

LREC-COLING 2024

So Hateful! Building a Multi-Label Hate Speech Annotated Arabic Dataset

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Outlines

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- Our Contribution
- Data Annotation Guidelines
- Results
- Limitation
- Conclusion

Background

Social media revolutionized peoples thoughts, opinions, and experiences

Information is shared through social media spread easily

However, the downside is that negative information and opinions are also easily spread

Background

Hate speech is one negative form of expression that is prevalent on social media

Hate speech often arises with the emergence of events around the world

Hateful content is known to spread faster than other content on social media

- No more Hate Speech
- The German government secured an agreement from social media platforms, to **delete all hate speech** targeting refugees within **24 hours** of its occurrence on the platform



Motivation for this Research



Identifying and removing hate speech from social media is challenging in Arabic Language due to the diverse nature



The language has various dialects that differ from each other and from Modern Standard Arabic



Various studies explored different approaches to annotating and detecting hate speech on Arabic Twitter



These studies collectively underscore the importance of developing hate speech detection within the challenges of informal dialectal social media posts

Our Contribution



Created the largest multi-label, fine-grained Arabic hate speech dataset to date



Our dataset is unique and versatile, with each tweet annotated with nine labels, such as sentiments, emotions, and valence, etc.



Documented the dataset's collection and guidelines and reproduceable for future projects

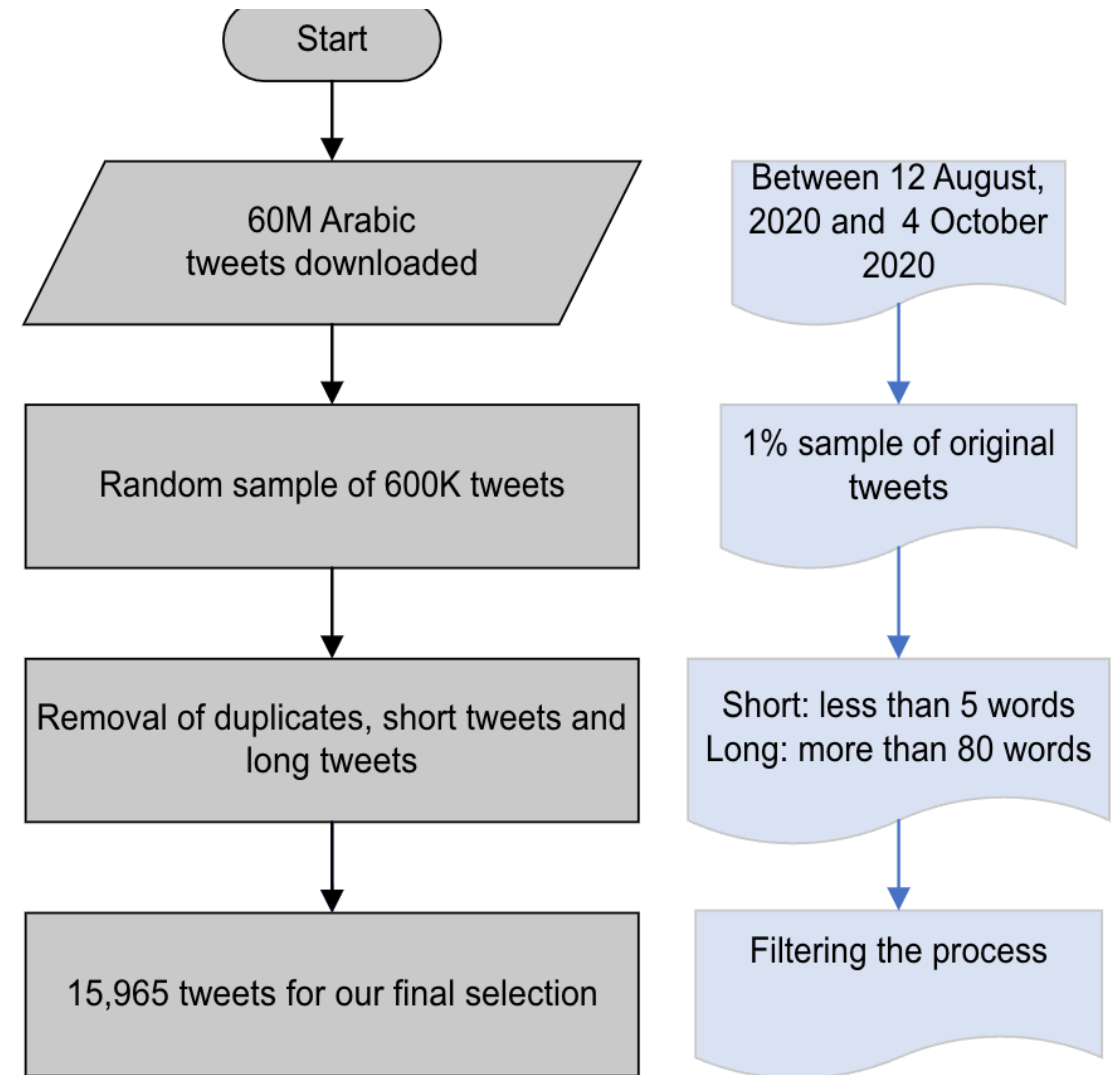


Comprehensive corpus analysis of the dataset on the distinct features of Arabic hate speech discourse



Carried out experiments with several classification techniques

Data Collection



Data annotation

- We chose multi-label dataset and focused not only on offensive discourse and hate speech
- Rather, we asked the annotators to label the tweets for the 13 categories



Data annotation categories

Categories	Subcategories
Q1. Emotions	choosing from 12 options like anger, anticipation, sadness etc. or neutral
Q2. Emotion intensity	no, small, moderate or large amount
Q3. Sentiment	very positive to very negative or neutral/mixed
Q4. Offensive content	whether tweet contains offensive language and if directed at a target
Q4.1 Hate speech type	individual, group, other entity
Q4.2 Hate speech target	race, religion, ideology, gender, social class
Q4.3 Vulgarity	whether offensive tweet contains profanity
Q4.4 Violence	if offensive tweet promotes violence
Q5. Effect	whether tweet is positive/inspiring or negative
Q6. Sarcasm/irony	whether directed at a target
Q7. Humor	not funny, somewhat funny or very funny
Q8. Factuality	if tweet contains verifiable information and is important
Q9. Spam	annoying advertising or requests

Annotators and training



Location: Middle East and North Africa



Language: Arabic



Tweet Evaluation: Initially 16 Annotators



Tweet Annotation: 1 to 5 annotators per tweet based on their ability to understand the dialect



Training: Every two-three weeks trained to understand of guidelines, procedures, complex concepts of Arabic tweet



Revision and Production

- Revision: Manager
 - analyzed errors,
 - unresolved cases,
 - feedback
 - updated guidelinesto maximize quality, consistency, and consensus in annotation decisions

Production: Annotators met regularly but usually worked independently

Annotation Interface

- MicroMappers is an online annotation management tool
- A screenshot from the Arabic version of the annotation interface (Showing the first three questions)

(Q1) السؤال الأول: تحديد المشاعر

قم باختيار مشاعر كاتب التغريدة (اختيار متعدد) [اقرأ المزيد...](#)

الثقة	التفجع	الاستمزاز	الخوف	الفرح	الحب	التفاؤل	التشاؤم	الحزن	الفجأة	الغلة
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(Q2) السؤال الثاني: تحديد قوة المشاعر

ما هي قوة المشاعر في التغريدة (اختيار وحيد) [اقرأ المزيد...](#)

لا مشاعر واضحة	كمية بسيطة من المشاعر	كمية متوسطة من المشاعر	كمية كبيرة من المشاعر
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(Q3) السؤال الثالث: تحديد الحالة العقلية للمستخدم

ما هي الحالة العقلية أو التقسيمية في التغريدة (اختيار وحيد) [اقرأ المزيد...](#)

إيجابية للغاية	إلى حد ما إيجابية	محايدة أو مزيج من حالات مختلفة	إلى حد ما سلبية	سلبية للغاية
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Annotation Guidelines

Category	Guidelines	Example
Emotions	Annotators selected emotions expressed from 12 options: neutral, anger, anticipation, disgust, fear, joy, love, optimism, pessimism, sadness, surprise, trust.	أشعر بالإحباط وخيبة الأمل من هذا الوضع "I'm frustrated and disappointed with this situation." - Labels: Anger, Pessimism
Emotion Intensity	Annotators judged the intensity of emotions in the tweet as: none, small, moderate or large.	أنا قلقة بعض الشيء بشأن امتحاني غدا "I'm a little worried about my exam tomorrow." - Label: Small amount

Category	Guidelines	Example
Emotions	Annotators selected emotions expressed from 12 options: neutral, anger, anticipation, disgust, fear, joy, love, optimism, pessimism, sadness, surprise, trust.	أشعر بالإحباط وخيبة الأمل من هذا الوضع "I'm frustrated and disappointed with this situation." - Labels: Anger, Pessimism
Emotion Intensity	Annotators judged the intensity of emotions in the tweet as: none, small, moderate, or large.	أنا قلقة بعض الشيء بشأن امتحاني غدا "I'm a little worried about my exam tomorrow." - Label: Small amount
Sentiment	Annotators labeled overall sentiment of the tweet as: very positive, somewhat positive, neutral/mixed, somewhat negative, very negative.	أنا قلقة بعض الشيء بشأن امتحاني غدا "I'm a little worried about my exam tomorrow." - Label: Small amount
Offensive Language	Annotators noted if tweet contains offensive language and if it's directed at a target.	أنت حقير ومثير للاشمئزاز. "You are despicable and loathsome." - Labels: Yes offensive, Yes directed
Hate Speech Type	Annotators identified the group targeted by hate speech: race, religion, ideology, gender, social class.	هؤلاء الرياضيون قمامة لا قيمة لهم "Those athletes are worthless trash." - Label: Group
Hate Speech Target	If offensive, annotators specified if hate speech targets an individual, group, or other entity.	كل الرجال حثالة ويجب أن يلقوا في السجن "All men are scum and should be thrown in jail." - Label: Gender
Vulgarity	Annotators marked if offensive tweet contains profanity	وأنت مال أمك يا ابن الوخنة "And you are your mother's money, you son of a bitch" - Label: Yes profanity
Violence	Annotators noted if offensive tweet promotes violence.	أيها الرئيس، اقتل كل المعارضين "Boss, kill all opponents." - Label: Yes violence
Effect	Annotators judged if tweet has a positive/inspiring or negative effect.	ابقي رأسك مرفوعا. المستقبل مشرق "Keep your head up. The future is bright." - Label: Positive effect
Sarcasm Irony	Annotators identified if tweet contains sarcasm/irony directed at a target.	اتفضل قول يا أذكي إخوانك! "You're so incredibly smart and talented!" - Label: Yes sarcasm
Humor	Annotators marked if tweet contains humor: not funny, somewhat funny, very funny.	اتفضل قول يا أذكي إخوانك! "You're so incredibly smart and talented!" - Label: Yes sarcasm
Factuality	Annotators noted if tweet contains verifiable information and if it's important.	سيتم إعلان نتائج الانتخابات اليوم "The election results will be announced today." - Labels: Verifiable information, Important
Spam	Annotators identified if tweet contains spam like advertising or requests.	بحاجة الى المال على وجه السرعة. الرجاء المساعدة من خلال التبرع على هذا الموقع "Need money urgently. Please help by donating on this site." - Label: Spam

Annotation Guidelines



Annotation Analysis

	N	%
Offensive language?		
Yes	2793	17.5%
No	13171	82.5%
Directed?		
Yes, directed	2348	84.7%
No, not directed	445	15.93%
Target2		
Individual	963	41.01%
People with common features	1090	46.42%
Organization, company, situation, or topic	299	13.9%
What do they have in common?		
Ideology, politics, sports	747	68.5%
Class, social status, or profession	83	7.6%
Religion or sect	144	13.3%
Gender	43	3.9%
Origin, race, or nationality	382	3.5%
Obscene language?		
Yes	874	31.29%
No	1919	69.71%
Advocates for violence?		
Yes	201	7.1%
No	2592	92.8%



Sentiment Results from Manual Annotation

Effect of the Tweet on the Reader's Wellbeing/Annotator's Sentiment	N	%
Frustrating	4237	26.54
Motivating	2637	15.51
Neither frustrating nor motivating	9089	56.93




Fact-Checking

Fact-Checking	N	%
No information	10190	63.83%
Contains information, but not verifiable	2987	18.71%
Contains information that is verifiable	2787	17.45%
Important to the public?		
Yes, important	1594	57.19%
No, not important	981	35.2%



Data Annotation Categories

Data Annotation Categories	N
Emotion (Pessimism,sadness,confidence, joy and others)	12301
Emotion Intensity (Small, large, average amount of feelings and others)	9075
Sarcasm, irony, and ridicule (yes, no)	9036
Humor, joking (Yes, but not funny and others)	9257
Spam Detection (Yes, specific product or service, and others)	9004
Valence (positive, negative. Neutral, and others)	9059



Annotation Evaluation

Label	Cohen's Kappa
Emotions	0.4396
Emotion intensity	0.5632
Sentiment	0.9289
Offensive content	0.8863
Hate speech type	0.7664
Hate speech target	0.8972
Vulgarity	0.9024
Violence	0.7304
Effect	0.4896
Sarcasm/irony	0.6377
Humor	0.7010
Factuality	0.8545
Spam	0.9499
Overall	0.7497

Example of Annotation Disagreement

Example tweet	Annotator Disagreement
<p>الله يبارك فيك حبيبي ي احمد عقبال عندك يارب 🍀❤️❤️ God bless you [in response to “congratulations”], my dear Ahmed. I wish you the same 🍀❤️❤️</p>	<p>One annotator opted for a "neutral" emotional classification, and the two others selected "joy#optimism#confidence" and "optimism#confidence".</p>

Offensive Language Detection Using Machine and Deep Learning

Model	Label	Precision	Recall	F1-Score	Accuracy
LR	Yes	0.58	0.34	0.43	0.84
	No	0.87	0.95	0.91	
RF	Yes	0.70	0.25	0.37	0.85
	No	0.86	0.98	0.91	
GB	Yes	0.81	0.14	0.23	0.84
	No	0.84	0.99	0.91	
SVM	Yes	0.49	0.40	0.44	0.82
	No	0.87	0.91	0.89	
DT	Yes	0.42	0.38	0.40	0.79
	No	0.87	0.88	0.88	
Ara-bert	All	0.49	1.00	0.65	0.82

HateSpeech Detection Using Machine and Deep Learning

Model	Label	Precision	Recall	F1-Score	Accuracy
LR	Yes	0.38	0.14	0.20	0.93
	No	0.94	0.98	0.96	
RF	Yes	0.47	0.05	0.09	0.93
	No	0.93	1.00	0.96	
GB	Yes	0.41	0.04	0.07	0.93
	No	0.93	1.00	0.96	
SVM	Yes	0.27	0.24	0.25	0.90
	No	0.94	0.95	0.95	
DT	Yes	0.22	0.19	0.21	0.90
	No	0.94	0.95	0.95	
Arabert	All	0.57	0.47	0.66	0.83



Limitations

The annotators originated from certain Arabic-speaking regions. This could introduce labeling biases based on regional dialects and interpretations of hate speech.

Hate speech involves inherent subjectivity which may have impacted the annotation accuracy

As Twitter was the sole data source, the findings might not reflect other social media platforms

Potential sampling bias may occur as we randomly sampled on our dataset

Majority were non-hate class which causes class imbalance, inflate model performance metrics

Conclusion

Multiple annotators from various Arabic-speaking regions ensure diversity on dataset annotation

Improvement and enhancement of the dataset is ongoing as well as refinement of the guidelines and annotations

This project serves as a valuable resource for researchers and practitioners in the field of Arabic language processing and analysis

Thank you
Q & A

