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Machine Learning with Python-From Linear Models to Deep Learning

<u>Help</u>



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<u> 1</u>

<u>Unit 1 Linear Classifiers and</u>

Course > Generalizations (2 weeks)

Project 1: Automatic Review

> <u>Analyzer</u>

> 6. Automative review analyzer

6. Automative review analyzer

Now that you have verified the correctness of your implementations, you are ready to tackle the main task of this project: building a classifier that labels reviews as positive or negative using text-based features and the linear classifiers that you implemented in the previous section!

The Data

The data consists of several reviews, each of which has been labeled with -1 or +1, corresponding to a negative or positive review, respectively. The original data has been split into four files:

- reviews train.tsv (4000 examples)
- reviews_validation.tsv (500 examples)
- reviews_test.tsv (500 examples)

To get a feel for how the data looks, we suggest first opening the files with a text editor, spreadsheet program, or other scientific software package (like <u>pandas</u>).

Translating reviews to feature vectors

We will convert review texts into feature vectors using a **bag of words** approach. We start by compiling all the words that appear in a training set of reviews into a **dictionary**, thereby producing a list of d unique words.

We can then transform each of the reviews into a feature vector of length d by setting the $i^{\rm th}$ coordinate of the feature vector to 1 if the $i^{\rm th}$ word in the dictionary appears in the review, or 0 otherwise. For instance, consider two simple documents "Mary loves apples" and "Red apples". In this case, the dictionary is the set $\{\text{Mary}; \text{loves}; \text{apples}; \text{red}\}$, and the documents are represented as (1;1;1;0) and (0;0;1;1).

A bag of words model can be easily expanded to include phrases of length m. A **unigram** model is the case for which m=1. In the example, the unigram dictionary would be (Mary; loves; apples; red). In the **bigram** case, m=2, the dictionary is

(Mary loves; loves apples; Red apples), and representations for each sample are (1;1;0), (0;0;1) In this section, you will only use the unigram word features. These functions are already implemented for you in the bag of words function.

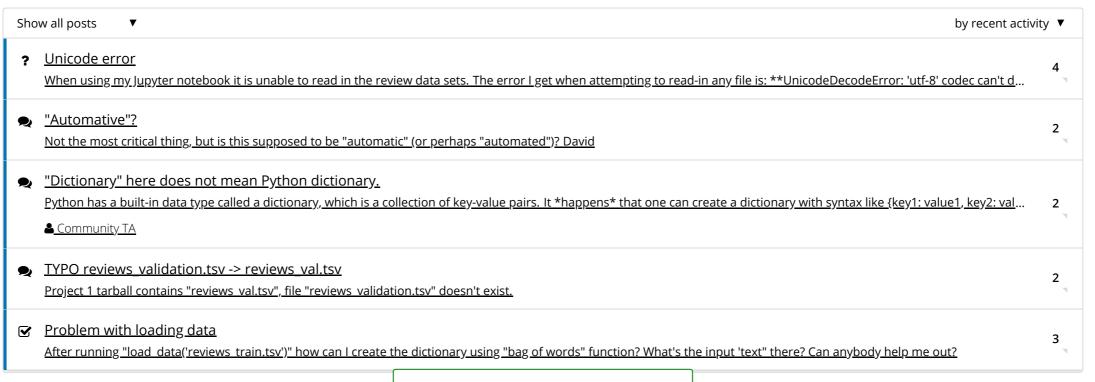
In utils.py, we have supplied you with the load data function, which can be used to read the .tsv files and returns the labels and texts. We have also supplied you with the bag_of_words function in project1.py, which takes the raw data and returns dictionary of unigram words. The resulting dictionary is an input to extract_bow_feature_vectors which computes a feature matrix of ones and zeros that can be used as the input for the classification algorithms. Using the feature matrix and your implementation of learning algorithms from before, you will be able to compute θ and θ_0 .

Discussion

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