

Applications of Lightweight Stochastic Context Free Grammars for RNA Analysis

BMI/CS 776

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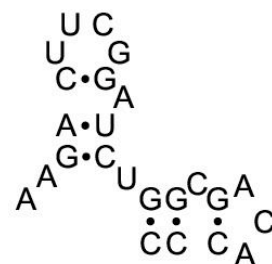
Searching Sequence for a Secondary Structure

Given

- a single RNA sequence with its secondary structure
- another RNA query sequence

ACGGCUUCGGCCUGGCGAGACCC

Determine if the query sequence has
“same” secondary structure



- this is analogous to pairwise alignment with primary sequences
- we take into account substitutions, insertions/deletions, and base-pair substitutions

The RIBOSUM Matrices [Klein & Eddy]

observed frequency of i aligned to j
in homologous RNAs

$$s_{ij} = \log_2 \frac{f_{ij}}{\dots}$$

background frequency of i

$$\xrightarrow{\quad} g_i g_j$$

AA	-2.49									A	2.22								
AC	-7.04	-2.11								C	-1.86	1.16							
AG	-8.24	-8.89	-0.80							G	-1.46	-2.48	1.03						
AU	-4.32	-2.04	-5.13	4.49						U	-1.39	-1.05	-1.74	1.65					
CA	-8.84	-9.37	-10.41	-5.56	-5.13														
CC	-14.37	-9.08	-14.53	-6.71	-10.45	-3.59													
CG	-4.68	-5.86	-4.57	1.67	-3.57	-5.71	5.36												
CU	-12.64	-10.45	-10.14	-5.17	-8.49	-5.77	-4.96	-2.28											
GA	-6.86	-9.73	-8.61	-5.33	-7.98	-12.43	-6.00	-7.71	-1.05										
GC	-5.03	-3.81	-5.77	2.70	-5.95	-3.70	2.11	-5.84	-4.88	5.62									
GG	-8.39	-11.05	-5.38	-5.61	-11.36	-12.58	-4.66	-13.69	-8.67	-4.13	-1.98								
GU	-5.84	-4.72	-6.60	0.59	-7.93	-7.88	-0.27	-5.61	-6.10	1.21	-5.77	3.47							
UA	-4.01	-5.33	-5.43	1.61	-2.42	-6.88	2.75	-4.72	-5.85	1.60	-5.75	-0.57	4.97						
UC	-11.32	-8.67	-8.87	-4.81	-7.08	-7.40	-4.91	-3.83	-6.63	-4.49	-12.01	-5.30	-2.98	-3.21					
UG	-6.16	-6.93	-5.94	-0.51	-5.63	-8.41	1.32	-7.36	-7.55	-0.08	-4.27	-2.09	1.14	-4.76	3.36				
UU	-9.05	-7.83	-11.07	-2.98	-8.39	-5.41	-3.67	-5.21	-11.54	-3.90	-10.79	-4.45	-3.39	-5.97	-4.28	-0.02			
	AA	AC	AG	AU	CA	CC	CG	CU	GA	GC	GG	GU	UA	UC	UG	UU			

$$s'_{ijkl} = \log_2 \frac{f'_{ijkl}}{g_i g_j g_k g_l}$$

observed frequency of two base pairs
 i - j and k - l aligned to each other in
homologous RNAs

Using a Lightweight SCFG to Search for Secondary Structure

given a structure



can construct a simple grammar characterizing it

$$\begin{aligned} s &\rightarrow C s_1 G \\ s_1 &\rightarrow A s_2 U \\ s_2 &\rightarrow b_1 l_1 \\ l_1 &\rightarrow b_2 b_3 \\ b_1 &\rightarrow U \\ b_2 &\rightarrow U \\ b_3 &\rightarrow C \end{aligned}$$

can add productions to allow for variation

$$\begin{aligned} s &\rightarrow U s_1 A \\ s &\rightarrow A s_1 U \\ s &\rightarrow G s_1 C \end{aligned}$$

base pair substitutions

$$s_1 \rightarrow s_1 A$$

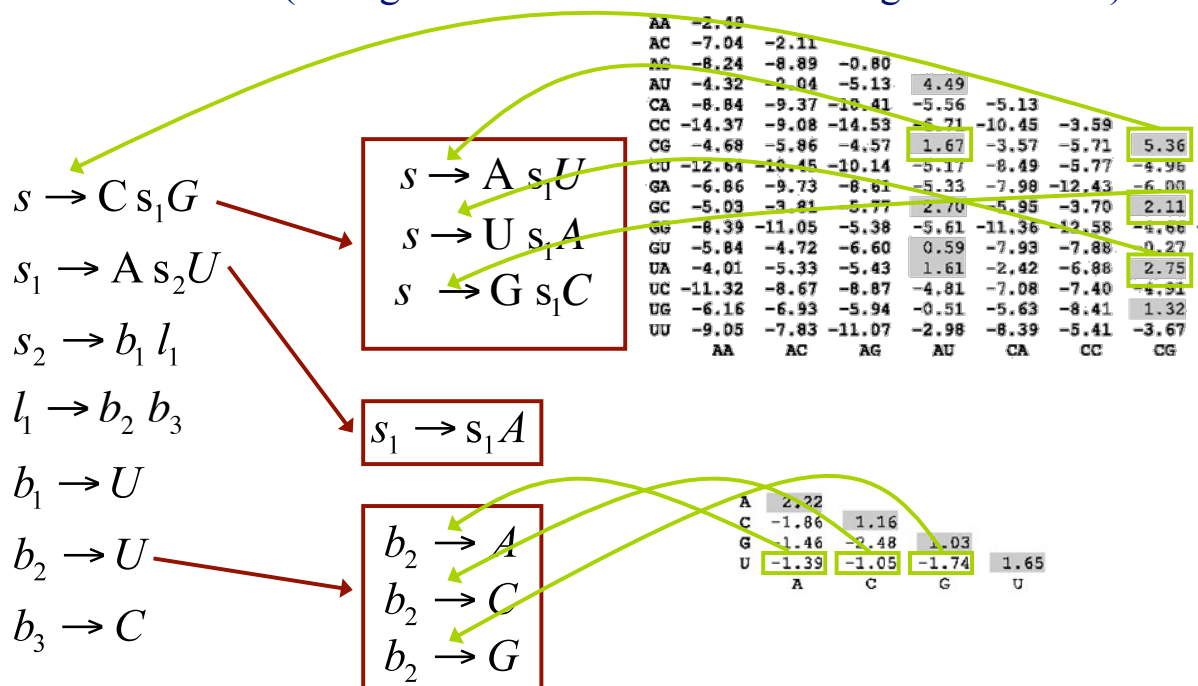
insertions

$$\begin{aligned} b_2 &\rightarrow A \\ b_2 &\rightarrow C \\ b_2 &\rightarrow G \end{aligned}$$

single base substitutions

Setting the Parameters in the Grammar

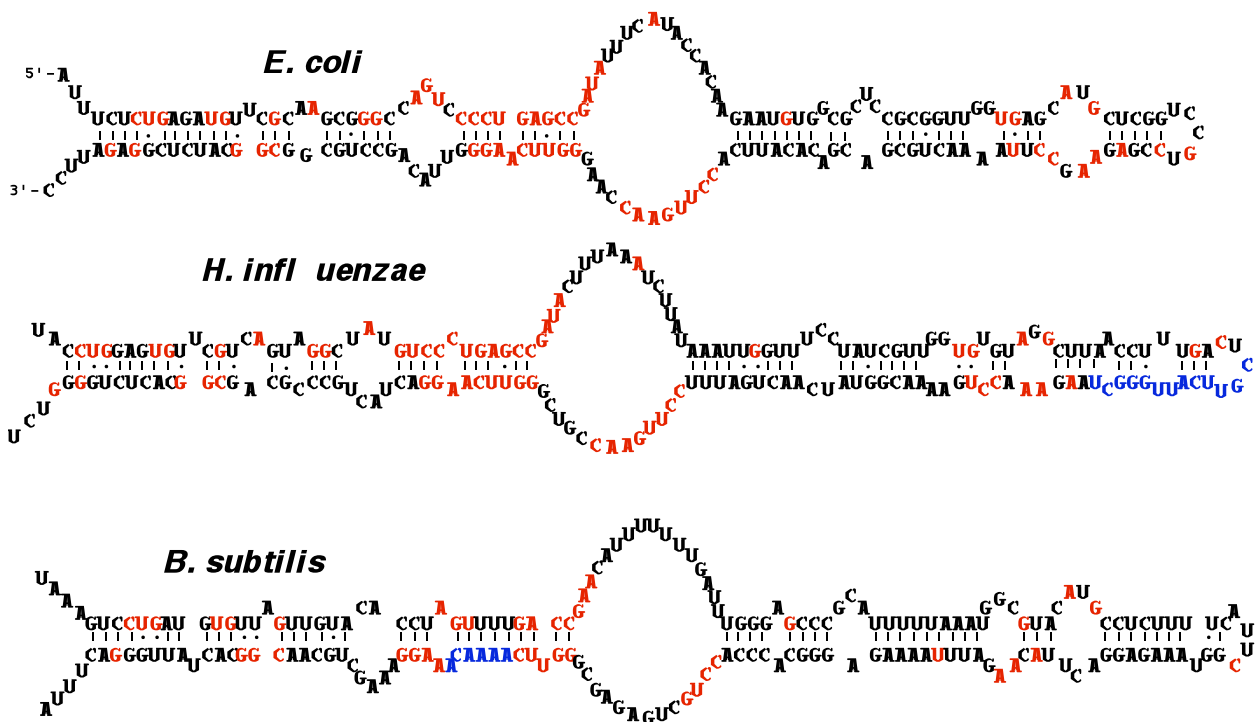
- Infer them from the parameters from the RIBOSUM matrices (taking into account the latter are log-odds scores)



RSEARCH: Searching Sequence for a Secondary Structure

- the RSEARCH algorithm [Klein & Eddy, *BMC Bioinformatics* 2003] implements this idea
- but uses a somewhat different SCFG formulation – covariance models (see section 10.3 in Durbin et al.)

6S RNA Secondary Structure



An RSEARCH Case Study

- finding 6S genes in bacterial genomes
 - we used E. coli 6S as the query structure
 - searched 14 other genomes with known 6S genes
 - ~ 5,000 intergenic sequences on average
 - the top-scoring RSEARCH hit in all 14 genomes was the known 6S gene

RNA Gene Detection

[Rivas & Eddy, BMC Bioinformatics 2001]

Given: a pair of putatively homologous sequences

Identify novel RNA genes in the sequences

```
TAGTCATGCAGTCAGCTATCATCAGCATCGATCGATCGACTAGCTACCTACGACTAGGACTAGCTACGTACGACTAGGACTAGCTACGTACGAACTGACTGACTAGGGGGGATATTCCTCTGGGCCCTCATCTACTGAGCTATCATCATCGTACTA  
TCAAACCTGACGTACTAGCTAGTCATGCAGTCAGCTATCATCAGCATCGATAACCTGACGTACTAGCTAGTCATGCAGTCAGCTATCATCAGCATCGATGCTATCATCAGCATCGATCCTATCGACTAGCTACGTACGACTAGGACTAGCTACGTACGA
```

RNA Gene Detection

key idea: the pattern of substitutions in the two sequences provides evidence about the role of the sequence

position-independent

```

| | | | | | | | | | | | | |
G T T A A C T G A G T A A C G
| x x | x | | | | | x | |
G C A A G C T G A G T T A C G
  
```

$P(G-G)*P(T-C)*P(T-A)...$

substitutions tend to be in the 3rd codon (wobble) position

coding

```

      G      Q      K      V      L
    [ ] [ ] [ ] [ ] [ ]
G G T C A G A A G T A C T T
| | | | | | | | | | | |
G G A C A G A A G G T T C T C
  
```

$P(GGT-GGA)*P(CAG-CAG)*...$

substitutions tend to preserve complementary base pairings

structural RNA

```

| | | | | | | | | | | |
T T G T T C G A A G A A C G
| | | x x | | | | | x x |
T T G A C C G A A G G T C G
  
```

$P(T-T)*P(T-T)*P(GC-GC)*P(TA-AT)*...$

Figure from Rivas & Eddy, *BMC Bioinformatics*, 2001

RNA Gene Detection

- illustrative examples of emission scores for three models (numbers before parens are log-odds with respect to a model of no alignment)

OTH	$P^{OTH}\left(\begin{smallmatrix} G \\ G \end{smallmatrix}\right)$ +0.76(-3.20)	$P^{OTH}\left(\begin{smallmatrix} C \\ C \end{smallmatrix}\right)$ +0.72(-3.52)	$P^{OTH}\left(\begin{smallmatrix} U \\ C \end{smallmatrix}\right)$ -0.19(-4.41)	$P^{OTH}\left(\begin{smallmatrix} A \\ U \end{smallmatrix}\right)$ -0.53(-4.45)
COD	$P^{COD}\left(\begin{smallmatrix} A & A & C \\ & & \\ A & A & C \end{smallmatrix}\right)$ +3.31(-8.19)	$P^{COD}\left(\begin{smallmatrix} A & A & C \\ & & \\ A & A & U \end{smallmatrix}\right)$ +3.31(-8.19)	$P^{COD}\left(\begin{smallmatrix} A & A & C \\ & & \\ A & U & C \end{smallmatrix}\right)$ -0.52(-12.31)	$P^{COD}\left(\begin{smallmatrix} U & C & U \\ & & \\ A & G & C \end{smallmatrix}\right)$ +1.29(-10.95)
RNA	$P^{RNA}\left(\begin{smallmatrix} G \cdots C \\ \quad \\ G \cdots C \end{smallmatrix}\right)$ +3.81(-4.37)	$P^{RNA}\left(\begin{smallmatrix} G \cdots U \\ \quad \\ G \cdots C \end{smallmatrix}\right)$ +1.36(-6.82)	$P^{RNA}\left(\begin{smallmatrix} G \cdots A \\ \quad \\ G \cdots A \end{smallmatrix}\right)$ -8.82(-16.42)	$P^{RNA}\left(\begin{smallmatrix} G \cdots C \\ \quad \\ C \cdots G \end{smallmatrix}\right)$ +2.43(-5.76)

both code for Asn

both code for Ser

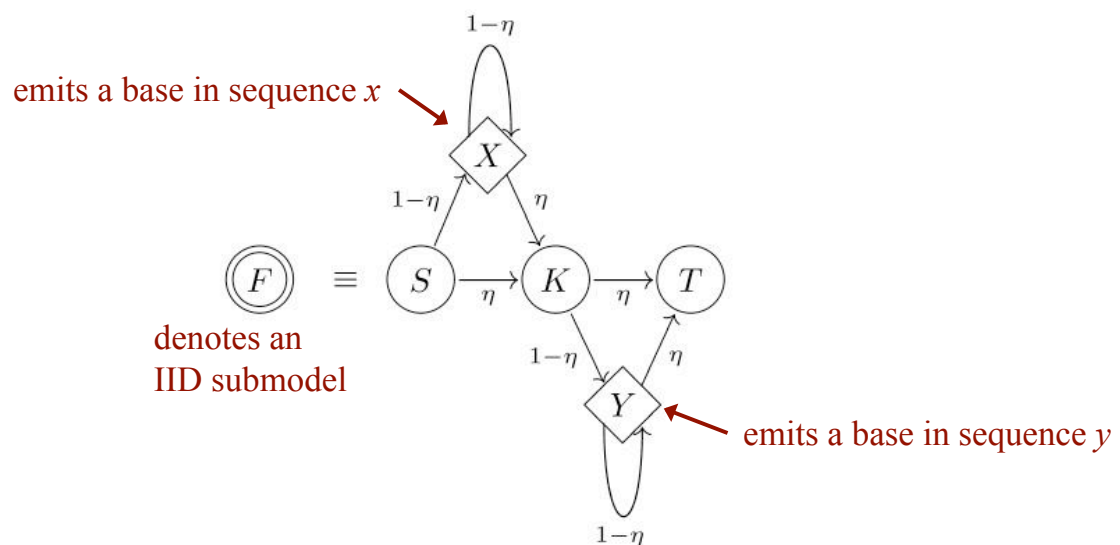
Figure from Rivas & Eddy, *BMC Bioinformatics*, 2001

RNA Gene Detection via Comparative Sequence Analysis

- given sequences x and y , want a model that can distinguish
 - homologous RNA subsequences
 - homologous coding subsequences
 - “other” homologous subsequences
 - non-homologous subsequences
- allow these to be interleaved, have gaps

RNA Gene Detection: The IID Model

- models non-homologous sequences, x and y



- S , K and T are silent states

RNA Gene Detection: The “Other” Homologous Sequence Model

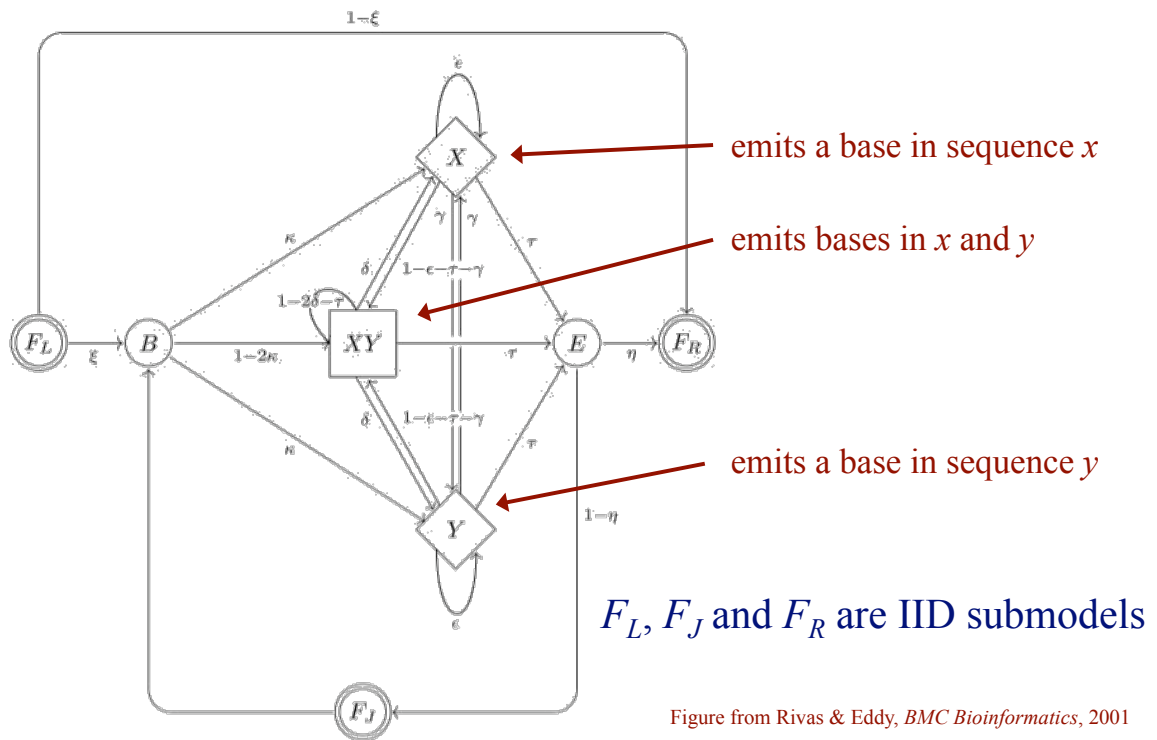


Figure from Rivas & Eddy, *BMC Bioinformatics*, 2001

RNA Gene Detection: The Coding Sequence Model

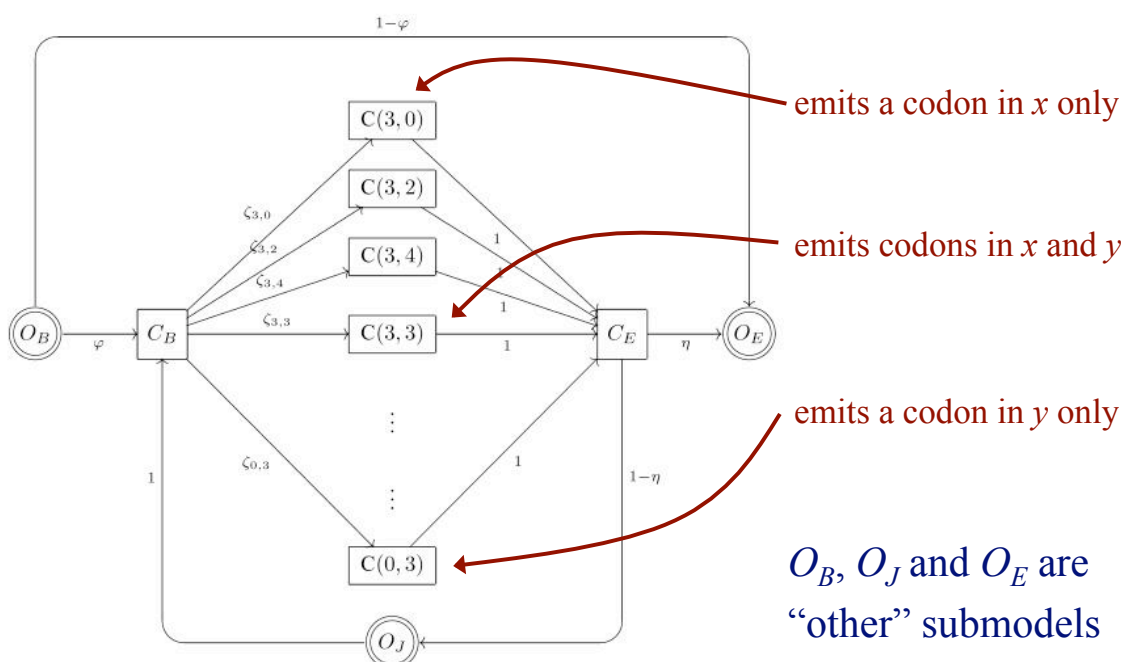
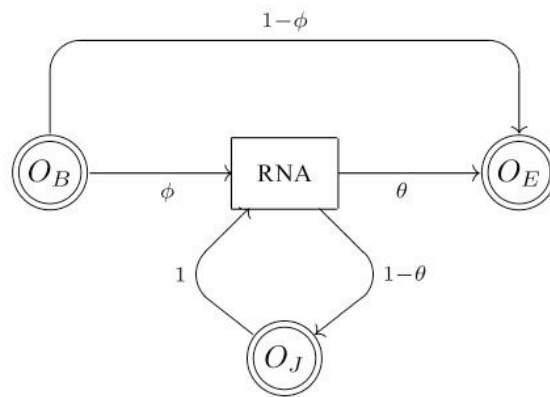


Figure from Rivas & Eddy, *BMC Bioinformatics*, 2001

RNA Gene Detection: The RNA Model



O_B , O_J and O_E are
“other” submodels

Figure from Rivas & Eddy, *BMC Bioinformatics*, 2001

- here, the RNA box is a “lightweight” pairwise SCFG

Summary of RNA Analysis Tasks

- given a sequence, predict its secondary structure
- given a set of related RNA sequences, construct a model of the set
 - parameter learning (Inside-Outside)
 - structure refinement
- given a model of an RNA class, find sequences that belong to the class (Inside or CYK)
- given a sequence/structure, find other sequences with similar structure
- given a pair of related genomic sequences, find subsequences that seem have similar secondary structure (RNA gene finding)