

Question # 1

- (a) Index encodings: Assign unique indices to each word. What are the indices for "Cat", "Dog", and "Bird"?

"Cat" : 0

"Dog" : 1

"Bird" : 2

- (b) One-hot encoding: Assume one-hot encoding with vectors of length 3 (for simplicity). Provide the one-hot encoding representation for each of the three words: "Cat", "Dog", "Bird".

"Cat" : $[1, 0, 0]$

"Dog" : $[0, 1, 0]$

"Bird" : $[0, 0, 1]$

- (c) Encode the following documents using Bag of words (BOW) technique:

Documents:-

- 1) "I like natural language processing. It is fascinating to see how computers can understand and generate human-like text".
- 2) "Natural language processing involves the interaction between computers and humans using natural language. It is a field of study that focuses on making computers understand and process language".

④

Vocabulary: { "I", "like", "natural", "language", "processing",

"It", "is", "fascinating", "to", "see", "how", "computers",
"can", "understand", "and", "generate", "human-like",
"text", "involves", "the", "interaction", "between",
"humans", "using", "a", "field", "of", "study", "that",
"focuses", "on", "making", "process", "find", "very",
"interesting", "ability", "machines", "comprehend",
"significant", "advancement", "in", "technology"}.

Document 1: [1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0]

Document 2: [0, 0, 2, 3, 1, 1, 1, 0, 0, 0, 0, 2, 0, 1, 2, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0]

Document 3: [1, 0, 1, 2, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 2, 0, 0, 0, 0, 1, 1, 2, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1]

(d) Encode the following documents using term frequency - inverse Document frequency (TF - IDF) scheme:

Documents:-

- 1). "I like natural language processing."
- 2). "Natural language processing is interesting".
- 3). "I find natural language processing fascinating".

Word	Document 1	Document 2	Document 3
I	1	0	1
like	1	0	0
natural	1	1	1
language	1	1	1
processing	1	1	1
is	0	1	0
interesting	0	1	0
find	0	0	1
fascinating	0	0	1

Question #2:-

1) Lisa is my pet dog

Lisa : $[1, 0, 0, 0, 0]$

is : $[0, 1, 0, 0, 0]$

my : $[0, 0, 0, 1, 0]$

pet : $[0, 0, 1, 0, 0]$

dog : $[0, 0, 0, 0, 1]$

weights : $[0.2, 0.3, 0.1, -0.2, 0.5]$

output layer weights

$= [0.6, -0.4, 0.2, 0.1, 0.3]$

→ Forward Pass

"is" target word

"Lisa" "my" context word

Input vector = $[1, 0, 0, 0, 0] \parallel [0, 0, 0, 1, 0] = [1, 0, 0, 1, 0]$

Hidden Layer Activation = $(\sum i w)$

$$= [0.13, 0.49, 0.14, 0, 0.93]$$

$$= \text{sigmoid}(\sum h_i w_i)$$

Output vector = $[0.74, 0.15, 0.35, 0.13, 0.61]$

→ Loss

$$= -\sum (\text{target-vector} \times \log(\text{output-vector}))$$

$$= -\sum (1 \times \log(0.74) + 0 + 0 + 0 + 0)$$

$$= -\log(0.74)$$

$$= -0.322$$

Back Propagation:

$$\begin{aligned}\frac{\partial L}{\partial w_{\text{output}}} &= (T - O) \times \text{hidden-layer-activation} \\ &= ([1, 0, 0, 0, 0] - [0.74, 0.15, 0.35, 0.13, 0.61]) \\ &\quad \times [0.13, 0.49, 0.14, 0, 0.93] \\ &= [-0.21, 0.11, -0.22, 0.00, -0.09]\end{aligned}$$

$$\begin{aligned}\frac{\partial L}{\partial w_{\text{hidden}}} &= [1, 0, 0, 0, 0] - [0.74, -0.15, 0.35, 0.13, 0.67] \\ &\quad \times [1, 0, 0, 1, 0] \times [0.2, 0.3, 0.1, -0.2, 0.5] \\ &= [0.148, 0, 0, -0.26, 0]\end{aligned}$$

$$\begin{aligned}w_{\text{output}} &= w_{\text{output}} - \alpha \frac{\partial L}{\partial w_{\text{output}}} \\ &= [0.6, -0.4, 0.2, 0.1, 0.3] - 0.01[-0.21, 0.11, -0.22, 0, 0.93] \\ &= [0.5, -0.4, 0.2, 0.1, 0.274]\end{aligned}$$

$$\begin{aligned}w_{\text{hidden}} &= w_{\text{hidden}} - \alpha \frac{\partial L}{\partial w_{\text{hidden}}} \\ &= [0.2, 0.3, 0.1, -0.2, 0.5] - 0.01[0.148, 0, 0, -0.26, 0] \\ &= [0.186, 0.3, 0.1, -0.226, 0.5]\end{aligned}$$

Forward:-

target "my"

context "is", "pet"

$$\begin{aligned}\text{input} &= [0, 1, 0, 0, 0] \parallel [0, 0, 1, 0, 0] \\ &= [0, 1, 1, 0, 0]\end{aligned}$$

$$\begin{aligned} \text{hidden} &= \text{ReLU}(w_i^o) \\ &= \text{ReLU}[0, 0.3, 0.1, 0, 0] \\ &= [0, 0.3, 0.1, 0, 0] \end{aligned}$$

$$\begin{aligned} \text{output} &= \text{softmax}(hw_{\text{out}}^o) \\ &= \text{softmax}([0, 0.3, 0.1, 0, 0] \cdot [0.5, 0.4, 0.2, 0.1, 0.2]) \\ &= [0, -0.12, 0.102, 0, 0] \end{aligned}$$

2) Skip Gram:

$$w_{ih} = [[0.4, 0.2, 0.3], [0.5, 0.6, 0.7]]$$

$$w_{ho} = [0.8, 0.9, 0.1]$$

\Rightarrow forward

$$h = \tanh(w_{ih} + b)$$

For "Lisa"

$$\begin{aligned} h_{\text{-Lisa}} &= \tanh([0.4, 0.2, 0.3] \cdot [0.5, 0.6, 0.7] + [1, 0, 0, 0, 0] \\ &\quad + [0, 0, 0]) \end{aligned}$$

$$= \tanh([0.4, 0.5])$$

$$= (0.481, 0.631)$$

$$h_{\text{-is}} = \tanh([0.46, 0.674])$$

$$h_{\text{-my}} = [0.517, 0.687]$$

$$h_{\text{-pet}} = [0.562, 0.731]$$

$$h_{\text{-dog}} = [0.557, 0.719]$$

Calculate Output:

$$Y = \text{softmax}(w \cdot h_0 \times h + b - \theta)$$

For Lisa

$$\begin{aligned} Y_{\text{Lisa}} &= \text{softmax}(w \cdot h_0 \times h_{\text{Lisa}} + b - \theta) \\ &= \text{softmax}([0.8, 0.9, 0.1] \times [0.481, 0.613] + [0, 0]) \\ &= [0.376, 0.624] \end{aligned}$$

$$Y_{\text{is}} = [0.382, 0.618]$$

$$Y_{\text{my}} = [0.418, 0.582]$$

$$Y_{\text{pet}} = [0.413, 0.587]$$

$$Y_{\text{dog}} = [0.386, 0.614]$$

Calculate Loss:-

$$\text{Loss} = -\log(Y_{\text{is}})$$

For Lisa

$$\begin{aligned} &= -\log(0.624) \\ &= -0.523 \end{aligned}$$

$$\text{Loss}_{\text{is}} = 0.535$$

$$\text{Loss}_{\text{my}} = -0.604$$

$$\text{Loss}_{\text{pet}} = -0.599$$

$$\text{Loss}_{\text{dog}} = -0.519$$

$$\text{Total Loss} = \sum L = 2.780$$

Back Propagation:-

For Lisa

$$\frac{\partial L}{\partial b} = [-0.376, -0.624]$$

$$\begin{aligned} \frac{\partial \text{Loss}}{\partial h} &= [0.8, 0.9, 0.1] \times [-0.376, -0.624] \\ &= [-0.28, -0.534] \end{aligned}$$

$$\frac{\partial \text{loss}}{\partial \text{in}} = \delta(\text{in}) \times \delta 1$$

$$= \text{sigmoid}([0.481, 0.613]) \times [-0.28, -0.534]$$

$$= [0.135, 0.188]$$

Question #3:-

It is raining

It: $[1, 0, 0, 0, 0]$

is: $[0, 1, 0, 0, 0]$

raining: $[0, 0, 0, 1, 0]$

Hidden State Initialization

$h_0 = [0, 0]$

Time step 1:

$$h_1 = \tanh(U \times [1, 0, 0, 0, 0] + w \times h_0)$$

$$= \tanh[0.1, 0.2, 0.1, 0.5]$$

$$= [0.1026, 0.1973]$$

Time step 2 ~~(0.1026, 0.1973)~~

$$h_2 = \tanh(U \times [0, 1, 0, 0, 0] + w \times h_0)$$

$$= [0.99, 0.96]$$

$$h_3 = [0.87, 0.50]$$

$$h_4 = [0.94, 0.61]$$

$$h_5 = [0.975, 0.69]$$

Output:

$$[0.10 \times 0.1, 0.19 \times 0.6] + [0.9 \times 0.1, 0.096 \times 0.6] \\ + [0.87 \times 0.1, 0.6 \times 0.5]$$

$$= [0.1, 0.54] + [0.09, 0.34] + [0.087, 0.30]$$

$$= [0.29, 0.66]$$

$$= 0.29 + 0.66$$

$$= \frac{1}{1 + e^{-0.95}}$$

$$= 0.72$$

Loss:-

$$-y \log(\hat{y}) - (1-y) \log(1-\hat{y})$$

$$= 0.024$$

Backpropagation:-

Question 3

is:

$$\text{Input: } [0, 1, 0, 0, 0]$$

$$\begin{aligned}w \cdot \text{ht} - 1 &= [0.1, 0.2, 0, 0.2] [0.76, 0.76] \\&= [0.76, 0.152]\end{aligned}$$

$$\begin{aligned}\text{ht} &= \tanh[1, 1] + [0.76, 0.152] \\&\approx [0.8, 0.8]\end{aligned}$$

warm:

$$\begin{aligned}v \cdot \text{xt} &= [1, 1] [0, 0, 0, 1] \\&= [0, 1]\end{aligned}$$

$$w \cdot \text{ht} - 1 = [0.16, 0.87]$$

Outside:

$$\begin{aligned}v \cdot \text{xt} &= [1, 1] [0, 0, 1, 0, 0] \\&= [0, 1]\end{aligned}$$

$$\begin{aligned}w \cdot \text{ht} - 1 &= [0.1, 0.2, 0, 0.2] [0.16, 0.87] \\&= [0.18, 0.37]\end{aligned}$$

$$\begin{aligned}\text{ht} &= \tanh([0, 1] + [0.18, 0.374]) \\&\approx [0.179, 0.88]\end{aligned}$$

Output :-

$$\begin{aligned}\text{Sigmoid}(v \cdot h_3) &= [0.1, 0.6] [0.179, 0.88] \\&= 0.557\end{aligned}$$

Loss:-

$$\begin{aligned}\text{Loss} &= -y \log \hat{y} - (1-y) \log (1-\hat{y}) \\&= 0.856\end{aligned}$$

$$= -y \log(\hat{y}) - (1-y) \log(1-\hat{y})$$

$$= 0.024$$

Back Propagation:-

b)

$$It: [1, 0, 0, 0, 0]$$

$$is: [0, 1, 0, 0, 0]$$

$$warm: [0, 0, 0, 0, 1]$$

$$outside: [0, 0, 1, 0, 0]$$

It:

$$U \cdot xt = [1, 1] [1, 0, 0, 0, 0]$$

$$= [1, 1]$$

$ht-1 = [0, 0]$
for time step 1

$$w \cdot ht-1 = [0.1, 0.2, 0, 0.2] [0, 0]$$

$$= [0, 0]$$

$$ht = \tanh [1, 1] + [0, 0]$$

$$\approx [0.76, 0.76]$$

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$$V_{\text{new}} = V - \text{learning rate} \cdot \frac{\partial \text{Loss}}{\partial V} = [0.10039, 0.60]$$

$$U_{\text{new}} = U - \text{learning rate} \cdot \frac{\partial \text{Loss}}{\partial U} = [1.0014, 0.0002, \\ 0.1684, 0.2012]$$

$$W_{\text{new}} = W - \text{learning rate} \cdot \frac{\partial \text{Loss}}{\partial W} = [0.12, 0.6, 0.347, 0.21]$$