Combination of measures distinguishes pre-miRNAs from other stem-loops in the genome of the newly sequenced *Anopheles darlingi*

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Supplementary Materials Efficient identification of candidate hairpins

The problem of identifying candidate stem-loop structures in a genome can be cast as the problem of finding an imperfect palindrome with a central intervening sequence. Scanning the entire genome of interest in an attempt to seek imperfect palindromes directly proved to be computationally unfeasible, especially considering the irregular nature of these stem-loop structures with potentially numerous and large bulges as well as non-canonical base pairings.

We adopted, instead, a filtering approach based on the observation that a segment of a genome which upon transcription can adopt a stem-loop conformation should exhibit a higher degree of potential pairing between its two halves (if the midpoint of the segment falls within the region corresponding to the terminal loop) than a segment that either does not contain a stem-loop or only partially contains such a structure. Let S be a string over an alphabet Σ , |S| denotes the length of the string, and S_i denotes the *i*th position of the string, with $0 < i \leq |S|$, \overleftarrow{S} denotes the reversed string, i.e., $\overleftarrow{S} = S_{|S|} \dots S_1$. We denote by $L^S = S_1 \dots S_{\lfloor |S|/2 \rfloor}$ and $R^S = S_{\lceil |S|/2 \rceil} \dots S_{|S|}$, respectively, the left and right halves of string S so that $S = L^S R^S$. Let $\Phi \subset \Sigma^2$ be the set of accepted pairings of characters in Σ . The set Φ induces the predicate

 $F_{\Phi}: \Sigma^* \times \Sigma^* \mapsto \{0,1\} \text{ defined as follows: } F_{\Phi}(a\alpha, b\beta) = 1 \quad \text{iff} \quad (a,b) \in \Phi \land (F_{\Phi}(\alpha, \beta) = 1 \quad \lor \quad \alpha = \beta = \varepsilon),$ with $a, b \in \Sigma, \alpha, \beta \in \Sigma^*$, and ε being the empty string.

In order to consider acceptable RNA pairings, including all canonical pairs along with G:U basepairs, we have $\Phi = \{(A, U), (U, A), (C, G), (G, C), (G, U), (U, G)\}.$

The best local alignment over a split string S is calculated using a DP matrix where each cell, H(i, j), is computed with the following recurrence:

$$\max \left\{ \begin{array}{ll} 0 \\ H(i-1,j-1) + \xi_0 & \text{if } F_{\Phi}(L^S{}_i,\overleftarrow{R^S}{}_j) = 1 \\ H(i-1,j-1) - \xi_1 & \text{if } F_{\Phi}(L^S{}_i,\overleftarrow{R^S}{}_j) = 0 \\ H(i-1,j) - \xi_2 \\ H(i,j-1) - \xi_2 \end{array} \right\}$$

where ξ_0 , ξ_1 , and ξ_2 represent the contribution of matches, mismatches, and gaps, respectively, to the alignment score.

Using the Smith-Waterman algorithm on a split string, S, one can determine the best alignment in $O(|S|^2)$. Consider a genome with k chromosomes, seen as a collection of sequences $S = \{S_1, S_2, \ldots, S_k\}$. The algorithm will slide a window of length w along each chromosome of the genome determining, for each position, the best local alignment under a model $M = (\xi_0, \xi_1, \xi_2)$.

Using the described sliding-window procedure, the best alignments for all windows in the genome can be computed in $O(w^2 \sum_i (|S_i| - w + 1))$.

An alignment Λ of two sequences S_1, S_2 is a tuple (e_1, e_2, σ) where $0 \le e_1 \le |S_1|, 0 \le e_2 \le |S_2|$, and $\sigma \in \{\uparrow, \leftarrow, \diagdown\}^*$.

If $\Lambda = (e_1, e_2, \sigma)$ is a best local alignment of a split string, then:

- ∀i, j H(i, j) ≤ H(e₁, e₂), where H(i, j) is the value of the *i*th row, *j*th column of the DP matrix of the Smith-Waterman algorithm.
- σ represents a path from (e_1, e_2) to a cell in the DP matrix containing the value 0 such that if the kth cell in the path is (i_k, j_k) and $H(i_k, j_k) \neq 0$ then the (k + 1)th cell in the path is:

$$- (i_{k} - 1, j_{k} - 1) \text{ if } H(i_{k}, j_{k}) = H(i_{k} - 1, j_{k} - 1) + \xi_{0} \text{ and } \sigma_{k} = \langle i_{k} - 1, j_{k} - 1 \rangle \text{ if } H(i_{k}, j_{k}) = H(i_{k} - 1, j_{k} - 1) - \xi_{1} \text{ and } \sigma_{k} = \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k}, j_{k}) = H(i_{k} - 1, j_{k}) - \xi_{2} \text{ and } \sigma_{k} = \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k}, j_{k}) = H(i_{k} - 1, j_{k}) - \xi_{2} \text{ and } \sigma_{k} = \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k}, j_{k}) = H(i_{k} - 1, j_{k}) - \xi_{2} \text{ and } \sigma_{k} = \langle i_{k} - 1, j_{k} \rangle \text{ and } \sigma_{k} = \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k}, j_{k}) = H(i_{k} - 1, j_{k}) - \xi_{2} \text{ and } \sigma_{k} = \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k}, j_{k}) = H(i_{k} - 1, j_{k}) - \xi_{2} \text{ and } \sigma_{k} = \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k}, j_{k}) = H(i_{k} - 1, j_{k}) - \xi_{2} \text{ and } \sigma_{k} = \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k}, j_{k}) = H(i_{k} - 1, j_{k}) - \xi_{2} \text{ and } \sigma_{k} = \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k}, j_{k}) = H(i_{k} - 1, j_{k}) - \xi_{2} \text{ and } \sigma_{k} = \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) = H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if } H(i_{k} - 1, j_{k}) + \langle i_{k} - 1, j_{k} \rangle \text{ if }$$

$$(i_k, j_k - 1)$$
 if $H(i_k, j_k) = H(i_k, j_k - 1) - \xi_2$ and $\sigma_k = \uparrow$

The rationale of the procedure is to take the best alignment of the two halves of each genome window considered and identify the windows where the pairing potential is locally maximal with respect to a normalised score.

The normalised score for a best local alignment $\Lambda = (e_1, e_2, \sigma)$ of a split string is defined as:

$$s(\Lambda) = \begin{cases} \frac{2H(e_1, e_2)}{e_1 + e_2} & \text{if } e_1 + e_2 > 0\\ 0 & \text{otherwise} \end{cases}$$

The adopted score not only normalises the score of the best alignment with respect to the alignment length, but it also privileges a base pairing closer to the midpoint of the genome window under consideration. We can now define what is a candidate position in the genome.

Consider a chromosome of a given genome. Let S_p be the sequence of length w starting at position p of the said chromosome, and let Λ_p be the best local alignment in S_p . We have that S_p is a candidate sequence iff the normalised score is locally maximal at S_p , i.e.,

1. $\exists \hat{p} : \forall p' : \hat{p} < p' \le p \implies s(\Lambda_{\hat{p}}) < s(\Lambda_{p}) = s(\Lambda_{p'})$ 2. $\exists \hat{p} : \forall p' : \hat{p} > p' \ge p \implies s(\Lambda_{\hat{p}}) < s(\Lambda_{p}) = s(\Lambda_{p'})$

having $s(\Lambda_{\hat{p}}) = 0$ for every $\hat{p} < 0$ or $\hat{p} > |S| - w + 1$.

As several candidate positions may be identified in contiguous co-ordinates in the genome, presumably for each window whose midpoint falls within the terminal loop portion of the stem-loop, we aggregate them together in candidate regions as they will refer to the same stem-loop structure.

Let R_p^l be a region of length $l \ge w$ starting at position p of a chromosome S of a given genome. R_p^l is a candidate region iff S_p, \ldots, S_{p+l-w} are candidate positions and $S_{p-1}, S_{p+l-w+1}$ are not.

We have chosen a window length of 200, which approximately corresponds to the length of the largest annotated metazoan precursor sequence and is wide enough to accommodate the vast majority of known animal pre-miRNAs, and we have adopted a scoring model such that $\xi_0 = \xi_1 = \xi_2 = 1$. The choice of parameters for the model is important since it may affect the identity and amount of candidate regions identified. Our scoring model was based on three observations. First, most DNA alignment methods prefer a linear model for gaps and an equal penalty for gaps and mismatches [1]. Second, miRNA precursors necessarily exhibit gaps and mismatches when aligning their stem portion due to the ubiquitous yet small bulges and inner loops which justifies that the penalty for a gap/mismatch is the same as the contribution of matches. Finally, small variations in the scoring model did not produce significantly different results, whereas more radical departs from the adopted model, such as having mismatches or gaps negatively contributing to the alignment score more than twice the contribution of a match, did have an impact on the sensibility (data not shown).

Having identified the candidate regions, these are folded using RNAfold with standard parameters and the largest stem-loop structure contained therein is extracted and re-folded. The final set of precursor candidates is made up of these refolded stem-loops restricted to those which exhibit a minimum free energy no higher than -20 Kcal/mol and with both stem arms at least 16-nt long, since these parameters will capture the vast majority of known pre-miRNAs while significantly reducing the number of candidate stem-loops. The set of candidates is subjected to an additional filtering step in order to identify different candidates with identical terminal loop co-ordinates in which case only the longest candidate is retained.

Table 1 - Homologs to pre-miRNAs of A. gambiae identified amongst the precursor candidates of A. darlingi

The Table shows the homologs to pre-miRNAs of *A. gambiae* identified amongst the precursor candidate sequences of *A. darlingi*, their position in the *A. darlingi* dataset and the E-value and identity percentage of the Blastn hit.

miRNA	Contig	Strand	Start	Stop	Length	E-value	Identity
aga-mir-281	ctg7180000455710	F	12304	12396	93	6e-44	98.92
aga-mir-137	ctg7180000423045	\mathbf{F}	24973	25062	90	3e-42	98.89
aga-mir-125	ctg7180000436522	R	30069	30161	93	2e-41	97.85
aga-mir-9c	ctg7180000436657	\mathbf{F}	27474	27563	90	8e-40	97.78
aga-mir-iab-4	ctg7180000409079	R	9295	9378	84	1e-38	98.81
aga-mir-278	ctg7180000393996	R	2139	2222	84	3e-36	98.81
aga-mir-8	ctg7180000296739	R	14269	14350	82	5e-35	97.56
aga-mir-957	ctg7180000502071	\mathbf{F}	10129	10209	81	2e-34	97.53
aga-mir-1175	ctg7180000395096	R	20239	20316	78	1e-32	97.44
aga-mir-305	ctg7180000409513	R	69	156	88	5e-32	95.51
aga-mir-9a	ctg7180000394369	R	31986	32065	80	2e-31	97.50
aga-mir-79	ctg7180000436657	\mathbf{F}	29361	29431	71	8e-31	98.59
aga-mir-263b	ctg7180000380439	\mathbf{F}	18574	18666	93	8e-31	94.74
aga-mir-927	ctg7180000364658	R	2214	2301	88	1e-29	94.32
aga-mir-1891	ctg7180000436657	\mathbf{F}	69525	69616	92	5e-29	92.39
aga-mir-1000	ctg7180000409240	\mathbf{F}	26262	26333	72	5e-29	97.22
aga-mir-929	ctg7180000436895	R	35399	35472	74	7e-28	95.95
aga-mir-993	ctg7180000380779	R	2989	3093	105	1e-27	90.57

Table 1:

Continued on Next Page...

miRNA	Contig	Strand	Start	Stop	Length	E-value	Identity
aga-mir-307	ctg7180000501812	F	90845	90913	69	3e-27	97.10
aga-mir-7	ctg7180000456148	R	10858	10934	77	3e-27	94.81
aga-mir-283	ctg7180000394200	R	11363	11450	88	1e-26	92.05
aga-mir-14	ctg7180000394624	\mathbf{F}	29822	29905	84	1e-26	94.05
aga-mir-210	ctg7180000325517	R	378	447	70	1e-25	95.71
aga-mir-92b	ctg7180000502202	\mathbf{R}	52014	52091	78	2e-25	93.59
aga-mir-190	ctg7180000409440	\mathbf{F}	26813	26893	81	8e-25	93.83
aga-mir-184	ctg7180000395192	\mathbf{F}	12643	12726	84	2e-24	94.05
aga-bantam	ctg7180000380411	\mathbf{F}	3920	4019	100	3e-24	93.07
aga-mir-263	ctg7180000422962	\mathbf{R}	6187	6272	86	2e-22	95.35
aga-mir-277	ctg7180000436725	\mathbf{F}	2857	2947	91	3e-21	90.11
aga-mir-124	ctg7180000394913	\mathbf{F}	3659	3737	79	9e-21	90.36
aga-mir-10	ctg7180000299625	\mathbf{F}	88710	88792	83	7e-19	89.16
aga-mir-13b	ctg7180000296969	\mathbf{F}	18851	18926	76	3e-18	92.21
aga-mir-988	ctg7180000423020	\mathbf{R}	24356	24424	69	3e-18	92.86
aga-mir-276	ctg7180000394910	\mathbf{R}	4307	4390	84	1e-17	89.41
aga-mir-219	ctg7180000456051	\mathbf{R}	64129	64209	81	4e-17	89.41
aga-mir-282	ctg7180000297228	\mathbf{F}	31847	31926	80	6e-16	90.12
aga-mir-9b	ctg7180000436657	\mathbf{F}	29840	29918	79	4e-14	87.21
aga-mir-1890	ctg7180000297175	\mathbf{R}	19896	19966	71	4e-14	90.14
aga-mir-317	ctg7180000381136	\mathbf{F}	4237	4319	83	2e-12	86.90
aga-mir-275	ctg7180000409513	\mathbf{R}	6089	6155	67	2e-12	89.71
aga-mir-87	ctg7180000358126	\mathbf{R}	296	383	88	1e-11	85.56
aga-mir-308	ctg7180000296848	\mathbf{R}	3648	3718	71	1e-11	88.73
aga-mir-279	ctg7180000436869	\mathbf{R}	24027	24088	62	2e-09	88.89
aga-mir-92a	ctg7180000502202	\mathbf{R}	73930	73990	61	9e-09	88.52

Table 2 - Alignment of mature miRNAs from $A. \ gambiae$ against precursor homologs identified amongst pre-miRNA candidates from $A. \ darlingi$

The Table shows the alignment of mature miRNAs from *A. gambiae* to the precursor homologs identified amongst the precursor candidate sequences of *A. darlingi*.

miRNA	Alignment	Identity
aga-mir-281	5 ⁷ auccaaugugaaaauaagagagcuauccgucgacaguagggauauaauucacugucauggaauugcucucuuuauguacaauucgauauaa ³⁷ 5, ugucauggaauugucau ₃ 3,	100.00
aga-mir-137	⁵ [,] AAAACUUGGUUGGCCACGCGUAUUCUUGGGUUACUAACACACUAUUUAUGUUGUUGGUGAGAAUACACGCUAGUUGACUAGUGU ³ , ₅ , UAUUGCUUGAGAAUACGUGAGAUAG3,	100.00
aga-mir-125	⁵ [,] GUAUCUGCUGAUUCCCUGAGACCCUAACUUGUGACUAUCGUUACAAGUUUUGAUCUCCGGUAUUAGCGGUUGAGAUGCGACGG ³⁷ 5,UCCCUGAGACCCUAACUUGUGA ₃ /	100.00
aga-mir-9c	⁵ / UUUCCGGCUGUGUGUUUGGUAUUCUAGCUGUAGAAUGUUGUUUGAUUGUAUAUUCUCUAAAGCUUUAGUACCAGGGGGGAA ^{3/} 5. UCUUUGGUAUUCUAGCUGUAGA3.	100.00
aga-mir-iab-4	⁵ / GUGCCGCUUCAUGAACGUAUGUAUGUAUCCUGGUGCUACUUAUCCGGUAUACGUAUACGUACAGGAGGGGGGGG	100.00
aga-mir-278	5' geuaceguaceguacegacegauagucuucaaceguucaceguucaceguegegaceguegegaceguuceguegegaceguugaagece $3'$	100.00
aga-mir-8	⁵ [,] guouguucacaucuuaccegecacauacuuuuaucegauacuucuaauacueucageuaaagaugucegecegacce ³ [,] ₅ , uaauacuegeucageuaagauguc ₃ ,	100.00
aga-mir-957	⁵ / acuacaaacauuauuuuaaacaaauuuuaauuaauuaaaauuuaaaaacaaucaaaacuaaaacaaaaaa	100.00
aga-mir-1175	5' GAUAUGGAAUAAGUGGAGUAGUCGUCAUCGCUUAGUUUUAGAAAAGUGAGAUUCUCGCGACUUAAUUCAUA $3'5'$ UGAGAUUCUAGUUCCGGACUUAA $3'$	100.00
aga-mir-305	^{5/} UUUGUCACAUGUCUAUUGUACUUCAGGUGCUCUGGUGGAUUUCAGAAACCCGGCACAUGUUGGAGUACACUUAAUGUGCUGACAA ^{3/} ₅ / AUUGUACUUCAUCAGGUGCUCUG ₃ /	100.00
aga-mir-9a	⁵ / GUCAAUGUUCCUUUGGUUAUCUAGCUGUAUGAGUGUAUUUAAAAACGUCAUAAAGCUAGCAUACCGAAGUUAAUAAUG ³ / ₅ / UCUUUGGUUAUCUAGCUGUAUGA ₃ /	100.00
aga-mir-79	⁵ , gcuuuggcgcguunagcuguugaugauuugaaguuuugaagguuggaugauuaccaaaggauagacgaa ³ ₅ , uaaagguagauuaccaaaggau ₃ ,	100.00
aga-mir-263b	$^{5'}$ ucaccaauuaucgaccuucgocacaaaaauucacgugaucguacguucuucuucgugaaucuuuucgugccaucguucgu	100.00
aga-mir-927	⁵ / GUUAAUGGUUCGUUUUAGAAUUCCUACGCUUUAACUUAAAGUAGUGCGGCAAAGCGUUUGGAUUCUGAAACGAACAUUAA ³ / 5/ UUUAGAAUUCCUACGCUUUACC3/	100.00
aga-mir-1891	^{5,} UCUUUUUCUGUCAUGUGAGUUAAUUUGGGUGUUUUUGCAUAGGAUUAACAGGUCCAUUAACUCGUACAUGAUGGAAAAACGGAGC ³⁷ ₅ , UGAGGAGUUAAUUUGGGUGUUUU ₃ ,	100.00

miRNA	Alignment	Identity
ga-mir-1000	⁵ / GUCGAUGAUAUNGUCCUGUCACAGCAGUACUAUUUCGCUAGCUUACUGUUGUUUCGGGACAUUUUCCAUCGAC ³ /	100.00
ga-mir-929	5, ADAUUGACUAUAUAUAUAUAUAUAUAU ⁵ UGGGAUUAAAUUGACUCUAGGGAGUCCUUCCAUACGGAGAGUCCCUAACGGAGUCAGAUUGAUU	100.00
ga-mir-993	$5'\ldots$ ceueaccueuceureureegecuuureugegeuugaaauacaaaacauguaaauucauauucueuuuuaucagaageuceuuucuauagageuaucuca $3'$	100.00
ga-mir-307	⁵ ucucuceauuacucaaccuegeugugaugauugaaucaucaaccuccuugagugagega ³ ₅ , ucacaaccucgugaggagaga	100.00
ga-mir-7	5' uuguauggaagagagugauuuuguuguuguuguuguuguu	100.00
ga-mir-283	⁵ / UUCGACUGAAGGUAAAUAUCAGCUGGUAAUUCUAGGCUAUCUAAACUUCGUGCACCCCGGAAUUUCAGCUGAUAUCCGU ³ / ₅ , UAAAUAUCAGCUGGUAAUUCU ₃ /	100.00
ga-mir-14	⁵ ccccgauaagccugugggagcgagauuaaggcuuguugaaguuuugaacuuuuggaguuuuugucucucuc	100.00
ga-mir-210	⁵ / cauugcagcuggaccacugcacaagauuagaauaugacucuugugggugacaacgggugugggg ^{3/} 5/ Uugugggugacaagggguaa;	100.00
ga-mir-92b	⁵ GGGCUCCGGAUGUAGGGCGUGACUUGUGCAUAAUUUGCUGAUGUCAAUGUCAAUUGCACUUGUCCCGGCCUGCAGC ³ 5, AAUUGCACUUGUCCGGGCUGC3,	100.00
ga-mir-190	⁵ / UUUCGGUAAGAUAUGUUUGAUAUUCUUGGUUGUUAAAUUGUUCAAUUAUCACCCAGGAAUCAAACAUAUUAUUACUGUGAC ³ / ₅ , AGAUAUGUUUGAUAUUCUUGGUUG3,	100.00
ga-mir-184	⁵ GGUGCAACCCUUAUCAUUCUUUCGCCCCGUGUGCAUUGCGAACCGAACCGGAGAACUGAUAAGGGCCCGGGUCACC ³ 5 · UGGACGGAGAACUGAUAGGGCCCGGGUGUGCAUUGCGAACGGAGAACUGAUAAGGGCCCGGGUCACC ³	100.00
ga-bantam	⁵ [/] AAAUGUAAUCACAGAACCGGUUUUCAUUUUCGAUCUUAUUCAUUUUACAACGAGUGAGAUCACUUUGAAAGCUGAUUUUGUACAGGUUAACUCAACG ^{3/} ₅ , UGAGAUCACUUUGAAAGCUGAUU3,	100.00
ga-mir-263	5' cccugguacauguadauggcacuggaagaauucacgggauuucaauacuccguguucucuuaguggcauaccggga $3'$ $5'$ uguaauggcacugggagaagaauucac $3'$	100.00
ga-mir-277	5 (suuuugegeuaceueucagaagugcauuuacaucgecauuucgageuauuuuguaaugeacuaucugeuacgacacauuccagaau $^{3'}$	100.00
ga-mir-124	⁵ / counturcedequenteredecededatecededatecededecedecedecedededededecededededede	100.00
ga-mir-10	⁵¹ UNAUGUUCUACAUCUACAUCUAGAUCCGAAUUUGUUUGAAAUUUAACAAGCGACAAAUUCGGUUCUAGAGAGGUUUGUGUGG ³⁷ ₅₁ ACCCUGUAGAUCCGAAUUUGU ₃₁	100.00
ga-mir-13b	⁵ , uccucedeucegauceuaaaaauceuucuceuceauacuuaceaaaguucaucaucaceceauuuuucaceaguu ³ , ₅ , uaucacegeceau ₃ ,	100.00

miRNA	Alignment	Identity
aga-mir-988	5' cocoucueucaceuraucaceuraceuraceuraceuraceura	100.00
aga-mir-276	⁵ ⁷ GOUGAUUGCCAUCAGCGAGGUAUAGAGUUCCUACGUUGUUCAUAUGAAAUCUGUAGGAACUUCAUACCGUGCUCUUGGAUAGCC ³⁷ ₅ , UAGGAACUUCAUGCGUGUUGAUAGGUUCCUGUGUUCAUAGGAACUUCAUACCGUGCUCU ₃ ,	100.00
aga-mir-219	⁵ / UUUCUAGCUCUGAUUGUCCAAACGCAAUUCUUGUUGAUACCAUUAGCUACUCAAGAGUUGUGACUGGACAUCCGUGGCUCG ^{3/} ₅ , UGAUUGUCCAAACGCAAUUCUUG ₃ /	100.00
aga-mir-282	⁵ / cuaaucuaeccucuccuaeecuuueucuguaauucacaauccaeacaacaeacuaeccueacaeaeuuaeeucaeaaucue ^{3/} 5/ aaucuaeecucuueeecuuueucueu ₃ /	96.43
aga-mir-9b	$5'$ cacuvauregeucuvurgeugauvuvuageuguvuvucuvucacavavageuvuaucaccaaaaaccuaaveug $3'$ $_5'$ cacuvurgeugaugeugaugageugaugageugaugageuga	95.65
aga-mir-1890	5' cagageuranuugeageanuucuugaaganauuuuuuuuuuuuuuu $5'$ cagageurageuuageuu $3'$	100.00
aga-mir-317	5' cucueccecuescauacaccuususcauuuscaauuesaauacucaagueaacacaucuesuusucucaeuesces $3'$ $5'$ ucaecacuesuscucacuesus $3'$	100.00
aga-mir-275	5' cececuaagcaggaaccoggacuugauccauuuugcaaacaguaccugagguaccugagguagggaggg	100.00
aga-mir-87	5' gauugcucccggcccrggccugaaauuugcuaaaccugcugcauaugaggaaaaggugaggaaaauauuucaggugugugguggug $3'$ $5'$ gauuggugaggaaaauauucaggugug $3'$	100.00
aga-mir-308	5' uguuucgcaguauauuuuugugaguuuguuguuguuuuuuuu	100.00
aga-mir-279	$^{5'}$ aaugegugaaucuagugguucacauguuuuugcuaguggguggacuagauccacacucauuaa $^{3'}$	100.00
aga-mir-92a	$5'$ ucgecugaaucaaggecaaaauuguguuuugauaccaauauuugacacuuguccegeccuau $3'$ $_5'$ ucgecugaugi	100.00

References

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