Sentiment Classification

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Dataset

Dataset

About dataset

tweet_id	▲ sentiment	=	▲ content =
1748453174.20 - 1762077424.00 Count: 11,487	neutral worry Other (22903)	22% 21% 57%	39827 unique values
1956967341	empty		@tiffanylue i know i was listenin to bad habit earlier and i started freakin at his part =[
1956967666	sadness		Layin n bed with a headache ughhhhwaitin on your call
1956967696	sadness		Funeral ceremonygloomy friday
1956967789	enthusiasm		wants to hang out with friends SOON!
1956968416	neutral		@dannycastillo We want to trade with someone who has Houston tickets, but no one will.

- ✔ 트윗을 작성한 사람이 주석 형태로 작성한 감정 정보
- tweet_id
- 💉 sentiment : 13가지의 감정 범주
- ✓ content : 주석 형태로 작성된 텍스트
- 데이터의 크기: (40000, 3)

Remove unnecessary features

	0	1	2
0	tweet_id	sentiment	content
1	1956967341	empty	@tiffanylue i know i was listenin to bad habi
2	1956967666	sadness	Layin n bed with a headache ughhhhwaitin o
3	1956967696	sadness	Funeral ceremonygloomy friday
4	1956967789	enthusiasm	wants to hang out with friends SOON!
39996	1753918954	neutral	@JohnLloydTaylor
39997	1753919001	love	Happy Mothers Day All my love
39998	1753919005	love	Happy Mother's Day to all the mommies out ther
39999	1753919043	happiness	@niariley WASSUP BEAUTIFUL!!! FOLLOW ME!! PEE
40000	1753919049	love	@mopedronin bullet train from tokyo the gf
40001 rows × 3 columns			



1	1	
empty @tiffanylue i know i was listenin to bad hab	empty	1
sadness Layin n bed with a headache ughhhhwaitin o	sadness	2
sadness Funeral ceremonygloomy friday	sadness	3
thusiasm wants to hang out with friends SOO	enthusiasm	4
neutral @dannycastillo We want to trade with someone w	neutral	5
neutral @JohnLloydTay	neutral	39996
love Happy Mothers Day All my lo	love	39997
love Happy Mother's Day to all the mommies out the	love	39998
appiness @niariley WASSUP BEAUTIFUL!!! FOLLOW ME!! PEE	happiness	39999
love @mopedronin bullet train from tokyo the qf	love	40000

Change feature name

1 empty @tiffanylue i know i was listenin to bad habi.		
empty @tillariyide i kilow i was listerilii to bad flabi.	empty	1
2 sadness Layin n bed with a headache ughhhhwaitin o.	sadness	2
3 sadness Funeral ceremonygloomy friday.	sadness	3
4 enthusiasm wants to hang out with friends SOON	enthusiasm	4
5 neutral @dannycastillo We want to trade with someone w.	neutral	5
6 neutral @JohnLloydTaylo	neutral	39996
7 love Happy Mothers Day All my lov	love	39997
8 love Happy Mother's Day to all the mommies out ther.	love	39998
9 happiness @niariley WASSUP BEAUTIFUL!!! FOLLOW ME!! PEE.	happiness	39999
0 love @mopedronin bullet train from tokyo the gf.	love	40000



	sentiment	content	
1	empty	@tiffanylue i know i was listenin to bad habi	
2	sadness	Layin n bed with a headache ughhhhwaitin o	
3	sadness	Funeral ceremonygloomy friday	
4	enthusiasm	wants to hang out with friends SOON!	
5	neutral	@dannycastillo We want to trade with someone w	
39996	neutral	@JohnLloydTaylor	
39997	love	Happy Mothers Day All my love	
39998	love	Happy Mother's Day to all the mommies out ther	
39999	happiness	@niariley WASSUP BEAUTIFUL!!! FOLLOW ME!! PEE	
40000	love	@mopedronin bullet train from tokyo the gf \dots	
40000 rows × 2 columns			

Text preprocessing

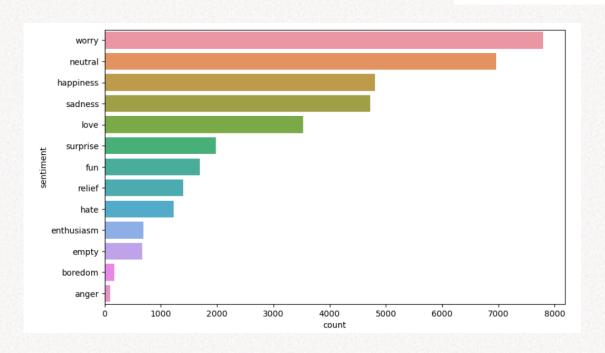
사용자명 제거	HTML 엔티티 제거
@tiffanylue know i was listenin to bad habit earlier and i started freakin at his part =[
Layın n bed with a headache ughhhhwaitin on your call	
Funeral ceremonygloomy friday	
wants to hang out with friends SOON!	
@dannycastillo Wo want to trade with someone who has Houston tickets, but no one will.	
Re-pinging @ghostridah1: why didn't you go to prom? BC my bf didn't like my friends	
I should be sleep, but im not! thinking about an old friend who I want. but he's married now	v. dan n, & b) wants me 2! scandalous!
Hmmm (http://www.dj) ero.com/ is down	
@charviray Charlene my lov. Umiss you	
@kelcouch I'm sorry at least it's Friday?	

URL 제거

→ 소문자 변환, 특수 문자 제거, 토큰화, 불용어 제거, 표제어 추출 등 진행

Check data structure

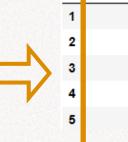
✔ 타입과 결측치 확인



✔ 감정의 종류와 분포 확인



sentiment	content
empty	know listenin bad habit earlier started freaki
sadness	layin n bed headache ughhhhwaitin call
sadness	funeral ceremonygloomy friday
enthusiasm	want hang friend soon
neutral	want trade someone houston ticket one
happiness	succesfully following tayla
love	happy mother day love
love	happy mother day mommy woman man long youre mo
happiness	wassup beautiful follow peep new hit single ww
love	bullet train tokyo gf visiting japan since thu
	empty sadness sadness enthusiasm neutral happiness love love happiness



	CALCADO DA SA	
content	sentiment	
know listenin bad habit earlier started freaki	2	1
layin n bed headache ughhhhwaitin call	10	2
funeral ceremonygloomy friday	10	3
want hang friend soon	3	4
want trade someone houston ticket one	8	5

Loading the pre-trained DistilBert model

```
import torch
import transformers as ppb
import warnings
warnings.filterwarnings('ignore')

model_class, tokenizer_class, pretrained_weights = (ppb.DistilBertModel, ppb.DistilBertTokenizer, 'distilbert-base-uncased')

tokenizer = tokenizer_class.from_pretrained(pretrained_weights)
model = model_class.from_pretrained(pretrained_weights)
```

Stratified sampling

```
from sklearn.model_selection import train_test_split

x = df["content"]
y = df["sentiment"]

x_train, _, y_train, _ = train_test_split(x, y, train_size=10000, stratify=y, random_state=42)

df_sampled = pd.DataFrame({'content':x_train, 'sentiment':y_train})
```

? 데이터가 40000개로 너무 많아서 deep learning을 진행하는데 무한 로딩 & 메모리 초과로 멈춤 발생
 ✓ Stratified sampling을 통해 10000개의 데이터만 추출하여 데이터 크기 축소

Check df_sample

✔ (10000,2)로 축소된 df_sample

	content	sentiment	
31383	omg found thnx	5	
33253	oops im watching mom son sleeping ing	8	
14926	kno im sad evry leavin horrible im supposed b \dots	10	
1860	ever come across something reminds alot one pe $% \label{eq:come} % eq:co$	10	
37840	also try friendly fire havent already heard gr	5	
18914	aww dude fair thought point thing	6	
38589	enjoying mommy day	4	
2121	word counting hand hurt	12	
21935	face mask hehe	4	
24321	may th starwarsday via	8	
10000 rows × 2 columns			

✓ 축소된 sentiment categories

Preparing data for DistilBert

√ Tokenization

```
tokenized = df_sampled['content'].apply((lambda x: tokenizer.encode(x, add_special_tokens=True))) tokenized
```

```
31383 [101, 18168, 2290, 2179, 16215, 26807, 102]
33253 [101, 1051, 11923, 10047, 3666, 3566, 2365, 57...
14926 [101, 14161, 2080, 10047, 6517, 23408, 2854, 1...
1860 [101, 2412, 2272, 2408, 2242, 15537, 2632, 414...
37840 [101, 2036, 3046, 5379, 2543, 4033, 2102, 2525...
```

```
✓ Padding
```

```
\max len = 0
for i in tokenized.values:
    if len(i) > max len:
       max_len = len(i)
#가장 긴 문장의 len을 구하고, 짧은 문장들은 뒤에 0을 추가해서 padding 진행
padded = np.arrav([i + [0]*(max len-len(i)) for i in tokenized.values])
padded
array([[ 101, 18168, 2290, ...,
                                              0],
      [ 101, 1051, 11923, ...,
                                              0],
                                        0,
      [ 101, 14161, 2080, ...,
                                              01.
      [ 101, 2773, 10320, ...,
      [ 101, 2227, 7308, ...,
      [ 101, 2089, 16215, ...,
                                              011)
```

Masking

```
attention_mask = np.where(padded != 0, 1, 0)
attention_mask.shape
(10000, 38)
```

Deep learning

✔ 전처리된 입력을 사용해 모델 실행

```
import time
start_time = time.time()
with torch.no_grad():
    last_hidden_states = model(input_ids, attention_mask=attention_mask)
end_time = time.time()
execution_time = end_time - start_time
print("실행시간 : {:.2f}초".format(execution_time))
실행시간 : 516.25초
```

```
BaseModelOutput(last_hidden_state=tensor([[[-0.2119, -0.0429, 0.0134, ..., 0.0016, 0.2919, 0.3347],
        [-0.2285, -0.0771, 0.2305, ..., -0.2790, 0.4037, 0.6434],
        [-0.2066, -0.2901, 0.3833, ..., -0.2589, 0.0803, 0.5664],
        [-0.1814, -0.3061, 0.0914, ..., 0.4309, -0.2672, 0.2827],
        [-0.1648, -0.1960, 0.0909, ..., 0.3828, -0.1762, 0.3305],
        [-0.1868, -0.1554, 0.1009, ..., 0.3504, -0.1224, 0.3267]]
       [[-0.2241, 0.2844, 0.1016, ..., -0.1495, 0.3913, 0.4542]
        [-0.6740, 0.8077, 0.4433, ..., 0.0297, 0.6494, 0.2652],
        [-0.2993, 0.4834, 0.5944, ..., -0.2159, 0.2872, 0.3982]
        [ 0.0412, 0.0990, 0.5274, ..., -0.2102, 0.0435, 0.1350]
        [-0.0976, 0.2543, 0.4701, ..., -0.1508, 0.0746, 0.1171],
        [ 0.1394, 0.2586, 0.4294, ..., -0.1421, -0.0988, -0.0767]],
       [[-0.3373, 0.0586, 0.1206, ..., -0.1060, 0.2444, 0.5299],
        [-0.4679, 0.2288, 0.7018, ..., 0.0090, 0.5424, 0.1029]
       [-0.8925, 0.2072, 0.4522, ..., -0.4207, -0.2298, 0.3141],
        [-0.2086, 0.0719, 0.3592, ..., -0.0648, 0.1394, 0.1804],
        [-0.0529, 0.0275, 0.2426, ..., -0.0929, -0.0309, 0.1999],
        [-0.0697, -0.0651, 0.1502, ..., -0.0025, -0.0025, 0.2161]]
```

✓ 입력 시퀀스에 대한 모델 처리 결과 (last_hidden_states)

LogisticRegression

✓ train_test_split으로 분리 후 LogisticRegression에 입력을 넣어 예측 진행 LogisticRegression(solver='liblinear')

Evaluating

from sklearn.metrics import accuracy score, confusion matrix, precision score, recall score, f1 score #accuracy accuracy = accuracy_score(y_test, pred) #precision precision = precision_score(y_test, pred, average='macro') #recall recall = recall_score(y_test, pred, average='macro') #f1 score f1 = f1_score(y_test, pred, average='macro') print(f"Accuracy: {accuracy}") print(f"Precision: {precision}") print(f"Recall: {recall}") Accuracy: 0.322 Precision: 0.1647537115056242 Recall: 0.15962835318492952 F1 Score: 0.15301426150447464

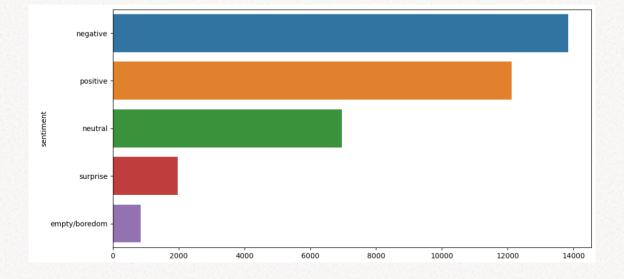
Performance Improvement

Performance improvement

Reducing categories

```
def categorization(sentiment):
    if sentiment in ['happiness', 'love', 'fun', 'relief', 'enthusiasm']:
        return 'positive'
    elif sentiment in ['worry', 'sadness', 'hate', 'anger']:
        return 'negative'
    elif sentiment == 'neutral':
        return 'neutral'
    elif sentiment in ['empty', 'boredom']:
        return 'empty/boredom' #두 감정 모두 감정적인 반응을 보이지 않음 (흥미가 없다거나 감정적으로 무감각)
    elif sentiment == 'surprise':
        return 'surprise' #surprise는 긍정이 될 수도, 부정이 될 수도 있으므로 단독
    else:
        return 'unknown'
```

✔ 감정의 종류와 분포 확인



Performance improvement

LogisticRegression

✓ train_test_split으로 분리 후 LogisticRegression에 입력을 넣어 예측 진행

Evaluating

Accuracy: 0.322

Precision: 0.1847581 115056242

Recall: 0.15962835318492952 F1 Score: 0.15301426150447464



Accuracy: 0.5424

Recall: 0.31795772639027414 F1 Score: 0.304735659483237

THANK YOU

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