

Jarvis - A Smart Bot for Piazza

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ABSTRACT

In this work, we explore some of the challenges faced with Piazza, specifically for large classes. We present a smart, interactive bot, Jarvis, that can tackle the problems by using state-of-the-art IR techniques. We show the Jarvis can greatly increase the efficiency of the instructors in managing large classes and also serve as a useful aid for students to master the material.

KEYWORDS

IR, Piazza, search, similarity, FAQ generation, question answer system

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1 INTRODUCTION

1.1 Piazza

Piazza is a free online gathering place where students can ask and answer questions 24/7 under the guidance of their instructors. Collaboration occurs in real time, which encourages participation and fuels discussion. Students can ask questions, which fellow students and instructors can attempt to answer. Students can post questions, answers, and notes, publicly or anonymously. Instructors can post notes, answer student questions, endorse good questions and answers, and collect student feedback about this course. Piazza provides students participation report which helps instructors to see students engagement in the class. Instructors can view the class report to see when in the term the most questions are being asked but not the other way around. Leading campuses across North America have a significant Piazza presence, especially for computer science courses. Larger courses typically have over 500 students actively using Piazza to assist with their coursework.

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1.2 Chat Bots

A chat bot is a computer program which conducts conversational question answering with users in a certain domain or on a certain topic with natural language sentence. Such procedure is aiming at simulating a persons behavior as a conversation partner and designed to pass the Turing test. Chat bots are commonly used in various dialog system for information acquisition and other customer services. Some natural language processing techniques are used in building chat bots. There are various different levels at which chatbots attempt to simulate a human - some are little more than guided menus, while others incorporate deep learning and Artificial Intelligence to answer questions.

2 PROBLEM

Piazza is a great tool for students and instructors to interact with each other. It allows students to post questions on a class wide forum, where other students and instructors can answer the questions. For small classes, where the question volume is very low, the system works well. Instructors can respond quickly, and scale isn't an issue. However, as the departments grow and class sizes increase, it becomes difficult to answer all the questions in a timely manner. For this project, the authors wanted to focus on a few specific issues related to Piazza at scale. They are interested in this research topic not only to gain insights, but also to help themselves! The authors are instructors for undergraduate courses at UofM - these course have roughly 400 and 600 students and only a handful of staff members to manage the class. They are interested in solving these problems to assists in the courses they teach. The goal is to free up instructor's time so they don't have to spend all of their time answer questions on Piazza, and provide a better experience for students using the platform.

2.1 Duplicate Questions

As one of the side effects of the number of students posting questions, it becomes difficult for other students in the class to know if their question has already been asked. Given the question volume, they may not have time to read every question that is asked nor use the search facilities that Piazza provides. As a result, students repost a question that is very similar to one that has already been asked. For instructors, this means that they have to answer the same question over and over, or link students to the previous question. This is not an easy task as a similar question might have been answered by another instructor which requires instructors to search in the first place. This requires instructors to know all of the questions that have been posted by all students, which

clearly is not possible. For students, it could provide confusion, as many similar questions have very similar answers, but students are quite sure if its the same question (and therefore has the same answer). There is also the consistency issue where two instructors might not concur exactly on the duplicate questions which confuses students even more.

2.2 Lack of a Weekly Summary

A related issue is that students aren't getting the most they could out of Piazza because of the number of questions. Students can't dedicate hours every day to read through all questions posted every day. Also, many of the question are very student-specific (they won't relate to other students, so the whole class doesn't benefit from reading them). However, there are some questions that are very common, or exceptionally helpful, and it would be beneficial to aggregate these high quality questions. Ideally, students want to see a weekly report of 10 questions that they do not want to miss out on. This should allow students to get much more from Piazza, and allow them to use it as the interactive learning tool it was designed to be.

2.3 Poor Search

Piazza's search features leave something to be desired. The largest problem seems to be that the results are very heavily skewed towards the most recent results, when that may not actually be what the student is looking for. In fact, we have seen posts where students will actually post a question, asking to be linked to a specific post that they are having trouble finding. We wanted to provide a method for students to more accurately search Piazza for their questions (or answers), that isn't so heavily skewed by one factor. This feature will hopefully cut down on duplicate questions if it works well, as a good search feature should help students find questions similar to ones that have already been posted, so they can avoid posting duplicates.

2.4 Information Retrieval

How do we formulate the above problems as information retrieval problems?

Finding duplicate questions is a natural IR problem. We have to find the most similar posts in the collection, and then see if similarity score is above the threshold we have chosen. If so, we can flag this question as a duplicate.

Finding the top questions for the week for the FAQ is a question where both traditional IR techniques come into play, and we can use specific information made available from the Piazza platform.

Improve Piazza's search feature is directly an IR task. Piazza's search isn't very good at IR, and we want to make it better!

3 METHOD

We designed and implemented Jarvis, a bot written in python that can accomplish these goals.

3.1 System Overview

Jarvis works by indexing all the contents of the course on Piazza. On start-up, it bulk downloads all of the questions and answers posted previously and constructs an index from there. After that, it periodically scrapes new contents and does an incremental update to the index. It can be run locally on an instructors laptop to quickly index and answer all available questions, or left running as a daemon on a dedicated server. For most of our experiments, we ran Jarvis locally, but it will be significantly more useful to students and staff if it is run on an always-on computer.

3.2 Tools and packages used

We took advantage of several python packages and libraries to make the development of Jarvis smoother and more stable. We used python's whoosh library for indexing the documents. Whoosh is a full index library that supports many different operations - creating a schema, search index or certain columns, quick searches (using BM25) and also date range queries [2]. We used NLTK implementation of the Porter stemmer to store our tokens in a standardized form [5]. We also took advantage of sklearn to help create our TF-IDF vectors [6]. Finally, numpy makes working with large arrays and matrices much easier than raw python [3].

3.3 Data Crawling

Piazza does not provide an elegant or easy way to programmatically interact with its posts. To get the questions and answers on the site, we had to use an unofficial API. This API has been reversed engineered [4] and most of the actions that one can perform on the piazza website can be done through the API. We started from this API, and then built our own abstraction on top of it that would read in posts and then store them in our whoosh on disk index. To speed up processing, testing, and to prevent Piazza from potentially rate limiting us, we also developed a feature to bulk download and store course data, which we could then treat as our "Piazza" input data for indexing and pre-processing purposes. We extracted several past semesters of the course EECS 281, and were able to use that as our training set.

3.4 Feature Extraction

The raw data that we get through the APIs aren't fully sufficient for the purposes of our bot. For instance, we require features such the # of endorsements from students and the instructors separately since they help in improving the quality of results. We also define a feature 'activity score' of a post which looks at the number of updates made to a post, the number of student and instructors involved, and also the number of follow-ups in the post. We had to convert, combine, and transform many of the fields that we were given into the actual fields and values that we were looking to extract.

Question History: # edits to post 173 views

Final_Practice Complexity Classes

In the best/worst/average complexity table question, the average complexity of both hash table searches is listed as $O(n/m)$. It feels as though in most cases m will equal n , making this an $O(1)$ complexity. It also just feels strange using a separate variable when using big- O complexity. When is it appropriate or necessary to use additional n -related variables for the complexity of an algorithm?

the students' answer, where students collectively construct a single answer

M is the number of buckets. If the number of buckets is equal to n , the number of items in the hash table, then an average case of $O(1)$ should be expected. If there are a lot more elements than buckets, you would expect a worse complexity. You should use a number other than n when the complexity is dependent on more than just one variable. Usually, input size is all that really matters, but there are cases where complexity can vary dramatically dependent on a different variable. If the number of buckets is 1, wouldn't you expect a vastly different complexity that if the number was infinite?

the instructors' answer, where instructors collectively construct a single answer

Another quick example of where complexity depends on multiple variables: searching for an element in a 2-D matrix of size n by m is $O(nm)$. If n where on the same order as m , we could simplify it to $O(n^2)$ but we don't always know this.

Figure 1: A Piazza question, containing both a student and instructor answer. The highlighted boxes show some of the many features that we can extract from the post.

3.5 Data Preprocessing

The subject and body of the post in its raw form is HTML format. We remove all the irrelevant HTML tags from the post body. We concatenate the title with the body together to make it easier for searching purposes. For each document, we perform some basic text cleaning by removing special characters through regex and also convert all text to lowercase. We also remove the stop words from documents and apply Porter stemmer [1] package to further shrink the vocabulary size.

3.6 Duplicate Detection

We implemented a function called `find_top_N_similar_docs()`, which achieve following functionality: input a new post, and a collection of existing posts, and a integer n , we can return the top n posts with highest similarity scores against new input post.

By representing the text document using vector space model, we can calculate relevance between the documents. We calculate the TF*IDF (term frequencyinverse document frequency) score for each word in the documents by using sklearn's built-in function. Then, we calculate the similarity between new posted documents with all documents in the whole collection. We have a variety of methods to use to

calculate similarity. We could use cosine, dice, or jaccard similarity scores. Jaccard similarity scores normally work well on shorter documents (shorter than the average Piazza post), so we decided not to use that [7]. In this case, Cosine similarity measure or Dice similarity measure would be more efficient since these measures tend not to over penalize non-overlapping terms. In our implementation, we used Cosine similarity measure for reliability. By comparing pairwise similarity scores of new input post and all posts in the collections with input parameters N , we can return top N most similar documents with highest scores in the collections.

3.7 Weekly FAQ

Every week on a Sunday, Jarvis makes an attempt to post the top 10 questions that could be relevant for every student in the course. To do this, Jarvis retrieves all the posts made one week prior to that Sunday from the whoosh index. Posts that do not contain an instructor answer are filtered out since they would not be of much use. With the remaining posts, Jarvis looks at various features of the posts and scores them. The key features that contribute to the score are :

- (1) # good question endorsements
- (2) # good question endorsements by TAs
- (3) # good answer endorsements

- (4) # good answer endorsements by TAs
- (5) # unique views
- (6) # student follow-ups

For each of the above features, we assign weights based on their importance. We finally rank the results and then return the top 10 results. We also threshold the scores so that during dry weeks, we don't return stale/irrelevant posts.

3.8 Improved Search Feature

For the search task, we are able to improve upon piazza's search facility in the following ways:

- (1) treat recency as a metric
- (2) support typos made in query
- (3) OR based query on each term rather than AND based one
- (4) importance to finding query term on subject/body rather than answer

A recency based search has the issue of having important posts go down the search results. We make recency as one of the features among the others. We use a decay function to make sure that old posts don't get a very low score. From our experience, typically posts which are older than a month get the same recency score. We run the query through a spell correction engine which replaces query terms when a high confidence is present. We do not perform an AND based search as sometimes, students include an extra term which returns no results. Hence, we score results that are able to match all terms more, but when no post matches all terms, we see the number of terms matched and also the maximum n-gram match for the query.

4 RESULTS

Here we present the results of our system. To test our system, we set up a mock Piazza course where we could post fake questions to Jarvis. Jarvis has read only access to previous semesters of questions, and read/write access on the mock course. When new questions are posted, Jarvis can read those questions, and write answers to them, but it cannot write to the past semesters course. We didn't want Jarvis answering real student questions quite yet.

4.1 Duplicate Detection

To test our system, we found all questions in previous semesters that had been marked with an '@'. This means that an instructor read the question, and decided that it was similar enough to a previous question, that, instead of answering it, they decided to simply link to a previous question. We used this set of questions to test our system. We found that our system performed very well on this data set, responding to more than 90 percent of known duplicate posts correctly.

Perhaps more exciting is the fact that our system was able to identify some duplicate questions that instructors did not find themselves. When we fed our system questions that specifically had not been marked with an '@', we found that some of them were still marked by our bot. Many of these were, in fact, duplicates that had been answered twice, by

two separate instructors at two separate occasions. However, since this data is not labeled, there is no easy way for us to determine our systems efficacy without looking at every duplicate it flags and checking to see if it seems logical. For the vast majority of the new duplicates flagged by our system, it seems to work well.

4.2 Weekly FAQ

To evaluate the 'best' or 'top' questions is an extremely subjective measure. We do not have a golden list of the best questions to compare our results with. We looked through several weeks worth of questions, and tried to find a few, extremely important posts that we knew *must* show up in the final list. These included instructor clarification posts, notes about when and where the exams were, and other important course logistic posts. For the questions we found, Jarvis always had these in the top ten. The other way we were able to evaluate is by manually looking at the questions that Jarvis has found, and attempt to determine if the questions reflect the important questions posted that week. We notice that 9.4/10 questions on average are important questions. One point to note is that during tight weeks, typically more than just 10 important questions are present. Hence, we have a good precision number (94%). However, calculating recall wouldn't do justice to the task since Jarvis trims the rest of the results. One other experiment we conducted was that we tried to find the 5 most crucial questions for the week and checked if Jarvis had these 5 questions in the top 10 list. We noticed that Jarvis was able to get 4.6 of them on average.

4.3 Better Search Features

To take advantage of our improved tag feature, students can post private questions with their search query, under a specific tag. When Jarvis sees questions that are posted with this tag, it searches for all similar questions and responds to the post with links to the most similar questions. While this is a more involved process then using the built in search feature in Piazza, we feel that the improved accuracy of the results is worth the additional time investment.

To evaluate our search results, we prepared a list of posts that we were looking for, and some keyword queries to try to locate those posts. We found that, when the post we were looking for was very recent, our improved search did not perform any better than the built in search. However, this is expected, as Piazza's search feature is heavily skewed towards the most recent posts. Once we searched for posts that were further in the past, our system performed much better. Unfortunately, since we are trying to improve on Piazza's ranking, we cannot use there results as a source of truth against which to calculate Kendalls Tau or other useful metrics. Therefore, most of our measurements were extremely subjective, and involved entering in identical queries into both systems and manually comparing the results. The results returned were almost always the same, but our results ranked the documents we were actually looking for much higher than Piazza's default search.

5 DISCUSSION

5.1 Conclusion

We design and implemented Jarvis bot, a python implementation of a chat bot meets information retrieval system that allows us to more effectively manage Piazza with large classes. We explored 3 problems in this work, but plan to work on some more problems. We find this tool effective to an extent where we can use it for the mentioned courses in the coming semesters. We are sure this tool will improve the efficiency of instructors and also serve as a useful aid for students.

5.2 Deliverables

We have open-sourced on implementation of Jarvis, as well as the code to download and index Piazza. All of the Jarvis bot code is available on Github, here

5.3 Challenges and Lessons

A large portion of our work had to do with dealing with messy input data scraped from Piazza. Because the API is unofficial, there were many additional steps that we had to take before we had a clean dataset of questions and answers for each course.

Incremental indexing wasn't trivial - we had to extract updated posts. We eventually found a hack to do this by using Piazza's read/unread filter UI to make Jarvis read updated posts. This way, incremental indexing could scale based on the number of updated posts and not on the total number of posts.

With respect to the quality of results for a course, we noticed that each course requires it's own set of weights and features for the tasks we solve. For instance, the course we tested on was EECS 281 (Data Structures and Algorithms). A good number of questions on this course are more private in nature - for example, compile time issues where the warning log is pasted on piazza. These posts add little value to the community as a whole.

On the other hand, courses such as EECS 482 (Operating Systems) has a lot more conceptual questions. When we applied similar models to this course, we noticed that it wasn't doing the best job. However, upon tweaking some of the weights and features, we noticed a big boost in quality.

We also noticed that code snippets pollute search quality. Piazza supports rich formatting and not utilizing them proved problematic in certain cases.

5.4 Future Work

In the future, we can improve our algorithm by improving on term selection and similarity calculation. For term selection, we can improve by using n-gram, for example bi-gram or trigram, which may better capture the relationship in each document. Also, we could try implement grammar structure analysis of a sentence using a set of input transformation rules to represent grammar rules. For similarity calculation, we can try applying different similarity functions and use cross validation to find a better model. Besides, we could

also try to applying applying new techniques, for example word2vec, to reduce the calculation.

To improve our weekly Top Questions feature, we have a few plans. For one, we want to expand, so that we aren't just looking at questions, but also answers. It is entirely possible that many different questions all have the same common problem, and are given a common answer. It would be great for our FAQ to be able to list these very common answers, and all of the questions that they are associated with.

A goal that is a bit more far-fetched, but certainly within the realm of possibility, is to actually add full question answering to Jarvis. Jarvis could look at recently posted questions, and instead of simply linking to a previous post, it could try to answer the question. This would involved adding new layers to Jarvis, but the infrastructure is set up to be easy to expand.

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