Improving an Exact Solution to the (I,d) Planted Motif Problem

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Introduction DNA motif finding

- motifs are repeated sub-sequences in DNA that have some biological significance
- DNA motif finding searches for motifs over a set of DNA sequences, allowing for mismatches due to mutation
- known as a difficult problem in computational biology and CS (proven NP-complete)

The (I,d) planted motif problem

Find a motif of length l=8 across these 5 DNA sequences, each containing the motif with at most d=2 mismatches.

```
S_1 at cactcgttctcctctaatgtgtaaagacgtactaccgacctta
```

 $S_2 \quad \mathtt{acgccgaccggtc} \\ \mathbf{cgatcctt} \\ \mathbf{gtatagctcctaacgggcatcagc}$

 S_4 ggccctcagcatcgtgcgtcctgctaacacattcccatgcagctt

 S_5 tgaaaagaatttacggtaaaggatccacatc ${ t caatcgtgt}$ gaaag

Planted motif: ccatcgtt



Solutions to the (I,d) planted motif problem)

There are two types of methods used by motif search algorithms:

- heuristic methods (ex. probabilistic sampling, projection) perform an iterative local search which is efficient, but not guaranteed to find all motifs
- exact methods (ex. combinatorial search, tree pruning)
 perform an exhaustive search which will find all possible motifs, at the cost of time/space efficiency

I-mers, Hamming distances, and *d*-neighborhoods

- ► /-mer
- ► Hamming distance d_H
- ► *d*-neighbor

I-mers, Hamming distances, and *d*-neighborhoods

- ► /-mer
 - sequence of length /

 $S_1 = \mathtt{atcactcgtt}$ ctcctctaatgtgtaaagacgtactaccgacctta

- ► Hamming distance d_H
- ► *d*-neighbor

I-mers, Hamming distances, and *d*-neighborhoods

- ► /-mer
- ► Hamming distance *d_H*
 - number of mismatches between I-mers x_1 and x_2

$$x_1 = \text{cgatcctt}$$
 $d_H(x_1, x_2) = 2$
 $x_2 = \text{ccatcgtt}$

► *d*-neighbor

I-mers, Hamming distances, and *d*-neighborhoods

```
▶ /-mer
\triangleright Hamming distance d_H
▶ d-neighbor
  - two I-mers x and x' are d-neighbors if d_H(x,x') < d
  N(\text{ccatcgtt}, 2) \rightarrow d\text{-neighborhood of ccatcgtt}, d=2
      = { ccatcgtt,
           acatcgtt,gcatcgtt,tcatcgtt,catcgtt,cgatcgtt,ctatcgtt,
           ...all /-mers with 1 mismatch
           aaatcgtt,agatcgtt,atatcgtt,gaatcgtt,ggatcgtt,gtatcgtt,
           taatcgtt,tgatcgtt,ttatcgtt,acctcgtt,acgtcgtt,acttcgtt,
           ...all /-mers with 2 mismatches
```

I-mers, Hamming distances, and *d*-neighborhoods

- ► /-mer
- ► Hamming distance d_H
- ► *d*-neighbor

EMS-GT Nabos, 2015

- ▶ an exact motif search (EMS) algorithm based on the candidate generate-and-test (GT) principle
- ▶ solves any (I,d) planted motif problem instance, $I \leq 17$
- operates on a bit-based representation of the search space

Generate-and-test approach

EMS-GT proceeds in two steps:

1. Generate the set C of candidate motifs: find the common d-neighbors of the first n' sequences $S_1, S_2, ..., S_{n'}$.

$$C = \mathcal{N}(S_1, d) \cap \mathcal{N}(S_2, d) \cap ... \cap \mathcal{N}(S_{n'}, d), \quad n' \leq n$$

2. Test every candidate $c \in C$: if a d-neighbor of c appears in each of the remaining sequences $S_{n'+1}, S_{n'+2}, ... S_n$, accept c as a motif.

Generate-and-test approach

$$(1,d) = (8,2)$$

- \mathcal{S}_1 atcactcgttctcctctaatgtgtaaagacgtactaccgacctta
- S_2 acgccgaccggtccgatccttgtatagctcctaacgggcatcagc
- S_3 tcctgactgcatcgcgatctcggtagtttcctgttcatcattttt

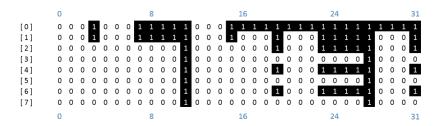
- S_4 ggccctcagcatcgtgcgtcctgctaacacattcccatgcagctt
- S_5 tgaaaagaatttacggtaaaggatccacatccaatcgtgtgaaag

Bit-based representation of the search space

- ▶ The search space contains all 4^I possible I-mers that can be formed with $\Sigma = \{a, c, g, t\}$.
- ➤ To represent sets in this space, EMS-GT assigns each of the 4¹ I-mers a bit flag, which is 1 if the I-mer is a member of the set, 0 otherwise.
- ► For efficiency, EMS-GT stores the 4¹ bits as $\frac{4^1}{32}$ 32-bit integers.

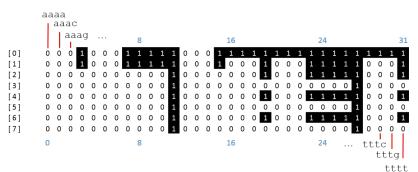
Bit-based representation of the search space

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: $4^{I} = 256$, $\frac{4^{I}}{32} = 8$



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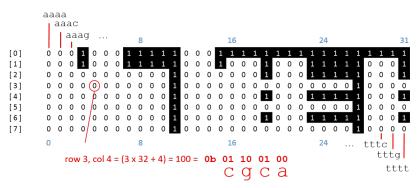
Bit-based representation of the search space

N(acgt, 1): $4^{1} = 256$, $\frac{4^{1}}{32} = 8$

```
aaaa
          aaac
                                                16
                                                                     24
[0]
[1]
[2]
[3]
[4]
[5]
[6]
[7]
                            8
                                                16
                                                                               tttc
                                                                                 ttta
            row 3, col 4 = (3 \times 32 + 4) = 100 = 0b 01 10 01 00
                                                                                   tttt
                                               cqca
```

Bit-based representation of the search space

► N(acgt, 1): $4^{I} = 256$, $\frac{4^{I}}{32} = 8$



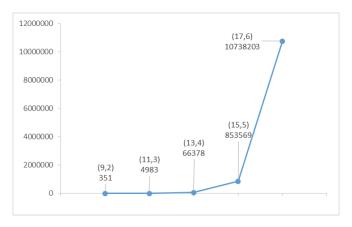
► EMS-GT generates a neighborhood bit-array by generating each individual neighbor, then finding and setting its bit flag.

Performance comparison

► EMS-GT is competitive against exact motif search algorithms PMSprune, qPMS7 and PMS8; however, for larger (*I*,*d*) values, it no longer outperforms the state-of-the-art.

```
img/emsgt.png
```

Neighborhood sizes for challenging (I,d)



▶ *I*-mer neighborhoods grow very quickly with (*I*,*d*), meaning that EMS-GT must spend more time locating and setting bits.

Research objectives Improving EMS-GT

The main objectives of this research are:

- To develop a speedup technique for EMS-GT that takes advantage of distance-related patterns in the search space;
- 2. To evaluate the speedup technique with regard to improvement in runtime; and
- 3. To evaluate the improved version of the EMS-GT algorithm against state-of-the-art motif search algorithms.

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the 4^k I-mers represented in a block all begin with the same prefix (first I − k characters), and will differ only in the k-suffix (last k characters);

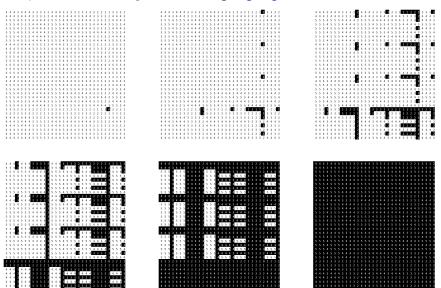
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- ▶ the 4^k I-mers represented in a block all begin with the same prefix (first I - k characters), and will differ only in the k-suffix (last k characters);
- ▶ and each block conforms to one of (k+2) patterns.

Block patterns in the d-neighborhood of acgtacgtacgt, d=5, k=5



Key observation

▶ In a *d*-neighbor of *x*, the *d* allowable mismatches from *x* are distributed between the prefix and the *k*-suffix.

$$x =$$
acgtacg tacgt $x' =$ acgaaaa tccga

▶ If a block's prefix already has p mismatches from x's prefix, then for an l-mer in the block to be a neighbor of x, it must have a suffix with at most d-p mismatches from x's suffix.

Key observation

▶ The (k + 2) block patterns thus correspond to k + 2 cases for the p mismatches between x's prefix and a block's prefix:

prefix mismatches	suffix mismatches
more than <i>d</i>	_
d	0
d - 1	0 or 1
d - 2	0, 1, or 2
d - (k-1)	0, 1, 2,, or (<i>k</i> -1)
d - k or less	0, 1, 2,, (k-1), or k

aaaaa \rightarrow

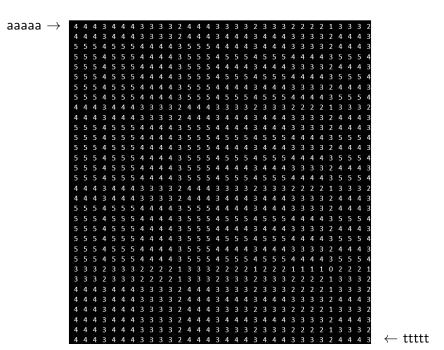
 \leftarrow ttttt

aaaaa \rightarrow 444344433332444333322333222213332 3 5 5 5 4 5 5 5 4 5 5 5 4 4 4 4 3 5 5 5 4 4 4 4 3 4 4 4 3 3 3 3 3 4 4 4 3 4 4 4 3 4 4 5 4 4 4 4 3 4 4 4 3 3 3 3 3 2 4 4 4 3 3 3 3 2 3 3 3 2 2 2 2 1 3 3 3 2 2 2 3 3 3 2 3 3 3 2 3 3 3 2 2 4 4 4 3 4 4 4 3 3 3 3 2 4 4 4 3 4 4 4 3 4 4 \leftarrow ttttt

aaaaa \rightarrow 44434443333244433332333222213332 4 4 4 3 4 4 4 3 3 3 3 3 4 4 4 3 4 4 4 3 4 4 4 3 3 3 3 3 4 4 4 3 3 3 3 2 3 3 3 2 2 2 2 1 3 3 3 2 2 2 2 1 1 1 3 3 3 2 3 3 3 2 2 2 2 1 3 3 3 2 3 3 3 2 3 3 3 2 2 2 \leftarrow ttttt

5 5 5 4 5 5 5 4 4 4 4 3 5 5 5 4 5 5 5 4 5 5 5 4 4 4 4 4 4 3 4 4 4 3 3 3 3 2 4 4 4 3 3 3 3 2 3 3 3 2 2 4 3 4 4 4 3 3 3 3 2 4 4 4 3 3 3 3 2 3 3 3 4 4 4 3 4 4 4 3 3 3 3 2 4 4 4 3 4 4 4 3 4 4 4 3 4 4 4 3 3 3 3 2 4 4 4 3 3 3 3 2 2 2 4 4 4 3 4 4 4 3 3 3 3 2 4 4 4 3 4 4 3 4 4 4 4 4 3 4 4 4 3 3 3 3 2 4 4 4 3 3 3 3 2 2 2 \leftarrow ttttt 5 4 4 4 4 3 5 5 5 5 5 4 5 5 2 2 2 1 3 3 4 4 3 3 3 3 \leftarrow ttttt aaaaa ightarrow \leftarrow ttttt aaaaa \rightarrow

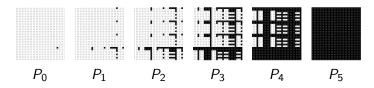




Generate and apply patterns

To generate N_x for x = yz, we perform two steps:

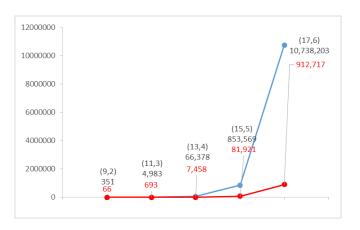
1. From x's suffix z, generate P, the set of block patterns.



2. From x's prefix y, recursively generate each d-neighbor y', and apply $P_{(d-d_H(y,y'))}$ to the block whose prefix is y'.

Results

Reduction in recursive neighbor generation



We still generate individual d-neighbors, but for x's prefix y; if we set k=5, y's neighborhood is only 10-20% the size of x's.

Results

Runtime comparison of PMS8, qPMS9, and EMS-GT

(I, d)	PMS8	qPMS9	EMS-GT	EMS-GT
				with speedup
9,2	0.74 s	0.47 s	0.06 s	0.11 s
11,3	1.58 s	1.06 s	0.22 s	0.20 s
13,4	5.39 s	4.52 s	1.98 s	1.04 s
15,5	36.45 s	24.63 s	25.06 s	15.51 s
17,6	3.91 min	1.96 min	5.14 min	2.93 min

Average runtime over 20 synthetic datasets per (I,d) on Intel Xeon 2.10 GHz processor (single core execution).

Results EMS-GT without vs. with the speedup technique

Runtime in seconds, averaged over 20 synthetic datasets per (I,d)

Results

PMS8 vs. qPMS9 vs. EMS-GT with speedup technique

Runtime in seconds, averaged over 20 synthetic datasets per (I,d)

Conclusions

We introduced a novel speedup technique for the EMS-GT algorithm, which takes advantage of suffix-related patterns in EMS-GT's bit-based representation of the motif search space.

Conclusions

➤ Our speedup technique reduces EMS-GT's runtime for most instances; however, for some (I,d) values our technique increases runtime due to overhead.

(I, d)	(9,2)	(11,3)	(13,4)	(15,5)	(17,6)
Runtime reduction	_	6.7%	47.5%	38.1%	43.0%

Conclusions

▶ EMS-GT with the speedup technique is highly competitive with the state-of-the-art: for all tested (*I*,*d*), it is faster than PMS8, and faster than qPMS9 except for (17,6).

(I, d)	(9,2)	(11,3)	(13,4)	(15,5)	(17,6)
RR from qPMS9	76.6%	81.1%	77.0%	37.0%	-
RR from PMS8	85.1%	87.3%	80.7%	57.6%	25.0%