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Frame Sync. To Phone Sync. **Overview** Frame synchronous Viterbi beam search in CTC Motivation: CTC model shows peaky posterior property and ignoring *blank* frames will not $\mathbf{w}^* =$ introduce additional search errors. w • Approach: A novel *phone synchronous* decoding framework and compact acoustic space representation, CTC lattice are proposed. \cong Experiments & Discussion: Experiments on both English and Mandarin show an extra 2-3 times speed up compared to the traditional $\pi_{1:T} = (\pi_1, \ldots, \pi_T)$ is the frame-wise decoding path frame synchronous CTC implementation. l_w is phone sequence corresponding to w in dictionary $l \in L$ and L is the phone set $\pi \in L'$ and $L' = L \cup \{\texttt{blank}\}$ **ASR Decoding & its Weakness** Frame synchronous to phone synchronous decoding Difference in model granularity → Decoder \mathbf{w}^* • AM, LM, HMM, Lexicon... • Prior arts of decoding Offline WFST based optimization and online viterbi search and beam prune • Variable frame rate (VFR) : from equal interval search to unequal (by feature analysis) • Weakness $U = \{u : y_{plank}^u \simeq 1\}$ is the set of common *blank* time indexes • Huge search space Search errors from pruning is the number of output phone labels J = T - |U| Feature level VFR shows limited improvement Different information rate From HMM to CTC model Acoustic information processing: frame by frame Linguistic information processing: phone by phone From HMM to CTC: do better in sequential modeling Adjustable search interval ► (q(t+1)) →(q(t))-→(q(t-1))-▶ (q(t)) →(q(t+1) ▶(q(t-1)} • WFST search interval is self-adjusted but not equal interval Compared with VFR • CTC model: learn the many-to-one function of \mathcal{B} Frame rate analysis on model rather than feature level \mathcal{B} : $L' \mapsto L$ $\prod y_{\pi_t}^t$ $P(\mathbf{l}|\mathbf{x}) = \sum P(\boldsymbol{\pi}|\mathbf{x}) =$ $L' = L \cup \{\texttt{blank}\}$ $\pi \in \mathcal{B}^{-1}(\mathbf{l})$ Analysis on Search Space Compression $\lambda = \frac{1}{N} \sum_{n=1}^{N} \frac{\#\{U^{(n)}\}}{T^{(n)}}$ peaky distribution and concentrated information output Network Traversal Reduction λ is the average of blank frame percentages of test utterances Theoretical Compression Rate $R = 1 - (1 - \lambda) \times \beta$ Waveform β is the percentage of active phones with respect to all phones for a given set of test utterances R is the overall measure of the search space compression yielded by PSD CTC

time axis ->

Phone Synchronous Decoding with CTC Lattice

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$$\operatorname{argmax}\{P(\mathbf{w})p(\mathbf{x}|\mathbf{w})\} = \operatorname{argmax}\{P(\mathbf{w})p(\mathbf{x}|\mathbf{l}_{\mathbf{w}})\}$$

$$= \operatorname{argmax}_{\mathbf{w}} \left\{ P(\mathbf{w}) \max_{\mathbf{l}_{\mathbf{w}}} \frac{P(\mathbf{l}_{\mathbf{w}} | \mathbf{x})}{P(\mathbf{l}_{\mathbf{w}})} \right\}$$

$$\operatorname{argmax}_{\mathbf{w}} \left\{ P(\mathbf{w}) \max_{\pi:\pi\in L', \mathcal{B}(\pi_{1:T})=\mathbf{l_w}} \frac{1}{P(\mathbf{l_w})} \prod_{t=1} y_{\pi_t}^t \right\}$$

$$\cong \underset{\mathbf{w}}{\operatorname{argmax}} \left\{ P(\mathbf{w}) \max_{\substack{\pi:\pi \in L', \mathcal{B}(\pi_{1:T}) = \mathbf{l}_{\mathbf{w}}}} \frac{1}{P(\mathbf{l}_{\mathbf{w}})} \right\}$$
$$= \underset{\mathbf{w}}{\operatorname{argmax}} \left\{ P(\mathbf{w}) \max_{\substack{\pi':\pi' \in L, \mathcal{B}(\pi'_{1:J}) = \mathbf{l}_{\mathbf{w}}}} \frac{1}{P(\mathbf{l}_{\mathbf{w}})} \prod_{j=1}^{J} y_{\pi'_{j}}^{t_{j}} \right\}$$

Experimental Setup

- Training stage

- 2-3M parameters
- Test stage

Task

Switchboa

CellPhone

- similar
- state-HMM

Decoding Speed-up



•3X speed-ups with no CER deterioration (similar speedup rate in CD-phone-CTC in our recent work)

paper

Experiments

English: SWB 300h 3-gram LM from SWB without interpolation • Mandarin: 300h & 5000h 3-gram LM (1.7GB with 118K words) • Procedure similar to *EESEN* (miao et al. 2015)

• On Intel(R) Xeon(R) CPU E5-2690 v2 @ 3.00GHz.

• Hub5e00 testset from Switchboard and a Mandarin testset, CellPhone, is used, which is recorded in several speech scenarios and with about 25 hours

Baseline CER/WER & RTF performance

	Context	Acoustic	CER /	DTE
	Dependency	Model	WER	NIF
ard	CD	dnn-hmm	18.3	0.27
	CI	lstm-ctc	20.7	0.044
2	CD	dnn-hmm	13.30	0.32
	CI	lstm-ctc	10.20	0.044

With 300 hours, CI-phone-CTC and CD-state-HMM are

With 5000 hours, CI-phone-CTC outperforms CD-

• CTC is faster than HMM by 7 times

Search Space Compression

All gotten by force-aligned CTC paths

testset	$\lambda(\%)$	$\beta(\%)$	R(%)
Switchboard	88	5	99.4
CellPhone	87	11	98.6

 Phone synchronous decoding remaining 10% network traversal in WFST search

 CTC lattice remaining 1% acoustic information from acoustic posterior distribution

model	search step	CER	RTF
HMM	frame	13.3	0.32
СТС	frame	10.2	0.044 (7.3X)
CIC	phone	10.1	0.016 (20X)

•Result of English corpus is similar and listed in our

- Speed robustness





- LM size $\uparrow \rightarrow CER \downarrow$
- linguistic search space

CTC lattice was proposed

- sources



• Extendibility of more complex linguistic search space • LM size $\uparrow \rightarrow$ linguistic search space \uparrow • Active Tokens $\uparrow \rightarrow \mathsf{RTF} \uparrow \rightarrow \mathsf{Speed} \downarrow$ LM Size • Extendibility: CTC PSD > CTC FSD >> DNN FSD CER____ CTC PSD CER ··· * · · DNN CER ··· * · CTC FSD CER --- CTC 2-Pass LM Size (MB) 10 100 1000 10000

• CTC PSD is suitable for combining with complex

Conclusions

Frame synchronous decoding was transformed into phone synchronous decoding Self-adjusting decoding interval Model level variable frame rate Removing tremendous search redundancy

Extremely compact acoustic information preserver Extensibility of combining with other knowledge