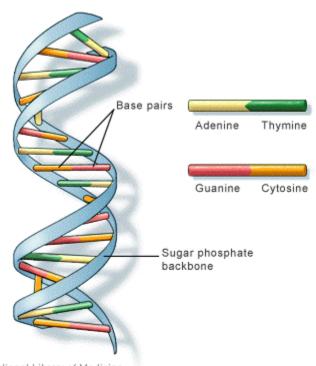
A problem from Computational Biology

Basic Biology

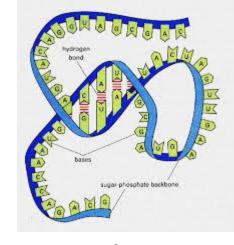
- Watson and Crick gave the double stranded DNA model where two strands are zipped together by complementary base-pairing
- Bases {A,C, G, T}
- Pairing A-T, C-G

Courtesy: ttp://www.chemguide.co.u

ttp://www.chemguide.co.uk/organicprops/aminoacids/dna1.html



RNA



- RNA is a basic biological molecule. It is single stranded.
- RNA molecules fold into complex secondary structures.
- Secondary structure often governs the behaviour of an RNA molecule.
- Various rules govern secondary structure formation:

Problem

- Pairs of bases match up;
 each base matches with
 1 other base.
- Adenine always matches with Uracil.
- Cytosine always matches with Guanine.
- There are no kinks in the folded molecule.
- Structures are knot-free.

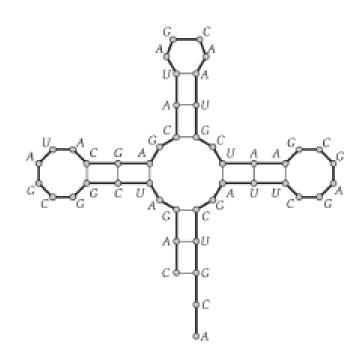


Figure 6.13 An RNA secondary structure. Thick lines connect adjacent elements of the sequence; thin lines indicate pairs of elements that are matched.

Problem: given an RNA molecule, predict its secondary structure.

Formulation

An RNA molecule is a string B = b₁b₂ ... b_n;
 each

$$b_i \in \{A, C, G, U\}$$

An secondary structure on B is a set of pairs S
 = {(i,j)}, where 1 ≤ i, j ≤ n and satisfies the
 following rules.

Rules

- (No sharp turns) The ends of each pair are separated by at least 4 intervening bases i.e. if (i , j) ϵ S, then i < j 4.
- The elements in each pair in S consist of either {A,U} or {C,G} (in either order).
- S is a matching: no base appears in more than one pair.
- (No knots) If (i,j) and (k,l) are two pairs in S, then we cannot have i < k < j < l.

The energy of a secondary structure is proportional to the number of base pairs in it.

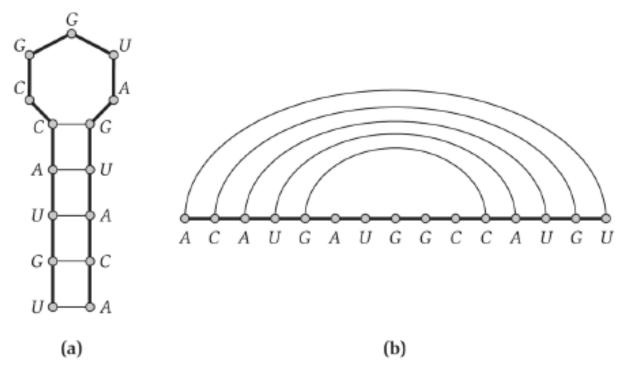


Figure 6.14 Two views of an RNA secondary structure. In the second view, (b), the string has been "stretched" lengthwise, and edges connecting matched pairs appear as noncrossing "bubbles" over the string.

First approach

- OPT(j) is the maximum number of base pairs in a secondary structure for b₁b₂ ... b_j. OPT(j) = 0, if j ≤ 5.
- In the optimal secondary structure on b_1b_2 ... b_i
 - if j is not a member of any pair, use OPT(j-1).
 - if j pairs with some t < j 4, knot condition yields two independent sub-problems! OPT(t -1) and ???

Pictorially

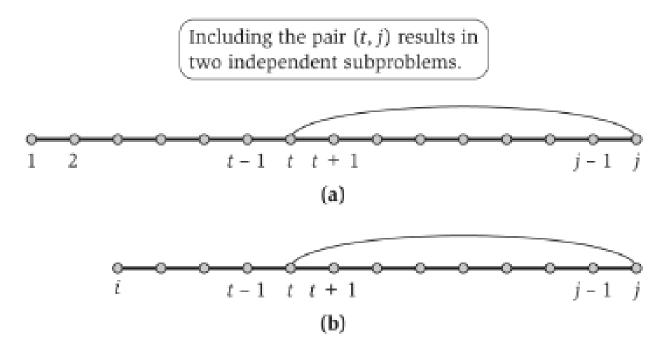


Figure 6.15 Schematic views of the dynamic programming recurrence using (a) one variable, and (b) two variables.

Correct approach

- OPT(i,j) is the maximum number of base pairs in a secondary structure for b_ib_{i+1} ... b_j. OPT(i,j) = 0, if i≥j-4.
- In the optimal secondary structure on $b_i b_{i+1} \dots b_i$
 - if j is not a member of any pair, use OPT(i,j-1).
 - if j pairs with some t < j 4, knot condition yields two independent sub-problems! OPT(i,t -1) and OPT(t+1,j-1)

Recurrence

$$OPT(i, j) = \max \begin{cases} OPT(i, j-1) \\ \max_{t} (1 + OPT(i, t-1) + OPT(t+1, j-1)) \end{cases}$$

t ranges from I to j-5 such that it is allowed to pair with j

There are $O(n^2)$ sub-problems. How do we order them from "smallest" to "largest"? Note that computing OPT(i , j) involves sub-problems OPT(l ,m) where m - l < j - i.

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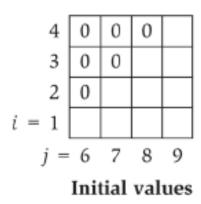
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Algorithm

```
Initialize \mathrm{OPT}(i,j)=0 whenever i\geq j-4 For k=5,\ 6,\dots,n-1 For i=1,2,\dots n-k Set j=i+k Compute \mathrm{OPT}(i,j) using the recurrence in (6.13) Endfor Endfor Return \mathrm{OPT}(1,n)
```

Lets do an example

Input ACCGGUAGU



Filling in the values for
$$k = 5$$

Filling in the values for k = 6

Filling in the values

for k = 7

Filling in the values

for k = 8

Reference

 Algorithm Design by Eva Tardos and Jon Kleinberg.