

Introduction to RNA Bioinformatics

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Objectives

- Understanding the limitations of traditional bioinformatics tools w.r.t. RNA molecules
- Providing an overview of the bioinformatics tools that are specific to RNA research



Bioinformatics

- **Database search**, in the form of sequence comparison, is the workhorse of bioinformatics
- “Basic Local Alignment Search Tool (**BLAST**) is one of the most heavily used sequence analysis tools available in the public domain”
- In 2004, on average, NCBI was running **140,000 blast runs per weekday**, on a farm consisting of 200 CPUs (running Linux)
- In 2008, “BLAST is the most popular bioinformatics tool and is used to run millions of queries each day”



Database search

Find all GenBank gene's that are similar to *Clostridium botulinum's* toxin gene



```
>gi|27867582(fragment of the known Clostridium botuninum toxin gene)
GTGAATCAGCACCTGGACTTTCAGATGAAAAATTAAATTTAACTATCCAAAATGATGCTT
ATATACCAAATATGATTCTAATGGAACAAGTGATATAGAACAACATGATGTTAATGAAC
TTAATGTATTTTTCTATTTAGATGCACAGAAAGTGCCCGAAGGTGAAAATAATGTCAATC
TCACCTCTTCAATTGATACAGCATTATTAGAACAACCTAAAATATATACATTTTTTTTCAT
CAGAATTTATTAATAATGTCAATAAACCTGTGCAAGCAGC
```

Result of a database search

>[gi|49138|emb|X68262.1|CBBONTF](#) C.barati gene for type F neurotoxin

Length=4073 Score = 81.8 bits (41), Expect = 1e-12
Identities = 99/121 (82.82%), Gaps = 2/121 (0.02%)
Strand=Plus/Plus

```
Query 48 CAAAATGATGCTTATATAACCAAATATGATTCTAATGGAACAAGTGATATAGAACAACAT 107
      ||| ||| | ||| ||| ||| ||| ||| ||| ||| ||| ||| ||| ||| ||| |||
Sbjct 1712 CAAAATGATTCTTACGTTCCAAAATATGATTCTAATGGTACAAGTGAAATAAA-GAATAT 1771

Query 108 GATGTTAATGAACTTAATGTATTTTTCTATTTAGATGCACAGAAAGTGCC-GAAGGTGAA 167
      ||| || ||| ||| ||| ||| ||| ||| ||| ||| ||| ||| ||| ||| |||
Sbjct 1772 ACTGTTGATAAACTAAATGTATTTTTCTATTTATATGCACAAAAGCTCCTGAAGGTGAA 1831

Query 168 A 168 |
Sbjct 1832 A 1832
```

...



How does it work?

Pairwise Sequence Alignment (Algorithm)

- An optimal alignment is obtained by extending:
 - An optimal alignment with one more residue from each sequence (**match** or mismatch);
 - An optimal alignment with one residue from the first sequence and a gap symbol (**deletion**);
 - An optimal alignment with one residue from the second sequence and a gap symbol (**insertion**).



Algorithm

Alignment cost $\text{aln}(\text{ATATAGAACAC, AATAAAGGAT })$ is

The maximum of:

$\text{aln}(\text{ATATAGAACAA, AATAAAGGAA}) + \text{cost of substituting } \underline{\text{C}} \text{ by } \underline{\text{T}}$

ATATAGAACAA C
AATAAAGGAA T

$\text{aln}(\text{ATATAGAACAA, AATAAAGGAAT}) + \text{cost of deleting } \underline{\text{C}}$

ATATAGAACAA C
AATAAAGGAAT -

$\text{aln}(\text{ATATAGAACAAC, AATAAAGGA}) + \text{cost of inserting } \underline{\text{T}}$

ATATAGAACAAC -
AATAAAGGAA T

Molecular Sequence Alignment Assumptions

- *i.i.d.*
- Positions along the sequence are **independent and identically distributed**
- Independence is necessary for the development of efficient exact algorithms (Smith-Waterman) or heuristics (such as BLAST)
- The **execution time** of the exact algorithms grows proportionally to the **product** of the **size of the database times the size of input sequence**



RNA Sequence Alignment

```
1  GUCGAGAGAC
   * * * * *
2  GUCGAAGCUG
   * * * * *
3  CAGAGAGCUG
```

1 and 2 are 50% identical (similarly for 2 and 3),
however, 1 and 3 don't seem to have anything in common

G A
A G
G-C
A-U
C-G

A A
G G
C C
U U
G G

A G
G A
C-G
U-A
G-C

CAGAGAGCUG
1

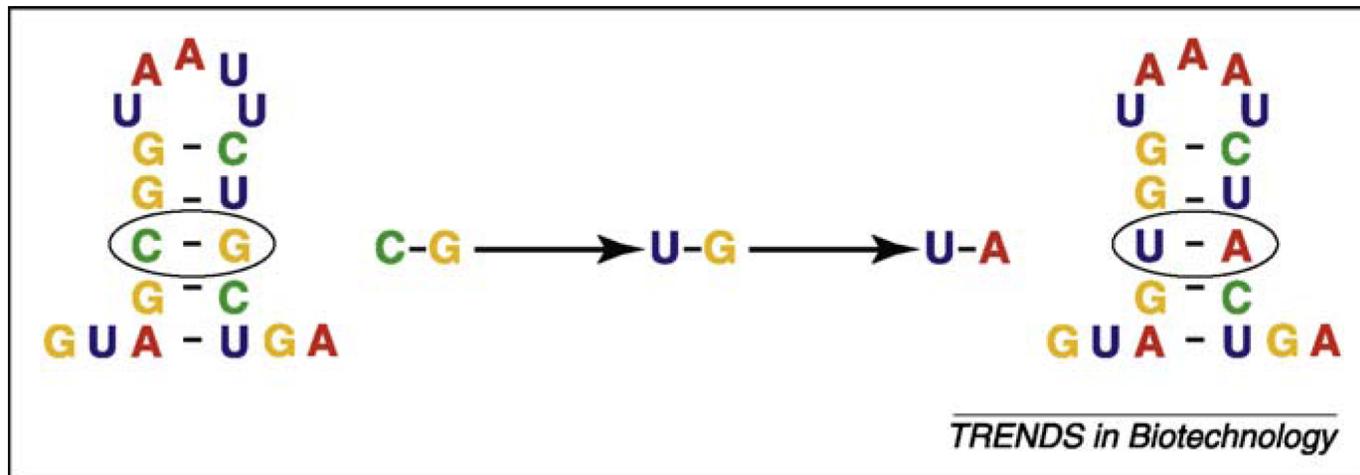
GUCGAAGCUG
2

GUCGAGAGAC
3

Yes, but sequences 1 and 3 share the same secondary structure!

Caveat

- RNAs conserve secondary structure interactions more than they conserve their sequence
- Traditional bioinformatics tools, assuming that positions are independent, perform poorly





Paradigms

1. Inference
2. Searching

Bias

- **Secondary structure** plays an important role in the elements that are sought

Time and space complexity

- Should we worry about the time and space complexity of the methods?
- After all, we can always buy a faster computer, right?
- Computer scientists use mathematical approaches to analyze the execution time and memory requirements



Time and space complexity

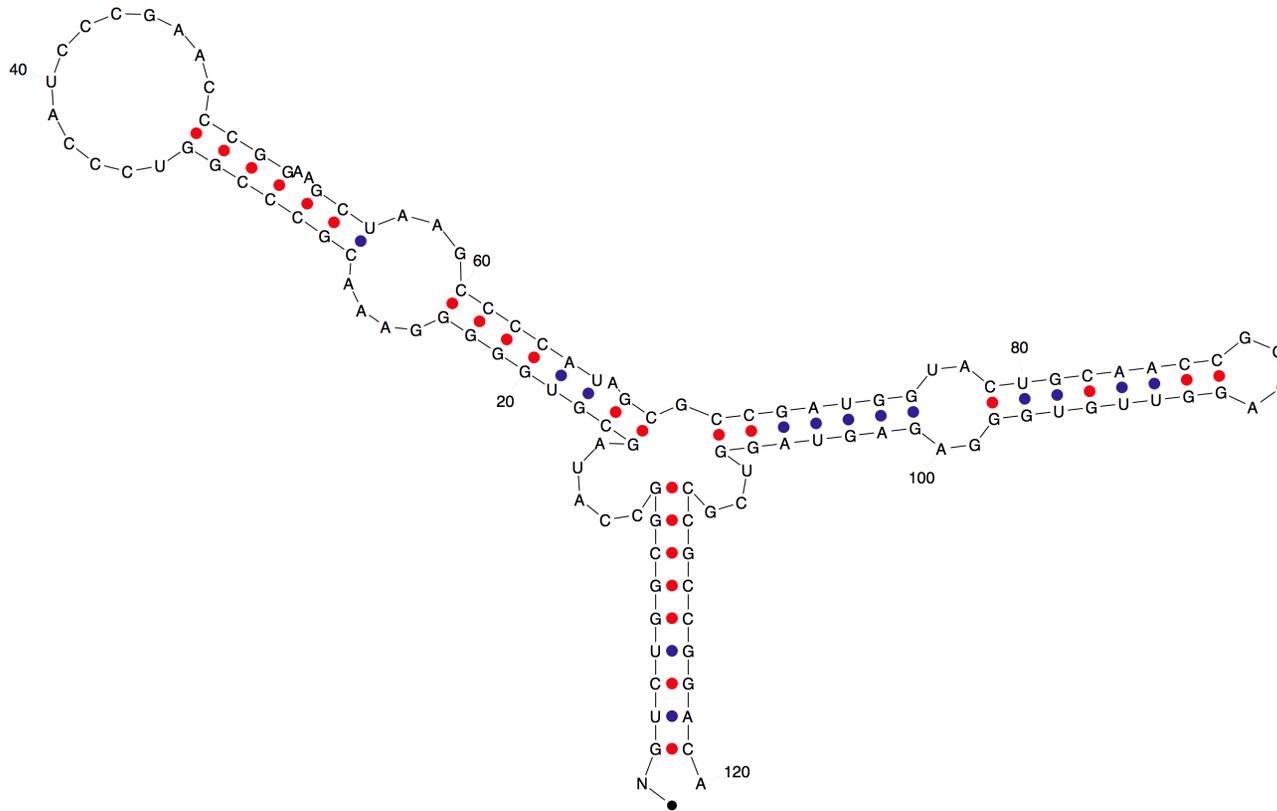
- Some algorithms require a **linear** amount of resources
- Some require **polynomial** amounts of resources
- Some always require **exponential** resources, these are **NP-hard**





Part I: Inference

Stems, hairpins, interior loops, bulges, and multi-branch loops



Definitions

Given an RNA **sequence** $S = s_1, s_2, \dots, s_n$ where s_i is the i^{th} nucleotide.

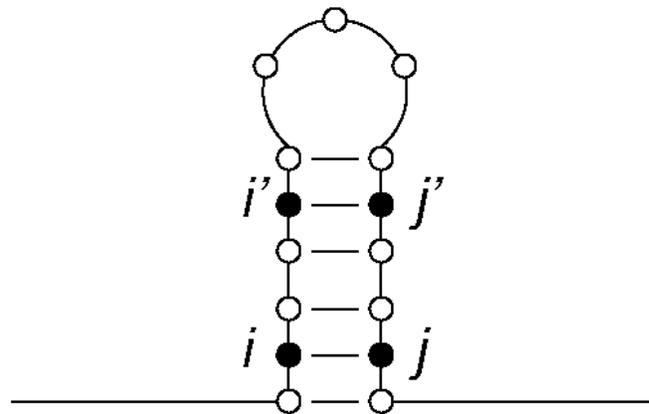
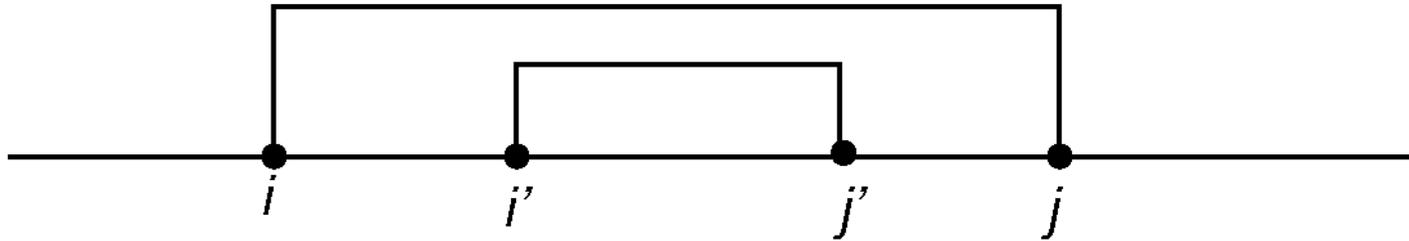
A **secondary structure** is an ordered list of pairs, $i.j$,

$1 \leq i < j \leq n$ such that:

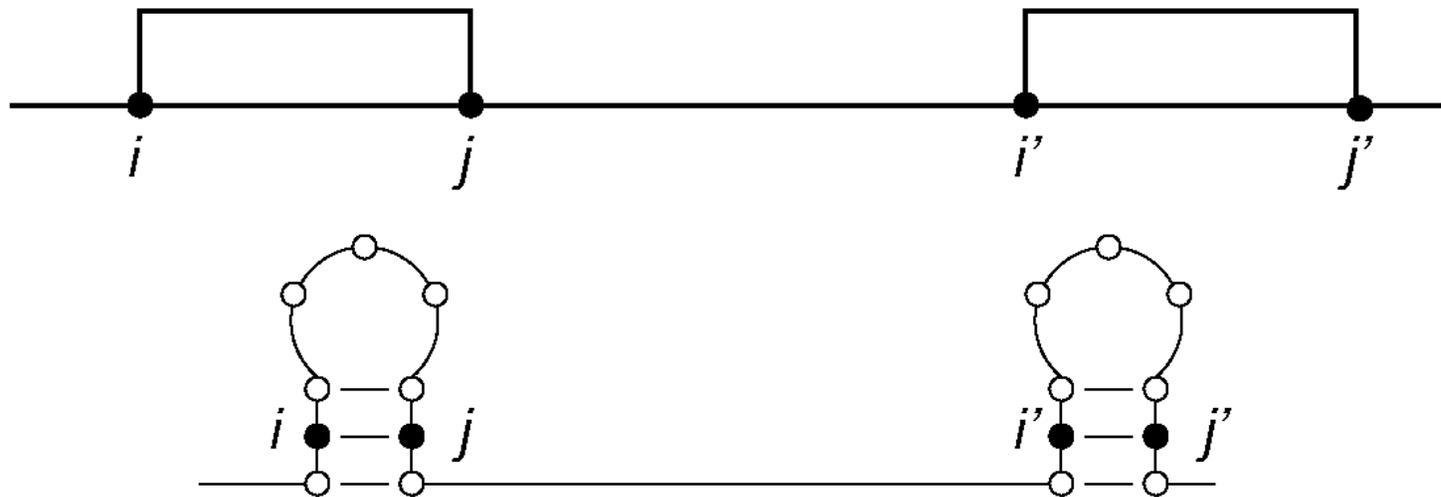
- $j - i \geq 4$
- Given $i.j$ and $i'.j'$, two base pairs, then either:
 - $i = i'$ and $j = j'$ (they are the same)
 - $i < j < i' < j'$ ($i.j$ precedes $i'.j'$)
 - $i < i' < j' < j$ ($i.j$ includes $i'.j'$)
 - $i < i' < j < j'$ (pseudoknot)



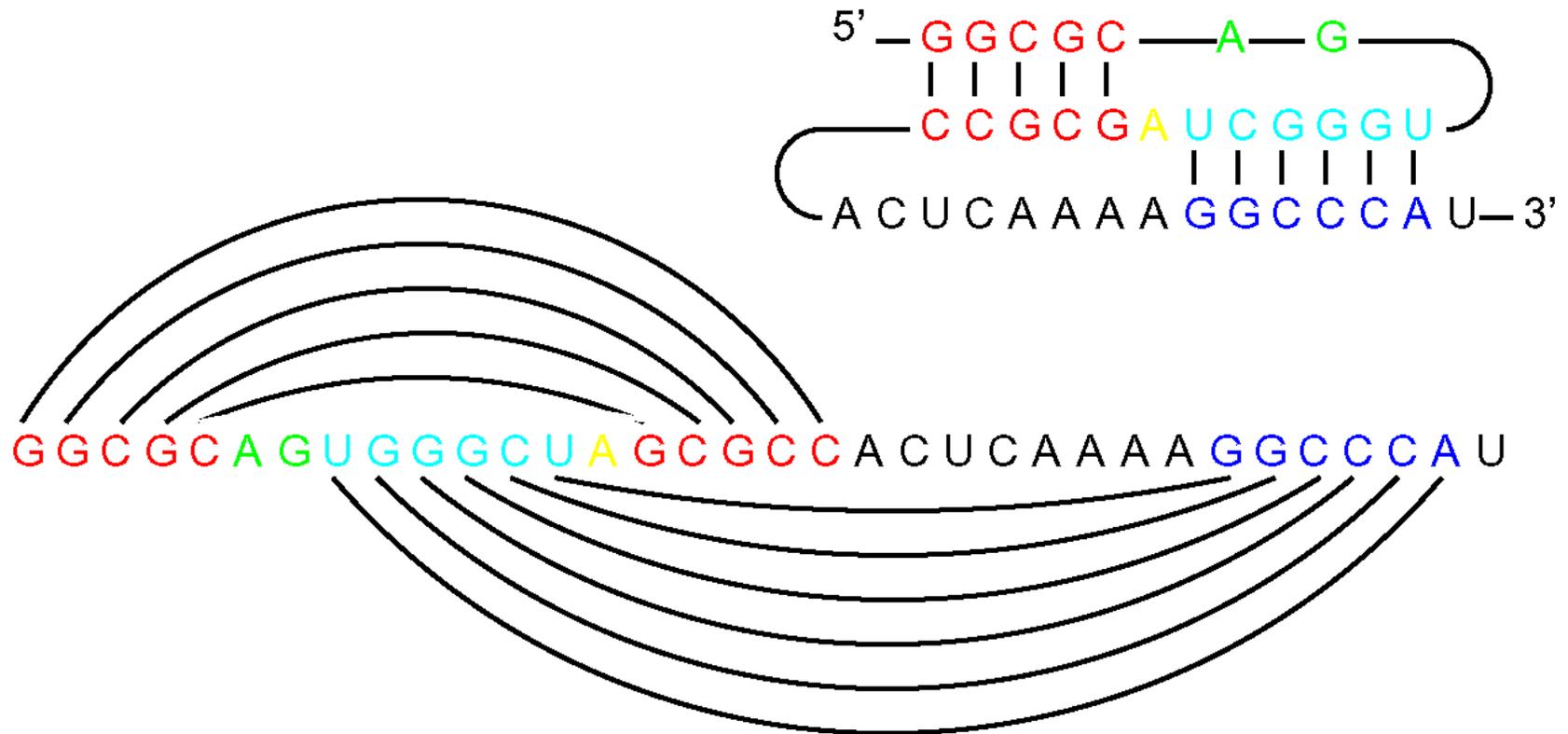
$i < i' < j' < j$ ($i..j$ includes $i'..j'$)

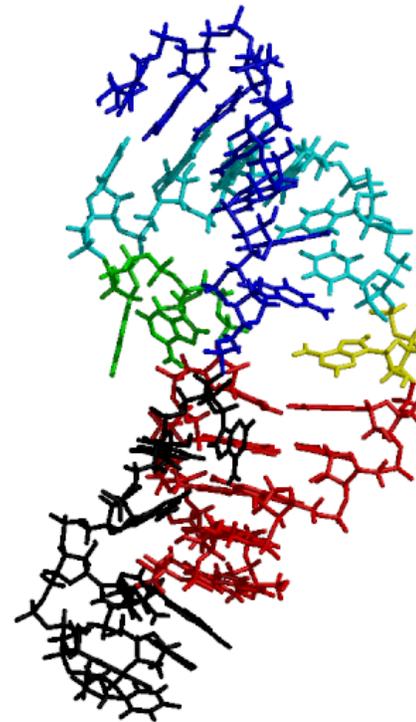
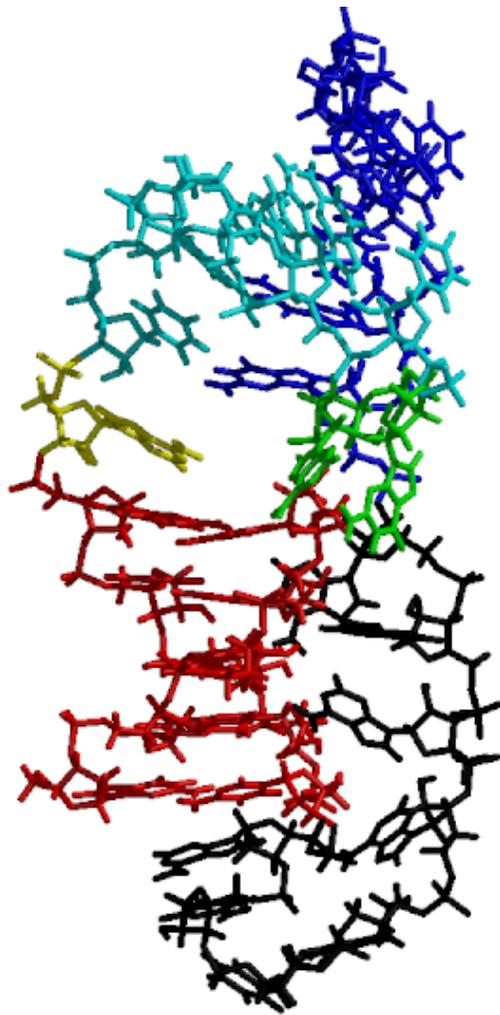
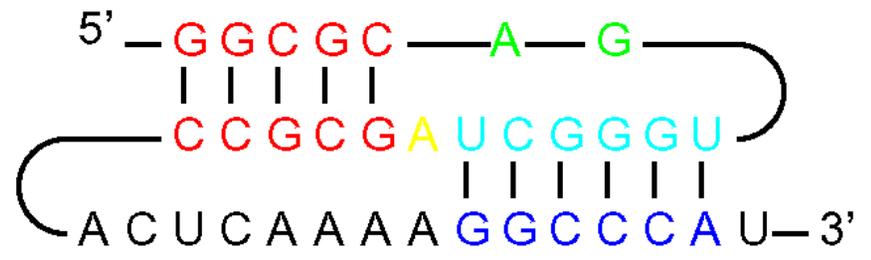


$i < j < i' < j'$ ($i.j$ precedes $i'.j'$)



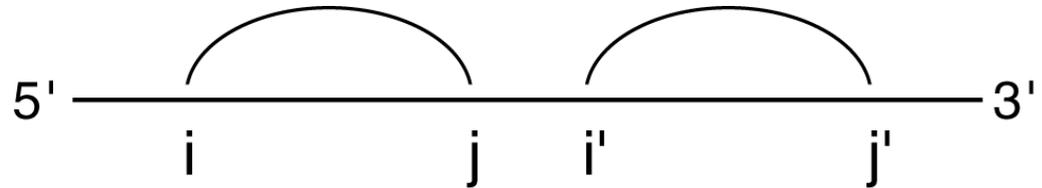
$i < i' < j < j'$ (pseudoknot)



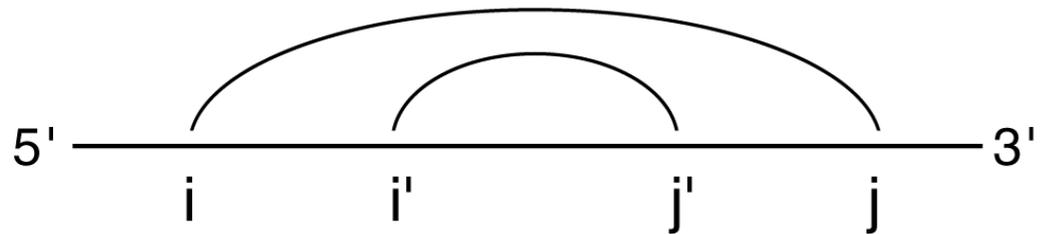


The three cases

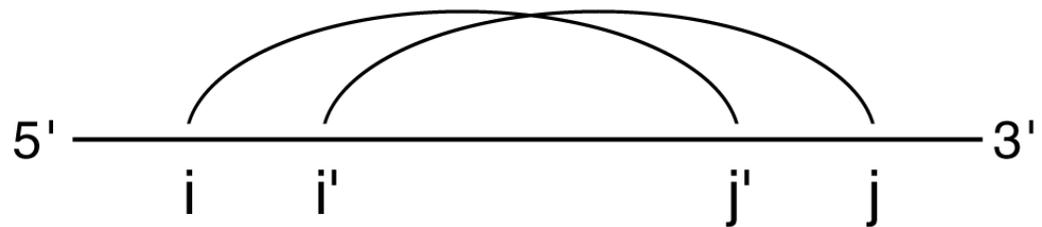
a. $i < j < i' < j'$



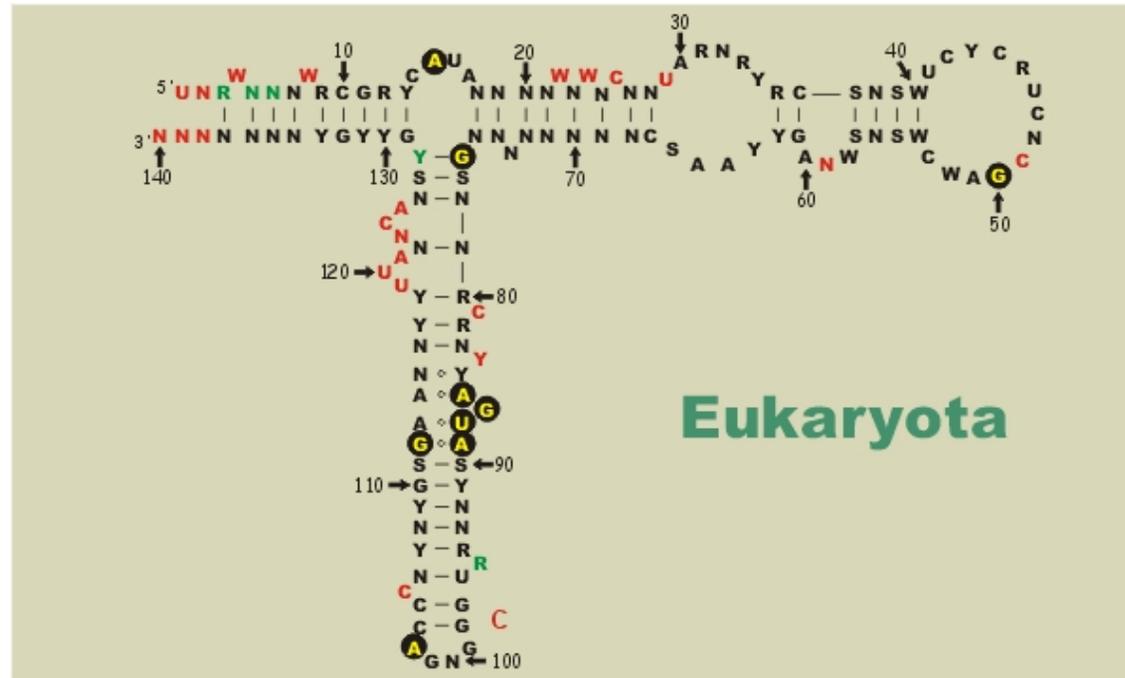
b. $i < i' < j' < j$



c. $i < i' < j < j'$

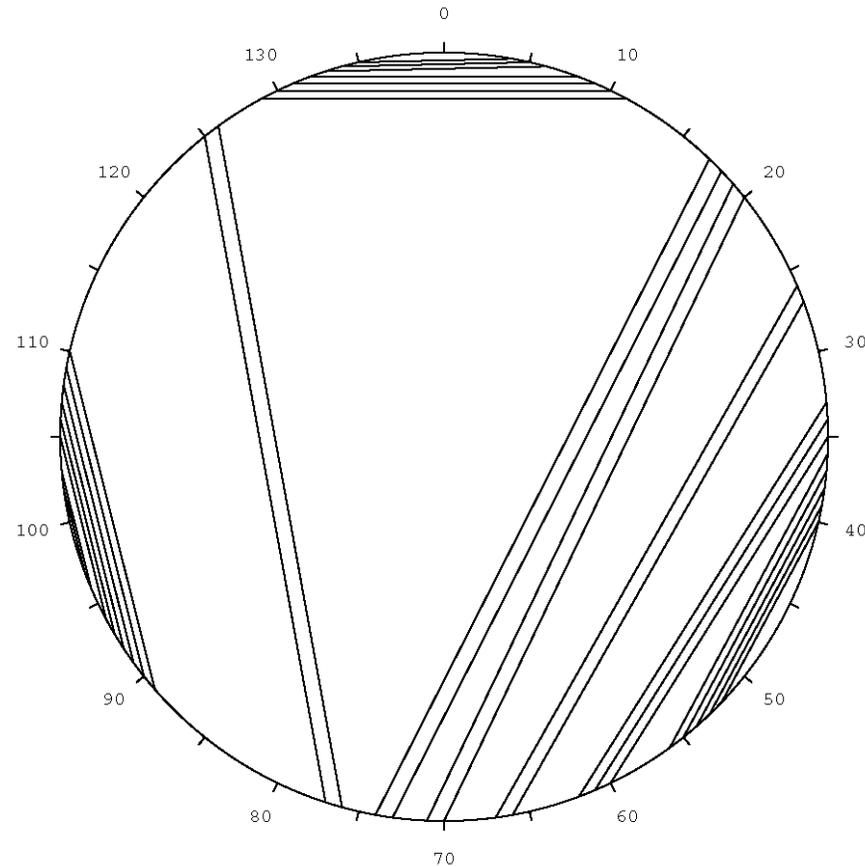


5S rRNA

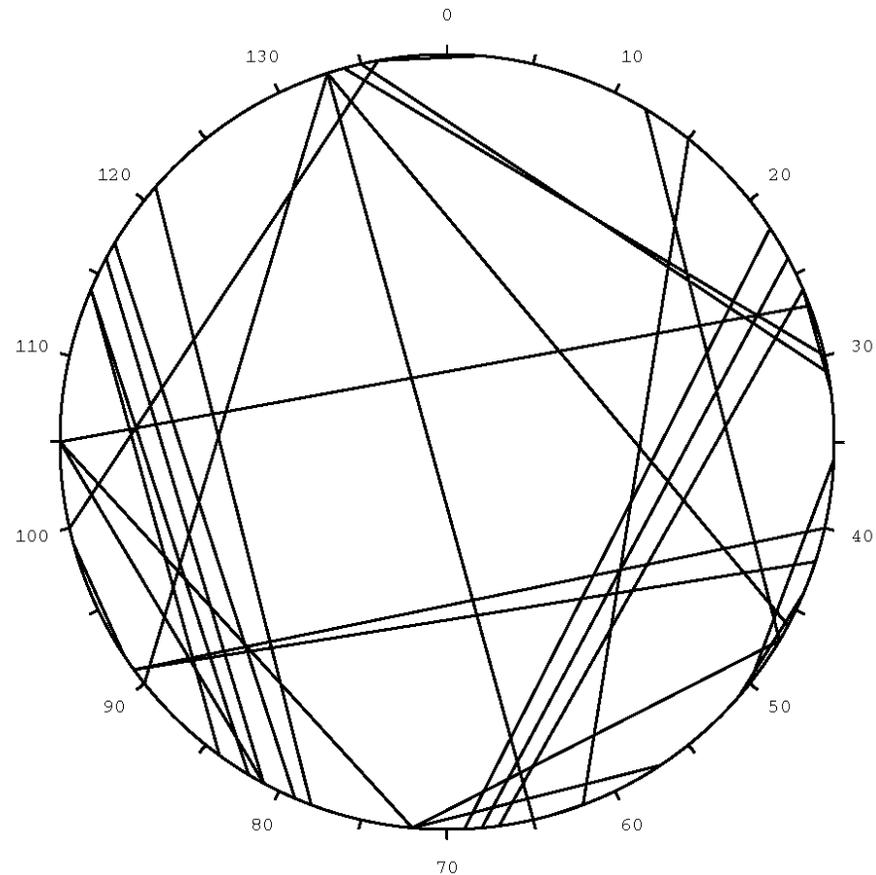


from <http://rose.man.poznan.pl/5SData/>

Eukaryotic 5S RNA sequences secondary structure interactions



Eukaryotic 5S RNA sequences (possible 3D interactions)



Secondary Structure Determination

- X ray crystallography, N.M.R.
- Chemical and enzymatic probing, cross-linking
- **Comparative sequence analysis**
- **Minimum free energy (MFE) methods**
- **Comparative sequence analysis + MFE**

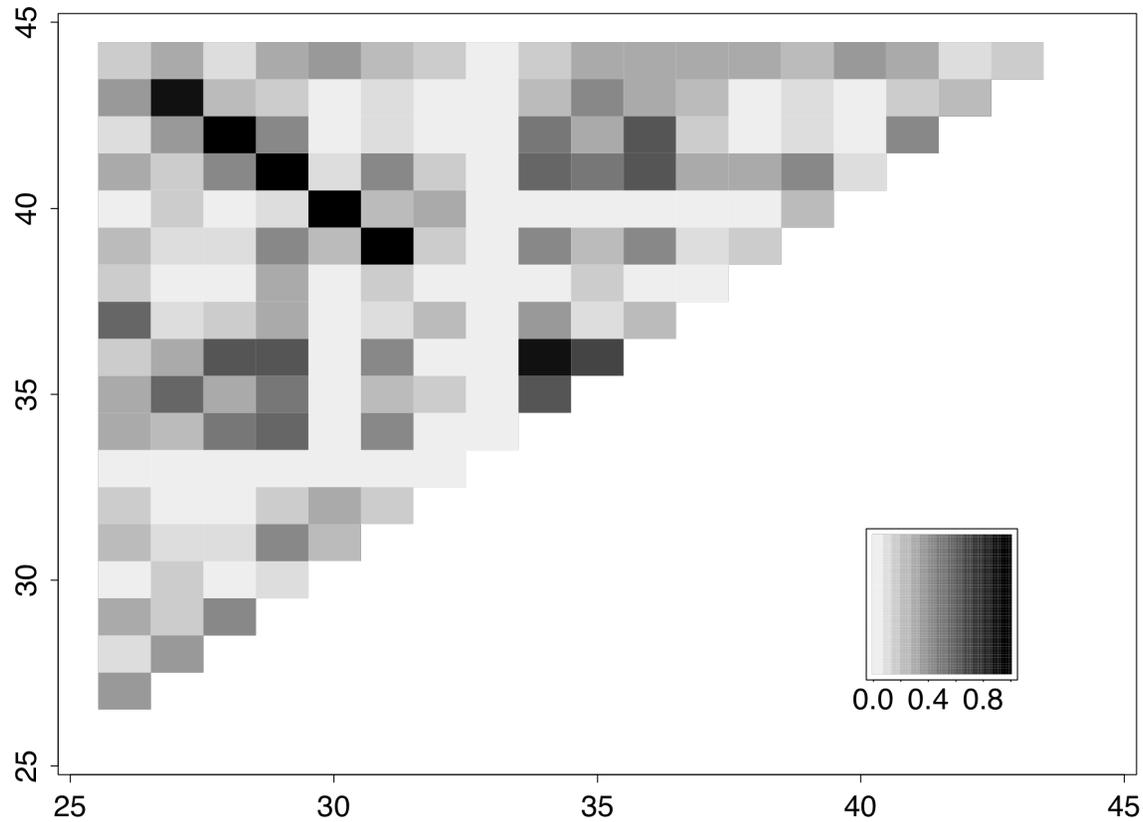


Comparative Sequence Analysis

“Today, comparative analysis has become the method of choice for establishing higher-order structure for large RNA”
Pace, Thomas, Woese (1999) In The RNA World. Cold Spring Harbor.

ACGU**C**AUCAGUCAUGUCAGUCA**G**UAGCUGA
ACGU**C**AAGG--AAUGUCAGUCA**G**UAGCUGA
ACGU**C**AUCAAGGUUGUCAGUCA**G**UAGCUGA
ACGU**G**AUCAGUCAUGGG--ACA**C**UAGCUGA
ACGU**C**AAGGGUUU--GGAGUCA**G**UAGCUGA





Saccharomyces cerevisiae
Spiroplasma meliferum
Mycoplasma capricolum
Mycoplasma mycoides
Spiroplasma meliferum
Streptomyces lividans

...**CCAGACUGAAGAUCUGG**...
CCUGCCUUGCACGCAGG
CCUCCCUGUCACGGAGG
CACGGUUUCAUCCGUG
UUUGAUUGAAGCUCAAA
ACGGCCUGCAAAGCCGU
 30 35 40

Comparative Sequence Analysis

- Starts with the alignment of a set of homologous sequences (computer-assisted, but manually refined)
- Detecting correlated pairs
- Analyzing correlated pairs:
 - Parallel chords implies helices
 - Others are tertiary structure interactions



Detecting Correlated Pairs

- Chi-square test of independence
- **Mutual information**

$$M(I, J) = H(I) + H(J) - H(I, J)$$

where

$$H(I, J) = - \sum_{\alpha\beta} P(i = \alpha, j = \beta) \log P(i = \alpha, j = \beta)$$

$$H(I) = - \sum_{\alpha} P(i = \alpha) \log P(i = \alpha)$$



Analyzing Correlated Pairs

- Detecting secondary structure elements:
 - Mostly canonical base pairs (Watson-Crick)
 - Parallel ($i:j, i+1:j-1$)
 - Wobble (G:U) and A:G are occurring frequently
- Non-canonical (isosteric)
- Detecting tertiary structures (including pseudoknot)
- Tetraloop: UNCG, CUYG, GMRA (GNRA)
- Base-triples



What are the main difficulties?

- Needs an alignment, but sequence alignment techniques are not well adapted for RNA sequences
- To produce a high quality alignment, the sequences should be similar
- If the sequences are similar, there will be few observed compensatory changes

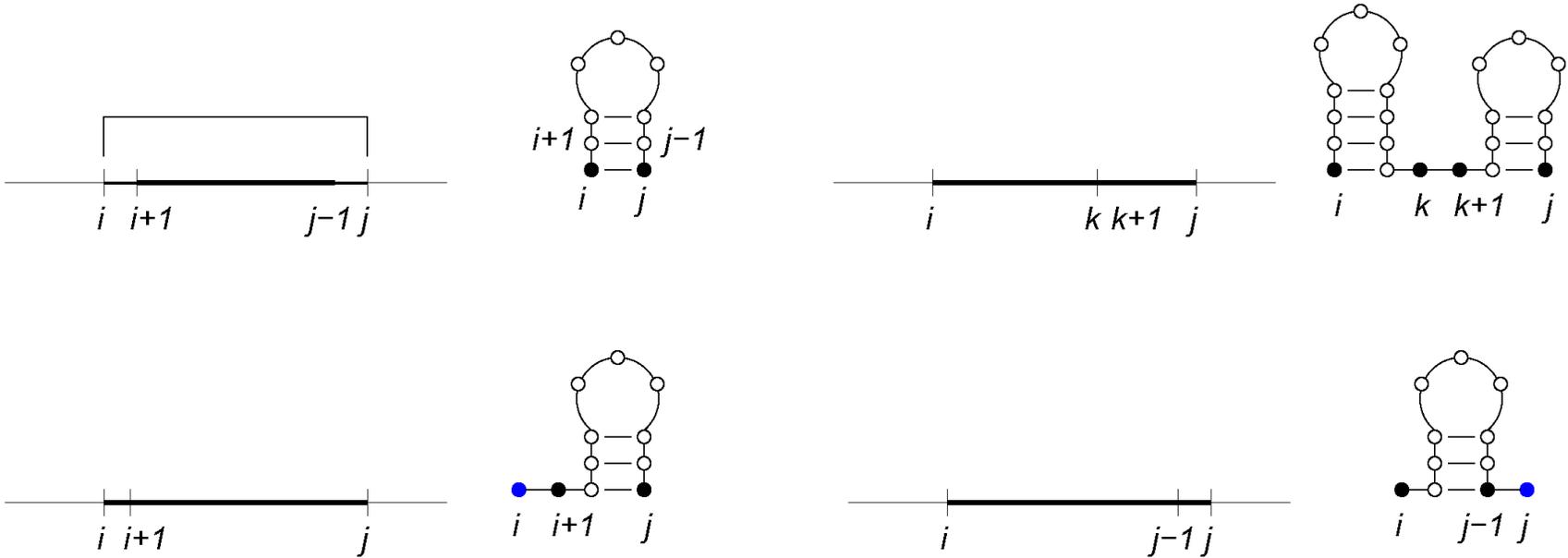


RNA folding

- How to search the space of all possible secondary structures?
- How to select the best structure?
 - Maximizing the number of base-pairs (Nussinov)
 - Maximizing the number of hydrogen bonds
 - Minimizing the free energy (Zuker/*mfold*)



What is the maximum number of base pairs that can be formed for the segment $i .. j$?



Putting it all together

- We know that for $j-i \leq 4$ **fold**(s, i, j) = 0
- Otherwise, **fold**(s, i, j) is the maximum of
 - **1 + fold**($s, i+1, j-1$) if $s(i)$ and $s(j)$ form a canonical base pair;
 - **fold**($s, i+1, j$);
 - **fold**($s, i, j-1$);
 - **fold**(s, i, k) + **fold**($s, k+1, j$) for some k s.t. $i \leq k \leq j$.
- The answer we're looking for is **fold**($s, 1, n$).



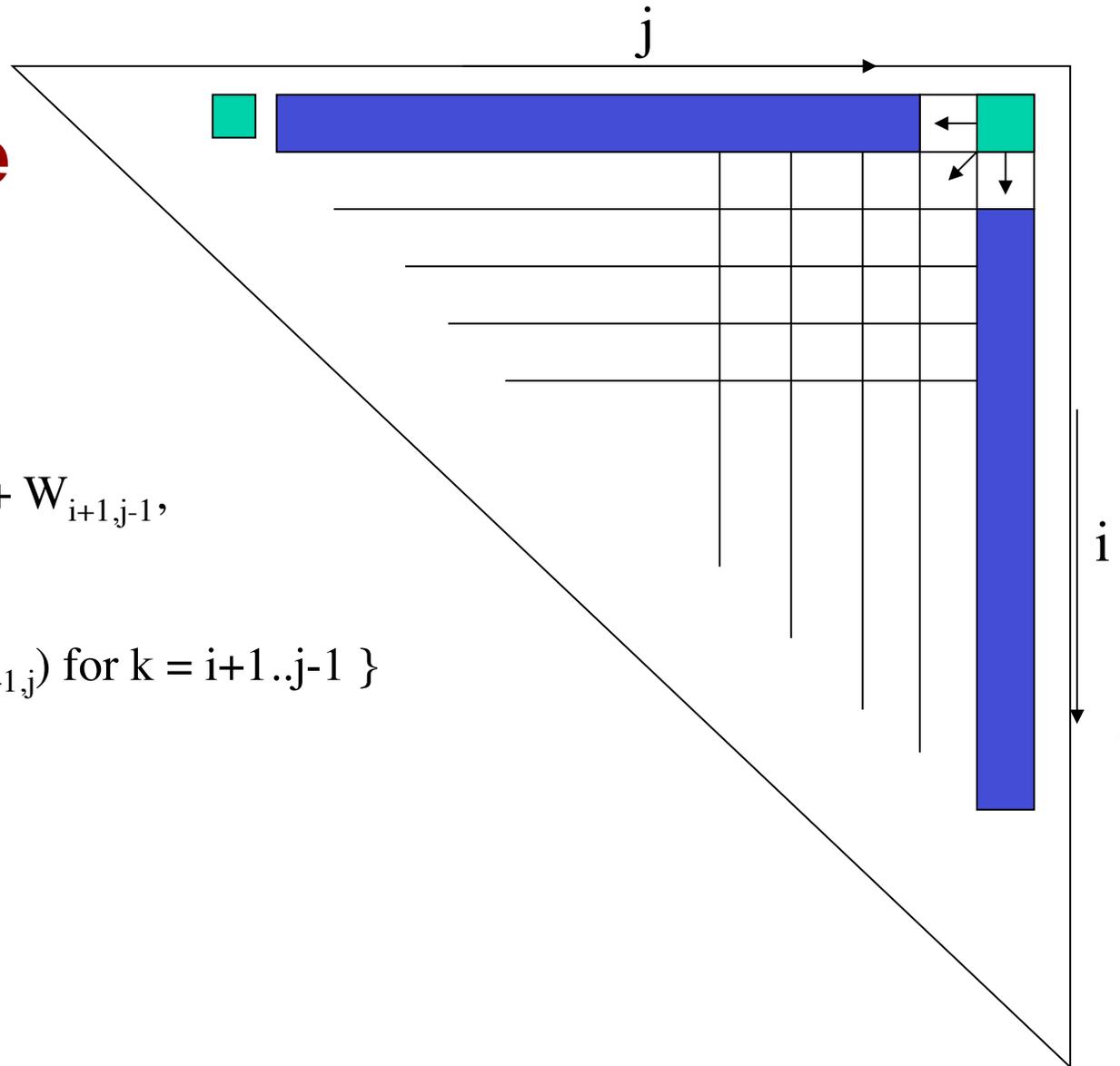
Remarks

- The proposed algorithm is not practical, it requires an **exponential** number of calls to **fold(s, i, j)**
- However, there is a maximum of $n \times n$ distinct values of **fold(s, i, j)**
- This suggests a caching strategy (tabular computation)



Filling the DP table

$$W_{ij} = \max \left\{ \begin{array}{l} \delta(s(i), s(j)) + W_{i+1, j-1}, \\ W_{i+1, j}, \\ W_{i, j-1}, \\ (W_{i, k} + W_{k+1, j}) \text{ for } k = i+1 \dots j-1 \end{array} \right\}$$



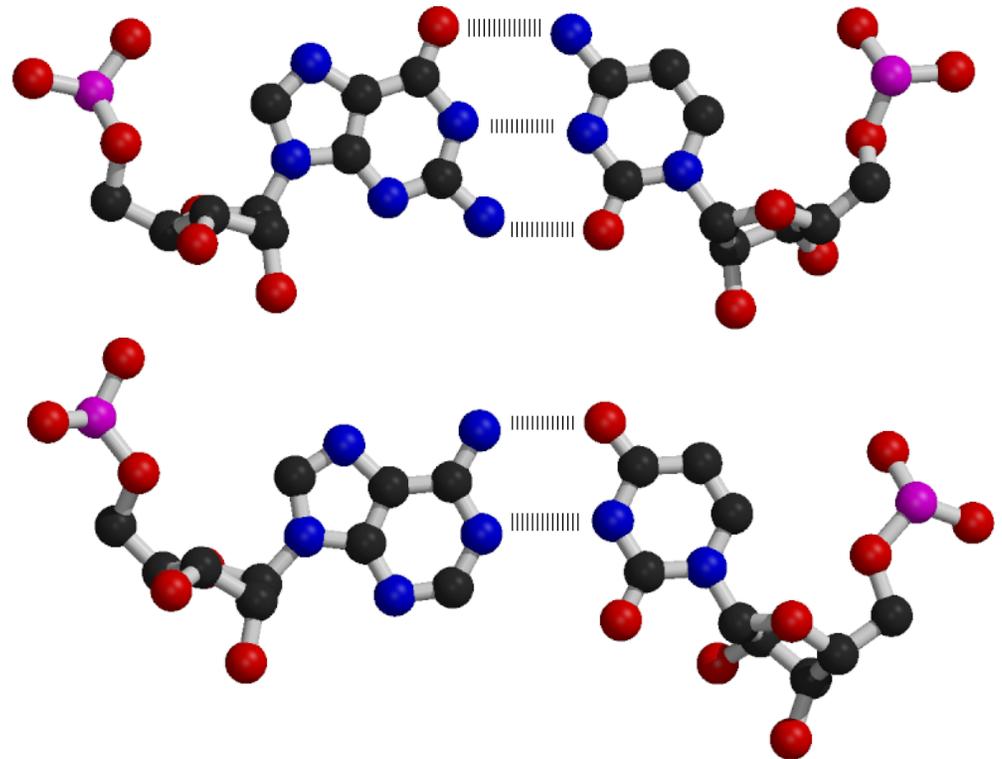
**Maximizing the number of
base pairs is not a good
strategy**





Maximizing the number of hydrogen bonds: A better cost function?

+ 3 for a G:C base pair
+2 for an A:U
+1 for a Wobble (G.U)



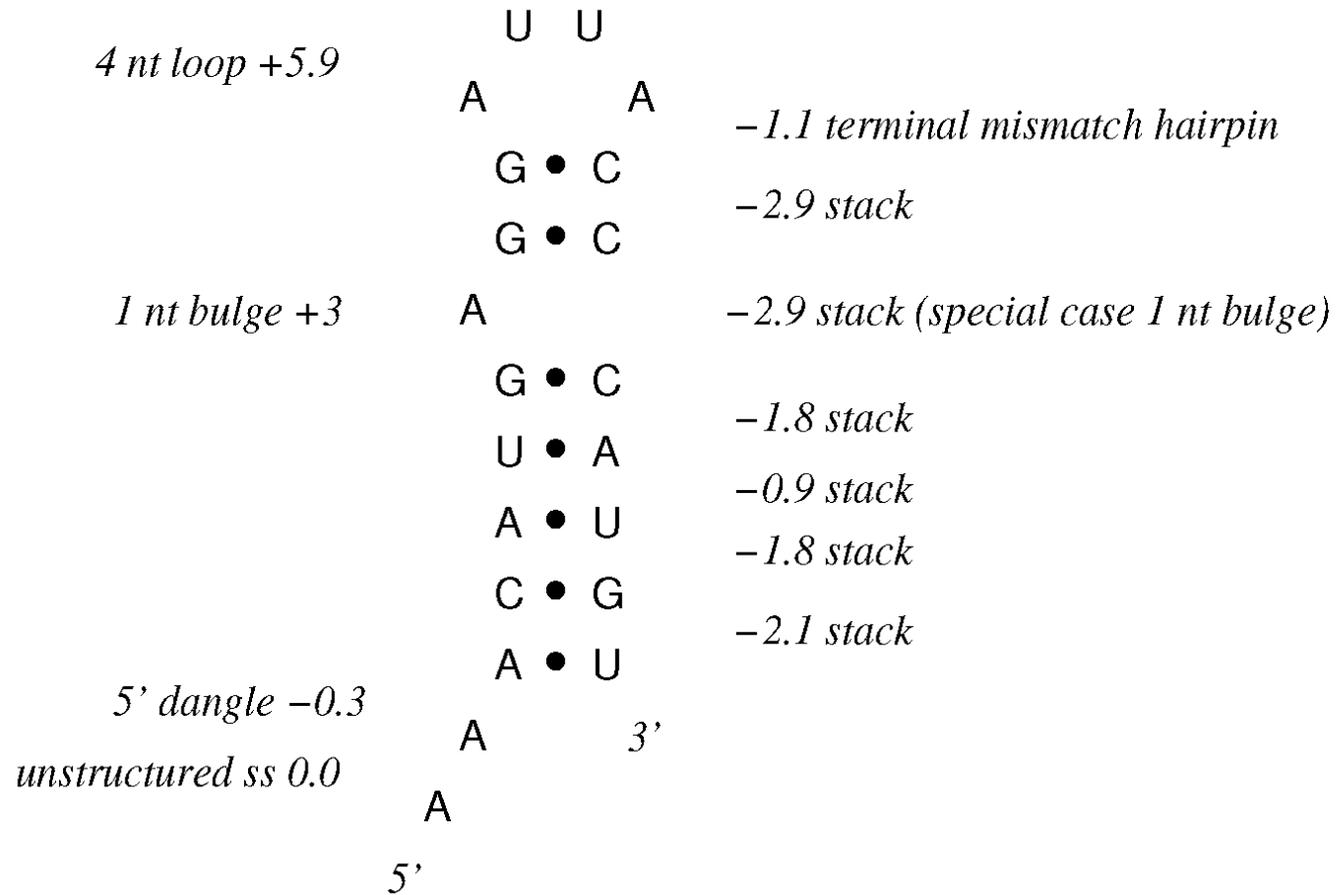


Better cost functions

- It turns out that maximizing the number of base pairs, or the number of hydrogen bonds, is not what Nature has favored
- The **stacking** contributions from the interface between neighboring base pairs seem to be preferred



$\Delta G = -4.9 \text{ kcal/mol}$



www.bioinfo.rpi.edu/~zukerm/rna/energy

From Durbin *et al* (1998) Cambridge Press.

MFOLD

- Sophisticated energy minimization program developed by **Mike Zuker**
- Finds the structure with the minimum equilibrium free energy (ΔG), as approximated by **neighboring base pair contributions**
- **Takes into account:** stacking, hairpin loop lengths, bulge loop lengths, interior loop lengths, multi-branch loop lengths, single dangling nucleotides and terminal mismatches on stems



MFOLD and PKNOTS (Implementation)

- MFOLD does not include pseudoknots
- MFOLD and the dynamic programming algorithm is in $O(N^3)$
- PKNOTS is an implementation of the dynamic programming that includes pseudoknots
- PKNOTS with pseudoknots is in $O(N^6)$



Some recent developments

- Dynalign is an algorithm that **simultaneously align two RNA sequences and finds a common secondary structure** with minimum free energy:
 $\Delta G_1 + \Delta G_2 + \Delta G_{gap}$ (*number of gaps*)
- Computationally intensive! $O(M^3 N^3)$, where N is the length of the shortest sequence and M is maximum insertion size



Practical Remarks

- MFOLD was benchmark on a set of 955 structures of 700 nt or less:
 - Before 1999, 64% of the known base pairs were correctly predicted
 - 1999+, **73%**
- Dynalign (a standalone program)
 - 13 tRNAs: Dynalign = **86.1%**, MFOLD = 59.7 %
 - 7 5S rRNA: Dynalign = **86.4%**, MFOLD = 47.8 %



Further extensions

- **eXtended Dynalign** takes three input sequences and produces 1) alignment as well as 2) a consensus secondary structure
- **Profile-Dynalign** takes as input an arbitrarily large number of input sequences, applies a **progressive alignment strategy** akin to CLUSTAL and produces 1) a multiple sequence alignment as well as 2) a consensus secondary structure



eXtended and Profile-Dynalign

- See PDF document.



Practical Remarks (contd)

- MFOLD requires a single sequence;
- MFOLD allows for constraints;
- MFOLD reports sub-optimal solutions;

Seed

- See PDF document.

Part II

- **Database search**
 - Traditional bioinformatics tools
 - Specialized tools
 - Specialized databases

S

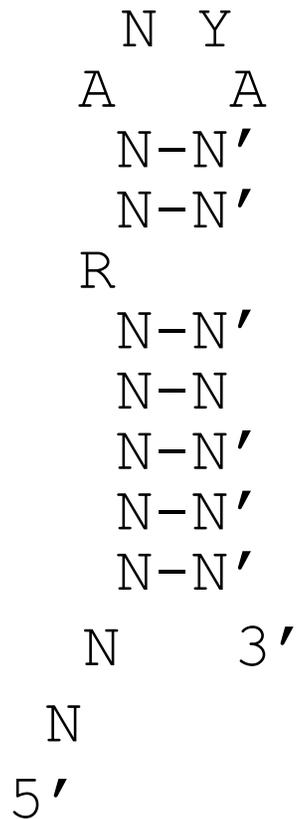
- See Backhofen's Garfield the fat and old cat vs Garfield the cat and the old hat

Important Observations

- Many RNAs conserve their (secondary) structure more than their sequence
- Consequently, sequence alignment techniques (such as blast) fail to detect homologues
- More sophisticated tools are required



R17 virus coat protein binding site



IUPAC ambiguity codes

R = [GA]

D = [^C]

Y = [CT]

H = [^G]

M = [AC]

V = [^T]

K = [GT]

N' is the

S = [GC]

complement of N

W = [AT]

N = [ACGT]



***i.i.d.* sequence model**

- Under the assumptions that positions are independent and identically distributed (*i.i.d.*), and all 4 nucleotide types are equiprobable;
- i.e. the sequence motif
NNNNNNNRNNANYNNNNNNN;
- The probability that a random sequence matches the **sequence motif** of the R17 coat protein binding site is,

$$\left(\frac{1}{2}\right) \times \left(\frac{1}{4}\right) \times 1^{17} = \frac{1}{64} = 0.015625$$

- You would expect 56 hits in the 3,569 nts of the R17 virus genome.

***i.i.d.* structural model**

- Under the assumptions that positions are independent, **except for paired positions**, and identically distributed (*i.i.d.*), and all 4 nucleotide types are equiprobable;
- The probability that a randomly selected sequence matches the secondary structure motif of the R17 virus coat protein binding site is,

$$\left(\frac{1}{4}\right)^7 \times \left(\frac{1}{4}\right)^2 \times \left(\frac{1}{2}\right)^2 = \left(\frac{1}{2}\right)^{20} \approx 9.5 \times 10^{-7}$$

– Would occur 0.003 times by chance in R17 virus genome.



Searching for Structural Motifs

- General purpose tools
 - Generation 1: pattern
 - Generation 2: built-in scoring mechanisms
 - Generation 3: built-in covariance model
 - Future: automatic inference
- Specialized programs
 - tRNA-scan-SE
 - snoRNA



Searching for Structural Motifs: A first generation of algorithms

The input of general motif search procedures, such as RNAMOT or RNABOB, requires a description of the motif in terms of its secondary and tertiary structure: the **descriptor** or **pattern**



RNAMOT Descriptor

H1 s1 H2 s2 H2 s3 H3 s4 H3 s5 H1

H1 3:5 0

H2 4:5 1 AGC:GCU

H3 4:5 1

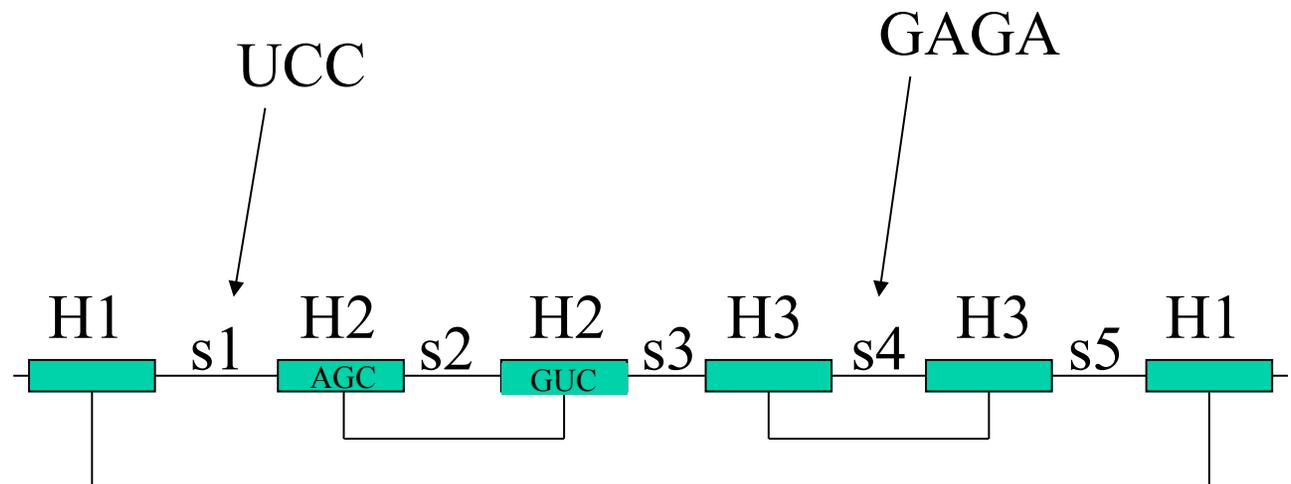
S1 3:6 UCC

S2 5:7

S3 0:3

S4 5:8 GAGA

S5 3:5



R H2 H3 H1

M 1

RNAMOT execution

- RNAMOT -s -s mydb.fa -d mystery.mot

```
--- HUM7SLR1 Human 7SL RNA pseudogene, clone p7L30.1. --- (110 bases)
|SCO: 201.40|POS:6-56|MIS: 0|WOB: 0|
|CAGCU|GAUGCU|AGCU|GAUGCU|AGCU|-|GAUCG|UAGCUAGU|CGAUC|CGU|AGCUG|
...
```

RNAMOT Descriptor

Secondary structure description

H1 s1 H2 s2 H2 s3 H3 s4 H3 s5 H1

Length range

H1 3:5 0 ← Number of allowed mismatches

H2 4:5 1 AGC:GCU ← Sequence pattern

H3 4:5 1

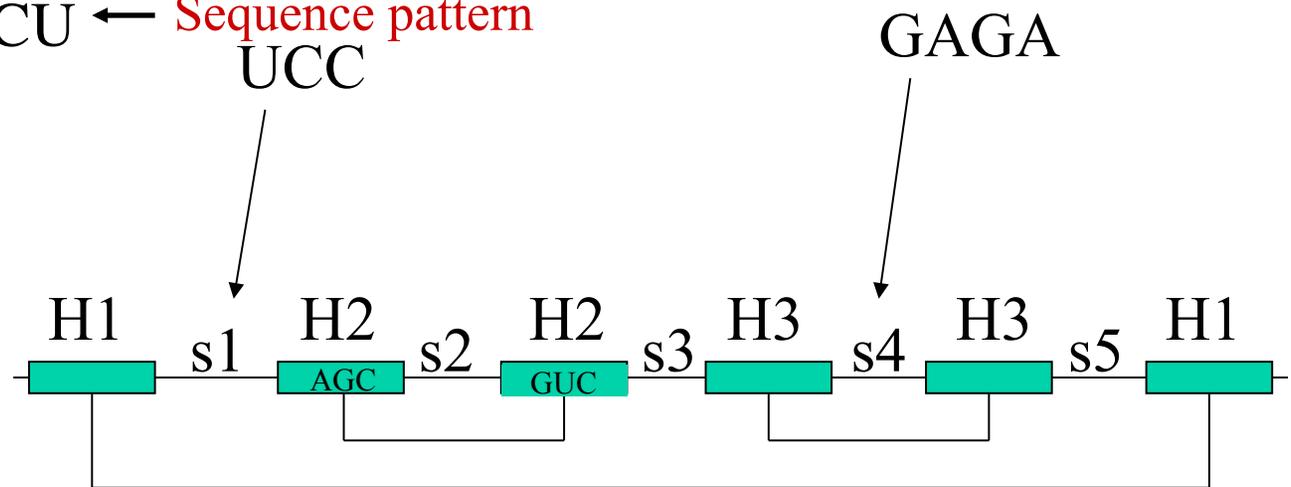
S1 3:6 UCC

S2 5:7

S3 0:3

S4 5:8 GAGA

S5 3:5



R H2 H3 H1 ← Search order information

M 1 ← Total number of mismatches

Similar tools

- RNABOB

<http://www.genetics.wustl.edu/eddy/software/>

- PatScan

- <http://www-unix.mcs.anl.gov/compbio/PatScan/>

- scan_for_matches (stand alone program)

- p1=4...7 3...8 ~p1

(p1 contains 4 to 7 characters, it is followed by 3 to 8 characters, followed by the reverse complement of p1)



Remarks

- These computer programs are practical and can be applied to large data-sets
- One of the major difficulties arises from the **subjectivity in deriving the best descriptor** for a family of sequences



Second Generation of Pattern Matching Engines

- 10+ years after RNAMOT was published, RNAMOTIF was released;
- It has all the functionalities of RNAMOT + the ability for the user to define a scoring function!
- It also features a powerful scripting language.

- Macke *et al.* (2001) *Nuc. Acids. Res.* **29(22)**: 4724-4735.

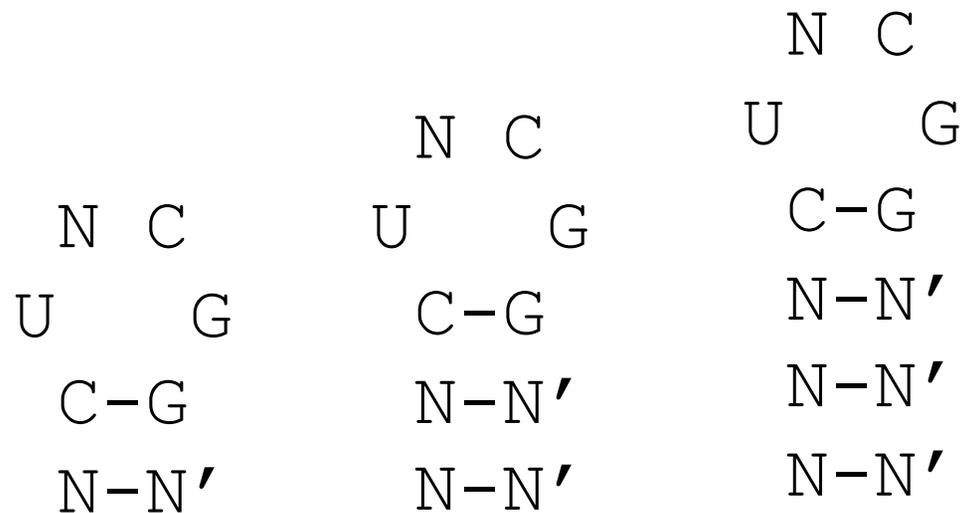
UNCG loop

descr

h5(minlen=2,maxlen=4,seq="C\$")

ss(len=4,seq="UNCG")

h3(seq="^G")



```

$ rnamotif -descr uncg.descr 16S_E_Co1i.fa
uncg.descr: complete descr length: min/max = 8/12
#RM scored
#RM descr h5 ss h3
>rRNA
rRNA          0.000 0      206      8 cc ttcg gg
>rRNA
rRNA          0.000 0      339     12 ctcc tacg ggag
>rRNA
rRNA          0.000 0      340     10 tcc tacg gga
>rRNA
rRNA          0.000 0      341      8 cc tacg gg
>rRNA
rRNA          0.000 0      418      8 cc ttcg gg
>rRNA
rRNA          0.000 0     1027      8 cc ttcg gg
>rRNA
rRNA          0.000 0     1448      8 cc ttcg gg

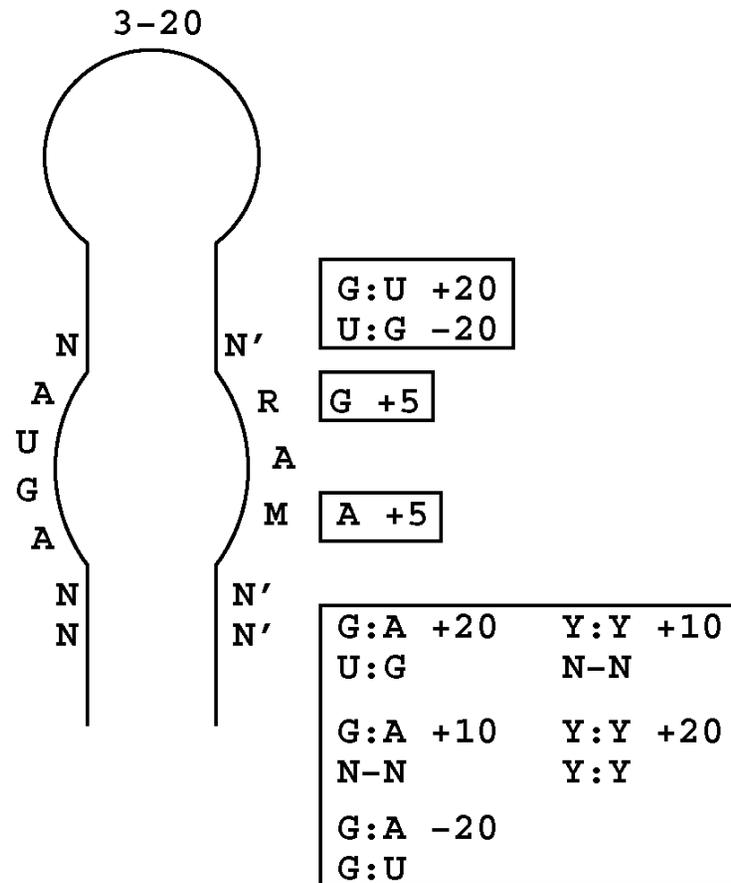
```

GNRA

```
Parms
  wc +=gu;
descr
  h5( len=3 )
  ss( len=4, seq="GNRA" )
  h3
```

Allowing for Wobble (GU) base pairs

E-loop



E-loop: defining new base pairs

```
parms ##Define global parameters  
  
wc += gu;  
ga = {"G:A","A:G"};  
all = {"g:a","g:c","g:u","g:g","u:c","u:u","u:a","u:g","c:c","c:u","c:g",  
       "c:a","a:a","a:c","a:g","a:u"}
```

E-loop: pattern description

```
descr #Core structure and sequence definition
h5(tag='lower_stem',minlen=0,maxlen=10, pair+=ga, pairfrac=0.8) #1
h5(tag='2',len=2, pair += all) #2
ss(len=4, seq="AGUA") #3 No variation allowed
h5(tag='3',len=1, pair += all) #4
h5(tag='upper_stem',minlen=0,maxlen=10,pair+=ga,pairfrac=0.8) #5
ss(minlen=3,maxlen=10, tag='stem_loop') #6 Bonus for GNRA +100, UNCG +100
h3(tag='upper_stem') #7
h3(tag='3') #8
ss(len=3,seq="RAM") #9 Bonus, R=G, +5, M=A +5
h3(tag='2') #10
h3(tag='lower_stem') #11
```

E-loop: score

```
score{ # User-controlled scoring section
motif_score=0;
## Element 2 bonus rules
### 5'-UG, AG-3' +20
### 5'-NG, AN-3' +10
### 5'-GG, AU-3' -20
## 5'-YY, YY-3' +20
### 5'-NY, YN-3' +10

### Good score for G:A in Start:End under some conditions
if (h5[2,2,1]:h3[10,1,1] in {"g:a"} ){
  if (h5[2,1,1]:h3[10,2,1] in {"u:g"} )
    motif_score += 20;
  else if (h5[2,1,1]:h3[10,2,1] in {"g:u"})
    motif_score -=20;
  else if (h5[2,1,1]:h3[10,2,1] in {"g:c","c:g","u:a","a:u"})
    motif_score +=10;
}
```

```

else if( h5[2,2,1]:h3[10,1,1] in {"u:u","u:c","c:u","c:c"} ){
  if (h5[2,1,1]:h3[10,2,1] in {"u:u","u:c","c:u","c:c"})
    motif_score +=20;
  else if (h5[2,1,1]:h3[10,2,1] in {"g:c","c:g","u:a","a:u"})
    motif_score +=10;
}

## Element 4 bonus rules
## Bonus GU +20, Penalty UG -20
if (h5[4,1,1]:h3[8,1,1] in {"g:u"})
  motif_score +=20;
else if (h5[4,1,1]:h3[8,1,1] in {"u:g"})
  motif_score -=20;

### Element 9 bonus rules
### Bonus M=A +5

if ( ss[9,3,1] =~ "a")
  motif_score +=5;

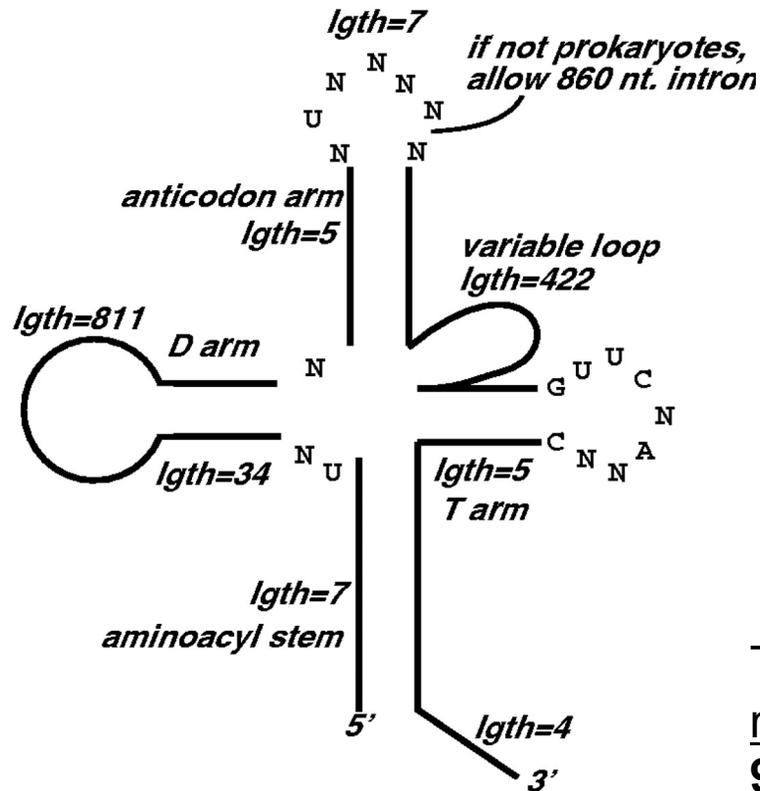
### Bonus R=G +5
if ( ss[9,1,1] =~ "g")
  motif_score +=5;

###Reject poor matches to the E-loop descriptor
if (motif_score < 0)
  REJECT;
SCORE = motif_score;
}

```

tRNA

A.



Tsui, Macke and Case (2003) [A novel method for finding tRNA genes.](#) *RNA* **9**:507-517.

B.

```
parms
  wc += gu;

descr
  h5(tag='h1',len=7,mispair=1,ends='mm')
  ss(tag='s1',len=2)
  h5(tag='h2',minlen=3,maxlen=4,mispair=1,ends='mm')
  ss(tag='s2',minlen=8,maxlen=11)
  h3(tag='h2')
  ss(tag='s3',len=1)
  h5(tag='h3',len=5,mispair=1,ends='mm')
  ss(tag='s4',len=7)
  h3(tag='h3')
  ss(tag='s5',minlen=4,maxlen=22)
  h5(tag='h4',len=5,mispair=1,ends='mm')
  ss(tag='s6',len=7)
  h3(tag='h4')
  h3(tag='h1')
  ss(tag='s7',len=4)

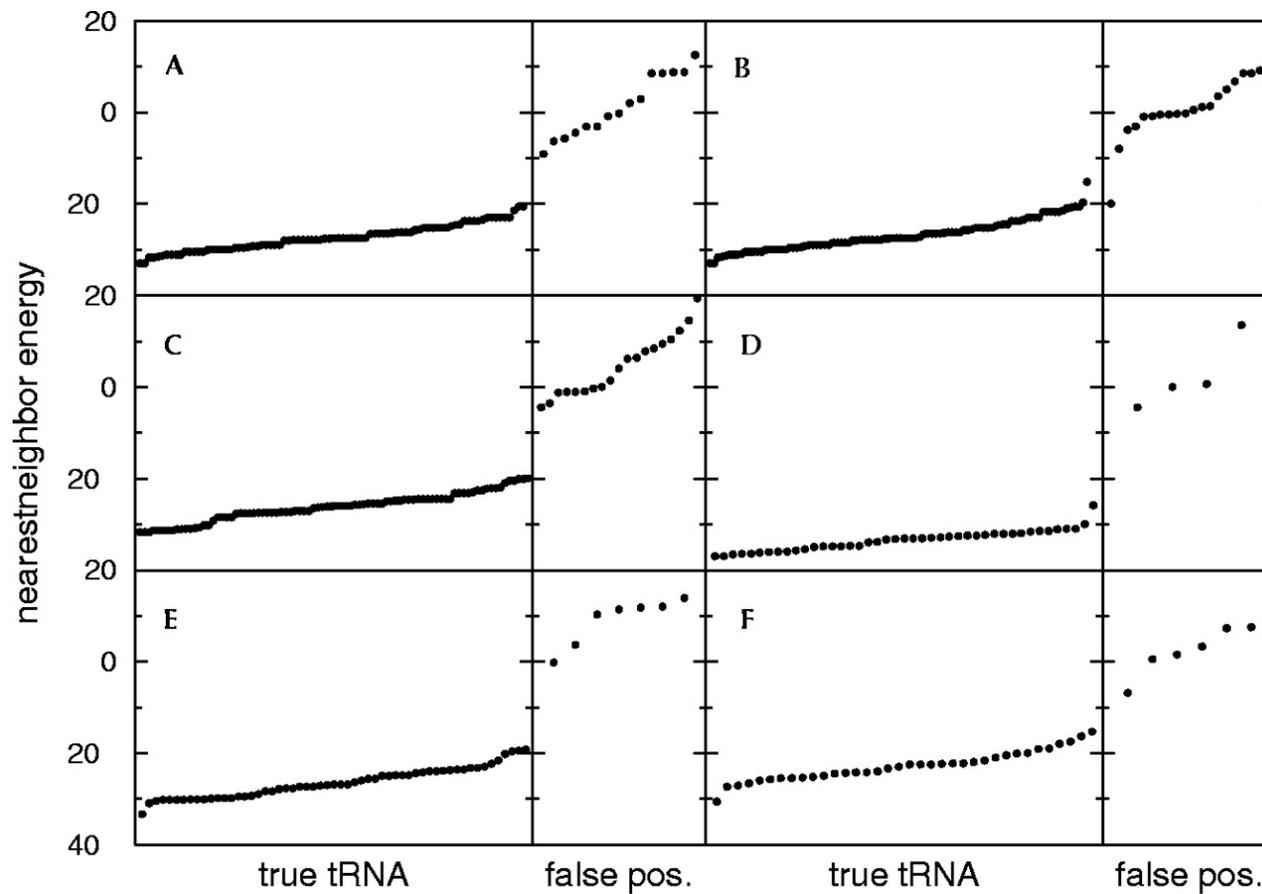
score
{
  n = 0;
  if (ss['s1',1,1] != "u") n++;
  if (ss['s4',2,1] != "u") n++;
  if (h5['h4',5,1] != "g") n++;
  if (ss['s6',1,1] != "u") n++;
  if (ss['s6',2,1] != "u") n++;
  if (ss['s6',3,1] != "c") n++;
  if (ss['s6',5,1] != "a") n++;
  if (h3['h4',1,1] != "c") n++;

  if (n > 1) REJECT;

  SCORE = efn( h5['h1'],ss['s7'] );
}
```



RNA “threading”



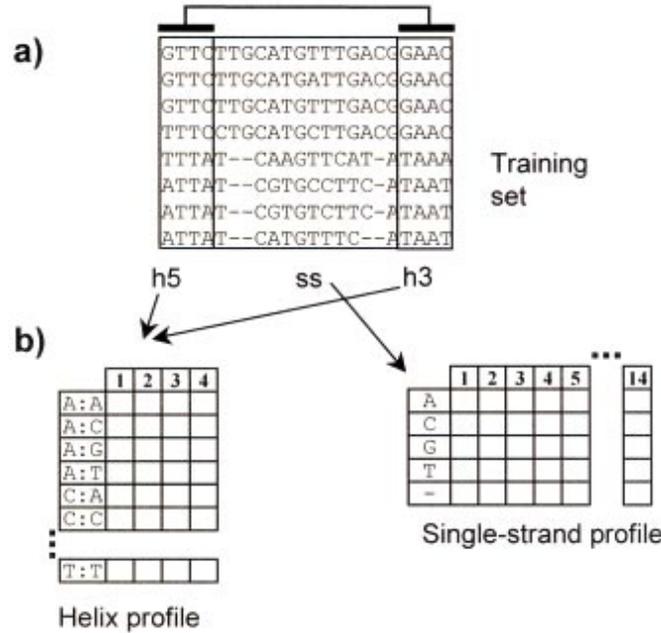
Recent Software Developments

- Profiles
 - ERPIN (Gautheret & Lambert, 2001)
- Stochastic Context-Free Grammars (SCFG)
 - Cove (Eddy & Durbin, 1994)
 - Rfam

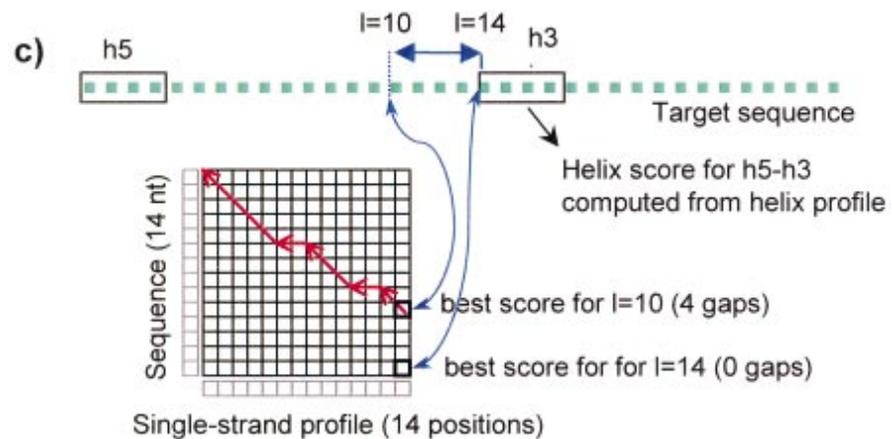
ERPIN

- **Problem:** Pattern matchers, such as RNAMOT, are “hit of fail”;
- The solution to this problem for proteins has been to use profiles, which are a probabilistic representation of the sequence;
- ERPIN generalizes this idea to “structural” profiles.

$$S_{i,j} = \log \frac{O_{i,j}}{E_i \times E_j}$$



$$S_i = \log \frac{O_i}{E_i}$$



Gautheret & Lambert (2001) *JMB* 313, 103-101.



Remarks

- Limitation: gaps are not allowed in helical regions;
- Initial version only allows searching for one hairpin (Hp), one helix (Hx), one strand (St) or two helices (H2);
- Fast enough to scan entire genomes;
- Iterative search; *à la* PSI-BLAST;
- tRNA benchmark: sensitivity = 95%, 0.2 false positive per *E.coli* genome



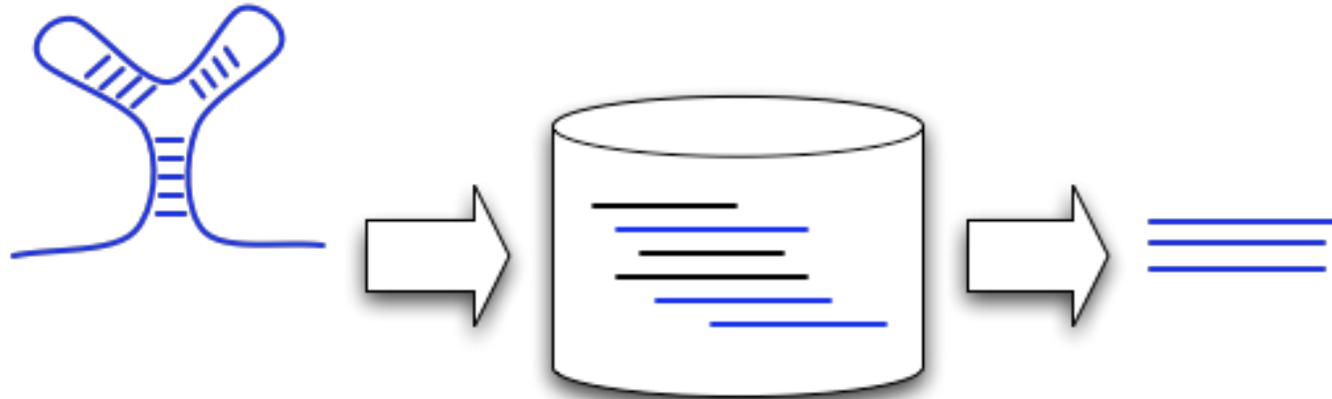


RSEARCH

- R.J. Klein and S.R. Eddy (2003) RSEARCH: Finding homologs of single structured RNA sequences. *BMC Bioinformatics* 2003, 4:44
(doi:10.1186/1471-2105-4-44)
- **Input:** an RNA sequence and its secondary structure
- **Output:** similar RNAs on the basis of both primary sequence and secondary structure



RSEARCH (contd)



RSEARCH Input

```
# STOCKHOLM 1.0

#=GS Holley DE tRNA-Ala that Holley sequenced from Yeast genome

Holley
  GGGCGTGTGGCGTAGTCGGTAGCGCGCTCCCTTAGCATGGGAGAGGtCTCCGGTTCGATTCCGGACTCGTCCA
#=GR Holley SS
  (((((.(..(((.....))))).(((.....))))).(((.....))))).(((.....))))).
  //
```

RSEARCH (contd)

- RIBOSUM substitution matrices (analogous to residue substitution scores such as PAM and BLOSUM but for base pairs)
- Reports the statistical significance of all the matches
- Execution time is $O(NM^3)$ where N is the size of the database and M is the length of the input sequence
- **“(...) a typical single search of a metazoan genome may take a few thousand CPU hours”**



Specialized Programs: tRNAs

- tRNAscan-SE
 - tRNAscan and EufindtRNA identify candidates that are subsequently analysed by Cove.
 - 1 false positive per 15 billion nt
 - Detect 99% of true tRNA
 - www.genetics.wustl.edu/eddy/tRNAscan-SE/
 - rna.wustl.edu/GtRDB/ (Genomic tRNA database)
- FAStRNA (El-Mabrouk and Lisacek)
- tRNAscan (Fichant & Burks, 1991)



Specialized Programs: others

- tmRNA genes
 - BRUCE
 - Laslett, Canback, Andersson (2002) *NAR* **30**, 344903453.

Database search: summary

- Specialized programs: high specificity/sensitivity, fast
- SCFG-based approaches (such as INFERNAL): good specificity/sensitivity, work best if some sequence conservation is observed, slooow
- General motif searching tools (such as RNABOB): fast, writing descriptors is an art

RNA Motif Databases: Rfam

- A database of **multiple sequence alignments** and **covariance models**
- Rfam 9.1 contains 1372 families
- Search a query sequence to find instances of known motifs
- rfam.wustl.edu/ (database)
- infernald.wustl.edu/ (software)



RNA families database of alignments and CMs

[Home](#)[Keyword Search](#)[Sequence Search](#)[Browse Rfam](#)[ftp](#)[Help](#)[miRNA](#)[HCV_IRES family](#)

seed alignment for HCV_IRES

U89019/1-390	GCCA GCCCC CGAUUG GGGGC GACACUCCACCAUAGAUCACUCC CCUGUGAGG AACUACUGUCUU CACGCA GAA AGCGUC	Next
AF356827/1-391	GCCA GCCCC CGAUUG GGGGC GACACUCCACCAUAGAUCACUCC CCUGUGAGG AACUACUGUCUU CACGCA GAA AGCGUC	Next
D50466/1-389	ACCC GCCCC UUAUU . GGGGC GACACUCCACC . AUGAUCACUCC CCUGUGAGG AACUACUGUCUU CACGCA GAA AGCGUC	Next
D45193/1-390	ACCU GCUCU CUAUG . AGAGC AACACUCCACCAUGAACCGCUC CCUGUGAGG AACUUCUGUCUU CACGCA GAA AGCGUC	Next
AF290978/1-379 UUG GGGGC GACACUCCACCAUGAAUCACUCC CCUGUGAGG AACUACUGUCUU CACGCA GAA AGCGUC	Next
AF165047/1-379 UUG GGGGC GACACUCCACCAUAGAUCACUCC CCUGGAGG AAUACUGUCUU AACGCA GAA AGCGUC	Next
X61595/1-374 CGC GACACUCCACCAUAGAUCACUCC CCUGUGAGG AACUACUGUCUU CACGCA GAA AGCGUC	Next
D63822/1-388	GCCA GCCCC UUAAC . GGGGC GACACUCCACC . AUGAUCACUCC CCUGUGAGG AACUACUGUCUU CACGCA GAA AGCGUC	Next
D38078/1-388	GCCA GCCCC UAAU . GGGGC GACACUCCACC . AUGAUCACUCC CCUGUGAGG AACUACUGUCUU CACGCA GAA AGCGUC	Next
AF165050/1-379 UUG GGGGC GACAUUCCACCAUAGAUAUUCC CCUGUGAGG AAUACUGUUUU AACGCA GAA AGCGUU	Next
AF177037/1-391	GCCA GCCCC CUGAUG GGGGC GACACUCCACCAUGAAUCACUCC CCUGUGAGG AACUACUGUCUU CACGCA GAA AGCGUC	Next
D37841/1-392	GCCA GCCCC UUAAC . GGGGC GACACUCCACC . AUGAUCACUCC CCUGUGAGG AACUACUGUCUU CACGCA GAA AGCGUC	Next
D37843/1-390	GCCA GCCCC UUAAC . GGGGC GACACUCCACC . AUGAUCACUCC CCUGUGAGG AACUACUGUCUU CACGCA GAA AGCGUC	Next
D84263/1-388	GCCA GCCCC UAAU . GGGGC GACACUCCACC . AUGAUCACUCC CCUGUGAGG AACUACUGUCUU CACGCA GAA AGCGUC	Next
D84264/1-388	GCCA GCCCC UAAU . GGGGC GACACUCCACC . AUGAUCACUCC CCUGUGAGG AACUACUGUCUU CACGCA GAA AGCGUC	Next
AF208024/1-379 UUG GGGGC GACAUUCCACCAUAGAUAUUCC CCUGUGAGG AACUACUGUCUU CACGCA GAA AGCGUC	Next
D45172/1-391	GCCA GCCCC CUGAUG GGGGC GACACUCCACCAUAGAUCACUCC CCUGUGAGG AACUACUGUCUU CACGCA GAA AGCGUC	Next
D31971/1-388	GCCA GCCCC UAAAC . GGGGC GACACUCCACC . AUGAUCACUCC CCUGUGAGG AACUACUGUCUU CACGCA GAA AGCGUC	Next
SS_cons <<<< >>>> <<<< . <<<< <<<<< <<<<<<	Next

RNA Motif Databases: UTRdb and UTRsite

Pesole G., Liuni S., Grillo G., Licciulli F., Mignone F., Gissi C., and Saccone C. - "*UTRdb and UTRsite: specialized database of sequences and functional elements of 5' and 3' untranslated regions of eukaryotic mRNAs. Update 2002*".

[Nucleic Acids Res \(2002\)](#), 30(1):335-340.

<http://bighost.area.ba.cnr.it/BIG/UTRHome/>



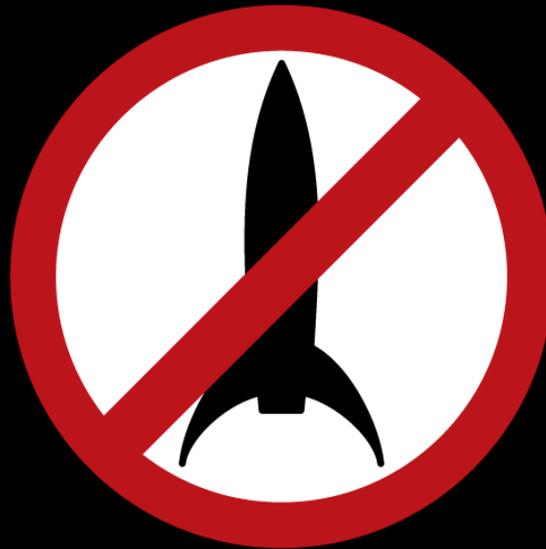
Specialized Motif Databases

- Methylation Guide snoRNA Database
 - snoscan (Lowe & Eddy, 1999)
 - <http://rna.wustl.edu/snoRNAdb/>
- tRNA databases
 - rna.wustl.edu/GtRDB/
- European Large Subunit Ribosomal RNA Database
- SRP database
- uRNA database
- Comparative RNA Web
- ...



Summary

- Sequence alignment methods are not appropriate for comparing divergent RNA sequences
- Tools such as RNAMOT, RNABOB and RNAMOTIF allows to describe and find RNA structure motifs in sequence databases
- RSEARCH finds all the sequences having a similar sequence and secondary structure to that of an input sequence and structure
- Homologous sequences and structures can be represented as a covariance model. The software program INFERNAL allows to find all the sequences that are likely to share the same overall fold (secondary structure)



**Rocket science is for kids
Bioinformatics is for scientists**



