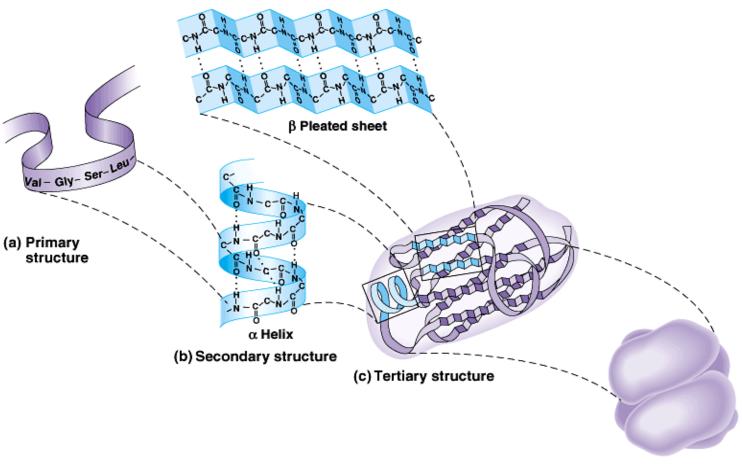
Nucleic Acid Secondary Structure

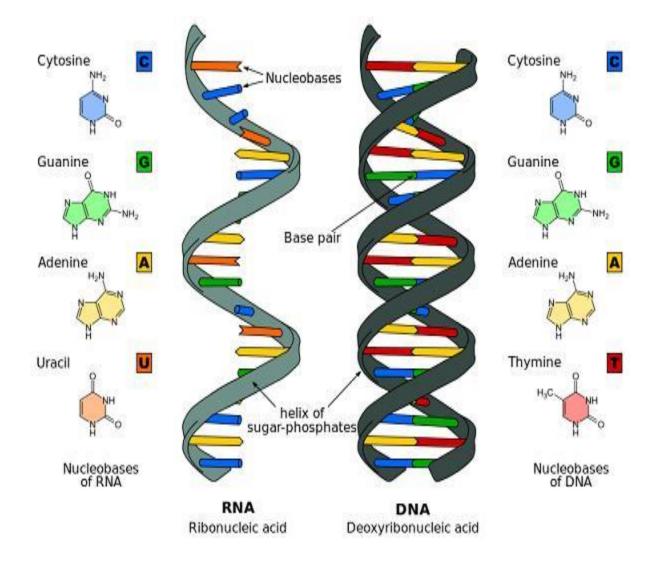
Structure of protein



(d) Quaternary structure

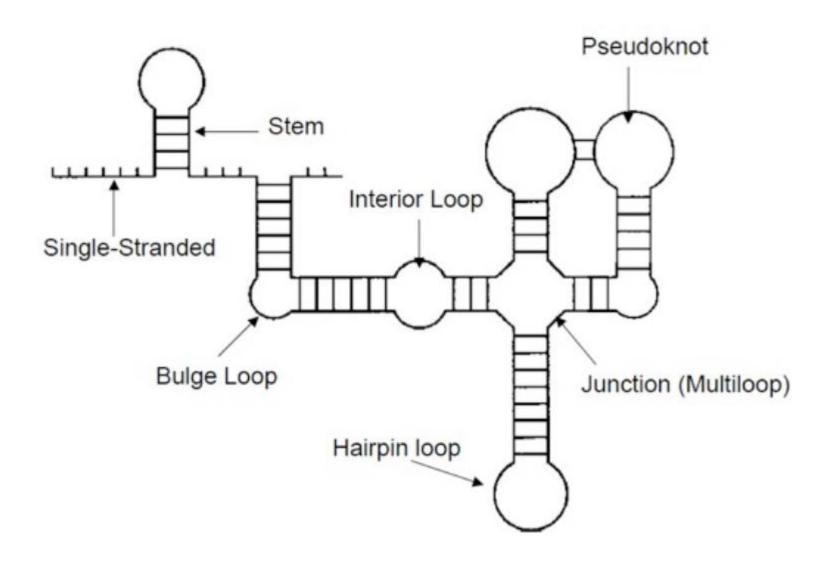
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- The secondary structure of a nucleic acid molecule refers to the base paring interactions within a single molecule or set of interacting molecules.
- > DNA molecules exist predominantly in the double stranded form.
- ➤ On the other hand, RNA molecules exist mostly as single strands.
- ➤ However, the single strand of RNA can twist and fold back on itself to form a well defined three dimensional structure.

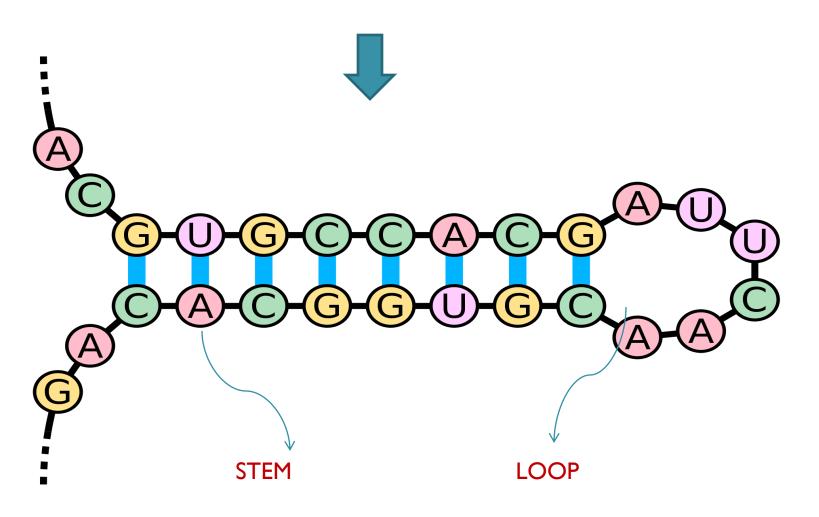


- Perhaps the most common structure that RNA molecules fold into is called Stem Loop.
- This structure is formed when the ribonucleic acid folds onto itself to form a double helical structure that consist of two complementary sequences.
- It also commonly contains mismatched nucleotides that bulge out of the structure.

Secondary structure elements



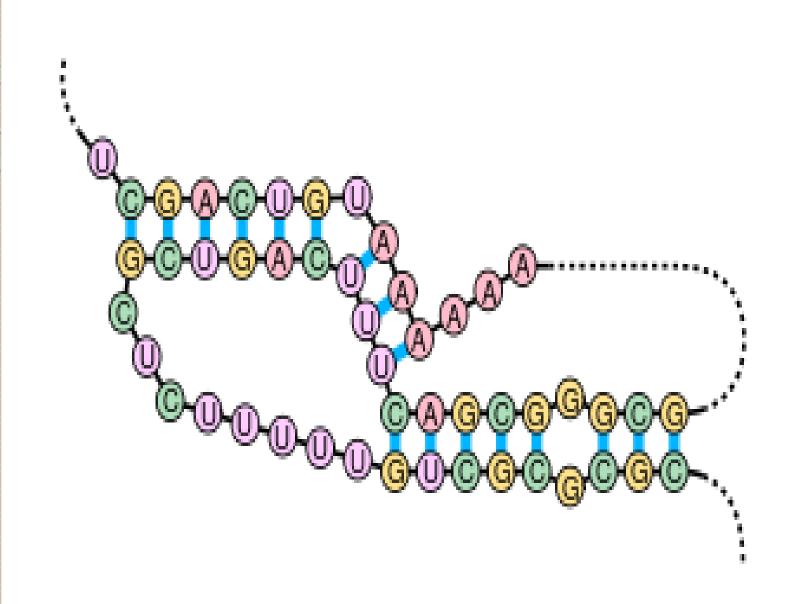
5' ACGUGCCACGAUUCAACGUGGCACAG3'



Pseudoknots

- A pseudoknot is a nucleic acid secondary structure containing at least two stem-loop structures in which half of one stem is intercalated between the two halves of another stem.
- ➤ Pseudoknots fold into knot-shaped three-dimensional conformations but are not true topological knots.



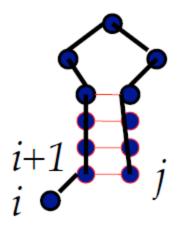


Secondary structure prediction

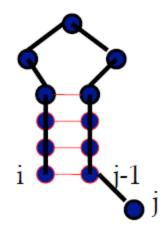
- Nucleic acid structure prediction is a computational method to determine nucleic acid secondary and tertiary structure from its sequence.
- > Secondary structure can be predicted from a single or from several nucleic acid sequences.
- > Tertiary structure can be predicted from the sequence, or by comparative modeling.
- The problem of predicting nucleic acid secondary structure is dependent mainly on base pairing and base stacking interactions; many molecules have several possible three-dimensional structures.

There are only four possible ways of getting the best structure for subsequence (i,j) from the best structures of the smaller subsequences

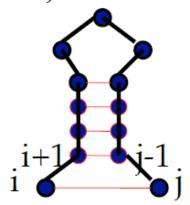
1) Add unpaired position i onto best structure for subsequence (i+1,j)



2) Add unpaired position j onto best structure for subsequence (i,j-1)

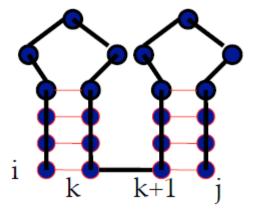


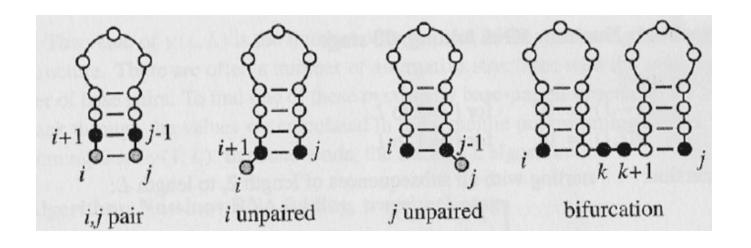
3) Add (i,j) pair onto best structure for subsequence (i+1,j-1)



4) Combine two optimal substructures (i, k) and

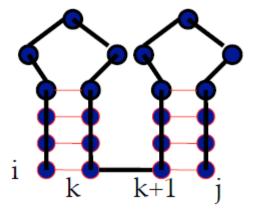
(k+1, j)

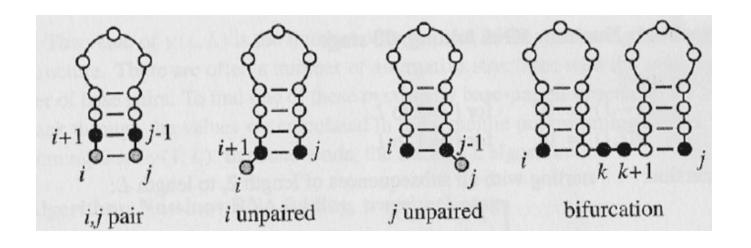


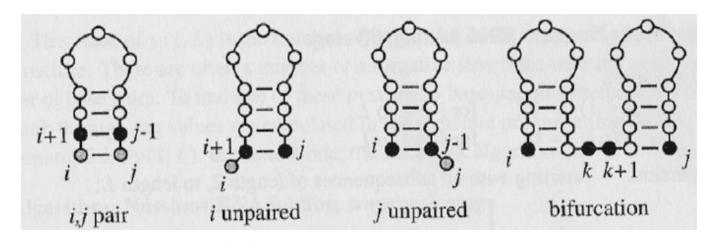


4) Combine two optimal substructures (i, k) and

(k+1, j)







$$S(i,j) = max \begin{cases} S(i+1,j-1) + w(i,j) \\ S(i+1,j) \\ S(i,j-1) \\ max_{i < k < j} S(i,k) + S(k+1,j) \end{cases}$$

Initialization

- S(i,i) = 0

Ruth Nussinov and co-workers who used the dynamic programming method for maximizing the number of base-pairs.

Nussinov folding algorithm (I)

- Let S[1..n] be the RNA sequence
- Let V(i,j) be the maximum number of base pairs in S[i..j].
- Base case:
 - V(i,i)=0 since the sequence has only one base!
 - V(i+1,i)= 0 since the sequence is empty!

Nussinov folding algorithm (II)

- When i<j, we have four cases:</p>
 - No base pair attached to j
 - V(i, j) = V(i, j-1)
 - 2. No base pair attached to i
 - V(i,j) = V(i+1, j)
 - 3. (i, j) form a base pair
 - $V(i, j) = V(i+1, j-1) + \delta(S[i], S[j])$ where $\delta(x, y)=1$ if $(x,y)\in\{(a,u), (u,a), (c,g), (g,c), (g,u), (u,g)\}$; and 0, otherwise
 - Both I and j attached to some base pairs both (i,j) is not a base pair
 - $V(i, j) = \max_{i \le k < i} \{V(i, k) + V(k+1, j)\}$

Traceback

```
Input: Matrix S and positions i, j.
Output: Secondary structure maximizing the number of
base pairs.
Initial call: traceback(i = 1, j = L).
if i < j then
  if S(i, j) = S(i + 1, j) then
                                                   // \operatorname{case}(1)
     traceback(i + 1, j)
  else if S(i, j) = S(i, j - 1) then
                                                   // \operatorname{case}(2)
     traceback(i, j - 1)
  else if S(i, j) = S(i + 1, j - 1) + w(i, j) then // case (3)
     print base pair (i, j)
     traceback(i+1, j-1)
  else for k = i + 1 to j - 1 do
                                                   // \operatorname{case} (4)
     if S(i,j) = S(i,k) + S(k+1,j) then
       traceback(i, k)
       traceback(k + 1, j)
       break
end
```

Example: base case

S[1..7]=ACCAGCU

$$S(i,j) = max \begin{cases} S(i+1,j-1) + w(i,j) \\ S(i+1,j) \\ S(i,j-1) \\ max_{i < k < j} S(i,k) + S(k+1,j) \end{cases}$$

- Initialization
 - S(i,i-1) = 0

	1	2	3	4	5	6	7
1	0						
2	0	0					
3		0	0				
4			0	0			
5				0	0		
6					0	0	
7						0	0

Example: recursive case (I)

• S[1..7]=ACCAGCU

$$S(i,j) = max \begin{cases} S(i+1,j-1) + w(i,j) \\ S(i+1,j) \\ S(i,j-1) \\ max_{i < k < j} S(i,k) + S(k+1,j) \end{cases}$$

V (3,5)=max number of base pairs in S[3..5].

By the recursive formula, max

$$V(3,5)=\max\{V(4,4)+\delta(S[3],S[5]),\ V(3,k)+V(k+1,5)\}=\max\{V(4,4)+1,\ V(3,3)+V(4,5),\ V(3,4)+V(5,5)\}=1$$

C		1	2	3	4	5	6	7
1		0	0	0				
2		0	0	0	0			
3			0	0	0	*		
4				0	0	0		
5					0	0	1	
6						0	0	0
1	,						0	0

Example: recursive case (II)

S[1..7]=ACCAGCU

$$S(i,j) = max \begin{cases} S(i+1,j-1) + w(i,j) \\ S(i+1,j) \\ S(i,j-1) \\ max_{i < k < j} S(i,k) + S(k+1,j) \end{cases}$$

V (4,7)=max number of base pairs in S[4..6].

By the recursive formula, max

$$V(4,7) = \max\{V(5,6) + \delta(S[4],S[7]), V(4,k) + V(k+1,7)\} = \max\{V(5,6) + 1, V(4,4) + V(5,7), V(4,5) + V(6,7), V(4,6) + V(7,7)\} = 2$$

С	1	2	3	4	5	6	7
1	0	0	0	0			
2	0	0	0	0	1		
3		0	0	0	1	1	
4			0	0	0	1	•
5				0	0	1	1
6					0	0	0
7						0	0

Example: recursive case (III)

S[1..7]=ACCAGCU

C	1	2	3	4	5	6	7
1	0	0	0	0	1	1	2
2	0	0	0	0	1	1	2
3		0	0	0	1	1	2
4			0	0	0	1	2
5				0	0	1	1
6					0	0	0
7						0	0

Nussinov folding algorithm (IV)

- Time analysis:
 - We need to fill-in O(n²) V(i,j) entries
 - Each V(i,j) entry can be computed in O(n) time.
 - Thus, Nussinov algorithm can be solved in O(n³) time.

THANK YOU