

Q1) a)  $\langle s \rangle$  she dances gracefully  $\langle /s \rangle$   
 $\langle s \rangle$  Gracefully she dances  $\langle /s \rangle$   
 $\langle s \rangle$  She dances  $\langle /s \rangle$

(i)  $P(\text{she} | \langle s \rangle) = \frac{2}{3} = 0.6667$

(ii)  $P(\langle /s \rangle | \text{dances}) = \frac{2}{3} = 0.6667$

(iii)  $P(\text{gracefully} | \text{dances}) = \frac{1}{3} = 0.3332$

(iv)  $P(\text{gracefully} | \langle s \rangle) = \frac{1}{3} = 0.3332$

(v)  $P(\text{dances} | \text{she}) = \frac{3}{3} = 1$

(vi)  $P(\text{she} | \text{gracefully}) = \frac{1}{2} = 0.5$

Q1b) i) "The cat is under the table"

closed class words  $\rightarrow$  The, is, under  
 open class words  $\rightarrow$  cat, table

ii) "One kilobyte is equivalent to 1024 bits"

closed class words  $\rightarrow$  One, is, to, 1024

open class words  $\rightarrow$  Kilobyte, equivalent, bits

Q1c) (i) Both document clustering & topic modelling are used for organising documents of text into topics.

Topic model  $\rightarrow$  eg. LDA Generative models, generates words using documents multinomial distribution, learn topic representation using word distributions, uses probabilistic distributions to identify similar words, must know number of topics, cannot capture correlation between topics (assume independence).

Document clustering: Also unsupervised, don't need to know number of topics, Doesn't rely on priors or conjugates, scalable

Q1c) ii) 1) Cataloguing topics of products in e-commerce  
 2) creating documents & topics for wikipedia catalogue  
 3) Recommendation by item & customer feedback



Q3) a) I left the umbrella in the car

→ Lesk algorithm is a knowledge based word sense disambiguation algorithm, considering the dictionary gloss or definitions provided for "left".

the steps are

(1) For the given definitions & the required word "left", calculate the number of non stopwords overlapping with the given context.

(2) whichever definition has highest overlapping non-stop words with context, choose that one for the disambiguation.

For this case,

(v) or (verb) has 2 overlapping non-stopwords

i.e. umbrella & car, whereas (n) & (adj) have only one.

Therefore, (v) or verb will be chosen.

Q3b) The way humans "see" a webpage or how a machine sees a webpage is very different. Semantic content is understandable to humans but not to computers as they interpret machine language.

→ Semantic web technologies help with interlinking data over the web

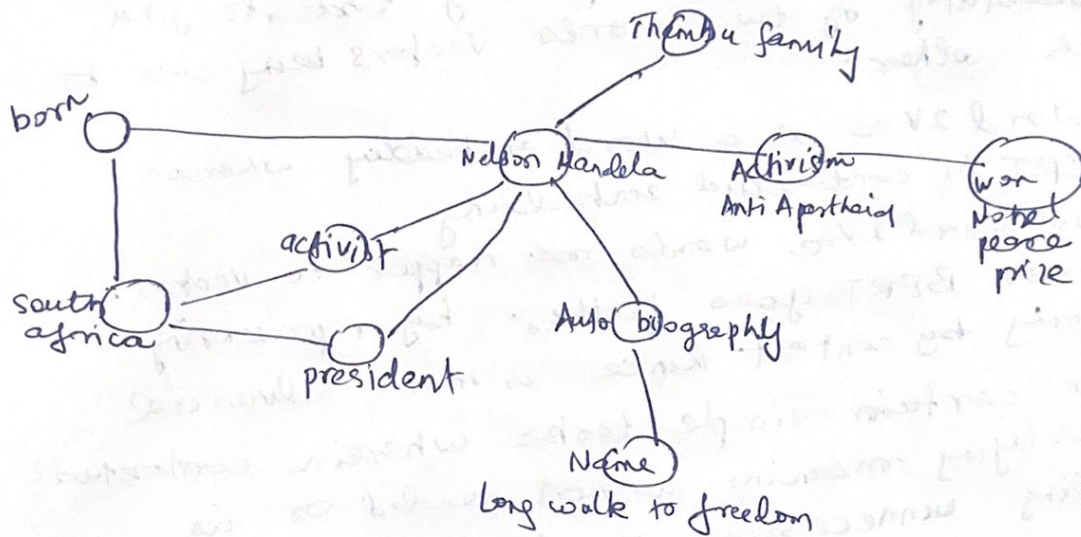
→ This creates relations between content giving context based searches possible for users.

→ This semantic web technologies make this possible for more expressive representation of this knowledge.

→ This major application can be put to use for information retrieval improving keyword matching, cross language retrieval, semantic, context based information retrieval etc.

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Q3) c) we can use concept of triple (subject, predicate, object).





Q4) a) Word2Vec is a popular embedding method.

It is fast to train, using techniques like skip gram.

A classifier is trained on a binary prediction task to find a word closest to a given context word.

→ Uses a large corpus of text, to find probability of two words vectors being close to each other.

→ Word2Vec is a word embedding whereas BERT is contextual embedding

→ In word2Vec words are mapped to vectors, whereas BERT goes further by representing meaning by context hence is more advanced

→ For certain simple tasks wherein contextual underlying meaning is not needed or is adding unnecessary complexity, Word2Vec offers a good or sometimes better approach

→ It reduces training & deployment complexity as word2vec corpus are readily available & require much less computational capacity.

Q4 b) Matrix for word-word cooccurrence

	bank 1	bank 2	river	money	bug
bank 1					
bank 2			0.1	0.2	0.3
river	0.2	0.5			
money			0.7	0.1	0.5
bug	0.4	0.7			

$$P(\text{bank 1, river \& bug}) = P(\text{river \& bug})P(\text{bank 1})$$

$$P(\text{bank 1, river \& bug}) = 0.1 + 0.3 = 0.4$$

$$P(\text{bank 2, river \& bug}) = 0.7 + 0.5 = 1.2$$

Hence, bank 2 is more related to river & bug.



Q1(c) Query  
Nile  
River  
Bug.

$$\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$$

Value

$$\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$$

Key

$$\begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$$

Query  $\rightarrow$  Nile

Attention weights =  $\text{softmax}(\text{Query} \cdot \text{Key})$

Attention weights =  $\text{softmax}([1 \times 1 + 0 \times 1 + 1 \times 0]) = \text{softmax}([1])$

Since  $\text{softmax}([1]) = [0.737; 0.269]$  are the attention values

Now we compute weighted sum of values using these attention weights

Output = Attention weights  $\times$  Value:

$$\text{Output} = [0.737 \times 1 + 0.269 \times 0, 0.737 \times 0 + 0.269 \times 0, 0.737 \times 1 + 0.269 \times 1]$$

$$\text{Output} = [0.737, 0, 0.737 + 0.269]$$

$$\text{Output} = [0.737, 0, 1]$$

Therefore output for query, Nile will be:

$$[0.737, 0, 1]$$

Q5) Query: 'Who is the friend of bug?'

$$\text{Similarity}(\text{document } i, \text{document } j) = 10 \times \frac{\text{number of overlapping words in } d_i \& j}{\text{length of } d_i \& j}$$

$$d1 = 10(\text{bug, the, of}) = 10(3) = 30$$

$$d2 = 10(\text{the, of, bug}) = 10(3) = 30$$

$$d3 = 10(\text{the, is, friend, bug, of}) = 10(5) = 50$$

$$d4 = 10(\text{bug}) = 10$$

$$d5 = 10(\text{the, bug, of}) = 10(3) = 30$$

Using the given function for similarity, we see that  $d3$  has maximum number of overlapping words (5), & a score of 50.

	Who is the friend of bug
$d1$	30
$d2$	30
$d3$	50
$d4$	10
$d5$	30

(b) Sort  $D$ , any max  $\left[ \sum \text{Rel}(i) - \sum \text{Rel}(i, j) \right]$   
 $\therefore D1 \& D3$ , Summary, "Bug placed in computer is a friend of the bug"

$$(c) \text{Range} = \frac{\sum \sum \text{count}(i, x) \text{count}(i, x)}{\sum \sum \text{count}(i, i)}$$

$$= \frac{1+1}{3+5} = \frac{2}{8} = \frac{1}{4} = 0.25$$