

CSE 52 I: Algorithms

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Dynamic Programming:
String alignment and RNA Folding

Outline

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

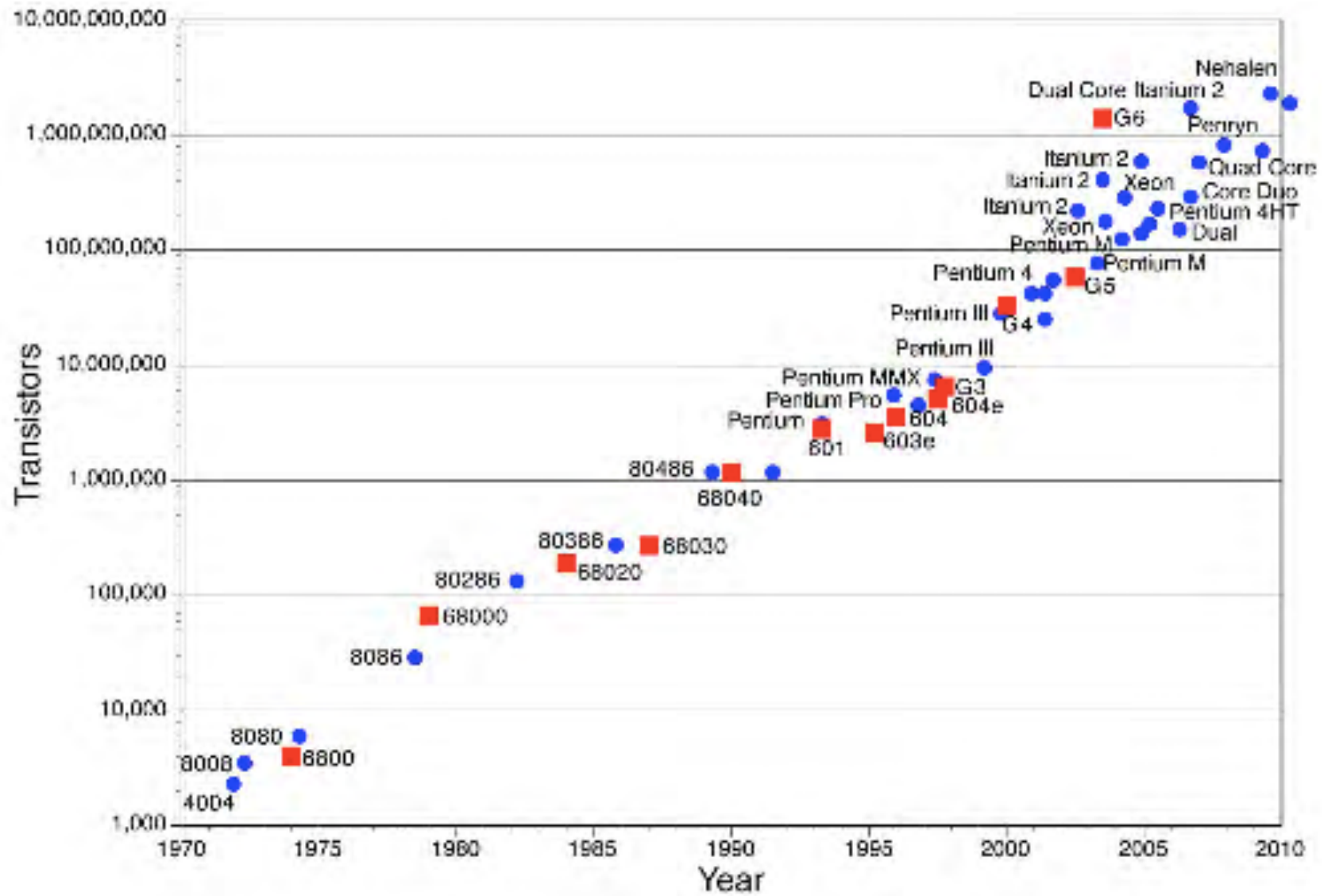
Sequence alignment

RNA structure

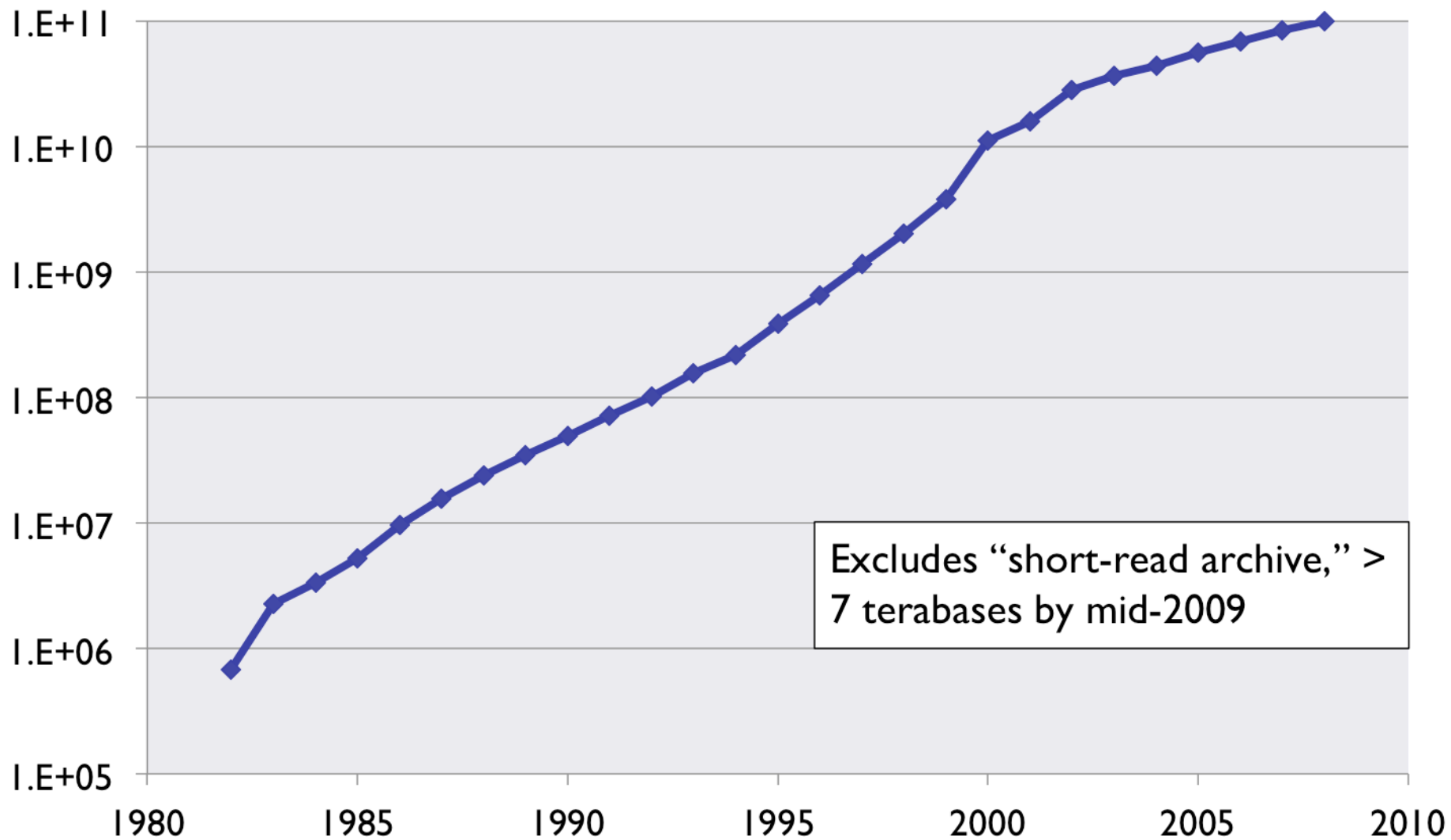
Algorithms for RNA structure

Application: Sequence Search

Moore's Law

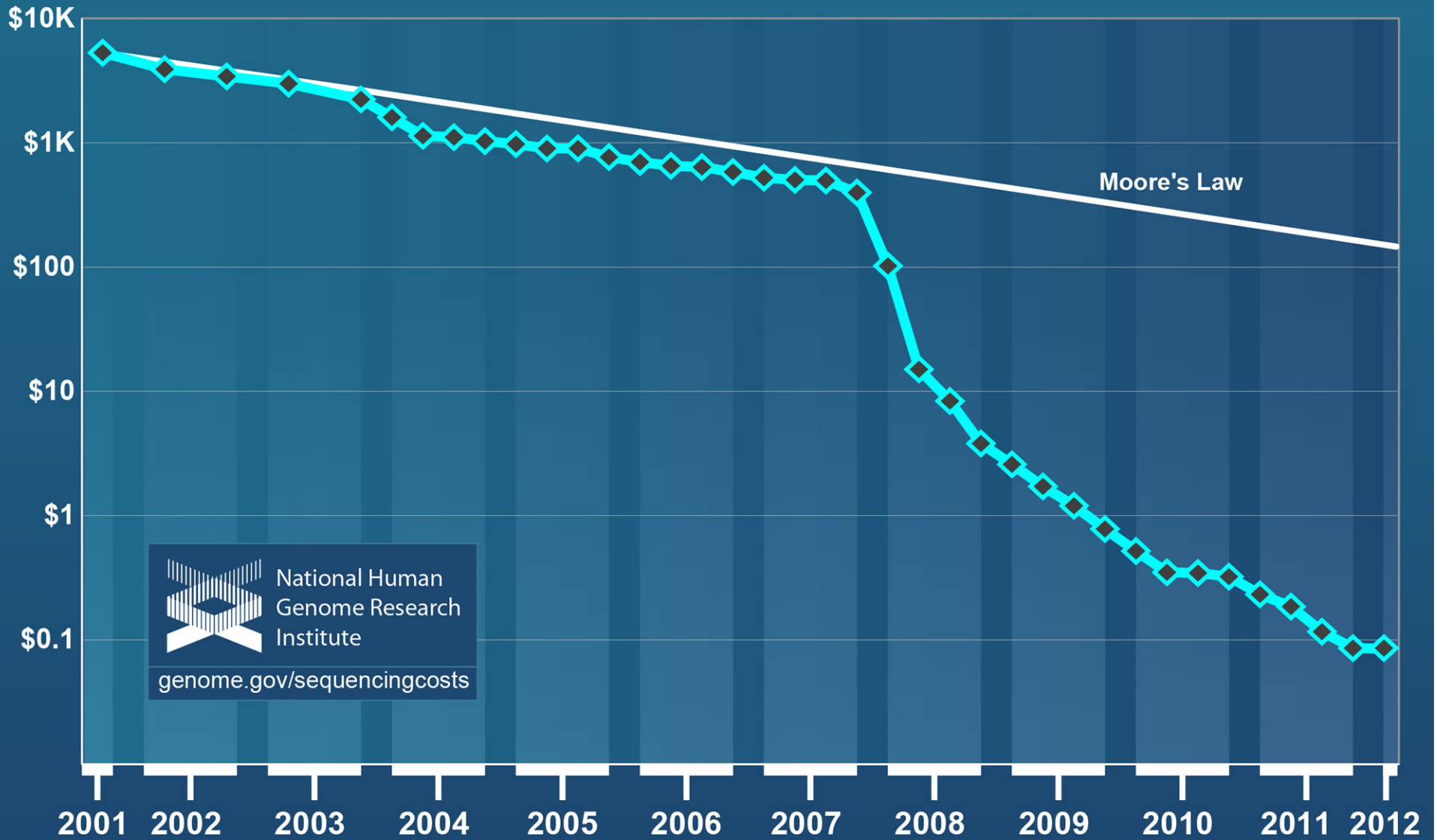


Growth of GenBank (Base Pairs)

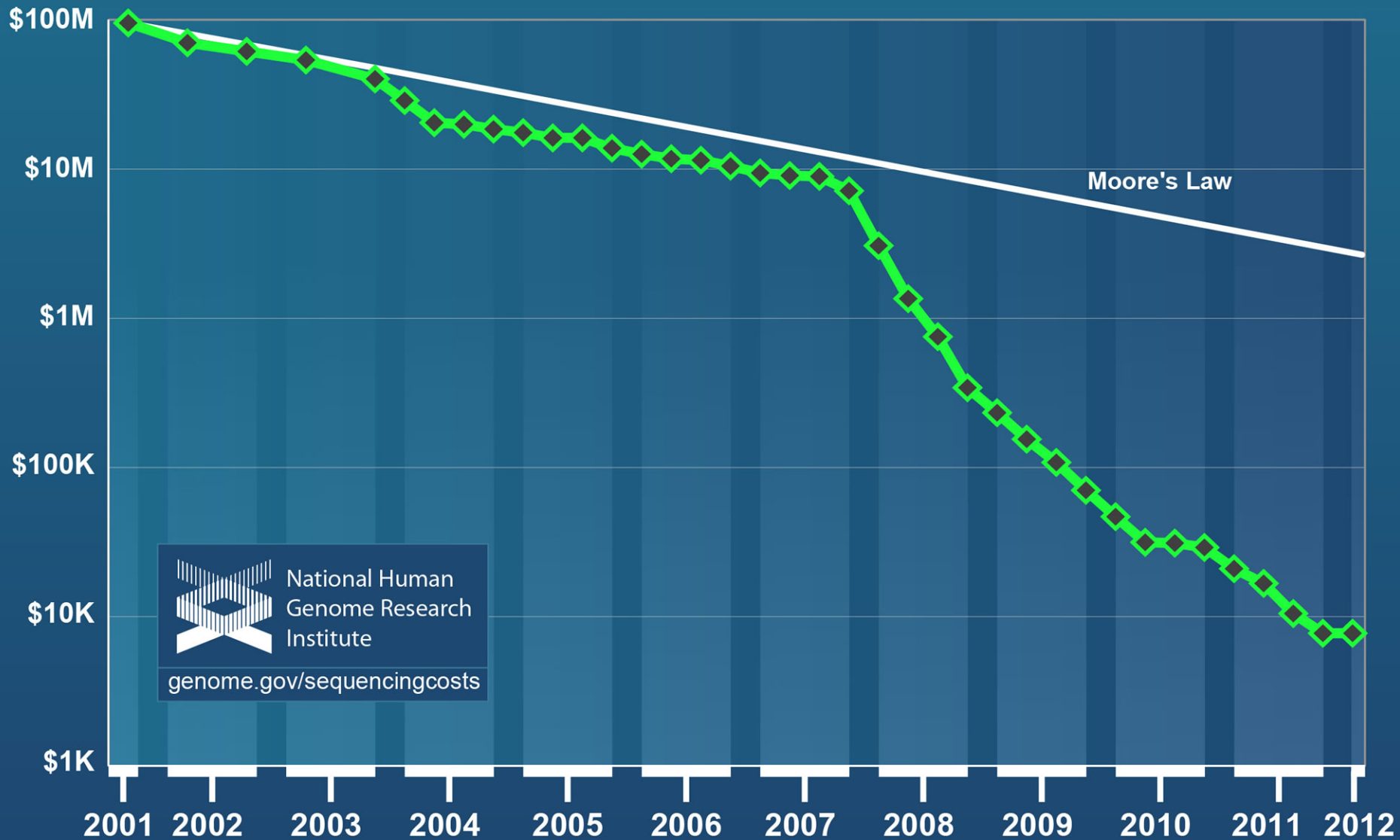


Source: <http://www.ncbi.nlm.nih.gov/Genbank/genbankstats.html>

Cost per Raw Megabase of DNA Sequence



Cost per Genome



 National Human
Genome Research
Institute
genome.gov/sequencingcosts

A Database Search

go to, e.g., <http://www.uniprot.org/>, “blast” tab, and paste in this:

```
>sp|P15172|MYOD1_HUMAN Myoblast determination protein 1 OS=Homo
  sapiens GN=MYOD1 PE=1 SV=3
MELLSPPLRDVDLTAPDGSLCSFATTDDFYDDPCFDSPDLRFFEDLDPRLMHVGALLKPE
EHSHFPAAVHPAPGAREDEHVRAPSGHHQAGRCLLWACKACKRKTTNADRRKAATMRERR
RLSKVNEAFETLKRCTSSNPNQRLPKVEILRNAIRYIEGLQALLRDQDAAPPGAAAFYA
PGPLPPGRGGEHYSGDSDASSPRSNCS DGMMDYSGPPSGARRRNCYEGAYYNEAPSEPRP
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			human Myoblast determination protein 1 (Homo sapiens)
<input type="checkbox"/> B2RC72	B2RC72_HUMAN			human cDNA, FLJ95884, highly similar to Hom... (Homo sapiens)
<input type="checkbox"/> E2RT59	E2RT59_CANFA			dog Uncharacterized protein (Canis familiaris)
<input type="checkbox"/> P49811	MYOD1_PIG			pig Myoblast determination protein 1 (Sus scrofa)
<input type="checkbox"/> D2KPI9	D2KPI9_PIG			pig Myogenic differentiation 1 (Sus scrofa)
<input type="checkbox"/> F1S9A9	F1S9A9_PIG			pig Uncharacterized protein (Sus scrofa)
<input type="checkbox"/> D2I0V4	D2I0V4_AILME			panda Putative uncharacterized protein (Ailuropoda melanoleuca)
<input type="checkbox"/> P29331	MYOD1_SHEEP			sheep Myoblast determination protein 1 (Ovis aries)
<input type="checkbox"/> D2SP11	D2SP11_BUBBU			water buffalo Myogenic factor MYOD1 (Bubalus bubalis)
<input type="checkbox"/> Q0VBX9	Q0VBX9_BOVIN			cow Myogenic differentiation 1 (Bos taurus)
<input type="checkbox"/> Q7YS82	MYOD1_BOVIN			cow Myoblast determination protein 1 (Bos taurus)
<input type="checkbox"/> Q8C6B1	Q8C6B1_MOUSE			mouse Myogenic differentiation 1 (Mus musculus)
<input type="checkbox"/> A0JPK9	A0JPK9_RAT			rat Myogenic differentiation 1 (Rattus norvegicus)
<input type="checkbox"/> Q02346	MYOD1_RAT			rat Myoblast determination protein 1 (Rattus norvegicus)
<input type="checkbox"/> P10085	MYOD1_MOUSE			mouse Myoblast determination protein 1 (Mus musculus)
<input type="checkbox"/> Q6DTY5	Q6DTY5_PIG			pig Eukaryotic myogenic factor MYF-3 (Sus scrofa)
<input type="checkbox"/> P21572	MYOD1_COTJA			quail Myoblast determination protein 1 homolog (Coturnix coturnix japonica)
<input type="checkbox"/> Q6DV59	Q6DV59_MELGA			turkey MyoD (Meleagris gallopavo)
<input type="checkbox"/> P16075	MYOD1_CHICK			chicken Myoblast determination protein 1 homolog (Gallus gallus)
<input type="checkbox"/> C5J072	C5J072_CHICK			chicken Myogenic differentiation 1 (Gallus gallus)
<input type="checkbox"/> C3U0I1	C3U0I1_ANAPL			duck Myogenic differentiation 1 (Anas platyrhynchos)
<input type="checkbox"/> F1NHM3	F1NHM3_CHICK			chicken Uncharacterized protein (Gallus gallus)
<input type="checkbox"/> F1NXM5	F1NXM5_CHICK			chicken Uncharacterized protein (Gallus gallus)
<input type="checkbox"/> P13904	MYODA_XENLA			frog Myoblast determination protein 1 homolog A (Xenopus laevis)
<input type="checkbox"/> Q8AVZ0	Q8AVZ0_XENLA			frog Myod1-a protein (Xenopus laevis)
<input type="checkbox"/> Q7T109	Q7T109_XENTR			frog MyoD protein (Xenopus tropicalis)

...And 100's 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	MELLSPPLRDVDLTAPDGLCSFATDDDFYDDPCF DSPDLRFFEDLDPRLMHVGALLKPE	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	RLSKVNEAFETLKRCTSSNP NQRLPKVEILRN A IRYIEGLQALLRDQDAAPP GAAA AFYA	180	P15172
	RLSKVNEAFETLKRCTS+NPNQRLPKVEILRN A IRYIE LQ+LLR Q+ +FY		
107	RLSKVNEAFETLKRCTSTNPNQRLPKVEILRN A IRYIESLQSLLRGQE-----ESFY-	158	Q7T109

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

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

Alignments	Entry	Entry name	Status	Protein names	Organism	Length	Identity	Score	E-value	Gene names
	B3LY60	B3LY60_DROAN	★	GF18746	Drosophila ananassae (Fruit fly)	334	42.0%	344	4.0×10 ⁻³⁰	GF18746 Dana\GF18746 Dana_GF18746
	Q4RGJ6	Q4RGJ6_TETNG	★	Chromosome undetermined SCAF15099, whole geno...	Tetraodon nigroviridis (Spotted green pufferfish) (Chelonodon nigroviridis)	126	57.0%	343	5.0×10 ⁻³⁰	GSTENG00034775001
	F1NFS6	F1NFS6_CHICK	★	Uncharacterized protein	Gallus gallus (Chicken)	237	46.0%	343	5.0×10 ⁻³⁰	Gga.378
	Q91151	Q91151_NOTVI	★	Myogenic regulatory factor; transcription fac...	Notophthalmus viridescens (Eastern newt) (Triturus viridescens)	219	44.0%	342	6.0×10 ⁻³⁰	MRF-4
	Q29BN7	Q29BN7_DROPS	★	GA10192	Drosophila pseudoobscura pseudoobscura (Fruit fly)	330	42.0%	342	6.0×10 ⁻³⁰	GA10192 Dpse\GA10192 Dpse_GA10192
	B4GP81	B4GP81_DROPE	★	GL13832	Drosophila persimilis (Fruit fly)	330	42.0%	342	6.0×10 ⁻³⁰	GL13832 Dper\GL13832 Dper_GL13832
	Q92020	MYF6_XENLA	★	Myogenic factor 6	Xenopus laevis (African clawed frog)	240	44.0%	340	1.0×10 ⁻²⁹	myf6 mrf4
	B7ZQB0	B7ZQB0_XENLA	★	MRF4a	Xenopus laevis (African clawed frog)	240	44.0%	340	1.0×10 ⁻²⁹	MRF4
	A7UCI1	A7UCI1_XENLA	★	MRF4a	Xenopus laevis (African clawed frog)	240	44.0%	340	1.0×10 ⁻²⁹	MRF4A MRF4
	F7FIX8	F7FIX8_MONDO	★	Uncharacterized protein	Monodelphis domestica (Gray short-tailed opossum)	243	47.0%	339	1.0×10 ⁻²⁹	MYF6
	D9IV56	D9IV56_EPICO	★	Myogenin	Epinephelus coioides (Orange-spotted grouper) (Epinephelus nebulosus)	250	47.0%	338	2.0×10 ⁻²⁹	
	Q6SYV5	MYF6_TAKRU	★	Myogenic factor 6	Takifugu rubripes (Japanese pufferfish) (Fugu rubripes)	239	46.0%	337	2.0×10 ⁻²⁹	myf6 mrf4
	G3WJP0	G3WJP0_SARHA	★	Uncharacterized protein	Sarcophilus harrisii (Tasmanian devil) (Sarcophilus lanarius)	243	47.0%	337	2.0×10 ⁻²⁹	MYF6
	G7Y452	G7Y452_CLOSI	★	Transcription factor SUM-1	Clonorchis sinensis (Chinese liver fluke)	946	64.0%	337	2.0×10 ⁻²⁹	CLF_100742
	G4LXJ1	G4LXJ1_SCHMA	★	Myogenic factor, putative	Schistosoma mansoni (Blood fluke)	864	68.0%	337	2.0×10 ⁻²⁹	Smp_167400
	G1EN33	G1EN33_DUGJA	★	Myogenic determinant factor	Dugesia japonica (Planarian)	498	42.0%	336	3.0×10 ⁻²⁹	MDF
	Q8MWP6	Q8MWP6_SCHMD	★	MyoD-like protein	Schmidtea mediterranea (Freshwater planarian flatworm)	466	69.0%	335	4.0×10 ⁻²⁹	
	Q6PUV5	MYF6_TETNG	★	Myogenic factor 6	Tetraodon nigroviridis (Spotted green pufferfish) (Chelonodon nigroviridis)	239	45.0%	333	7.0×10 ⁻²⁹	myf6 mrf4 GSTENG00021536001
	Q2PZ12	Q2PZ12_SALSA	★	Myogenin	Salmo salar (Atlantic salmon)	254	55.0%	333	7.0×10 ⁻²⁹	
	Q91207	Q91207_ONCMY	★	TMyogenin protein	Oncorhynchus mykiss (Rainbow trout) (Salmo gairdneri)	254	45.0%	332	9.0×10 ⁻²⁹	TMyogenin

hits at rank ~ 250 still extremely good matches,
even though very distantly related organisms

The foregoing search capability is a *huge* deal

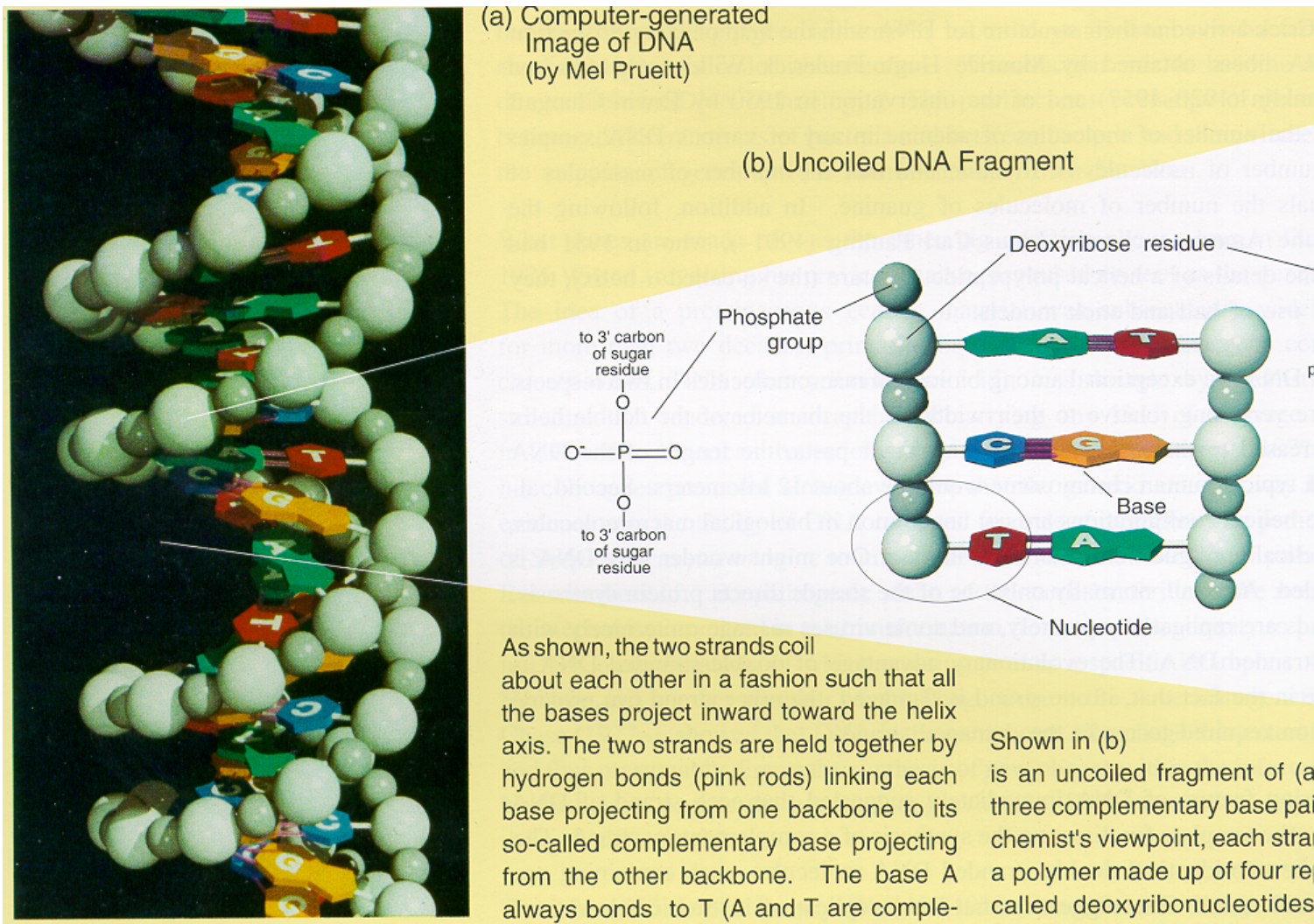
millions of searches

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

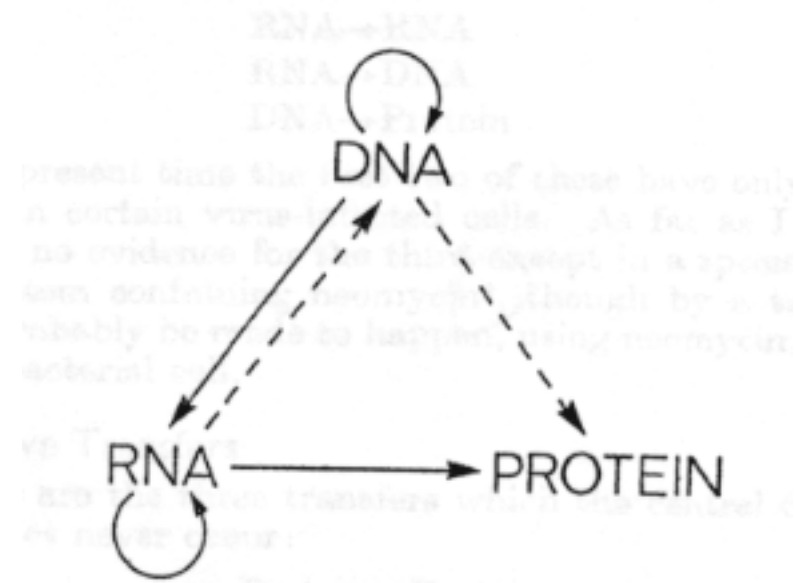
FRANCIS CRICK

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

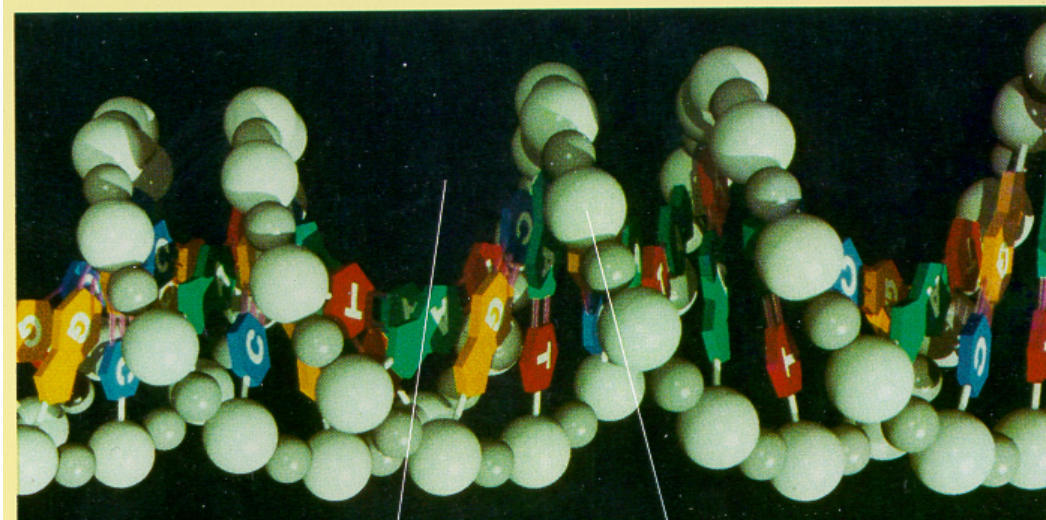
Regulation, transport, stability/degradation

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

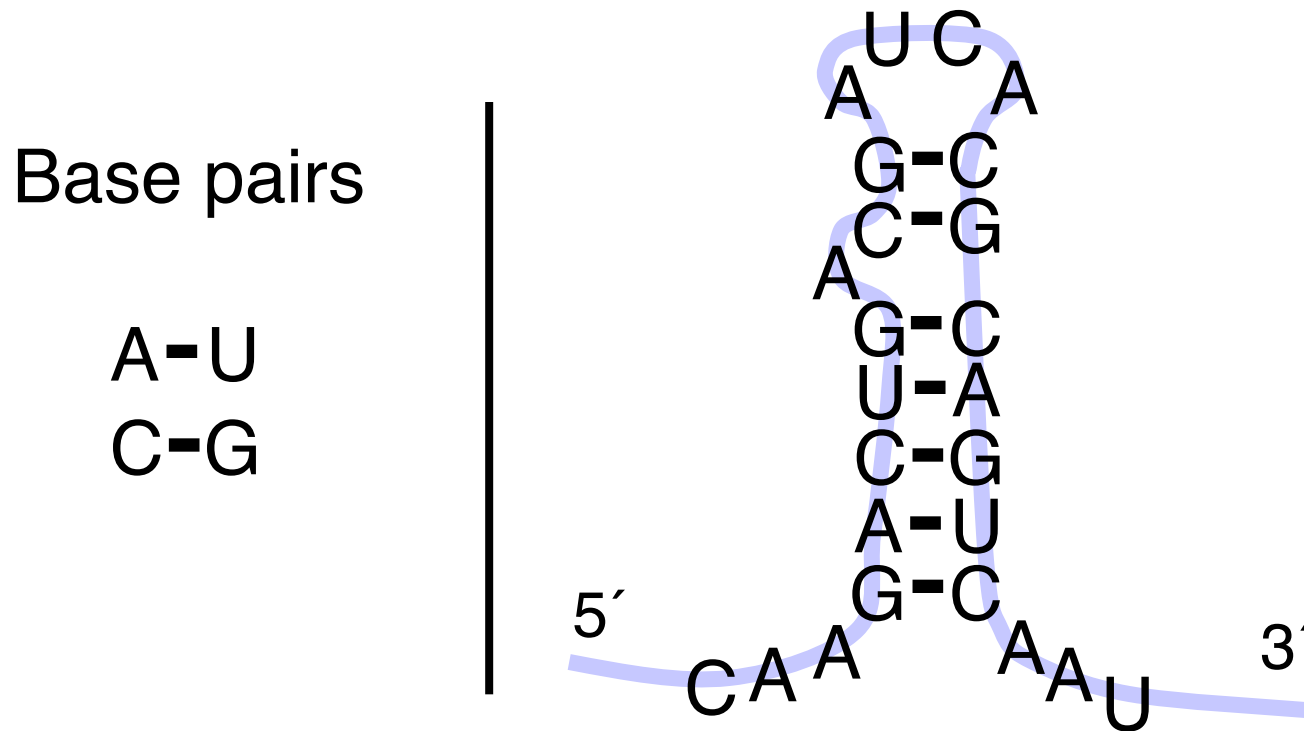
E.g. "riboswitches": 1000s in bacteria

DNA structure: dull

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



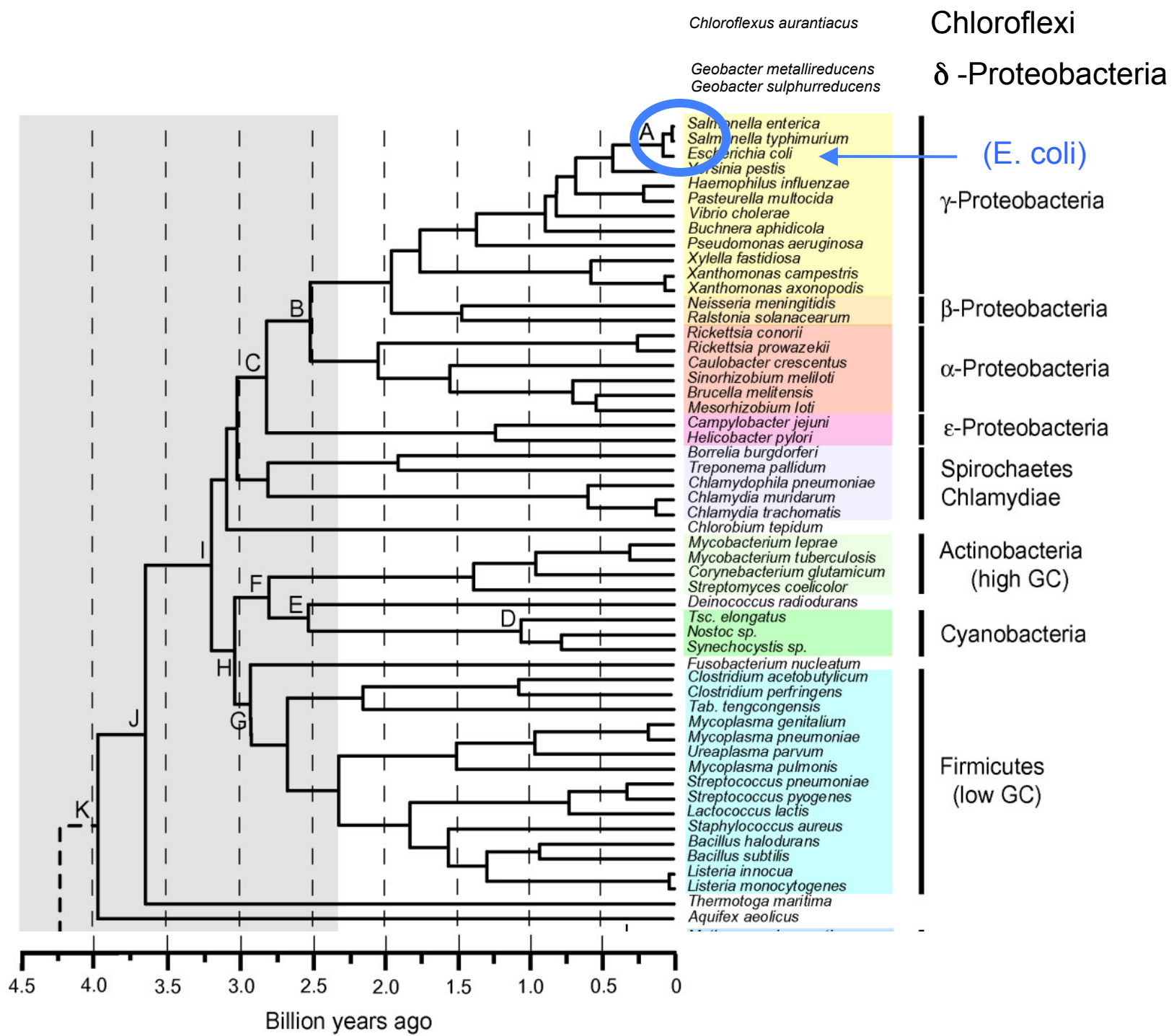
RNA Secondary Structure: RNA makes helices too



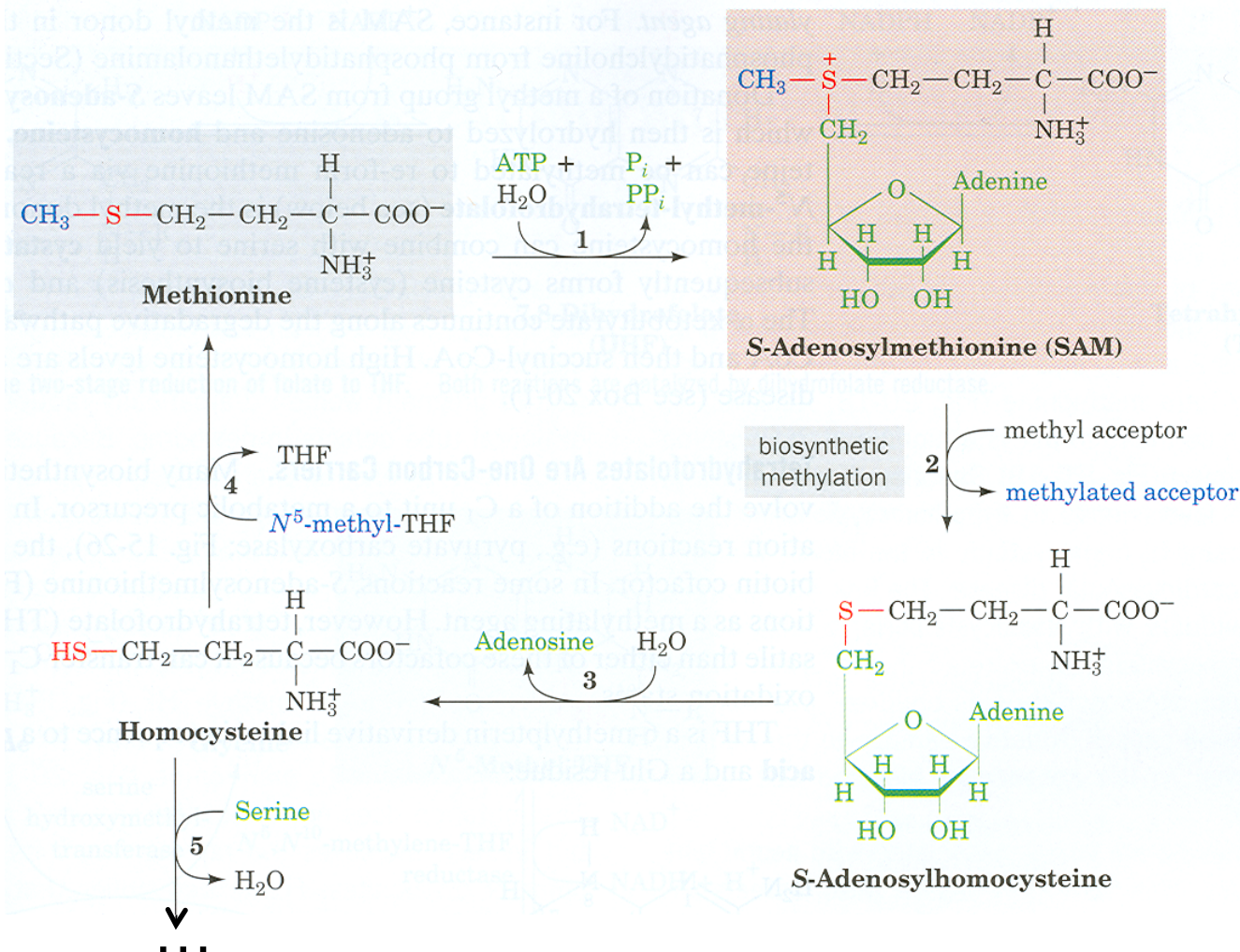
Usually *single* stranded

Why is structure important?

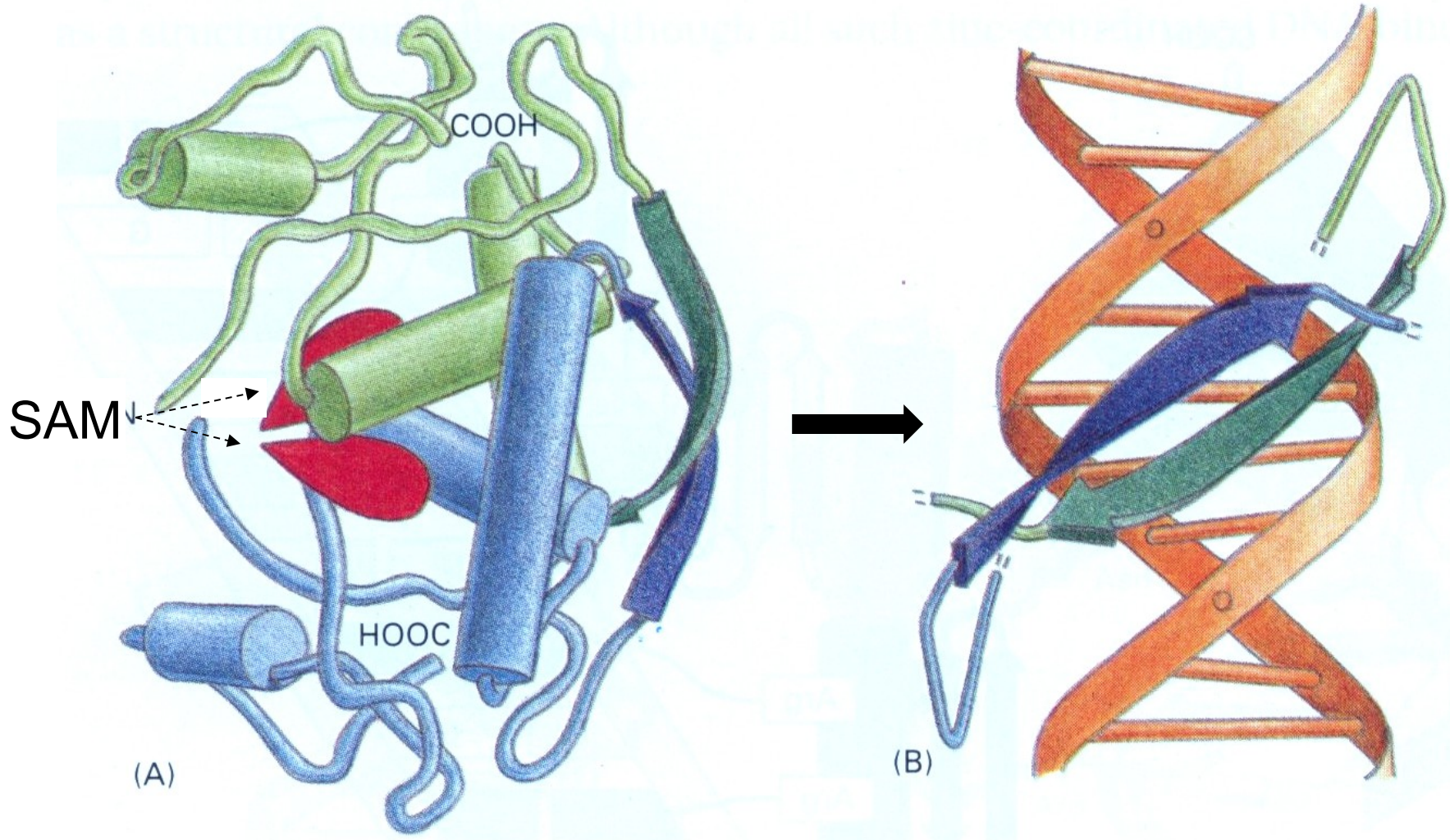
- For protein-coding, similarity in sequence is a powerful tool for finding related sequences
 - e.g. “hemoglobin” is easily recognized in all vertebrates
- For non-coding RNA, many different sequences have the same structure, and structure is most important for function.
 - So, using structure plus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



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



Gene Regulation: The MET Repressor

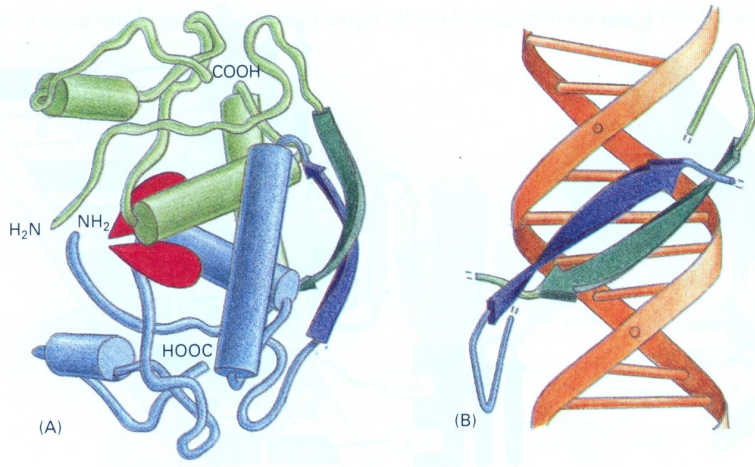


Protein

Alberts, et al, 3e.

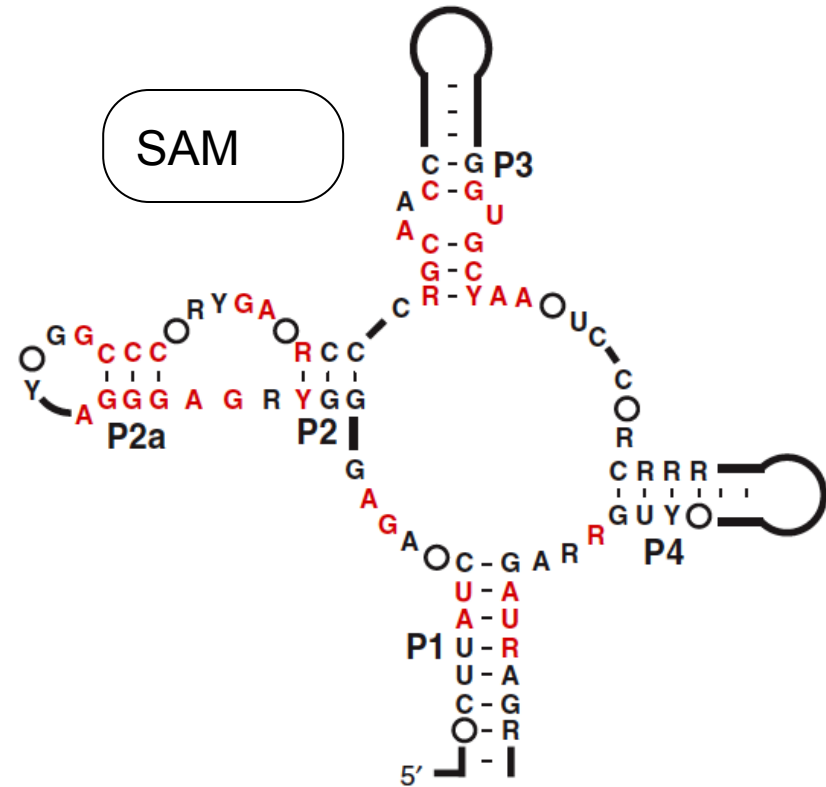
DNA

Alberts, et al, 3e.



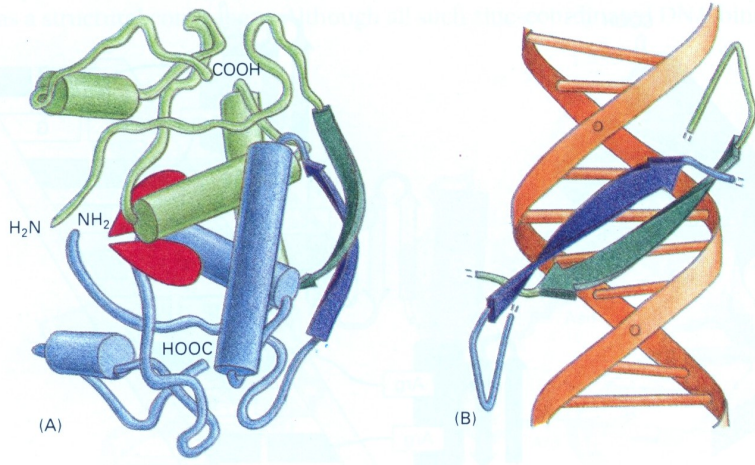
← The protein way

Riboswitch alternative



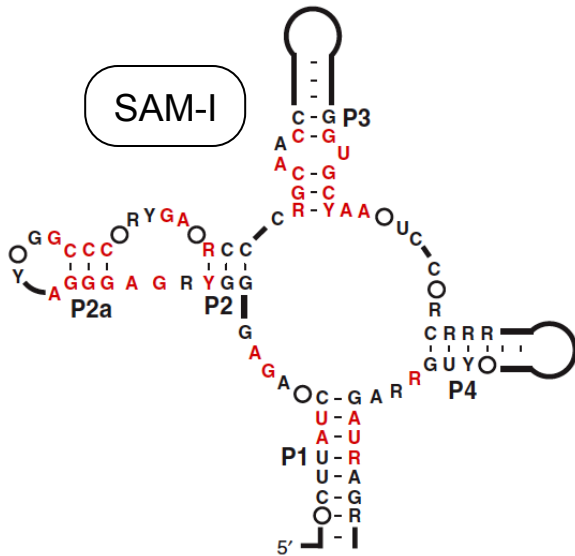
Grundy & Henkin, Mol. Microbiol 1998
Epshtein, et al., PNAS 2003
Winkler et al., Nat. Struct. Biol. 2003

Alberts, et al, 3e.

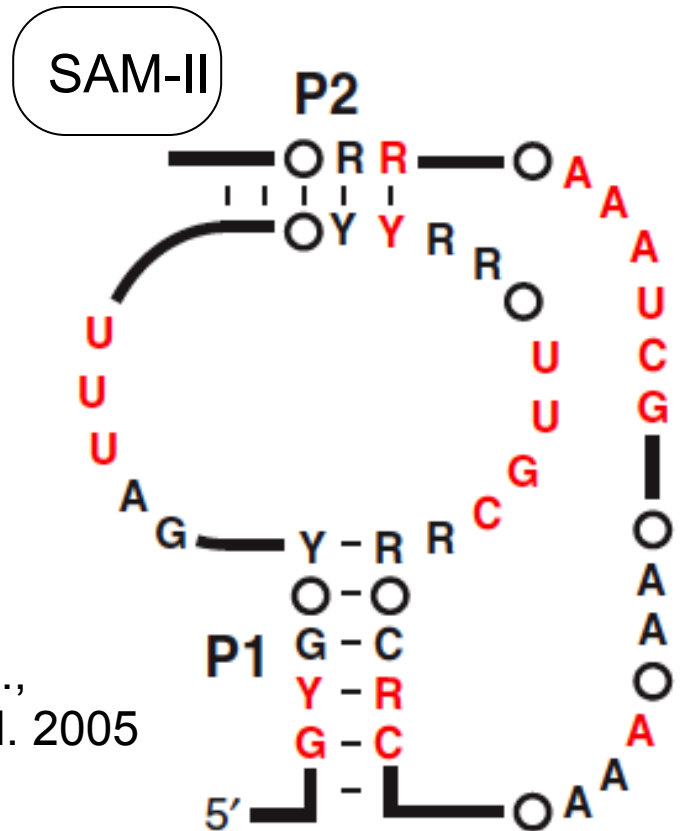


The protein way

Riboswitch alternatives

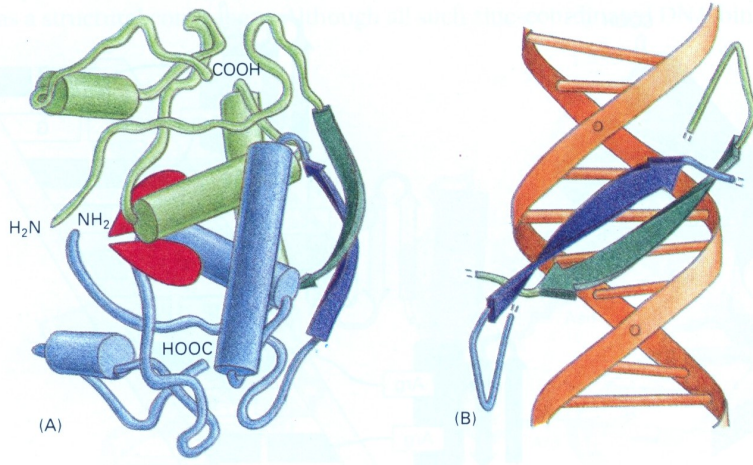


Grundy, Epshtein, Winkler et al., 1998, 2003



Corbino et al.,
Genome Biol. 2005

Alberts, et al, 3e.



← The protein way

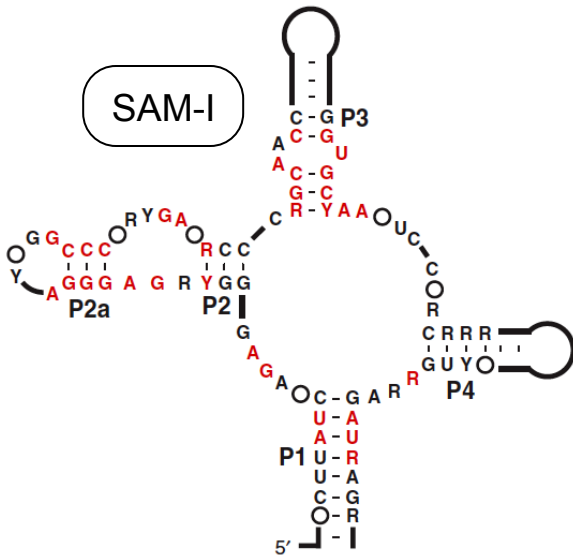
Riboswitch alternatives



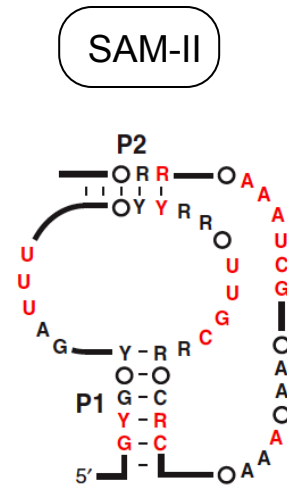
SAM-III



Fuchs et al., NSMB 2006

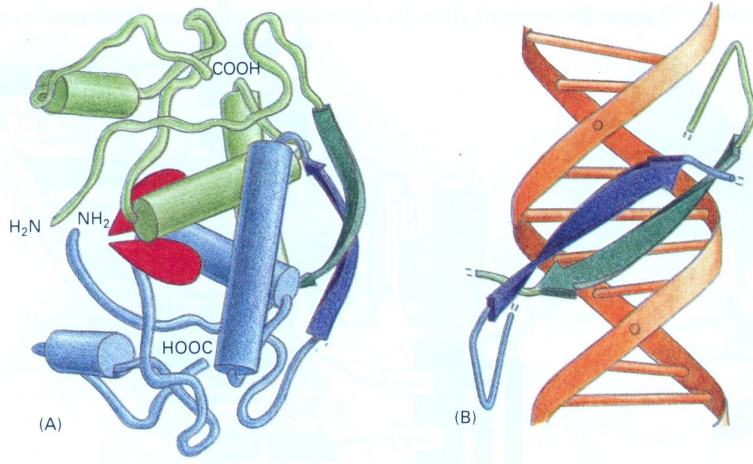


Grundy, Epshtein, Winkler et al., 1998, 2003



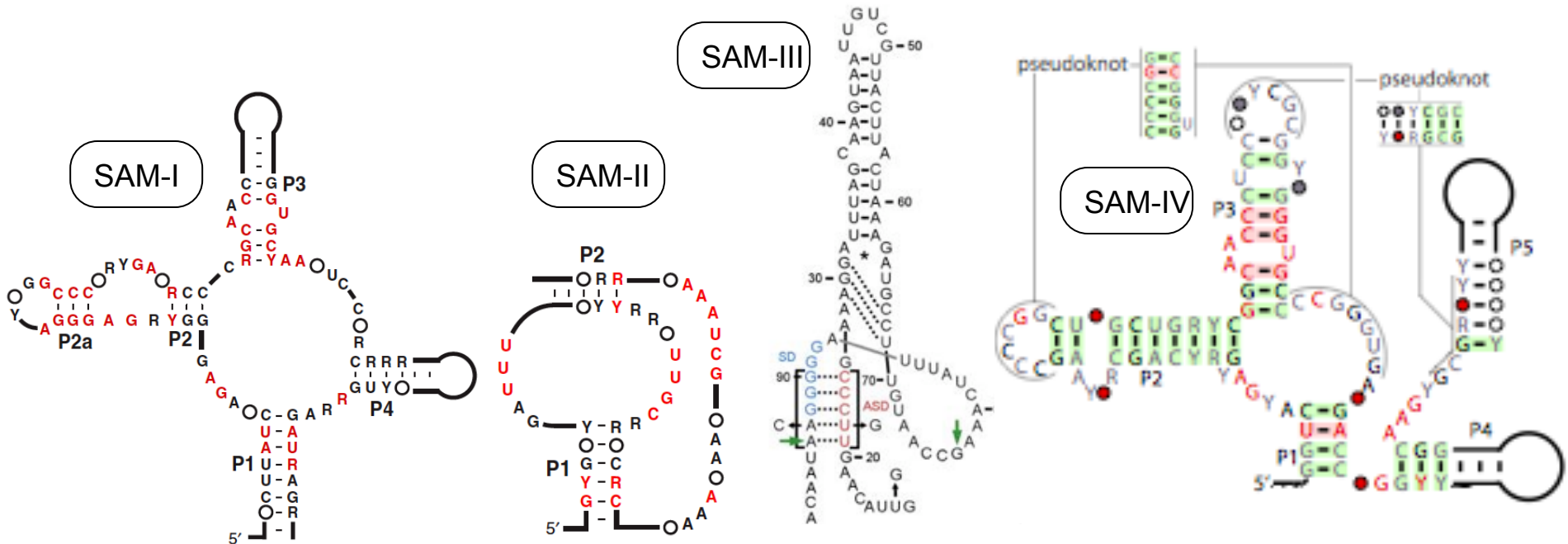
Corbino et al., Genome Biol. 2005

Alberts, et al, 3e.



The protein way

Riboswitch alternatives



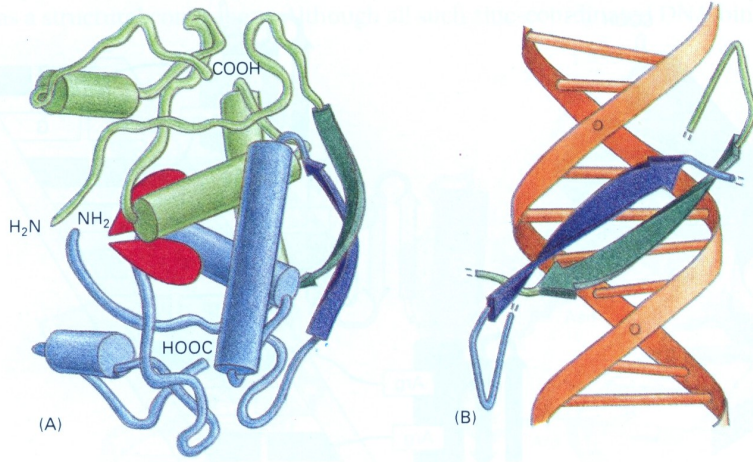
Grundy, Epshtein, Winkler et al., 1998, 2003

Corbino et al., Genome Biol. 2005

Fuchs et al., NSMB 2006

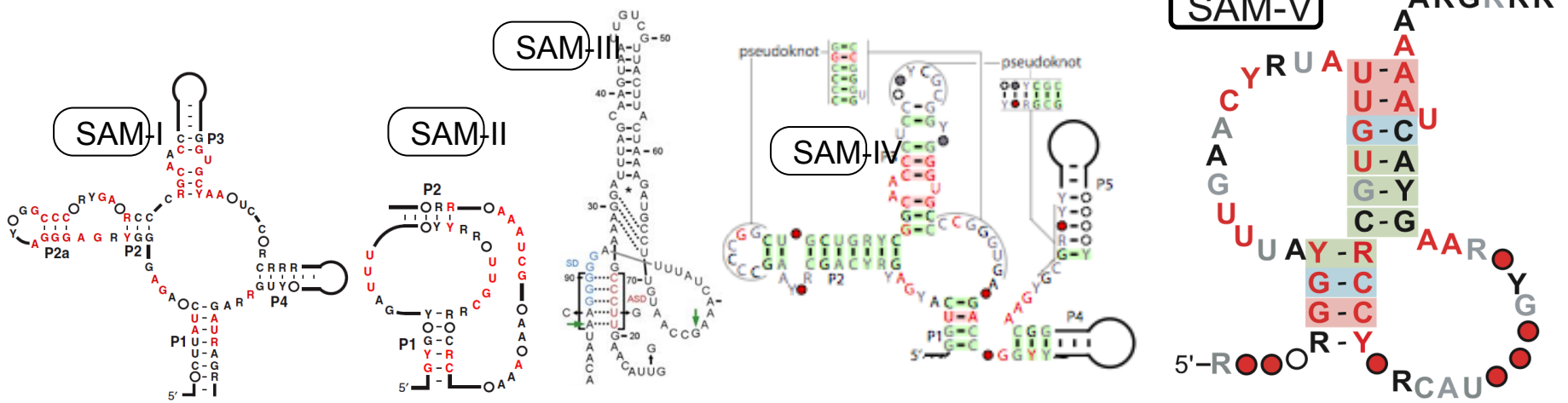
Weinberg et al.,³⁵ RNA 2008

Alberts, et al, 3e.



The protein way

Riboswitch alternatives



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

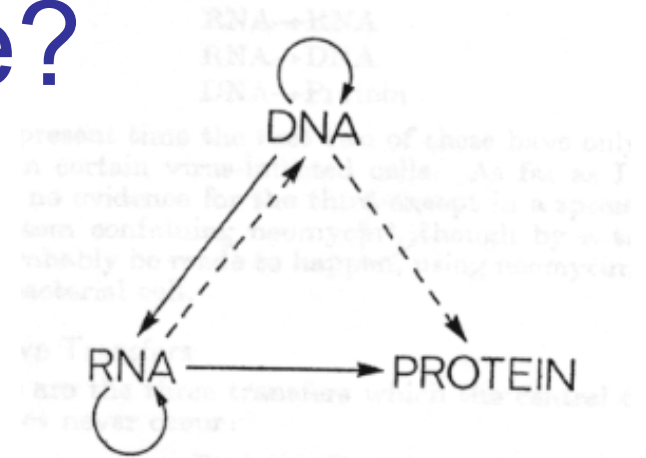
Corbino et al., Genome Biol. 2005

Fuchs et al., NSMB 2006

Weinberg et al., RNA 2008

Meyer, et al., BMC Genomics 2009

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 (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*.)

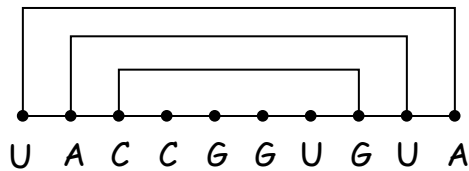
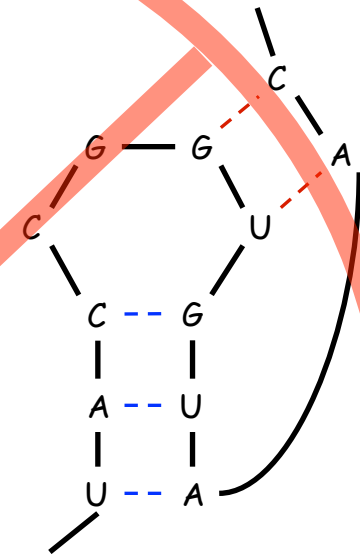
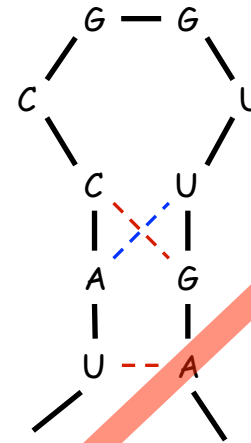
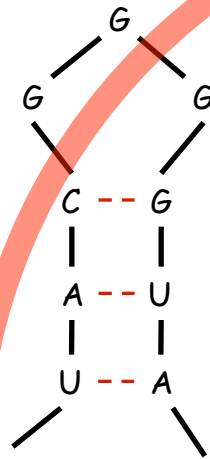
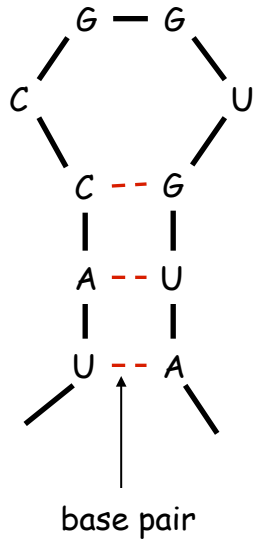
Free energy. Usual hypothesis is that an RNA molecule will form the secondary structure with the optimum total free energy.

↑
approximate by number of base pairs

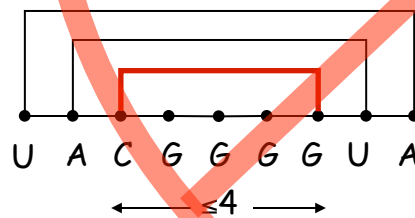
Goal. Given an RNA molecule $B = b_1b_2\dots b_n$, find a secondary structure S that maximizes the number of base pairs.

RNA Secondary Structure: Examples

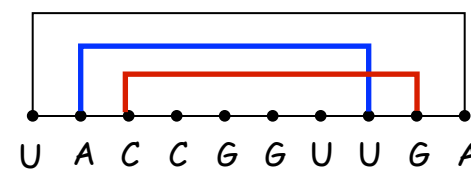
Examples.



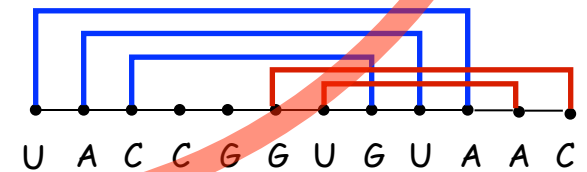
ok



sharp turn



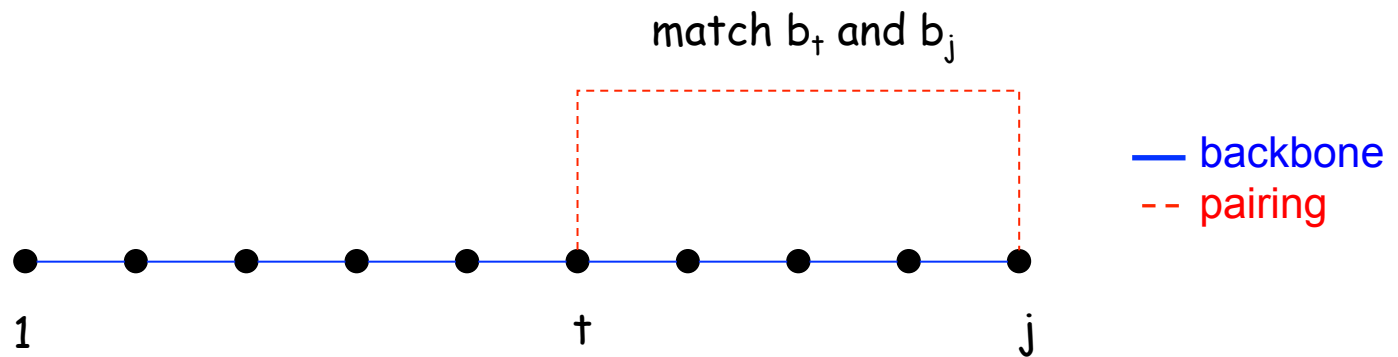
crossing



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

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.

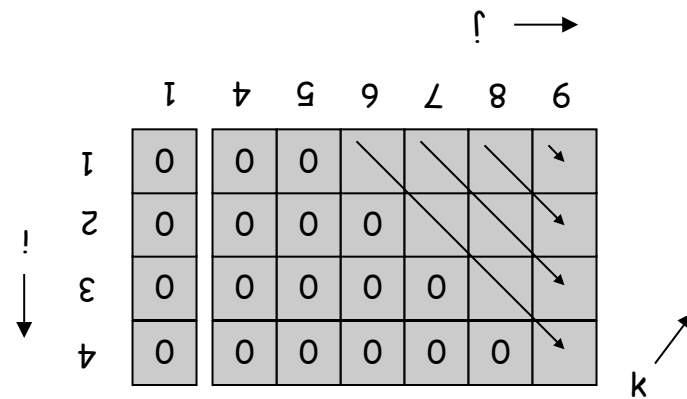
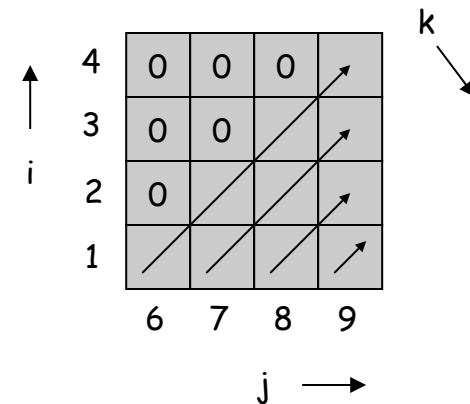
Bottom Up Dynamic Programming Over Intervals

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. $O(n^3)$.

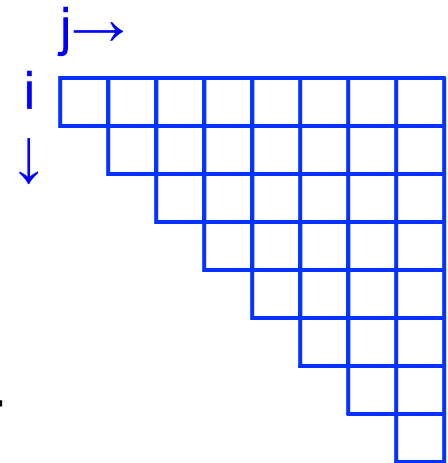
Nussinov: Max Pairing

$\text{opt}(i,j)$ = # pairs in optimal pairing of $r_i \dots r_j$

$\text{opt}(i,j) = 0$ for all i, j with $i \geq j-4$; otherwise

$\text{opt}(i,j) = \max$ of:

$$\left\{ \begin{array}{l} \text{opt}(i,j-1) \\ \max \{ \text{opt}(l,t-1)+1+\text{opt}(t+1,j-1) \mid \\ \quad i \leq t < j-4 \text{ and } r_t-r_j \text{ may pair} \} \end{array} \right.$$

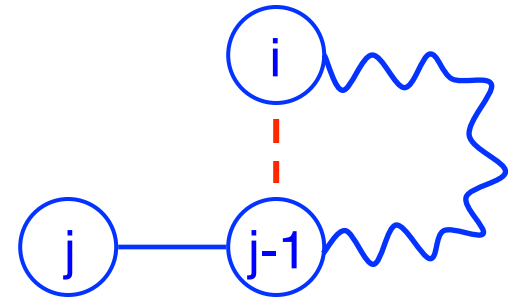


“Optimal pairing of $r_i \dots r_j$ ”

Two possibilities

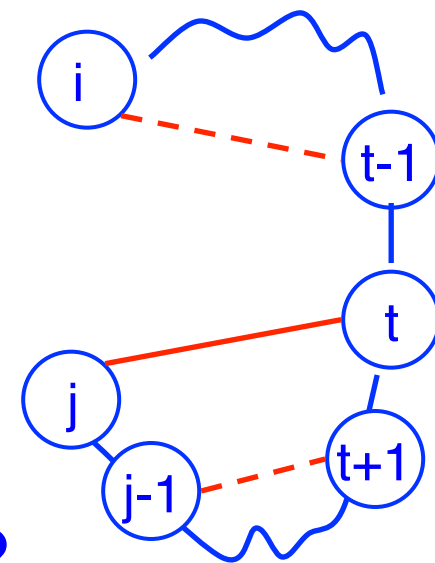
j Unpaired:

Find best pairing of $r_i \dots r_{j-1}$



j Paired (with some t):

Find best $r_i \dots r_{t-1}$ +
best $r_{t+1} \dots r_{j-1}$ **plus 1**



Why is it slow?

Why do pseudoknots matter?

— backbone
- pairing

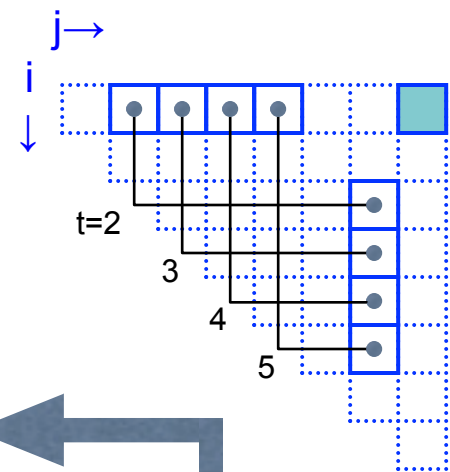
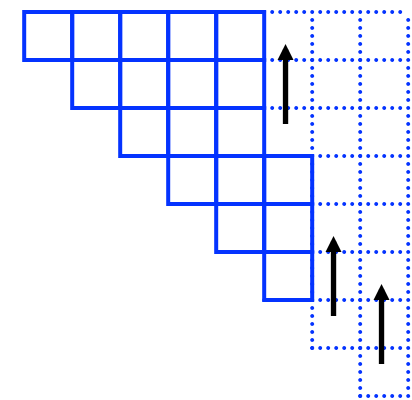
Nussinov: Computation Order

$\text{opt}(i,j)$ = # pairs in optimal pairing of $r_i \dots r_j$

$\text{opt}(i,j) = 0$ for all i, j with $i \geq j-4$; otherwise:

$\text{opt}(i,j) = \max$ of:

$$\left\{ \begin{array}{l} \text{opt}(i,j-1) \\ \max \{ \text{opt}(l,t-1)+1+\text{opt}(t+1,j-1) \mid \\ \quad i \leq t < j-4 \text{ and } r_t-r_j \text{ may pair} \} \end{array} \right.$$



Time: $O(n^3)$

Which Pairs?

Usual dynamic programming “trace-back” tells you *which* base pairs are in the optimal solution, not just how many

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	0	0	0	1
								0	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	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

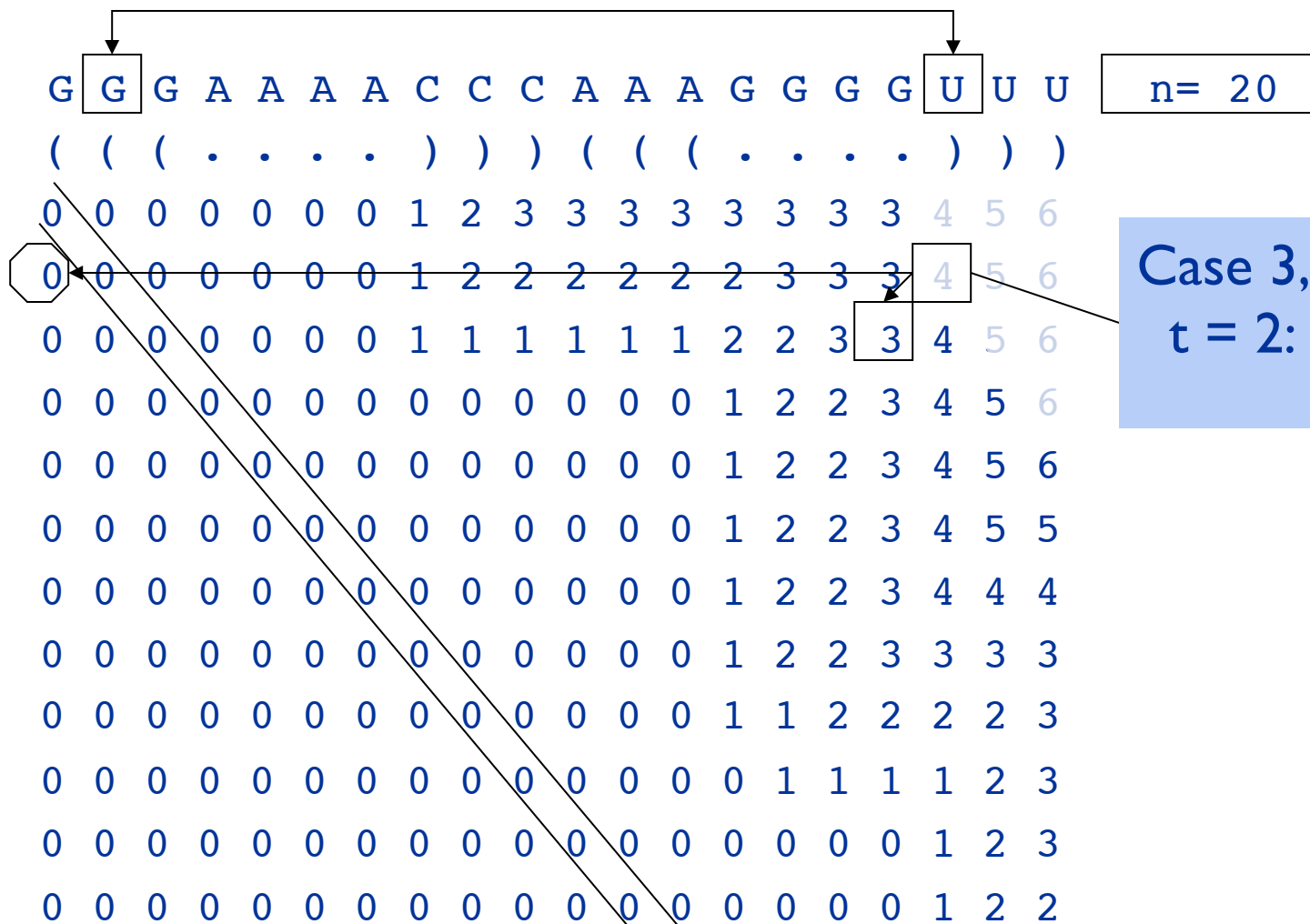
Case 1:
 $2 \geq 18-4?$ no.
 Case 2:
 B_{18} unpaired?
 Always a possibility;
 then $OPT[2,18] \geq 3$

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 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	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, $2 \leq t < 18-4$:
 $t = 3$: 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 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] = ?



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] = ?



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

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

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



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

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

GGAAA**ACC**CAAAGGGGU
((((.....))))

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

0 0

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

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, $2 \leq t < 18-4$:
 $t = 8$: 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

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

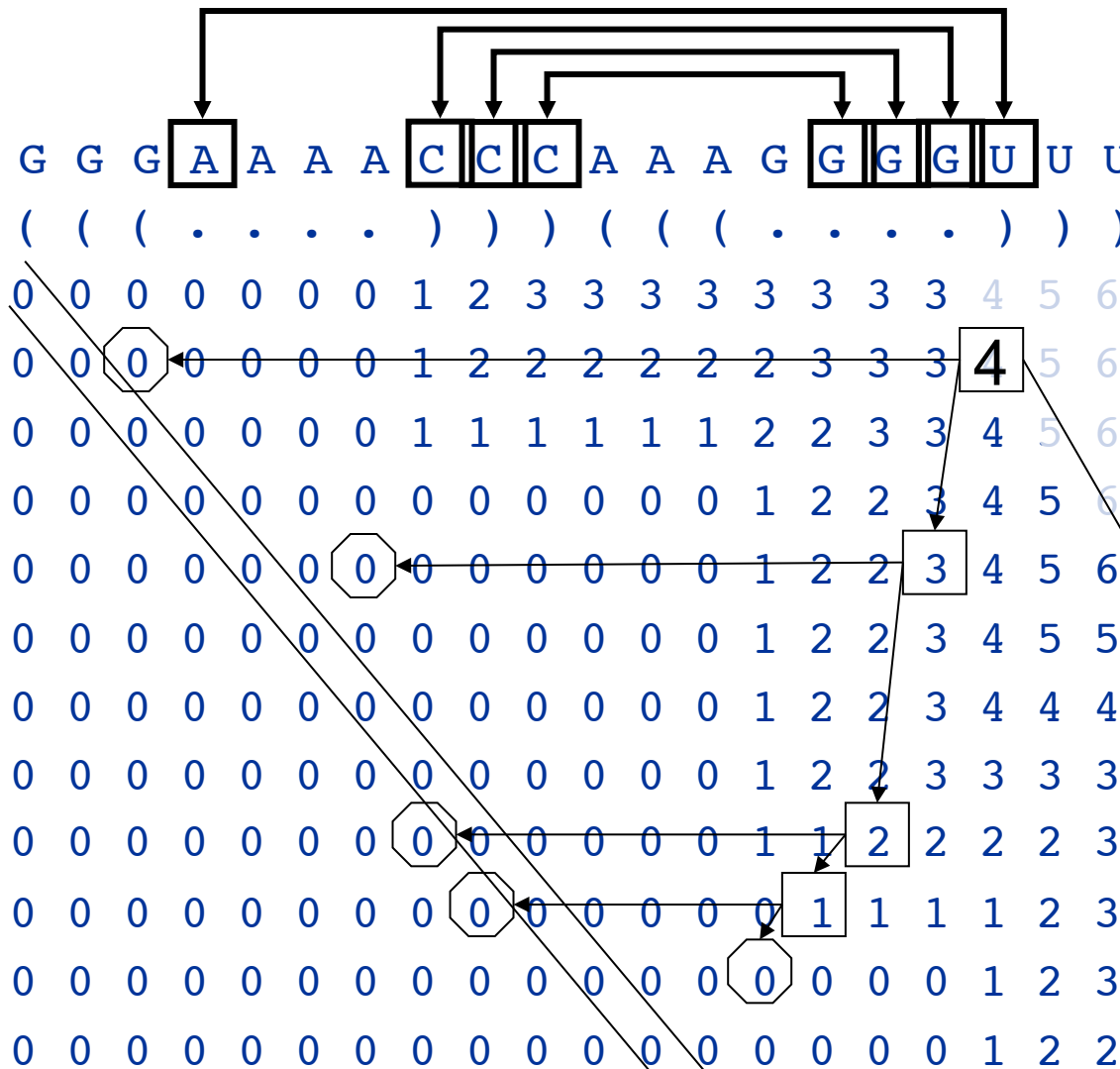
(not shown:
 $t=9,10,12,13$)

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

0
 0

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

$n = 20$



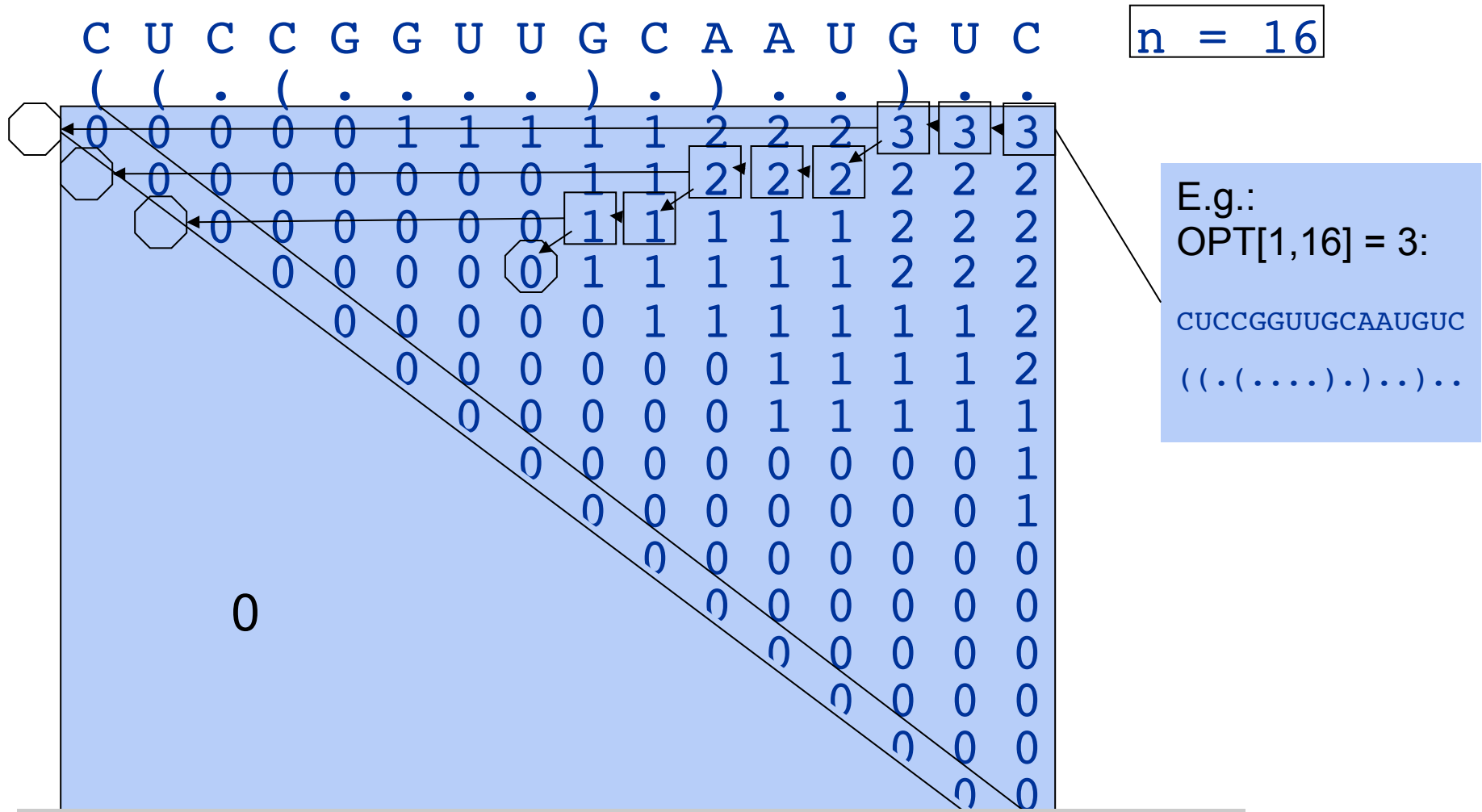
Overall, Max = 4
several ways, e.g.:

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

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



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

