

keywords selection

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0.0.1 Airbnb's description of all aspects.

- Overall Experience What was your guest's overall experience?
- Cleanliness Did your guests feel that your space was clean and tidy?
- Accuracy How accurately did your listing page represent your space?
- Value Did your guest feel your listing provided good value for the price?
- Communication How well did you communicate with your guest before and during their stay?
- Check-in How smoothly did their check-in go?
- Location How did guests feel about your neighborhood?

```
In [1]: # use wordnet find keywords
        from nltk.corpus import wordnet as wn
        import re

In [2]: # functions for extracting the keywords of an aspect.
        def synset_description(word):
            for synset in wn.synsets(word):
                print("%s : %s" % (synset.name(), synset.definition()))

        def synonym_words_in(word, include_list):
            thesaurus = []
            for synset in wn.synsets(word):
                if synset.name() in include_list:
                    thesaurus += synset.lemma_names()

            return list(set(thesaurus))

        def synonym_words_ex(word, exclude_list):
            thesaurus = []
            for synset in wn.synsets(word):
                if synset.name() not in exclude_list:
                    thesaurus += synset.lemma_names()

            return list(set(thesaurus))

        def anton_list(syn_include_list):
```

```

anton_list = []
for synset in syn_include_list:
    for l in wn.synset(synset).lemmas():
        if l.antonyms():
            anton_list += l.antonyms()

return anton_list

def synonyms_of_antonyms(anton_list):
    antonyms = []
    pattern = r"Lemma\('[a-z]+\.[a-z]+\d+)"

    for anton in anton_list:
        antonyms += wn.synset(re.findall(pattern, str(anton))[0]).lemma_names()

    return list(set(antonyms))

```

1 Determine the keywords relevant to aspect location.

The Airbnb officially determine the aspect location as: * Location How did guests feel about your neighborhood?

So, I will use the two main keywords, location and neighborhood, to find synonyms for further review text processing.

1.0.1 Firstly, I will look at the synset description of location.

This is to remove the irrelevant synset and then increase the accuracy of the set of synonyms.

```
In [3]: synset_description('location')
```

```

location.n.01 : a point or extent in space
placement.n.03 : the act of putting something in a certain place
localization.n.01 : a determination of the place where something is
location.n.04 : a workplace away from a studio at which some or all of a movie may be made

```

1.0.2 Then, exclude not necessarily relevant synset and get synonyms.

From above, we can see the location.n.04, localization.n.01, placement.n.03 are not necessarily in accord to the Airbnb's definition of aspect location.

So, I will ignore synonyms in those synset.

As a result, the synonyms I will use later will be:

```
In [4]: synw1 = synonym_words_in('location', ['location.n.01'])
```

1.0.3 Do the same thing for word neighborhood.

```
In [5]: synset_description('neighborhood')
```

vicinity.n.01 : a surrounding or nearby region
 neighborhood.n.02 : people living near one another
 region.n.04 : the approximate amount of something (usually used prepositionally as in `in the
 neighborhood.n.04 : an area within a city or town that has some distinctive features (especial

```
In [6]: synw2 = synonym_words_ex('neighborhood', ['neighborhood.n.02'])
```

1.04 Find antonyms and their synonyms

```
In [7]: # this is a list of synsets that are selected as useful synset in finding synonyms.
        syn_include_list = ['location.n.01', 'vicinity.n.01', 'region.n.04', 'neighborhood.n.04']

        # a list of antonym synsets for above list.
        anton_list(syn_include_list)

        # all lemma words in upper antonym synsets.
        antw = synonyms_of_antonyms(anton_list(syn_include_list))
```

1.05 Finalize the aspect synonyms set of location aspect.

```
In [8]: location_synonyms = list(set(synw1 + synw2 + antw))
        location_synonyms
```

```
Out[8]: ['location',
        'neighbourhood',
        'neighborhood',
        'neck_of_the_woods',
        'region',
        'locality',
        'vicinity']
```

2 Determine the keywords relevant to aspect cleanliness.

The Airbnb officially determine the aspect cleanliness as: * Cleanliness Did your guests feel that your space was **clean** and **tidy**.

So, I will use the three main keywords, clean, tidy and cleanliness, to find synonyms for further review text processing.

The process will be the same as location.

2.01 cleanliness

```
In [9]: synset_description('cleanliness')
```

cleanliness.n.01 : the habit of keeping free of superficial imperfections
 cleanliness.n.02 : diligence in keeping clean

```
In [10]: synw1 = synonym_words_ex('cleanliness', '')
```

2.0.2 tidy

```
In [11]: synset_description('tidy')
```

```
tidy.n.01 : receptacle that holds odds and ends (as sewing materials)
tidy.v.01 : put (things or places) in order
tidy.a.01 : marked by order and cleanliness in appearance or habits
kempt.s.01 : (of hair) neat and tidy
goodly.s.01 : large in amount or extent or degree
```

```
In [12]: synw2 = synonym_words_in('tidy', ['tidy.v.01', 'tidy.a.01'])
```

2.0.3 clean

```
In [13]: # synset_description('clean')
```

```
In [14]: synw3 = synonym_words_in('clean', ['clean.v.01', 'houseclean.v.01', 'clean.v.05', 'cl
        'scavenge.v.04', 'clean.a.01'])
```

2.0.4 Find antonyms and their synonyms

After I used the synonyms as keywords to extract sentences that are relevant to cleanliness aspect from all review text, I found that the accuracy is pretty high.

However, it neglects some sentences that used the antonyms of the keywords like "the kitchen needed a good scrubbing - it was **dirty**".

So, I decided to also add antonyms and their synonyms into keywords list using the `anton_list` and `synonyms_of_antonyms` function.

```
In [15]: # this is a list of synsets that are selected as useful synset in finding synonyms.
        syn_include_list = ['cleanliness.n.01', 'cleanliness.n.02', 'tidy.v.01', 'tidy.a.01',
        'clean.v.01', 'houseclean.v.01', 'clean.v.05', 'clean.v.08',
        'scavenge.v.04', 'clean.a.01']
```

```
In [16]: # a list of antonym synsets for above list.
        anton_list(syn_include_list)
```

```
Out[16]: [Lemma('uncleanliness.n.01.uncleanliness'),
        Lemma('untidy.a.01.untidy'),
        Lemma('dirty.v.01.dirty'),
        Lemma('dirty.a.01.dirty')]
```

```
In [17]: # all lemma words in upper antonym synsets.
        antw = synonyms_of_antonyms(anton_list(syn_include_list))
```

2.0.5 Finalize the aspect synonyms set of cleanliness aspect.

```
In [18]: cleanliness_synonyms = list(set(synw1 + synw2 + synw3 + antw))
        cleanliness_synonyms
```

```
Out[18]: ['colly',
          'neaten',
          'straighten',
          'soil',
          'tidy',
          'make_clean',
          'tidy_up',
          'clean',
          'begrime',
          'straighten_out',
          'bemire',
          'square_away',
          'unclean',
          'clean_up',
          'untidy',
          'grime',
          'dirty',
          'cleanliness',
          'scavenge',
          'soiled',
          'clean_house',
          'houseclean',
          'uncleanliness']
```

```
In [200]: # #
          # wn.synset('goodly.s.01').hypernyms()
          # #
          # wn.synset('cleanliness.n.02').lemmas()[0].antonyms()

          # def word_similarity(w1, w2):
          #     token1 = nlp(w1)
          #     token2 = nlp(w2)
          #     return token1.similarity(token2)

          # antonym_words('clean.a.01')
```