- pargriment - 2 4 To 9 dentify the important node in the network by determining the rank of each node pregent in the melwork. - Step-1: Corpeter the network with 4 modes [A,B, C.D) and It has 6 discerted lanks b/w the nodes * Node A lanke to Mode B * Node B links to node C * Node c Janks to node D * Hode D tonte to node A * Alode A drake to node C lants to mode A * Node 93 8 tep 8: - Arrange she rank ir valuer for each mode ice, Ra, Rb, Rc, Rd * Mode A has a outbound Imbérée, B&C * Node B has 2 outbound donks lie, Agc * Hode c has a outbound lonks (cc, D, * Hode D has a outbound long loc, A Hence, we can wrote 4 equations: Ro = 0.5 * Rb + Rd > 0

"Kb= 0.5 * Ra -> 3 Rc = 0.5 x Ra + 0.5 x Rb -> 3 Rd = Rc -> 1 Henu, there are a outbounds lanks from mode A and T node B, Both will be sharing 9to half of the Influence of respectively 2° egn O gegn O gegn O Step-3: - Solving mathematically now refrequent the equation? on the form of matrix Ra 0 0.50 0 1. Rb 0.50 0 .0 0 Rc 0.50 0.50 0 Rd 0 Step-42- Samplification of the problem of the using the gnotal rank values & then Compute new rank values bell they &tabelize no of nodes Insteal Value Idiration=1 Ra=0.5* Rb+Rd=0.375 0.250 Ra Rb2 0.5x 0.250= 0.125 0,250 Rb Rc = 0.5 + 0.250 = 0.250 0.250 Rc Rd 2 RL 2 0. 250. 0,250 Rd

Scanned with CamScanne

	I		IV	I	VI
Ra Rb Rc Rd	0.8185	0.84875 0.15625 0.25 0.25	0. 328125 0.171875 0. 25 0.25	0.2559375 0.164062 0.25 0.25	0.35205125 50.167968 0.25 0.25
2			VIII		
Ra Rb RC Rd	0.333	984375	0.333		3trongest node 18 Ra
ale	To 9 dents	Hy the ex	nportant node	ode m the present ?	network by
		AX DX	(C)	E	
	3tcb -1:	- Conside	or the netwo	oork weth	5 nodest
3	C. D. E. * Node * Node * Node * Node * Node * Node	Et hase A de	8 derected	node Tode	The nodes ?
				Scanned u	uth Lambcanne

Holde B looke to node A Holde B looke to node E Holde C looke to node C				
Help All Appenn the rank R' values for each node &. Ha Rb Rc, Rd, Re H Node A has 2 outbound longs are BEC.				
Whole B has 2 outbound lonks rec, B&C Whole C has 1 outbound lonks rec, A, C&E H Node B has 1 outbound lonks rec, A H Node B has 1 outbound lonks rec, A H Node B has 1 outbound lonks rec, C				
Hend, we can with 5 equations Re = 0.34 Rb + Rd > 0 Re = 0.34 Ra -> 2				
Re = 0, 54Rb -> (5) Re = 0, 34Rb -> (5)				
Both 32 dolving mathematically,				
Re 0.5 0.3 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0				
(Re) 0 000				

anne s	Ala balaze.	/n =	moder no	0.2	Valuetall
rarable Ra Re Rd	0.2	Value	0.54 Ra+	0.3×2612 0.3×2612	2001 2001 2001 20036.
Re	0.5		-		×
Ra RA RA RA RA	5 0.399 6 3 0.116 2 0.184 36 0.22	0.0348	0.2216	0.21995	0.2200

पाा	IX	I	X	XU
Ra= 0.2667 Rb= 0.1291 Rc= 0.2124 Rd= 0.2304 Rd= 0.0467	0.2691 0.1334 0.2188 0.2124 0.0384	0.2524 0.1346 0.2133 0.2188	0.1262	0.2512 0.1292 0.2079 0.2066 0.0579

XIII	XIV	XV	XVI	XVII	XVIII
	0.2456 0.1227 0.1992 0.2023		0-2360 0-1196 0-1932 0-1973 0-0368	0.1932	

XIX	XX	XXI
Ra = 0.2257	0-2222	0.2186
Rbz 0.1143	0.1129	0. 1111
Rc 20:1847	0.1882	0.1793
Rd 2 0.1879	0.1847	0-1822
Rez 0.0350	0.0343	0.0339
	and the second	

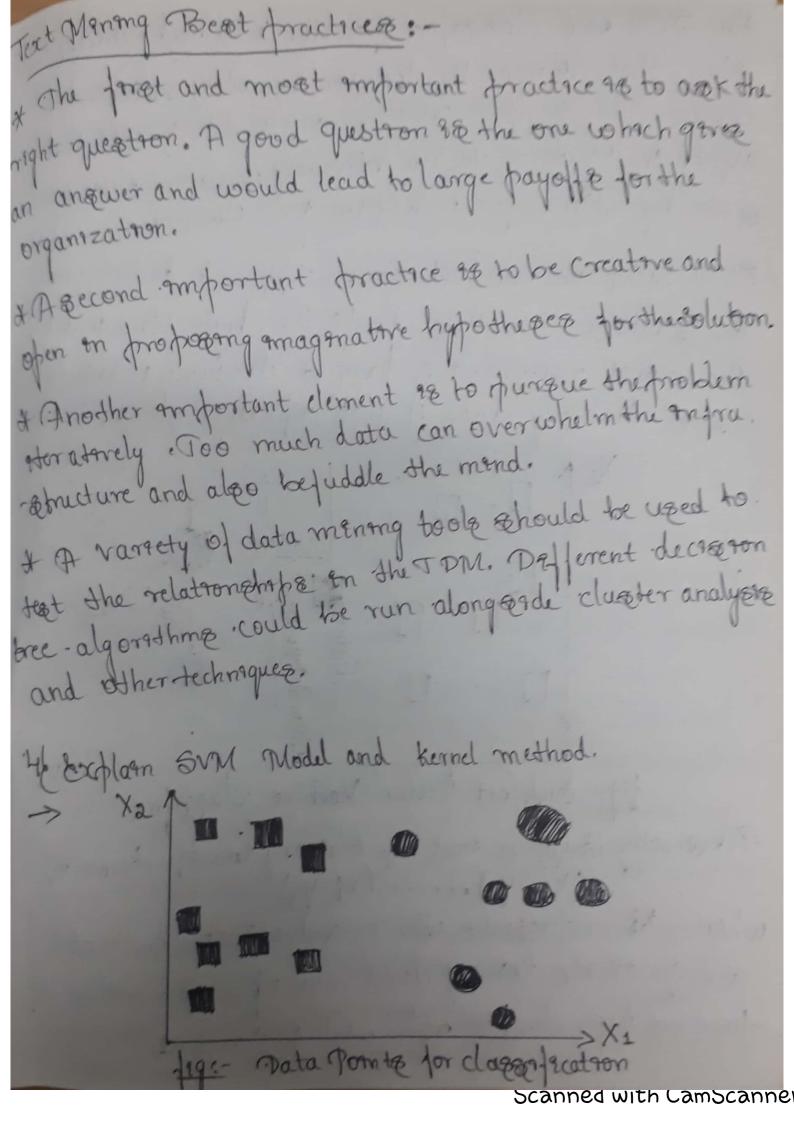
node se Ra

30 Define Text mining ? Brief out the applications and explains and explain the beat fractices of text mining.

Frest mining so the art and ocsence of drocovering Knowledge, another and patterns from an organized.

Collection of Sectual databases. Text Maning applacations: Marketing: The vorce of the customer can be captured on the native and raw format and then analyzed for customer spreferences and Complaints. at Social personas are a clustering techniques to develop cuatomer segments of order cat. Consumer enfut from the fortal medra fources, such as reviews, blogs, and tweets, contain numerous leading indicators that can be used towards anticipating and predicting consumer behaviour. bl A' hætening platform' es a text mining application, that In real time, guthers social media, blogs and other textual feedback and felters Out the chatter to extract true consumer spentements. of the customer call center conversation and records can be analyzed for patterne of customer complaints. * Bugmes operation: Many appects of bustness function -nong can be accurately guarged from analyzing text.
af Social network analyzing and text moneng can be applied to Emarle, blogs, social media and other data to maquire the emotronal Estates and the mood of employee & Studying people of emotional inventors and wing first

Prychology can belf en obtarning superior investment returns. * Ligal: - In legal applications, lawyers and paraligate con more capity gearch cape histories and laws for relevant documents en a particular cape to improve their chances of winning. aftert mining qualque embedded en e-diacovery. platforme that help on manamazing rack in the processe of charing legally mandated documents ble Case hostories, testimonies, and client meeting notes can reveal addetronal enformation ; ouch as morbidates In healthcare & stuatrong that can help better fredset hagh- coat injurier and prevent coatre. * Governance and Politica: Government can be overturned based on a tweet originating from a self-immolating fruitvender en Jungara. af Social network analyses and text mining of large-scale Eoceal medra data can be used for measuring the emotional states and the mood of constituent populations. ble In geopolatical occurrity, antomet chatter can be processed for real-time and formation and to connect the dots. Co In academace, regearch streams could be mota- analyzed for underlying research trends.



An SVM rea clarger ser function in a high-dimensional Eface that defines the decision boundary between two classes. The Support rectors one the data formts that of the hyperplane; for each of the two classes. SVM taken the wideret so treet affroach to democrate the 2 classes and thus finds the hyperplane that has the widest margen, i.e. langest distance to the nearest braining data ipoints of either class. Winder Wingin jegt Suftfort Vector Machine classofrer Abotractly, Suppose that the training data of n fromto re. where, 12 represente the of-value vector for front? and ye are stre tomany clare value of 1 00 - 1. Thuz There are suo classes represented as 1" and -1

paruming that the data 95 mided brearly expanded M. X+ p=0. where, we normal vector to the hyperplane. The hard marging can be defined by the following W. X+b= 1 and w. X+b=-1 the wealth of the hard margen on (2/11/1). The SVM algorithm finds the weights vector (w) for the features rouch that there is widest margin brothe à Slote. The Kernel methods. * The heart of an 6VM algorithm 92 the Kernel mother! * Most kernel algorithms are based on optimization on a convex space and are statistically well-founded. X. Kernel Stands for the core of the germ on a frest. * Kernel methode oferate using what & called the 'Kernel trick'. The Kernel trick mushing completing and working with the inner products of only theredirant parre of data on the feature stace, they do not need to Compute all the data on high - dominational feature of hace. * Kernel methods achieve these by learning from motance. They do not apply some standard Computational doger to all the Jeanures of each white

of putelly people of the map united structure. (Meneng Web Content - Web Structure - Meb Usinge - Structure -Web Content -129: Neb monting-Structure. Of Web Content Mining: A weberte to dissigned in the form of pages with a distinct URLEUniversal Presource Locator). A large website may contain theusands elfage These pages are managed using specialized so beare system Called Content Management Systems. by Web Structure maning: - The web works through a System of hyperlanks using the hypertext protocol (http) There are 2 basic strategic models for successful webate - Hube and Authorstone. * Hube: - These are pages with a large number of Interesting lanks. They serve as hub of a gather my tomit + Authorities: - Ultimately, people would gratitable wante · pages that · provide to most Complete and a withoutatine Information on a farticular subject.

ce Neb Vægge minnings- De a uper cheke anywhere on a web-page @ application, the action of recorded by many entitives on many locations. The browger at the chent machine will record the click and the web percentrovid - my the content would also make a record of the frages Bonied and upor activity on those pages. Mars Data Gadher and Web Logor for Patterna prepare Data -Websett Web chake - Upage Patterny To for Analyzie Vaglore, 3 Streams - Vagage profiles - Identy users Upers, - Identify vigite - Web spage astoniers profiles -Mep offens -Summarge activity -zatron 199: Web Væge Mening Architecture. analysis we, analyzing web activity for fatterns of sequence of clacks and the location and duration of vagate on websites * Textual Information accepted. On the pages retrieved by uzere could be analyzed uzing text mining techniques. at Deferentiate between SNA and TDA