#### Stochastic Models for DNA Tandem Duplication

Farzad Farnoud, with M. Schwartz, J. Bruck

Jan 18, 2016, University of Washington

GCTCCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGGG GCTGCGTTACAGGTGGGCAGGGGAGGGG GCTCCGTTACAGGTGGGCAGGGGAGGGG GCTGCGTTACAGGTGGGCGGGGGGAGGCG GCTGCGTTACAGGTGGGCGGGGGGAGGC GCTGCGTTACAGGTGGGCGGGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGGG 

GCTGCGTTACAGGTGGGGCAGGGG GCTGCGTTACAGGTGGGCAGGGGA GCTCCGTTACAGGTGGGCAGGG GCTGCGTTACAGGTGGGCAGG GCTCCGTTACAGGTGGGCAGG GCTGCGTTACAGGTGGGCAG GCTGCGTTACAGGTGGGCGGGGGGA GCTGCGTTACAGGTGGGCGGGGGA **GCTCCGTTACAGGTGGGGC**GGGGG GCTGCGTTACAGGTGGGCGGGGGG GCTCCGTTACAGGTGGGGCGGGGGGA GCTGCGTTACAGGTGGGCGGGGGG GCTGCGTTACAGGTGGGCGG GCTGCGTTACAGGTGGGCGG GCTGCGTTACAGGTGGGCAGGGG GCTGCGTTACAGGTGGGCGGGGG



### Mutations



### Mutations



#### 8.7 million species



### Mutations





Data storage in *live* DNA

#### 8.7 million species



TGATGCA Point Mutation

TCATGCA

TGATGCA Point Mutation

TCATGCA



D





Ch1: 933,911–935,015

Ch1: 933,911–935,015

GCTCCGTTACAGGTGGGCAGGGGGGGGGGGGG GCTGCGTTACAGGTGGGCAGGGGAGGCG 

Ch1: 933,911–935,015

GCTCCGTTACAGGTGGGCAGGGGGGGGGGGGGG GCTGCGTTACAGGTGGGCAGGGGGGGGGGGGG GCTGCGTTACAGGTGGGCAGGGGAGGCG 

repeat unit length

GCTCCGTTACAGGTGGGCAGGGGGGGGGGGGG

Ch1: 933,911–935,015

GCTCCGTTACAGGTG GGTGGGCGGGGGGGGGG Point mutation (PM) GGTGGGCGGGGGGAGGCG GCTGCGTTACAGGTG GCTGCGTTACAGGTGGGCAGGGGGGGGGGGGG GCTCCGTTACAGGTGGGCGGGGGGGGGGGGGGG 

## Point mutations are in the same positions

GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTCCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG **GCTGCGTTACAGGTGGGCAGGGGAGGCG** 

GCTCCGTTACAGGTGGGCAGGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTCCGTTACAGGTGGGCCGGGGGGAGGCT GCTGCGTTACAGGTGGGCAGGGGAGGCG

- Tandem duplications (TD) of one or more repeat units
- Point mutations (PM)

Start from one repeat unit (*seed*). Random mutations:

- Tandem duplications (TD) of one or more repeat units
- Point mutations (PM)

ACGT

- Tandem duplications (TD) of one or more repeat units
- Point mutations (PM)



- Tandem duplications (TD) of one or more repeat units
- Point mutations (PM)



- Tandem duplications (TD) of one or more repeat units
- Point mutations (PM)



- Tandem duplications (TD) of one or more repeat units
- Point mutations (PM)



- Tandem duplications (TD) of one or more repeat units
- Point mutations (PM)



Start from one repeat unit (*seed*). Random mutations:

- Tandem duplications (TD) of one or more repeat units
- Point mutations (PM)

Parameters of the model:

- Prob. of PM
- Prob. of TDs of different lengths

Can we learn them?



Start from one repeat unit (*seed*). Random mutations:

- Tandem duplications (TD) of one or more repeat units
- Point mutations (PM)

#### Parameters of the model:

- Prob. of PM
- Prob. of TDs of different lengths

#### Can we learn them?

#### ACGT ACTT GCGT ACTT GCGT ACTT

ACGT ACTT GCGT ACTT GCGT ACTT

ACGT ACTT GCGT ACTT GCGT ACTT













TD1, 2 TD2, 2 PM

Maximum Parsimony Thought to be NP-hard [Gascuel et al., 2005]





# Given the final sequence, can we efficiently estimate the parameters?

GCTCCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG **GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTCCGTTACAGGTGGGCAGGGGAGGCG** GCTGCGTTACAGGTGGGCGGGGGGGGGGGCT **GCTGCGTTACAGGTGGGCAGGGGAGGCG** 

GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTCCGTTACAGGTGGGCAGGGGAGGCG **GCTGCGTTACAGGTGGGCAGGGGAGGCG** GCTCCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCGGGGGGAGGCG GCTGCGTTACAGGTGGGCGGGGGGAGGCG GCTCCGTTACAGGTGGGCCGGGGGGGAGGCG GCTCCGTTACAGGTGGGCCGGGGGGAGGCT GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTCCGTTACAGGTGGGCCGGGGGGAGGCG

# How to extract information from point mutations?

Autocorrelation function:

 $\star$   $r(\delta) = fraction of symbols at distance <math>\delta$  units that are the same



### Stochastic Approximation

\* Suppose a discrete random process *x* satisfies:

$$x_{n+1} - x_n = \frac{1}{n}(h(x_n) + M_{n+1})$$

for a Lipschitz function *h*, and martingale difference *M*.

 Then x<sub>n</sub> converges almost surely to a compact connected internally chain transitive invariant set of the ode

 $\dot{x}_t = h(x_t).$ 

# Stochastic Approximation for Autocorrelation



### Stochastic Approximation for Autocorrelation

\*  $r_n$ : autocorr. after *n* mutations

The stochastic approximation equation for r<sub>n</sub>:

$$\frac{d}{dt}r_t = Ar_t$$

*A*: a matrix that depends on the parameters: P(PM), P(TD1), P(TD2), ...

\* As n increases,  $r_n$  tends to a point in the null space of A

P(PM) = 0.250P(TD1) = 0.525P(TD2) = 0.225



δ

P(PM) = 0.250P(TD1) = 0.525P(TD2) = 0.225



P(PM) = 0.250P(TD1) = 0.525P(TD2) = 0.225



P(PM) = 0.250P(TD1) = 0.525P(TD2) = 0.225



δ

### Estimation Algorithm

GCTCCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTCCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTCCGTTACAGGTGGGCAGGGGAGGCG GCTCCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG 

1. Calculate autocorrelation *r* of *s*.

2. Find mutation probs such that the  $l_2$ -norm  $||Ar||_2$  is minimized.

### Estimation Algorithm

GCTCCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTCCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTCCGTTACAGGTGGGCAGGGGAGGCG GCTCCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG 



- 1. Calculate autocorrelation *r* of *s*.
- 2. Find mutation probs such that the  $l_2$ -norm  $||Ar||_2$  is minimized.

### Estimation Algorithm

GCTCCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTCCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTCCGTTACAGGTGGGCAGGGGAGGCG GCTCCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG GCTGCGTTACAGGTGGGCAGGGGAGGCG 





Start with a short random seed over {A,C,G,T}



- Start with a short random seed over {A,C,G,T}
- Choose the parameters
  q = (PM1, TD1, TD2, TD3)
  randomly

TGAATGT

#### $\mathbf{q} = (0.24, 0.33, 0.34, 0.09)$

- Start with a short random seed over {A,C,G,T}
- Choose the parameters
  q = (PM1, TD1, TD2, TD3)
  randomly
- Apply *n* random mutations

#### TGAATGT

$$\mathbf{q} = (0.24, 0.33, 0.34, 0.09)$$

#### 200 mutations: TGAATGTGCGT...

- Start with a short random seed over {A,C,G,T}
- Choose the parameters
  q = (PM1, TD1, TD2, TD3)
  randomly
- Apply *n* random mutations
- Estimate the parameters

$$\min_{\hat{\mathbf{q}}} \|Ar\|_2$$

#### TGAATGT

$$\mathbf{q} = (0.24, 0.33, 0.34, 0.09)$$

#### 200 mutations: TGAATGTGCGT...







Stochastic estimation algorithm (NP-Hard(?) combinatorial problem).

- Stochastic estimation algorithm (NP-Hard(?) combinatorial problem).
- Point mutation enables estimation of duplication lengths.

- Stochastic estimation algorithm (NP-Hard(?) combinatorial problem).
- Point mutation enables estimation of duplication lengths.



Capacity? Error-correcting codes?
 Synthetic biology algorithms using storage, performing encoding/decoding?

- Stochastic estimation algorithm (NP-Hard(?) combinatorial problem).
- Point mutation enables estimation of duplication lengths.



- Capacity? Error-correcting codes?
  Synthetic biology algorithms using storage, performing encoding/decoding?
- Application to phylogenetics (estimation of # mutations of each type)