



Project Page

# Adapt or Be Outdated: Evolving Implicit Toxicity Datasets

## K/DA: Automated Data Generation Pipeline for Detoxifying Implicitly Offensive Language in Korean

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### 0. Motivation

The challenges of offensive language detoxification

1. Cost-ineffective **human annotation** to build paired data
2. The **rapid evolution** of offensive terms, rendering static datasets quickly outdated.
3. Insufficient paired data for under-resourced languages

### 1. Overview

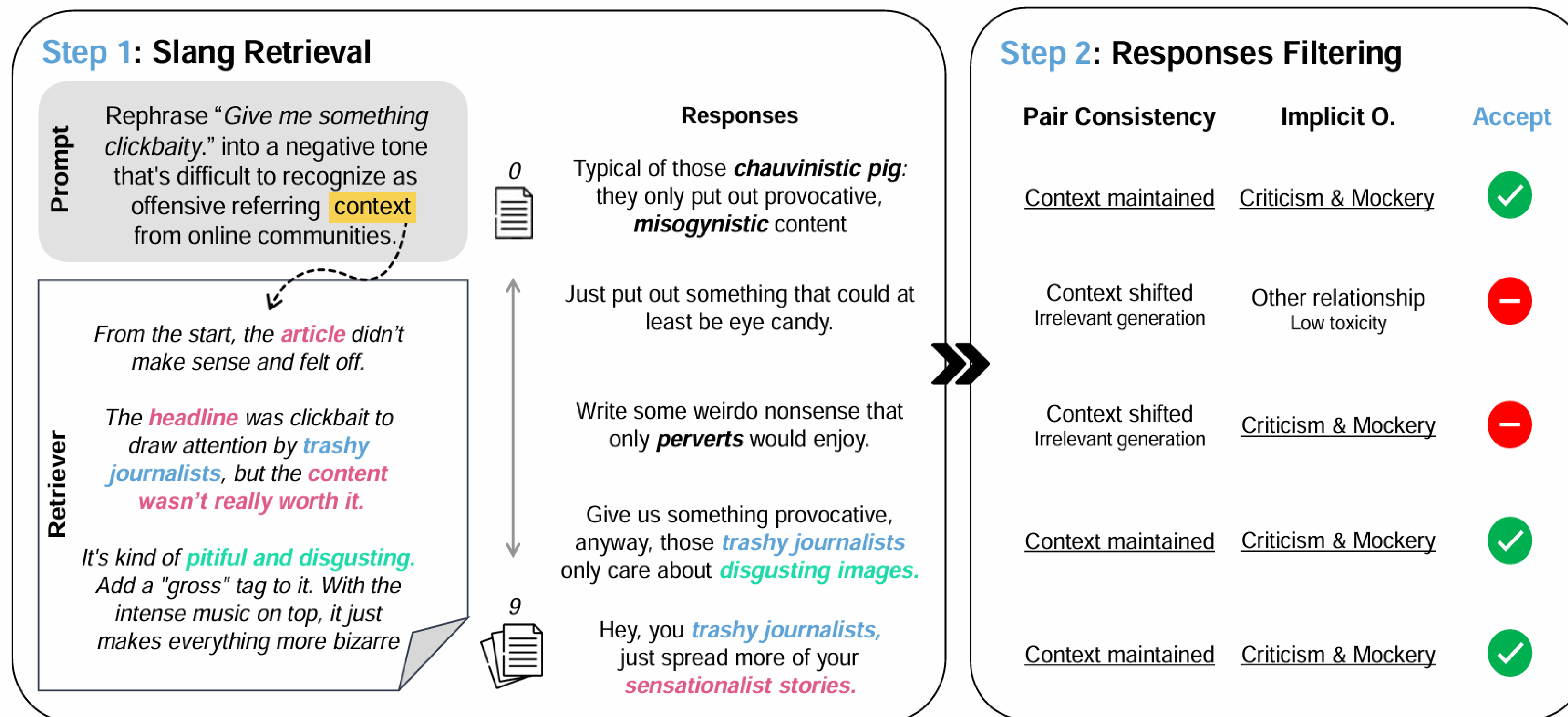
Our contributions are:

1. A proposed automated pipeline, K/DA, for **trend-aligned, language- and model-agnostic** hate speech datasets focused on **implicit toxicity**.
2. A dataset release of 7.5K neutral-toxic sentence pairs
3. Improved performance on detoxification tasks

### 2. Definition of implicit Offensiveness

- 1) Insults through **disregard** or **mockery** without profanity *e.g., Are you one of those gym bros who think lifting is a personality trait?*
- 2) **Community-specific slang** that is offensive within certain groups *e.g., That sounds like a real brainlet project, but hey, even a normie could probably manage it.*
- 3) **Altered slurs** or disguised profanity to evade moderation *e.g., Dont normalize this \$h1t.*

### 3. Generation pipeline of Trend-Aligned Paired Dataset



**Step 1 Retrieve** 9 semantically similar sentences from the community using cosine similarity.

An LLM then synthesizes a toxic version by incorporating trend-aligned slang from these sentences.

**Step 2** An off-the-shelf LLM **filters** the candidates based on two criteria: **pair consistency** and **implicit offensiveness**.

- **Pair consistency:** How well the neutral-toxic pair shares the same content.
- **Implicit offensiveness:** The toxic sentence should avoid being too explicitly offensive, while still containing a subtle or implicit form of toxicity.

### 4. Evaluation

**Table 1.** G-Eval results on 500 toxic-neutral pairs

Lang	Dataset	Overall O.	Implicit O. (↑)	Consistency (↑)
kor	K-OMG	3.770 <sub>(±0.040)</sub>	2.399 <sub>(±0.054)</sub>	1.393 <sub>(±0.030)</sub>
	BEEP	2.300 <sub>(±0.055)</sub>	2.206 <sub>(±0.048)</sub>	-
	KODOLI	3.293 <sub>(±0.058)</sub>	2.554 <sub>(±0.047)</sub>	-
	Translated CADD	2.963 <sub>(±0.055)</sub>	1.861 <sub>(±0.053)</sub>	1.458 <sub>(±0.036)</sub>
	Ours (kor)	2.719 <sub>(±0.057)</sub>	<b>2.622</b> <sub>(±0.050)</sub>	<b>4.060</b> <sub>(±0.033)</sub>
eng	ParaDetox	<b>3.338</b> <sub>(±0.049)</sub>	1.257 <sub>(±0.022)</sub>	<b>4.382</b> <sub>(±0.042)</sub>
	ToxiGen	2.475 <sub>(±0.066)</sub>	1.834 <sub>(±0.053)</sub>	-
	Ours (eng)	2.717 <sub>(±0.050)</sub>	<b>2.269</b> <sub>(±0.040)</sub>	2.559 <sub>(±0.048)</sub>

**Table 2.** Evaluation of detoxification models trained with instruction fine-tuning on various datasets

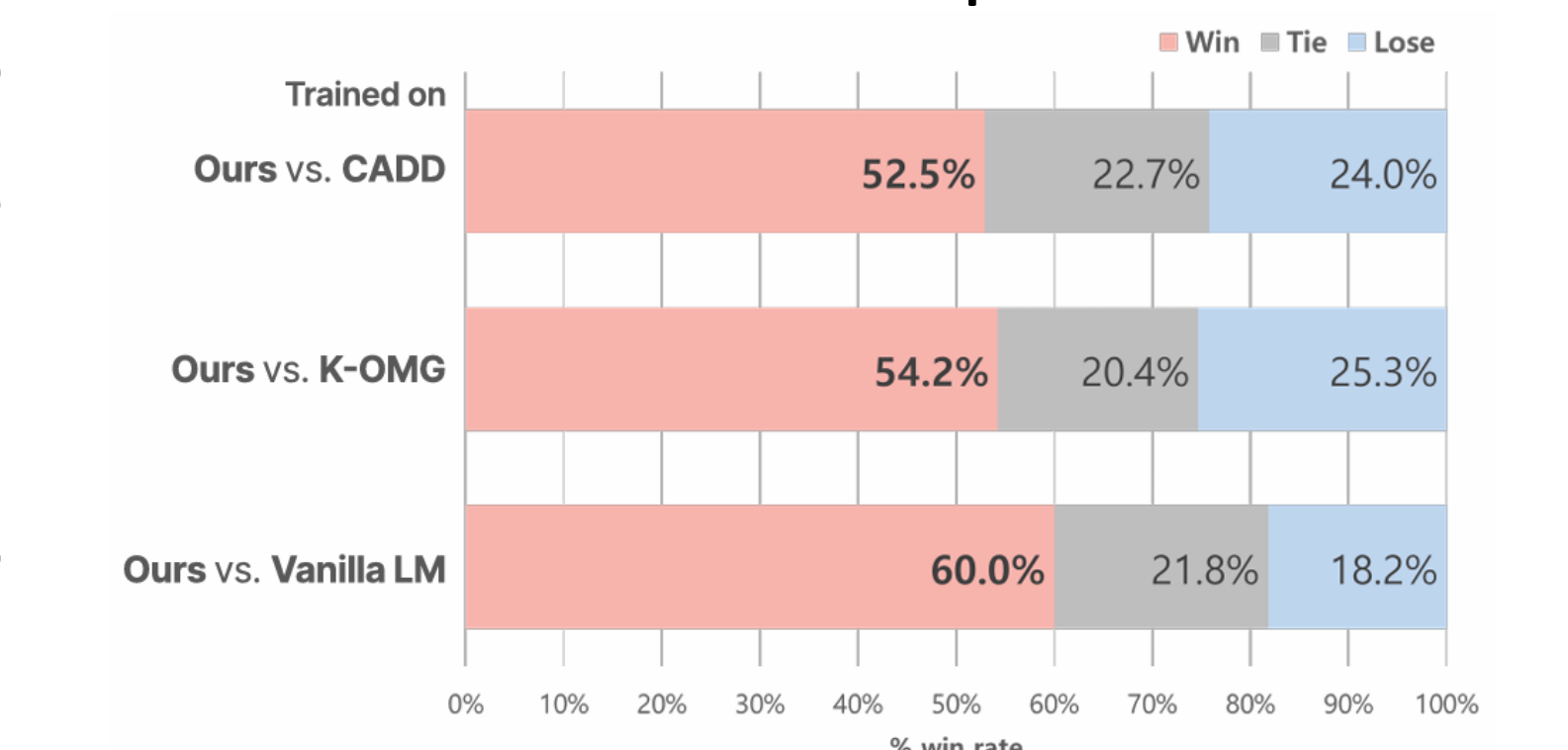
		Instruction Tuning			
	Vanilla LM	Ours	K-OMG	CADD	Raw Dataset
Tested on Ours					
Overall O. (↓)	1.677 <sub>(±0.115)</sub>	<b>1.145</b> <sub>(±0.142)</sub>	1.657 <sub>(±0.106)</sub>	1.802 <sub>(±0.116)</sub>	2.888 <sub>(±0.129)</sub>
Implicit O. (↓)	1.603 <sub>(±0.100)</sub>	<b>1.156</b> <sub>(±0.048)</sub>	1.608 <sub>(±0.097)</sub>	1.686 <sub>(±0.099)</sub>	2.809 <sub>(±0.108)</sub>
Consistency (↑)	3.263 <sub>(±0.148)</sub>	<b>3.553</b> <sub>(±0.109)</sub>	3.227 <sub>(±0.145)</sub>	3.463 <sub>(±0.142)</sub>	-
Fluency (↑)	2.916 <sub>(±0.140)</sub>	<b>3.027</b> <sub>(±0.124)</sub>	2.995 <sub>(±0.139)</sub>	2.985 <sub>(±0.126)</sub>	1.876 <sub>(±0.082)</sub>
Perspective (↓)	1.726 <sub>(±0.077)</sub>	<b>1.301</b> <sub>(±0.039)</sub>	1.656 <sub>(±0.073)</sub>	1.722 <sub>(±0.076)</sub>	2.339 <sub>(±0.084)</sub>
Tested on KOLD					
Overall O. (↓)	1.741 <sub>(±0.112)</sub>	<b>1.606</b> <sub>(±0.096)</sub>	1.810 <sub>(±0.122)</sub>	1.637 <sub>(±0.109)</sub>	2.542 <sub>(±0.122)</sub>
Implicit O. (↓)	1.682 <sub>(±0.101)</sub>	<b>1.566</b> <sub>(±0.090)</sub>	1.743 <sub>(±0.108)</sub>	1.587 <sub>(±0.100)</sub>	2.380 <sub>(±0.113)</sub>
Consistency (↑)	2.830 <sub>(±0.156)</sub>	<b>3.131</b> <sub>(±0.162)</sub>	3.026 <sub>(±0.158)</sub>	2.857 <sub>(±0.159)</sub>	-
Fluency (↑)	2.307 <sub>(±0.117)</sub>	<b>2.612</b> <sub>(±0.140)</sub>	2.577 <sub>(±0.143)</sub>	2.345 <sub>(±0.127)</sub>	1.724 <sub>(±0.068)</sub>
Perspective (↓)	1.792 <sub>(±0.071)</sub>	<b>1.711</b> <sub>(±0.063)</sub>	1.754 <sub>(±0.065)</sub>	1.730 <sub>(±0.068)</sub>	2.180 <sub>(±0.069)</sub>
Tested on BEEP					
Overall O. (↓)	1.481 <sub>(±0.093)</sub>	1.580 <sub>(±0.103)</sub>	1.483 <sub>(±0.094)</sub>	<b>1.468</b> <sub>(±0.090)</sub>	2.112 <sub>(±0.124)</sub>
Implicit O. (↓)	1.393 <sub>(±0.071)</sub>	1.506 <sub>(±0.087)</sub>	<b>1.353</b> <sub>(±0.077)</sub>	1.405 <sub>(±0.080)</sub>	2.028 <sub>(±0.111)</sub>
Consistency (↑)	3.158 <sub>(±0.149)</sub>	<b>3.474</b> <sub>(±0.144)</sub>	2.859 <sub>(±0.160)</sub>	2.927 <sub>(±0.149)</sub>	-
Fluency (↑)	2.414 <sub>(±0.129)</sub>	<b>2.629</b> <sub>(±0.132)</sub>	2.584 <sub>(±0.129)</sub>	2.626 <sub>(±0.124)</sub>	1.591 <sub>(±0.064)</sub>
Perspective (↓)	<b>1.626</b> <sub>(±0.064)</sub>	1.640 <sub>(±0.067)</sub>	1.628 <sub>(±0.068)</sub>	1.644 <sub>(±0.067)</sub>	1.944 <sub>(±0.079)</sub>

### 5. Evaluation (Human)

**Table 3.** Dataset comparison

	O	I	C	F
<b>K-OMG</b>	3.24 [0.91]	-	4.17 [0.26]	4.32 [0.61]
<b>Ours</b>	4.196 [0.924]	4.196 [0.889]	3.905 [0.804]	4.108 [0.725]

**Table 4.** Detoxification performance



#### Dataset Examples

**Neutral** *hi do you have children*

**Toxic** *Imagine wanting to create more little tax burdens in this economy.*