Minimum Hypothesis Phone Error as a Decoding Method for Speech Recognition

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Maximum a Posteriori (MAP)

The standard decoding formula normally used in speech recognition is Maximum A Posteriori (MAP) as follows:

$$W^* = \operatorname{argmax}_W P(W|\mathcal{O})$$

= $\operatorname{argmax}_W P(W) p(\mathcal{O}|W)$

Known limitation

Such a criterion guarantees a sentence that minimizes sentence error can be decoded; however, it is usually the Word Error Rate (WER) not the sentence error used as the evaluation criterion for the recognition system performance. To make up for this mismatch, Minimum Bayes Risk (MBR) criterion is a natural alternative.

Minimum Bayes Risk (MBR)

$$W^* = \operatorname{argmin}_{W_i} \sum_{j=1}^{N} P(W_j | \mathcal{O}) E(W_i | W_j)$$

where $E(W_i|W_j)$ is the number of errors (Levenshtein distance) given W_i as a reference.

Problems

Direct calculating the criterion in a subspace of W(generally represented as word graph/lattice) is prohibitive, thus many approximated strategies are attempted.

Known approaches to WER minimization

- N-best sentence list based decoding scheme proposed by A.Stockle et al.
- Consensus network proposed by L.Mangu et al.
- ► Time-frame word error proposed by F.Wessel et al. al.

MPE/MWE as a criterion for lattice rescoring

We approximate MBR with Minimum Phone Error (MPE) discriminative training criterion as decoding criterion, the approach is

$$W^* = \operatorname{argmax}_W \sum_{W'} P^{\kappa}(W'|\mathcal{O})Acc(W'|W)$$

where $\operatorname{argmax}_{W}$ is taken over an N-best list that we derive from the decoding lattice, and so is $\sum_{W'}$. In other words we find the hypothesis W^* to maximize the objective criterion.

Advantages of the proposed method

- Explicitly optimizing the objective criterion, the correctness of which has been proofed by MPE/MWE discriminative training.
- Conceptually simple and clear, as simplified forward-backward algorithm is performed on the decoding lattice.
- Much flexible on the accuracy criterion Acc(W'|W) calculating, such as on phone error, time-frame phone error, time-frame word error, and word error criteria can be implemented under the same framework.

How large N for a desired W^* ? (1)



Figure: WER versus N on MSRA data (trained with MPE)

How large N for a desired W^* ? (2)



Figure: WER versus N on broadcast data(trained with MAP+MPE)

How large N for a desired W^* ? (3)

As illustrated from the figures, the desired WER can be gained when N is ranged from 20 to 40. Therefore with a very limited N, we can approximate WER minimization. Similar experiments has also been performed on English test data, and we reach the same conclusions.

Recognition system (A) trained with the MLE criterion

| Methods | MSRA (MLE, N=40) | | | | |
|---------|------------------|------|------|--------|--------|
| Methous | #ins | #sub | #del | SER | WER |
| Base | 24 | 2333 | 171 | 95.40% | 26.41% |
| MHPE | 51 | 2319 | 107 | 95.40% | 25.88% |
| Δ | +27 | -12 | -64 | -0.0% | -0.53% |

Table: Baseline vs. MHPE on MSRA test data.

Recognition system (B) trained with the MAP criterion

| Methods | | BD | C (MA | P,N=40) | |
|---------|------|------|-------|---------|--------|
| Methous | #ins | #sub | #del | SER | WER |
| Base | 114 | 7065 | 963 | 98.02% | 33.83% |
| MHPE | 187 | 7014 | 651 | 98.18% | 32.62% |
| Δ | +73 | -51 | -312 | +0.16% | -1.21% |

Table: Baseline vs. MHPE on BDC test data.

Recognition system (C) trained with the MPE criterion

| Methods | | MSF | RA (MP | E, N=40) | |
|---------|------|------|--------|----------|--------|
| Methous | #ins | #sub | #del | SER | WER |
| Base | 26 | 2074 | 183 | 94.60% | 23.85% |
| MHPE | 47 | 2035 | 108 | 93.40% | 22.90% |
| Δ | +21 | -39 | -75 | -1.20% | -0.95% |

Table: Baseline versus MHPE on MSRA test data.

Recognition system (D) trained with MAP+MPE criterion

| Methods | | BD | C (MPE | E,N=40) | |
|---------|------|------|--------|---------|--------|
| Methous | #ins | #sub | #del | SER | WER |
| Base | 109 | 6296 | 959 | 96.45% | 30.60% |
| MHPE | 178 | 6254 | 568 | 96.20% | 29.08% |
| Δ | +69 | -42 | -391 | -0.25% | -1.52% |

Table: Baseline versus MHPE on BDC test data.

MHPE versus Consensus Network on lattice decoding

| l able: | MHPE | versus CN | |
|---------|------|-----------|--|
| | | | |

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| Test sets | Base | CN | MHPE |
|-----------|--------|--------|--------|
| MSRA(MLE) | 26.41% | 25.92% | 25.88% |
| BDC(MAP) | 33.83% | 32.80% | 32.62% |
| MSRA(MPE) | 23.85% | 23.42% | 22.90% |
| BDC(MPE) | 30.60% | 29.41% | 29.08% |

Conclusions and future work

We have introduced a new decoding method for lattice rescoring that aims to get closer to the Minimum Bayes Risk decision rule with respect to the Word Error Rate. Future work will be focused on

- To have a full comparison on other criteria to implement $Acc(W_i, W)$.
- More sophisticated approach will be investigated to take the place of N-best sentence list.
- Based on the proposed criterion, new system combination scheme will be studied, not just Confusion Network Combination.

Thanks !