

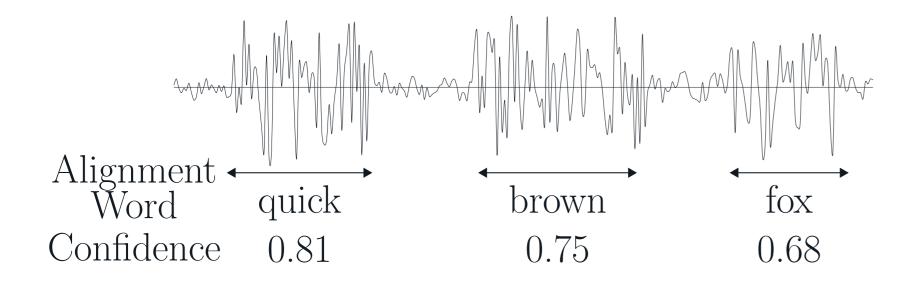
CONFIDENCE SCORES FOR SEQUENCE DATA

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Introduction

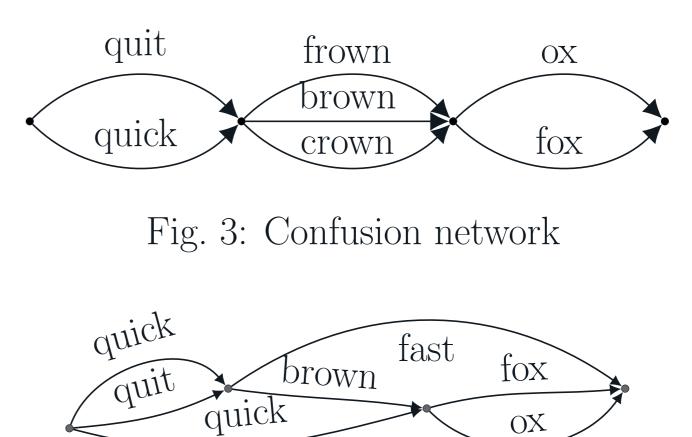
• Automatic speech recognition aims to generate a transcription for a given speech recording.



- Fig. 1: Transcription and confidence score prediction for the phrase: quick brown fox.
- A good confidence score is able to predict errors in a hypothesis transcription.

Representing Hypotheses

- Simplest output from an ASR is a 1-best hypothesis.
- Confusion networks and lattices represent the N-best competing hypotheses efficiently.



• The aim of this research is to improve confidence scores in the context of speech processing.

LatticeRNN for Confidence

- Generalisation of BiLSTM to graph-like structures
- Incoming arcs are merged using attention

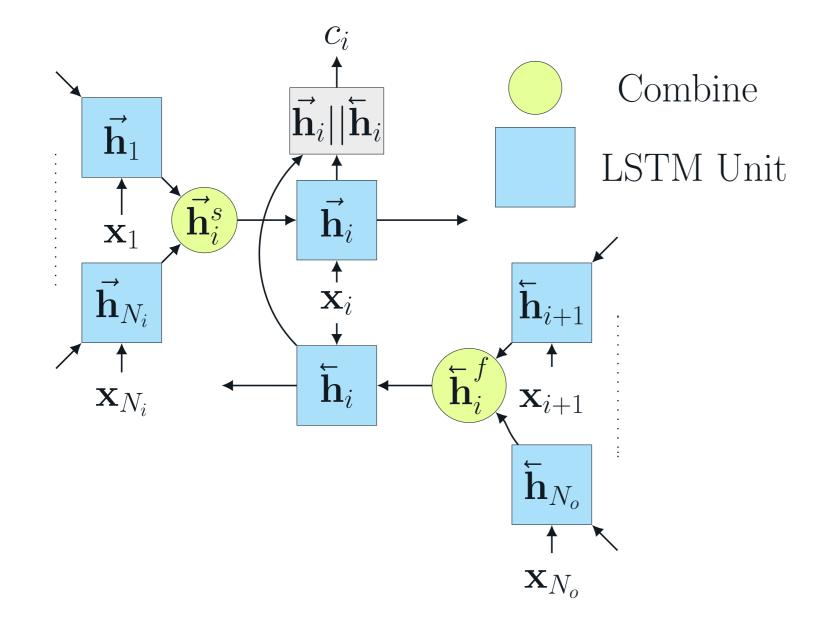


Fig. 2: Bi-Directional LatticeRNN for confidence estimation.

Results



Fig. 4: Word lattice

Sub-word Level Features

- Typically, word-level features such as the word duration (t_i^d) , the language model score (LM_i) , acoustic model score (AM_i) , and word posterior (p_i) are used.
- This research investigates methods for incorporating subword level features - specifically grapheme-level features.

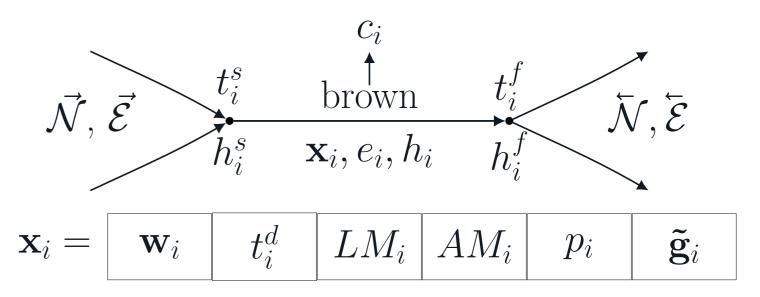


Fig. 5: Lattice edge with word and sub-word level features.

- Fixed representation of grapheme-level features is required.
- A BiGRU with additive attention operates as a grapheme encoder to produce $\tilde{\mathbf{g}}_i$.

Conclusion

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Table 1: Improving confidence score estimate	s for one-	best sequ	ences
Model	NCE	AUC	

Unmapped word posteriors	-0.1978	0.9081
Decision tree mapped word posteriors	0.2755	0.9081
BiLSTM	0.2911	0.9121
BiLSTM + grapheme encoder	0.2978	0.9139

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Table 2:	Improving	confidence	score estimates	tor	contusion	networks
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	One-be	est arcs	All arcs		
Features	NCE	AUC	NCE	AUC	
Word-based	0.2934	0.9201	0.4959	0.8406	
+ grapheme features	0.2998	0.9228	0.4993	0.8432	
+ LM and AM	0.3004	0.9231	0.5013	0.8444	

- Sub-word level features can be incorporated into existing LatticeRNN models using a grapheme encoder.
- The grapheme embedding and grapheme duration are complimentary features for confidence estimation.

References

[1] Qiujia Li et al. "Bi-Directional Lattice Recurrent Neural Networks for Confidence Estimation". In: arXiv preprint arXiv:1810.13024 (2018). [2] Anton Ragni et al. "Confidence Estimation and Deletion Prediction Using Bidirectional Recurrent Neural Networks". In: SLT. 2018.