CSCW 2020

Exploring Antecedents and Consequences of Toxicity in Online Discussions: A Case Study on Reddit

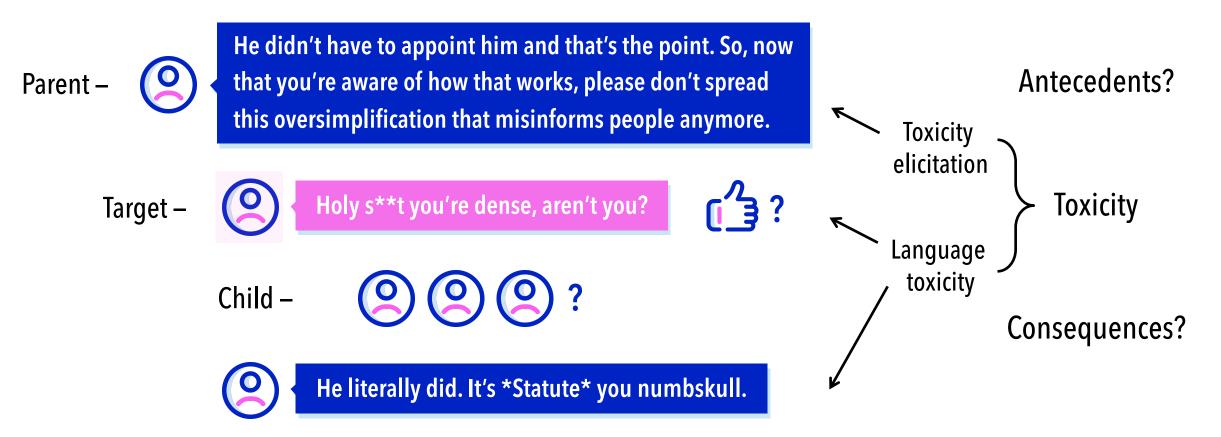
Yan Xia Fudan University Haiyi Zhu Carnegie Mellon University **Tun Lu** Fudan University

Peng Zhang Fudan University **Ning Gu** Fudan University

Icons made by Freepik from Flaticon.com

Motivation





Related Work & Hypotheses

- Antecedents of language toxicity
 - H1: Author's propensity toward toxicity will increase the language toxicity of his/her text.
 - H2: Author's experience in the community will <u>reduce</u> the language toxicity of his/her text.
 - H3: **Toxicity in discussion context** will <u>increase</u> the language toxicity of text.
 - H4: Polarity in discussion context will increase the language toxicity of text.
- Consequences of language toxicity
 - Q1: How will the language toxicity of text influence the volume of discussion?
 - Q2: How will the language toxicity of text influence the **evaluation of discussion**?
- Antecedents and consequences of toxicity elicitation
 - Q3: How will the antecedents and consequences of toxicity elicitation differ from those of language toxicity?

Method

- Reddit comments
 - r/announcements, r/worldnews, r/politics, r/todayilearned, r/AskReddit
 - 19,682/152,632/200,635/139,353/265,893 comments => preprocessed
- Quantifying comment features: NLP tools
 - Toxicity: Perspective API (* with human validation)
 - Polarity: TextBlob library
- Regression analyses
 - Study I, Antecedents of Toxicity: What author/parent comment features predict target comment toxicity?
 - Study II, Consequences of Toxicity: What child comment features are predicted by target comment toxicity?



Study I: Antecedents of Toxicity

		Dependent Variable: Target-language-toxicity				-toxicity	
Factor (Expected Correlation)	Independent Variable	Regression Coefficient in r/annc r/wn r/pol r/til r/ar					
Author's propensity (+)	Target-author-toxicity	.207 **	.133 ***	.125 ***	.103 ***	.160 ***	
Author's experience (-)	Target-author-age Target-author-karma	029 031	055 * .008	029 .004	035 .030	.024 003	
Toxicity in context (+)	Parent-language-toxicity	.210 **	.182 ***	.156 ***	.184 ***	.246 ***	
Polarity in context (+)	Parent-positivity Parent-negativity	.135 * .017	000 038	.005 020	.000 027	018 014	

H1: Author's propensity toward toxicity will increase the language toxicity of his/her text.



H2: **Author's experience in the community** will <u>reduce</u> the language toxicity of his/her text.

H3: **Toxicity in discussion context** will <u>increase</u> the language toxicity of text.

Supported

H4: **Polarity in discussion context** will <u>increase</u> the language toxicity of text.



Study I: Antecedents of Toxicity

		Depen	-toxicity-el	-elicitation			
Factor (Expected Correlation)	Independent Variable	Regression Coefficient in r/annc r/wn r/pol r/til r/ar					
Author's propensity (?)	Target-author-toxicity	.011	.068 **	.024	.033	.041 *	
Author's experience (?)	Target-author-age Target-author-karma	.013 032	.010 .039	.012 008	.005 .006	.013 019	
Toxicity in context (?)	Parent-language-toxicity	.173 *	.118 ***	.082 ***	.151 ***	.136 ***	
Polarity in context (?)	Parent-positivity Parent-negativity	.052 .049	.029 .025	021 .007	031 037	016 .007	
	Control Variable						
/	Target-language-toxicity	.260 ***	.113 ***	.115 ***	.168 ***	.142 ***	

Q3: How will the antecedents (and consequences) of **toxicity elicitation** differ from those of language toxicity?



Factor (Expected Correlation): Volume of Discussion (?)								
Dependent	Independent	Regression Coefficient in						
Variable	Variable	r/annc	r/wn	r/pol	r/til	r/ar		
Target-#children	Target-language-toxicity	.055	.065 **	.056 **	.068 **	.052 **		
	Target-toxicity-elicitation	128	027	008	009	.019		
Target-#descendants	Target-language-toxicity	.005	.083 ***	.113 ***	.107 ***	.080 ***		
	Target-toxicity-elicitation	144 **	055 **	.012	.021	.032 *		
Target-height	Target-language-toxicity	040	.022	.026	.049 *	.009		
	Target-toxicity-elicitation	.016	024	.001	.048 *	.044 *		

Q1: How will the language toxicity of text influence the **volume of discussion**?



Factor (Expected Correlation): Volume of Discussion (?)								
Dependent	Independent	Regression Coefficient in						
Variable	Variable	r/annc	r/wn	r/pol	r/til	r/ar		
Target-#children	Target-language-toxicity	.055	.065 **	.056 **	.068 **	.052 **		
	Target-toxicity-elicitation	128	027	008	009	.019		
Target-#descendants	Target-language-toxicity	.005	.083 ***	.113 ***	.107 ***	.080 ***		
	Target-toxicity-elicitation	144 **	055 **	.012	.021	.032 *		
Target-height	Target-language-toxicity	040	.022	.026	.049 *	.009		
	Target-toxicity-elicitation	.016	024	.001	.048 *	.044 *		

Q3: How will the (antecedents and) consequences of **toxicity elicitation** differ from those of language toxicity?



Factor (Expected Correlation): Evaluation of Discussion (?)								
Dependent	Independent	Regression Coefficient in						
Variable	Variable	r/annc	r/wn	r/pol	r/til	r/ar		
Target-score	Target-language-toxicity	.019	.011	.023	.025	.034 *		
	Target-toxicity-elicitation	050	018	002	016	.014		
Children-score-max	Target-language-toxicity	.015	.018	.028 *	.012	.049 **		
	Target-toxicity-elicitation	029	021	.009	025	.062 ***		

Q2: How will the language toxicity of text influence the **evaluation of discussion**?



Factor (Expected Correlation): Evaluation of Discussion (?)								
Dependent	Independent	Regression Coefficient in						
Variable	Variable	r/annc	r/wn	r/pol	r/til	r/ar		
Target-score	Target-language-toxicity	.019	.011	.023	.025	.034 *		
	Target-toxicity-elicitation	050	018	002	016	.014		
Children-score-max	Target-language-toxicity	.015	.018	.028 *	.012	.049 **		
	Target-toxicity-elicitation	029	021	.009	025	.062 ***		

Q3: How will the (antecedents and) consequences of **toxicity elicitation** differ from those of language toxicity?

Discussion & Design Implications

- From within and without: Triggers of toxicity
 - Toxicity in discussion context
 - <u>Design implication</u>: To interfere with this toxicity generation process
- Flames in disguise: **Toxicity elicitation**
 - Strong-toned / Sarcastic / Digressive / Against common sense or values
 - <u>Design implication</u>: To regulate toxicity-eliciting comments
- The multi-faced devil: **Complexity of toxicity**
 - Different target / emotion / intention => Different consequence
 - <u>Design implication</u>: To distinguish different types of toxic comments in detection and regulation

Limitations & Future Work

- Bias of NLP tools
- Specific time, community and regulation settings
- Limited modeling of toxicity dynamics
- Further questions:
 - Other ways to vitalize a discussion without toxicity?
 - How will the resulted discussions differ?
 - How to stop toxicity from offending people while retaining the "edge" of discussion?

Thanks to the anonymous reviewers and Ge Gao, Diyi Yang, Dakuo Wang, Beisi Zhou, Xiaofeng Zhao for helping us with the study.

Thank you for listening!

2020.09.23