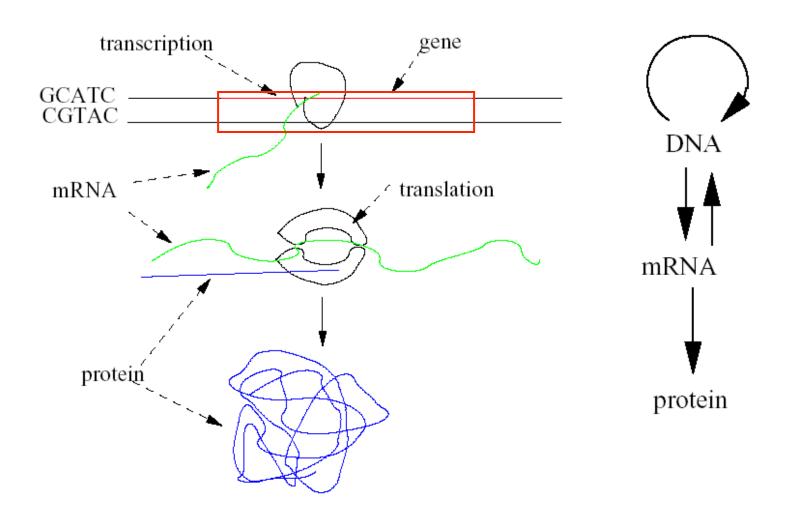
Take-home Problem: the number of alignments

- Let A(n,m) = the number of alignments of one sequence with length i and one sequence with length j.
- A(n,m)=A(n-1,m)+A(n-1,m-1)+A(n,m-1).

Non-coding RNA gene finding problems

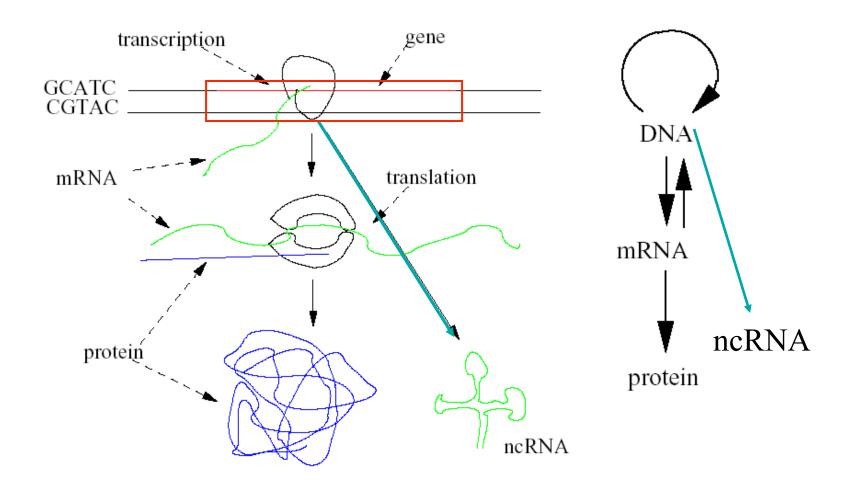
Central dogma cont' d



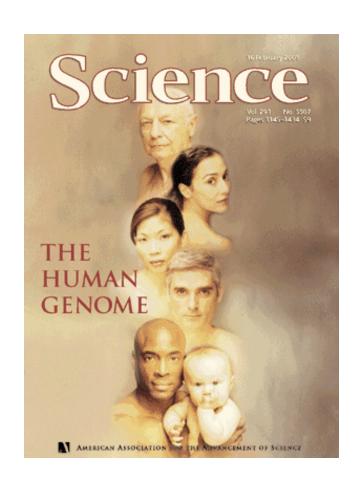
Central dogma cont' d

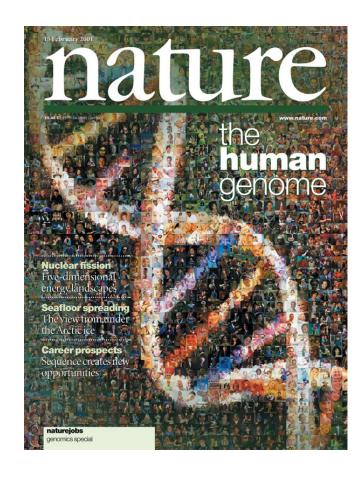
- Non-coding RNA (ncRNA)
 - RNA acting as functional molecule.
 - Not translated into protein.
- Non-coding RNA gene
 - The region of DNA coding ncRNA.

Central dogma cont' d



Human genome





How many genes do we have?

- Only about 30,000 to 40,000 protein-coding genes in the human genome [Lander et al. Nature (2001), Venter et al. Science (2001)].
- Total protein coding gene length is only about 1.5 percent of the human genome. (3*109 bases)

What did we miss out?

- Current gene prediction methods only work well for protein coding genes.
- Non-coding RNA genes are undetected because they do not encode proteins.
- Modern RNA world hypothesis:
 - There are many unknown but functional ncRNAs. [Eddy Nature Reviews (2001)]
 - Many ncRNAs may play important role in the unexplained phenomenon.[Storz Science (2002)]

Question:

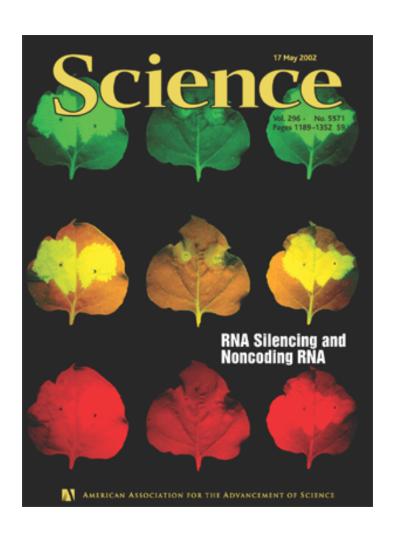
If there are many ncRNAs, what are they doing?

Question:

Biologically, why do we need functional ncRNAs in addition to protein?

Why do we need ncRNAs?

- ncRNAs involve sequence specific recognition of other nucleic acids (e.g. mRNAs, DNAs).
- ncRNA is an ideal material for this role.
 - DNA is big and packaged and can do this job.
- Base complementary allows ncRNA to be sequence specific!
- For example:
 - small interfering RNAs (siRNA) is used to protect our genome.
 - It recognizes invading foreign RNAs/DNAs based on the sequence specificity.
 - And helps to degrade the foreign RNAs.





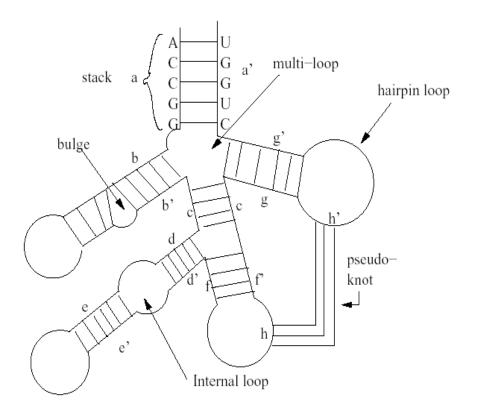
How can we find such ncRNA genes in the genome?

RNA secondary structure

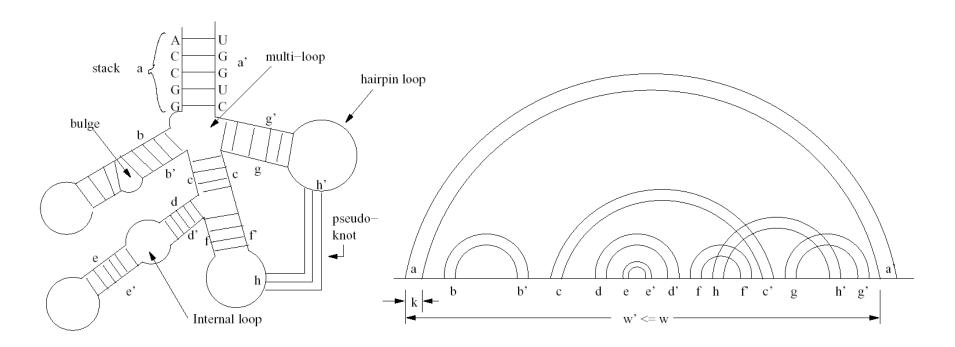
- ncRNA is not a random sequence.
- Most RNAs fold into particular base-paired secondary structure.
- Canonical basepairs:
 - Watson-Crick basepairs:
 - G C
 - A U
 - Wobble basepair:
 - G U

RNA secondary structure cont'd

- Stacks: continuous nested basepairs. (energetically favorable)
- Non-basepaired loops:
 - Hairpin loop.
 - Bulge.
 - Internal loop.
 - Multiloop.



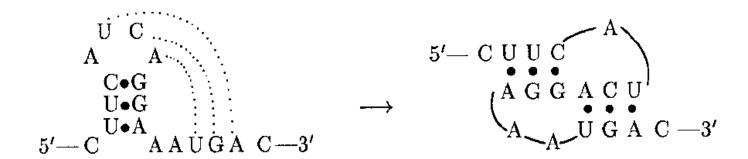
RNA secondary structure cont'd



- Most basepairs are non-crossing basepairs.
 - Any two pairs (i, j) and $(i', j') \rightarrow i < i' < j' < j$ or i' < i < j < j'
- Pseudoknots are the crossing basepairs.

Pseudoknots

- Pseudoknots are important for certain ncRNAs
- Violate the non-crossing assumption.
- Pseudoknots make most problems harder
- We assume there are no pseudoknots otherwise noted.



[Rivas and Eddy (1999)]

RNA secondary structure prediction

- It is a basic issue in ncRNA analysis
- It is important information to the biologists.
- Searching and alignment algorithms are based on these models.
- RNA secondary structure -- a set of noncrossing base pairs.

Base pair maximization problem

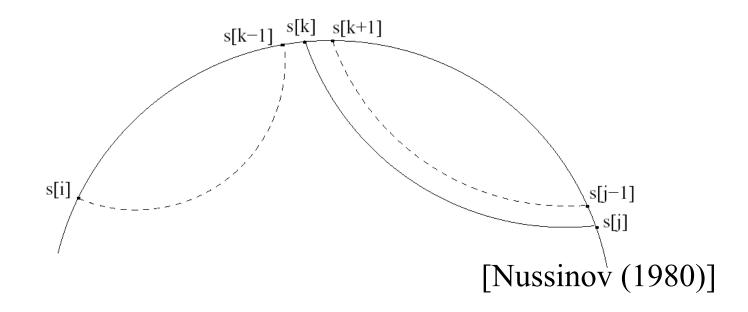
- A simple energy model is to maximize the number of basepairs to minimize the free energy. [Waterman (1978), Nussinov et al (1978), Waterman and Smith (1978)]
- G C, A U, and G U are treated as equal stability.
- Contributions of stacking are ignored.

Problem 1: [Base pair maximization problem]

Given an RNA sequence, determine a set of base pairs in a RNA sequence such that the number of base pairs is maximal and no base pairs cross each other.

A dynamic programming solution

- Let s[1...n] be an RNA sequence.
- $\delta(i,j) = 1$ if s[i] and s[j] form a complementary base pair, else $\delta(i,j) = 0$.
- M(i,j) is the maximum number of base pairs in s[i...j].



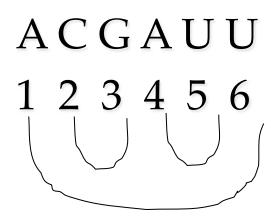
A dynamic programming solution

$$M(i,j) = \max \begin{cases} M(i,j-1), \\ M(i,k-1) + M(k+1,j-1) + \delta(k,j) \\ \text{for } i \leq k < j. \end{cases}$$

- M(1,n) is the number of base pairs in the optimal basepaired structure for s[1...n].
- All these basepairs can be found by tracing back through the matrix M.
- Filling M needs $O(n^3)$ time.

RNA structure: example

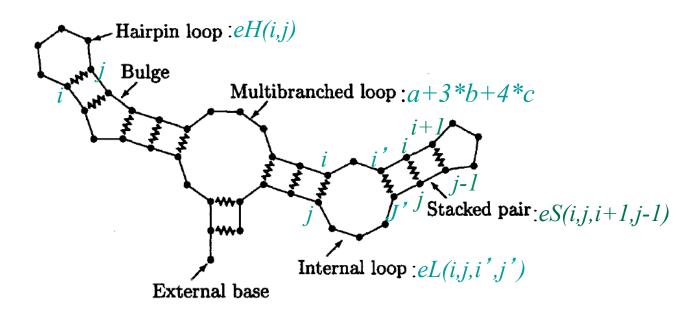
i	į	1	2	3	4	5	6
J	2	0					
	3	1	1				
	4	1	1	0			
	5	2	2	1	1		
	6	3	2	1	1	0	



Zuker-Sankoff minimum energy model

- Stacks (contiguous nested base pairs) are the dominant stabilizing force – contribute the negative energy
- Unpaired bases form loops contribute the positive energy.
 - Hairpin loops, bulge/internal loops, and multiloops.
- Zuker-Sankoff minimum energy model. [Zuker and Sankoff (1984), Sankoff (1985)]
- Mfold and ViennaRNA are all based on this model.
 (this model is also called mfold model)

Zuker-Sankoff minimum energy model



[Lyngsø (1999)]

RNA minimum energy problem

Problem 2: [RNA minimum energy problem]

Given an RNA sequence s[1...n], and four energy functions eH, eS, eL, a, b, and c, determine a noncrossing secondary structure for this RNA sequence such that the sum of the energy over all the loops and stacks in the secondary structure is minimized.

- This problem can be solved by a dynamic programming algorithm in $O(n^4)$ time.
- Lyngsø et al. (1999) revise the energy function for internal loop, proposed an $O(n^3)$ time solution.

Zuker-Sankoff model

- **Hairpin loop:** eH(i, j) is the energy of the hairpin loop from i + 1 to j 1, which *closed* by base pair (i, j).
- Stacked base pairs: eS(i, j, i + 1, j 1) is the energy of the stacking base pairs (i, j) and (i + 1, j 1).
- Bulge and internal loop: eL(i, j, i', j') is the energy of the the bulge or internal loop starting from i+1 to i'-1 and from j'+1 to j'-1 which is *closed* by base pairs (i, j) and (i', j').
- Multi-loop: a is the energy of generating a multi-loop, b is the energy of one base pair that closes the multi-loop, and c is the energy of one unpaired base in the multi-loop.

Recursive functions

- W(i) holds the minimum energy of a structure on s[1...i].
- V(i,j) holds the minimum energy of a structure on s[i...j] with s[i] and s[j] forming a basepair.
- WM(i,j) holds the minimum energy of a structure on s[i...j] that is part of multiloop.

```
W(i) = \min\{W(i-1), \\ \min_{0 \le k < i} \{W(k) + V(k+1, i)\}\}.
```

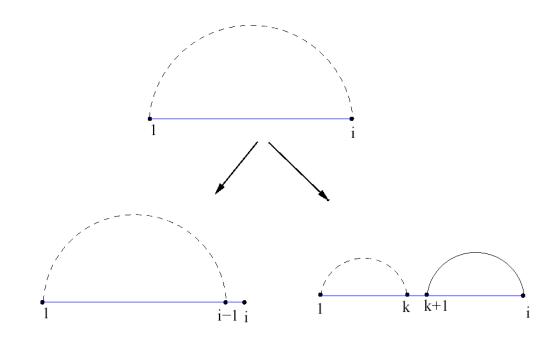
```
\begin{split} V(i,j) &= \min\{eH(i,j), \\ eS(i,j,i+1,j-1) + V(i+1,j-1), \\ &\underset{i < i' < j' < j \text{ and } i'-i+j-j' > 2}{\min} \{eL(i,j,i',j') + V(i',j')\}, \\ &\underset{i+1 < k < j}{\min} \{WM(i+1,k-1) + WM(k,j-1) + a\}\}, \end{split}
```

$$\begin{split} WM(i,j) &= \min\{V(i,j) + b, \\ WM(i,j-1) + c, \\ WM(i+1,j) + c, \\ &\min_{i < k \le j} \{WM(i,k-1) + WM(k,j)\}\}, \end{split}$$

Recursive functions (Zuker)

- W(i) holds the minimum energy of a structure on s[1...i].
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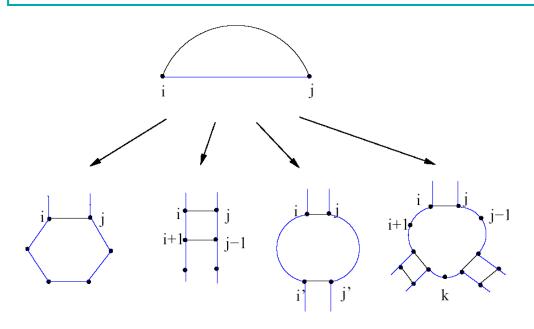
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A recursive solution

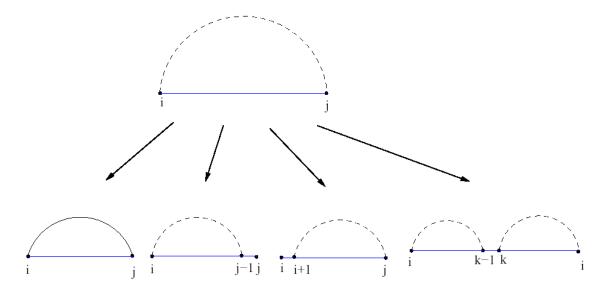
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$$\begin{split} V(i,j) &= \min\{eH(i,j), \\ &eS(i,j,i+1,j-1) + V(i+1,j-1), \\ &\underset{i < i' < j' < j \text{ and } i'-i+j-j' > 2}{\min} \{eL(i,j,i',j') + V(i',j')\}, \\ &\underset{i+1 < k < j}{\min} \{WM(i+1,k-1) + WM(k,j-1) + a\}\}, \end{split}$$



A recursive solution

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Prediction with pseudoknots

- Base pair maximization allowing crossing pairs can be solved in polynomial time.
- leong et al. (2003) proved that base pairing maximization problem allowing crossing pairs in a *planar* secondary structure is NP-hard.



Prediction with pseudoknots

- Prediction allowing generalized pseudoknots with energy functions depending on adjacent basepairs is NP-hard.
 - Akutsu (2000) (longest common subsequence for multiple sequences (LCS)).
 - Lyngsø and Pedersen (2000) (3SAT).
 - similar to Zuker-Sankoff minimum energy model.
- Pseudoknots in structure-known RNAs.
 - Biologists are not interested in the approximation solutions.
 - Most pseudoknots are planar.
 - Not too many variations.
- Rivas and Eddy (1999) presented a $O(n^6)$ solution allowing most types of pseudoknots in known ncRNAs.