- matrix(colMeans(pdata[,2:11]), nrow=1)
    pfcov <- cov(pdata[,2:11])
    if(toshorts == TRUE){
      pfopt.sol <- portfolio.optim(x = pfmu, pm = mu,
        covmat = pfcov, shorts = toshorts)
    } else {
      pfopt.sol <- portfolio.optim(x = pfmu, pm = mu,
        covmat = pfcov, shorts = toshorts,
        reslow = rep(0,10), reshight = rep(1,10))
    }
    wt <- pfopt.sol$pw
    SD <- sqrt((t(wt) %*% pfcov %*% wt)[1])
    return(c(mu, SD, wt))
  }
  mu.list <- seq(mu.min, mu.max, length.out = nobs)
  effpf <- t(sapply(mu.list, pfoptFun, pdata, toshorts))
  colnames(effpf) <- c("Mean", "SD", paste("W", 1:10, sep=""))
  rownames(effpf) <- rep("", nrow(effpf))
  return(effpf)
}
#####
#### (AEW) Mean-Variance Optimal Portfolio
#### Mean of each monthly portfolio return
pfmu.e <- matrix(colMeans(pdata.e[,2:11]), nrow=1)
#### Max and min of the mean
pmin.e <- min(pfmu.e) + 1e-4
pmax.e <- max(pfmu.e) - 1e-4
#### Mean-Variance Optimal Portfolio
effpfT.e <- effpfFun(pdata.e, 0, 0.5, 1000, TRUE)
effpfF.e <- effpfFun(pdata.e, pmin.e, pmax.e, 100, FALSE)
#### Subset for nicer layout

```

```

min_x.e <- min(effpfT.e[,2], effpfF.e[,2])
max_x.e <- max(effpfF.e[,2])
effpfT.e <- effpfT.e[effpfT.e[,2] <= max_x.e,]
#####
#### (AVW) Mean-Variance Optimal Portfolio
#### Mean of each monthly portfolio return
pfmu.v <- matrix(colMeans(pdata.v[,2:11]), nrow=1)
#### Max and min of the mean
pmin.v <- min(pfmu.v) + 1e-4
pmax.v <- max(pfmu.v) - 1e-4
#### Mean-Variance Optimal Portfolio
effpfT.v <- effpfFun(pdata.v, 0, 0.5, 1000, TRUE)
effpfF.v <- effpfFun(pdata.v, pmin.v, pmax.v, 100, FALSE)
#### Subset for nicer layout
min_x.v <- min(effpfT.v[,2], effpfF.v[,2])
max_x.v <- max(effpfF.v[,2])
effpfT.v <- effpfT.v[effpfT.v[,2] <= max_x.v,]
#####
#### (Shorts Allowed) Mean-Variance Optimal Portfolio: AEW vs AVW
pmin <- 0
pmax <- 0.03
effpf.e <- effpfFun(pdata.e, pmin, pmax, 31, TRUE)
effpf.v <- effpfFun(pdata.v, pmin, pmax, 31, TRUE)
min_x.ev <- min(effpf.e[,2], effpf.v[,2])
max_x.ev <- max(effpf.e[,2], effpf.v[,2])
#####
#### The End of R codes for functions and variables
#####
#### The Beginning of R codes for Chapter
#####
#### Chapter 1: Introduction
#####
#### OLS Assumptions
#### Residual Plots
#### (e.g. AEW Portfolio 2)
residplotFun(data.l, 1, 2)
#### P-Values of Lilliefors Test for Normality
normtestpval.A
#### Mean of Residuals
meanresid.A

```

```

#### P-values of Non-constant Variance Score Test
ncvtestpval.A
#####
#### Chapter 2: Identification of Influential Observation
#####
#### Results from Cook's Distance and Covariance Ratio
#####
#### Plot of Cook's Distance and Covariance Ratio
#####
#### (e.g. Portfolio 2)
toplotFun(data.1, 2, TRUE)
#####
#### Histogram of Cook's Distance and Covariance Ratio
#####
#### (e.g. Portfolio 2)
toplotFun(data.1, 2, FALSE)
#####
#### Plot of MOM vs Monthly Portfolio Return
#####
#### (e.g. AEW Portfolio 2)
momreturnplotFun(data.1, 1, 2)
#####
#### Chapter 3: Influence of Influential Observation
#####
#### (AEW) Plots of Beta
par(mfrow=c(3,2))
#####
#### Expand right side of clipping rect to make room for the legend
par(xpd = T, mar = par()$mar + c(0,0,0,5))
min_y <- min(beta.Ae[,2], beta.Be[,2])
max_y <- max(beta.Ae[,2], beta.Be[,2])
plot(beta.Ae[,1], beta.Ae[,2], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "Intercept",
main = "(AEW) Intercept")
axis(1, at=1:10, lab=1:10)
lines(beta.Be[,1], beta.Be[,2], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel B"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta.Ae[,3], beta.Be[,3])
max_y <- max(beta.Ae[,3], beta.Be[,3])
plot(beta.Ae[,1], beta.Ae[,3], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "Beta(RMRF)",
main = "(AEW) Beta of RMRF")
axis(1, at=1:10, lab=1:10)
lines(beta.Be[,1], beta.Be[,3], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel B"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta.Ae[,4], beta.Be[,4])

```

```

max_y <- max(beta.Ae[,4], beta.Be[,4])
plot(beta.Ae[,1], beta.Ae[,4], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "Beta(SMB)",
main = "(AEW) Beta of SMB")
axis(1, at=1:10, lab=1:10)
lines(beta.Be[,1], beta.Be[,4], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel B"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta.Ae[,5], beta.Be[,5])
max_y <- max(beta.Ae[,5], beta.Be[,5])
plot(beta.Ae[,1], beta.Ae[,5], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "Beta(HML)",
main = "(AEW) Beta of HML")
axis(1, at=1:10, lab=1:10)
lines(beta.Be[,1], beta.Be[,5], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel B"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta.Ae[,6], beta.Be[,6])
max_y <- max(beta.Ae[,6], beta.Be[,6])
plot(beta.Ae[,1], beta.Ae[,6], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "Beta(MOM)",
main = "(AEW) Beta of MOM")
axis(1, at=1:10, lab=1:10)
lines(beta.Be[,1], beta.Be[,6], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel B"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta.ABe[,2:6])
max_y <- max(beta.ABe[,2:6])
plot(beta.ABe[,1], beta.ABe[,2], type = "o", col = "red",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "%",
main = "(AEW) % Change from Panel A to Panel B")
axis(1, at=1:10, lab=1:10)
lines(beta.ABe[,1], beta.ABe[,3], type="o", col="green")
lines(beta.ABe[,1], beta.ABe[,4], type="o", col="blue")
lines(beta.ABe[,1], beta.ABe[,5], type="o", col="purple")
lines(beta.ABe[,1], beta.ABe[,6], type="o", col="brown")
lines(1:10, rep(0, 10), lty = 3)
legend(10.5, max_y,
c("Intercept", "Beta(RMRF)", "Beta(SMB)", "Beta(HML)", "Beta(MOM)"),
cex=0.6, col=c("red", "green", "blue", "purple", "brown"),
lty=1, bty="n")

```

```

# Restore default clipping rect
par(mar = c(5, 4, 4, 2) + 0.1)
par(mfrow=c(1,1))

#####
#### (AVW) Plots of Beta
par(mfrow=c(3,2))

#### Expand right side of clipping rect to make room for the legend
par(xpd = T, mar = par()$mar + c(0,0,0,5))
min_y <- min(beta.Av[,2], beta.Bv[,2])
max_y <- max(beta.Av[,2], beta.Bv[,2])
plot(beta.Av[,1], beta.Av[,2], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "Intercept",
main = "(AVW) Intercept")
axis(1, at=1:10, lab=1:10)
lines(beta.Bv[,1], beta.Bv[,2], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel B"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta.Av[,3], beta.Bv[,3])
max_y <- max(beta.Av[,3], beta.Bv[,3])
plot(beta.Av[,1], beta.Av[,3], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "Beta(RMRF)",
main = "(AVW) Beta of RMRF")
axis(1, at=1:10, lab=1:10)
lines(beta.Bv[,1], beta.Bv[,3], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel B"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta.Av[,4], beta.Bv[,4])
max_y <- max(beta.Av[,4], beta.Bv[,4])
plot(beta.Av[,1], beta.Av[,4], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "Beta(SMB)",
main = "(AVW) Beta of SMB")
axis(1, at=1:10, lab=1:10)
lines(beta.Bv[,1], beta.Bv[,4], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel B"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta.Av[,5], beta.Bv[,5])
max_y <- max(beta.Av[,5], beta.Bv[,5])
plot(beta.Av[,1], beta.Av[,5], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "Beta(HML)",
main = "(AVW) Beta of HML")
axis(1, at=1:10, lab=1:10)

```

```

lines(beta.Bv[,1], beta.Bv[,5], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel B"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta.Av[,6], beta.Bv[,6])
max_y <- max(beta.Av[,6], beta.Bv[,6])
plot(beta.Av[,1], beta.Av[,6], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "Beta(MOM)",
main = "(AVW) Beta of MOM")
axis(1, at=1:10, lab=1:10)
lines(beta.Bv[,1], beta.Bv[,6], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel B"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta.ABv[,2:6])
max_y <- max(beta.ABv[,2:6])
plot(beta.ABv[,1], beta.ABv[,2], type = "o", col = "red",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "%",
main = "(AVW) % Change from Panel A to Panel B")
axis(1, at=1:10, lab=1:10)
lines(beta.ABv[,1], beta.ABv[,3], type="o", col="green")
lines(beta.ABv[,1], beta.ABv[,4], type="o", col="blue")
lines(beta.ABv[,1], beta.ABv[,5], type="o", col="purple")
lines(beta.ABv[,1], beta.ABv[,6], type="o", col="brown")
lines(1:10, rep(0, 10), lty = 3)
legend(10.5, max_y,
c("Intercept", "Beta(RMRF)", "Beta(SMB)", "Beta(HML)", "Beta(MOM)"),
cex=0.6, col=c("red", "green", "blue", "purple", "brown"),
lty=1, bty="n")
# Restore default clipping rect
par(mar = c(5, 4, 4, 2) + 0.1)
par(mfrow=c(1,1))
#####
### Plots of Intercept t-statistic
par(mfrow=c(2,2))
### Expand right side of clipping rect to make room for the legend
par(xpd = T, mar = par()$mar + c(0,0,0,5))
min_y <- min(beta_t.Ae[,2], beta_t.Ce[,2])
max_y <- max(beta_t.Ae[,2], beta_t.Ce[,2])
plot(beta_t.Ae[,1], beta_t.Ae[,2], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "T(Intercept)",
main = "(AEW) T-Stats of Intercept")
axis(1, at=1:10, lab=1:10)

```

```

lines(beta_t.Ce[,1], beta_t.Ce[,2], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta_t.Av[,2], beta_t.Cv[,2])
max_y <- max(beta_t.Av[,2], beta_t.Cv[,2])
plot(beta_t.Av[,1], beta_t.Av[,2], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "T(Intercept)",
main = "(AVW) T-Stats of Intercept")
axis(1, at=1:10, lab=1:10)
lines(beta_t.Cv[,1], beta_t.Cv[,2], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta_nwt.Ae[,2], beta_nwt.Ce[,2])
max_y <- max(beta_nwt.Ae[,2], beta_nwt.Ce[,2])
plot(beta_nwt.Ae[,1], beta_nwt.Ae[,2], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "NWT(Intercept)",
main = "(AEW) NW T-Stats of Intercept")
axis(1, at=1:10, lab=1:10)
lines(beta_nwt.Ce[,1], beta_nwt.Ce[,2], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta_nwt.Av[,2], beta_nwt.Cv[,2])
max_y <- max(beta_nwt.Av[,2], beta_nwt.Cv[,2])
plot(beta_nwt.Av[,1], beta_nwt.Av[,2], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "NWT(Intercept)",
main = "(AVW) NW T-Stats of Intercept")
axis(1, at=1:10, lab=1:10)
lines(beta_nwt.Cv[,1], beta_nwt.Cv[,2], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
# Restore default clipping rect
par(mar = c(5, 4, 4, 2) + 0.1)
par(mfrow=c(1,1))
#####
### Plots of RMRF t-statistic
par(mfrow=c(2,2))
### Expand right side of clipping rect to make room for the legend
par(xpd = T, mar = par()$mar + c(0,0,0,5))
min_y <- min(beta_t.Ae[,3], beta_t.Ce[,3])
max_y <- max(beta_t.Ae[,3], beta_t.Ce[,3])
plot(beta_t.Ae[,1], beta_t.Ae[,3], type = "o", col = "blue",

```

```

ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "T(RMRF)",
main = "(AEW) T-Stats of RMRF")
axis(1, at=1:10, lab=1:10)
lines(beta_t.Ce[,1], beta_t.Ce[,3], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta_t.Av[,3], beta_t.Cv[,3])
max_y <- max(beta_t.Av[,3], beta_t.Cv[,3])
plot(beta_t.Av[,1], beta_t.Av[,3], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "T(RMRF)",
main = "(AVW) T-Stats of RMRF")
axis(1, at=1:10, lab=1:10)
lines(beta_t.Cv[,1], beta_t.Cv[,3], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta_nwt.Ae[,3], beta_nwt.Ce[,3])
max_y <- max(beta_nwt.Ae[,3], beta_nwt.Ce[,3])
plot(beta_nwt.Ae[,1], beta_nwt.Ae[,3], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "NWT(RMRF)",
main = "(AEW) NW T-Stats of RMRF")
axis(1, at=1:10, lab=1:10)
lines(beta_nwt.Ce[,1], beta_nwt.Ce[,3], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta_nwt.Av[,3], beta_nwt.Cv[,3])
max_y <- max(beta_nwt.Av[,3], beta_nwt.Cv[,3])
plot(beta_nwt.Av[,1], beta_nwt.Av[,3], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "NWT(RMRF",
main = "(AVW) NW T-Stats of RMRF")
axis(1, at=1:10, lab=1:10)
lines(beta_nwt.Cv[,1], beta_nwt.Cv[,3], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
# Restore default clipping rect
par(mar = c(5, 4, 4, 2) + 0.1)
par(mfrow=c(1,1))
#####
#### Plots of SMB t-statistic
par(mfrow=c(2,2))
### Expand right side of clipping rect to make room for the legend
par(xpd = T, mar = par()$mar + c(0,0,0,5))

```

```

min_y <- min(beta_t.Ae[,4], beta_t.Ce[,4])
max_y <- max(beta_t.Ae[,4], beta_t.Ce[,4])
plot(beta_t.Ae[,1], beta_t.Ae[,4], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "T(SMB)",
main = "(AEW) T-Stats of SMB")
axis(1, at=1:10, lab=1:10)
lines(beta_t.Ce[,1], beta_t.Ce[,4], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta_t.Av[,4], beta_t.Cv[,4])
max_y <- max(beta_t.Av[,4], beta_t.Cv[,4])
plot(beta_t.Av[,1], beta_t.Av[,4], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "T(SMB",
main = "(AVW) T-Stats of SMB")
axis(1, at=1:10, lab=1:10)
lines(beta_t.Cv[,1], beta_t.Cv[,4], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta_nwt.Ae[,4], beta_nwt.Ce[,4])
max_y <- max(beta_nwt.Ae[,4], beta_nwt.Ce[,4])
plot(beta_nwt.Ae[,1], beta_nwt.Ae[,4], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "NWT(SMB)",
main = "(AEW) NW T-Stats of SMB")
axis(1, at=1:10, lab=1:10)
lines(beta_nwt.Ce[,1], beta_nwt.Ce[,4], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta_nwt.Av[,4], beta_nwt.Cv[,4])
max_y <- max(beta_nwt.Av[,4], beta_nwt.Cv[,4])
plot(beta_nwt.Av[,1], beta_nwt.Av[,4], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "NWT(SMB",
main = "(AVW) NW T-Stats of SMB")
axis(1, at=1:10, lab=1:10)
lines(beta_nwt.Cv[,1], beta_nwt.Cv[,4], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
# Restore default clipping rect
par(mar = c(5, 4, 4, 2) + 0.1)
par(mfrow=c(1,1))
#####
#### Plots of HML t-statistic

```

```

par(mfrow=c(2,2))
### Expand right side of clipping rect to make room for the legend
par(xpd = T, mar = par()$mar + c(0,0,0,5))
min_y <- min(beta_t.Ae[,5], beta_t.Ce[,5])
max_y <- max(beta_t.Ae[,5], beta_t.Ce[,5])
plot(beta_t.Ae[,1], beta_t.Ae[,5], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "T(HML)",
main = "(AEW) T-Stats of HML")
axis(1, at=1:10, lab=1:10)
lines(beta_t.Ce[,1], beta_t.Ce[,5], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta_t.Av[,5], beta_t.Cv[,5])
max_y <- max(beta_t.Av[,5], beta_t.Cv[,5])
plot(beta_t.Av[,1], beta_t.Av[,5], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "T(HML)",
main = "(AVW) T-Stats of HML")
axis(1, at=1:10, lab=1:10)
lines(beta_t.Cv[,1], beta_t.Cv[,5], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta_nwt.Ae[,5], beta_nwt.Ce[,5])
max_y <- max(beta_nwt.Ae[,5], beta_nwt.Ce[,5])
plot(beta_nwt.Ae[,1], beta_nwt.Ae[,5], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "NWT(HML)",
main = "(AEW) NW T-Stats of HML")
axis(1, at=1:10, lab=1:10)
lines(beta_nwt.Ce[,1], beta_nwt.Ce[,5], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta_nwt.Av[,5], beta_nwt.Cv[,5])
max_y <- max(beta_nwt.Av[,5], beta_nwt.Cv[,5])
plot(beta_nwt.Av[,1], beta_nwt.Av[,5], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "NWT(HML)",
main = "(AVW) NW T-Stats of HML")
axis(1, at=1:10, lab=1:10)
lines(beta_nwt.Cv[,1], beta_nwt.Cv[,5], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
# Restore default clipping rect
par(mar = c(5, 4, 4, 2) + 0.1)

```

```

par(mfrow=c(1,1))
#####
#### Plots of MOM t-statistic
par(mfrow=c(2,2))
### Expand right side of clipping rect to make room for the legend
par(xpd = T, mar = par()$mar + c(0,0,0,5))
min_y <- min(beta_t.Ae[,6], beta_t.Ce[,6])
max_y <- max(beta_t.Ae[,6], beta_t.Ce[,6])
plot(beta_t.Ae[,1], beta_t.Ae[,6], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "T(MOM)",
main = "(AEW) T-Stats of MOM")
axis(1, at=1:10, lab=1:10)
lines(beta_t.Ce[,1], beta_t.Ce[,6], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta_t.Av[,6], beta_t.Cv[,6])
max_y <- max(beta_t.Av[,6], beta_t.Cv[,6])
plot(beta_t.Av[,1], beta_t.Av[,6], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "T(MOM",
main = "(AVW) T-Stats of MOM")
axis(1, at=1:10, lab=1:10)
lines(beta_t.Cv[,1], beta_t.Cv[,6], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta_nwt.Ae[,6], beta_nwt.Ce[,6])
max_y <- max(beta_nwt.Ae[,6], beta_nwt.Ce[,6])
plot(beta_nwt.Ae[,1], beta_nwt.Ae[,6], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "NWT(MOM)",
main = "(AEW) NW T-Stats of MOM")
axis(1, at=1:10, lab=1:10)
lines(beta_nwt.Ce[,1], beta_nwt.Ce[,6], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),
lty=1, bty="n")
min_y <- min(beta_nwt.Av[,6], beta_nwt.Cv[,6])
max_y <- max(beta_nwt.Av[,6], beta_nwt.Cv[,6])
plot(beta_nwt.Av[,1], beta_nwt.Av[,6], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "NWT(MOM",
main = "(AVW) NW T-Stats of MOM")
axis(1, at=1:10, lab=1:10)
lines(beta_nwt.Cv[,1], beta_nwt.Cv[,6], type="o", col="red")
legend(10.5, max_y, c("Panel A", "Panel C"), cex=0.6, col=c("blue", "red"),

```

```

lty=1, bty="n")
# Restore default clipping rect
par(mar = c(5, 4, 4, 2) + 0.1)
par(mfrow=c(1,1))
#####
#### Plots of % change in t-statistic
par(mfrow=c(2,2))
#### Expand right side of clipping rect to make room for the legend
par(xpd = T, mar = par()$mar + c(0,0,0,5))
min_y <- min(beta_t.ACe[,2:6])
max_y <- max(beta_t.ACe[,2:6])
plot(beta_t.ACe[,1], beta_t.ACe[,2], type = "o", col = "red",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "%",
main = "(AEW) % Change in T-Stats\nFrom Panel A to Panel C")
axis(1, at=1:10, lab=1:10)
lines(beta_t.ACe[,1], beta_t.ACe[,3], type="o", col="green")
lines(beta_t.ACe[,1], beta_t.ACe[,4], type="o", col="blue")
lines(beta_t.ACe[,1], beta_t.ACe[,5], type="o", col="purple")
lines(beta_t.ACe[,1], beta_t.ACe[,6], type="o", col="brown")
lines(1:10, rep(0, 10), lty = 3)
legend(10.5, max_y,
c("Intercept", "T(RMRF)", "T(SMB)", "T(HML)", "T(MOM)",
cex=0.6, col=c("red", "green", "blue", "purple", "brown"),
lty=1, bty="n")
min_y <- min(beta_t.ACv[,2:6])
max_y <- max(beta_t.ACv[,2:6])
plot(beta_t.ACv[,1], beta_t.ACv[,2], type = "o", col = "red",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "%",
main = "(AVW) % Change in T-Stats\nFrom Panel A to Panel C")
axis(1, at=1:10, lab=1:10)
lines(beta_t.ACv[,1], beta_t.ACv[,3], type="o", col="green")
lines(beta_t.ACv[,1], beta_t.ACv[,4], type="o", col="blue")
lines(beta_t.ACv[,1], beta_t.ACv[,5], type="o", col="purple")
lines(beta_t.ACv[,1], beta_t.ACv[,6], type="o", col="brown")
lines(1:10, rep(0, 10), lty = 3)
legend(10.5, max_y,
c("Intercept", "T(RMRF)", "T(SMB)", "T(HML)", "T(MOM)",
cex=0.6, col=c("red", "green", "blue", "purple", "brown"),
lty=1, bty="n")
min_y <- min(beta_nwt.ACe[,2:6])
max_y <- max(beta_nwt.ACe[,2:6])

```

```

plot(beta_nwt.ACe[,1], beta_nwt.ACe[,2], type = "o", col = "red",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "%",
main = "(AEW) % Change in NW T-Stats\nFrom Panel A to Panel C")
axis(1, at=1:10, lab=1:10)
lines(beta_nwt.ACe[,1], beta_nwt.ACe[,3], type="o", col="green")
lines(beta_nwt.ACe[,1], beta_nwt.ACe[,4], type="o", col="blue")
lines(beta_nwt.ACe[,1], beta_nwt.ACe[,5], type="o", col="purple")
lines(beta_nwt.ACe[,1], beta_nwt.ACe[,6], type="o", col="brown")
lines(1:10, rep(0, 10), lty = 3)
legend(10.5, max_y,
c("Intercept", "NWT(RMRF)", "NWT(SMB)", "NWT(HML)", "NWT(MOM)"),
cex=0.6, col=c("red", "green", "blue", "purple", "brown"),
lty=1, bty="n")
min_y <- min(beta_nwt.ACv[,2:6])
max_y <- max(beta_nwt.ACv[,2:6])
plot(beta_nwt.ACv[,1], beta_nwt.ACv[,2], type = "o", col = "red",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "%",
main = "(AVW) % Change in NW T-Stats\nFrom Panel A to Panel C")
axis(1, at=1:10, lab=1:10)
lines(beta_nwt.ACv[,1], beta_nwt.ACv[,3], type="o", col="green")
lines(beta_nwt.ACv[,1], beta_nwt.ACv[,4], type="o", col="blue")
lines(beta_nwt.ACv[,1], beta_nwt.ACv[,5], type="o", col="purple")
lines(beta_nwt.ACv[,1], beta_nwt.ACv[,6], type="o", col="brown")
lines(1:10, rep(0, 10), lty = 3)
legend(10.5, max_y,
c("Intercept", "NWT(RMRF)", "NWT(SMB)", "NWT(HML)", "NWT(MOM)"),
cex=0.6, col=c("red", "green", "blue", "purple", "brown"),
lty=1, bty="n")
# Restore default clipping rect
par(mar = c(5, 4, 4, 2) + 0.1)
par(mfrow=c(1,1))
#####
### OLS Assumptions
### Panel B
#### P-Values of Lilliefors Test for Normality
normtestpval.B
#### P-values of Non-constant Variance Score Test
ncvtestpval.B
### Panel C
#### P-Values of Lilliefors Test for Normality
normtestpval.C

```

```

#### P-values of Non-constant Variance Score Test
ncvtestpval.C

#####
#### Chapter 4: Portfolio Optimization
#####
#### Mean and SD of Monthly Portfolio Returns
#####
#### Plots of Mean and SD
#####
#### Fill by column
par(mfcol=c(3,2))

#### Expand right side of clipping rect to make room for the legend
par(xpd = T, mar = par()$mar + c(0,0,0,3))
min_y <- min(meansd.Ae[,2], meansd.Be[,2], meansd.Ce[,2])
max_y <- max(meansd.Ae[,2], meansd.Be[,2], meansd.Ce[,2])
plot(meansd.Ae[,1], meansd.Ae[,2], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "Mean",
main = "(AEW) Mean of Monthly Portfolio Return")
axis(1, at=1:10, lab=1:10)
lines(meansd.Be[,1], meansd.Be[,2], type="o", col="red")
lines(meansd.Ce[,1], meansd.Ce[,2], type="o", col="green")
legend(10.5, max_y, c("Panel A", "Panel B", "Panel C"), cex=0.6,
col=c("blue", "red", "green"), lty=1, bty="n")
min_y <- min(meansd.Ae[,3], meansd.Be[,3], meansd.Ce[,3])
max_y <- max(meansd.Ae[,3], meansd.Be[,3], meansd.Ce[,3])
plot(meansd.Ae[,1], meansd.Ae[,3], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "SD",
main = "(AEW) SD of Monthly Portfolio Return")
axis(1, at=1:10, lab=1:10)
lines(meansd.Be[,1], meansd.Be[,3], type="o", col="red")
lines(meansd.Ce[,1], meansd.Ce[,3], type="o", col="green")
legend(10.5, max_y, c("Panel A", "Panel B", "Panel C"), cex=0.6,
col=c("blue", "red", "green"), lty=1, bty="n")
min_y <- min(meansd.ABe[,2:3], meansd.ACe[,2:3])
max_y <- max(meansd.ABe[,2:3], meansd.ACe[,2:3])
plot(meansd.ABe[,1], meansd.ABe[,2], type = "o", col = "skyblue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "%",
main = "(AEW) % Change from Panel A to Panels B & C")
axis(1, at=1:10, lab=1:10)
lines(meansd.ACe[,1], meansd.ACe[,2], type="o", col="orange")
lines(meansd.ABe[,1], meansd.ABe[,3], type="o", col="purple")
lines(meansd.ACe[,1], meansd.ACe[,3], type="o", col="brown")
legend(10.5, max_y, c("Mean (B)", "Mean (C)", "SD (B)", "SD (C)"), cex=0.6,

```

```

col=c("skyblue", "orange", "purple", "brown"), lty=1, bty="n")
min_y <- min(meansd.Av[,2], meansd.Bv[,2], meansd.Cv[,2])
max_y <- max(meansd.Av[,2], meansd.Bv[,2], meansd.Cv[,2])
plot(meansd.Av[,1], meansd.Av[,2], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "Mean",
main = "(AVW) Mean of Monthly Portfolio Return")
axis(1, at=1:10, lab=1:10)
lines(meansd.Bv[,1], meansd.Bv[,2], type="o", col="red")
lines(meansd.Cv[,1], meansd.Cv[,2], type="o", col="green")
legend(10.5, max_y, c("Panel A", "Panel B", "Panel C"), cex=0.6,
col=c("blue", "red", "green"), lty=1, bty="n")
min_y <- min(meansd.Av[,3], meansd.Bv[,3], meansd.Cv[,3])
max_y <- max(meansd.Av[,3], meansd.Bv[,3], meansd.Cv[,3])
plot(meansd.Av[,1], meansd.Av[,3], type = "o", col = "blue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "SD",
main = "(AVW) SD of Monthly Portfolio Return")
axis(1, at=1:10, lab=1:10)
lines(meansd.Bv[,1], meansd.Bv[,3], type="o", col="red")
lines(meansd.Cv[,1], meansd.Cv[,3], type="o", col="green")
legend(10.5, max_y, c("Panel A", "Panel B", "Panel C"), cex=0.6,
col=c("blue", "red", "green"), lty=1, bty="n")
min_y <- min(meansd.ABv[,2:3], meansd.ACv[,2:3])
max_y <- max(meansd.ABv[,2:3], meansd.ACv[,2:3])
plot(meansd.ABv[,1], meansd.ABv[,2], type = "o", col = "skyblue",
ylim = c(min_y,max_y), xlab = "Portfolio", ylab = "%",
main = "(AVW) % Change from Panel A to Panels B & C")
axis(1, at=1:10, lab=1:10)
lines(meansd.ACv[,1], meansd.ACv[,2], type="o", col="orange")
lines(meansd.ABv[,1], meansd.ABv[,3], type="o", col="purple")
lines(meansd.ACv[,1], meansd.ACv[,3], type="o", col="brown")
legend(10.5, max_y, c("Mean (B)", "Mean (C)", "SD (B)", "SD (C)"), cex=0.6,
col=c("skyblue", "orange", "purple", "brown"), lty=1, bty="n")
# Restore default clipping rect
par(mar = c(5, 4, 4, 2) + 0.1)
par(mfrow=c(1,1))
#####
### Plots of Mean vs SD
par(mfrow=c(3,2))
min_x <- min(meansd.Ae[,3], meansd.Be[,3], meansd.Ce[,3])
max_x <- max(meansd.Ae[,3], meansd.Be[,3], meansd.Ce[,3])
min_y <- min(meansd.Ae[,2], meansd.Be[,2], meansd.Ce[,2])

```

```

max_y <- max(meansd.Ae[,2], meansd.Be[,2], meansd.Ce[,2])
plot(meansd.Ae[,3], meansd.Ae[,2], type = "o", col = "blue",
ylim = c(min_y,max_y), xlim = c(min_x,max_x),
xlab = "SD", ylab = "Mean", main = "(AEW) Mean vs SD")
lines(meansd.Be[,3], meansd.Be[,2], type="o", col="red")
lines(meansd.Ce[,3], meansd.Ce[,2], type="o", col="brown")
legend("topright", c("Panel A", "Panel B", "Panel C"), cex=0.6,
col=c("blue", "red", "brown"), lty=1, bty="n")
min_x <- min(meansd.Av[,3], meansd.Bv[,3], meansd.Cv[,3])
max_x <- max(meansd.Av[,3], meansd.Bv[,3], meansd.Cv[,3])
min_y <- min(meansd.Av[,2], meansd.Bv[,2], meansd.Cv[,2])
max_y <- max(meansd.Av[,2], meansd.Bv[,2], meansd.Cv[,2])
plot(meansd.Av[,3], meansd.Av[,2], type = "o", col = "green",
ylim = c(min_y,max_y), xlim = c(min_x,max_x),
xlab = "SD", ylab = "Mean", main = "(AVW) Mean vs SD")
lines(meansd.Bv[,3], meansd.Bv[,2], type="o", col="purple")
lines(meansd.Cv[,3], meansd.Cv[,2], type="o", col="orange")
legend("topright", c("Panel A", "Panel B", "Panel C"), cex=0.6,
col=c("green", "purple", "orange"), lty=1, bty="n")
min_x <- min(meansd.Ae[,3], meansd.Av[,3])
max_x <- max(meansd.Ae[,3], meansd.Av[,3])
min_y <- min(meansd.Ae[,2], meansd.Av[,2])
max_y <- max(meansd.Ae[,2], meansd.Av[,2])
plot(meansd.Ae[,3], meansd.Ae[,2], type = "o", col = "blue",
ylim = c(min_y,max_y), xlim = c(min_x,max_x),
xlab = "SD", ylab = "Mean", main = "(Panel A) Mean vs SD")
lines(meansd.Av[,3], meansd.Av[,2], type="o", col="green")
legend("topright", c("AEW", "AVW"), cex=0.6, col=c("blue", "green"),
lty=1, bty="n")
min_x <- min(meansd.Be[,3], meansd.Bv[,3])
max_x <- max(meansd.Be[,3], meansd.Bv[,3])
min_y <- min(meansd.Be[,2], meansd.Bv[,2])
max_y <- max(meansd.Be[,2], meansd.Bv[,2])
plot(meansd.Be[,3], meansd.Be[,2], type = "o", col = "red",
ylim = c(min_y,max_y), xlim = c(min_x,max_x),
xlab = "SD", ylab = "Mean", main = "(Panel B) Mean vs SD")
lines(meansd.Bv[,3], meansd.Bv[,2], type="o", col="purple")
legend("topright", c("AEW", "AVW"), cex=0.6, col=c("red", "purple"),
lty=1, bty="n")
min_x <- min(meansd.Ce[,3], meansd.Cv[,3])
max_x <- max(meansd.Ce[,3], meansd.Cv[,3])

```

```

min_y <- min(meansd.Ce[,2], meansd.Cv[,2])
max_y <- max(meansd.Ce[,2], meansd.Cv[,2])
plot(meansd.Ce[,3], meansd.Ce[,2], type = "o", col = "brown",
ylim = c(min_y,max_y), xlim = c(min_x,max_x),
xlab = "SD", ylab = "Mean", main = "(Panel C) Mean vs SD")
lines(meansd.Cv[,3], meansd.Cv[,2], type="o", col="orange")
legend("topright", c("AEW", "AVW"), cex=0.6, col=c("brown", "orange"),
lty=1, bty="n")
min_x <- min(meansd.Ae[,3], meansd.Av[,3],
meansd.Be[,3], meansd.Bv[,3],
meansd.Ce[,3], meansd.Cv[,3])
max_x <- max(meansd.Ae[,3], meansd.Av[,3],
meansd.Be[,3], meansd.Bv[,3],
meansd.Ce[,3], meansd.Cv[,3])
min_y <- min(meansd.Ae[,2], meansd.Av[,2],
meansd.Be[,2], meansd.Bv[,2],
meansd.Ce[,2], meansd.Cv[,2])
max_y <- max(meansd.Ae[,2], meansd.Av[,2],
meansd.Be[,2], meansd.Bv[,2],
meansd.Ce[,2], meansd.Cv[,2])
plot(meansd.Ae[,3], meansd.Ae[,2], type = "o", col = "blue",
ylim = c(min_y,max_y), xlim = c(min_x,max_x),
xlab = "SD", ylab = "Mean", main = "Mean vs SD")
lines(meansd.Av[,3], meansd.Av[,2], type="o", col="green")
lines(meansd.Be[,3], meansd.Be[,2], type = "o", col = "red")
lines(meansd.Bv[,3], meansd.Bv[,2], type="o", col="purple")
lines(meansd.Ce[,3], meansd.Ce[,2], type = "o", col = "brown")
lines(meansd.Cv[,3], meansd.Cv[,2], type="o", col="orange")
legend("topright",
c("AEW (A)", "AVW (A)", "AEW (B)", "AVW (B)", "AEW (C)", "AVW (C)"),
cex=0.6,
col=c("blue", "green", "red", "purple", "brown", "orange"),
lty=1, bty="n")
par(mfrow=c(1,1))
#####
### Mean-Variance Optimal Portfolio
par(mfrow=c(3,1))
### (AEW) Mean-Variance Optimal Portfolio
plot(effpfT.e[,2], effpfT.e[,1], type = "o", col = "red",
xlim = c(min_x.e,max_x.e), xlab = "Standard Deviation",
ylab = "Expected Monthly Return",

```

```

  main = "(AEW) Mean-Variance Optimal Portfolio")
lines(effpfF.e[,2], effpfF.e[,1], type="o", col="blue")
legend("right", c("Short Allowed", "Long Only"), cex=0.6,
col=c("red", "blue"), lty=1, bty="n")
### (AVW) Mean-Variance Optimal Portfolio
plot(effpfT.v[,2], effpfT.v[,1], type = "o", col = "red",
      xlim = c(min_x.v,max_x.v), xlab = "Standard Deviation",
      ylab = "Expected Monthly Return",
      main = "(AVW) Mean-Variance Optimal Portfolio")
lines(effpfF.v[,2], effpfF.v[,1], type="o", col="blue")
legend("right", c("Short Allowed", "Long Only"), cex=0.6,
col=c("red", "blue"), lty=1, bty="n")
### (Shorts Allowed) Mean-Variance Optimal Portfolio: AEW vs AVW
plot(effpf.e[,2], effpf.e[,1], type = "o", col = "red",
      xlim = c(min_x.ev,max_x.ev), xlab = "Standard Deviation",
      ylab = "Expected Monthly Return",
      main = "(Shorts Allowed) Mean-Variance Optimal Portfolio: AEW vs AVW")
lines(effpf.v[,2], effpf.v[,1], type="o", col="blue")
legend("right", c("AEW", "AVW"), cex=0.6,
col=c("red", "blue"), lty=1, bty="n")
par(mfrow=c(1,1))
### Mean-Variance Optimal Portfolio: Solution and Correlation
print(
xtable(effpf.e,
caption = "(AEW) Mean-Variance Optimal Portfolio: Solution",
label = "table:t4.1", digits = c(0, rep(3, 2), rep(2, 10)),
display = c('e', rep('f', 12))),
table.placement = "htbp", caption.placement = "top",
include.rownames = FALSE, latex.environment = "center")
print(
xtable(effpf.v,
caption = "(AVW) Mean-Variance Optimal Portfolio: Solution",
label = "table:t4.2", digits = c(0, rep(3, 2), rep(2, 10)),
display = c('e', rep('f', 12))),
table.placement = "htbp", caption.placement = "top",
include.rownames = FALSE, latex.environment = "center")
print(
xtable(cor(effpf.e),
caption = "(AEW) Mean-Variance Optimal Portfolio: Correlation",
label = "table:t4.3", digits = c(0, rep(1, 12)),
display = c('e', rep('f', 12))),
```

```

table.placement = "htbp", caption.placement = "top",
latex.environment = "center")
print(
xtable(cor(effpf.v),
caption = "(AVW) Mean-Variance Optimal Portfolio: Correlation",
label = "table:t4.4", digits = c(0, rep(1, 12)),
display = c('e', rep('f', 12))),
table.placement = "htbp", caption.placement = "top",
latex.environment = "center")
### Scatterplot of the Portfolio's Mean, SD, and Weight
pairs(effpfT.e, col="blue")
pairs(effpfT.v, col="blue")
#####
### Chapter 5: Conclusion
#####
### None
#####
### The End of R codes for Chapter
#####
### The Beginning of R codes for Appendix
#####
### Appendix A: Residual Plots
#####
### (AEW) Portfolio 1
portfolio <- 1
residplotFun(data.l, 1, portfolio)
### (AVW) Portfolio 1
portfolio <- 1
residplotFun(data.l, 2, portfolio)
### (AEW) Portfolio 2
portfolio <- 2
residplotFun(data.l, 1, portfolio)
### (AVW) Portfolio 2
portfolio <- 2
residplotFun(data.l, 2, portfolio)
### (AEW) Portfolio 3
portfolio <- 3
residplotFun(data.l, 1, portfolio)
### (AVW) Portfolio 3
portfolio <- 3

```

```
residplotFun(data.l, 2, portfolio)
### (AEW) Portfolio 4
portfolio <- 4
residplotFun(data.l, 1, portfolio)
### (AVW) Portfolio 4
portfolio <- 4
residplotFun(data.l, 2, portfolio)
### (AEW) Portfolio 5
portfolio <- 5
residplotFun(data.l, 1, portfolio)
### (AVW) Portfolio 5
portfolio <- 5
residplotFun(data.l, 2, portfolio)
### (AEW) Portfolio 6
portfolio <- 6
residplotFun(data.l, 1, portfolio)
### (AVW) Portfolio 6
portfolio <- 6
residplotFun(data.l, 2, portfolio)
### (AEW) Portfolio 7
portfolio <- 7
residplotFun(data.l, 1, portfolio)
### (AVW) Portfolio 7
portfolio <- 7
residplotFun(data.l, 2, portfolio)
### (AEW) Portfolio 8
portfolio <- 8
residplotFun(data.l, 1, portfolio)
### (AVW) Portfolio 8
portfolio <- 8
residplotFun(data.l, 2, portfolio)
### (AEW) Portfolio 9
portfolio <- 9
residplotFun(data.l, 1, portfolio)
### (AVW) Portfolio 9
portfolio <- 9
residplotFun(data.l, 2, portfolio)
### (AEW) Portfolio 10
portfolio <- 10
residplotFun(data.l, 1, portfolio)
### (AVW) Portfolio 10
```

```
portfolio <- 10
residplotFun(data.1, 2, portfolio)
#####
#### Appendix B: Plot of Cook's Distance and Covariance ratio
#####
##### Portfolio 1
toplotFun(data.1, 1, TRUE)
##### Portfolio 2
toplotFun(data.1, 2, TRUE)
##### Portfolio 3
toplotFun(data.1, 3, TRUE)
##### Portfolio 4
toplotFun(data.1, 4, TRUE)
##### Portfolio 5
toplotFun(data.1, 5, TRUE)
##### Portfolio 6
toplotFun(data.1, 6, TRUE)
##### Portfolio 7
toplotFun(data.1, 7, TRUE)
##### Portfolio 8
toplotFun(data.1, 8, TRUE)
##### Portfolio 9
toplotFun(data.1, 9, TRUE)
##### Portfolio 10
toplotFun(data.1, 10, TRUE)
#####
#### Appendix C: Histogram of Cook's Distance and Covariance ratio
#####
##### Portfolio 1
toplotFun(data.1, 1, FALSE)
##### Portfolio 2
toplotFun(data.1, 2, FALSE)
##### Portfolio 3
toplotFun(data.1, 3, FALSE)
##### Portfolio 4
toplotFun(data.1, 4, FALSE)
##### Portfolio 5
toplotFun(data.1, 5, FALSE)
##### Portfolio 6
toplotFun(data.1, 6, FALSE)
##### Portfolio 7
```

```
toplotFun(data.l, 7, FALSE)
### Portfolio 8
toplotFun(data.l, 8, FALSE)
### Portfolio 9
toplotFun(data.l, 9, FALSE)
### Portfolio 10
toplotFun(data.l, 10, FALSE)
#####
##### Appendix D: Information for the Excluded Influential Observations
#####
#### (AEW) Portfolio 1
portfolio <- 1
exclusioninfoFun(data.e, portfolio)
### (AVW) Portfolio 1
portfolio <- 1
exclusioninfoFun(data.v, portfolio)
### (AEW) Portfolio 2
portfolio <- 2
exclusioninfoFun(data.e, portfolio)
### (AVW) Portfolio 2
portfolio <- 2
exclusioninfoFun(data.v, portfolio)
### (AEW) Portfolio 3
portfolio <- 3
exclusioninfoFun(data.e, portfolio)
### (AVW) Portfolio 3
portfolio <- 3
exclusioninfoFun(data.v, portfolio)
### (AEW) Portfolio 4
portfolio <- 4
exclusioninfoFun(data.e, portfolio)
### (AVW) Portfolio 4
portfolio <- 4
exclusioninfoFun(data.v, portfolio)
### (AEW) Portfolio 5
portfolio <- 5
exclusioninfoFun(data.e, portfolio)
### (AVW) Portfolio 5
portfolio <- 5
exclusioninfoFun(data.v, portfolio)
### (AEW) Portfolio 6
```

```

portfolio <- 6
exclusioninfoFun(data.e, portfolio)
### (AVW) Portfolio 6
portfolio <- 6
exclusioninfoFun(data.v, portfolio)
### (AEW) Portfolio 7
portfolio <- 7
exclusioninfoFun(data.e, portfolio)
### (AVW) Portfolio 7
portfolio <- 7
exclusioninfoFun(data.v, portfolio)
### (AEW) Portfolio 8
portfolio <- 8
exclusioninfoFun(data.e, portfolio)
### (AVW) Portfolio 8
portfolio <- 8
exclusioninfoFun(data.v, portfolio)
### (AEW) Portfolio 9
portfolio <- 9
exclusioninfoFun(data.e, portfolio)
### (AVW) Portfolio 9
portfolio <- 9
exclusioninfoFun(data.v, portfolio)
### (AEW) Portfolio 10
portfolio <- 10
exclusioninfoFun(data.e, portfolio)
### (AVW) Portfolio 10
portfolio <- 10
exclusioninfoFun(data.v, portfolio)
#####
##### Appendix E: Excluded Months By Panel and Weighting Method #####
#####
list(
"Panel A's Number of Monthly Observations" = nrow(data.e),
"(AEW) Panel B's Excluded Months" = exclusion.Be,
"(AEW) Panel B's Number of Excluded Months" = length(exclusion.Be),
"(AVW) Panel B's Excluded Months" = exclusion.Bv,
"(AVW) Panel B's Number of Excluded Months" = length(exclusion.Bv),
"(AEW) Panel C's Excluded Months" =
matrix(c(as.character(exclusion.Ce), rep("",3)), ncol = 5, byrow = TRUE),
"(AEW) Panel C's Number of Excluded Months" = length(exclusion.Ce),

```

```

"(AVW) Panel C's Excluded Months" =
matrix(c(as.character(exclusion.Cv), rep("",3)), ncol = 5, byrow = TRUE),
"(AVW) Panel C's Number of Excluded Months" = length(exclusion.Cv)
)

#####
#### Appendix F: Plot of MOM vs Monthly Portfolio Return
#####
##### (AEW) Portfolio 1
portfolio <- 1
momreturnplotFun(data.l, 1, portfolio)
##### (AVW) Portfolio 1
portfolio <- 1
momreturnplotFun(data.l, 2, portfolio)
##### (AEW) Portfolio 2
portfolio <- 2
momreturnplotFun(data.l, 1, portfolio)
##### (AVW) Portfolio 2
portfolio <- 2
momreturnplotFun(data.l, 2, portfolio)
##### (AEW) Portfolio 3
portfolio <- 3
momreturnplotFun(data.l, 1, portfolio)
##### (AVW) Portfolio 3
portfolio <- 3
momreturnplotFun(data.l, 2, portfolio)
##### (AEW) Portfolio 4
portfolio <- 4
momreturnplotFun(data.l, 1, portfolio)
##### (AVW) Portfolio 4
portfolio <- 4
momreturnplotFun(data.l, 2, portfolio)
##### (AEW) Portfolio 5
portfolio <- 5
momreturnplotFun(data.l, 1, portfolio)
##### (AVW) Portfolio 5
portfolio <- 5
momreturnplotFun(data.l, 2, portfolio)
##### (AEW) Portfolio 6
portfolio <- 6
momreturnplotFun(data.l, 1, portfolio)
##### (AVW) Portfolio 6

```

```
portfolio <- 6
momreturnplotFun(data.l, 2, portfolio)
### (AEW) Portfolio 7
portfolio <- 7
momreturnplotFun(data.l, 1, portfolio)
### (AVW) Portfolio 7
portfolio <- 7
momreturnplotFun(data.l, 2, portfolio)
### (AEW) Portfolio 8
portfolio <- 8
momreturnplotFun(data.l, 1, portfolio)
### (AVW) Portfolio 8
portfolio <- 8
momreturnplotFun(data.l, 2, portfolio)
### (AEW) Portfolio 9
portfolio <- 9
momreturnplotFun(data.l, 1, portfolio)
### (AVW) Portfolio 9
portfolio <- 9
momreturnplotFun(data.l, 2, portfolio)
### (AEW) Portfolio 10
portfolio <- 10
momreturnplotFun(data.l, 1, portfolio)
### (AVW) Portfolio 10
portfolio <- 10
momreturnplotFun(data.l, 2, portfolio)
#####
##### The End of R codes for Appendix
#####
```

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