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## Customer Segmentation Using Machine Learning

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**ABSTRACT:** The emergence of many competitors and entrepreneurs has caused a lot of tension among competing businesses to find new buyers and keep the old ones. As a result of the predecessor, the need for exceptional customer service becomes appropriate regardless of the size of the business. Furthermore, the ability of any business to understand the needs of each of its customers will provide greater customer support in providing targeted customer services and developing customized customer service plans. This understanding is possible through structured customer service. Each segment has customers who share the same market features . Big data ideas and machine learning have promoted greater acceptance of automated customer segmentation approaches in favor of traditional market analytics that often do not work when the customer base is very large. In this paper, the k-means clustering algorithm is used for this purpose.

KEYWORDS: ML, Kmeans Clustering, VS code.

#### I. INTRODUCTION

Over the years, increased competition among businesses and the availability of largescale historical data has resulted in widespread use of data mining techniques to find critical and strategic information that is hidden in organizations' information. Data mining is the process of extracting logical information from and so on. The importance of customer segmentation includes, inter alia, the ability of a business to customize market plans that would be appropriate for each segment of its customers; Support for business decisions based on risky environments such as credit relationships with its customers; Identify products related to individual components and how to manage demand and supply power; Interdependence and interaction between consumers, between products, or between customers and products are revealed, which the business may not be aware of; The ability to predict customer declines, and which customers are likely to have problems and raise other market research questions and provide clues to find solutions. Buried in a database of integrated data proved to be effective for detecting subtle but subtle patterns or relationships. This mode of learning is classified under supervised learning. Integration algorithms include the K-Means algorithm, K-nearest algorithm, sorting map (SOM), and more. These algorithms, without prior knowledge of the data, are able to identify groups in them by repeatedly comparing input patterns, as long as static aptitude in training examples is achieved based on subject matter or process.

#### **II. EXISTING WORK**

Customer Data stored in database is of less use as we cannot sort-it-out or make any meaningful conclusion about customer needs. It does not helps the business to understand the market situation and leads to decrease in revenue generated and consumer dissatisfaction. The existing system is unable to predict the current situation of market which in turn we lose competitive advantage.

#### **III. LITERATURE SURVEY**

E-commerce transactions are no longer a new thing. Many people shop with e- commerce and many companies use ecommerce to promote and to sell their products.



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Because of that, overloading information appears on the customers' side. Overloading information occurs when customers get too much information about product then feel confused. Personalization will become a solution tooverloading problem. In marketing, personalization technique can be used to getpotential customers in a case to boost sales. The potential customer is obtained from customer segmentation or market segmentation. This paper will reviewcustomer segmentation using data, methods and process from a customersegmentation research. The data for customer segmentation were divided into internal data and external data. Customer profile and purchase history were treated as the internal data while server log, cookies, and survey data were as the external data. These data can be processed usingone of several methods. Over the years, as there is very strong competition in the business world, theorganizations have to enhance their profits and business by satisfying the demands of their customers and attract new customers according to their needs. Theidentification of customers and satisfying the demands of each customer is a verycomplex and tedious task. This is because customers may be different according totheir demands, tastes, preferences and so on. Instead of "one-size-fits-all" approach, customer segmentation clusters the customers into groups sharing thesame properties or behavioral characteristics. According to customersegmentation is a strategy of dividing the market into homogeneous groups. Thedata used in customer segmentation technique that divides the customers intogroups depends on various factors like, data geographical conditions, economicconditions, demo-graphical conditions as well as behavioral patterns. Thecustomer segmentation technique allows the business to make better use of theirmarketing budgets, gain a competitive edge over their rival companies, demonstrating the better knowledge of the needs of the customer. It also helps an organization in, increasing their marketing efficiency, determining new marketopportunities, making better brand strategy, identifying customers retention. Marketing techniques are entirely based on mutual consumer-retailer relationship. One way to increase profits is to determine customer requirements, through communication with the consumers. Communicating with consumers on personallevel is practically not an easy task and without building communication, marketing disasters are inevitable. To tackle this problem retailers cancommunicate through the data generated by consumers. Retailers can group their consumers according their habits and later on develop business strategiesaccording to it. Customer segmentation is a manner to enhance communique with the consumer, to realize the desires of the consumer hobby in order that suitable communique may be built. Also, strategies of Customer Segmentation may becategorized into Simple technique, RFM technique, Target technique, andUnsupervised technique. In the aggressive marketplace of e-commerce, the hassleof figuring out capacity consumer is gaining increasingly more attention. Thispaper proposes the solution to identify potential customers using RFM analysistool and k-means algorithm. One of the important steps for customer segmentationis clustering. A MATLAB implementation of k-Means clustering set of rules forconsumer segmentation primarily based totally on records accrued from megaenterprise retail outfit, has a purity of 0.95 indicating 95% accurate segmentation of the customers. Over the years, the commercial world is becoming more competitive, as suchorganizations have to satisfy the needs and wants of their customers, attract newcustomers, and hence enhance their businesses . The task of identifying andsatisfying the needs and wants of each customer in a business is a very complextask. This is because customers may be different in their needs, wants, demography, geography, tastes and preferences, behaviours and so on. As such, it is a wrong practice to treat all the customers equally in business. This challengehas motivated the adoption of the idea of customer segmentation or marketsegmentation, in which the customers are subdivided into smaller groups orsegments wherein members of each segment show similar market behaviours orcharacteristics. According to customer segmentation is a strategy of dividing themarket into homogenous groups posits that -the purpose of segmentation is the concentration of marketing energy and force on subdivision (or market segment) to gain a competitive advantage within the segment. It's analogous to the militaryprinciple of concentration of force to overwhelm energy. | Customer or Marketsegmentation includes geographic segmentation, demographic segmentation, media segmentation, price segmentation, psychographic or lifestyle segmentation, distribution segmentation and time segmentation. The zeitgeist of modern era is innovation, where everyone is embroiled intocompetition to be better than others. Today's business run on the basis of suchinnovation having ability to enthral the customers with the products, but with such a large raft of products leave the customers confounded, what to buy and what tonot and also the companies are nonplussed about what section of customers totarget to sell their products. This is where machine learning comes into play, various algorithms are applied for unravelling the hidden patterns in the dataforbetter decision making for the future. This elude concept of which segment totarget is made unequivocal by applying segmentation. The process of segmentingthe customers with similar behaviour into the same segment and with different patterns into different segments is called customer segmentation. In this paper, 3different clustering algorithms (k-Means, Agglomerative, and Meanshift) are been implemented to segment the customers and finally compare the results of clusters betained from the algorithms. A python program has been developed and theprogram is been trained by applying standard scaler onto a dataset having

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twofeatures of 200 training sample taken from local retail shop. Both the features arethe mean of the amount of shopping by customers and average of the customer'svisit into the shop annually. By applying clustering, 5 segments of cluster havebeen formed labelled as Careless, Careful, Standard,Target and Sensiblecustomers. However, two new clusters emerged on applying mean shift clusteringlabelled as High buyers and frequent visitors and High buyers and occasional visitors.

#### **IV. METHODOLOGY**

Data Collection :This is a data arrangement stage. The component generally serves to refine all information things at a standard rate to work on the execution of clustering algorithms and customer classification is done.

Group Analysis : Group analysis is an incorporation or unification, way to deal with purchasers dependent on their likeness. There are 2 primary sorts of unmitigated gathering examination in market strategy: hierarchical group analysis, and characterization. Meanwhile, we will examine how to characterize gatherings, called k-means.

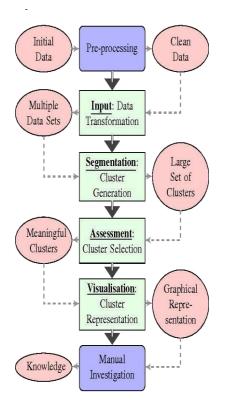


Fig. 3.1 Model Architecture

#### K-means clustering

K-means clustering algorithm is one of the grouping calculations dependent on division. It's anything but a heuristic iterative cycle to re-partition information articles and re-update group focuses.

The essential thought of the calculation is: assume a set with component objects and the quantity of bunches to be generated.

In the first cycle, an example component is arbitrarily chosen as the underlying bunch centre , and the distance between other example components and the middle point is broke down the groups are individually isolated by the distance.

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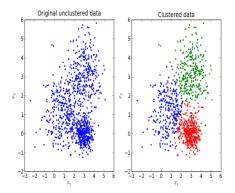
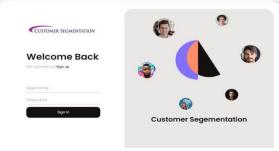


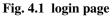
Fig. 3.2 K-means clustering

#### Data and Sources of Data

	A	8	C	D	E	F	G	н	1.	1	K	1	M	N	0	P	Q	R	5	т	U	v	w	×
Row	ID	Order ID	Order Date	Ship Date	Ship Mode	Customer ID	Customer Na	Segment	City	State	Country	Postal Code	Market	Region	Product ID C	ategory	Sub-Categor	Product Nam	sales	Quantity	Discount	Profit	Shipping Cost	Order Priorit
	32298	CA-2012-124	31-07-2012	31-07-2013	2 Same Day	RH-19495	Rick Hansen	Consumer	New York Cit	New York	United States			East	TEC-AC-1000 T	echnology	Accessories	Plantronics C	2309.65		7 1	762.1845	933.57	Critical
	26341	IN-2013-778	05-02-2013	07-02-201	Second Clas	s JR-16210	Justin Ritter	Corporate	Wollongong	New South V	Australia		APAC	Oceania	FUR-CH-100CF	urniture	Chairs	Novimex Exe	3709.395		9 0.:	-288.765	923.63	Critical
	25330	IN-2013-7124	17-10-2013	18-10-201	B First Class	CR-12730	Craig Reiter	Consumer	Brisbane	Queensland	Australia		APAC	Oceania	TEC-PH-1000 T	echnology	Phones	Nokia Smart	5175.171		9 0.:	919.971	915.49	Medium
	13524	E5-2013-157	28-01-2013	30-01-2013	B First Class	KM-16375	Katherine Mu	Home Office	Berlin	Berlin	Germany		EU	Central	TEC-PH-1000 T	echnology	Phones	Motorols Sm	2892.51		5 0.:	-96.54	910.16	Medium
	47221	\$G-2013-432	05-11-2013	06-11-201	3 Same Day	RH-9495	<b>Rick Hansen</b>	Consumer	Dakar	Dakar	Senegal		Africa	Africa	TEC-SHA-100 Te	echnology	Copiers	Sharp Wireler	2832.96		8 1	311.52	903.04	Critical
	22732	IN-2013-4230	28-06-2013	01-07-201	B Second Clas	s JM-15655	Jim Mitchum	Corporate	Sydney	New South V	Australia		APAC	Oceania	TEC-PH-1000 T	echnology	Phones	Samsung Sma	2862.675		5 0.1	763.275	897.35	Critical
	30570	IN-2011-818	07-11-2011	09-11-201	First Class	T5-21340	Toby Swindel	Consumer	Porirua	Wellington	New Zealand		APAC	Oceania	FUR-CH-100CF	urniture	Chairs	Novimex Exe	1822.08		4 1	564.84	894,77	Critical
	31192	IN-2012-8630	14-04-2012	18-04-2013	2 Standard Cl	ai MB-18085	Mick Brown	Consumer	Hamilton	Walkato	New Zealand		APAC	Oceania	FUR-TA-1000 Fi	urniture	Tables	Chromcraft C	5244.84		6 1	996.48	878.38	High
	40155	CA-2014-135	14-10-2014	21-10-2014	4 Standard Cl	ad JW-15220	Jane Waco	Corporate	Sacramento	California	United States	95823	US	West	OFF-BI-1000EO	ffice Suppli	Binders	Fellowes PB5	5083.96		5 0.3	1906.485	867.69	Low
	40936	CA-2012-116	28-01-2012	31-01-2012	Second Clas	s JH-15985	Joseph Holt	Consumer	Concord	North Carolin	United States	28027	US	South	FUR-TA-1000 F	urniture	Tables	Chromcraft B	4297.644	10	13 0.4	-1862.3124	865.74	Critical
	34577	CA-2011-102	05-04-2011	09-04-201	Second Clas	s GM-14695	Greg Maxwel	Corporate	Alexandria	Virginia	United States	22304	US	South	OFF-SU-1000 O	tfice Suppli	Supplies	Martin Yale C	4164.05		5 0	83.281	846.54	High
	28879	ID-2012-2840	19-04-2012	22-04-2012	2 First Class	AJ-10780	Anthony Jaco	Corporate	Kabul	Kabul	Afghanistan		APAC	Central Asia	FUR-TA-1000 Fi	urniture	Tables	Bevis Confere	4626.15		5 1	647.55	835.57	High
	45794	SA-2011-1834	27-12-2011	29-12-201	Second Clas	a MM-7260	Magdelene N	Consumer	Jizan	Jizan	Saudi Arabia		EMEA	EMEA	TEC-CIS-100CT	echnology	Phones	Cisco Smart F	2616.96		4 1	1151.4	832.41	Critical
	4132	MX-2012-130	13-11-2012	13-11-2012	2 Same Day	VF-21715	Vicky Freyma	Home Office	Toledo	Parana	Brazil		LATAM	South	FUR-CH-100CF	urniture	Chairs	Harbour Crea	2221.8		7 1	622.02	810.25	Critical
	27704	IN-2013-7395	06-06-2013	08-05-201	B Second Clas	s PF-19120	Peter Fuller	Consumer	Mudanjiang	Heilongiang	China		APAC	North Asia	OFF-AP-1000 O	ffice Suppli	Appliances	KitchenAid M	3701.52	1	12 1	1036.08	804.54	Critical
	13779	E5-2014-5091	31-07-2014	03-08-201	Second Clas	s 89-11185	Ben Peterma	Corporate	Paris	Ile-de-France	France		EU	Central	OFF-AP-1000 O	ffice Suppli	Appliances	Breville Refrig	1869.588		4 0.:	186.948	801.66	Critical
	36178	CA-2014-143	03-11-2014	06-11-201	Second Clas	s TB-21175	Thomas Bola	Corporate	Henderson	Kentucky	United States	42420	US .	South	TEC-AC-1000 To	echnology	Accessories	Logitech diNo	2249.91		9 1	517.4793	780.70	Critical

#### **V. RESULTS & DISCUSSION**





This snapshot shows the welcome page. If a user has not registered earlier has to register before logging in. Once the user is registered, he/she can login to the page with credentials such as username and password.

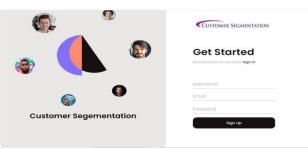


Fig. 4.2 Registration page

This snapshot shows there user registration page for new users. User has to provide details such as username, email id and password for registration.

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CUSTOMER SEGMENTATION	Home	Analysis	Log Out
CUSTOMER SEGMENTATION FOR GROWTH		2	6 20
Manage Analysis		{	
The more segmentation you do, the more groups will be formed and your sample size will increase in response to that.			
			- D

Fig. 4.3 Home page

This snapshot shows the dashboard of our website. Dashboard consists of options including Home, Analysis and Log Out.



This snapshot shows customer analysis bar graph where segmentation of category based on county and the profit is selected. Here, categories and sub-categories are displayed. User can view the respective analysis graph by selecting categories.

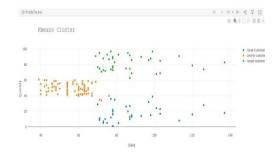


Fig. 4.5 Analysis 2

This figure shows K-means cluster graph. Here, usual customer, priority customer and target customers are shown on graph with respect to sales and quantity of product.

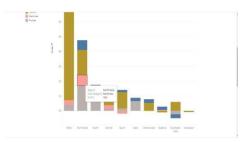


Fig. 4.6 Analysis 3

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This snapshot shows bar graph representing the profit in various regions.

#### Fig. 4.7 Analysis 4

This snapshot shows profit analysis graph. Graph is generated with respect to month of order date, profit and shipping cost.

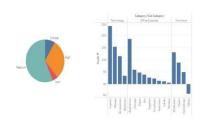


Fig. 4.8 Product categories and shipment

This snapshot shows pie chart to present product categories and their shipment priorities and bar graph to show profit in multiple categories like technology, office supplies and furniture.

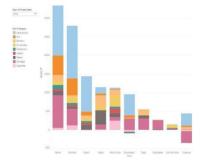


Fig. 4.9 Bar graph of profit of all categories

This snapshot shows bar graph representing the profit of all categories in various regions and the year of order date is 2012.

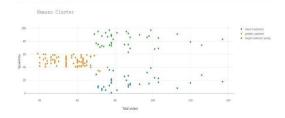


Fig. 5.0 K-means cluster on quality of products

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This snapshot shows K-means cluster graph for quality of products with respect to total orders.

#### VI. CONCLUSION

We used the K-means algorithm to segment our customer in various clusters having similar similarity. Client segmentation in shopping malls is achievable despite the fact that this form of machine learning application is highly useful in the market, a manager can concentrate all of his or her attention on each cluster that has been discovered and meet all of their requirements. Mall managers must be able to understand what customers require and, more importantly, how to meet those needs. analyze their purchasing habits, and establish frequent encounters with customers that make them feel comfortable in order to satisfy their demands.

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