

**Analyzing Topics in Classroom of the Elite:**

*How Japanese Light Novels Unveil the Shortcomings of Topic Modelling*

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Course: Computational Literary Studies

Academic Year: 2021-2022

# Abstract

# This paper will make use of topic models to measure the statistical relationship between topic clusters for each school trimester discussed in the light novel *Classroom of the Elite*. The research found that while some statistically meaningful or unambiguous trends in the generated topic distributions are present, others are hard to interpret. The paper goes on to discuss the limits of the study and those of topic modelling in general before reaching a conclusion.

# Introduction

“Though equality is a complete lie, we should not accept inequality either” (Syougo, 2015). *Classroom of the Elite* is a Japanese novel that philosophizes about the role education has in enforcing equality as well as inequality. In short, it tells the story of Japanese high school students striving to graduate school as the top class or ‘class a’ of their year. The students that manage to achieve this goal are, after all, promised to gain any job or stature they would like after graduation. The story revolves around protagonist Ayanokouji Kiyotaka and his classmates starting their first year of high school in the bottom class or ‘class d’. While facing challenges measuring their physical, academic and mental strength they too, aim to reach the famed ‘class a’- graduation (ibid, 2015). This series of light novels thus tells a story spanning over a three-year time period. During this time the characters mature and get to know more people. This paper will aim to reveal if that growth can be derived from a change in topics discussed as the main characters experience each consecutive school trimester. In order to attain sets of comparable topics, a mixed-membership model, better known as a topic model will be utilized as the main tool of analysis (Karsdorp, Riddell and Kestemont, 2021).

# Datasets

The *Classroom of the Elite* is an ongoing series of Japanese light novels written by Kinugasa Syougo and illustrated by Tomose Shunsaku that started being published in 2015. So far, the series includes 22 volumes. Light novels are a specific type of visual novel originating in Japan, their illustrations commonly cohere to manga art styles. These novels tend to cater to a young adult audience around the age of high school students (Enomoto, 2018). When this study was executed only 15 volumes of these light novels were translated to English. In addition, many of these translations are fan-made since the official translations tend to come out rather slowly. Therefore, international fans who cannot wait to read the story have collectively asked for fan-made translations. These translations get posted online and are deleted once the official translations come out (Graze et al. , 2021).

In this study, the words *main data* will refer to a self-made tab separated dataset. It contains fan-made translations of all separate chapters of each volume of *Classroom of the Elite* up to the 15th (Syougo, 2015, Graze et al. , 2021). These translations were copy pasted from the internet and their usage falls under the concept of fair use (Stim, 2021). Each translator was credited by means of their pen name in the *translator* column. The following table (1) exemplifies the first two rows of the dataset:

|  |  |  |  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- |
| Table 1 | | | | | | | | | | | |
| ID | Volume | Volume  \_Type | Chapter  \_Type | Chapter  \_Num | Publishing  Date | Title | Text | Translator | Editor | Proof-  reader | Trimester |
| 1\_P | 1 | Main | Prologue | 0 | 2015-05-25 | Title of chapter 1 volume 1 | Text in chapter 1 volume 1 | qbomb | NaN | NaN | 1 |

Each chapter was attributed to a unique ID number referring to the volume number and chapter type. The volume type clarifies whether a chapter belongs to a main or sub volume and the trimester column indicates what school trimester the protagonist finds himself in for each chapter. A last column worth mentioning is the chapter type column which can be divided into the following categories: Prologue, Chapter, Epilogue, Afterword and Bonus. For this study, each sub volume and each chapter belonging to the Afterword and Bonus category were separated from the main data and added to the *pretraining data*.

Pretraining data is needed in order to pretrain any topic model. This pretraining data should be large in size and as similar to the main data, in terms of content and vocabulary used, as possible (Bianchi, Terragni and Hovy, 2021). The topic model will be able to identify a set of topics based on the pretraining data. Afterwards, the model can assess whether these identified topics are present in the main data and if so, it detects to what degree they are present (Karsdorp, Riddell and Kestemont, 2021, Newman, Chemudugunta, Smyth and Steyvers, 2006). In this case, the pretraining data contains the aforementioned bonus chapters, afterwords of *Classroom of the Elite*. Though they are not part of the main story entailed in the volumes, they are very closely related in terms of characters and setting. Thus, they are fit to be used as pretraining data. The sub volumes of Classroom of the Elite will also be used as pretraining data since they entail content regarding vacation periods which do not take place during a school semester. However, their total token count is on the small side (193,177 tokens). Thus, on their own they do not suffice as pretraining data. As a solution to this problem, a selection of data from the Project Gutenberg Corpus was added to the pretraining data (Gerlach and Font-Clos, 2020). This selection contains 23 boarding school related books (1,402,919 tokens) which were copied from the Project Gutenberg Corpus and transformed into a tab separated format (tsv) making it easier to merge with the bonus and afterword chapters from *Classroom of the Elite*. The resulting tsv file is exemplified in the table

|  |  |  |  |
| --- | --- | --- | --- |
| Table 2 | | | |
| Filename | Author | Title | Text |
| PG24025\_tokens | Angela Brazil | Title of the book | Text included in the book |

(2) below:

As mentioned, a topic model needs to be pretrained on a large set of data similar to the data the study will be performed on. In this case, the best pretraining data would have been a large dataset containing Japanese light novels, mostly revolving around Japanese teens. This dataset however, does not exist in a publicly accessible environment, if it even exists at all. Time- and work limitations prevented me from making this type of dataset myself so an attempt was made to find the next-best alternative. While the bonus chapters, afterwords and sub volumes used as pretraining data were excellent in terms of comparability to the main data, they were limited to 193,177 tokens. Extra data was needed. However, adding more data was not simple. As mentioned, there is a lack of data similar to the contents of *Classroom of the Elite*. It is not a simple ‘school story’. It has a fictive Japanese school setting with unique school rules that lead to more than one survival-like exam staged on an island. When adding more generic data to pretraining data, the topics generated failed to represent topic trends in *Classroom of the Elite*. The only accessible data that could be seen as suitable was a selection of boarding school related data from the Project Gutenberg Corpus (Gerlach and Font-Clos, 2020). While still not the best fit, since they do not take place in a Japanese school system, this data could still provide topics befitting a boarding school setting which is present in the *Classroom of the Elite* series. Merging these two data caused the pretraining data to count 1,596,096 tokens instead. While the amount of pretraining data was almost quadrupled, questions could still be raised as to whether this is enough data for the topic model to be trained on. However, I opted not to add more data that would not suit the vocabulary or content of the main data.

# Hypotheses

Before delving into topics that can be derived computationally, this short section aims to explain which topics and topic trends are expected to show up. The *Classroom of the Elite* revolves mostly around how each separate class and its leaders handle obstacles on the path towards becoming or staying ‘the best’. Being ‘the best’ in this case means garnering the largest amount of class points by the time of their graduation ceremony. Gaining class points can been done in multiple ways but the most notable one is by outperforming other classes in exams. In this type of school system, class atmospheres and cooperation are of the utmost importance. At the start of the story, the students of class D have yet to realize this. They do not know each other yet and refuse to take the school system seriously. In the first trimester of school, the protagonist is mostly concerned with solving internal class issues while ensuring his class does not fall behind any more than it already has. Aside from this goal, the protagonist also aims to stay hidden as a ‘leader behind the scenes’. He does not have a deep relationships with other students yet at this point. During the second trimester, the protagonist establishes a study group that quickly grows into his first friend group. The reader can see him grow closer to his fellow students. In the third trimester, most 1st year class leaders have realized that Ayanokouji is the ‘shadow’ leader of class D. He gets challenged by the leader of class A, amongst others on multiple occasions. The 1st years start to interact more with each other and with senior students. Soon after, the oldest students or the 3rd years graduate after taking part in an exam in which all students in the school were involved. The fourth trimester mostly revolves around the introduction of new 1st year students and a survival exam including students from all school years that takes place on an island. Ayanokouji also gets negative attention from the new headmaster (Syougo, 2015). Topic clusters related to these events, specific to each trimester are expected to be portrayed within the following research.

# Methodology

This section will concisely describe what steps were taken in the coding stage of this project. First of all, the main data and the pretraining data were preprocessed so they could optimally be interpreted by the topic modelling algorithm. All stop words present in the English nltk stop words list were removed from the text data for both datasets (Bird, Klein and Loper, 2009). Markers of politeness typical of the Japanese language such as ‘san’ and ‘kun’ were added to this list. Next to that, all text data was lowercased and special characters such as brackets, ampersands and hashtags in addition to punctuation(', !, ", -, ., :, ;, etc.) were removed.

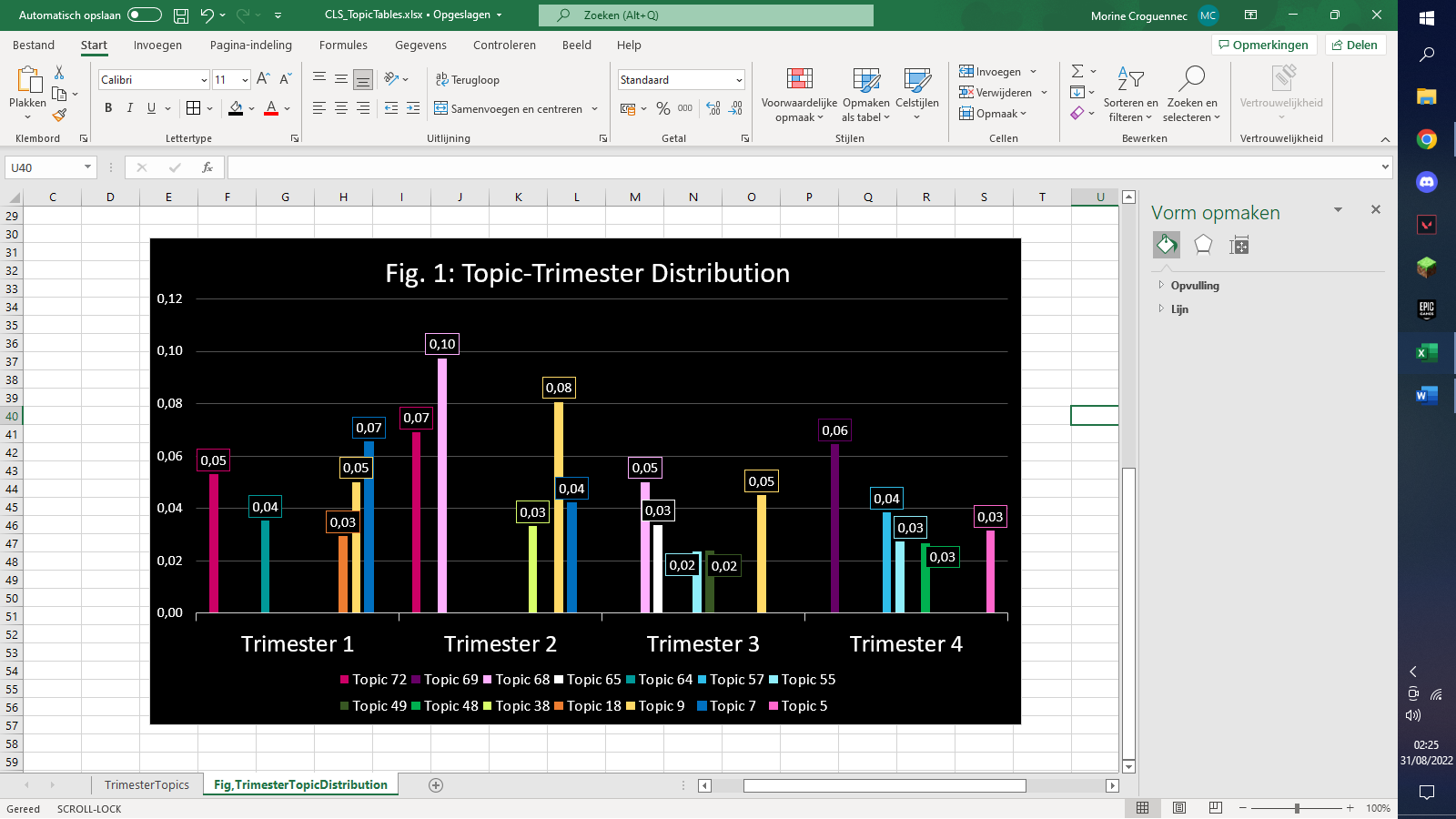
After this process was executed, a Non-Negative Matrix Factorization (NMF) model was trained and fitted on the pre-training data. This model was chosen for its interpretably and effectivity as a topic model (Zhang, 2012). Any word present in the pretraining data that occurred in at least 9 of the 84 documents (10 percent) and appeared in less than 50 percent or 42 of those 84 documents were viable to be included in the vocabulary list for the model. In this case, a document refers to either one of the 23 books in the Gutenberg selection or one of the 61 chapters of the pretraining selection of *Classroom of the Elite*. If a word only occurred in 10 percent of the documents, it was assumed that it would not be a marker of the pretraining data’s general content. They were not suited for the formation of topic clusters based on the topics generally present in this data. The 50 percentile was chosen after a process of trial and error, if multiple topics included the same set of words combined differently, the maximum document frequency was lowered. The resulting vocabulary list counted a total of 5,000 words, an amount suitable to form an extensive vocabulary of content words out of the relatively small pretraining dataset.

Once the NMF model was trained, a total of 75 topics had been derived from the training data. This parameter as well, was chosen through a process of trial and error. Choosing a larger amount like 100 or 150 topics mostly lead to nonsensical outcomes where the words inside a topic cluster could not be related to one another. Generating a large amount of topics would also not be suitable based on the fact that the pretraining data only counts 1,596,096 tokens. It is important to mention that if any of the mentioned variables would have been adjusted, the results of the research would also be influenced. Subsequently, any results concluded from this research should be taken with a grain of salt.

Each topic can be seen as a cluster of words that are used in a similar context within the pretraining data (Newman, Chemudugunta, Smyth and Steyvers, 2006). Once these topics were defined, the NMF model was fitted on the main project data. Resulting from this process, a data frame containing a measure of *presence* of each topic in each chapter of *Classroom of the Elite* was devised. Thus, it became possible to measure which topics were most present in each chapter as well as in each set of chapters present in the main *Classroom of the Elite* dataset (Meeks and Weingart, 2012). According to the proposed hypotheses, four separate groupings of chapters were made. They cluster all volumes covering each separate school trimester, excluding sub volumes which cover vacation periods. These volume clusters can be seen in the ‘Volumes’ row in the table (3) below. Once the most present topics and the value of their presence for each separate cluster were generated, it became possible to analyze and compare them to the contents of the *Classroom of the Elite*.

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
|  | Table 3 | | | |
| Trimester | 1 | 2 | 3 | 4 |
| Volumes | 1 - 4 | 5 - 7 | 8 - 11 | 12 - 15 |
| Sub Volume  (marking the end  of each trimester) | 4.5 | 7.5 | 11.5 | / |

## Results and Discussion

This section will discuss each generated trimester-based topic cluster. In order to investigate these topic clusters in a clear, interpretable way, the top-5 most present topics in each trimester were selected to then be compared to one another. Their degree of presence was taken into account as well. The following figure (Fig. 1) was prepared to visualize the results found. Each topic will be discussed in order of presence per trimester. The order will be set from most present to least present. This discussion will be followed up by a concluding section linking the results found to the above-mentioned hypotheses.

### Trimester 1

Topic 7(sudou, ike, yamauchi, council, ~~birthday~~, …): Sudou, Ike and Yamauchi are three guys from the protagonist’s class that quickly grew into a close friend group. On the first day of classes they form a group of boys to discuss different matters. Amongst other matters, they discuss how stern the student council president looked at the welcoming ceremony for the first years. The word birthday does not occur in the volumes covering the 1st trimester.

Topic 9(karuizawa, ~~nagumo~~, hirata, date, love, …): At the start of the schoolyear, it quickly becomes clear that Karuizawa and Hirata start dating. They are two of the three most popular people in class D and end up taking a leading role during class activities and discussions. The name nagumo does not show up in any of the volumes included in the 1st trimester. This name refers to the second year who later becomes the new student council president. It is unclear why his name is included in this topic.

Topic 72(sensei, reception, ayanokouji, sakayanagi, permission, …): This topic is hard to make sense of since the words included do not seem related to one another in a clear way. The Japanese word ‘sensei’ refers to a teacher. However, teachers were not involved in the club receptions at the start of the year. Of course, sometimes students such as the protagonist Ayanokouji or Sakayanagi(the leader of class A) ask teachers for permission in order to do something but there is no such specific moment to refer to within the contents of the 1st trimester.

Topic 64(changing, ike, ~~car~~, peeping, pool, …): The boys in class D, led by Ike, show interest in what the girls will look like after changing into their bathing suits for a swimming class. Ike even contemplates going into their changing room.

Topic 18(hair, image, kushida, sudou, love, …): This topic could refer to Kushida, the most popular girl in class D and her secret personality. While she has the image of a goody-two-shoes who tries to get along with everyone, secretly she curses and complains about how much everyone bothers her. She often tends to twirl her hair when exposing this secret side of hers to others. However, the words Sudou and love do not seem to have a connection with Kushida’s secret personality.

Overall, the topics discussed here seem to correlate to the content present in the volumes covering the 1st trimester. As predicted in the hypothesis section, the first trimester mostly covered topics depicting the internal relations within class D (the protagonist’s class). The protagonist, Ayanokouji has more of an observing function in this trimester. As a consequence, the most present topics revolve more around his classmates’ actions rather than his own.

### Trimester 2

Topic 68(ryuuen, nagumo, council, older, points) At the end of the second trimester, Ayanokouji comes into contact with the now ‘former’ student council president Horikita Manabu. The former president is concerned about the new president; nagumo’s modus operandi, he fears Nagumo intends to change the school too extremely. He hopes to entrust Ayanokouji with the task of revoking nagumo from his position. While Ayanokouji is not particularly interested he owes Horikita Manabu a favor. While Horikita Manabu makes his request, Ryuuen is listening in on the conversation.

Topic 9(karuizawa, nagumo, hirata, date, love):

Topic 72(sensei, reception, ayanokouji, sakayanagi, permission): As mentioned earlier the words in this topic do not have a clear connection to one another, thus this cluster is hard to contextualize.

Topic 7(sudou, ike, yamauchi, council, birthday)

Topic 38(months, importantly, reliable, kinugasa, refrain)

the protagonist establishes a study group that quickly grows into his first friend group. The reader can see him grow closer to his fellow students.

### Trimester 3

Topic 68(ryuuen, nagumo, council, older, points)

Topic 9(karuizawa, nagumo, hirata, date, president)

Topic 65(sakayanagi, ichinose, ryuuen, watch, theater)

Topic 55(council, nagumo, brother, join, president)

Topic 49(ayanokouji, boy, amazing, cup, formidable)

most 1st year class leaders have realized that Ayanokouji is the ‘shadow’ leader of class D. He gets challenged by the leader of class A, amongst others on multiple occasions. The 1st years start to interact more with each other and with senior students. Soon after, the oldest students or the 3rd years graduate after taking part in an exam in which all students in the school were involved

### Trimester 4

Topic 69(senpai, understood, detail, human, ayanokouji)

Topic 57(senpai, ayanokouji, answering, partners, okay)

Topic 55(council, nagumo, brother, join, president)

Topic 48(senpai, climbing, catch, definitely, ayanokouji)

Topic 5(senpai, ayanokouji, beach, win, flag)

The fourth trimester mostly revolves around the introduction of new 1st year students and a survival exam including students from all school years that takes place on an island. Ayanokouji also gets negative attention from the new headmaster (Syougo, 2015).

# Limitations

Topic modelling is a unsupervised method for labeling texts that solely relies on computer algorithms to decide which content words could be grouped together and labelled as one topic (Karsdorp, Riddell and Kestemont, 2021, Newman, Chemudugunta, Smyth and Steyvers, 2006). However, this does not mean no human intervention is present when running a mixed-membership or topic model. The human running the model decides what data to use during the pretraining stage, how many words can be included, and how many topics will be generated amongst other variables. Therefore, topic modelling is never without bias and it is important as a researcher to remain aware of this fact (Haverals and Geybels, 2021, Meeks and Weingart, 2012).

Topic modelling is not without its downsides. Aside from the need for a large set of pretraining data comparable to the main data mentioned in the *dataset* section of this paper, researchers have pointed out a multitude of weaknesses the research method has. Meeks and Weingart in their article “The Digital Humanities Contribution to Topic Modeling” summarize some of these weaknesses. They refer to topic modelling as:

*distant reading in the most pure sense: focused on corpora and not individual texts, treating the works themselves as unceremonious “buckets of words” and providing seductive but obscure results in the forms of easily interpreted (and manipulated) “topics” (2012).*

This quote reveals different critiques that have come up in regards to topic modelling. First of all, when applying a topic model algorithm, the narrative context of a written work gets disregarded. A text indeed, gets treated as a set of different words each occurring with a different frequency. The practice has been therefore also been labelled as ‘simply counting words’ (Da, 2019). It is also argued that topic modelling distracts literary scholars from their true purpose of interpreting language, instead they are tempted to simply interpret topics, simple clusters of out-of-context words (Schmidt, 2012). It is true that topic modelling leads to interpreting the topics the model provides, it is important to link the interpretations of those topics to interpretations of each text as a whole. Where the topic model discards context to provide topic clusters, the literary scholar should interpret those topics by relaying them back to that very context the algorithm discarded. Interpreting topics cannot only lead to a loss of literary interpretations but to ‘manipulating’ topics as well. This is another problem which Meeks and Weingart hint at in their article (2012). As shortly addressed in the methodology section of this paper, a topic model can reveal different results based on the variables the coder choses to include. Comparing the same models performed on the same texts, only varying in terms a variables can lead to largely differentiating results (Goldstone and Underwood, 2012). A researcher could easily be tempted to choose variables which provide results that best adhere to their hypotheses. Not all of these problems can be prevented though they can be reduced to a minimum as long as the researcher stays aware of them. When choosing variables I attempted to have the topics make sense when related back to the story as a whole before looking at the clustered results for each hypothesis made. The aim was to minimize result manipulation while still generating topic clusters that made sense. The main condition for a topic to be sensible was for model’s vocabulary list to contain mostly content words, thus reducing the presence of other words. As the Merriam-Webster online dictionary states: a topic is “the subject of a discourse or of a section of a discourse” (2022), topics contain subject matter which contain meaning it would be hard to argue that function words should make part of a topic. To prevent the other problems Meeks and Weingart point out from appearing within this study, all topics were interpreted in context of the whole story as much as possible and if no specific relation could be found, it was explicitly mentioned. As McCarty states, it is important to understand what we are doing while modeling topics since that awareness is the key to unlocking the true potential of the practice or in her words: “No one doubts the usefulness of the practice. Rather it's the *intellection of praxis* to which the next stage in the argument I have begun here must turn”(2014).

# Conclusion

Topic models are a helpful aid when it comes to literary analysis though they should not be utilized mindlessly. Research based on topic modelling is useful only when possible caveats are understood and taken into account. This case study exemplifies that topic models need to be pretrained well for them to create understandable topic clusters. Some topic clusters generated during this project were not as intuitively comprehensible as others. However, considering the scarcity of pretraining data suited for Japanese visual novel research the topic model did a fairly good job at interpreting the given input. Most returned topics were easily related to the content of the *Classroom of the Elite*.

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