Project 内容

1: 用 MapReduce 算法实现贝叶斯分类器的训练过程并输出训练模型

1.1: 将本地文件读入 hdfs 文件系统

将本地文件系统上的 data 文件夹递归复制到 hdfs 文件系统, 方法的参数分别为 本地路径"/home/hadoop/workspace/data/NBCorpus/Industry",与 hdfs 文件系 统下指定的路径 usr/hadoop/data。

Class Path: tools/InputFromLocal.java

```
public static void inputFromLocal(String src,String dst) throws Exception{
      //fs.mkdirs(new Path(dst));
     File srcPath=new File(src);
     File []srcFiles=srcPath.listFiles();
     if(srcFiles!=null){
          for(File file:srcFiles){
              if(file.isDirectory()){
                  inputFromLocal(file.getAbsolutePath(),dst+"/"+file.getName());
                  String inputFile=src+"/"+file.getName();
                  String dstFile=Global Env.hdfsPath+"/"+dst+"/"+file.getName();
                  //System.out.println(dstFile);
                  InputStream in=new BufferedInputStream(new FileInputStream(inputFile));
                  OutputStream out=fs.create(new Path(dstFile));
                  IOUtils.copyBytes(in, out, 1);
             }
         }
     }
 }
```

导入结果如下:

▼ ■ DFS Locations

▼ MapReduce Location

```
▼ (1)
Country (169)
 ▶ (217)
```

1.2: 从文件中分出测试集与训练集

Industry 目录下的文件总数为 4999, 随机从数据集中随机抽取 10%作为测试文 件,90%作为训练文件

Class Path: tools/DataInitialize. java

public static void dataInitial(String src) throws Exception{ //fs.mkdirs(new Path(dst)); File srcPath=new File(src); //System.out.println(src); File []srcFiles=srcPath.listFiles(); if(srcFiles!=null){ for(File file:srcFiles){ if(file.isDirectory()){ dataInitial(file.getAbsolutePath()); } }else{ String inputFile=src+"/"+file.getName(); String dstFile; if(isTestFile()){ dstFile=Global Env.dataTextPath+"/"+file.getParentFile().getName()+"/"+file.getName(); System.out.println("choosen test file"+dstFile); dstFile=Global Env.dataTrainPath+"/"+file.getParentFile().getName()+"/"+file.getName(); InputStream in=new BufferedInputStream(new FileInputStream(inputFile)); OutputStream out=fs.create(new Path(dstFile)); IOUtils.copyBytes(in, out, 1); } } } 其中选择方法使用随机数实现 public static boolean isTestFile(){ Random random=new Random(); int randomNumber=random.nextInt(5002); if(randomNumber<=500){</pre> return true; return false; } 结果如下: [main] DEBUG org.apache.hadoop.ipc.Client - getting client out of cache: org.apache.hadoop.ipc.Client@5 [main] DEBUG org.apache.hadoop.util.PerformanceAdvisory - Both short-circuit local reads and UNIX domai [main] DEBUG org.apache.hadoop.hdfs.protocol.datatransfer.sasl.DataTransferSaslUtil - DataTransferProtocol.datatransfer.sasl.DataTransferSaslUtil - DataTransferSaslUtil - DataTran

1.3: 实现 SequenceFile

the number of test file 527

finished!!

由于 Industry 目录是由许多小文件组成,为了加快处理速度先将小文件合成为一个 SequenceFile。

首先,定义 WholeFileRecodReader,读取整个文件的值作为一个 record 的值。 其主要方法 nextKeyValue 实现如下:

Class Path: tools/WholeFileRecodReader.java

其次,将多个文件打包成一个 Sequence File, 其中 Sequence File Mapper 实现如下。

map 函数中将 key 设置为父路径名+文件名,即类名+文件名。

Class Path: tools/SmallFilesToSequenceFileConverter.java

获取的 SequenceFile 可通过命令行或代码读取, 这里实现 read 函数读取输出的 SequenceFile 如下:

```
Console 23
<terminated > CountDocNum (2) [Java Application] /usr/lib/jvm/java-8-openjdk-amd64/bin/java (Dec 18, 2019 6:50:06 PM)
113080/48595AnewsML. txt 61 6 6 6 5 6 6 6 7 0 d 0 a 71 75 65 73 74 0 d 0 a 62 75 79 0 d 0 a 74 72 69 6e 69 64 61 64 0 d 0 a 66 65 74 68 61 6e 6f 6c 0 d 0 a 71 13080/48595AnewsML. txt 61 6c 65 65 72 74 61 0 d 0 a 65 6e 67 22 67 79 0 d 0 a 73 61 69 64 0 d 0 a 74 68 75 72 73 64 61 79 0 d 0 a 73 58 0 d 0 a 65 72 73 64 61 79 0 d 0 a 73 58 75 0 d 0 a 70 65 72 69 66 69 67 0 d 0 a 73 61 69 64 0 d 0 a 74 68 75 72 73 64 61 79 0 d 0 a 73 58 0 d 0 a 76 57 26 35 65 67 4 40 0 a 65 66 65 63 64 74 68 75 67 65 72 73 64 61 79 0 d 0 a 73 61 69 64 0 d 0 a 74 68 75 72 73 64 61 79 0 d 0 a 73 58 0 d 0 a 76 57 27 63 65 67 13 0 d 0 a 67 65 72 65 66 74 4 0 d 0 a 65 66 65 63 34 72 69 69 63 0 d 0 a 63 66 67 37 56 66 57 27 30 d 0 a 67 72 65 75 65 65 74 4 0 d 0 a 65 66 65 63 34 72 69 69 63 0 d 0 a 63 66 67 37 55 66 65 72 33 0 d 0 a 67 65 72 65 65 64 0 d 0 a 75 67 65 72 73 64 61 79 0 d 0 a 73 61 0 69 64 0 d 0 a 74 68 75 72 73 64 61 79 0 d 0 a 73 64 61 79 0 d 0 a 73 61 69 64 0 d 0 a 74 68 75 72 73 64 61 79 0 d 0 a 73 64 61 79 0 d 0 a 73 61 69 64 0 d 0 a 74 68 75 72 73 64 61 79 0 d 0 a 73 64 61 79 0 d 0 a 73 61 69 64 0 d 0 a 74 68 75 72 73 64 61 79 0 d 0 a 73 64 61 79 0 d 0
```

1.4: 训练先验概率

首先,定义 DocTypeInputFormat.java 与 DocTypeRecordReader.java, 在DocTypeInputFormat 的 createRecordReader 函数中生成 DocTypeRecordReader 的实例, 代码如下:

在 DocTypeRecordReader 对于每个文件,将<文件所在的目录类别,1>作为输出, 具体实现在 DocTypeRecordReader 的 nextKeyValue()方法中。

由于在 Sequence File 将 key 设置为了父路径名+文件名,这里直接获取父路径即可。

最后在 CountDocType. java 中实现 Map 与 Reduce

在内部类CountTypeNumMapper. java的map方法中直接将<key, value>原样输出。

在内部类 CountTypeNumReduce. java 的 Reduce 方法中对 map 中的 value 进行求和:

最终, 作业输出文件中每一行的格式为类名+文档总数。输出如下:

```
hdfs://localhost:9000/user/hadoop/DocTypeNumOutput/part-r-00000 🛭
 T01000
 I01001
         180
 I01002
         45
 102000
         20
 I03000
         5
 I11000
         16
 I13000
         325
 T14000
         65
 T15000
```

1.5: 训练条件概率

首先定义 TermCNumInputFromat. java 和 TermCseqRecordReader. java

在 TermCNumInputFromat. java 中生成 TermCsegRecordReader 的实例如下:

在 TermCseqRecordReader 中,将<文件父路径名,文件二进制流>作为输出。 TermCseqRecordReader 的 nextKeyValue()函数实现如下:

```
@Override
public boolean nextKeyValue() throws IOException, InterruptedException {
    // TODO Auto-generated method stub
    if(!processed) {
        Writable key s = (Writable) ReflectionUtils.newInstance(reader.getKeyClass(), conf);
        Writable value s = (Writable) ReflectionUtils.newInstance(reader.getValueClass(), conf);

    boolean unFinished=reader.next(key s, value s);
    if(key s.toString().trim().equals("")) {
        processed=true;
        return true;
    }
    key.set(new Text((new Path(key s.toString())).getParent().toString()));

    value=((BytesWritable)value s);
    if(!unFinished) {
        processed=true;
    }
}
return !processed;
}
```

最后在 CountTermC.java 中实现 CountTermcSeqNumMapper.java 与CountTermCNumReduce.java。

在 CountTermcSeqNumMapper. java 中将读取 value 中的文件二进制流,将<类名-单词, IntWritable(1)>作为输出。

map 函数代码如下:

writeKVInLine()函数读取每一行,在上下文中写入 key-value 对: <<classNmae,word>,count>

```
private void writeKVInLine(Mapper<Object,Object,Text,IntWritable>.Context context) throws IOException, Interru
// TODO Auto-generated method stub
byte[] filebytes=value l.getBytes();
InputStream in = new ByteArrayInputStream(filebytes);
BufferedReader brIn=new BufferedReader(new InputStreamReader(in));

String key word=null;

while((key word=brIn.readLine())!=null){
    if(key word-toString().trim().equals(""))
        continue;
    Text key w=new Text(key l.toString()+"\t"+new String(key word));
    context.write(key w, new IntWritable(1));
}
```

在 CountTermCNumReduce. java 中将相同 className-word 的 key 对应的 value 累加, 计算出每个单词在每一类中的出现频率, reduce 方法代码如下:

最终, 输出格式为〈类名 单词名 单词频数〉, 如下图:

hdfs://localhost:9000/user/hadoop/TermCNumOutput/part-r-00000 🛭

```
I01000
        0.6 1
I01000
        1
             2
        1,000
I01000
                 1
        1,250
I01000
                 1
        1,500
I01000
I01000
        1.1-million 1
        1.7 1
I01000
        10
I01000
I01000
        10.4
                 1
        100 2
I01000
        100,000 1
I01000
```

至此,条件概率训练完毕。

2: 用输出的模型对测试集文档进行分类测试。

2.1: 计算每个单词属于文档 C 的概率: $\log \hat{P}(t_k|c)$

在 ConditionalProbility. java 中实现 getT_C()方法,其返回值为 HashMap<HashMap<String, String>, Double>,对应着<<类名,单词>,概率>。 其代码如下:

```
//get the prior Conditional Probility of docC--term
//HaspMap<<类名,单词>,概率>
public static HashMap<HashMap<String,String>,Double> getT_C(){
    int TOATAL TERM=0;
   HashMap<HashMap<String,String>,Double> hashMap=new HashMap();
   HashMap<String,Integer> total_count=new HashMap();
   String path="hdfs://localhost:9000/user/hadoop/TermCNumOutput/part-r-00000";
   Configuration conf=new Configuration();
        FileSystem fs=FileSystem.get(URI.create(path),conf);
        InputStream in=null;
        in=fs.open(new Path(path));
       BufferedReader br=new BufferedReader(new InputStreamReader(in));
       String line=null;
       String curDocType=null;
        //totalWordOfC record the word number of the current classC
        int totalWordOfC=0:
       while((line=br.readLine())!=null){
            totalWordOfC++;
            String[] split=line.split("\\s+");
            String docType=split[0];
            String term=split[1];
            Double value=Double.parseDouble((split[2]))+1;
           HashMap<String,String> key=new HashMap();
            key.put(docType, term);
            hashMap.put(key,value);
      if(!curDocType.equals(docType)){
           System.out.println("curDocType:"+curDocType+" doctype "+docType+" "+totalWordOfC);
           total count.put(curDocType, totalWordOfC+Global Env.TOATAL TERM);
           TOATAL TERM+=totalWordOfC;
           curDocType=docType;
           totalWordOfC=0;
  }
       total_count.put(curDocType, totalWordOfC+Global_Env.TOATAL_TERM);
        for(Map.Entry<HashMap<String, String>, Double> entry: hashMap.entrySet())
            Double value=entry.getValue();
            HashMap<String,String> hashKey=entry.getKey();
            for(String key:hashKey.keySet())
                dev=total_count.get(key);
            entry.setValue(value/dev);
          // System.out.println("Key: "+ entry.getKey()+ " Value: "+entry.getValue());
       System.out.println(" TOATAL TERM"+TOATAL TERM);
   } catch (IOException e) {
       // TODO Auto-generated catch block
       e.printStackTrace();
   return hashMap;
```

2.2: 计算文档 C 的先验概率: $\log \hat{P}(c)$

在 priorProbability. java 中实现 getPriorPrb()方法,该方法返回一个 HashMap<String, Double>,对应<类名,该类的先验概率>,其代码如下:

```
//HashMap<类名,该类的先验概率>
 public static HashMap<String,Double> getPriorPrb(){
     int count=0:
     HashMap<String,Double> hashmap=new HashMap();
     String path="hdfs://localhost:9000/user/hadoop/DocTypeNumOutput/part-r-00000";
     Configuration conf=new Configuration();
     try {
         FileSystem fs=FileSystem.get(URI.create(path),conf);
         InputStream in=null;
         in=fs.open(new Path(path));
         BufferedReader br=new BufferedReader(new InputStreamReader(in));
         String line=null:
         while((line=br.readLine())!=null){
             String[] split=line.split("\\s+");
             String key=split[0];
             Double value=Math.log(Double.parseDouble(split[1])/Global Env.TOTAL DOC);
             hashmap.put(key, value);
             count++;
         }
     } catch (IOException e) {
         // TODO Auto-generated catch block
         e.printStackTrace();
     return hashmap;
```

2.3: 实现 conditionalProbabilityFroClass()函数

在 Prediction. java 中实现 conditional Probability Fro Class () 函数。 该函数的输入参数为一个文件的二进制文件流与类名, 返回该文件属于该类的概率 P(class | doc)。

```
//Parameter 1--文件的二进制文件流
//parameter 2--测试类名
//return value--文件属于该类的概率
static double conditionalProbabilityFroClass(BytesWritable content,String className){
    double conditionalProbability=priorProbilityMap.get(className);
    byte[] filBytes=content.getBytes();
    InputStream in=new ByteArrayInputStream(filBytes):
    BufferedReader br=new BufferedReader(new InputStreamReader(in));
    String term=null;
       while((term=br.readLine())!=null){
            if(term.toString().trim().equals("")){
               continue:
            HashMap<String,String> key=new HashMap();
            key.put(className, term);
            if(hashMap.get(kev)!=null)
            conditionalProbability+=Math.log(hashMap.get(key));
                conditionalProbability+=Math.log(1.0/Global Env.TOATAL TERM);
       }
    } catch (IOException e) {
        // TODO Auto-generated catch block
        e.printStackTrace();
    return conditionalProbability;
}
```

2.4: 实现 PredictionMapper. class

PredictionMapper 的输入为<文件名,文件二进制流>,输出为<文件名,〈类名,文件属于该类的概率〉>,对于文档类集合中的每一个类,计算文档属于该类的概率,其实现如下。

```
public static class PredictionMapper extends Mapper<Object,Object,Text,Text>{
      private Text key l=new Text();
      private BytesWritable value l=null;
      //InputFormat <文件名,文件二进制流>
      //OutputFormat<文件名,类名,文件属于该类的概率>
      @Override
      public void map(Object docName, Object docContent,
              Mapper<Object,Object,Text,Text>.Context context)
              throws IOException, InterruptedException {
          // TODO Auto-generated method stub
          key l=new Text();
          key l.set(docName.toString());
          value l=(BytesWritable)docContent;
          if(key_l.toString().trim()!="")
          //对于文档类集合中的每一个类,计算文档属于该类的概率
          for (int i = 0; i < classC.size(); i++) {</pre>
              double probility=conditionalProbabilityFroClass(value l,classC.get(i));
              Text val=new Text(classC.get(i)+"\t"+probility);
              context.write(key l, val);
      }
  }
```

2.5: 实现 PreductionReduce. class

PreductionReduce 中 reduce 的输入为<文件名, list<类名,文件属于该类的概率>>,对于每个文件,找到最大的 prob,输出<文件名,最大的 prob 对应的类名>。

```
//InputFormat---<文件名,list<类名,文件属于该类的概率>>
 //OutputFormat---<文件名,最大的prob对应的类名>
 @Override
 public void reduce(Text key, Iterable<Text> values,
         Reducer<Text,Text,Text,Text >.Context context) throws IOException, InterruptedException {
     // TODO Auto-generated method stub
     double maxP=Double.NEGATIVE INFINITY;
     String doctype="";
     for(Text value:values){
         String split[]=value.toString().trim().split("\\s+");
         if(split.length<2)</pre>
             continue;
         double currentP=Double.parseDouble(split[1]);
         if(currentP>maxP){
             maxP=currentP:
             doctype=split[0];
 // System.out.println("reduce key:"+key+" value:"+count);
     context.write(key,new Text(doctype));
 }
```

2.6: 测试文档的分类结果如下

hdfs://localhost:90	00/user/hadoop/DocTyp	*ConditionalProbility.java	T.
477891newsML.txt	I21000		
477896newsML.txt	I21000		
477900newsML.txt	I21000		
477902newsML.txt	I21000		
477905newsML.txt	I33020		
477908newsML.txt	I81402		
477909newsML.txt	I81402		
477917newsML.txt	I81402		
477919newsML.txt	I33020		
477926newsML.txt	I81402		
477927newsML.txt	I81402		
477928newsML.txt	I81402		
477929newsML.txt	I33020		
477935newsML.txt	I81402		
477936newsML.txt	175000		
477940newsML.txt	I81402		
477945newsML.txt	I33020		
477946newsML.txt	I81402		
477947newsML.txt	113000		
477959newsML.txt	I81402		
477965newsML.txt	179020		
477974newsML.txt	181402		
477977newsML.txt	181402		
477978newsML.txt	133020		
477979newsML.txt	181402		
477980newsML.txt	133020		
477987newsML.txt	181402		
477988newsML.txt	181402		
477989newsML.txt	133020		
477990newsML.txt	I13000		
478000newsML.txt	133020		
478002newsML.txt	125000		
478004newsML.txt	181402		
478015newsML.txt	113000		
478021newsML.txt	I81402		
478028newsML.txt	I81402		
478029newsML.txt	I13000		
478040newsML.txt	135101		
478046newsML.txt	I81402		
478052newsML.txt	135101		
478054newsML.txt	I33020		
478055newsML.txt	135101		
478057newsML.txt	181402		
478060newsML.txt	181402 181402		
478062newsML.txt	I33020		
478063newsML.txt	133020 133020		
	181402		
478067newsML.txt			
478071newsML.txt	I13000		
478072newsML.txt	I75000		
478073newsML.txt	I81402		
478076newsML.txt	181402		
478078newsML.txt	I16100		
478081newsML.txt	181402		
478085newsML.txt	181402		
478087newsML.txt	I35101		
Problems 🙆 Tasks	@ Javadoc % Type Hier	archy ‡ Call Hierarchy ∺	

3: 利用测试文档的真实类别, 计算分类模型的 Precision, Recall 和 F1 值。

3.1: microPrecision

在 Judge. java 中实现 microPrecision () 计算其 Micoraverage Precision

```
static void microPrecision(){
     double TP=0;
     double TN=0;
     double FP=0;
     double FN=0;
    Iterator<String> iterator = classCSet.iterator();
    while(iterator.hasNext()){
        String c=iterator.next();
        for(Map.Entry<String, String> testedDoc:resMap.entrySet()){
            String docName=testedDoc.getKey();
            String resClass=testedDoc.getValue();
            String oriClass=classOfDoc.get(docName);
            if(c.equals(oriClass)&&c.equals(resClass))
                TP++:
            else if(c.equals(oriClass)&&!c.equals(resClass))
                FN++;
            else if(!c.equals(oriClass)&&c.equals(resClass))
            else if(!c.equals(oriClass)&&!c.equals(resClass))
                TN++;
        }
    double precision=TP/(TP+FP);
    double recall=TP/(TP+FN);
    double f1=2*precision*recall/(precision+recall);
    System.out.println("microPrecision----- "+precision*100+"%");
Micoraverage Precision 输出如图:
   lastLocatedBlock=LocatedBlock{BP-728634802-192.168.5
   isLastBlockComplete=true}
 [main] DEBUG org.apache.hadoop.hdfs.DFSClient - Connec
 [main] DEBUG org.apache.hadoop.hdfs.protocol.datatrans
 microPrecision----- 39.28365106874639%
 [Thread-3] DEBUG org.apache.hadoop.ipc.Client - stoppi
3.1: macroPrecision
在 Judge. java 中实现 macroPrecision() 计算其 Macoraverage Precision
static void macroPrecision(){
    double precision=0;
    for(Map.Entry<String, Double> entry:f1Map.entrySet()){
        precision+=entry.getValue();
    precision/=f1Map.size();
    System.out.println("macroPrecision----- "+precision*100+"%");
}
其中 f1Map 为每个类所对应的 f1 值, 其键值对结构为<类名, f1>, 代码实现如下:
```

```
static void evaluationForC(){
    Iterator<String> iterator = classCSet.iterator();
    while(iterator.hasNext()){
         double TP=0;
         double TN=0;
         double FP=0;
         double FN=0;
         String c=iterator.next();
         for(Map.Entry<String, String> testedDoc:resMap.entrySet()){
            String docName=testedDoc.getKey();
            String resClass=testedDoc.getValue();
            String oriClass=classOfDoc.get(docName);
            if(c.equals(oriClass)&&c.equals(resClass))
                TP++;
            else if(c.equals(oriClass)&&!c.equals(resClass))
                FN++;
            else if(!c.equals(oriClass)&&c.equals(resClass))
                FP++:
            else if(!c.equals(oriClass)&&!c.equals(resClass))
                TN++:
            // System.out.println(c+"docName="+docName+" oriClass="+oriClass+" resClass="-
        double precision=TP/(TP+FP);
        double recall=TP/(TP+FN);
        double f1=2*precision*recall/(precision+recall);
        if(TP!=0)
           f1Map.put(c, f1);
           f1Map.put(c, 0.0);
        System.out.println(c+"---TP="+TP+" TN="+TN+" FP="+FP+" FN="+FN+" precision="+preci
Macoraverage Precision 输出如图:
<terminated>Judge [Java Application] /usr/lib/jvm/java-8-openjdk-amd64/bin/java (Dec 19, 2019 5:41:45 PM)
log4j:WARN No appenders could be found for logger (org.apache.hadoop.metrics2.lib.MutableMetricsFactory).
log4j:WARN Please initialize the log4j system properly.
log4j:WARN See http://logging.apache.org/log4j/1.2/faq.html#noconfig for more info.
f1=0.222222222222222
f1=0.2873563218390805
                                                                                  f1=0.38095238095238093
                                                                                  f1=0.449438202247191
f1=0.72000000000000001
macroPrecision-----
                    53.05630989177219%
以上为 project 的全部内容。
```

二 贝叶斯分类器理论介绍

首先假设已经有分好类的N篇文档: (d1, c1)、(d2, c2)、(d3, c3) ······ (dn, cn)。di 表示第 i 篇文档, ci 表示第 i 个类别。目标是: 寻找一个分类器, 这个分类器能够: 当丢给它一篇新文档 d, 它就输出 d 最有可能属于哪个类别 c。

朴素贝叶斯分类器是一个概率分类器。假设现有的类别 C={c1, c2, ·····cm}。 给定一篇文档 d. 这个问题用数学公式表示如下:

$$\hat{c} = \operatorname*{argmax}_{c \in C} P(c|d)$$

 c^{n} 就是:在所有的类别 $C=\{c1, c2, \dots cm\}$ 中,使得:条件概率 P(c|d) 取最大值的类别。使用贝叶斯公式,将转换成如下形式:

$$\hat{c} = \operatorname*{argmax}_{c \in C} P(c|d) = \operatorname*{argmax}_{c \in C} \frac{P(d|c)P(c)}{P(d)}$$

$$\hat{c} = \operatorname*{argmax}_{c \in C} P(c|d) = \operatorname*{argmax}_{c \in C} P(d|c) P(c)$$

这个公式由两部分组成,前面那部分P(d|c) 称为似然函数,后面那部分P(c) 称为先验概率。

文档 d, 文档 d 的每个特征表示为: d={f1, f2, f3······fn}, 那么这里的特征 fi 其实就是单词 wi 出现的频率(次数),公式转化成如下形式:

$$\hat{c} = \underset{c \in C}{\operatorname{argmax}} \underbrace{P(f_1, f_2,, f_n | c)}^{\text{likelihood}} \underbrace{P(c)}^{\text{prior}}$$

对文档 d 做个假设:假设各个特征之间是相互独立的。那么 p(f1, f2······ fn|c)=p(f1|c)*p(f2|c)*······*p(fn|c),公式转化成如下形式:

$$c_{NB} = \operatorname*{argmax}_{c \in C} P(c) \prod_{f \in F} P(f|c)$$

由于每个概率值很小若干个很小的概率值直接相乘,得到的结果会越来越小。为了避免计算过程出现下溢,引入对数函数 Log,然后使用每个单词 wi 出现频率作为特征,得到如下公式:

$$c_{NB} = \underset{c \in C}{\operatorname{argmax}} \log P(c) + \sum_{i \in positions} \log P(w_i|c)$$

训练朴素贝叶斯的过程其实就是计算先验概率和似然函数的过程。

①先验概率 P(c) 的计算

P(c)的意思是:在所有的文档中,类别为 c 的文档出现的概率有多大?假设训练数据中一共有 Ndoc 篇文档,只要数一下类别 c 的文档有多少个就能计算 p(c)了,类别 c 的文档共有 Nc 篇,先验概率的计算公式如下:

$$\hat{P}(c) = \frac{N_c}{N_{doc}}$$

②似然函数 P(wi | c) 的计算

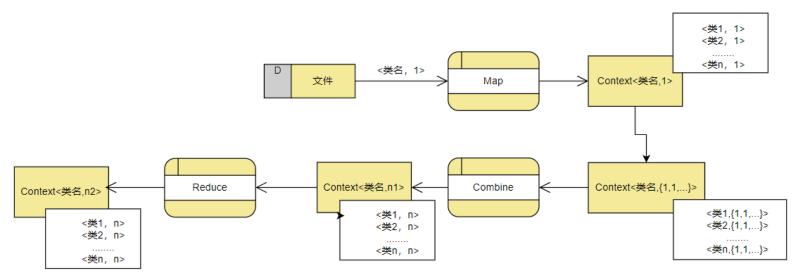
对于文档 d 中的每个单词 wi, 找到训练数据集中所有类别为 c 的文档, 数一数单词 wi 在这些文档 (类别为 c) 中出现的次数: count (wi, c), 然后再数一数训练数据集中类别为 c 的文档一共有多少个单词。计算 二者之间的比值, 就是似然函数的值。似然函数计算公式如下:

$$\hat{P}(w_i|c) = \frac{count(w_i,c)}{\sum_{w \in V} count(w,c)}$$

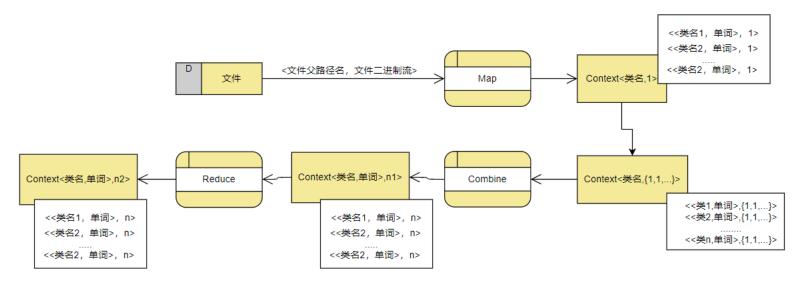
将"出现次数加1"。似然函数公式变成如下形式:

$$\hat{P}(w_i|c) = \frac{count(w_i,c) + 1}{\sum_{w \in V} (count(w,c) + 1)} = \frac{count(w_i,c) + 1}{\left(\sum_{w \in V} count(w,c)\right) + |V|}$$

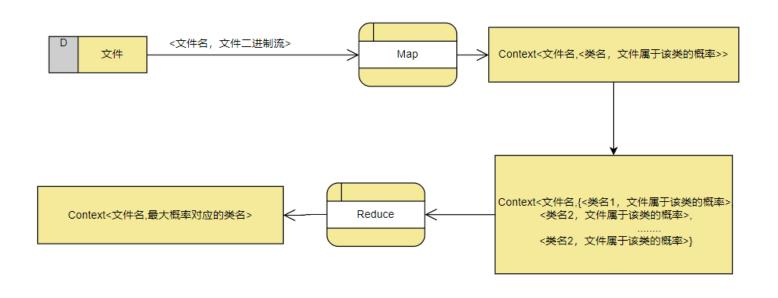
三 贝叶斯分类器训练的 MapReduce 算法设计



计算每个类型中文档数目的 MapReduce job 数据流图上如图所示, RecordReader 输出的键值为<类名, 1>。Map 的输入为<类名, 1>,Map 中不做处理, 直接输出< 类名, 1>。Reduce 的输入为<类名, {文件数 1, 文件数 2, ..., 文件数 n}>, Reduce 中对其进行求和, 输出为<类名, 文件总数>。



计算每个文件中每个单词出现频率的 MapReduce job 的数据流图如上图所示,使用 SequenceFile 作为输入,将每个文件作为一个 record 处理,Record Reader输出的键值为<文件父路径名,文件二进制流>即<文件类名,文件二进制流>,即为 Map 的输入,map 的输出为<<类名,单词>,1>。Reduce 的输入为<<类名,单词>, {count1,count2,...,countn}>,Reduce 对其求和,输出为<<类名,单词>,单词数量>。



预测 Map Reduce job 的数据流图如上图所示,使用 SequenceFile 作为输入,将 每个文件作为一个 record 处理,Record Reader 输出的键值为<文件名,文件二进制流>,即为 Map 的输入,map 的输出为<文件名,<类名,文件属于该类的概率>>。 Reduce 的输入为<文件名,<类名,list{文件属于该类的概率>}>,最终输出为<文件名,最大概率对应的类名>。

四:源代码清单

1: 统计每类文件数目

```
//CountDocType.java
//计算每类文件的数目
public class CountDocType extends Configured implements Tool{
private CountDocType() {}
private static String trainoutputPath="DocTypeNumOutput";
private static String wholeOutputPath="DocTypeWholeOutput";
private static String testDocTypeNumOutput="testDocTypeNumOutput";
public static void main(String[] args) throws Exception {``
   int res=ToolRunner.run(new Configuration(), new CountDocType(),null);
   System.exit(1);
public int run(String[] arg0) throws Exception {
   // TODO Auto-generated method stub
   Configuration conf=getConf();
   Job job=Job.getInstance(conf);
   FileSystem fileSystem=new Path(Global_Env.hdfsPath).getFileSystem(conf);
   try{
       job.setJobName("Count-Doc-Type");
        job.setJarByClass(CountDocType.class);
       FileInputFormat.addInputPath(job, new Path(Global_Env.seqOutputPath));
       new LoadPath(fileSystem, job).addOutPath(outputPath);
        job.setInputFormatClass(DocTypeInputFormat.class);
        job.setMapperClass(CountTypeNumMapper.class);
        job.setCombinerClass(CountTypeNumReduce.class);
       job.setReducerClass(CountTypeNumReduce.class);
       job.setOutputKeyClass(Text.class );
       job.setOutputValueClass(IntWritable.class);
       System.out.println("___start_
       boolean res=job.waitForCompletion(true);
       System.out.println("finished!! res="+res);
   finally{
   return 0;
//InputFormat---<文件所在的目录类别,1>
//OnputFormat---<文件所在的目录类别,1>
//该map函数对key-value不做处理
public static class CountTypeNumMapper extends Mapper<Text,IntWritable,Text,IntWritable>{
   @Override
   public void map(Text key, IntWritable value,
           Mapper<Text, IntWritable, Text, IntWritable>.Context context)
           throws IOException, InterruptedException {
       // TODO Auto-generated method stub
       //System.out.println("map key:"+key+" value:"+value);
       context.write(key, value);
   }
}
```

```
//InputFormat---<文件所在的目录类别,1>
//OnputFormat---<文件所在的目录类别,该类别下文件总数>
public static class CountTypeNumReduce extends Reducer<Text,IntWritable,Text,IntWritable>{
   @Override
   public void reduce(Text key, Iterable<IntWritable> values,
           Reducer<Text,IntWritable,Text,IntWritable >.Context context) throws IOException, InterruptedException
       // TODO Auto-generated method stub
       int count=0;
       for(IntWritable value:values){
           count+=value.get();
   // System.out.println("reduce key:"+key+" value:"+count);
       context.write(key, new IntWritable(count));
   }
}
//读取输出文件,统计文件类型,返回list<文件类型>
public static List readDocType(){
   List<String> list=new ArrayList();
    Configuration conf = new Configuration();
   String path = trainoutputPath+"/part-r-00000";
   FileSystem fs;
   int totalNum=0;
   try {
       fs = FileSystem.get(URI.create(path),conf);
       InputStream in=null;
       in=fs.open(new Path(path));
       BufferedReader br=new BufferedReader(new InputStreamReader(in));
       String line=null;
       while((line=br.readLine())!=null){
           String[] split=line.split("\\s+");
           String docType=split[0];
           String docNum=split[1];
           totalNum+=Integer.parseInt(docNum);
           list.add(docType);
   } catch (IOException e) {
       // TODO Auto-generated catch block
       e.printStackTrace();
   System.out.println("total doc number="+totalNum);
   return list;
```

```
//DocTypeRecordReader.java
public class DocTypeRecordReader extends RecordReader<Text, IntWritable>{
   private IntWritable value=new IntWritable(0);
    private Text key=new Text();
    private FileSplit fileSplit;
    private Configuration conf;
    private boolean processed=false;
    SequenceFile.Reader reader=null;
    @Override
    public Text getCurrentKey() throws IOException, InterruptedException {
       // TODO Auto-generated method stub
        return key;
    @Override
    public IntWritable getCurrentValue() throws IOException,
           InterruptedException {
        // TODO Auto-generated method stub
        return value;
    @Override
    public float getProgress() throws IOException, InterruptedException {
       // TODO Auto-generated method stub
        return processed?1.0f:0.0f;
    @Override
    public void initialize(InputSplit split, TaskAttemptContext context)
           throws IOException, InterruptedException {
        // TODO Auto-generated method stub
       this.fileSplit=(FileSplit)split;
       this.conf=context.getConfiguration();
        this.reader = new SequenceFile.Reader(fileSplit.getPath().getFileSystem(conf), fileSplit.getPath(), conf);
    public boolean nextKeyValue() throws IOException, InterruptedException {
        // TODO Auto-generated method stub
        if(!processed){
           Writable key_s = (Writable) ReflectionUtils.newInstance(reader.getKeyClass(), conf);
           Writable value_s = (Writable) ReflectionUtils.newInstance(reader.getValueClass(), conf);
            boolean unFinished=reader.next(key_s, value_s);
            if(key_s.toString().trim().equals("")){
               processed=true;
               return true;
            //返回文档父路径即类名
            key.set(new Text((new Path(key_s.toString())).getParent().toString()));
            value=new IntWritable(1);
            if(!unFinished){
               processed=true;
        return !processed;
    }
```

2: 统计类中单词频率

```
//统计类中单词频率
//CountTermC.java
public class CountTermC extends Configured implements Tool{
    private static String outputPath="TermCNumOutput";
   public static void main(String[] args) throws Exception {
       \verb|int res=ToolRunner.run(new Configuration(), new CountTermC(), null);\\
        System.exit(1);
    //统计测试集合内所有单词数
    public static int countTermNumber(){
       String path="hdfs://localhost:9000/user/hadoop/"+outputPath+"/part-r-00000";
       int count=0;
       Configuration conf=new Configuration();
        try {
            FileSystem fs=FileSystem.get(URI.create(path),conf);
            InputStream in=null;
            in=fs.open(new Path(path));
            BufferedReader br=new BufferedReader(new InputStreamReader(in));
           String line=null;
            while((line=br.readLine())!=null){
                String[] splited = line.split("\\s+");
                int termNum=Integer.parseInt(splited[2]);
                count+=termNum;
        } catch (IOException e) {
            // TODO Auto-generated catch block
            e.printStackTrace();
        System.out.println("Term num="+count);
        return count;
    public int run(String[] arg0) throws Exception {
        // TODO Auto-generated method stub
        Configuration conf=getConf();
        Job job=Job.getInstance(conf);
        FileSystem fileSystem=new Path(Global_Env.hdfsPath).getFileSystem(conf);
        trv{
            job.setJobName("Count-TermC-Type");
            job.setJarByClass(CountTermC.class);
            FileInputFormat.addInputPath(job, new Path(Global\_Env.seqOutputPath));\\
            new LoadPath(fileSystem, job).addOutPath(outputPath);
            job.setInputFormatClass(TermCNumInputFromat.class);
            job.setMapperClass(CountTermC.CountTermcSeqNumMapper.class);
            job.setCombinerClass(CountTermC.CountTermCNumReduce.class);
            job.setReducerClass(CountTermC.CountTermCNumReduce.class);
            job.setOutputKeyClass(Text.class );
            job.setOutputValueClass(IntWritable.class );
            System.out.println("___start____");
            boolean res=job.waitForCompletion(true);
            System.out.println("finished!! res="+res);
        finally{
        }
        return 0;
```

```
//Mapper
   //输入<文件父路径名,文件二进制流>
   //输出<类名-单词,IntWritable(1)>
   public static class CountTermcSeqNumMapper extends Mapper<Object,Object,Text,IntWritable>{
       private Text key l=new Text();
       private BytesWritable value_l=null;
       //InputFormat <className,binary file stream>
       //OutputFormat<<classNmae,word>,count>
       @Override
       public void map(Object line, Object value,
               Mapper<Object,Object,Text,IntWritable>.Context context)
               throws IOException, InterruptedException {
           // TODO Auto-generated method stub
           key l=new Text();
           key_l.set(line.toString());
           value l=(BytesWritable)value;
           //read each line of file write key-value
           writeKVInLine(context);
       private void writeKVInLine(Mapper<Object,Object,Text,IntWritable>.Context context) throws IOException, InterruptedException {
           // TODO Auto-generated method stub
           byte[] filebytes=value_1.getBytes();
           InputStream in = new ByteArrayInputStream(filebytes);
           {\tt BufferedReader} \ brIn = new \ BufferedReader(new \ InputStreamReader(in)); \\
           String key_word=null;
           while((key_word=brIn.readLine())!=null){
               if(key_word.toString().trim().equals(""))
                   continue;
               Text key_w=new Text(key_1.toString()+"\t"+new String(key_word));
               context.write(key_w, new IntWritable(1));
       }
//Reduce
//输入<<类名,单词>,{count1, count2,...,countn}>
//输出<<类名,单词>,单词数量>
public static class CountTermCNumReduce extends Reducer<Text,IntWritable,Text,IntWritable>{
       @Override
       public void reduce(Text key, Iterable<IntWritable> values,
               Reducer<Text,IntWritable,Text,IntWritable >.Context context) throws IOException, InterruptedException {
           // TODO Auto-generated method stub
           int count=0;
           for(IntWritable value:values){
               count+=value.get();
       // System.out.println("reduce key:"+key+" value:"+count);
           context.write(key, new IntWritable(count));
```

3: 预测文件所对应的最大概率的类

```
//TermCseqRecordReader.java
public \ class \ TermCseqRecordReader \ extends \ RecordReader < Text, BytesWritable > \{ Continuous and Conti
        private FileSplit fileSplit;
        private Configuration conf;
       private Text key=new Text();
        private BytesWritable value=new BytesWritable();
        private boolean processed=false;
        Configuration job;
        SequenceFile.Reader reader=null;
        public \ Text \ getCurrentKey() \ throws \ IOException, \ InterruptedException \ \{
                // TODO Auto-generated method stub
                return key;
        @Override
        public BytesWritable getCurrentValue() throws IOException,
                        InterruptedException {
                // TODO Auto-generated method stub
                return value;
        public float getProgress() throws IOException, InterruptedException {
                // TODO Auto-generated method stub
                return processed==true?1.0f:0.0f;
//TermCsegRecordReader.java
public class TermCseqRecordReader extends RecordReader<Text,BytesWritable>{
        @Override
        public void initialize(InputSplit split, TaskAttemptContext context)
                        throws IOException, InterruptedException {
                // TODO Auto-generated method stub
                this.fileSplit=(FileSplit)split;
                this.conf=context.getConfiguration();
                this.reader = new SequenceFile.Reader(fileSplit.getPath().getFileSystem(conf), fileSplit.getPath(), conf);
        }
        @Override
        public boolean nextKeyValue() throws IOException, InterruptedException {
                // TODO Auto-generated method stub
                if(!processed){
                        Writable key_s = (Writable) ReflectionUtils.newInstance(reader.getKeyClass(), conf);
                        Writable value_s = (Writable) ReflectionUtils.newInstance(reader.getValueClass(), conf);
                        boolean unFinished=reader.next(key_s, value_s);
                        if(key_s.toString().trim().equals("")){
                                processed=true;
                                return true;
                        key of CountTermC key-value <文件类名,文件二进制流>
                        key.set(new Text((new Path(key_s.toString())).getParent().toString()));
                        key of Prodection key-value <文件名,文件二进制流>
                        key.set(new Text((new Path(key_s.toString()).getName()).toString()));
                        value=((BytesWritable)value_s);
                        if(!unFinished){
                                processed=true;
                return !processed;
```

```
//Prediction.java
public class Prediction extends Configured implements Tool{
    static HashMap<HashMap<String>,Double> hashMap=ConditionalProbility.getT_C();
    static HashMap<String,Double> priorProbilityMap=priorProbability.getPriorPrb();
    static List<String> classC=CountDocType.readDocType();
    public static void main(String[] args) {
       try {
            int res=ToolRunner.run(new Configuration(), new Prediction(),null);
        } catch (Exception e) {
            // TODO Auto-generated catch block
            e.printStackTrace();
        }
        System.exit(1);
    }
    //Parameter 1--文件的二进制文件流
    //parameter 2--测试类名
    //return value--文件属于该类的概率
   static double conditionalProbabilityFroClass(BytesWritable content,String className){
        double conditionalProbability=priorProbilityMap.get(className);
        byte[] filBytes=content.getBytes();
        InputStream in=new ByteArrayInputStream(filBytes);
        BufferedReader br=new BufferedReader(new InputStreamReader(in));
       String term=null;
       try {
           while((term=br.readLine())!=null){
               if(term.toString().trim().equals("")){
                   continue;
               }
               HashMap<String,String> key=new HashMap();
               key.put(className, term);
               if(hashMap.get(key)!=null)
               conditionalProbability+=Math.log(hashMap.get(key));
                   conditionalProbability+=Math.log(1.0/Global_Env.TOATAL_TERM);
        } catch (IOException e) {
           // TODO Auto-generated catch block
           e.printStackTrace();
       return conditionalProbability;
   public int run(String[] arg0) throws Exception {
        // TODO Auto-generated method stub
        Configuration conf=getConf();
        Job job=Job.getInstance(conf);
       FileSystem fileSystem=new Path(Global_Env.hdfsPath).getFileSystem(conf);
        job.setJobName("Prediction");
       job.setJarByClass(Prediction.class);
        FileInputFormat.addInputPath(job, new Path(Global_Env.seqTestOutputPath));
        new LoadPath(fileSystem, job).addOutPath(Global_Env.testResultPath);
        job.setInputFormatClass(TermCNumInputFromat.class);
        job.setMapperClass(Prediction.PredictionMapper.class );
        job.setReducerClass(Prediction.PreductionReduce.class);
        job.setOutputKeyClass(Text.class );
        job.setOutputValueClass(Text.class );
        System.out.println("___start____");
        boolean res=job.waitForCompletion(true);
        System.out.println("finished!! res="+res);
        return 0;
```

```
public static class PredictionMapper extends Mapper<Object,Object,Text,Text>{
    private Text key_l=new Text();
    private BytesWritable value_l=null;
    //InputFormat <文件名,文件二进制流>
    //OutputFormat<文件名,<类名,文件属于该类的概率>>
    @Override
    public void map(Object docName, Object docContent,
           Mapper<Object,Object,Text,Text>.Context context)
            throws IOException, InterruptedException {
        // TODO Auto-generated method stub
        key_l=new Text();
        key_l.set(docName.toString());
        value_l=(BytesWritable)docContent;
        if(key_l.toString().trim()!="")
        //对于文档类集合中的每一个类, 计算文档属于该类的概率
        for (int i = 0; i < classC.size(); i++) {
           double probility=conditionalProbabilityFroClass(value_l,classC.get(i));
           Text val=new Text(classC.get(i)+"\t"+probility);
           context.write(key_l, val);
public static class PreductionReduce extends Reducer<Text,Text,Text,Text>{
        //InputFormat---<文件名,list<类名,文件属于该类的概率>>
       //OutputFormat---<文件名,最大的prob对应的类名>
       @Override
       public void reduce(Text key, Iterable<Text> values,
               Reducer<Text,Text,Text,Text >.Context context) throws IOException, InterruptedException {
           // TODO Auto-generated method stub
           double maxP=Double.NEGATIVE_INFINITY;
           String doctype="";
           for(Text value:values){
               String split[]=value.toString().trim().split("\\s+");
               if(split.length<2)
                   continue;
               double currentP=Double.parseDouble(split[1]);
               if(currentP>maxP){
                   maxP=currentP;
                   doctype=split[0];
        // System.out.println("reduce key:"+key+" value:"+count);
           context.write(key,new Text(doctype));
```

```
//ConditionalProbility.java
public class ConditionalProbility {
   public static void main(String[] args) {
       getT_C();
   //get the prior Conditional Probility of docC--term
    //HaspMap<<类名,单词>,概率>
   public static HashMap<HashMap<String,String>,Double> getT_C(){
       int TOATAL_TERM=0;
       HashMap<HashMap<String>,Double> hashMap=new HashMap();
       HashMap<String,Integer> total_count=new HashMap();
       String path="hdfs://localhost:9000/user/hadoop/TermCNumOutput/part-r-00000";
       Configuration conf=new Configuration();
       try {
           FileSystem fs=FileSystem.get(URI.create(path),conf);
            InputStream in=null;
           in=fs.open(new Path(path));
           BufferedReader br=new BufferedReader(new InputStreamReader(in));
           String line=null:
           String curDocType=null;
            //totalWordOfC record the word number of the current classC
            int totalWordOfC=0:
```

```
while((line=br.readLine())!=null){
       totalWordOfC++;
       String[] split=line.split("\\s+");
       String docType=split[0];
       String term=split[1];
       Double value=Double.parseDouble((split[2]))+1;
       HashMap<String,String> key=new HashMap();
       key.put(docType, term);
       hashMap.put(key,value);
       if(curDocType==null){
           curDocType=docType;
       if(!curDocType.equals(docType)){
           System.out.println("curDocType:"+curDocType+" doctype "+docType+" "+totalWordofC);
           total count.put(curDocType, totalWordOfC+Global Env.TOATAL TERM);
           TOATAL TERM+=totalWordOfC;
           curDocType=docType;
           totalWordOfC=0;
   }
    total count.put(curDocType, totalWordOfC+Global Env.TOATAL TERM);
    for(Map.Entry<HashMap<String, String>, Double> entry: hashMap.entrySet())
        Double value=entry.getValue();
        HashMap<String,String> hashKey=entry.getKey();
        int dev=0;
        for(String key:hashKey.keySet())
        {
        dev=total_count.get(key);
        }
        entry.setValue(value/dev);
} catch (IOException e) {
    // TODO Auto-generated catch block
    e.printStackTrace();
return hashMap;
```

```
//priorProbability.java
public class priorProbability {
    public static void main(String[] args) {
        getPriorPrb();
    //HashMap<类名,该类的先验概率>
    public static HashMap<String,Double> getPriorPrb(){
        int count=0;
        HashMap<String,Double> hashmap=new HashMap();
        String path="hdfs://localhost:9000/user/hadoop/DocTypeNumOutput/part-r-00000";
        Configuration conf=new Configuration();
        try {
            FileSystem fs=FileSystem.get(URI.create(path),conf);
            InputStream in=null;
            in=fs.open(new Path(path));
            {\tt BufferedReader\ br=new\ BufferedReader(new\ InputStreamReader(in));}
            String line=null;
            while((line=br.readLine())!=null){
                String[] split=line.split("\\s+");
                String key=split[0];
                Double value=Math.log(Double.parseDouble(split[1])/Global_Env.TOTAL_DOC);
                hashmap.put(key, value);
                count++;
        } catch (IOException e) {
            // TODO Auto-generated catch block
            e.printStackTrace();
        return hashmap;
```

```
//Judge.java
public class Judge {
    static HashMap<String,String> classOfDoc=new HashMap<String, String>();
    static HashMap<String,String> resMap=new HashMap<String, String>();
    static HashSet<String> classCSet=new HashSet<String>();
    static HashMap<String,Double> precisionMap=new HashMap<String,Double>();
    static HashMap<String,Double> recallMap=new HashMap<String,Double>();
    static HashMap<String,Double> f1Map=new HashMap<String,Double>();
    public static void main(String[] args) {
        readClassOfDoc();
        readResult();
        evaluationForC();
        macroPrecision();
        microPrecision();
    //读取文件的真实类型
    static void readClassOfDoc(){
      Configuration conf = new Configuration();
      String path = "docCountOutput/part-r-00000";
      FileSystem fs;
       try {
            fs = FileSystem.get(URI.create(path),conf);
           InputStream in=null;
           in=fs.open(new Path(path));
           BufferedReader br=new BufferedReader(new InputStreamReader(in));
           String line=null:
            while((line=br.readLine())!=null){
                String[] split=line.split("\\s+");
                String classC=split[0];
                String doc=split[1];
                classOfDoc.put(doc,classC);
           }
        } catch (IOException e) {
            // TODO Auto-generated catch block
            e.printStackTrace();
    static void evaluationForC(){
        Iterator<String> iterator = classCSet.iterator();
        while(iterator.hasNext()){
             double TP=0;
             double TN=0;
             double FP=0;
             double FN=0;
             String c=iterator.next();
             for(Map.Entry<String, String> testedDoc:resMap.entrySet()){
                 String docName=testedDoc.getKey();
                String resClass=testedDoc.getValue();
                String oriClass=classOfDoc.get(docName);
                 if(c.equals(oriClass)&&c.equals(resClass))
                 else if(c.equals(oriClass)&&!c.equals(resClass))
                 else if(!c.equals(oriClass)&&c.equals(resClass))
                 else if(!c.equals(oriClass)&&!c.equals(resClass))
                    TN++;
            double precision=TP/(TP+FP);
           double recall=TP/(TP+FN);
           double f1=2*precision*recall/(precision+recall);
           if(TP!=0)
                f1Map.put(c, f1);
           else
                f1Map.put(c, 0.0);
```

```
static void macroPrecision(){
       double precision=0;
       for(Map.Entry<String, Double> entry:f1Map.entrySet()){
           precision+=entry.getValue();
       precision/=f1Map.size();
       System.out.println("macroPrecision----- "+precision*100+"%");
   static void microPrecision(){
        double TP=0;
        double TN=0;
        double FP=0;
        double FN=0;
       Iterator<String> iterator = classCSet.iterator();
       while(iterator.hasNext()){
            String c=iterator.next();
            for(Map.Entry<String, String> testedDoc:resMap.entrySet()){
                String docName=testedDoc.getKey();
                String resClass=testedDoc.getValue();
                String oriClass=classOfDoc.get(docName);
                if(c.equals(oriClass)&&c.equals(resClass))
                else if(c.equals(oriClass)&&!c.equals(resClass))
                else if(!c.equals(oriClass)&&c.equals(resClass))
                else if(!c.equals(oriClass)&&!c.equals(resClass))
                    TN++;
            }
       double precision=TP/(TP+FP);
       double recall=TP/(TP+FN);
       double f1=2*precision*recall/(precision+recall);
       System.out.println("microPrecision----- "+precision*100+"%");
   }
}
```

五:数据集说明

本项目中使用了 data/Industry 目录下的 4999 个文件。

其中随机选择的测试文件 469 个, 训练文件 4533 个。

对训练集使用 CountDocType 执行文件类型的 MapReduce 统计,输出的文件 类型与个数如下:

```
hdfs://localho
                                                              I83954
                                                                       20
                 I22400
                                I34420
                                         23
                          10
                                              174000
 101000
          11
                                                       48
                                I34430
                                         7
                                                              I83960
                                                                       69
                 I22450
                          15
                                              175000
                                                       198
 I01001
          180
                                I34440
                                         6
                                                              I84000
                                                                       3
                 I22460
                          6
                                              175100
                                                       2
 T01002
          45
                                T34520
                                         8
                                                              I84200
                 I22470
                          4
                                              176300
                                                       23
                                I34531
                                         31
 I02000
          20
                                              I76400
                                                       10
                 I22471
                                                              I85000
                                                                      49
                                I34532
 I03000
          5
                                              177002
                                                       7
                          7
                 I22472
                                                              I92110
                                                                       13
                                T34540
                                         26
                                              179010
 I11000
          16
                 I23000
                          11
                                I34600
                                         4
                                                              I92120
                                              179020
                                                       171
 I13000
          325
                                I34700
                                         5
                 I24000
                          10
                                                              I92300
                                                                      1
                                              I81400
                                                       1
                                I35000
                                         18
 I14000
          65
                 I24100
                          1
                                                              I95100
                                                                       53
                                              I81401
                                                       11
                                I35101
                                         143
                 I24200
 I15000
                          17
          4
                                              T81402
                                                       494
                                                              I97100
                                I35102
                 I24300
          33
 I16000
                                         30
                                              I81403
                                                       7
                                T35300
                                                              I97400
                 I24400
                          1
                                              I81501
                                                       70
 I16100
          113
                                I36101
                                         12
                                                              I9741102
                                                                           11
               [124500
                          1
                                              I81502
                                                       128
                                I36102
                                         1
 I1610107
               124700
                                                              19741105
                          24
                                I36200
                                              I82000
                                                       30
                                         4
 I1610109
                                              I82001
                                                              19741110
                                                                           3
                                I36400
                                         31
                                                       26
                 I24794
                          2
          23
 I16200
                                              182002
                                                       14
                                I37000
                                         3
                                                              I9741112
                 I24800
                          5
 I16300
          2
                                         3
                                I37100
                                              I82003
                                                       13
                 I25000
                          72
                                                              I97412
                                I37200
                                         45
                                              I83100
                                                       96
 I17000
          30
                 I25110
                          3
                                                              I97911
                                                                       14
                                137300
                                         11
                                              I83200
                                                       2
 I21000
          121
                 I25120
                          14
                                                              I97912
                                I37330
                                         3
                                                                       18
                                              I83400
                                                       1
 I22000
          1
                 I25130
                          5
                                         5
                                T37400
                                              I83500
                                                       2
                                                              I98100
          79
I22100
                 I25140
                          8
                                         29
                                I41000
                                              I83600
```

对测试集使用 CountDocType 执行文件类型的 MapReduce 统计,输出的文件类型与个数如下:

```
hdfs://localhost
 101000
          1
                   T22450
 I01001
           16
                                   I34200
                                             2
                                                   174000
                                                             2
20
 I01002
           4
                   I22471
                             1
                                                                  I83954 2
                                   I34330
                                             1
 I02000
           3
                   T22472
                             1
                                                    T75000
                                   I34400
                                                                  I83960
                                                    176300
 T13000
           29
                   T24000
                             1
                                                             1
                                   I34420
                                                    179010
                                                                  I84000
                                                                          2
 I14000
                   T24200
                             4
                                   I34440
                                                    179020
                                                             14
 I15000
           2
                   I24700
                             3
                                                                  I84200
                                                                          1
                                   I34520
                                                    I81401
 I16000
                   I25000
                             4
                                   I34531
                                                                  I85000
                                                                          4
                                                    I81402
                                                             51
          13
 I16100
                   I25120
                             2
                                             3
                                   I34540
                                                    I81403
                                                             1
 I16200
                                                                  I95100
                                                                          4
           1
                   T25510
                             1
                                                    I81501
                                   I34600
 I17000
                                             1
           6
                   I25670
                             1
                                                                  I97100
                                                                          1
                                                    I81502
                                   T34700
                                             1
 I21000
           8
                   T25700
                             14
                                                    I82000
                                   I35000
                                                                  I97400
 I22100
                   T25800
                             1
                                                    I82001
 122400
                                   I35101
                                             14
                                                                  I9741102
                   I32000
                             4
                                                    T82002
                                   I35102
                                             1
 122450
                   I32200
                             2
                                                    I82003
                                                                  I9741110
                                                                              1
 I22471
                                   I35300
                                             1
                   T32840
                             1
                                                    183100
                                                                  197911 1
 I22472
                                   I36400
                                             1
                   T33010
                             1
                                   I37100
 I24000
                             24
                                                                  I97912
                   I33020
                                                   I83700
```

在全部类型中随机选择测试集时是按照 10%的比例,可以看出在具体到某一具体类型时,其训练集与测试集的比例也在 10:1 左右。

六:程序运行说明

Cluster Metrics

Apps Submitted	Apps Pending	Apps Running	Apps Completed	Containers Running	Memory Used	Memory Total	Memory Reserved	VCores Used	VCores Total	VCores Reserved	Active Nodes	Decommissioned Nodes	Lost Nodes	Unhealthy Nodes	Rebooted Nodes
10	0	0	10	0	0 B	8 GB	0 B	0	8	0	1	<u>0</u>	<u>0</u>	<u>0</u>	0
Show 20 🛨 e	Show 20 • entries Search:														
	ID	▼ Us	ser 🗢 Name		ation Type	≎ Queue ∜	StartTime	≎ Finis	hTime \$	State \$	FinalStatus	♦ Progress ♦	Tracking U	≎ Blacklis	ted Nodes
application_1	57683111796	4_0010 had	doop Count-Do Type	oc- MAPRED	UCE	default	Fri, 20 Dec 2019 10:40:4 GMT	,	Dec 2019 :25 GMT	FINISHED	SUCCEEDED		<u>History</u>	N/A	
application_1	57683111796	4_0009 had	doop Predictio	n MAPRED	UCE	default	Fri, 20 Dec 2019 10:36:1 GMT	,	Dec 2019 :05 GMT	FINISHED	SUCCEEDED		<u>History</u>	N/A	

```
19/12/20 02:38:03 INFO mapreduce.Job: map 100% reduce 100%
19/12/20 02:38:07 INFO mapreduce.Job: Job job_1576831117964_0009 completed succ
19/12/20 02:38:07 INFO mapreduce.Job: Counters: 49
          File System Counters
                      FILE: Number of bytes read=4634336
                      FILE: Number of bytes written=9483515
                      FILE: Number of read operations=0
FILE: Number of large read operations=0
                      FILE: Number of write operations=0
                      HDFS: Number of bytes read=3719174
HDFS: Number of bytes written=11256
                      HDFS: Number of read operations=12
                      HDFS: Number of large read operations=0
                      HDFS: Number of write operations=2
          Job Counters
                      Launched map tasks=1
Launched reduce tasks=1
                      Data-local map tasks=1
                      Total time spent by all maps in occupied slots (ms)=72984
Total time spent by all reduces in occupied slots (ms)=11390
                      Total time spent by all map tasks (ms)=72984
                      Total time spent by all reduce tasks (ms)=11390
Total vcore-milliseconds taken by all map tasks=72984
                      Total vcore-milliseconds taken by all reduce tasks=11390
Total megabyte-milliseconds taken by all map tasks=74735616
Total megabyte-milliseconds taken by all reduce tasks=11663360
          Map-Reduce Framework
                      Map input records=470
                      Map output records=101990
                      Map output bytes=4430350
                      Map output materialized bytes=4634336
                      Input split bytes=142
                      Combine input records=0
                      Combine output records=0
                      Reduce input groups=469
Reduce shuffle bytes=4634336
Reduce input records=101990
                      Reduce output records=469
                      Spilled Records=203980
Shuffled Maps =1
                      Failed Shuffles=0
                      Merged Map outputs=1
                      GC time elapsed (ms)=51814
                      CPU time spent (ms)=61570
                      Physical memory (bytes) snapshot=391086080
Virtual memory (bytes) snapshot=3898028032
Total committed heap usage (bytes)=251527168
          Shuffle Errors
                      BAD_ID=0
                      CONNECTION=0
                      IO ERROR=0
                      WRONG_LENGTH=0
                      WRONG_MAP=0
          WRONG_REDUCE=0
File Input Format Counters
                      Bytes Read=471579
          File Output Format Counters
                      .
Rvtes Written=11256
```

七:实验结果分析

对于 Macoraverage Precision, 输出如图:

```
<terminated>Judge [Java Application] /usr/lib/jvm/java-8-openjdk-amd64/bin/java (Dec 19, 2019 5:41:45 PM)
  log4j:WARN No appenders could be found for logger (org.apache.hadoop.metrics2.lib.MutableMetricsFactory).
log4j:WARN Please initialize the log4j system properly.
 f1=0.222222222222222
| International Content | Inte
                                                                                                                                                                                                                                                                f1=0.333333333333333333
                                                                                                                                                                                                                                                                                                                                                                                       f1=0.2873563218390805
                                                                                                                                                                                                                                                                                                                              f1=0.799999999999999
                                                                                                                                                                                                                                                                                                                                                                             f1=0.38095238095238093
                                                                                                                                                                                                                                                                                                                          f1=0.7894736842105262
                                                                                                                                                                                                                                                                                                                                                                                  f1=0.449438202247191
                                                                                                                                                                                                                                                                                                                                     f1=0.4761904761904762
                                                                                                                                                                                                                                                                                                                                                                         f1=0.72000000000000001
                                                                                                                                                                                                                                                                                                                       f1=0.26666666666666667
  I72300---TP=1.0 TN=468.0 FP=0.0 FN=0.0 precision=1.0 recall1.0
 macroPrecision-----
                                                                                         53.05630989177219%
```

如上图,对于 I21000, 其训练集中有 121 个文件,测试集中有 8 个文件,其 f1=66.6%;

对于 181502, 其训练集中有 128 个文件, 测试集中有 4 个文件, 其 f1=40%; 对于 113000, 其训练集中有 325 个文件, 测试集中有 29 个文件, 其 f1=44.9% 可以看出由于训练集与测试数据集的数量不同, 每一类的 precision、recall 与 f1 的值都有较大差异。

对所有类计算 Micoraverage Precision, 结果如图:

```
lastLocatedBlock=LocatedBlock{BP-728634802-192.168.5
isLastBlockComplete=true}
[main] DEBUG org.apache.hadoop.hdfs.DFSClient - Connec
[main] DEBUG org.apache.hadoop.hdfs.protocol.datatrans
microPrecision------ 39.28365106874639%
[Thread-3] DEBUG org.apache.hadoop.ipc.Client - stoppi
```