

CSE 42I: Intro Algorithms

W. L. Ruzzo

Dynamic Programming:
String alignment and RNA Folding

Outline

A few slides on *applications* of dynamic programming in biology

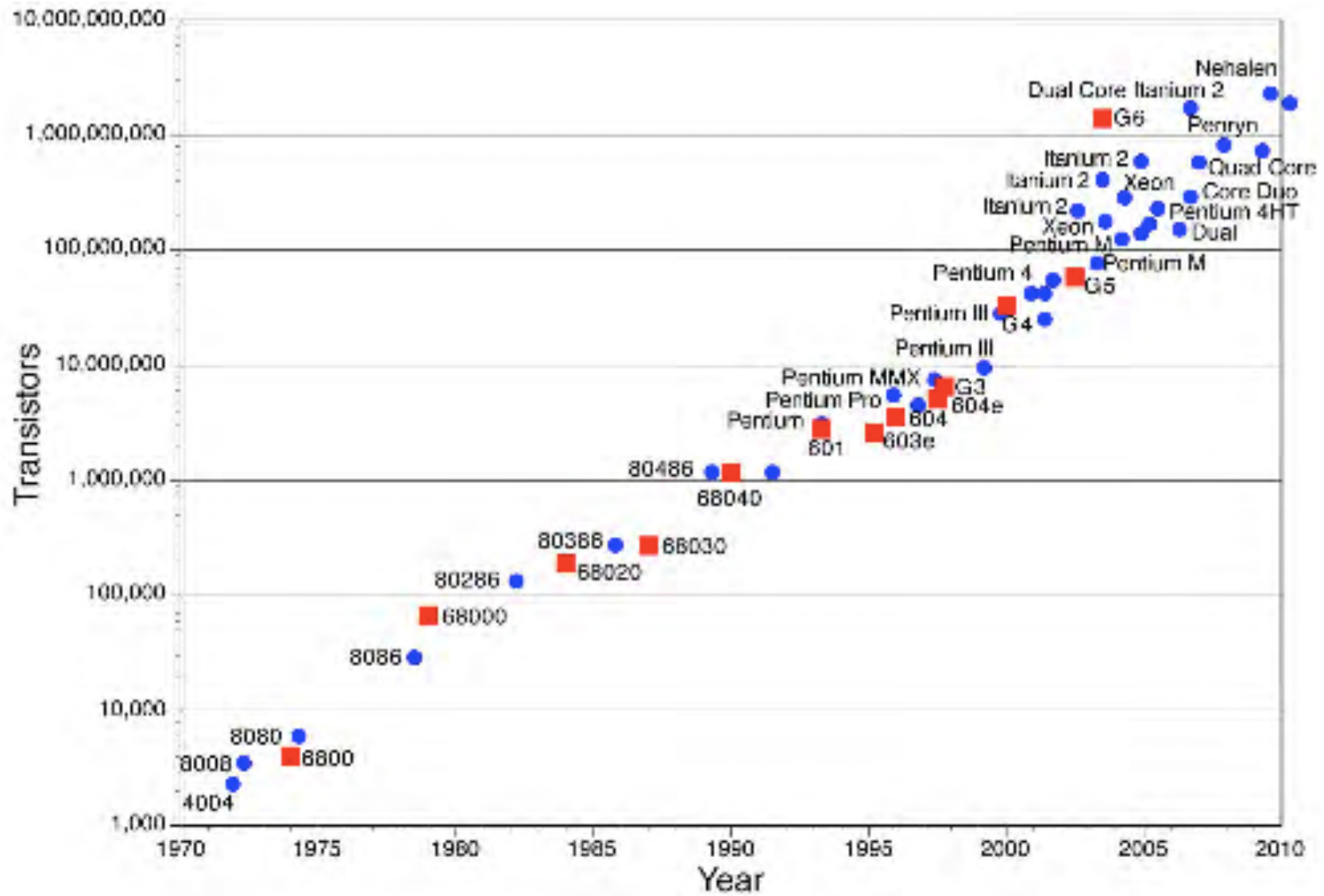
Sequence alignment

RNA structure

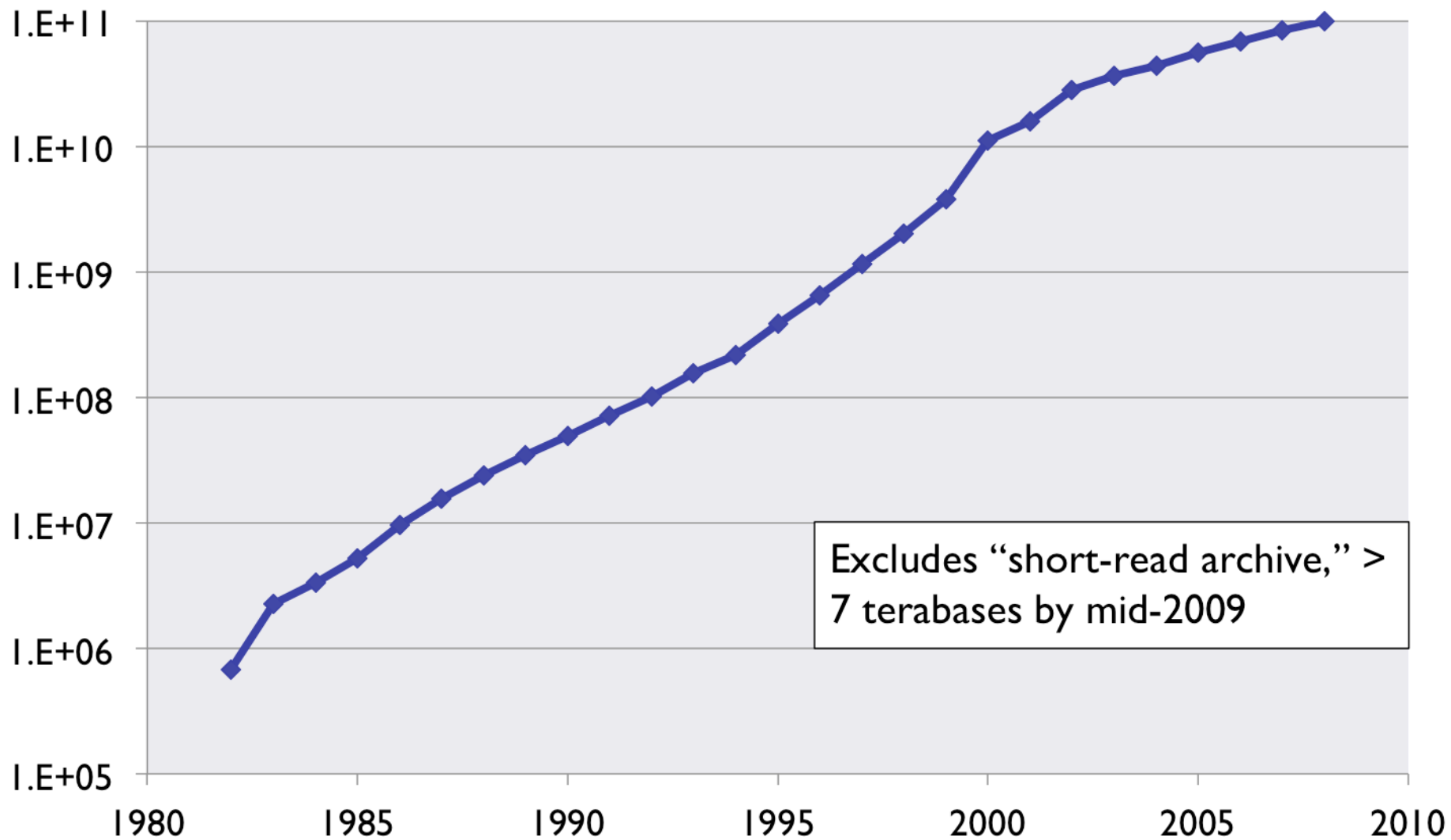
Algorithms for RNA structure

Application: Sequence Search

Moore's Law

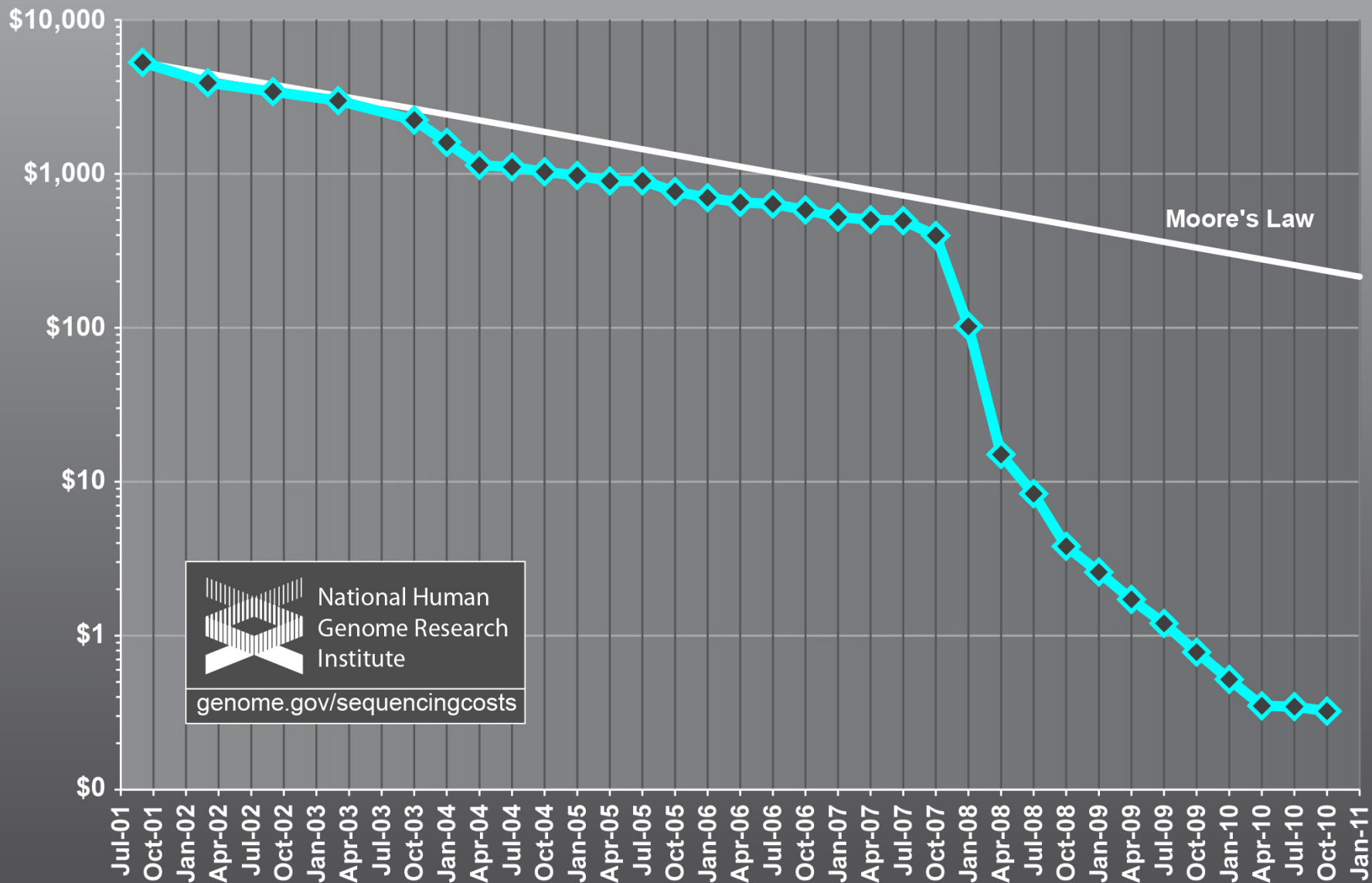


Growth of GenBank (Base Pairs)



Source: <http://www.ncbi.nlm.nih.gov/Genbank/genbankstats.html>

Cost per Megabase of DNA Sequence



 National Human
Genome Research
Institute
genome.gov/sequencingcosts

A Database Search

go to, e.g., <http://www.uniprot.org/>, paste in this:

```
>sp|P15172|MYOD1_HUMAN Myoblast determination protein 1 OS=Homo  
sapiens GN=MYOD1 PE=1 SV=3  
MELLSPPLRDVDLTAPDGSLCSFATTDDFYDDPCFDSPLRFFEDLDPRLMHVGALLKPE  
EHSHFPAAVHPAPGAREDEHVRAPSGHHQAGRCLLWACKACKRKT TNADRRKAATMRERR  
RLSKVNEAFETLKRCTSSNP NQRLPKVEILRN A IRYIEGLQALLRDQDAAPPGAAAFYA  
PGPLPPGRGGEHYSGDSDASSPRSNCS DGMMDYS GPPSGARRRNCYEGAYYNEAPSEPRP  
GKSAAVSSLDCLSSIVERISTESPAAPALLLADV PSESPPRRQEAAAPSEGESSGDPTQS  
PDAAPQCPAGANPNPIYQVL
```

A Few seconds Later...

Graphical overview

Color code for identity 0-100% =



Accession	Entry name	0Query hit320	0Match hit (sqrt scale)17392	Name (Organism)
<input type="checkbox"/> P15172	MYOD1_HUMAN			Myoblast determination protein 1 (Homo sapiens)
<input type="checkbox"/> B2RC72	B2RC72_HUMAN			cDNA, FLJ95884, highly similar to Hom... (Homo sapiens)
<input type="checkbox"/> E2RT59	E2RT59_CANFA			Uncharacterized protein (Canis familiaris)
<input type="checkbox"/> P49811	MYOD1_PIG			Myoblast determination protein 1 (Sus scrofa)
<input type="checkbox"/> D2KPI9	D2KPI9_PIG			Myogenic differentiation 1 (Sus scrofa)
<input type="checkbox"/> F1S9A9	F1S9A9_PIG			Uncharacterized protein (Sus scrofa)
<input type="checkbox"/> D2IOV4	D2IOV4_AILME			Putative uncharacterized protein (Ailuropoda melanoleuca)
<input type="checkbox"/> P29331	MYOD1_SHEEP			Myoblast determination protein 1 (Ovis aries)
<input type="checkbox"/> D2SP11	D2SP11_BUBBU			Myogenic factor MYOD1 (Bubalus bubalis)
<input type="checkbox"/> Q0VBX9	Q0VBX9_BOVIN			Myogenic differentiation 1 (Bos taurus)
<input type="checkbox"/> Q7YS82	MYOD1_BOVIN			Myoblast determination protein 1 (Bos taurus)
<input type="checkbox"/> Q8C6B1	Q8C6B1_MOUSE			Myogenic differentiation 1 (Mus musculus)
<input type="checkbox"/> A0JPK9	A0JPK9_RAT			Myogenic differentiation 1 (Rattus norvegicus)
<input type="checkbox"/> Q02346	MYOD1_RAT			Myoblast determination protein 1 (Rattus norvegicus)
<input type="checkbox"/> P10085	MYOD1_MOUSE			Myoblast determination protein 1 (Mus musculus)
<input type="checkbox"/> Q6DTY5	Q6DTY5_PIG			Eukaryotic myogenic factor MYF-3 (Sus scrofa)
<input type="checkbox"/> P21572	MYOD1_COTJA			Myoblast determination protein 1 homolog (Coturnix coturnix japonica)
<input type="checkbox"/> Q6DV59	Q6DV59_MELGA			MyoD (Meleagris gallopavo)
<input type="checkbox"/> P16075	MYOD1_CHICK			Myoblast determination protein 1 homolog (Gallus gallus)
<input type="checkbox"/> C5J072	C5J072_CHICK			Myogenic differentiation 1 (Gallus gallus)
<input type="checkbox"/> C3U0I1	C3U0I1_ANAPL			Myogenic differentiation 1 (Anas platyrhynchos)
<input type="checkbox"/> F1NHM3	F1NHM3_CHICK			Uncharacterized protein (Gallus gallus)
<input type="checkbox"/> F1NXM5	F1NXM5_CHICK			Uncharacterized protein (Gallus gallus)
<input type="checkbox"/> P13904	MYODA_XENLA			Myoblast determination protein 1 homolog A (Xenopus laevis)
<input type="checkbox"/> Q8AVZ0	Q8AVZ0_XENLA			Myod1-a protein (Xenopus laevis)
<input type="checkbox"/> Q7T109	Q7T109_XENTR			MyoD protein (Xenopus tropicalis)

...And 975 more...

Accession	Entry name	Status	Protein names	Organism	Length
Q7T109	Q7T109_XENTR	★	MyoD protein	Xenopus tropicalis (Western clawed frog) (<i>Xenopus tropicalis</i>)	288

Some Details from #25

Alignment 1 against Q7T109

Score	964	E-value	1.0 × 10 ⁻¹⁰²
Identity	64.0%	Positives	74.0%
Query length	320	Match length	288

Position Q7T109 matches from 1 to 288 (288AA), in the query sequence from 1 to 320 (320AA)

Graphical



1	MELLSPPLRDVDLTAPDGLCSFATDDFYDDPCFDSPLRFFEDLDPRLMHVGALLKPE	60	P15172
	MELL PPLRD+++T +GSLCSF T DDFYDDPCF++ D+ FFEDLDPRL+HV ALLKPE		
1	MELLPPPLRDMEVT--EGSLCSFPTPDDFYDDPCFNTSDMSFFEDLDPRLVHV-ALLKPE	57	Q7T109

61	EHSHFPAAVHPAPGAREDEHVRAPSGHHQAGRCLLWACKACKRKT TNADRRKAATMRERR	120	P15172
	+ H EDEHVRAPSGHHQAGRCLLWACKACKRKT TNADRRKAATMRERR		
58	DPHH-----NEDEHVRAPSGHHQAGRCLLWACKACKRKT TNADRRKAATMRERR	106	Q7T109

121	RLSKVNEAFETLKRCTSSNPQRLPKVEILRNAI RYIEGLQALLRDQDAAPPGAAAFYA	180	P15172
	RLSKVNEAFETLKRCTS+NPNQRLPKVEILRNAI RYIE LQ+LLR Q+ +FY		
107	RLSKVNEAFETLKRCTSTNPNQRLPKVEILRNAI RYIESLQSLLRGQE-----ESFY-	158	Q7T109

181	PGPLPPGRGGEHYSGDS DASSPRSNCS DGMMDYSGPPSGARRRNCYEGAYYNEAPSEPRP	240	P15172
	P+ EHYSGDS DASSPRSNCS DGM DYS PP G+RRRN Y+ ++Y+++P+ R		
159	--PVL-----EHYSGDS DASSPRSNCS DGMTDYS-PPCGSRRRNSYDSSFYS DSPNGLRL	210	Q7T109

241	GKSAAVSSLDCLSSIVERISTESPAAPALLLADV PSESPPRRQEAAAPSEGES---SGDP	297	P15172
	GKS+ +SSLDCLSSIVERISTESP P + AD SE P +P +GE+ SG		
211	GKSSVISSLDCLSSIVERISTESPVC PVIPAADSGSEGSP-----CSPLQGETLSESGII	265	Q7T109

The foregoing search capability is a *huge* deal

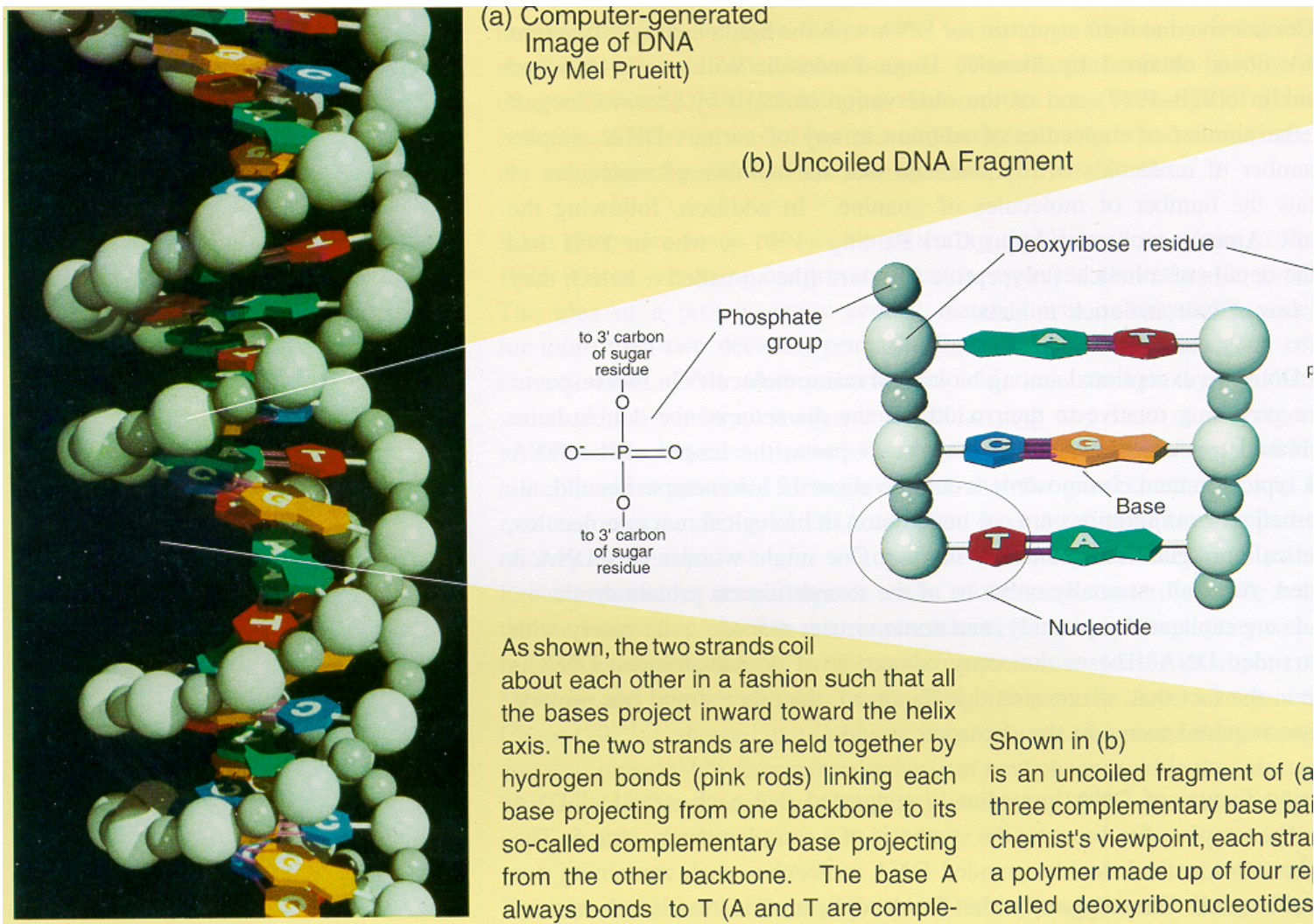
millions of searches

nearly all biologists (not just “computational
biologists”) use this routinely

It connects information about *all* living things

Application: RNA structure

The Double Helix



Central Dogma of Molecular Biology

by

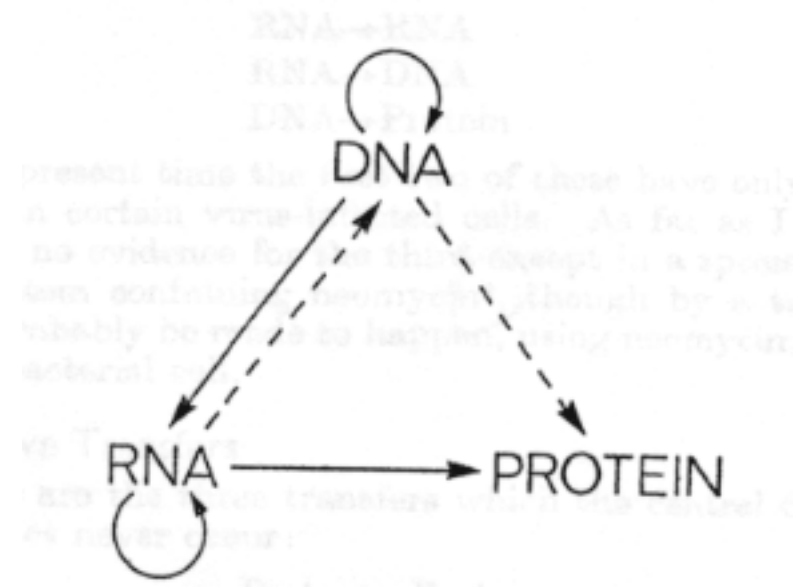
FRANCIS CRICK

MRC Laboratory
Hills Road,
Cambridge CB2 2QH

The central dogma of molecular biology deals with the detailed residue-by-residue transfer of sequential information. It states that such information cannot be transferred from protein to either protein or nucleic acid.

“The central dogma, enunciated by Crick in 1958 and the keystone of molecular biology ever since, is likely to prove a considerable over-simplification.”

Fig. 2. The arrows show the situation as it seemed in 1958. Solid arrows represent probable transfers, dotted arrows possible transfers. The absent arrows (compare Fig. 1) represent the impossible transfers postulated by the central dogma. They are the three possible arrows starting from protein.



Non-coding RNA

Messenger RNA - codes for proteins

Non-coding RNA - all the rest

Before, say, mid 1990' s, 1-2 dozen known
(critically important, but narrow roles)

Since mid 90' s dramatic discoveries

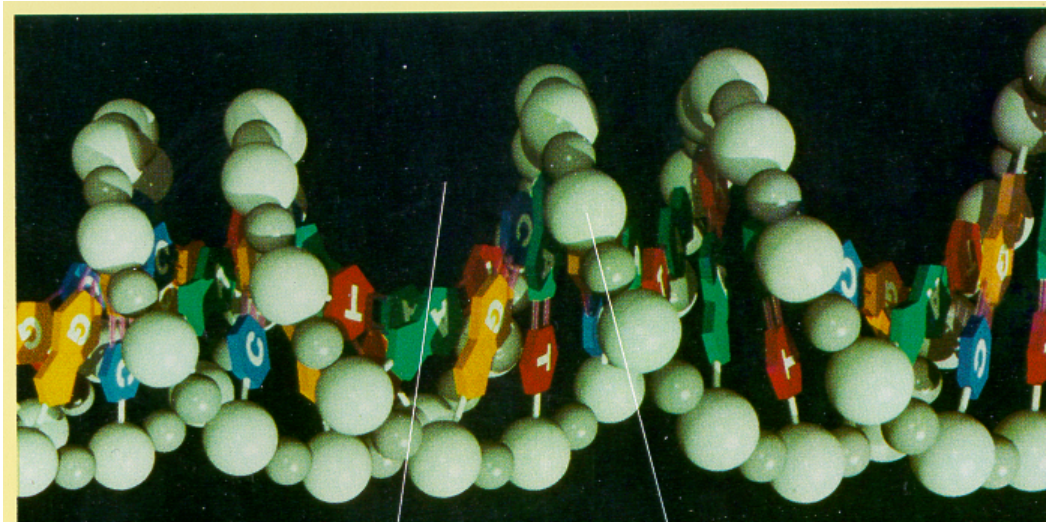
Regulation, transport, stability/degradation

E.g. “microRNA”: 100s in humans => 50% of genes

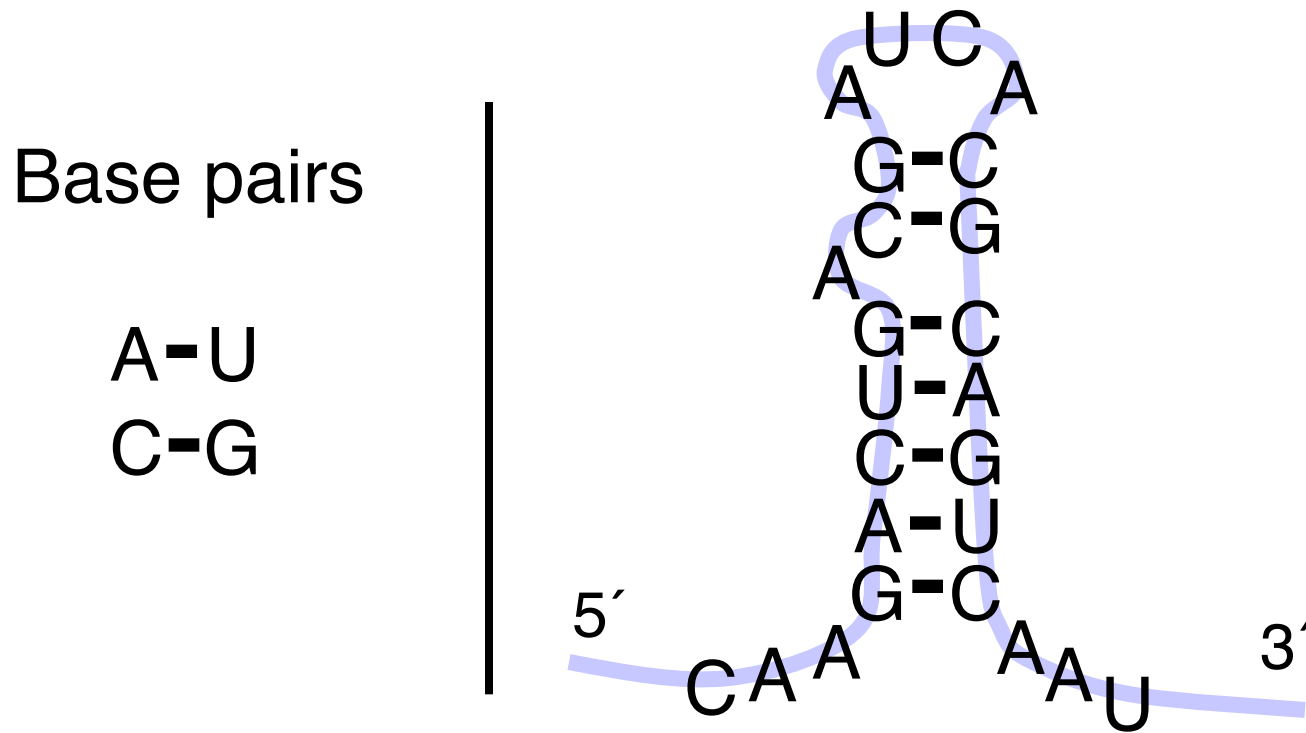
E.g. “riboswitches”: 1000s in bacteria

DNA structure: dull

5' ...ACCGCTAGATG...3'
| | | | | | | | | |
3' ...TGGCGATCTAC...5'



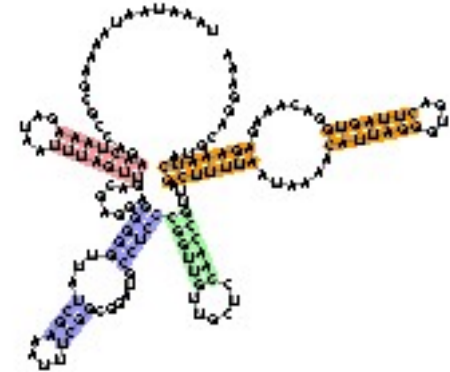
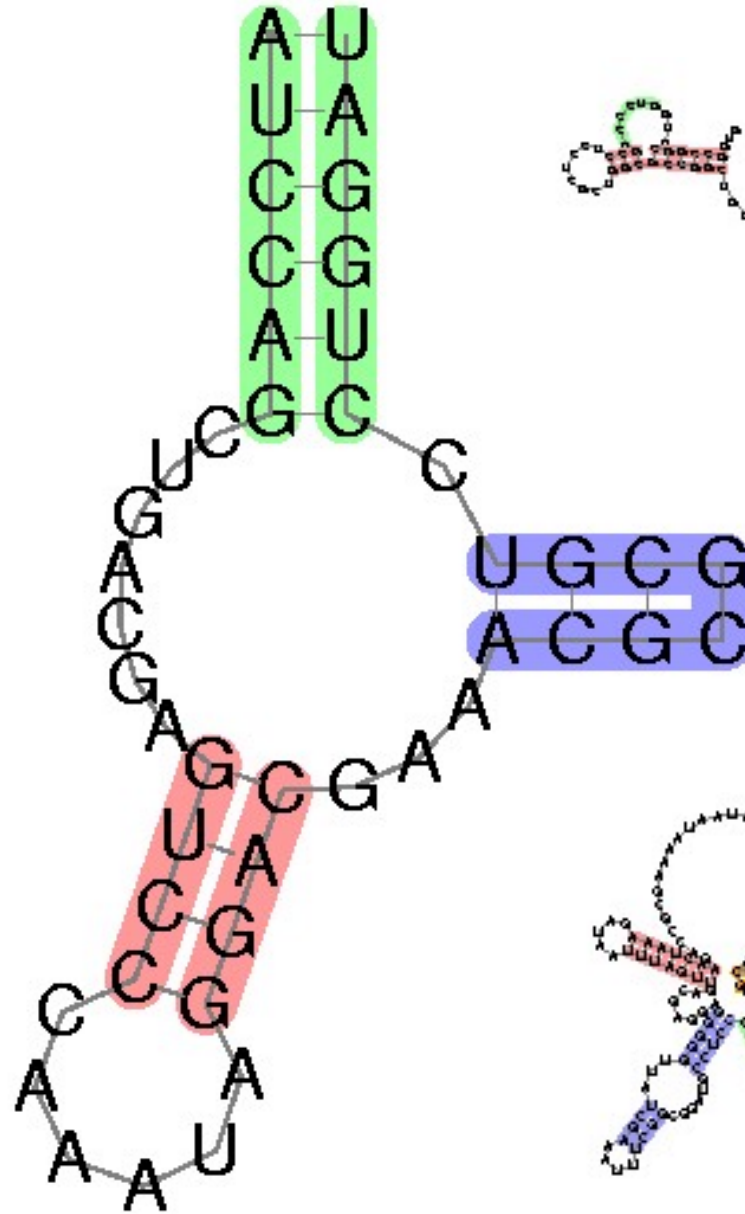
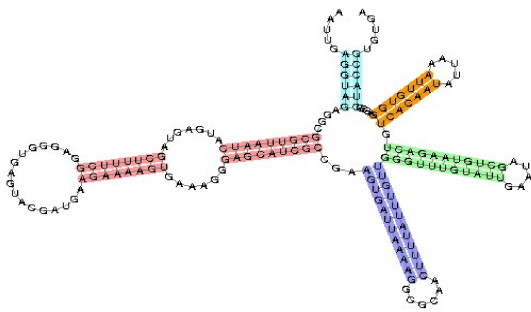
RNA Secondary Structure: RNA makes helices too



Usually *single* stranded

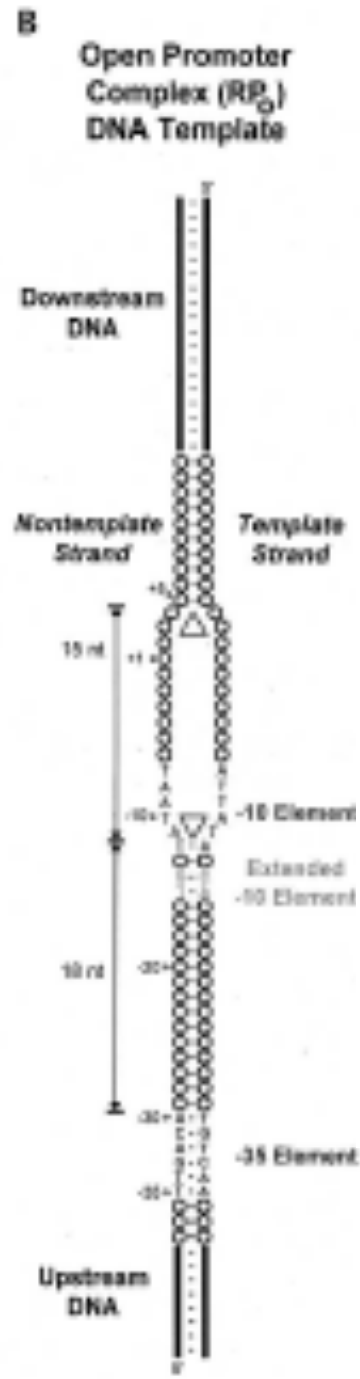
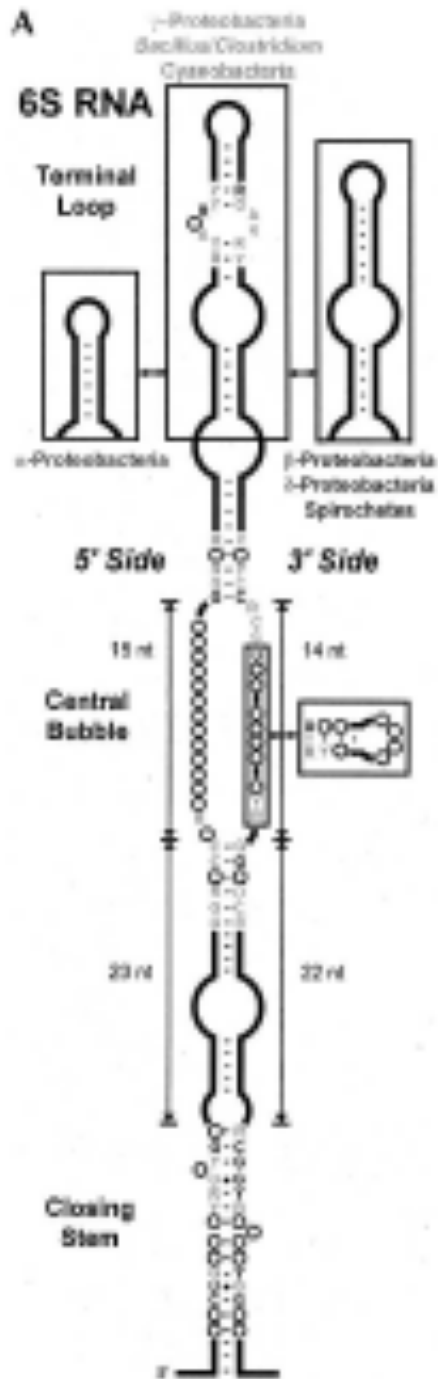
RNA Secondary Structure:

Not everything,
but important,
easier than 3d

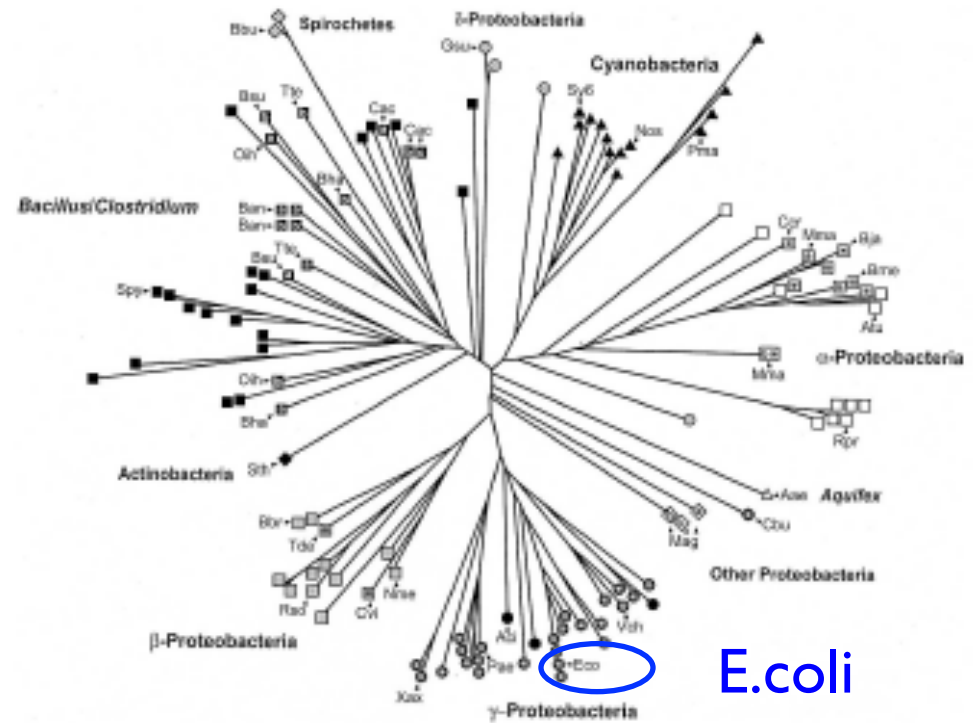


Why is structure important?

- For protein-coding, similarity in sequence is a powerful tool for finding related sequences
 - e.g. “hemoglobin” is easily recognized in all vertebrates
- For non-coding RNA, many different sequences have the same structure, and structure is most important for function.
 - So, using structure plus sequence, can find related sequences at much greater evolutionary distances



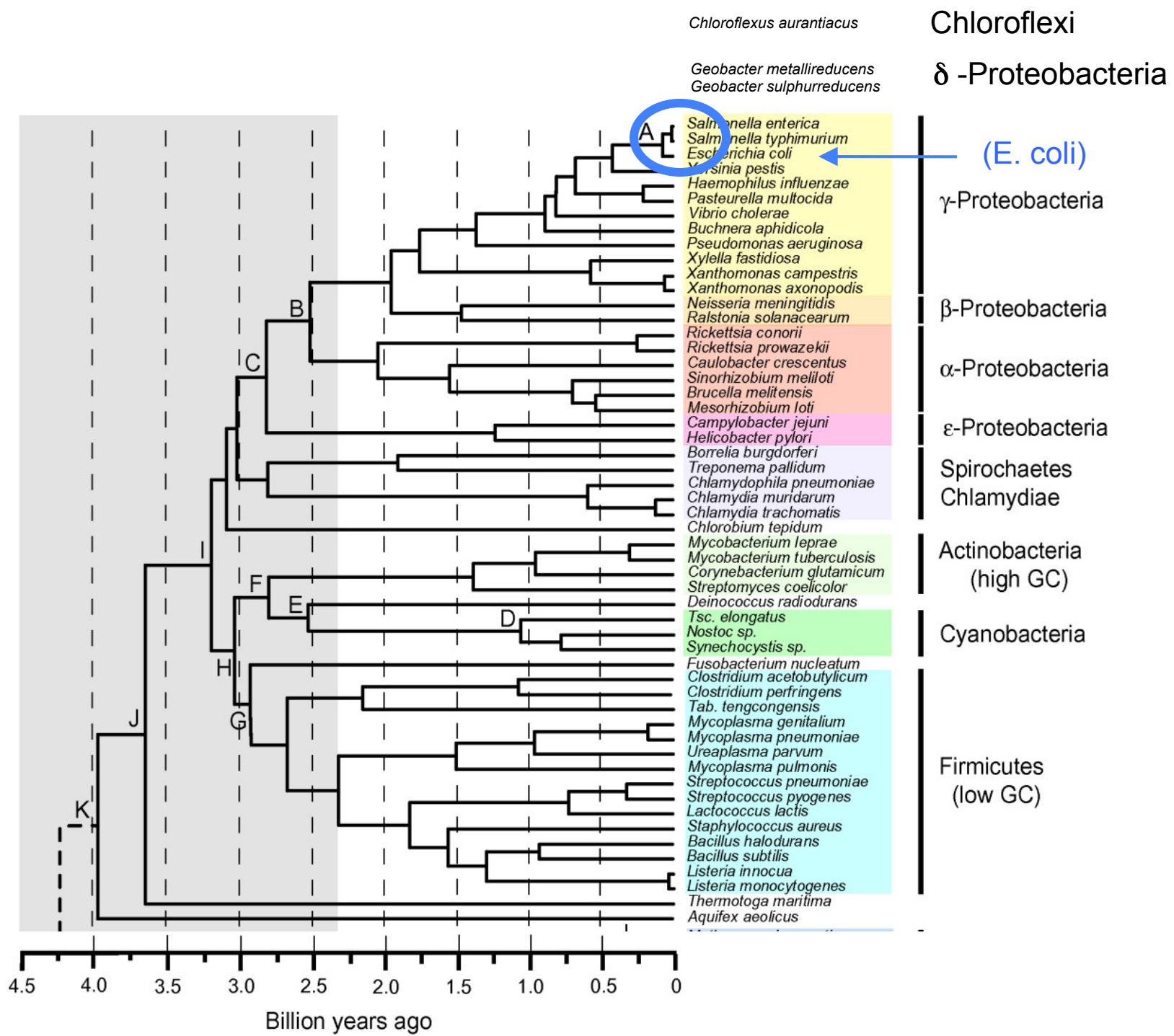
6S mimics an open promoter



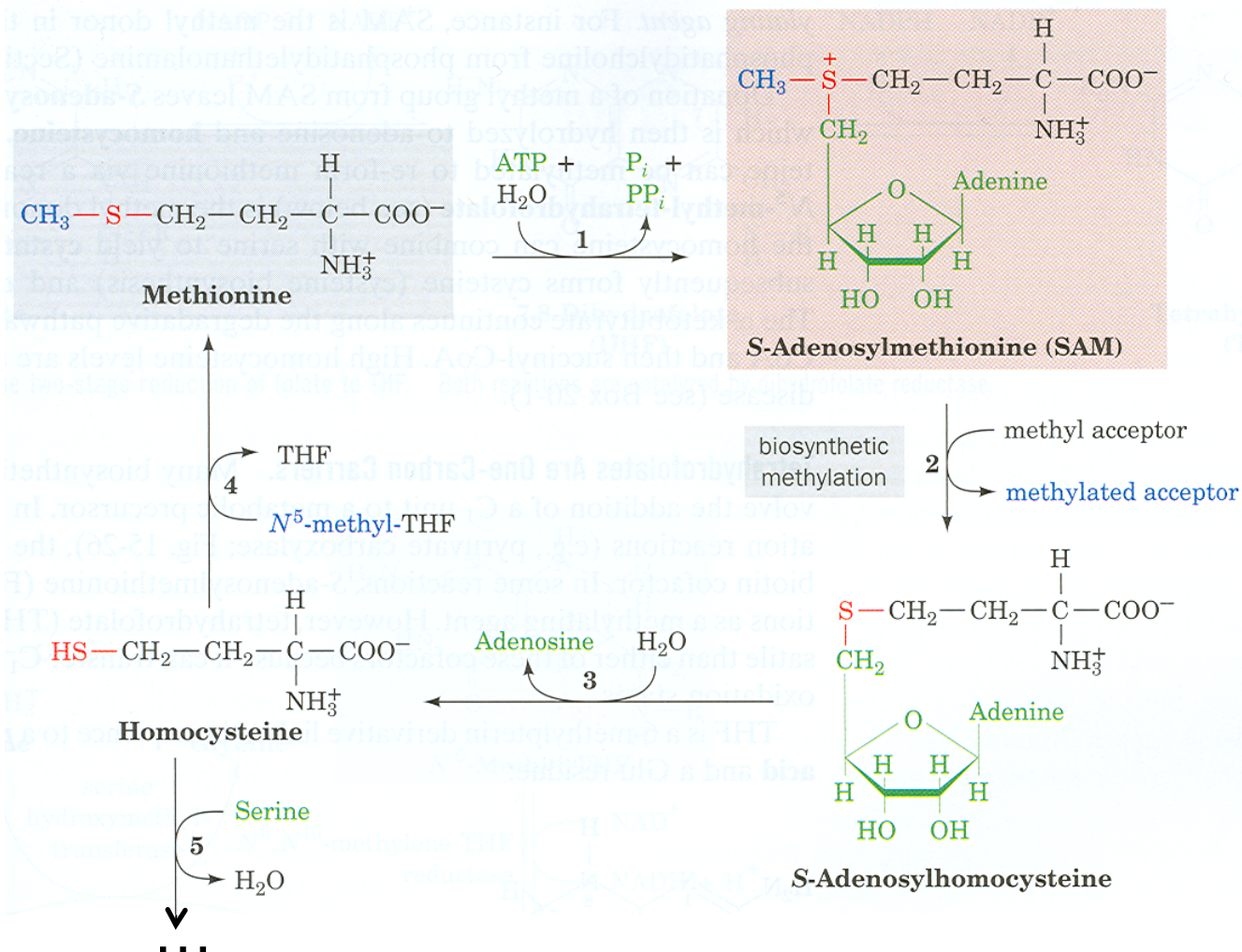
Barrick et al. *RNA* 2005

Trotochaud et al. *NSMB* 2005

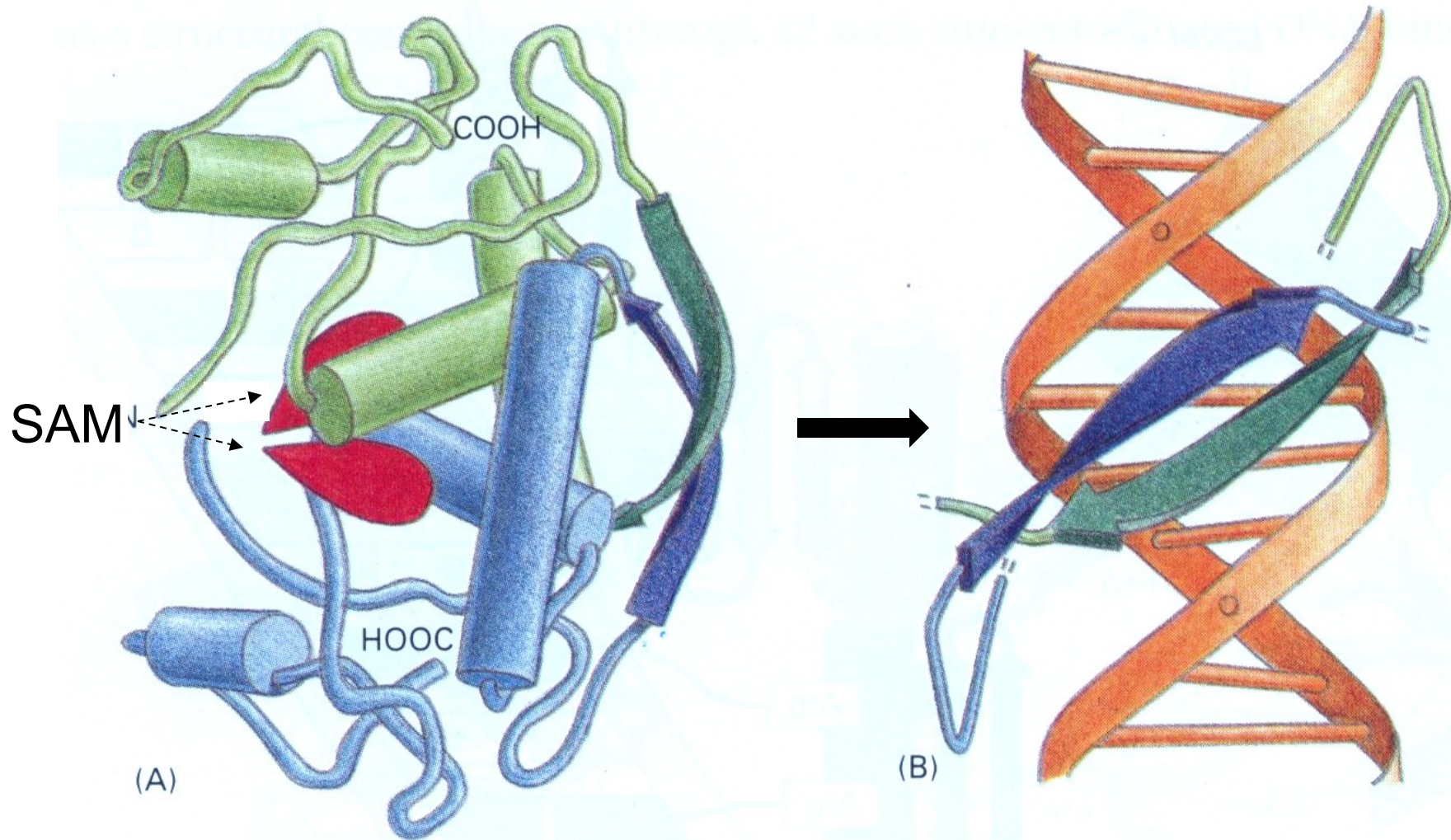
Willkomm et al. *NAR* 2005



In Bacteria: A typical biosynthetic cycle around a critical metabolite (“SAM”)



Gene Regulation: The MET Repressor



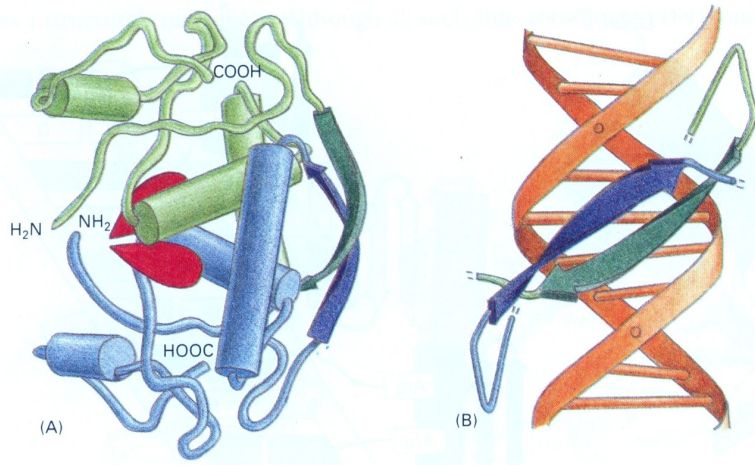
Protein

Alberts, et al, 3e.

DNA

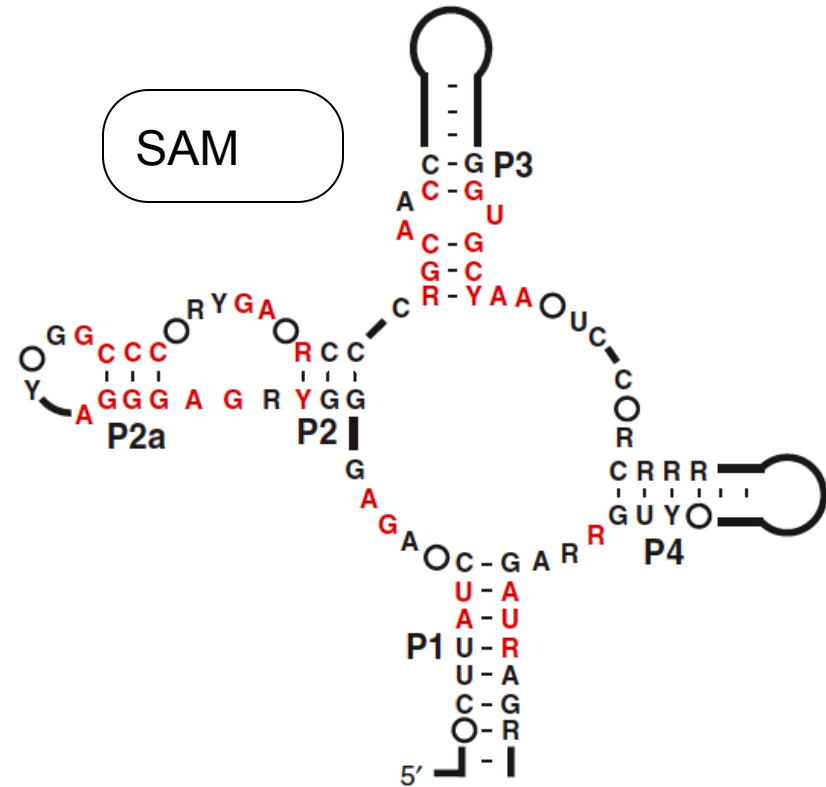
26

Alberts, et al, 3e.



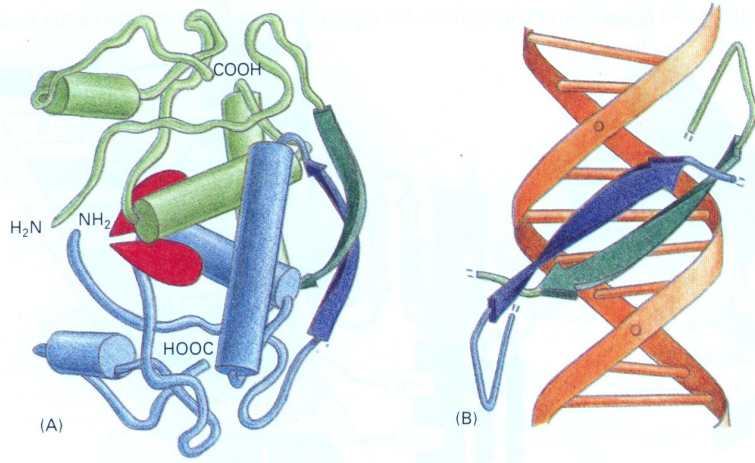
← The protein way

Riboswitch alternative



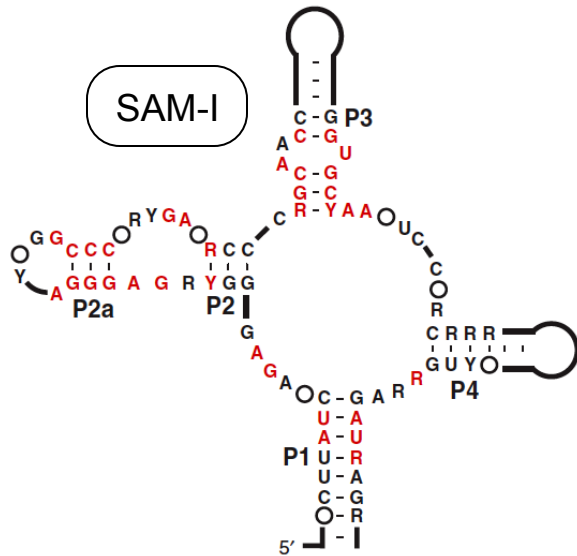
Grundy & Henkin, Mol. Microbiol 1998
Epshtein, et al., PNAS 2003
Winkler et al., Nat. Struct. Biol. 2003

Alberts, et al, 3e.

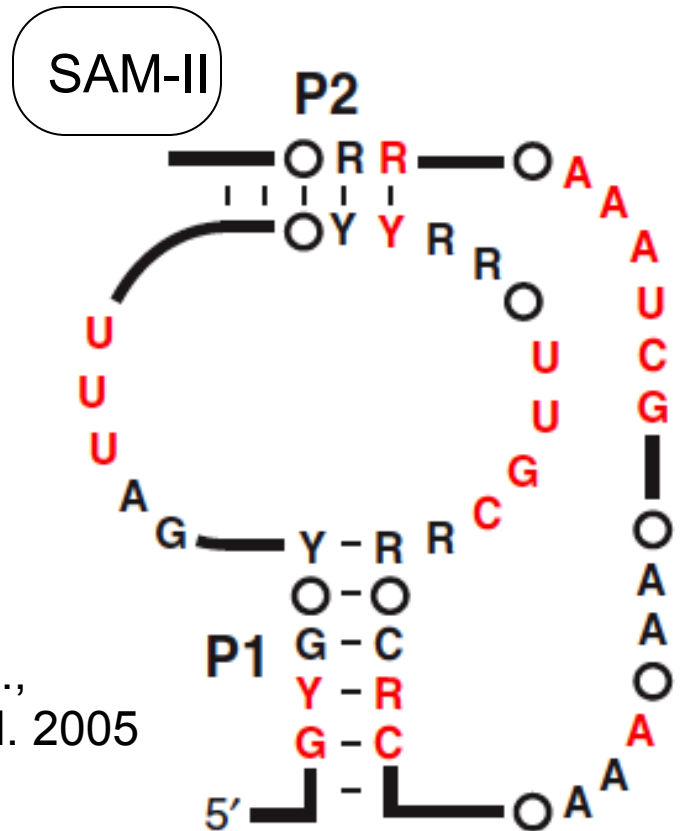


The protein way

Riboswitch alternatives

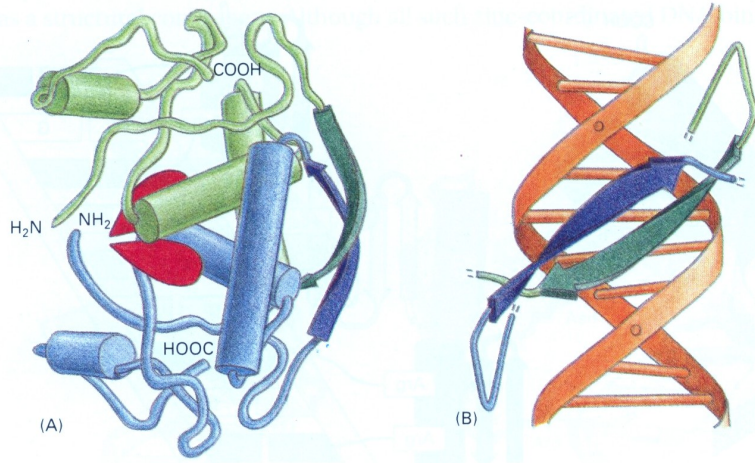


Grundy, Epshtein, Winkler et al., 1998, 2003



Corbino et al.,
Genome Biol. 2005

Alberts, et al, 3e.



← The protein way

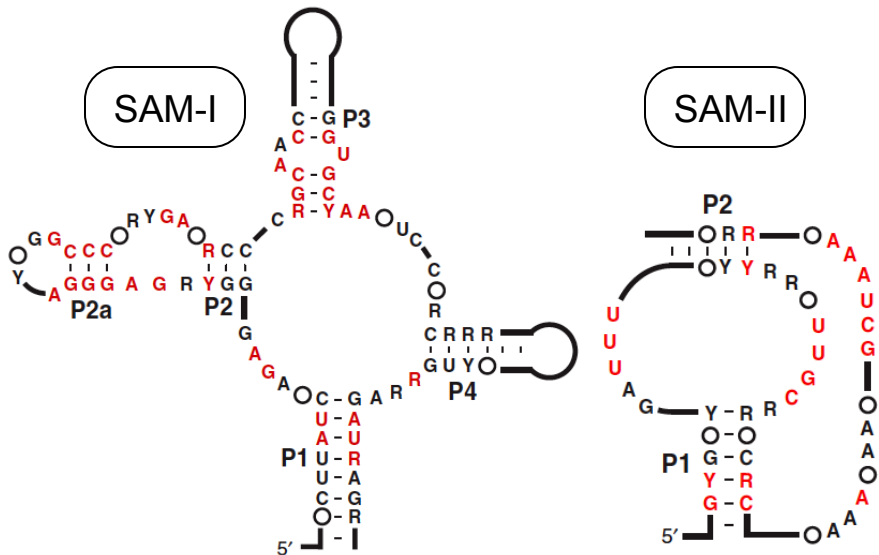
Riboswitch alternatives



SAM-III



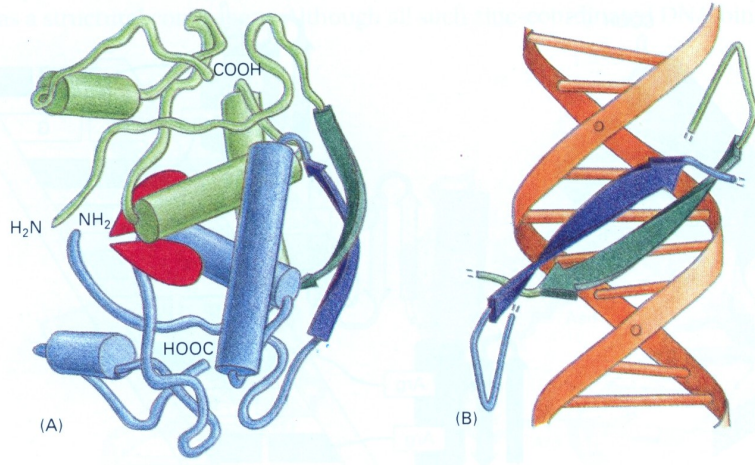
Fuchs et al., NSMB 2006



Grundy, Epshtein, Winkler et al., 1998, 2003

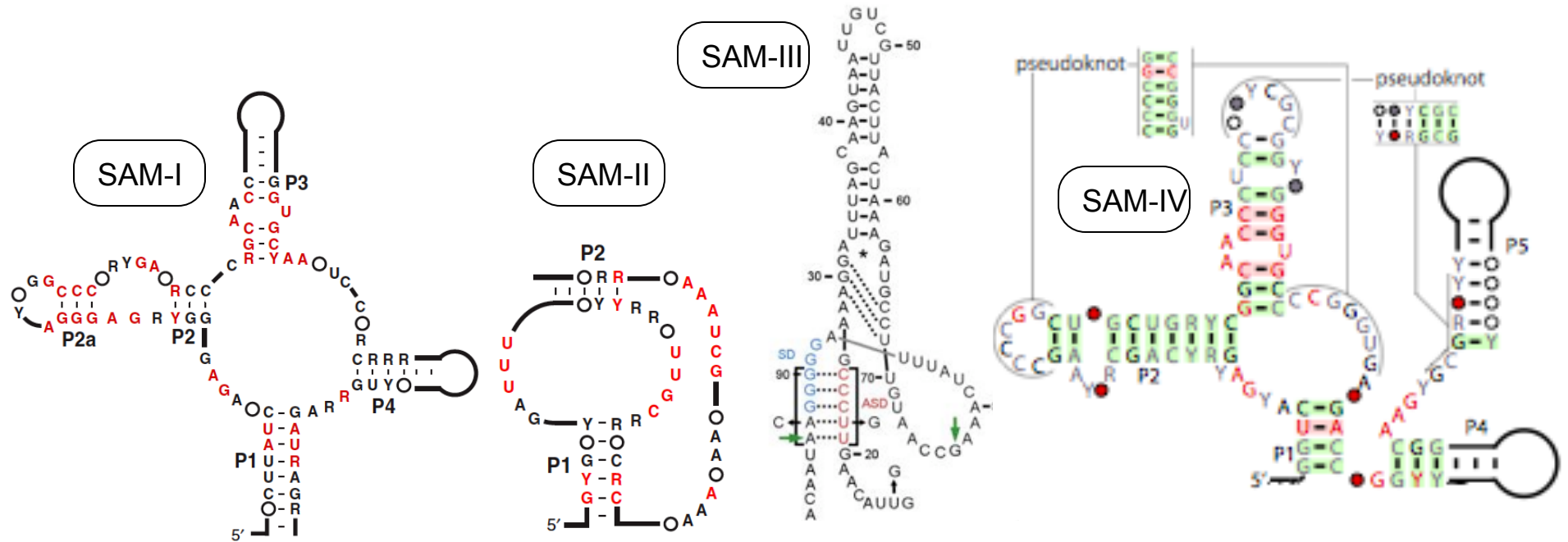
Corbino et al., Genome Biol. 2005

Alberts, et al, 3e.



The protein way

Riboswitch alternatives

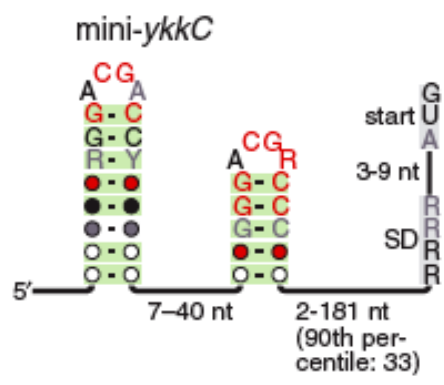
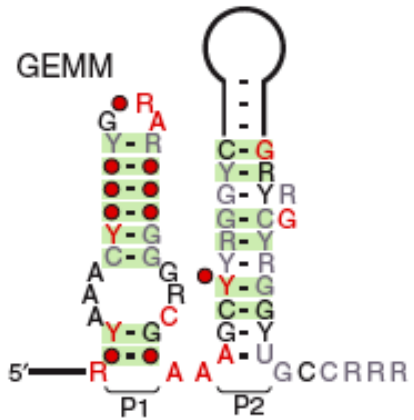


Grundy, Epshtein, Winkler et al., 1998, 2003

Corbino et al., Genome Biol. 2005

Fuchs et al., NSMB 2006

Weinberg et al.,³⁰ RNA 2008

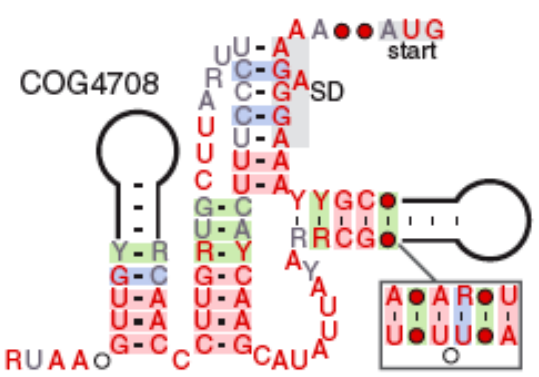
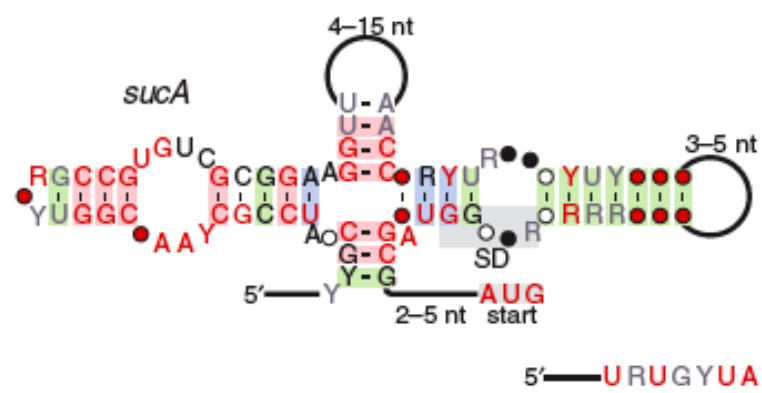
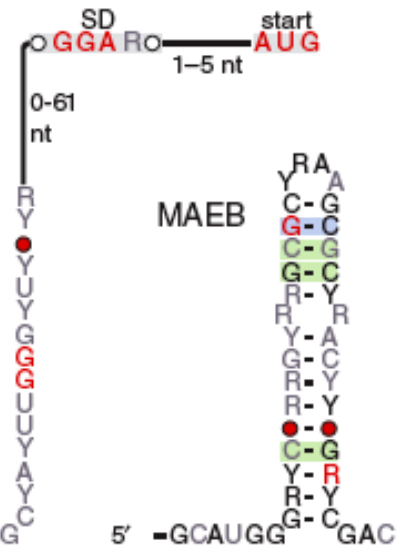
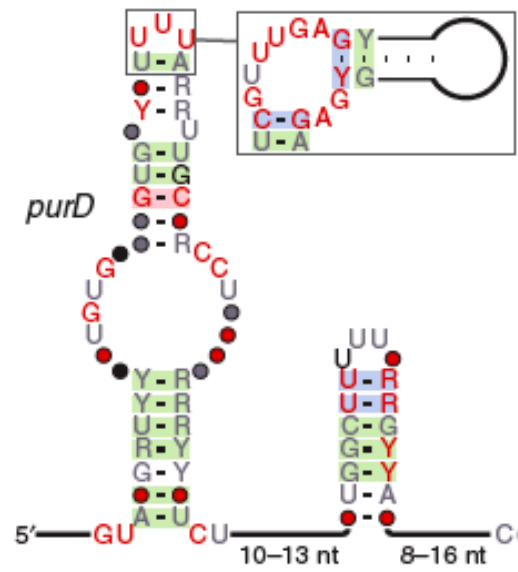
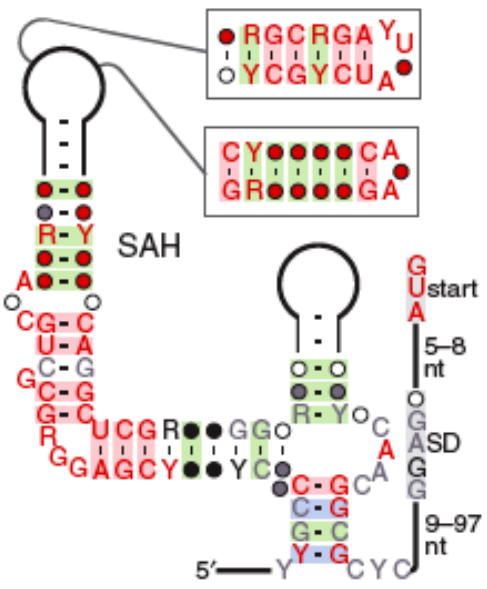


Legend

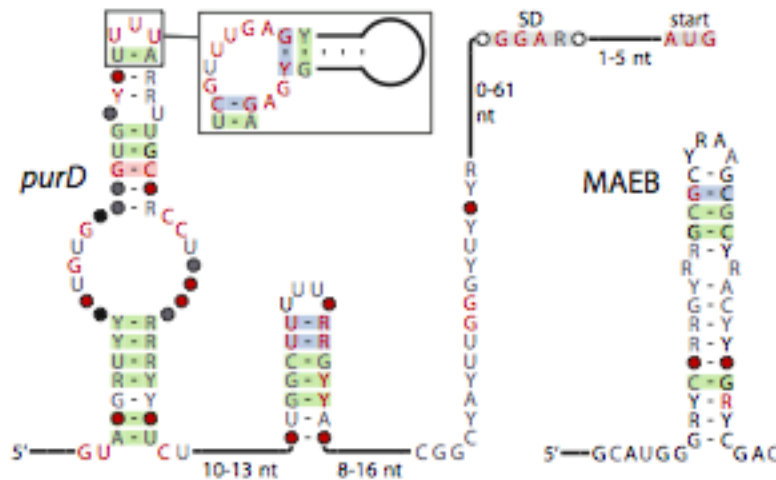
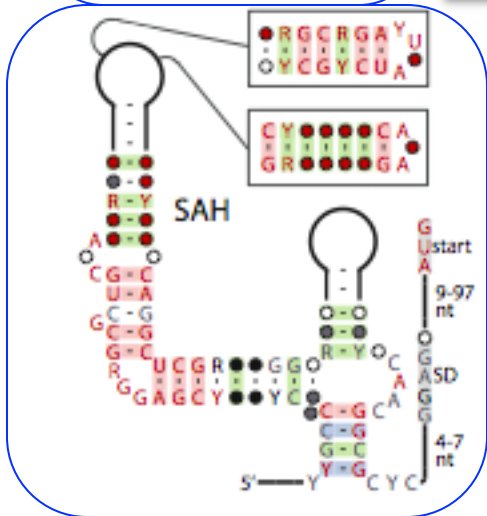
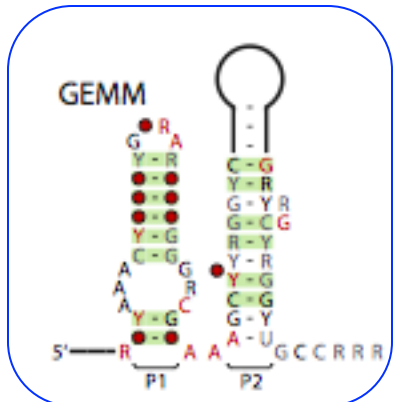
nt: nucleotides, R: A/G, Y: C/U
 For gray-shaded nucleotides, SD: Shine-Dalgarno, start: start codon

nucleotide identity	N 97%	has covarying mutations
N 90%	has compatible mutations	
N 75%	no mutations observed	

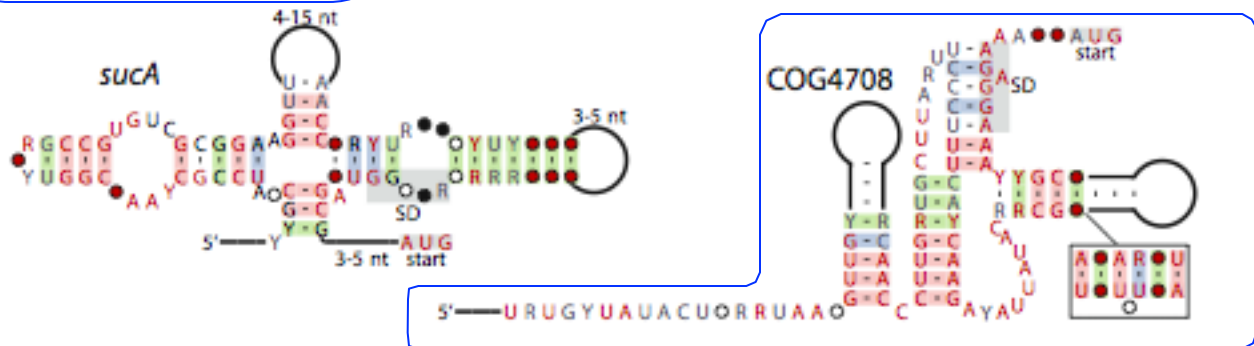
nucleotide present	● 97%	○ variable hairpin
● 90%	○ variable loop	
● 75%	□ modular structure	
○ 50%		



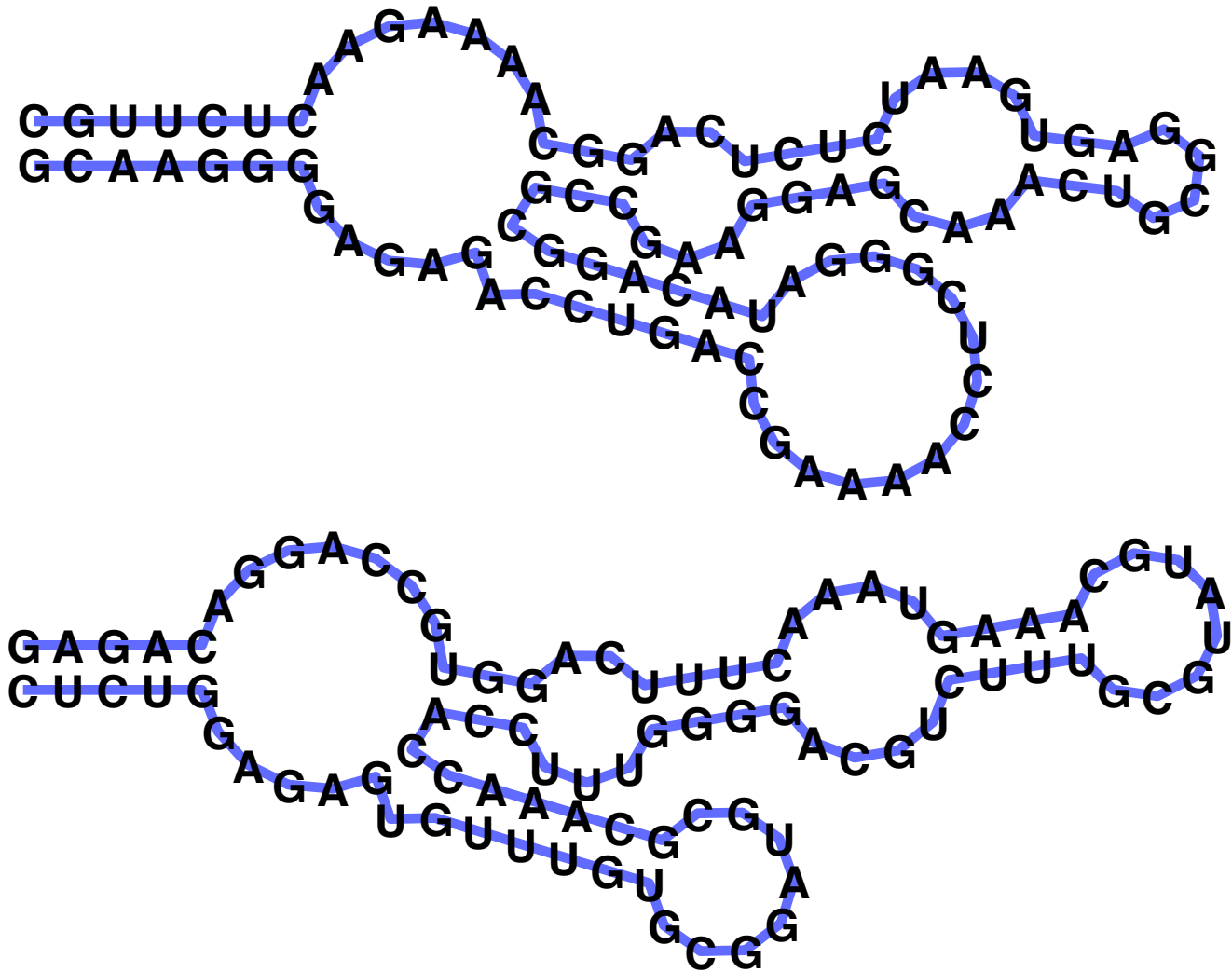
And many other examples. Widespread, deeply conserved, structurally sophisticated, functionally diverse, biologically important uses for ncRNA throughout prokaryotic world.



boxed = confirmed riboswitch (+2 more)



Q: What's so hard?



A: Structure often more important than sequence

Origin of Life?

Life needs

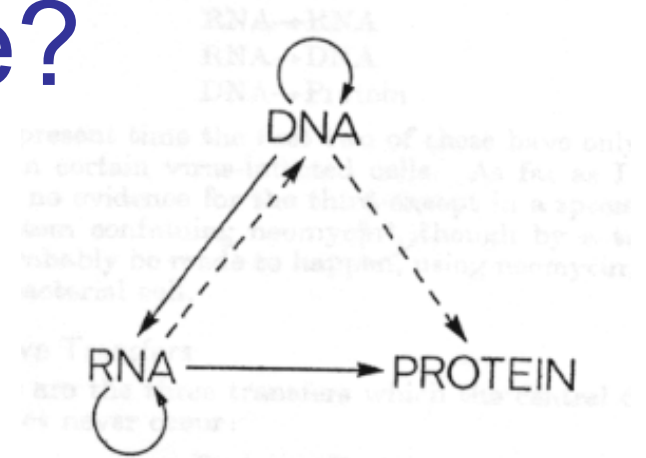
information carrier: DNA

molecular machines, like enzymes: Protein

making proteins needs DNA + RNA + proteins

making (duplicating) DNA needs proteins

Horrible circularities! How could it have arisen in an abiotic environment?



Origin of Life?

RNA can carry information, too

RNA double helix; RNA-directed RNA polymerase

RNA can form complex structures

RNA enzymes exist (ribozymes)

RNA can control, do logic (riboswitches)

The “RNA world” hypothesis:
1st life was RNA-based

6.5 RNA Secondary Structure

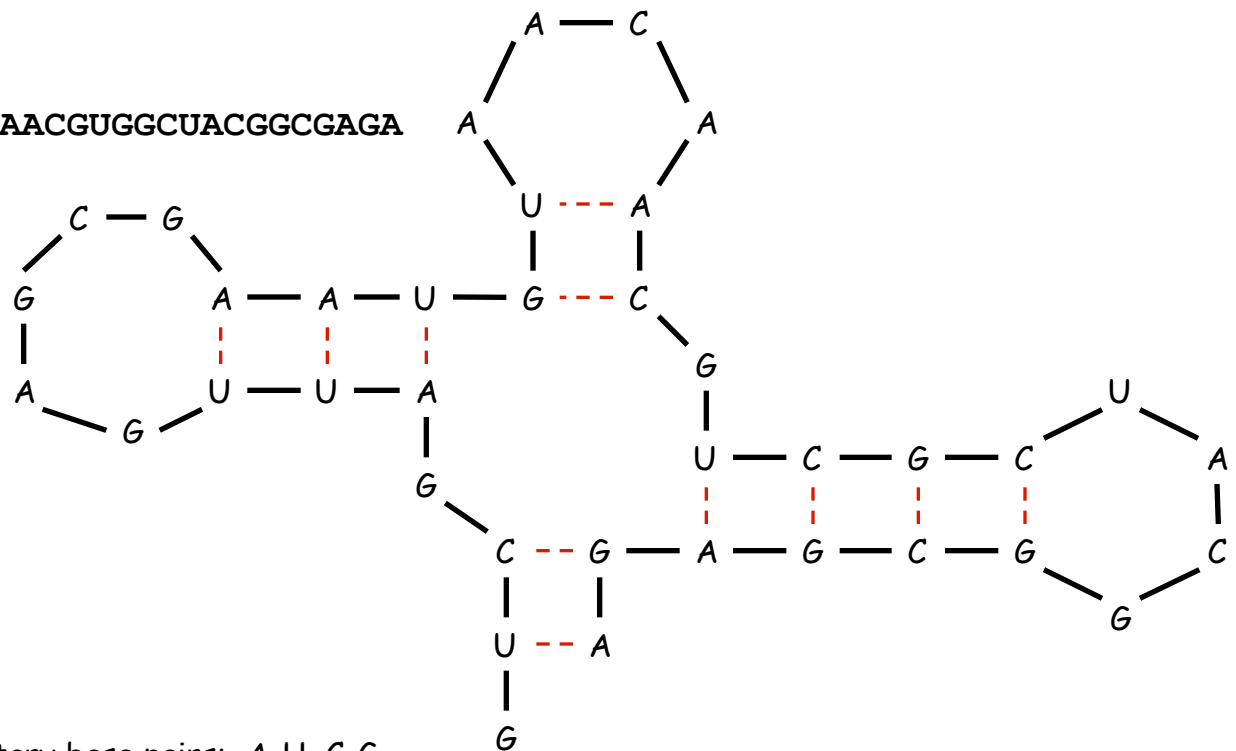
Nussinov's Algorithm – core technology
for RNA structure prediction

RNA Secondary Structure

RNA. String $B = b_1b_2\dots b_n$ over alphabet $\{ A, C, G, U \}$.

Secondary structure. RNA is usually single-stranded, and tends to loop back and form base pairs with itself. This structure is essential for understanding behavior of molecule.

Ex: GUCGAUUGAGCGAAUGUAACAACGUGGCUACGGCGAGA



complementary base pairs: A-U, C-G

RNA Secondary Structure (somewhat oversimplified)

Secondary structure. A set of pairs $S = \{ (b_i, b_j) \}$ that satisfy:

- [Watson-Crick.]
 - S is a *matching*, i.e. each base pairs with at most one other, and
 - each pair in S is a Watson-Crick pair: A-U, U-A, C-G, or G-C.
- [No sharp turns.] The ends of each pair are separated by at least 4 intervening bases. If $(b_i, b_j) \in S$, then $i < j - 4$.
- [Non-crossing.] If (b_i, b_j) and (b_k, b_l) are two pairs in S , then we cannot have $i < k < j < l$. (Violation of this is called a *pseudoknot*.)

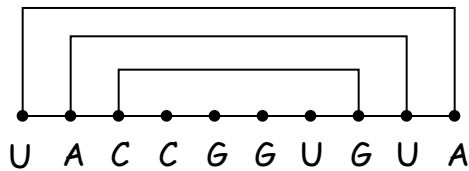
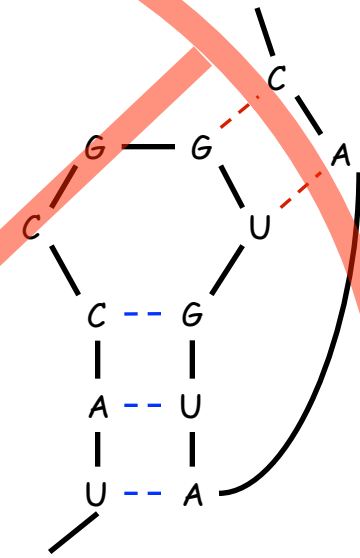
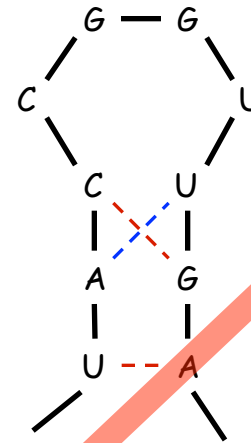
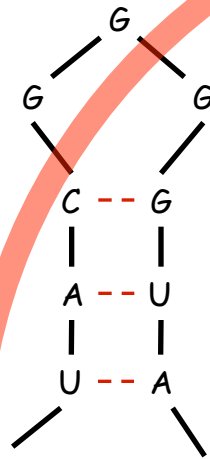
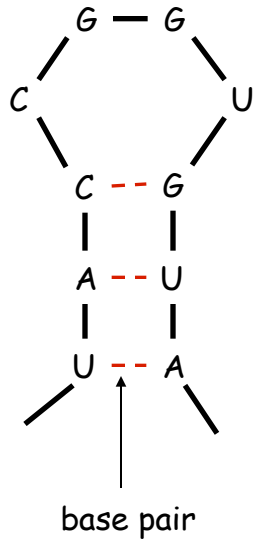
Free energy. Usual hypothesis is that an RNA molecule will form the secondary structure with the optimum total free energy.

↑
approximate by number of base pairs

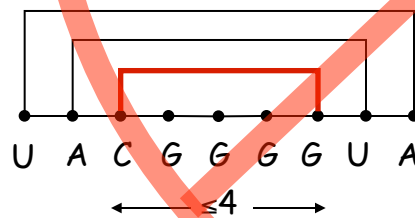
Goal. Given an RNA molecule $B = b_1b_2\dots b_n$, find a secondary structure S that maximizes the number of base pairs.

RNA Secondary Structure: Examples

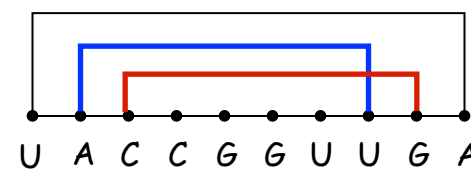
Examples.



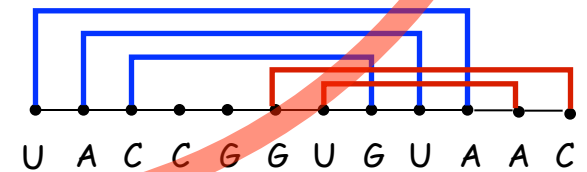
ok



sharp turn

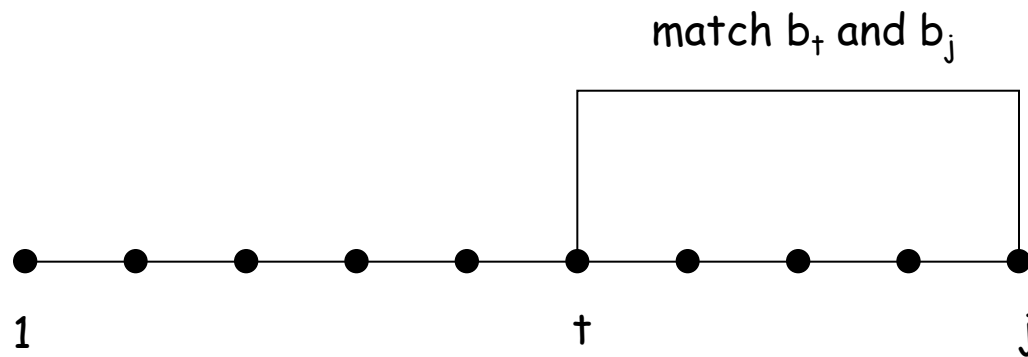


crossing



RNA Secondary Structure: Subproblems

First attempt. $OPT[j]$ = maximum number of base pairs in a secondary structure of the substring $b_1b_2\dots b_j$.



Difficulty. Results in two sub-problems.

- Finding secondary structure in: $b_1b_2\dots b_{t-1}$. ← $OPT(t-1)$
- Finding secondary structure in: $b_{t+1}b_{t+2}\dots b_{j-1}$. ← not OPT of anything; need more sub-problems

Dynamic Programming Over Intervals: (R. Nussinov's algorithm)

Notation. $OPT[i, j]$ = maximum number of base pairs in a secondary structure of the substring $b_i b_{i+1} \dots b_j$.

- Case 1. If $i \geq j - 4$.

$$OPT[i, j] = 0 \text{ by no-sharp turns condition.}$$

- Case 2. Base b_j is not involved in a pair.

$$OPT[i, j] = OPT[i, j-1]$$

- Case 3. Base b_j pairs with b_t for some $i \leq t < j - 4$.
non-crossing constraint decouples resulting sub-problems

$$OPT[i, j] = 1 + \max_t \{ OPT[i, t-1] + OPT[t+1, j-1] \}$$

↑
take max over t such that $i \leq t < j-4$ and
 b_t and b_j are Watson-Crick complements

Key point:
Either last base
is unpaired
(case 1,2) or
paired (case 3)

Remark. Same core idea in CKY algorithm to parse context-free grammars.

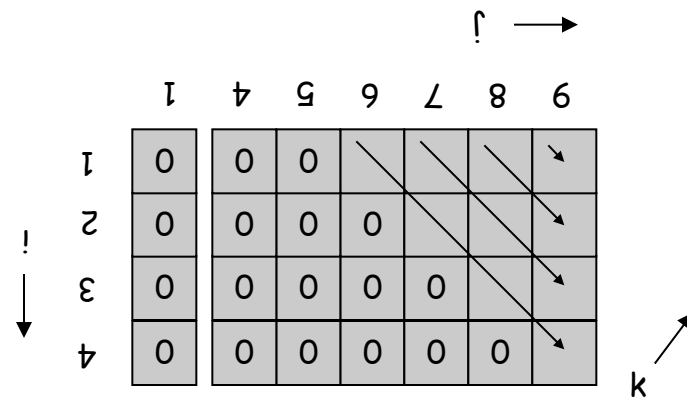
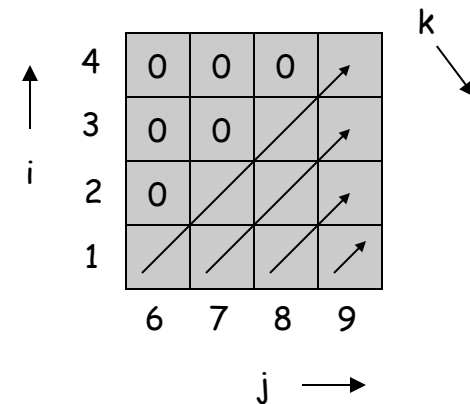
Bottom Up Dynamic Programming Over Intervals

Q. What order to solve the sub-problems?

A. Do shortest intervals first.

```

RNA( $b_1, \dots, b_n$ ) {
  for  $k = 5, 6, \dots, n-1$ 
    for  $i = 1, 2, \dots, n-k$ 
       $j = i + k$ 
      Compute  $OPT[i, j]$ 
    return  $OPT[1, n]$  using recurrence
}
    
```



Running time. $O(n^3)$.

C U C C G G U U G C A A U G U C
 ((. (. . .) .) . .) . .

n = 16

0	0	0	0	0	1	1	1	1	1	2	2	2	3	3	3
	0	0	0	0	0	0	0	1	1	2	2	2	2	2	2
		0	0	0	0	0	0	1	1	1	1	1	2	2	2
			0	0	0	0	0	1	1	1	1	1	2	2	2
				0	0	0	0	0	1	1	1	1	1	1	2
					0	0	0	0	0	1	1	1	1	1	2
						0	0	0	0	0	1	1	1	1	1
							0	0	0	0	0	1	1	1	1
								0	0	0	0	0	0	0	1
									0	0	0	0	0	0	1
										0	0	0	0	0	0
											0	0	0	0	0
												0	0	0	0
													0	0	0
														0	0
															0

E.g.:
 OPT[1,6] = 1:
 CUCCGG
 (.....)

E.g.:
 OPT[6,16] = 2:
 GUUGCAAUGUC
 ((.....).....)

Computing one cell: OPT[2,18] = ?

G G G A A A A C C C A A A G G G G U U U n= 20

(((. . .))) (((. . .)))

0	0	0	0	0	0	0	1	2	3	3	3	3	3	3	3	4	5	6	
0	0	0	0	0	0	0	1	2	2	2	2	2	2	3	3	4	5	6	
0	0	0	0	0	0	0	1	1	1	1	1	1	2	2	3	3	4	5	6
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	5	6
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	5	5
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	4	4
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	3	3	3
0	0	0	0	0	0	0	0	0	0	0	0	0	1	1	2	2	2	2	3
0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	1	1	1	2	3
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	3
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2

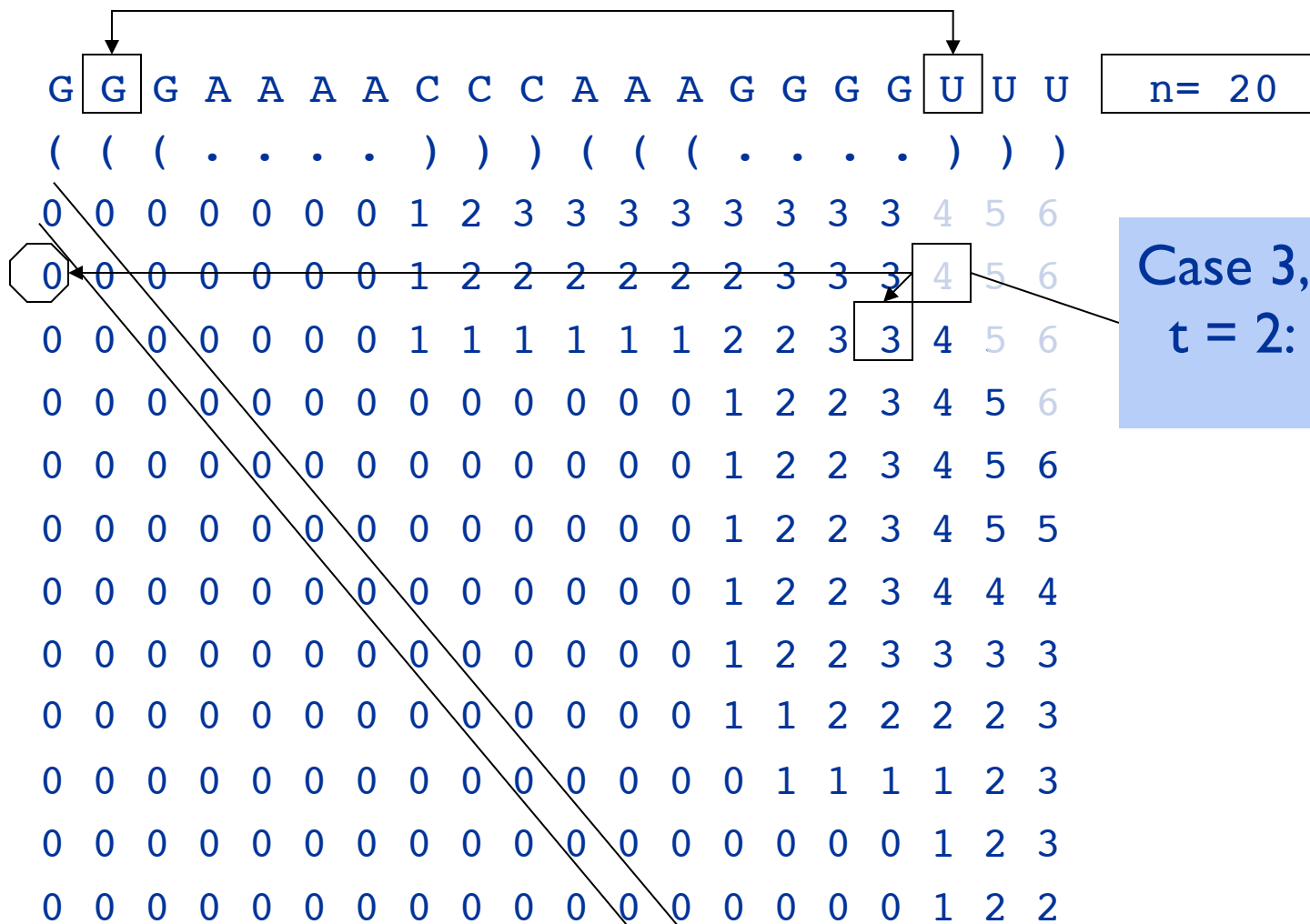
Case 1:
 $2 \geq 18-4?$ no.
 Case 2:
 B_{18} unpaired?
 Always a possibility;
 then $OPT[2,18] \geq 3$

GGAAAACCCAAAGGGGU
 ((...))(...)...

$$OPT(i,j) = \begin{cases} 0 & \text{if } i \geq j - 4 \\ \max \left\{ \begin{array}{l} OPT[i,j-1] \\ 1 + \max_t (OPT[i,t-1] + OPT[t+1,j-1]) \end{array} \right\} & \text{otherwise} \end{cases}$$

0
 0

Computing one cell: OPT[2,18] = ?



Case 3, $2 \leq t < 18-4$:
 $t = 2$: no pair

$$OPT(i,j) = \begin{cases} 0 & \text{if } i \geq j - 4 \\ \max \left\{ \begin{array}{l} OPT[i,j-1] \\ 1 + \max_t (OPT[i,t-1] + OPT[t+1,j-1]) \end{array} \right\} & \text{otherwise} \end{cases}$$

0
 0

Computing one cell: OPT[2,18] = ?

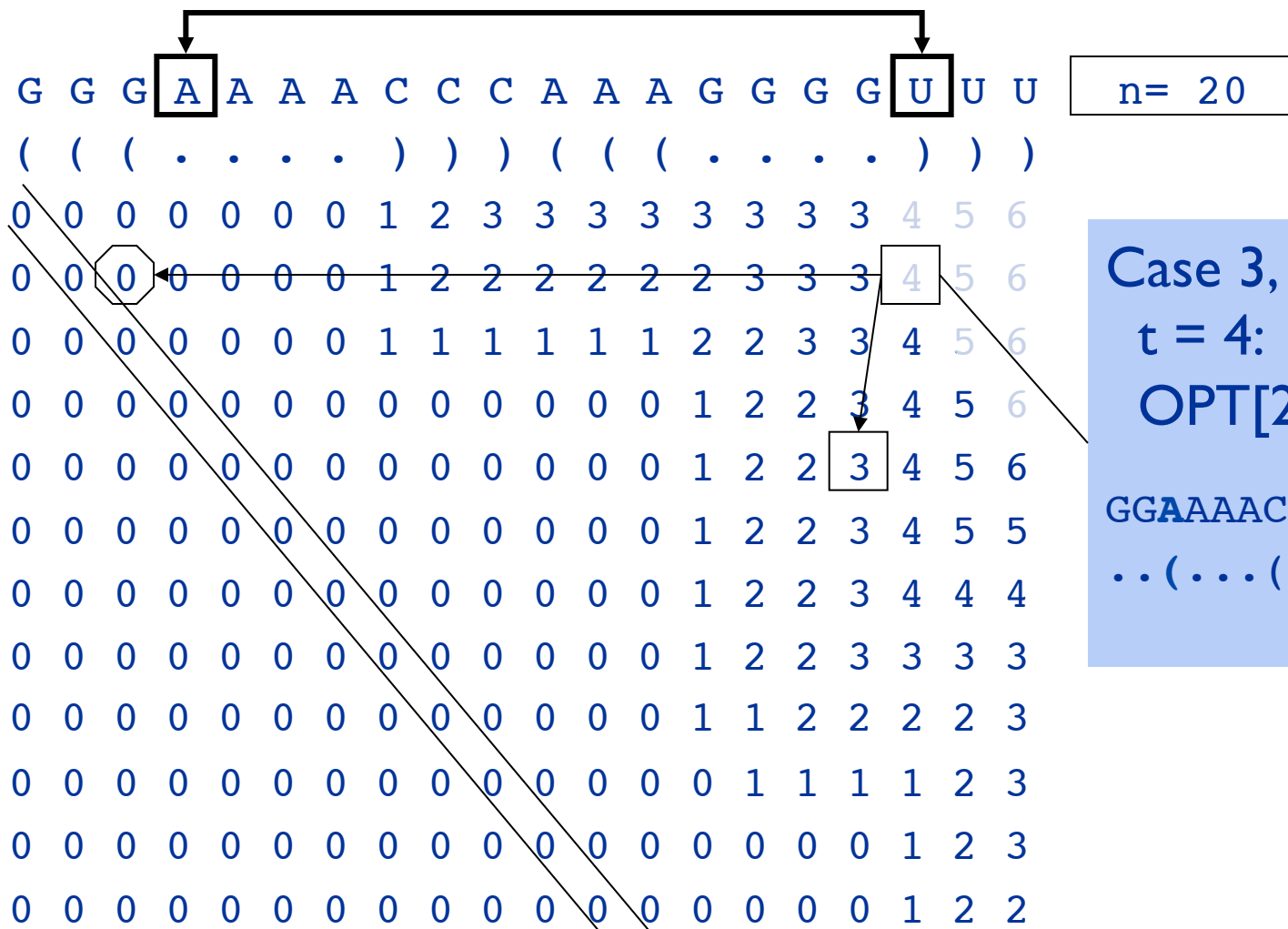


Case 3, $2 \leq t < 18-4$:
 $t = 3$: no pair

$$OPT(i, j) = \begin{cases} 0 & \text{if } i \geq j - 4 \\ \max \left\{ \begin{array}{l} OPT[i, j - 1] \\ 1 + \max_t (OPT[i, t - 1] + OPT[t + 1, j - 1]) \end{array} \right\} & \text{otherwise} \end{cases}$$

0
 0

Computing one cell: OPT[2,18] = ?



Case 3, $2 \leq t < 18-4$:
 $t = 4$: yes pair
 $OPT[2,18] \geq 1 + 0 + 3$

GGAAAACCCAAAGGGGU
 $\dots((\dots(((\dots))))))$

$$OPT(i,j) = \begin{cases} 0 & \text{if } i \geq j - 4 \\ \max \left\{ \begin{array}{l} OPT[i,j-1] \\ 1 + \max_t (OPT[i,t-1] + OPT[t+1,j-1]) \end{array} \right\} & \text{otherwise} \end{cases}$$

0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0

0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0

Computing one cell: OPT[2,18] = ?

G G G A **A** A A C C C A A A G G G G **U** U U n = 20

(((. . .))) (((. . .)))

0	0	0	0	0	0	0	1	2	3	3	3	3	3	3	3	3	4	5	6
0	0	0	0	0	0	0	1	2	2	2	2	2	2	3	3	3	4	5	6
0	0	0	0	0	0	0	1	1	1	1	1	1	2	2	3	3	4	5	6
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	5	6
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	5	6
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	5	5
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	4	4
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	3	3	3
0	0	0	0	0	0	0	0	0	0	0	0	0	1	1	2	2	2	2	3
0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	1	1	1	2	3
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	3
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2

Case 3, $2 \leq t < 18-4$:
 $t = 5$: yes pair
 $OPT[2,18] \geq 1+0+3$

GGAAACCCAAAGGGGU
 $\dots(\dots(((\dots))))$

$$OPT(i,j) = \begin{cases} 0 & \text{if } i \geq j - 4 \\ \max \left\{ \begin{array}{l} OPT[i,j-1] \\ 1 + \max_t (OPT[i,t-1] + OPT[t+1,j-1]) \end{array} \right\} & \text{otherwise} \end{cases}$$

0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0

Computing one cell: OPT[2,18] = ?

G G G A A **A** A C C C A A A G G G G **U** U U n = 20

(((. . .))) (((. . .)))

0	0	0	0	0	0	0	0	1	2	3	3	3	3	3	3	3	3	4	5	6
0	0	0	0	0	0	0	0	1	2	2	2	2	2	2	3	3	3	4	5	6
0	0	0	0	0	0	0	0	1	1	1	1	1	1	2	2	3	3	4	5	6
0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	5	6
0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	5	6
0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	5	5
0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	4	4
0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	3	3	3
0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	1	2	2	2	2	3
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	1	1	1	2	3
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	3
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2

Case 3, $2 \leq t < 18-4$:
 $t = 6$: yes pair
 $OPT[2,18] \geq 1+0+3$

GGAAAACCCAAAGGGGU
 $\dots(((((\dots))))))$

$$OPT(i,j) = \begin{cases} 0 & \text{if } i \geq j - 4 \\ \max \left\{ \begin{array}{l} OPT[i,j-1] \\ 1 + \max_t (OPT[i,t-1] + OPT[t+1,j-1]) \end{array} \right\} & \text{otherwise} \end{cases}$$

0 0

0 0

Computing one cell: OPT[2,18] = ?

G G G A A A **A** C C C A A A G G G G **U** U U n = 20

(((. . .))) (((. . .)))

0	0	0	0	0	0	0	1	2	3	3	3	3	3	3	3	3	4	5	6	
0	0	0	0	0	0	0	0	1	2	2	2	2	2	2	3	3	3	4	5	6
0	0	0	0	0	0	0	1	1	1	1	1	1	2	2	3	3	4	5	6	
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	5	6	
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	5	6	
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	5	5	
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	4	4	
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	3	3	3	
0	0	0	0	0	0	0	0	0	0	0	0	0	1	1	2	2	2	2	3	
0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	1	1	1	2	3	
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	3	
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	

Case 3, $2 \leq t < 18-4$:
 $t = 7$: yes pair
 $OPT[2,18] \geq 1+0+3$

GGAAA**A**CCCAAAGGGGU
((((.....))))

$$OPT(i,j) = \begin{cases} 0 & \text{if } i \geq j - 4 \\ \max \left\{ \begin{array}{l} OPT[i,j-1] \\ 1 + \max_t (OPT[i,t-1] + OPT[t+1,j-1]) \end{array} \right\} & \text{otherwise} \end{cases}$$

0
 0

Computing one cell: OPT[2,18] = ?



Case 3, $2 \leq t < 18-4$:
 $t = 8$: no pair

$$OPT(i, j) = \begin{cases} 0 & \text{if } i \geq j - 4 \\ \max \left\{ \begin{array}{l} OPT[i, j - 1] \\ 1 + \max_t (OPT[i, t - 1] + OPT[t + 1, j - 1]) \end{array} \right\} & \text{otherwise} \end{cases}$$

0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0

Computing one cell: OPT[2,18] = ?

G G G A A A A C C C **A** A A G G G G **U** U U n = 20

(((. . .))) (((. . .)))

0	0	0	0	0	0	0	1	2	3	3	3	3	3	3	3	3	4	5	6		
0	0	0	0	0	0	0	1	2	2	2	2	2	2	3	3	3	4	5	6		
0	0	0	0	0	0	0	1	1	1	1	1	1	2	2	3	3	4	5	6		
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	5	6		
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	5	6		
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	5	5		
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	4	4	4		
0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2	3	3	3	3		
0	0	0	0	0	0	0	0	0	0	0	0	0	1	1	2	2	2	2	3		
0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	1	1	1	2	3		
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	3	
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	2	2

Case 3, $2 \leq t < 18-4$:
 $t = 11$: yes pair
 $OPT[2,18] \geq 1+2+0$

GGAAAACCC**AA**AGGGGU
 ((.....))(.....)

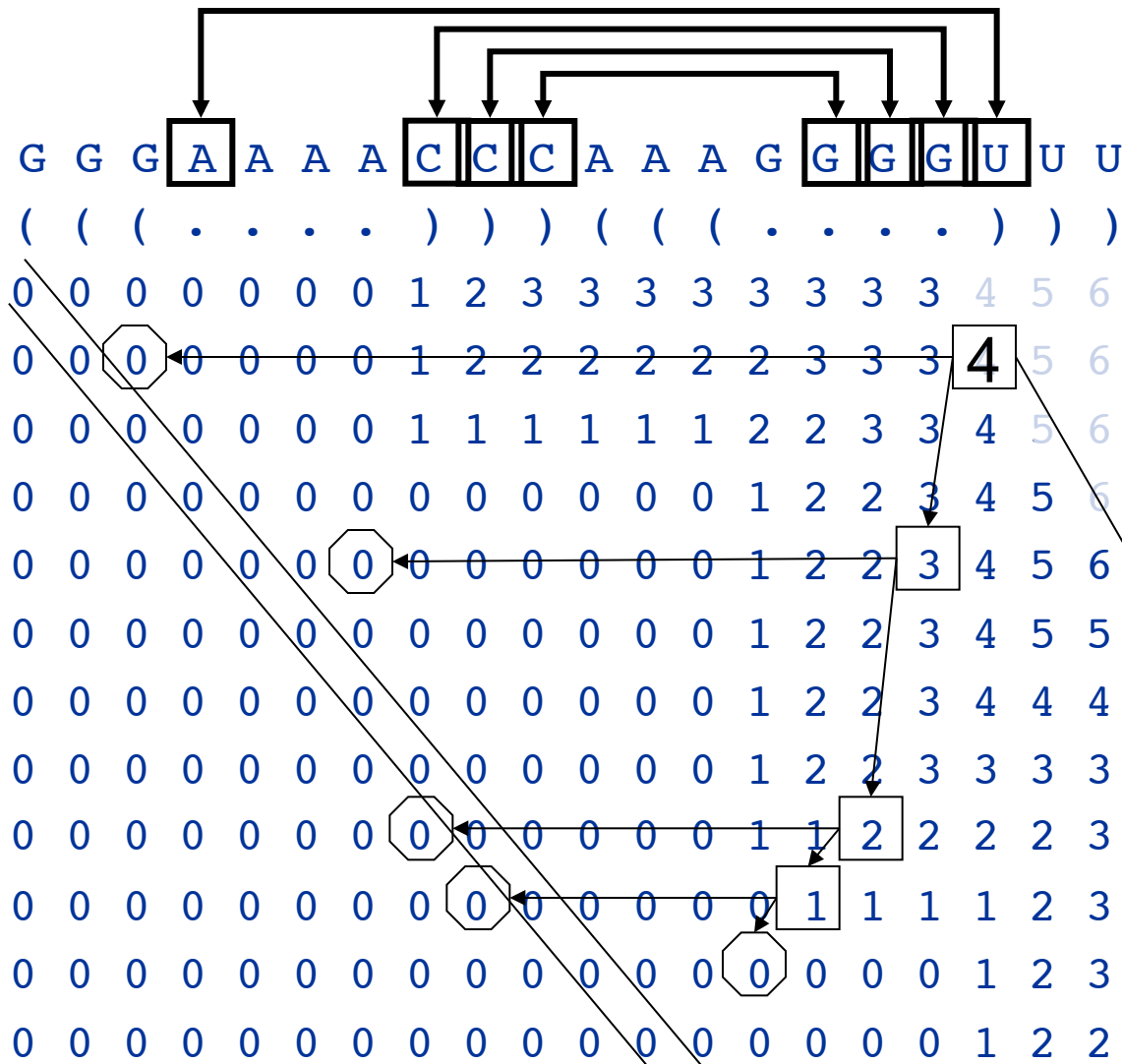
(not shown:
 $t=9,10,12,13$)

$$OPT(i,j) = \begin{cases} 0 & \text{if } i \geq j - 4 \\ \max \left\{ \begin{array}{l} OPT[i,j-1] \\ 1 + \max_t (OPT[i,t-1] + OPT[t+1,j-1]) \end{array} \right\} & \text{otherwise} \end{cases}$$

0 0

Computing one cell: $OPT[2,18] = 4$

$n = 20$



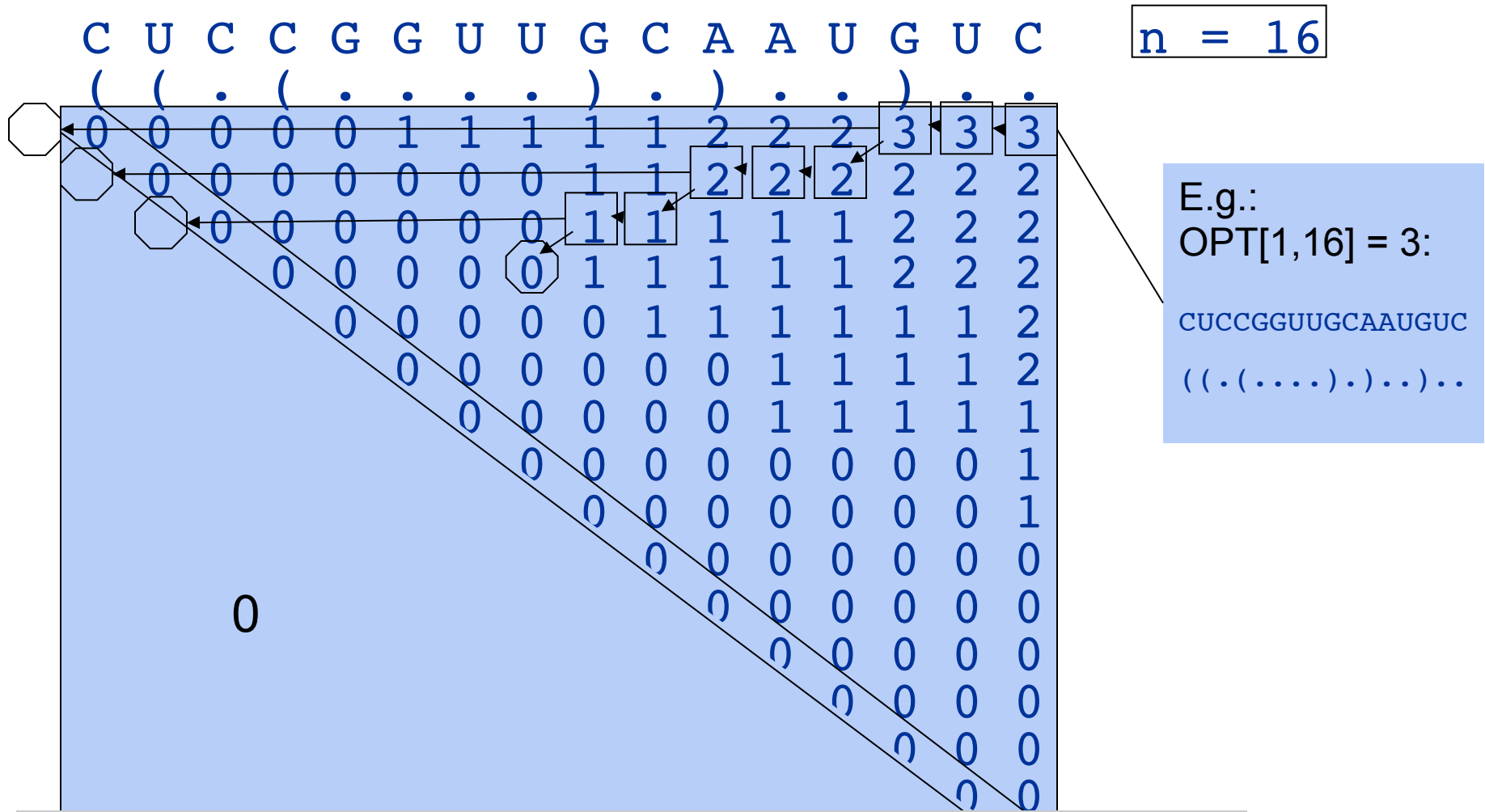
Overall, Max = 4
several ways, e.g.:

GGAAAACCCAAAGGGGU
..(....((((.....))))

tree shows trace back:
square = case 3
octagon = case 1

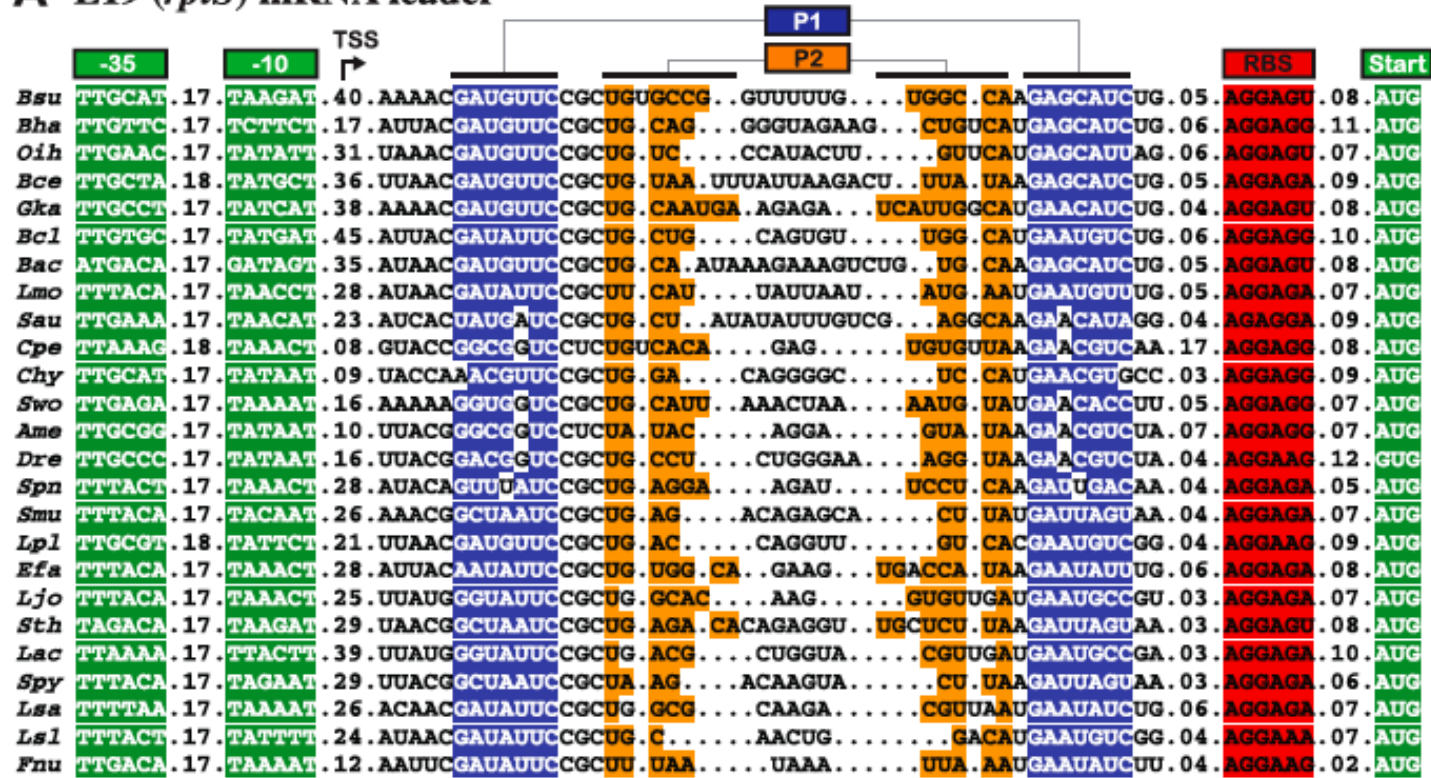
$$OPT(i,j) = \begin{cases} 0 & \text{if } i \geq j - 4 \\ \max \left\{ \begin{array}{l} OPT[i,j-1] \\ 1 + \max_t (OPT[i,t-1] + OPT[t+1,j-1]) \end{array} \right\} & \text{otherwise} \end{cases}$$

Another Trace Back Example

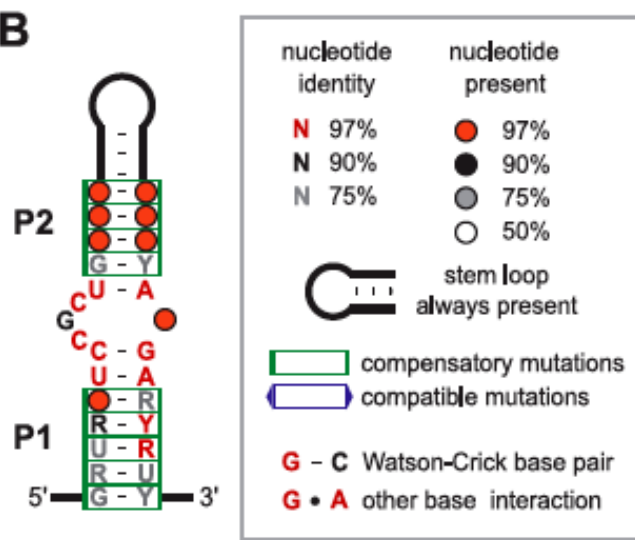


$$\text{OPT}(i,j) = \begin{cases} 0 & \text{if } i \geq j - 4 \\ \max \left\{ \begin{array}{l} \text{OPT}[i,j-1] \\ 1 + \max_t (\text{OPT}[i,t-1] + \text{OPT}[t+1,j-1]) \end{array} \right\} & \text{otherwise} \end{cases}$$

A L19 (*rplS*) mRNA leader



B



C

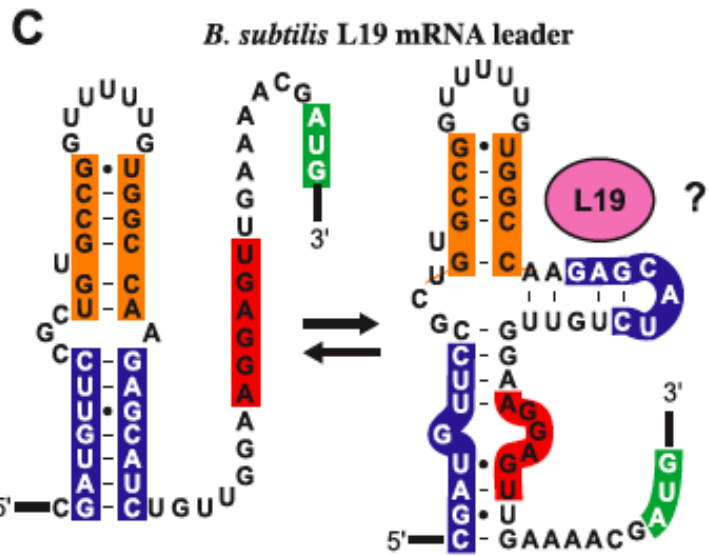


Figure 3. Putative Autoregulatory Structure in L19 mRNA Leaders