# Term project presentation (Korean SAT-English part topic modeling)

소프트웨어학부 20186663

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#### Title

• The subject of the project is to guess the correct answer to the English topic(Korean SAT topic problems).

22. 다음 글의 주제로 가장 저절한 것은?

When we hear a story, we look for beliefs that are being commented upon. Any story has many possible beliefs inherent in it. But how does someone listening to a story find those beliefs? We find them by looking through the beliefs we already have. We are not as concerned with what we are hearing as we are with finding what we already know that is relevant. Picture it in this way. As understanders, we have a list of beliefs, indexed by subject area. When a new story appears, we attempt to find a belief of ours that relates to it. When we do, we find a story attached to that belief and compare the story in our memory to the one we are processing. Our understanding of the new story becomes, at that point, a function of the old story. Once we find a belief and connected story, we need no further processing; that is, the search for other beliefs stops.

- ① the use of a new story in understanding an old story
- 2 the limits of our memory capacity in recalling stories
- ③ the influence of new stories on challenging our beliefs
- the most efficient strategy to improve storytelling skills
- (5) the role of our existing beliefs in comprehending a new story



To guess topic/answer

#### Data

• I used title, subject problems in Korea English SAT

1. 다음 글의 주제로 가장 적절한 것은? [3월

During the last two decades many developing countries have joined the global tourism market the Iron Curtain. These countries had suffered from negative public and media image which made it challenging for them to compete over is a major obstacle in attracting tourists. the case of destinations suffering from prolonged expect any target audience to visit a destination and "put aside" these long-lasting negative images and stereotypes, just because of an advertising campaign or other promotional effort. Tackling prolonged negative place images is crucial for developing tourism in Africa, the Middle East, Latin America, Eastern Europe and Asia Although these destinations differ greatly in the eyes of many potential tourists they all suffer from weak place images, negative stereotypes and problematic perceptions.

- ① growing conflicts between tourists and local
- 2 roles of media in shaping the global trend in tourism
- 3 necessity of global cooperation for sustainable tourism (4) importance of the tourism industry in
- national economy
- destination images
- In the last few years cartography has been knowing she is an American girl living in a slipping from the control of the powerful elites certain specific community, with parents of such

several hundred years. You probably already have noticed this with the emergence of fantastically popular mapping applications such the great map houses of the West, national and local governments, the major mapping and GIS companies and to a lesser extent academics have been confronted by important dominance. For example, as Google Earth has shown the actual business of manmaking of collecting spatial data and mapping it out is high-quality residents and investors. However, in passing out of the hands of the experts. The interactive 3D map, is now available to anyone with a home computer and a broadband Internet

- 1 various ways of collecting geographic data 2 various technologies involved in map-making

- 3 diminishing dominance of cartographic elites and its cause

다음 글의 주제로 가장 적절한 것은? [3D]<sup>3)</sup>

If we look out of a window in winter we might see millions of identical snowflakes fluttering by. However, if we took a magnifying glass and looked at the flakes separately, we would soon (5) developing nations' need to improve that each flake had a distinct shape that no other flake duplicated exactly. The same is true of human beings. We can tell quite a lot about what Susan will experience just by the fact that 2. 다음 글의 주제로 가장 적절한 것은? [3J] 2) she is human. We can tell even more by 1. 다음 글의 제목으로 가장 적절한 것은? [3월

With the general accessibility of photocopiers in student libraries, students tend to copy the in humans in the past 60,000 years. Rather, relevant material for later use. In such cases the students are not always selective about seems to require consistent environmental what they copy. Often useless material is pressures over tens of thousands of years. the night before an exam or essay due date. In evolution but also of what our collective future addition, when most people photocopy material might hold. For a number of the challenges from books, they feel as if they have actually currently facing our species — global climate accomplished something. After all, a few change and many infectious diseases, for photocopied pages in their notebook now represent information that used to be in a big. thick book. The reality of the situation is that nothing significant has been accomplished yet. The student only has the information in a transportable form. He or she has not learned anything from the material. The information content of the photocopied sheets is just as foreign as if it had been left on the library

- Information Accessibility Leads to Intellectual
- 2 Reasons You Should Keep Study Material After Exams 3 Photocopied Material: Not a Sign of
- @ Careless Photocopying May Be Considered
- D Photocopier: A Contributor to Information

As an evolutionary biologist, I am often asked

Our data suggests that the classic natural selection scenario, in which a single beneficial population, has actually occurred relatively rarely this mechanism of evolutionary change usually understanding not only of recent human example - natural selection probably occurs too slowly to help us much.

- ① Effect of Social Pressure on Mar

- @ How to Solve Environmental Problems
- ® Power of Evolution to Overwhelm Culture

다음 글의 제목으로 가장 적절한 것은?

books designed to help you get more money. The books say that greater wealth often provides more happiness. New research shows that the law of diminishing marginal utility also applies to money. In a national sample of Americans, individuals thought that their satisfaction with life would double if they made \$50,000 rather than \$25,000 twice as much money, twice as much happy. But the data 2. 다음 글의 제목으로 가장 적절한 것은? [3J] only 9 percent more satisfied than those making \$25,000 Around the world income has surprisingly little influence on whether people whether humans are still evolving today. We smile, laugh, and experience enjoyment on a certainly are. But the answer to the question of typical day! And in the United States, once

Subject problem

Title problem

# Data preprocessing

 I used title, subject problems in Korea English SAT 1. first all, I processed .pdf data to .txt data

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1. 다음 교의 주제로 가장 적절한 것은? [3월 서울시]<sup>1)</sup> fantastically popular mapping applications such During the last two decades many developing as Google Earth. The elites — the map experts. 

in tourism

1 necessity of global cooper
sustainable tourism

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2 Natural Selection Visil Save Humans substance Evolution
Institution of the Surface Institution of Surface Insti foreign as if it had been left on the library 3. 다음 글의 제목으로 가장 적절한 것은?

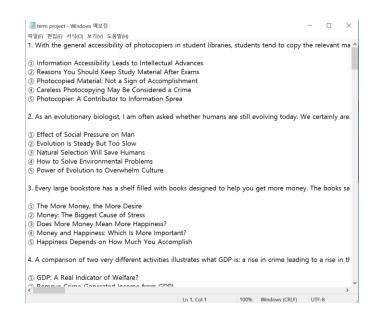
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a Crime
Photocopier: A Contributor to Information

attribute on with life would double if they made \$50,000 rather than \$25,000; twice as much revealed that people who earned \$50,000 were 2. CIEST AND MINED 2019 MINED 2019 TO As an evolutionary belogist. I am other saled whether humans and stell evolving below; the certainty are but the certainty are but the certainty are but the certainty are but the sale voice was an elementary in the more consideration of to only we are changing in the more consideration of complete are sense ground \$75,000 per year. Sense of the contract of the contract of complete are sense ground \$75,000 per year. Sense of the contract of the contract of contract of the contract of sense of the contract of the contract of the contract of contract of the contract of sense of the contract of the contract of sense of the contract of the contract of sense of the contract of the contract of sense of



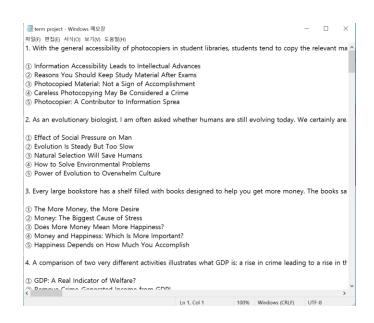


.pdf data

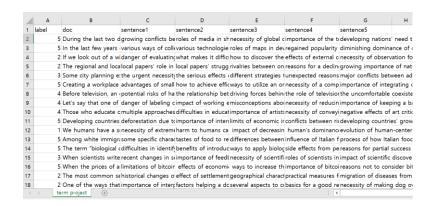
.txt data

# Data preprocessing

• I used title, subject problems in Korea English SAT 2. And then, I processed .txt data to .csv data

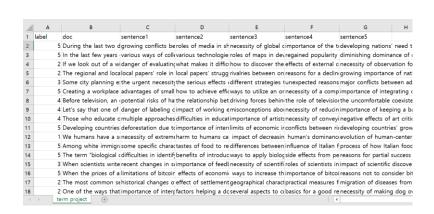




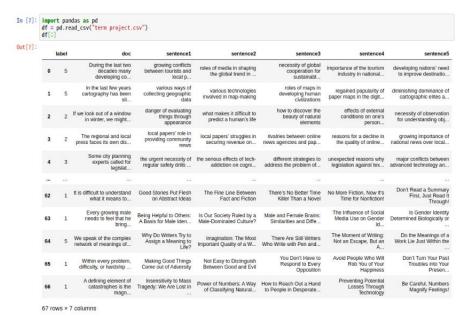


### Data preprocessing

• I used title, subject problems in Korea English SAT 3. Finally, I used .csv data in pandas for data analysis







.csv data

Data by pandas

#### Vectorization

- I used Count Vectorization and TF-IDF Vectorization
  - Count Vectorizer: Simply create a vector with the frequency of words used.
  - TF IDF Vectorizer: Weight is given by comparing the number of words used in the document with the number of words used in the entire document.

$$w_{x,y} = tf_{x,y} \times log(\frac{N}{df_x})$$



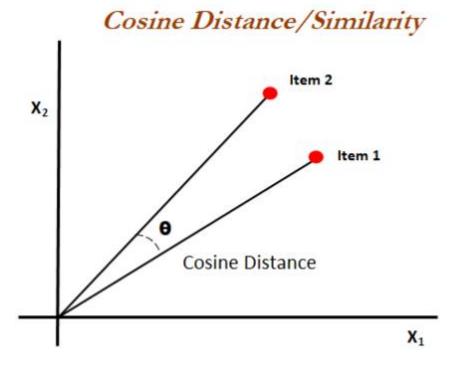
 $tf_{x,y}$  = frequency of x in y

 $df_x$  = number of documents containing x

N = total number of documents

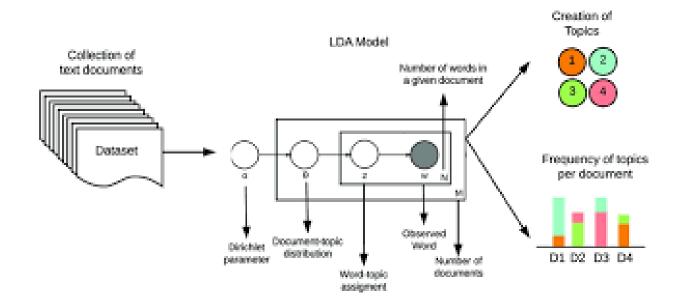
### Cosine Similarity

• I used cosine similarity to measure the similarity between two vectorized documents.



#### Latent Dirichlet Allocation

- Latent Dirichlet allocation(LDA) is one of the probabilistic topic model techniques for describing what topics exist in each document for a given document.
- In this project, I used LDA to extract words related to the topic.



#### **TF-IDF Vectorization**

TF-IDF Vectorizer → cosine similarity

#### TF-IDF Vectorization

```
my label = []
                 for i in range(len(df doc)):
                     dfdf = df doc.iloc[i]
                     tfidf vect = TfidfVectorizer(stop words='english',ngram range=(1,2), max df=100)
        Vectorizer | dd = tfidf_vect.fit_transform(dfdf)
                     dd = dd.todense()
                     a num = 0
                     vect0 = np.array(dd[0]).reshape(-1,)
                     for j in range(1,6):
                         vect1 = np.array(dd[j]).reshape(-1,)
                         b = cos similarity(vect0, vect1)
Cosine similarity
                         if b > a:
                             a = b
                             a num =
                     my label.append(a num)
                 print('accuarcy', metrics.accuracy score(df label, my label))
                 accuarcy 0.47761194029850745
     accuracy
```

#### Count Vectorization

Count Vectorizer → cosine similarity

#### Count Vectorization

```
my label = []
              for i in range(len(df doc)):
                   dfdf = df doc.iloc[i]
                  tfidf vect = CountVectorizer(stop words='english',ngram range=(1,2), max df=100, min df=1)
      Vectorizer | dd = tfidf_vect.fit_transform(dfdf)
                  dd = dd.todense()
                   a num = 0
                   vect0 = np.array(dd[0]).reshape(-1,)
                  for j in range(1,6):
                      vect1 = np.array(dd[j]).reshape(-1,)
                      b = cos similarity(vect0, vect1)
Cosine similarity
                           a = b
                           a num = j
                  my label.append(a num)
              print('accuarcy', metrics.accuracy score(df label, my label))
```

accuracy 0.4925373134328358

### TF-IDF Vectorization & LDA

TF-IDF Vectorizer → LDA → cosine similarity

TF-IDF Vectorization & Latent Dirichlet Allocation

```
from sklearn.decomposition import LatentDirichletAllocation
                   my label = []
                   for i in range(len(df doc)):
                       dfdf = df doc.iloc[i]
                       tfidf vect = TfidfVectorizer(stop words='english',ngram range=(1,2), max df=100)
        vectorizer
                       dd = tfidf vect.fit transform(dfdf)
                       lda = LatentDirichletAllocation(n components=30,random state=0)
              LDA
                       lda.fit(dd)
                       dd = lda.components .T
                       a num = 0
                       vect0 = np.array(dd[0]).reshape(-1,)
                       for j in range(1,6):
                           vect1 = np.array(dd[j]).reshape(-1,)
                           b = cos similarity(vect0, vect1)
Cosine similarity
                           if b > a:
                               a = b
                               a num = j
                       my label.append(a num)
                   print('accuarcy', metrics.accuracy score(df label, my label))
                   accuarcy 0.3880597014925373
 accuracy
```

#### Count Vectorization & LDA

Count Vectorizer → LDA → cosine similarity

#### Count Vectorization & Latent Dirichlet Allocation

```
from sklearn.decomposition import LatentDirichletAllocation
                   my label = []
                   for i in range(len(df doc)):
                       dfdf = df doc.iloc[i]
                       tfidf vect = CountVectorizer(stop words='english',ngram range=(1,2), max df=100)
        vectorizer
                       dd = tfidf vect.fit transform(dfdf)
                       lda = LatentDirichletAllocation(n components=10,random state=0)
               I DA
                       lda.fit(dd)
                       dd = lda.components .T
                        a = 0
                       vect0 = np.array(dd[0]).reshape(-1,)
                       for | in range(1,6):
                           vect1 = np.array(dd[j]).reshape(-1,
                           b = cos similarity(vect0, vect1)
Cosine similarity
                                a num = j
                       my label.append(a num)
                   print('accuarcy', metrics.accuracy_score(df_label, my_label))
  accuracy
                   accuarcy 0.26865671641791045
```

#### Conclusion

- If using this programming, we can get about 50 points in the English test.
- It was difficult to find a meaningful one in each sentence because the amount of each sentence was too small length compared to the document. Therefore, the value of count vectorization, which simply vectorizes into the number of words, came out the highest.
- In short sentences, the meaning of the sentence being LDA was difficult to find, so the process of LDA was hindered.

# Thanks