

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

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ABSTRACT

Smart phone has various utilizations 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.

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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 understudies 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 bartler trial of sphericity 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.

Table 1. Comparison with other approaches (continue)

Ref	Title	Methodology	Summary
[23]	The Welfare Classification of Indonesian National Civil Servant Using TOPSIS and K-Nearest Neighbor (KNN)	<p>Ranking criteria by using TOPSIS</p> <p>↓</p> <p>Define welfare criteria association</p> <p>↓</p> <p>Classifying dataset by using KNN</p>	<p>1. Public services to citizen must improve</p> <p>2. Define welfare criteria and classify civil servant data based on welfare measurement by utilizing k-NN</p> <p>Algorithm (or k-NN for short) and TOPSIS.</p>
[24]	Classifying rubber breed based on rough set feature selection	<p>Collect the rubber data</p> <p>↓</p> <p>Find the minimal feature set with rough set-based feature selection: (feature dependency, and the reduct are calculated from such decision table.)</p> <p>↓</p> <p>Generate a rubber breed classifier (uses 10 cross fold validation to iterative partition the dataset and KNN algo for test and training)</p> <p>↓</p> <p>Classify a breed for given rubber seedling</p> <p>↓</p> <p>(Evaluate the effectiveness of breed classifier using Chi-Square, Spearman's Correlation, and Pearson's Correlation)</p>	<p>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.</p> <p>2. Generate rubber breeds classifier by using KNN technique based on selected set of features of rubber.</p>

Table 1. Comparison with other approaches (continue)

Ref	Title	Methodology	Summary
[25] MATLAB s/w	Automated web usage data mining and recommendation system using KNN classification method	<p>Developed online real time recommendation expert system</p> <p>↓</p> <p>Data acquisition (Collection of data RSS user dbase)</p> <p>↓</p> <p>Data cleansing (removal of entries with error ,failure status ,unsuccessful HTTP status code....)</p> <p>↓</p> <p>Data mart development (dsmart logical subset of data warehouse)</p> <p>↓</p> <p>Transaction identification (distinguish different user so as to analyze user access behavior)</p> <p>↓</p> <p>Pattern discovery (grouping user based on similarity using KNN algo)</p>	<p>1. Major problem of many online websites is the presentation of many choices to the client at a time. Time consuming tasks</p> <p>2 Automatic web usage data mining & recommendation system based on current user behavior.</p>
[26]	A Fuzzy K nearest neighbor algorithm	<p>Fuzzy set is introduced into KNN</p> <p>↓</p> <p>Three methods of assigning fuzzy memberships to the labeled samples are proposed, and experimental results and comparisons to crisp version are presented.</p> <p>↓</p> <p>A fuzzy analog of the nearest prototype algorithm is also developed</p>	<p>1. Theory of fuzzy set is introduced into KNN to develop fuzzy algorithm.</p>
[27]	A TOPSIS based method for Gene selection for cancer classification	<p>Technique combines TOPSIS (Techniques for Order Preference by Similarity to an Ideal Solution) and F-score method to select subset of relevant genes</p> <p>↓</p> <p>Output is fed into 4 different classifiers (KNN,DT,SVM,NB)</p> <p>↓</p> <p>Select most informative gene that gives better classification accuracy</p>	<p>1. The goal of this proposed approach is to select most informative subset of features/genes that give better classification accuracy.</p>
[28]	Relationship between customer sentiment and online customer ratings for hotels - An empirical analysis	<p>Ms excel 2010 spreadsheet used to collect data regarding 110 hotels for budget category and 138 for premium category.</p> <p>↓</p> <p>Hotels belong to sampling frame placed in 1st column</p> <p>↓</p> <p>Adjacent column was generated by rand() function</p> <p>↓</p> <p>hotels in first 20 rows for sampling is selected</p> <p>↓</p> <p>SRS requires that sample size less than 10% of population size</p> <p>↓</p> <p>To compensate this ,we multiply the standard errors with finite population correction(FPC) $FPC = \frac{(N-n)}{N} * 0.5$ where N is population size and n is sample size</p> <p>↓</p> <p>Implemented naive bayes method for classification in R</p>	<p>1. Sentiment analysis of online hotel reviews for explaining customer ratings.</p> <p>2. Premium and budget segment hotels in goa considered for study.</p> <p>3. Statistically significant variation in ratings explained by sentiment polarity.</p> <p>4. Sentiments are less positive for premium hotels that budget hotel in goa.</p> <p>5. Premium hotel fair better than in terms of staff performance.</p> <p>Data collection->Preprocessing->Exploratory data analysis->Predictive data analysis>Finding>Removal of punctuation, numbers, stop words>Converting all letters to lowercase.</p>

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. Nunnally 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.

Table 4. Reliability Analysis

Cronbach's Alpha	Cronbach's Alpha Based on Standardized Items		N of Items
.866	.880		32

Table 5. Reliability Statistics

Cronbach's Alpha	N of Items
.879	32

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.

Table 6. KMO and Bartlett's Test

Kaiser-Meyer-Olkin Measure of Sampling Adequacy.		.852
Bartlett's Test of Sphericity	Approx. Chi-Square	5224.678
	df	496
	Sig.	.000

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.

Table 7. Total Variance explained

Component	Initial Eigenvalues			Extraction Sums of squared loadings			Rotations sums of squared loadings		
	Total	% of variance	Cumulative %	Total	% of variance	Cumulative %	Total	% of variance	Cumulative %
1	7.097	22.179	22.179	7.097	22.179	22.179	3.090	9.657	9.657
2	2.778	8.682	30.861	2.778	8.682	30.861	2.999	9.372	19.029
3	1.990	6.220	37.081	1.990	6.220	37.081	2.667	8.336	27.365
4	1.777	5.554	42.635	1.777	5.554	42.635	2.526	7.895	35.260
5	1.266	3.955	46.590	1.266	3.955	46.590	2.086	6.518	41.778
6	1.232	3.849	50.438	1.232	3.849	50.438	1.723	5.383	47.161
7	1.205	3.766	54.205	1.205	3.766	54.205	1.658	5.182	42.343
8	1.036	3.239	57.443	1.036	3.239	57.443	1.632	5.100	57.443

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 of 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.

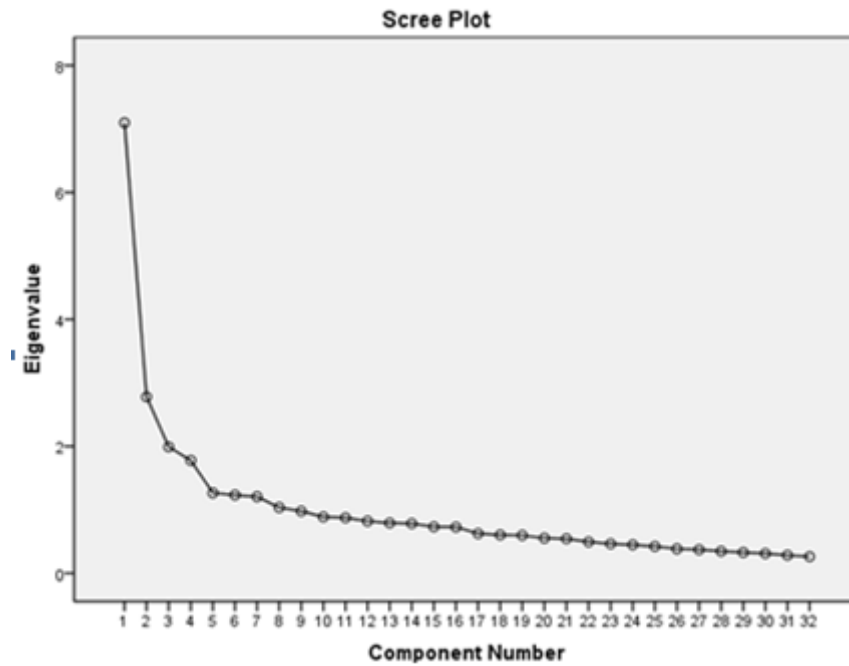


Figure 1. Scree plot

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						
FriendsandColleague recommendation			0.797					
Neighbor recommendation			0.793					
Familymembersopinion			0.673					
Domestic product			0.607					
KeypadHindior 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	FriendsandColleague recommendation	0.797
		10	Neighbor recommendation	0.793
		11	Familymembersopinion	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
F5	Appearance	16	Model Style	0.802
		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	DualSimOption	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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REFERENCES

- 1]. Kenney M, Pon B., "Structuring the smartphone industry: Is the mobile Internet OS platform the key?" *J Ind Compet Trade*, 2011; 11(3), 61-239.
- 2]. UtrechtUniversity. Chapter 7 – Factor Analysis – SPSS. Analysis [Internet]. 2012;1–5. Available from: <http://www.cs.uu.nl/docs/vakken/arm/SPSS/spss7.pdf>
- 3]. Min M, Hong L, Ai J, Wah P., "Conceptual Paper : Factors Affecting the Demand of Smartphone among Young Adult ," Factors Affect Demand Smartphone among Youth Adult, 2012, 2(2), 9–44.
- 4]. Cornelis PCM. Effects of co-branding in the theme park industry: a preliminary study. *Int J Contemp Hosp Manag* [Internet]. 2010, 22(6), 96–775 Available from: <http://www.emeraldinsight.com/doi/10.1108/09596111011063089>
- 5]. Khasawneh, K. and ABIH., "The effect of familiar brand names on consumer behaviour: A Jordanian perspective," *Int Res J Financ Econ [Internet]*, 2010, 43(1), 34–57, Available from: https://www.researchgate.net/publication/289792582_The_effect_of_familiar_brand_names_on_consumer_behaviour_A_Jordanian_perspective
- 6]. Moosylvania, "The shopping experience in a smartphone world," [Internet], 2013, Available from: <http://www.brickmeetsclick.com/shopping-in-a-smartphone-world>
- 7]. Lisa Rashotte, "Social Influence," *Blackwell Encycl Sociol*. 1966, 9 (1985), 9–4426.
- 8]. For C, Systems G. a Study on Consumer Behavior and Opportunities for Nokia Smart-Phones in India. 2(1), 2014.
- 9]. Karjaluoto H, Karvonen J, Kesti M, Koivumäki T, Manninen M, Pakola J, et al., "Factors Affecting Consumer Choice of Mobile Phones: Two Studies from Finland," *J Euromarketing*, 14(3), 59–82, 2005.
- 10]. Yun MH, Han SH, Hong SW, Kim J., "Incorporating user satisfaction into the look-and-feel of mobile phone design," *Ergonomics* [Internet], 46(13–14), 1423–40, 2003. Available from: <http://www.tandfonline.com/doi/abs/10.1080/00140130310001610919>
- 11]. Işıklar G, Büyüközkan G. Using a multi-criteria decision making approach to evaluate mobile phone alternatives. *Comput Stand Interfaces* [Internet], 29(2), 265–74, 2007. Available from: <http://linkinghub.elsevier.com/retrieve/pii/S0920548906000663>
- 12]. Nair LR, Shetty SD, Shetty SD, "Streaming big data analysis for real-time sentiment based targeted advertising," *Int J Electr Comput Eng*, 7(1), 7– 402, 2017.
- 13]. Dhokrat A, Khillare S, Mahender CN., "Review on Opinion Mining for Fully Fledged System," *Indones J Electr Eng Informatics* [Internet], 4(2), 8–141, 2016. Available from: <http://section.iaesonline.com/index.php/IJEEI/article/view/225>
- 14]. Eze UC, Lee CH., "Conceptualizing factors that influence consumers' mobile phone purchase," *ICIMTR 2012 - 2012 Int Conf Innov Manag Technol Res*, 43–139, 2012.
- 15]. Rahim A, Safin SZ, Kheng LK, Abas N, Ali SM., "Factors Influencing Purchasing Intention of Smartphone among University Students," *Procedia Econ Financ [Internet]*, Elsevier B.V., 37(16), 53–245, 2016. Available from: [http://dx.doi.org/10.1016/S2212-5671\(16\)30121-6](http://dx.doi.org/10.1016/S2212-5671(16)30121-6)
- 16]. Lu H, Lin P, Lin Y.. "A Study of the Factors Affecting the Purchase Intention on Mobile Game Apps," 7(4), 44–239, 2016.

- [17]. Jamal A., The International Journal Of Business & Management, Consumer Behavior towards Buying of Smart Phone A Case of Jinnah University for Women, 4(1) 35–226, 2016.
- [18]. Konok V, Pogány Á, Miklósi Á., "Mobile attachment: Separation from the mobile induces physiological and behavioural stress and attentional bias to separation-related stimuli," *Comput Human Behav [Internet]*, 71, 39–228, 2017. Available from: <http://linkinghub.elsevier.com/retrieve/pii/S0747563217300754>
- [19]. Shahzad khan(city university of science and I-T, peshawar, pakistan), sobia rohi(city university of science and I-T, Peshawar P. Investigating The Factors Affecting Youth Brand Choice For Mobile Phones Purchase – A Study Of Private Universities, 8(2), 84–369, 2013.
- [20]. Lim YJ, Osman A, Salahuddin SN, Romle AR, Abdullah S. Factors Influencing Online Shopping Behavior: The Mediating Role of Purchase Intention. *Procedia Econ Financ [Internet]*. Elsevier B.V., 35(October 2015),10–401, 2016. Available from: [http://dx.doi.org/10.1016/S2212-5671\(16\)00050-2](http://dx.doi.org/10.1016/S2212-5671(16)00050-2)
- [21]. Abdar M, Kalthori SRN, Sutikno T, Much I, Subroto I, Arji G. Comparing Performance of Data Mining Algorithms in Prediction Heart Diseases. 5(6), 76–1569, 2015.
- [22]. Reddy RVK, Raju KP, Kumar MJ., "Comparative Analysis of Common Edge Detection Algorithms using Pre-processing Technique," 7(5), 80–2574, 2017.
- [23]. Sari WP., "The Welfare Classification of Indonesian National Civil Servant Using TOPSIS and K-Nearest Neighbour (KNN)," 2016.
- [24]. Classifying rubber breed based on rough set feature selection. June, 2016.
- [25]. Adeniyi DA, Wei Z, Yongquan Y., "Automated web usage data mining and recommendation system using K-Nearest Neighbor (KNN) classification method," *Appl Comput Informatics [Internet]*. King Saud University, 12(1), 90–108, 2016. Available from: <http://www.sciencedirect.com/science/article/pii/S221083271400026X>
- [26]. Keller JM, Gray MR., "Ieee transactions on systems, man, and cybernetics," vol. smc-15, no. 4, july/august 1985, (4), 5–580, 1985.
- [27]. Fattah IMA, Khedr WI, Sallam KM., "A TOPSIS based Method for Gene Selection for Cancer Classification," *Int J Comput Appl*, 67(17), 39–44, 2013.
- [28]. Geetha M, Singha P, Sinha S., "Relationship between customer sentiment and online customer ratings for hotels - An empirical analysis," *Tour Manag [Internet]*. Elsevier Ltd; 2017;61:43–54. Available from: <http://dx.doi.org/10.1016/j.tourman.12.022.2016>.
- [29]. Huang J, Sun H, Guo P, Zhao M, Niu K., "Fine-Grained Sentimental Tendency Analysis Based on Chinese Online Commentary of Mobile Phone A," Segment words Segment sentence Algorithm o / Sentiment Analysis Preliminary sentiment analysis a) Feature extraction, 23–319, 2016.
- [30]. Nunnally JC. Psychometry theory. In: Psychometric theory. 2nd ed. McGraw-Hill, Inc. New York, NY, USA ©1986; 1978. p. 701.

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