

# miRNA Discovery & Prediction Algorithms

Sergei Lebedev

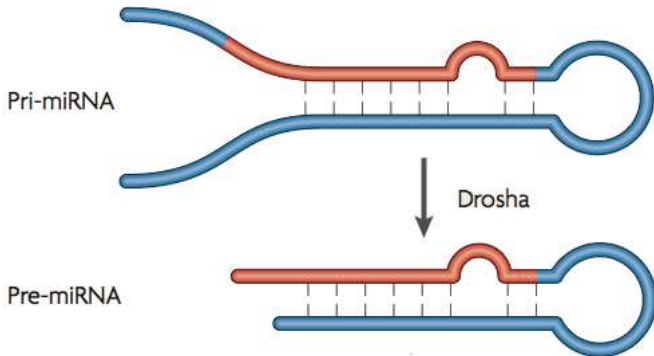
October 13, 2012

- microRNA or miRNA,  $\approx$  22 nucleotide-long **non-coding** RNA;
- mostly expressed in a tissue-specific manner and play crucial roles in cell proliferation, apoptosis and differentiation during cell development;
- thought to be involved in post-transcriptional control in plants and animals;
- linked to disease<sup>1</sup>, for example *hsa-miR-126* is associated with retinoblastoma, breast cancer, lung cancer, kidney cancer, asthma etc.

---

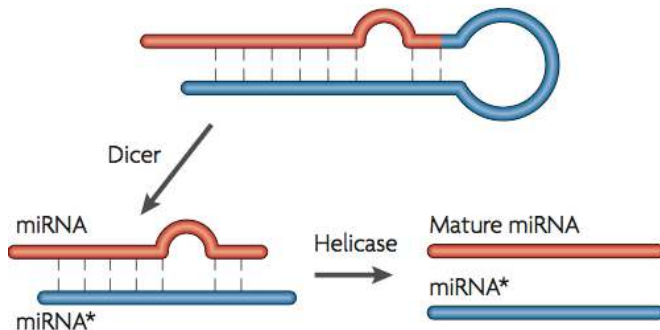
<sup>1</sup>See <http://www.mir2disease.org> for details.

## miRNA in action: nucleus [1]



- **pri-miRNA** is transcribed by RNA polymerase II and seem to possess promoter and enhancer regions, similar to protein coding genes;
- pri-miRNA is then cleaved into (possibly multiple) **pre-miRNA** by an enzyme complex *Drosha*.

## miRNA in action: cytoplasm [1]

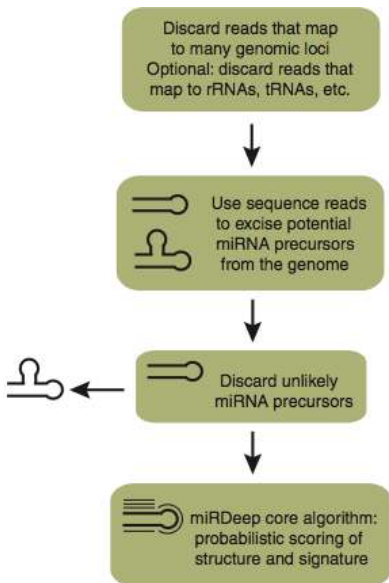


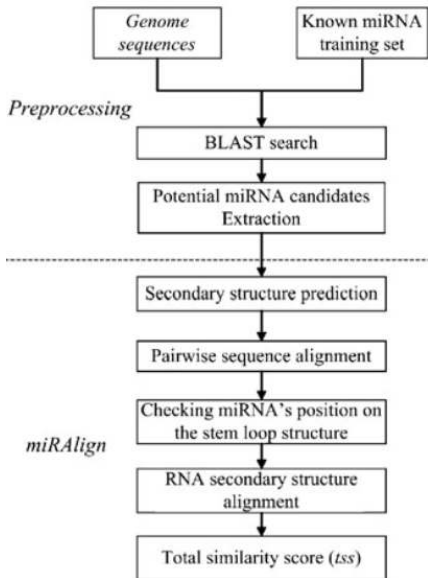
- *Dicer* removes the stem-loop, leaving two complementary sequences: miRNA and miRNA\*, the latter is not known to have any regulatory function.
- Mature miRNA base-pairs with 3' UTR of target mRNAs and blocks protein synthesis or causes mRNA degradation.

- Biological methods: northern blots, qRT-PCR<sup>2</sup>, micro arrays, RNA-seq or miRNA-seq.
- Bioinformatics to the rescue! the usual strategy: first sequence everything, RNA-seq in this case, then try to make sense of whatever the result is.
- In this talk: miRDeep [2], MiRAlign [3], MiRank [4].
- A lot of existing tools out of scope, most can be described with a one liner: *“We’ve developed a novel method for miRNA identification, based on machine learning approach, SVM, HMM!”*.

---

<sup>2</sup>RT for reverse transcription, not real-time.





- Treat miRNA identification problem as a problem of information retrieval, where novel miRNAs are to be retrieved from a set of candidates by the known query samples – “true” miRNAs.
- More formally, given a set of known pre-miRNAs  $X_Q$  as *query samples* and a set of putative candidates  $X_U$  as *unknown samples*, rank  $X_U$  with respect to  $X_Q$ .
- To do so, compute the relevancy values  $f_i \in [0, 1]$  for all unknown samples, assuming  $f_i = 1$  for query samples.
- After that, simply select  $n$  ranked samples, which constitute to predicted pre-miRNA.
- Makes sense, right?



## miRank: how does it work?

- miRank models belief propagation process by doing Markov random walks on a graph, where each vertex corresponds to either known pre-miRNA or a putative candidate and two vertices are connected by an edge if the two vertices are *“close to each other”*.
- Each edge on the graph is assigned a weight  $w_{ij}$ , proportional to the Euclidean distance between the samples  $v_i$  and  $v_j$  (see next slide on how samples are represented).
- When a random walker transits from  $v_i$  to  $v_j$  it transmits the relevancy information of  $v_i$  to  $v_j$  by the following update rule:

$$f_i^{(k+1)} = \alpha \sum_{x_j \in X_U} p_{ij} f_j^{(k)} + \sum_{x_j \in X_Q} p_{ij} f_j \quad p_{ij} = \frac{w_{ij}}{\text{deg}(v_{ij})}$$

## Global

- normalized minimum free energy of folding (MFE);
- normalized no. of paired nucleotides on both arms;
- normalized loop length.

## Local – RNAFold

GUAGCACUAAAGUGCUUAUAGUGCAGGUAGUGUUUAGUUUACUACUGCAUUUAGAGCACUAAAAGUACUGC  
 ((((. (((. (((((((((((((((((( (. (((((.....)). )))))))))))))))))))))). )))). ))))

- Each nucleotide is either paired, denoted by a bracket (– 5' arm, )– 3' arm, or unpaired – .;
- Each local feature is a “word” of length 3, further distinguished by the nucleotide in the middle position, examples: ((. .((.

## miRank: good parts, bad parts & magic

- The method doesn't require any genomic annotations, except for the set of query samples.
- $\approx 75\%$  precision and  $\approx 70\%$  recall even with **very** few query samples (1, 5) – hard to validate, because the source code was never released.
- The notion of *similarity* between query samples, which defines the graph structure is unclear, even though it looks critical for algorithm performance.
- Two user-specified parameters,  $n$  – number of predicted samples and  $\alpha$  – the weight of unknown samples in the relevancy value. How do they affect precision-recall and how to choose them?
- Overall, it seems like miRank isn't used much by biologists<sup>3</sup>.

---

<sup>3</sup>[http://www.ncbi.nlm.nih.gov/pubmed?linkname=pubmed\\_pubmed\\_citedin&from\\_uid=18586744](http://www.ncbi.nlm.nih.gov/pubmed?linkname=pubmed_pubmed_citedin&from_uid=18586744)



K. Chen and N. Rajewsky.

The evolution of gene regulation by transcription factors and microRNAs.

*Nat. Rev. Genet.*, 8(2):93–103, Feb 2007.



M. R. Friedlander, W. Chen, C. Adamidi, J. Maaskola, R. Einspanier, S. Knespel, and N. Rajewsky.

Discovering microRNAs from deep sequencing data using miRDeep.

*Nat. Biotechnol.*, 26(4):407–415, Apr 2008.



X. Wang, J. Zhang, F. Li, J. Gu, T. He, X. Zhang, and Y. Li.

MicroRNA identification based on sequence and structure alignment.

*Bioinformatics*, 21(18):3610–3614, Sep 2005.



Y. Xu, X. Zhou, and W. Zhang.

MicroRNA prediction with a novel ranking algorithm based on random walks.

*Bioinformatics*, 24(13):i50–58, Jul 2008.