Assignment:

Your assignment is to run a number of machine learning experiments on a set of news data and describe your experiments and findings. The task involves classifying news stories in two different ways:

1. **By Source Type:** The training and testing material will include both newswire (NWIRE) text digests and broadcast news (BN) transcripts. This is a binary classification: BROADCAST vs. TEXT.

2. **By Topic:** The data has been manually annotated for broad class of topics comprising general topics like “Accidents” or “Sports News”. There are thirteen broad categories; some helpful information about the classes can be found in Appendix B. Bear in mind, not every story is annotated for topic. You are only asked to classify those that are. Therefore, only construct feature vectors for a subset of the stories. You are only required to classify broad topics. You may not use the narrow topic labels referring to specific events (for example, “Deadly Fire in Bangladeshi Garment Factory”) as features!

You are to use the Machine Learning toolkit *weka* in order to run your classification experiments. To this end, one part of your submission will be a program that generates *weka* `.arff` formatted files. As discussed in class and in the *weka* documentation, these files describe your data set as a series of class-labeled feature vectors. Your program should read the data set and produce one `.arff` file for each classification, for a total of 2 files.

A good part of your assignment will be determining the features for learning. The feature set that you extract for use in these classification experiments should use at least three of the following types of features: words, n-grams (e.g., bigrams), POS tags, dates identified by your program for HW1. You may derive features from these feature types in different ways (e.g., you could use POS frequency for the different tags, you could use the POS tags themselves). You should experiment with the impact of these different features on your accuracy by using each of the three feature types separately and then in combination. For example, suppose you decided to use: 1. all words, 2. the 50 most frequent bigrams, and 3. frequency of dates. You would run the classifier four times, first with just words, second with just bigrams, third with date frequency, and fourth with all features. You may receive extra credit for additional features if they are well justified and if they increase your accuracy or for running and analyzing feature selection.
Obviously, you must not use any of the document labels (<SOURCE_TYPE>, <SOURCE_LANG>, <DOC_DATE>, <NARROW_TOPIC> etc.) as features in your feature vector.

You should try two different classification algorithms for each task so you can see how they operate on different tasks. We suggest J48 and Naïve Bayes.

For these classification experiments you should use 10-fold cross-validation. It is essential that you use the weka package that can be found at /home/cs4705/bin/weka.jar to run your experiments. If you do not, there is no guarantee that it will be possible to evaluate your final models. You must also export the model that yielded the best results for each task, and submit it along with your feature extractor code – if you do not, evaluating your submission will be impossible. Also, it is essential that you indicate the classifier and parameters that generated the submitted model.

You may find that you want to use features that are calculated relative to the entire data set. For example, “does this story have more or less words that the average story in the training data?” These types of features can be very useful. However, you need to be careful when using them in a cross-validation setting. These features should never be calculated using any testing material. This may force you to run the cross validation evaluation “manually”. That is, randomly dividing the training data into training and testing sets for feature extraction and evaluation. For your submission you may build a model on your entire training set.

For the classification by Source Type every story in a document will have the same class. However, the classification should still operate on the story level, not the document level. Therefore, it might make sense for every story from a document to have an identical feature vector.

Submission:

Submit your files together in a file named: <uni>HW2.tar.gz. Your submission should require as little human interaction as possible to test; therefore you MUST follow these instructions:

In your submission you must include the following files generated by your system:
1) sourceType.arff and sourceType.model
2) topicBroad.arff and topicBroad.model

The following are crucial scripts:
1) Submit one script to compile your code: make.sh
2) Submit **two** additional scripts, one for each classification task. Each of these scripts generates an arff file and runs weka on a given a directory that contains the input files.
These scripts will be used to test your models on unseen data, for example:

`./runSourceType.sh sourceType.model /home/nlp/hw2-testfiles`

- It will extract features from all *.input files in /home/nlp/hw2-testfiles => generates sourceTypeTest.arff file
- This script will also run weka using sourceType.model and sourceTypeTest.arff => weka result report. To get these results from the command line you can use the following:

```
java -Xmx1G -cp /home/cs4705/bin/weka.jar weka.classifiers.trees.J48 –l sourceType.model -T sourceType.arff
```

(assuming that J48 algorithm was used when you built your model)

2) `./runTopicBroad.sh topicBroad.model /home/nlp/hw2-testfiles`

You must also produce a write-up of your experiments. This write-up should describe your experiments and also must include a discussion of the processes you took and the results you obtained. Some questions that should be addressed can be found in the grading section below. This write-up should definitely include the cross-validation result report of the experiments you ran. Make your discussion empirical rather than impressionistic (i.e. refer to specific results/statistics, performance comparisons) whenever possible.

Materials:

You are provided with 1,409 files to develop your classifiers. This data set can be found at: `/home/cs4705/corpora/tdt4/ENG`. POS tagged data can be found at: `/home/cs4705/corpora/tdt4/ENG_POSTAGGED`. The dataset is made up of a mix of broadcast news transcripts and newswire digests (a set of newswire stories concatenated together). These files will have class information embedded in them that you will need to extract in order to construct the .arff files for training and classification. The file format is as follows:

```xml
<DOC>
  <SOURCE_TYPE> {BN|NWIRE} </SOURCE_TYPE>
  <SOURCE_LANG> {ARB|ENG|MAN} </SOURCE_LANG>
  <SOURCE_ORG> {News org. identifier from Appendix A.} </SOURCE_ORG>
  <DOC_DATE> YYYYMMDD </DOC_DATE>
  <BROAD_TOPIC> {Broad Topic Index from Appendix B.} </BROAD_TOPIC>
  <NARROW_TOPIC> {Narrow Topic Index from Appendix B.} </NARROW_TOPIC>
  <TEXT>
    all text news material.
  </TEXT>
</DOC>
```
...as above ...

If there ever is any material outside <DOC> tags, this can be ignored. Additionally, text information outside of <TEXT> tags can also be disregarded. The class information can identified from within the corresponding tags in the training material.

Source Type <SOURCE_TYPE>
Broad Topic <BROAD_TOPIC>

Requirements:
Note: same language restrictions hold as in HW 1.

Functionality (25pts)
- Does the feature extractor compile?
- Does the feature extractor produce well-formed arff files?
- Did you include trained model files for each classification task?
- Submit any supporting scripts.
- How much do they limit the required human interaction?

Results (25pts)
- How well does the submission classify the supplied training data?
  - Document Type (Broadcast News or Newswire)
  - Major Topic
- How well does the submission classify the unseen testing data?

Write-up (25pts)
- Include the cross-validation accuracy for each experiment
- Which classifiers did you use on each of the tasks? Why?
- Which were the fastest? Most accurate? Easiest to use?
- Which do you prefer and why?
- Which classification task was the easiest/hardest? Why?
- Within tasks, were some classes easier to classify than others? (NB: Examine the weka output for this information.) Why?
- Which classifications were the most similar/most different? Why?
- What features did you use? Why?
- Did they perform better or worse than expected?
- Did early experiments guide your thinking for your final submission? How?
- Which features were the most/least useful? Why?
- If you used any external resources, which did you use? How did they contribute to the success of your submission?

Documentation (15pts)
• README file: This must include the following.
  o How to compile the feature extractor (if necessary)
  o How to run the feature extractor
  o What features are extracted? How, in broad strokes?
  o Which submitted model corresponds to which classification task?
  o Which machine learning algorithm (and parameters) generated the model?
  o Any particular successes or limitations of the submission should be highlighted here.

Within-Code Documentation
• Every method should be documented.

Coding Practices (10pts)
• Informative method/variable names
• Efficient implementation
• Programmer, Memory, and Processor efficiency – don't sacrifice one unless another is improved

Extra Credit may be awarded for particularly inventive or successful approaches to the assignment.

Academic Integrity:

Copying or paraphrasing someone's work (code included), or permitting your own work to be copied or paraphrased, even if only in part, is not allowed, and will result in an automatic grade of 0 for the entire assignment or exam in which the copying or paraphrasing was done. Your grade should reflect your own work. If you believe you are going to have trouble completing an assignment, please talk to the instructor or TA in advance of the due date.
(Appendices on the following pages)