SemEval 2023 Task 9: Multilingual Tweet Intimacy Analysis

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Introduction:

In this task, we are going to predict **textual intimacy of multilingual tweets**.

Dataset:

This dataset is **multilingual** text data which was collected from **twitter**. The tweets are sampled from **2018 to 2022**.

The dataset is called **MINT** (Multilingual INTimacy dataset) which covers tweets in **6** languages as the **training** data, including **English**, **Spanish**, **French**, **Portuguese**, **Italian**, and **Chinese**, covering major languages used in The Americas, Europe, and Asia.

A total of **12,000 tweets** are annotated for **6 languages**. For testing the **generalization** of model under **zero-shot** setting, they also annotated small **test sets** for **Dutch**, **Korean**, **Hindi**, and **Arabic** (**500 tweets for each** of those).

For collecting this dataset, they use the *lang_id* in the tweet object to select **English** and **Chinese**. For **other** languages they use **fastText for language identification** and assign labels when the **model confidence** is **larger than 0.8**. All the mentions of **unverified users** are replaced with a special token "@user" during pre-processing to remove noise from random and very infrequent usernames.

For annotating tweets they recruited some **annotators**. They set a "**first language**" requirement during annotator pre-screening.

Intimacy is annotated using a **5-point Likert Scale** where **1** indicates "**not intimate at all**" and **5** indicates "**very intimate**". For each language in the training set, they collected annotations for 2,000 tweets. **Each tweet was annotated by 7 annotators and each**

annotator was shown 50 tweets. Also the annotators were balanced by sex and they were from 73 unique counties and regions.

For post-processing they remove annotations from users who failed the attention test. They also remove potential noise in the crowdsourcing setting, similar to trimmed mean. They removed one highest score and one lowest score for tweets with at least five labels. Then they kept the tweets with at least two valid scores.

For external test languages (i.e. Dutch, Hindi, Korean, and Arabic), they only kept tweets with a relatively low label diversity (i.e. standard deviation lower than 1) to ensure a good golden test set for the zero-shot setting. The final intimacy score is calculated as the mean score of all the remaining labels for each tweet.

The final dataset includes 13,384 tweets annotated with the textual intimacy score.

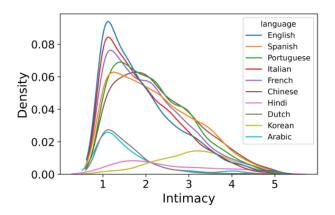


Figure 1: The distribution of intimacy scores for each language

To verify the quality of the annotations, they conduct a split-half-reliability test (SHR) which randomly splits labels into two groups and calculates the Pearson correlation between the aggregated scores from the two groups. All the SHR scores are above 0.63 with an average of 0.68. Suggesting that the final aggregated scores are reliable. Then split all dataset (13,384 tweets) into training, validation, and test sets following a ratio of 7:1:2.

| Language | α | SHR | Amount |
|------------|----------|------|--------|
| English | 0.48 | 0.69 | 1,984 |
| Spanish | 0.52 | 0.72 | 1,991 |
| Portuguese | 0.45 | 0.66 | 1,996 |
| Italian | 0.43 | 0.63 | 1,916 |
| French | 0.47 | 0.67 | 1,981 |
| Chinese | 0.44 | 0.64 | 1,996 |
| Hindi | 0.61 | 0.68 | 280 |
| Korean | 0.53 | 0.67 | 411 |
| Dutch | 0.48 | 0.68 | 413 |
| Arabic | 0.58 | 0.74 | 416 |

Table 2: Statistics for the annotated dataset

Since the dataset is collected from Twitter, it is **conversational** and **informal**.

Since recognizing the intimacy of text can serve as an important benchmark to test the ability of computational models to understand social information, we are interested in this project.

Since the dataset was **ready**, we **don't use any API** for collecting it, we just download it and use it.

Since annotating textual intimacy is **challenging** because of the **subjective nature of intimacy perception** and **potential individual rating bias**, we have limitations for collecting more data.

Examples for analysis:

Work work work, 1.0, English:

• We agree with this example. Because it's not intimate at all.

• We also agree with this example. It's a bit more intimate with respect to the previous example.

he wants to have a partyyy 😭 😭 😭 ,2.0,English

• We agree with this. It's a bit more intimate with respect to the previous example.

@user @user Come and join! @user @user @user,3.0,English:

• We disagree. It's not so intimate.

i would love to be dead rn,4.0,English:

• We disagree. It's not very intimate.

You'll miss me when I'm gone,3.333333333333333,English

• We agree. It's quite intimate.

@user @user Love you ♥ ♥,4.0,English:

• We agree. It's very intimate.

I loved u,4.2,English

• We agree. It's very intimate.

need a kiss ,4.75,English:

• We agree. It's very intimate.

@user I think I fell in love with you,4.8,English

• We agree.