Data Article

Benchmark data for identifying N6-methyladenosine sites in the *Saccharomyces cerevisiae* genome

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**Abstract**

This data article contains the benchmark dataset for training and testing iRNA-Methyl, a web-server predictor for identifying N6-methyladenosine sites in RNA (Chen et al., 2015 [15]). It can also be used to develop other predictors for identifying N6-methyladenosine sites in the *Saccharomyces cerevisiae* genome.

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RNA sample was formulated by combining its dinucleotide composition (DNC) [1,2] and the pseudo components [3] since nearly all the machine-learning algorithms can only handle vectors [4]. The concept of pseudo components was originally introduced to reflect the sequence patterns of protein sequences via a series of vector components [5,6] and has been widely used in computational proteomics [7]. Recently, it has been successfully extended to cover DNA [8–11] and RNA sequences [12,13] as well. For the detailed development process in this regard, see a recent review article [14].

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In Appendix A of this paper and at the web-site http://lin.uestc.edu.cn/server/iRNAMethy/data

1. Background

The benchmark dataset for developing computational methods to identify the methylation sites in DNA (see, e.g., [16]) is available [17], and the information thus obtained is very useful for both basic research and drug development. But so far no existing benchmark dataset whatsoever is available for developing computational methods to identify N6-methyladenosine in RNA. The present study was initiated in an attempt to construct a benchmark dataset for the later based on the experimental observations reported by Schwartz et al. [18] recently.

2. Data, experimental design, materials and methods

The data presented here are the benchmark dataset for training and testing iRNA-Methyl [15] (http://lin.uestc.edu.cn/server/iRNAMethy), a web-server predictor for identifying m6A sites in the S. cerevisiae genome. By means of the m6A-seq technique, Schwartz et al. [18] first identified 1,307 methylated adenine (m6A) sites in the S. cerevisiae genome. They have observed that most of the m6A sites share a consensus motif GAC where its center base may be methylated [18]. To construct the corresponding negative benchmark dataset, we used the flexible sliding window approach [19,20] to search the S. cerevisiae genome, and obtained 33,280 RNA segments with exactly the same GAC consensus motif that, however, were not detected by the m6A-seq technique as methylated sites. Furthermore, it had been observed via preliminary tests that when the length of the RNA segments thus derived was 51 bp, the corresponding outcomes were most promising [15]. Accordingly, the 1,307 and 33,280 RNA segments each having 51 bp long were designated as positive and negative samples, respectively. Also, since the size of the negative samples thus obtained is overwhelmingly larger than that of the positive samples, to minimize the false prediction caused by such a highly skewed benchmark dataset, we randomly picked out 1,307 RNA segments from the 33,280...
negative samples to form a negative subset that has the same size with the positive one. The final benchmark dataset thus obtained contains 1,307 positive samples and 1,307 negative samples. Their detailed sequences are given in Appendix A. They can also be downloaded at the web-site http://lin.uestc.edu.cn/server/iRNAMethyl/data.

Conflict of interest

None of the authors claims conflicting interest.

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Appendix A. Supplementary material

Supplementary data associated with this article can be found in the online version at http://dx.doi.org/10.1016/j.dib.2015.09.008.

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