# Nanopore automata

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## 1 Abstract

State machine algorithms for aligning Nanopore reads. Initial goal is simple reusable code for aligning a nanopore read to a reference sequence. No attempt at optimization yet.

# 2 Specification

## 2.1 Parameterization algorithm

Given the following inputs

- Reference genome (FASTA)
- Segment-called reads (FAST5/HDF5)

Perform the following steps

• Perform Baum-Welch to fit a rich model

Rich model incorporates segment statistics.

## 2.2 Reference search algorithm

Given the following inputs

- Reference genome
- Segment-called reads (FAST5/HDF5)
- Parameterized rich model

Perform the following steps

• Perform Viterbi alignment

## 2.3 Implementation

Libraries etc.

 $\mathrm{HDF}5...$ 

### 2.4 Evaluation

Strategy...

Data sets...

## 3 Methods

Model & inference algorithms.

#### 3.1 Model

- $\bullet$  Order-N transducer.
- Input: nucleotide
- Output: nucleotide, segment mean, duration
- Emissions:
  - categorical (base k-mer)
  - mixture of Normal/gamma (mean/duration)
- Transitions:
  - Match: emit single segment, absorb 1 base
  - Insert: affine gap insertion of bases: emits segments, absorbs no bases
  - Delete: affine gap deletion of bases: emits no segments, absorbs bases
  - Merge: emit single segment, absorb 2 or 3 bases

- Split: emit single segment, absorb 0 bases
- Skip: emit single segment, absorb  $2 \dots K$  bases (large K, low extension penalty)

This can be achieved by a Mealy transducer with  $4 \times 4^N$  states. The factor of  $4^N$  accounts for the order-N context. For each such context, the four states are MAT, INS, DEL and SKP.

Parameters:

- Gap opening & extension probabilities  $\lambda_{qo}$ ,  $\lambda_{qx}$
- Merge probability  $\lambda_{mo}$ , probability that it's a 3-merge is  $\lambda_{mx}$
- Split probability  $\lambda_s$
- Skip probability  $\lambda_{ko}$ , skip extension probability  $\lambda_{kx}$

In general the emissions are of the form

$$(y, m, d) \sim \text{CNE}(L)$$

where L is a "label" indexing the appropriate emission distribution

$$y \sim \text{Categorical}(\mathbf{p}_L)$$

$$m \sim \text{Normal}(\mu_L, \tau_L)$$

$$d \sim \text{Exponential}(\kappa_L)$$

where  $\Omega$  is the nucleotide alphabet,  $y \in \Omega$  is the nucleotide as decoded by the basecaller,  $m \in \Re$  is the mean segment current, and  $d \in \Re^+$  is the segment duration.

The transition table is as follows:

Transition	From	То	Weight	Input	Output
Match	MAT	MAT	$(1 - \lambda_{go})(1 - \lambda_{mo})(1 - \lambda_s)(1 - \lambda_{ko})$	$x \in \Omega$	$(y, m, d) \sim \text{CNE}(\text{match}, x, c)$
Insert	MAT	INS	$\lambda_{go}/2$	none	$(y, m, d) \sim \text{CNE(insert)}$
	INS	INS	$\lambda_{gx}$	none	$(y, m, d) \sim \text{CNE(insert)}$
	INS	MAT	$1 - \lambda_{gx}$	none	none
Delete	MAT	DEL	$\lambda_{go}/2$	$x\in \Omega$	none
	DEL	DEL	$\lambda_{gx}$	$x\in \Omega$	none
	DEL	MAT	$1 - \lambda_{gx}$	none	none
Merge	MAT	MAT	$\lambda_{mo}(1-\lambda_{mx})$	$x\in\Omega^2$	$(y, m, d) \sim \text{CNE}(\text{merge2}, x)$
	MAT	MAT	$\lambda_{mo}\lambda_{mx}$	$x\in\Omega^3$	$(y, m, d) \sim \text{CNE}(\text{merge3}, x)$
Split	MAT	MAT	$\lambda_s$	none	$(y, m, d) \sim \text{CNE}(\text{split}, x)$
Skip	MAT	SKP	$\lambda_{ko}$	$x\in \Omega$	none
	SKP	SKP	$\lambda_{kx}$	$x\in \Omega$	none
	SKP	MAT	$1 - \lambda_{kx}$	none	none

Here  $c \in \Omega^N$  is the input context.

# 3.2 Baum-Welch algorithm

# 3.3 Viterbi algorithm

- 4 Results
- 5 Discussion

# 6 Acknowledgments

# 7 Figure Legends

# 8 Appendix

## 8.1 Exponential distribution

$$x \sim \text{Exponential}(\kappa)$$
 
$$P(x|\kappa) = \kappa \exp(-\kappa x)$$
 
$$\text{E}[x] = \kappa^{-1}$$
 
$$\text{Var}[x] = \kappa^{-2}$$

Rate parameter  $\kappa$ .

### 8.2 Gamma distribution

$$\begin{array}{rcl} x & \sim & \operatorname{Gamma}(\alpha,\beta) \\ P(x|\alpha,\beta) & = & \frac{x^{\alpha-1}\beta^{\alpha}\exp(-x\beta)}{\Gamma(\alpha)} \\ & \operatorname{E}[x] & = & \alpha/\beta \end{array}$$
 
$$\operatorname{Var}[x] & = & \alpha/\beta^2$$

Shape parameter  $\alpha$ , rate parameter  $\beta$ .  $\Gamma()$  is the gamma function

$$\Gamma(\alpha) = \int_0^\infty z^{\alpha - 1} \exp(-z) dz$$

Note  $\Gamma(n) = (n-1)!$  for positive integer n.

### 8.3 Normal distribution

$$x \sim \text{Normal}(\mu, \tau)$$

Mean  $\mu$ , precision  $\tau$  (precision is reciprocal of variance).

$$P(x|\mu,\tau) = \sqrt{\frac{\tau}{2\pi}} \exp\left(-\frac{\tau}{2}(x-\mu)^2\right)$$