WFSTs for ASR

Peter Bell

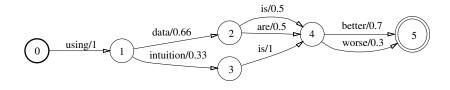
Automatic Speech Recognition – ASR Lecture 9 13 February 2023

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- Weighted finite state automaton that transduces an input sequence to an output sequence (Mohri et al 2008)
- States connected by transitions. Each transition has
 - input label
 - output label
 - weight
- Weights use the *log semi-ring* or *tropical semi-ring* with operations that correspond to multiplication and addition of probabilities
- There is a single start state. Any state can optionally be a final state (with a weight)
- Used by Kaldi

Weighted Finite State Acceptors



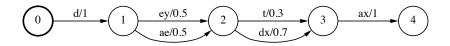
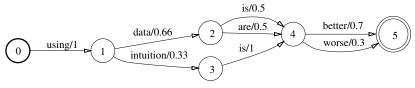


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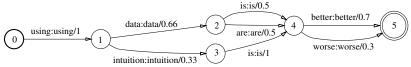
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Weighted Finite State Transducers

Acceptor



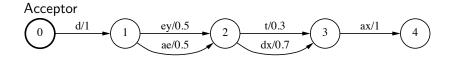
Transducer



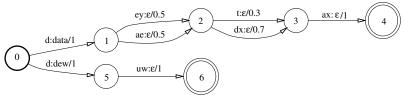
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Weighted Finite State Transducers



Transducer



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A (1) > A (2)

FST Weights

- Formally, WFST weights must be members of a semiring
- This defines special operations for mutiplication ("Times", $\otimes)$ and addition ("Plus", $\oplus)$
- You can think of the weights as negative log-probabilities, so that:

$$w_1 \otimes w_2 = w_1 + w_2$$

 $w_1 \oplus w_2 = -\log(e^{-w_1} + e^{-w_2})$

corresponding to the normal multiplication/addition operations in the probability domain. This is the *log semiring*

• You may also encounter the *tropical semiring* (the default in OpenFst), which is the same as above, except

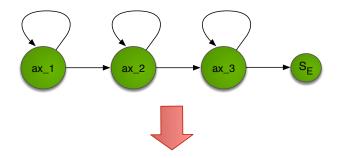
$$w_1 \oplus w_2 = \min(w_1, w_2)$$

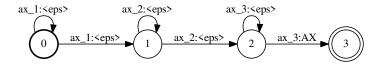
which can be interpreted as taking the best of two probabilities, rather than summing them. Composition Combine transducers T_1 and T_2 into a single transducer acting as if the output of T_1 was passed into T_2 .

Determinisation Ensure that each state has no more than a single output transition for a given input label

Minimisation Transforms a transducer to an equivalent transducer with the fewest possible states and transitions Weight pushing Push the weights towards the front of the path

The HMM as a WFST





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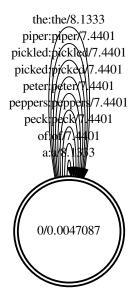
Applying WFSTs to speech recognition

• Represent the following components as WFSTs

| | transducer | input sequence | output sequence |
|---|-----------------------|----------------|-----------------|
| G | word-level grammar | words | words |
| L | pronunciation lexicon | phones | words |
| С | context-dependency | CD phones | phones |
| Н | HMM | HMM states | CD phones |

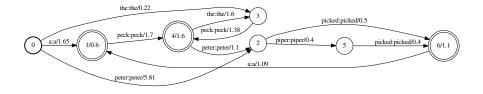
- Composing *L* and *G* results in a transducer *L* \circ *G* that maps a phone sequence to a word sequence
- $H \circ C \circ L \circ G$ results in a transducer that maps from HMM states to a word sequence

Grammar - unigram



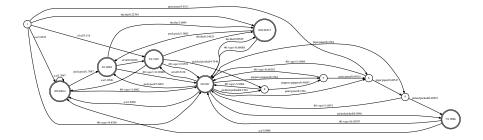
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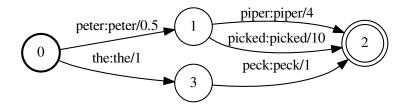


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Bigram with back-off

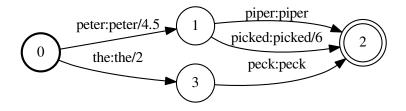


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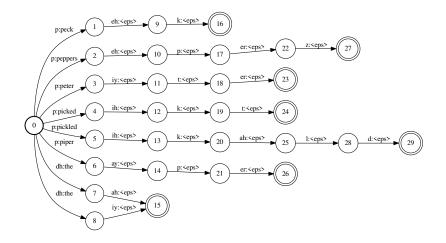
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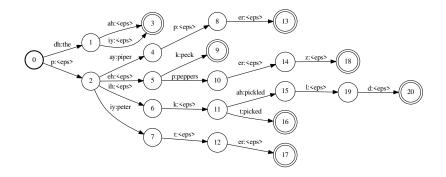


For clarity, this figure omits loops back to the start state

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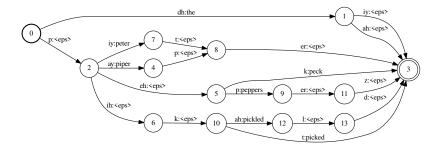
Determinization - det(L)



For clarity, this figure omits loops back to the start state

Image: A math a math

Minimization -min(det(L))

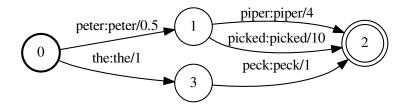


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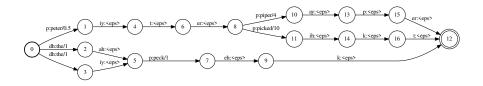
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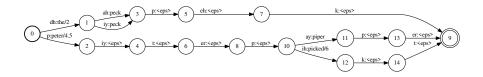
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Composition: $L \circ G$



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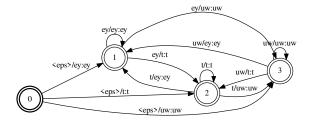
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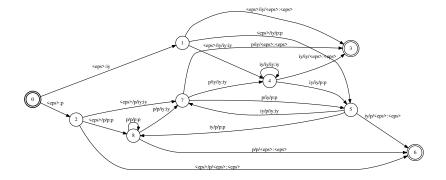
Context-dependency: left biphones



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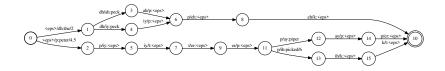
Context-dependency: triphones



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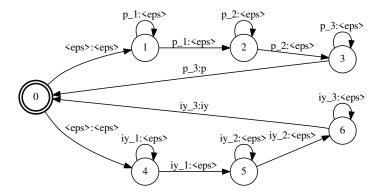
$C \circ L \circ G$ – biphones



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HMM transducer, H



- We can also use a version that outputs context-dependent phones
- *H* can be used to encode state-tying

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- Combining the transducers gives an overall HMM structure for the ASR system – but minimisation and determination operations on the WFSTs means it is much smaller than naively combining the HMMs
- But it is important in which order the algorithms are combined otherwise the transducers may "blow-up"
- standard approach is to determinize and minimize after each composition
- In Kaldi, ignoring one or two details

 $HCLG = \min(\det(H \circ \min(\det(C \circ \min(\det(L \circ G)))))))$

- Mohri et al (2008). "Speech recognition with weighted finite-state transducers." In Springer Handbook of Speech Processing, pp. 559-584. Springer. http://www.cs.nyu.edu/~mohri/pub/hbka.pdf
- WFSTs in Kaldi. http://danielpovey.com/files/Lecture4.pdf