

Mutual Fund Performance Evaluation: A Unifying Approach

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Abstract: In a market with declining interest rates and narrowing alternative investment opportunities, mutual funds do constitute an important segment of the Indian financial market. Mutual fund schemes have a pass through structure i.e. investor return depends entirely on the income generated by the corpus of the scheme in the capital market which in turn depends heavily on the market conditions and also (albeit to a lesser extent) on the fund management abilities of relative mutual fund. While the risk emanating from adverse market conditions is an unavoidable risk, the second kind of risk can be mitigated (to certain extent) through portfolio reallocation which in turn depends on fund management skills. Performance evaluation of mutual funds is therefore an area of research which has generated considerable interest in the minds of both academicians and industry experts.

The methodology of mutual fund evaluation has two distinct strands. The traditional approach is essentially a ratio based approach which evaluates observed portfolio performance relative to the market portfolio. Thus this approach uses an external benchmark for performance evaluation. The non-parametric approach, on the other hand, uses an endogenous benchmark as it generates a risk-return frontier (from the observed data) which is then used for evaluating the performance of any observed fund. The present study adopts a hybrid path as it attempts to use the market portfolio risk-return data for performance evaluation in a non-parametric setting.

Keywords: Mutual Fund, Data envelopment analysis, Performance evaluation, Hybrid approach, Fixed benchmarking, Variable benchmarking.

JEL Classification: G-110, C-61, D-21.

Introduction

For common investors with limited knowledge about the market fundamentals, mutual funds provide an importance vehicle of investment in the capital market. Depending on the return-risk expectation of the investor, the market offers pure equity, pure debt and hybrid (balanced) schemes of investment in mutual funds. In a market with declining interest rates and narrowing alternative investment opportunities, mutual funds do constitute an important segment of the financial market.

Unlike the commercial banks, the mutual fund schemes have a pass through structure i.e. investor return depends entirely on the income generated by the corpus of the scheme in the capital market which in turn depends heavily on the market conditions and also (albeit to a lesser extent) on the fund management abilities of relative mutual fund. While the risk emanating from adverse market conditions is an unavoidable risk, the second kind of risk can be mitigated (to certain extent) through portfolio reallocation which in turn depends on fund management skills. Performance evaluation of mutual funds is therefore an area of research which has generated considerable interest in the minds of both academicians and industry experts.

Broadly speaking, there are two approaches for the evaluation of mutual fund performance. The traditional approach is a ratio based approach which evaluates observed portfolio performance relative to the market excess return i.e. it uses an external benchmark. Thus the Capital Asset Pricing Model links the portfolio excess return with the excess return available on the market portfolio. The Fama–French (1992) approach adds two other explanatory factors –differential return available on small caps (over big caps) and differential return on firms with high book value/market value ratio (over low book value/market value ratio).

The non-parametric approach, on the other hand, uses an endogenous benchmark as it generates a risk-return frontier (from the observed data) which is then used for evaluating the performance of any observed fund. While this approach uses multi-criteria evaluation method and is suitable for intra-fund comparison, the approach does not link fund performance with market return.

The present study adopts a unifying approach. To be more specific, it attempts to use the market portfolio risk-return data for performance evaluation in a non-parametric setting. Thus, the current approach retains the flavor of external benchmarking in the context of multi-criteria decision making. The paper has four sections and proceeds as follows. Section 1 reviews the extant literature. Section 2

outlines the methodological developments. Section 3 describes the variables and the study outcomes. Section 4 concludes.

1. Review of Related Literature

Murthi, Choi, and Desai (1997) estimated the performance of 2083 mutual fund schemes for October-December 1993 using DPEI (DEA Portfolio Efficiency Index). The study considered a two input (standard deviation and transaction loads) –one output (excess return) model. The second stage analysis tried to identify the sources of efficiency variation.

Basso and Funari (2001) evaluated performance of fifty Italian mutual fund schemes (comprising of 24 stock funds, 10 balanced funds and 16 balanced funds) for January 1997 to June 1999. The study included two DEA measures. The first measure included mutual fund return as the output and two inputs (standard deviation of return and transaction costs). The second DEA measure included a stochastic dominance indicator as well. The second study by Basso and Funari (2003) used an ethical score of mutual funds in lieu of the stochastic dominance indicator.

Gregoriou, Sedzro, Zhu (2005) pointed out that standard performance measures like Sharpe ratio is not suitable for evaluating hedge fund performance. They accordingly evaluated (using DEA) eight categories of hedge funds for two separate runs spanning 1997-2001 and 1999-2001. They initially utilised the Banker, Charnes-Cooper (1984) model to categorise the hedge funds for the identification of efficient and inefficient funds. Further, the study utilised cross and super-efficiency models for estimating peer-appraisal scores and ranking of in-sample funds.

Daraio and Simar (2006) applied a variety of measures (conditional input oriented order-m efficiency, Free Disposal Hull (FDH) method and DEA, Jensen's α and Sharpe Index) for performance benchmarking of six mutual funds groups (asset allocation, aggressive growth, balanced, equity income, growth and growth income). The study used a three input (turnover ratio, expense ratio and fund loads)-one output (total return) framework. The study estimated correlation of traditional indicators (Jensen's α and Sharpe Index) with the non-parametric estimates (order m efficiency, DEA and FDH). The results indicated that while intra-group correlations are high, inter-group correspondence is weak.

Zhao, Wang and Lai (2011) applied two DEA models (involving quadratic constraints) for evaluating 25 open-ended Chinese mutual funds for 2005 and 2006. The study identified two key factors influencing mutual fund performance: excess

return and risk. Two measures of return indicating asset allocation and excess over market benchmark return are used. The two quadratic-constrained DEA models measure efficiency from output and input perspectives. The results show that although the market environment in year 2006 was much better than that in 2005, mean efficiency score decreased during 2006 due to the relaxation of system risk control. Further, most of the observed mutual funds did show fluctuations in efficiency ranking.

Goel and Mani (2018) investigated efficiency performance of Indian mutual funds and tried to identify the corrective measures for improving their performance. Their study included 143 mutual fund schemes for 11 years (April 2006 to March 2017). The study includes two outputs (Sharpe ratio and Jensen's alpha) and five inputs (load fee, expense ratio, minimum initial investment needed and risk (β and α)). The outcome showed that only a few mutual fund schemes performed efficiently and the inefficient schemes need to reduce their load and expense ratio. Most of the efficient funds belong to the equity linked savings scheme category followed by income, growth and balanced funds.

2. Methodological Developments

1.1. CAPM, Mean-Variance and Stochastic Dominance

The capital asset pricing model proposed by Sharpe (1964) and Lintner (1965) showed that in equilibrium the excess return of an observed portfolio (over the risk free rate) is linked to the excess return of the market portfolio. This theoretical development provided the basis for the evaluation of portfolio risk-return performance. Thus the Sharpe ratio used excess return earned by a portfolio over the risk free rate normalized by the standard deviation as the indicator performance. The Treynor (1965) measure considered the ratio of the excess return of the portfolio and the portfolio beta as the measure of performance. Jensen (1968) measured the difference between the observed and risk adjusted rate of return (computed from capital asset pricing model).

The seminal contribution of Markowitz (1952) in terms of the mean-variance approach was behind the development of capital asset pricing model. The framework of analysis, however, critically depends upon the assumption of normality of portfolio returns which is unlikely to materialize in most of the cases. Thus many researchers expressed reservations about the use of CAPM framework for portfolio performance

evaluation. Roy (1952) pointed out that an investor would be concerned primarily by the safety of principal in case of his investments. The investor would thus prefer the investment with the smallest probability of going below the target return. By maximizing a reward to variability ratio, $\frac{(R_e - T_r)}{\sigma}$, the investor will choose the portfolio with the lowest probability of going below the target level, T_r , given an expected mean return, R_e , and a standard deviation, σ . Klemkosky (1973) and Ang and Chua (1979) pointed out that CAPM based measures can lead to incorrect rankings. Thus they suggested the use of reward to semi-variability (R/SV) ratio instead of portfolio reward to standard deviation. However, significant development regarding the use of downside risk measures occurred with the development of the Lower Partial Moment (LPM) risk measure contributed by Bawa (1975) and Fishburn (1977). Bawa (1975) expressed the lower partial moment (LPM) as a general family of below-target risk measures. Further, he proved that the LPM measure is mathematically connected to the measure of stochastic dominance for risk tolerance values of 0, 1, and 2. Fishburn (1974,1978) further extended the model and formulated the conditions for identifying optimal and dominated choice sets. i.e. Conditional Stochastic Dominance which enables the decomposition of the choice set into optimal and dominated sets. Bawa, Lindenberg & Rafsky (1979) and Bawa, Bothurda Jr., Rao and Suri (1985) proposed and implemented exact linear programming algorithms (with the application of Fishburn's conditions for convex stochastic dominance) for the assignment of discrete return distributions into the first- and second-order stochastic dominance optimal sets. For third-order stochastic dominance, Bawa et.al. (1985) defined a superconvex stochastic dominance approach which permits classification of choice elements into superdominated, mixed, and superoptimal sets. The downside approach to the measurement of risk got the necessary support from the practitioners as well. Sortino (1991) pointed out that in situations with some identifiable Minimum Acceptable Return (MAR), it is essential to segregate good and bad volatility where good volatility is dispersion above MAR and bad volatility is dispersion below MAR. Sortino (1994) thus proposed a downside risk ratio (an alternative to the Sharpe ratio) which includes portfolio return net of MAR on the numerator and downside deviation in the denominator. Sortino, Meer and Platinga (1999) introduced the Upside Potential Ratio which contains the expected return above the MAR in the numerator and the denominator is downside risk.

The discussion so far mostly includes the refinements in the ratio based approach. However, with the growing application of non-parametric approaches in the context of the financial services industry, portfolio performance evaluation became an interesting application area for Data Envelopment Analysis. In the non-parametric approach, the performance of a portfolio is evaluated on the basis of its distance from a risk-return frontier. Thus the evaluation is based on the concept of distance function which was originally formalised in the context of a multi-input and multi-output technology. The input distance function relates to the input set of the production technology while the output distance function relates to the output set.

For providing a brief description of the concept of distance function, Let us consider a production set P_s with a nonnegative vector of inputs $x = (x_1, x_2, \dots, x_n) \in R_+^n$ which is used to produce a nonnegative vector of outputs $y = (y_1, y_2, \dots, y_m) \in R_+^m$. In functional terms, they can be related as: $y = P(x)$ and $x = L(y)$ where $P(x)$ and $L(y)$ represent the output and input requirement set respectively. In input distance function can be defined as $D_{IN} = \text{Min}[\lambda: X/\lambda \in L(Y)]$. Thus an input distance function provides the minimum amount by which the producer's input vector can be radially deflated and yet remain feasible for the given output vector. The inverse of the input distance function (λ) can be considered as the radial measure of input oriented technical efficiency.

In an similar vein, the output distance function is defined as: $D_{OUT} = \text{Max}[\mu: Y/\mu \in P(X)]$. Intuitively speaking, an output distance function gives the maximum amount by which the producer's output vector can be deflated and yet remain feasible for a given input vector. The radial measure of output oriented technical efficiency coincides with the output distance function.

The benefit of the distance function approach is that we are able to accommodate multi-inputs and multiple outputs which is not possible in the ratio approach. When we apply the non-parametric method for the estimation of the distance function, we need not assume a specific parametric relationship. However, the conventional non-parametric approach computes efficiency by comparing the observed portfolio with a frontier which is constructed from the observed set of data. In the present context, we have avoided this by taking the commonly used market portfolio indices as the benchmark. The technology is global i.e. estimation is made under the assumption of operation of constant returns to scale.

Continuing with the previously mentioned production possibility set P_s , the objective of an observed decision making unit (fund manager in the present context) for output maximization under the Banker, Charnes and Cooper model (1984) is:

$$\begin{aligned} & \text{Max } \theta \\ & \text{Subject to: } x_0 \geq \lambda X, y_0 \leq \lambda Y, \lambda \geq 0 \end{aligned}$$

Where x_0 and y_0 represent the input and output vectors of the observed decision making unit (DMU). The Charnes, Cooper and Rhodes (1978) approach towards benchmarking is special case of the more generalized Banker, Charnes and Cooper model where we do not permit local variations in technology. In both cases, the decision making units are evaluated on the basis of reference set formed output of the input and output data for the observed productive units. In the context of mutual fund performance evaluation, both models use benchmarks which are endogenous in nature.

In the present study, extended the conventional DEA based benchmarking by following the procedures suggested by Cook, Seiford and Zhu (2004). To be more specific, we have used the fixed and variable benchmarking models suggested by we have used two external benchmarks –CNX Nifty and BSE 100. The benchmarks have been used in both fixed and variable benchmark models. In the fixed benchmark model the mathematical program is:

$$\begin{aligned} & \text{Max } \theta_F \\ & \text{Subject to: } x_0 \geq \lambda X_B, \theta_F y_0 \leq Y_B, \lambda \geq 0 \end{aligned}$$

Where X_B and Y_B represent the input and output corresponding to the external benchmark used for performance evaluation. Efficiency of the observed DMU is the inverse of θ_F . θ_F is the Farrell (1957) distance function in the fixed benchmarking case.

In the variable benchmark model, let the two benchmarks be (X_1, Y_1) and (X_2, Y_2) respectively. Then the optimization program for the observed decision making unit is:

$$\begin{aligned} & \text{Max } \theta_V \\ & \text{Subject to: } x_0 \geq \sum \lambda_i X_i, \theta_V y_0 \leq \lambda, \sum \lambda_i Y_i, \lambda \geq 0 \end{aligned}$$

Efficiency of the observed DMU is the inverse of which is the Farrell (1957) distance function in the context of variable benchmarking.

3. Variables, Results and Discussion

3.1. Description of Variables

For explaining the working of the two benchmarking models (fixed and variable), we have used the AMFI supplied data relating to Net Asset Values of 50 equity oriented mutual fund schemes. The equity oriented schemes include 27 sectoral and 23 diversified mutual funds. The data pertain to the financial years 2010-11, 2011-12 and 2012-13.

For the evaluation of fund efficiency, it is essential to specify inputs and outputs. In the present context, we have used one input (value at risk) and two outputs-mean return on the portfolio and probability of excess return over the mean. The input and output measures have been calculated from the net asset values. In order to avoid the problem of handling negative data, rate of return is calculated as where and represent the net asset values of the observed fund for time periods t and $t+1$ respectively.

3.2. Results and Discussion

In the present study we have applied fixed and variable benchmark models for the in-sample funds for the period under observation. In addition to the aforementioned estimates, we have also estimated fund efficiency scores using the radial envelopment model and computed the returns to scale exhibited by the funds.

3.2.1. Fixed and Variable Benchmark Models

Table 1 provides the mean efficiency scores obtained by the application of the two benchmarks (CNX Nifty and BSE 100) one at a time. The table includes the scores separately for the sectoral and diversified equity funds as well as the overall scores. The sectoral equity funds concentrate their equity investments to a few or one identified high growth sectors. On the the other hand, diversified mutual funds follow the strategy of portfolio diversification. Obviously the sectoral funds are more risky as compared to the diversified funds. However, they also provide higher prospects of return in a booming market. In our case study, the sectoral funds (in terms of mean efficiency) have out-performed the diversified equity oriented mutual funds for all the three years under observation. The fund wise efficiency scores are included in appendix tables A1 through A4. Figure 1 provides a graphical presentation of mean efficiency scores for the years 2010-11, 2011-12 and 2012-13.

Table 1: Mean efficiency scores-fixed benchmark model

<i>Fund Category</i>	<i>Benchmark</i>	<i>2010-11</i>	<i>2011-12</i>	<i>2012-13</i>
Sectoral equity schemes	CNX Nifty	1.0723	1.0515	1.0542
Diversified equity schemes	CNX Nifty	1.0467	1.0144	1.0253
Overall	CNX Nifty	1.0605	1.0344	1.0409
Sectoral equity schemes	BSE 100	1.0352	1.0321	1.0404
Diversified equity schemes	BSE 100	1.0132	0.9991	1.0145
Overall	BSE 100	1.0251	1.0169	1.0284

Source: Calculated.

Figure 1: Mean efficiency-fixed benchmark models

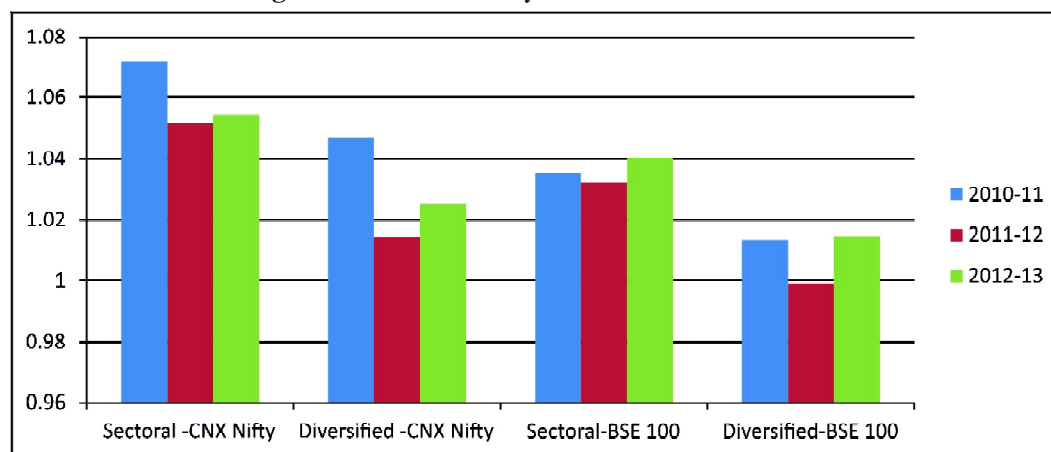


Table 2 provides the mean efficiency scores for three in-sample years pertaining to the variable benchmarking model where weighted combination of CNX Nifty and BSE 100 is taken as the benchmark. In this case also, the mean efficiency score for the sectoral mutual fund schemes is found to be higher than the diversified schemes. Figure 2 gives a graphical presentation of the mean efficiency scores. Tables A 5 and A 6 provides the fund wise mean efficiency scores for the sectoral and diversified equity oriented schemes respectively. Tables A 7 through A 12 present the optimal weights corresponding to the two benchmarks (CNX Nifty and BSE 100) for the observed years.

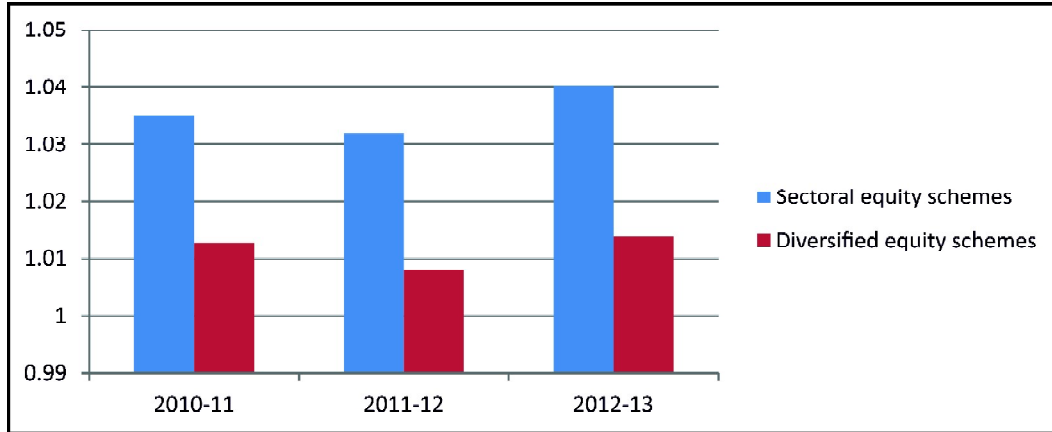
Table 2: Mean efficiency scores-variable benchmark model

<i>Fund Category</i>	<i>Benchmark</i>	<i>2010-11</i>	<i>2011-12</i>	<i>2012-13</i>
Sectoral equity schemes	Weighted average of CNX Nifty and BSE 100	1.0351	1.0321	1.0401
Diversified equity schemes	Weighted average of CNX Nifty and BSE 100	1.0130	1.008	1.0141
Overall	Weighted average of CNX Nifty and BSE 100	1.0249	1.0177	1.0281

Source: Calculated.

3.2.2. Endogenous benchmarking and returns to scale

The previous results are obtained from the application of NSE-50 and BSE-100 as the two external benchmarks. The results can be supplemented by the use of

Figure 2: Mean efficiency-variable benchmark

endogenous benchmark in the conventional DEA framework. Presently, we provide the outcomes refer table 3) from the application of radial CCR (1978) and BCC (1984) envelopment model. Table 4 provides the summary information regarding returns to scale exhibited by the observed funds. The fund wise efficiency scores under constant and variable returns to scale are available from appendix tables A 13 through A 16.

Table 3: Mean efficiency scores (envelopment model)

<i>Fund type</i>	<i>Benchmark</i>	<i>2010-11</i>	<i>2011-12</i>	<i>2012-13</i>
Sectoral equity schemes	Global (CRS)	0.9873	0.9859	0.9836
Sectoral equity schemes	Local (VRS)	0.9992	0.9991	0.9993
Diversified equity schemes	Global (CRS)	0.9812	0.9716	0.9716
Diversified equity schemes	Local (VRS)	0.9988	0.9989	0.9996
All schemes	Global (CRS)	0.9845	0.9794	0.9781
All schemes	Local(VRS)	0.9990	0.9990	0.9994

Source: Calculated.

Table 4: Returns to scale composition

<i>Fund type</i>	<i>Returns to scale</i>	<i>2010-11</i>	<i>2011-12</i>	<i>2012-13</i>
Sectoral	Constant	3	4	2
Sectoral	Increasing	24	23	25
Sectoral	Decreasing	0	0	0
Diversified	Constant	0	0	0
Diversified	Increasing	0	0	0
Diversified	Decreasing	23	23	23

Source: Calculated.

Table 3 shows that the sectoral funds have outperformed (in terms of mean efficiency) the diversified funds for the observed years (2010-11, 2011-12 and 2012-13) under the observation of constant returns to scale. However, if we assume local returns to scale (vrs), then the diversified funds exhibited slightly higher mean efficiency than the diversified funds for 2012-13. However, for the previous two years, the mean efficiency for the sectoral funds was higher than the diversified equity oriented mutual funds.

4. Concluding Observations

Performance evaluation of mutual fund schemes is important from the point of view of all stakeholders including investors, asset management companies and capital market regulators. The efficacy of the traditional ratio based evaluation of performance is critically dependent on the validity of the Capital Asset Pricing Model. None of the research studies supported the normality of return distribution. Thus it is essential to move beyond normality. In the present study, we have used an extended model by incorporating both downside measure of risk and upside measure of return. However, the estimation of efficiency in the non-parametric setting is now linked to the commonly used indices of market portfolio. For the interest of the reader, the results so obtained are supplemented by efficiency estimates obtained through endogenous benchmarking. The study is limited to three years only and this can be extended to longer period to see equity oriented schemes have performed in India relative to the various market portfolio indices.

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Table A 1: Sectoral equity fund wise efficiency performance (fixed benchmark-CNX Nifty)

<i>Fund Name</i>	<i>2010-11</i>	<i>2011-12</i>	<i>2012-13</i>
Baroda Pioneer Infrastructure Fund	1.1230	1.0275	1.0101
BOI AXA Focused Infrastructure Fund	1.0506	1.0180	1.0271
Canara Robeco Infrastructure	1.0622	1.0317	1.0738
DSP BlackRock World Energy Fund	1.1078	1.0677	1.1093
DSP BlackRock World Gold Fund	1.1036	1.0264	1.1862
DSP BlackRock World Mining Fund	1.0826	1.0576	1.0301
DWS Global Agribusiness offshore Fund	1.0551	1.1334	1.0691
Fidelity Global Real Assets Fund	1.0314	1.0368	1.1095
HDFC Core and Satellite Fund	1.1068	1.0282	1.0618
HDFC Infrastructure Fund	1.0615	1.0545	1.0302
ICICI Prudential Banking and Financial Services Fund	1.0180	1.1090	1.0169
ICICI Prudential Exports and Other Services Fund	1.0262	1.0424	1.0183
ICICI Prudential Technology Fund	1.0645	1.0966	1.0384
IDFC Strategic Sector (50-50) Equity Fund	0.9997	0.9998	1.0069
ING Global Commodities Fund	1.1201	1.0959	1.0271
JPMorgan India Smaller Companies Fund	1.0865	1.0959	1.1047
Kotak PSU Bank ETF	1.1442	1.1008	1.0591
L&T Infrastructure Fund	1.1066	1.0615	1.0791
LIC Nomura MF Infrastructure Fund	1.1461	1.0044	1.0221
Mirae Asset Global Commodity Stocks Fund	1.0057	1.0408	1.1088
PineBridge Infrastructure & Economic Reform Fund	1.0560	1.0590	1.0347
PineBridge World Gold Fund	1.0460	1.0244	1.0885
Reliance Banking Fund	1.1013	1.0566	1.0195
SBI Infrastructure Fund	1.0436	1.0003	1.0576
Sundaram Energy Opportunities Fund	1.0964	1.0734	1.0619
Tata Growing Economies Infrastructure Fund	1.0453	0.9851	0.9935
UTI Banking Sector Fund	1.0624	1.0616	1.0194

Source: Calculated.

Table A 2: Diversified equity fund efficiency performance (fixed benchmark-CNX Nifty)

<i>Fund Name</i>	<i>2010-11</i>	<i>2011-12</i>	<i>2012-13</i>
UTI Equity Fund	1.0433	0.9941	1.0018
Tata Equity Fund	1.0879	0.9940	1.0601
SBI Equity Fund	1.0667	0.9949	1.0027
Religare Invesco Equity Fund	1.0074	1.0566	1.0817
Reliance Equity Fund	1.0229	1.0166	1.0588
Quantum Equity Fund	1.0578	0.9978	1.0149
Pramerica Equity Fund	1.0972	0.9999	1.0076
PineBridge Equity Fund	1.0205	0.9930	1.0346
Kotak Mahindra Equity Fund	1.0546	1.0253	0.9845
JPMorgan Equity Fund	1.0458	0.9965	1.0382
JM Financial Equity Fund	1.0218	1.0670	1.0086
ING Equity Fund	1.0627	1.0028	1.0598
IDFC Equity Fund	1.0006	0.9995	1.0166
ICICI Prudential Equity Fund	0.9985	0.9846	0.9846
HSBC Equity Fund	1.0433	0.9955	1.0033
Franklin Templeton Equity Fund	1.0213	1.0184	1.0276
DSP BlackRock Equity Fund	1.0429	1.0574	1.0662
Deutsche Equity Fund	1.0946	1.0370	1.0211
Canara Robeco Equity Fund	1.0697	1.0178	1.0185
BOI AXA Equity Fund	1.0372	1.0141	1.0231
Birla Sun Life Equity Fund	1.0707	0.9976	1.0148
Baroda Pioneer Equity Fund	1.0268	1.0181	1.0106
Axis Equity Fund	1.0789	1.0533	1.0414

Source: Calculated.

Table A 3: Sectoral equity fund wise efficiency performance (fixed benchmark-BSE 100)

<i>Fund Name</i>	<i>2010-11</i>	<i>2011-12</i>	<i>2012-13</i>
Baroda Pioneer Infrastructure Fund	1.0793	1.0052	1.0109
BOI AXA Focused Infrastructure Fund	1.0098	0.9958	1.0097
Canara Robeco Infrastructure	1.0210	1.0093	1.0521
DSP BlackRock World Energy Fund	1.0647	1.0444	1.0870
DSP BlackRock World Gold Fund	1.0607	1.0206	1.1624
DSP BlackRock World Mining Fund	1.0405	1.0346	1.0309
DWS Global Agribusiness offshore Fund	1.0140	1.1088	1.0476
Fidelity Global Real Assets Fund	1.0006	1.0142	1.0871
HDFC Core and Satellite Fund	1.0638	1.0057	1.0404
HDFC Infrastructure Fund	1.0202	1.0316	1.0126
ICICI Prudential Banking and Financial Services Fund	1.0070	1.0848	1.0177
ICICI Prudential Exports and Other Services Fund	0.9982	1.0198	1.0103
ICICI Prudential Technology Fund	1.0230	1.0727	1.0213
IDFC Strategic Sector (50-50) Equity Fund	1.0007	0.9942	1.0078
ING Global Commodities Fund	1.0765	1.0720	1.0280
JPMorgan India Smaller Companies Fund	1.0442	1.0720	1.0825
Kotak PSU Bank ETF	1.0996	1.0768	1.0378
L&T Infrastructure Fund	1.0635	1.0384	1.0574
LIC Nomura MF Infrastructure Fund	1.1016	0.9988	1.0134
Mirae Asset Global Commodity Stocks Fund	0.9950	1.0181	1.0865
PineBridge Infrastructure & Economic Reform Fund	1.0148	1.0358	1.0138
PineBridge World Gold Fund	1.0105	1.0158	1.0666
Reliance Banking Fund	1.0585	1.0336	1.0158
SBI Infrastructure Fund	1.0030	0.9947	1.0363
Sundaram Energy Opportunities Fund	1.0537	1.0500	1.0406
Tata Growing Economies Infrastructure Fund	1.0046	0.9795	0.9934
UTI Banking Sector Fund	1.0210	1.0384	1.0202

Source: Calculated.

Table A 4: Diversified equity fund efficiency performance (fixed benchmark-BSE 100)

<i>Fund Name</i>	<i>2010-11</i>	<i>2011-12</i>	<i>2012-13</i>
UTI Equity Fund	1.0028	1.0116	1.0026
Tata Equity Fund	1.0457	1.0117	1.0389
SBI Equity Fund	1.0252	1.0108	1.0034
Religare Invesco Equity Fund	0.9959	0.9675	1.0599
Reliance Equity Fund	1.0026	1.0054	1.0375
Quantum Equity Fund	1.0167	1.0078	1.0064
Pramerica Equity Fund	1.0545	1.0058	1.0084
PineBridge Equity Fund	0.9927	1.0127	1.0137
Kotak Mahindra Equity Fund	1.0136	0.9970	0.9853
JPMorgan Equity Fund	1.0052	1.0092	1.0174
JM Financial Equity Fund	1.0018	0.9581	1.0094
ING Equity Fund	1.0213	1.0029	1.0384
IDFC Equity Fund	1.0016	1.0061	1.0083
ICICI Prudential Equity Fund	0.9793	0.9675	0.9854
HSBC Equity Fund	1.0027	1.0054	1.0041
Franklin Templeton Equity Fund	0.9938	1.0078	1.0194
DSP BlackRock Equity Fund	1.0023	1.0058	1.0447
Deutsche Equity Fund	1.0521	1.0127	1.0047
Canara Robeco Equity Fund	1.0282	0.9970	1.0019
BOI AXA Equity Fund	1.0008	1.0092	1.0064
Birla Sun Life Equity Fund	1.0290	0.9581	1.0063
Baroda Pioneer Equity Fund	0.9984	1.0029	1.0096
Axis Equity Fund	1.0369	1.0061	1.0204

Source: Calculated.

Table A 5: Sectoral equity fund wise efficiency performance (variable benchmark model)

<i>Fund Name</i>	<i>2010-11</i>	<i>2011-12</i>	<i>2012-13</i>
Baroda Pioneer Infrastructure Fund	1.0793	1.0052	1.0101
BOI AXA Focused Infrastructure Fund	1.0098	0.9959	1.0096
Canara Robeco Infrastructure	1.0210	1.0092	1.0521
DSP BlackRock World Energy Fund	1.0647	1.0444	1.0870
DSP BlackRock World Gold Fund	1.0607	1.0207	1.1624
DSP BlackRock World Mining Fund	1.0405	1.0346	1.0301
DWS Global Agribusiness offshore Fund	1.0140	1.1088	1.0475
Fidelity Global Real Assets Fund	1.0004	1.0142	1.0871
HDFC Core and Satellite Fund	1.0638	1.0058	1.0404
HDFC Infrastructure Fund	1.0202	1.0316	1.0125
ICICI Prudential Banking and Financial Services Fund	1.0063	1.0849	1.0169
ICICI Prudential Exports and Other Services Fund	0.9979	1.0198	1.0098
ICICI Prudential Technology Fund	1.0231	1.0728	1.0212
IDFC Strategic Sector (50-50) Equity Fund	0.9997	0.9942	1.0070
ING Global Commodities Fund	1.0765	1.0721	1.0272
JPMorgan India Smaller Companies Fund	1.0442	1.0721	1.0824
Kotak PSU Bank ETF	1.0997	1.0768	1.0378
L&T Infrastructure Fund	1.0635	1.0384	1.0574
LIC Nomura MF Infrastructure Fund	1.1015	0.9988	1.0129
Mirae Asset Global Commodity Stocks Fund	0.9943	1.0181	1.0865
PineBridge Infrastructure & Economic Reform Fund	1.0149	1.0359	1.0138
PineBridge World Gold Fund	1.0104	1.0158	1.0666
Reliance Banking Fund	1.0585	1.0336	1.0152
SBI Infrastructure Fund	1.0031	0.9947	1.0363
Sundaram Energy Opportunities Fund	1.0537	1.0500	1.0406
Tata Growing Economies Infrastructure Fund	1.0046	0.9796	0.9927
UTI Banking Sector Fund	1.0210	1.0385	1.0194

Source: Calculated.

Table A 6 : Diversified equity fund efficiency performance (variable benchmark model)

<i>Fund Name</i>	<i>2010-11</i>	<i>2011-12</i>	<i>2012-13</i>
UTI Equity Fund	1.0028	0.9885	1.0018
Tata Equity Fund	1.0457	0.9884	1.0388
SBI Equity Fund	1.0252	0.9893	1.0027
Religare Invesco Equity Fund	0.9953	1.0336	1.0599
Reliance Equity Fund	1.0021	0.9947	1.0375
Quantum Equity Fund	1.0166	0.9922	1.0060
Pramerica Equity Fund	1.0545	0.9942	1.0076
PineBridge Equity Fund	0.9923	0.9874	1.0137
Kotak Mahindra Equity Fund	1.0136	1.0030	0.9846
JPMorgan Equity Fund	1.0052	0.9909	1.0174
JM Financial Equity Fund	1.0013	1.0437	1.0086
ING Equity Fund	1.0214	0.9971	1.0384
IDFC Equity Fund	1.0006	0.9939	1.0078
ICICI Prudential Equity Fund	0.9789	0.9719	0.9846
HSBC Equity Fund	1.0027	0.9900	1.0033
Franklin Templeton Equity Fund	0.9936	1.0052	1.0189
DSP BlackRock Equity Fund	1.0023	1.0343	1.0447
Deutsche Equity Fund	1.0520	1.0144	1.0045
Canara Robeco Equity Fund	1.0281	0.9956	1.0017
BOI AXA Equity Fund	1.0007	0.9923	1.0063
Birla Sun Life Equity Fund	1.0290	0.9920	1.0059
Baroda Pioneer Equity Fund	0.9981	0.9959	1.0088
Axis Equity Fund	1.0370	1.0304	1.0204

Source: Calculated.

Table A 7 : Sectoral fund reference weights (variable benchmark model)-2010-11

<i>Fund Name</i>	<i>CNX Nifty</i>	<i>BSE 100</i>
Baroda Pioneer Infrastructure Fund	0	0.997
BOI AXA Focused Infrastructure Fund	0	0.994
Canara Robeco Infrastructure	0	1.007
DSP BlackRock World Energy Fund	0	0.994
DSP BlackRock World Gold Fund	0	0.990
DSP BlackRock World Mining Fund	0	0.985
DWS Global Agribusiness offshore Fund	0	1.006
Fidelity Global Real Assets Fund	0.234	0.767
HDFC Core and Satellite Fund	0	1.004
HDFC Infrastructure Fund	0	1
ICICI Prudential Banking and Financial Services Fund	0.710	0.285
ICICI Prudential Exports and Other Services Fund	0.303	0.699
ICICI Prudential Technology Fund	0	0.997
IDFC Strategic Sector (50-50) Equity Fund	1.001	1
ING Global Commodities Fund	0	0.997
JPMorgan India Smaller Companies Fund	0	1
Kotak PSU Bank ETF	0	0.985
L&T Infrastructure Fund	0	0.996
LIC Nomura MF Infrastructure Fund	0	0.996
Mirae Asset Global Commodity Stocks Fund	0.721	0.285
PineBridge Infrastructure & Economic Reform Fund	0	1.005
PineBridge World Gold Fund	0.126	0.864
Reliance Banking Fund	0	0.994
SBI Infrastructure Fund	0	1.001
Sundaram Energy Opportunities Fund	0	0.998
Tata Growing Economies Infrastructure Fund	0	1.007
UTI Banking Sector Fund	0	0.991

Source: Calculated.

Table A 8 : Sectoral fund reference weights (variable benchmark model)-2011-12

<i>Fund Name</i>	<i>CNX Nifty</i>	<i>BSE 100</i>
Baroda Pioneer Infrastructure Fund	0	1.003
BOI AXA Focused Infrastructure Fund	0	1.004
Canara Robeco Infrastructure	0	1.007
DSP BlackRock World Energy Fund	0	0.982
DSP BlackRock World Gold Fund	0	0.980
DSP BlackRock World Mining Fund	0	0.983
DWS Global Agribusiness offshore Fund	0	0.996
Fidelity Global Real Assets Fund	0	0.994
HDFC Core and Satellite Fund	0	1.010
HDFC Infrastructure Fund	0	1.001
ICICI Prudential Banking and Financial Services Fund	0	0.997
ICICI Prudential Exports and Other Services Fund	0	1.005
ICICI Prudential Technology Fund	0	0.993
IDFC Strategic Sector (50-50) Equity Fund	0	1.006
ING Global Commodities Fund	0	0.986
JPMorgan India Smaller Companies Fund	0	0.986
Kotak PSU Bank ETF	0	0.989
L&T Infrastructure Fund	0	1.002
LIC Nomura MF Infrastructure Fund	0	1.001
Mirae Asset Global Commodity Stocks Fund	0	0.990
PineBridge Infrastructure & Economic Reform Fund	0	1.005
PineBridge World Gold Fund	0	0.985
Reliance Banking Fund	0	0.999
SBI Infrastructure Fund	0	1.005
Sundaram Energy Opportunities Fund	0	1.002
Tata Growing Economies Infrastructure Fund	0	1.021
UTI Banking Sector Fund	0	0.994

Source: Calculated.

Table A 9 : Sectoral fund reference weights (variable benchmark model)-2012-13

<i>Fund Name</i>	<i>CNX Nifty</i>	<i>BSE 100</i>
Baroda Pioneer Infrastructure Fund	0.989	0
BOI AXA Focused Infrastructure Fund	0.152	0.837
Canara Robeco Infrastructure	0	0.993
DSP BlackRock World Energy Fund	0	0.968
DSP BlackRock World Gold Fund	0	0.966
DSP BlackRock World Mining Fund	0.969	0
DWS Global Agribusiness offshore Fund	0.982	0.982
Fidelity Global Real Assets Fund	0.980	0.980
HDFC Core and Satellite Fund	0.996	0.996
HDFC Infrastructure Fund	0.143	0.843
ICICI Prudential Banking and Financial Services Fund	0.983	0
ICICI Prudential Exports and Other Services Fund	0.587	0.404
ICICI Prudential Technology Fund	0.179	0.800
IDFC Strategic Sector (50-50) Equity Fund	0.992	0
ING Global Commodities Fund	0.973	0
JPMorgan India Smaller Companies Fund	0	0.972
Kotak PSU Bank ETF	0	0.975
L&T Infrastructure Fund	0	0.988
LIC Nomura MF Infrastructure Fund	0.551	0.436
Mirae Asset Global Commodity Stocks Fund	0	0.976
PineBridge Infrastructure & Economic Reform Fund	0	0.990
PineBridge World Gold Fund	0	0.971
Reliance Banking Fund	0.781	0.204
SBI Infrastructure Fund	0.991	0
Sundaram Energy Opportunities Fund	0.988	0
Tata Growing Economies Infrastructure Fund	0.965	0.042
UTI Banking Sector Fund	0.981	0

Source: Calculated.

Table A 10 : Diversified fund reference weights (variable benchmark model)-2010-11

<i>Fund Name</i>	<i>CNX Nifty</i>	<i>BSE 100</i>
UTI Equity Fund	0	1.001
Tata Equity Fund	0	0.998
SBI Equity Fund	0	1.003
Religare Invesco Equity Fund	0.704	0.300
Reliance Equity Fund	0.487	0.510
Quantum Equity Fund	0	1.003
Pramerica Equity Fund	0	1.000
PineBridge Equity Fund	0.301	0.707
Kotak Mahindra Equity Fund	0	1.022
JPMorgan Equity Fund	0	0.999
JM Financial Equity Fund	0.495	0.503
ING Equity Fund	0	0.999
IDFC Equity Fund	0.999	0
ICICI Prudential Equity Fund	0.515	0.506
HSBC Equity Fund	0	1.001
Franklin Templeton Equity Fund	0.311	0.696
DSP BlackRock Equity Fund	0	1.002
Deutsche Equity Fund	0	1.000
Canara Robeco Equity Fund	0	1.007
BOI AXA Equity Fund	0.099	0.900
Birla Sun Life Equity Fund	0	0.999
Baroda Pioneer Equity Fund	0.292	0.709
Axis Equity Fund	0.999	0.999

Source: Calculated.

Table A 11 : Diversified fund reference weights (variable benchmark model)-2011-12

<i>Fund Name</i>	<i>CNX Nifty</i>	<i>BSE 100</i>
UTI Equity Fund	1.012	1.012
Tata Equity Fund	1.012	1.012
SBI Equity Fund	1.011	1.011
Religare Invesco Equity Fund	1.015	1.015
Reliance Equity Fund	1.005	1.005
Quantum Equity Fund	1.008	1.008
Pramerica Equity Fund	1.006	1.006
PineBridge Equity Fund	1.013	1.013
Kotak Mahindra Equity Fund	1.030	1.030
JPMorgan Equity Fund	1.009	1.009
JM Financial Equity Fund	1.005	1.005
ING Equity Fund	1.005	1.005
IDFC Equity Fund	1.006	1.006
ICICI Prudential Equity Fund	1.030	1.030
HSBC Equity Fund	1.010	1.010
Franklin Templeton Equity Fund	0.995	0.995
DSP BlackRock Equity Fund	1.006	1.006
Deutsche Equity Fund	1.010	1.010
Canara Robeco Equity Fund	1.012	1.012
BOI AXA Equity Fund	1.008	1.008
Birla Sun Life Equity Fund	1.008	1.008
Baroda Pioneer Equity Fund	1.004	1.004
Axis Equity Fund	1.010	1.010

Source: Calculated.

Table A 12 : Diversified fund reference weights (variable benchmark model)-2012-13

<i>Fund Name</i>	<i>CNX Nifty</i>	<i>BSE 100</i>
UTI Equity Fund	0.998	0
Tata Equity Fund	0	0.998
SBI Equity Fund	0.997	0
Religare Invesco Equity Fund	0	1
Reliance Equity Fund	0	0.991
Quantum Equity Fund	0.563	0.431
Pramerica Equity Fund	0.992	0
PineBridge Equity Fund	0	0.999
Kotak Mahindra Equity Fund	1.015	0
JPMorgan Equity Fund	0.995	0
JM Financial Equity Fund	0.991	0
ING Equity Fund	0	0.990
IDFC Equity Fund	0.572	0.420
ICICI Prudential Equity Fund	1.015	0
HSBC Equity Fund	0.996	0
Franklin Templeton Equity Fund	0.569	0.413
DSP BlackRock Equity Fund	0.000	0.992
Deutsche Equity Fund	0.194	0.802
Canara Robeco Equity Fund	0.182	0.816
BOI AXA Equity Fund	0.182	0.812
Birla Sun Life Equity Fund	0.566	0.428
Baroda Pioneer Equity Fund	0.903	0.087
Axis Equity Fund	0	0.996

Source: Calculated.

Table A 13: Sectoral equity fund wise efficiency performance (constant returns to scale)

<i>Fund Name</i>	<i>2010-11</i>	<i>2011-12</i>	<i>2012-13</i>
Baroda Pioneer Infrastructure Fund	0.9877	0.9777	0.9779
BOI AXA Focused Infrastructure Fund	0.9892	0.9765	0.9768
Canara Robeco Infrastructure	0.9773	0.9745	0.9741
DSP BlackRock World Energy Fund	0.9913	1.0000	1.0000
DSP BlackRock World Gold Fund	0.9946	1.0000	1.0000
DSP BlackRock World Mining Fund	1.0000	0.9982	0.9973
DWS Global Agribusiness offshore Fund	0.9791	1.0000	0.9856
Fidelity Global Real Assets Fund	0.9845	0.9877	0.9868
HDFC Core and Satellite Fund	0.9813	0.9707	0.9707
HDFC Infrastructure Fund	0.9840	0.9813	0.9795
ICICI Prudential Banking and Financial Services Fund	0.9908	0.9938	0.9845
ICICI Prudential Exports and Other Services Fund	0.9822	0.9769	0.9773
ICICI Prudential Technology Fund	0.9876	0.9950	0.9880
IDFC Strategic Sector (50-50) Equity Fund	0.9846	0.9746	0.9749
ING Global Commodities Fund	0.9888	1.0000	0.9944
JPMorgan India Smaller Companies Fund	0.9847	1.0000	0.9951
Kotak PSU Bank ETF	1.0000	0.9982	0.9907
L&T Infrastructure Fund	0.9876	0.9813	0.9784
LIC Nomura MF Infrastructure Fund	1.0000	0.9787	0.9803
Mirae Asset Global Commodity Stocks Fund	0.9790	0.9907	0.9907
PineBridge Infrastructure & Economic Reform Fund	0.9789	0.9789	0.9762
PineBridge World Gold Fund	0.9943	0.9960	0.9946
Reliance Banking Fund	0.9913	0.9835	0.9826
SBI Infrastructure Fund	0.9825	0.9749	0.9759
Sundaram Energy Opportunities Fund	0.9858	0.9831	0.9783
Tata Growing Economies Infrastructure Fund	0.9769	0.9601	0.9610
UTI Banking Sector Fund	0.9933	0.9881	0.9869

Source: Calculated.

Table A 14: Diversified equity fund efficiency performance (constant returns to scale)

<i>Fund Name</i>	<i>2010-11</i>	<i>2011-12</i>	<i>2012-13</i>
UTI Equity Fund	0.9830	0.9689	0.9699
Tata Equity Fund	0.9859	0.9691	0.9697
SBI Equity Fund	0.9810	0.9698	0.9707
Religare Invesco Equity Fund	0.9799	0.9707	0.9673
Reliance Equity Fund	0.9865	0.9756	0.9762
Quantum Equity Fund	0.9809	0.9727	0.9736
Pramerica Equity Fund	0.9839	0.9741	0.9755
PineBridge Equity Fund	0.9767	0.9678	0.9687
Kotak Mahindra Equity Fund	0.9631	0.9548	0.9532
JPMorgan Equity Fund	0.9853	0.9712	0.9723
JM Financial Equity Fund	0.9857	0.9781	0.9764
ING Equity Fund	0.9855	0.9769	0.9768
IDFC Equity Fund	0.9855	0.9740	0.9753
ICICI Prudential Equity Fund	0.9636	0.9531	0.9532
HSBC Equity Fund	0.9828	0.9706	0.9714
Franklin Templeton Equity Fund	0.9779	0.9857	0.9861
DSP BlackRock Equity Fund	0.9824	0.9776	0.9750
Deutsche Equity Fund	0.9846	0.9716	0.9719
Canara Robeco Equity Fund	0.9769	0.9692	0.9692
BOI AXA Equity Fund	0.9847	0.9732	0.9736
Birla Sun Life Equity Fund	0.9849	0.9725	0.9735
Baroda Pioneer Equity Fund	0.9824	0.9766	0.9766
Axis Equity Fund	0.9852	0.9740	0.9719

Source: Calculated.

Table A 15: Sectoral equity fund wise efficiency performance (variable returns to scale)

<i>Fund Name</i>	<i>2010-11</i>	<i>2011-12</i>	<i>2012-13</i>
Baroda Pioneer Infrastructure Fund	0.9988	0.9984	0.9989
BOI AXA Focused Infrastructure Fund	0.9983	0.9984	0.9989
Canara Robeco Infrastructure	0.9987	0.9989	0.9992
DSP BlackRock World Energy Fund	1.0000	1.0000	1.0000
DSP BlackRock World Gold Fund	0.9993	1.0000	1.0000
DSP BlackRock World Mining Fund	1.0000	0.9994	0.9983
DWS Global Agribusiness offshore Fund	0.9996	1.0000	0.9998
Fidelity Global Real Assets Fund	0.9998	0.9997	0.9993
HDFC Core and Satellite Fund	0.9992	0.9986	0.9991
HDFC Infrastructure Fund	0.9986	0.9986	0.9987
ICICI Prudential Banking and Financial Services Fund	0.9994	0.9992	1.0000
ICICI Prudential Exports and Other Services Fund	0.9989	0.9989	1.0000
ICICI Prudential Technology Fund	0.9994	0.9996	0.9996
IDFC Strategic Sector (50-50) Equity Fund	0.9993	0.9986	0.9987
ING Global Commodities Fund	1.0000	1.0000	0.9991
JPMorgan India Smaller Companies Fund	0.9989	1.0000	0.9999
Kotak PSU Bank ETF	1.0000	0.9992	0.9986
L&T Infrastructure Fund	0.9984	0.9987	0.9991
LIC Nomura MF Infrastructure Fund	1.0000	0.9983	0.9993
Mirae Asset Global Commodity Stocks Fund	0.9995	0.9991	0.9994
PineBridge Infrastructure & Economic Reform Fund	0.9987	0.9987	0.9991
PineBridge World Gold Fund	0.9995	0.9995	0.9976
Reliance Banking Fund	0.9995	0.9990	1.0000
SBI Infrastructure Fund	0.9982	0.9982	0.9990
Sundaram Energy Opportunities Fund	0.9985	0.9989	0.9990
Tata Growing Economies Infrastructure Fund	0.9990	0.9990	0.9996
UTI Banking Sector Fund	0.9993	0.9992	0.9997

Source: Calculated.

Table A 16: Diversified equity fund efficiency performance (variable returns to scale)

<i>Fund Name</i>	<i>2010-11</i>	<i>2011-12</i>	<i>2012-13</i>
UTI Equity Fund	0.9991	0.9990	0.9996
Tata Equity Fund	0.9987	0.9990	0.9996
SBI Equity Fund	0.9989	0.9989	0.9996
Religare Invesco Equity Fund	0.9988	0.9991	1.0000
Reliance Equity Fund	0.9983	0.9988	0.9998
Quantum Equity Fund	0.9989	0.9987	0.9995
Pramerica Equity Fund	0.9983	0.9989	0.9996
PineBridge Equity Fund	0.9987	0.9989	0.9995
Kotak Mahindra Equity Fund	0.9990	0.9995	0.9995
JPMorgan Equity Fund	0.9990	0.9987	0.9996
JM Financial Equity Fund	0.9986	0.9965	0.9996
ING Equity Fund	0.9990	1.0000	0.9995
IDFC Equity Fund	0.9990	0.9988	0.9997
ICICI Prudential Equity Fund	0.9990	0.9993	0.9995
HSBC Equity Fund	0.9989	0.9987	0.9994
Franklin Templeton Equity Fund	0.9992	0.9995	0.9998
DSP BlackRock Equity Fund	0.9989	0.9989	0.9995
Deutsche Equity Fund	0.9988	0.9988	0.9997
Canara Robeco Equity Fund	0.9990	0.9991	0.9995
BOI AXA Equity Fund	0.9988	0.9989	0.9995
Birla Sun Life Equity Fund	0.9986	0.9987	0.9995
Baroda Pioneer Equity Fund	0.9977	0.9984	0.9986
Axis Equity Fund	0.9988	0.9991	1.0000

Source: Calculated.