# Indel rates and probabilistic alignments

**Gerton Lunter** 

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#### Alignment accuracy



Simulation:

Jukes-Cantor model Subs/indel rate = 7.5 Aligned with Viterbi + true model

CGACATTAA--ATAGGCATAGCAGGACCAGATACCAGATCAAAGGCTTCAGGCGCA CGACGTTAACGATTGGC---GCAGTATCAGATACCCGATCAAAG----CAGACGCA

CGACATTAA--ATAGGCATAGCAGGACCAGATACCAGATCAAAGGCTTCAGGCGCA CGACGTTAACGATTGGC---GCAGTATCAGATACCCGATCAAAG----CAGACGCA

Look at *inter-gap segments* Pr(length = L)?

CGACATTAA--ATAGGCATAGCAGGACCAGATACCAGATCAAAGGCTTCAGGCGCA CGACGTTAACGATTGGC---GCAGTATCAGATACCCGATCAAAG----CAGACGCA

*i i*+1

• Look at *inter-gap segments* 

Pr(length = L)?

**Def:**  $p_i = Pr(\text{ column } i+1 \text{ survived } | \text{ column } i \text{ survived})$ 

Assumption: indels are *independent* of each other

CGACATTAA--ATAGGCATAGCAGGACCAGATACCAGATCAAAGGCTTCAGGCGCA CGACGTTAACGATTGGC---GCAGTATCAGATACCCGATCAAAG----CAGACGCA

*i i*+1

• Look at *inter-gap segments* 

 $Pr(length = L) \propto p_i p_{i+1} \dots p_{i+L-2}$ 

**Def:**  $p_i = Pr(\text{ column } i+1 \text{ survived } | \text{ column } i \text{ survived})$ 

Assumption: indels are *independent* of each otherAssumption: indels occur *uniformly* across the genome

```
CGACATTAA--
ATAGGCATAGCAGGACCAGATACCAGATCAAAGGCTTCAGGCGCA
CGACGTTAACGATTGGC---GCAGTATCAGATACCCGATCAAAG----
CAGACGCA
```

```
i i+1
```

• Look at *inter-gap segments* 

 $Pr(length = L) \propto p^{L}$ 

**Def:**  $p_i = Pr(\text{ column } i+1 \text{ survived } | \text{ column } i \text{ survived})$ 

Assumption: indels are *independent* of each otherAssumption: indels occur *uniformly* across the genome

**Prediction:** Inter-gap distances follow a *geometric distribution* 

## Inter-gap distances in alignments



# Inter-gap distances in alignments (simulation)



#### **Biases in alignments**



A: gap wander (Holmes & Durbin, JCB 5 1998)

- **B,C**: gap attraction
- **D**: gap annihilation

# **Biases in alignments**

![](_page_10_Figure_1.jpeg)

#### Influence of alignment parameters

![](_page_11_Figure_1.jpeg)

- De-tuning of parameters away from "truth" does not improve alignments
- Accuracy of parameters (within ~ factor 2) does not hurt alignments much

#### Influence of model accuracy

![](_page_12_Figure_1.jpeg)

Improved model (for mammalian genomic DNA):

- Better modelling of indel length distribution
- Substitution model & indel rates depend on local GC content
- Additional variation in local substitution rate

Parameters: BlastZ alignments of human and mouse

#### Influence of model accuracy

![](_page_13_Figure_1.jpeg)

![](_page_13_Picture_2.jpeg)

#### Simulation:

- 20 GC categories
- 10 substitution rate categories
- 100 sequences each = 20.000 sequences
- Each ~800 nt, + 2x100 flanking sequence

## Summary so far

- Alignments are biased
  - Accuracy depends on position relative to gap
  - Fewer gaps than indels
- Alignments can be quite inaccurate
  - For 0.5 subs/site, 0.067 indels/site:
     accuracy = 65%, false positives = 15%
- Choice of parameters does not matter much
- Choice of MODEL does not matter much...

#### Alignments: Best scoring path

(Needleman-Wunsch, Smith-Waterman, Viterbi)

![](_page_15_Figure_2.jpeg)

#### Alignments: Posterior probabilities

(Durbin, Eddy, Krogh, Mitchison 1998)

![](_page_16_Figure_2.jpeg)

ACC GTT CACAATGGAT

#### Posterior probabilities

![](_page_17_Figure_1.jpeg)

![](_page_17_Figure_2.jpeg)

#### Posteriors: Good predictors of accuracy

![](_page_18_Figure_1.jpeg)

#### Posterior decoding: better than Max Likelihood

![](_page_19_Figure_1.jpeg)

...leading to lower 'asymptotic accuracy'...

![](_page_20_Figure_1.jpeg)

Posteriors & estimating indel rates

The inter-gap histogram slope estimates the indel rate, and is not affected by gap attraction...

![](_page_20_Figure_4.jpeg)

.. but is influenced by gap annihilation...

![](_page_20_Figure_6.jpeg)

...which cannot be observed – but posteriors can be...

![](_page_20_Figure_8.jpeg)

#### ...and they are identical in the mean:

![](_page_20_Picture_10.jpeg)

#### Indel rate estimators

![](_page_21_Figure_1.jpeg)

<b>Density:</b>	Alignment gaps per site
Inter-gap:	Slope of inter-gap histogram
BW:	Baum-Welch parameter estimate
Prob:	Inter-gap histogram with posterior probability correction

#### Human-mouse indel rate estimates

![](_page_22_Figure_1.jpeg)

#### Simulations: inferences are accurate

![](_page_23_Figure_1.jpeg)

#### Second summary

- Alignments are biased, and have errors
- Posterior accurately predicts local alignment quality
- Posterior decoding **improves** alignments, **reduces** biases
- With posterior decoding: modelling of indel lengths and sequence content **improves alignments**
- Indel rates (human-mouse) 60-100% higher than apparent from alignments

## Neutral indel model: Whole genome

![](_page_25_Figure_1.jpeg)

![](_page_26_Picture_0.jpeg)

#### Model: • Genome is mixture of "conserved" and "neutral" sequence

- "Conserved" sequence accepts no indel mutations
- "Neutral" sequence accepts any indel mutation
- Indels are point events (no spatial extent)

Account for "neutral overhang":

![](_page_26_Figure_6.jpeg)

Correction depends on level of clustering of conserved sequence:

- Low clustering: conserved segment is flanked by neutral overhang neutral contribution = 2 x average neutral distance between indels
- High clustering: indels "sample" neutral sequence neutral contribution = 1 x average neutral distance between indels

Lower bound: ~79 Mb, or ~2.6 % Upper bound: ~100 Mb, or ~3.25 %

# How much of our genome is under purifying selection?

![](_page_27_Picture_1.jpeg)

2.56 – 3.25% indel-conserved (79-100 Mb)

![](_page_27_Figure_3.jpeg)

![](_page_27_Picture_4.jpeg)

![](_page_27_Picture_5.jpeg)

#### Inferences are not biased by divergence

#### Inferred from data:

#### Simulation (100 Mb conserved)

![](_page_28_Figure_3.jpeg)

# Conclusions

- Alignment is an inference problem; don't ignore the uncertainties!
- Posterior decoding (heuristic) can be better than Viterbi (exact)
- Indel rates are high. Useful for identifying functional regions, since indels can be more disruptive of function than substitutions.
- Up to 10% of our genome may be functional, and a large proportion is rapidly turning over.