# CSEP 590A Summer 2006 Lecture 8

RNA Secondary Structure Prediction

## Outline

Biological roles for RNA

What is "secondary structure?

How is it represented?

Why is it important?

**Examples** 

**Approaches** 

### RNA Structure

Primary Structure: Sequence

Secondary Structure: Pairing

Tertiary Structure: 3D shape

## RNA Pairing

Watson-Crick Pairing

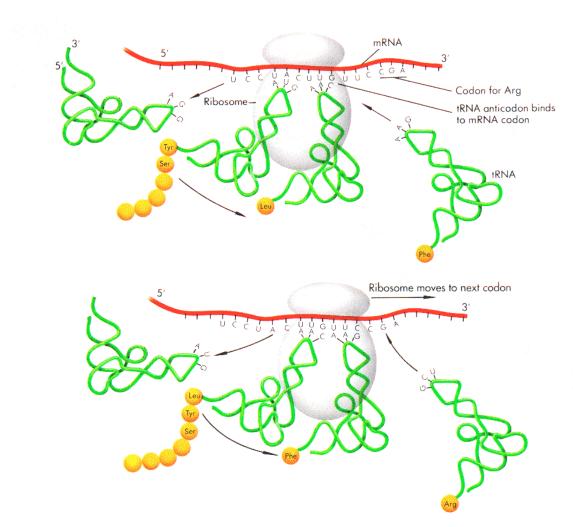
C - G ~ 3 kcal/mole

A - U ~ 2 kcal/mole

"Wobble Pair" G - U ~1 kcal/mole

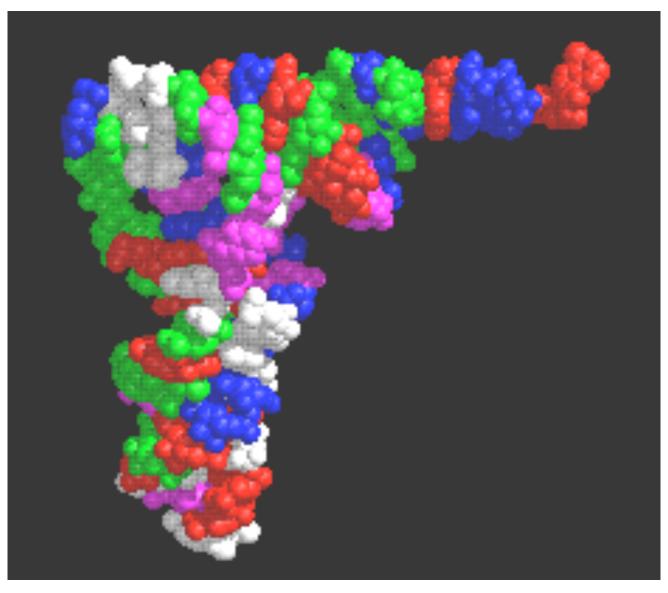
Non-canonical Pairs (esp. if modified)

### Ribosomes



Watson, Gilman, Witkowski, & Zoller, 1992

## tRNA 3d Structure



## tRNA - Alt. Representations

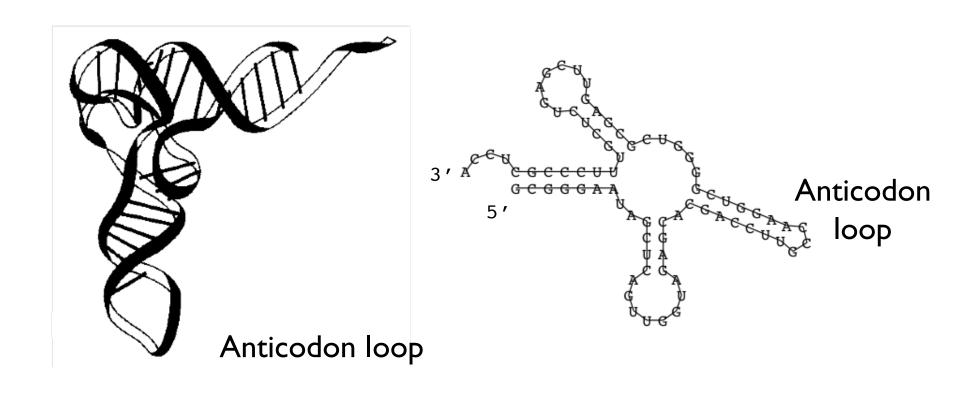
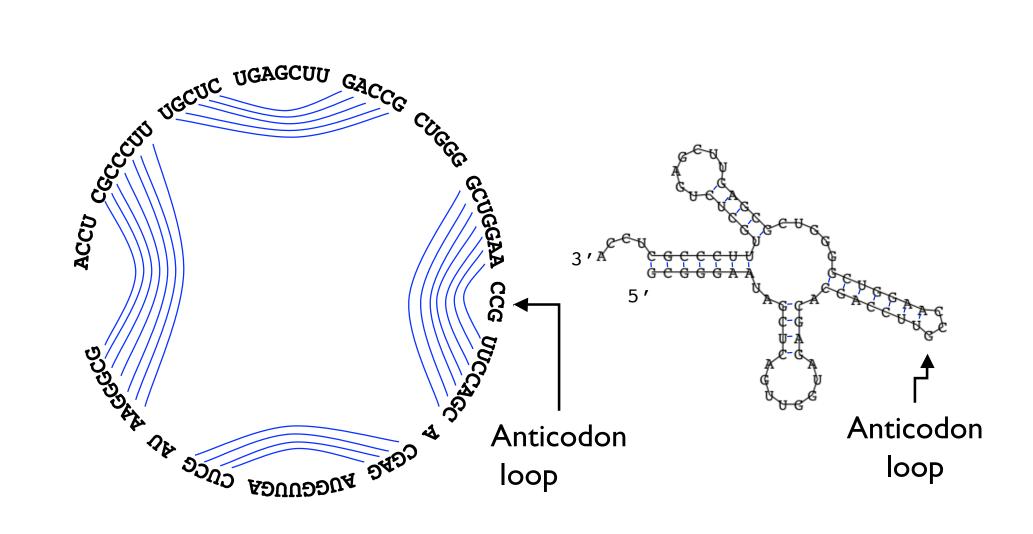


Figure 1: a) The spatial structure of the phenylalanine tRNA form yeast

b) The secondary structure extracts the most important information about the structure, namely the pattern of base pairings.

## tRNA - Alt. Representations



### "Classical" RNAs

```
tRNA - transfer RNA (~61 kinds, ~ 75 nt)

rRNA - ribosomal RNA (~4 kinds, 120-5k nt)

snRNA - small nuclear RNA (splicing: U1, etc, 60-300nt)

RNaseP - tRNA processing (~300 nt)

RNase MRP - rRNA processing; mito. rep. (~225 nt)

SRP - signal recognition particle; membrane targeting
 (~100-300 nt)

SECIS - selenocysteine insertion element (~65nt)

6S - ? (~175 nt)
```

## Semi-classical RNAs (discovery in mid 90's)

tmRNA - resetting stalled ribosomes

Telomerase - (200-400nt)

snoRNA - small nucleolar RNA (many varieties; 80-200nt)

### Recent discoveries

microRNAs

riboswitches

many ribozymes

regulatory elements

. . .

#### Hundreds of families

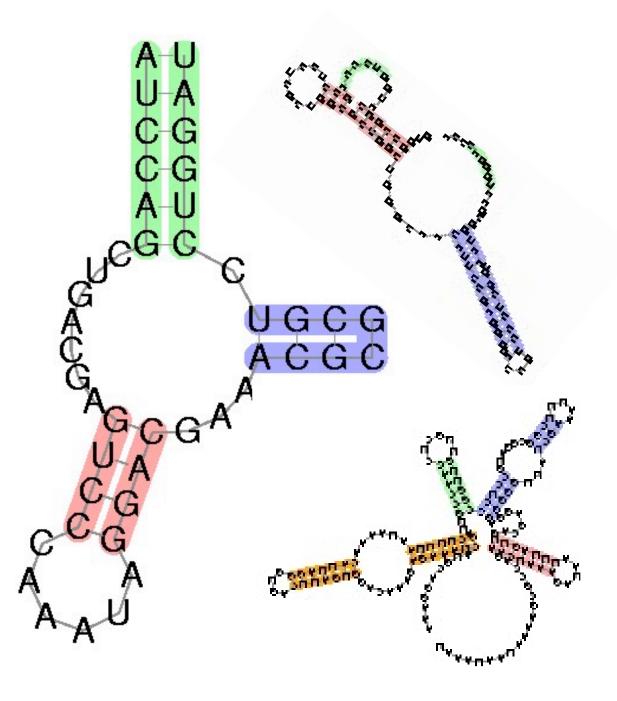
Rfam release I, I/2003: 25 families, 55k instances

Rfam release 7, 3/2005: 503 families, 300k instances

## Why?

RNA's fold, and function

Nature uses what works



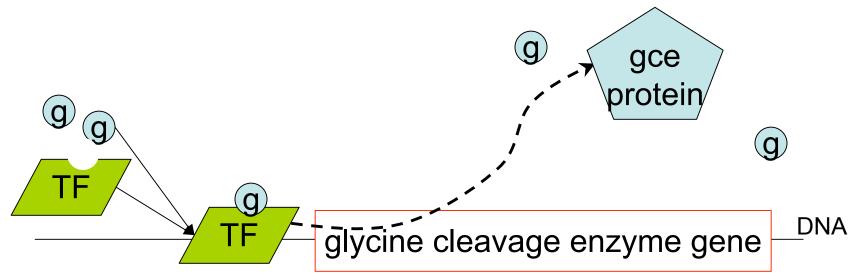


## Noncoding RNAs

### Example: Glycine Regulation

How is glycine level regulated?

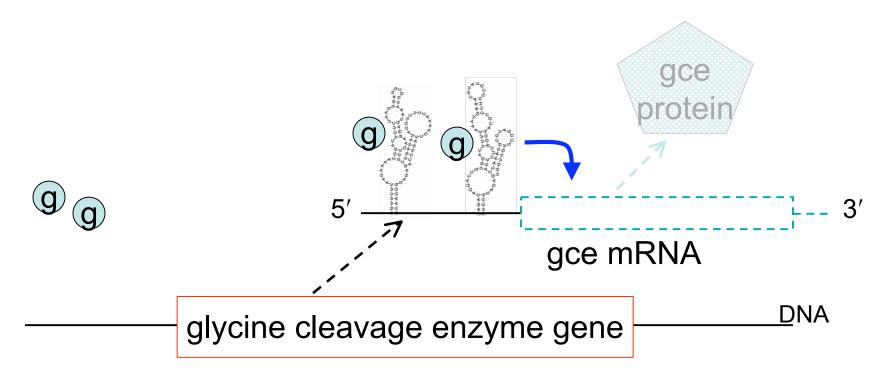
Plausible answer:



transcription factors (proteins) bind to DNA to turn nearby genes on or off

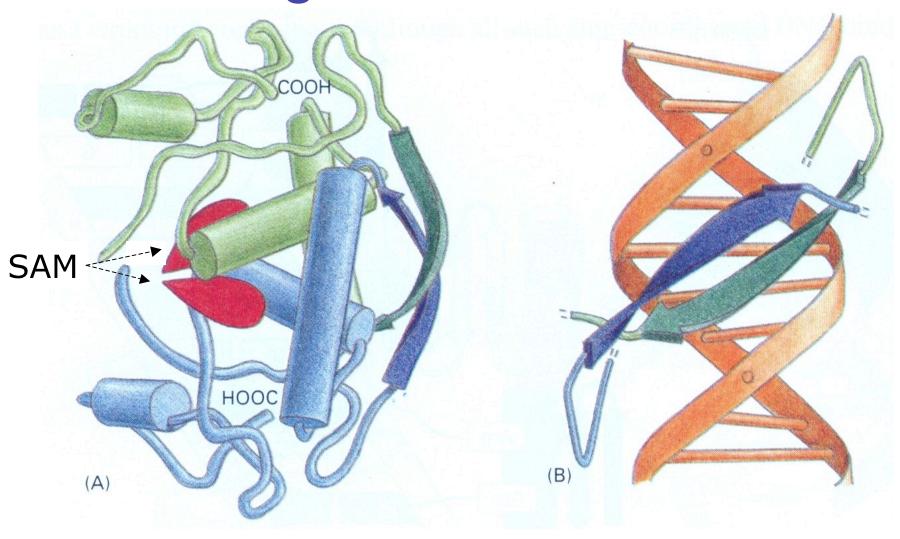
### The Glycine Riboswitch

Actual answer (in many bacteria):



Mandal et al. Science 2004

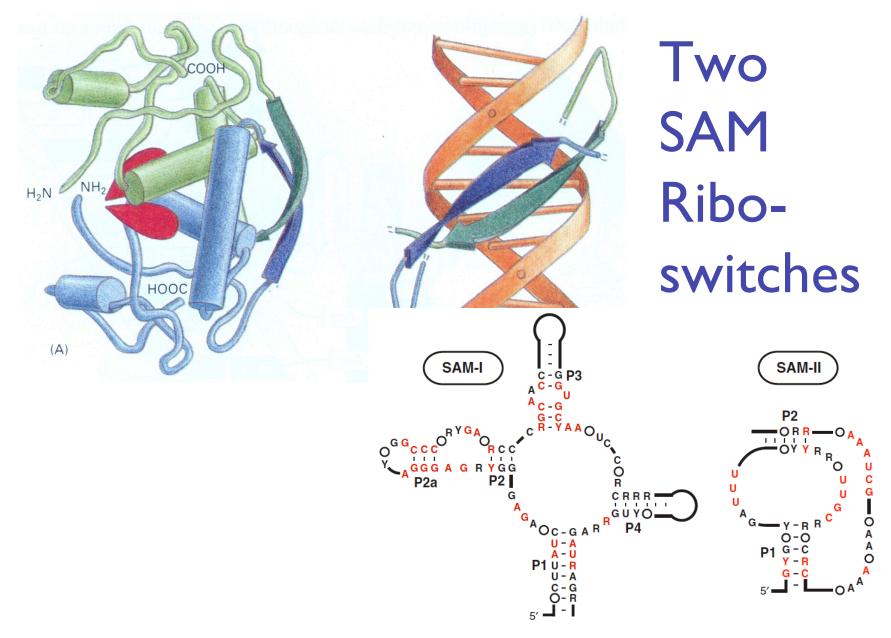
## Gene Regulation: The Met Repressor



Protein

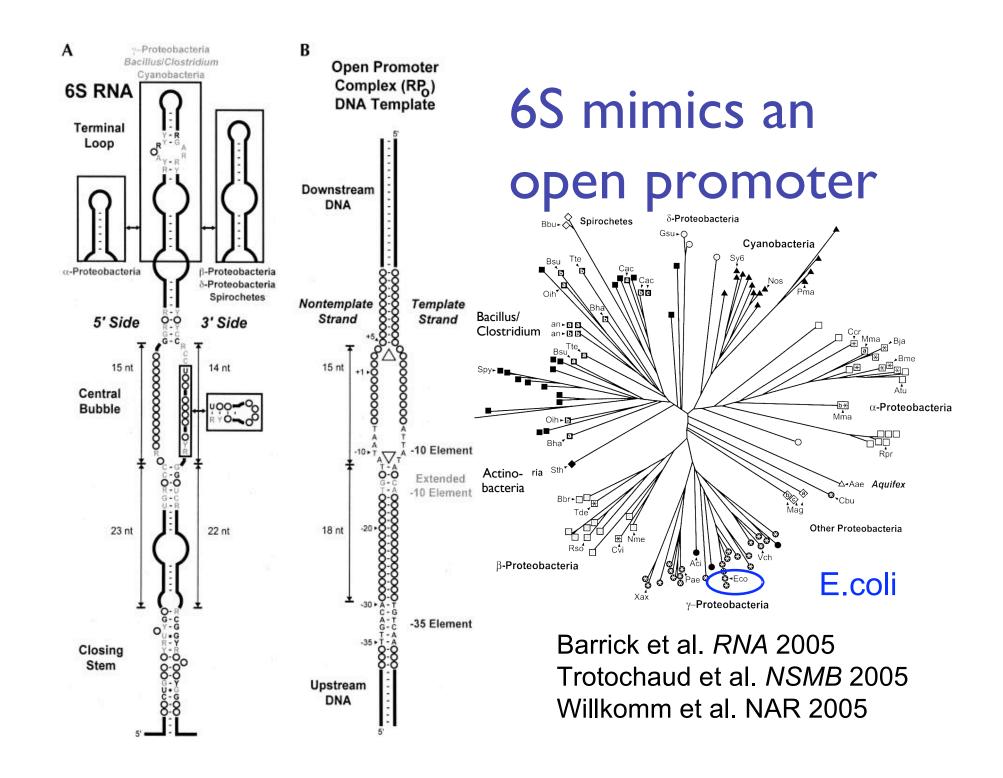
Alberts, et al, 3e.

DNA



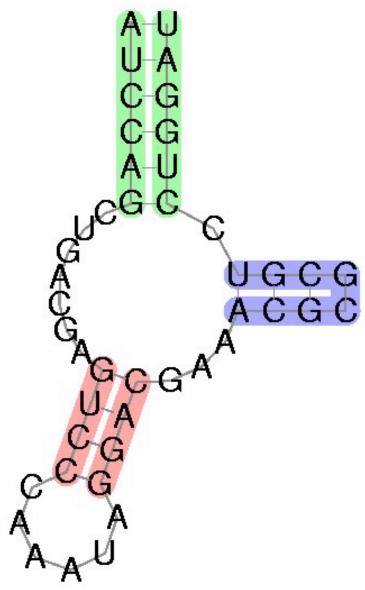
Alberts, et al, 3e.

Corbino et al., Genome Biol. 2005



# The Hammerhead Ribozyme

Involved in "rolling circle replication" of viruses.

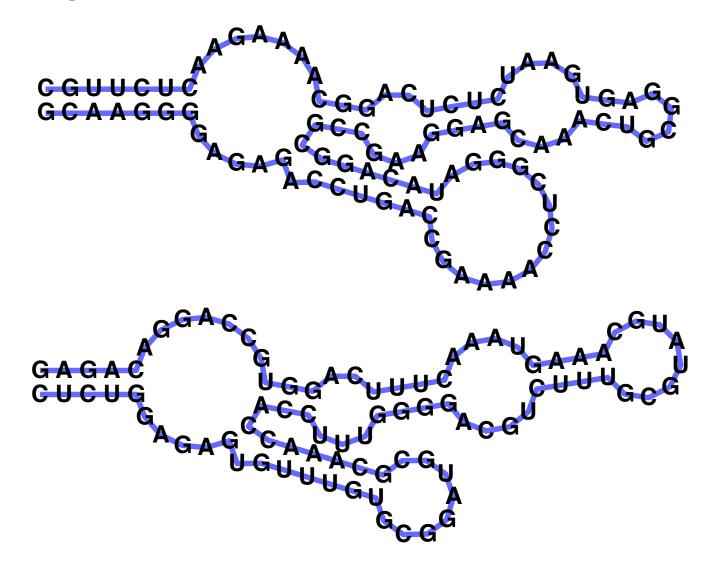


#### Wanted

Good structure prediction tools
Good motif descriptions/models
Good, fast search tools
("RNA BLAST", etc.)
Good, fast motif discovery tools
("RNA MEME", etc.)

Importance of structure makes last 3 hard

### Why is RNA hard to deal with?



A: Structure often more important than sequence

# Task I: Structure Prediction

## RNA Pairing

Watson-Crick Pairing

C - G ~ 3 kcal/mole

A - U ~ 2 kcal/mole

"Wobble Pair" G - U ~ 1 kcal/mole

Non-canonical Pairs (esp. if modified)

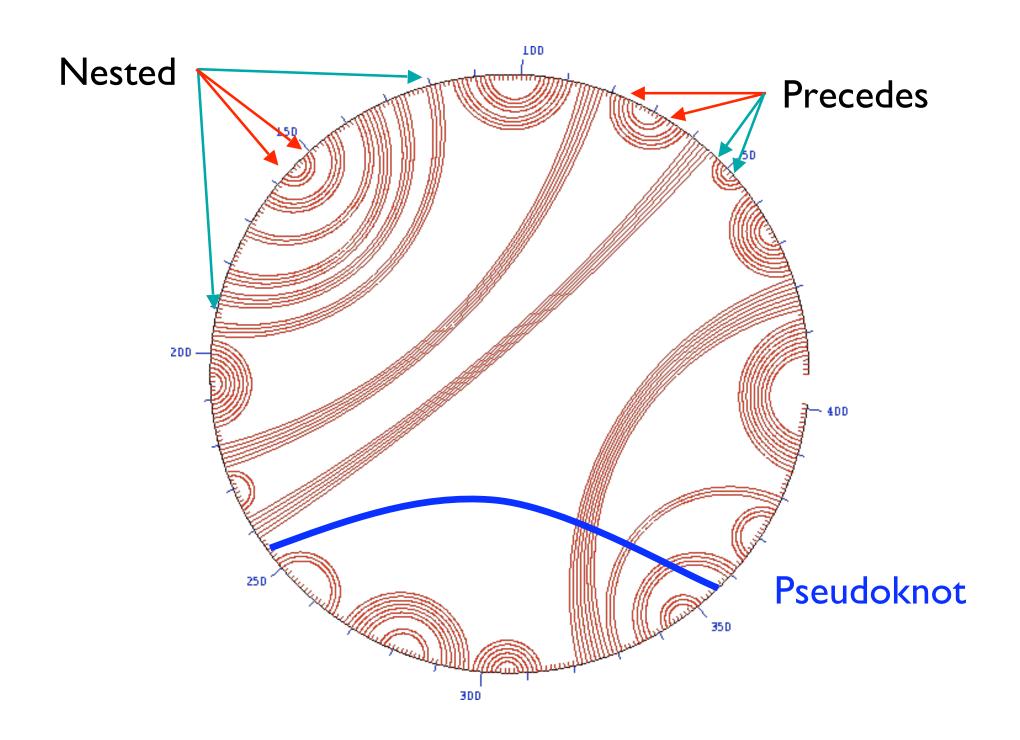
### **Definitions**

```
Sequence 5 r_1 r_2 r_3 ... r_n 3 in {A, C, G, T}

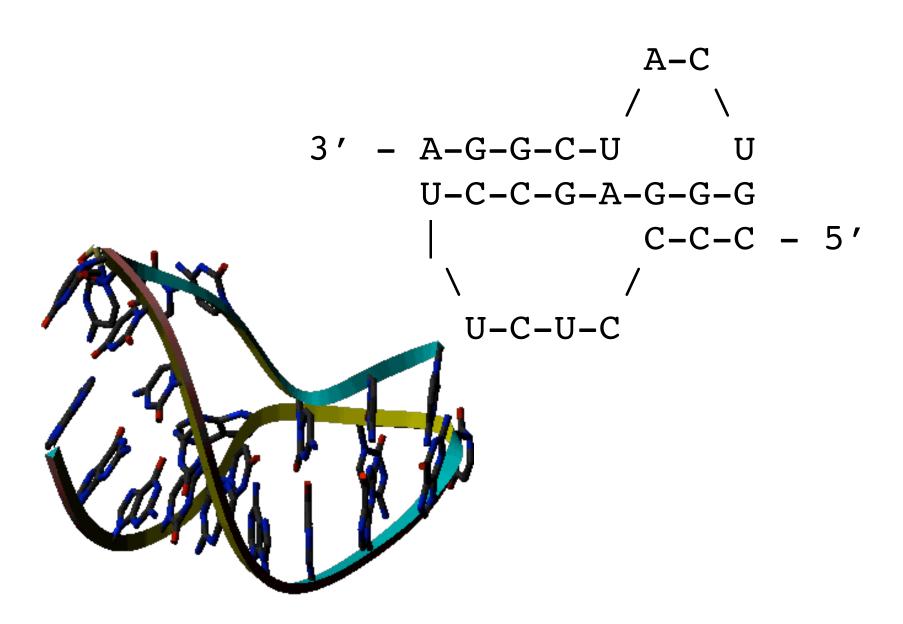
A Secondary Structure is a set of pairs i \cdot j s.t.

i < j-4, and

i < j-4
```



### A Pseudoknot



## Approaches to Structure Prediction

#### Maximum Pairing

- + works on single sequences
- + simple
- too inaccurate

#### Minimum Energy

- + works on single sequences
- ignores pseudoknots
- only finds "optimal" fold

#### Partition Function

- + finds all folds
- ignores pseudoknots

## Approaches, II

#### Comparative sequence analysis

- + handles all pairings (incl. pseudoknots)
- requires several (many?) aligned, appropriately diverged sequences

Stochastic Context-free Grammars

Roughly combines min energy & comparative,
but no pseudoknots

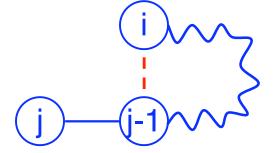
Physical experiments (x-ray crystalography, NMR)

## Nussinov: Max Pairing

```
B(i,j) = \# \ pairs \ in \ optimal \ pairing \ of \ r_i \ ... \ r_j
B(i,j) = 0 \ for \ all \ i, \ j \ with \ i \ge j-4; \ otherwise
B(i,j) = \max \ of:
\begin{cases} B(i,j-1) \\ \max \ \{ \ B(i,k-1)+1+B(k+1,j-1) \ | \\ i \le k < j-4 \ and \ r_k-r_j \ may \ pair \end{cases}
Time: O(n^3)
```

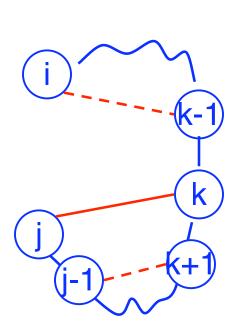
## "Optimal pairing of $r_i$ ... $r_j$ " Two possibilities

J Unpaired: Find best pairing of  $r_i \dots r_{i-1}$ 



J Paired:
Find best  $r_i$  ...  $r_{k-1}$  +
best  $r_{k+1}$  ...  $r_{i-1}$  plus I

Why is it slow?
Why do pseudoknots matter?



## Pair-based Energy Minimization

```
E(i,j) = \text{energy of pairs in optimal pairing of } r_i \dots r_j
E(i,j) = \infty \text{ for all } i, j \text{ with } i \ge j-4; \text{ otherwise}
E(i,j) = \min \text{ of:}
\begin{cases} E(i,j-1) & \text{energy of } j-k \text{ pair} \\ \min \left\{ E(i,k-1) + e(r_k, r_j) + E(k+1,j-1) \mid i \le k < j-4 \right\} \end{cases}
Time: O(n^3)
```

Loop-based Energy Minimization

Detailed experiments show it's more accurate to model based on loops, rather than just pairs

#### Loop types

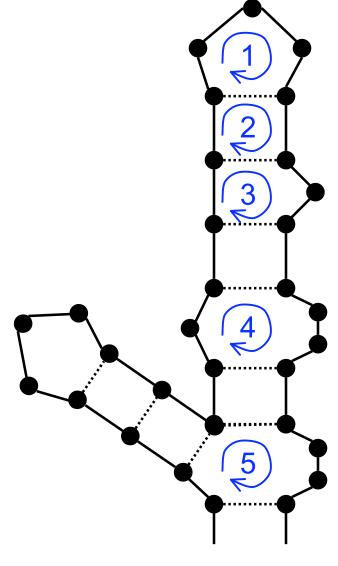
Hairpin loop

Stack

Bulge

Interior loop

Multiloop



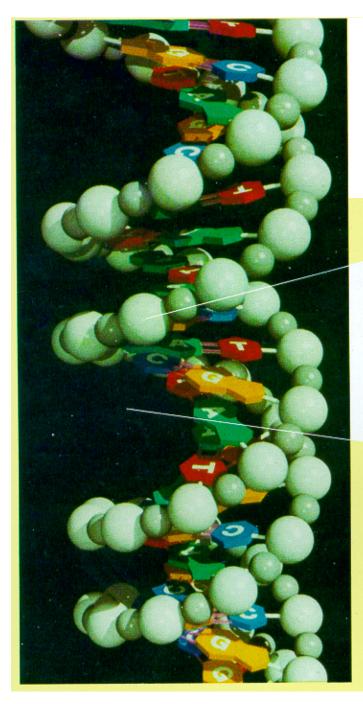
## Base Pairs and Stacking

uracil The state of the state o

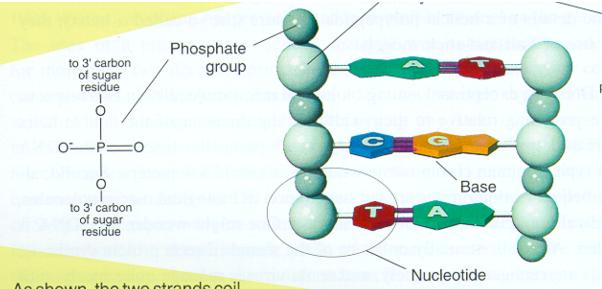
thymine

guanine

adenine



## The Double Helix



As shown, the two strands coil about each other in a fashion such that all the bases project inward toward the helix axis. The two strands are held together by hydrogen bonds (pink rods) linking each base projecting from one backbone to its so-called complementary base projecting from the other backbone. The base A always bonds to T (A and T are comple-

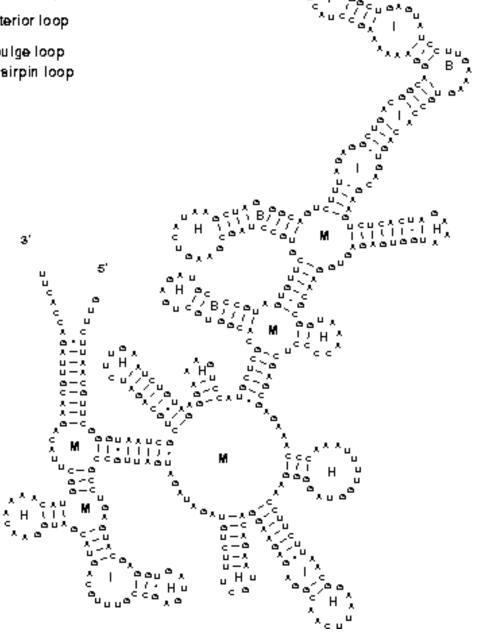
#### Shown in (b)

is an uncoiled fragment of (a three complementary base pai chemist's viewpoint, each stra a polymer made up of four re called deoxyribonucleotides



- I interior loop
- B bulge loop
- H-hairpin loop

# Loop Examples



# Zuker: Loop-based Energy, I

```
\begin{split} W(i,j) &= \text{energy of optimal pairing of } r_i \dots r_j \\ V(i,j) &= \text{as above, but forcing pair } i \bullet j \\ W(i,j) &= V(i,j) = \infty \text{ for all } i, j \text{ with } i \geq j\text{-}4 \\ W(i,j) &= \min(W(i,j\text{-}1), \\ \min \big\{ W(i,k\text{-}1) + V(k,j) \mid i \leq k \leq j\text{-}4 \big\} \\ \big) \end{split}
```

# Zuker: Loop-based Energy, II

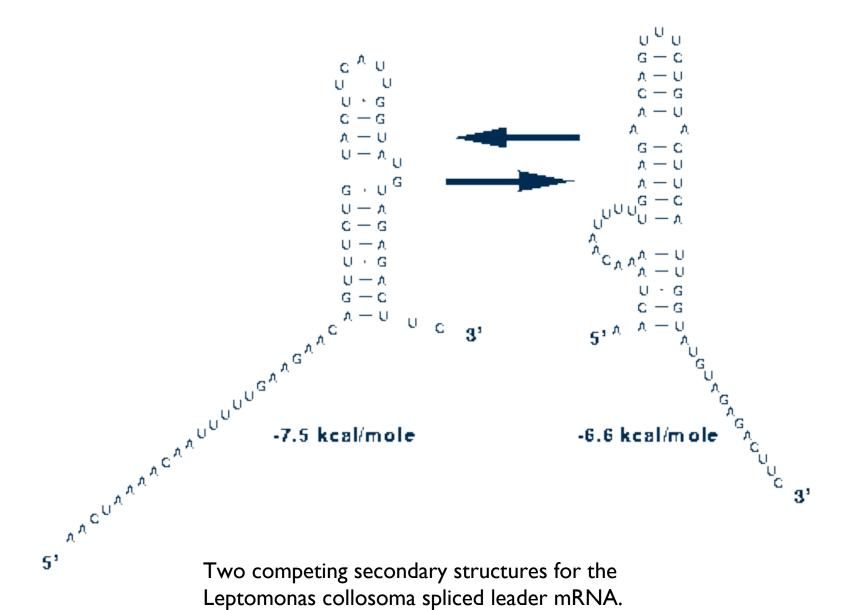
```
bulge/
                                                         multi-
                hairpin
                          stack
                                                interior
                                                          loop
V(i,j) = min(eh(i,j), es(i,j)+V(i+1,j-1), VBI(i,j), VM(i,j))
VM(i,j) = min \{ W(i,k)+W(k+1,j) \mid i < k < j \}
VBI(i,j) = min \{ ebi(i,j,i',j') + V(i', j') \}
                         i < i' < j' < j \& i'-i+j-j' > 2
                                                    Time: O(n^4)
          bulge/
         interior
                           O(n<sup>3</sup>) possible if ebi(.) is "nice"
```

## Suboptimal Energy

There are always alternate folds with near-optimal energies. Thermodynamics: populations of identical molecules will exist in different folds; individual molecules even flicker among different folds

Mod to Zuker's algorithm finds subopt folds

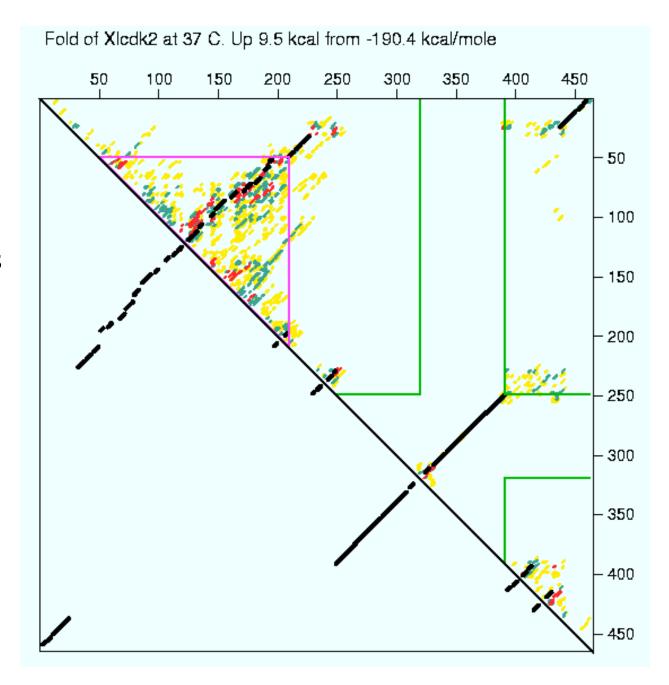
McCaskill: more elaborate dyn. prog. algorithm calculates the "partition function," which defines the probability distribution over all these states.



# Example of suboptimal folding

Black dots: pairs in opt fold

Colored dots: pairs in folds 2-5% worse than optimal fold



## Accuracy

Latest estimates suggest ~50-75% of base pairs predicted correctly in sequences of up to ~300nt

Definitely useful, but obviously imperfect

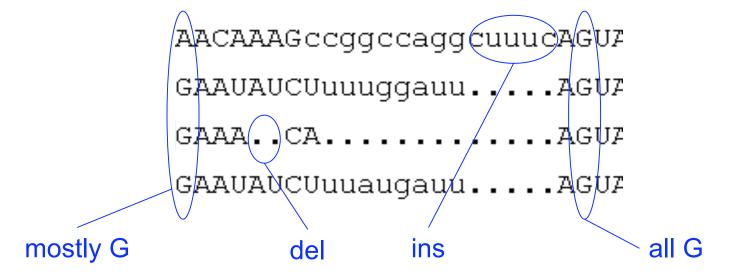
# Task 2: Motif Description

#### How to model an RNA "Motif"?

#### Conceptually, start with a profile HMM:

from a multiple alignment, estimate nucleotide/ insert/delete preferences for each position

given a new seq, estimate likelihood that it could be generated by the model, & align it to the model



#### How to model an RNA "Motif"?

Add "column pairs" and pair emission probabilities for base-paired regions

cuuuugc	AAACAAAGccggccaggcuuucAGUA.	.G	UGAAA	<b>√</b> G
GGAAUGU	GAAUAUCUuuuggauuAGUA	λG	CAUUC	:C
UUCAUUA	GAAACAAGUA	JU.	AAUGG	βA
GGAAUGU	GAAUAUCUuuaugauuAGUA	A	CAUUC	C
<<<<<	paired columns		>>>>	

#### RNA Motif Models

"Covariance Models" (Eddy & Durbin 1994)

aka profile stochastic context-free grammars
aka hidden Markov models on steroids

Model position-specific nucleotide
preferences and base-pair preferences

Pro: accurate

Con: model building hard, search sloooow

### Summary

RNA has important roles beyond mRNA Many unexpected recent discoveries

Structure is critical to function

True of proteins, too, but they're easier to find, due, e.g., to codon structure, which RNAs lack

RNA secondary structure can be predicted (to useful accuracy) by dynamic programming

RNA "motifs" (seq + 2-ary struct) well-captured by "covariance models"