

# Comparisons of Asset Pricing Models in the Egyptian Stock Market

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## Abstract

This paper employs GRS test to empirically compare the applicability of five alternatives of asset pricing models for 55 shares listed on the EGX100 for the Egyptian stock market: 1) the CAPM, 2) the Fama-French three factor model, 3) the Cahart four factor model, 4) liquidity-augmented four factor model, 5) and the five factor model (liquidity and momentum-augmented Fama-French three factor model). The sample is split into six portfolios sorted on size and book-to market ratio and 45 shares are excluded due to data unavailability. Our results based on GRS (1989) show evidence that Fama-French model is the best and reject the other models.

**Keywords:** Fama-French, Three factor model, Emerging markets, Asset pricing models

## 1. Introduction

The poor empirical performance of the capital asset pricing model in developed markets (Fama & French, 2004), and the violations of the key assumptions of standard asset pricing theory caused by emerging markets' s characteristics like the Egyptian stock market (Harvey, 2001) raise the question whether that may affect asset pricing or not. A good asset pricing model can help investors making dependable investment decisions when they develop investment strategies and construct an investment portfolio.

While there are plenty of studies that investigate asset prices in emerging markets, only a small number of them care about African emerging markets in general and Egyptian emerging market in particular. In other words, Egyptian stock market has a faint presence in the literature of asset pricing theory. The popular research papers are: Appiah-Kusi and Menyah (2003) study 11 emerging markets in Africa to test the weak-form pricing efficiency by using weekly index returns adjusted for thin trading which is considered as nonlinear autoregressive process with conditional heteroscedasticity, Cheng, Janhan-Parvar and Rothman (2010) provide an evidence on excess return on market portfolio in Middle East and North Africa countries within the framework of three extensions of the CAPM: the static international CAPM; the constant-parameter intertemporal CAPM; and a Markov-switching intertemporal CAPM, Omran and Pointon (2004) attempt to bridge the gap of the relationship between cost of capital and firm characteristics, Alagidede and Panagiotidis (2009) analyze the movements of stock returns in the biggest African stock markets including Egypt. Alagidede (2011) investigates the predictability of returns in African stock market and analyzes the first two moments of return variations.

The study sheds some light on the Egyptian stock market that differs from developed markets in terms of informational efficiency and shows different return patterns accordingly (Sakr, Ragheb, Ragab & Abdou, 2014). as an Emerging market, Egyptian stock market is of importance because of its number of listed securities (213), number of investors and volume of trading (2.3 billion) (Abd El Aal, 2011). Furthermore, it is deeply ingrained in comparison with the other markets in the MENA region and has attracted many investors over the past few years, resulting in a significant increase in the aggregate capitalization and in the market indices (Billmeier and Massa, 2008).

Nonetheless, Fama-French three factor model (1993) as a tool for describing the relationship between risk and return has been given insufficient attention in the Egyptian stock market. According to the authors' knowledge, we are the first to deal with Fama-French three factor model and its variants, comparing them with the CAPM in Egypt in order to provide the most plausible model. Hence, the key question of the paper is which asset pricing model works properly in the Egyptian stock market.

The rest of the paper is organized as follows. In section 2, we provide the background of related works. Section 3 provides descriptions of data and sample. Section 4 focuses on methodology including the method of constructing portfolios used for empirical analysis and construction of factor portfolios. Section 5 shows the empirical study and analysis. Section 6 reports the concluding remarks and recommendations for future research.

## 2. Literature review

In recent years, there are several asset pricing models suggested to guide investment practice as a result of widespread criticisms of the capital asset pricing model. Thus, comparisons become more justifiable than before to pick the best one out in a business environment.

Drew and Veerarghavan (2002) compare the CAPM with the Fama-French three factor model in terms of explanatory power for Hong Kong, Korea, Malaysia and Philippines. They find that the Fama-French three factor model outperforms the CAPM in explaining the variation of stock returns. Durack, Durand and Maller (2004) point out the CAPM is better than Arbitrage pricing model presented by Chen, Roll & Ross (1986) in terms of  $R^2$  but it is slightly worse than the Fama-French three factor model. Bartholdy and Peare (2005) compare the performance of the CAPM and the Fama-French three factor model. They conclude that Fama-French model does not do much better than the CAPM. Grauer and Janmaat (2010) explore that cross-sectional test results provide little or no support for either model. However, when size and book-to-market factors in the Fama-French framework are added to the CAPM, the performance improves. Rossi (2001) uses monthly data to compare Fama-French three factor model and CAPM in the Italian stock market and discerns that risk-return relationship cannot be described by beta, hence putting size and book-to-market ratio into market return leads to improved description of the returns. Gregory, Tharyan and Christidis (2013) follow new methodology to form factor portfolios using CPZ-style market capitalization weighting of SMB, HML and UMD to liken Fama-French three factor model to Cahart four-factor model in the London stock exchange. Their results do not strongly support value weighting and decomposing factors. Nguyen, Fafaa and Gharghor (2009) depend on all securities in the Australian stock exchange during the period 1990 to 2005, aiming at testing GDP-augmented Fama-French three factor model. They conclude adding GDP to Fama-French three factor model does not make a significant difference. Trimech, Kortas, Benammou and Benammou (2009) investigate the ability of Fama-French three factor model to predict stock returns at multiple time scales in the French stock market. The more wavelet scale surges, the more predictability of the model explained. Tsuji (2012) argues incorporation of momentum and reversal factors in the Fama-French three factor model causes positive alpha in the Japanese stock market.

Some studies are done in emerging markets. For example, Rogers and Securato (2007) implement comparative study of three alternative asset pricing models in Brazilian stock market. They chase up Fama-French methodology to construct test portfolios and explanatory factors. The findings prefer Fama-French three factor model to the CAPM and the reward beta model developed by Bornholt (2007) in describing expected returns.

Iqbal and Brooks (2007) apply the CAPM and the Fama-French three factor model in the Pakistani stock market using daily, weekly and monthly data. In daily data, the Fama-French model performs well in explaining expected returns, while in case of weekly and monthly data neither estimation approach offers significant explanations of returns.

Iqbal, Brooks and Galagedera (2010) also conduct another study in the Pakistani emerging market and they reject unconditional CAPM. However, when size and book-to-market factors are included in the CAPM, the empirical findings sound better. Kim, Kim and Shin (2012) compare the CAPM; APT motivated model (FF5 is constructed by adding liquidity factor and long-term reversal factors to Fama-French three factor model); the Consumption based CAPM; Intertemporal CAPM and Conditional CAPM. The findings support FF5 as most satisfactorily among the models considered in the study.

Considering the literature analysis above, we can imply that there are bounded literatures studying the Fama-French framework in Africa setting.

## 3. Description of data and sample

The study uses a sample of 55 firms out of 100 stocks listed on EGX 100, the Index of the most 100 active stocks. We use monthly data ranging from January 2003 to December 2007. The market data is gathered from EGID (Egypt for Information Dissemination) that is the official and sole aggregator and authorized distributor of the Egyptian Exchange' Listed Companies' Information. Yet, the annual accounting data is collected from Kompass Financial year book (2004, 2005, 2006, 2007, 2008) issued by Coface Egypt. Finally, the global financial crisis in 2008 has an

enormous effect on the Egyptian market; it made the Egyptian stock market index “EGX30” drops from 12000 points to 3380 points (Abd El Aal, 2011). Therefore the data belongs to 2008 and 2009 are cut away.

The dataset used includes monthly closing price, 3-months Treasury Bills rates as a proxy for risk-free rates of returns, monthly return on market index as a proxy for return on market portfolio. However, the company annual data consists of a number of shares outstanding, trading volume, market value of equity (ME) and M/B ratio (converted to B/M ratio).

Market return is monthly return of Egyptian market index, market capitalization is used as a proxy for size, B/M is total shareholders' equity divided by total book value. Momentum is one-year prior return behavior and liquidity measured by stock turnover that is equal to annually trading volume divided by number of outstanding shares.

#### 4. Methodology

##### 4.1 Construction of size - B/M sorted portfolios

The dependent variables are the excess return on the six test portfolios that are “B/L, B/M, B/H, S/L, S/M and S/H”. The excess return on each portfolio equals the monthly value-weighted average rate of return minus the risk-free interest rate ( $R_i - R_f$ ).

In each year of the sample period, the stocks are divided into two groups- big (B) and small (S)- based on their market capitalization. The sample size is also divided into three groups based on their BE/ME ratio-high (H) with top 30%, medium (M) with middle 40% and low (L) with bottom 30%. The six constructing intersection portfolios are B/L, B/M, B/H, S/L, S/M, S/H - that are formed from the intersection of the two size-and three-BE/ME groups. For example, S/L consists of stocks that are in the small size group and the low BE/ME group.

##### 4.2 Factor Portfolios

The explanatory returns to  $R_m - R_f$ , SMB, HML, WML and IMV are formed in accordance with Fama-French methodology (1993) that uses returns on zero investment portfolios as explanatory variables.

#### 5. The models

The time series regressions of Black, Jensen and Scholes (1972) are used for estimating the following models:

- The Basic CAPM
- The Basic Fama-French three factor model (1993)
- The Cahart four factor model (1997) : Momentum-augmented Fama-French three factor model
- The Chan and Faff model (2005): Liquidity-augmented Fama-French three factor model
- The Five-factor model: Momentum and Liquidity-augmented Fama-French three factor model

Those models are estimated using the stepwise procedure by adding one or more explanatory variables at a time and chase  $R^2$  to decide whether it increases with addition of variables. Moreover, the ordinary least square “OLS” is used to estimate the parameters.

#### 6. The empirical analysis

This analysis is split into two different parts as follows:

##### 6.1 Characteristics of portfolios

The results presented in this paper are from portfolios of stocks selected at different points in time. Table 1 provides summary information for the 6 size-B/M portfolios.

Panel A presents the average number of stocks in each of the six portfolios. Overall, the biggest portfolio is S/H, with an average of 13.6 firms. Whereas the smallest portfolio is B/H, with an average of 3.6 firms only. Compared with prior studies in almost all developed markets, the number of stocks in each portfolio is considerably small. However, this is a remarkable limitation in emerging stock markets. Also, it is observed that the biggest size portfolios (B/H, B/H, B/M) include about 27 stocks (49% from sample size on average), while the smallest size portfolios (S/H, S/M, S/L) compose 28 stocks (51% from sample size on average). According to B/M ratio portfolios, the highest B/M portfolios (B/H, S/H) contain 17 stocks (on average, 31% from sample size), while the lowest B/M portfolios (B/L, S/L) consist of 16 stocks (29% from sample size on average).

Panel A: Average number of stocks in each portfolio by year

Portfolios Year	B / H	B / M	B / L	S / H	S / M	S / L	Total
2003	2	14	11	15	8	5	55
2004	4	10	13	13	12	3	55
2005	3	11	13	14	11	3	55
2006	5	13	10	13	8	6	55
2007	4	13	10	13	9	6	55
Average	3.6	12.2	11.4	13.6	9.6	4.6	55

Panel B shows the average size (market capitalization) in each of size-B/M sorted portfolios. It is noted that the biggest portfolio concerning size (market capitalization) is B/L with 53.8% from the market capitalization of the whole sample, while S/L is the smallest average market capitalization with 7.6% from the overall sample's market capitalization. The biggest size portfolios (B/H, B/M) capture up to 96% from sample size, the smallest size portfolios only capture (S/H, S/M, S/L) 4%. According to B/M ratio portfolios, the highest B/M portfolios (B/H, S/H) consist of 11% of the sample's market capitalization but the lowest B/M portfolios (B/L, S/L) contain up to 54.6% of the sample's market capitalization.

Panel B: Average market capitalization ME (in millions)

This table represents the market capitalization as a proxy for size for our sample. It is measured at the end of year for all companies in the sample.

Portfolios Year	B / H	B / M	B / L	S / H	S / M	S / L	Total
2003	653.5	29432.2	24715.2	960.9	480.7	487.8	56730.3
2004	1331.5	17546.3	61834.5	1305.1	1259.2	191.5	83468.1
2005	2931	92630.7	175994.4	3438.2	2895.5	605.3	278795.1
2006	11012.9	51559	14918.9	3379.7	2683.8	2098.1	85652.4
2007	60114	66574.4	148195	4966	3601.7	2651.1	286102.2
Average	15208.58	51608.52	85131.6	2809.98	2184.18	1206.76	158149.6

## 6.2 GRS test

Gibbons, Ross and Shanken (1989, GRS F-statistic) offer one of the most famous tests for asset pricing models. It uses to examine the empirical fit of models and to test the hypothesis that  $\alpha_i = 0 \forall_i$

$$GRS = \left( \frac{T}{N} \right) \left( \frac{T - N - L}{T - L - 1} \right) \left[ \frac{\hat{\alpha} \hat{\Sigma}^{-1} \hat{\alpha}}{1 + \bar{\mu} \hat{\Omega}^{-1} \bar{\mu}} \right] \sim F(N, T - N - L)$$

$\hat{\alpha}$  = N x 1 vector of estimated intercept.

$\hat{\Sigma}$  = an unbiased estimate of the residual covariance matrix.

$\bar{\mu}$  = L x 1 vector of the factor portfolios' sample means.

$\hat{\Omega}$  = an unbiased estimate of the factor portfolios' covariance matrix.

T = number of observations.

N = number of regressions.

L = number of explanatory factors in the regression.

Because of the significance of intercepts in time series regressions, it is required to use GRS test as a tool to test the precision of each model. Table 2 reviews the performance of the five alternative models. The results show that the Fama-French three factor model outperforms the other models and the GRS test for Fama-French three factor model is smallest figure with a value of 2.336, Hence it cannot be rejected as an asset pricing model due to its ability to describe the variation in cross-section of average returns as well as its P-value is bigger than 1%.

On the other hand, the results do not support the CAPM, the Cahart model, the Chan & Faff model and the five factor model because their P-values are less than 1%. Accordingly, they all fail to explain the cross-section of average returns.

Table 2. GRS results

<b>Models</b>	<b>GRS Test</b>	<b>P-value</b>
CAPM	4.877	0.00058
Fama-French	2.336	0.047
Cahart	4.241	0.0048
Chan & Faff	5.76	0.000164
Five-factor model	6.155	0.00009591

Statistical significance at the 1% level

The table exhibits the GRS test, which is F-statistic. It tests the hypothesis that all six are equal 0. P-value is the probability of obtaining a value at least as big as GRS. CAPM is capital asset pricing model, Fama-French is Fama-French three-factor model, Cahart is momentum augmented Fama-French three-factor model, Chan & Faff is liquidity augmented Fama-French three factor model and Five-factor model is momentum and liquidity augmented Fama-French three factor model.

## 7. Conclusion and Future Research

This study conducted empirical comparison of asset pricing models in the Egyptian stock market, including the CAPM, the Fama-French three factor model, the Cahart model, the Chan & Faff model and the five factor model to identify the model of the best performance for the valuation of equities.

In general, the findings confirm that the use of Fama-French three factor model is the best to capture the cross-section of average returns than the other models do during the period 2003-2007. The market factor obviously has the most predictability power for some size-B/M sorted portfolios but the incorporation of SMB and HML factors in the Fama-French three factor model shows the model superiority over the CAPM and the three variants of the Fama-French three-factor model.

Contrary to popular belief in financial research, one-year momentum surges nothing to the explanatory capability of the Fama-French three factor model for all six size-B/M sorted portfolios. It does not reinforce the applicability of the Cahart model in Egyptian market. Therefore, the validation of the momentum factor seems under questioning in emerging markets. It would be interesting to investigate three-month and six-month momentum in the framework of Fama-French three factor model in the future.

Our results do not provide strong support to the literature regarding the essential role of liquidity in asset pricing models and its compatibility with Fama-French model. This may be due to the defective measure of liquidity "turnover" which involves only one dimension "trading quantity". However, liquidity is multi-dimensional phenomenon consisting of trading quantity, trading cost, trading speed and price impact. Hence, using multi-dimensional measure of liquidity in the framework of Fama-French three factor model is suggested for future works.

These findings have contributed to the growing number of boosts to the Fama-French three factor model. In practice, the Fama-French three factor model can be used for all applications that require estimates of expected stock returns.

The applications include activities such as selecting portfolios, evaluating the performance of managed portfolios, estimating the cost of capital and measuring abnormal return in event studies.

The other recommendations are: 1) investigate new risk factors such as dividend yield, E/P (earning to price ratio), CF/P (Cash flow to price ratio) and Past sales growth in the context of Fama & French methodology in Egyptian stock market. 2) Re-compare the same models using high-frequency data like daily data and weekly data

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