

RNA secondary structure prediction with *deep neural networks*

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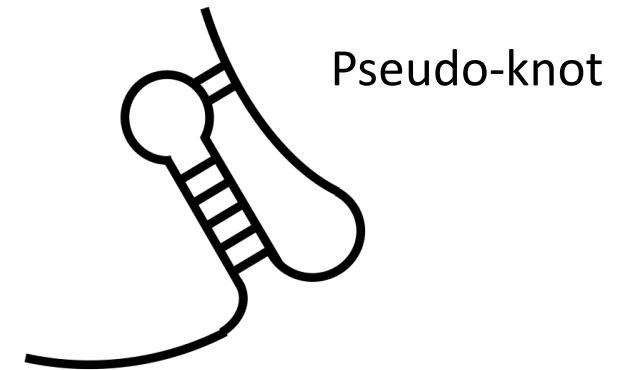
Presenter: Zudi Lin

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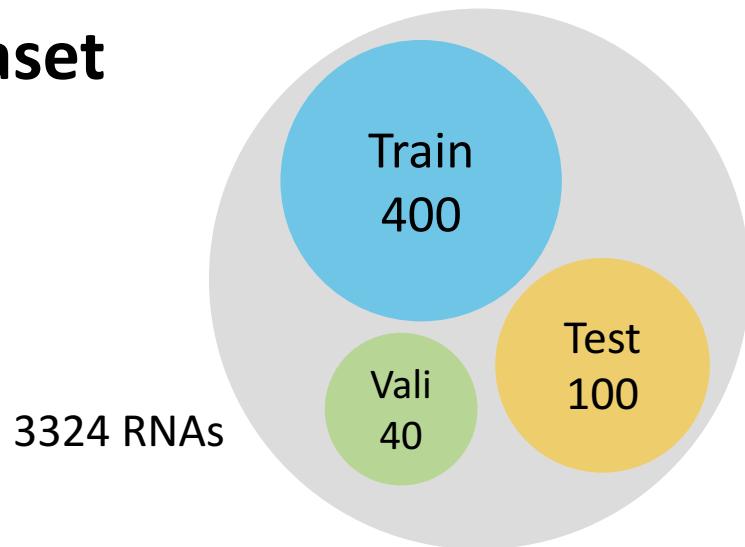
Background

▪ Dominant structure prediction methods

- 1) Based on thermodynamic parameters
- 2) Nearest Neighbor Assumption
- 3) Dynamic Programming



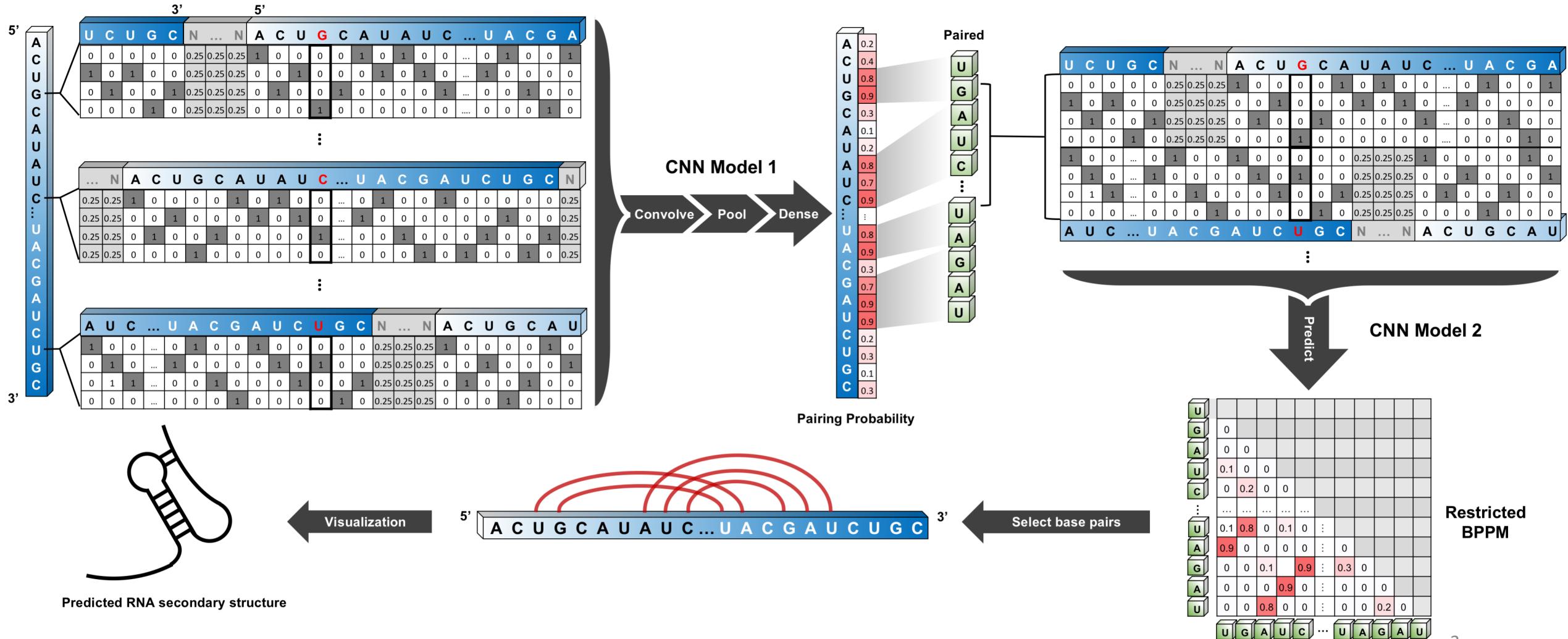
▪ Dataset



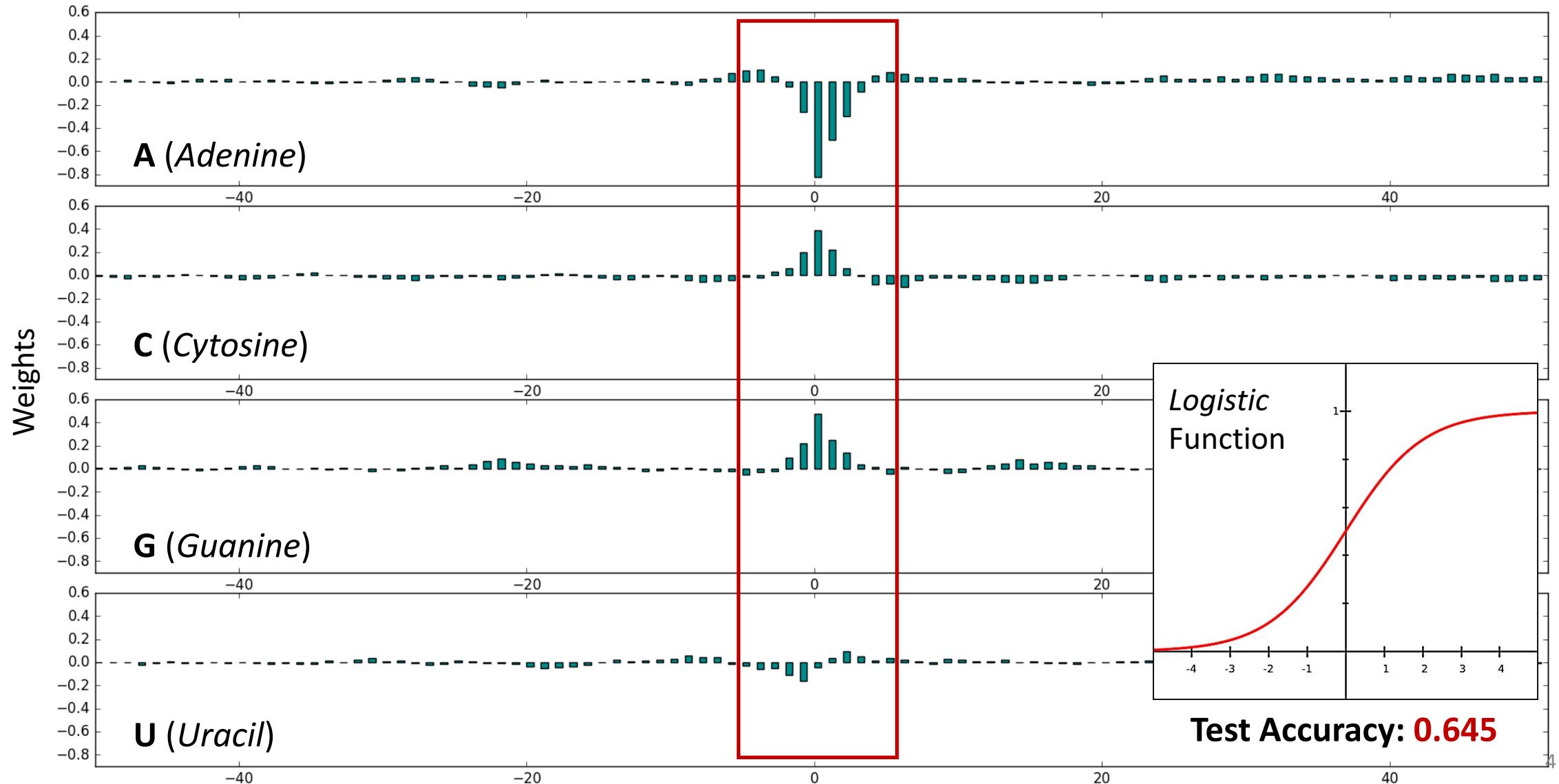
[1] Totally 3324 different RNA sequences were collected, including tRNA, rRNA, telomerase RNA, RNase etc..

[2] Filtered with *EMBOSS Needle* global alignment tool. Each pair of sequences have a similarity < **60%**.

Prediction Pipeline

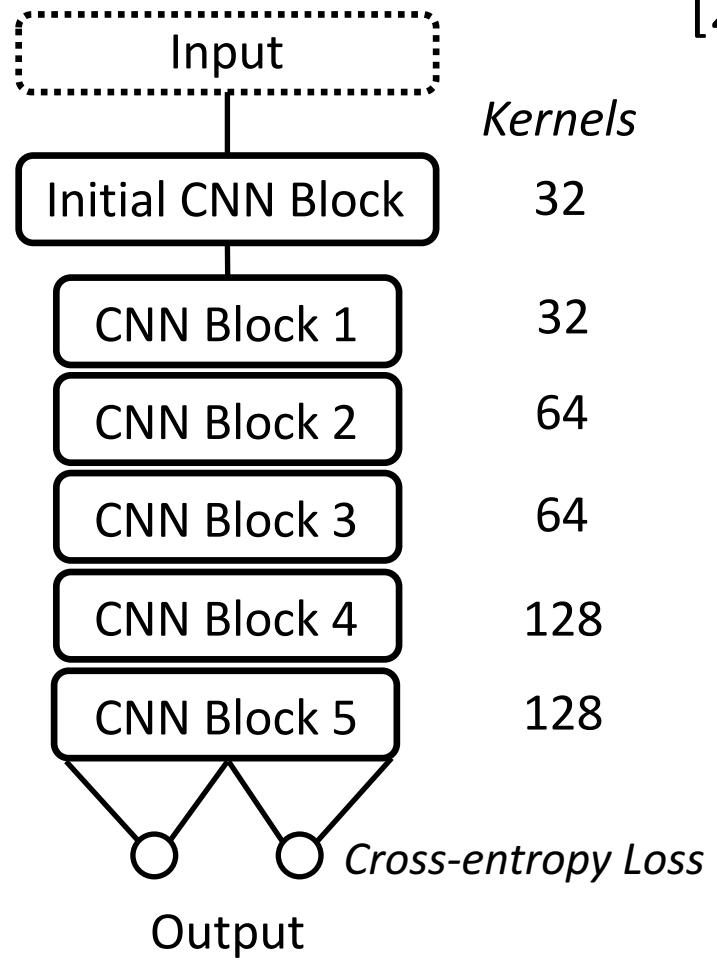
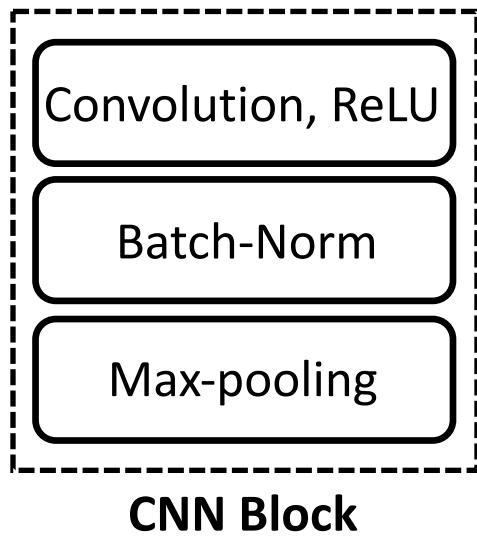


Logistic regression is *not* effective

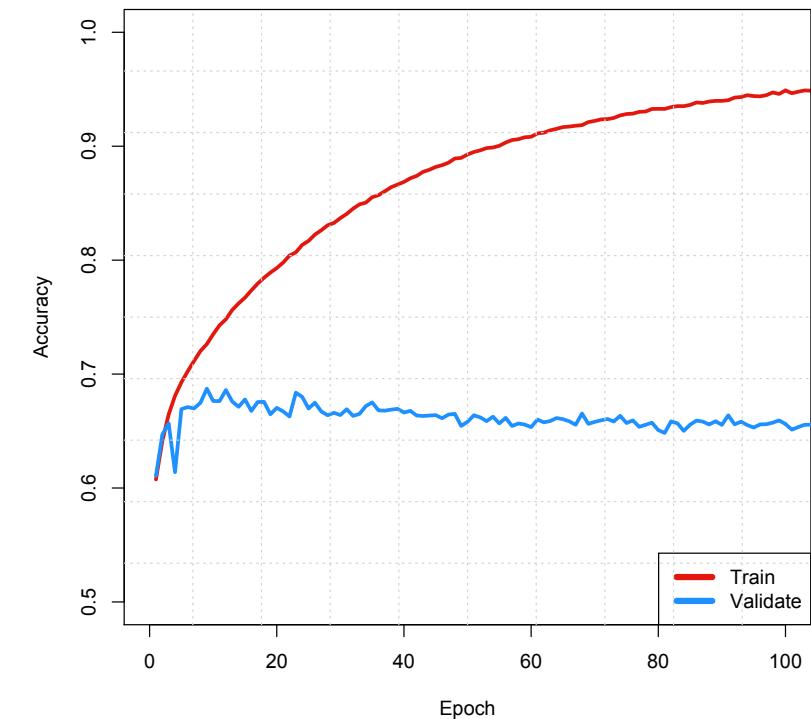


Network Architecture & Training Curve

[1]



[2]



Training & Validation Accuracy

Solving the Over-fitting Problem

[1]

3'					5'				
U C U G C					N ... N A C U G C A U A U C ... U A C G A				
0	0	0	0	0	0	0	0	1	0
1	0	1	0	0	0	0	0	0	1
0	1	0	0	1	0	0	0	1	0
0	0	0	1	0	0	0	0	1	0



3'					5'				
U C U G C					N ... N A C U G C A U A U C ... U A C G A				
0	0	0	0	0	0	0	0	1	0
1	0	1	0	0	0	0	0	0	1
0	1	0	0	1	0	0	0	1	0
0	0	0	1	0	0	0	0	1	0
0	0	0	0	0	0	0	0.25	0.5	1
0	0	0	0	1	0	0	0.25	0.5	0

Target nucleotide
Pairing Partners

Circulative Representation

[2]

$$C = C_0 + \frac{\lambda}{2n} \sum_w w^2$$

L2 Regularization

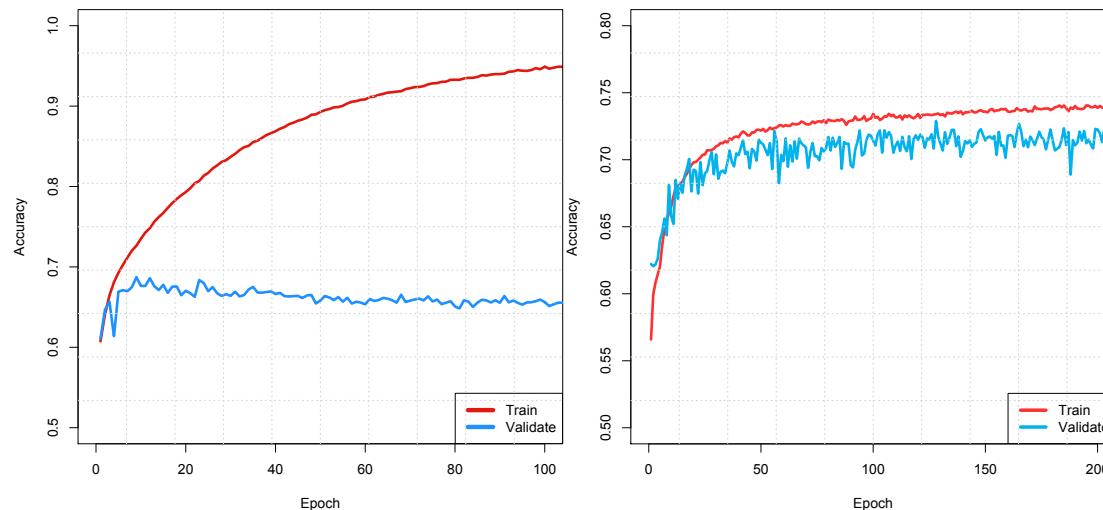
$$w \rightarrow \left(1 - \frac{\eta\lambda}{n}\right)w - \frac{\eta}{m} \sum_x \frac{\partial C_x}{\partial w}$$

$$b \rightarrow b - \frac{\eta}{m} \sum_x \frac{\partial C_x}{\partial b},$$

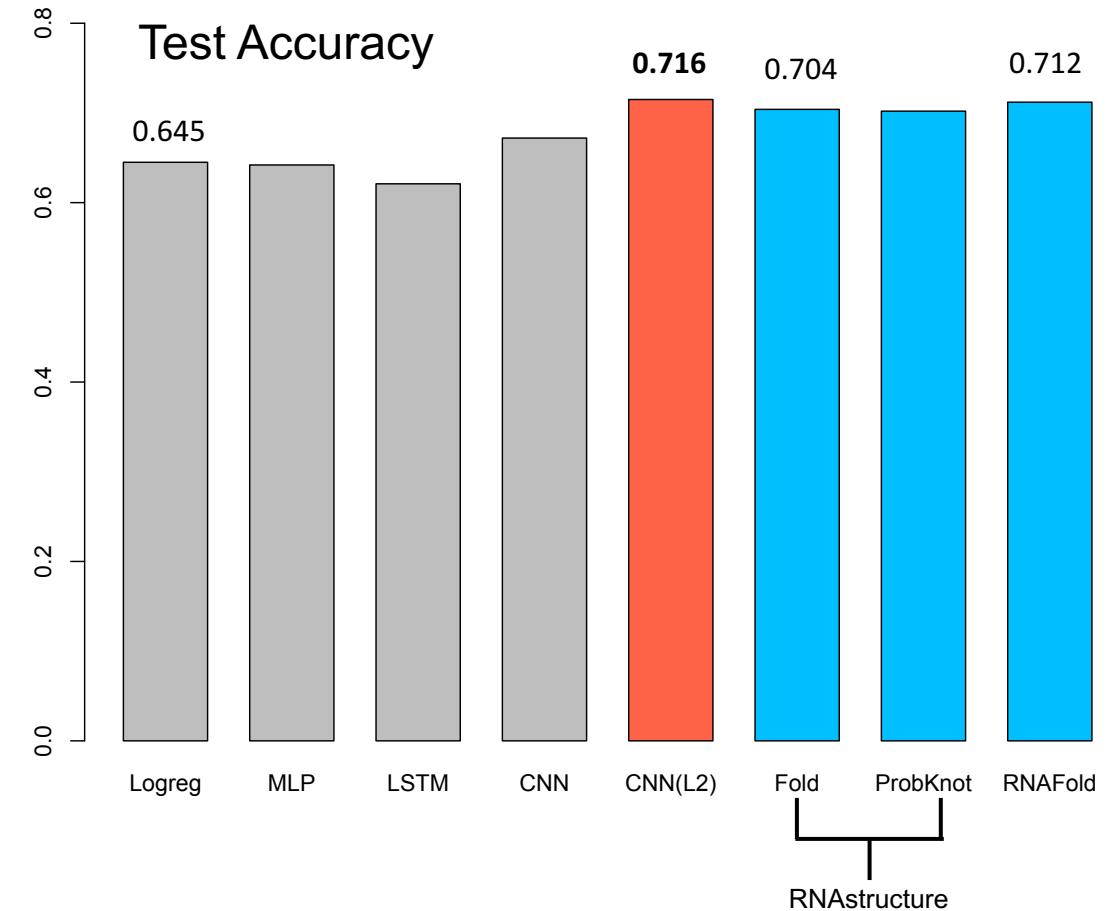
Stochastic Gradient Descent

One-dimensional prediction results

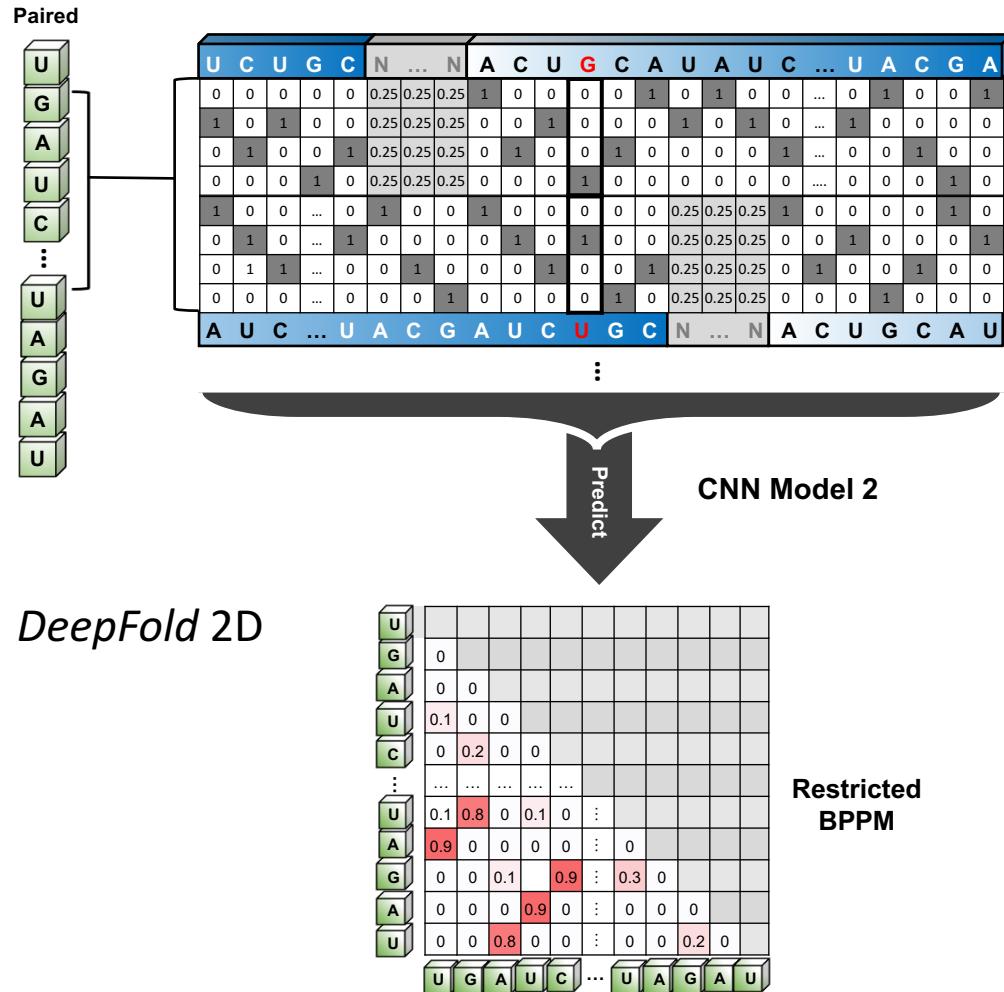
Training Curve



By utilizing L2 regularization and Gaussian noise, the validation accuracy is increased from **0.682** to **0.738**.



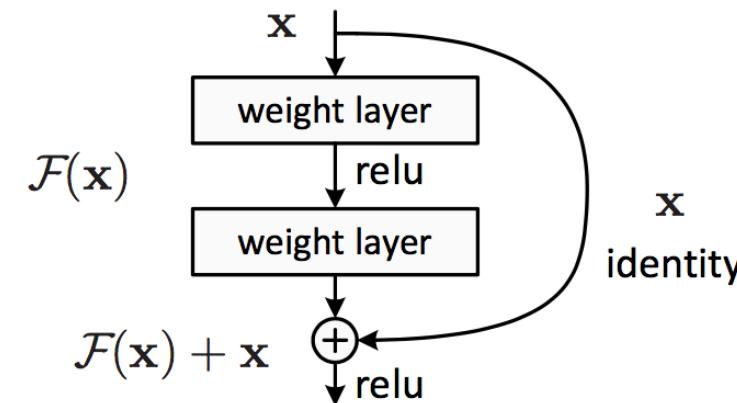
Two-dimensional structure prediction



[1] Severe Class Imbalance

- a. Only consider canonical case pairing. Decrease neg-to-pos ratio from over 200 to 78.
 - b. Data-balanced training approach.

[2] Under-fitting problem



Residual learning framework

Two-dimensional structure prediction

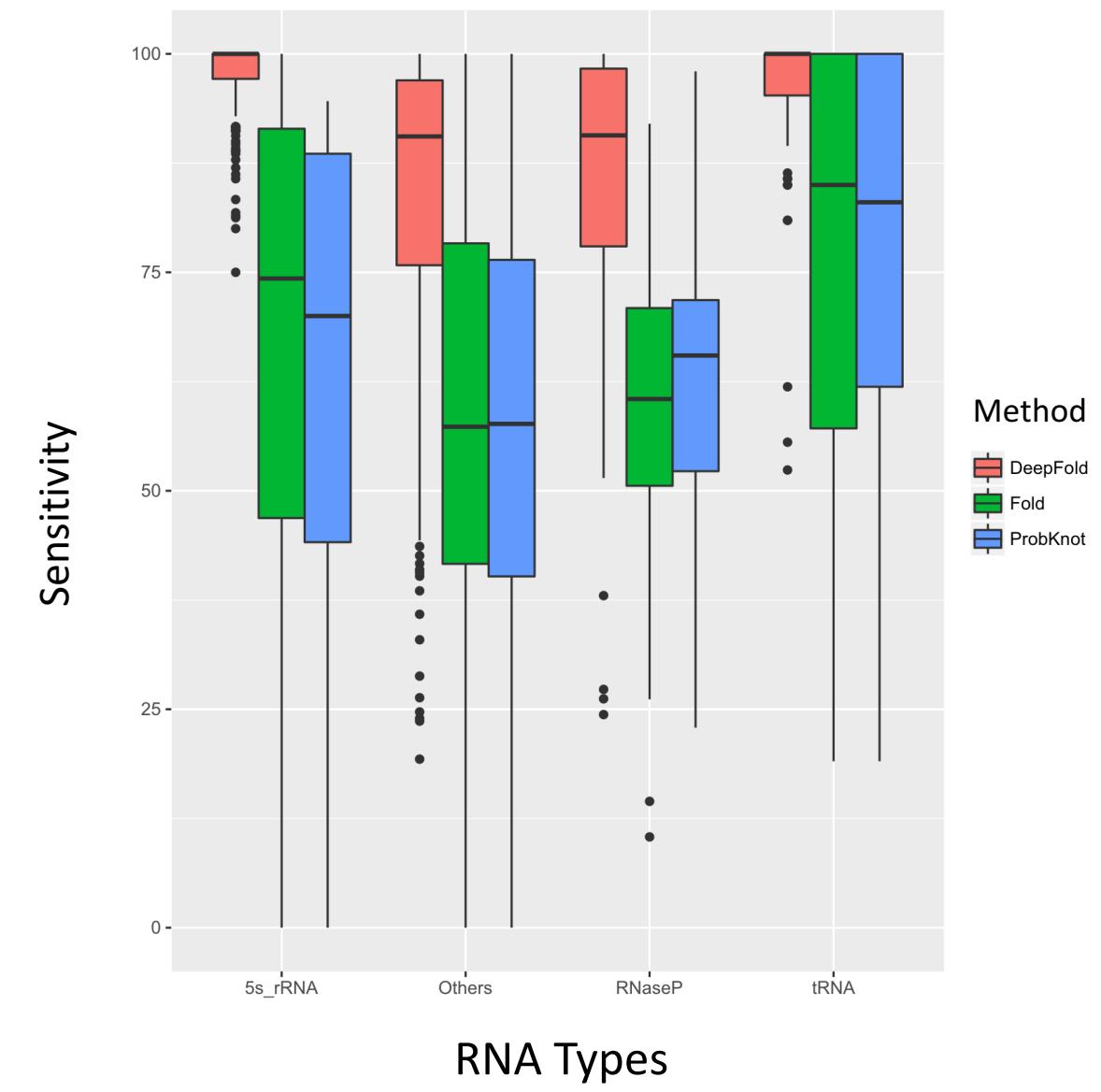
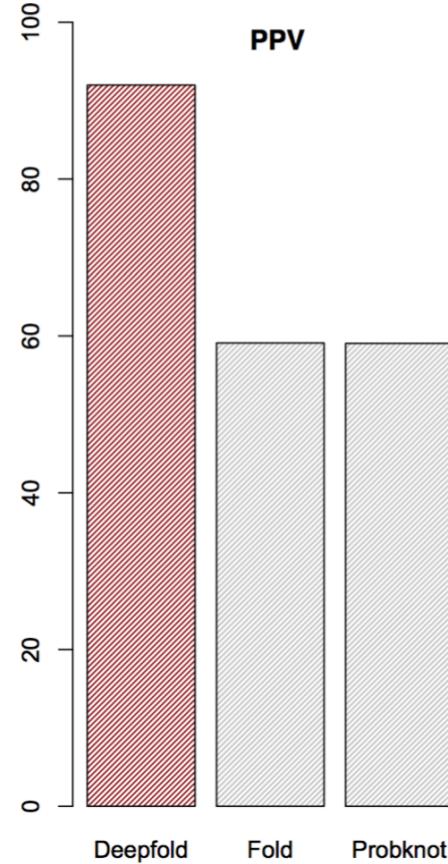
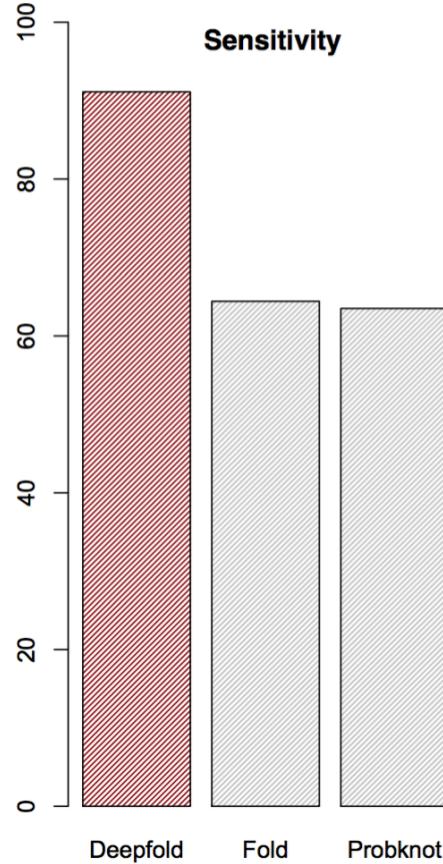
	Test Accuracy on positive samples	Test Accuracy on negative samples	$P(S^+ T^+)$
MLP	0.75	0.81	0.06
CNN	0.89	0.70	0.04
Residual CNN	0.72	0.98	0.31

Stabilize the performance on negative samples and improve the performance on positive samples.

$$P(S^+|T^+) = \frac{P(T^+|S^+)P(S^+)}{P(T^+)}$$

$$P(S^+|T^+) = \frac{P(T^+|S^+)P(S^+)}{P(T^+|S^+)P(S^+)+P(T^+|S^-)P(S^-)} = \frac{P(T^+|S^+)\frac{1}{79}}{P(T^+|S^+)\frac{1}{79}+P(T^+|S^-)\frac{78}{79}}$$

Predictive Power



All RNA sequences, do not filter out sequences with high similarities.

Acknowledgement

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Thank you all for listening!