[Objective]

Your model should generate word vectors.

[Requirements]

- 1. Implement fastText model with Gensim.
- 2. You should experiment with settings stated in the evaluation report, and report the result of each settings.
- 4. You should report the experimental results.

[Code structure]

- fasttext_train.py

Pre-trained fastText

- https://fasttext.cc/docs/en/english-vectors.html
- https://dl.fbaipublicfiles.com/fasttext/vectors-english/wiki-news-300d-1M-subword.bin.zip

Download pre-trained word vectors

Pre-trained word vectors learned on different sources can be downloaded below:

- 1. wiki-news-300d-1M.vec.zip: 1 million word vectors trained on Wikipedia 2017, UMBC webbase corpus and statmt.org news dataset (16B tokens).
- 2. wiki-news-300d-1M-subword.vec.zip: 1 million word vectors trained with subword infomation on Wikipedia 2017, UMBC webbase corpus and statmt.org news dataset (16B tokens).
- 3. crawl-300d-2M.vec.zip: 2 million word vectors trained on Common Crawl (600B tokens).
- 4. crawl-300d-2M-subword.zip: 2 million word vectors trained with subword information on Common Crawl (600B tokens).

fastText training

```
from gensim.models import FastText
from gensim.models.word2vec import PathLineSentences

sentences = PathLineSentences("./data/1billion/")
model = FastText(sentences=sentences, size=100, window=5, min_count=5,
workers=4, sg=0, hs=0, negative=5, ns_exponent=0.75, alpha=0.01,
min_alpha=0.0001, iter=1, word_ngrams=1, min_n=3, max_n=6)
model.save("fastText.model")
print(len(model.wv.vocab))
score, predictions = model.wv.evaluate_word_analogies('./data/questions-words.txt')
print(score)
```

fastText training

Hyperparameters

- size: 단어 벡터의 차원
- window size: context 단어 수 / 2
- min_count: 최소 빈도수 기준, 단어사전에 포함 여부 결정
- wokers: 스레드 수
- sg: 1이면 skip-gram 사용
- hs: 1이면 hierarchical soft, 0이면 negative sampling 사용
- negative: negative sample의 개수
- ns_exponent: unigram distribution에 적용될 지수 값
- cbow_mean: 1이면 context 단어의 평균을 사용, 0이면 합을 사용
- alpha: learning rate
- min_alpha: learning rate decay 시에 최소 learning rate
- max_vocab_size: 단어 사전의 최대 크기
- iter: epoch 수
- sorted_vocab: 1이면 사전의 단어들을 빈도수 기준 내림차순 정렬
- batch_words: batch size
- word_ngrams: subword 정보 사용여부, 1이면 사용, 0이면 그냥 word2vec
- min_n: character n-gram의 최소 길이
- max_n: character n-gram의 최대 길이

Experiments: Effect of the size of n-grams

• Taking a large range such as 3-6 provides a reasonable amount of subword information

| | 2 | 3 | 4 | 5 | 6 | | 2 | 3 | 4 | 5 | 6 | | 2 | 3 | 4 | 5 | 6 |
|---|----|--------|---------------|----------------------|----------------|-----|----|---------|---------------|---------------|----------------|------|-------|---------|---------------|----------------------|---------|
| 2 | 57 | 64 | 67 | 69 | 69 | 2 | 59 | 55 | 56 | 59 | 60 | 2 | 45 | 50 | 53 | 54 | 55 |
| 3 | | 65 | 68 | 70 | 70 | 3 | | 60 | 58 | 60 | 62 | 3 | | 51 | 55 | 55 | 56 |
| 1 | | | 70 | 70 | 71 | 4 | | | 62 | 62 | 63 | 4 | | | 54 | 56 | 56 |
| 5 | | | | 69 | 71 | 5 | | | | 64 | 64 | 5 | | | | 56 | 56 |
| 5 | | | | | 70 | 6 | | | | | 65 | 6 | | | | | 54 |
| | | | | | | | | | | | | | | | | | |
| | | (a) Di | E-GUF | 350 | | | (| b) DE | Sema | antic | | | (c |) DE | Synta | ctic | |
| | 2 | (a) Di | E-GUF | 5 | 6 | _ | 2 | b) DE | Sema 4 | antic 5 | 6 | Ď. | 2 | 3 | Synta 4 | ctic 5 | 6 |
| 2 | | 3 | 4 | 5 | 6 48 | | | | 29 | | -11 | | 27.07 | 20.2 | | 3257 | 6 73 |
| 2 | 2 | 3 | 4 46 | 5 47 | 48 | 2 3 | 2 | 3 | 4 | 5 | 76 | 2 3 | 2 | 3 | 4 | 5 | 5475 |
| | 2 | 3 42 | 4 46 | 5 47 | 48 48 | | 2 | 3 76 | 4 75 | 5 76 | 76 77 | 1000 | 2 | 3 71 | 4 73 | 5 74 | 73 |
| 3 | 2 | 3 42 | 4 46 46 | 5 47 48 | 48 48 48 | 3 | 2 | 3 76 | 4 75 77 | 5 76 78 | 76 77 79 | 3 | 2 | 3 71 | 4 73 74 | 5 74 75 | 73 |

[Evaluation report]

| fastText Evaluation Report | | | | | | | | | | | | |
|----------------------------|-------|---------|-------------------|-----------------------|---------------|--------------------------|-----------|-----------|---------------|----------|----------|--------------|
| | Model | n-gram | Negative Sampling | # of negative samples | Learning rate | Learning rate decay(O/X) | dimension | iteration | training time | Accuracy | OOV word | most_similar |
| setting #1 | SG | 2,3 | 0 | 15 | 0.01 | 0 | 100 | 5 | | | | 1 |
| setting #2 | SG | 3,4,5,6 | 0 | 15 | 0.01 | 0 | 100 | 5 | | | | 1 |
| setting #3 | SG | 2,3 | 0 | 15 | 0.01 | 0 | 300 | 5 | | | | |
| setting #4 | SG | 3,4,5,6 | 0 | 15 | 0.01 | 0 | 300 | 5 | | | | |
| [결과 정리] | | | | | | | | | | | | |
| 1 | | | | | | | | | | | | |

```
model = FastText.load("fastText.model")
score, predictions = model.wv.evaluate_word_analogies('./data/questions-words.txt')
print(score)
print(model.wv.most_similar("thank____you", topn=20))
```

• Evaluation Criteria

| Simplicity | How concisely did you write the code? - 배점 2점 | | | | | |
|---------------------|--|--|--|--|--|--|
| Performance | How well did the results of the code perform? - 배점 4점 - acc 55%이상 달성: 3점 - OOV word 생성 및 유사단어 확인: 1점 | | | | | |
| Brevity and Clarity | How concisely and clearly did you explain the results? - 배점 4점 | | | | | |

- Due to : ~ 11.1 (Sun)
- Submission: Online submission on blackboard
- Your submission should contain
 - 1) The whole code of your implementation
 - 2) The evaluation report
- You must implement the components yourself!
- File name : StudentID_Name.zip