# Efficient Pattern-Matching with Don't Cares

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### Abstract

We present a randomized algorithm for the *string matching* with don't cares problem. Based on the simple fingerprint method of Karp and Rabin for ordinary string matching [4], our algorithm runs in time  $O(n \log m)$  for a text of length n and a pattern of length m and is simpler and slightly faster than the previous algorithms [3, 5, 1].

#### 1 Introduction.

We extend the simple randomized fingerprinting algorithm of Karp and Rabin [4] to the problem of string matching with don't cares. Our algorithm uses a single, simple convolution. This is optimal in the sense that the string matching with don't cares problem is at least as hard as the boolean convolution problem [6]. Thus, to improve our run time of  $O(n \log m)$  on text of length n and pattern of length m, one would have to improve on the Fast Fourier Transform.

Fischer and Paterson's algorithm [1] runs in time  $O(n \log m \log |\Sigma|)^1$ . Since their deterministic algorithm in 1974, the only improvements were by Muthukrishan and Palem [5], who reduced the constant factor, and Indyk [3], who gave a randomized algorithm that also involved convolutions, running in time  $O(n \log n)$ . In addition to the small time improvement  $O(n \log n)$  over  $O(n \log n)$ , we hope the conceptually simple algorithm is also of interest. Determining the complexity of this problem was on a list of Galil's [2] as an open problem.

## 2 Problem and Solution

Let text  $X = x_1 x_2 \cdots x_n$ , with  $x_i \in \Sigma = \{1, 2, \dots, s\}$ , and a pattern  $Y = y_1 y_2 \cdots y_m$ , with  $y_i \in \Sigma \cup \{*\}$ . We think of \* as a "don't care." Then, we say that  $X(j) = x_j x_{j+1} \cdots x_{j-1+m}$  matches Y if

$$x_{j-1+i} = y_i \text{ or } y_i = *, \text{ for } 1 \le i \le m.$$

The following algorithm finds all positions j where X(j) matches Y.

Input: Text  $X = \overline{x_1 \dots x_n}$ , pattern  $Y = y_1 \dots y_m$ , and parameter N.

- 1. For  $1 \le i \le m$ , set  $r_i = \begin{cases} 0 \text{ if } y_i = * \\ \text{random from } \{1, 2, \dots, N\} \text{ otherwise.} \end{cases}$
- **2.** Compute  $t = \sum_{i=1}^{m} y_i r_i$ .
- **3.** For  $1 \le j \le n m$ , compute  $s(j) = \sum_{i=1}^{m} x_{j-1+i} r_i$  using FFT.
- **4.** Output MATCH for those j's where s(j) = t.

If X(j) matches Y, then the above algorithm will certainly output MATCH. However, if X(j) does not match Y, then for some i,  $y_i \neq x_{j-1+i}$  and  $y_i \neq *$ . Fixing the remaining variables, the equation s(j) = t then has a unique solution for  $r_i$ , which we will choose with probability at most 1/N. For say  $N = n^2$ , we would have at most a probability 1/n of having any false matches.

Runtime is  $O(n \log m)$  assuming we can do calculations of the size  $\log(N\Sigma n)$  in constant time, since Fast Fourier Transforms can be done in time  $O(n \log m)$ . All calculations could also be done modulo a prime p > N.

# 3 Don't Cares in Pattern and Text

The above algorithm can be extended to handle the case of don't cares in the text X as well. In this case, a \* in the text matches any symbol in the pattern. We use a second convolution to compute what s(j) should be for a match, which depends on j,

$$t(j) = \sum_{1 \le i \le m \mid x_{j-1+i} \ne *} y_i r_i.$$

This is an FFT of the vectors  $\langle y_i r_i \rangle_{i=1}^m$  with  $\langle \delta_{x_j \neq *} \rangle_{j=1}^n$ .  $(\delta_{x_j \neq *})$  is 1 if  $x_j \neq *$  and 0 if  $x_j = *$ .) If we replace \*'s with 0's for step 3, then simple calculation shows we will certainly have s(j) = t(j) if X(j) matches Y. And for the same reason as above, we will err with probability at most 1/N.

**Acknowledgements.** I thank Piotr Indyk and the anonymous SODA referees for helpful comments.

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<sup>&</sup>lt;sup>1</sup>Like Indyk [3], we use the RAM model. The run time given by Fischer and Patterson [1],  $O(n\log^2 m\log\log m\log|\Sigma|)$ , is slightly higher because they do not use the RAM model.

<sup>&</sup>lt;sup>2</sup>In personal communications, Indyk has indicated that his algorithm may also be improved to  $O(n \log m)$ .

## References

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