A Comparative Analysis of Factor Effecting the Buying Judgement of Smart Phone

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ABSTRACT
Smart phone has various utilisations to various clients as per their necessities. With sensational rise in the usage of smart phone the individuals are considering different factors while purchasing a smart phone. This paper has put endeavor to reveal the fundamental factors which effect clients in picking up of the smart phone. A sample of 512 responses was taken through questionnaire. An organized questionnaire was planned with five point Likert scale was utilized to meeting respondent’s .Factor analysis and descriptive statistical tools were applied to extricate the basic variables influence cell phone acquiring choice. The result shows that the most important factors are physical attributes, apps and sounds while the less importance is given to other factors such as convenience, price which can also vary by age, service and gender. The future scope of this paper lies in the fact that whether age, occupation, gender makes any difference in purchasing decision of smart phone.

1. INTRODUCTION
Smart phones can be portrayed as preparing contraptions which are a mix of remote voice organizations, and programming applications. As they are related with web they can run a couple of virtual universes based organizations, for instance, messages, video-streaming, web based systems administration, geo area et cetera and besides give unprecedented customer encounter [1]. Smart phones today have supplanted different other littler and critical gadgets in our lives. It normally acts and helps an individual like an individual colleague, a savvy guide, plan planner, personal performer et cetera. It runs a progressed portable working framework enabling outsider applications to keep running on the same with an exact touch screen user interface. Accordingly of the alluring looks and capable elements, smart phones have turned into the prime decision for roughly 900 million smart phone clients in India for remaining associated with the outside world. With the happening to advancement there has been a reduction in the cost of contraption and moreover the cost of data which has elevate driven the customer to trade their traditional handsets to front line PDAs with incredible number of segments.

1.1. Status of Smartphone
India has the most astounding rate of smart phone clients all inclusive. After reviews we found that purchasers in India spend a typical of 3 hours on their phone on general schedule, check their phones no under 77 times to 100 times every day. According to [2] Nielsen India study 62% of customers who are later purchasers purchased android phones. In this study, various age groups have been taken.85% respondents fall in the age group from 19-25.
The ascent in use of smart phones by youthful purchasers counting undergraduates has made it fundamental for smart phone producers to comprehend the particular factors, applications, working framework and so on., which this portion search for while settling on their buy choice. Henceforth the target of this review is to comprehend the factors influencing youthful understudy's decision of Smartphone handsets.

The factors influencing youthful understudy's decision of a smart phone are perplexing henceforth it is exceptionally hard to relegate a solitary reason or factor as it varies for various people [3]. Branding assumes an exceptionally pivotal part in a customers' buy design. Better the ubiquity of the brand better is the acknowledgment it accomplishes. In states [4] that associations or organizations have understood that Brand name is one of the most critical resources for organizations. Brands are utilized by purchasers as methods for self-expression as they have a tendency to distinguish themselves with the brands that they buy. According to the discoveries of mark name impacts customer's assessment and in addition their [5] buy choice. According to the review by [6], different channels are being utilized by proprietors of smart phones for correlation of items/ costs. They utilize channels, for example, notices, content messages, QR codes, and connections to data recordings, portable coupons, and applications. As per social impacts [7] concerning an individual can bring about other individuals to change their attitude, conduct, or expectations. So clients might be worried that the smart phone they utilize ought to be enjoyed by their companions or they would get one just to fit in the associate gathering. A review led in [7] Philippines by on factors influencing decision of smart phones reasoned that new innovation elements were more imperative than the size. A review directed by on Nokia Lumia smart phone demonstrates that [8] Connectivity through 3g; Wi-Fi, and so forth are too the factors that are the essential necessities that are expected by people.

A review was directed by on Finnish [9] buyers what's more, their expectation to get smart phones and the factors influencing change of smart phones. According to this think about, there were sure fundamental factors like brand, cost, properties and interface which influence shopper decision of smart phones. As indicated by the review led by, outline of the phones [10], (for example, simplicity, pleasant appearance, striking nature, inflexibility, colorfulness and so on..) are critical factors influencing shopper decision of smart phones. Agreeing to [11] usefulness is the most critical factor influencing shopper decision while stretch that product traits like factors and feel are most vital while picking a cell phone. Nair LR et. al. [12] In this paper, a cloud based framework for continuous focused on publicizing in view of tweet assessment investigation is outlined and actualized utilizing the enormous information handling motor Apache Spark, using its spilling library. Application is intended to advance strategically pitching and give better client bolster [13]. This paper considers the premise prerequisites of sentiment mining to investigate the present procedures used to build up a completely fledged system. Is features the open doors or arrangement and research of such frameworks.

1.2. Conceptual Framework and Formulation of Idea

Author utilizes inferential examination, for example, correlation, analysis of difference and relapse to recognize the significance of brand picture and brand dedication in customer buy goal [14] and furthermore to comprehend the market pattern and buyer inclinations in the business to lift deals. Rahim A et. al. [15] outcome demonstrates item highlights, mark name and social impact having extraordinary relationship between them. It directed Pearson correlation coefficient to known acquiring goal of PDA.Creator utilized UTAUT (Unified hypothesis of acknowledgment and utilization of innovation) to research buy expectation of cell phone. Lu H et. al. [16] clients in acquiring versatile diversion apps.Firstly filled poll review lastly directed quantitative measurable examination. To know shopper conduct towards buying of smart phone creator taken just these five components price, product features, brand image, durability and after deals benefit and connected correlation, multiple relapse and directed dependability test [17]. Result indicates strength and cost is principle factor. Least corresponded variables are after deals administration and brand. Look at the significance of various components influencing buyer thought processes and furthermore [9]. Research principle motivation to change versatile phone. Survey of 196 respondents taken. Correlation grid and bartter trial of sphercity utilized for examining purchaser motives. Result indicates specialized issue is the motivation to change cell phone. Encourage more research led to know how much individuals append with mobile. Appropriate change connected to standardize distribution. Answers from question prepared utilizing thing determination procedure.

Konok V et. al. [18] PCA with varimax pivot led on remaining items. Random assignments of subjects to the trial gatherings were analyzed utilizing one way annova, kruskalwalis and chi-square test. GLMM (Generalize direct blended model) is led to know how partition from the cell phone actuates physiological and behavioral anxiety and consideration predisposition to detachment related boosts [19]. To investigate the elements Affecting youth mark Choice for cell phones Purchase 110 poll were conveyed ,out of which 70 were usable while rest were discarded. Collected information is broke down in view of ANOVA,
correlation and relapse examination by utilizing SPSS. Additionally descriptive, pie table likewise used to break down the reactions. Creator indicates association between buy goal and internet shopping having the most grounded relationship [20]. Pilot concentrate done on 7-point likert scale questionnaire, from which 600 respondent escaped 800 questionnaire. It figured mean, standard deviation to known interceding part of procurement aim.

Smart phones can be recognized as cutting edge phones that offer propelled advances with comparative usefulness to that of a PC. There has been an expansion sought after of smart phone as of late with the advancement of innovation. Thus it is essential to distinguish the fundamental factors from the perspective of a client. This review intends to inspect the relationship of hardware, technology, application, family conclusions and so on related variables that impact when we think to buy new smart phone. These factors are all around considered with the assistance of the reactions gathered and henceforth best outcomes are acquired. In this paper [21] [22] creator utilize choice tree (C5.0), neural system, bolster vector machine (SVM), calculated relapse and k-closest neighborhood (KNN) and common edge detection algorithm were produced and approved to think about various information mining calculation on forecast of heart diseases. In future, we can likewise utilize this innovation for expectation of elements and can contrast technologies and each other. In Table 1 represent how our approach is better than other approaches. In our approach we are using factor analysis and descriptive statistical tools were applied to extricate the basic variables influence cell phone acquiring choice.

Need to decide the incentive for parameter K, if we use KNN techniques. But in our purpose technique, don’t need to determine any value for parameter K. Moreover computation cost is quite high using KNN technique. Separation based learning isn’t clear which kind of separation to utilize and which credits to use to deliver the best results. Shall we utilize all traits or certain properties as it were? In arrangement errands you require a major informational collection so as to make dependable estimations of the likelihood of each class. You can utilize Naïve Bayes characterization calculation with a little informational index yet accuracy and review will keep low. So, it is working well with a small dataset.

| Ref | Title | Methodology | Summary |
|-----|-------|-------------|---------|
| [23] | The Welfare Classification of Indonesian National Civil Servant Using TOPSIS and K-Nearest Neighbor (KNN) | | 1. Public services to citizen must improve 2. Define welfare criteria and classify civil servant data based on welfare measurement by utilizing k-NN Algorithm (or k-NN for short) and TOPSIS. |
| [24] | Classifying rubber breed based on rough set feature selection | | 1. Rubber is the economic crop that is planted widely in almost all regions of Thailand and makes a lot of income for the export of this country. 2. Generate rubber breeds classifier by using KNN technique based on selected set of features of rubber. |

Table 1. Comparison with other approaches (continue)
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| Ref  | Title                                                                 | Methodology                                                                 | Summary                                                                                                                                                                                                 |
|------|----------------------------------------------------------------------|-----------------------------------------------------------------------------|---------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| [25] | Automated web usage data mining and recommendation system using KNN classification method | Developed online real time recommendation expert system                     | 1. Major problem of many online websites is the presentation of many choices to the client at a time. Time consuming tasks 2. Automatic web usage data mining & recommendation system based on current user behavior. |
|      |                                                                      | Data acquisition (Collection of data RSS user data)                          |                                                                                                                                             |
|      |                                                                      | Data cleaning/removal of entries with error/failure status, unsuccessful HTTP status code. |                                                                                                                                             |
|      |                                                                      | Data mart development (smart logical subset of data warehouse)               |                                                                                                                                             |
|      |                                                                      | Transaction identification (distinguish different user so as to analyze user access behavior) |                                                                                                                                             |
|      |                                                                      | Pattern discovery (grouping user based on similarity using KNN algo)        |                                                                                                                                             |
| [26] | A Fuzzy K nearest neighbor algorithm                                  | Fuzzy set is introduced into KNN                                             | 1. Theory of fuzzy set is introduced into KNN to develop fuzzy algorithm. 2. Other methods for assigning fuzzy memberships. |                                                                                                                                             |
|      |                                                                      | Three methods of assigning fuzzy memberships to the labeled samples are proposed and experimental results and comparisons to crisp version are presented. |                                                                                                                                             |
|      |                                                                      | A fuzzy analog of the nearest prototype algorithm is also developed         |                                                                                                                                             |
| [27] | A TOPSIS based method for Gene selection for cancer classification    | Output is fed into 4 different classifiers (kNN, LDT, SVM, NB)              | 1. The goal of this proposed approach is to select most informative subset of features/genes that give better classification accuracy. |
|      |                                                                      | Select most informative gene that gives better classification accuracy       |                                                                                                                                             |
|      |                                                                      | Ms excel 2010 spreadsheet used to collect data regarding 110 hotels for budget category and 138 for premium category. |                                                                                                                                             |
| [28] | Relationship between customer sentiment and online customer ratings for hotels - An empirical analysis | To compensate this, we multiply the standard errors with finite population correction (FPC) 
FPC = \( \frac{(N-n)(N-1)}{N(n-1)} \times 0.5 \) where N is population size and n is sample size | 1. Sentiment analysis of online hotel reviews for explaining customer ratings. 2. Premium and budget segment hotels in goa considered for study. 3. Statistically significant variation in ratings explained by sentiment polarity. 4. Sentiments are less positive for premium hotels than budget hotel in goa. 5. Premium hotel fair better than in terms of staff performance. | Data collection->Preprocessing->Exploratory data analysis->Predictive data analysis->Finding->Removal of punctuation, numbers, stop words->Converting all letters to lowercase. |
2. RESEARCH METHOD

Questionnaire is created that is filled by number of persons of different age-groups. We get 512 total responses [29]. To transfer data from Google form to spss. We need to recode all variables. Recode into same variable will permanently overwrite the original variable. This we can do by going transform -> recode into same variables. Out of which 16 are females and 97 are male members. Recode variable before changing over sort of factor from String to Numeric else reactions will run off. We got responses according to age is given in Table 2. Out of which maximum response came from students whose age come in between 19-25. We can analyze that we got maximum responses from students as shown in Table 3.

Table 2. Responses according to the age

| Frequency | Percent | Valid Percent | Cumulative Percent |
|-----------|---------|---------------|--------------------|
| 35        | 6.8     | 6.8           | 6.8                |
| 437       | 85.0    | 85.0          | 91.8               |
| 22        | 4.3     | 4.3           | 96.1               |
| 8         | 1.6     | 1.6           | 97.7               |
| 6         | 1.2     | 1.2           | 98.8               |
| 6         | 1.2     | 1.2           | 100.0              |
| 514       | 100.0   | 100.0         |                    |

Table 3. Responses according to Occupation

| Frequency | Percent | Valid Percent | Cumulative Percent |
|-----------|---------|---------------|--------------------|
| 21        | 4.1     | 4.1           | 4.1                |
| 484       | 94.2    | 94.2          | 98.2               |
| 4         | .8      | .8            | 99.0               |
| 3         | .6      | .6            | 99.6               |
| 2         | .4      | .4            | 100.0              |
| 514       | 100.0   | 100.0         |                    |

2.1. Scales of Measurement

There are four sizes of estimation in SPSS. Nominal, ordinal, interim and ratio. Nominal factors are utilized to distinguish the individual or question uniquely. For delineation enlistment number of understudies, shirt number of cricker is measures at nominal level. A player with number 30 is not a more prominent measure of anything than a player with number 15. This level of estimation depicts some asked for relationship among the variable's discernments. Accept an understudy scores the most bewildering assessment of 100 in the class. For this circumstance, he would be selected the foremost rank. At that point, another cohort scores the second most astounding evaluation of a 92; she would be allocated the second rank. A third understudies scores and 81 and he would be allotted the third rank, et cetera. The ordinal level of estimation demonstrates a requesting of the estimation. The interval level of estimation gatherings and solicitations the estimations, and in addition discovers that the partitions between each interval on the scale are indistinguishable along the scale from low interim to high interim. In our data gender, age, occupation, monthly income and education variable be considered as nominal variable. Because there is no characteristic request between categories. Put just, one can't state that a specific class is prevalent / superior to another. Other variables like camera, video, Bluetooth, multimedia option, color display, attractive color, model style etc are measured as ordinal because the different classifications can be intelligently masterminded in an important request. In any case, the distinction between the classifications is not "significant". The requesting is adaptable the request can undoubtedly be switched without influencing the translation.

Pilot study was led by consummation of poll by 514 respondents to test the reliability and legitimacy of the review plan. At that point, the information was deciphered utilizing analytical devices including SPSS. Cronbach's alpha is the most generally perceived measure of internal consistency ("unwavering quality"). It is for the most part used when you have diverse Likert request in a review/overview that edge a scale and you wish to choose whether the scale is strong. Analyze->Scale-Reliability analysis.

The nearer the alpha is to 1.00 the more prominent the interior consistency of things in the instrument being evaluated. As shown in Table 4 value of alpha is .866 that shows high consistency between variables. Nunnaly JC [30] offered a dependable guideline of 0.7. As we can see in Table 5 that Cronbach's alpha is 0.879, which demonstrates a high state of interior consistency for our scale with this particular example.
2.2. Factor Analysis

Kaiser-Meyer-Olkin (KMO) Test is a measure of how suited your information is for Factor Analysis. The estimation is a measure of the degree of contrast among elements that might be standard change. The lower the degree, the more suited your data is to Factor Analysis. KMO returns values in the vicinity of 0 and 1. KMO values in the vicinity of 0.8 and 1 demonstrate the examining is sufficient. KMO values near zero implies that there are substantial fractional relationships contrasted with the whole of connections. As such, there are far reaching connections which are a huge issue for component examination. KMO score .852 shows satisfactory for testing as shown in Table 6.

2.3. Total Variance Explained

Eigen values of their connection grid are appeared in the left area "Introductory eigenvalues" as shown in Table 7. These eigen values relate to the differences of Principal segments (i.e. PCA was performed), not of components. Descriptive word "introductory" signifies "at the start purpose of the examination" and does not infer that there must be some "last" eigen values. The (default in SPSS) Kaiser run "eigenvalues>1" was utilized to choose what number of elements to concentrate, along these lines, 8 components will come. Extraction of them was done by Principal axis method and matrix of loading obtained. In Varimax rotation we exclude java (0.3), Complexity of Operating System (0.369) and reliability (0.321) as they are not strong factors.

The quantity of columns in this table (Extraction aggregate of squared loadings) relates to the quantity of elements held. In this illustration, we asked for that eight variables be held, so there are eight columns, one for each held element. The qualities in this board of the table are ascertained in an indistinguishable route from the qualities in the left board, with the exception that here the qualities depend on the normal difference. The qualities in this board of the table will dependably be lower than the qualities in the left board of the table, since they depend on the normal change, which is constantly littler than the aggregate difference. The Extraction Sums of Squared Loadings is indistinguishable to the Initial Eigen esteems with the exception of elements that have eigenvalues under 1 are not appeared. This implies the initial eight variables together record for 57.443% of the aggregate fluctuation.
As shown in above Figure 1. Scree plot is valuable for deciding what number of variables to retain. From the eighth component onward, you can see that line is level implying that each progressive elements is representing littler and littler measures of the aggregate fluctuation. The below table shows the quantity of elements those influence clients to pick portable mobile brands. From the revolution strategy the accompanying eight components Table 8 might be gotten. Next step is to move to questionnaire and read all items belonging to one factor and name the factor or identification of variables.

![Scree Plot](image)

**Table 8. Rotated Factor matrix**

| Component | 1     | 2     | 3     | 4     | 5     | 6     | 7     | 8     |
|-----------|-------|-------|-------|-------|-------|-------|-------|-------|
| CPU       | 0.799 |       |       |       |       |       |       |       |
| Operating System | 0.7   |       |       |       |       |       |       |       |
| Internal  | 0.626 |       |       |       |       |       |       |       |
| Card Slot | 0.623 |       |       |       |       |       |       |       |
| Browser   | 0.788 |       |       |       |       |       |       |       |
| Messaging | 0.768 |       |       |       |       |       |       |       |
| Alert types | 0.646 |       |       |       |       |       |       |       |
| Loudspeaker | 0.641 |       |       |       |       |       |       |       |
| Sensors   | 0.453 |       |       |       |       |       |       |       |
| Friends&Colleagues recommendation | 0.797 |       |       |       |       |       |       |       |
| Neighbor recommendation | 0.793 |       |       |       |       |       |       |       |
| Family members opinion | 0.673 |       |       |       |       |       |       |       |
| Domestic product | 0.607 |       |       |       |       |       |       |       |
| KeypadHindio English | 0.437 |       |       |       |       |       |       |       |
| Multimedia option |       | 0.686 |       |       |       |       |       |       |
| Color display |       | 0.685 |       |       |       |       |       |       |
| Bluetooth |       | 0.663 |       |       |       |       |       |       |
| Model Style |       |       | 0.802 |       |       |       |       |       |
| Brand Value |       |       | 0.644 |       |       |       |       |       |
| Attractive color |       |       |       | 0.586 |       |       |       |       |
| Weight |       |       |       |       | 0.817 |       |       |       |
| Size |       |       |       |       | 0.773 |       |       |       |
| Camera |       |       |       |       |       | 0.689 |       |       |
| Video |       |       |       |       |       | 0.597 |       |       |
| Battery Backup |       |       |       |       |       | 0.593 |       |       |
| Charging hour |       |       |       |       |       | 0.418 |       |       |
| Product price |       |       |       |       |       |       | 0.653 |       |
| Special offers |       |       |       |       |       |       | 0.6   |       |
| DualSimOption |       |       |       |       |       |       |       | 0.575 |
3. RESULTS AND ANALYSIS

Despite the fact that many works with respect to opinion investigation on PDA some time recently, this work uses continuous advanced mobile phone supposition examination for constant DATASET gathered from LPU understudies, which was not endeavored some time recently. These eight factors affect customer purchase decision of mobile phone. Distinguished components are recorded underneath in Table 9.

a. **Physical attributes**: It is the most critical variable. It can clarify 22.179 percent of aggregate difference in client choices of portable obtaining. Physical properties incorporate all the physical qualities of cell phone like CPU, internal slot, card slot and others.

b. **Apps and sounds**: It is the second component which is able to clarify 8.682 percent of aggregate change. The second component incorporate messaging, alert types and sound which impact client purchasing choice of mobile phone.

c. **Companions Colleagues and family assessment**: This is third most basic element. It can illuminate 6.22 percent of total change. A few respondents take the suggestions from their partners and partners before securing mobile phones.

d. **Features and technology**: It is the fourth element that clarifies 5.554 percent of aggregate variance. There are numerous respondents who by and large take the elements and innovation bolster as imperative issues in purchasing cell phone.

e. **Appearance**: This is the fifth element which clarifies 3.955 percent of aggregate variance. It demonstrates that clients likewise settle on their acquiring choice in view of looks and style.

f. **Size and weight**: It is the 6th component that clarifies 3.849 percent of aggregate change. There are numerous respondents who by and large take the size and weight as critical issues in purchasing cell phone.

g. **Equipment properties**: The seventh element can clarify 3.766 percent of aggregate change in client choices of portable buying. Equipment properties incorporate all the equipment attributes of cell phone like camera, video, battery reinforcement and charging hour.

h. **Pricing**: It is the eighth element which is competent to clarify 3.239 percent of aggregate difference. This eighth element of valuing incorporates all the cost related variables that the clients consider before purchasing. Most extreme information part is gathered from students, it is stun impact to known cost does not have extraordinary effect in buy choice of cell phone.

| Table 9. Identification of Factors |
|-----------------------------------|
| Factor No. | Name of dimension | Item no | Variables | Factor loading |
|------------|-------------------|--------|-----------|----------------|
| F1         | Physical Attributes | 1      | CPU       | 0.799          |
|            |                   | 2      | Operating System | 0.7            |
|            |                   | 3      | Internal | 0.626          |
|            |                   | 4      | Card Slot | 0.623          |
|            |                   | 5      | Browser  | 0.788          |
| F2         | Apps and Sounds   | 6      | Messaging | 0.768          |
|            |                   | 7      | Alert types | 0.646         |
|            |                   | 8      | Loudspeaker | 0.641        |
| F3         | Companions Colleagues and family assessment | 9 | Friends and Colleagues recommendation | 0.797 |
|            |                   | 10     | Neighbor recommendation | 0.793 |
|            |                   | 11     | Family members opinion | 0.673 |
|            |                   | 12     | Domestic product | 0.607 |
| F4         | Features and Technology | 13 | Multimedia option | 0.686 |
|            |                   | 14     | Color display | 0.685 |
|            |                   | 15     | Bluetooth | 0.663          |
|            |                   | 16     | Model Style | 0.802          |
| F5         | Appearance        | 17     | Brand Value | 0.644          |
|            |                   | 18     | Attractive color | 0.586          |
| F6         | Size and Weight   | 19     | Weight | 0.817          |
|            |                   | 20     | Size | 0.773          |
|            |                   | 21     | Camera | 0.689          |
| F7         | Equipment properties | 22 | Video | 0.597 |
|            |                   | 23     | Battery Backup | 0.593 |
|            |                   | 24     | Charging hour | 0.418 |
|            |                   | 25     | Product price | 0.653 |
| F8         | Pricing           | 26     | Special offers | 0.6 |
|            |                   | 27     | Dual Sim Option | 0.575 |
4. CONCLUSION

The focus of the work was to understand the factors effecting the purchase decision of smart phone users. In order to understand the same, a total number of 512 responses were taken from different people having different age groups. On the basis of questionnaire, we find out that physical attribute, apps and sounds are the most crucial factor in purchasing a smart phone. We always say that price is always a main factor in choosing a right smart phone. But we found that price has less effect on purchasing. As a rivalry in the smart phone industry gets stiffer, particularly with the capability of new competitors into the smart phone, it is essential for firms to comprehend the market pattern and shoppers' inclinations into their business. Further this can help the companies in remaking their technique with a specific end goal to push forward of the opposition bend in the market. This review will likewise give a few thoughts to firms in understanding the market patterns and in building up the proper estimating methodology to lift deals. There are other critical factors such as friends, colleagues and family opinions which effect the buying decision. People are more mindful about technology. So, kind of innovation upheld by smart phone is additionally imperative variable. Thus, organizations ought to approach valuing systems with alert, and watch the vital necessities through useful statistical surveying before applying them. In future, we can look whether age, service and gender has any effect on buying the smart phone.

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