Learning Visual Voice Activity Detection with an Automatically Annotated Dataset

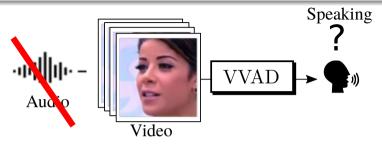
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Inria Grenoble Rhône-Alpes and Univ. Grenoble Alpes, France LTCI, Télécom Paris, Institut polytechnique de Paris, France Andalusian Research Institute in Data Science and Computational Intelligence (DaSCI), University of Granada, Spain.

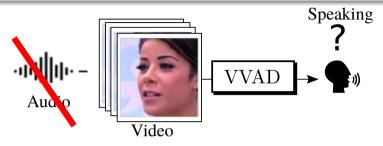


ICPR 2021

Visual Voice Activity Detection(VVAD)



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Why do we need VVAD?



(c) Audio unavailable



(d) Noisy Audio

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Learning VVAD



(f) Silent examples

Figure: MVAD dataset.



(a) Speaking examples

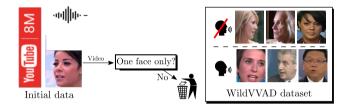
(b) Silent examples

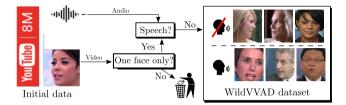
Figure: CUAVE dataset.

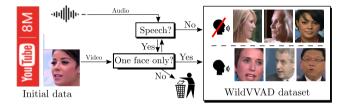
Existing datasets are too simple and too constrained.

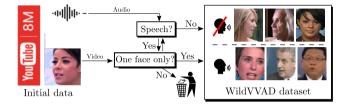










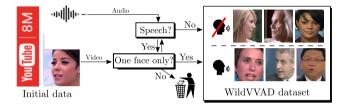




(a) Speaking examples



(b) Silent examples



WildVVAD:

- 13000 videos
- High diversity
- Manually cleaned test set
- Percentage of mislabeled speaking and silent videos are of 12% and 8.6%, respectively.

Proposed architectures

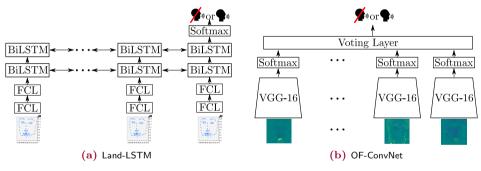


Figure: Architectures of the two proposed models.

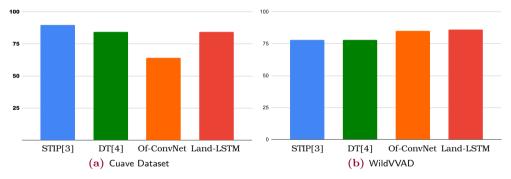
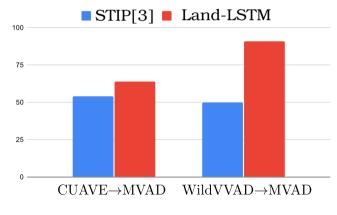


Figure: Experimental evaluation.

Two questions:

- Which method has better generalization features?
- Which is the best suited dataset to learn a general purpose VVAD model?



Contributions

• We propose a method for automatically collecting a dataset for VVAD.



(a) Speaking examples



(b) Silent examples

- We introduce and compare two deep architectures for VVAD
- We show a better generalization ability of VVAD models when they are trained on our dataset.