String searching algorithms: summary

Lecture 5.5
Biological sequence databases

- Databases contain a systematically organized data about:
  - genome sequences,
  - c-DNAs,
  - protein sequences,
  - promoters,
  - short motives,
  - repeats,
  - metabolic networks

- Search in these databases is the main application area of string searching algorithms
Example 1: Cancer

- A disease of multi-cellular organisms caused by unregulated abnormal cell proliferation

- 2 classes of genes are responsible for normal cell growth and division:
  - Growth factors which induce proliferation. Their enhanced activity causes cancer
  - Suppressor factors which inhibit proliferation of a cell. Their reduced activity causes cancer
Genes and cancer

Normal cell growth and division

Cell division involves the DNA replication, each 1 out of $10^6$ nucleotides mutates

The repair proteins fix somatic mutations

Growth factor

P53 – growth suppressor
Genes and cancer
Programmed cell death – extreme repair of somatic mutations

If the repair was unsuccessful, p53 causes cell death – apoptosis – removing it from the next cell division cycle.
Sequence similarity

viral oncogene – protein in viruses known to cause cancer
growth factor – protein in humans responsible for cell growth
Simian sarcoma viral oncogene v-sis sequence was similar to the PDGF growth factor sequence, as discovered by Doolittle from comparing sequences by eye.

This lead to a hypothesis that an oncogene is just a transformed growth factor.

The hypothesis proved to be correct: cancer is caused by mutation or displacement of a normal cellular regulatory gene.
Proto-oncogenes

- The normal growth-related genes were named *proto-oncogenes*

- The virus becomes oncogenic when it incorporates (cuts off by mistake?) the proto-oncogenetic sequence of PDGF. This sequence then mutates or it is moved by a virus close to a strong enhancer or away from the strong repressor.

- The uncontrollable cell growth causes cancer.
Genes and cancer

Mutation or misplacement of a growth factor gene

Growth factor with the enhanced function – moved under the control of a strong promoter

Mutation in a growth factor

Cells divide with an increasing speed, repair mechanisms are insufficient and cell division goes out of control – the tumor grows
Genes and cancer

Mutation in the apoptosis genes

Mutation in the apoptosis genes leads to the proliferation of the mutated cell

Continues – cell with somatic mutation

Continues
Computer programs discover new oncogenes searching for similarities in the genes database

<table>
<thead>
<tr>
<th>Oncogene</th>
<th>Normal gene</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sis</td>
<td>B chain of PDGF</td>
</tr>
<tr>
<td>Erb-B</td>
<td>truncated EGF receptor</td>
</tr>
<tr>
<td>Ras</td>
<td>small GTPase</td>
</tr>
<tr>
<td>Src</td>
<td>tyrosine kinase</td>
</tr>
<tr>
<td>E-myc</td>
<td>DNA-binding protein</td>
</tr>
</tbody>
</table>
Example 2: Large-scale discoveries

How to discover new proteins and their functions

- Extract m-RNAs from the cell
  - Reverse transcription into c-DNAs
  - Sequence c-DNAs

ACCGCACCA
GCCTACCAA
AATTTAACA

m-RNA
Reverse transcription into c-DNAs

ACCGCACCA
GCCTACCAA
AATTTAACA
Large-scale discoveries: pipeline

Protein sequences
- PAP
- GPA
- WTA

Translate into aminoacid sequences
- ACCGCACCA
- GCCTACCAA
- AATTTAACA

Protein-coding sequences

Search SWISS-PROT database for similar known proteins
- GPA -> GPA - actin
- GPA - myosin
- PPA – p53

Search Riley database for protein functions

Discovery
Large-scale discoveries

- Instead of extracting m-RNA, obtaining c-DNA and cloning, try to identify protein-coding regions from the genome sequence.
- Genes discovered this way are called putative\(^1\) genes.
- The database search on 1743 putative coding regions of the first sequenced organism Haemophilus Influenzae resulted in finding similar proteins and supposed function for 1000 of them.
- The discovery \textit{in silico}\(^1\) accepted by supposition.

\(^1\) accepted by supposition
Example 3: Multiple Sclerosis and database search

► MS is a lethal neurological disease of an *autoimmune* origin

Specific clone of T-cells is always produced in response to a foreign antigen

Sometimes specific T-cells start misbehave and give signals to destroy the organism’s own proteins – autoimmune disease

MS is caused by T-cells targeting the protein (myelin sheet receptor) of neurons
Multiple Sclerosis and database search

- The specific T-cells that pseudo-recognize myelin receptors were found and isolated

- These T-cells were previously mobilized by an unknown foreign protein

- We can try to search for a protein similar to the myelin sheet receptor in a database of viral or bacterial proteins, to find this mobilazer
Multiple Sclerosis and database search

► The search was conducted and about 100 candidate proteins were found to be closely similar to the myelin receptor

► Then it was confirmed that the proteins found by the database search are indeed recognized by the specific MS T-cells

► If we further study what is common to all these proteins, we can better understand what particular features of the receptor lead to be mistakenly recognized as a foreign protein
Example 4: BRCA1 and granins

- The database search requires a biological and statistical expertise to interpret the results.

- The hereditary predisposition for the breast and ovarian cancer is due to the mutation in 2 genes, BRCA1 and BRCA2, which have both been precisely located, cloned and sequenced.

- But how these genes are involved in the development of a cancer?
Hypothesis: The protein coded by BRCA1 is similar to granins

- The database search reported a substring (of length 10) of BRCA1 which almost perfectly matched the consensus sequence of the family of secretory extracellular proteins called granins, 70-80 aminoacids in length.

- The more successful treatment strategies can be applied to the extracellular than to the intracellular proteins. The discovery that BRCA1 can be an extracellular protein lead to optimistic speculations.
The search was statistically not reliable

- Originally, researchers have reported that the probability of finding the similarity of the BRCA1-coded protein and granins by chance is 0.00175.

- They concluded that the resulting search reflected an important biological phenomenon.

- But the correct question should have been: what is the probability that any 10-letters long motif from the PROSITE database, will occur in BRCA1 by chance?

- That probability was reported to be 0.87 due to 3 factors:
  - There are about 1000 motifs
  - The granin motif is only 10 letters long, and 3 positions are wild cards
  - The BRCA1 sequence is rather long (about 1,800 residues)
The correct search strategy for BRCA1

Later, the researchers identified 6 regions in BRCA1 corresponding to the globular structure.

By matching each of these 6 regions separately to the protein database, they found a moderately conserved 202-residue long region in a human protein 53BP1, known to bind to the universal tumor suppressor p53.

The BRCA1 gene is associated with a well established cancer agent and not with a granin.
Do we need to develop new algorithms for the sequence database search

- The string searching tools presented in this part of the course work just fine for the similarity search in the database.

- The important details such as:
  - local vs global alignment,
  - gap model,
  - scoring matrix,
  - threshold for reporting similarity,
  - statistical significance of obtained results

are *modeling and statistical* rather than *algorithmic* issues.
Algorithmic future of the biological sequence comparison

► Specialized algorithmic strategy for solving specific problems, for example
  ▪ Finding known motives
  ▪ Approximate pattern discovery from a set of biosequences
  ▪ Finding semi-characterized features in the query string

► More sensitive, selective and efficient algorithms can be developed for each particular problem
Summary of tools
Exact matching 1

- Linear-time exact pattern search based on a pre-processing of a pattern

Diagram showing a sequence of characters and a matching process.
Summary of tools

Exact matching 2

Exact pattern search using suffix trees

T = cacao
Summary of tools
Exact matching 3

Finding repetitions using suffix trees

parent of MR

bananas

mark a

mark MR

MR

1ε

2b

4n

6n

7a

5a

3a

nas

na

a

s

s

s

s

s

s

b a n a n a s
1 2 3 4 5 6 7
Summary of tools
Exact matching 3

Finding common substrings using suffix trees
Summary of tools
Approximate matching 1

Dynamic programming with different scoring matrices

```
algorithm cheepestCost ( array diagonalCost, N, M )
    return cost ( N, M )

algorithm cost ( i, j )
    if i=0 then
        return j
    if j=0 then
        return i
    return min ( cost ( i-1, j ) +1, cost ( i, j-1)+1, cost ( i-1, j-1)+diagonalCost [i] [j] )
```
Summary of tools
Approximate matching 1

- Dynamic programming with different scoring matrices

```python
algorithm cheepestCost ( array diagonalCost, N, M )

    return cost ( N, M )

algorithm cost ( i, j )

    if i=0 then
        return j
    if j=0 then
        return i
    return min ( cost ( i-1, j ) +1, cost ( i, j-1)+1, cost ( i-1, j-1)+diagonalCost [i] [j] )
```
Summary of tools
Approximate matching 2

► Divide and conquer – dynamic programming in linear space

```
 a t c a t g

 a  
 c  
 a 3 2 2 1 2 3
 3  
 2  
 1  
 6 5 4 3 2 1 0
 a t c a t g
```
Summary of tools
Approximate matching 3

- A threshold search around the main diagonal

<table>
<thead>
<tr>
<th></th>
<th>S2</th>
<th>a</th>
<th>t</th>
<th>c</th>
<th>a</th>
<th>c</th>
<th>g</th>
</tr>
</thead>
<tbody>
<tr>
<td>S1</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>a</td>
<td>0</td>
<td>1</td>
<td>2</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>c</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>2</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>a</td>
<td>2</td>
<td>2</td>
<td>2</td>
<td>1</td>
<td>2</td>
<td></td>
<td></td>
</tr>
<tr>
<td>t</td>
<td>2</td>
<td>3</td>
<td>2</td>
<td>2</td>
<td>2</td>
<td>3</td>
<td>3</td>
</tr>
<tr>
<td>a</td>
<td>3</td>
<td>3</td>
<td>3</td>
<td>3</td>
<td>3</td>
<td>3</td>
<td>3</td>
</tr>
<tr>
<td>g</td>
<td>3</td>
<td>3</td>
<td>3</td>
<td>3</td>
<td>3</td>
<td>3</td>
<td>3</td>
</tr>
</tbody>
</table>

Find a pattern similar to atcagc with an additional threshold – not more than 2 errors

Compute dynamic programming table values around in a 2*2+1=5 cells strip around the main diagonal.

If all values in the row are >2, abort the computation.
Some problem solving

Each of these problems can be used for a final project of type 3: original research
1. Comparing very similar sequences to reveal polymorphism

- Biological data: DNA sequences of 2 genomes of different E. Coli strains:
  - 1 can process cellulose
  - 1 cannot process cellulose

- How to locate putative genes responsible for the processing of the cellulose?
2. Recognizing DNA contamination

► 70s – the dinosaur DNA was sequenced and through the database search was found that the dinosaur DNA contains the sequences which are more similar to the mammals than to the amphibians !!!

► This is a general problem of *DNA contamination*

► If you have the databases of sequenced genomes for main biological groups, how would you check for a possible contamination?
3. Circular string linearization

Given an ordering of characters in the alphabet, a string $S_1$ is lexicographically smaller than a string $S_2$ if $S_1$ would appear before $S_2$ in a normal dictionary for these strings.

A circular string linearization problem is: find the place where to cut string $S$ of $M$ characters such that the resulting string would be lexicographically the smallest among all possible cuts of $S$. 
3. Circular string linearization

► This can be used, for example, for the canonical representation of circular bacterial genomes

► Try on the following input: A C A C T

Lexicographic ordering: A < C < T
4. Selecting a unique substring of length 26 for each of 100 DNAs

- This can be used for instance in the DNA microarrays to reduce the cross-binding in case that the same representative substring has been chosen for different genes
5. Finding maximal tandem repeats

► A tandem repeat is a sequence of two or more DNA bases that is repeated numerous times in a head-to-tail manner on a chromosome.

► In some cases, tandem repeats can serve as genetic markers to track inheritance in families.

► How to discover maximal tandem repeats? Consider a hashing versus the suffix tree solution.