A Strategic Approach in Handling Information Retrieval Course for Attaining Course Outcomes – A Case Study

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Abstract:

Background:
The major challenge in engineering education is to educate and train the student’s community with pedagogical practices in order to determine the notable skills to solve more complex and fragile structured problems. Innovative pedagogical practices make the students to adhere to the complex formulations of the corresponding domain and its applications. The realm of good understanding of knowledge and its deliverables can be ascertained through pedagogical practices with a focus on learner-centric activities in the classroom teaching.

The actual target in teaching learning process is to make all the students to have a good exploration on the domain knowledge with success ratio. The learner success is considered to be the core metric with which we can judge the success of learner-centric activities with creativity and quality processes. Confirming the process in which the learners are engaged with the key ideas of the course to be taught makes the students to have practical implementations in Teaching-Learning process.

Objective:
This research work focus on the assessment and evaluation of learner-centric techniques for outcome based education upon statistical evaluation. Students of third year (VI Semester) of two consecutive batches have been analyzed for the course on Information Retrieval (14ITPS0). Two set of batches 2015-19 and 2016-20 have been considered for the assessment and analysis of Active Learning Strategies (ALS). The incorporation has been processed using the strategic approach based on daily, weekly and monthly assessments focusing on student learning criteria and their responding behaviors for 2016-20 batch. Each paradigm has been measured corresponding to the course outcomes at each level. Significant statistical analysis has been made for validating the process behind the teaching learning process.

Real time case study:
Considering the batch 2015-19 summary assignments, presentations has been used for the evaluation and assessment of interim assignments. The observed response from the students went well among the student’s community. All the students have been registered for the course and they too have completed the same. But, the time and success ratio in learning mechanism varied from student to student upon completion of the course. Considering the batch 2016-20 active learning strategies such as Quiz by Kahoot!, Flipped classroom activity and MOOCs course has been incorporated. Also, for MOOC’S online courses a set of two courses has been identified relevant to the subject of study such as Text Retrieval and Search engines & Text mining and Analytics. For the Research problem identification the students are allowed to choose the domain specific topics for the course on information retrieval. In this process the time and success ratio has also been observed it also varied from one student to another.

Keywords: Education Technology, Course Outcomes, Target Measures, Interim Assessment, Learner-centric activities, Outcome-based Education.

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1. Introduction
The domain of educational technology is moving its wings from teacher-centric to learner centric activities in order to make the students to have an effective content delivery and understanding of each subject due course of study. The impact of learning among the students community must make them to have an adaptable environment in the teaching learning process. This will make changes in the process behind the student learning schemes. Accessing and evaluating the student deliverables must be made accordingly to that of the student’s acceptable patterns which then surely reflects some fruitful course outcomes.

Learning by doing some activities will surely make the student’s environment more flexible and adaptable with
specific constraints. These constraints will vary accordingly from student to student learning process and their learning environment. Activities concerning to ICT in OBE with a keen focus on Active Learning Strategies (ALS) will significantly have an impact in the teaching learning process.

This research paper deals with the applicability of ALS for the course on Information retrieval with the motto of making significant change in the teaching learning process. Strategies focusing on quiz, flipped classroom, online learning along with research-based exploration have been considered for the assessment and evaluation of course outcomes with the incorporation of ICT in action.

2. Literature Review

The work by the authors (Sheik Abdullah, et.al. 2018) proposed a target fixing scheme for the attainment of course outcomes. The target fixing has been made accordingly in two ways with upper and lower control limits. The proposed target fixing scheme by the authors made a significant improvement in the attainment of course outcomes for the course on information systems.

3. Technical Details and Implementation

Self-learning is considered to be an important part in student learning process (Parkavi et.al. 2018). The intention of self-learning varies significantly from student to student based on their ability and learning criteria (Baker et.al. 2012). In this case study we have considered two batches for measuring the effectiveness of learning with the proposed strategic approach. The course on information retrieval consists of 5 modules such as:

1. Data modeling
2. Querying
3. Text operations
4. Web search
5. Applications

The following Figure 1 depicts the course contents and its structure in the form of Cmap.

The following are the course outcomes in which the student will acquire upon completion of the course. Upon completion of the course the students will be able to:

- Use information retrieval modeling techniques for Corpus documents
- Apply query processing techniques to locate relevant information from the large collection of data
- Apply information retrieval techniques for textual data
- Evaluate different information retrieval systems for web search tasks
- Develop simple information retrieval system for applications like personalization and recommender systems, search engines, etc

The assessment of course evaluation lies at remember level of about 20%, understand level of about 20%, apply level 50% and analyze at 10%. The level of each module has to be mapped in accordance with the understanding level of the students (Ouafae et.al. 2019). The contents have been formulated in such a way that it coincides with the understanding of each and every student (Gloria et.al. 2013). The Active Learning Strategies (ALS) included is:

1. Quiz by Kahoot! (Daily Assessment)
2. Flipped classroom (Weekly Assessment)
3. MOOCs online courses (Monthly Assessment)

Apart from this the student with more adverse exposure will be given an opportunity to formulate research proposals as an activity for making them to focus on research-based applications (Duarte et al., 2020). The following Figure 2 provides the daily evaluation by using ICT tool (Kahoot!). The students with correct answers will be ranked and the same will be displayed through podium. This mechanism makes the students to participate enthusiastically with interest and understanding of leaning the concepts (Ounjit, et.al. 2011). The attainment is evaluated with the following calculation process:

**CO, PO, PSO Attainment Calculation Process:**
Use historical data for fixing the targets
Expected Proficiency (EP)
Expected Level of Attainment (ELA)

**Target Fixing**

- Count the number of Students Grade wise
- Calculate Average of two batches for each Grade
- Find cumulative % of Average
- Set EP, in which the cumulative Average crosses 50%
- Set ELA with 30% increase (use nearest 10s)
The weekly assessment is made through another ICT mechanism with a keen focus on flipped classroom activity (Rhode Island, 2016). This is pedagogical approaches which express the instructions from group learning to an individual learning environment (Parkavi, et.al. 2017). This situation enables the students to have a creative learning atmosphere where they can apply the concepts related to the flipped classroom activity (Friess, et.al. 2020). The modules has been segregated into three components one focusing on Quiz, another on flipped classroom, and finally on MOOCs and preparation of research report. The following Table 1 depicts the contents identified for preparation of flipped classroom activity.

Table 1. Flipped classroom weekly assessment

| Week  | Topic                                           |
|-------|-------------------------------------------------|
| 1-4   | 1. Classical Information retrieval models       |
|       | 2. Retrieval Evaluation                         |
| 5-8   | 3. Querying                                     |
|       | 4. Structural queries                           |
|       | 5. User Relevance Feedback                      |
|       | 6. Local and Global Analysis                    |
| 9-12  | 7. Text operations                              |
|       | 8. Text Compression                             |
|       | 9. Indexing and Searching                       |
|       | 10. Inverted files                              |
| 13-16 | 11. Web search                                  |
|       | 12. Hyperlink search                            |
|       | 13. Browsing                                    |

The assessment is made in such a way that each of the student is assigned to concentrate on an individual topic with the preparation of flipped classroom activity (Naif et.al. 2013; Nikkie, 2016). The recordings are made using ScreenCAST OMatic tool with recording of about 20 minutes and the same has been published in you tube channel. Some of the students prepared most admirable video lectures and liked by most of the public users in you tube channel. One of the most viewed video as prepared by M.Divya which has got 3662 views and likes for the topic on inverted files corresponding to module 3.

Similarly the monthly assessment is made by using MOOCs online course in any of the online learning platform (Karthikeyan, P, 2018). This will makes the students to have a learning environment which makes them to explore their research ideas and practical applications (Larson et.al., 2020). Here we have used coursera as another platform to make them to disseminate their ideas/views to have practical implementations (Jennifer, 2016). The 2016-20 batch has been provided with these set of ICT learning mechanisms for exploring their learning practice and the attainment of course outcomes (Lekha, 2020; Levine, 2016).

4. Experimental Results and Evaluation

Experimental results have been carried out by considering the marks corresponding to interim assignment along with the improvements in internal and its terminal examination marks (Sun et.al., 2020). The ratio given to internal marks was 0.3 and to that of interim assignment is 0.3 and for the terminal examination is 0.4 accordingly. The data is formalized to evaluate the normal distribution in order to identify the statistical relevance. Upper Control Limit (UCL) and Lower Control Limit (LCL) have been set to determine the level of attainments and calculations (Jodi, 2016). The following Table 2 describes about the sample questions and its corresponding blooms level.

Table 2. Sample questions during practice

| Questions                                                                 | Blooms level |
|--------------------------------------------------------------------------|--------------|
| Recall the process of information retrieval system.                      | Remember     |
| Explain the roles and responsibilities of database producers and vendors for online IR systems. | Understand   |
| Apply the concept of vector space model and determine the following cases: | Apply        |
| 1. Construct tf-idf values.                                              |              |
| 2. Given the query “gold silver truck” calculate if-idf vector for the query. |              |
| 3. Construct the score of each document relative to the query.           |              |
| 4. Calculate the length of each document and query with accordance to cosSim(d1.q), cosSim(d2.q) and cosSim(d3.q). |              |
| Experiment the key expression for ranking computation in probabilistic model. Provide the methods for computing the probabilities initially | Analyze      |

https://www.youtube.com/watch?v=8VjULlboO34
In this course we have 5 course outcomes the measurement is made in such a way that the direct attainment for CA is 60% and to that of TE is 30%. The following Table 3 describes the attainment values observed for the 2015-19 batch without the applicability of ALS.

Table 3. CO attainment for 2015-19 batch

| Course Outcomes | CO 1 | CO 2 | CO 3 | CO 4 | CO 5 |
|-----------------|------|------|------|------|------|
| Direct CO Attainment (CA) EP 'B' | 43 | 1 | 13 | 21 | 12 |
| Direct CO Attainment in % (CA) | 68.2 | 1.5 | 20.6 | 33.3 | 19.0 |
| Normalized (ELA - 60) | 113.7 | 2.6 | 34.3 | 55.5 | 31.7 |
| Direct CO Attainment in % (CA - 0.6 wt) | 68.25 | 1.5 | 20.6 | 33.3 | 19.0 |
| Direct CO Attainment (TE) | 33 | 33 | 33 | 33 | 33 |
| Direct CO Attainment in % (TE) | 52.3 | 52.3 | 52.3 | 52.3 | 52.3 |
| Normalized (ELA - 60) | 87.3 | 87.3 | 87.3 | 87.3 | 87.3 |
| Direct CO Attainment in % (TE - 0.3 wt) | 34.9 | 34.9 | 34.9 | 34.9 | 34.9 |
| Average Indirect CO in % | 67 | 55 | 68 | 47 | 45 |
| Indirect CO Attainment (0.1 wt) | 6.7 | 5.5 | 6.8 | 4.7 | 4.5 |
| CO Attainment | 100 | 42.0 | 62.3 | 72.9 | 58.4 |

In Indirect measurement we have made 10% which is got through means of feedback (Kim & Strimel, 2020). The following Table 4 describes the attainment values observed for the 2016-20 batch with the applicability of ALS.

Table 4. CO attainment for 2016-20 batch

| Course Outcomes | CO 1 | CO 2 | CO 3 | CO 4 | CO 5 |
|-----------------|------|------|------|------|------|
| Direct CO Attainment (CA) EP 'B' | 43 | 21 | 34 | 39 | 25 |
| Direct CO Attainment in % (CA) | 68 | 33 | 53 | 61 | 39 |
| Normalized (ELA - 60) | 113 | 55 | 89 | 103 | 66 |
| Direct CO Attainment in % (CA - 0.6 wt) | 68 | 33 | 53 | 61 | 39 |
| Direct CO Attainment (TE) | 43 | 43 | 43 | 43 | 43 |
| Direct CO Attainment in % (TE) | 68 | 68 | 68 | 68 | 68 |
| Normalized (ELA - 60) | 113 | 113 | 113 | 113 | 113 |
| Direct CO Attainment in % (TE - 0.3 wt) | 45 | 45 | 45 | 45 | 45 |
| Average Indirect CO in % | 77 | 82 | 96 | 76 | 81 |
| Indirect CO Attainment (0.1 wt) | 7.7 | 8 | 9.6 | 7.6 | 8.1 |

Table 5. Descriptive statistics for 2016-20 batch

| Statistic | CO1 | CO2 | CO3 | CO4 | CO5 |
|-----------|-----|-----|-----|-----|-----|
| No. of observations | 63 | 63 | 63 | 63 | 63 |
| Minimum | 21.7 | 0.0 | 9.0 | 22.0 | 23.4 |
| Maximum | 100.0 | 51.8 | 95.4 | 96.0 | 82.9 |
| 1st Quartile | 65.2 | 11.1 | 30.3 | 51.5 | 48.9 |
| Median | 80.4 | 18.5 | 42.4 | 69.0 | 56.3 |
| 3rd Quartile | 94.5 | 31.4 | 61.3 | 85.5 | 63.8 |
| Mean | 77.2 | 21.6 | 46.6 | 67.5 | 56.6 |
| Variance (n-1) | 392.8 | 176.6 | 456.3 | 476.2 | 174.9 |
| Standard deviation (n-1) | 19.8 | 13.2 | 21.3 | 21.8 | 13.2 |

Table 6. Descriptive statistics for 2015-19 batch

| Statistic | CO1 | CO2 | CO3 | CO4 | CO5 |
|-----------|-----|-----|-----|-----|-----|
| No. of observations | 65 | 65 | 65 | 65 | 65 |
| Minimum | 21.7 | 0.0 | 9.0 | 22.0 | 23.4 |
| Maximum | 100.0 | 100.0 | 100.0 | 100.0 | 100.0 |
| 1st Quartile | 65.2 | 11.1 | 30.3 | 51.7 | 48.9 |
| Median | 80.4 | 18.5 | 43.1 | 69.5 | 56.3 |
| 3rd Quartile | 95.6 | 33.3 | 63.6 | 86.2 | 64.0 |
| Mean | 77.6 | 22.8 | 47.4 | 68.0 | 57.2 |
| Variance (n-1) | 394.6 | 269.8 | 493.6 | 485.1 | 201.6 |
| Standard deviation (n-1) | 19.8 | 16.4 | 22.2 | 22.0 | 14.1 |

The observed results show that learning based on the proposed method for the student’s batch of 2016-20 shown an overall improvement in success ratio with improvements over time complexity measures. Also, the measurement of student’s performance against CLO threshold and PLO threshold found to be efficient with statistical evaluation of P value <0.0001 (Sheik Abdullah et al., 2018). Hence this effectiveness can be implemented towards all courses semester-wise (Marco et al. 2015). From the results observed with the Table 5 and Table 6 it has been came to notice that the data corresponding to 2015-19 is lower in statistical performance that its preceding batch of students. Hence it has been statistically proved that the performance of the students improved with regard to the incorporation of ALS and its remedial measures. Significant improvisations are also noticed for the batch of 2016-20 batch of students in the computation of variance and SD analysis. Also, from Table 4 for the 2016-20 batch all the Cos has been attained with the Expected Proficiency of level ‘B’ and ALS also been verified accordingly.
The descriptive statistics has been made for the 2016-20 batch of students the SD values has been found to be nominal with the CO values and its preferable outcomes (Thayer-Hart, et.al., 2010). The values corresponding to CO2 has little extent to lower deviations in box plot analysis meanwhile the other COs addressed to the desired extent (William et.al., 2015). The following Table 7 describes the confusion matrix that has been observed as a part of predictive analysis for the 5 COs that has been used for evaluation.

| class | recall | precision |
|-------|--------|-----------|
| true Zero | 93.75  | 71.43     |
| true One | 86.67  | 71.43     |
| true Two | 84.62  | 71.43     |
| true Three | 75.00  | 71.43     |
| true Four | 71.43  | 71.43     |
| pred. Three | 100.00 | 71.43     |
| pred. Two | 73.33   | 71.43     |
| pred. Zero | 100.00 | 71.43     |
| pred. One | 76.47   | 71.43     |
| pred. Four | 71.43   | 71.43     |

5. Limitations

The systematic approach that we have adopted has been applied to the course on Information Retrieval (14ITPS0). The success rate depends upon the practice and the way of learning that has been adhered for other courses. Some of the courses will be adaptable to other such ALS techniques and ICT tools that best suits the learning process.

6. Conclusion and Future Work

The major challenge in engineering education is to evaluate and examine the students in an efficient way of Teaching-Learning process. The process has to be defined in such a way that it should make the students to explore their knowledge with adaptable practical and thinking skills. This research work focus towards the applicability and non-applicability of ALS strategies in a strategic way with an intention to measure the learning performance of the students for each of the component, with an intention to propose a remedial measure to increase the learning process.

In considering the batch 2015-19, this follows the traditional approach the final level of CO attainment has been addressed by the CO1 and CO5 only. But for the batch 2016-20 the entire COs has been attained with the applicability of ALS in action. We have considered the course on Information Retrieval for the effective implementation of ALS strategies among the student’s community. The results proved that research based learning with online MOOC’s courses improved the learning process among the students with improved accuracy of about 84.38% respectively. The statistical measure has also been evaluated with P value <0.0001 accordingly. Hence this will ensure the improvement of performance of learning among the student’s by targeting it with a threshold value. Thereby the strength’s, pitfalls, of the learning process can be efficiently observed with the proposal of new/changes in the Teaching-Learning process, or the course content. The different capabilities with regard to e-mental health interventions are the future aspects of this proposed scheme. The data corresponding to virtual/online learning is observed for a batch of students in order to measure the effectiveness in learning with regard to the modern technology.

Acknowledgement

On behalf of our Management we would like to thank our beloved principal, deans, faculty members and students for making us to explore our views in education technology and its practice.

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