First GOP Debate Twitter Sentiment

Data Preparation

Data is downloaded from

https://www.kaggle.com/crowdflower/first-gop-debate-twitter-sentiment

It contains 14k tweets.

It contains the following fields:

- 1. id
- 2. candidate
- 3. candidate_confidence
- 4. relevant_yn
- 5. relevant_yn_confidence
- 6. sentiment
- 7. sentiment_confidence
- 8. subject_matter
- 9. subject_matter_confidence
- 10. candidate_gold
- 11. name
- 12. relevant_yn_gold
- 13. retweet_count
- 14. sentiment_gold
- 15. subject_matter_gold
- 16. text

- 17. tweet coord
- 18. tweet created
- 19. tweet_id
- 20. tweet_location
- 21. user_timezone

| id | | randidate i | candidate | relevant | relevant | sentimen | sentimen | subject n | subject m | candidate | name | relevant | retweet | csentiment | subject i | rtext | tweet co | tweet cre | tweet id | tweet In | cuser t |
|----|---|-------------|------------|----------|----------|----------|----------|-----------|-----------|------------|-----------|-----------|----------|------------|-----------|-----------|-------------|------------|----------|----------|---------|
| 10 | | Janaraate | carraraate | relevant | relevant | Jentimen | | | | carraraate | Harric | relevant_ | retweet_ | Cochimen | Jubject_i | | _ | | _ | tweet_lo | cusci_ |
| | 1 | No candid | 1 | yes | 1 | Neutral | 0.6578 | None of t | 1 | | I_Am_Ker | ızi | 5 | | | RT @Nand | yLeeGrahi | ######## | 6.3E+17 | | Quito |
| | 2 | Scott Wall | 1 | yes | 1 | Positive | 0.6333 | None of t | 1 | | Peaceful | Quest | 26 | i | | RT @Scott | tWalker: D | ######## | 6.3E+17 | | |
| | 3 | No candid | 1 | yes | 1 | Neutral | 0.6629 | None of t | 0.6629 | | PussssyCr | oook | 27 | , | | RT @TJMS | Show: No n | ######## | 6.3E+17 | | |
| | 4 | No candid | 1 | yes | 1 | Positive | 1 | None of t | 0.7039 | | MattFrom | Texas31 | 138 | | | RT @Rob0 | George: Th | ######## | 6.3E+17 | Texas | Centra |
| | 5 | Donald Tr | 1 | yes | 1 | Positive | 0.7045 | None of t | 1 | | sharonDa | y5 | 156 | j | | RT @Dans | Scavino: #G | ######### | 6.3E+17 | | Arizon |
| | 6 | Ted Cruz | 0.6332 | yes | 1 | Positive | 0.6332 | None of t | 1 | | DRJohnso | n11 | 228 | | | RT @Greg | Abbott_TX | ********** | 6.3E+17 | | Centra |

So firstly we have create schema:

val schema = StructType(Array(StructField("id", IntegerType, true),

StructField("candidate", StringType, true),

StructField("candidate_confidence", DoubleType, true),

StructField("relevant_yn", StringType, true),

StructField("relevant_yn_confidence", DoubleType, true),

StructField("sentiment", StringType, true),

StructField("sentiment_confidence", DoubleType, true),

StructField("subject_matter", StringType, true),

StructField("subject_matter_confidence", DoubleType, true),

StructField("candidate_gold", StringType, true),

StructField("name", StringType, true),

StructField("relevant_yn_gold", StringType, true),

StructField("retweet_count", DoubleType, true),

StructField("sentiment_gold", StringType, true),

```
StructField("subject matter gold", StringType, true),
                  StructField("text", StringType, true),
                  StructField("tweet_coord", StringType, true),
                  StructField("tweet_created", StringType, true),
                  StructField("tweet_id", StringType, true),
                  StructField("tweet_location", StringType, true),
                  StructField("user_timezone", StringType, true)
                     ))
Now, we load the data in a dataframe using above schema:
val df = spark.read.
            format("csv")
            .schema(schema)
            .option("header","true")
            .load("/home/harsh/Desktop/gop debate
twitter/Sentiment.csv")
Selecting only those columns which are useful:
val df1 = df.select("candidate", "candidate_confidence",
"relevant yn", "relevant yn confidence", "sentiment",
"sentiment_confidence", "subject_matter",
"subject_matter_confidence", "retweet_count", "text", "tweet_coord")
```

Combining text (String) columns into a single column:

```
val o = df1.select( concat($"candidate", lit(" "), $"subject_matter",
lit(" "), $"text" ).alias("text"), $"candidate_confidence",
$"relevant_yn", $"relevant_yn_confidence", $"sentiment",
$"sentiment_confidence", $"subject_matter_confidence",
$"retweet_count" )
```

Removing records having null values:

```
val p = o.na.drop
```

Column "relevant_yn" consist of two string values either yes or no. To convert its values in numerical values, we create a udf:

```
def lo(i:String) : Double = { if(i.equals("no")){0} else{1} }
val rel = udf(lo _)
```

Column "sentiment" consist of three string values Negative, Nuetral, Positive. To convert its values in numerical values, we create a udf:

```
val sent = udf(pr _)
```

For pre-processing (removing website links, emojis, special characters, unnecessary spaces) we create a udf:

```
def prep(d:String) :String = { d.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()
}
val preProcess = udf(prep _)
Applying all udfs on columns:
val data = p.select(preProcess($"text" ).alias("text"),
$"candidate_confidence", rel($"relevant_yn").alias("relevant_yn"),
$"relevant yn confidence", sent($"sentiment").alias("label"),
$"sentiment confidence", $"subject matter confidence",
$"retweet count")
```

Then to use nlp stemmer, we need to text have to go through document, then through token, then through normalizer, then through stemmer, then finally through finisher:

```
val document = new DocumentAssembler()
  .setInputCol("text")
  .setOutputCol("document")
val d1 = document.transform(data)
val token = new com.johnsnowlabs.nlp.annotator.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)
After loading, we need to convert tweets into feature vectors.
val hashingTF = new HashingTF()
.setInputCol("final").setOutputCol("rawFeatures").setNumFeatu
res(10000)
val featurizedData = hashingTF.transform(f1)
val idf = new
IDF().setInputCol("rawFeatures").setOutputCol("features")
val idfModel = idf.fit(featurizedData)
```

val rescaledData = idfModel.transform(featurizedData)

Then we need to assemble all numeric columns and categorical feature vectors to form final Features vector:

val assembler = new VectorAssembler()

.setInputCols(Array("candidate_confidence", "relevant_yn",
"sentiment_confidence", "subject_matter_confidence", "retweet_
count", "features"))

.setOutputCol("finalFeatures")

val output = assembler.transform(rescaledData)

Extracting label and Final Features (only they are needed to proceed further)

val limited =
output.select(\$"label",\$"finalFeatures".alias("features"))

Then we split the transformed data into two subsets i.e. training and test(ratio 0.8:0.2)

val Array(training, test) =
limited.randomSplit(Array[Double](0.8,0.2))

Model Selection and Model Tuning

We tried Logistic Regression for classification.

val lr = new

LogisticRegression().setMaxIter(10).setRegParam(0.01).setLab elCol("label").setElasticNetParam(0.5)

val model = lr.fit(training)

```
val preTr = model.transform(training)
val preTs = model.transform(test)
```

Conclusion

We evaluated accuracy for model using MultiClassClassification Evaluator and got 75 % accuracy for training and 68 % for testing.

val evaluator = new MulticlassClassificationEvaluator()

.setLabelCol("label")

.setPredictionCol("prediction")

.setMetricName("accuracy")

val train_accuracy = evaluator.evaluate(preTr)
val test_accuracy = evaluator.evaluate(preTs)