

Natural Language Processing

Assignment 4

Type of Question: MCQ

Number of Questions: 8

Total Marks: $(6 \times 1) + (2 \times 2) = 10$

1. Baum-Welch algorithm is an example of -

[Marks 1]

- A) Forward-backward algorithm
- B) Special case of the Expectation-maximization algorithm
- C) Both A and B
- D) None

Answer: A

Solution: Theory.

2. Once a day (e.g. at noon), the weather is observed as one of
state 1 : rainy state 2: cloudy state 3: sunny
The state transition probabilities are :

0.4	0.3	0.3
0.2	0.6	0.2
0.1	0.1	0.8

Given that the weather on day 1 ($t = 1$) is sunny (state 3), what is the probability that the weather for the next 7 days will be “sun-sun-rain-rain-sun-cloudy-sun”?

[Marks 2]

- A) 1.54×10^{-4}
- B) 8.9×10^{-2}
- C) 7.1×10^{-7}
- D) 2.5×10^{-10}

Answer: A

Solution:

$O = \{S3, S3, S3, S1, S1, S3, S2, S3\}$

$P(O \mid \text{Model})$

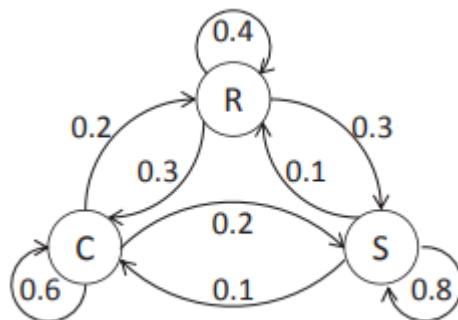
$= P(S3, S3, S3, S1, S1, S3, S2, S3 \mid \text{Model})$

$= P(S3) P(S3|S3) P(S3|S3) P(S1|S3) P(S1|S1) P(S3|S1) P(S2|S3) P(S3|S2)$

$= Q3 \cdot a33 \cdot a33 \cdot a31 \cdot a11 \cdot a13 \cdot a32 \cdot a23$

$= (1)(0.8)(0.8)(0.1)(0.4)(0.3)(0.1)(0.2)$

$= 1.536 \times 10^{-4}$



3. In the above question, the expected number of consecutive days of sunny weather is:

- A) 2
- B) 3
- C) 4
- D) 5

[Marks 1]

Answer: D

Solution:

$\text{Exp}(i) = 1/(1-p_{ii})$ So for sunny the $\text{exp} = 1/(1-0.8) = 5$

4. Let us define an HMM Model with K classes for hidden states and T data points as observations. The dataset is defined as $X = \{x_1, x_2, \dots, x_T\}$ and the corresponding hidden states are $Z = \{z_1, z_2, \dots, z_T\}$. Please note that each x_i is an observed variable and each z_i can belong to one of classes for hidden state. What

will be the size of the state transition matrix, and the emission matrix, respectively for this example. **[Marks 1]**

- A) $K \times K, K \times T$
- B) $K \times T, K \times T$
- C) $K \times K, K \times K$
- D) $K \times T, K \times K$

Answer: A

Solution: Since there are K hidden states, the state transition matrix will be of size $K \times K$. The emission matrix will be of size $K \times T$, as it defines the probability of emitting an observed state from a hidden state.

5. You are building a model distribution for an infinite stream of word tokens. You know that the source of this stream has a vocabulary of size 1000. Out of these 1000 words you know of 100 words to be stop words each of which has a probability of 0.0019. With only this knowledge what is the maximum possible entropy of the modelled distribution. (Use log base 10 for entropy calculation) **[Marks 2]**

- A) 5.079
- B) 0
- C) 2.984
- D) 12.871

Answer: C

Solution: There are 100 stopwords with each having an occurrence probability of 0.0019. Hence,

$$P(\text{Stopwords}) = 100 * 0.0019 = 0.19$$

$$P(\text{non-stopwords}) = 1 - 0.19 = 0.81$$

For maximum entropy, the remaining probability should be uniformly distributed.

For every non-stopword w , $P(w) = 0.81/(1000 - 100) = 0.81/900 = 0.0009$. Finally, the value of the entropy would be,

$$H = E(\log(1/p))$$

$$= -100(0.0019 * \log(0.0019)) - 900(0.0009 \log(0.0009))$$

$$= -2.9841$$

6. For an HMM model with N hidden states, V observable states, what are the dimensions of parameter matrices A, B and π ? A : Transition matrix, B : Emission matrix, π : Initial Probability matrix. **[Marks 1]**

- A) $N \times V, N \times V, N \times N$
- B) $N \times N, N \times V, N \times 1$
- C) $N \times N, V \times V, N \times 1$
- D) $N \times V, V \times V, V \times 1$

Answer: B

Solution: Matrix A contains all the transition probabilities and have dimension $N \times N$. Similarly, matrix B contains all the emission probabilities and dimension $N \times V$. Similarly, π contains initial probability for all hidden states and have dimension $N \times 1$.

7. Suppose you have the input sentence "Death Note is a great anime". And you know the possible tags each of the words in the sentence can take.

- Death: NN, NNS, NNP, NNPS
- Note: VB, VBD, VBZ
- is: VB
- a: DT
- great: ADJ
- anime: NN, NNS, NNP

How many possible hidden state sequences are possible for the above sentence and States? **[Marks 1]**

- A) $4 \times 3 \times 3$
- B) 4^{3^3}
- C) $2^4 \times 2^3 \times 2^3$
- D) $2^{4 \times 3 \times 3}$

Answer: A

Solution: Each possible hidden sequence can take only one POS tag for each of the words. Hence the total possibility will be a product of the number of candidates for each word.

8. In Hidden Markov Models or HMMs, the joint likelihood of an observed sequence O with a hidden state sequence Q , is written as $P(O, Q; \theta)$. In many applications, like POS tagging, one is interested in finding the hidden state sequence Q , for a given observation sequence, that maximizes $P(O, Q; \theta)$. What is the time required to compute the most likely Q using an exhaustive search? The required notations are, N : possible number of hidden states, T : length of the observed sequence. **[Marks 1]**

- A) Of the order of TN^T
- B) Of the order of N^2T
- C) Of the order of T^N
- D) Of the order of N^2

Answer: A

Solution: We will need to compute $P(O, Q|\theta)$ for all possible Q . There are a total of N^T possible hidden sequences Q for a sequence of length T . Each individual probability calculation also requires T multiplications.