A Details on Preparation for Style Datasets in XSLUE

Formality. Appropriately choosing the right formality in the situation is the key aspect for effective communication (Heylighen and Dewaele, 1999). We use GYAFC dataset (Rao and Tetreault, 2018) that includes both formal and informal text collected from the web.⁷

Humor. Humor (or joke) is a social style to make the conversation more smooth or make a break (Rodrigo and de Oliveira; Kiddon and Brun, 2011). We use the two datasets widely used in humor detection: ShortHumor (CrowdTruth, 2016) and ShortJoke (Moudgil, 2017), where both are scraped from several websites. We randomly sample negative (i.e., non-humorous) sentences from random text from Reddit corpus (Jung et al., 2019) and literal text from Reddit corpus (Khodak et al., 2017).

Politeness. Encoding (im)politeness in conversation often plays different roles of social interactions such as for power dynamics at workplaces, decisive factor, and strategic use of it in social context (Chilton, 1990; Clark and Schunk, 1980). We use Stanford's politeness dataset StanfPolite (Danescu et al., 2013) that includes request types of polite and impolite text scraped from Stack Exchange question-answer community.

Sarcasm. Sarcasm acts by using words that mean something other than what you want to say, to insult someone, show irritation, or simply be funny. We choose two of them for XSLUE: SarcGhosh (Ghosh and Veale, 2016) and SARC v2.0 (Khodak et al., 2017)⁸. For SARC, we use the same preprocessing scheme in Ilić et al. (2018).

Metaphor. Metaphor is a figurative language that describes an object or an action by applying it to which is not applicable. We use two benchmark datasets:⁹ Trope Finder (TroFi) (Birke and Sarkar, 2006) and VU Amsterdam VUA Corpus (Steen, 2010) where metaphoric text is annotated by human annotators.

Offense. Hate speech is a speech that targets disadvantaged social groups based on group char-

acteristics (e.g., race, gender, sexual orientation) in a manner that is potentially harmful to them (Jacobs et al., 1998; Walker, 1994). We use the HateOffenssive dataset (Davidson et al., 2017) which includes hate text (7%), offensive text (76%), and none of them (17%).

Romance. Since we could not find any dataset containing romantic texts, we crawl and preprocess them from eleven different web sites, then make a new dataset ShortRomance. We make the same number of negative samples from the literal Reddit sentences (Khodak et al., 2017) as the romantic text. ShortRomance text are crawled from the following websites. The copyright of the messages are owned by the original writer of the websites.

- http://www.goodmorningtextmessages.com/2013/ 06/romantic-text-messages-for-her.html
- https://www.travelandleisure. com/travel-tips/ romantic-love-messages-for-him-and-her
- https://www.amoramargo.com/en/ sweet-text-messages-for-her/
- https://www.techjunkie.com/
 best-romantic-text-messages-for-girlfriend/
- https://liveboldandbloom.com/10/ relationships/love-messages-for-wife
- https://www.marriagefamilystrong.com/ sweet-love-text-messages/
- https://pairedlife.com/love/ love-messages-for-him-and-her
- https://truelovewords.com/ sweet-love-text-messages-for-him/
- https://www.serenataflowers.com/ pollennation/love-text-messages/
- https://www.greetingcardpoet.com/ 73-love-text-messages/
- https://www.wishesmsg.com/ heart-touching-love-messages/

Sentiment. Identifying sentiment polarity of opinion is challenging because of its implicit and explicit presence in text (Kim and Hovy, 2004; Pang et al., 2008). We use the annotated sentiment corpus on movie reviews; Sentiment Tree Bank (Socher et al., 2013) (SentiBank).

Emotion. Emotion is more fine-grained modeling of sentiment. (degree of control). We use two datasets: DailyDialog (Li et al., 2017) from the Ekman (1992)'s six categorical model, and EmoBank (Buechel and Hahn, 2017) from (Warriner et al., 2013)'s three-dimensional VAD

⁷The dataset requires an individual authorization for access, so we only provide a script for preprocessing.

⁸SARC_{pol} is a sub-task for the text from politics subreddit. ⁹we did not include Mohler et al. (2016)'s dataset because

the labels are not obtained from human annotators.

model. We also include a large but noisy emotionannotated corpus CrowdFlower (CrowdFlower, 2016), including in addition to Ekman's categories, additional seven categories: enthusiasm, worry, love, fun, hate, relief, and boredom.

Persona. Persona is a pragmatics style in group characteristics of the speaker. We use the stylistic language dataset written in parallel called PASTEL (Kang et al., 2019) where multiple types of the author's personas are given in conjunction. PASTEL has six different persona styles (i.e., age, gender, political view, ethnicity, country, education) where each has multiple attributes.

B Details on Annotation Schemes

Figure 6 shows snapshots of our annotation platform with the detailed instructions. We estimate the execution time of a task as 4 minutes, so paying 9/(60minutes/3minutes) = 0.4 per task. We make 10 size of batches multiple times and incrementally increase the size of batches up to 400 samples. For each batch, we manually checked the quality of outputs and blocked some bad users who abusively answered the questions.

C Details on Hyper-Parameters

For our BERT classifier, we use the uncased BERT English model. We did parameter sweeping on learning rates [2,5e-5,4] and batch sizes [32,16,8] on the validation set. For the BiLSTM baseline, we use 32^{10} sizes of batching for both training and testing and 256 hidden size for LSTM layer with 300 sizes of word embedding from GloVe (Pennington et al., 2014). The vocabulary size of BiLSTM is the same as the maximum vocabulary of the BERT model; 30522. For both BERT and BiLSTM models, we use the same maximum input length 128. Both training use 2e - 5 learning rate and 1.0 maximum gradient clipping with Adam optimizer with 1e - 8 epsilon. Also, we use early stopping until the maximum training epochs of 5.

 $^{^{10}}$ 32 batch size shows slightly better performance than smaller sizes like 8 or 16.

| Read the instruction given below carefully | Read | the | instruction | given | below | carefull | y: |
|--|------|-----|-------------|-------|-------|----------|----|
|--|------|-----|-------------|-------|-------|----------|----|

Your goal is to read given the text and predict its (A) stylistic features (e.g., humorous, politeness, sentiment, emotion) and the writer's (B) demographic features (e.g., gender, age). For demographic questions, if you can't guess the appropriate category, choose Don't Know button (but, try your best to make less number of it). For example,

Age (25-34)

Input Text : "It was really cool spkeaing with you Today I look forward to working for you!"

| Output (| Choices : | |
|----------|-----------|--|
| - | | |

Please carefully read the input text first. Then, click the appropriate category of button for each style (no multi-choice is allowed). Once you choose answers on every question, you can click the submit button at the bottom to end the task. Otherwise, you can't end the task. The estimated time of a task is 3-4 minutes.

NOTE: If one makes <u>random responses</u> or <u>inappropriate answers</u> detected on our validation samples, they will be entirely blocked from our future studies.

| Input Text : <u>\${input_text}</u> | (Recap) Input Text : <u>\${input_text}</u> | | |
|---|---|--|--|
| Part A: Choose appropriate stylistic features of the text | Part B: Choose appropriate demographic features of the text | | |
| A-1) How polite is the input text ? Polite Neural Impolite/Rude A-2) How humorous is the input text ? Humorous Non-Humorous | B-1) Choose the gender type of the writer based on the text? Male Female Non-binary / Third Gender Don't Know B-2) Choose the age range of the writer based on the text? <17 18-34 35-54 55-74 2-75 Don't Know | | |
| A-3) How formal is the input text ? Formal Neutral Informal/Casual | B-3) Choose the ethnic group of the writer based on the text? Caucasian African Hispanic Or Latino Native American Asian Middle Eastern Other Don't Know | | |
| A-4) How sarcastic is the input text ? Sarcastic Non-sarcastic A-5) How metaphoric is the input text ? | B-4) Choose the education level of the writer based on the text? | | |
| Metaphoric Non-Metaphoric | Doctorate Degree Other Don't Know | | |
| A-6) How offensive/hateful is the input text ? Offensive Hateful Neither | B-5) Choose the country of the writer based on the text? | | |
| A-7) How romantic is the input text ? Romantic Non-romantic | B-6) Choose the political ideaology type of the writer based on the text? Left Wing Centrist Right Wing Don't Know | | |
| A-8) What sentiment (e.g., positive vs negative) is contained in the input text ? Positive Negative Positive Negative | Sudamit | | |
| A-9) What valence (e.g., pleasant; calm vs unpleasant; upset) is contained in the input text ? Pleasant Unpleasant Unpleasant Unpleasant | | | |

Figure 6: Snapshots of our annotation tasks: general instruction (top) and annotation tasks on each style (bottom).