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# Market Efficiency and Day of the Week Effect in the Romanian Stock Market

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Abstract: The presence of calendar anomalies and other such effects in stock markets constitute 9 evidence of an inefficient market on the short term and investors could take advantage of these 10 effects to further increase their profits. Therefore, the main objective of this paper is to determine if 11 calendar anomalies exist in the Romanian stock market. This research began with a review of the 12 literature surrounding the Efficient Market Hypothesis (EMH) followed by an empirical study 13 regarding the day of the week effect in the case of the Romanian Stock Market. The dataset consists 14 of daily closing values for the analyzed indicators, from July 31st, 2013, until August 18th, 2023. 15 We investigated the day of the week effect, in the Bucharest Stock Exchange (BSE) market using the 16 daily closing values of Bucharest Exchange Trading Index (BET) and as proxy for the world market 17 portfolio, we chose the Dow & Jones industrial average and the S&P 500. The initial research results 18 suggested a statistically significant Tuesday effect that is in accordance with the literature. Upon 19 further analysis, and after considering the seasonality component of the global markets, estimated 20 by an ARCH & GARCH model, we obtained that the Tuesday coefficient is, in this instance, 21 decreased. Thus, indicating that although the day of the week effect is present in the BSE market, it 22 is caused by the global market's seasonality component rather than by an endogenous anomaly in 23 the Romanian stock market. These findings suggest that the efficiency of the BSE market might be 24 affected on the short-term. 25

Keywords: efficient market hypothesis; calendar anomalies; Romanian stock market

1. Introduction

In this paper we discussed the Efficient Market Hypothesis (EMH) in all its forms as 29 formulated by Fama (1970). We explored the literature regarding EMH and equilibrium 30 models alongside other statistical and mathematical methods for estimating market 31 efficiency. Furthermore, we proposed to test for the day of the week effect in the Romanian 32 Stock Market. If these kinds of anomalies are present in markets, their efficiency might be 33 affected in the short-term. Such effects are known in the literature as calendar anomalies. 34 This study's primary goal is to determine whether the day of the week exists in the 35 Romanian stock market. Calendar anomalies and other such effects constitute evidence of 36 an inefficiency in the market on the short-term. If such anomalies are present in stock 37 markets, investors can take advantage of these effects to further increase profits. The 38 alternative to buy and hold or dollar cost averaging (DCA) is speculative in nature. Investors 39 shall find and exploit market price variation for their own advantage. This topic's 40implications are of great importance for academics as well. Calendar anomalies such as: 41 January effect, day of the week effect, turn of the month effect, Halloween effect and others, 42 only appear in inefficient markets. 43

An inefficient market can be described as a market in which all available information 44 regarding the price of a security is not fully and freely available to all investors and investors 45 do not all agree on the implications of this information. As we can see from the literature, 46 the day of the week patterns have been exhaustively discussed and studied. Mean returns 47 and distributions of stock prices varies according to the day (see Cross, 1973; French, 1980; 48 Keim & Stambaugh, 1984; Rogalski, 1984; Ajayi et al., 2004; Berument & Kiymaz, 2001; 49 Enescu Student, n.d.; Plastun et al., 2019). The common finding of these above-mentioned 50 studies are mean negative returns on Monday in the historical perspective. The reasons 51 behind such low returns on Monday have led psychologists to hypothesize that this could 52 be because of a pessimistic outlook following the weekend and the beginning of a new work 53 week. Investors might be stressed due to the work that awaits them ahead. This 54 phenomenon has been observed on the first trading day of the week, which is typically 55 Monday; however, the effect is still evident in most Middle Eastern countries where the first 56 day of the week is Sunday (see Shehadeh & Zheng, 2023). More realistically speaking, lower 57 returns on Monday are due to the weekend effect. Investors who are averse to risk prefer to 58 close positions on Friday and exist the market during the weekend. News that are not 59 usually optimistic appear during the weekend causing the market to react Monday morning 60 (as it is closed during the weekend). 61

An anomaly in which the month of January has statistically significant mean returns 62 higher than the rest of the year. Emotion is which drives stock prices more than anything 63 else. These anomalies arise due to the collective emotions, spirit, and mental state of the 64 investors. For example, investors are more likely to have elaborate investment plans, 65 purchase more securities, and drive-up prices in the first month of the year when they are 66 optimistic about the new year and its prospects. This kind of optimism and pessimism 67 among investors is of quite importance in market fluctuations. Another importance of this 68 topic is sensibility to the global market risk and seasonality. Emerging markets are more 69 prone to be influenced by the seasonality of developed markets, this is quite perfectly 70 reflected in the saying "If the USA sneezes, the whole world catches a cold" In the literature 71 this phenomenon was studied using models that account for time-varying volatility such as 72 ARCH and GARCH models. 73

Considering all this, the objective of our research was to empirically test the day of the week effect in the Romanian stock market.

We used the daily closing values of Bucharest Exchange Trading Index (BET) from 76 2013 to 2023, the closing values for the same period for both the S&P500 and The Dow & 77 Jones industrial average from www.dow.com, to serve as a proxy for the world market 78 portfolio and to account for the seasonality in returns. We calculated the one period 79 returns for the entire data set and proceeded with an OLS regression model at first, 80 followed by goodness of fit tests such as Breusch-Pagan test for heteroskedasticity, Breusch-81 Godfrey test for autocorrelation, and elaboration of a correlogram to further inspect the 82 heteroskedasticity of residuals. An autoregressive conditional heteroskedasticity model 83 was applied to the BET daily return index values as a function of the day of the week and 84 global market risk. Elaboration of ARCH & GARCH model was done by considering the 85 time-varying volatility of our residuals. 86

The rest of the paper is structured as follows: Literature Review, Methodology and data, Results and discussion and Conclusions. 88

#### 2. Literature Review

The study of calendar anomalies and market efficiency have been already 90 exhaustively discussed in the literature from the 1960's onwards (see DIMA et al., 2021; 91 Dima & Milo, 2009; Dimson & Mussavian, 1998; Fama, 1970; Fama et al., 1969; Pleşoianu 92 et al., 2012). For discussions on the random walk hypothesis and measurements of the 93 efficiency of capital markets see Beran, 1992; Dittrich & Srbek, 2020; Fama, 1965, 1970b; 94 Fama et al., 1969; Mandelbrot', n.d.; Mandelbrot & Hudson, n.d. 95

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Fama (1970) has defined that the capital market's main concern is ownership 96 allocation of the economic capital in the most efficient, direct, and transparent way 97 possible. An efficient market described by Fama (1970) is a market in which all necessary 98 information regarding a security is freely available and accessible to all participants, and 99 at the same time, it requires that all investors agree on the implications of this freely 100 available information. Everyone must believe that a security's price is determined 101 equitably by the market, that it is independent of its own historical values, and that 102 nothing and no one can alter how future prices are distributed. Fama identified and 103 described three levels of market efficiency, the weak-form, the semi-strong form, and the 104 strong-form efficiency. 105

According to Fama (1970), the principal role of capital markets around the world is ownership allocation of economic capital. The ideal market is one in which prices of securities fully reflect all available information at any given time, and all investors agree with the information available. He provided economists with three simple conditions that are sufficient to consider any market efficient: 100

- (i) there are no trading costs, commissions and other expenses for investors 111 trading securities, 112
- (ii) all information is available free of charge and easily accessible to all 113 investors, at any given time; 114
- (iii) all investors agree with the implications of current information for the current price and distributions of future prices for each security.

In a market that satisfies all the above conditions, obviously, is a market in which security prices fully reflect all given information, but such a frictionless market is not characteristic of markets met in practice.

The weak form of the Efficient Market Hypothesis (EMH) states that the previous 120 prices of a given security are the sole relevant data set for determining its present price. 121 Investors shouldn't anticipate seeing any trends in the security's past pricing that may be 122 used to build complex trading strategies and increase profits. The random walk model 123 was employed in most of the literature to investigate this weak form of EMH. Changes in 124 pricing resulting from new information will appear random since new information is 125 assumed to arrive in an unforeseen way. As a result, price changes in a weak-form efficient 126 market are independent of one another and happen at random. Studies of past prices of a 127 given asset cannot reliably produce excess returns if a market is weak-form efficient since 128 there is no correlation among subsequent prices. Because it is based solely on analyzing 129 historical price patterns without considering any additional contextual information, this 130 type of research is known as technical or chart analysis. Eugene Fama claimed in 1965 that 131 the random walk theory could not fully capture reality, in his study on the behavior of 132 stock market prices (Fama, 1965). 133

Moving forward to the semi-strong form of EMH, it asserts that a market is efficient 134 if the price of a security quickly reflects every relevant piece of information that is 135 available to the public, regarding that security. According to this form of efficiency, the 136 market will swiftly adjust prices to a new equilibrium level that reflects the shift in supply 137 and demand brought about by the release of pertinent new information. The semi-strong 138 form of the EMH gains empirical strength despite its possible lack of intellectual rigor 139 because it is easier to test than the strong form. When considering all publicly available 140 information regarding the risk and return of an investment, the current market price is 141 the best available unbiased predictor of a fair price in a market that is semi-strong efficient. 142 Analysis of any available data, not just historical pricing, cannot produce a steady stream 143 of excess returns. This is a slightly more contentious conclusion than the weak- form EMH 144 because it implies that fundamental analysis, or the methodical examination of businesses, 145 industries, and the economy, is incapable of yielding returns that are greater than those 146 that are warranted by chance. Such a theory cast doubt on the importance and usefulness 147 of investment research and analysis. 148

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In its most robust version, the strong form of the Efficient Market Hypothesis (EMH) 149 states that a market is efficient when the market price promptly and accurately reflects all 150 relevant information regarding a share's value, regardless of whether it is generally 151 available to current or potential investors. For instance, the holders of confidential 152 information will purchase shares to take advantage of the pricing anomaly if the current 153 market price is less than the value supported by that information. They intend to persist 154 in this manner until the price of the shares reaches the level substantiated by their 155 confidential information, given the positive demand for them. They will then be 156 unmotivated to keep purchasing, leaving the market, and the price will level off at this 157 new equilibrium. Calendar anomalies are another crucial subject in the EMH debate. 158 These anomalies, which cannot occur when the market behaves as an efficient one, 159 provide evidence against the EMH. 160

The most studied anomalies are: the month-of-the-year effect, the day-of-the-week 161 effect, the turn-of-the-month effect, the holiday effect, the week-of-the-year effect, January 162 effect, the Halloween effect, etc. It is important to conduct research on these effects and 163 how they affect investors and academics. Rather than employing buy and hold as a long-164 term strategy, investors who base their strategy on speculation and are not risk-averse can 165 take advantage of these effects to further increase returns, in short periods of time. As the 166 literature demonstrates, these effects can be used to take advantage of variations in market 167 prices. Scholars can gain insight into how global markets function by examining these 168 anomalies. 169

Less visible anomalies like the Halloween effect as in (Bouman & Jacobsen, 2002) 170 imply that investors elaborate medium term strategies to gain from these kinds of 171 anomalies. When the bear market begins, usually around the end of May and lasts until 172 October, investors sell all their assets and securities, a tactic known as the "Halloween 173 effect" or "Sell in May and go away". This implies holding cash in the period of May -174 October when stock market returns are significantly lower than the rest of the year and 175 rejoin the stock market in November. Of course, this kind of behavior is exactly what 176 Dimson & Marsh (1998) describe as Murphy's law in the stock market. In this scenario, 177 the bear market that begins in May is caused by investors who decide to sell because they 178 think the market may experience a downturn rather than by some outside "force.". 179 Because enough investors chose to sell and think that the bear market would begin, the 180 market will now actually experience a downturn rather than the other way around. 181 Dimson and Marsh (1998) say that such behavior is the one causing these kinds of 182 anomalies. 183

Tilică & Oprea (2014) studied the day of the week effect in the Romanian stock market 184 using daily closing data of BET from 2005 until 2011 and found the presence of the Friday 185 effect and that seasonality in the stock market returns is closely correlated with the global 186 market risk. Diaconasu et al. (2012) studied the presence of the day of the week effect and 187 month of the year effect on the Bucharest Stock Exchange market for the period of 2000 -188 2011. They observed a statistically significant Thursday effect for the period analyzed but 189 did not find any traditional Monday effect. For the entire sample period, the January effect 190 was not present, thus concluding that the Romanian equity market was reasonably 191 efficient from 2000 - 2011. Enescu (2022) also studied the day of the week effect and 192 January effect in BSE from 2002 until 2022 dividing the series into three subperiods and 193 employing and OLS dummy regression. The Thursday effect was confirmed for this 194 period, but the January effect was only confirmed for the subperiod 2003 - 2007. 195 Concluding that the changes in market conditions for this period have drastically reduced 196 the possibility of speculating from market anomalies. 197

From a historical perspective, Berument & Kiymaz (2001) have studied the day of the week effect on the S&P 500 from 1973 until 1997. For this period the day of the week effect was present with the highest and lowest returns observed on Wednesday and Monday. Patev & Kanaryan (2003) reported negative Monday returns in Romania between 1997 to 2000 thus confirming the presence of a Monday effect. Results seem contradictory at times 202 depending on the approach of the researcher. Ajayi et al. (2004) tested for the day of the 203 week effect in emerging markets of Eastern Europe and did not find the presence of such 204 anomaly. Heininen (2008) reported the disappearance of such anomalies in Central and 205 Eastern Europe after joining the EU. Borges (2009) observed the presence of the day of the 206 week effect only in the Slovenian stock market after joining the EU. 207

## 3. Methodology and data

The research started with identifying the calendar anomalies such as the day of the week effect, which if present suggests that the efficiency of the market might be affected 210 in the short term. 211

The dataset used in this paper was created using the following variables: (1) 212 Bucharest Exchange Trading Index - BET which is a market capitalization weighted index, 213 and its value reflect the price movement of all companies listed on the I<sup>st</sup> and II<sup>nd</sup> 214 category of the Bucharest Stock Exchange (BSE) and was obtained from www.bvb.ro; (2) 215 The Dow & Jones industrial average; (3) And the S&P 500 which both were obtained from 216 www.dow.com and serve as a proxy for the world market portfolio. The applied dataset 217 consists of daily closing values for all afore mentioned indices, from July 31st 2013 until 218 August 18th 2023. 219

The efficient market hypothesis tells us that the price of a security J at time T fully reflects all available information, thus Fama (1970) wrote the following equation to describe the relation:

$$E(p_{j,t+1}|\Phi_t) = [1 + E(r_{j,t+1}|\Phi_t)]p_{jt}$$

Where E is the expected value;  $p_{jt}$  is the price of security j at time t;  $p_{j,t+1}$  is the price at time t+1;  $r_{i,t+1}$  is the one period change in returns;  $\Phi_t$  denotes the information that is assumed to fully reflect in the price of the security at time t. It is assumed however that the information contained in  $\Phi_t$  at time t is fully utilized in determining current equilibrium of expected returns.

The relationship described above has a major implication in empirical work, if current equilibrium of returns fully reflect all given information in  $\Phi_t$ , this completely rules out the elaboration of trading systems based solely on information  $\Phi_t$  that have expected profits more than equilibrium expected profits. If

$$x_{j,t+1} = p_{j,t+1} - E(p_{j,t+1} | \Phi_t)$$

 $E(x_{j,t+1}|\Phi_t) = 0$ The sequence  $\{x_{j,t}\}$  is a "fair game" process in respect to the information sequence

 $\{\Phi_t\}, x_{i,t+1}$  is the excess market value of security j at time t, which by definition of the 237 efficient market, it has to be equal to zero as described above. Thus, lastly, we must assume 238 that: 239

$$E(p_{j,t+1}|\Phi_t) \ge p_{jt} \tag{240}$$

the price sequence  $\{p_{it}\}\$  of security j at time t follows a submartingale with respect to the information denoted in  $\{\Phi_t\}$ , and this tells us that the expected value of next period's price, as projected based on information  $\{\Phi_t\}$  is equal to or greater than the current price.

If calendar anomalies are present in markets met in practice, the price of security j at 244 time t does not fully reflect the information sequence  $\{\Phi_t\}$ , and thus the price sequence 245  $\{p_{it}\}$  will not be equal to or greater than the current price of the security. 246 The expected value operator in respect to the information sequence  $\{\Phi_t\}$ , 247

 $E(x_{j,t+1}|\Phi_t) \neq 0$ 248

will not equal to zero, and thus it renders the market inefficient.

A market that is conditional value operator incorporates beside the information sequence denoted as  $\{\Phi_t\}$ , an additional specific set of information regarding a particular day of the week (i.e., day-of-the-week effect) can be written as follows:

$$E(p_{j,t+1}|\Phi_t + D_{it}) = [1 + E(r_{j,t+1}|\Phi_t + D_{it})]p_{jt}$$
<sup>253</sup>

$$x_{j,t+1} = p_{j,t+1} - E(p_{j,t+1}|\Phi_t + D_{it})$$
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$E(x_{i,t+1} \Phi_t + D_{it}) \neq 0$ or $E(x_{i,t+1} \Phi_t + D_{it}) \geq p_{it}$	257

 $E(x_{j,t+1}|\Phi_t + D_{it}) \neq 0 \text{ or } E(x_{j,t+1}|\Phi_t + D_{it}) \geq p_{jt}$ 

We have calculated the one period returns for the indices using the following 258 formula: 259

$$R_{j,t} = ln\left(\frac{P_{j,t}}{P_{j,t-1}}\right) * 100$$
 260

Where  $R_{j,t}$  is the return of index j at time t;  $P_{j,t}$  is the price of index j at time t;  $P_{j,t-1}$ 261 is the price of index j at time t-1. 262

After computing the one period returns for each index included in our analysis, we 263 proceed with tabulating the dummy variables for each day of the week. Let  $\{D_{it}\}$  be  $\{D_{it}\}$ 264  $\epsilon$  [1, 2, 3, 4,5] the sequence of days in one business week (Monday through Friday), and 265 if  $D_{it} = Monday$  then  $\{D_{it}\} \in [1, 0, 0, 0, 0]$ ; if  $D_{it} \neq Monday$  (lets say it is a Tuesday) then 266  $\{D_{it}\} \in [0, 1, 0, 0, 0]$ , and so forth for each day of the week. Lastly, we have calculated the 267 proxy for world market risk (return) using the Dow & Jones industrial average and the 268 S&P500. After calculating the returns, we determined the RWM proxy such as: 269

$$Rwm = R_{Dija} + R_{Sp500}$$
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Where, RWM is return of world markets;  $R_{Dija}$  is the one period return of DIJA;  $R_{Sp500}$  is the one period return of the S&P500.

To proceed further with our analysis, as we are dealing with time series financial 275 data, we must test the stationarity of our series. To model this kind of data, unit-root must 276 be assumed. Let  $p_t = p_{t-1} + u_t$  where  $u_t \sim \text{IID}(0, \sigma_t^2)$ , then  $p_t$  would be a random walk. 277 Indeed, some economists believe that stock market prices and indices follow a random 278 walk, meaning that the price of a stock today is the price of the stock from yesterday plus 279 a random shock. If this assumption is true, a random walk is a non-stationary time-series, 280 and thus we cannot obtain an estimable model. Any shock received by the stock price will 281 be permanent and will not decay rapidly like other AR (1) processes. We can say that if 282 the initial price of the stock is 283

 $p_0 = u$ , then  $p_1 = u + u_1$ , then  $p_2 = u + u_1 + u_2$ ,..., and so on  $p_t = u + \sum_{j=1}^t u_j$ . As we 284 pointed out earlier, the variance of  $p_t$  is dependent on t and  $p_t$  is not covariance 285 stationary. Fixing this problem implies first differencing  $p_t$  to get  $u_t$  which is stationary 286 (see Baltagi, 2011).

The Augmented Dickey-Fuller test has been applied to our dataset in the following 288 way: First, we tested the stationarity of all indices closing values for the entire period, 289 which of course, are non-stationary, and the p-values are all greater than 0.05 thus 290 rejecting the null. Applying first differences such as  $\Delta p_t = p_t - p_{t-1}$  we get a stationary 291 series. The Augmented Dickey-Fuller test conducted on first differences  $\Delta p_t$  made us 292 accept the null (all p-values < 0.001) and conclude that our series is stationary at first level 293 .As we move further to testing the stationarity of our return variable for each index, as we 294 calculated it with the above mentioned formula and obtained the one-period % change in 295 returns, the Dickey – Fuller test confirmed that the data is stationary at level and has unit-296 root (p-value < 0.001). 297

To test if the day-of-the-week effect is present in the Bucharest Stock Exchange market, we continue with a simple OLS regression. Let  $\{R_{it}\}$  be a function of  $\{D_{it}\}$  $\epsilon$  [1, 2, 3, 4, 5] such as  $f_{(R_{i,t})} = \sum_{i=1}^{5} \beta_i D_{it}$  Or  $R_{j,t} = f(\sum_{i=1}^{5} \beta_i D_{it})$ , then we obtain:

 $R_{i,t} = \alpha_0 + \sum_{i=1}^5 \beta_i D_{it} + \epsilon_i$ 

Where  $R_{i,t}$  – return index of BET;  $\alpha_0$ - mean Monday return;  $\{D_{it}\}$  sequence of days; and  $\epsilon_i$  is our error term which satisfies  $\epsilon_i \sim N(0 | \sigma^2)$ .

If we estimate the above model, we will not get a Monday coefficient like the other 304 days, thus we removed the constant  $\alpha_0$  from the equation and obtained: 305

 $R_{i,t} = \sum_{i=1}^{5} \beta_i D_{it} + \epsilon_i$ 

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Where  $R_{j,t}$  – return index of BET;  $\{D_{it}\}$  sequence of days; and  $\epsilon_i$  is our error term 307 which satisfies  $\epsilon_i \sim N(0 | \sigma^2)$ . 308

After estimating the model using OLS, we run the Breusch-Pagan Test for 310 heteroskedasticity, with a p-value less than 0.05 we cannot accept the null and must 311 assume that heteroskedasticity is present among the residuals. To further inspect the 312 heteroskedastic potential of our residuals, we computed a correlogram and observed that 313 constant variance was not achieved. 314

The data might be autocorrelated due to the nature of financial time-series, therefore 315 we tested it using Breusch-Godfrey LM test for autocorrelation. The p-value of the test 316 was less than 0.05 making us reject the null and declare that autocorrelation is present. To 317 fix this issue we added the lag operator of our dependent variable such as  $\Delta R_{j,t}$  = 318  $R_t - R_{t-1}$ , and added the world market returns to the initial equation and obtained the 319 following model: 320

$$R_{j,t} = \sum_{i=1}^{5} \alpha_i D_{it} + \sum_{i=1}^{5} \beta_i RWM_t + \sum_{j=1}^{n} \gamma_j R_{t-j} + \epsilon_t$$

Where  $R_{j,t}$ - return index of BET;  $\{D_{it}\}$  sequence of days ;  $RWM_t$  – return of world market;  $R_{t-j}$  – lag of BET;  $\alpha_i, \beta_i, \gamma_j$  – regression coefficients;  $\epsilon_t$  – error term which is  $\epsilon_t \sim IID(0,1)$ .

The error term must satisfy  $\epsilon_i \sim N(0 | \sigma^2)$  however the variance of  $\epsilon_i$  is dependent on *t*, this may result in incorrect estimations. Because heteroskedasticity exists, we must include the time-varying variance into the equation to model the volatility of our series.

The expected value operator in respect to the sequence of days  $\{D_{it}\}$  and as a timevarying function of world market return and of its own stochastic process can be written as follows:

$$E(r_{j,t+1}|\mathsf{D}_{it}) = f(x) \left[ \sum_{i=1}^{5} \beta_i \, RWM_t + \sum_{j=1}^{n} \gamma_j \, R_{t-j} + \sigma_t^2 \right]$$
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Where E is the expected value of BET returns at time t+1 regarding the days sequence  $\{D_{it}\}$  and as a time-varying function.

To obtain a testable model that can be estimated using feasible GLS or maximum 334 likelihood, the days sequence  $\{D_{it}\}$  regarded as the bases of the expected value of r at 335 time t + 1 has been included as a variable in the function of the world market portfolio 336 and past returns values such as 337

$$E(r_{j,t+1}|f_{(X)}) = \sum_{i=1}^{5} \alpha_i D_{it} + \sum_{i=1}^{5} \beta_i RWM_t + \sum_{j=1}^{n} \gamma_j R_{t-j} + \sigma_t^2$$
338  
thus, we can say that
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$$R_{j,t+1} = 1 + \left[\sum_{i=1}^{5} \alpha_i D_{it} + \sum_{i=1}^{5} \beta_i RWM_t + \sum_{j=1}^{n} \gamma_j R_{t-j} + \sigma_t^2\right]$$
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$$R_{jt} = \sum_{i=1}^{5} \alpha_i D_{it} + \sum_{i=1}^{5} \beta_i RWM_t + \sum_{j=1}^{n} \gamma_j R_{t-j} + \sigma_t^2$$
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And

 $\sigma_t^2 = \gamma_0 + \gamma_1 u_{t-1}^2 + \delta_1 \sigma_{t-1}^2$ 

Where the conditional variance of  $u_t$  depends upon q of its lagged values and as well as on p of the squared lagged values of  $u_t$  346

With the basic GARCH model specification elaborated above, we tested for any remaining ARCH effects using the Lagrange Multiplier test for ARCH effects as in Engle. 348

# 4. Results and discussion

We have defined earlier that calendar anomalies constitute evidence against the efficient market hypothesis. For those who are making use of such phenomena to generate additional profits, calendar anomalies and other trading systems based on day-to-day price fluctuations in securities provide an unfair advantage. 351

The summary statistics of the analyzed indicators is presented in Table 1.

Table 1. Summary statistics.

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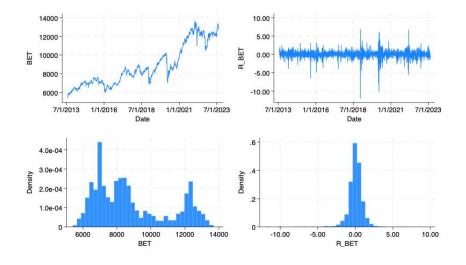
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	Ν	Mean	Median	SD	Min	Max	Kurt.	Skew.
BET	2515	8838.748	8274.540	2149.476	5403.6	13681.92	2.114	.657
R BET	2515	.034	0.060	.979	-11.89	6.82	23.542	-1.579
DIJA	2515	24512.037	24706.350	6545.099	14776.13	36799.65	1.703	.227
R DIJA	2514	.033	0.056	1.107	-13.842	10.764	25.839	954
SP500	2515	2889.468	2723.060	881.534	1630.48	4796.56	1.973	.53
R	2514	.039	0.059	1.118	-12.765	8.968	19.257	815
SP500								
Day	2515	3.011	3.000	1.408	1	5	1.712	007

Source: own computing using STATA 18 software

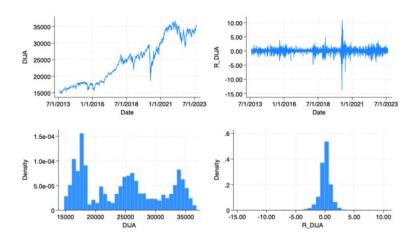
In all cases kurtosis is greater than 3, having leptokurtic distributions we can expect 358 high volatility and risk potential. This high level of kurtosis (greater than 20) is to be 359 expected when working with financial data such as stock prices.

From Figures 1 through 3 we can see the distributions of all indices and their 361 respective daily returns. 362



Source: own computing using STATA 18 software

Figure 1: Trend & Volatility and distributions of BET between 2013-2023

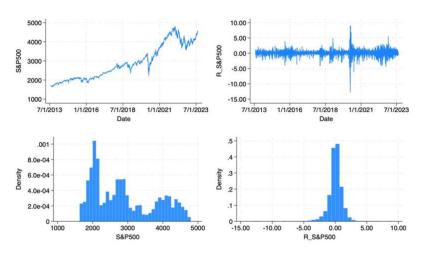


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# Figure 2: Trend & Volatility and distributions of DIJA between 2013-2023

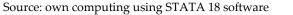


Figure 3: Trend & Volatility and distributions of S&P500 between 2013-2023

Fatter tails are found in all three histograms. If we pay attention to the returns and 373 their according histograms, we might say that they follow a normal distribution. The 374 histograms for returns are dense in the center with no visible tails. This kind of histogram 375 follows a bell-shaped Gaussian distribution suggesting that returns are indeed random 376 and IID ~ N(0 |  $\sigma$ ^2). This independence does not hold however when computing the 377 autocorrelation and partial autocorrelation of each index and their respective returns. 378

The dynamic of BET, DIJA and S&P500 in period 2013-2023 can be seen in Figure 4. 379



Figure 4: Dynamics of BET, DIJA and S&P500 (non-logarithmic scale) 2013-2023 Source: own computing using STATA 18 software

The autocorrelations of BET, DIJA and S&P500 are presented in Figure 5 through Figure 7. By including 40 lags, we see that the present value of BET is autocorrelated with its own past values over the 40-day period. This implies that the weak form of EMH 387

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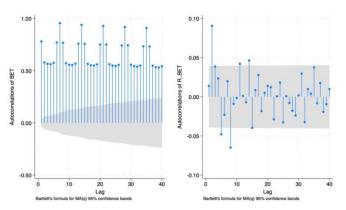
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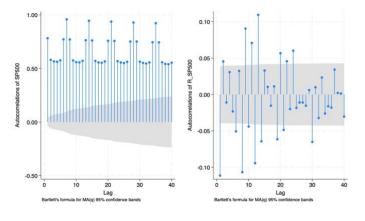
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asserts relevancy in determining current prices of a security only on historical prices of 388 that security that are indeed random as new information arises randomly in the market. 389 390



Source: Own computing using STATA 18 software

Figure 5: Autocorrelation of BET & BET Returns including 40 lags



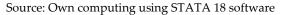
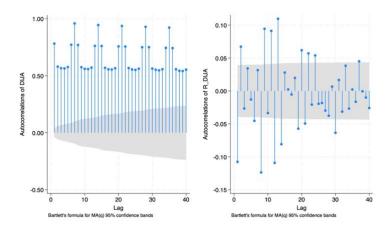


Figure 6: Autocorrelation of SP500 & SP500 Returns including 40 lags



Source: Own computing using STATA 18 software

Figure 7: Autocorrelation of DIJA & DIJA Returns including 40 lags

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Present values of BET are influenced by their own past values and do not follow a 400 random walk. RWH tells us that tomorrow's price is equal to today's price plus a random 401 shock. Nor does a "fair game" model that states the impossibility of elaborating trading 402 systems to obtain higher returns than equilibrium returns hold in this case. For the "fair 403 game" model to be valid, all past lag values and serial covariances must be zero which in 404our case, they are not (Fama, 1970). 405

The normality of our data was tested using the Ljung-Box test, Jarque-Bera test and 406 the Skewness and Kurtosis test for normality. All test results and their according p-values 407 are found in Table 2. All tests reject the hypothesis of normality, the data is not normally 408 distributed. 409

	Ljung-Box	Pr(skew.)	Pr(kurt.)	chi2(2)	Prob>chi2	Jarque-Bera (Chi-squared)
BET	45607.936	0.000	0.000	468.430	0.000	263.4
	(p-value=0.000)					
R BET	82.591	0.000	0.000	1198.610	0.000	4.5e+04
	(p-value=0.000)					
DIJA	46385.052	0.000	0.000	7024.360	0.000	198.0
	(p-value=0.000)					
R DIJA	294.405	0.000	0.000	941.150	0.000	5.5e+04
	(p-value=0.000)					
SP500	46367.652	0.000	0.000	752.070	0.000	228.2
	(p-value=0.000)					
R SP500	249.266	0.000	0.000	792.470	0.000	2.8e+04
	(p-value=0.000)					

Table 2. Ljung-Box, Jarque-Bera and Skewness and Kurtosis tests for normality

Source: own computing using STATA 18 software

From the results reported in Table 3 we can see a statistically significant coefficient 413 for Tuesday, this means that in the analyzed period, Tuesday had abnormally high returns 414 in comparison with other days.

Table 3: OLS Regression Results and	ARCH & GARCH Estimations
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	(1)	(2)	(3)
	OLS	OLS with RWM	ARCH / GARCH
Monday	-0.0165	-0.0195	-0.0142
	(0.0442)	(0.0442)	(0.0253)
Tuesday	0.145***	0.147***	-0.0350
	(0.0435)	(0.0434)	(0.0234)
Wednesday	-0.00490	-0.0102	-0.0282
	(0.0433)	(0.0434)	(0.0243)
Thursday	0.0105	0.00847	-0.00624
	(0.0435)	(0.0435)	(0.0237)
Friday	0.0341	0.0318	0.0131
	(0.0436)	(0.0436)	(0.0239)
$R_{DIJA}$		-0.0918	
		(0.0652)	
<i>RS&amp;P</i> 500		0.132*	
		(0.0646)	
RWM			0.504***

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			(0.00353)
$R_{BET,t-1}$			0.0406***
			(0.0114)
ARCH			
$\mu_{t-1}^2$			0.339***
			(0.0212)
$\sigma_{t-1}^2$			0.565***
			(0.0419)
$\alpha_0$			0.0415**
			(0.0146)
Ν	2515	2514	2514
$R^2$	0.005	0.008	

Standard errors in parentheses, \* *p* < 0.05, \*\* *p* < 0.01, \*\*\* *p* < 0.001

Source: own computing using STATA 18 software

This finding is in line with the Tuesday Effect in the literature (Berument & Kiymaz 2001). Other studies conducted on the Bucharest Stock Exchange found the presence of Thursday and Friday effect (Diaconasu et al. 2012; Tilica & Oprea 2014; Enescu 2022).

All studies mentioned found that between 2000 and 2011 there were negative returns 423 for Tuesday and abnormally high returns for Thursday and Friday. This finding suggests 424 that investors (for the period of 2000 – 2011) should have bought securities on Tuesday 425 and sold on Thursday or Friday for extra profits. But indeed, nature can be perverse and 426 according to Dimson & Marsh 1998, if enough investors discover that stock prices are 427 lower on Tuesday and higher on Friday, they will buy on Tuesday, driving up prices, and sell on Friday, driving down prices. 429

Global markets are subject to Murphy's law, which states that if enough investors 430 take advantage of these calendar anomalies, the markets will reverse, and the opposite 431 will occur. (i.e. higher prices on Tuesday and lower on Friday). 432

This theory is consistent with the study's findings. Between 2013 and 2023 (our sample period) the returns have flipped. Tuesday has the highest mean return of the week, and Thursday and Friday lower than Tuesday. Monday and Wednesday have negative mean returns, indicating that traders may create complex plans to buy on Monday and Wednesday and sell on Tuesday of the following week. However, if enough traders take advantage of this anomaly, it will eventually reverse, making Tuesday returns negative and producing unusually high returns on Monday and Wednesday. 433

In model (2) we also see a significance for the Return of S&P 500 coefficient, thus 440suggesting that besides a relationship between Tuesday returns and BET, we can assume 441 that a positive change in the S&P500 returns will transfer to an increase in BET returns, 442 this makes the BSE market sensible to the S&P500 (this findings are also validated by 443 Diaconasu et al. 2012; Tilica & Oprea 2014; Enescu 2022; ). This kind of relationship 444 between the BSE market and other more influential indices such as the S&P500 makes the 445 Romanian stock market sensible to changes in the global market. This sensibility is 446 primarily the result of market integrations brought about by globalization. 447

The hypothesis of seasonality in returns and lack of market efficiency in the analyzed 448 period was accepted, as the  $\alpha_i$  coefficients of the tested model were statistically 449 significant. The initial statistically significant coefficient for Tuesday returns decreases 450 and becomes insignificant as we estimate the ARCH / GARCH equation. 451

Model (3) suggests that the day of the week effect in the Romanian stock market is due to seasonality in the world market returns, and to the sensibility of BSE to exogenous shocks and not due to endogenous anomalies. The findings of this study serve as evidence 454

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against the efficient market hypothesis in the case of Romania and are in line with studies such as Dragotă & Țilică (2014); Dragota et al. (2009);

Further inspection of seasonality in returns can be conducted using an additive457decomposition method of the time series such in obtaining the trend (or cycle) the seasonal458component and the noise. Such decomposition can be done by employing an unobserved459component model and predicting the seasonal component of that model.460

The result show the cycle (a period of 12 month for example) and the remaining noise 461 of the time series that has been decomposed. This decomposition can tell us if the day of 462 the week effect (and other calendar anomalies) are in fact an outcome of the global market 463 or are there any other factors that influence the returns according to the day of the week. 464 Indeed, our results suggest that there is a seasonal component causing the disturbances 465 in returns according to the days of the week. The risk-return relationship between the 466 world market portfolio and BET is significant in explaining the causality of these 467 anomalies. Of course, the other important aspect of these anomalies is investors behavior. 468

#### 5. Conclusions

This paper studied the Efficient Market Hypothesis in its weak, semi-strong and470strong form as formulated by Eugene Fama in his 1970 paper. Furthermore, it examined471the calendar anomalies in the case of the Romanian Stock market.472

The findings of this study suggest that a Tuesday effect is present in the Bucharest Stock 473 Exchange market but when considering the seasonality of market returns, the Tuesday 474 coefficient fades until it becomes unsignificant. This leads to the conclusion that the day of 475 the week effect in the Romanian stock market is present due to the influence of the global 476 market, and not due to some intrinsic cause. 477

In both models (1) and (2) we obtained a statistically significant coefficient for 478 Tuesday. This coefficient suggests that in the studied period Tuesday had abnormally 479 high returns in comparison to the other days of the week. This finding tells us that 480 investors should have bought on Monday and Wednesday (the days with a negative 481 coefficient) and sold on Tuesdays for an extra profit. 482

The behavior and exploitation of this effect led investors to buy on Tuesday (because 483 prices were lower) and sell on Thursday or Friday (when prices were higher). This 484 behavior determined in the next period (2013-2023 period of our analysis), a statistically 485 significant raise in Tuesday returns (because enough investors decided to buy, they 486 ramped up prices) and lower returns on Thursday and Friday (when investors decided to 487 sell, and so prices went lower). The study's conclusions supports Murphy's Law which 488 Dimson & Marsh (1998) proposed, and show that the efficiency of the BSE market might 489 be affected in the short term. 490

To further inspect the relationship of the world market portfolio and BET, an ARCH 491 & GARCH model was elaborated to take the time varying component into model 492 estimation. The informational shock of a financial time series data such as stock prices 493 does not rapidly decay like an AR (1) model, and in this instance the time-varying 494 volatility of the series must be considered. Since the initial shock might not become 495 apparent in the data for some time, an autoregressive conditional heteroskedastic model 496 was employed to account for this evolution. 497

The estimation of model (3) determined the Tuesday coefficient to fade until it 498 became unsignificant. This finding suggests that the day of the week effect is present in 499 the BSE market due to the seasonality component of the global market portfolio and not 500 due to some endogenous anomaly. 501

Further research of calendar anomalies in the Romanian stock market can be done by using more indices and not just BET. This analysis can also be extended to other Romanian markets such as the Bond Market and with individual listed companies as well. The seasonality component can be further elaborated on by conducting individual studies of different economic sectors in the same period to see if one is more prone to higher returns that the other according to the time of the year or the current economic cycle. Estimation 507

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of efficiency for the Romanian market can be done by employing R/S analysis as in Mandelbrot 1972, estimating the fractal dimension for short term memory and Hurst exponent for long term memory of the series. 510

One of our study's limitations is that historical BET closing values are not readily 511 available and information about the Bucharest Stock Exchange is not easily accessible. 512 Other limitations regarding the dataset used are in relation to the global market portfolio 513 proxy we elaborated. One could extend this proxy by incorporating indices from stock 514 markets across the world such as the New York Stock Exchange, London Stock Exchange, 515 Tokyo Stock Exchange, Deutsche Börsa; Singapore Stock Exchange, Australian Securities 516 Exchange, and many others. 517

Supplementary Materials:The following supporting information can be downloaded at:518www.mdpi.com/xxx/s1, Figure S1: title; Table S1: title; Video S1: title.519

Author Contributions: For research articles with several authors, a short paragraph specifying their 520 individual contributions must be provided. The following statements should be used 521 "Conceptualization, X.X. and Y.Y.; methodology, X.X.; software, X.X.; validation, X.X., Y.Y. and Z.Z.; 522 formal analysis, X.X.; investigation, X.X.; resources, X.X.; data curation, X.X.; writing-original draft 523 preparation, X.X.; writing-review and editing, X.X.; visualization, X.X.; supervision, X.X.; project 524 administration, X.X.; funding acquisition, Y.Y. All authors have read and agreed to the published 525 version of the manuscript." Please turn to the CRediT taxonomy for the term explanation. 526 Authorship must be limited to those who have contributed substantially to the work reported. 527

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