

Prediction of miRNA Target Genes in Various Organisms through Computer-Based Bioinformatics Tools

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Abstract

The current article discusses miRNAs called 'killer RNAs' and the method of their gene regulation, which are highly specific to target genes. There are various bioinformatics tools that can be used to find out the miRNA. Furthermore, the discovery of miRNAs has opened up new possibilities for therapeutic interventions in diseases such as cancer, where dysregulation of miRNAs has been implicated in tumor development and progression. By manipulating the expression of specific miRNAs, researchers hope to develop novel treatments that can target cancer cells with precision and minimal side effects. Overall, the study of miRNAs and their regulatory functions represents a promising area of research with far-reaching implications for understanding gene expression, disease mechanisms, and potential therapeutic strategies. As technology continues to advance, the field of miRNA research is likely to uncover even more insights into the complex world of gene regulation.

Key words: miRNA, MIRGENE, siRNAs, miRNA target, Complementary, Animal miRNAs

miRNA are small RNAs (21-24 nucleotides) long that regulate gene expression in animals. Like small interfering RNAs (siRNAs), miRNAs get into the post-transcriptional gene silencing pathway, leading either to degradation of the target mRNA or to translational repression. miRNA is responsible for controlling protein expression by binding with mRNA [1]. There are many weak binding sites present in mRNA [2]. The complementary target region for miRNA binding is contained within the gene. miRNAs are also termed 'killer RNAs' that have emerged as an answer for deciphering hitherto unknown ways of gene regulation, which are highly specific to target genes [3]. The miRNAs are encoded by a distinct class of genes, called MIRGENE; the transcripts of which do not encode proteins. The transcribed RNA is partially self-complementary and folds into a characteristic structure that includes imperfect double-stranded regions from which the miRNAs originate [4]. The excision of the mature miRNA is a multi-step reaction that involves trimming the initial precursor several times. However, plant miRNAs only show homology to other plant miRNAs and animal miRNAs to animal miRNAs indicating that this mechanism of gene regulation has evolved separately in animals [5].

In silico identification of miRNA targets is possible because animal miRNAs generally exhibit a high degree of complementarity to their target sites [6]. Given the small size of miRNAs, computer-aided target prediction is complicated [7]. To refine the search for miRNA targets, it is important to test and update the known rules for miRNA and target interaction. Since the first discovery of miRNAs *lin-4* and *let-7* by genetic analysis of *C. elegans* developmental timing [8], numerous computational models and tools have been developed to

complement biological experiments to design to study the diverse regulatory roles of miRNAs. Computational as well as experimental analysis shows that most of the human protein-coding genes are regulated by miRNAs [9]. The bioinformatics approaches have been useful in understanding the complexity of miRNA genes and their targets [10]. Most computational methods used for miRNA studies can be classified into two broad categories.

Methods do vary depending on whether a method is for animals, bacteria, or viruses since the biology of miRNAs is somewhat different in each case [11]. A miRNA target detection algorithm can be trained using the properties of known miRNA-mRNA duplexes which can then be used for finding new miRNA-mRNA duplexes [12]. In animals, miRNAs are evolutionarily conserved; many miRNAs, such as miR7 and miR18, are conserved from worm to human [13]. It has been shown that miRNAs in *Arabidopsis* evolved from random sequences by inverted duplication events [14].

These miRNAs can be grouped into 42 miRNA families according to their nucleotide sequence similarity. A majority of these miRNA families have also been found in other species, such as rice, maize, and sorghum [15]. Based on conservation, miRNAs are classified into 4 groups: highly, moderately, lowly, and non-conserved. For some miRNAs, for example, miR163 and miR158, the recent discovery of a much larger set of weakly expressed, less conserved miRNAs in animals which intern provided insights into the origin of miRNAs and hypotheses explaining the birth, selection, and death of miRNAs in land animals [16]. Searching of similar miRNA-miRNA relationships by homology-based computer searching is more useful in plants than animals [17].

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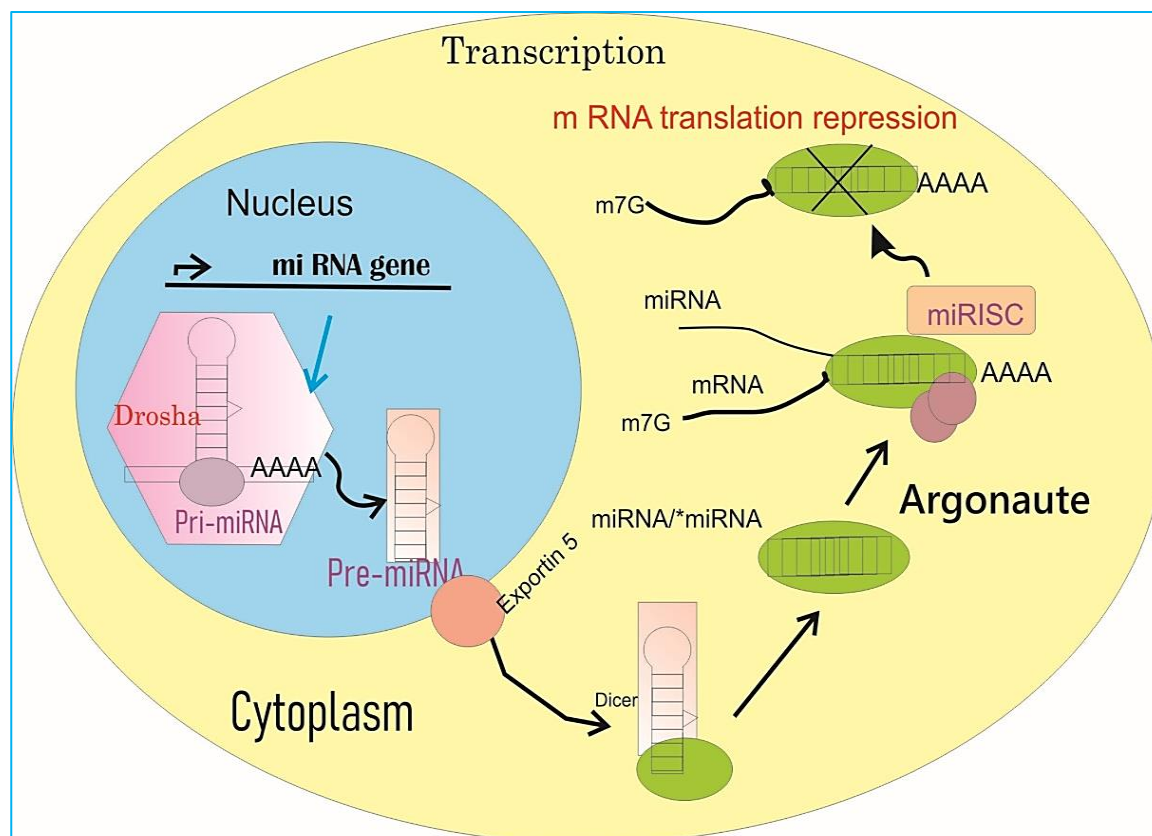


Fig 1 Pre-processing of miRNA (A diagram illustrating the process of miRNA formation and function in animals)

The biogenesis of miRNA initiates in the nucleus, where a large capped and polyadenylated transcript called primary miRNA (pri-miRNA) is transcribed by RNA-polymerase II (RNAPII). The pri-miRNA is then processed by the RNase III endonuclease, Drosha, along with its cofactor, Dgcr8, resulting in the formation of smaller stem-loop structures known as precursor miRNAs (pre-miRNA) [18]. These pre-miRNAs are transported from the nucleus to the cytosol by Exportin 5. In the cytosol, a second RNase III enzyme called Dicer further processes the pre-miRNA, leading to the production of mature miRNA. The mature miRNA then associates with the miRNA-induced silencing complex (miRISC) [19]. Within the miRISC, the seed sequence of the mature miRNA forms Watson-Crick base pairs with complementary sequences primarily located within the 3'-UTRs of mRNA molecules. This base-pairing interaction ultimately results in post-transcriptional gene silencing [20].

MicroRNAs are produced from either their own genes or from introns. It is expected that some non-conserved or lowly conserved miRNAs may be reclassified as more miRNAs are discovered. Conservation is an indicator of miRNA function; conserved miRNAs play an important role in conserved gene regulation, such as leaf and flower morphology or signal transduction [21].

miRNA targets include a large number of transcription factors, suggesting their role in the control of regulatory networks, cellular growth, and development [22]. Conserved miRNAs do not necessarily exhibit the same levels of expression, patterns of the stage of expression in different species. Therefore, sequence and expression divergence in miRNAs between species may affect miRNA accumulation and target regulation leading to developmental changes and phenotypic variation. The conservation of miRNAs and other small RNAs between the conifers and the angiosperms indicates important RNA silencing processes were highly developed in the earliest spermatophytes [23].

A total of 1642 animal miRNAs have been deposited in the miRBase, a database of miRNAs. MicroRNAs (miRNAs) are observed as important post-transcriptional gene regulators, having diverse impacts on diverse pathological and physiological pathways [24]. Experimental identification of miRNA targets is a difficult and time-consuming process. So, there is still a great demand to understand more clearly the total number of existing miRNAs and their full impact on regulatory pathways [25]. As a consequence, several computational prediction methods have been devised to predict targets for follow-up experimental validation such as the Stacking Binding Matrix (SBM) that uses both the information about the miRNAs as well as experimentally validated target sequences in the search for candidate target sequences [26]. Recently, 79 miRNA candidates were discovered by using wheat EST sequences. Similarly, deep sequencing has proved to be an effective strategy for discovering miRNAs of low-abundance and candidate miRNAs and to recover some of the hits missed by existing methods 454 sequencing can be a valuable tool to assign probability-based scores to them [27].

MATERIALS AND METHODS

Identification of miRNA

Downloading EST data sets

Nucleotide sequences (ffn) of *Mycobacterium leprae*, *Mus musculus*, *H. sapiens*, *Drosophila melanogaster*, *C. brigasse*, and *C. elegans* were downloaded from the National Centre for Biotechnology Information (NCBI) (www.ncbi.nlm.nih.gov) in FASTA format.

miRNA data source

The miRNAs were taken from miRBase (<http://microrna.sanger.ac.uk>), a growing repository of all published miRNAs of viruses, animals, and plants [28]. The

miRBase Sequence Database is a searchable database of published miRNA sequences and annotations [29]. The miRBase Registry continues to provide gene hunters with unique names for novel miRNA genes prior to the publication of results. Every entry in the miRBase gives a predicted hairpin portion of a miRNA transcript (termed mir in the database), with information on the location and sequence of the mature miRNA sequence (called miR) [30]. Both hairpin and mature sequences are available for BLAST, and entries can also be retrieved by name, keyword, references, and annotation. Starting from 120-200 bases stem-loop precursor to mature 20-21 bases long sequences are available in FASTA format for download [31]. Along with miRNA ID, other information like accession number, chromosomal location, sense or antisense strand, start and end point, etc. are available. These miRNA sequences were downloaded in FASTA format and saved in a text file to find their potential targets [32]. In animals, these target sites are usually present at the 3' untranslated region (3'UTR) of the mRNA [33]. Genomic sequences of six species were used in this work, including *C. briggsae*, *C. elegans*, *D. melanogaster*, *Mus musculus*, and *Homo sapiens*. Most of these miRNAs were identified or verified by experiments, and others were computationally predicted as their close homologs [34]. In this comparative approach, we tested the algorithms by using a publicly available miRNA knockdown [35]. Prediction of miRNA targets get affected by improper sequence specificity, easy availability of target site, and mRNA structure thermodynamic [36].

Construction of local database for standalone BLAST

To identify miRNA targets, features just as seed matching, site conservation, free energy, and site accessibility are checked [37] (Wen et al., 2018). The FASTA formatted text file of *Mycobacterium leprae* was uploaded to create a local nucleotide database using the "Accessory application" available in BioEdit software [38]. The database was constructed separately for *Mycobacterium leprae* as a "local nucleotide database" which was used for target identification by standalone blasts [39]. Then we did-

BLAST Local BLAST

Prediction of miRNA targets

Despite continued progress, bioinformatics prediction of microRNA targets remains a challenge since many software lack accuracy and sensitivity [40]. Several preprocessing steps are taken to draw miRNA candidate sequences from the genome [41]. The reverse complementary miRNAs were used one by one as a query sequence and a standalone blast was performed against the respective "local nucleotide database" of *Mycobacterium leprae* [42].

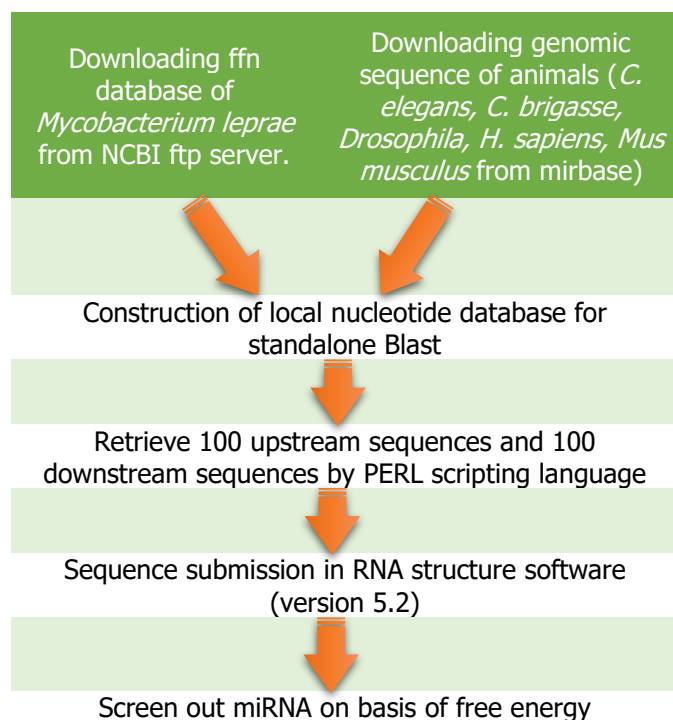
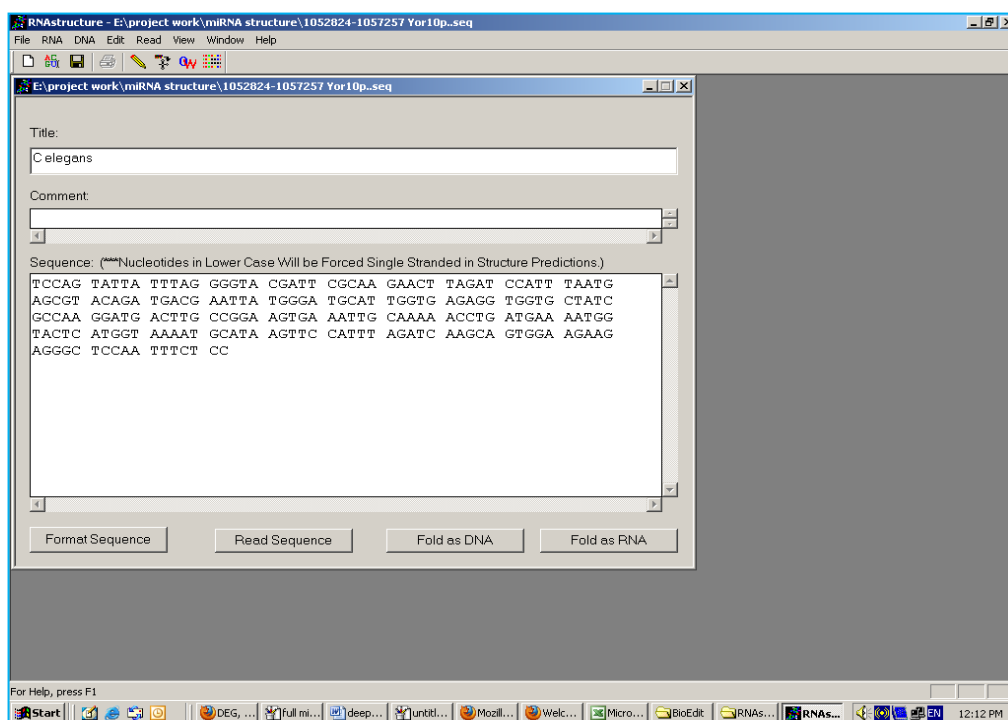


Fig 2 Flow chart depicting screening of miRNA on the basis of free energy



<https://rna.urmc.rochester.edu/RNAstructure.html>

Fig 3 Sequence submission in RNA structure software

RESULTS AND DISCUSSION

Number of miRNA target genes predicted in *C. elegans* = 9
 Number of miRNA target genes predicted in *C. brigasse* = 5
 Number of miRNA target genes predicted in *Drosophila melanogaster* = 4
 Number of miRNA target genes predicted in *Homo sapiens* = 19

Number of miRNA target genes predicted in *Mus musculus* = 78

Number of miRNA target genes predicted in *Mycobacterium leprae* = 14

The (Table 1) presents data on predicted microRNA (miRNA) target genes across different species, including *Caenorhabditis elegans*, *C. brigasse*, *Drosophila melanogaster*, *Homo sapiens*, and *Mus musculus*.

Table 1 Prediction of miRNA target genes in various organisms

Gene id	Name of protein	mi-RNA ID	Length	Bit score	Identity (%)	E-value	Contig sequence
<i>Caenorhabditis elegans</i>							
NC_002677.1 c251441-5-2513870	Hypothetical protein ml2113	cel-miR-2	23	28.2	100	0.06	CAGCTTTGATGTGC
NC_002677.1 2145165-2146505	Sugar transport periplasmic binding protein	cel-miR-60*	22	32.2	100	0	ACTGGAAGAGTGCCAT
NC_002677.1 c288343-4-2882646	Putative cytochrome c-type biogenesis protein	cel-miR-70	23	28.2	100	0.06	GTCGTTGGTGTTC
NC_002677.1 c65112-63319	Hypothetical protein ml00052	cel-miR-358*	23	28.2	100	0.06	CCTGGCCAGGCATT
NC_002677.1 218458-219639	Succinyl-co synthetase subunit beta	cel-miR-796	23	28.2	100	0.06	ATGTAGTTGAGGTT
NC_002677.1 c659721-660932	S-denosylmethionine synthetase	cel-miR-1832	21	30.2	100	0.01	GGCGGAGCGAATCGA
NC_002677.1 c253290-9-2531614	Type ii citrate synthase	cel-miR-2209c	23	28.2	100	0.06	ACCGGTTACACTAC
NC_002677.1 1339580-1340836	Udp-n-acetylglucosamine 1-carboxyvinyltransferase	cel-miR-1832b	23	28.2	100	0.06	GGCAGAGCGATTTCG
NC_002677.1 1417029-1417634	Lipoprotein signal peptidase	cel-miR-2217	24	30.2	94	0.02	TGGGAAGTCGGTGTCGATC
<i>C. brigasse</i>							
NC_002677.1 c192573-7-1924844	Site-specific tyrosine recombinase xerc	cbr-miR-124	21	30.2	100	0.01	CACGCGGTGAATGCC
NC_002677.1 c707588-706461	Transaldolase	cbr-miR-40	22	28.2	100	0.05	GGTGTCAATCAGCT
NC_002677.1 2457996-2460191	Malate synthase g	cbr-miR-76	20	28.2	94	0.04	TTCGTTGTTGACGAAGTC
NC_002677.1 c280487-2-2799404	Polyketide synthase	cbr-miR-235b*	24	28.2	100	0.06	GCTCTGGTCATGTG
NC_002677.1 c488641-486263	Sugar phosphorylase	crm-miR-786	23	28.2	100	0.06	CTTGCTGAGATTCC
<i>Drosophila melanogaster</i>							
NC_002677.1 c313170-0-3128833	Integral membrane protein	dme-miR-306*	22	30.2	94	0.01	GGTCGCTCTGTGCCTGTGC
NC_002677.1 1073832-1075859	Penicillin-binding protein 2	dme-miR-985	22	28.2	100	0.05	CCAATGGTCGGGCA
NC_002677.1 c193999-1-1939452	16s rna-processing protein rim	dme-miR-252	22	28.2	100	0.05	AAGTACTAGTGCCG
NC_002677.1 303740-305263	Lysyl-trna synthetase	dme-miR-1012	23	28.2	100	0.06	GTCAAAGATTTTCC
<i>H. sapiens</i>							
NC_002677.1 c276904-3-2767652	Hypothetical protein ml2336	hsa-miR-21	21	28.2	100	0.05	CAGTCGATGGGCTG
NC_002677.1 1443804-1444400	Hypothetical protein ml1222	hsa-miR-16-1*	22	28.2	100	0.05	TATTAAGTGTGCTG
NC_002677.1 c3090766-3091569	Hypothetical protein ml2587	hsa-miR-24	22	28.2	100	0.05	TGGCTCAGTTCAGC
NC_002677.1 1699648-1700454	Cdp-diacylglycerol pyrophosphatase	hsa-miR-33a	22	30.2	100	0.01	TTTCCACAGTGCATC
NC_002677.1 1352912-1354093	Acetyl-coa acetyltransferase	hsa-miR-27b*	22	28.2	100	0.05	GCTGATTGGTGAAC
NC_002677.1 c192573-7-1924844	Site-specific tyrosine recombinase xerc	hsa-miR-124	20	30.2	100	0.01	CACGCGGTGAATGCC
NC_002677.1 c798689-797292	Acyl-coa dehydrogenase	hsa-miR-150*	22	32.2	100	0	ACAGGCCTGGGGGACA
NC_002677.1 c313170-0-3128833	Integral membrane protein	hsa-miR-185	22	28.2	100	0.05	AAGGCAGTTCCTGA
NC_002677.1 c218947-5-2188468	Hypothetical protein ml1808	hsa-miR-194	22	28.2	100	0.05	CAGCAACTCCATGT
NC_002677.1 c180547-4-1803621	Hypothetical protein ml1497	hsa-miR-106b	21	28.2	100	0.05	GTGCTGACAGTGCA

NC_002677.1 c270449-268857	Hypothetical protein ml0201	hsa-miR-200a*	22	28.2	100	0.05	CCGGACAGTGCTGG
NC_002677.1 c1858031-1856520	Hypothetical protein ml1539	hsa-miR-323-5p	22	28.2	100	0.05	GGTCCGTGGCGCGT
NC_002677.1 188551-189252	Lipoprotein	hsa-miR-338-3p	22	28.2	100	0.05	CAGTGATTTTGTG
NC_002677.1 c88719-88246	Ribonuclease activity regulator	hsa-miR-425	23	28.2	100	0.06	AATGACACGATCAC
NC_002677.1 2068541-2069857	Transferase	hsa-miR-425*	22	28.2	100	0.05	TGTCGTGTCCGCC
NC_002677.1 c2086347-2084623	Cytochrome c oxidase, polypeptide	hsa-miR-486-5p	22	28.2	100	0.05	ACTGAGCTGCCCCG
NC_002677.1 2021235-2021810	Hypothetical protein ml1677	hsa-miR-493	22	28.2	94	0.05	GGTCTACTGTGTACCA GG
NC_002677.1 2476953-2478869	Multidrug resistance pump	hsa-miR-497	21	28.2	94	0.05	GCAGCACACTGTTGTTT G
NC_002677.1 1352912-1354093	Acetyl-coaacyltransferas	mmu-miR-27b	22	28.2	100	0.05	GCTGATTGGTGAAC
		<i>Mus musculus</i>					
NC_002677.1 c1925737-1924844	Site-specific tyrosine recombinase xerc	mmu-miR-124	20	30.2	100	0.01	CACGCGGTGAATGCC
NC_002677.1 c798689-797292	Acyl-coa dehydrogenase	mmu-miR-150*	22	28.2	100	0.05	ACAGGCCTGGGGGA
NC_002677.1 c2443504-2442641	Cell surface protein	mmu-miR-181a	23	28.2	100	0.06	CAACGCTGTCTGGTG
NC_002677.1 c3131700-3128833	Integral membrane protein	mmu-miR-185	22	28.2	100	0.05	AAGGCAGTTCCTGA
NC_002677.1 3090766-3091569	Hypothetical protein ml2587	mmu-miR-24	22	28.2	100	0.05	TGGCTCAGTTCAGC
NC_002677.1 c2561489-2561097	Hypothetical protein ml2158	mmu-miR-297a	22	28.2	100	0.05	GTGCATGTGCATGT
NC_002677.1 c1805474-1803621	Hypothetical protein ml1497	mmu-miR-106b	21	28.2	100	0.05	GTGCTGACAGTGCA
NC_002677.1 c1915424-1914867	Ribosome recycling factor	mmu-miR-30c-2* M	22	28.2	100	0.05	GGAGAAGGCTGTTT
NC_002677.1 2925762-2926499	Hypothetical protein ml2450	mmu-miR-148a*	22	28.2	100	0.05	GAGACACTCCGACT
NC_002677.1 c270449-268857	Hypothetical protein ml0201	mmu-miR-200a*	22	28.2	100	0.05	CCGGACAGTGCTGG
NC_002677.1 c2769043-2767652	Hypothetical protein ml2336	mmu-miR-21*	22	30.2	100	0.01	GCAGTCGATGGGCTG
NC_002677.1 c1858031-1856520	Hypothetical protein ml1539	mmu-miR-323-5p	22	28.2	100	0.05	GGTCCGTGGCGCGT
NC_002677.1 c1124822-1123017	Pyruvate phosphate dikinase	mmu-miR-328*	22	30.2	100	0.01	GGCAGGAGGGGCTCA
NC_002677.1 188551-189252	Lipoprotein	mmu-miR-338-3p	22	28.2	100	0.05	CAGTGATTTTGTG
NC_002677.1 1668337-1671246	Excinnucleaseabc subunit A	mmu-miR-223	22	28.2	100	0.05	TTGTCAAATACCCC
NC_002677.1 1699648-1700454	Cdp-diacylglycerolpyrophosphatase	mmu-miR-33	22	30.2	100	0.01	TTTCCACAGTGCATC
NC_002677.1 411592-412068	2-c-methyl-d-erythritol	hsa-miR-194*	22	28.2	100	0.05	GTGGGGCTGCTGTT
NC_002677.1 3082923-3084725	Hypothetical protein ml2582	mmu-miR-382	22	28.2	100	0.05	CGGACAACACTTTT
NC_002677.1 c88719-88246	Ribonuclease activity regulator	mmu-miR-425	23	28.2	100	0.06	AATGACACGATCAC
NC_002677.1 c2922708-2921383	Lipoprotein	mmu-miR-431	21	30.2	100	0.01	TGTCTTGACAGGCCGT
NC_002677.1 918368-920137	Lipoprotein lpqb	mmu-miR-470	23	28.2	100	0.06	TTCTTGGACTIONGCA
NC_002677.1 c2086347-2084623	Cytochrome c oxidase, polypeptide i	mmu-miR-486	22	28.2	100	0.05	ACTGAGCTGCCCCG
NC_002677.1 582395-583444	Holliday junction dna helicase ruvb	mmu-miR-546	16	26.3	100	0.1	ATGGTGGCACGGA
NC_002677.1 c902902-901763	2-phospho-l-lactate transferase	mmu-miR-543	22	32.2	100	0	ATTGCGGTGCACTTC
NC_002677.1 1668337-1671246	Excinnucleaseabc subunit A	mmu-miR-539-5p	22	28.2	100	0.05	TTATCCTTGGTGTG
NC_002677.1 2383320-2387525	Peptide synthase	mmu-miR-494	22	28.2	100	0.05	TTGTCCGTGTTGTC
NC_002677.1 182173-188484	Polyketide synthase	mmu-miR-592*	22	34.2	100	0.06	CATCACGTGGTGACGC A

NC_002677.1 :3058964-3060328	Acyl-carrier-protein	mmu-miR-551b*	23	28.2	100	0.06	GCTTGGGTGAGACC
NC_002677.1 :2081-3280	Dna polymerase iii subunit beta	mmu-miR-668	24	28.2	100	0.06	TCGGCTCGGCCCAC
NC_002677.1 :c1187205-1186213	Integral membrane protein	mmu-miR-665*	26	30.2	100	0.02	TGCCTCTATCCAGGA
NC_002677.1 :c1734472-1733639	Glycosyltransferase	mmu-miR-667*	25	28.2	100	0.07	CGGTGCTGGTGGAG
NC_002677.1 :c1466688-1465612	Hypothetical protein ml1232	mmu-miR-667	23	28.2	100	0.06	CTGCCACCCAGCCC
NC_002677.1 :c2214166-2212448	Ppe-family protein	mmu-miR-761	22	28.2	100	0.05	CAGCAGGGTGAAC
NC_002677.1 :1694884-1696296	Gininosuccinatelyase	mmu-miR-344d-1*	24	30.2	100	0.02	CTGCTGGCTATACAC
NC_002677.1 :194221-200571	Mycocerosic synthase	mmu-miR-666-3p	22	28.2	100	0.05	CGTGATCGCCTGCT
NC_002677.1 :627741-628964	Chorismate synthase	mmu-miR-760-5p	23	32.2	100	0	GGCCACCAGAGCCCGG
NC_002677.1 :c425727-424867	Abc-transporter	mmu-miR-674	22	28.2	100	0.05	TGAGATGGGAGTGG
NC_002677.1 :1630435-1631208	Pseudouridine synthase	mmu-miR-677 M	22	28.2	100	0.05	GATGATTAGCTTCT
NC_002677.1 :2476953-2478869	Multidrug resistance pump	mmu-miR-497	22	28.2	94	0.05	GCAGCACACTGTTGTTTG
NC_002677.1 :c2561489-2561097	Hypothetical protein ml2158	mmu-miR-669b	22	28.2	100	0.05	GTGCATGTGCATGT
NC_002677.1 :118819-120768	Hypothetical protein ml0096	mmu-miR-501-5p	22	28.2	100	0.05	GTCCCTGGGTGAAA
NC_002677.1 :1086005-1087030	Ftsq-family protein	mmu-miR-676	21	28.2	100	0.05	CGTCCTGAGGTTGT
NC_002677.1 :c2372198-2369802	Isocitratelase	mmu-miR-877	20	28.2	100	0.04	GTAGAGGAGATGGC
NC_002677.1 :c147327-144115	Arabinosyltransferase	mmu-miR-598	23	30.2	94	0.01	CGGTGATGCCGATGATGCG
NC_002677.1 :386629-388932	Serine-threonine protein kinase	mmu-miR-598	22	30.2	100	0.01	GTCATCGTCGTCATC
NC_002677.1 :c2561489-2561097	Hypothetical protein ml2158	mmu-miR-669f-5p	24	28.2	100	0.06	GTGCATGTGCATGT
NC_002677.1 :c2561489-2561097	Hypothetical protein ml2158	mmu-miR-466j	23	28.2	100	0.06	GTGCATGTGCATGT
NC_002677.1 :c2561489-2561097	Hypothetical protein ml2158	mmu-miR-466j	23	28.2	100	0.06	GTGCATGTGCATGT
NC_002677.1 :1548392-1549180	Hypothetical protein ml1300	mmu-miR-1197*	21	34.2	100	0.06	GGTTGACCATGGTGTGT
NC_002677.1 :c212814-211150	Glucose-6-phosphate isomerase	mmu-miR-1897-3p	22	30.2	100	0.01	CTCGTTCTGTCCGGT
NC_002677.1 :503217-504401	Hypothetical protein ml0405	mmu-miR-1903	22	28.2	100	0.05	CTTCTTCTTCTTCC
NC_002677.1 :c899638-897971	Dnamethylase	mmu-miR-1899	22	28.2	100	0.05	GCGATGGCCGAATC
NC_002677.1 :124599-126506	Long-chain-fatty-acid--coa ligase	mmu-miR-1906	22	32.2	100	0	GCAGCAGCCTGAGGCA
NC_002677.1 :3204710-3209539	Cation transport atpase	mmu-miR-1893	22	30.2	100	0.01	CGGGCGCTGGACGCC
NC_002677.1 :c1194348-1193875	Hypothetical protein ml1027	mmu-miR-1945	22	28.2	100	0.05	GCGGGTACTGTCGG
NC_002677.1 :c3264167-3263112	Cell division protein	mmu-miR-1306	21	30.2	100	0.01	TTGGCTCTGGTGGTG
NC_002677.1 :3065268-3065441	Hypothetical protein	mmu-miR-1946a	27	30.2	100	0.02	CCGGGCAGTGGTGGC
NC_002677.1 :c2561489-2561097	Hypothetical protein ml2158	mmu-miR-669m-5p	23	28.2	100	0.06	GTGCATGTGCATGT
NC_002677.1 :c2811474-2804869	Polyketide synthase	mmu-miR-1955-3p	21	28.2	100	0.05	GAGCATTGCATGCT
NC_002677.1 :c2475105-2472769	Preproteintranslocase subunit seca	mmu-miR-1966	25	28.2	100	0.07	GGAGCTGGCTCAGG
C_002677.1 :3065268-3065441	Hypothetical protein	mmu-miR-1946b	26	30.2	100	0.02	CCGGGCAGTGGTGGC
NC_002677.1 :3065557-3069774	Integral membrane protein	mmu-miR-2861	19	28.2	100	0.04	GGCCTGGCGGCGGG
NC_002677.1 :2617849-2618970	Hypothetical protein ml2203	mmu-miR-3057-5p	24	28.2	100	0.06	GCTGAGATTCTGCG
NC_002677.1 :c526941-526294	Oligoribonuclease	mmu-miR-3057-5p	24	28.2	100	0.06	GGAGCTGAGATTCT

NC_002677.1 c587947-584381	Acyl-co synthetase	mmu-miR-3064-5p	22	28.2	100	0.05	TGGCTGTTGTGGTG
NC_002677.1 188551-189252	Lipoprotein	mmu-miR-3065	22	28.2	100	0.05	
NC_002677.1 c2039801-2038215	D-3-phosphoglycerate dehydrogenase	mmu-miR-3067*	23	30.2	100	0.01	GGCTGCCCTGGGAGA
NC_002677.1 c2861605-2861300	Hypothetical protein ml2390	mmu-miR-3068*	22	28.2	100	0.05	GGTGAATTGCAGTA
NC_002677.1 c3170139-3169324	Hypothetical protein ml2649	mmu-miR-3073-3p	22	28.2	100	0.05	TGTCCACTGTGACC
NC_002677.1 3090766-3091569	Hypothetical protein ml2587	mmu-miR-3074-5p	22	30.2	100	0.01	GCTGAACTGAGCCAG
NC_002677.1 2303405-2304298	Dnaglycosylase	mmu-miR-3080-3p	22	28.2	100	0.05	CGGGCAAAGCGCTT
NC_002677.1 700214-703018	Phosphoenolpyruvate carboxylase	mmu-miR-3084	21	28.2	100	0.05	GCCAGTCTCCTTCA
NC_002677.1 219659-220561	Succinyl-co synthetase subunit	mmu-miR-3086-5p	20	30.2	100	0.01	AGATTGTAGGCCCAT
NC_002677.1 c2561489-2561097	Hypothetical protein ml2158	mmu-miR-466m-5p	23	28.2	100	0.06	GTGCATGTGCATGT
NC_002677.1 1877242-1878684	Prolyl-trnasynthetase	mmu-miR-3093-5p	22	28.2	100	0.05	ACCCCGCGGAGCTC
NC_002677.1 c2229420-2228980	50s ribosomal protein l15	mmu-miR-3105-5p	22	28.2	100	0.05	AAGCCCGTAAGCAG
NC_002677.1 1519218-1519658	Hypothetical protein ml1276	mmu-miR-3105-5p	22	28.2	100	0.05	AGCCCGTAAGCAGC
NC_002677.1 c2086347-2084623	Cytochrome c oxidase, polypept	mmu-miR-3107	22	28.2	100	0.05	ACTGAGCTGCCCCG

Overall, the (Table 1) provides a comprehensive view of miRNA-target gene interactions across various species, highlighting high-confidence predictions with 100% identity and significant bit scores. The E-values indicate the statistical significance of these interactions, with lower values representing higher confidence.

Hence prediction of miRNA drug target genes involves availability of latest software's and well-versed data sets. Source of miRNA, miRNA ID, number of mismatch sequences, length of contig sequence, Bit score, E value was easily determined by above methods. In this study we have analyzed the genomic nucleotide sequence of six species *Mus musculus*, *Homo sapiens*, *Drosophila melanogaster*, *C. brigasse*, *Caenorhabditis elegans* and *Mycobacterium leprae*. From the

above study it is clear that number of miRNA predicted genes are found highest in *Mus musculus* than *H sapiens* and also proteins coded by them were also predicted. 42% proteins of *Mycobacterium leprae* are homologous with Human genome while rest are non-homologous. Non-homologous sequences are selected as target genes for miRNA. Protein encoded by these drug target gene searched by ExPaSy proteome server to find out their functions. 51% genes were involved in endoplasmic reticulum while 36% used in periplasmic process. However, 13% (34) genes are helpful in the synthesis of cell membrane and identified as target, which can be targeted in future. The sequences work as a magic bullet for anti-bacterial and antifungal drug. A total of 115 miRNA predicted drug target genes are identified by computational method.

Table 2 miRNA of M leprae NM: Number of mismatch; LM: Length of mature miRNA; LP: Length of precursor; MFEI: Minimal folding free energy index

S. No.	New miRNA	Source	miRNA sequence	NM/nU	LM/nU	LP/nU	MFEI
1	micro-MIR 59	GSS	GCAGUCGAUGGGCUG	0	15	215	-65.4
2	micro-MIR 8	GSS	GGCGGAGCGAAUCGA	0	15	215	-92.1
3	micro-MIR 79	GSS	CGGUGCUGGUGGAG	0	14	214	-83.5
4	micro-MIR 39	GSS	CCGGACAGUGCUGG	0	14	214	-105.1
5	micro-MIR 58	GSS	CCGGACAGUGCUGG	0	14	214	-105.1
6	micro-MIR 11	GSS	UGGGCAGUCGGUGUCGAUC	6	18	218	-77.6
7	micro-MIR 31	GSS	UGGCUCAGUUCAGC	0	14	214	-88.4
8	micro-MIR 70	GSS	ACUGAGCUGCCCCG	0	14	214	-83.4
9	micro-MIR 23	GSS	CCAAUGGUCGGGCA	0	14	214	-69.9
10	micro-MIR 25	GSS	GUCAAAGAUUUUCC	0	14	214	-62
11	micro-MIR 82	GSS	CUGCUGGCUAUACAC	0	15	215	-95
12	micro-MIR 43	GSS	UGUCGUGUCCGCC	0	15	215	-82.6
13	micro-MIR 22	GSS	GGUCACUCUGGCCUGUGC	6	18	218	-10.8
14	micro-MIR 24	GSS	AAGUACUAGUGCCG	0	14	214	-67.8

CONCLUSION

Hence from the above study it has been concluded that prediction of miRNA target genes was easily achieved through computer-based software and web-based bioinformatics tools. For prediction of miRNA target genes, we have to identify certain features such as miRNA binding sites and target down regulation by various methods. All the web-based bioinformatics tools, databases and software provides valuable miRNA target genes in various species of rat, dog and chicken also. By use of PERL scripting language and sequence submission in RNA structure software, 115 miRNA drug target genes have been predicted in various organisms like *Mus musculus*, *Homo sapiens*, *Drosophila melanogaster*, *C. brigasse* and *Caenorhabditis elegans* and 14 miRNA targets

have been identified in *Mycobacterium leprae* which appeared to be related to the development, growth, metabolism and other physiological processes such as stress response of *Mycobacterium leprae*. The combined usage of nucleotide databases of various organisms and their sequence submission in RNA mediated software will increase the true prediction and decrease the availability of false positive results. The findings from this study will contribute to further researches in miRNAs function and regulatory mechanisms in *Mus musculus* which will be very beneficial to study some essential features in *Mus musculus* and also studies the confirmation of interaction between predicted miRNA-miRNA duplexes of various model and novel organisms. These findings will lead to open new avenues in computational research and miRNA will act as potential therapeutic targets for human disease.

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