## Big Data Analytics

## ESSEC

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HomeWork 4 Solution: Mining Data Streams, part 2

- 1. (Exercise 4.2.1 MMDS book) Suppose we have a stream of tuples with the schema (university, courseID, studentID, grade). Assume universities are unique, but a courseID is unique only within a university (i.e., different universities may have different courses with the same ID, e.g., "CS101") and likewise, studentID's are unique only within a university (different universities may assign the same ID to different students). Suppose we want to answer certain queries approximately from a 1/20th sample of the data. For each of the queries below, indicate how you would construct the sample. That is, tell what the key attributes should be.
  - (a) For each university, estimate the average number of students in a course.
  - (b) Estimate the fraction of students who have a GPA of 3.5 or more.
  - (c) Estimate the fraction of courses where at least half the students got "A."

## Solution:

- (a) The query wants to generate average number of students in a course. For each tuple, the "university" field is unique, then we chose "university" as the key. To take a sample of 1/20th, we hash the key for each tuple to an integer from 0 to 19, and accept the tuple for the sample if the hash value is 0. Thus, we store only 1/20th of the tuples as the sample, and discard others. For each university in the sample, we can easily count the average number of students in a course.
- (b) The query wants to estimate the fraction of students who have a GPA of 3.5 or more. Since the "studentID" is unique only within a university, it cannot be alone used to identify one tuple in the stream. Here we need to build a composite key, (university, studentID) to identify each couple. We hash the composite key to an integer from 0 to 19, and accept the tuple for the sample if the hash value is 0. Thus, we store only 1/20th of the tuples as the sample, and discard others.
- (c) Same solution as for (b) taking as composite key (university, courseID).
- 2. (Exercise 4.3.1: MMDS book) For the situation of our running example (8 billion bits, 1 billion members of the set S), calculate the false-positive rate if we use three hash functions? What if we use four hash functions?
  - **Solution:** As we discussed during the lecture, the probability that a given bit will be 1, using one hash function, is  $1-e^{-1/8}\approx 0.1175$ . Now, suppose that we use three different hash functions. This situation corresponds to throwing three billion darts at eight billion targets, and the probability that a bit remains 0 is  $e^{-3/8}$ . In order to be a false positive, a nonmember of S must hash thrice to bits that are 1, and this probability is  $(1-e^{-3/8})^3\approx 0.03$ . Adding a fourth hash function we will get  $(1-e^{-1/2})^4\approx 0.024$ .
- 3. (Exercise 4.4.1 MMDS book ) Suppose our stream consists of the integers 3, 1, 4, 1, 5, 9, 2, 6, 5. Our hash functions will all be of the form  $h(x) = ax + b \mod 32$  for some a and b. You should treat the result as a 5-bit binary integer. Determine

the tail length for each stream element and the resulting estimate of the number of distinct elements if the hash function is:

- (a)  $h(x) = 2x + 1 \mod 32$ ;
- (b)  $h(x) = 3x + 7 \mod 32$ ;
- (c)  $h(x) = 4x \mod 32$ ;

## Solution:

| (a) | Element | Hashed value | Binary representation | Tail length |
|-----|---------|--------------|-----------------------|-------------|
|     | 3       | 7            | 00111                 | 0           |
|     | 1       | 3            | 00011                 | 0           |
|     | 4       | 9            | 01001                 | 0           |
|     | 1       | 3            | 00011                 | 0           |
|     | 5       | 11           | 01011                 | 0           |
|     | 9       | 19           | 10011                 | 0           |
|     | 2       | 5            | 00101                 | 0           |
|     | 6       | 13           | 01101                 | 0           |
|     | 5       | 11           | 01011                 | 0           |

For the maximum tail length, R, we have R=0 and the number of distinct elements is estimated to be  $2^0=1$ .

| (b) | Element | Hashed value | Binary representation | Tail length |
|-----|---------|--------------|-----------------------|-------------|
|     | 3       | 16           | 10000                 | 4           |
|     | 1       | 10           | 01010                 | 1           |
|     | 4       | 19           | 10011                 | 0           |
|     | 1       | 10           | 01010                 | 1           |
|     | 5       | 22           | 10110                 | 1           |
|     | 9       | 2            | 00010                 | 1           |
|     | 2       | 13           | 01101                 | 0           |
|     | 6       | 25           | 11001                 | 0           |
|     | 5       | 22           | 10110                 | 1           |

For the maximum tail length, R, we have R=4 and the number of distinct elements is estimated to be  $2^4=16$ .

| (c) | Element | Hashed value | Binary representation | Tail length |
|-----|---------|--------------|-----------------------|-------------|
|     | 3       | 12           | 01100                 | 2           |
|     | 1       | 4            | 00100                 | 2           |
|     | 4       | 16           | 10000                 | 4           |
|     | 1       | 4            | 00100                 | 2           |
|     | 5       | 20           | 10100                 | 2           |
|     | 9       | 4            | 00100                 | 2           |
|     | 2       | 8            | 11000                 | 3           |
|     | 6       | 24           | 11000                 | 3           |
|     | 5       | 20           | 10100                 | 3           |

For the maximum tail length, R, we have R=4 and the number of distinct elements is estimated to be  $2^4=16$ .

If we take the mean of these estimators, we get 11 as an estimate of the number of distinct elements which overestimates the true number, 6.