Birla Institute of Technology & Science, Pilani Work-Integrated Learning Programmes Division Second Semester 2020-2021 M.Tech (Data Science and Engineering)

End-Semester Test (EC-3 Makup)

Course No. : DSECLZG525

Course Title : Natural Language Processing

Nature of Exam : Open Book

Weightage : 50%

Note: Assumptions made if any, should be stated clearly at the beginning of your answer.

Question 1.

a) Consider the training set:

(4 marks)

No. of Pages

No. of Questions = 5

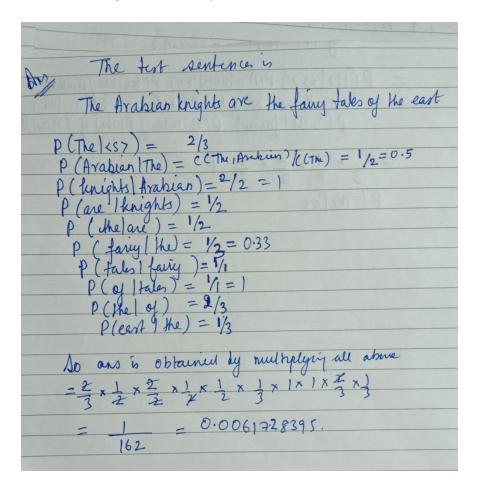
The Arabian knights

These are the fairy tales of the east

The stories of the Arabian knights are translated in many languages

Compute using the bigram model the probability of the sentence. Include start and end symbol in your calculations.

The Arabian knights are the fairy tales of the east



- b) Using Penn Tree bank, find the POS tag sequence for the following sentences: [6 Marks]
 - 1. The actor was happy he got a part in a movie even though the part was small. [2 marks]
 - 2. I am full of ambition and hope and charm of life. But I can renounce everything at the time of need [3 marks]
 - 3. When the going gets tough, the tough get going. [1 mark] Solution

The/DT actor/NN was/VB happy/JJ he/PRP got/VB a/DT part/NN in/IN a/DT movie/NN "even though"/CC the/DT part/NN was/VB small/ADV. [2 marks]

I//PRP am/VB full/JJ of/IN ambition/NN and/CC hope/NN and/CC charm/JJ of/IN life/NN. But/CC I/PRP can/VB renounce/VB everything/JJ at/IN the/DT time/NN of/IN need/NN [3 marks]

When/WDT the/DT going/NN gets/VB tough/RB, the/DT tough/NN get/VB going/RB.[1 mark]

Tag	Description	Example	Tag	Description	Example
CC	coordin. conjunction	and, but, or	SYM	symbol	+,%, &
CD	cardinal number	one, two, three	TO	"to"	to
DT	determiner	a, the	UH	interjection	ah, oops
EX	existential 'there'	there	VB	verb, base form	eat
FW	foreign word	mea culpa	VBD	verb, past tense	ate
IN	preposition/sub-conj	of, in, by	VBG	verb, gerund	eating
JJ	adjective	yellow	VBN	verb, past participle	eaten
JJR	adj., comparative	bigger	VBP	verb, non-3sg pres	eat
JJS	adj., superlative	wildest	VBZ	verb, 3sg pres	eats
LS	list item marker	1, 2, One	WDT	wh-determiner	which, that
MD	modal	can, should	WP	wh-pronoun	what, who
NN	noun, sing. or mass	llama	WP\$	possessive wh-	whose
NNS	noun, plural	llamas	WRB	wh-adverb	how, where
NNP	proper noun, singular	IBM	\$	dollar sign	\$
NNPS	proper noun, plural	Carolinas	#	pound sign	#
PDT	predeterminer	all, both	"	left quote	or "
POS	possessive ending	's	,,	right quote	or "
PRP	personal pronoun	I, you, he	(left parenthesis	[, (, {, <
PRP\$	possessive pronoun	your, one's)	right parenthesis],), }, >
RB	adverb	quickly, never	,	comma	,
RBR	adverb, comparative	faster		sentence-final punc	.!?
RBS	adverb, superlative	fastest	:	mid-sentence punc	: ;
RP	particle	up, off			

Question 2.

a) Build a parse tree for the sentence "She loves to visit Goa" using Probabilistic Parsing [5marks]

 $S \rightarrow NP VP 1.0$

 $VP \rightarrow V PP 0.4$

 $VP \rightarrow V NP 0.6$

 $PP \rightarrow P NP 1.0$

NP -> V NP 0.1

 $NP \rightarrow NP PP 0.3$

 $NP \rightarrow N 0.3$

 $N \rightarrow visit 0.3$

 $V \rightarrow visit 0.6$

 $N \rightarrow Goa 0.3$

 $N \rightarrow She 0.5$

 $V \rightarrow loves 1$

 $P \rightarrow to 1$

 $DT \rightarrow a 1$

a) State the correct sequence of actions that generates the following parse tree of the sentence "She baught her a dress" using Arc-Eager Parsing [5marks]



Solution:

Transitions: SH-LA-SH-RA-SH-LA-RE-RA-RE-RA

Arcs:

She <- baught

baught > her

a <- dress

baught -> dress

baught -> .

Question 3. Word sense disambiguation and ontology-

b) What are lexical sample task and all word task in word sense disambiguation? How can sources like Wikipedia be used for word sense disambiguation [2 marks]
 Solution

What are lexical sample task and all word task in word sense disambiguation?

Lexical sample task and all word task are 2 variants of word sense disambiguation

- Lexical sample task -Small pre-selected set of target words
- All-words task System is given an all-words entire texts and lexicon with an inventory of senses for each entry. We have to disambiguate every word in the text (or sometimes just every content word).

How can sources like Wikipedia be used for word sense disambiguation

Wikipedia can be used as training data for word sense disambiguation using supervised learning techniques

- Concept is mentioned in a Wikipedia: article text may contain an explicit link to the concept's Wikipedia page, which is named by a unique identifier (can be used as a sense annotation)
- These sentences can then be added to the training data for a supervised system.

How can WordNet relations be used for word sense disambiguation in following sentences:

[3 marks]

- 1. A bat is not a bird, but a mammal.
- 2. Jaguar reveals its quickest car ever
- 3. Raghuram Rajan was the 23rd Governor of the Reserve Bank of India

Solution

Nouns and verbs can be extracted from the sentences. The senses in wordnet can be extracted for these words and senses with close relations can be extacted as correct sense.

- 1. Bat can be sports bat or mammal. But looking at nouns bat, bird and mammal, correct sense of bat as MAMMAL can be found using WordNet relations.
- 2. Jaguar can be a car or animal. Looking at nouns Jaguar, correct sense of Jaguar as CAR can be found using WordNet relations.
- 3. Bank can be river bank or financial bank.: Search senses of nouns Bank,"Raghuram Rajan",
 Governer. The correct sense of BANK as FINANCIAL sense can be found using WordNet relations.
 - How is Syntactic web different from the Semantic web? What is URI in semantic web ontology? [2 marks]

Syntactic web consist of huge data on net connected by hyperlinks which is rendered by machines but machines cannot process it due to inability to understand the meaning of the content.

The semantic Web identifies a set of technologies, tools, and standards which form the basic building blocks of an infrastructure to support the vision of the Web associated with meaning.

A Universal Resource Identifier (URI) is a formatted string that serves as a means of identifying abstract or physical resource. A URI can be further classified as a locator, a name, or both. Every resource is identified with unique URI in ontology.

Develop an OWL ontology using the following for animal kingdom for classes like carnivorous, herbivorous and omnivorous. Use following Property characteristics, restrictions and Class expressions [3 marks]

- inverseOf
- domain
- range
- Cardinality
- disjointWith
- subClassOf

```
<rdfs:Class rdf:ID="Carnivorous">
  <rdfs:subClassOf rdf:resource="#Animal"/>
</rdfs:Class>
<rdfs:Class rdf:ID="Herbivorous">
  <rdfs:subClassOf rdf:resource="#Animal"/>
</rdfs:Class>
<rdfs:Class rdf:ID="Omnivorous">
  <rdfs:subClassOf rdf:resource="#Animal"/>
</rdfs:Class>
<rdfs:Class rdf:ID="Carnivorous">
  <owl:disjointWith rdf:resource="#Herbivorous"/>
</rdfs:Class>
<owl:Restriction>
   <owl:onProperty rdf:resource="#hasLegs" />
   <owl:cardinality</pre>
 rdf:datatype="&xsd;nonNegativeInteger">4</owl:cardinality>
 </owl:Restriction>
 <owl:ObjectProperty rdf:ID="Eats">
   <rdfs:domain rdf:resource="#Carnivorous"/>
```

```
<rdfs:range rdf:resource="#Animal"/>
</owl:ObjectProperty>
```

Question 4.

a) In this modern age where the internet is growing rapidly, the existence of the internet can make it easier for tourist to find information about hotels. Tourists usually tell the experience during the hotel by writing reviews on the internet. Hence many hotel's reviews are found on the internet. With the availability of reviews on the internet with large numbers, tourists can't understand all the reviews they read whether they contain positive or negative opinions. It takes a sentiment analysis to quickly detect if the reviews is a positive or negative reviews. Using the Multinomial Naïve Bayes Classifier method find out that the given hotel reviews are positive or negative.

D1	The hotel is clean and great	Positive
D2	The hotel owner is very helpful	Positive
D3	Overall Aston Hotel's experience was great	Positive
D4	The condition of the hotel was very bad	Negative
D5	A HORRIBLE EXPERIENCE FOR ONE WEEK	Negative
D6	The hotel view was great	?
D7	My holiday experience stay in usa so horrible	?
D8	Overall the hotel in aston very clean and great	?

Soln:

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b. Compute the BLEU score for the below translations (candidate1, candidate2). Consider 1 gram, 2 gram, 3 gram, 4 gram and Brevity-Penalty for calculating BLUE score .

Reference: The teacher arrived late because of the traffic

Candidate 1: The teacher was late due to the traffic

Candidate 2: A teacher arrived late because of transportation

B) en Score

Cardidade 1

Uni gram = 4

Bigram = 0

Four gram = 0

Candidate 2

Unigram = 6

Bigram = 4

Tri gram = 3

Four gram = 3

Four gram = 3

Four gram = 3

Four gram = 1

BP for cardidate 1 = 0.867

Blen Score for cardidate 1 6 0

BP fro cardidate = 0.615

Blen Oceretor cardidate 2 6 0 58

1. Identify the type of the dialogue agent from the following conversation. Also explain the working of that type of dialogue agent. (5 marks)

System: How may I help you?

User: I want to book dinner at a restaurant that serves good Chinese food

System: Berkos is a restaurant that serves good Chinese food.

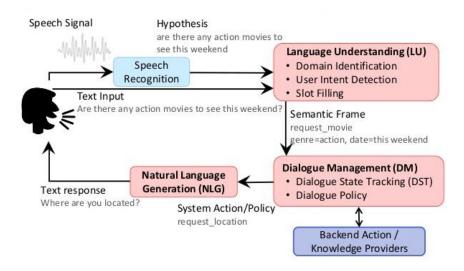
User: ok

System: Should I book a dinner?

User: yes, please System: ok.

Soln: Above is the task oriented dialogue system

The diagram shows the working of the task oriented dialogue system. The student should explain briefly each step as mentioned in the slides



2. Find the intent, domain and slots for the following:

Book an appointment on 12th Feb 2021 at 10 am for a ECG Test.

(2 marks)

SOLN:

DOMAIN: Medical

INTENT: Book an Appointment

Slots

Services: ECG TEST
 Date: 12th Feb 2021

> Time: 10 AM

3. In a collection of 10000 document, the following words occur in the following number of documents: (3 marks)

Oasis occurs in 400 documents, Place occurs in 3500 documents, Desert occurs in 800 documents, Water occurs in 800 documents, Comes occur in 800 documents
Beneath occurs in 200 documents, Ground occurs in 900 documents
Calculate TF-IDF term vector for the following document:
Oasis Place Desert Water Comes Beneath Ground Place

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	Tenn freg.	IDF	TENIOF
Oasio	1/8	log (10009400)	0,1747
Place	2/8	dog (10000/3500)	0-11398
Desert	1/8	log (1000/800)	0.137114
Water	1/8	leg(10000/god)	0.137114
Cemes	1/8	log(1000/800)	0-137114
Beneath	1/8	dog(woo/200)	07212371
ground	48	log (10000/900)	0.13072
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TF-10P	veetor	(0.1747,0.11398,0	0,137114,0,137114
			17114, 0,212371,
			(072)