

A Four Decade Analysis of Equity Returns in the Japanese Financial Market (2020)

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Abstract

The purpose of this study is to examine returns on Japanese equities over nearly a four-decade period and to compare results among the four decades and the entire period of the study. “Long memory” modeling of time series developed to predict slowly moving time series is a method to predict longtime components of time series data. Previously, other studies indicated some progress in producing results of predictability by these “long memory” analyses. The authors examined statistically for some of the reasons why long memory forecasting may not be very suitable for predicting equity returns over lengthy periods. Data secured from a source that collects information on Japanese equity returns enabled a study of possible explanations of why lengthy predictions are difficult. The analysis is of an application to financial time series and does not dispute the use of long memory modeling in other applications. The conclusions made are therefore not universal but only to the use in financial engineering and time series analysis. Future work should consider the cost effectiveness of long-memory modeling in other forms of financial time series analysis.

Key Words: *Long memory modeling, Time series components, Japanese equity market, Long-run dependence*

Introduction

One studies Japanese equity markets because of the great growth in its equity markets in the 1980s and subsequent fluctuations in later years and the problem of explaining fluctuations in the predictability of monthly stock prices in Japan. Ziemba and Schwartz (1991) and Ziemba and Ziemba (2012) produced evidence indicating that the dynamics of the Japanese markets had significant effects on the prices of Japanese securities. Many of these include Rothlein and Jarrett (2002), Jarrett and Kyper (2005a, 2005b, 2006), Caporale and Gil-Alana (2002), and Kubata and Takehara (2003).

Studies of changes in the Japanese economy in the 1980s through the 1990s and into the new century produce evidence to identify shocks in the Japanese markets leading to changes in the predictability in prices, thus reducing the accuracy of listed Japanese equities. Nagayasu (2000) documented the era when Japan went from great growth in its asset prices to virtual stagnation producing the worst crises in Japan since the outcome of World War II. Furthermore, Ray, Jarrett, and Chen (1997) produced evidence of both temporary and permanent components in the time series of a sample of listed Japanese equities. The last study, using ARFIMA time series methods, identified these components but also indicated some of the great difficulties in predicting prices of Japanese securities. They showed that the inclusion of the temporary component in a sample of 15 individual listed Japanese firms. Japanese equities contain 5% to 15% of permanent components, and thus, there may be a small amount of predictability in

Japanese equity prices. Nagayasu (2003), using the ARFIMA-FGARCH model, studied the efficiency of the Japanese equity market by examining the statistical properties of the return and volatility of the Nikkei 225. He found that there is a long-range dependence. This differs from the notion of the efficient market hypothesis (EMH) and is valid for the sample periods studied. This suggests that the equity market reform of the early 2000s did not produce major efficiency gains.

In addition to the Japanese study noted above, others have made similar studies in other equity markets. These studies include Agiakloglou, Newbold, and Wohar (1992), Baillie (1996), Baillie, Bollerslev, and Mikkelsen (1996), Barkoulas, Baum, and Travlos (2000), Bollerslev and Mikkelsen (1996), Lo (1991, 1997), Mills (1993), Sadique and Silvapulle (2001), Davidson and Li (2015), and hi and Ho (2015). In all, these studies conclude that forecasting stock returns is a very difficult and exceptionally rare case when the future is barely predictable over the long term. This is not surprising but is often the case.

Data Analysis

In this study, we examine the evidence concerning the lack of powerful long memory permanent components in a lengthy period of time series data on returns to the Japanese equity market. As a number of previous studies, we collected data over four decades from the PACAP databases on Japanese equity markets kept at the University of Rhode Island/CBA (e-mail: pacapd@etal.uri.edu).

The purpose of this study is to determine what factors in the time series of Japanese equity prices may cause this great difficulty in prediction. We propose to study the value weighted and equal weighted monthly return over a lengthy period to explain the inability of predicting accurately Japanese equity prices.

In particular, we study the history of equity prices for the Japanese (Tokyo) stock exchange over a lengthy period on what may have caused these prices to have changed during the time span.

Stated differently, long-run dependence (long-run memory) is very important in explaining equity behavior. For example, if long-run memory is present such as in predicting the overflow of a river causing floods beyond its banks, one may take action by increasing the height of the riverbanks to prevent future overflow of the river. This analogy, if present in the time series of equity returns, may permit one to adjust his/her financial decision making to reduce the disturbing effects of serial long-term dependence. Hence, our goal is to examine if serial dependence produced changes that differ in several time periods.

In addition, because we have collected data on Japanese equity over a lengthy period in Japanese currency, we transformed the data in two ways. One set of time series over a four-decade time horizon applies the principle of “value-weighted monthly returns,” identified as MVRMWD in the data set of the time series. A second time series is “equal-weighted monthly” returns identified by the variably name MERMWD. We distinguish the two data sets because the original data in Japanese currency are transformed into U.S. currency and we wish to see if the two methods yield similar results and there is no transformation bias.

Big Data Analytics

First, we analyze and compare the returns for the Japanese stock market for the entire roughly four-decade sample of value weighted monthly returns denoted in the following table and figure as variable MVRMWD. The table consists of five panels—that is, Moments, Descriptive Statistics, and Statistical Tests for Location, Quantiles, and Extreme Observations. In the first panel (Moments), observe that the mean is very small in comparison to the standard deviation. Skewness is slightly negative and kurtosis—that is, the state or quality of flatness or peakedness of the curve describing a frequency distribution in the region about its mode. In this case, a

value of 1.353 or so indicates that there are a few extremes in the data than would be seen in a normal curve whose measure would be about 3. The coefficient of variation of 981.354 (the standard deviation divided by the mean) would indicate that the mean would not be a good predictor because the standard deviation is large with respect to the mean.

The second panel contains the descriptive statistics. Note that the mean and median differ by a factor of about 10:1, indicating that the data are highly skewed, as noted in the first panel on moments. The standard deviation relative to the mean would also be very large and corroborates the findings of the coefficient of variation noted in the first panel. In addition, the other measures of variation—the variance, range, and interquartile range—show also that the data are widely dispersed.

The third panel contains the Tests for Location: $\mu_0 = 0$, which were the t-test, M, and signed rank. All three tests indicated that the null hypothesis should be reject at small p-values of 0.3030, 0.0147, and 0.0102. Hence, the means were nonzero and positive whether they were tested by parametric or rank statistics. Finally, the last two panels of Table 1 show the wide distribution of the data based on quantiles, quartiles, and extreme observations.

For the entire period of the study, the return on Japanese equities tended to be positive but widely distributed and at no time tend to be easily predictable. The question remains, how can one observe the wide distribution over a lengthy period? Hence, in the next sections, we observe the pattern in the roughly four decades of the study to determine why the permanent component of the time series repeats in each new decade.

Table 1
Comparing Japan Stock Market Return by Decade
Value Weighted Monthly Return (MVRMWD)

Moments (Univariate Procedure)			
N	455	Sum Weights	455
Mean	0.00525402	Sum Observations	2.390581
Standard Deviation	0.05156059	Variance	0.00265849
Skewness	0.151716	Kurtosis	1.35303696
Corrected Sum of Squares	1.20695631		
Coefficient of Variation	981.354192	Standard Error of the Mean	0.0024172

Descriptive Statistics			
Location		Variability	
Mean	0.005254	Std. Deviation	0.05156
Median	0.0005576	Variance	0.00266
Interquartile Range	0.05635	Range	0.37645

Tests for Location: $\mu_0 = 0$			
Test	Statistic	P-value	Heading
<i>t</i>	2.1736	Pr > abs (t)	0.0303
M	26.5	Pr \geq abs (M)	0.0147
Signed Rank (S)	7,199	Pr \geq abs (S)	0.0102

Quantiles	
Quantile	Estimate
Maximum	0.1791
99%	0.1367
95%	0.0898
90%	0.0638
75% (Q3)	0.0342
50% (Median, Q2)	0.0056
25% (Q1)	-0.0222
10%	-0.0585
5%	-0.0792

Note: Q is for quartile.

Extreme Observations			
Lowest		Highest	
Value	Observation	Value	Observation
-0.1973	405	0.1367	228
-0.1963	188	0.1410	290
-0.1571	226	0.1464	193
-0.1388	408	0.1579	134
-0.1266	404	0.1791	189

Analysis by Decade

Presented in the Tables 2–5 is the same analysis for each decade referred in the table as a group. Group = 1 refers to the first decade of data collected from the source noted before. In turn, Group = 2, 3, and 4 refer to each new decade. The analysis for each decade is the same as in Table 1. Note that the number of observations is smaller than the entire sample studied; hence, with less degrees of freedom, the significant tests for location may vary. We expect the moments and descriptive statistics to vary as well. By observing the mean rates of return, we observe in Tables 2–5 that the mean rates of return were extremely small but with a declining trend with the largest value in Group = 2. The coefficients of variation again were very large indications of the wide diversity in the mean rates for each firm listed on the exchange. Furthermore, for all the groups, 1, 2, 3, and 4 indicated that the test of hypothesis of mean equals zero could not be rejected. The p-values tended to be very large regardless of whether the test was a t-test or nonparametric analysis was performed.

In addition, the same set of tables indicated the wide diversity in the returns from decade to decade. At no time did we observe a pattern of growth as exhibited in the tables for each decade. The evidence seems to suggest that the Japanese equity market did not appear to have a positive growth during the decades but did exhibit wide variation, skewness, and kurtosis in the distribution of returns. This may suggest that a permanent component in the time series data was not present. Analysis of data suggests that long memory modeling may not be available as a panacea for predicting future returns in the Japanese equity market.

Table 2
Comparing Japan Stock Market Return by Decade Value Weighted Monthly Return (MVRMWD) Group = 1

Moments –(Univariate Procedure)			
N	119	Sum Weights	119
Mean	0.0121775	Sum Observations	1.449122
Standard Deviation	0.03212	Variance	0.00103169
Skewness	0.29980506	Kurtosis	2.02395426
Corrected Sum of Squares	0.12173992		
Coefficient of Variation	263765208	Standard Error of the Mean	0.00294444

Mean	0.012177	Std. Deviation	0.03212
Median	0.009441	Variance	0.00103
		Range	0.22198
		Interquartile Range	0.03269

Tests for Location: $\mu_0 = 0$			
Test	Statistic	P-value	Heading
<i>t</i>	4.135766	Pr > abs (t)	S.0001
M	18.5	Pr abs (M)	0.0009
Signed Rank	1,592	Pr abs (SJ)	0.0001

Quantiles	
Quantile	Estimate
Maximum	0.1235
99%	0.1082
95%	0.0662
90%	0.0509
75% (Q3)	0.0282
50% (Median, Q2)	0.0094
25% (Q1Qi)	-0.0045
10%	-0.0251
5%	-0.392
1%	-0.0618
Minimum	-0.0985

Note: Q is for quartile

Extreme Observations			
Lowest		Highest	
Value	Observation	Value	Observation
-0.0984	112	0.0732	115
-0.0618	80	0.0838	9
-0.0481	85	0.0870	1
-0.0428	6	0.1082	23
-0.0392	114	0.1235	110

Table 3
Comparing Japan Stock Market Return by Decade Value Weighted Monthly Return (MVRMWD)

Group = 2

Moments (Univariate Procedure)			
N	120	Sum Weights	120
Mean	0.0006542	Sum Observations	0.78504
Std. Deviation	0.06311431	Variance	0.00398342
Skewness	0.0226193	Kurtosis	0.98950479
Corrected Sum of Squares	0.47402645		
Coefficient of Variation	964.755513	Std. Error Mean	0.00576152

Descriptive Statistics			
Location		Variability	
Mean	0.006542	Std. Deviation	.06311
Median	0.005635	Variance	0.00398
		Range	0.37541
		Interquartile Range	0.07417

Tests for Location: $\mu_0 = 0$

Test	Statistic	P value	
<i>t</i>	1.135464	Pr > abs (<i>t</i>)	0.2585
M	6	Pr !: abs (M)	0.37541
Signed Rank (S)	456	Pr !: abs (SJ)	0.2339

Quantile	
Quantile	Estimate
Maximum	0.1791
99%	0.1579
95%	0.1290
90%	0.0763
75% (Q3)	0.0409
50% (Median, Q2)	0.0056
25% (Q1)	0.0333
10%	0.0697
5%	0.0976
1%	0.1571
Minimum	0.1963

Note: Q is for quartile.

Extreme Observations

Least		Highest	
Value	Observation	Value	Observation
0.196303	188	0.131426	211
0.157149	226	0.136679	228
0.124507	182	0.0146419	193
0.124503	187	0.157948	134
0.112875	190	0.179105	189

Table 4
Comparing Japan Stock Market Return by Decade Value Weighted Monthly Return (MVRMWD)

Group = 3

Moments (Univariate Procedure)			
N	120	SUM Weights	120
Mean	0.0010967	Sum Observations	0.131604
Std. Deviation	0.05101579	Variance	0.00260261
Skewness	0.19039713	Kurtosis	0.3175241
Corrected Sum of Squares	0.309855		0.30971067
Coefficient of Variation	4651.75421	Std. Error of the Mean	0.00465708

Descriptive Statistics			
Location		Variability	
Mean	0.00110	Std. Deviation	0.05102
Median	0.00309	Variance	0.00260
		Range	0.26458
		Interquartile Range	0.08118

Tests for Location: $\mu_0 = 0$			
Test	Statistic	P-Value	Heading
<i>t</i>	0.235491	Pr > abs (t)	0.8142
M		Pr abs(M)	0.6483
Signed Rank (S)		Pr abs (S)	0.9668

Quantiles	
Quantile	Estimate
Maximum	0.1410
99%	0.1149
95%	0.0838
90%	0.0628
75% (Q3)	0.0405
50% (Median, Q2)	0.0031
25% (Q1Qi)	0.0407
10%	0.0643
5%	0.0776
1%	0.0856
Minimum	0.1235

Note: Q is for quartile

Extreme Observations			
I Lowest		Highest I	
Value	Observation	Value	Observation
0.123537	283	0.102495	286
0.085585	306	0.109559	350
0.084380	318	0.113935	246
0.084048	329	0.774880	293
0.079367	241	0.141038	290

Table 5
Japan Stock Market Return by Decade Value Weighted Monthly Return (MVRMWD)
Group = 4

Moments (Univariate Procedure)			
N	96	Sum Weights	96
Mean	0.000258	Sum Observations	0.02481
Std. Deviation	0.05536	Variance	0.003064
Skewness	0.57034	Kurtosis	1.13855
Coefficient Variation	21,415.1414	Corrected sum of square	0.29111
		Std. Error of the Mean	0.0056497

Descriptive Statistics			
Location		Variability	
Mean	0.000258	Std. Deviation	0.05536
Median	0.003503	Variance	0.00306
Interquartile Range	0.06946	Range	0.31095

Tests for Location: $\mu_0 = 0$

Test	Statistic	P_value	Heading
t	0.045752	Pr > abs (t)	0.9636
M	5	Pr abs (M)	0.3584
Signed Rank	107	Pr abs (SJ)	0.6979

Quantile	Quantiles	
		Estimate
Maximum		0.0036
99%		0.1136
95%		0.0898
90%		0.0720
75% (Q3)		0.0350
50% (Median)		0.0035
25% (Q1)		0.0344
10%		0.0629
5%		0.1027
1%		0.1973
Minimum		0.1973

Note: Q is for quartile

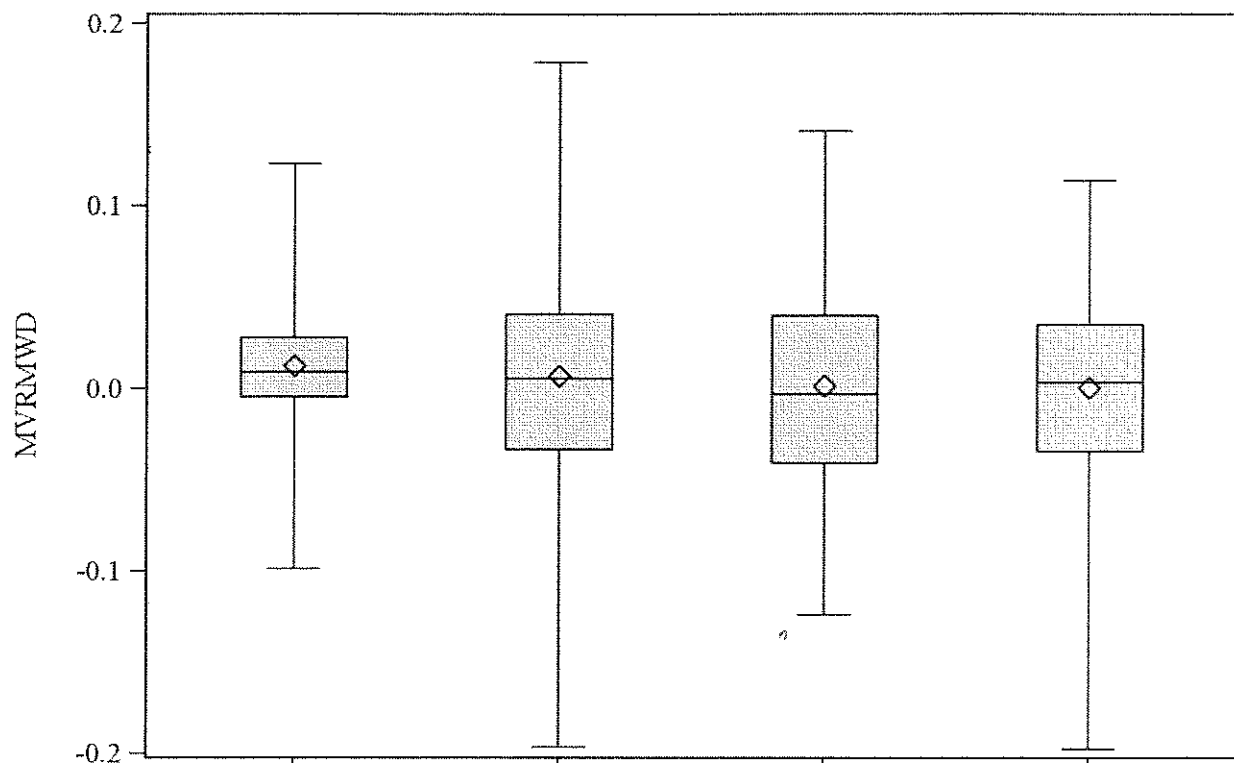
Extreme Observations			
Lowest		Highest	
I			I
Value	Observation	Value	Observation
0.197343	405	0.089771	455
0.138821	408	0.102213	445
0.126562	404	0.103183	422
0.104901	424	0.109199	399
0.102674	448	0.113605	368

Graphical Analysis

Finally, we observe boxplots of the data on Japanese equity returns by decade in Figure 1. From left to right, the boxplot represents mean (diamond), median (the horizontal line through the box), the limits of the interquartile range (upper and lower limits of the box), and the upper lower range of the data for each decade. Note that for decade 1, the box tends to have less variation than for the remaining three decades. The range is also narrower in decade 1 than for the remaining decades. The mean and median return are very near to each other, with decade 2 having a mean and median that are approximate. The boxplot shows that the Japanese equity returns have not shown any growth and are near zero. The boxplot itself shows visually the same conclusion that we observe in the moments, descriptive statistics, hypothesis tests, quantiles, and extreme observation noted in Tables 2–5. We may conclude that the Japanese

equity returns have stagnated over the lengthy study, and most important, the results appear to verify the conclusions from previous studies of long memory modeling. If we were to buy the entire market, we would observe that wealth creation would be very slow. However, investors do not usually buy an entire equity market but instead purchase and sell individual securities or market baskets of securities. The like result is that these market baskets would be difficult to predict their returns if the selection process was random. Forecasting such a market basket of securities may be undesirable. We must be extremely careful in selecting equities to purchase in this market, and this has ramifications for mutual funds and exchange-traded funds.

Figure 1
Boxplot to Compare Japan Stock Market by Decades MVRMWD Variable
MVRMWD by Group 2, 3, 4, and 5



Data Analysis for Equal Weighted Monthly Returns

The analysis of the equal weighted monthly returns (MERMWD) is performed in the same manner as in the previous sections. First, we examine the data for the entire four decades; in turn, we examine the data decade by decade as denoted by Groups 1–4. Finally, we compare the boxplots of the four decades. We begin with Table 6 where the data for the entire four-decade time series sample characteristics appear.

Table 6
Equal Weighted Monthly Return (MERMWD)

Moments (Univariate Procedure)			
N	455	Sum Weights	455

Quantiles			
Quantile		Estimate	
Maximum		0.2563	
99%		0.1509	
95%		0.1011	
90%		0.0742	
75% (Q3)		0.4138	
50% (Median, Q2)		0.0112	
25% (Q1)		0.0215	
10%		0.0583	
5%		0.0855	
1%		0.4167	
Minimum		0.1891	

Mean	0.00878095	Sum Observations	3.995334
Std. Deviation	0.05696745	Variance	0.00324529
Skewness	0.06207574	Kurtosis	1.56248128
Corrected Sum of Squares	1.4733619		
Coefficient of Variation	648.76	Standard Error Of the Mean	0.00267068

Descriptive Statistics			
Location		Variability	
Mean	0.008781	Std. Deviation	0.05697
Median	0.11278	Variance	0.00325
Interquartile Range	0.06286	Range	0.44543

Tests for Location: $\mu_0 = 0$			
Test	Statistic	P_value	P-value
t	3.287915	Pr > abs (t)	0.0011
M	36.5	Pr ≥ abs (M)	0.0007
Signed Rank (S)	10,312	Pr ≥ abs (S)	0.0002

Note: Q is for quantile.

Extreme Observations			
I Lowest		Highest I	
Value	Observation	Value	Observation
0.1891	188	0.1509	228
0.1674	405	0.1663	350
0.1669	226	0.1768	189

0.1581	187	0.2294	193
0.1467	190	0.2563	276

Again, the number of observations is 455 as in Table 1, but the mean rate of return is 0.0088 (rounded to four decimal places), indicating less than a 1-percent rate of return compared with 0.00525 for the value weighted monthly return in Table 1. The coefficient of variation is again very large (648.76); the skewness coefficient is much smaller for the equal weighted monthly returns (0.0621), but the kurtosis is larger at 1.5625 than for the value weighted monthly returns. Hence, this distribution tends to be less skewed but more flattened. The descriptive statistics indicate that this distribution contains less variation about the mean even though it is still large, but the skewness is no longer negative—that is, the median is larger than the mean. The tests for location (t-test, M, and S) all reject the hypothesis that the mean equals zero at very small levels (0.0011, 0.0007, and 0.0002).

The quantiles and quartiles of the data show large distances between each border of the quantiles and quartiles. Last, the extreme observation at their lowest and highest values are a large number in comparison to the size of the time series data set. Hence, the range of the data and interquartile range tend also to be very large, indicating wideness in the data set.

Data Analysis by Decade of Equal Weighted Monthly Returns

Tables 7–10 contain the same analysis as appearing in Tables 2–5 for the value weighted monthly returns. The sizes of the sample for each decade correspond to the analysis in each decade. Group = 1 and 2 (the first and second decades) possess the largest mean and median rates of return with very large coefficients of variation but not as large for the same statistic in Tables 2 and 3 (the same groups by value weighted data). Skewness and kurtosis statistics tend to indicate similar results in Tables 7 and 8 as they do in Tables 2 and 3. Tables 9 and 10 again show similar results as in Tables 4 and 5. Hence, the method of weighting may show different values for the moments and descriptive statistics; the statistical tests of $\mu_0 = 0$ do indicate different results. However, the most stunning result refers to the comparison in the results in Tables 7 and 8 with Tables 9 and 10. Again, the results indicate that for decades (groups) 1 and 2, the mean rates are statistically different from zero, whereas in decades 3 and 4, this is not statistically evident. This corroborates the results in Tables 2 and 3 in comparison to Tables 4 and 5. Hence, the methods of weighting the monthly returns did not affect the interpretation of the results. Japanese equity returns by months deferred as time passed. The result is not an anomaly but indicate that time components did change and returns were probably influenced by changes in the Japanese economy.

Table 7
Comparing Japan Stock Market Return by Decades
Group 1

Moments (Univariate Procedure)			
N	119	Sum Weights	119
Mean	0.0152	Sum Observations	1.8102
Std. Deviation	0.0299	Variance	0.00089
Skewness	0.0764	Kurtosis	0.0957

Corrected Sum of Squares	0.1053		0.1053
Coefficient of Variation	196.3424	Std. Error Of the Mean	0.0027

Descriptive Statistics			
Location		Variability	
Mean	0.0152	Std. Deviation	0.0299
Median	0.0299	Variance	0.0009
Interquartile Range	0.0417	Range	0.1601

Tests for Location: $\mu_0 = 0$			
Test	Statistic	P-value	P-value
t	5.5560	Pr > abs (t)	≤.0001
M	25.5	Pr ≥ abs (M)	≤.0001
Signed Rank (S)	1923	Pr ≥ abs (S)	≤.0001
Quantiles			
Quantile	Estimate		
Maximum	0.0964		
99%	0.0858		
95%	0.0596		
90%	0.0535		
75% (Q3)	0.0351		
50% (Median, Q2)	0.0172		
25% (Q1)	0.0066		
10%	0.0228		
5%	0.0388		
1%	0.0568		
Minimum	0.0636		

Note: Q is for quantile.

Extreme Observations			
I Lowest		Highest I	
Value	Observation	Value	Observation
0.0636	57	0.0672	115
0.0568	112	0.0678	108
0.0431	7	0.0788	9
0.0409	76	0.0858	36
0.0396	34	0.0964	12

Table 8
Comparing Japan Stock Market Return by Decades
Group 2

Moments (Univariate Procedure)			
N	120	Sum Weights	120
Mean	0.0118	Sum Observations	1.4180
Std. Deviation	0.0671	Variance	0.0045
Skewness	0.1101	Kurtosis	1.3312
Corrected Sum of Squares	0.5357		
Coefficient of Variation	567.7790	Std. Error Of the Mean	0.0061

Descriptive Statistics			
Location		Variability	
Mean	0.0118	Std. Deviation	120
Median	0.147	Variance	0.0045
Interquartile Range	0.0620	Range	0.4186

Tests for Location: $\mu_0 = 0$			
Test	Statistic	P--value	P-value
t	3.287915	Pr > abs (t)	0.0011
M	36.5	Pr ≥ abs (M)	0.0007
Signed Rank (S)	10,312	Pr ≥ abs (S)	0.0002

Quantiles	
Quantile	Estimate
Maximum	0.2295
99%	0.1768
95%	0.1302
90%	0.0841
75% (Q3)	0.0443
50% (Median, Q2)	0.0147
25% (Q1)	0.0175
10%	0.0713
5%	0.1013
1%	0.1669
Minimum	0.1891

Note: Q is for quantile.

Extreme Observations			
Lowest		Highest	
Value	Observation	Value	Observation
0.1891	188	0.1351	184
0.1669	226	0.1410	219
0.1581	187	0.1509	228
0.1467	190	0.1768	189
0.1039	199	0.2295	193

Table 9
Comparing Japan Stock Market Return by Decades
Group 3

Moments (Univariate Procedure)			
N	120	Sum Weights	120
Mean	0.0038	Sum Observations	0.4502
Std. Deviation	0.0665	Variance	0.0044
Skewness	0.5862	Kurtosis	0.9239
Corrected Sum of Squares	0.5259		
Coefficient of Variation	1,772.0605	Std. Error Of the Mean	0.0061

Descriptive Statistics			
Location		Variability	
Mean	0.00375	Std. Deviation	0.06648
Median	0.00079	Variance	0.00442
Interquartile Range	0.08963	Range	0.39386

Tests for Location: $\mu_0 = 0$			
Test	Statistic	P-value	P-value
<i>t</i>	0.6181	Pr > abs (t)	0.5376
M	1	Pr ≥ abs (M)	0.9273
Signed Rank (S)	57	Pr ≥ abs (S)	8821
Quantile	Estimate		
Maximum	0.2563		
99%	0.1663		
95%	0.1145		
90%	0.0958		
75% (Q3)	0.0473		
50% (Median, Q2)	0.0008		
25% (Q1)	0.0424		
10%	Quantile		
5%	0.0855		
1%	0.1296		
Minimum	0.1375		

Note: Q is for quantile.

Extreme Observations			
Lowest		Highest	
Value	Observation	Value	Oberservation
0.1375	275	0.1280	348
0.1296	272	0.1374	286
0.1219	283	0.1381	290
0.1178	274	0.1663	350
0.0910	263	0.2563	276

Table 10
Comparing Japan Stock Market Return by Decades
Group 4

Moments (Univariate Procedure)			
N	96	Sum Weights	96
Mean	0.0033	Sum Observations	0.3169
Std. Deviation	0.0557	Variance	0.0031
Skewness	0.3094	Kurtosis	0.1209

Corrected Sum of Squares	0.2946		
Coefficient of Variation	1686.8887	Std. Error Of the Mean	0.0057

Descriptive Statistics			
Location		Variability	
Mean	0.0033	Std. Deviation	0.0557
Median	0.0027	Variance	0.0031
Interquartile Range	0.0771	Range	0.4392

Tests for Location: $\mu_0 = 0$			
Test	Statistic	P-value	P-value
<i>t</i>	0.5808	Pr > abs (t)	0.5627
M	0	Pr \geq abs (M)	1.0000
Signed Rank (S)	213	Pr \geq abs (S)	0.0771

Quantile	Estimate
Maximum	0.1181
99%	0.1181
95%	0.0876
90%	0.0742
75% (Q3)	0.0452
50% (Median, Q2)	0.0027
25% (Q1)	0.0320
10%	0.0678
5%	0.0921
1%	0.1674
Minimum	0.1674

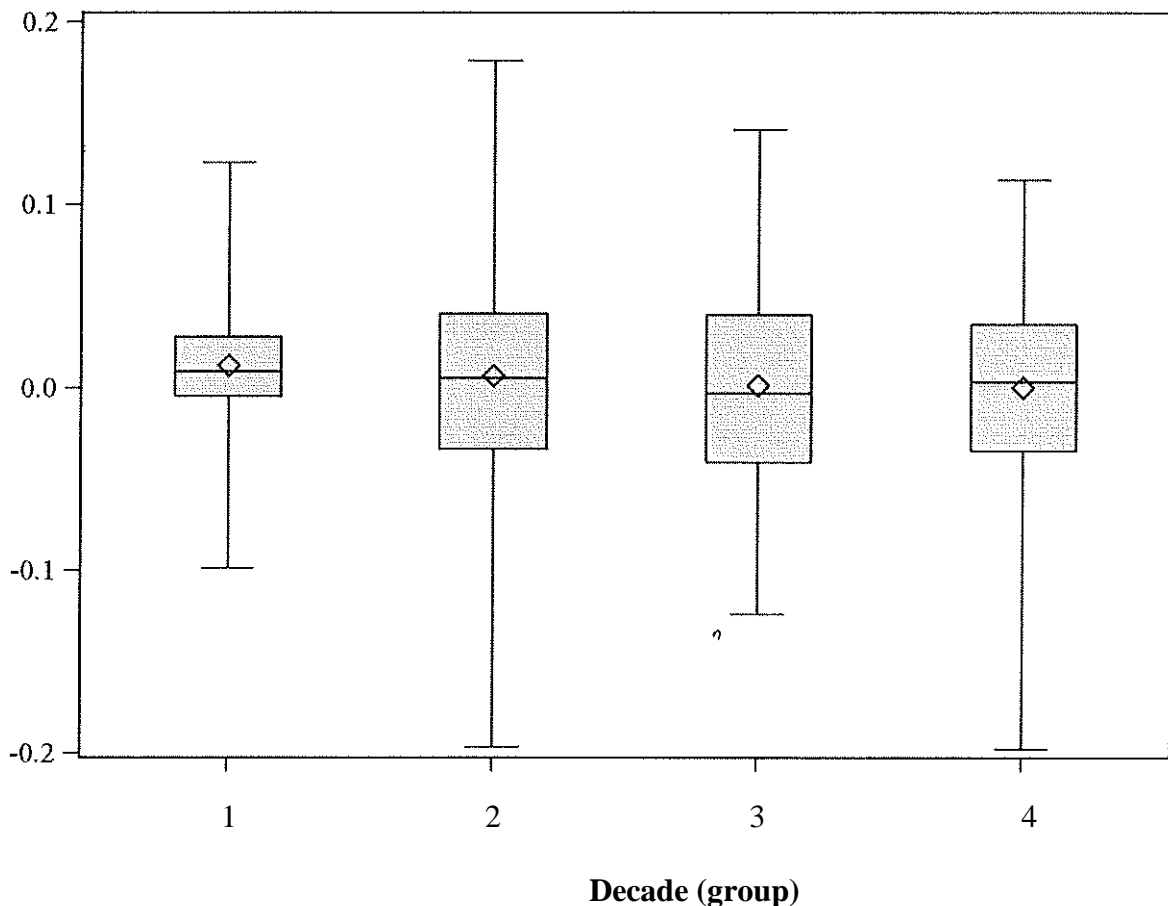
Note: Q is for quantile.

Extreme Observations			
Lowest		Highest	
Value	Observation	Value	Observation
	405		445
	448		371
	424		413
	404		412
	418		422

Boxplot Analysis of Equal Weighted Monthly Returns

By observing the boxplot by decade in Figure 2, we note its similarity to Figure 1, indicating that the method of weighting had little effect on the sample statistics developed to analytically observe the patterns in the data over the four-decade intervals. However, some results should be noted. The limits of the first decade's central box are the narrowest in decade 1 in Figure 1, and the same is true for Figure 2. The middle 50 percent boxes in decades 2–4 are larger than in decade 1 in both figures. The relation of the mean and median are also the same. Hence, it is likely that both methods of weighting did not change the results and conclusions that one can draw.

Figure 2
Boxplot to Compare Japan Stock Market Return by Decades MERMWD



Although the two figures were drawn using different time series, we observe the great similarity in the boxplots of monthly returns. Visual representation appears to indicate that the results are similar, not the same, and conclusions drawn from this visual information should be very similar.

Conclusions

The purpose of this study was to study if long memory modeling could possibly improve prediction of returns in at least one significant nationwide equity market. Previous studies indicated some predictability in producing results that would aid financial analysts and economists in forecasting returns to equity markets. This study does not dispute earlier results but clarifies some of the results of previous studies employing long memory modeling. We conclude that earlier studies may have opened the enthusiasm of utilizing these modern and

well-respected time series methods. However, for the purpose of forecasting returns to equities, long memory modeling may simply not be enough to be useful. Although producing long memory modeling of equity returns may be economically cost-effective in managing one's portfolio decisions, alternatives may be more accurate and useful. This analysis concludes that there are great many uses of long memory modeling, but predicting future returns over a lengthy period may not be one of them.

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