SOLUTIONS **Exercises on Block2: Finding Frequent Item Sets Finding Similar Items Searching in Data Streams**

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Frequent Item Sets (1) – Assignment

- Suppose 100 items (numbered 1 to 100) and 100 baskets (numbered 1 to 100)
 - Item i is in basket b if and only if i divides b with no remainder, i.e., item 1 is in all the baskets, item 2 is in all fifty of the even-numbered baskets, etc.
- Consider that the support threshold is 5:
 - 1) Identify the frequent items
 - 2) Compute the confidence of these association rules
 - a) $\{5, 7\} \rightarrow 2$
 - b) $\{2, 3, 4\} \rightarrow 5$

Frequent Item Sets (1) – Recap

- Simplest question: Find sets of items that appear together "frequently" in baskets
- Support for item set I: Number of baskets containing all items in I
 - (Often expressed as a fraction of the total number of baskets)
- Given a support threshold s, then sets of items that appear in at least s baskets are called frequent itemsets

TID	Items
1	Bread, Coke, Milk
2	Beer, Bread
3	Beer, Coke, Diaper, Milk
4	Beer, Bread, Diaper, Milk
5	Coke, Diaper, Milk

Support of {Beer, Bread} = 2

Frequent Item Sets (1) – Recap

Association Rules:

- If-then rules about the contents of baskets
- $\{i_1, i_2, ..., i_k\} \rightarrow j$ means: "if a basket contains all of $i_1, ..., i_k$ then it is *likely* to contain j''
- In practice there are many rules, want to find significant/interesting ones!
- **Confidence** of this association rule is the probability of *j* given $I = \{i_1, ..., i_k\}$

$$\operatorname{conf}(I \to j) = \frac{\operatorname{support}(I \cup j)}{\operatorname{support}(I)}$$

Frequent Item Sets (1) – Solution

- 1) 20 frequent items: **1–20**
- 2) Association rules:
 - a) The baskets containing both items 5 and 7 are baskets 35 and 70, in which only basket 70 also contains item 2. Hence, the confidence of the rule {5, 7} → 2 is 1/2.
 - b) The baskets whose numbers are the multiples of 12 contain item set {2, 3, 4} as a subset – there are 8 such baskets. The baskets whose numbers are the multiples of 60 contain item set {2, 3, 4, 5} as a subset – there is 1 such basket. Hence, the confidence of the rule {2, 3, 4} → 5 is 1/8.

Frequent Item Sets (2) – Assignment

- Consider the following twelve baskets, each of them contains 3 of 6 items (1 through 6):
 - {1, 2, 3} {2, 3, 4} {3, 4, 5} {4, 5, 6}
 - {1, 3, 5} {2, 4, 6} {1, 3, 4} {2, 4, 5}
 - {3, 5, 6} {1, 2, 4} {2, 3, 5} {3, 4, 6}
- Suppose the support threshold is 4. On the first pass of the PCY algorithm, a hash table with 11 buckets is used, and the set {*i*, *j*} is hashed to bucket *i*×*j* mod 11:
 - 1) Compute the support for each item and each pair of items
 - 2) Which pairs hash to which buckets?
 - 3) Which buckets are frequent?
 - 4) Which pairs are counted on the second pass?

Frequent Item Sets (2) – Recap

PCY Algorithm – First Pass

```
FOR (each basket) :
    FOR (each item in the basket) :
        add 1 to item's count;
        FOR (each pair of items) :
        hash the pair to a bucket;
        add 1 to the count for that bucket;
```

Few things to note:

- Pairs of items need to be generated from the input file; they are not present in the file
- We are not just interested in the presence of a pair, but we need to see whether it is present at least s (support) times

Frequent Item Sets (2) – Recap

- Observation: If a bucket contains a frequent pair, then the bucket is surely frequent
- However, even without any frequent pair, a bucket can still be frequent ³
 - So, we cannot use the hash to eliminate any member (pair) of a "frequent" bucket
- But, for a bucket with total count less than s, none of its pairs can be frequent ⁽³⁾
 - Pairs that hash to this bucket can be eliminated as candidates (even if the pair consists of 2 frequent items)

Pass 2:

Only count pairs that hash to frequent buckets

Frequent Item Sets (2) – Solution 1/4

 Compute the support for each item and each pair of items

Support for each item:

item	1	2	3	4	5	6
support	4	6	8	8	б	4

Support for each pair of items:

pair support		{1, 4} 2			{2, 5} 2
		{3, 5} 4		{5, 6} 2	

Frequent Item Sets (2) – Solution 2/4

2) Which pairs hash to which buckets?

The set {*i*, *j*} is hashed to bucket no.: *i*×*j* mod 11

	{1, 3} 3			
	{3, 4} 1			

Frequent Item Sets (2) – Solution 3/4

- 3) Which buckets are frequent?
 - Bucket support sum of supports of pairs belonging to the given bucket:

bucket	0	1	2	3	4	5	б	7
support	0	5	5	3	б	1	3	2
bucket	8	9	10					

The frequent buckets are those with support above 4, i.e., buckets: 1, 2, 4, 8

Frequent Item Sets (2) – Solution 4/4

- 4) Which pairs are counted on the second pass
 - As only pairs in frequent buckets will be counted on the second pass of PCY, they are:

 $\{1, 2\}, \{1, 4\}, \{2, 4\}, \{2, 6\}, \{3, 4\}, \{3, 5\}, \{4, 6\}, \{5, 6\}$

Shingling (1) – Assignment

- Consider two documents A and B
 - If their 3-shingle resemblance is 1 (using Jaccard similarity), does that mean that A and B are identical?
 - If so, prove it. If not, give a counterexample.

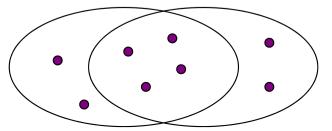
Shingling (1) – Recap

- A k-shingle (or k-gram) for a document is a sequence of k tokens that appears in the doc
 - Tokens can be characters, words or something else, depending on the application
 - Assume tokens = characters for examples
- Example: k=2; document D₁ = abcab Set of 2-shingles: S(D₁) = {ab, bc, ca}
 - Option: Shingles as a bag (multiset), count ab twice: S'(D₁) = {ab, bc, ca, ab}

Shingling (1) – Recap

- Document D₁ is a set of its k-shingles C₁=S(D₁)
- Equivalently, each document is a 0/1 vector in the space of k-shingles
 - Each unique shingle is a dimension
 - Vectors are very sparse
- A natural similarity measure is the Jaccard similarity:

 $sim(D_1, D_2) = |C_1 \cap C_2| / |C_1 \cup C_2|$



Shingling (1) – Solution

- No, the documents A and B need not be identical
 - Counterexample:
 - A: abab
 - 3-shingles: S(A) = {aba, bab}
 - B: baba
 - 3-shingles: S(B) = {bab, aba}
 - $sim(A, B) = | S(A) \cap S(B) | / | S(A) \cup S(B) | = 1$

Shingling (2) – Assignment

- Consider two documents A and B
 - Each document's number of token is O(n)
 - It does not matter whether tokens are characters or words
 - What is the runtime complexity of computing A and B's k-shingle resemblance (using Jaccard similarity)?
 - Assume that comparison of two k-shingles to assess their equivalence is O(k)
 - Express your answer in terms of n and k, where n >> k

Shingling (2) – Solution

Time to create shingles: O(n)

- Time to find intersection (using the brute force algorithm): O(k·n²)
 - n shingles in each document
- Time to find union (using the intersection): O(n)
- Total time: O(k·n²)

Finding Similar Items (1) – Assignment

- Compute the Jaccard similarities of each pair of the following three sets:
 - A = {1, 2, 3, 4}
 - B = {2, 3, 5, 7}
 - C = {2, 4, 6}

Finding Similar Items (1) – Solution

- sim(A, C) = 2/5
- sim(B, C) = 1/6

Finding Similar Items (2) – Assignment

For the matrix

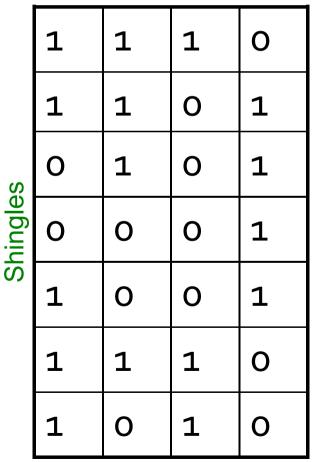
Element	D1	D2	D3	D4
0	0	1	0	1
1	0	1	0	0
2	1	0	0	1
3	0	0	1	0
4	0	0	1	1
5	1	0	0	0

- Compute the minhash signature for each column (document) using the following hash functions:
 - $h_1(x) = 2x + 1 \mod 6$
 - $h_2(x) = 3x + 2 \mod 6$
 - $h_3(x) = 5x + 2 \mod 6$
- 2) Which of these hash functions are true permutations?
- 3) How close are the estimated Jaccard similarities for the six pairs of columns to the true Jaccard similarities?

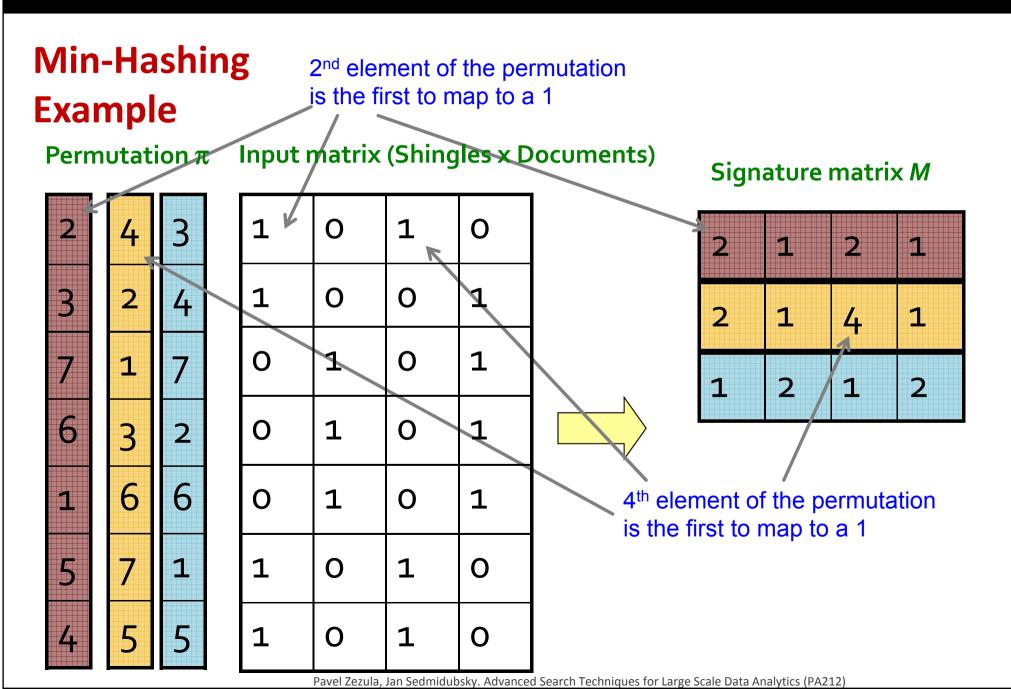
Finding Similar Items (2) – Recap

- Rows = elements (e.g., shingles)
- Columns = sets (e.g., documents)
 - 1 in row *e* (shingle) and column *s* (document) if and only if *e* is a member of *s*
 - Column similarity is the Jaccard similarity of the corresponding sets (rows with value 1)
 - Typical matrix is sparse!
- Each document is a column:
 - Example: sim(C₁,C₂) = ?
 - Size of intersection = 3; size of union = 6, Jaccard similarity (not distance) = 3/6
 - d(C₁,C₂) = 1 (Jaccard similarity) = 3/6

Documents



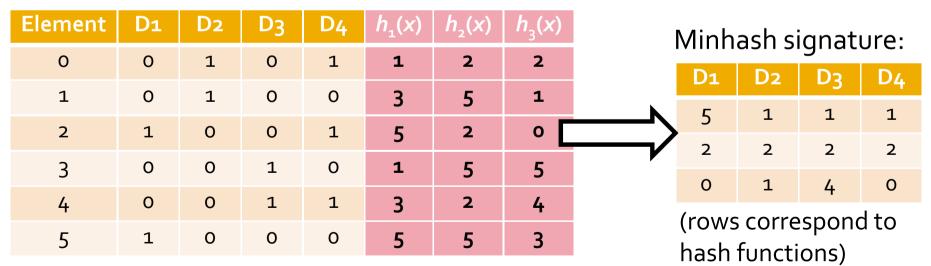
Finding Similar Items (2) – Recap



Finding Similar Items (2) – Solution 1+2/3

- 1) Compute the minhash signature for each column using the following hash functions:
 - $h_1(x) = 2x + 1 \mod 6$
 - $h_2(x) = 3x + 2 \mod 6$
 - $h_3(x) = 5x + 2 \mod 6$

Hashes are computed on element IDs:



2) Which of these hash functions are true permutations: h_3 only

Finding Similar Items (2) – Solution 3/3

3) How close are the estimated Jaccard similarities for the six pairs of columns (documents) to the true Jaccard similarities?

Jaccard similarities on	D1 / D2	D1/D3	D1 / D4	D2 / D3	D2 / D4	D3 / D4
Original documents	0	0	0.25	0	0.25	0.25
Minhash signatures	0.33	0.33	0.67	0.67	0.67	0.67

- => the estimated Jaccard similarities are not close to the true ones at all
 - To make the estimated similarity closer to the true one, there is a need of more and better (i.e., resulting in true permutations) hash functions

Data Streams (1) – Assignment

- Suppose we are maintaining a count of 1s using the DGIM method
 - Each bucket is represented by (*i*, *t*)
 - i the number of 1s in the bucket
 - t the bucket timestamp (time of the most recent 1)
- Consider the following properties:
 - Current time is 200
 - Window size is 60
 - Current buckets are:
 - (16, 148) (8, 162) (8, 177) (4, 183) (2, 192) (1, 197) (1, 200)
 - At the next ten clocks (201 through 210), the stream has 0101010101
- What will the sequence of buckets be at the end of these ten inputs?

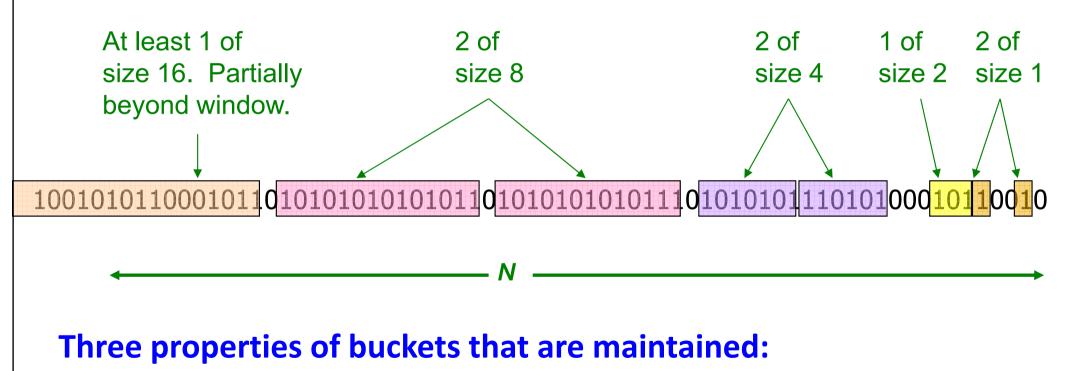
DGIM: Buckets

- A *bucket* in the DGIM method is a record consisting of:
 - (A) The timestamp of its end [O(log N) bits]
 - (B) The number of 1s between its beginning and end [O(log log N) bits]
- Constraint on buckets:
 Number of 1s must be a power of 2
 - That explains the O(log log N) in (B) above

Representing a Stream by Buckets

- Either one or two buckets with the same power-of-2 number of 1s
- Buckets do not overlap in timestamps
- Buckets are sorted by size
 - Earlier buckets are not smaller than later buckets
- Buckets disappear when their end-time is > N time units in the past

Example: Bucketized Stream



- Either one or two buckets with the same power-of-2 number of 1s
- Buckets do not overlap in timestamps
- Buckets are sorted by size

Updating Buckets (1)

- When a new bit comes in, drop the last (oldest) bucket if its end-time is prior to N time units before the current time
- **2 cases:** Current bit is **0** or **1**

If the current bit is 0: no other changes are needed

Updating Buckets (2)

- If the current bit is 1:
 - (1) Create a new bucket of size 1, for just this bit
 - End timestamp = current time
 - (2) If there are now three buckets of size 1, combine the oldest two into a bucket of size 2
 - (3) If there are now three buckets of size 2, combine the oldest two into a bucket of size 4
 - (4) And so on ...

Updating buckets (example):

Current state of the stream: Bit of value 1 arrives Two orange buckets get merged into a yellow bucket Next bit 1 arrives, new orange bucket is created, then 0 comes, then 1: Buckets get merged... State of the buckets after merging

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Data Streams (1) – Solution

- There are 5 occurrences of 1s in the stream. Each one updates the buckets to be:
 - (1) Combine the oldest two buckets of size 1

 (16, 148)
 (8, 162)
 (4, 183)
 (2, 192)
 (1, 197)
 (1, 200)
 (1, 202)
 - => (16, 148) (8, 162) (8, 177) (4, 183) (2, 192) (2, 200) (1, 202)
 - (2) No combination needed
 (16, 148) (8, 162) (8, 177) (4, 183) (2, 192) (2, 200) (1, 202) (1, 204)
 - (3) Combine the oldest two buckets of size 1, and then oldest two buckets of size 2 (16, 148) (8, 162) (8, 177) (4, 183) (2, 192) (2, 200) (1, 202) (1, 204) (1, 206)
 - => (16, 148) (8, 162) (8, 177) (4, 183) (2, 192) (2, 200) (2, 204) (1, 206)
 - => (16, 148) (8, 162) (8, 177) (4, 183) (4, 200) (2, 204) (1, 206)
 - (4) No combination needed; window size is 60, so (16, 148) should be dropped (16, 148) (8, 162) (8, 177) (4, 183) (4, 200) (2, 204) (1, 206) (1, 208)
 - => (8, 162) (8, 177) (4, 183) (4, 200) (2, 204) (1, 206) (1, 208)
 - (5) Combine the oldest two buckets of size 1

 (8, 162)
 (8, 177)
 (4, 183)
 (4, 200)
 (2, 204)
 (1, 206)
 (1, 208)
 (1, 210)
 - => (8, 162) (8, 177) (4, 183) (4, 200) (2, 204) (2, 208) (1, 210)