9.2	Discuss challenges at the MLP and how to avercome them
	Matural language processing faces several challenges due to the complexity and number of thuman languages Here are some major challenges and potential ways to overcome them. Ambiguity: Ambiguity in a language can lead to confusion one way to tackle this is by using context based approaches. Machine learning models that earlies aumander and
	Machine learning models that eonsider sumording word or sentence can often disumbiguate the intended meaning
b)	Moriability and Intormality: - Language is incredibily diverse with slong colleiguialisms and veration across cultures. Increporting more extension and diverse training dates can be help models better understand and generate different language style.
9	Lack of context: Understanding context is crucial for accorde comprehension. Models like transformer based architecture leverge attention mechanism to copture context over larger sequence of the aiding in our coming challenge:

93

AN

Takenization is a fund amental preprocessing step

m natural language processing (MLP) that involves

breaking down at a text into individual words or

unito called tokens. These tokens senier as the building

blocks for the various MLP task, such a language

modeling text classification and machine translation.

Takenization is important for many reasons.

a) Text understanding :-

Tokenization divides text into mongable units allowing machine to undestand the structure of settlence paragraphs and documents.

b) Feature extraction -

In this task like text classification each is taken con be trated as a feature republing model to learns pattern and relationships with in the text.

c) Language modelling:

Takenization aid in the building language model by breaking textinto smaller units for predicting the next word or generating coherent text.

Code

import te

text = "Tokenization is good"

takens = te. findall (r'/b/w+/b',text)

print (tokens).

0-4		
3		
	Stemming	Lemmatization
		Lt. III anzane
	Stemming is the faster herauce	Lemmatization, es lower
		as composed to stemming
120		but it know the critical
	gren sentence.	of the word before
		processing.
	This a rule - based :	Itis a dictionary -based
	appraich	It is a dictionary - based approach.
		des per Verdensellie de la
	Accuracy is less	Accuracy is more
	chilin (1 1)	tor example
	For example studies = 'studi'	for example studies = study'.
THE LOT	Porter Stemmer algorithm:	The spring of the
	-71	in malatorial
	the parter stemmer algorith	m iswidely used
	The porter stemmer algorithes stemming algorithm developed	by Martin Parter in 1980.
	7+40 law a set of heuristic	rules to transform
	words in to their stems. The a	genthm consist of
	several phases, each tagether	tegerting specific
	suffixer. The phrases aim to	progressively remove
	Common suffixe simple word	to their root forms.

DATE: Explain Edit distance algorithm 9.5 The edit distance algorithm means the similarly hetween two strings by calculating the minimum numbers of single characters edib required to transform one string into the An The edib can be insertion, deletions or substitutions. Itis widely invovious applications such as spell checking, DMA sequence analyon and plagiarism detection Alagnthm Auration edit Distance (str 1, str2): m= length of str 1 n = length of str2 Create adp array of the (mx1) x(n+1) for i from o to m: de [i][o] = i for j from o ton: dp [a][j]=j for i from I to m: for j from I ton: if str1[:-1] = str2[;-1] delisted = deli-13 [j-1]. dp [i][j]= min (dp [i-1]+1, dp[i][j-1]+1, dp[i-1][j-1]+1). tetum de [m][n].

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Forexample:cat and hat edit distance is 1 subsitute the 'c' with 'h'. Use of Amite automata and Amite state transducers Ans 1) Takenization. FST can he used for segmenting text into tokens
for example in the english you can used on FST
to split a setence into individual words or a sub word. 2) Port of speech tagging: FST can help assign parts of speech to words masentene By constructing and for that recognites word sequence and assigns Pos tags you can develop Pas tagging system. 3) Morophological analysis

FSTs are extensively used in morphological analysis

to break down words in to their constituent morophemes. 4) Spell checking and correction: Theon he used to do spell checking and correction system.

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ASSE	DATE:
9.7	Importance of Morphological analysis.
-	
1)	Language Understanding
	understanding the structure ofwards is fundamented
	tounder stand language.
	Shring an amend and continued
2)	MLP tasks:
	The integral to various MLP tasks, including machine translation, text sumonization information retrival and
0	translation, text sumonization information tetrival and
	Sentment analysis.
3)	Resource-Score e Longuage
	In languages with rich marphological complexity morphological analysis vital for a coura te processing
	marphological oraysis viras 10. accorde a final son
41	Storch engine
)	for each search engines morphological analysis enables searching for variations of words.
	searching for variations of words.
0	
5)	Sentiment and yous:
	In sentiment analysis, morphological analysis helps
	identitying negation or words form that change
	the polony of a sentence.
	A-alia tians
3-36 F.J.	Information Retrincal
1	Machine translation
3	Sentiment analysis
	Speech recognition.

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	DATE:
0.0	N
9.8	N-grams.
An	Ngram are contingous sequences of M-Herns words, letter or symbol from a given sample of text or speech
	TI A A A A A A A A A A A A A A A A A A A
	They are a fundamental concept in the HLP M-gram are used to analyze and model the
	telationship hetween elements ma sequence
	They are classified as
508	
1)	Unigrams (1-grams):
	There are sing le item in a sequence latter individual
	word.
	Committee of the commit
2)	Bigram (2-gram):
	These consist at pair at adjacent tem ma sequence
	otwords.
3	Trigrams (3-grams):
	Trigram are composed of the consective item.
4)	M-grams (H>3):
	N-gram con he ony sequence of the Mitems.
	Example:
	This is an group otwords.
	Unigrams: - 'This', 'is', an' group', of', 'words'. Bigrams: - 'This is', (an group', lot words'. Trigrams: - 'This is an', 'group of words'.
	Bigrams: This is', (an group', of words'.
	Trigrams: - This is an , group of words.
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	DATE:
9.9	Penn Trebank
N	the property to the party that the said that it was to
An	Penn Treebonk Pos tagset is a widely used tagging
	Scheme for English that assign toos to word based on
	their orannatical toles in a sentence.
	Penn Treebonk Pos tagset is a widely used tagging Scheme For English that assign tags to word based on their grammatical toles in a sentence. Following are the same of the tags:
	Jen Control of
1)	Co (coordinating conjuction):
7	connect words praises or clauses
	Commes Comment of Comment of the Com
9)	colordinal number)
-)	Represent number like 'one', two',
	TO COME THAT WE STATE OF THE ST
2 }	DT (Determiner)
3)	modifie the noun and specifices their references
	modified the this come?
	such as the this', some'.
41	(77) aditative
4)	Describes noun like, happy', 'blue', 'tall'.
	Describe from the property
-1	The colour singular
5)	HM (Moun, singular) Lepresent singular or mass noun like cat'
	teptstat singulor or has her
-1	000 1,001
- 6)	RB (adverb) modifies verb, adjective, or other adverb.
	modumo yero, anjecting in a second
-1	T. (1)
7)	10 (to)
	Port of the infinitive form of the verbs like to' in' togo'
	7090
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	DATE:
9-10	Rule based part at speech (Pas) tagging
)	Tokenization The first step is to takenize the input text into individual token
2)	Linguistic Rules Develop a set of linguistic rules that describes the characteristic and patterns.
3)	Ambiguity Handling Implementing rule based tegging may lead to ambiguits rules should be designed to handle such cases
4)	Expection Handling:- Rules should account for exception and imagular's inthe language.
5)	The involve analyzing.
	dants and a second seco
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