

**BerkeleyX: CS110x Big Data Analysis with Apache Spark**

Bookmarks

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Lecture 3: Apache Spark Resilient Distributed Datasets

Quizzes



Lab3a - RDD Tutorial

Lab due Sep 13, 2016 at 04:30 IST

Lab 3b - Text Analysis and Entity Resolution

Lab due Sep 13, 2016 at 04:30 IST

Lab 3b Quiz Questions

Quizzes



Week 3 - Programming with Resilient Distributed Datasets > Lab 3b Quiz Questions > Lab 3b Quiz Questions



Bookmark

You should complete Lab 3b before answering these quiz questions.

Lab 3b Part (2d) Tokens with the smallest IDF

(1/1 point)

In part (2d), do you think the 11 terms (tokens) are useful for entity resolution?

☐ Yes

☒ No ✓

EXPLANATION

Answer the next quiz question for the explanation.

Lab 3b Part (2d) Explanation

(1/1 point)

In part (2d), why do you think the terms are useful or not useful for entity resolution?

- ☐ These terms are useful for entity resolution because they describe distinguishing tokens in product descriptions
- ☒ These terms not useful for entity resolution because they are generic terms for marketing, prices, and product categories. ✓

EXPLANATION

For this question, the answer is the explanation - the terms are too generic to be useful in entity resolution.

Lab 3b Part (2e) IDF Histogram

(1/1 point)

Using the plot in (2e), what conclusions can you draw from the distribution of weights?

- ☐ The distribution of IDF values is very dense.
- ☐ You cannot draw any conclusions from the histogram.

☒ There is a long tail of rare words in the corpus - these have large IDF values. ✓

☐ The distribution of IDF values is very flat.

EXPLANATION

There are gaps between IDF values because IDF is a function of a discrete variable, i.e., a document count.

Lab 3b Part (3e) Perform a Gold Standard evaluation

(1/1 point)

In part (3e) you used the "gold standard" data to answer the following questions:

- * How many true duplicate pairs are there in the small data sets?
- * What is the average similarity score for true duplicates?
- * What about for non-duplicates?

Based on the answers to the questions in part (3e), is cosine similarity doing a good job, qualitatively speaking, of identifying duplicates?

☒ Yes ✓

☐ No

EXPLANATION

Cosine similarity looks useful, because duplicates on average are 250X more similar than non-duplicates. As long as variance isn't too high, that's a good signal.

Lab 3b Part (5c) Line Plots - Part 1

(1/1 point)

Using the plots in (5c), what is the optimal threshold value to maximize the F-measure?

☐ 0

☐ 0.1

☒ 0.2 ✓

☐ 0.5

☐ 0.85☐ 1.0**EXPLANATION**

F-measure is maximized with the threshold equal to ~ 0.2 , so that is the optimal threshold if we value precision and recall equally.

Lab 3b Part (5c) Line Plots - part 2

(1/1 point)

If false-positives are considered much worse than false-negatives, how does that change your answer?

☐ 0☐ 0.1☐ 0.2☒ 0.5 ✓

☒ 0.85 ✓☐ 1.0**EXPLANATION**

If we wanted to really avoid false positives, that means we want higher precision at the cost of lower recall, in which case ~ 0.5 offers the best trade-off. If we didn't care at all about recall, ~ 0.85 has peak precision, and would be the best choice.

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