

# README

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This is the guidance on how to load the vectorized data.

## 0.0.1 Load the data

```
[1]: import pickle
import numpy as np

def load_data(file_name):
    with open(f'{file_name}.pkl', 'rb') as f:
        loaded_data = pickle.load(f)
    return loaded_data

[2]: german_sen_emb = load_data("german_sen_emb")
socc_sen_emb = load_data("socc_sen_emb")
german_word_emb = load_data("german_word_emb")
socc_word_emb = load_data("socc_word_emb")
german_label_expert1 = load_data("german_label_expert1")
german_label_expert2 = load_data("german_label_expert2")
socc_label = load_data("socc_label")
```

## 0.0.2 Data intro

There are total 7 data files

german represent for the IWG\_hatespeech\_public dataset ([https://github.com/UCSM-DUE/IWG\\_hatespeech\\_public](https://github.com/UCSM-DUE/IWG_hatespeech_public))

socc represent for the SFU Opinion and Comments Corpus dataset (<https://github.com/sfu-discourse-lab/SOCC>)

*german\_sen\_emb* and *socc\_sen\_emb* are 2-D numpy.array which are the sentences embedding of the text.

*socc\_word\_emb* and *german\_word\_emb* are the word embedding of the text.

*german\_label\_expert1*, *german\_label\_expert2* are numpy.array which are the labels annotated by two experts.

*socc\_label* is a numpy.array which is the **toxicity\_level** line in **SFU\_constructiveness\_toxicity\_corpus.csv**.

### german dataset

```
[3]: german_sen_emb.shape #469 is the size of dataset, 768 is the embedding dim of
↳ the sent2vec model
```

[3]: (469, 768)

```
[4]: len(german_word_emb) #469 is the size of dataset, each of the german_sen_emb
      ↪list is a numpy.array, which in shape of (num_words, 300).
```

[4]: 469

```
[5]: german_word_emb[0].shape # 300 is the dim of the word2vec model, 11 is the
      ↪length of the first text (11 word vectors in total)
```

[5]: (11, 300)

**NOTE: EACH ELEMENTS IN THE LIST ARE NOT IN THE SAME SHAPE**

```
[6]: german_word_emb[1].shape #different shape with german_word_emb[0].shape
```

[6]: (9, 300)

```
[7]: german_label_expert1.shape #labels annotated by the first expert
```

[7]: (469,)

```
[8]: german_label_expert2.shape #labels annotated by the second expert
```

[8]: (469,)

**socc dataset**

```
[9]: socc_sen_emb.shape #1043 is the size of dataset, 768 is the embedding dim of
      ↪the sent2vec model
```

[9]: (1043, 768)

```
[10]: len(socc_word_emb) #1043 is the size of dataset, each of the german_sen_emb
      ↪list is a numpy.array, which in shape of (num_words, 300).
```

[10]: 1043

```
[11]: socc_word_emb[0].shape #300 is the dim of the word2vec model, 120 is the length
      ↪of the first text (120 word vectors in total)
```

[11]: (120, 300)

```
[12]: socc_word_emb[1].shape #300 is the dim of the word2vec model, 223 is the length
      ↪of the second text (223 word vectors in total)
```

[12]: (223, 300)

**NOTE: socc\_word\_emb contains np.nan element**

This is because there are some samples which all the words of this sample are not included in the word2vec model.

```
[13]: for i in range(len(socc_word_emb)):
      try:
          if np.isnan(socc_word_emb[i]):
              print(f"The {i} element of socc_word_emb is Nan")
      except:
          pass
```

The 123 element of socc\_word\_emb is Nan

The 364 element of socc\_word\_emb is Nan

The 422 element of socc\_word\_emb is Nan

```
[14]: socc_label.shape #labels, the original toxicity_level has 4 levels, I simply
      ↪make level 1 as label 0 and levels 2, 3, 4 as label 1
```

```
[14]: (1043,)
```

```
[ ]:
```