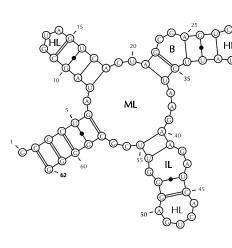
RNA-RNA Interaction Prediction with Stochastic Grammars

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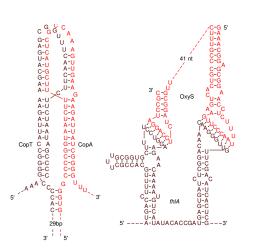
5. October 201210. Herbstseminar der Bioinformatik

RNA secondary structure: model



- primary structure: word over $\{a, c, g, u\}$
- secondary structure: parentheses word {(, |,)}

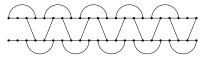
RNA also interacts!



- bacterial antisense RNA
- interact non-trivally, "knotted" structure
- goal: predict whole interaction structure, not only interaction sites

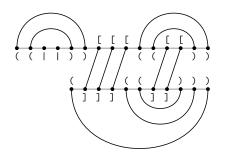
RNA-RNA interaction problem (RIP)

- two RNA molecules interact, i. e. form a *joint* secondary structure
- predict the "best" one possible
- RIP in general \mathcal{NP} -complete \rightsquigarrow restrict structure
 - Exclude pseudo knots (internal & external)
 - 2 Exclude Zig-Zags:



Example joint secondary structure

- → joint structure can be encoded as "2D-word":
 - pair of upper and lower word
 - matching () → internal bond
 - matching [] → external bond (between two molecules)
 - I → unpaired base



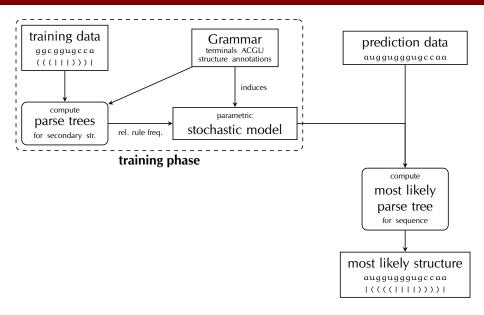
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Stochastic CFG

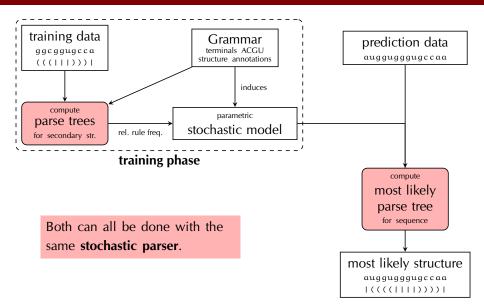
Stochastic Context-free Grammars

- $G = (N, \Sigma, R, S, P)$
 - nonterminals N
 - terminal alphabet Σ
 - rule set $R \subset N \times (N \cup \Sigma)^*$
 - start nonterminal $S \in N$
 - $P: R \rightarrow [0, 1]$: rule probabilities
- Probability of a derivation tree:
 Product of used rules' probabilities
- For structure prediction:
 - $\Sigma = \{|b, (b, b)| : b \in \{a, c, g, u\}\}$
 - unambiguous w. r. t. structure

Structure Prediction with Formal Grammars



Structure Prediction with Formal Grammars



Stochastic CFG

Stochastic Context-free Grammars

- $G = (N, \Sigma, R, S, P)$
 - nonterminals N
 - terminal alphabet Σ
 - rule set $R \subset N \times (N \cup \Sigma)^*$
 - start nonterminal $S \in N$
 - $P: R \rightarrow [0, 1]$: rule probabilities

2-dimensional CFG

2-dimensional Context-free Grammars

- $G = (N, \Sigma, R, S, P)$
 - nonterminals N
 - terminal alphabet Σ
 - rule set $R \subset N \times (N \cup (\Sigma^*)^2)^*$
 - start nonterminal $S \in N$
 - $P: R \rightarrow [0, 1]$: rule probabilities

$$s \to (\begin{array}{c} \begin{array}{c} \begin{array}{c}$$

Example:

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Example:

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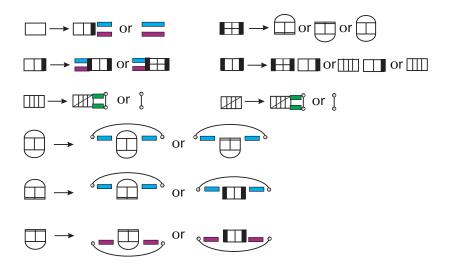
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Used Grammar



Implementation

- stochastic parser, independent of grammar
- mixture of Earley-Parser and dynamic programming
- fast manual implementation in C++

RNA pair	n	m	runtime	memory
DIS DIS	35	35	2 min	300 MB
CopA CopT	56	57	1 h	2 GB
ompA MicA	137	72	2 d	18 GB
U2 and U6 snRNAs in yeast 21	144	95	1 week	34 GB

Summary

This Talk:

- 2D-CFGs give stochastic model for RNA joint structures (only slight extension of CFGs needed)
- Earley parsing can be used to train model and compute most likely structures
- efficient implementation available

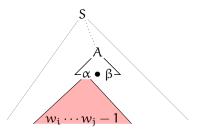
Open Problems:

- get good training data
- full empirical evaluation of prediction quality

Earley-Parsing

- does not need normal form
- here: as formal calculus
- defined in terms of **items** $(i \ j, A \rightarrow \alpha \bullet \beta)$ can be derived iff

$$S \Rightarrow^* w_{1,i-1} A \gamma \Rightarrow w_{1,i-1} \alpha \beta \gamma \Rightarrow^* w_{1,j-1} \beta \gamma$$



Earley-Parser for SCFGs

- Items: $(i \ j, A \to \alpha \bullet \beta)$ derivable **iff** $S' \Rightarrow^* w_{1,i-1}A\gamma \Rightarrow w_{1,i-1}\alpha\beta\gamma \Rightarrow^* w_{1,j-1}\beta\gamma$
- \bullet Start-Item: (1 1, S' \rightarrow \bullet S) , Goal-Item (1 n + 1, S' \rightarrow S \bullet)
- Derivation Rules:

• Scanner
$$\frac{ \left(\text{i } \text{j} - 1 \text{, } A \rightarrow \alpha \bullet w_{j-1}\beta \right) }{ \left(\text{i } \text{j } \text{, } A \rightarrow \alpha w_{j-1} \bullet \beta \right) }$$

• Predictor:

$$\frac{\left(\mathfrak{i}\ \mathfrak{j},\,\mathsf{A}\to\alpha\bullet\mathsf{B}\gamma\right)}{\left(\mathfrak{j}\ \mathfrak{j},\,\mathsf{B}\to\bullet\,\boldsymbol{\beta}\right)}$$

Completer:

$$\frac{(i r, A \to \alpha \bullet B\gamma) \quad (r j, B \to \beta \bullet)}{(i j, A \to \alpha B \bullet \gamma)}$$

Earley-Parser for 2D-CFGs

- Items: $\begin{pmatrix} \mathbf{i} & \mathbf{j} \\ \mathbf{k} & \mathbf{l} \end{pmatrix}$, $A \to \begin{pmatrix} \alpha_1 \bullet \beta_1 \\ \alpha_2 \bullet \beta_2 \end{pmatrix}$
 - $\text{derivable iff } S' \Rightarrow^* \ ^{\mathfrak{u}_{1,i-1}}_{\mathfrak{v}_{1,k-1}} A \gamma \Rightarrow \ ^{\mathfrak{u}_{1,i-1}}_{\mathfrak{v}_{1,k-1}} \ \alpha \beta \gamma \Rightarrow^* \ ^{\mathfrak{u}_{1,j-1}}_{\mathfrak{v}_{1,l-1}} \ \beta \gamma$
- $\bullet \text{ Start-Item: } \left(\begin{smallmatrix} 1 & 1 \\ 1 & 1 \end{smallmatrix}, S' \to \begin{smallmatrix} \bullet S_1 \\ \bullet S_2 \end{smallmatrix}\right), \text{ Goal-Item } \left(\begin{smallmatrix} 1 & n+1 \\ 1 & m+1 \end{smallmatrix}, S' \to \begin{smallmatrix} S_1 \bullet \\ S_2 \bullet \end{smallmatrix}\right)$
- Derivation Rules:
 - Scanner (upper) (lower similar):

$$\frac{\begin{pmatrix} \mathfrak{i} & \mathfrak{j}-1 \\ k & l \end{pmatrix}, A \to \begin{pmatrix} \alpha_1 \bullet \mathfrak{u}_{\mathfrak{j}-1} \beta_1 \\ \alpha_2 \bullet & \beta_2 \end{pmatrix}}{\begin{pmatrix} \mathfrak{i} & \mathfrak{j} \\ k & l \end{pmatrix}, A \to \begin{pmatrix} \alpha_1 \mathfrak{u}_{\mathfrak{j}-1} \bullet \beta_1 \\ \alpha_2 & \bullet \beta_2 \end{pmatrix}}$$

$$\frac{\begin{pmatrix} i & j \\ k & l \end{pmatrix}, A \to \begin{pmatrix} \alpha_1 \bullet B_1 \gamma_1 \\ \alpha_2 \bullet B_2 \gamma_2 \end{pmatrix}}{\begin{pmatrix} j & j \\ l & l \end{pmatrix}, B \to \begin{pmatrix} \bullet \beta_1 \\ \bullet \beta_2 \end{pmatrix}}$$

$$\frac{\begin{pmatrix} \mathfrak{i} & \mathfrak{r} \\ k & s \end{pmatrix}, A \to \begin{pmatrix} \alpha_1 \bullet B_1 \gamma_1 \\ \alpha_2 \bullet B_2 \gamma_2 \end{pmatrix} \begin{pmatrix} \mathfrak{r} & \mathfrak{j} \\ s & \mathfrak{l} \end{pmatrix}, B \to \begin{pmatrix} \beta_1 \bullet \\ \beta_2 \bullet \end{pmatrix}}{\begin{pmatrix} \mathfrak{i} & \mathfrak{j} \\ k & \mathfrak{l} \end{pmatrix}, A \to \begin{pmatrix} \alpha_1 B_1 \bullet \gamma_1 \\ \alpha_2 B_2 \bullet \gamma_2 \end{pmatrix}}$$

Implementation Note: Dense Grammars

- Number of items: $\Theta(\frac{1}{4}n^2m^2)$ for n and m lengths of two RNA sequences
- structure prediction grammars are dense: most items are derivable
- ~> compute values for all items à la dynamic programming