Hashing and its applications

1 Pattern matching

Suppose we are trying to find a pattern string *P* in a long document *D*. How can we do it quickly and efficiently?

Hash the pattern P into, say, a 16-bit value. Then run through the file, hashing each set of |P| consecutive characters into a 16-bit value. If we ever get a match for a pattern, we can check to see if it corresponds to an actual pattern match. (In this case, we want to double-check and not report any false matches!) Otherwise we can just move on. We can use more than 16 bits, too; we would like to use enough bits so that we will obtain few false matches.

This scheme is efficient, as long as hashing is efficient. Of course hashing can be a very expensive operation, so in order for this approach to work, we need to be able to hash quickly on average. In fact, a simple hashing technique allows us to do so in constant time per operation!

The easiest way to picture the process is to think of the file as a sequence of digits, and the pattern as a number. Then we move a pointer in the file one character at a time, seeing if the next |P| digits gives us a number equal to the number corresponding to the pattern. Each time we read a character in the file, the number we are looking at changes is a natural way: the leftmost digit *a* is removed, and a new rightmost digit *b* is inserted. Hence, we update an old number *N* and obtain a new number *N'* by computing

$$N' = 10(N - a \cdot 10^{|P| - 1}) + b.$$

When dealing with a string, we will be reading characters (bytes) instead of numbers. Also, we will not want to keep the whole pattern as a number. If the pattern is large, then the corresponding number may be too large to do effective comparisons! Instead, we hash all numbers down into say 16 bits, by reducing them modulo some appropriate prime p. We then do all the mathematics (multiplication, addition) modulo p, i.e.,

$$N' = [10(N - a \cdot 10^{|P|-1}) + b] \mod p.$$

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P = 17935

p = 251

P \mod p = 114

D = 6386179357342...

63861 \mod p = 117

38617 \mod p = 214

86179 \mod p = 86

61793 \mod p = 47

17935 \mod p = 41

93573 \mod p = 41

93573 \mod p = 201

35734 \mod p = 92

57342 \mod p = 114

...
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Figure 1: A fingerprinting example. The pattern *P* is a 5 digit number. Note successive calculations take constant time: $38617 \mod p = ((63861 \mod p) - (60000 \mod p)) \cdot 10 + 7 \mod p$. Also note that false matches are possible (but unlikely); $57432 = 17935 \mod p$.

All operations mod *p* can be made quite efficient, so each new hash value takes only constant time to compute!

This pattern matching technique is often called fingerprinting. The idea is that the hash of the pattern creates an almost unique identifier for the pattern – like a fingerprint. If we ever find two fingerprints that match, we have a good reason to expect that they must come the same pattern. Of course, unlike real fingerprints, our hashing-based fingerprints do not actually uniquely identify a pattern, so we still need to check for false matches. But since false matches should be rare, the algorithm is very efficient!

One question remains. How should we choose the prime *p*? We would like the prime we choose to work well, in that it should have few false matches. The problem is that for every prime, there are certainly some bad patterns and documents. If we choose a prime in advance, then someone can try to set up a document and pattern that will cause a lot of false matches, making our fingerprinting algorithm go very slowly.

A natural approach is to choose the prime *p* randomly. This way, nobody

can set up a bad pattern and document in advance, since they are not sure what prime we will choose.

Let us make this a bit more rigorous. Let $\pi(x)$ represent the number of primes that are less than or equal to x. It will be helpful to use the following fact:

$$\frac{x}{\ln x} \le \pi(x) \le 1.26 \frac{x}{\ln x}.$$

Consider any point in the algorithm, where the pattern and document do not match. If our pattern has length |P|, then at that point we are comparing two numbers that are each less than 10|P|. In particular, their difference (in absolute value) is less than 10|P|. What is the probability that a random prime divides this difference? That is, what is the probability that for the random prime we choose, the two numbers corresponding to the pattern and the current |P| digits in the document are equal modulo p.

First, note that there are at most $\log_2 10|P|$ distinct primes that divide the difference, since the difference is at most 10|P| (in absolute value), and each distinct prime divisor is at least 2. Hence, if we choose our prime randomly from all primes up to Z, the probability we have a false match is at most

$$\frac{\log_2 10^{|P|}}{\pi(Z)}$$

Now, the probability that we have a false match anywhere is at most |D| times the probability that we have a false match in any single location, by the union bound. Hence the probability that we have a false match anywhere is at most

$$|D|\frac{\log_2 10^{|P|}}{\pi(Z)}.$$