## Nanopore automata

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#### 1 Abstract

State machine algorithms for aligning Nanopore reads.

## 2 Specification

Initial goal (Preliminary Results) is simple reusable code for aligning a segmented nanopore read (with segment currents summarized) to a reference sequence.

Longer-term goals (Specific Aims) include

- quasi-hierarchical series of models for processed→raw data (raw, FAST5, FASTQ, FASTA)
- transducer intersection-style models for read-pair alignment
- systematic strategies for approximation/optimization algorithms, climbing the hierarchy (starting with k-mer or FM-index approaches)
- transducer intersection models for aligning reads from different sequencing technologies

#### 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

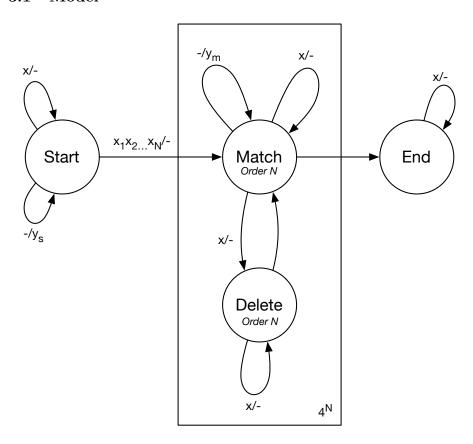
Strategy...

Data sets...

### 3 Methods

Model & inference algorithms.

#### 3.1 Model



- $\bullet$  Order-N Mealy transducer.
- Input alphabet:  $\Omega = \{A, C, G, T\}$  (nucleotides)
- $\bullet$  Output alphabet:  $\Re$  (real numbers signifying current levels)
- States: Start, End, { Match\_{x\_1...x\_N}, Delete\_{x\_1...x\_N}: x\_1...x\_N \in \Omega^N}
- Parameters:  $p^{\text{StartEmit}}, p^{\text{BeginDelete}}, p^{\text{ExtendDelete}}, \mu^{\text{Start}}, \tau^{\text{Start}},$   $\{p^{\text{MatchEmit}}_{x_1...x_N}, \mu^{\text{Match}}_{x_1...x_N}, \tau^{\text{Match}}_{x_1...x_N} : x_1 ... x_N \in \Omega^N\}$

Transitions:

TI WILDIUIOID.				
Source	Destination	Weight	Absorbs	Emits
Start	Start	$p^{ ext{StartEmit}}$		$y_s \sim \text{Normal}(\mu^{\text{Start}}, \tau^{\text{Start}})$
		$\times P(y_s \mu^{ ext{Start}},  au^{ ext{Start}})$		
Start	Start	1	$x\in\Omega$	
Start	$\mathrm{Match}_{x_1x_N}$	$1 - p^{\text{StartEmit}}$	$x_1 \dots x_N \in \Omega^N$	
$\mathrm{Match}_{x_1x_N}$	$\mathrm{Match}_{x_1x_N}$	$p_{x_1x_N}^{ ext{MatchEmit}}$		$y_m \sim \text{Normal}(\mu_{x_1x_N}^{\text{Match}}, \tau_{x_1x_N}^{\text{Match}})$
		$\times P(y_m \mu_{x_1x_N}^{\text{\tiny Match}},\tau_{x_1x_N}^{\text{\tiny Match}})$		
$\mathrm{Match}_{x_1x_N}$	$\mathrm{Match}_{x_2x_{N+1}}$	$(1-p_{x_1x_N}^{{}^{\mathrm{MatchEmit}}})$	$x_{N+1} \in \Omega$	
		$\times (1 - p^{\text{BeginDelete}})$		
$\mathrm{Match}_{x_1x_N}$	$\mathrm{Delete}_{x_2x_{N+1}}$	$(1-p_{x_1x_N}^{{}^{\mathrm{MatchEmit}}})$	$x_{N+1} \in \Omega$	
		$ imes p^{ ext{BeginDelete}}$		
$\mathrm{Match}_{x_1x_N}$	End	$1-p_{x_1\dots x_N}^{{}^{\scriptscriptstyle{\mathrm{MatchEmit}}}}$		
$\mathrm{Delete}_{x_1x_N}$	$\mathrm{Delete}_{x_2x_{N+1}}$	$p^{\scriptscriptstyle  m Extend Delete}$	$x_{N+1} \in \Omega$	
$\mathrm{Delete}_{x_1x_N}$	$\mathrm{Match}_{x_1x_N}$	$1 - p^{\text{ExtendDelete}}$		
End	End	1	$x \in \Omega$	

## 3.2 Baum-Welch algorithm

As usual.

## 3.3 Viterbi algorithm

As usual.

- 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).

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