Introduction to Bioinformatics

JTMS-19

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First examples of algorithms, probability models

What is this session about?

Boyer-Moore algorithm and suffix trees are introduced. The concept of a probability model is discussed and key types of probabilities are described: conditional, joint, marginal, posterior, prior probabilities and likelihood.

How can you revise the material after the session?

Read Durbin et al. chapter 1.3 Look at the two algorithms on

https://en.wikipedia.org/wiki/Boyer-Moore_string-search_algorithm

https://en.wikipedia.org/wiki/Suffix_tree

alternative reading: Hütt/Dehnert chapters 2.1 – 2.4

(A few comments on) Algorithms and data structures

Basic definitions

- String
- Graph
- Tree
- Path

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• Algorithm

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Run time of an algorithm

Examples:

operation on a sequence of numbers

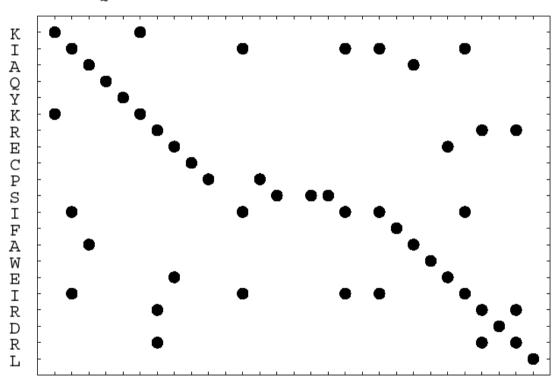
dot plot

paths on graphs

Run time of an algorithm

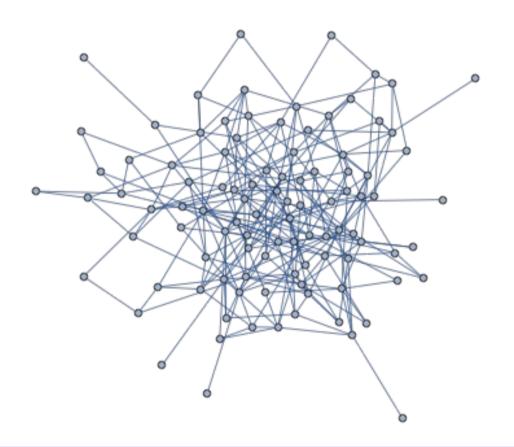
Example: dot plot

KIAQYKRECP**NIPSVSSIN**IFAWEIRDRL



Run time of an algorithm

Example: paths on a graph



n	n!
1	1
2	2
3	6
4	24
5	120
6	720
7	5040
8	40 320
9	362 880
10	3 628 800
11	39916800
12	479 001 600
13	6 227 020 800
14	87178291200
15	1 307 674 368 000

- target (text) sequence t
- pattern sequence p

Example of a naive algorithm:

```
Input: pattern p = p_1 p_2 ... p_m
text t = t_1 t_2 ... t_n
```

Algorithm:

```
\begin{split} I &= \{ \}, \\ \text{For } j = 0 \text{ to } n\text{--m do} \\ &\quad i = 1 \\ &\quad \text{While } p_i = t_j\text{+-1 and } i \text{-- m do} \\ &\quad i := i\text{+-1} \\ &\quad \text{If } i = m\text{+-1 then} \\ &\quad I := I \ U \ \{ j\text{+-1} \} \end{split} End(for)
```

- target (text) sequence t
- pattern sequence p

Naive algorithm:

- shifts p along t from left to right by single symbols
- compares p with t from left to right

Boyer-Moore algorithm:

- shifts by more than one symbol
- compares p with t from right to left
- preprocessing of p is needed for that!
- two rules for determining the shift size
 - bad character rule (BCR)
 - good suffix rule (GSR)

- target (text) sequence t
- pattern sequence p

Programming Techniques

G. Manacher, S.L. Graham

Editors

A Fast String Searching Algorithm

Robert S. Boyer Stanford Research Institute J Strother Moore Xerox Palo Alto Research Center

Communications of

the ACM

October 1977

Volume 20 Number 10

- target (text) sequence t
- pattern sequence p

Boyer-Moore algorithm: small example

• bad character rule (BCR)

$$t_{j+m} \leftrightarrow p_m$$
 $t_{j+m-1} \leftrightarrow p_{m-1}$
 $t_{j+m-2} \leftrightarrow p_{m-2}$
 $\rightarrow i = m-2$
 $\rightarrow store t_{j+m-2} = c$
 $\rightarrow shift to the rightmost c in p$

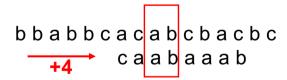
preprocessing of p:

for all s from the alphabet compute b(s), the rightmost position of s

- target (text) sequence t
- pattern sequence p

Boyer-Moore algorithm: small example

- good suffix rule (GSR)
 - $p_{m-1}p_m$: good suffix
 - \rightarrow shift to the rightmost occurrence of $p_{m-1}p_m$ in p

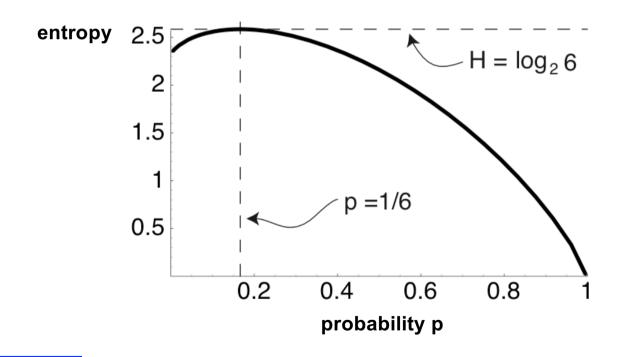


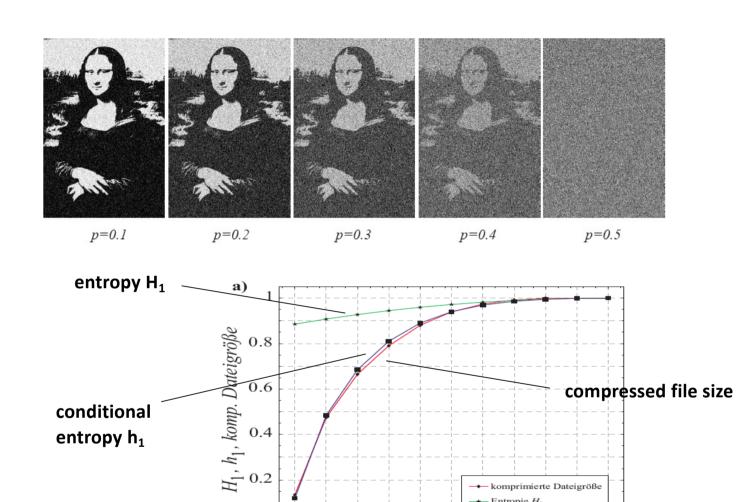
Brief digression:

What determines, if the Boyer -Moore algorithm runs fast?

→ entropy of a string

entropy
$$H = -\sum_{i \in \varSigma} p_i \, \log_2 p_i$$
 .





0.2

0.1

0

0

→ komprimierte Dateigröße

0.4

0.5

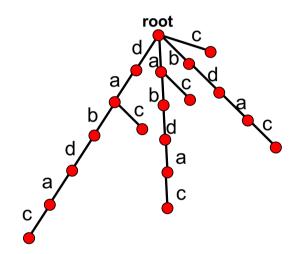
★ Entropie H₁ bedingte Entropie h₁

0.3

Parameter p

Suffix trees

- preprocessing of text t
- efficient storing of the suffix information of t
- (1) dabdac

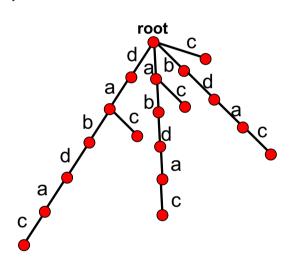


Suffix tree for $t = t_1 t_2 \dots t_n$

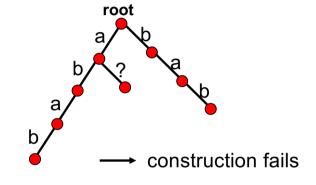
- n leaves
- edges in the tree are labeled from the alphabet
- all edges leaving a node carry different symbols
- path from root to leaf i has labels $t_i \; t_{i+1} \; ... \; t_n$

Suffix trees

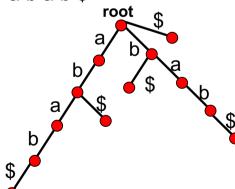
- preprocessing of text t
- efficient storing of the suffix information of t
- (1) dabdac



(2) abab



(3) abab\$



	From a res	search	article to data	a sets – a sa	ample pat	
Basic idea of algorithms, pro	bability models	•	Marc-Thorsten Hütt, Feli	x Jonas	•	IntroBioinfo – Session 2

Comprehensive analysis of CpG islands in human chromosomes 21 and 22

Daiva Takai* and Peter A. Jones

Department of Biochemistry and Molecular Biology, University of Southern California/Norris Comprehensive Cancer Center, Keck School of Medicine of the University of Southern California, 1441 Eastlake Avenue, Los Angeles, CA 90033

CpG islands are useful markers for genes in organisms containing 5-methylcytosine in their genomes. In addition, CpG islands located in the promoter regions of genes can play important roles in gene silencing during processes such as X-chromosome inactivation, imprinting, and silencing of intragenomic parasites. The generally accepted definition of what constitutes a CpG island was proposed in 1987 by Gardiner-Garden and Frommer [Gardiner-Garden, M. & Frommer, M. (1987) J. Mol. Biol. 196, 261–282] as being a 200-bp stretch of DNA with a C+G content of 50% and an observed CpG/expected CpG in excess of 0.6. Any definition of a CpG island is somewhat arbitrary, and this one, which was derived before the sequencing of mammalian genomes, will include many sequences that are not necessarily associated with controlling regions of genes but rather are associated with intragenomic parasites. We have therefore used the complete genomic sequences of human chromosomes 21 and 22 to examine the properties of CpG islands in different sequence classes by using a search algorithm that we have developed. Regions of DNA of greater than 500 bp with a G+C equal to or greater than 55% and observed CpG/expected CpG of 0.65 were more likely to be associated with the 5' regions of genes and this definition excluded most Alu-repetitive elements.

3740-3745 | PNAS | March 19, 2002 | vol. 99

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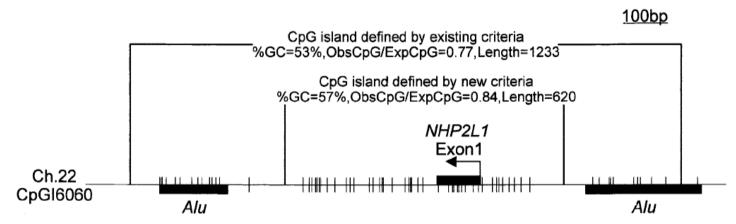
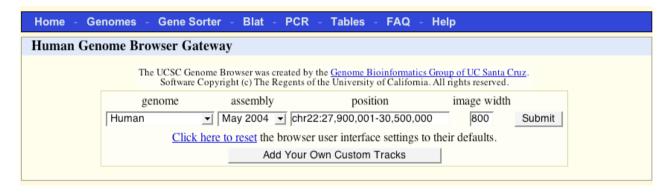
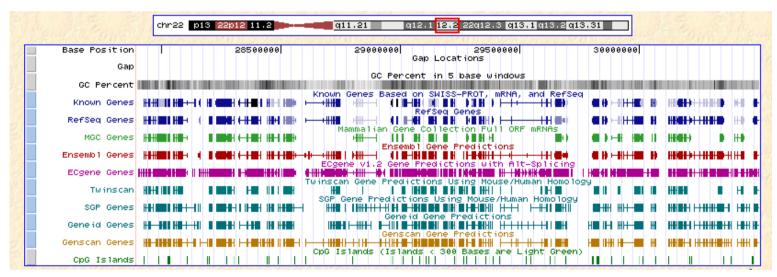


Fig. 3. The modified criteria also helped remove Alu sequences previously identified as part of 5' region CpG islands. In this example, a 1,233-bp fragment originally extracted by the algorithm included two Alu sequences with some CpG suppression associated with the nonhistone chromosome protein 2 like 1 (NHP2L1). The modified stringent criteria reduced the size of the island to 620 bp and excluded the Alu sequences.

```
CpG island searcher command line version
Ver 1.3 released 05/21/03
by Takai D. & Jones PA.
```

```
Selected lower limits: %GC=55, ObsCpG/ExpCpG=0.65, Length=500 human_ch, CpG island 1, start=83409, end=85437, %GC=64.6, ObsCpG/ExpCpG=0.888, Length=2029 human_ch, CpG island 2, start=199701, end=200609, %GC=55, ObsCpG/ExpCpG=0.756, Length=909 human_ch, CpG island 3, start=224154, end=224658, %GC=55.2, ObsCpG/ExpCpG=0.65, Length=505 human_ch, CpG island 4, start=261555, end=262391, %GC=56.8, ObsCpG/ExpCpG=0.65, Length=837 human_ch, CpG island 5, start=262849, end=263596, %GC=60.1, ObsCpG/ExpCpG=0.65, Length=748 human_ch, CpG island 6, start=353407, end=353937, %GC=63.6, ObsCpG/ExpCpG=0.652, Length=531 human_ch, CpG island 7, start=511685, end=512510, %GC=59.2, ObsCpG/ExpCpG=0.663, Length=826 human_ch, CpG island 8, start=924779, end=925798, %GC=61, ObsCpG/ExpCpG=0.672, Length=1020 human_ch, CpG island 9, start=981212, end=981828, %GC=58.1, ObsCpG/ExpCpG=0.651, Length=617 human_ch, CpG island 10, start=996188, end=996688, %GC=68, ObsCpG/ExpCpG=0.656, Length=501 human_ch, CpG island 11, start=1012831, end=1013415, %GC=63.4, ObsCpG/ExpCpG=0.651, Length=585 human_ch, CpG island 12, start=1020111, end=1021323, %GC=64.7, ObsCpG/ExpCpG=0.81, Length=1213
```

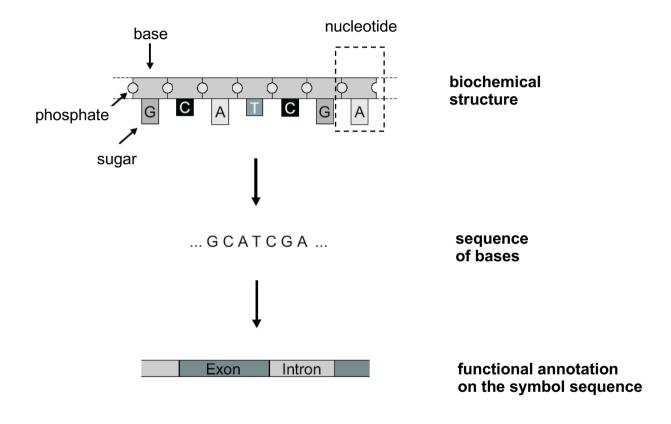




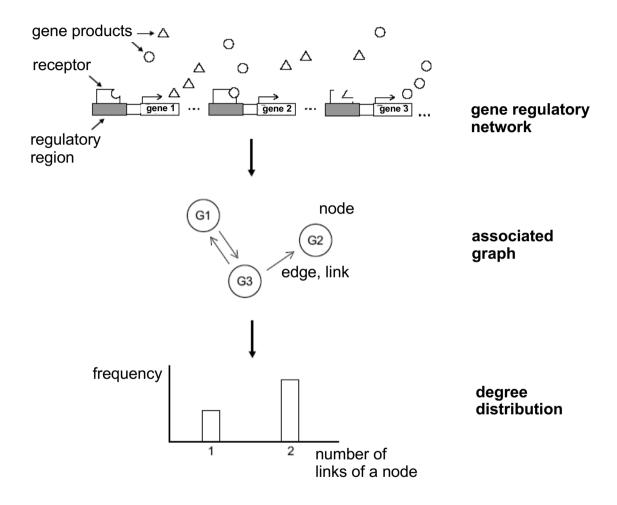
Proba	bi	litv	m	od	е	S
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•



structure → abstraction → analysis



structure → abstraction → analysis

How to quantify the match between data and a probability model?

