

## **The Factor Model for Determining the Individual Investment behavior in India.**

**Dr.S.Jayaraj**

*Associate Professor, Department of Accounts and Finance, College of Business and Economics  
Bahirdar University Bahirdar Ethiopia.*

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**Abstract:** *Globalization of financial markets has been increasing the retail investors' community over the past two decades by providing a wide variety of market and investment options. However, it makes much more complex in their investment decisions process. Expected utility theory views the individual investment decision as a tradeoff between immediate consumption and deferred consumption. But individuals do not always prefer according to the classical theory of economics. Recent studies on individual investor behavior have shown that they do not act in a rational manner, rather several factors influence their investment decisions in stockmarket. It will be useful to examine whether some psychological and contextual factors affect individual investor behavior. Based on the previous literature on economics, finance and psychology, individual investors were surveyed to find what and to what extent affects their investment behaviour. It is widely believed that investment behavior of individual investors influence the prices of stocks rarely. Therefore majority of the trading strategies and stock market policies are formulated with a focus only on institutional investors. This paper aims at identifying the factors influencing the retail investor's attitude. In this study principal component analysis is used to find out the determinants of individual investment behavior. The psychological factors that may drive the individual investors are identified by using Principal Component Analysis. The present study provides five major factors that can influence investor behavior in Indian stock market. The findings may be helpful to design relevant investment strategies according to their personal characteristics so as to realize optimum return on their investment.*

**Keywords:** *Behavior Finance, Factor analysis, investors' attitude, riskJEL: GO2.*

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### **I. Introduction.**

According to economic theorists, investors think and behave “rationally” when buying and selling stocks. Generally investors are presumed to use all available information to form “rational expectations” in investment decision making. In reality, driven by greed and fear, investors speculate stocks between unrealistic highs and lows. They are misled by extremes of emotion, subjective thinking and the herd mentality. Indian stock market is considered to be highly volatile, sensitive and reactive to unanticipated shocks and news and it takes no time to impact the market activities. However at the same time, Indian stock market is resilient and recovers soon after shocks. The role and importance of individual investors and their trading behavior in Indian stock market is also very crucial. Unlike institutional investors, individual investors are believed to be less informed, have psychological biases and also thought of as the proverbial noise traders in the stock market. It is believed that trading behavior of individual investors rarely influences the stock prices. With this perception about the individual investors, majority of trading strategies and stock market policies are designed and focused to their institutional counterparts, thereby ignoring the individual investor's interests to some extent. The purpose of this paper is to analyze the determinants of individual investor behavior in Indian stock market. An empirical study is conducted to analyze the investment behavior and decision making style of individual investors.

### **II. Text**

#### **1. Review of Literature**

There are large number of researches in behavioral finance covering the issue of dynamic relationship between individual investment behavior, trading volume, variation in stock prices and volatility and returns. Major studies include Odeon(1999), Choe, Kho and Stultz (1999), Barber and Odeon(2000, 2001, 2005), Grinblatt and Keloparju (2000) Coval, Hirshleifer and Smway(2002), Goetzmann and Kumar (2008), Griffin, Harris and Topalogu (2003), Jackson (2003), Andrade, Change and Seasholes(2005), Barber et al (2005), Barber, Odean and Zhu (2009), Hvidkjaer (2005), Richards (2005) and San (2005). Assessing individual behavior through questionnaire survey is a well adopted approach in behavioral sciences research. A large number of researchers adopted this approach to study the importance of cognitive and other factors on individual behavior. Nagy and Obenberger (1994) examined factors influencing investor behavior. They developed questionnaire incorporating 34 variables. Their findings suggested that classical wealth maximization criteria are important to investors even though they employ diverse criteria when choosing stock for investment.

Researchers have acknowledged the significance of information factor in individual investment behavior. Epstein (1994) examined the demand for social information by individual investors. The results indicate the usefulness of annual reports to corporate shareholders. The result also indicates a strong demand for information about product safety and quality and about company's environmental activities. Further the majority of the stock holders surveyed wanted information about corporate ethics, employee relations and community involvement. Krishnan and Brooker (2002) analyzed factors influencing the decisions of investor who use the recommendation of analyst to arrive at short term investment decision to hold or sell a stock. The results indicate that a strong analyst's report reduces disposition error for gains and also reduces the disposition error for loss. Merikas et al (2003) adopted a modified questionnaire to analyze Greek investor behavior on the Athens Stock Exchange. The results indicated that individuals base their stock purchase decisions on economic criteria combined with other diverse variables. The authors did not rely on a single integrated approach, but rather on many categories of factors. The results also revealed that there is a certain degree of correlation between the factors that behavioral finance theory and previous empirical evidence identify as the influencing factors for the average equity investor, and the individual behavior of active investors in the Athens Stock Exchange (ASE) influenced by the overall trends prevailing at the time of the survey in the ASE. Fisher and Statman (2000) revealed that the sentiment of Wall Street strategists is unrelated to the sentiment of individual investors or that of newsletter writer (another category of investors provided by them), although the sentiment of the individual investors and newsletter writers groups is closely related. They concluded that sentiment can be useful for tactical asset allocation, and that a negative relationship between the sentiment of each of these three groups and future stock returns, and the relationship is strategically significant for Wall Street strategists and individual investors. Malmendier and Shanthikumar (2003) tried to answer the question: Are small investors' naive? They found that large investors generate abnormal volumes of buyer initiated trades after a positive recommendation only if the analyst is unaffiliated. Small traders exert abnormal buy pressure after all positive recommendations, including those of affiliated analysts. Hodge (2003) analyzed investors' perceptions of earnings quality, auditor independence, and the usefulness of audited financial information. He concluded that lower perceptions of earnings quality are associated with greater reliance on a firm's audited financial statements and fundamental analysis of those statements when making investment decisions. Korniotis and Kumar (2009a) predict cognitive ability using a host of demographic variables (e.g., age, education, and social networks). Using the LDB dataset, they show that smarter investors outperform others by about 30 bps per month (or 3.6% annually) both before and after accounting for transaction costs. Smarter investors earn returns net of trading costs that are on par with appropriate benchmark returns; they make good stock picks, but only good enough to cover their trading costs. Other investors underperform appropriate benchmarks by a bit more than 30 basis points per month (or 3.6% annually) after costs, with about half of the shortfall being traced to trading costs and half to bad stock selection.

Several papers have developed theoretical models based on the observation that investors are overconfident (Benos (1998), Caballe and Sakovics (2003), Daniel, Hirshleifer, and Subramanian (1998), Gervais and Odean (2001), Hong, Scheinkman, and Xiong (2006), Kyle and Wang (1997), Odean (1998), Peng and Xiong (2006), Scheinkman and Xiong (2003), and Wang (2001)). Generally, these models assume investors suffer from the miscalibration type of overconfidence. Empirical work has attempted to tease out the type of overconfidence, miscalibration or better-than-average that is linked to excessive trading. Combining survey evidence with trades and positions for 1,345 German investors, Dorn and Huberman (2005) document those investors who think themselves more knowledgeable than average churn their portfolios more. Similarly, Glaser and Weber (2007) use survey evidence and trading records for 215 German investors to document a link between the "better-than-average" type of overconfidence and trading activity.

Closely related to the notion of overconfidence are self-assessments of competence, which are studied by Graham, Harvey, and Huang (2009). They use survey responses from 475 U.S. investors to study the impact of self-assessed competence on trading. Graham et al. document a strong link between self-assessed competence and the propensity to trade. They find weak evidence that this measure of overconfidence is linked to trading activity. The disposition effect has been documented for individual investors in several countries, for some groups of professional investors, and for different types of assets. Shapiro and Venezia (2001) analyze the trading of 4,330 investors with accounts at an Israeli brokerage in 1994. Objectives of the study.

The present study has the following specific objectives.

- To identify the factors which determine individual investor investment behavior.
- To know whether any psychological biases exist among Indian investors.

The present study considers eight behavioral factors suggested by various analysts. These behavioral factors are Representativeness, Overconfidence, Gambler's fallacy, Availability bias, regret aversion and Mental accounting.

In addition to these behavioral factors generated from academic sources the following are also considered in the study. They are contextual factors:

- a. Market share and reputation of the firm.
- b. Accounting and financial information.
- c. Publically available information through various media.
- d. Recommendation by relatives, brokers, friends
- e. Personal financial needs.

Therefore totally 13 factors are included in this study.

## **2. Method and Analysis:**

The study is based on primary data collected through questionnaire. The study adopted the five point Likert scales which is ideal for survey instrument. The respondents were furnished with scenario based questions to which they were asked to mark their response in a range from 1 (strongly disagree) to 5 (strongly agree). The target population for the survey was Indian individual equity investors. Data were collected through a survey of about 300 individual retail investors based in selected cities of Tamilnadu region in India. The study was conducted during the period between October 2012 to March 2013.

**Sampling Size and Procedure:** In order to collect information from retail investors the sampling design has been carefully decided and properly chosen for study. The sample size covered 300 retail investors who were spread through different investment centers in Tamilnadu. The cities where large investors are available were identified as investment centers for this study using purposive sampling method. Chennai, Coimbatore, Trichy, Erode, Tiruppur, Madurai were identified as investment centres. From each identified investment center, five approved stock brokers were chosen and ten investors were contacted with the help of each broker. However, on scrutiny of the filled in questionnaire, it was found that 25 of them had provided incomplete information and hence responses could not be used for further analysis. Thus, the study is based on 275 selected respondents of retail investors. The researcher made alteration in the questions by consulting three academicians and two industry professionals.

## **3. Analysis.**

An analysis of the result of correlation suggests that correlation among variable are not statistically significant enough to indicate any problem among variables themselves. Using the test suggested by Anderson the correlation coefficient that exceeds the value of 0.8 is searched for as there may be potential problem among variables. The coefficient table did not show any such problem. The reliability of the survey instrument was tested with the help of Cronbach's Alpha method. This method allows the researcher to measure the reliability of different categories. It consists of estimates of how much variation in scores of different variables is attributable to chance or random errors. As a general rule, a coefficient greater than or equal to 0.5 is considered acceptable and a good indication of construct reliability (Nunnally, 1976). The overall Cronbach's Alpha for eight categories is 0.756. It shows that these categories for survey instruments are valid and reliable. This suggests that scales used in survey instruments are unidimensional.

## **Univariate Analysis.**

The present study considers eight behavioral factors. These are: Representativeness (Kahneman and Tversky, 1973), Overconfidence (Tversky, 1990; Gervais et al., 2001a, 2001b) Gamblers fallacy (Kahneman and Tversky, 1974), Anchoring bias (Gervais et al., 2001a), Availability bias, Regret aversion, Prospect theory, and Mental accounting

In addition to these behavioral factors generated from academic sources the following contextual factors are also considered.

- a. Market share and reputation of the firm
- b. Accounting and financial information
- c. Publically available information through various media
- d. Recommendation by the relatives, brokers, friends
- e. Personal financial needs.

Therefore thirteen behavioral and contextual factors are included in the study. Investors, even professionals, fall prey to important logical fallacies and psychological failings. Some of them are relatively new; others have been known for decades. These psychological pressures impact the decisions of the people under conditions of uncertainty in a very predictable manner, not only in the marketplace, but in virtually every aspect of our lives. The bottom line is that these powerful forces lead most people to make the same mistakes time and again. Understanding them is their best protection against stampeding with the crowd, and may help the investors to profit from their mistakes instead.

Heuristics is defined as the process by which people reach conclusions, usually by what they find for themselves, from available information. This often led them to develop the thumb rules, which are not always accurate. These heuristics cause investors commit errors in particular situations. A review of the heuristics identified as tested with the help of survey questionnaire is helpful in understanding the individual investment behavior in Indian stock market. A total of five heuristics are considered with seven variables. The influence of each heuristics on the investors who were surveyed is furnished below:

Influence of 'Representative biases.

#### ***A Representative Heuristic***

It is a cognitive bias in which an individual categorizes a situation based on a pattern of previous experiences or beliefs about the scenario. It can be useful when trying to make a quick decision but it can also be limiting because it leads to close-mindedness such as in stereotypes. In the market, this means labeling two companies, or two market environments, as the same when the actual resemblance is superficial. Give people a little information and, click!, they pull out a picture they're familiar with, though it may only remotely represent the current situation. First, it may give too much emphasis to the similarities between events (or samples), but not to the probability that they will occur. Second, representativeness may reduce the importance of variables that are critical in determining the event's probability.

The analysis provides evidences both for the presence and absence of representative bias among Indian investors. The representative bias is captured by variables V1 and V2. A large number of the respondents were found to make investment on the performance of stock in the recent past. Nearly 53.1 % of sample investors disagreed that they follow the performance of stock in the past. The mean is 2.567 (SD=0.857). Others stated that they ignore the recent past performance because they felt that the past performance is not a guarantee for future returns.

#### ***B. Influence of Overconfidence bias.***

The overconfidence bias is the tendency of people to be more confident than is objectively justified by their abilities and characteristics. There is a fine line between confidence and overconfidence. Confidence implies realistically trusting in one's abilities, while overconfidence usually implies an overly optimistic assessment of one's knowledge or control over a situation. In terms of investing, overconfidence can be detrimental to one's stock-picking ability in the long run. This bias is construed in V3. When the surveyed individual investors were asked whether they use their predictive skill which is presumed to have been obtained from their experience, in order to outperform the market, a large number of investors answered in the negative. Nearly 52.7% of the respondents answered in the negative. The mean value of this variable is 2.49 (SD=0.771). The overconfident tendency does not found to be popular among the sample investors.

#### ***C. Influence of Anchoring bias.***

Similar to how a house should be built upon a good, solid foundation, investors' ideas and opinions should also be based on relevant and correct facts in order to be considered valid. However, this is not always so. The concept of anchoring draws on the tendency to attach or "anchor" investors' thoughts to a reference point - even though it may have no logical relevance to the decision at hand. Anchoring can also be a source of frustration in the financial world, as investors base their decisions on irrelevant figures and statistics. This heuristics is found significantly among individual investors. The sample investors were asked whether trading was influenced by recent experiences about price in the market, and whether they use purchase price of stocks as a reference point while making trading decisions. Results indicate that they were likely to be influenced by their recent experiences.. The trading decision of about 55.6% sample investors is well influenced by their experiences in the recent past. On the other hand nearly 50.91% of respondents were suffering from anchoring bias as they used their purchase price as a reference point for their trading decisions.

#### ***d. Gamblers Fallacy.***

A gambler's fallacy is a heuristic in which a person thinks the probability of an outcome has changed, when in reality, it has stayed the same. In the gambler's fallacy, an individual erroneously believes that the onset of a certain random event is less likely to happen following an event or a series of events. This line of thinking is incorrect because past events do not change the probability that certain events will occur in the future. It is not hard to imagine that under certain circumstances, investors or traders can easily fall prey to the gambler's fallacy. For example, some investors believe that they should liquidate a position after it has gone up in a series of subsequent trading sessions because they don't believe that the position is likely to continue going up. Conversely, other investors might hold on to a stock that has fallen in multiple sessions because they view further declines as "improbable". Just because a stock has gone up on six consecutive trading sessions does not mean that it is less likely to go up on during the next session. In the present study this heuristics is reflected by



variable V6. The respondents were asked whether they should be able to anticipate the end of good or poor returns at the stock market. This question was specially referring to the anticipation of market returns. This variable assumed a mean of 2.63 (SD=0.725). The score shown in the univariate statistics indicates these investors are at least likely to anticipate about stock market returns. The frequency statistics also show that 51.64 % were of the view that they could successfully anticipate market returns whether it would be good or poor at the market index. This indicates that the Indian investors are influenced by gambler's fallacy.

***e. Availability Bias.***

The availability heuristic is a mental shortcut that occurs when people make judgments about the probability of events by how easy it is to think of examples. The availability heuristic operates on the notion that, "if you can think of it, it must be important." The availability of consequences associated with an action is positively related to perceptions of the magnitude of the consequences of that action. In other words, the easier it is to recall the consequences of something, the greater we perceive these consequences to be. But the frequencies that events come to mind are usually not accurate reflections of their actual probability in real life. Investors are more likely to act on this readily available information. Growth stock is a very hot story and everyone likes a stock that goes up very fast. Individuals as well as Institutional investors fall prey to this availability heuristics, start believing in the growth story and chase growth stocks. This availability heuristics comes into play while individuals taking trading decisions in the markets. Survey results show that sample investors do not hold it good for their investment decision-making. The above-average-score of this factor shows that they are likely to be swayed away by media reports and join the bandwagon. And frequency statistics also support this finding that about 57.7 per cent of individual investors did not seem to be taking their investment/trading decisions on the basis of the readily available information about their choices. This figure leads us to believe that majority of sample investors are not swayed away by the vividly publicized information about any stock and choose to buy that stock above those with less media attention.

***f. Influence of Mental accounting.***

Mental accounting refers to the tendency for people to separate their money into separate accounts based on a variety of subjective criteria like the source of money and intent for each account. According to this heuristics, individuals assign different functions to each asset group, which has an often irrational and detrimental effect on their investment decisions. The variable V12 captures the influence of mental accounting bias. The sample investors were asked whether they would treat it as overall loss if two out of ten stocks in their investment portfolio are reduced in their investment value significantly. The mean score of this variable is 2.80 (SD=0.603) indicates that individual investors seem to taking the losses separately, not as overall loss on the entire portfolio. Most of the sample investors consider loss on stocks separately rather calculating it as loss of the portfolio. About 71.9 % of sample investors believe that if any stock in their portfolio has depreciated in its value, their loss must be attributed to that asset only and it cannot be adjusted against the value of the entire portfolio. The data from this survey provides strong evidence for mental accounting effect on Indian individual investors. Lack of proper awareness and access to right source of information make them more conservative and hence they get influenced by mental accounting error.

***g. Prospect Theory.***

Kahneman and Tversky (1979) contented that people values gains and losses differently, and as such, will base decisions on the perceived gains rather than perceived losses. According to this theory, losses have more emotional impact than an equivalent amount of gains. During the survey the sample investors were asked whether they decide their buying on the basis of perceived gains rather than perceived losses. The frequency analysis showed nearly 53.4% of the respondents expressed that take only the overall gain but combination of gains and losses. The results indicate that the Indian investors are influenced by Prospect theory.

***h. Testing of "Regret Aversion Bias" influence.***

According prospect theory, losses are three times painful than the pleasure of gain. The reason is people hate regret and losses produce regret. Often, people weigh only negatives in order to protect themselves against future losses and subsequent regret. Hence the potential of regret influence their decision making. This behavior tendency is reflected in the variables V11 and V12. The V11 violate the individual investors' tendency to delay in selling stocks that have decreased in their value. The mean for this variable is 2.93 (with SD=0.712). Most of the investors, nearly 71.3 % of the surveyed nesters showed the tendency of delaying the selling of stocks. They believe the prices to recover and delay the selling of such stocks till the prices bounce back to their reference point. This view is supported by the data on variable V12 (Mean 2.83, SD=0.804). This variable traduces investors' tendency to book profits on the stocks that have increased in value even if the prospect of gaining in near future is significant.

**Multivariate analysis of Indian individual investment behavior.**

The principal component analysis approach suggests that the number of components extracted is equal to number of variables analyzed, necessitating that it is to be decided just how many of these components are truly meaningful and worthy of being retained for rotation and interpretation. In general the first few components will account for meaningful amounts of variance and that later components will tend to account for only trivial variance. The next step of analysis is therefore, to determine how many meaningful components should be retained for interpretation. The following three criteria are generally used for the purpose:

- a. Eigen value-one criterion
- b. the Scree test
- c. the proportion of variance accounted for.

The Eigen value-one criterion, which is also known as Kaiser criterion is one of the most commonly used criterion to retain the number of components for rotation and interpretation. Secondly, with Scree test, the Eigen values associated with each component are plotted in a graph and observed for a 'break' between components with relatively large Eigen values and those with small Eigen values. The components that appear before the break are assumed to be meaningful and retained for interpretation. According to the third criterion, the components that account for a specified proportion of variance in the data set.

On the basis of the criteria mentioned above, the component matrix is formed for further orthogonal rotation using varimax rotation algorithm which is standard rotation method. The multivariate analysis extracts obviously 18 behavioral components, but only six components were judged sufficient to explain the significant data variance and also qualified the above mentioned criteria for solving the number of components to be retained problem. In fact, all the six components so selected seem to explain 61% of total variance and the remaining variance is explained by other variables as given below.

S.No	Components	Eigen Value	% of Variance Explained	Cumulative variance
1	Component1	3.907	21.706	21.706
2	Component2	2.029	11.272	32.977
3	Component3	1.922	10.681	43.658
4.	Component4	1.240	6.890	50.548
5	Component5	1.215	6.748	57.296
6	Component6	1.053	5.849	63.145

Each of the principal components selected for rotation and interpretation are given suitable label based on statement loaded under each component. Rotation maximizes the loadings of each variable on one of the extracted factors while minimizing the loading on all other factors. To interpret and to give a title to each behavioral factor, initial variables were examined carefully along with their respective correlations with concerned factors. Then only common interpretation of the components could be carried on.

***The Behavioral Factors Defined.***

According to the extracted coefficients, those six pertinent behavioral components were named as follows: Conservatism, Diligent and Discreet, Remorse Abhorrence, Cognition, Prudence and Precaution, under confidence. Following are the detailed analysis for each of the six components extracted from the principal component analysis.

***1. Conservatism***

According to the multivariate analysis results, the sample investors seem to be conservative rather than adoptive. This behavioral component accounts for about 21.706% of data variance explained. The rotation sums of squared loadings show that this component accounts for about 21.45% of data variance after orthogonal rotation of component matrix. Hence this is the most influential factor in case of individual investment behavior in the stock market. The underlying variables signify the investor behavior tendency to look into all publically available information through different media. Investors are more amenable to the recommendations and suggestions of the acquaintances. One of the factors which affect them is their personal financial needs. Therefore Indian individual investors seem to be significantly influenced by conservatism psychological bias.

***2. Diligent and Discreet.***

The second component is diligent and discreet. It is related to anchoring and accounts. This component accounts for 11.272% of data variance. After rotation sums of squared loadings it accounts for 10.151% of variation in data. Each variable included under this factor is associated with a different kind of behavioral attitude. The variables underline calculated trading decision. Their decisions are found to be based upon the accounting information of the company. Indian individual investors' trade behavior, it seems, is based on trend

analysis. Their trading decision depends upon recent experiences about prices in the market. This diligent attitude shown by the surveyed investors may be justified by anchoring their decisions on the basis of their recent experience about prices in the market and their inquisitiveness to look into the accounting information.

### **3. Remorse Abhorrence.**

The third is remorse abhorrence. The investors do not want to repent their decisions. The loss is bad for investors and this loss is underlying factor behind regret. Indian investors seem not to be subject to this to this psychological upset. This component accounts for nearly 10.681% of data variance. Even after rotation sums of squared loadings this component accounts for 10.073% of data variance. The investors tend to book profits in stocks that have increased in their value even if there is prospect of higher gains in future. The surveyed investors do not want to regret their hasty decision of selling a stock because the stock may recover.

### **4. Cognition.**

The fourth factor extracted by principal component analysis is cognition. People most often underweight long-term averages. Investors tend to put too much weight on recent experiences. This component accounts for 8.990% of data variance and after the rotation sums of squared loadings it account for 9.404% of data variance. More often than not, an investor will simply recognize an event that is recent performance and conclude, often incorrectly, that this event will be the same. The investors who were surveyed seem to be afflicted with this bias because they base their investment decision on the basis of performance of the stock in recent past.

### **5. Prudence and Precaution.**

The fifth component extracted by the factor analysis is prudence and precaution attitude as the variable included under this component are related to it. This component accounts for more than 6% of data variance .After rotation sums of squared loadings it accounts for 8.710 % of variance in the data. The behavioral attitude of risk aversion and calculated trading decision. The investors' decision depends on the combination of both high returns and losses and they do not seem to rely on overall gains only. This prudent attitude displayed by the surveyed investors may be justified by more than the optimal risk aversion in the stock market.

### **6. Under confidence**

The sixth factor extracted by the factor analysis is under confidence as the variables included under this factor are related to it. This factor accounts for only 5.849 and after rotation sums of squared loadings it accounted for 8.405% of data variance. Infact the Indian individual investors seems to be trusting their predictive skills and their conceived ability to see the end of return in the market. The investors who were surveyed seem to be having undue trust in their forecasting ability and they also tend to base their trading decision on their ability to foresee the stock prices.

## **III. Conclusion.**

The present paper carried out principal component analysis on the data collected from survey of sample individual investors, to extract the factors influencing their investment behavior. Especially, the psychological biases which may drive their trading behavior were identified. This paper makes an attempt to identify the psychological biases which may influence individual investment behavior in Indian stock market. To be more specific it tried to explore the investor trading behavior by directly addressing to investor. To realize this objective the questionnaire technique was used .The questionnaire contained scenario based questions relating to psychological and contextual biases. Cronbach Alpha test was used to measure internal consistency and reliability of the questionnaire. The data were collected from a sample of 300 individual investors spread across strategic investment centers in Tamilnadu. The collected survey were put to univariate and multivariate analysis. The principal component analysis technique was primarily used for multivariate analysis of the data. Statistical tests such as KMO test and Bartlett's Test of Sphericity were used to test the suitability of the techniques. The results of the principal component analysis reveal the six underlying psychological axes that appear driving the individual investor behavior. These six pertinent axes on the basis of underlying variable are named as *conservatism, diligent and discreet, remorse abhorrence, cognition, prudence and precaution, under confidence*. The results reveal that the psychological axes conservatism, diligent and discreet, remorse abhorrence fall in line with the earlier research to some extent. But prudence and under confidence are the contrary behavior axes reported by the multivariate analysis. These psychological components seem to be influencing individual investors' trading behavior in Indian Stock Market.

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**Table.1.Summary Statistics of Survey Respondents.**

Summary Statistics - Demographic Variables (Total 275)				
Age Group		Count	%	Mean
	< 25	13	4.733	36.48
	25-35	75	27.273	
	35-45	108	39.273	
	45-55	58	21.091	
	>55	21	7.636	
Income group( in Indian INR)				1059713
	< 5lakhs	32	11.636	
	5-10lakhs	103	37.455	
	10-15 lakhs	70	25.455	
	15-20 lakhs	54	19.636	
	>20 lakhs	16	5.818	
Investment Group(in INR)				457605
	<1 lakh	32	11.636	
	1-3 lakhs	87	31.633	
	3-5 lakhs	98	35.636	
	5-7 lakhs	46	16.727	
	>7 lakhs	12	4.364	
Education				
	Schooling	16	5.818	
	Non-Graduates	56	20.364	
	Graduates			
	PG	79	28.727	
	Others	36	13.091	
		88	4.364	
Gender				
	Male	238	86.545	
	Female	37	13.455	

**Table.2.Correlation Matrix.**

	V1	V2	V3	V4	V5	V6	V7	V8	V9	V10	V11	V12	V13	V14	V15	V16	V17	V18	
V1	1	0.623329	-0.04586	-0.08645	-0.13128	-0.04419	0.022725	0.007582	-0.21572	0.161457	0.018006	-0.0064	-0.09648	0.144434	-0.25015	0.064809	0.073849	0.012271	
V2		1	-0.03802	0.067579	-0.01097	-0.02138	0.122683	-0.05984	-0.12338	0.11303	0.12251	-0.02602	-0.11338	0.069594	-0.08856	0.081796	0.046465	0.123406	
V3			1	0.212547	0.069438	0.366549	0.072213	-0.04337	0.084534	0.021919	-0.03444	-0.07141	-0.07644	0.14354	0.013714	0.153031	0.036429	0.093009	
V4				1	0.150094	0.1168	0.267326	0.040201	0.061357	0.221716	0.187947	-0.1348	0.106901	0.343626	0.305783	0.327386	0.185644	0.419784	
V5					1	0.104015	0.047987	-0.0038	0.046862	0.092025	0.155039	-0.19223	0.027409	0.211993	0.053823	0.066626	0.094505	0.237637	
V6						1	0.051336	-0.14339	0.084787	0.067722	-0.10017	-0.12756	0.112862	0.146603	-0.02783	0.192624	-0.08312	-0.05206	
V7							1	0.039531	-0.08608	0.218443	0.223323	-0.13644	0.035049	0.396976	0.117165	0.285546	0.400571	0.41472	
V8								1	-0.12275	0.033898	0.425577	0.255974	0.241469	0.086687	0.080591	0.114921	0.17097	0.228299	
V9									1	-0.27333	0.01152	-0.00393	0.027948	-0.08554	0.226072	0.111953	-0.04884	0.062326	
V10										1	0.253571	-0.12906	0.104411	0.511532	-0.08556	0.272449	0.253454	0.311027	
V11											1	0.217239	0.348321	0.300933	0.194096	0.325796	0.364342	0.563818	
V12												1	0.187648	-0.12499	0.077228	-0.13355	-0.11568	0.002295	
V13													1	0.111305	0.067618	0.346207	0.247249	0.290163	
V15														1	0.064008	0.417498	0.358239	0.518379	
V16															1	0.079707	0.079982	0.231806	
V17																1	0.566083	0.505393	
V18																	1	0.559649	
																			1

Determinant = 0.007

**Table .3.Descriptive Statistics.**

Variables	Mean	Standard Deviation	Valid Percentages		Analysis N
			Agree	Do not agree	
V1	2.567	0.857	46.9	53.1	275
V2	2.629	0.824	50.9	49.1	275
V3	2.491	0.771	47.3	52.7	275
V4	2.681	0.943	55.6	44.4	275
V5	2.556	0.797	50.9	49.1	275
V6	2.556	0.824	51.6	48.4	275
V7	2.400	0.904	42.45	57.5	275
V8	2.771	0.731	66.5	33.5	275
V9	2.513	0.686	52.4	47.6	275
V10	2.784	0.835	64.6	35.4	275

V11	2.898	0.799	71.3	28.7	275
V12	2.800	0.837	66.2	33.8	275
V13	2.866	0.924	70.9	29.1	275
V14	2.287	0.989	40.4	59.6	275
V15	2.255	0.616	31.6	68.4	275
V16	2.807	1.037	64.7	35.3	275
V17	2.909	0.821	70.2	29.8	275
V18	2.949	0.836	69.1	30.9	275

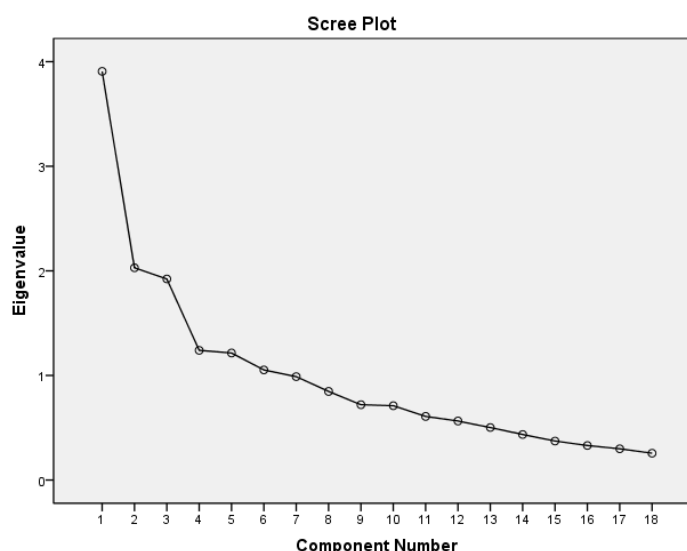
Method: Principal Component Analysis.

**Table.4.Kaiser-Meyer Olkin and Bartlett's test**

KMO and Bartlett's Test		
Kaiser-Meyer-Olkin Measure of Sampling Adequacy.		.721
Bartlett's Test of Sphericity	Approx. Chi-Square	1308.983
	df	153
	Sig.	.000

**Table.5.Total Variance Explained**

	Initial Eigen Values			Extraction Sums of Squared Loadings			Rotation Sums of Squared Loadings		
	Total	% of Variance	Cumulative %	Total	% of Variance	Cumulative %	Total	% of Variance	Cumulative %
1	3.907	21.706	21.706	3.907	21.706	21.706	2.952	16.401	16.401
2	2.029	11.272	32.977	2.029	11.272	32.973	1.827	10.151	26.553
3	1.922	10.681	43.658	1.922	10.681	43.658	1.813	10.073	36.626
4	1.240	6.890	50.548	1.240	6.890	50.548	1.693	9.404	46.030
5	1.215	6.748	57.296	1.215	6.748	57.296	1.568	8.710	54.740
6	1.053	5.849	63.145	1.053	5.849	63.145	1.573	8.405	63.145
7	0.989	5.493	68.638						
8	0.847	4.707	73.345						
9	0.720	4.000	77.345						
10	0.710	3.947	81.292						
11	0.608	3.378	84.670						
12	0.564	3.135	87.805						
13	0.502	2.784	90.593						
14	0.435	2.418	93.012						
15	0.373	2.072	95.083						
16	0.330	1.833	96.916						
17	0.299	1.660	98.576						
18	0.256	1.424	100.00						



**Table.6.Component Matrix.**

Variable	Components					
	1	2	3	4	5	6
V1		-0.777				
V2		-0.661			0.529	
V3			0.517			
V4	0.538					
V5						
V6			0.570			
V7	0.559					
V8			-0.570			
V9		0.500				
V10	0.520					
V11	0.640					
V12			-0.594	0.517		
V13						
V14	0.708					
V15		0.515				
V16	0.713					
V17	0.697					

Extraction Method: Principal Component Analysis

a.6components extracted.

**Table.7.Rotated Component Matrix.**

Variable	Components					
	1	2	3	4	5	6
V1				0.844		
V2				0.890		
V3						0.730
V4		0.645				
V5						
V6						0.836
V7						
V8			0.709			
V9					-0.754	
V10					0.663	
V11			0.612			
V12			0.740			

V13						
V14						
V15		0.607				
V16	0.822					
V17	0.818					
V18	0.668					

a. Extraction Method: Principal Component Analysis.    b. Rotation Method: Varimax with Kaiser Normalization.

**Table.8.Component Transformation Matrix**

Variable	1	2	3	4	5	6
1	0.816	-0.459	0.235	0.088	0.214	0.122
2	-0.002	0.220	0.326	-0.734	-0.551	.050
3	-0.005	0.281	-0.733	-0.161	-0.054	0.595
4	0.048	-0.545	0.358	-0.176	0.289	0.677
5	-0.029	0.032	0.215	0.624	-0.653	0.358
6	-0.576	0.602	0.356	0.084	0.360	0.204

Extraction Method: Principal Component Analysis.

Rotation Method: Varimax with Kaiser Normalization.

**Table Captions:**

**Table1.Summary Statistics of Survey Respondents.**

**Table.2.Correlation Matrix**

**Table.3. Descriptive Statistics**

**Figure.1.Scree diagram**

**Table.4.KMO and Bartlett's test**

**Table.5.Total Variance Explained**

**Table.6Component Matrix**

**Table.7.Rotated Component Matrix**

**Table.8.Component Transformation Matrix.**