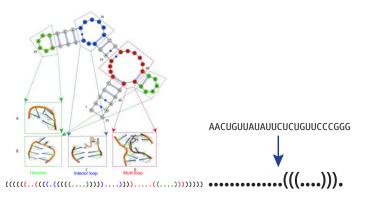
Learning the RNA inverse folding problem.

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Structure Prediction a.k.a. Folding



- ▶ Input: RNA sequence $x_1, x_2, ..., x_N$ where $x_i \in \{A, U, C, G\}$
- ▶ Output: RNA 2D structure $\omega = \{(i, j), ...\}$ where i and j are indices in the sequence x that are paired.
- ► Solution: Dynamic Programming + physics based rules.
- ▶ Current approaches achieve high accuracy in $\mathcal{O}(n^3)$ time for sequences of length < 200.



Structure Design a.k.a. Inverse Folding



- Input: 2D Structure
- ▶ Output: sequence whose minimum energy fold corresponds to the input.
- Key problem for synthetic biology and drug design.
- Current solutions: local search strategies
- ▶ No linear time algorithms exist to solve this problem.

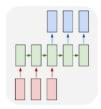
Data

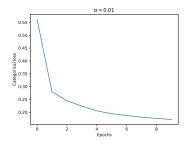
Two types of sequence-structure data:

- Real world
 - ▶ Rfam databse (hand curated sequences for structural families): $\mathcal{O}(100K 1M)$ sequences per family
- Artificial
 - ▶ Local search software for design: unlimited data size

Approach: One model \rightarrow one structure

- Generate sequences belonging to one of 5 Rfam structural families.
- Given set of member sequences, generate novel likely members.
- ► Model: RNN + LSTM
- Evaluation: Covariance models, GC content, base pair distance, discriminator NN.

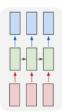






Approach: One model \rightarrow all structures

- ► Goal: generate a likely sequence for a given structure.
- ▶ Input: vector representations $s_i \in \{0,1\}^{|\omega|}$ for each index i in structure ω where



$$s_i[j] = \begin{cases} 1 & i \text{ paired with } j \\ 0 & \text{else} \end{cases}$$

- ▶ Output: 1 of 4 encoding of the nucleotide in $\{A, U, C, G\}$ belonging to position i in structure.
- Approach: sequence to sequence RNN, LSTM. Recursive NN.
- Long term goal