

Total Marks of Question no.		Examiner 1
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		<u>NLP MSE-2</u>
		Q1(a) Discuss Lesk algorithm Ans Lesk algorithm is an overlap approach used for word sense disambiguation. Steps involved are -
	1)	Preprocessing - Tokenize the input text into individual words.
	2)	Remove stop words & punctuation marks to focus on content words.
	3)	Retrieval definition - Retrievive dictionary definition for the ambiguous word from a lexical resource like wordnet
	4)	Calculate overlap - Compare the definition of the ambiguous word with the definitions of its content words. - calculate the overlap between the glosses (definition) of the ambiguous word & the glosses of its content words. - The overlaps can be calculated using various techniques, such as counting words or using semantic similarity measures
	5)	Select the best sense - Choose the sense with the highest overlap as the disambiguated sense for the ambiguous word.

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		word.
		- The sense with the highest overlap indicates the most relevant meaning it gives context.
		Eg:-
		On Burning coal we get ash
		Sense Bag - It contains the words in the definition of a candidate sense of the ambiguous word.
		Context Bag - It contains the words in the definition of each sense of each context word.
		<u>Ash</u>
		<u>Sense 1</u> - Tree of the olive family.
		<u>Sense 2</u> - Solid residue left when combustible material is thoroughly burnt or oxidized.
		<u>Sense 3</u> - To convert into ash.
		<u>Coal</u>
		<u>Sense 1</u> - A piece of glowing carbon or burnt wood.
		<u>Sense 2</u> - Charcoal
		<u>Sense 3</u> - A black solid combustible substance.
		In this case Sense 2 of ash will be winner " it has maximum overlapping words.

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		Q1(b) Discuss the applications of WordNet & BabbieNet Lexical database
Ans		<p>Application of WordNet</p> <ul style="list-style-type: none">- Synonym & Antonym Extraction- word sense disambiguation.- Text classification & sentiment analysis- Query expansion in information retrieval- Ontology building & knowledge representation
		<p>Application of BabbieNet</p> <ul style="list-style-type: none">- Multilingual NLP- Cross language Information Retrieval- Multilingual knowledge representation- Name Entity Recognition- Semantic Text similarity- Cross language Plagiarism Detection.
		Q1(c) Explains how a supervised learning algorithm can be applied for word sense disambiguation.
Ans		<p>Supervised algorithm assumes that context can supply enough evidence to disambiguate words on its own.</p> <ul style="list-style-type: none">- It involves training a model using a labeled dataset of word senses. The model is then used to disambiguate the sense of a target word in a new text.

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		<ul style="list-style-type: none">- <u>Decision List</u> - A decision list is a set of rules that assign to a target word based on context in which it appears
		<ul style="list-style-type: none">- <u>Neural Network</u> - Neural Network such as feedforward, recurrent NN & Transformer NN are used to model the context-sense relationships.
		<ul style="list-style-type: none">- <u>Support Vector Machine</u> - word feature vectors in the context of wSD, words & their context are represented as features like part of speech, word embedding linguistic information. The SVM learns to find a hyperplane that best separates the data points representing the different sense.
		<ul style="list-style-type: none">- <u>Naive Bayes</u> - It is probabilistic algorithm that uses Bayes theorem to classify text into predefined categories.
		<ul style="list-style-type: none">- <u>Decision Tree</u> - It is flowchart-like structures in which an internal node represents feature (or attribute) the branch represents a decision rule & each leaf node represents the outcome.

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(Q2/a) Convert the following rfa grammar into CNF

$$S \rightarrow A \mid cB \mid c$$

$$A \rightarrow Bc \mid b$$

$$B \rightarrow cAA \mid \epsilon$$

Sol) CNF

$$V \rightarrow V, V_2$$

$$V \rightarrow T$$

Remove Null Production

$$S \rightarrow A \mid cB \mid c$$

$$A \rightarrow Bc \mid b \mid c$$

$$B \rightarrow cAA$$

Remove Unit Production

A.

$$S \rightarrow Bc \mid b \mid c \mid cB$$

$$A \rightarrow Bc \mid b \mid c$$

$$B \rightarrow cAA$$

Let $X \rightarrow AA$ $Y \rightarrow C$

$$S \rightarrow BY \mid b \mid c \mid YB$$

$$A \rightarrow BY \mid b \mid c$$

$$B \rightarrow YX$$

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		Q2(b) What are the challenges of NLP & how to resolve them
	Ans 1)	<u>Ambiguity</u> Lexical (word having multiple meaning) syntactic (different ways to parse) semantic (same word interpreted differently) <u>Solution</u> - wSD - use contextual clues & statistical models to determine the correct sense of a word in a given context. - Deep learning models like transformers (BERT) which takes the entire sentence or paragraph into account can reduce ambiguity.
	2)	<u>Understand Content & Pragmatics</u> Challenge - understanding context, especially in complex texts or dialogues remains difficult. - Context embedding - Models like BERT, GPT-4 generate dynamic word embeddings that change depending on the surrounding words. - Reinforcement learning - Systems can be trained using real-world interactions to better interpret pragmatics, intent & tone.

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Space for Marks	Question No.	START WRITING HERE	
	3)	<p><u>Handling Long dependencies</u> <u>challenge - understanding long-range relationships b/w words.</u> <u>Solⁿ - Transformers, BERT, GPT.</u></p>	
	4)	<p><u>Date sparsity</u> <u>challenge - for low resource or specialized domain there must be labelled date for training NLP models</u> <u>Solⁿ - Transfer learning, data augmentation</u></p>	
	5)	<p><u>Domain adaptation</u> <u>Legal, medical applicatⁿ trained on general dataset is a challenge</u> <u>Solⁿ - fine-tuning of dataset</u></p>	
	6)	<p><u>Cultural & linguistic diversity</u> <u>Solⁿ - multilingual model - mBERT</u> <u>Crowdsourcing & low resource model</u></p>	

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		D2(k) Define various POS Tagging methods Ans. POS Tagging methods 1) Rule Based POS Tagging 2) Stochastic POS Tagging 3) Transformation Based Tagging
		<u>Rule Based POS Tagging</u> - They use dictionary or lexicons for getting possible tags for tagging each word. - If the word has more than one possible tag then it uses handwritten rules to identify the correct tag. - The car was noticed free car is either noun or modal verb But since the is Determinant so car should be noun.
		<u>Stochastic POS Tagging</u> - The model that includes frequency or probability is called stochastic. - Stochastic taggers generally resolve tagging ambiguity by using a training corpus to compute the probability of a given word having a given tag in a given context.

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		<p>It uses four main taggers -</p> <ol style="list-style-type: none">1) <u>Word frequency approach</u>: stochastic tagger disambiguates the words based on the probability that a word occurs with a particular tag.2) <u>Tag sequence probability</u>: The tagger calculates the probability of a given sequence of tags occurring. <ul style="list-style-type: none">- Transformation Based tagging:<ul style="list-style-type: none">• A hybrid approach.- Like rule-based taggers - this tagger is based on rules.- Also automatically induced from hand-tagged data.

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Q2(a) Using HMM pos tagging tag
"The park is a book"

Copies

<s> book a car </s>

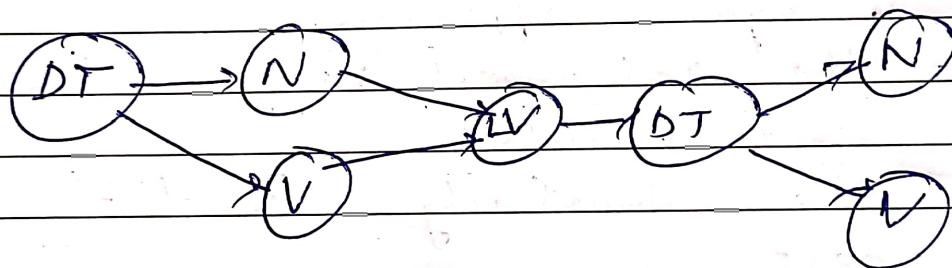
<s> Park the car </s>

<s> The book is in the car </s>

<s> The car is is a park </s>

Step 1

The park is a book



<s> Book a car </s>

V DT N

<s> park the car </s>

V DT N

<s> The Book is in the car </s>

DT N LV IN DT N

<s> The car is in a park </s>

DT N LV IN DT N

DT - Determinant

IN - Preposition

N - Noun

V - Verb

LV - Linking verb

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		Step 2 = Emission Probability Matrix																																										
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		Park as verb $\Rightarrow (\text{Park} \text{ is LV}) \times (\text{Verb})$ $= \frac{1}{2} \times \frac{1}{6} = \frac{1}{12}$ $\therefore \text{Park is Noun}$	is is LV $\Rightarrow \frac{1}{36} \times \frac{1}{6} = \frac{1}{216}$ $= \frac{1}{124}$ $\therefore \text{Park is Noun}$
		Book as Noun $\Rightarrow \frac{1}{324} \times \frac{1}{6} = \frac{1}{1944}$	Book as Verb $\Rightarrow \frac{1}{324} \times \frac{1}{2} \times \frac{1}{6} = \frac{1}{124}$ $\therefore \text{Book is Noun}$

(Q6) Explain -

- (1) Homonymy - same word diff meaning
Bank - financial bank
Bank - river bank
- (2) Heterophone - same sound diff meaning
Right & White
- (3) Polysemy - multiple related meaning
by bank
- (4) Synonym - same meaning different word
sofa & couch
- (5) Antonym - opposite
light / dark
- (6) Hypernym
Car is hypernym of animal vehicle
- (7) Hyponym - Vehicle is hyponym of car