

Department of Information Engineering and Computer Science

Introduction

- Valence indicates the emotional value (on a scale from positive to negative) associated with an event, situation or object.
- Automated prediction of valence may provide crucial information for mental healthcare
- Role of context in the task of valence prediction?

The ULM state of mind dataset (USoMS)

- Personal narratives with self-reported affect information
- Participants recount two negative and two positive events from their life for 5 mins
- Self assessment of valence using 10 point Likert scale, at five stages
- Scores grouped into three classes: Low (0-4), Medium (5-7) and High (8-10)



Data collection process: N1, N2 are Negative nar N4 are positive narratives; while At0-4 are self re affect scores

- A part of USoMS was used in the INTERSPEECI Computational Paralinguistics Challenge (ComPa
 - Task: automatically predicting self-reported at
 - Input: 8 seconds speech fragment
 - Acoustic features

Motivation

- Context provided by the full history of the:
 - Current narrative
 - All previous narratives by the same individual
- ComParE challenge does not allow exploring con
- Generalizability: utilizing solely the textual inform

Modeling user context for valence prediction from narratives

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Methodology

time

• Incorporate context :

- Feature Engineering
- ML and DNN architectures



DNN based sequence tagging architecture

ratives; N3,	Are previous				<u>Discuss</u>	sion				
eported	narratives	N ₁ -	die	schlimmsten	also eiger	ntlich der	tag	an		
	recounted by an	N 2 -	abireise	und ich	wollte	es	ich	habe u		
H 2018 arE)	individual useful	Nat	mit	zufrieden		noch da.				
ffect	for predicting	N3 +	mit	zutrieden	war	nach der	n abi	weils		
	his/her current	N ₄ +	war	uhm als	es	drum	ging	das		
	mental state?		Dis	tribution of	fattentio	n weights	on four ((fragmen		
	• Senti	ment carry	ving words	s and phrase	es (e.g. h	appy " <i>zuf</i>	rieden")			
 Emotion triggering concepts 										
	• people (e.g. grandfather "opa")									
l ntext	• events (e.g. high school exam " <i>abi</i> ", trip after the high school exam									
	• places (e.g. university "uni")									
mation	• Distiuencies ("uhm", "ahm")									

Experiments and Results

	Model	Narratives used	Features	Accuracy				
		N _t	μ word emb	55.5 ±5.0				
nce class	Linear SVM	N _t	μ word emb, pol	57.8 ±4.8				
layer		N _t	tf-idf, pol	57.8 ±4.5				
ence		N_{t-1} , N_{t}	tf-idf, pol	59.7 ±5.9				
19901	biRNN + attn	N _t	word emb, pol	58.2 ±6.8				
tive ings ion	Encoder (biRNN + attn) + RNN (context pair)	N _{t-1} , N _t	word emb, pol	62.4 ±8.7				
Bi-directional GRU	Encoder (biRNN + attn) + RNN (sequence tagging)	N ₀ ,, N _t	word emb, pol	61.8 ±6.4				
coder) bedding	 Previous context seems to be helpful for valence prediction 							
layer	 In general Neural models outperform Linear SVM 							

nts of) consecutive narratives









