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Predictive Analytics

Assignment Three Classifying Textual Data

Learning Outcomes:

- 1. Developing KNN data classification multi-class algorithm.
- 2. Applying KNN to classify textual data using distance metric.
- 3. Develop k-fold cross-validation algorithm to validate data.
- 4. Measuring performance of a supervised learning model.

In this assignment, you will apply a supervised learning algorithm to text data that you preprocessed in homework one and two. You will need to develop KNN (K- nearest neighbor) algorithm that finds similar documents to the document in question and predict its label (cluster/topic).

You must write your own <u>KNN algorithm</u> that takes, a raw document (not a vector): a file name or path, and the data matrix of TF/IDF along with an appropriate labels (e.g.: Predictive Analytics, Irma and Harvey, etc) you had in HW#1 and an integer K that specifies the nearest neighbors.

You may want to create a class or a method that takes care of taking a document and preprocesses that document into a vector (You can re-use code from HW#1 or 2).

See chapter data classification to learn more about K nearest neighbor algorithm.

You will need another KNN pre-built in a library *such as* to validate and compare you results: Java Machine Learning Library in order for your to make sure that you implemented KNN correctly.

It is highly recommended to use an Integrated Development Environment such as Eclipse to be able to import the <u>Java library with no issues</u>.

Tasks:

1. Develop a routine (method) that takes a raw document as an input, a similarity measure, and return the K most similar/nearest documents to that document and **the label (what topics this document is about)**. The similarity measure is **the distance you discover works** better according to your HW2.

We will be testing with new raw documents and test the predicted labels.

Please note that the following topics of the original documents:

C1: Airline Safety

C2: Amphertamine

C3: China and Spy Plan and Captives

C4: Hoof and Mouth Desease

C5: Iran Nuclear

C6: Korea and Nuclear Capability

C7: Mortrage Rates

C8: Ocean and Pollution

C9: Satanic Cult

C10: Store Irene

C11: Volcano

C12: Saddam Hussein

C13: Kim Jong-un

C14: Predictive Analytics

C15: Irma & Harvey

- **2.** Develop a 10 fold cross-validation function which **randomly** splits the training data and calculate the accuracy of your algorithm. Find the best k using your results.
- **3.** You will need to use another KNN pre-built library such <u>Java Machine Learning Library</u>, or Weka to compare your results. (It has cross-validation utility, performance measure)

Note: For selecting K, plot/compare average cross-validation errors on ranging the value of K.

For getting the value of cross-validation errors, call the utility which you created in clustering assignment to calculate the precision/accuracy.

4. Measuring Model Performance

Use the k chosen from 2 and 3 to generate *Confusion Matrix* from clustering assignment and print out the accuracy of the model on the **Test Dataset** along with the confusion matrix. You might also use <u>precision and recall/F-measure /confusion matrix</u> to measure the performance of your classification algorithm.

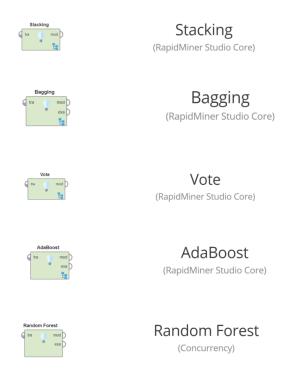
5. (Bonus + 10pts) Apply another classification algorithm (e.g. SVM) and compare your results to KNN for two documents of your choice. Document your algorithm that you used and the comparation with KNN.

6. Ensembles with Rapid Miner

In this part you will be using RapidMiner as a tool where you will you should run <u>at least</u> <u>one</u> of the following ensemble. It is known in the literature that ensembles produce better prediction. An ensemble method uses multiple predictive algorithms to obtain better predictive performance than could be obtained from any of the constituent learning algorithms alone.

You will need to use the matrix or the reduce version of matrix (e.g SVD, PCA) you produced in HW#1 into one of these ensembles.

You will need to build a process that will take new document with unknown label to be processed and fed to your ensemble. Summarize your results in a word document with snapshots of the processes used in RapidMiner and the results. Save the processes and attach them in the homework folder under a subfolder (essemblesRapidMiner). Click on the pictures bellow to learn more about the ensemble operators.



Include your code and your report in a Zip file. Make sure that you have all the following deliverables in your code and your report:

Deliverables	Points
ReadMe document (details how to export your project into Eclipse or JAVA IDE)	10
(7) Ensembles	20
Implement YOUR OWN KNN algorithm.	20
Implement cross-validation function on Training-data	20
Selecting the right K value in KNN. Attach snippet of plot/print Avg. Cross-validation errors vs. K value	10
Compare results with Java ML KNN library, or Weka or any other library	10
Measuring Model Performance on Test-data (see attached test data – you can select some of the documents from test data, run your algorithms and record the results in the report)	10
Total	100

The report should include and not limited to:

- 1. For all questions: Screenshot of each methods/classes (code) you wrote (for all question), simple explanation of each method (if you include comments in the code, that would be sufficient as a snapshot)
- 2. question 3 and 4: The plot of your metric with different K (let's say five different K are tried) (either 5 curves with x-axis as ten folders, or one curve with x-axis with different K)
- 3. question 5: K chosen, your confusion matrix and overall accuracy.
- 4. For the report on question 7 you will need to include all the snapshots of the processes of rapid miner and add short comments on your learning outcomes from this exercise.