

## Abstract

Describing soundscapes in sentences allows better understanding of the acoustic scene than a single label indicating the acoustic scene class or a set of audio tags indicating the sound events active in the audio clip. We study how much the collection of audio captions can be guided by the instructions given in the annotation task, by analysing the possible bias introduced by auxiliary information provided in the annotation process. In this work, we address the diversity obtained when collecting descriptions of soundscapes using crowdsourcing, studying how differently annotators describe the same soundscape. We also release a new dataset of audio captions and audio tags produced by multiple annotators, publicly available, which we call MACS<sup>1</sup>.

#### MACS

Multi-Annotator Captioned Soundscapes. Audio from TAU Urban 2019 development dataset.

- Scenes: Airport, public square and park.
- Files: total of 3930 audios, each one 10 seconds long.
- Annotators: 133 students, assigned into 30 groups, each file anno ent annotators.

## **Diversity, bias and similarity metrics**

### Lexical diversity

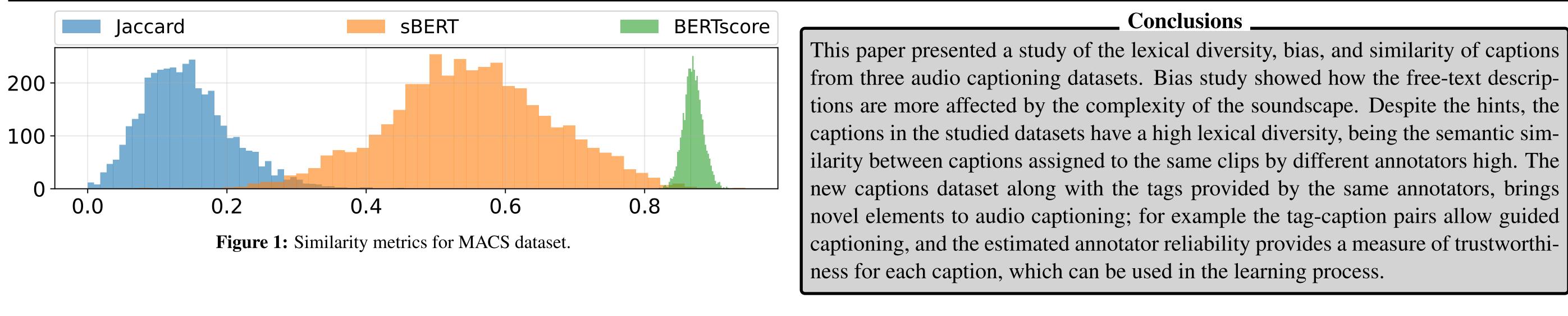
Measured with Type-Token ratio, often used in measuring language acquisition i a second language.

> #Tokens TTR = -

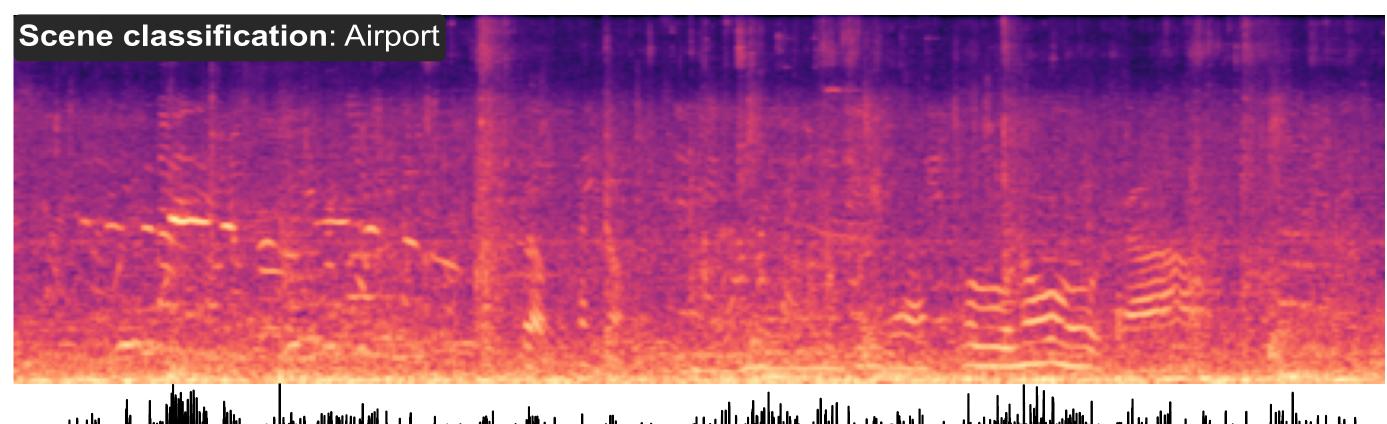
It ranges from a theoretical 0 (infinite repetition of a single word) and 1 (no rep considering the whole dataset (overall) and descriptions belonging to the same ite

S	L	AudioCaps	Clotho		MACS	
		overall	overall	local	overall	local
_	_	1.09%	1.30%	56.52%	1.80%	69.37%
-	$\checkmark$	0.79%		52.08%		
$\checkmark$	$\checkmark$	1.27%	1.66%	60.43%	2.17%	71.02%

Table 2: Global and local lexical diversity of captions. S: removal of stopwords; L: lemmatization. • AudioCaps  $\rightarrow$  hints are the tags associated to the clip in AudioSet. AudioCaps has only a single caption per clip, thus we do not calculate local lexical diversity for it.



# Sound event envelope estimation in polyphonic mixtures Irene Martín-Morató, Annamaria Mesaros Computing Sciences, Tampere University, Finland



			1 .1		
n Acoustic Scenes		Audio tagging	Aud		
	A1	music adults talking	a person whistli		
	A2	adults talking	people are talkir		
	A3	footsteps adults talking	whistling and si		
notated by 5 differ-					
			Most used words		
in infants or learners of	• MA(	<b>CS</b> : talk, people, adult, noise and	bird.		
	• AudioCaps: man, speak, follow, talk and engine.				
	• Clotho: bird, water, background, chirp and someone.				
	Vocabu	lary bias			
epetition at all). Results tem (local):	Defined as the proportion of hinted sounds with respect to the nuttion. AudioSet vocabulary, a total of 722 labels, is used to identify				
			Tag bias (std) Word bi		

	lag blas (std)	word b
AudioCaps	0.33 (0.35)	0.35 (
MACS	0.38 (0.36)	0.49 (

• MACS  $\rightarrow$  hints are the 10 pre-defined tags

### Similarity

To study how similar are the descriptions by different annotators from the same item, we use metrics from machine translation, calculated for every pair of captions. Basic approaches to compare similarity



- udio captioning
- ling and singing
- ing whistling and singing
- singing many people talking

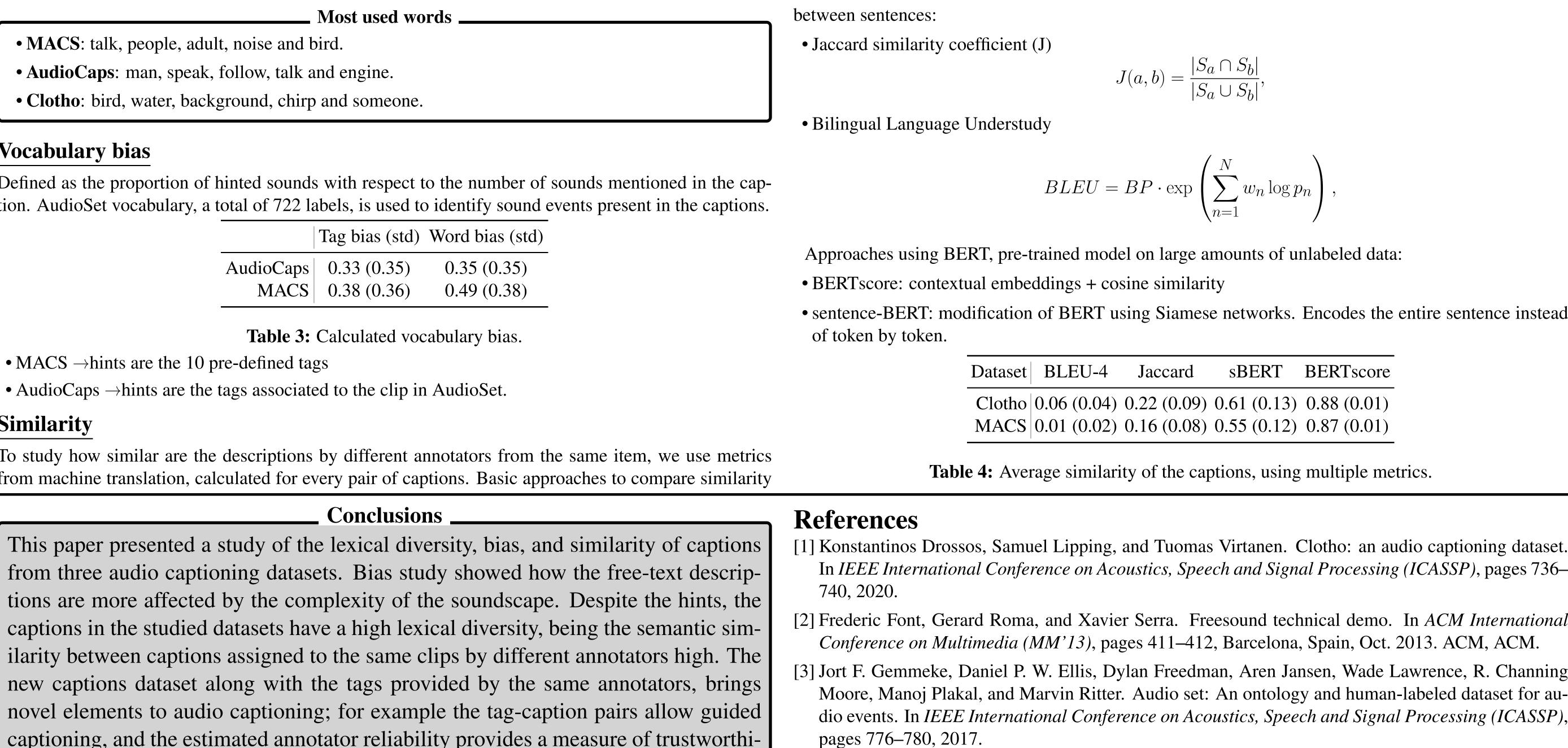
### **Other datasets**

AudioCaps [4] is a collection of sentence-long descriptions for a subset of AudioSet [3]. The video was provided to be played if necessary, and the AudioSet tags were presented to the annotator as hints. The dataset contains over 46k files of 10 seconds each, and one caption per file, collected using MTurk.

**Clotho** [1] is collected using MTurk and contains five captions per clip, for audio clips 15 to 30 seconds long that were collected from Freesound [2]. We consider this dataset as having no bias, since the captions are based solely on the audio clip provided, and no additional information regarding the possible active sounds or clip content was available to annotators.

				Sentence length (std)
AudioCaps	57188	5218	52198	9.17 (4.27)
Clotho	5929	4373		11.34 (2.78)
MACS	3930	2775	16262	9.46 (3.89)

**Table 1:** Statistics of the studied datasets.



[4] Chris Dongjoo Kim, Byeongchang Kim, Hyunmin Lee, and Gunhee Kim. AudioCaps: Generating captions for audios in the wild. In Proceedings of the 2019 Conference of the NAACL HLT, Vol. 1, pages 119–132, Minneapolis, Minnesota, June 2019. Association for Computational Linguistics.

# **J** Tampere University

$$) = \frac{|S_a \cap S_b|}{|S_a \cup S_b|},$$

$$\cdot \exp\left(\sum_{n=1}^N w_n \log p_n\right),$$

• sentence-BERT: modification of BERT using Siamese networks. Encodes the entire sentence instead

ccard	sBERT	BERTscore
	0.61 (0.13) 0.55 (0.12)	

Table 4: Average similarity of the captions, using multiple metrics.

In IEEE International Conference on Acoustics, Speech and Signal Processing (ICASSP), pages 736–

[2] Frederic Font, Gerard Roma, and Xavier Serra. Freesound technical demo. In ACM International Conference on Multimedia (MM'13), pages 411–412, Barcelona, Spain, Oct. 2013. ACM, ACM.

[3] Jort F. Gemmeke, Daniel P. W. Ellis, Dylan Freedman, Aren Jansen, Wade Lawrence, R. Channing Moore, Manoj Plakal, and Marvin Ritter. Audio set: An ontology and human-labeled dataset for audio events. In IEEE International Conference on Acoustics, Speech and Signal Processing (ICASSP),