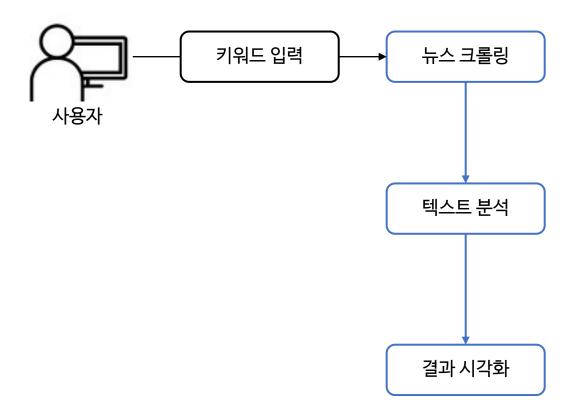
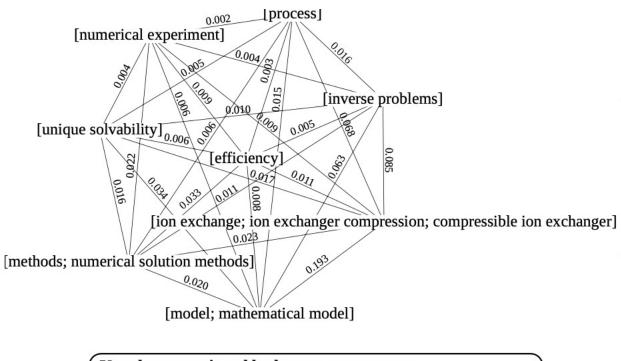
Minicycle

20210723

Goal





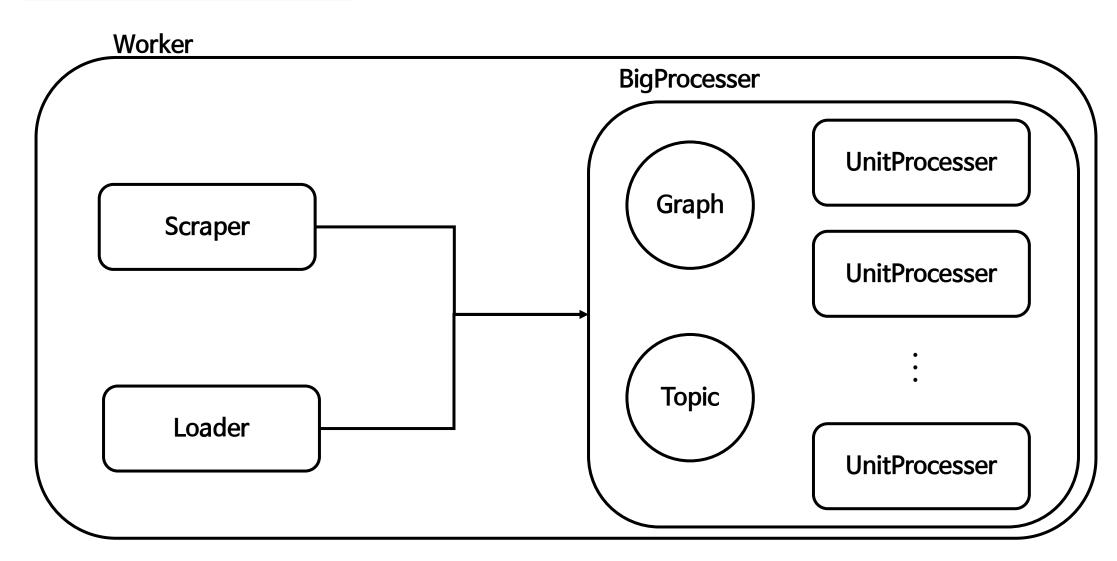
Keyphrases assigned by human annotators:

ion exchange; mathematical model; inverse problems; numerical solution methods; unique solvability; compressible ion exchanger; ion exchanger compression

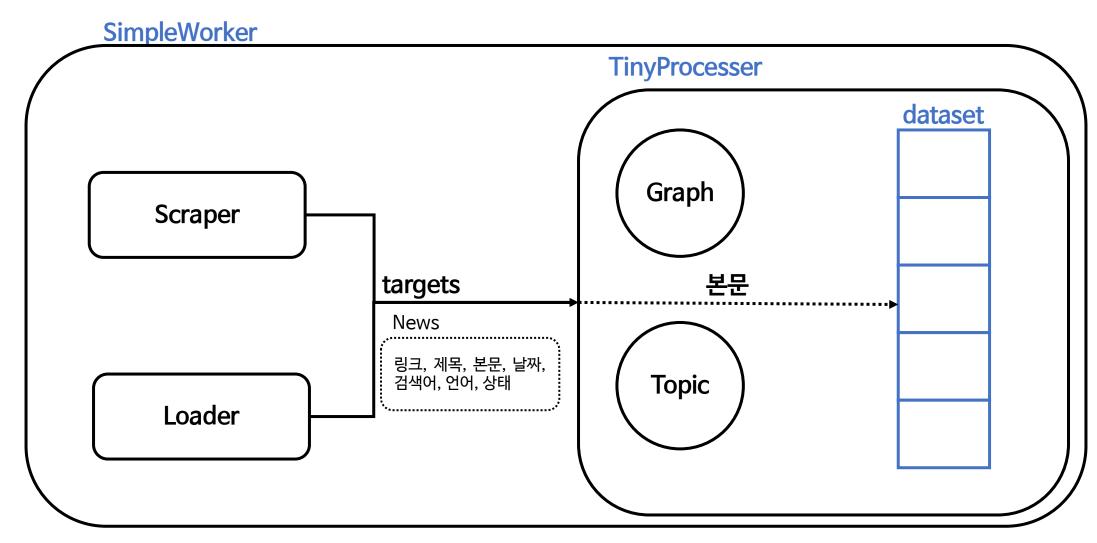
Keyphrases assigned by TopicRank:

ion exchange; mathematical model; inverse problems; numerical solution methods; process; unique solvability; efficiency; numerical experiment

Architecture



What changed?



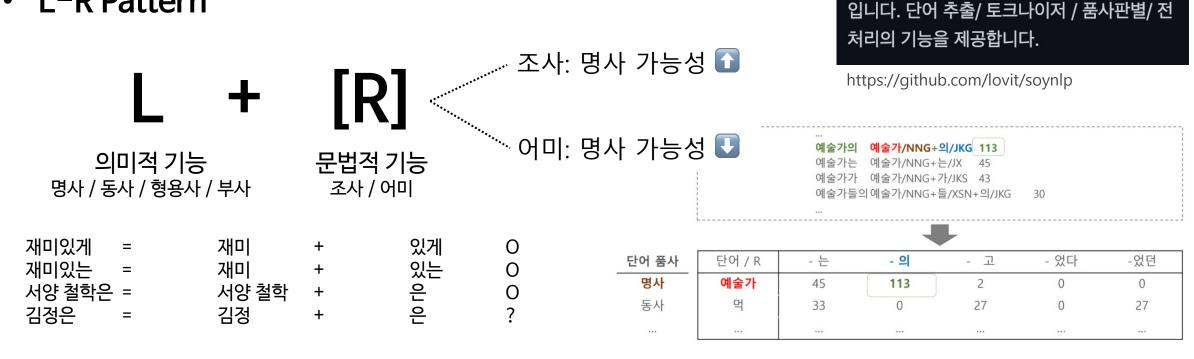
Workflow

Candidate Candidate Data Load Preprocessing **Filtering** Extraction Count-based Association Visualization analysis Ranking

Candidate Selection

```
def select_candidates(self):
    noun_extractor = LRNounExtractor_v2(verbose=True)
    nouns = noun_extractor.train_extract(self.dataset)
    self.candidates = nouns
```

L-R Pattern



soynlp

About

한국어 분석을 위한 pure python code 입니다. 학습데이터를 이용

하지 않으면서 데이터에 존재하는 단어를 찾거나, 문장을 단어열로 분

한국어 자연어처리를 위한 파이썬 라이브러리

해, 혹은 품사 판별을 할 수 있는 비지도학습 접근법을 지향합니다.

Count-based Ranking

```
	ext{tf}(t,d) = rac{f_{t,d}}{\sum_{t' \in d} f_{t',d}}, term frequency f_{t,d} igg/\sum_{t' \in d} f_{t',d}
def simpleTfldf(self):
      c = list(self.candidates.keys())
      f dataset = list()
      for i, in enumerate (self.dataset):
             f_dataset.append(helper.filter_text(self.dataset[i], c))
                                                                                            \operatorname{idf}(t,D) = \log rac{N}{|\{d \in D : t \in d\}|} \quad egin{array}{c} \log rac{N}{n_t} = -\log rac{n_t}{N} \end{aligned}
      total_words = sum([len(text.split()) for text in f_dataset])
      total num text = len(f dataset)
      for each cinc:
             for i, in enumerate (self.dataset):
                                                                                                                      tfidf(t, d, D) = tf(t, d) \cdot idf(t, D)
                          cnt = len(self.dataset[i].split(each_c)) - 1
                          num_words = len(self.datset[i].split()) - 1
                          self.tf_score[each_c] = self.tf_score.get(each_c, 0) + cnt / num_words
             self.idf_score[each_c] = sum([ 1 if each_c in dataset else 0 for dataset in f_dataset])
      self.tf_score.update([[candidate, tf/total_words] for candidate, tf in self.tf_score.items()])
      self.idf_score.update([[candidate, math.log(total_num_text / idf)] for candidate, idf in self.idf_score.items()])
      self.tf_idf_score = {key: self.tf_score[key] * self.idf_score[key] for key in self.tf_score.keys()}
```

inverse document frequency

Association Analysis

```
def pmi(self):
     corpus = list()
     c = list(self.candidates.keys())
     for doc in self.dataset:
           sents = kss.split_sentences(doc)
           for i, _ in enumerate(sents):
                      corpus.append(helper.filter_text(sents[i], c))
     x, idx2vocab = sent_to_word_contexts_matrix(
           corpus,
           windows=5.
           min tf=10,
           dynamic_weight=False,
           verbose=True
     pmi_dok, px, py = pmi(
           min_pmi=0,
           alpha=0.0001
     return idx2vocab, pmi_dok
```

	from	swerve	of	shore	to	bend	of	bay	,	brings
Window: 3	4	3	2	1	0	1	2	3	4	5
Scaling: flat	0	1	1	1	1	1	1	1	0	0
Scaling: $\frac{1}{n}$	0	$\frac{1}{3}$	$\frac{1}{2}$	$\frac{1}{1}$	1	$\frac{1}{1}$	$\frac{1}{2}$	$\frac{1}{3}$	0	0

Example corpus:

- I like deep learning.
- I like NLP.
- I enjoy flying.

counts	1	like	enjoy	deep	learning	NLP	flying	
1	0	2	1	0	0	0	0	0
like	2	0	0	1	0	1	0	0
enjoy	1	0	0	0	0	0	1	0
deep	0	1	0	0	1	0	0	0
learning	0	0	0	1	0	0	0	1
NLP	0	1	0	0	0	0	0	1
flying	0	0	1	0	0	0	0	1
	0	0	0	0	1	1	1	0

$$\mathrm{pmi}(x;y) \equiv \log rac{p(x,y)}{p(x)p(y)} = \log rac{p(x|y)}{p(x)} = \log rac{p(y|x)}{p(y)}.$$

Visualization





2016년 2021년

Visualization

