Mutation Systems

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- Biological evolution proceeds by variation and selection
- A model of evolution of strings

Variation:

A function μ mapping a string to possible mutations Selection:

A function f deciding whether a string is fit

• Evolvability:

Can s evolve to t via stepwise mutations to fit strings?

- A mutation system (Σ, μ, f) has
 - an alphabet Σ
 - a mutator $\mu: \Sigma^* \to 2^{\Sigma^*}$ μ maps a string to the set of its mutations
 - a fitness function $f:\Sigma^* \to \{0,1\}$
 - f decides whether a string is fit (1) or not (0)

Let $S = (\Sigma, \mu, f)$ be a mutation system

•
$$s \rightarrow_{\mu} t$$
 if $t \in \mu(s)$

s can mutate to t in one step

s can mutate to t if
$$s \rightarrow^*_{\mu} t$$

s can evolve to t if $s \rightarrow^*_{S} t$

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Example S:

• Σ = {*a*, *b*, *c*}

• $\mu(s) =$ strings obtained by swapping adjacent symbols in s

• f(s) = 1 if no two adjacent symbols are equal

Mutation steps:

 $abcbc
ightarrow_{\mu} abccb
ightarrow_{\mu} acbcb
ightarrow_{\mu} cabcb$

But *abccb* is not fit!

Alternative evolution steps:

 $abcbc \rightarrow_S bacbc \rightarrow_S bcabc \rightarrow_S bcacb \rightarrow_S cbacb \rightarrow_S cabcb$

The Mutation Graph: $(\Sigma^*, \rightarrow_{\mu})$

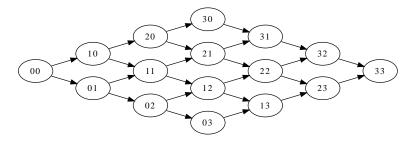
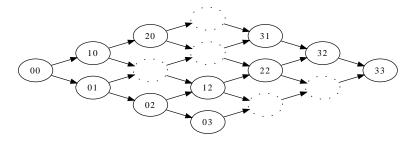


Image: A image: A

Nodes: all strings of length 2 over $\{0, 1, 2, 3\}$ Directed edges: \rightarrow_{μ}

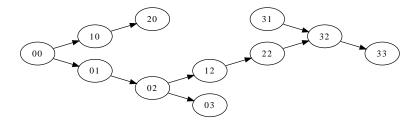
Unfit Strings are Removed



Remove unfit strings to get evolvability graph

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The Evolvability Graph: $(f^{-1}(1), \rightarrow_S)$



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Nodes: fit strings of length 2 over $\{0, 1, 2, 3\}$ Directed edges: \rightarrow_S The problem

Input: A mutation system and two strings s and t

Output: Can s evolve to t?

is undecidable for

 $\mu = {\rm point\ mutations}$

f = a strictly 2-testable predicate

A point mutation of s is obtained by

- replacing one symbol in s
- or deleting one symbol from s
- or inserting one symbol in s

So $\mu(bcb)$ contains

- acb, ccb, bab, bbb, bca, bcc
- cb, bb, bc
- abcb, bbcb, cbcb, bacb, bccb, bcab, bcbb, bcba, bcbc

Point mutations are reversible: the mutation graph is undirected

Strictly k-Testable Fitness Functions

A strictly k-testable L given by (PRE, MID, SUF)

- PRE contains strings of length k-1
- MID contains strings of length k
- SUF contains strings of length k-1

L contains all strings s such that

- length k 1 prefix of s in PRE
- every length k substring of s in MID
- length k 1 suffix of s in SUF

Fitness function $f_L(s) = 1$ iff $s \in L$

Fitness function f with

 $PRE = \{a, b\}$ MID = {aa, ac, bb, bd, cc, dd} SUF = {c, d}

has fit strings $a^+c^+ + b^+d^+$

How to control point mutations?

- Duplication map d(s) replaces each symbol x by x_1x_2
- Define a fitness function:

PRE contains all symbols x_1 MID contains all pairs of symbols x_1x_2 and y_2x_1 SUF contains all symbols x_2

- Fit strings are d(s)
- Point mutations of fit strings are unfit

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Evolvability \leftrightarrow computational reachability
      Issue of reversibility? Use computation histories
     Annotate symbol read with state (x if unread)
Example M
     \Sigma = \{a, b\}
     \delta(s) = \text{parity of } a's
Histories of M on input abaa
      Initial history: a_x b_x a_x a_x
      History after first step: a_1 b_x a_x a_x
      . . .
      Final history: a_1 b_1 a_0 a_1
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Duplicate history symbols: a_a^1, a_a^2
PRE: index 1, unread or correct transition from q_0
MID: indices 1,2
     main input symbols equal
     both unread or both read and states equal
     first read and second unread
MID: indices 2,1
     both unread
     first read and second unread
     both read and state transition to second correct
SUF: index 2, unread or read
```

Example of Mutations Simulating *M* on *abaa*

Initial history, all unread

$$a^1_{\scriptscriptstyle X} \quad a^2_{\scriptscriptstyle X} \quad b^1_{\scriptscriptstyle X} \quad b^2_{\scriptscriptstyle X} \quad a^1_{\scriptscriptstyle X} \quad a^2_{\scriptscriptstyle X} \quad a^1_{\scriptscriptstyle X} \quad a^2_{\scriptscriptstyle X}$$

First symbol read

 $a_1^1 \quad a_x^2 \quad b_x^1 \quad b_x^2 \quad a_x^1 \quad a_x^2 \quad a_x^1 \quad a_x^2$ Duplicate of first symbol updated $a_1^1 \quad a_1^2 \quad b_x^1 \quad b_x^2 \quad a_x^1 \quad a_x^2 \quad a_x^1 \quad a_x^2$ Second symbol read $a_1^1 \quad a_1^2 \quad b_1^1 \quad b_x^2 \quad a_x^1 \quad a_x^2 \quad a_x^1 \quad a_x^2$ Duplicate of second symbol updated

$$a_1^1 a_1^2 b_1^1 b_1^2 a_x^1 a_x^2 a_x^1$$

 a_{x}^{2}

A 1-dimensional reversible asynchronous cellular automaton:

An alphabet Σ Transition rules Substitutions: $axb \leftrightarrow ayb$ Insertions/Deletions: $axb \leftrightarrow ab$

Example:

Rules { $abc \leftrightarrow adc, dce \leftrightarrow dfe, fe \leftrightarrow fge$ } Reachable from abce are {abce, adce, adfe, adfge} From a cellular automaton C to a mutation system S:

Symbols ${}_{u}a_{v}^{i}$ main symbol component *a* from Σ index *i* from $\{1, 2, *\}$ left neighbor information *u* right neighbor information *v*

Rules from substitution and deletion/insertion rules of C

Application of a rule of C becomes a sequence of mutations:

Symbol "locks" its neighbors

Symbol then changes

Symbol "unlocks" its neighbors

Up to 14 mutations for 1 rule application

Starting with d(abcde):

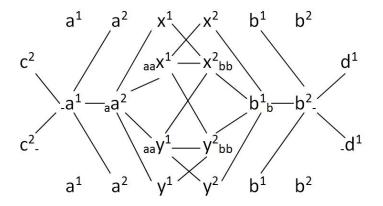
$$a^1 \cdot a^2 \cdot b^1 \cdot b^2 \cdot c^1 \cdot c^2 \cdot d^1 \cdot d^2 \cdot e^1 \cdot e^2$$

After several mutations:

$$a^1 \cdot a^2 \cdot _b^1 \cdot _b^2 \cdot _bb}c^1 \cdot c^2_{dd} \cdot d^1_d \cdot d^2_- \cdot e^1 \cdot e^2$$

symbol *c* has locked its left and right neighbors and is prepared for a rule application

Fitness Pairs for $axb \leftrightarrow ayb$

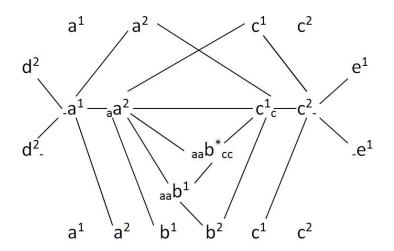


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Fitness Pairs for $abc \leftrightarrow ac$



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Thus the problem

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is undecidable for

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Random point mutations?

FSM simulation becomes a random walk: $O(n^3)$ steps Can be biased forward: $O(n^2)$ steps More generally?

Learnability?

Mutation process known & fitness function unknown? *k*-testable languages POS limit learnable [GV 1990] Also concatenations of *k*-testable languages [KY 1994] Stochastic results?

Thank you!

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