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in the state of th		Participle	Conjuction	Adverb - Europeen Languege	Basis of n	Ponous	Yeab	NOW)		Tagging & Content Analysis.	How to input the content & issues faced while doing POS	Host also are statistical in nature with occuracy ~ 97%	V ta	or word - Noun!	coll of mords - santings coll of book at word	Pos Tagging [Part of Speech] - small part	NLP:-	Other languages by the other by the control of the other control of the	tweety	we wher something from a colly or	SNA can also be based on cont	_ (1 hecture	O sich an any print of time (depends how me

He'll be booking at Rule-based approach.	
2) -> we've HMM	
1.) - Not very papular ( we've very good MI models)	
discumbiguate words accordinaly.	
in will tell it to be a o	
Ex. A student	6
s band:	
decided by set of relles (disembiguate)  No leaving bearing is done by human being	
Rule based: People have read articles & Jabetted the words manually & classify accordingly. They've made their own rules.	*
Now, we	
Objective of this : to understand work, we need to the know its	Objectin
PAGE NO.:	

	any adjectus on work we don't have
	(Ale:
	challenges behind building Rule based approach
	d
	English word classes
	/
	Open Closed
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1 good	2 water,
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							Bu Rused.												*	JT	4	م		ريع	1	9	
Lymn R R. mark	Jointon 2 acoust the verb	The rest for that	dinating	<u></u>	dossify accordingly)	(In state based - it'll work at provious mage of word &	-> Militaria me see down, book poer.	mean	both words (	Ex. rule out - eliminate , algo has to look at	1 Avegan	to Brains & words + maticle &	vent particle	Ex turned down reject	n combi	- Proposition always occur before nours	Net	one, two Arst, second	- up, down, on, off, in, out (of and	ubs - can, may, should	<b>L</b>	Pronouns - he, she, who, I	a, an the, thus, that	Determines - Articles determine	_	Closed Classes	DATE: / /

			0		25			, h	* *	-	1	1		L	
Seben Spete	So, they give 14% ambiguity in surtences.	7	Algo is more effecient when it works with lex no of tags.	from 87 - tag Brown Corpus (1960-70), during Browns	45-tag come penn Treesant tagset (1993) → more popular	Topmar teg is:	jud set of tags for POS Tags I to m	marker: please, thankyon	s in net	interviention: Oh, ah un ala	Ausillary verbs be , do , ean , have , had one	no, ha	Personal Possessium who pronouns	Prometure	SAGE NO:

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Rule based approach ;	
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