

## 1.1 Introduction

### 1.1.1 Motivation

This work has been motivated by recent advances of molecular genetics. The human genome has been sequenced in 2001. Also mouse, drosophila, etc. Nowadays # reference model genomes are available in genbank.

Next-generation sequencing has been the second revolution. NGS produces billions of reads for 1000\$ dollars. Why should one re-sequence a known genome? Resequencing applications include variant calling, etc. So NGS impacts biomedicine.

Given a set of reads, two approaches are possible: assembly and mapping.

Assembly methods are based on overlaps, de brujin graphs, or...

Read mapping methods work on a previously assembled reference genome.

The typical SNPs analysis pipeline ?? consists of...

In this work we focus on read mapping, although many core algorithms considered are also applicable to assembly, as well as to later pipeline stages.

*Figure 1.1: NGS pipeline.*

### 1.1.2 Problem definition

We define formally the basic objects alphabet and word. We call alphabet a finite set of symbols, we denote it as  $\Sigma$ . We call word a sequence of symbols over an alphabet. We denote a word  $w$  over  $\Sigma^*$ .

Define distance. Let  $d$  be a distance function over words, defined as  $d : (\Sigma^*, \Sigma^*) \rightarrow \mathcal{N}$ . Common distance metrics are the Hamming and Levenshtein or edit distance.

$$d_H =$$

$$d_E =$$

Define occurrence. We say that a word  $p$  of length  $m$  occurs in another word  $t$  at position  $e$  if  $t_{e-m,e} = p$ . The notion of occurrence can be generalized to metric distances...

Define approximate string matching. Given two words  $t$  and  $p$  such that  $t \gg p$ , respectively called text and pattern. Given a distance threshold  $k \in \mathcal{N}$ . The approximate string matching problem is to find all occurrences of  $p$  into  $t$  within the error threshold  $k$ .

These definitions allow us to model basic biological.

## 1.2 Overview of existing methods

Existing methods can be classified in three categories: online, indexed and filtering. Online and filtering in [?]. Indexed in [?].

### 1.2.1 Online methods

#### Dynamic Programming

The edit distance between two strings can be efficiently computed via dynamic programming.

The recurrence relation is as follows:  $D[i, j] = \dots$

Answering the question whether the distance  $d(x, y) \leq k$  is an easier problem: a band of size  $k + 1$  is sufficient.

Finding all occurrences of a pattern  $p$  in a text  $t$ ...

Change initialization.

Change traceback to start from all  $j : D[m, j] \leq k$ .

#### DP Bit-parallelism

#### Automata

#### NFA

### 1.2.2 Indexed methods

#### Suffix tree

#### Backtracking

### 1.2.3 Filtering methods

#### Why filtering

#### Pigeonhole principle

#### $q$ -Gram lemma

## 1.3 Related problems

### 1.3.1 Local similarity search

Define score and scoring scheme.

Define local similarity.

**Online methods**

Give dynamic programming solution.

**Indexed methods**

Backtracking over substring index. BWT-SW.

**Filtering methods**

SWIFT/Stellar is based on the q-gram lemma. Lastz resembles a suffix filter.

**1.3.2 Dictionary search**

Define problem.

**Trie****Backtracking****1.3.3 Overlaps computation**

Define problem.

DP solution.

Indexed solution, exact and approximate.



**2.1 Myers' bit-vector algorithm**

**2.2 Banded Myers' bit-vector algorithm**

**2.3 Increased bit-parallelism using SIMD instructions**



## **3.1 Classic Full-Text Indices**

### **3.1.1 Trie**

### **3.1.2 Suffix trie and suffix tree**

### **3.1.3 Suffix array**

### **3.1.4 $q$ -Gram index**

## **3.2 Compressed Full-Text Indices**

### **3.2.1 Burrows-Wheeler transform**

### **3.2.2 FM-index**

### **3.2.3 Rank dictionaries**

## **3.3 Backtracking**

### **3.3.1 Pruning methods**

### **3.3.2 Multiple backtracking**





**4.1  $q$ -Gram filters**

**4.1.1 Exact seeds**

**4.1.2 Gapped seeds**

**4.2 Factor filters**

**4.2.1 Exact seeds**

**4.2.2 Approximate seeds**

**4.3 Suffix filters**



**5.1 Related work**

**5.1.1 Best mappers**

**5.1.2 All mappers**

**5.2 The Masai mapper**

**5.2.1 Single-end mapping**

**5.2.2 Paired-end mapping**

**5.3 The Masai 2 mapper**

**5.3.1 Single-end mapping**

**5.3.2 Paired-end mapping**

**5.3.3 Parallelization**

**5.3.4 Hardware acceleration**

**5.4 Experimental results**

**5.4.1 Comparison of filtration strategies**

**5.4.2 Rabema benchmark results**

**5.4.3 Variant detection results**

**5.4.4 Runtime results**

**5.5 Discussion**



**6.1 The competition**

**6.2 Our method**

**6.2.1 Implementation**

**6.2.2 Parallelization**

**6.3 Related methods**

**6.4 Experimental results**

**6.4.1 Search**

**6.4.2 Join**

**6.5 Discussion**