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EVALUATING THE SELECTION AND TIMING ABILITIES OF A MUTUAL FUND

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Abstract: The paper presents the methodology and a case study to evaluate the performance of a mutual fund by taking a look at the timing and selection abilities of a portfolio manager. Separating the timing and selection abilities of the fund manager is taken into consideration by two major models. The data about the mutual fund chosen for study is the German blue chip fund "DWS Deutsche Aktien Typ O", which includes most of the DAX 30 companies. The data consists of 117 monthly observations of the fund returns from January 1999 to September 2008. We used EViews to analyse the data.

Key words: selection ability, timing ability, portfolio risk, regression analysis.

1. Methodology

The literature discusses three major models to evaluate timing and selection abilities. At first we considered taking a look at the overall performance of the fund manager. Therefore we decided to use Jensen's Alpha (1968) model:

 $R_{pt} - R_{ft} = \alpha_J + \beta^* (R_{mt} - R_{ft}) + u_{pt}$

Although Jensen assumes stationarity in systematic risk, which is not the case in an actively managed fund over a long period of time, we used it to provide an image of the overall performance.

In a next step we wanted to separate the timing and the selection abilities of the fund manager by taking into consideration two major models: Treynor and Mazuy (1966) and Henriksson and Merton (1981). As a result of several empirical studies about the reliability of the Treynor and Mazuy (1966) model that showed that its beta estimates are biased (see e.g. Grinblatt and Titman (1991)), we decided not to use this model in our analysis. Hence, we

decided to choose the model of Henriksson and Merton (1981):

$$\begin{split} R_{pt} - R_{ft} &= \alpha_T + \beta_u {}^*X_{ut} + \beta_d {}^*X_{dt} + u_{pt} \\ \text{where} \\ X_{ut} &= max \; [0, \, R_{mt} - R_{ft}]; \end{split}$$

$$X_{dt} = \min[0, R_{mt} - R_{ft}];$$
 and

 u_{pt} = random error term.

 $(R_{pt} - R_{ft})$ is the excess return of the fund p over the risk-free rate f. $(R_{mt} - R_{ft})$ is the excess return of the market portfolio *m* over the risk-free rate *f*.

The main advantage of using this model is that it clearly separates the fund manager's timing and selection abilities.

The selection ability is shown by the intercept α_T , while β_u represents the timing ability in an up-market, β_d in a down-market, respectively. In order for the fund manager to have selection ability, α_T should be statistically significant and above zero.

As for the timing ability, the up-market β_u and the down-market β_d should be significantly different from each other (H₀:

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 $\beta_u = \beta_d$) and for a good market timer β_u should be greater than β_d .

In this case a fund increases its advantages in an up-market by increasing its systematic risk and reducing the negative effects in a down-market by reducing its systematic risk.

2. Data

For the mutual fund we chose the German blue chip fund "DWS Deutsche Aktien Typ O" (ISIN: DE0008474289) [1] which includes most of the DAX 30 companies. The data consists of 117 monthly observations of the fund returns from January 1999 to September 2008.

For the market portfolio, we chose the DAX 30 PERFORMANCE index because it is representative for the German market's blue chips and it includes the same equities as in the fund's portfolio. We calculated the returns using the

continuous compound returns formula $R_t=100 * LN(Pt/P_{t-1}).$

For the risk-free rate we chose the 3 months EURIBOR, which is generally used in the Euro-zone. We divided the annualised EURIBOR data by 12 to be consistent with the monthly returns of the fund and market portfolio. We collected our data from Datastream.

3. Empirical Results

Estimating the Jensen regression we came to the following results:

 $R_{pt} - R_{ft} = 0.001152 + 1.004246^{*}(R_{mt} - R_{ft})$ (0.001701) (0.025269)

Running the t-test on the coefficients shows that the estimated Jensen Alpha of 0.001152 – although positive - is not significantly different from zero.

The Beta coefficient is highly significant, as it may be seen in the regression from Table 1.

JENSEN REGRESSION

Table 1

Variable	Coefficient	Std. Error	t-Statistic	Prob.
EX_RET_MK C	1.004246 0.001152	0.025269 0.001701	39.74172 0.677452	0.0000 0.4995
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood Durbin-Watson stat	0.932130 0.931539 0.018399 0.038929 302.4630 2.405850	Mean deper S.D. depend Akaike info Schwarz cri F-statistic Prob(F-stati	lent var criterion terion	9.01E-05 0.070318 -5.136120 -5.088903 1579.404 0.000000

HENRIKSSON AND MERTON REGRESSION Table 2

Dependent Variable: EX Method: Least Squares Date: 10/01/08 Time: 1 Sample: 1999M01 2008 Included observations: 1	 5:44 M09			
Variable	Coefficient	Std. Error	t-Statistic	Prob.
DUMMY1*EX_RET_MK		0.057585	18.38335	0.0000
DUMMY2*EX_RET_MK		0.040198	24.16515	0.0000
C		0.002700	-0.389195	0.6979
R-squared	0.932780	Mean dependent var		9.01E-05
Adjusted R-squared	0.931601	S.D. dependent var		0.070318
S.E. of regression	0.018390	Akaike info criterion		-5.128661
Sum squared resid	0.038556	Schwarz criterion		-5.057836
Log likelihood	303.0267	F-statistic		790.9658
Durbin-Watson stat	2.478246	Prob(F-statistic)		0.000000

Not going into more detail with Jensen's model we now analyze the results of our main model, Henriksson and Merton (1981). We modelled the min/max-operators by using 2 dummy variables: Dummy1 $\begin{cases} 1 \text{ if } R_{mt} > R_{ft} \end{cases}$ $\begin{array}{c} 0 \ otherwise \\ Dummy2 & \left\{ \begin{array}{l} 1 \ if \ R_{nt} < R_{ft} \\ 0 \ otherwise \\ and \ so \ estimated \ the \ regression, \ from \\ Table 2. \end{array} \right. \end{array}$

$$\begin{array}{l} R_{pt} \mbox{-} R_{ft} = -0.001051 \mbox{+} 1.058612 \mbox{*} (R_{mt} \mbox{-} R_{ft}) \mbox{*} Dummy1 \mbox{+} 0.971393 \mbox{*} (R_{mt} \mbox{-} R_{ft}) \mbox{*} Dummy2 \\ (0.002700) \mbox{-} (0.057585) \mbox{(} 0.040198) \end{array}$$

Having estimated this regression, we checked if the OLS assumptions hold for our model.

- $E[u_t] = 0$; this is true as we have an intercept in the regression α_T .
- Var(u_t) = $\sigma^2 < \infty$; the White test X^2 probability of 0.942216 shows that we cannot reject the H₀: Homoskedastic behavior – therefore we have no evidence for heteroskedasticity (Table 3):

WHITE TEST

Table 3

White Heteroskedasticity Test:

F-statistic	0.185879	Prob. F(4,112)	0.945326
Obs*R-squared	0.771585	Prob. Chi-Square(4)	0.942216

Test Equation: Dependent Variable: RESID*2 Method: Least Squares Date: 10/01/08 Time: 16:30 Sample: 1999M01 2008M09 Included observations: 117 Collinear test regressors dropped from specification

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C DUMMY1*EX_RET_MK (DUMMY1*EX_RET_MK)^2 DUMMY2*EX_RET_MK (DUMMY2*EX_RET_MK)^2	-0.000528	0.000119 0.004873 0.037938 0.003249 0.014143	2.845771 -0.448051 0.653120 -0.162591 -0.234044	0.0053 0.6550 0.5150 0.8711 0.8154
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood Durbin-Watson stat	0.006595 -0.028884 0.000551 3.40E-05 714.5316 1.764434	Mean deper S.D. depend Akaike info Schwarz cri F-statistic Prob(F-stati	dent var criterion terion	0.000330 0.000543 -12.12874 -12.01070 0.185879 0.945326

• $Cov(u_i, u_j) = 0$; at first we ran the Durbin Watson test. The result was inconclusive, because the DW test statistic was in the range of 2.42 (4-d_U) to 2.50 (4-d_L) – *see Table 2*. Then we ran the Breusch-Godfrey test with 12 lags because we used monthly data and any autocorrelation can appear within one year and therefore should be tested. We could not reject the H_0 : no autocorrelation at a 5% significance level because of a X^2 probability of 0.062404, as presented in Table 4:

BREUSCH GODFREY TEST Table 4
Breusch-Godfrey Serial Correlation LM Test:

	Prob. F(12,102) Prob. Chi-Square(12)	0.061351 0.062404
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Test Equation:	
Dependent Variable: RESID	
Method: Least Squares	
Date: 10/01/08	
Sample: 1999M01 2008M09	
Included observations: 117	
Presample missing value lagged residuals set to zero.	

Variable	Coefficient	Std. Error	t-Statistic	Prob.
DUMMY1*EX_RET_MI DUMMY2*EX_RET_MI C RESID(-1) RESID(-2) RESID(-3) RESID(-3) RESID(-4) RESID(-5) RESID(-5) RESID(-6) RESID(-6) RESID(-7) RESID(-7) RESID(-10) RESID(-11) RESID(-12)		0.059640 0.041147 0.002740 0.103246 0.104053 0.108813 0.108813 0.110832 0.111935 0.109575 0.112971 0.112971 0.112971 0.113605 0.112493 0.113992 0.110658	0.618107 -0.987935 -0.761772 -2.756682 1.608923 1.503440 -0.213993 -0.811823 0.476714 1.556024 0.097131 -0.064200 -1.744020 0.265928 1.196346	0.5379 0.3255 0.4480 0.0069 0.1107 0.1358 0.8310 0.4188 0.6346 0.1228 0.9228 0.928 0.9489 0.0842 0.7908 0.2343
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood Durbin-Watson stat	0.173124 0.059632 0.017679 0.031881 314.1476 1.914872	Mean dependent var S.D. dependent var Akaike info criterion Schwarz criterion F-statistic Prob(F-statistic)		7.12E-19 0.018231 -5.113633 -4.759509 1.525421 0.115185

 $1.x_t$ are non-stochastic, but discrete observations

2. u_t normal distributed ~N(0, σ^2); therefore we ran the Jarque-Bera normality test. We could not reject the H₀: normally distributed residuals at a 5% significance level because of the 0.180883 probability. The test is presented in the chart from Figure 1.

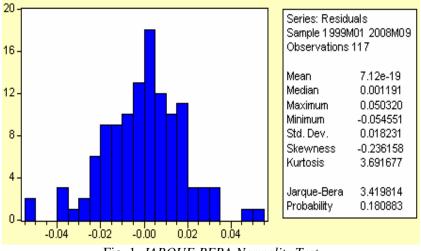


Fig. 1. JARQUE-BERA Normality Test

Further assumptions for correct estimation:

1. Multi-co-linearity; the correlationmatrix shows a coefficient of 0.423517 which is below the critical value of 0.8 for near multi-co-linearity. Therefore we conclude no multi-co-linearity. 2. Linearity; we conducted the Ramsey RESET test with two fitted variables and could not reject the H_0 : linearity (*t* probabilities for fitted values 0.5566 and 0.1771), presented in Table 5.

RAMSEY RESET TEST Ramsey RESET Test:			Table 5	
F-statistic Log likelihood ratio	1.221915 2.525476	Prob. F(2,112) Prob. Chi-Square(2)	0.298563 0.282878	
Test Equation:				

Dependent Variable: EX_RET_FUND Method: Least Squares Date: 10/01/08 Time: 16:58 Sample: 1999M01 2008M09 Included observations: 117

Variable	Coefficient	Std. Error	t-Statistic	Prob.
DUMMY1*EX_RET_MK DUMMY2*EX_RET_MK C FITTED*2 FITTED*3		0.152148 0.120932 0.004073 0.805379 1.989961	7.514614 8.092815 -0.372612 -0.589630 -1.358423	0.0000 0.0000 0.7101 0.5566 0.1771
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood Durbin-Watson stat	0.934216 0.931866 0.018355 0.037733 304.2894 2.487544	Mean deper S.D. depen Akaike info Schwarz cri F-statistic Prob(F-stati	dent var criterion terion	9.01E-05 0.070318 -5.116058 -4.998016 397.6336 0.000000

3.Parameter Stability; we considered the Chow break point test and the Predictive Failure test, however the excess returns of the fund graph show no obvious break points, as in the chart from Figure 2.

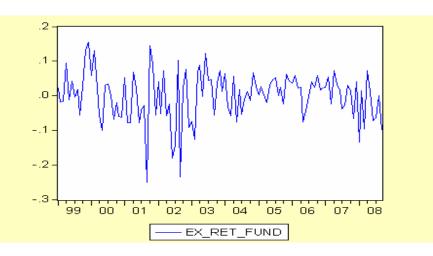


Fig. 2 Excess Returns of the Fund

Coming back to our original regression, hypotheses (H_0 : all coefficients are zero) of we conducted F- and t-tests. The null the F-test is rejected.

	WALD TEST		Tab	le 6
Wald Test: Equation: HENRIK	_REQ			
Test Statistic	Value	df	Probability	
F-statistic Chi-square	1.103656 1.103656	(1, 114) 1	0.2957 0.2935	
Null Hypothesis Su	ımmary:			
Normalized Restriction (= 0)		Value	Std. Err.	
C(1) - C(2)		0.087219	0.083022	

Restrictions are linear in coefficients.

Running the t-test shows that the β_u and β_d are highly significant, while α_T is statistically not significant, as shown in Table 6.

Finally we conducted the Wald test to determine whether β_u and β_d are statistically different from each other (H₀: $\beta_u - \beta_d = 0$). Taking a look at the test

statistics in Table 6, we failed to reject the null hypotheses at the 5% significance level, since the X^2 probability is 0.2935.

4. Conclusion

We found that the overall performance ability of the fund manager, as estimated in the α_J by the Jensen (1968) model is positive. However this coefficient is statistically insignificant, which means the fund is, in a statistical sense, not overperforming the market.

Separating the timing and selection abilities by using the Henriksson and Merton (1984) model, we found β_u greater than β_d , which could let us conclude that the fund manager has market timing ability.

However, we did not find the betas to be significantly different from each other. Therefore the fund manager is a poor market-timer. To evaluate his selection ability we took a look at the α_T of the regression. It is negative and not significantly different from zero. That shows the fund manager has no selection ability either.

A possible explanation for these results is the structure of the fund. The DWS Deutsche Aktien Typ O fund consists mainly of German blue chips and therefore is highly correlated with the German DAX 30 PERFORMANCE INDEX. This is also shown in the high R^2 (approximately 0.93 for both analysed models). That is one possible reason why it is difficult to over-perform the market.

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Notes

1. http://www.dws.de/DE/facts/FactSheet Holdings.aspx?FundId=286