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# **Udacity Reviews**

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 □ DISCUSS ON STUDENT HUB → Return to "Machine Learning Engineer Nanodegree" in the classroom Plagiarism Detector REVIEW CODE REVIEW HISTORY Meets Specifications Hi, your work is awesome! Keep it up with the hard work and you will make it for sure. I wish you the best in this program! All Required Files and Tests The submission includes complete notebook files as .ipynb : "2\_Plagiarism\_Feature\_Engineering" and "3\_Training\_a\_Model". And the test and helper files are included: "problem\_unittests.py", "helpers.py". The submission also includes a training directory source\_sklearn OR source\_pytorch . Nice work submitting all the notebooks and project files as required for evaluation. ]: +1: All the unit tests in project have passed. Splendid! All the unit tests have passed. 💯 Notebook 2: DataFrame Pre-Processing The function numerical\_dataframe should be complete, reading in the original file\_information.csv file and returning a DataFrame of information with a numerical Category column and new, class column. There is no code requirement here, just make sure you run all required cells to create a complete\_df that holds pre-processed file text data and Datatype information. Notebook 2: Features Created The function calculate\_containment should be complete, taking in the necessary information and returning a single, normalized containment value for a given answer file. Good job on the calculate\_containment function as it gets a single, normalized containment value for

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every given answer file. :: +1:

## **Pro Tips**

The CountVectorizer provides a simple way to both tokenize a collection of text documents and build a vocabulary of known words, but also to encode new documents using that vocabulary.

- How to Prepare Text Data for Machine Learning with scikit-learn
- The difference between CountVectorizer token counts and TfidfTransformer
- · Hacking Scikit-Learn's Vectorizers
- Provide an answer to the question about containment feature calculation.
- The function lcs\_norm\_word should be complete, taking in two texts and returning a single, normalized LCS value.

Impressive implementation of lcs\_norm\_word to return a single, normalized LCS value. 💥



#### **Pro Tips**

- · Longest Common Subsequence Pseudocode
- · Python Program for Longest Common Subsequence
- · Rosetta Code: Longest common subsequence
- Define an n-gram range to calculate multiple containment features. Run the code to calculate one LCS feature, and create a DataFrame that holds all of these feature calculations.

#### Notebook 2: Train and Test Files Created

- Complete the function train\_test\_data. This should return only a selection of training and test features, and corresponding class labels.
- Select a few features to use in your final training and test data.
- Provide an answer that describes why you chose your final features.
- Implement the make\_csv function. The class labels for train/test data should be in the first column of the csv file; selected features in the rest of the columns. Run the rest of the cells to create train.csv and test.csv files.

## Notebook 3: Data Upload

Upload the train.csv file to a specified directory in an S3 bucket.

## Notebook 3: Training a Custom Model

Complete at least one of the train.py files by instantiating a model, and training it in the main if statement. If you are using a custom PyTorch model, you will have to complete the model.py file, as well (you do not have to do so if you choose to use an imported sklearn model).

Well done completing the code in the train.py in source\_sklearn directory. 👍

#### Pro Tips

- StackShare: PyTorch vs scikit-learn
- Mhat are the differences in Scikit Learn Keras or Dutorch

~	Define a custom sklearn OR PyTorch estimator by passing in the required arguments.
~	Fit your estimator (from the previous rubric item) to the training data you stored in S3.
Notebo	ook 3: Deploying and Evaluating a Model
~	Deploy the model and create a predictor by specifying a deployment instance.
~	Pass test data to your deployed predictor and evaluate its performance by comparing its predictions to the true, class labels. Your model should get at least 90% test accuracy.
~	Provide an answer to the two model-related questions.
Notebo	ook 3: Cleaning up Resources
~	Run the code to clean up your final model resources.
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