06 sec preprocessing

September 29, 2021

1 Word vectors from SEC filings using Gensim: Preprocessing

In this section, we will learn word and phrase vectors from annual SEC filings using gensim to illustrate the potential value of word embeddings for algorithmic trading. In the following sections, we will combine these vectors as features with price returns to train neural networks to predict equity prices from the content of security filings.

In particular, we use a dataset containing over 22,000 10-K annual reports from the period 2013-2016 that are filed by listed companies and contain both financial information and management commentary (see chapter 3 on Alternative Data). For about half of 11K filings for companies that we have stock prices to label the data for predictive modeling

1.1 Imports & Settings

```
[1]: import warnings
     warnings.filterwarnings('ignore')
[2]: from dateutil.relativedelta import relativedelta
     from pathlib import Path
     import numpy as np
     import pandas as pd
     from time import time
     from collections import Counter
     import logging
     import spacy
     from gensim.models import Word2Vec
     from gensim.models.word2vec import LineSentence
     from gensim.models.phrases import Phrases, Phraser
[3]:
    np.random.seed(42)
[4]: def format time(t):
         m, s = divmod(t, 60)
         h, m = divmod(m, 60)
         return f'{h:02.0f}:{m:02.0f}:{s:02.0f}'
```

1.1.1 Logging Setup

1.2 Data Download

The data can be downloaded from here. Unzip and move into the data folder in the repository's root directory and rename to filings.

1.2.1 Paths

Each filing is a separate text file and a master index contains filing metadata. We extract the most informative sections, namely - Item 1 and 1A: Business and Risk Factors - Item 7 and 7A: Management's Discussion and Disclosures about Market Risks

The notebook preprocessing shows how to parse and tokenize the text using spaCy, similar to the approach in chapter 14. We do not lemmatize the tokens to preserve nuances of word usage.

We use gensim to detect phrases. The Phrases module scores the tokens and the Phraser class transforms the text data accordingly. The notebook shows how to repeat the process to create longer phrases.

```
[6]: sec_path = Path('..', 'data', 'sec-filings')
filing_path = sec_path / 'filings'
sections_path = sec_path / 'sections'
```

```
[7]: if not sections_path.exists():
    sections_path.mkdir(exist_ok=True, parents=True)
```

1.3 Identify Sections

```
txt = pd.Series(items).reset_index()
txt.columns = ['item', 'text']
txt.to_csv(sections_path / (filing.stem + '.csv'), index=False)
```

500 1000 1500 2000 2500 3000 3500 4000 4500 5000 5500 6000 6500 7000 7500 8000 8500 9000 9500 10000 10500 11000 11500 12000 12500 13000 13500 14000 14500 15000 15500 16000 16500 17000 17500 18000 18500 19000 19500 20000 20500 21000 21500 22000 22500

1.4 Parse Sections

Select the following sections:

```
[9]: sections = ['1', '1a', '7', '7a']
 [9]: clean_path = sec_path / 'selected_sections'
      if not clean_path.exists():
          clean_path.mkdir(exist_ok=True)
[14]: | nlp = spacy.load('en', disable=['ner'])
      nlp.max_length = 6000000
[16]: vocab = Counter()
      t = total_tokens = 0
      stats = []
      start = time()
      to_do = len(list(sections_path.glob('*.csv')))
      done = len(list(clean_path.glob('*.csv'))) + 1
      for text_file in sections_path.glob('*.csv'):
          file id = int(text file.stem)
          clean_file = clean_path / f'{file_id}.csv'
          if clean_file.exists():
          items = pd.read_csv(text_file).dropna()
          items.item = items.item.astype(str)
          items = items[items.item.isin(sections)]
          if done % 100 == 0:
              duration = time() - start
              to_go = (to_do - done) * duration / done
              print(f'{done:>5}\t{format_time(duration)}\t{total_tokens / duration:,.
       →0f}\t{format_time(to_go)}')
          clean doc = []
          for _, (item, text) in items.iterrows():
              doc = nlp(text)
              for s, sentence in enumerate(doc.sents):
```

```
clean_sentence = []
        if sentence is not None:
            for t, token in enumerate(sentence, 1):
                if not any([token.is_stop,
                             token.is_digit,
                             not token.is_alpha,
                             token.is_punct,
                             token.is_space,
                             token.lemma_ == '-PRON-',
                             token.pos_ in ['PUNCT', 'SYM', 'X']]):
                    clean_sentence.append(token.text.lower())
            total_tokens += t
            if len(clean_sentence) > 0:
                clean_doc.append([item, s, ' '.join(clean_sentence)])
(pd.DataFrame(clean_doc,
              columns=['item', 'sentence', 'text'])
 .dropna()
 .to_csv(clean_file, index=False))
done += 1
```

```
100
       00:02:38
                       18,125 09:53:45
 200
       00:05:36
                       17,183 10:28:08
 300
       00:08:30
                       16,514 10:32:46
 400
                       17,093 10:08:36
       00:10:57
 500
       00:13:21
                       17,482 09:50:42
 600
                       17,806 09:45:08
       00:15:56
 700
       00:18:33
                       18,003 09:41:23
 800
       00:20:46
                       18,139 09:26:55
                       18,262 09:18:06
 900
       00:23:07
                       18,342 09:12:43
1000
       00:25:33
1100
       00:27:51
                       18,425 09:05:06
1200
       00:30:27
                       18,486 09:03:41
1300
       00:33:05
                       18,536 09:02:49
1400
       00:35:36
                       18,579 08:59:47
1500
       00:38:15
                       18,621 08:58:47
1600
                       18,666 08:54:19
       00:40:39
1700
       00:42:57
                       18,714 08:48:44
1800
       00:45:36
                       18,759 08:47:41
1900
       00:47:52
                       18,805 08:42:17
2000
       00:50:14
                       18,853 08:38:10
2100
       00:52:23
                       18,879 08:32:06
2200
       00:54:43
                       18,908 08:28:11
2300
       00:57:17
                       18,908 08:26:20
2400
                       18,834 08:24:02
       00:59:48
2500
       01:01:56
                       18,868 08:18:43
2600
       01:04:21
                       18,898 08:15:43
2700
       01:06:23
                       18,924 08:10:03
```

| 2800 | 01:08:29 | 18,951 | 08:05:05 |
|------|----------|--------|----------|
| 2900 | 01:10:42 | 18,981 | 08:01:03 |
| 3000 | 01:12:49 | 19,008 | 07:56:28 |
| 3100 | 01:15:12 | 19,030 | 07:53:45 |
| 3200 | 01:17:46 | 19,052 | 07:52:11 |
| 3300 | 01:20:04 | 19,074 | 07:49:02 |
| 3400 | 01:22:17 | 19,098 | 07:45:27 |
| 3500 | 01:24:32 | 19,117 | 07:42:03 |
| 3600 | 01:26:45 | 19,134 | 07:38:35 |
| 3700 | 01:28:52 | 19,151 | 07:34:43 |
| 3800 | 01:31:00 | 19,167 | 07:30:58 |
| 3900 | 01:33:15 | 19,189 | 07:27:52 |
| 4000 | 01:35:44 | 19,204 | 07:25:53 |
| 4100 | 01:37:58 | 19,217 | 07:22:49 |
| 4200 | 01:40:21 | 19,227 | 07:20:22 |
| 4300 | 01:42:40 | 19,241 | 07:17:39 |
| 4400 | 01:45:03 | 19,250 | 07:17:05 |
| 4500 | 01:43:03 | 19,258 | 07:12:23 |
| 4600 | 01:47:19 | 19,269 | 07:12:23 |
| 4700 | 01:49:54 | 19,209 | 07:09:30 |
| 4800 | 01:53:55 | 19,277 | 07:00:43 |
| | | · · | |
| 4900 | 01:56:07 | 19,298 | 07:00:11 |
| 5000 | 01:58:29 | 19,305 | 06:57:47 |
| 5100 | 02:00:38 | 19,316 | 06:54:42 |
| 5200 | 02:02:47 | 19,322 | 06:51:34 |
| 5300 | 02:04:53 | 19,328 | 06:48:21 |
| 5400 | 02:07:18 | 19,336 | 06:46:11 |
| 5500 | 02:09:56 | 19,347 | 06:44:43 |
| 5600 | 02:12:05 | 19,354 | 06:41:42 |
| 5700 | 02:14:05 | 19,359 | 06:38:18 |
| 5800 | 02:16:29 | 19,368 | 06:36:04 |
| 5900 | 02:18:34 | 19,378 | 06:32:57 |
| 6000 | 02:20:58 | 19,382 | 06:30:43 |
| 6100 | 02:23:23 | 19,388 | 06:28:33 |
| 6200 | 02:25:33 | 19,396 | 06:25:44 |
| 6300 | 02:27:41 | 19,405 | 06:22:51 |
| 6400 | 02:30:02 | 19,412 | 06:20:31 |
| 6500 | 02:32:16 | 19,418 | 06:17:53 |
| 6600 | 02:34:22 | 19,424 | 06:14:57 |
| 6700 | 02:36:32 | 19,430 | 06:12:13 |
| 6800 | 02:38:56 | 19,432 | 06:10:01 |
| 6900 | 02:41:05 | 19,433 | 06:07:15 |
| 7000 | 02:43:38 | 19,436 | 06:05:23 |
| 7100 | 02:46:09 | 19,439 | 06:03:27 |
| 7200 | 02:48:23 | 19,443 | 06:00:53 |
| 7300 | 02:50:49 | 19,442 | 05:58:44 |
| 7400 | 02:53:09 | 19,444 | 05:56:24 |
| 7500 | 02:55:20 | 19,448 | 05:53:45 |
| | | | |

| 7600 | 02:57:48 | 19,452 | 05:51:38 |
|-------|----------|--------|----------|
| 7700 | 03:00:12 | 19,451 | 05:49:26 |
| 7800 | 03:02:39 | 19,453 | 05:47:18 |
| 7900 | 03:04:59 | 19,456 | 05:44:57 |
| 8000 | 03:06:55 | 19,460 | 05:41:51 |
| 8100 | 03:09:12 | 19,464 | 05:39:26 |
| 8200 | 03:11:20 | 19,469 | 05:36:44 |
| 8300 | 03:13:31 | 19,473 | 05:34:09 |
| 8400 | 03:15:45 | 19,478 | 05:31:38 |
| 8500 | 03:18:02 | 19,483 | 05:29:13 |
| 8600 | 03:20:21 | 19,488 | 05:26:52 |
| 8700 | 03:22:28 | 19,494 | 05:24:12 |
| 8800 | 03:24:48 | 19,498 | 05:21:53 |
| 8900 | 03:27:06 | 19,504 | 05:19:31 |
| 9000 | 03:29:28 | 19,511 | 05:17:16 |
| 9100 | 03:31:41 | 19,514 | 05:14:45 |
| 9200 | 03:33:53 | 19,518 | 05:12:14 |
| 9300 | 03:36:15 | 19,521 | 05:09:59 |
| 9400 | 03:38:35 | 19,528 | 05:07:41 |
| 9500 | 03:40:50 | 19,534 | 05:05:14 |
| 9600 | 03:43:02 | 19,539 | 05:02:45 |
| 9700 | 03:45:23 | 19,539 | 05:00:28 |
| 9800 | 03:47:45 | 19,541 | 04:58:12 |
| 9900 | 03:49:56 | 19,545 | 04:55:41 |
| 10000 | 03:51:60 | 19,549 | 04:53:02 |
| 10100 | 03:54:13 | 19,553 | 04:50:36 |
| 10200 | 03:56:37 | 19,558 | 04:48:22 |
| 10300 | 03:59:01 | 19,562 | 04:46:09 |
| 10400 | 04:01:29 | 19,566 | 04:44:00 |
| 10500 | 04:03:49 | 19,568 | 04:41:41 |
| 10600 | 04:06:03 | 19,573 | 04:39:16 |
| 10700 | 04:08:28 | 19,577 | 04:37:04 |
| 10800 | 04:10:45 | 19,581 | 04:34:41 |
| 10900 | 04:13:10 | 19,585 | 04:32:28 |
| 11000 | 04:15:13 | 19,588 | 04:29:51 |
| 11100 | 04:17:46 | 19,592 | 04:27:47 |
| 11200 | 04:20:06 | 19,593 | 04:25:27 |
| 11300 | 04:22:25 | 19,596 | 04:23:09 |
| 11400 | 04:24:29 | 19,599 | 04:20:34 |
| 11500 | 04:26:39 | 19,603 | 04:18:06 |
| 11600 | 04:29:06 | 19,605 | 04:15:54 |
| 11700 | 04:31:25 | 19,609 | 04:13:35 |
| 11800 | 04:33:43 | 19,612 | 04:11:14 |
| 11900 | 04:35:54 | 19,613 | 04:08:48 |
| 12000 | 04:38:21 | 19,617 | 04:06:36 |
| 12100 | 04:40:38 | 19,619 | 04:04:15 |
| 12200 | 04:43:09 | 19,621 | 04:02:05 |
| 12300 | 04:45:18 | 19,623 | 03:59:37 |
| | | | |

| 12400 | 04:47:37 | 19,626 | 03:57:18 |
|-------|----------|--------|----------|
| 12500 | 04:49:53 | 19,629 | 03:54:57 |
| 12600 | 04:51:59 | 19,631 | 03:52:27 |
| 12700 | 04:54:05 | 19,634 | 03:49:58 |
| 12800 | 04:56:16 | 19,636 | 03:47:33 |
| 12900 | 04:58:43 | 19,639 | 03:45:20 |
| 13000 | 05:01:04 | 19,643 | 03:43:02 |
| 13100 | 05:03:13 | 19,646 | 03:40:36 |
| 13200 | 05:05:34 | 19,648 | 03:38:19 |
| 13300 | 05:07:57 | 19,650 | 03:36:03 |
| 13400 | 05:10:23 | 19,652 | 03:33:49 |
| 13500 | 05:12:43 | 19,654 | 03:31:31 |
| 13600 | 05:14:42 | 19,657 | 03:28:59 |
| 13700 | 05:17:04 | 19,658 | 03:26:42 |
| 13800 | 05:19:11 | 19,662 | 03:24:15 |
| 13900 | 05:21:22 | 19,665 | 03:21:51 |
| 14000 | 05:23:52 | 19,668 | 03:19:40 |
| 14100 | 05:26:12 | 19,669 | 03:17:22 |
| 14200 | 05:28:30 | 19,671 | 03:15:03 |
| 14300 | 05:30:26 | 19,674 | 03:12:30 |
| 14400 | 05:32:47 | 19,676 | 03:10:13 |
| 14500 | 05:35:05 | 19,679 | 03:07:54 |
| 14600 | 05:37:30 | 19,682 | 03:05:39 |
| 14700 | 05:39:50 | 19,685 | 03:03:21 |
| 14800 | 05:41:54 | 19,689 | 03:00:54 |
| 14900 | 05:44:06 | 19,692 | 02:58:32 |
| 15000 | 05:46:16 | 19,694 | 02:56:10 |
| 15100 | 05:48:32 | 19,696 | 02:53:50 |
| 15200 | 05:50:45 | 19,698 | 02:51:29 |
| 15300 | 05:52:56 | 19,700 | 02:49:06 |
| 15400 | 05:55:12 | 19,702 | 02:46:47 |
| 15500 | 05:57:27 | 19,704 | 02:44:27 |
| 15600 | 05:59:48 | 19,704 | 02:42:10 |
| 15700 | 06:01:60 | 19,705 | 02:39:48 |
| 15800 | 06:04:10 | 19,707 | 02:37:27 |
| 15900 | 06:06:17 | 19,708 | 02:35:04 |
| 16000 | 06:08:18 | 19,710 | 02:32:38 |
| 16100 | 06:10:21 | 19,713 | 02:30:14 |
| 16200 | 06:12:48 | 19,713 | 02:27:60 |
| 16300 | 06:14:58 | 19,715 | 02:25:38 |
| 16400 | 06:17:09 | 19,714 | 02:23:18 |
| 16500 | 06:19:26 | 19,715 | 02:20:59 |
| 16600 | 06:21:36 | 19,717 | 02:18:38 |
| 16700 | 06:23:39 | 19,719 | 02:16:15 |
| 16800 | 06:25:57 | 19,720 | 02:13:57 |
| 16900 | 06:28:29 | 19,721 | 02:11:44 |
| 17000 | 06:30:55 | 19,723 | 02:11:44 |
| 17100 | 06:33:19 | 19,725 | 02:03:23 |
| 11100 | 30.00.10 | 10,720 | 02.01.10 |

| 45000 | 00.05.40 | 40 505 | 00 04 50 |
|-------|----------|--------|----------|
| 17200 | 06:35:40 | 19,725 | 02:04:56 |
| 17300 | 06:37:54 | 19,727 | 02:02:37 |
| 17400 | 06:40:24 | 19,728 | 02:00:22 |
| 17500 | 06:42:43 | 19,730 | 01:58:05 |
| 17600 | 06:44:57 | 19,733 | 01:55:45 |
| 17700 | 06:47:14 | 19,734 | 01:53:27 |
| 17800 | 06:49:27 | 19,735 | 01:51:08 |
| 17900 | 06:51:35 | 19,737 | 01:48:47 |
| 18000 | 06:53:39 | 19,738 | 01:46:25 |
| 18100 | 06:55:52 | 19,740 | 01:44:06 |
| 18200 | 06:58:01 | 19,741 | 01:41:46 |
| 18300 | 07:00:07 | 19,741 | 01:39:26 |
| 18400 | 07:02:40 | 19,739 | 01:37:11 |
| | | | |
| 18500 | 07:04:53 | 19,737 | 01:34:52 |
| 18600 | 07:07:11 | 19,736 | 01:32:35 |
| 18700 | 07:09:44 | 19,735 | 01:30:20 |
| 18800 | 07:12:01 | 19,733 | 01:28:02 |
| 18900 | 07:14:23 | 19,731 | 01:25:45 |
| 19000 | 07:16:51 | 19,730 | 01:23:29 |
| 19100 | 07:19:04 | 19,730 | 01:21:10 |
| 19200 | 07:21:27 | 19,728 | 01:18:53 |
| 19300 | 07:23:60 | 19,725 | 01:16:38 |
| 19400 | 07:26:18 | 19,726 | 01:14:20 |
| 19500 | 07:28:50 | 19,724 | 01:12:04 |
| 19600 | 07:31:13 | 19,724 | 01:09:47 |
| 19700 | 07:33:37 | 19,722 | 01:07:29 |
| 19800 | 07:35:47 | 19,721 | 01:05:10 |
| 19900 | 07:38:20 | 19,719 | 01:02:54 |
| 20000 | 07:40:44 | 19,719 | 01:00:37 |
| 20100 | 07:42:56 | 19,720 | 00:58:18 |
| 20200 | 07:45:08 | 19,720 | 00:55:59 |
| 20300 | 07:47:17 | 19,720 | 00:53:39 |
| 20400 | 07:49:17 | 19,720 | 00:51:19 |
| 20500 | 07:51:35 | 19,721 | 00:49:01 |
| 20600 | 07:53:55 | 19,721 | 00:45:01 |
| 20700 | 07:56:16 | 19,720 | 00:44:26 |
| | 07:58:14 | - | |
| 20800 | | 19,719 | 00:42:06 |
| 20900 | 08:00:33 | 19,719 | 00:39:48 |
| 21000 | 08:02:53 | 19,718 | 00:37:30 |
| 21100 | 08:04:54 | 19,718 | 00:35:11 |
| 21200 | 08:07:18 | 19,717 | 00:32:54 |
| 21300 | 08:09:40 | 19,716 | 00:30:36 |
| 21400 | 08:11:56 | 19,716 | 00:28:18 |
| 21500 | 08:14:09 | 19,716 | 00:25:60 |
| 21600 | 08:16:32 | 19,715 | 00:23:42 |
| 21700 | 08:18:49 | 19,715 | 00:21:24 |
| 21800 | 08:21:02 | 19,714 | 00:19:06 |
| 21900 | 08:23:18 | 19,714 | 00:16:48 |
| | | | |

```
22000
       08:25:44
                       19,713 00:14:30
22100
      08:28:19
                       19,711 00:12:13
22200
                       19,709 00:09:55
      08:30:33
22300
       08:33:16
                       19,698 00:07:37
                       19,696 00:05:19
22400
       08:35:44
22500
                       19,704 00:03:01
       08:38:06
                       19,712 00:00:43
22600
       08:39:59
```

1.5 Create ngrams

```
[10]: ngram_path = sec_path / 'ngrams'
stats_path = sec_path / 'corpus_stats'
for path in [ngram_path, stats_path]:
    if not path.exists():
        path.mkdir(parents=True)
```

```
[19]: unigrams = ngram_path / 'ngrams_1.txt'
```

```
[20]: def create_unigrams(min_length=3):
          texts = \Pi
          sentence_counter = Counter()
          vocab = Counter()
          for i, f in enumerate(clean_path.glob('*.csv')):
              if i % 1000 == 0:
                  print(i, end=' ', flush=True)
              df = pd.read_csv(f)
              df.item = df.item.astype(str)
              df = df[df.item.isin(sections)]
              sentence_counter.update(df.groupby('item').size().to_dict())
              for sentence in df.text.dropna().str.split().tolist():
                  if len(sentence) >= min_length:
                      vocab.update(sentence)
                      texts.append(' '.join(sentence))
          (pd.DataFrame(sentence_counter.most_common(),
                        columns=['item', 'sentences'])
           .to_csv(stats_path / 'selected_sentences.csv', index=False))
          (pd.DataFrame(vocab.most_common(), columns=['token', 'n'])
           .to_csv(stats_path / 'sections_vocab.csv', index=False))
          unigrams.write_text('\n'.join(texts))
          return [1.split() for 1 in texts]
```

```
[21]: start = time()
  if not unigrams.exists():
      texts = create_unigrams()
  else:
```

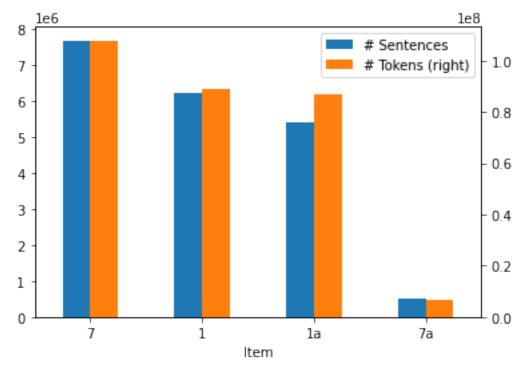
```
texts = [1.split() for 1 in unigrams.open()]
print('\nReading: ', format_time(time() - start))
```

0 1000 2000 3000 4000 5000 6000 7000 8000 9000 10000 11000 12000 13000 14000 15000 16000 17000 18000 19000 20000 21000 22000 Reading: 00:04:14

```
[22]: def create_ngrams(max_length=3):
          """Using gensim to create ngrams"""
          n_grams = pd.DataFrame()
          start = time()
          for n in range(2, max_length + 1):
              print(n, end=' ', flush=True)
              sentences = LineSentence(ngram path / f'ngrams {n - 1}.txt')
              phrases = Phrases(sentences=sentences,
                                min_count=25, # ignore terms with a lower count
                                threshold=0.5, # accept phrases with higher score
                                max_vocab_size=40000000, # prune of less common_
       →words to limit memory use
                                delimiter=b'_', # how to join ngram tokens
                                progress_per=50000, # log progress every
                                scoring='npmi')
              s = pd.DataFrame([[k.decode('utf-8'), v] for k, v in phrases.
       →export_phrases(sentences)],
                               columns=['phrase', 'score']).assign(length=n)
              n_grams = pd.concat([n_grams, s])
              grams = Phraser(phrases)
              sentences = grams[sentences]
              (ngram_path / f'ngrams_{n}.txt').write_text('\n'.join([' '.join(s) for_
       →s in sentences]))
          n_grams = n_grams.sort_values('score', ascending=False)
          n_grams.phrase = n_grams.phrase.str.replace('_', ''')
          n_grams['ngram'] = n_grams.phrase.str.replace(' ', '_')
          n_grams.to_parquet(sec_path / 'ngrams.parquet')
          print('\n\tDuration: ', format_time(time() - start))
          print('\tngrams: {:,d}\n'.format(len(n_grams)))
          print(n_grams.groupby('length').size())
```

```
[]: create_ngrams()
```

1.6 Inspect Corpus



```
[]: ngrams.head()
 []: ngrams.score.describe(percentiles=percentiles)
 []: ngrams[ngrams.score>.7].sort_values(['length', 'score']).head(10)
[15]: vocab = pd.read_csv(stats_path / 'sections_vocab.csv').dropna()
[16]: vocab.info()
     <class 'pandas.core.frame.DataFrame'>
     Int64Index: 200867 entries, 0 to 200868
     Data columns (total 2 columns):
          Column Non-Null Count
                                   Dtype
                  -----
          token
                  200867 non-null object
                  200867 non-null int64
      1
     dtypes: int64(1), object(1)
     memory usage: 4.6+ MB
[19]: vocab.n.describe(percentiles).astype(int)
[19]: count
                200867
     mean
                  1439
      std
                 22312
     min
      10%
                     1
      20%
                     2
      30%
                     3
      40%
                     4
      50%
                     7
      60%
                    12
      70%
                    24
      80%
                    61
      90%
                   260
     max
               2574572
      Name: n, dtype: int64
[20]: tokens = Counter()
      for 1 in (ngram_path / 'ngrams_2.txt').open():
          tokens.update(l.split())
[21]: tokens = pd.DataFrame(tokens.most_common(),
                           columns=['token', 'count'])
[22]: tokens.info()
```

```
<class 'pandas.core.frame.DataFrame'>
     RangeIndex: 230112 entries, 0 to 230111
     Data columns (total 2 columns):
         Column Non-Null Count
                                  Dtype
                 _____
      0
                 230112 non-null object
         token
      1
          count
                 230112 non-null int64
     dtypes: int64(1), object(1)
     memory usage: 3.5+ MB
[23]: tokens.head()
[23]:
           token
                    count
         million 2340187
     0
     1 business 1696732
     2 december 1512367
     3
         company 1490617
     4 products
                 1367413
[24]: tokens.loc[tokens.token.str.contains('_'), 'count'].describe(percentiles).
      →astype(int)
[24]: count
               29951
                 926
     mean
     std
                9611
     min
                   1
     10%
                  26
     20%
                  31
     30%
                  37
     40%
                  46
     50%
                  61
     60%
                  85
     70%
                 131
     80%
                 237
     90%
                 666
     max
              593859
     Name: count, dtype: int64
[25]: tokens[tokens.token.str.contains('_')].head(20).to_csv(sec_path /_
      [26]: tokens[tokens.token.str.contains('_')].head(20)
[26]:
                          token
                                 count
     46
                     year_ended
                                593859
             results_operations
     64
                                492047
     71
                 table_contents
                                436034
```

```
78
                company_s
                           412971
85
      financial_condition
                           396164
86
             common_stock
                           387629
107
                fair_value 341108
152
            united_states 276401
158
                cash_flows
                           266725
168
     financial_statements
                           255115
187
             interest_rate 234621
     approximately_million
188
                           234385
199
          adversely_affect
                           227984
223
                long_term 203600
238
               real_estate 192824
239
          material_adverse 192238
240
               fiscal_year 192189
243
            interest_rates 190754
248
                income_tax 186923
267
               natural_gas 178765
```

1.7 Get returns

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 22631 entries, 0 to 22630
Data columns (total 11 columns):

| # | Column | Non-Null Count | Dtype |
|---|--------------|----------------|----------------|
| | | | |
| 0 | cik | 22631 non-null | int64 |
| 1 | company_name | 22631 non-null | object |
| 2 | form_type | 22631 non-null | object |
| 3 | date_filed | 22631 non-null | datetime64[ns] |
| 4 | edgar_link | 22631 non-null | object |
| 5 | quarter | 22631 non-null | int64 |
| 6 | ticker | 22631 non-null | object |
| 7 | sic | 22461 non-null | object |
| 8 | exchange | 20619 non-null | object |
| 9 | hits | 22555 non-null | object |
| 10 | year | 22631 non-null | int64 |
| <pre>dtypes: datetime64[ns](1), int64(3), object(7)</pre> | | | |

```
[31]: idx = pd.IndexSlice
[32]: first = sec.date_filed.min() + relativedelta(months=-1)
     last = sec.date_filed.max() + relativedelta(months=1)
     prices = (prices
                .loc[idx[first:last, :]]
                .unstack().resample('D')
                .ffill()
                .dropna(how='all', axis=1)
                .filter(sec.ticker.unique()))
[33]: sec = sec.loc[sec.ticker.isin(prices.columns), ['ticker', 'date_filed']]
     price_data = []
     for ticker, date in sec.values.tolist():
         target = date + relativedelta(months=1)
          s = prices.loc[date: target, ticker]
         price_data.append(s.iloc[-1] / s.iloc[0] - 1)
     df = pd.DataFrame(price_data,
                       columns=['returns'],
                        index=sec.index)
[34]: df.returns.describe()
[34]: count
              11101.000000
                  0.022839
     mean
                  0.126137
     std
     min
                 -0.555556
     25%
                 -0.032213
     50%
                  0.017349
     75%
                  0.067330
                  1.928826
     max
     Name: returns, dtype: float64
[35]: sec['returns'] = price_data
     sec.info()
     <class 'pandas.core.frame.DataFrame'>
     Int64Index: 11375 entries, 0 to 22629
     Data columns (total 3 columns):
          Column
                     Non-Null Count Dtype
     ---
                     -----
      0
         ticker
                    11375 non-null object
      1
          date filed 11375 non-null datetime64[ns]
          returns
                  11101 non-null float64
```

memory usage: 1.9+ MB

```
dtypes: datetime64[ns](1), float64(1), object(1)
```

memory usage: 355.5+ KB

```
[36]: sec.dropna().to_csv(sec_path / 'sec_returns.csv', index=False)
```