Amazon Reviews for Sentiment Analysis

Data Preparation

Data is downloaded from https://www.kaggle.com/bittlingmayer/amazonreviews

It contains 2 files i.e. test and train in txt format.

Content:

Every Review is in format of:

" label x summaryText: Text"

Where x can be 1 or 2, 1 corresponds to negative and 2 corresponds to positive review.

__label__2 Great CD: My lovely Pat has one of the GREAT voices of her generation. I have listened to this CD fo other video game soundtracks. I must admit that one of the songs (Life-A Distant Promise) has brought tears to as hesitant due to unfavorable reviews and size of machines. I am weaning off my VHS collection, but don't want really did like this one... before it stopped working. The dvd player gave me problems probably after a year of school, and do not know all the requirements of admission, then this book may be a tremendous help. If you have label__1 A complete Bust: This game requires quicktime 5.0 to work...if you have a better version of quicktime s actualy came in the relaxer kit. I tried it and could not beleive the difference it made with one use. I coul ill. I have sent numerous emails to the company - I can't actually find a phone number on their website - and I sto do! My four year old daughter is in love with the many tasks to complete in this game, including dressing money on something you may not be able to use. label 2 one of the last in the series to collect! The magazi

So firstly we have create schema using case class.

case class AmzView(label : Int, view : String)

Then, we save paths of train and test data in String format.

val train_path = "/home/harsh/Desktop/amazon views/1305_800230_compressed_train.ft.txt.bz2/train.ft.txt"

val test_path = "/home/harsh/Desktop/amazon views/1305_800230_compressed_test.ft.txt.bz2/test.ft.txt"

```
Then, we create a method that load the data along with pre-processing
(remove links, emojis, special characters, unnecessary spaces,
converting text into vectors)
def pre_process(path:String) = {
val data = sc.textFile(path).map(attributes => AmzView(attributes(9),
attributes.substring(11,
attributes.length()).replace("\"","").toLowerCase()
    .replaceAll("\n", "")
    .replaceAll("rt\\s+", "")
    .replaceAll("\s+@\w+", "")
    .replaceAll("@\\w+", "")
    .replaceAll("\s+\#\w+", "")
    .replaceAll("#\\w+", "")
    .replaceAll("(?:https?|http?)://[\\w/%.-]+", "")
    .replaceAll("(?:https?|http?)://[\w/\%.-]+\s+", "")
    .replaceAll("(?:https?|http?)//[\\w/%.-]+\\s+", "")
    .replaceAll("(?:https?|http?)//[\\w/%.-]+", "")
    .replaceAll("[^\u0000-\uFFFF]","")
    .replaceAll("(\u00a9|\u00ae|[\u2000-\u3300]|\ud83c[\ud000-\u000ae]]
\dotudfff]\dotud000-\dotudfff]\dotud000-\dotudfff]\dotud000-\dotudfff]
   .trim()
 )).toDF()
```

 $def lo(d:Int) : Int = \{if(d=49)\{2\} else\{1\}\}\$

```
val lco = udf(lo _)
val p = data.select( (lco($"label")).alias("label"), $"view")
val tokenizer = new
Tokenizer().setInputCol("view").setOutputCol("words")
val wordsData = tokenizer.transform(p)
val hashingTF = new HashingTF()
. setInputCol("words"). setOutputCol("rawFeatures"). setNumFeatures("vords") \\
10000)
val featurizedData = hashingTF.transform(wordsData)
val idf = new
IDF().setInputCol("rawFeatures").setOutputCol("features")\\
val idfModel = idf.fit(featurizedData)
val rescaledData = idfModel.transform(featurizedData)
rescaledData
}
Loading data using data path and above method:
val training = pre process(train path)
```

```
val test = pre_process(test_path)
```

Model Selection

We tried Logistic Regression for classification.

val Ir = new LogisticRegression()

.setMaxIter(10)

.setRegParam(0.01)

.setLabelCol("label")

.setElasticNetParam(0.5)

val model = Ir.fit(training)

We evaluated model using MulticlassClassification Evaluator

val evaluator = new

MulticlassClassificationEvaluator()

.setLabelCol("label")

.setPredictionCol("prediction")

.setMetricName("accuracy")

val predict_train = model.transform(training)

val training_accuracy =
evaluator.evaluate(predict_train)

val predict_test = model.transform(test)

```
val test_accuracy =
evaluator.evaluate(predict_test)
```

We got an accuracy of 84% using above model.

Model Tuning

We got a slight increase in accuracy (87%) when we used nlp stemming.

val document = new DocumentAssembler()

.setInputCol("view")

.setOutputCol("document")

val d1 = document.transform(training)

val token = new Tokenizer()

.setInputCols("document")

.setOutputCol("token")

val t1 = token.fit(d1).transform(d1)

val normalizer = new Normalizer()

.setInputCols("token")

```
.setOutputCol("normal")
```

val n1 = normalizer.fit(t1).transform(t1)

val stemmer = new Stemmer()

.setInputCols("normal")

.setOutputCol("stem")

val s1 = stemmer.transform(n1)

val finisher = new Finisher()

.setInputCols("stem")

.setOutputCols("final")

val f1 = finisher.transform(s1)

val hashingTF = new HashingTF()

.setInputCol("filtered").setOutputCol("rawFeatures").setNumFe atures(10000)

val featurizedData = hashingTF.transform(f1)

```
val idf = new
IDF().setInputCol("rawFeatures").setOutputCol("features")
```

val idfModel = idf.fit(featurizedData)

val rescaledData = idfModel.transform(featurizedData)

val d2 = document.transform(testing)

val t2 = token.fit(d2).transform(d2)

val n2 = normalizer.fit(t2).transform(t2)

val s2 = stemmer.transform(n2)

val f2 = finisher.transform(s2)

val featurizedData2 = hashingTF.transform(f2)

val rescaledData2 = idfModel.transform(featurizedData2)

Conclusion

So, using nlp stemming and logistic Regression, we got accuracy at maximum.