Query-by-Example Spoken Term Detection Using Attention-Based Multi-Hop Networks

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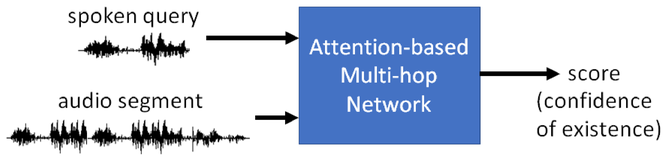
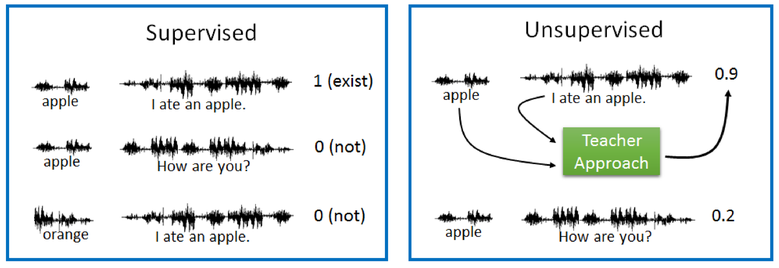
1. Introduction

- Task: query-by-example spoken term detection
  - Given a spoken query, detecting whether an audio segment contains the spoken query
  - Matching of signals directly on the acoustic level without transcribing them into text.
  - We propose an end-to-end attention-based multi-hop



- Supervised Learning
  - Given the true labels to learn, become a classification problem
- Unsupervised Learning
  - Generating the labels by teacher approach (e.g. Dynamic Time Warping)



2. Proposed Approach

- Query Representation (A)
  - Input MFCC feature sequence: X_1 , X_2 , … , X_T
  - Using LSTM encode to V_Q
- Audio Segment Representation (B)
  - Input MFCC feature sequence: X_1', X_2' , … , X_T'
  - Using the same LSTM encode each frame S_1 , S_2 , … , S_T.

Attention mechanism:

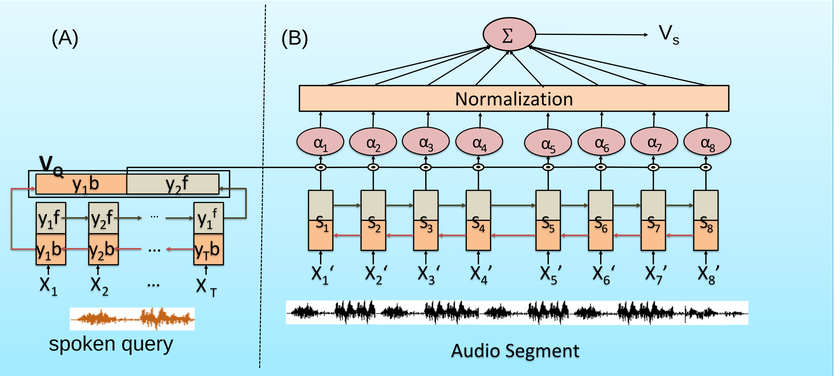
$$\alpha_t = S_t \odot V_Q \qquad \odot\text{: Cosine simiarity}$$

Normalization:

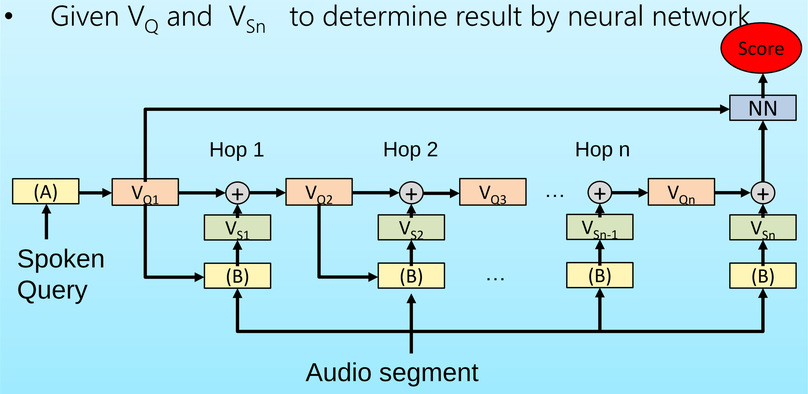
$$\alpha_t' = \frac{\exp(\alpha_t)}{\sum_{t=1}^{T}\exp(\alpha_t)}$$

Audio segment vector:

$$V_S = \sum_{t=1}^{T} \alpha_t' S_t$$



- Hopping
  - Using attention mechanism repeatedly to extract more relative information from audio segment.
- Keyword Detection
  - Given V_Q and V_{Sn} to determine result by neural network



3. Experiments

- Data set : LibriSpeech
- Training set :
  - Query set : 500
  - Query and Audio segment pair : 70,000

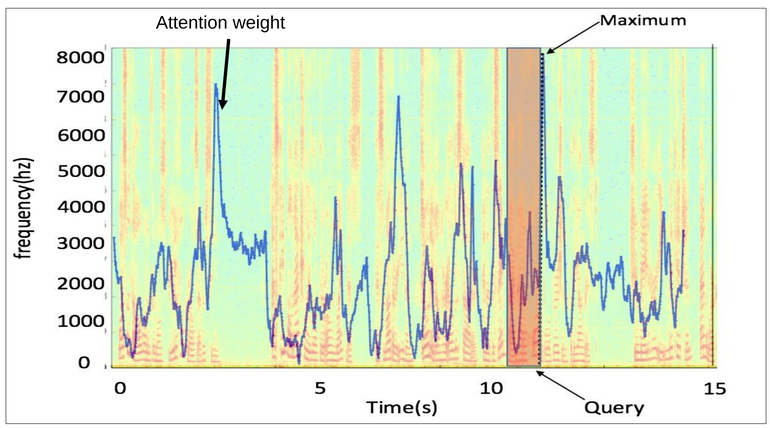
- Testing set 1 :
  - Query acoustic feature from training set.
  - Query set : 30
  - Query and Audio segment pair : 1,500

- Testing set 2 :
  - Query Acoustic feature is different.
  - Query set : 30
  - Query and Audio segment pair : 1,500

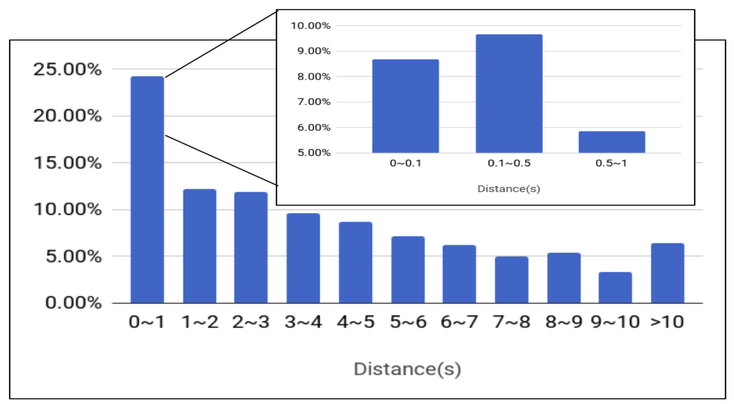
- Testing set 3:
  - Query keyword didn't present in training set
  - Query set : 100
  - Query and Audio segment pair :10,000

- Attention analysis

Blue line: attention value , Red box : query position.



Distance: difference between the position with the highest attention weight and the end of the query



		Test set1	Test set2	Test set 3
Supervised	(a): DTW	0.6173	0.5778	0.5678
	(b): 1-hop	0.6523	0.6246	0.5754
	(c): 2-hop	0.6472	0.6430	0.5842
	(d): 3-hop	0.6676	0.6404	0.5837
	(e): 4-hop	0.6417	0.6476	0.5792
	(f): (a) + (d)	0.6789	0.6430	0.5830
Unsupervised	(e): 1-hop	0.6128	0.5893	0.5548
	(g) 3-hop	0.6141	0.5964	0.5702