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| **Capstone Project**  **Machine Learning Engineer Nanodegree** | Jed Isom  April 18th, 2016 |

**Definition**

*(approximately 1 - 2 pages)*

**Project Overview**

In 2011, I was interested in learning Mandarin Chinese and as part of that interest I decided to read a book in Mandarin. The book I decided to read had 3 columns on each page. In the 1st column, it had traditional Chinese characters, in the 2nd column it had the pinyin pronunciation of the characters in the 1st column, and the 3rd column had an English translation. As I was reading, if I didn’t know a Chinese character, I would look to the pinyin to see if it would jog my memory. If that didn’t work, I would look up the character in [this iOS app](https://itunes.apple.com/ca/app/pleco-chinese-dictionary/id341922306?mt=8), write the character, its pinyin and English translation in a notebook.

A couple of days into this reading process, I decided to track my progress because I’m a data driven guy and thought it would be interesting. I did this by recording how long I read each day (in minutes) as well as how many Chinese characters I read. I expected the average seconds taken to read a character to follow the traditional logarithmic scale learning curve. As I plotted my progress, there it mostly followed the expected curve, but there were some very obvious exceptions to the curve. I found that there were certain sections of the book, or periods of time, where I read much slower than I had previously. There were also instances where I read quicker than I would have expected.

In this project I am looking to find a way to explain more of the variance in my reading speed performance. I’ll use the logarithmic learning curve as the baseline performance for this supervised learning project. Luckily, this book can also be found and read online for free [here](https://www.lds.org/scriptures/bofm?lang=zho). To create the dataset, I used my handwritten notes to create a raw dataset where each row represents a date/study session. The columns in the raw dataset include

* Date
* Time spent reading (in minutes)
* Text read; copied and pasted text from the book that represents the text read during the study session. This was recreated digitally for this project based on my detailed notes I took while reading the book.

I plan to use the information within the ‘Text read’ field to help improve the prediction of my reading speed. Some potential features I’ll explore to explain some of the error in the baseline model include

* % of characters in study session text not seen before in the book
* Get the number of days since each character in the text was last read, and then take the mean.
* Potentially use Latent Dirichlet allocation (LDA) to create topics and use the posterior probabilities of the text being in a certain topic as a predictor
* …

I **can’t** use the number of characters in the text read field or the time spent values as predictors in this project. These 2 features are what I used to create the actual output (y variable) and so would easily fit the data, but not solve the underlying question about why my reading speed did not follow the traditional learning curve rate.

* *Has an overview of the project been provided, such as the problem domain, project origin, and related datasets or input data?*
* *Has enough background information been given so that an uninformed reader would understand the problem domain and following problem statement?*

**Problem Statement**

I plan to determine what underlying variables in the text that I was reading could have better predicted my Chinese character reading speed while reading this book. I will do this by treating this project as a supervised learning problem, where my output (y) is my reading speed in seconds per Chinese character, and my inputs (X) will be features I will derive from the text data and cumulative time spent or cumulative characters read. These last 2 features will represent the baseline learning curve model I’m using to compare against the supervised learning performance.

Tasks required to complete this project include

1. Use hand-written notes to create the digital dataset for analysis
2. Import the MS Excel entered dataset into Python
3. Scrub the Chinese Unicode text by removing numbers, punctuation, non-Chinese characters, etc.
4. Calculate preliminary characters per study session and seconds per character metrics
5. Validate data entry against handwritten notes to make there are no big mistakes in the data entry, go back and correct #1 as necessary
6. Create X features from the dataset to predict seconds per character. Potential features to create that might help explain the variation are listed again here.
   1. % of characters in study session text not seen before in the book
   2. Get the number of days since each character in the text was last read, and then take the mean.
   3. Potentially use Latent Dirichlet allocation (LDA) to create topics and use the posterior probabilities of the text being in a certain topic as a predictor
   4. …
7. Segregate training and test set data. This segregation will be random and static so that the same examples are always in the training set and test set. This will prevent data snooping.
8. Perform supervised learning on training set with cross validation to estimate generalized model performance
9. If cross validated model performance is worse than current best, then go back to #6 or #8 and try to create new/better features or learning models for predicting reading speed.
10. I’m not sure how I’ll know when to stop this loop from 6 to 9, but I will likely start to get diminishing returns on new/different features and stop at that point. At this point, I’ll have a solid set of predictors that will help me understand what other features determined my reading speed through the book.

In the end, I believe that using supervised learning I will be able to determine what features of the text I read have the most predictive power for my Chinese character reading speed. This will be valuable because I can use this information to improve my study habits based on this information. In a future project, this type of information might also be used to help someone else learn to read Chinese characters faster.

* *Is the problem statement clearly defined? Will the reader understand what you are expecting to solve?*
* *Have you thoroughly discussed how you will attempt to solve the problem?*
* *Is an anticipated solution clearly defined? Will the reader understand what results you are looking for?*

**Metrics**

As mentioned previously, the output variable, y, is the seconds per character reading speed for the study session. Since this value is very large in the beginning of the book and decreases exponentially due to the learning curve effect, I plan on taking the natural logarithm of this value for model fitting purposes. This will help the model error to be more uniform across the study sessions and thus not over fit study sessions near the beginning of the book.

The project is a regression problem. Kaggle has a [list of metrics to consider using for regression problems](https://www.kaggle.com/wiki/Metrics):

* Mean absolute error
* Weighted mean absolute error
* Root mean squared error
* Root mean squared logarithmic error

Based on what I know about the problem either simple mean absolute error or root mean squared error should work fine. I’ll be using root mean squared error because it is more commonly used, and also because it penalizes predictions farther from the actual much more than points closer to the prediction due to the squared term. This helps incentivize the model to fit these data points farther from the predictive model better, which is essentially what this project is about; fitting the data that varies from the baseline model better.

* *Are the metrics you’ve chosen to measure the performance of your models clearly discussed and defined?*
* *Have you provided reasonable justification for the metrics chosen based on the problem and solution?*

**Analysis**

*(approximately 2 - 4 pages)*

**Data Exploration**

As described previously, I copied and pasted text that I read during each study session into a digital dataset for this project’s analysis. Below is a sample row/instance of one study session of data.

|  |  |  |
| --- | --- | --- |
| Date | Time\_Spent | Text\_Read |
| 2011/09/07 | 30 | 他​心中​愚蠢​的​幻想。   12 最年長​的​拉曼​和​雷米爾​這樣​抱怨​他們的​父親。​他們​抱怨，​因為​他們​​不​知道​創造​他們的​神​的​作為。   13 他們​也​不信​耶路撒冷​那​座​大​城 |

In Excel, I was able to create a custom date format YYYY/MM/DD that would make it easier to transfer the dataset to the Python datetime package that takes (YYYY, MM, DD). The “Time\_Spent” columns is the time spent reading that study session in minutes. It was too much effort to record this data in a more granular form like seconds, so this will be a somewhat lumpy source of error because the seconds read during a session will jump in increments of 60 seconds. The “Text\_Read” column includes the unformatted/filtered text I read during the study session and copied into the dataset.

At a high level, this dataset contains 974 records (study sessions) spanning 29,227 minutes of study time. After preprocessing the text data (see the discussion on how this was done in that section below) I was able to count 334,195 total characters read. This gives an overall reading speed average of 5.247 seconds per character.

In this section, you will be expected to analyze the data you are using for the problem. This data can either be in the form of a dataset (or datasets), input data (or input files), or even an environment. The type of data should be thoroughly described and, if possible, have basic statistics and information presented (such as discussion of input features or defining characteristics about the input or environment). Any abnormalities or interesting qualities about the data that may need to be addressed have been identified (such as features that need to be transformed or the possibility of outliers). Questions to ask yourself when writing this section:

* *If a dataset is present for this problem, have you thoroughly discussed certain features about the dataset? Has a data sample been provided to the reader?*
* *If a dataset is present for this problem, are statistics about the dataset calculated and reported? Have any relevant results from this calculation been discussed?*
* *If a dataset is****not****present for this problem, has discussion been made about the input space or input data for your problem?*
* *Are there any abnormalities or characteristics about the input space or dataset that need to be addressed? (categorical variables, missing values, outliers, etc.)*

**Exploratory Visualization**

In this section, you will need to provide some form of visualization that summarizes or extracts a relevant characteristic or feature about the data. The visualization should adequately support the data being used. Discuss why this visualization was chosen and how it is relevant. Questions to ask yourself when writing this section:

* *Have you visualized a relevant characteristic or feature about the dataset or input data?*
* *Is the visualization thoroughly analyzed and discussed?*
* *If a plot is provided, are the axes, title, and datum clearly defined?*

**Algorithms and Techniques**

In this section, you will need to discuss the algorithms and techniques you intend to use for solving the problem. You should justify the use of each one based on the characteristics of the problem and the problem domain. Questions to ask yourself when writing this section:

* *Are the algorithms you will use, including any default variables/parameters in the project clearly defined?*
* *Are the techniques to be used thoroughly discussed and justified?*
* *Is it made clear how the input data or datasets will be handled by the algorithms and techniques chosen?*

**Benchmark**

In this section, you will need to provide a clearly defined benchmark result or threshold for comparing across performances obtained by your solution. The reasoning behind the benchmark (in the case where it is not an established result) should be discussed. Questions to ask yourself when writing this section:

* *Has some result or value been provided that acts as a benchmark for measuring performance?*
* *Is it clear how this result or value was obtained (whether by data or by hypothesis)?*

**Methodology**

*(approximately 3 - 5 pages)*

**Data Preprocessing**

In this section, all of your preprocessing steps will need to be clearly documented, if any were necessary. From the previous section, any of the abnormalities or characteristics that you identified about the dataset will be addressed and corrected here. Questions to ask yourself when writing this section:

* *If the algorithms chosen require preprocessing steps like feature selection or feature transformations, have they been properly documented?*
* *Based on the****Data Exploration****section, if there were abnormalities or characteristics that needed to be addressed, have they been properly corrected?*
* *If no preprocessing is needed, has it been made clear why?*

**Implementation**

In this section, the process for which metrics, algorithms, and techniques that you implemented for the given data will need to be clearly documented. It should be abundantly clear how the implementation was carried out, and discussion should be made regarding any complications that occurred during this process. Questions to ask yourself when writing this section:

* *Is it made clear how the algorithms and techniques were implemented with the given datasets or input data?*
* *Were there any complications with the original metrics or techniques that required changing prior to acquiring a solution?*
* *Was there any part of the coding process (e.g., writing complicated functions) that should be documented?*

**Refinement**

In this section, you will need to discuss the process of improvement you made upon the algorithms and techniques you used in your implementation. For example, adjusting parameters for certain models to acquire improved solutions would fall under the refinement category. Your initial and final solutions should be reported, as well as any significant intermediate results as necessary. Questions to ask yourself when writing this section:

* *Has an initial solution been found and clearly reported?*
* *Is the process of improvement clearly documented, such as what techniques were used?*
* *Are intermediate and final solutions clearly reported as the process is improved?*

**Results**

*(approximately 2 - 3 pages)*

**Model Evaluation and Validation**

In this section, the final model and any supporting qualities should be evaluated in detail. It should be clear how the final model was derived and why this model was chosen. In addition, some type of analysis should be used to validate the robustness of this model and its solution, such as manipulating the input data or environment to see how the model’s solution is affected (this is called *sensitivity analysis*). Questions to ask yourself when writing this section:

* *Is the final model reasonable and aligning with solution expectations? Are the final parameters of the model appropriate?*
* *Has the final model been tested with various inputs to evaluate whether the model generalizes well to unseen data?*
* *Is the model robust enough for the problem? Do small perturbations (changes) in training data or the input space greatly affect the results?*
* *Can results found from the model be trusted?*

**Justification**

In this section, your model’s final solution and its results should be compared to the benchmark you established earlier in the project using some type of statistical analysis. You should also justify whether these results and the solution are significant enough to have solved the problem posed in the project. Questions to ask yourself when writing this section:

* *Are the final results found stronger than the benchmark result reported earlier?*
* *Have you thoroughly analyzed and discussed the final solution?*
* *Is the final solution significant enough to have solved the problem?*

**Conclusion**

*(approximately 1 - 2 pages)*

**Free-Form Visualization**

In this section, you will need to provide some form of visualization that emphasizes an important quality about the project. It is much more free-form, but should reasonably support a significant result or characteristic about the problem that you want to discuss. Questions to ask yourself when writing this section:

* *Have you visualized a relevant or important quality about the problem, dataset, input data, or results?*
* *Is the visualization thoroughly analyzed and discussed?*
* *If a plot is provided, are the axes, title, and datum clearly defined?*

**Reflection**

In this section, you will summarize the entire end-to-end problem solution and discuss one or two particular aspects of the project you found interesting or difficult. You are expected to reflect on the project as a whole to show that you have a firm understanding of the entire process employed in your work. Questions to ask yourself when writing this section:

* *Have you thoroughly summarized the entire process you used for this project?*
* *Were there any interesting aspects of the project?*
* *Were there any difficult aspects of the project?*
* *Does the final model and solution fit your expectations for the problem, and should it be used in a general setting to solve these types of problems?*

**Improvement**

In this section, you will need to provide discussion as to how one aspect of the implementation you designed could be improved. As an example, consider ways your implementation can be made more general, and what would need to be modified. You do not need to make this improvement, but the potential solutions resulting from these changes are considered and compared/contrasted to your current solution. Questions to ask yourself when writing this section:

* *Are there further improvements that could be made on the algorithms or techniques you used in this project?*
* *Were there algorithms or techniques you researched that you did not know how to implement, but would consider using if you knew how?*
* *If you used your final solution as the new benchmark, do you think an even better solution exists?*

Before submitting your report, ask yourself…

* Does the project report you’ve written follow a well-organized structure similar to that of the project template?
* Is each section (particularly Analysis and Methodology) written in a clear, concise and specific fashion? Are there any ambiguous terms or phrases that need clarification?
* Would the intended audience of your project be able to understand your analysis, methods, and results?
* Have you properly proof-read your project report to assure there are minimal grammatical and spelling mistakes?
* Are all the resources used for this project correctly cited and referenced?
* Is the code that implements your solution easily readable and properly commented?
* Does the code execute without error and produce results similar to those reported?