Optimization

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Optimization strategies:

- (1) Always operate data on rdd except that I have to collect data in the popularity map.
- (2) Persist the input rdd, in which case Spark will keep the elements around on the cluster for much faster access the next time you query it.

```
val input = sc.textFile(inputFile)
input.persist()
```

(3) Broadcast the popularityMap. Broadcast variables allow the programmer to keep a read-only variable cached on each machine rather than shipping a copy of it with tasks. They can be used, for example, to give every node a copy of a large input dataset in an efficient manner.

```
//broadcast popularity
val global_popularity = sc.broadcast(popularityMap)
```

(4)Apply PPjoin :length filter. For example, for set {A,B,C,D,E} and threshold=0.8, discard all sets with length smaller than 4.

```
if (math.ceil(candidatel.size * threshold) <= candidate2.size){ //length filter
```

(5) Only calculate intersection instead of both intersection and union when calculating similarity.

```
val intersection_size = (candidate1 intersect candidate2).size.toDouble
//calculate similarity
val similarity = intersection_size / (candidate1.size + candidate2.size - intersection_size).toDouble
```

(6)Use map-reduce to get distinct pairs instead of ".distinct".

val pairID_similarity = prefixToken_ridContenPair.flatMap ($x \Rightarrow my_filter(x)$) // get (rid1,rid2),similarity) by $my_filter(x)$.reduceByKey((a,b)=>a) //get distinct pairs

(7) I tried Positional Filter but the result is slower than before. It is probably because the data in filckr_london.txt are not suitble for it. Extra calculating time is more than the time saved.