Multilingual acoustic word embedding models for processing zero-resource languages

ICASSP 2020

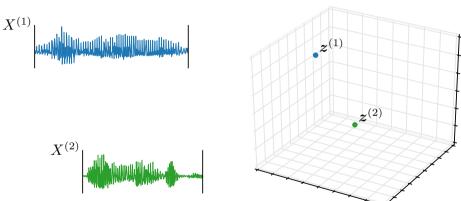
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Background: Why acoustic word embeddings?

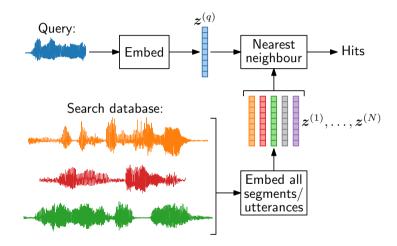
- Current speech recognition methods require large labelled data sets
- Zero-resource speech processing aims to develop methods that can discover linguistic structure from unlabelled speech [Dunbar et al., ASRU'17]
- Example applications: Unsupervised term discovery, query-by-example
- Problem: Need to compare speech segments of variable duration

Acoustic word embeddings



Embedding space with $\boldsymbol{z} \in \mathbb{R}^M$

Example application: Query-by-example search



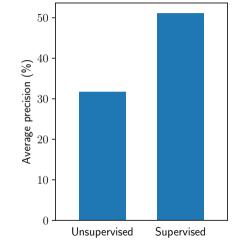
[Levin et al., ICASSP'15]

Supervised and unsupervised acoustic embeddings

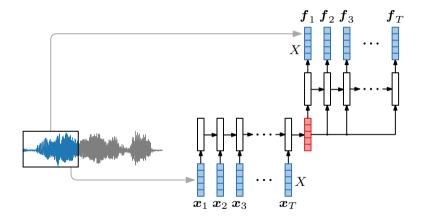
- Growing body of work on acoustic word embeddings
- Supervised and unsupervised methods
- Unsupervised methods can be applied in zero-resource settings
- But there is still a large performance gap

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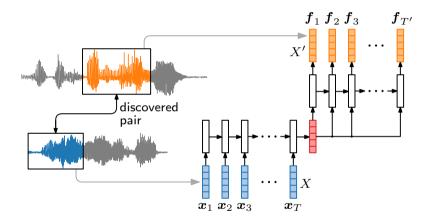


Unsupervised monolingual acoustic word embeddings



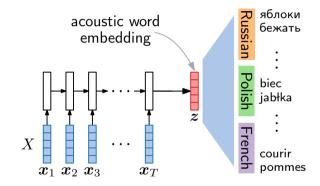
[Chung et al., Interspeech'16; Kamper, ICASSP'19]

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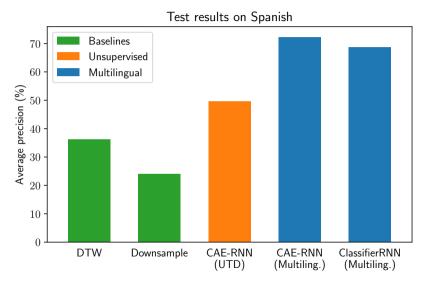
Supervised multilingual acoustic word embeddings



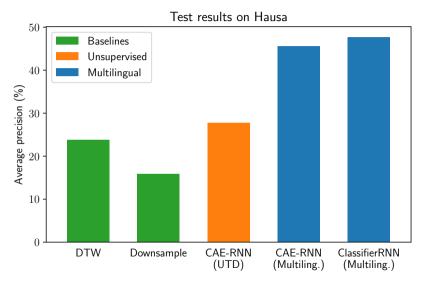
Experimental setup

- Training data: Six well-resourced languages Czech (CS), French (FR), Polish (PL), Portuguese (PT), Russian (RU), Thai (TH)
- Test data: Six languages treated as zero-resource Spanish (ES), Hausa (HA), Croatian (HR), Swedish (SV), Turkish (TR), Mandarin (ZH)
- Evaluation: Same-different isolated word discrimination
- Embeddings: M = 130 for all models
- Baselines:
 - Downsampling: 10 equally spaced MFCCs flattened
 - Dynamic time warping (DTW) alignment cost between test segments

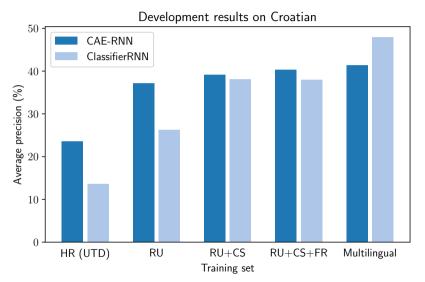
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2. Does training on more languages help?



3. Is the choice of training language important?

	ES	Ev: HA	aluatior HR	n langua SV	age TR	ZH
CS -	41.6	51.1	41.0	28.7	37.0	42.6
FR -	42.6	41.8	30.4	25.3	32.5	35.8
anguag - Td	41.1	43.7	35.8	25.5	33.7	39.5
- Taining language - TA	45.9	46.2	36.4	26.6	34.1	39.6
⊢ RU -	35.0	39.7	31.3	22.3	29.7	37.1
TH -	28.5	44.5	29.9	17.9	23.6	36.2

Conclusions and future work

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- Proposed to train a supervised multilingual acoustic word embedding model on well-resourced languages and then apply to zero-resource languages
- Multilingual CAE-RNN and ClassifierRNN consistently outperform unsupervised models trained on zero-resource languages

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Future work:

- Different models both for multilingual and unsupervised training
- Analysis to understand the difference between CAE-RNN and ClassifierRNN
- Does language conditioning help during decoding?

https://arxiv.org/abs/2002.02109

https://github.com/kamperh/globalphone_awe