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Teaching Tip

Using Text Analytics AI Insights in Microsoft Power BI Desktop to Score Sentiments, Extract Key Phrases, and Discover Unstructured Data Patterns

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ABSTRACT

This teaching tip describes using Microsoft Power BI Desktop in a class to analyze unstructured data from an exit survey of prior students from a Master of Science in Management Information Systems program. Results from a short survey administered to these students showed that the students, using the no-code Power BI, were able to accomplish their text analytics tasks in a shorter period, and with less overall effort, as compared to using traditional code-rich text analytics methods. The process is described in detail as to how the students, new to unstructured data analysis, can uncover these results. This process can be replicated by other instructors teaching text analytics.

Keywords: Text analytics, Unstructured data, Artificial intelligence, Sentiments, Key phrases, Power BI desktop

1. INTRODUCTION

Questions such as *what are the major strengths and weaknesses of a degree program* and *how could a degree program be improved* are among many survey questions each graduate student is requested to fill out when graduating from a large Midwestern university. Those questions allow students to answer in an open-ended style that is not structured in a predefined data model. Their responses are stored in native text format as qualitative data. Those text or unstructured data present a challenge for businesses wishing to utilize their data for analysis (Balducci & Marinova, 2018; Gentzkow et al., 2019). Per a recent study from International Data Communications (IDC), 80% to 90% of organizations' data will be unstructured by 2025 (Rydning, 2021). In the meantime, an overwhelming majority of businesses (90%) say analyzed data is essential to their organization's innovation strategy (Krammer, 2021). Businesses can take advantage of text analytics tools, leveraging artificial intelligence (AI) including machine learning and natural language processing, to find meanings from enormous amount of unstructured data. Enabling smarter and scalable AI was identified as the top data and analytics trend by Gartner research (Panetta, 2021).

At the same time, businesses need to solve their data analytics problems quickly and cost effectively, and the no-code method has been shown to support accomplishing this goal by allowing them to promptly collect data, perform data analysis, and make accurate data predictions (Bhattacharyya & Kumar, 2021). Applying the no-code approach is transforming

the traditional information system development paradigm (Wang & Wang, 2021; Wang & Wang, 2022). It makes it possible for both large and small businesses to deploy AI without hiring an army of developers, programmers, and data scientists (Reilly, 2021). Some even have suggested that the no-code AI platforms not only enable users to implement and test their ideas but also become the only way to ensure AI technology is adopted in every use case (Ataee, 2022; Smith, 2022). Therefore, there exists a strong appeal to teach students how to use AI insights without writing code to extract meanings, themes, and patterns from volumes of unstructured text data.

2. THE TEACHING ENVIRONMENT

The lesson described in this paper was based on a graduate business intelligence (BI) applications and tools course taught in Fall 2022 at a large, Midwestern university. All thirty-one students enrolled in this course were MIS majors. The lead author of this paper was the course instructor. The course was designed to introduce popular data analytics software tools from different vendors. Microsoft Power BI is one of those tools used by students for analyzing enterprise data. This was the second time this course covered the text analytics topic with a learning objective of applying the no-code method of AI technology to analysis of unstructured text data. (The text analytics topic was first introduced to the MIS and Data Analytics students in the Spring 2022 semester.) Twenty-nine out of thirty-one students were familiar with database management concepts such as entity relationship modeling,

database normalization, and SQL commands. Students either had or were taking a database course concurrently with the BI applications and tools course. However, twenty-two of the thirty-one students lacked solid programming skills, which is not unusual for MIS students in a business school.

Students were asked to use AI Insights of Text Analytics in Microsoft Power BI Desktop to score sentiments, extract key phrases, and discover data patterns from past MIS student exit surveys to present their findings on how the MIS degree program could be improved. Microsoft Power BI Desktop is an enterprise grade business intelligence (BI) software that lets users combine and analyze data from multiple sources to create interactive dashboards, reports, and apps (Ehrenmueller-Jensen, 2020). It is a popular data analytics tool used in academia (Techronus, 2021). Although students can download Power BI Desktop for free, a subscription fee was paid by the school for the Power BI Premium Capacity so that students could use the AI Insights to gain access to its Cognitive Services that contain pre-trained machine learning models developed by Microsoft's science team. The AI Insights in Microsoft Power BI Desktop uses a natural language processing (NLP) technique to determine whether the sentiment expressed in textual data is positive, negative, or neutral (Microsoft, 2021). Microsoft Power BI Desktop is a powerful tool to help academic programs monitor student sentiments in feedback through student surveys

and understand student needs. By using the pre-trained models, students can skip the steps for training, validating, and testing their models and move easily to analysis of their unstructured text data.

3. THE TEACHING APPROACH

Figure 1 shows the student exit survey source data loaded into Power Query Editor in Power BI Desktop. The data were stored in Excel. The data were edited by the course instructor to ensure no personally identifiable information was available to the students. With Power Query Editor, students learned to transform the data into the shape they wanted, e.g., by removing columns not needed for the study.

All software, including Power BI Desktop, used in this course has been installed in a virtual cloud computing environment called *Citrix*, or *Anywhereapps* as shown in Figure 2. This virtual cloud environment provides the students with access to the required software from anywhere at any time. This environment benefits the students during the COVID-19 pandemic since this approach allows the students to analyze data at their own pace, and from any location. Moreover, the instructor did not need to ask students to download the software and answer questions on software configurations.

Q	In your opinion, what are the major strengths/weakness of the MIS program?	In your opinion, how could the MIS program be improved?	How well has the MIS program
1	The MIS program has equipped me with the necessary skills and tenacity to be employable in the IT field. The program can be better improved by helping students secure internships as a way of gathering experience.	I believe the curriculum can be improved upon to capture more innovative topics like Cloud Computing, Machine Learning, etc.	Excellent
2	The dean and advisor has really help me out during my 5 years. The professors are really good and knowledgeable. This program really helps me with my career and career path.	I wish for more exams rather than just two. It feels like too much of the grades are dependent on two exams.	Excellent
3	Weakness: Not teaching coding to students that they will be required to learn after they finish the program.	Teach coding to students like Python, R and Sql that will be used in career after college. Have good quality teacher who knows what they are teaching. Do not get professor like Professor Onkar Malgon.	Fair
4	The major strength of the program was the specialization you are able to do. I was not able to finish...	In my opinion I would add a little more education and practice in the business analyst class on newer ways of doing things.	Excellent
5	The major strengths are exposure to various subjects. On the other hand the weakness is I think some of the courses are too basic.	Please try to add more advanced topics in the technical subjects. Also please update the syllabus based on the current industry trends.	Excellent
6	Major strengths are tuition, staff support, curriculum. I have compared the MIS program course offerings with other programs.	Offering programming courses for non-CIS Bachelors, and not just boot camps.	Excellent
7	Major Strengths : Curriculum, SAP course and the faculty. Weakness: Nothing for me.	I personally feel that it's already designed pretty well. I can't think about anything which is lacking in the program.	Excellent
8	The curriculum is the strength and MIS program has a really good faculty.	Classes can be more interactive.	Very Good
9	Well Structure. Great faculty. No Weakness.	It pretty perfect in my opinion.	Very Good
10	The course structure is the strength and the Internship should be made mandatory for all the students.	They should change the course structure.	Fair
11	Strengths : Supportive Faculty. Professors with Actual Knowledge.	No suggestions from my side.	Good
12	In the MIS program, I would say, OMIS 670 had a module that used to use python for topic modeling...	update the current curriculum as per the market requirements and skills. which will be really helpful when we...	Very Good

Figure 1. Student Exit Survey Data in Power BI's Power Query Editor

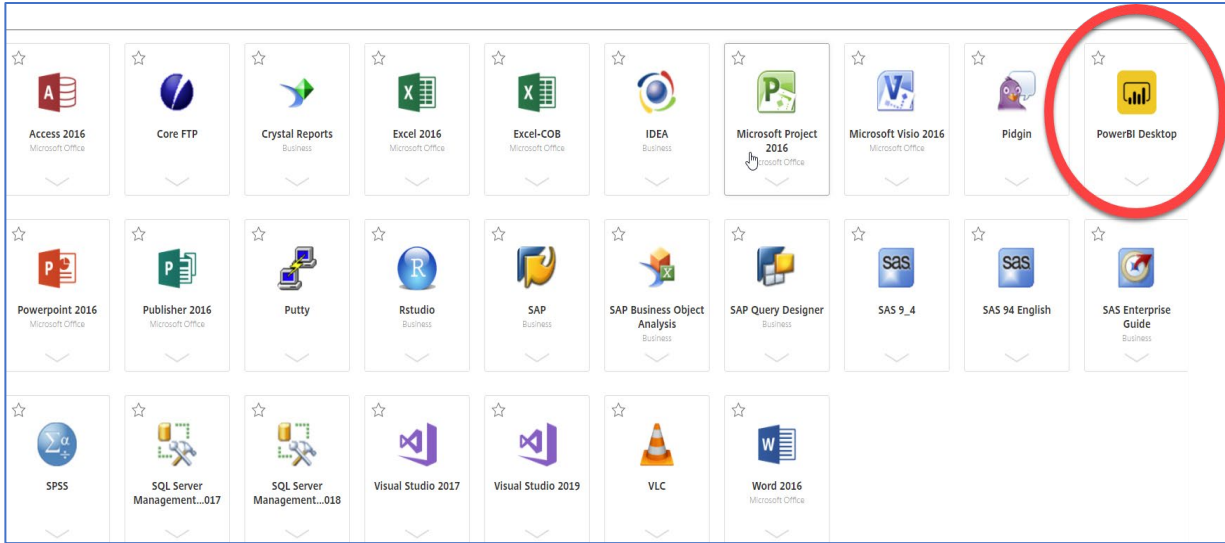


Figure 2. The Virtual Cloud Environment for Accessing Power BI Desktop

With the Text Analytics in Power BI, the students log into the Premium Capacity as shown in Figure 3. Then they can apply three different cognitive services, *i.e.*, Detect Language, Extract Key Phrases, and Score Sentiment as AI functions or insights shown in Figure 4. Figure 4 also shows that a premium license, *College of Business – OMIS*, was available for student use.

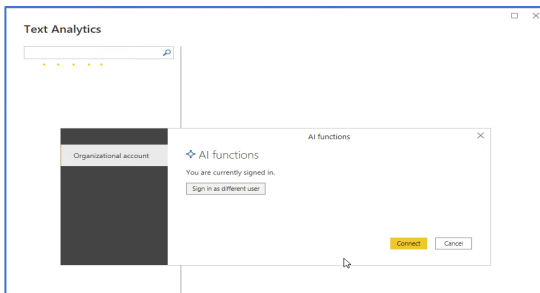


Figure 3. Login to Power BI Premium Capacity

3.1 Detect Language

The AI language detection function evaluates text input in the “How could the MIS graduate program be improved?” data column. After completing language detection, two new data columns, Language Name and Language Code Identifier (ISO), are created as shown in Figure 5. This function supports a wide range of national languages, variants, dialects, and some regional or cultural languages (Microsoft, 2021). The students then remove rows with *null* values based on rendered detected language name column. The instructor could have asked the students to delete those records with empty values for the analyzed text column in Excel since the original data were stored there. The deleting action after applying the AI function is a better way to help the students understand that the detected language serves as the input for other AI functions

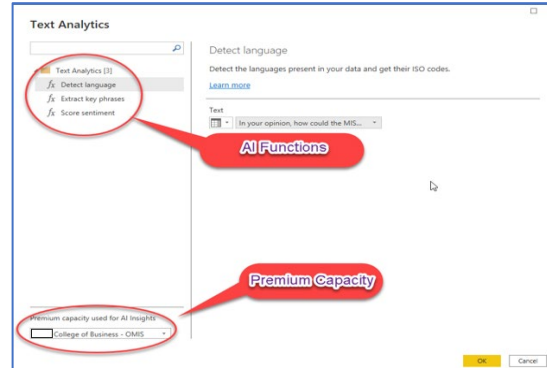


Figure 4: Three AI Functions for the Text Analytics in Power BI

3.2 Score Sentiment

Next, students continue to apply text analytics to score sentiment as described in Figure 4 for the question “How could the MIS graduate program be improved?”. The AI engine’s score sentiment evaluates the text input for the question and returns a sentiment score for each response, ranging from 0 to 1, *i.e.*, 0 represents extremely negative and 1 represents extremely positive. To be consistent with the school’s program assessment criteria, the students are asked to create a column named “Sentiment Text” using the conditional formatting feature as shown in Figure 6 to specify the following outcomes: if sentiment scores ≥ 0.80 , then “Very Positive”; if sentiment scores ≥ 0.60 but < 0.80 , then “Positive”; if sentiment scores ≤ 0.20 , then “Very Negative”; if sentiment scores > 0.20 but ≤ 0.40 , then “Negative”; else the sentiment text will display “Neutral”. Figure 7 displays the sentiment scores and the classified text values of “Sentiment Text” generated from this approach. The average sentiment was scored 0.68 or positive in this exercise.

following - Provid...	Detect language.Detected Language Name	Detect language.Detected Language
	English	en
	English	en
	English	en
	English	en
	English	en
	English	en
	English	en
	English	null
	English	en
	English	null
	English	en
	English	en
	English	en
	English	en
	English	en
	English	en
	English	en
	English	en
	English	en
	English	en
	English	en
	English	en
	English	en
	English	en

Figure 5. Language Detection Results

Column Name	Operator	Value	Output
Score sentiment	is greater than or...	0.80	Very Positive
Score sentiment	is greater than or...	0.60	Positive
Score sentiment	is less than or equ...	0.20	Very Negative
Score sentiment	is less than or equ...	0.40	Negative

Figure 6. Group Sentiment Scores to Very Positive, Positive, Neutral, Negative, and Very Negative

Score sentiment	Sentiment Text
0.99	Very Positive
0.88	Very Positive
0.50	Neutral
0.50	Neutral
0.50	Neutral
0.50	Neutral
0.00	Very Negative
0.15	Very Negative
0.99	Very Positive
0.91	Very Positive

Figure 7. Score Sentiment and Corresponding Sentiment Text

3.3 Extract Key Phrases

Next, the students extract key phrases as described in Figure 4 for the question "How could the MIS graduate program be improved?". The AI's key phrases extraction evaluates the unstructured text and returns a list of extracted key phrases as shown in Figure 8.

At this point, the students were excited to see sentiments scored and key phrases extracted from the survey question with this simple approach without any code writing. It was interesting to note that sentiment was scored as 0.50 or neutral if "not sure", "n/a", or "none" were provided as the feedback or input for the question. The students learned that if feedback is mostly objective there was no extracted sentiment, resulting in a 0.50 score. The students examined the extracted key phrases and they concluded that *Python programming, in-depth technical courses,* and *student-focused learning* emerged as the critical areas to be improved, based on the feedback from the exit survey.

These conclusions were drawn from a word cloud map students created which is presented in Figure 9. The word cloud visual has been used in various contexts to provide an overview of extracted text, indicating the frequency of occurrence of words (Heimerl et al., 2014). The more a specific word appears in a source of textual data, the bigger and bolder it appears in the word cloud.

3.4 Discover Data Patterns

It would not be perfect to end the exercise by simply applying the AI insights to detect language, score sentiments and extract key phrases. The students were challenged further to present their text analytics results using various visualized reports. Figure 9 shows an example of a dashboard generated from the class exercise to show the average sentiment score in percentage, options (checkbox or slicer) to display the score sentiment for each sentiment text category, a word cloud map on the extracted key phrases, and a summary table of the sentiment score, sentiment text, and key phrases based on individual student id.

4. DISCUSSION AND SUGGESTION

An online survey was conducted by the authors (see Appendix) after completing the Text Analytics exercises. The students strongly agreed (Mean=4.81, Std. Dev=0.32, Median=5.0) with the statement that the "no-code" approach of using *Text Analytics of AI Insights accurately detected sentiments on "how could the MIS program be improved?"*. Furthermore, they felt more confident in the use of the "no-code" approach to solve Text Analytics problems (Mean=4.36, Std. Dev=0.68, Median=4.0).

Score sentiment	Sentiment Text	Extract key phrases
0.99	Very Positive	real time examples, skill, students
1.00	Very Positive	program, final project, Project management, python, SAS Enterprise M...
0.98	Very Positive	remedial Python course
0.99	Very Positive	examples, real time
0.88	Very Positive	classes, programming software, SAS, great programming, SQL, languag...
0.50	Neutral	additional stats course
0.50	Neutral	real job examples
0.00	Very Negative	data analytics topics, classes, major overhaul, major waste, tools nece...
0.15	Very Negative	subsequent courses, students similar, prepared students, computer sci...
0.99	Very Positive	people, hands data analytics, semester long capstone project, finish, jo...
0.91	Very Positive	advanced knowledge of data analytical applications, python
0.30	Negative	data-related applications, depth, python
0.89	Very Positive	program, courses, foundational class, capstone project, foundational p...
0.95	Very Positive	topics, depth, baseline, couple, knowledge
0.78	Positive	Better profs, longer timeline
0.50	Neutral	intro, Python

Figure 8. Extracted Key Phrases with Score Sentiment and Corresponding Sentiment Text

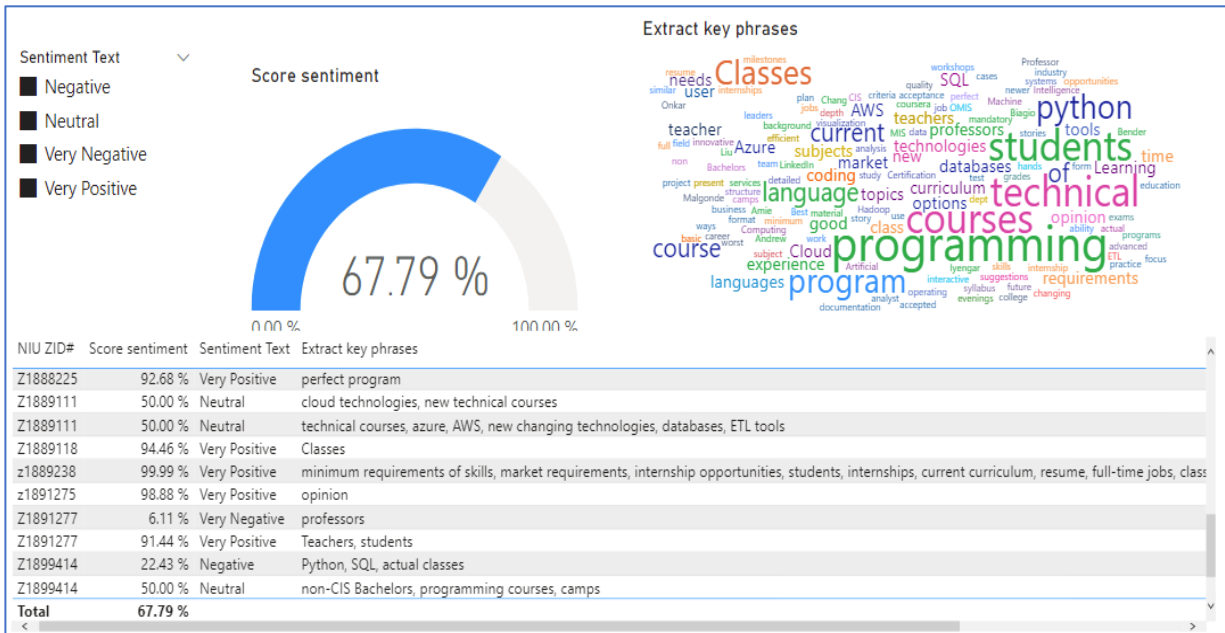


Figure 9. Discover Data Patterns From the Text Analytics

Teaching AI applications have become an integral part of machine learning and business intelligence courses in a MIS program, but many students and faculty have not completely explored it since familiarity with programming languages such as Python and R are often required. This is often particularly true in business schools where students would rather conduct

data analysis without writing code. This paper describes a faster and an easier way to teach the AI insights of text analytics, compared to using a code writing approach, by using Microsoft Power BI to analyze unstructured text data. Interested readers can request the sample data file and lab instructions on the exercise from the lead author. Students showed great

enthusiasm to learn and apply AI technology using a no code approach. For example, immediately after the described exercise, the students were asked in class to apply the same AI functions, i.e., *Detect Language*, *Score Sentiment*, and *Extract Key Phrases* for the “What are the major strengths and weakness of the MIS graduate program?” survey question. This is another example of text input of unstructured data. Results indicated that all students could correctly analyze the responses without any additional help from the instructor. A homework assignment was also given in which students analyzed customer review comments from a local restaurant website. Again, students finished the assignment easily by using the approach described in this paper.

Moreover, the students were given extra credits to watch a YouTube video “*Simple Sentiment Analysis in Python*” to compare with the no-code method of Text Analytics in Power BI (see Appendix). All students either agreed or strongly agreed with the statement that the “no-code” approach of using Text Analytics in Power BI Desktop “solves the issue” faster and easier compared to code writing for conducting text analytics as demonstrated in the video.

The opportunity for analysts to give meaning to unstructured data is substantial. Although programming knowledge and skills are important, this no-code technique can help provide graduates to fill this labor need. As a result of the lessons learned, the authors would like to provide the following suggestions for other faculty to teach Text Analytics using Microsoft Power BI Desktop:

- Prerequisites for students should include the knowledge of database concepts for building table relationships. Students can learn to go deeper into the analysis if they have these skills. For example, the exercise could be extended to include additional files with student profile data after the personal identification information has been removed. Once the relationship is established using student campus id, sentiment scores could be compared based on major, gender, ethnic background, direct vs. transfer admission, or other dimensions. This additional demographic information will enrich program assessment for each academic program.
- Prior to applying text analytics in Power BI, it is important for students to cleanse and shape the source data to ensure data quality. Invalid or incorrect inputs should be filtered out for the data analysis. “Null” values should be removed after the language detection AI function is applied. Students can learn a valuable lesson on data quality if a faculty member guides students using Power Query Editor in Power BI to transform and cleanse the source data.
- It would be helpful to ask students to examine the aggregation or summarization method and data format for the sentiment scores. After the numbers were derived from “score sentiment,” *Sum* was the default aggregation. Most students continued to use the default aggregation without realizing that using *Average* was appropriate instead. Students can enhance their learning if a faculty member guides them to review different types of summary statistics to aggregate data for analysis. Figure 10 shows the students were asked to use the *Average* and *Percentage* for the aggregated sentiment scores.

- Students should be challenged to further explore AI features in Power BI to include exercises on using *Q&A*, *Key Influencers*, *Decomposition Trees*, *Smart Narrative*, and *Vision Analytics* visual tools. Lab hands-on materials on these topics are available to share based on request. Along with text analytics, those exercises help students to gain a solid overview of AI features in Power BI.

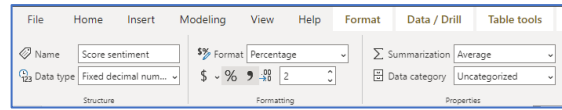


Figure 10. Change the Sentiment Score to Average and Percentage Aggregation

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APPENDIX

Text Analytics of Power BI Desktop Survey

Please rate your skills for the following:

	Poor	Fair	Average	Good	Excellent
(1) Programming – Writing Code	1	2	3	4	5
(2) Entity Relationship Modeling	1	2	3	4	5
(3) Database Table Normalization	1	2	3	4	5
(4) SQL Commands	1	2	3	4	5

Please watch the video of “Simple Sentiment Analysis in Python” at https://www.youtube.com/watch?v=tXuvh5_Xyrw (16 minutes)

Please complete the guided exercise of using Power BI Desktop for analyzing sentiments on “How could the MIS program be improved?” from past student exit survey.

Please share your opinion: **The “no code” approach of using Text Analytics of AI Insights accurately detects sentiments on “How could the MIS program be improved?”**

Strongly Disagree 1 2 3 4 5 Strongly Agree

Please share your opinion:

The “no code” approach of using Text Analytics in Power BI Desktop solves the issue in short period time with less effort compared to code writing for conducting text analytics as demonstrated in the video.

Strongly Disagree 1 2 3 4 5 Strongly Agree

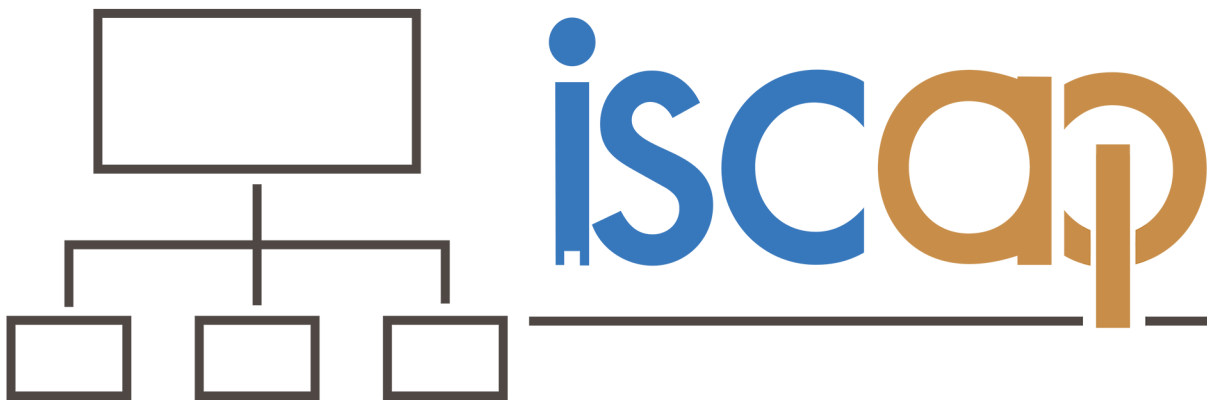
I prefer to use the “no code” approach to solve Text Analytics problems.

Strongly Disagree 1 2 3 4 5 Strongly Agree

I feel more confident to use the “no code” approach to solve Text Analytics problems.

Strongly Disagree 1 2 3 4 5 Strongly Agree

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