Bloom Filters

Anil Maheshwari

anil@scs.carleton.ca School of Computer Science Carleton University Canada

1

Outline

Bloom Filter

Data Structure

Queries

False-Positives

Analysis

Summary

Bloom Filter

Problem Definition

Let U be the universe. Input: A subset $S \subseteq U$. Query: For any $q \in U$, decide whether $q \in S$ quickly.

Objective

Answer queries quickly and use very little extra space.

SPAM Detection

- U = AII possible email addresses;
- S = My collection of non-junk email addresses.

Query: Given any $q \in U$, report whether $q \in S$?

History of Bloom Filters

- Bloom, *Space/Time tradeoffs in Hash Coding with Allowable Errors*, Communications of ACM 1970
- Space-Efficient Probabilistic Data Structure for Membership Testing
- May have false positives
- Numerous Variants: Counting Filters, Dynamic Filters with insertion/deletion of elements in *S*.
- Applications: Estimating size of union/intersection of sets, Avoid cashing 'one-hit wonders', Google Bigtable, Chrome's used it to detect malicious URLs,
- Refined Analysis in 2008 by members of our school.

Data Structure

Data Structure

An array B consisting of m bits and k hash functions $h_1,h_2,\ldots,h_k,$ where $h_i:U\to\{1,\ldots,m\}$

Initialization

 $B \leftarrow 0.$ For all $x \in S$, set $B[h_1(x)] = B[h_2(x)] = \cdots = B[h_k(x)] = 1.$

An Illustration

Queries

Queries

Answering Query

For any query $q \in U$, if $B[h_1(q)] = B[h_2(q)] = \cdots = B[h_k(q)] = 1$, report $q \in S$, else report $q \notin S$.

Observation

If $q \in S$, the queries are answered correctly.

False Positives

Suppose $q \notin S$ If $B[h_1(q)] = B[h_2(q)] = \cdots = B[h_k(q)] = 1$, we will report that $q \in S$.

False-Positives

Claim: Let n = |S|. After initializing Bloom filter *B* of size *m* with *k* hash-functions for elements of *S*, $Pr(B[l] = 1) = p = 1 - (1 - \frac{1}{m})^{nk}$, where $l \in \{1, ..., m\}$.

On query $q \notin S$, for False-Positive to occur, all of the *k* specified locations $B[h_1(q)], \ldots, B[h_k(q)]$ must be "1".

Bloom70

 $Pr(B[h_1(q)] = B[h_2(q)] = \dots = B[h_k(q)] = 1) = p^k.$

Analysis

An Example

Let
$$n = 1, m = 2, k = 2,$$

 $U = \{x, y\}, S = \{x\}$ and $q = y \neq x.$

Implicit assumption that $B[h_2(q)] = 1$ is independent of $B[h_1(q)] = 1$ may not be true . . .

We came up with a fairly technical proof and showed that

Theorem

2. L

Let $p_{k,n,m}$ be the false-positive rate for a Bloom filter that stores n elements of a set S in a bit-vector of size m using k hash functions.

1. We can express $p_{k,n,m}$ in terms of the Stirling number of second kind as follows:

$$p_{k,n,m} = \frac{1}{m^{k(n+1)}} \sum_{i=1}^{m} i^k i! \binom{m}{i} \begin{cases} kn \\ i \end{cases}$$

et $p = 1 - (1 - 1/m)^{kn}$, $k \ge 2$ and $\frac{k}{p} \sqrt{\frac{\ln m - 2k \ln p}{m}} \le c$ for some $c < 1$.
Upper and lower bounds on $p_{k,n,m}$ are given by

$$p^k < p_{k,n,m} \le p^k \left(1 + O\left(\frac{k}{p}\sqrt{\frac{\ln m - 2k\ln p}{m}}\right)\right)$$

Summary

- 1. A simple scheme for testing membership. Has one-sided error, i.e., false positives.
- 2. How to find the right number of hash functions and right size of the filter?
- 3. Implemented in various search engines, routers, SPAM filters, ...
- Unpleasant analysis in our work (Reference: P. Bose, H.Guo, E. Kranakis, A. Maheshwari, P. Morin, J. Morrison, M. Smid, Y. Tang: On the false-positive rate of Bloom filters. Inf. Process. Letters 108(4): 210-213 (2008))
- 5. Challenge: A nicer analysis. Hopefully, this will help with the analysis of variants of Bloom Filters.

A Toy Bloom Filter

Design a Bloom filter *B* for a small universe *U* and a subset $S \subseteq U$. Experiment with different sizes of *U*, *S*, and *k*. Evaluate the probability of false positives experimentally and compare with the quantity p^k . (Try to use some library for hash functions.)