UNCERTAINTY AND CONFIDENCE SCORES IN SEQUENCE DATA





Fig. 2: A 1-best hypothesis with the predicted word, confidence score, and hidden state h_i

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Lattice representation

• Represents N-best lists in an efficient and compact form. • Additional information from each confusion is considered



Fig. 3: A simple word-marked lattice

Lattice enrichment



Fig. 4: Edge e_i with enriched features

$$h_i = f(h_{\mathcal{N}}) = f(h_{\vec{\mathcal{N}}}, h_{\vec{N}}) \tag{1}$$

$$c_i = f(h_i, \mathbf{x}_i) \quad y_i = f(h_i, \mathbf{x}_i) \tag{2}$$

• Using grapheme-marked lattices allows subword level features to be embedded in the arcs.

• These features include the language model score (LM_i) , acoustic model score (AM_i) , and a fixed length representation of the grapheme information such as the duration (\mathbf{d}_i) .

• Arc combination and aggregation of grapheme information is achieved through attention.



Fig. 5: LatticeRNN model for confidence score prediction [3].



- lattice-based MLLR". In: 2001.
- 2
- IEEE. 2019, pp. 6755–6759.





LatticeRNN

• Bidirectional recurrent architecture which considers the forward and backward probabilities distinctly.

• The existing arc combination procedure can be improved by applying attention over arc neighbourhoods $\vec{\mathcal{N}}$ and $\vec{\mathcal{N}}$ rather than just the incoming and outgoing arcs.

References

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