Report

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Q1.

Our application can collect the tweets according to the keyword specified (here the keyword is "food"). Then class "TwitterSentimentalAnalysis" will collecte the tweets (count: 20) and class "SentimentAnalyzer" will do sentimental analysis for the collected 20 tweets. The analysis result will be printed out in the log.

Screenshots:

Analysis result

```
Adding annotator split
Adding annotator parse
Adding annotator sentiment
TheetWithSentIment [Line=Dealing with this chest pain, trying to eat fatty food, & watching Gotham with five of my mains., cssClass=sentiment : negative]
Adding annotator tokenize
Adding annotator parse
Adding annotator parse
Adding annotator sentiment
Adding annotator sentiment
Adding annotator sentiment
Adding annotator sentiment
TheetWithSentIment [Line=I'm happy asf because I just received food, cssClass=sentiment : positive]
Adding annotator sentiment
TheetWithSentIment [Line=I'm Food was as brilliant as it was involved, cssClass=sentiment : positive]
Adding annotator sentiment
Adding annotator tokenize
Adding annotator sentiment
Adding annotator sentiment
TheetWithSentIment [Line=Never had food from the strip club. Wonder if it's really good or not. They be advertising it like it is, cssClass=sentiment : negative]
Adding annotator tokenize
Adding annotator tokenize
Adding annotator tokenize
Adding annotator tokenize
Adding annotator parse
Adding annotator parse
Adding annotator parse
Adding annotator tokenize
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```

Step 1: Define 10 categories:

2::animal
3::art
4::book
5::food
6::movie
7::music
8::TV
9::sport
10::travel
11::other

Step 2: Collect tweets as training data to categorize the tweets into these 10 categories using keywords searching.

Step 3: Collect tweets again for recommendation training. In order to do rating, four items should be collected:

- 1) UserId. It's the tweet's userId which should be converted into integer.
- 2) Category. It is analyzed using feature extraction TF-IDF by the training data collected in step 2.
- 3) Rating. Use sentiment analysis to give the rating for each tweet.
- 4) Timestamp. Tweet creation time which should be converted into integer.

These four items should be written into one file called "rating.txt".

- **Step 4:** Get the category mapping file called "category.txt".
- **Step 5:** Get the recommendation for one particular user. The recommendation is the categories that the user prefers.

Step 6: Send the results to the smartphone/smartwatch

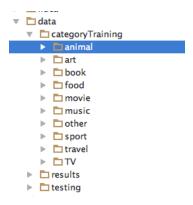
Screenshots:

Example of getting category training data.

```
* Created by Ting on 3/9/16.
late more...(幾F1) GetCategoryTraining {
    public List<Status> GetCategoryData(String keyword) {
        ConfigurationBuilder cb = new ConfigurationBuilder();
        \verb|cb.setDebugEnabled(true).setOAuthConsumerKey("R2v2WMKrF7UGipifRcMkOyjT1")| \\
                .setOAuthConsumerSecret("InkVklJfUsJPQyA17GzGks9uzFSwUnRY9HqsR9m4vZ5Et3sW2d")
                .setOAuthAccessToken("3630687739-9y2qw6YKOMgeApmq09DKOuYosm2piadUy8aa96n")
                .setOAuthAccessTokenSecret("IBjoDz21BTBaXwnJ13jy2A0h0FaYzCYHmNRxCrhLLJong");
        TwitterFactory tf = new TwitterFactory(cb.build());
        Twitter twitter = tf.getInstance();
        Query query;
        query = new Query(keyword +" -filter:retweets -filter:links -filter:replies -filter:images");
        query.setCount(100);
query.setLocale("en");
        query.setLang("en");
        try {
            QueryResult queryResult = twitter.search(query);
            return queryResult.getTweets();
        } catch (TwitterException e) {
            // ignore
            e.printStackTrace();
        return Collections.emptyList();
    }
   public static void main(String[] args) throws IOException {
       GetCategoryTraining getCategoryTraining = new GetCategoryTraining();
       List<Status> statuses = getCategoryTraining.GetCategoryData("food");
       int i = 0;
       for (Status status : statuses) {
           if (status.getText() != null) {
               File foodTextFile = new File("data/categoryTraining/food/" + i + ".txt");
               FileWriter fw = new FileWriter(foodTextFile);
                fw.write(status.getText());
               fw.close();
       }
```

Finally 100 txt files for each category was collected.

File structure:



Category mapping:

```
2::animal
3::art
4::book
5::food
6::movie
7::music
8::TV
9::sport
10::travel
11::other
```

Sentimental analysis and category analysis

Collect training data for recommendation according to above analyzes and write into

file.

```
try
{
    String filename= "data/results/rating.txt";
    FileWriter fw2 = new FileWriter(filename,true); //the true will append the new data
    fw2.write(a);//appends the string to the file
    fw2.close();
}
catch(IOException ioe)
{
    System.err.println("IOException: " + ioe.getMessage());
}
```

Recommendation of user's category preference(partial code):

```
object MakeRecommendation {
 def main(args: Array[String]) {
    System.setProperty("hadoop.home.dir","F:\\winutils")
    Logger.getLogger("org.apache.spark").setLevel(Level.WARN)
    Logger.getLogger("org.eclipse.jetty.server").setLevel(Level.0FF)
    if (args.length != 2) {
      println("Usage: /path/to/spark/bin/spark-submit --driver-memory 2g --class MovieLensALS " +
        "target/scala-*/movielens-als-ssembly-*.jar movieLensHomeDir personalRatingsFile")
      sys.exit(1)
    // set up environment
    val conf = new SparkConf()
      .setAppName("CategoryALS")
      .set("spark.executor.memory", "2g").setMaster("local[*]")
    val sc = new SparkContext(conf)
    // load personal ratings
    val myRatings = loadRatings(args(1))
    val myRatingsRDD = sc.parallelize(myRatings, 1)
    // load ratings and category titles
    val categoryHomeDir = args(∅)
 vac racings - screekerice(new rice(categorynomedir, racing.ext /.toscring/.map \ cine -/
  val fields = line.split("::")
   // format: (timestamp % 10, Rating(userId, categoryId, rating))
   (fields(3).toLong % 10, Rating(fields(0).toInt, fields(1).toInt, fields(2).toDouble))
 println(ratings)
 val categories = sc.textFile(new File(categoryHomeDir, "category.txt").toString).map { line =>
  val fields = line.split("::")
   // format: (categoryId, categoryName)
   (fields(0).toInt, fields(1))
 }.collect().toMap
 val numRatings = ratings.count()
 val numUsers = ratings.map(_._2.user).distinct().count()
 val numCategories = ratings.map(_._2.product).distinct().count()
 println("Got " + numRatings + " ratings from "
  + numUsers + " users on " + numCategories + " categories.")
```

Result log

```
16/03/10 00:58:07 INFO JniLoader: already loaded netlib-native_system-osx-x86_64.jnilib

RMSE (validation) = 0.16005871649500103 for the model trained with rank = 8, lambda = 0.1, and numIter = 10.

RMSE (validation) = 0.18066115598146681 for the model trained with rank = 8, lambda = 0.1, and numIter = 20.

RMSE (validation) = 3.692744729379982 for the model trained with rank = 8, lambda = 10.0, and numIter = 10.

RMSE (validation) = 3.692744729379982 for the model trained with rank = 12, lambda = 10.0, and numIter = 20.

RMSE (validation) = 0.15739262495617246 for the model trained with rank = 12, lambda = 0.1, and numIter = 10.

RMSE (validation) = 3.692744729379982 for the model trained with rank = 12, lambda = 0.1, and numIter = 20.

RMSE (validation) = 3.692744729379982 for the model trained with rank = 12, lambda = 10.0, and numIter = 10.

RMSE (validation) = 3.692744729379982 for the model trained with rank = 12, lambda = 10.0, and numIter = 20.

The best model was trained with rank = 12 and lambda = 0.1, and numIter = 10, and its RMSE on the test set is 1.8075950816211317.

The best model improves the baseline by -61.31%.

Categories recommended for you:

1: art

2: travel

3: TV

Process finished with exit code 0
```

Results sent to smartphone:

