Independent Study CS499

Art Grichine

December 9, 2015

Introduction

Every semester, The Computer Science department at California State University, Fullerton faces oversubscribed courses. This problem affects both the students and the staff in the department.

For the students of Cal State Fullerton, oversubscribed courses lead to uncertainty during registration, inability to fill required course load, and even delay in graduation. Sometimes additional courses are added or a student is chosen from a wait list. However, for every student that is selected from the waitlist, countless others are unable to take the necessary class. I have personally experienced this problem and have been at the mercy of the department to make room for me.

Currently, course demand is based on the department's intuition. This leads to difficult decisions regarding what courses the department should offer and how many sections to offer per course. Furthermore, these decisions must be made by the department several months before the semester to ensure staff availability thereby leading to difficulties with last minute additions due to demand.

By applying machine learning principles (e.g. regression, neural networks), Dr. Wortman and I believe that it should be possible to predict the demand for each course offered by the department.

Review of Literature

While researching the enrollment forecasting topic, it came as a surprise that there was not much literature covering this topic. Most research in this domain focuses on enrollment in the higher-education system rather than dealing with the individual department. Two articles were found from 1973 and 1975. However, a more recent paper from 2005 by Kraft and Jarvis [?] addresses these papers. Kraft and Jarvis attempt to solve the enrollment forecasting problem through a seat release system which attempts to predict the number of seats needed for a given section. They identify student characteristics that are significant in predicting course enrollment such as first-time vs. transfer student and pass/fail rates of courses. Data from the previous three years is used as a historical estimation and verification for the parameters in the seat release model. The model output reports the minimum, mean,

and maximum predicted values for an individual course. The authors argued that their model proved to be accurate and robust.

More recently, a lecture from University of Missouri-Columbia in 2010 by Felts and Ehlert [?] presented an approach which considers a ratio of student demand for a course to the total students in the available population. They use the historical calculated student demand correlation value and apply it to the new student population. The output is the new predicted student demand.

Methods

Course enrollment models not using department intuition consider regression analysis to determine course demand. However, regression models may have large error when department policies change regarding curriculum hours and courses required. Also, the student progression through the courses can vary significantly due to individual professor grading. The model used for this study considers historical course grades to predict the amount of students that will pass the course and continue and course demand which is gathered through a questionnaire.

Data

Student data was generated through a Python script called Dataset_Generator.py. The dataset generator used the libraries Pandas, Numpy, and MatPlotLib to generate, manipulate, and store data to a .csv file which is then read and used by other programs. This dataset generator was created due to the inability to timely gather needed data for this project. Instead, a Python dictionary of courses was created to hold all undergraduate courses offered by the department of Computer Science at Cal State Fullerton. The main program Course_Progression.py calls methods belonging to the Dataset_Generator.py program to generate data for its computation. The functions handle_grades() and handle_demand() are called from the main menu of Course_Progression.py. Depending on the user's menu selection, the Dataset_Generator.py methods are used to generate the appropriate data. The user also has the option of loading their own data into the program.

Demand Data

The handle_demand() function in Course_Progression.py asks the user whether they would like to load demand data or if they would like to generate a dataset. Loading data is a trivial action in which the Pandas read_csv() method is used and a Pandas data frame containing the file contents is returned to the main menu. The generate demand option calls the generate_demand() method from the Dataset_Generator.py program. This method uses the default value of students=1000 and creates a Pandas data frame of current courses and wanted courses for 1000 individual student. This is done to emulate a questionnaire which would ask students which courses they are currently enrolled in and which courses they would like to take next semester.

As the Pandas data frame for student demand is being created, list Python's list comprehension is used to generate individual samples of which courses a student is currently taking

and what they would like to take. In the list comprehension, the generate_samples(students) function is called to generate individual samples. The generate_samples(students) function iterates through the range of the amount of samples needed (in this case students=1000) and appends a random sample of courses from the cs_courses dictionary into a list of lists (a list of 1000 students each with their own preference). The Numpy random.choice() method is used to vary the amounts of courses a student is taking. This method takes a parameter p which can be set to a probability distribution for the values chosen. Here, we would like to select between one to six courses for each student (simulating a student course load). However, in most cases it is likely that the student is taking between three to five courses and not one or six courses. Thus, using np.random.choice() we can set a higher probability at choosing values of three to five and less at choosing the values of one and six. Once the choice of how many courses to sample from cs_courses is returned. The random.sample() function is called to choose a random sample of the cs_courses dictionary. The samples are then returned to the data frame of current courses and wanted courses in the generate_demand() method and the data frame is saved to a demand_data.csv file and returned to the caller.

Grade Data

Just as with demand data, the grade data must be loaded or generated for this program. When the user selects from the main menu the option generate or load student demand data, the handle_grades() function is called and the user is prompted to select if they would like to load or generate grade distribution data for their run. If the user chooses to load grades from their own dataset, they will be asked how many grade datasets they would like to load. The user will then enter the names for the .csv file in the working directory one by one until the grades are successfully loaded. However, if the user would like to generate grade distribution data, the user is asked how many datasets would they like to generate and then the users input is passed to the generate_grades() method from Dataset_Generator.py to generate and load the necessary data frame.

The generate_grades() method from Dataset_Generator.py program takes one parameter which contains the amount of records needed. It returns a list of data frames which contain grade distributions for each simulated semester. Each individual data frame contains all courses offered by the undergraduate program at the Computer Science Department at Cal State Fullerton. A loop which iterates over the range of records needed calls the assign_grades() function which returns a grade distribution for all courses. This result is then appended to the list of data frames and the file is saved to the working directory for record. Once the loop is complete, the list of data frames is returned to the caller.

The assign_grades() function creates a random Gaussian (normal) distribution by calling Numpy's np.random.normal() method. This method takes two parameters, mean or μ and standard deviation or σ . The mean grade value of $\mu = 75$ and the standard deviation value of $\sigma = 10$ was used to generate the course grades. However, with course grade distribution information this value can be trained to more adequately resemble each individual course distribution. In Figure ??, we see a histogram of the grade distribution generated for an individual course plotted with a probability density function (PDF) line to demonstrate ideal distribution. These grade values are recorded for a given course and then saved. This

process is repeated until all courses have grades. Note that each course has its own unique grade distribution as would be seen in practice. A data frame created from the distribution and then returned to the textttgenerate_grades() method. The data frame is saved for record and the loop continues. Once the needed amount of records is created the list of data frames is returned to the caller.

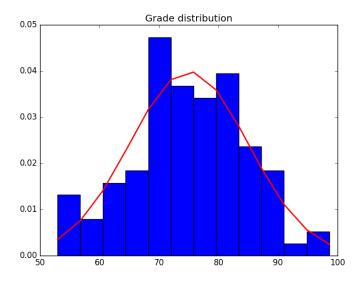


Figure 1: Grade Distribution For A Course

Forecast

To forecast data, the user must select the Forecast Enrollment option in the main menu. At least two files must be loaded before this option is possible. The demand data as well as the grade distribution data must be in loaded using the menu before the forecast enrollment option will work. If this option is selected before the necessary data is loaded, the user is asked to go back and load or generate the necessary data. When the necessary data is loaded, the forecast_enrollment() function is called and the forecast is computed. When the computation is complete, the result is saved to the working directory as forecast.csv and the head and tail of the data frame is shown to the user on the command line. The user is then directed back to the main menu where they may run another simulation or exit the program.

Forecast Enrollment

When the forecast_enrollement() function is called by the main menu, it takes the list of data frames containing the grade distributions for each course and a data frame which contains the current student courses and demand. The function then calls the grade_progression() function which takes the list of grade distributions to process the data.

In the grade_progression() function, we must extract the amount of passing grades for each course as well as the total amount of grades assigned. The list of data frames containing

the grade distributions for each simulated semester is enumerated through to extract the sum or passing grades (C- or better) and the sum of total grades assigned to each course. These records are aggregated into one data frame which contains all courses offered as well as the passing, total, and success ratio for each course. The success ratio is computed by

$$\frac{\sum \text{Passing Grades}}{\sum \text{Total Assigned Grades}} = \text{Success Ratio}.$$

This data frame is then saved to grade_stats.csv and returned to the forecast_enrollment() function.

Next, the forecast_enrollment() function must also extract the necessary information from the student records. The student_demand() function is called and the data frame containing the student records is passed as a parameter. Student demand must go through the records of every individual student generated and count the courses that the students are currently in and the ones that are demanded. The student_demand() function counts and aggregates all of the student demand information into a dataframe which contains the course number, course name, amount currently enrolled, and amount requesting course. The data frame is then saved as student_stats.csv and returned to the forecast_enrollment() function.

The forecast_enrollment() function then creates final data frame which will contain the forecasted contents as well as other important data. The course names and numbers are added to the data frame. The column containing the amount of students currently enrolled in each course is also added. Finally, the amount of students currently enrolled in each course is multiplied by the success ratio in the grade distribution data frame. The column containing the course demand is transferred to the data frame and the head and tail of the contents are displayed to the screen. The data frame contents are then saved to the working directory as forecast.csv and the user is directed back to the main menu.

Results

The program created works as intended. It is robust and did not crash during testing. The results of this study cannot be determined because of the inability to measure which courses are ultimately taken. This simulation would not only need to generate data for the students and grades but would have to also generated enrollment information for the following semester to compute error. However, there is no objective way to compute these values. Therefore, we cannot objectively judge the performance of this approach. An example of the output can be seen in Figure ?? as well as a full run in Figure ?? at the end of this document.

Discussion

The course enrollment forecasting topic affects most departments in higher education. Certainly the larger ones. When researching this topic, little research related to individual department course forecasting was found. This is surprising since the forecasting topic is

Figure 2: Full Run

researched at an institution level to compute the enrollment of the incoming class. The department offers a handbook which contains suggestions on which courses to take and at what time to be done in a timely fashion. However, many students follow their own path which is what makes enrollment forecasting for a department difficult.

There were two difficult parts when approaching this enrollment forecasting problem. First, the lack of research in this subject made it difficult to find an approach to improve on. Instead, it seems that to address this problem one will have to have an original approach which is only limited by the second difficulty, the data. As this study deals with student data there must be institutional review board (IRB) approval to access and use these records. For this study, sudo records were generated however this becomes a problem when approaching the testing phase of the model. Data is key when implementing any type of machine learning algorithm and the quality and quantity of data often dictates the model.

Though department intuition may play a role in the final decision of how many sections and which courses should be offered, there needs to be computational aid through some data analysis be it regression or a more advanced and adaptive model such as a artificial neural network.

References

- [1] Ehlert Felts. Prediction model for course demand at MU. 2010.
- [2] Christine R Kraft and James P Jarvis. An adaptive model for predicting course enrollment. Technical report, Technical Report, 2005.

```
bash-3.2# python Course_Progression.py
Welcome to the enrollement forcasting program. Please choose from the
following menu. Note: Data for student demand and course grades must
be loaded or generated before option 5 is available.
1 Generate or Load course grades.
2 Generate or Load student demand.
3 Forecast Enrollment.
4 Exit
Please Select: 1
1 Load Grades
2 Generate Grades
Please Select: 2
How many grade datasets would you like to generate? 2
1 Generate or Load course grades.
2 Generate or Load student demand.
3 Forecast Enrollment.
4 Exit
Please Select: 2
1 Load Demand
2 Generate Demand
Please Select: 2
Demand generated
-- dataset.csv found locally
-- writing to local demand_data.csv file
1 Generate or Load course grades.
2 Generate or Load student demand.
 Forecast Enrollment.
4 Exit
Please Select: 3
* df_forecast.head()
                     course name course number
                                                    enrolled
                                                               passing
                                                                         demand
   Introduction to Programming Programming Concepts
                                              120
                                                          88
                                                                    60
                                                          90
22
       Data Structures Concepts
       C Programming Visual BASIC Programming
                                                          87
 df_forecast.tail()
                               course name course number
                         Game Programming
                                                                   72
90
                Game Development Project
                                                       489
    Variable Topics in Computer Science
         Internship in Computer Science
                                                                             64
                        Independent Study
                                                                   68
    demand
        84
        78
77
File has been saved to 'forecast.csv' in the working directory
1 Generate or Load course grades.
 Generate or Load student demand.
 Forecast Enrollment.
4 Exit
Please Select: 4
Goodbye!
```

Figure 3: Full Run