



-ANALYSIS OF RELEVANCY USING RAPID MINER WITH AN EXPERIMENTAL STUDY ON POPULAR DISCUSSION FORUM

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Abstract: Massive Open Online Courses (MOOCs) is a model for delivering course content online independent of attendance and place .Online Discussion not only brings opportunities for innovation in education but also shifts focus from traditional educations. MOOCs in Computer Programming can help in attracting students because it personalizing their learning experiences at lower cost. In addition to watching videos, students also engage in discussion forums to share their understanding. On the other hand, online discussion forums provide a platform to ingenious students to share their ideas in a holistic way which is not possible through regular websites, videos, online courses. Both novice and expert users search the web exhaustively for their coding practises to learn gradations behind libraries, programming languages and frameworks.In community-based question-answering communities, where students ask questions there is no guarantee that students get their what they are searching for. This poses an unsatisfactory level in the student. In this paper, we present a machine learning model that predicts the relevancy of answers to the forum questions using historical forum data .The study attempted to identify the relevance criteria that people use when browsing a discussion forum

Keywords: MOOC's, Engagement, Discussion forums, Prediction, Relevancy.

1. INTRODUCTION

MOOCs have become a popular and significant source of distance-education due to the flexibility of course access and varied topics of courses. Online courses provide a good platform for learning across various places with these benefits ,there are million students who enroll in one or more MOOCs. Since MOOCs typically offer with only video-based lectures with no interactions with professor, students usually lack the opportunity to have interactions with other students and the instructor relative to traditional resident courses.

Anderson [3] discusses the features of successful forums , as well as the capacities and boundaries for online learning. Social network analysis is commonly applied to investigate the interactions on learning forums[4]. Dropout behaviour is identified as related to sentiment of students' forum posts based on sentiment analysis of MOOC forums discussions[5].Huang et al.[6] studied the behaviour of users who make great contributions across different MOOC forums.

Technology-based interactions, such as blogs and forums ,are introduced as potential solutions . The Q&A forum like StackOverflow, Stack exchange, quora are the most popular forums [7], [8], but a lot of niche forums exist for every feasible piece of programming technology. In addition, online documentation pages, technical blog posts, and open-source code repository websites often embed discussion

forums at the bottom of each webpage to allow people to discuss that page's contents.

Research using MOOC Forum

Regardless of the prevalent acceptance of discussion forums, the user interface of discussion forums has not changed much in the past years since early avatars such as Usenet newsgroups [9] and The WELL [10]. Most of the discussion forums are comprised of tree of text-based threads and all the other kinds of programming forums look almost identical. Some popular sites such as StackOverflow have incorporated features such as voting, answers ,reputation metrics, category tags, hot networking questions, searching, sorting, filtering, favorites, rich text formatting, active, oldest tags, recent tags and syntax highlighting for code. But in principal, most of them follow generic hierarchy of word-based discussions. When it comes to relevancy of answer to the question asked in discussion forum, most of them are accepted through upvotes. The answer with highest upvotes is treated as the most acceptable answer. However students cannot find the answer what they are searching for and this is one the reason for their drop out ratio from online courses.In this study we attempted to find the relevancy of an answer to the question asked in a popular discussion forum, stack exchange . The data from discussion form is taken and 1000 questions from programming on C is taken and a dataset is prepared out of it. We have applied

supervised classification algorithm “Decision Tree” on our dataset using Rapid Miner tool.

Prediction Model

The data provided by the discussion forum is used to define various features. By these features, we can provide many inclusive methods to get the predictions right. The data-set formation is based on the binary classification that is to provide the relevant discussion. These features help for the formation of the data-set that help to predict the relevancy. Features include user features and the thread features.

User features:

As on all the discussion forums, the candidates participate in answering different set of questions which help them to understand or clear the doubts of others. These answering to the different set of questions given in a thread that ultimately have an answer accepted or it may be opinion based which hence can be used for training model. These feature also provide further information about the last activity in the forum and the number of viewers visited that forum . In the formation of the data-set the number of accepted answers are considered to provide better and more accurate predictions.

Thread features:

It holds the information about the built-in thread post. Most of the thread post provide information about the answers and the comments that have been already posted.

The questions and the answers are taken from the most of the computer science discussion forums wherein the discussion in these forums are more frequent and the discussion goes on a related pace. The impact of these features are provide the information gain by the number of accepted answers posted by the users, sum of the votes received by an user overall threads ,and number of posted comments by users.

Each extracted feature is then represented into set of distinct values to facilitate better machine learning . For our predication purpose we used RAPID-MINER tool-kit without any other techniques involved to show that our features are sufficient enough to create meaningful predictions.

The programming languages like c is considered for the formation of the formation of the data-set. These languages are mostly discussed languages when compared to the other as they are the base roots for advanced computer programming languages to learn.

Thread ID	Thread Title	Answer
1	Difference: methods in C	726 good
2	Difference: functions in C	486 fair
3	Difference: functions in C	853 good
4	Difference: functions in C	527 fair
5	Difference: Methods in C	343 good
6	why do you use malloc in C	685 good
7	why do you use malloc in C	596 good
8	why do you use malloc in C	452 fair
9	why do you use malloc in C	564 excellent
10	why do you use malloc in C	283 poor
11	why do you use malloc in C	356 excellent
12	what is polymorphism in C	798 good
13	what is polymorphism in C	526 poor
14	what is polymorphism in C	473 fair
15	what is polymorphism in C	726 good
16	what is polymorphism in C	492 excellent
17	What is the use of getch() in C	736 good
18	What is the use of getch() in C	441 fair
19	What is the use of getch() in C	poor
20	What is the use of getch() in C	248 good
21	What is the use of getch() in C	826 excellent
22	What is the use of getch() in C	493 fair
23	What is the use of getch() in C	409 good
24	What is the use of getch() in C	542 excellent
25	What is the use of getch() in C	473 fair
26	What is the use of getch() in C	544 fair
27	What is the use of getch() in C	poor

Fig :1 Sample formation of data-set to provide the relevancy.

We used a classification algorithm called “Decision Tree” which supervised learning algorithm (having a pre-defined target variable) that is mostly used in classification problems. It works for both categorical and continuous input and output variables. In this technique, we split the population or sample into two or more homogeneous sets (or sub-populations) based on most significant splitter / differentiator in input variables.

2. EVALUATION

We took a whole set of 1000 questions from different programming languages and we browsed through the discussion forum to our evaluation we got with a total of 950 questions where answered and the rest of 350 questions where unanswered .Approximately about three-fourth of those questions were responded with at-least two to three answers and the rest with the single answers. Since we used historical data for the training of the model , we made sure to use only the the time stamped features and thus only the data was available when each answer was posted.

Evaluation measures

In a machine learning model, the prediction is based on the formation of the data-set . In the data-set the class label or the attribute (number of times accepted) that is to be used plays a vital role . The algorithms are used to provide the a class probability I.e., for the training data-set probabilities are calculated by using both the algorithms .After the comparison of the algorithms ,these calculations estimates the relevancy in the discussion forum. The analysis can provide which discussion forum is best to refer and discuss .The attributes like questions ,answers , number of times accepted and relevancy are used .However , the number of times accepted is formally divided and a grading format is assigned to make the better and accurate predictions. The customized grading system was based on the number of times the answer is accepted in the discussion forum and also the up-votes to the answer. The grading system includes four categories they are :

- Excellent : the range or the accepted points should be between 750-1000
- Good : the range or the accepted points should be between 500-749.
- Fair : the fair points are between 230-499
- Poor : The range or the accepted points should be between 0-229.

We have use this accepted points to give a proper definition rather than manually giving the grading format tag e.g.,if their is a thread for that class label can be placed manually as it is a implementation based we used the number of times the answer is accepted number suppose the answer is accepted in the range of 750 to 1000 then it is given the grade of “ Excellent ”. Similarly we used this formatting process for the grading in the other attribute’s class label section.

As training and evaluation was performed on historical data, we implemented ten fold cross validation to mitigate the impact of over-fitting . We can also use the percentages but the accuracy using it is less when compared to that of cross validation.

We used many features to get the more accurate predictions while training the model .They are:

Number of times accepted : As we discussed above , the answers of the user that had the most answers accepted is made into grade format and given the specific tag I.e., excellent , good, fair and poor.

Most votes received and most comments : The answers of the users from the discussion forums that have the most up-votes across all the forums and also the most comments respectively is considered.

Further , we trained machine learning models on all the features using “Decision Tree Algorithm” (a branching method to illustrate every possible outcome of a decision)

3. EXPERIMENTAL SETUP

We implemented the whole set-up in the Rapid-miner . Rapid-miner is required to be uploaded with a data-set for further operations .The data-set is manually prepared by using some specific formatting process. The available data-set can be saved as a CSV file / JSON file or excel file. For our proposed experiment we used a CSV file as shown in Fig.

The Rapid-miner provides different platforms to execute your model like Market Based Analysis ,Out-lier detection, Medical fraud detection, Web Analytics etc. We will use a blank process for this experiment and create process model or the design by putting different operators in Fig .2 like retrieve data-set, cross validation rule etc.

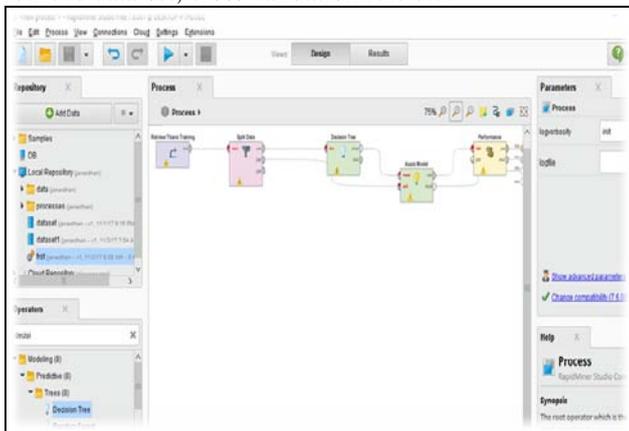


Fig.2 Rapid miner with different operators.

Methodology and Result:

The aim of designing in Rapid-miner is to supply the input along with the validation rule to the decision tree operator and see the results. This set-up give the performance evaluation of decision tree classifier. Following steps are to be in followed in Rapid-miner for setting up the process and getting desired result using decision tree classifier.

1. Create new Blank process in Rapid-miner.
2. Load data-set by using Rapid-miner.

The Rapid miner have different steps while loading the data-set . The data-set should contain a class label and that can be done using change role option that is provided in setting part below the each attribute part .Figure 3 is the loaded data-set into the Rapid-miner for operation.

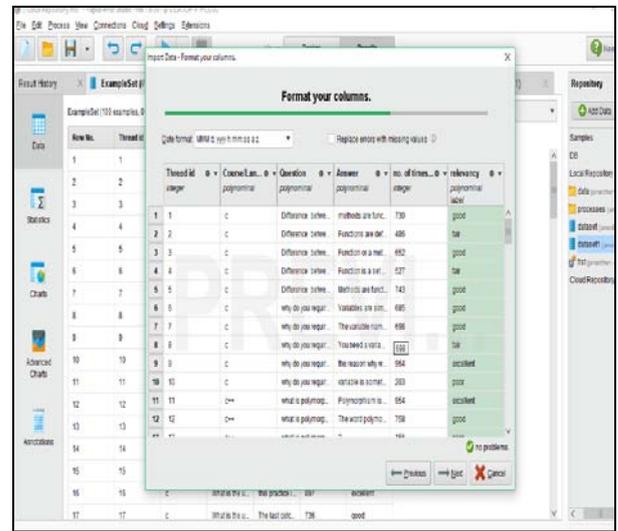


Fig 3: Loaded data-set into Rapid-miner.

The loaded data is available under ‘Local Repository in Rapid-miner.For plotting of data. A new blank process is added I Rapid-miner.

3. Save the process with the respective file name and
4. Run.
5. When the data-set gets loaded in the Rapid-miner provides some different statistical graphs.

For the proposed experiment we required to design a running model which is done in the designing section of the Rapid-miner. Following steps are taken in designing the process model.

- 1) Retrieves the MOOC data-set from the Repository.
- 2)Cross validation is connected to the data-set and it splits into two subsets. Stratified sampling is used to retain the distribution of label in both training and test stages.j

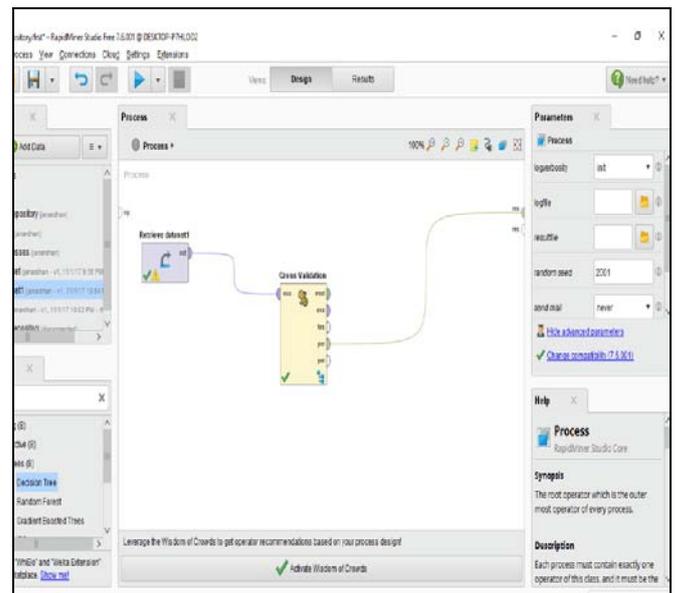


Fig 4: Process model When we double click on the cross validation we get stages like

- A) Training stage and
- B) Testing stage.

Training stage :

In this stage , we have to include retrieved data-set model connect it to the mod part of the training section . In the operator the classification algorithm that is decision tree is

selected and dragged on to the connected link (from retrieve data-set and mod part).The connection is established after this process.

Testing stage:

This stage is very much important to get the statistical measures where in apply model and performance models are connected and that link is given a final link to the res port in the testing stage. The input for this is taken form the connected port that mod port .

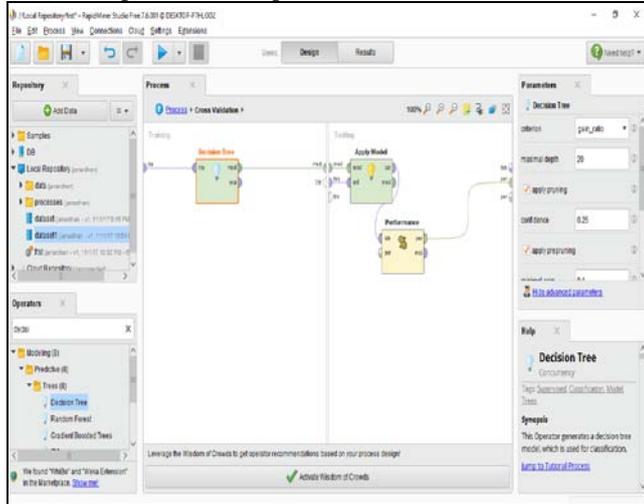


Fig 4: Cross Validation design.

3) Finally, go back to the process model, and make a final connection from cross validation to the res port .

4) Hit on run button.

The statistical measures are provided after the whole process have run successfully.

When the process is run, then the following output is generated in following:

- I. Simple Distribution
- II. Example-set(Apply-Model)
- III. Performance Vector.

The above three categories have further classification which are

I Simple Distribution

- a)Description
- b)Charts(Fig.6)
- c)Distribution Table

Ii)Example-set(Apply-Model)

- a)Data(Fig.7)
- b)Statistics(Fig.8)
- c)Charts
- d)Advanced Charts

III) Performance Vector

- a) %performance
- b) Description(Fig.9)
- c) Annotations.

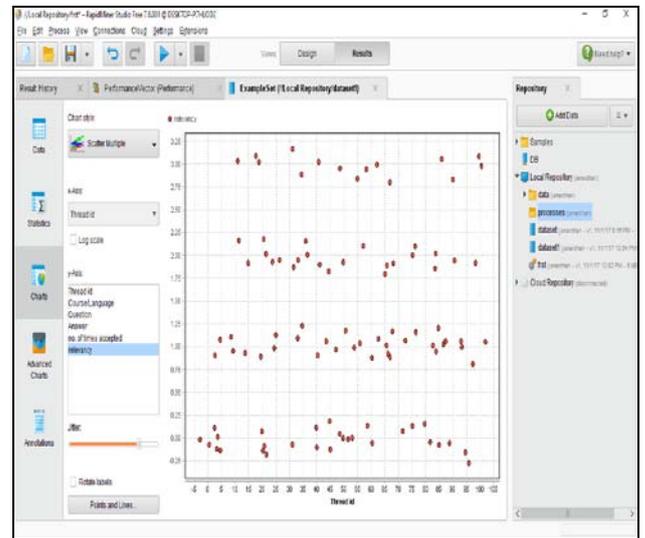


Fig.6: Simple Distribution (Charts)

Row No.	category	Threshold	CourseLang.	Question	Answer	no. of times accepted
1	good	1	c	Difference b...	maths are...	733
2	bad	2	c	Difference b...	Functions ar...	468
3	good	3	c	Difference b...	Functions ar...	652
4	bad	4	c	Difference b...	Functions ar...	527
5	good	5	c	Difference b...	Methods are...	743
6	good	6	c	why do you...	Variables are...	665
7	good	7	c	why do you...	The variable...	468
8	bad	8	c	why do you...	There are a...	422
9	excellent	9	c	why do you...	The reason is...	994
10	good	10	c	why do you...	variables are...	263
11	excellent	11	++	what's polym...	Polymorphic...	994
12	good	12	++	what's polym...	The most po...	750
13	good	13	++	what's polym...	?	100
14	bad	14	++	what's polym...	Polymorphic...	473
15	good	15	++	what's polym...	is call to m...	725
16	excellent	16	c	What is the...	this progra...	667
17	good	17	c	What is the...	The last opt...	750

Fig.7: Apply model(data)

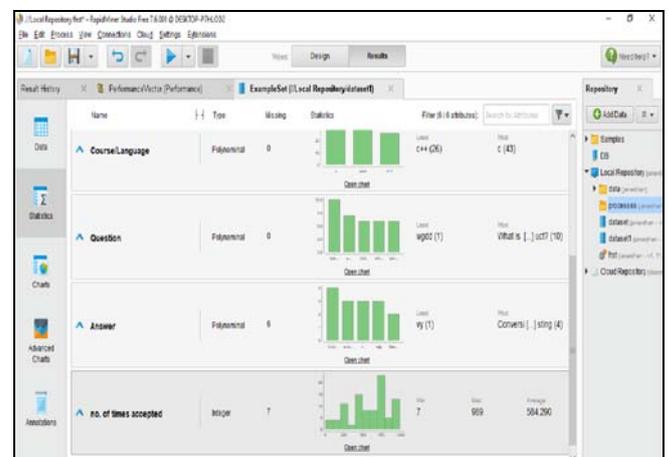


fig 8: statistics

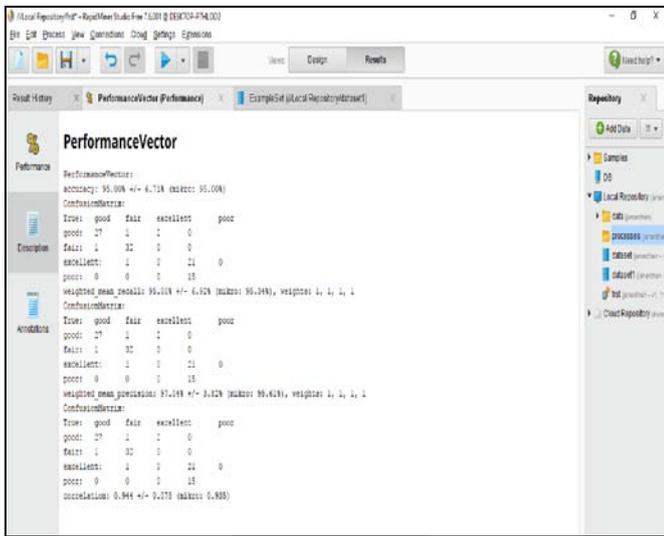


Fig:9 Description

4. CONCLUSION

Discussion forums offer a great platform for students to share their understanding of the subject. However, when it comes to relevancy of answers posted is relatively poor and some are left unanswered. This increases the dropout ratio of the students consistently. Our work presented here shows the analysis of relevancy of the posts to programming subjects of computer science. The relevancy obtained shows fair grade scale that indicates the discussion in the forum is average. This clearly states that by the formation of proper data-set using different discussion forum data we can find the relevancy also.

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