Indian Institute of Technology Jodhpur

CSL7110 Machine Learning with Big Data

Assignment 1: Map-Reduce and Similar Itemsets Mining

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1. Create k-Grams

A.

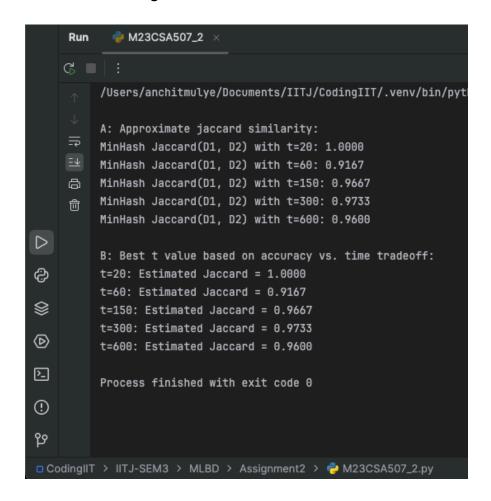
M23CSA507_1 B: Jaccard similarity between document pairs: Jaccard(D1, D2) [2-gram, char]: 0.9811 Jaccard(D1, D3) [2-gram, char]: 0.8157 ∃y Jaccard(D1, D4) [2-gram, char]: 0.6444 Jaccard(D2, D3) [2-gram, char]: 0.8000 Jaccard(D2, D4) [2-gram, char]: 0.6413 Jaccard(D3, D4) [2-gram, char]: 0.6530 Jaccard(D1, D2) [3-gram, char]: 0.9780 \triangleright Jaccard(D1, D3) [3-gram, char]: 0.5804 6 Jaccard(D1, D4) [3-gram, char]: 0.3051 Jaccard(D2, D3) [3-gram, char]: 0.5680 Jaccard(D2, D4) [3-gram, char]: 0.3059 Jaccard(D3, D4) [3-gram, char]: 0.3121 (D) Jaccard(D1, D2) [2-gram, word]: 0.9408 Jaccard(D1, D3) [2-gram, word]: 0.1823 2 Jaccard(D1, D4) [2-gram, word]: 0.0302 Jaccard(D2, D3) [2-gram, word]: 0.1737 (!) Jaccard(D2, D4) [2-gram, word]: 0.0303 ଫ Jaccard(D3, D4) [2-gram, word]: 0.0161

□ CodingIIT > IITJ-SEM3 > MLBD > Assignment2 > 🟺 M23CSA507_1.py

B.

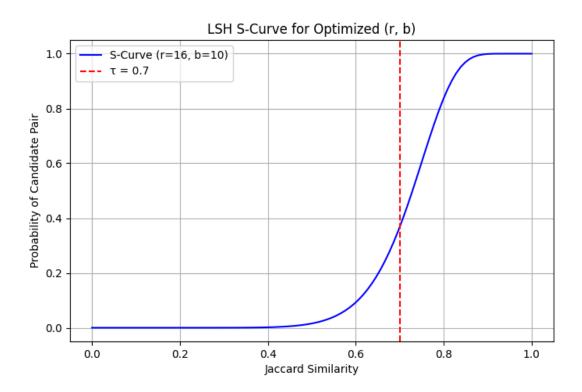
```
M23CSA507_1 ×
     Run
          A: Distinct k-grams for each document:
\triangleright
          D1 (2-gram, char): 263 unique k-grams
          D2 (2-gram, char): 262 unique k-grams
63
          D3 (2-gram, char): 269 unique k-grams
          D4 (2-gram, char): 255 unique k-grams
D1 (3-gram, char): 765 unique k-grams
          D2 (3-gram, char): 762 unique k-grams
⦸
          D3 (3-gram, char): 828 unique k-grams
          D4 (3-gram, char): 698 unique k-grams
2
          D1 (2-gram, word): 279 unique k-grams
          D2 (2-gram, word): 278 unique k-grams
①
          D3 (2-gram, word): 337 unique k-grams
ଫ୍ର
          D4 (2-gram, word): 232 unique k-grams
□ CodingIIT > IITJ-SEM3 > MLBD > Assignment2 > 👨 M23CSA507_1.py
```

2. Min-Hashing



3. LSH:

A.



В.

```
M23CSA507_3 ×
     Run
          /Users/anchitmulye/Documents/IITJ/CodingIIT/.venv/bin/python /Users/anchitmulye/Docum
          B: Best t value based on accuracy vs. time tradeoff:
          t=20: Estimated Jaccard = 1.0000
         t=60: Estimated Jaccard = 1.0000
69
         t=150: Estimated Jaccard = 1.0000
          t=300: Estimated Jaccard = 1.0000
寥
          t=600: Estimated Jaccard = 1.0000
          Optimal LSH parameters: r = 16, b = 10
Ø
          LSH Probability(D1, D2) > 0.7: 1.0000
          LSH Probability(D1, D3) > 0.7: 0.0671
兦
          LSH Probability(D1, D4) > 0.7: 0.0001
①
          LSH Probability(D2, D3) > 0.7: 0.0545
          LSH Probability(D2, D4) > 0.7: 0.0001
          LSH Probability(D3, D4) > 0.7: 0.0001
□ CodingIIT > IITJ-SEM3 > MLBD > Assignment2 > 🧖 M23CSA507_3.py
```

4. Min-Hashing on MovieLens dataset

```
M23CSA507_4 ×
     Run
         /Users/anchitmulye/Documents/IITJ/CodingIIT/.venv/bin/python /Users/anchitmulye/Docum
         Exact Jaccard pairs (≥ 0.5): 10
         MinHash t=50: False Positives = 34079, False Negatives = 5
     minHash t=100: False Positives = 29661, False Negatives = 5
    MinHash t=200: False Positives = 17763, False Negatives = 4
     🖶 | t=50: Avg False Positives = 6815.80, Avg False Negatives = 1.00
     亩 t=100: Avg False Positives = 5932.20, Avg False Negatives = 1.00
          t=200: Avg False Positives = 3552.60, Avg False Negatives = 0.80
          Process finished with exit code 0
6
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℗
2
①
안
□ CodingIIT > IITJ-SEM3 > MLBD > Assignment2 > 🦆 M23CSA507_4.py
```

5. LSH on MovieLens dataset

