

Identification and Expression of Small Non-Coding RNA, L10-Leader, in Different Growth Phases of *Streptococcus mutans*

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Streptococcus mutans is one of the major cariogenic bacteria in the oral environment. Small non-coding RNAs (sRNAs) play important roles in the regulation of bacterial growth, stress tolerance, and virulence. In this study, we experimentally verified the existence of sRNA, L10-Leader, in *S. mutans* for the first time. Our results show that the expression level of L10-Leader was growth-phase dependent in *S. mutans* and varied among different clinical strains of *S. mutans*. The level of L10-Leader in *S. mutans* UA159 was closely related to the pH value, but not to the concentrations of glucose and sucrose in culture medium. We predicted target mRNAs of L10-Leader bioinformatically and found that some of these mRNAs were related to growth and stress response. Five predicted mRNA targets were selected and detected by quantitative reverse transcriptase-polymerase chain reaction (qRT-PCR), and we found that the expression levels of these mRNAs were closely related to the level of L10-Leader at different growth phases of the bacteria. Our results indicate that L10-Leader may play an important role in the regulation of responses in *S. mutans*, especially during its growth phase and acid adaptation response.

Introduction

IN PROKARYOTES, SMALL RNAs include transfer-messenger RNA, 4.5S RNA, 6S RNA, and small non-coding RNA. In bacteria, small non-coding RNAs are generally referred to as small RNAs (sRNAs) and range from 50 to 500 nucleotides (nt) in length. These sRNAs are transcribed but not translated. They used to be hard to detect until recently, when a growing number of bacterial sRNAs have been found, some of which have been shown to regulate critical cellular processes. The first bacterial sRNA was found in *Escherichia coli* (Wassarman et al., 1999), and subsequently, sRNAs were also found in numerous Gram-positive bacteria (Mraheil et al., 2010), such as *Listeria monocytogenes*, *Staphylococcus aureus*, *Streptococcus pyogenes*, and many more. To date, sRNAs have been found to be growth phase dependent and to vary in response to growth medium (Christiansen et al., 2006). They have been implicated in the stress response, iron homeostasis, outer membrane protein biogenesis, sugar metabolism, and quorum sensing, suggesting they might play an essential and central role in the pathogenicity of many bacteria. In *L. monocytogenes*, a 70-nt sRNA (*sbrA*) was identified, which was reported to be under

the control of stress tolerance factor (Nielsen et al., 2008). In *S. aureus*, the expression of most of the sRNAs was growth phase dependent and transcription was in response to environmental changes, such as an acidic pH (Geissmann et al., 2009).

Major sRNAs interact by pairing with their target mRNAs, acting as antisense RNAs. Some sRNAs form parts of RNA-protein complexes with the involvement of chaperone protein, Hfq. A few sRNAs regulate by binding with a target protein (Vogel and Wagner, 2007). Many sRNAs, such as the MicF, OxyS, DsrA, Spot42, and RyhB RNAs, act by base pairing to activate or repress translation, or to destabilize mRNAs. Major groups are based on their mode of action; these are *cis*- and *trans*-acting regulatory sRNAs. Riboswitches and RNA thermometers (Narberhaus et al., 2006) belong to the class of *cis*-acting regulatory RNAs, which are located at the 5'-untranslated region (UTR) of their genes. Riboswitches modulate their regulatory structure in response to metabolite binding, which are present in the shared environment. RNA thermometers are involved in sensing global signals. Both *cis*- and *trans*- regulatory RNA structures are essential for fine regulation systems, to ensure an immediate physiological response within the bacterial cell to a variable

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environment. As the function of sRNAs becomes increasingly clear, there is also the prospect of broad applications of sRNAs, such as in the design of novel potential therapeutics based on sRNA-complementary peptide nucleic acids (Karkare and Bhatnagar, 2006) and developing a novel ultrasensitive diagnostic system, which has been conceived for the detection of small target samples (Mraheil et al., 2010).

Streptococcus mutans is a Gram-positive facultative anaerobic bacterium. It is the main etiological agent in the development of dental caries (Gibbons and van Houte, 1975). *S. mutans* possesses a variety of virulence factors that enable it to accumulate in large numbers, colonizing the tooth surface. It uses a wide array of carbohydrate sources, and is able to produce and tolerate acid, being able to survive at low pH. It adheres to the surfaces of teeth, which results in the demineralization of tooth surfaces (Burne, 1998). The regulatory mechanism of acid adaptation and the virulence factors required are not clear, despite much research having been performed. To date, there has not been confirmation of the sRNAs present in oral bacteria. With a greater knowledge of sRNAs in oral bacteria, especially in *S. mutans*, we will better understand oral infectious diseases, such as dental caries.

This study was carried out using bioinformatic approaches to predict sRNAs in *S. mutans*, followed by confirmation using experimental techniques. The sRNA sequence was found through the analysis of the Rfam website (<http://rfam.sanger.ac.uk/>), and annotated as L10-Leader (RF00557) family. We detected all 40 sRNAs predicted by at least 2 software application methods, including the software and the web page by quantitative reverse transcriptase-polymerase chain reaction (qRT-PCR). The transcript level of L10-Leader was relatively high compared with the other sRNAs and the changes were more significant. So L10-Leader was chosen for the further study. We detected the levels of L10-Leader in different growth phases, different clinical strains, and different growth media. Additionally, we predicted 5 target mRNAs for L10-Leader computationally. This study was the first to attempt to identify sRNA in oral bacteria, and the first to find sRNA-L10-Leader in *S. mutans*.

Materials and Methods

Strains and growth media

Streptococcus mutans UA159 and 2 clinical strains of *S. mutans* were used in this study. All *S. mutans* strains were routinely grown in brain heart infusion (BHI) medium (Becton Dickinson) or Todd-Hewitt broth (Becton Dickinson) supplemented with 1% yeast extract at 37°C in a 5% CO₂ atmosphere. For growth studies, cultures were grown overnight in BHI broth and then diluted 1:50 into fresh BHI broth and grown to the necessary growth phase [optical density at 600 nm (OD₆₀₀)=0.2, 0.4, 0.5, 0.6, 0.8, 1.0, and 1.5]. Tryptone yeast extract medium (TYE) (10% tryptone, 5% yeast extract, 17.2 mM K₂HPO₄) was utilized for acid shock. For acid shock assays, HCl was added to TYE medium to adjust final the pH to 5.5 or 7.5. For sugar concentration assays, 0.1% or 1% glucose and 0.5% or 1% sucrose were added to TYE.

For acid shock assays, overnight cultures of *S. mutans* UA159 were grown in THYE and diluted 10-fold in TYE supplemented with 0.5% glucose at pH 7.5 and incubated at 37°C until cells reached the mid-logarithmic phase (OD₆₀₀=0.4). Cell cultures were then divided into 2 aliquots, pelleted by centrifugation,

and resuspended in TYE supplemented with 0.5% glucose at pH 5.5 or 7.5. The cells were subsequently incubated for 0.5, 1.5, 2.5, 3.5, or 4.5 hours at 37°C, harvested by centrifugation, and utilized for RNA isolation.

For different sugar concentration assays, overnight cultures of *S. mutans* UA159 were grown in THYE medium and later diluted 1:20 using sterile TYE supplemented with 0.1% or 1% glucose and 0.5% or 1% sucrose for 24 hours at 37°C, harvested by centrifugation, and utilized for RNA isolation.

Bioinformatic identification of sRNAs in *S. mutans* UA159 genome

The genome of *S. mutans* UA159 (NCBI No.: NC_004350) was analyzed with sRNA Predict, sRNASVM, and SIPHT software for bacterial genomic sRNA prediction. SIPHT (sRNA identification protocol using high-throughput technology) and sRNAPredict predict candidate sRNA-encoding loci, identified based on the presence of putative Rho-independent terminators downstream of conserved intergenic sequences, and each locus is annotated for several features, including conservation in other species, association with one of several transcription factor binding sites and homology to any of over 300 previously identified sRNAs and *cis*-regulatory RNA elements (Livny, 2007; Livny et al., 2008). sRNASVM is a model for the prediction of small non-coding RNAs in *E. coli* using support vector machines (Saha and Raghava, 2006). The web page "Oral Pathogens Non-Coding Small RNA Prediction" (Los Alamos National Laboratory Oral Pathogen Sequences Databases; http://www.oralgen.lanl.gov/oralgen/bacteria/analysis/srna_result/index.html) also predicted the sRNA of *S. mutans*.

qRT-PCR

Protocols for the isolation of RNA from *S. mutans* call for the pre-treatment of bacteria with lytic enzymes in TE buffer. Total RNA was then extracted using Trizol Reagent (Invitrogen) according to manufacturer's instructions. Contaminated genomic DNA was removed by DNase I (Promega). RNA was reverse transcribed using M-MLV reverse transcriptase (Promega), according to the manufacturer's protocol. Quantitative PCR was performed according to the protocol of the SYBR Green qPCR Master Mix (2×) (Fermentas) with Mx3000p (Stratagene); 16S mRNA levels were used for normalization. The experiment was repeated at least 3 times.

The primer sequences were: 16S forward: CTTACCAG GTCTTGACATCCCG; 16S reverse: ACCCAACATCTCAC GACACGAG; L10-Leader forward: CTAAGACAGCAGG GGAGCGT; L10-Leader reverse: GTCTGGGTTTGGGTGC GATT; T7 promoter sequence: TAATACGACTCACTATA GGG; M1 forward: AAAAGCCCCCTCTTGCAATG; M1 reverse: TCTGTGACATCTGGTTGAAAAGT; M2 forward: GTGGCGATCCAACCTGGTACT; M2 reverse: TAGGCTGCT GCAATTTCTT; M3 forward: CGTCTTCTCAACAACC AGCA; M3 reverse: TTTGAACGCCCATAAAGTCC; M4 forward: TGACCGTGGTGACAGGTCTA; M4 reverse: ATC CGCCTTTCCGATAAGTT; M5 forward: CCGCAAATTC TAGCGACTC; M5 reverse: TCGTCCACACGCAAATACAT.

Northern blotting

Overall, 50 µg total RNA samples were analyzed by northern hybridization. The *in vitro* transcription was carried

out with T7-RNA polymerase (Fermentas) on PCR fragments predicted by bioinformatics, which were constructed to contain the T7 promoter as a 100-nt marker. To prepare the hybridization probe, the L10-Leader gene was labeled with [α - 32 P] using Random Primer DNA Labeling Kit Ver. 2 (TaKaRa). sRNA was run on a 2% agarose gel containing 2.2M formaldehyde and transferred to nylon membrane (Hybond). RNA was linked to the membrane under the ultraviolet light for 15 minutes. Hybridizations with strand-specific DNA probe in Perfect HybTM Hybridization Solution (TOYOBO) were carried out according to the hybridization solution manufacturer's instructions; 16S ribosomal RNA (rRNA) was used as a loading control. The experiment was repeated at least 3 times.

Bioinformatic prediction of target mRNAs of L10-Leader target in *S. mutans* UA159

The genome of *S. mutans* UA159 (NCBI No.: NC_004350) and L10-Leader sequence were analyzed with web services sTarPicker and RNAPredator. sTarPicker (<http://ccb.bmi.ac.cn/starpicker/index.php>) is based on complementary base-pairing and energy calculations (Ying et al., 2011). RNAPredator (http://rna.tbi.univie.ac.at/RNAPredator2/target_search.cgi) uses a dynamic programming approach to compute putative targets (Eggenhofer et al., 2011).

Results

Bioinformatic identification of sRNAs in the *S. mutans* UA159 genome

The genome of *S. mutans* UA159 (NCBI No.: NC_004350) was analyzed with sRNA Predict, sRNASVM, and SIPHT prediction software, and the web page "Oral Pathogens Non-Coding Small RNA Prediction" was also used for the prediction of the sRNAs. From the 4 methods of prediction, the total number of sRNAs was 334, of which 40 sRNAs were predicted by at least 2 of prediction methods, including the software and the web page (Table 1; Fig. 1).

Identification of L10-Leader in *S. mutans* UA159

We detected all 40 sRNAs predicted by at least 2 software applications by qRT-PCR. The transcript level of L10-Leader was relatively high compared with the other sRNAs and the changes were more significant. Therefore, L10-Leader was chosen for the further study.

The lowest-energy secondary-structure for the L10-Leader sRNA was predicted in *S. mutans* (Fig. 2a). The *S. mutans* UA159 was grown to 2 typical phases: mid-logarithmic growth phase ($OD_{600}=0.5$) and stationary growth phase ($OD_{600}=1.5$). We detected L10-Leader by northern blotting (Fig. 2b). The northern blot result was consistent with the qRT-PCR result (Fig. 2c). The transcript level of L10-Leader was lower in the stationary growth phase (Fig. 2c).

S. mutans UA159 was grown in BHI to different growth phases: early-logarithmic growth phase ($OD_{600}=0.2$), mid-logarithmic growth phase ($OD_{600}=0.4, 0.6$), late-logarithmic growth phase ($OD_{600}=0.8, 1.0$), and stationary growth phase ($OD_{600}=1.5$). The levels of L10-Leader were gradually downregulated from logarithmic growth phase to stationary growth phase (Fig. 2d).

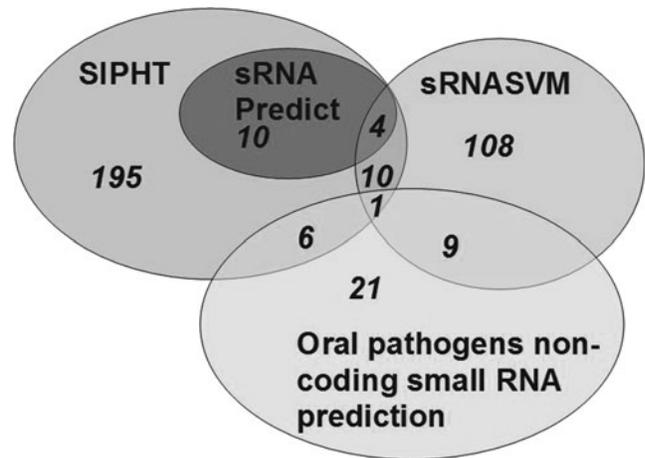


FIG. 1. Venn diagram shows the overlap between sRNA Predict, sRNASVM, SIPHT (small non-coding RNA identification protocol using high-throughput technology), and the web page (Oral Pathogens Non-Coding Small RNA Prediction) for the prediction of the small non-coding RNAs (sRNAs). The number of sRNAs is shown in the ellipses and overlapping sections.

L10-Leader in different clinical strains of *S. mutans*

Two clinical strains of *S. mutans* were selected to analyze L10-Leader. Strain 1 has stronger adherence, more synthesis of extracellular polysaccharides, and produces more acid than strain 2. We compared the expression of L10-Leader in these 2 strains at different growth phases: logarithmic growth phase ($OD_{600}=0.5$) and stationary growth phase ($OD_{600}=1.5$). The levels of L10-Leader were higher in strain 1 than strain 2, regardless of which stage they were in (Fig. 3a, b). The 2 clinical strains were cultured to early-logarithmic growth phase ($OD_{600}=0.2$), mid-logarithmic growth phase ($OD_{600}=0.5$), late-logarithmic growth phase ($OD_{600}=1.0$), and stationary growth phase ($OD_{600}=1.5$). Levels of L10-Leader were gradually downregulated from logarithmic growth phase to stationary growth phase (Fig. 3c, d), similar to the effect seen in *S. mutans* UA159.

Expression of L10-Leader under different environmental stresses

S. mutans is able to adapt to acidic environments, using glucose for acid production and sucrose to synthesize extracellular polysaccharides for adhesion to the tooth surfaces. The expression of sRNA is related to the environmental stress response, so we examined L10-Leader of *S. mutans* UA159 under stress conditions. For acid shock assays, *S. mutans* UA159 were grown in TYE medium at pH 7.5 until cells reached the mid-logarithmic phase ($OD_{600}=0.4$). Cell cultures were then divided into 2 aliquots, pelleted by centrifugation, and resuspended in TYE medium at pH 5.5 or 7.5. The cells were subsequently incubated for 0.5, 1.5, 2.5, 3.5, or 4.5 hours. The levels of L10-Leader were consistently higher in pH 5.5 than in pH 7.5, from 1.5 to 4.5 hours. There was little change in the level of L10-Leader at 0.5 hours (Fig. 4a).

When *S. mutans* UA159 was grown in 0.1% glucose or 1% glucose TYE medium for 24 hours, the levels of L10-Leader showed no significant differences (Fig. 4b). Again, there was

TABLE 1. THE NUCLEOTIDE SEQUENCES OF SMALL NON-CODING RNAs PREDICTED BY AT LEAST TWO OF THE PREDICTION METHODS, INCLUDING THE SOFTWARE AND THE WEB PAGE

Class	Nucleotide sequence
SIPHT, sRNA Predict, sRNASVM	TGCTCGGGAAAGAGGTGCGAGGAAGAGCTATTTCCGGTGGGCCTTTGGCCCTGACCGGTA GAAGTGGCTTACAGTCGCAGAACAAACCACAGTTCACACT TACCATAGCCCCTTTGAGAGGTAAGTGACGAGTCAAAAAGCAGTTAGGCTTGAACAAAGT GAAAGCCAGCGTCTTTAGGCGCTGGCTGGTGATGTGGGCTTATAGCCCTCGTCCAAAC CACCCGTTAGACGGGTGGTTATGA TAGATAAAAACGGATGGTACCGCGTGCAACGCTCCGCTTAAGGAGTTTTGGCACTTTTT GATTTTTTGAATAAAATGAGTTGCGGGTCTAGCCCCGAAACTGGGTGGTACCGCGGAGAT CATAGAAATGACATTCGTCCTGTCAATTGGCAGGGACGGTTTTTG GGCTCTAACGTAGCCACCGTTTTATGTTTACATAAAAAAATTAGGGAGTAGTATTAATGA TAGGCCTGCACTTTCTCAGCTGCCAAGTGCTCTCTTAAAG
SIPHT, sRNASVM, website ¹	TTTGAGACTTTCCTTAGGTGATACGGACGGTAGCGACTTTCCTTCGGAATTCCATACCTAA GATTTGAGCCTAAGGTCTCAAATCTCCGAGTGCCTGAAACTTTAAAGTTTCAGGCACT ACCACCCGCTATGCGGGTGCGCATTGAATGTTATACAAAAATCTCCAAACGCGATACAA TAAAGGTGTTCAAGCCTATTGTAAAGCGAAAGGAGCAAAAAATGGCACAAAAGGCACA TAGTTTATCACATACAA GAATATCGTTCACAAATGAACTGCACCCCAAAAGTTAGACAGAAAAAATCTAACT GTGTACTGACCCCAAAAAGTTAGACAAAAAATTTAAGTAAAG TTCAGAAAATTTCTGATTTCTATTGACATTCATTCTAAAAAAGGTATGATGTAAACAACA TACATCGAAAGGAAAAAGTCAATGTTTACACGAAATGACGCTTTGTTGTTGAGGTGT GAGGGCTAATGCAAAAAGCATTAGTCTGTTTGGCTTACCAAGCGGGAAGAAAAAA CATCTCGCTTGGGTTTCTGAGTGAGATGTTTTTT TTCAGAAAATTTCTGATTTCTATTGACATTCATTCTAAAAAAGGTATGATGTAAACAACA TACATCGAAAGGAAAAAGTCAATGTTTACACGAAATGACGCTTTGTTGTTGAGGTGT GAGGGCTAATGCAAAAAGCATTAGTCTGTTT AGAACTTCTGCTTTGGTGCTCAATATTAAGTGTAGTGGATGACTGCTAACTAAGCATGAGA GAGAATCAG CGTCCAAATTATTAGTCTTTCTGTAAACCTCGTTTCTCCGAAACGAGTTATGG TCATCAATTTTGTGAGTGTTATGATGTTGGAATTATAATAATTTCTAGTCTAGCTCTGGTC TACCACTAGGAGAGGCTCCTAGTGGAACTCATCATGTTGAAAAAGACAGCGTTTATT TAGGAAATACTTTAATAGTAGAAGCATTTCCTAAATAAACGCTTTTTG TAATCTGATAAAAATAAGGTCATTTCAGAGGTGTTTTGAATTCATTTGTTATAG GAAGCGGCGTTACTGGGAAATCCGCACAAATGTGCGAAACTGGTGTGATGT AATCAGCCTTTAGCTTTGATACAAGAGGTTGCGATACGCTCGGTTGCATTGCCACG CAAAAGCCTGTGCGTTTTCTGTGGAGCTAGCCTATTATTTAAATAGACGAGAGGA GAAAAGATGGCAAACAAAAAATCCGTATCCGTTTGA AAAATGCTGTATTGGCACCGATGATTCACGTTTTGTATTGAACCCGCACGCTTAAG CAGTGGGCAACTCGCTCTGACTTTAGTTGCTTCCGCGTGAAACGGCCTGCATACTGTCA GAAGACTTTTTGTTCCCGATAATACAAAAAATA TTTATCATGAGGAAAGTCCATGCTAGCACTGGCTGTGATGCCAGTAGTGTGTTGTGCTAGA CAAAAAAATAAGTCTAGGGATGTGCTTTGCGCATTACGGCGGATAAAATGGCTAAGTCTTT GATAGGCCGGAGTAATTCTGAAAGTGCCACAGTGACGTAGCTTTTATGGAAACA TAAAAGGTGGAACGCGGTAAACCCCTCAAGCTAGCAACCCAACTTTGGTAGGGGCATG GAAAAGCTGGAAAAAG ACTTGCATTTTTAACGTAATTTGTTAAAATAGTAAGGTAAGTTAGACTGTATGCC TACTGTCTATCTATAAAATATATTTTATTGGAGGCTTTTCCTAA CATCCTGTGCAACCGCACGAAAATCGCTTTAAGCAGCTTTCGTAGC <u>CTAAGACAGCAGGGGAGCGTGCTCATAATATTCTGCCGAGGCACAAAAACGTAAACT</u> <u>GAAAAACGTATTGTACTTTCTGTTCTGGGTTTGGGTGCGATT</u> TGACGGACTTCTGGAGAGACCTACTAGCGCCGAAGGGGCAAGGCTGTTTGTCAAACCTCT CAGGCAAAAAGGACAGAAAAGAAAAA TGTAAGACTGCGTTCAGTTGGAAGTGGTACAGGACCAGCAACGCTTGCACCTGTAC GAGTCCGTGTTTCTACGATTTTTCTGCCGCTGTGCAAGTGT TTTCTCCTCTCGTCTATTTAAAAATAATAGGCTAGCTCCACAAGAAAAACCGA CAGGCTTTTGGTGGCAATGCAACCGAGCGTATCGCAACCTCTTGTATC TGAATACGCCTACGACTGTGAAAAAGAGAAATACCTCCTAGATACCCCTGG TATCTTCGTTGGATTTCTAATTTTTCTTTGAGTTCGTTAACGGCTTAATATCTTA GGTATTGTTTCCCCATTTGAAAGGCCCGGAACCTTCAAATAATTTGTGGACCGGAA CATCCACATTGTAAACATAAA TCCCGTTAATTGAGGTGGCACCGTGTAGCTAGATTTAACGCCCTCACA
SIPHT, sRNA Predict	
sRNASVM, website	
SIPHT, website	

(continued)

TABLE 1. (CONTINUED)

Class	Nucleotide sequence
SIPH7, sRNASVM	TGATGTTGTCGTCAATCCGAAACTGTTTTCCGGCAACATATTAATAGCGAGACTTGTT
	TAATTATT
	TATTCCTTTAACACTGTCCCGTGAGGCAGGCAAGGAGCGACAAAGAAAAAAGTGCAGGAA
	CAACTGTACTTATATTTTGTGA
	AATAATTGTGTTGGAATCATTGAAACAACACAGCAAGTTAAAATAAGGCAGTGATTTT
	TAATCCAGTCCGTACACAACCTTGAAAAAGTGCGCACCGATTCCGGTGCTTTTTTAT
	GGAATCATTAAGATTCTAGGAGGACTCATCTTTTTGCCACAATCCCGAGCCCGTGGTCAGT
	TAAGATTTGAGCCTAAGGTCTCAAATCTTCCGAGTGCCT
	CTTTAGGCGCTGGCTGGTGATGTGGGCTTATAGCCCTCGTCCAACCACCCGTTA
	GACGGGTGGTTATGA
	TAGATAAAAACGGATGGTACCGCGTGTCAACGCTCCGCTTAAGGAGTTTTGGCACTTTTTT
	TGGCTCCTTGTCAACTGTAATAAGTGATTGATAAAGCGTGCGAGAGAATCAG
	GTGAAACAAATAAAAAAACTGGGAATTTAGATAAAAAAATCTAAAAAGATA
	CAAAGGCTGTTCTACTTGTGTCAATAACATCAAAAAGTCTGGAAAAAA
	ATACGAACCTTATAATTGCGAGCTGGCGAACAACGCGTAGCGGAAAGCTATCCATGAT
	GACGATATCGTCTAATCAAATGGCCCCGTCATGCCTTGAT
	ATAGCTTTTTTGATTTAAGAAACTCTTATTAGAAAATA
	ATGTAAAAAAATGTGTCCATAATAAAATTAATCATGATAAACTATATTATTATTTAAT
	GACATTAATAATAATAATTTATTTATGAGAGGAAAAA
	GGCTTATAGCCAGATAGCGTTCGTTCTGTGATGCTTTTCACCTCCTTGCATCGCAGCGCTG
CAGCCAGTACTGGAGTATGGCAAAACGAACACTAACGAAGT	
CTTTAGTTATAATACTATAACTTGTTTAAAAAATCAACACATTTTTGTTTATTTTTTAAA	
CATTTTTGGTATTATTTAAAAAATATTTCTTTTTCTG	

Sequence with bold and underline is L10-Leader.

¹⁷Oral Pathogens Non-Coding Small RNA Prediction" [Los Alamos National Laboratory Oral Pathogen Sequences Databases (ORALGEN); www.oralgen.lanl.gov/_index.html].

no significant change in levels of L10-Leader when the *S. mutans* strain UA159 was grown in 0.5% sucrose or 1% sucrose TYE medium for 24 hours (Fig. 4c). We propose that L10-Leader has little effect in glucose- and sucrose-starved environments.

Expression of predicted target mRNAs of L10-Leader target in *S. mutans* UA159

The possible target mRNAs of L10-Leader were predicted according to the bioinformatic prediction web services,

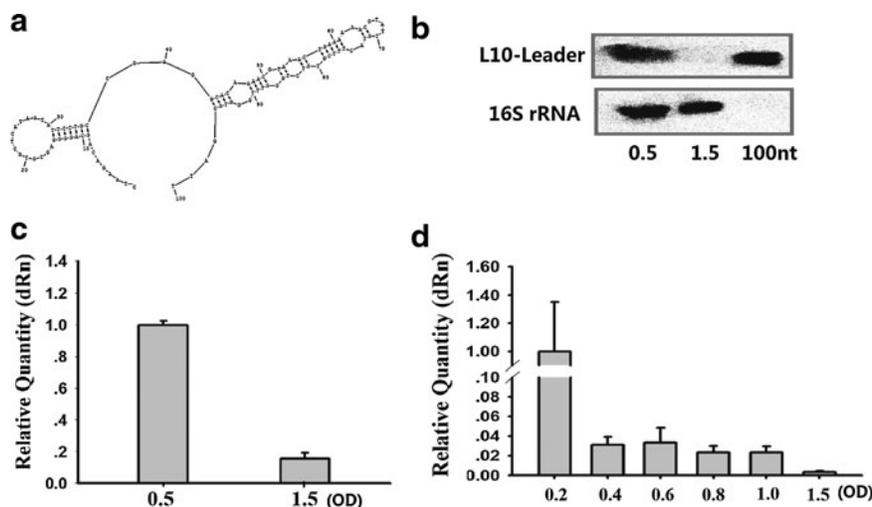
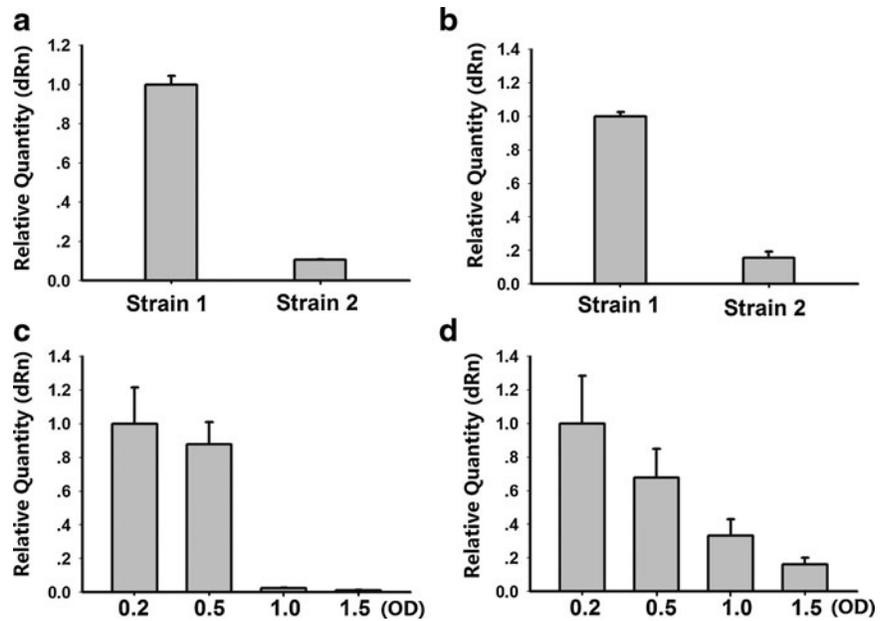


FIG. 2. Identification of sRNA L10-Leader in *Streptococcus mutans* UA159. **(a)** Lowest-energy secondary-structural predictions for the L10-Leader sRNA showing loop structure. **(b)** Northern blot analysis of L10-Leader at optical density at 600 nm (OD_{600}) = 0.5 and 1.5 during growth. The predicted size from the northern blot analyses, 100 nucleotides (nt), is shown on the right. 16S ribosomal RNA was used as a loading control. **(c)** qRT-PCR analyses of the transcript levels of L10-Leader in *S. mutans* UA159 grown to OD_{600} = 0.5 and 1.5. **(d)** *S. mutans* UA159 was grown in brain heart infusion to OD_{600} = 0.2, 0.4, 0.6, 0.8, 1.0, and 1.5, and qRT-PCR detected the transcript levels of L10-Leader. qRT-PCR, quantitative reverse transcriptase-polymerase chain reaction.

FIG. 3. qRT-PCR analysis of L10-Leader expression in clinical strains of *S. mutans*. **(a)** Two strains were grown to logarithmic growth phase, $OD_{600}=0.5$. **(b)** Two strains were grown to stationary growth phase, $OD_{600}=1.5$. The levels of L10-Leader were higher in strain 1 than strain 2, regardless of stage. **(c)** Strain 1 was grown to $OD_{600}=0.2, 0.5, 1.0,$ and 1.5 . **(d)** Strain 2 was grown to $OD_{600}=0.2, 0.5, 1.0,$ and 1.5 .



sTarPicker and RNAPredator. The numbers of mRNAs predicted by sTarPicker and RNAPredator were 100 and 4946, respectively. The 5 putative targets with highest scores in both prediction web services were chosen by us further investigation. They encode proteins SMU.276c and SMU.688, hydrolase SMU.488, thioesterase SMU.633 and transporter SMU.1149 respectively. L10-Leader and target sites of 5 target mRNAs predicted by sTarPicker (Fig. 5a). The results of qRT-PCR show that the transcription levels of mRNAs were lower in the stationary growth phase than in the logarithmic growth phase (Fig. 5b). These changes are similar to those of L10-Leader. For acid shock assays, *S. mutans* UA159 were grown in TYE medium at pH 7.5 until cells reached the mid-logarithmic phase ($OD_{600}=0.4$). Cell cultures were then divided into 2 aliquots, pelleted by centrifugation, and resuspended in TYE medium at pH 5.5 or 7.5. The cells were subsequently incubated for 0.5, 1.5, 2.5, 3.5, or 4.5 hours. The levels of 5 target mRNAs changed differently in pH 5.5 than in pH 7.5, from 0.5 hours to 2.5 hours. But the levels of 5 target mRNAs were consistently higher in pH 5.5 than in pH 7.5, at 3.5 hours and 4.5 hours (Fig. 5c-g), which were similar to L10-

Leader. The changes of mRNAs were smaller and slower than L10-Leader.

Discussion

In this study, we used 4 kinds of bioinformatic approaches to predict the sRNAs in *S. mutans*. Bioinformatic approaches have been developed to identify sRNA-encoding genes in intergenic regions of bacterial genomes by searching for the co-localization of genetic features. These include predicted Rho-independent transcription terminators, promoters and transcription factor binding sites, intergenic conservation among closely related species, and conserved secondary structures (Kulkarni and Kulkarni, 2007). While these approaches have proven to be effective in identifying sRNAs in several species, they present significant computational challenges, requiring the positional relationships of thousands of individual genetic features to be found for each genome-wide search. SIPHT and sRNAPredict software predict candidate sRNA-encoding loci, identified based on the presence of putative Rho-independent terminators downstream of

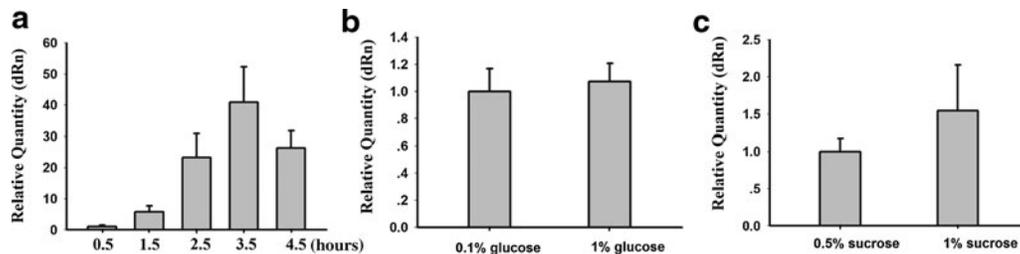


FIG. 4. qRT-PCR analysis of L10-Leader expression under different environmental stresses. **(a)** *S. mutans* UA159 in pH 5.5 and pH 7.5 TYE medium at 0.5h, 1.5h, 2.5h, 3.5h, and 4.5h. **(b)** *S. mutans* UA159 in 0.1% glucose and 1% glucose TYE medium for 24 hours. **(c)** *S. mutans* UA159 in 0.5% sucrose and 1% sucrose TYE medium for 24 hours.

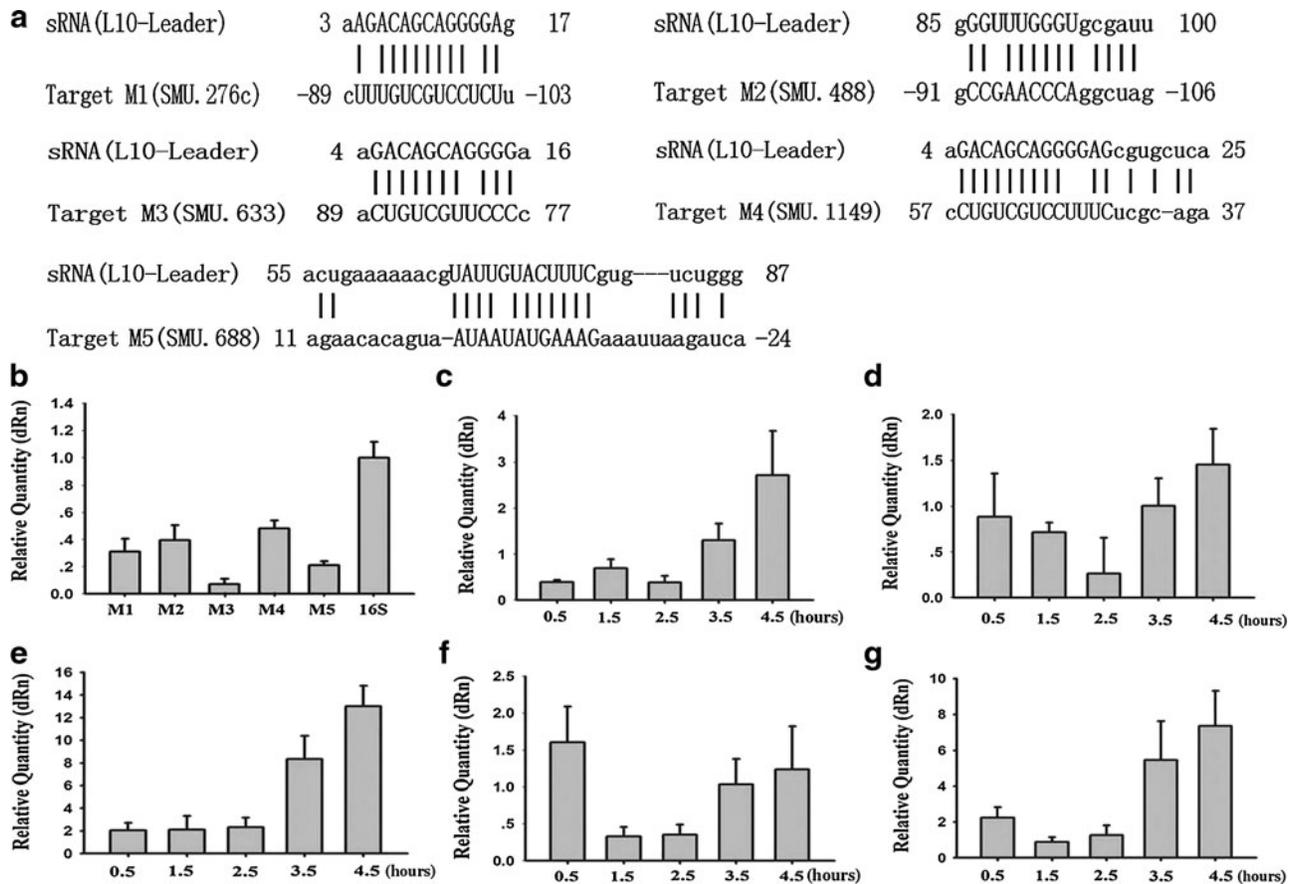


FIG. 5. Target sites and qRT-PCR of L10-Leader 5 target mRNAs. **(a)** L10-Leader and target sites of 5 target mRNAs of *S. mutans* UA159 predicted by sTarPicker. M1: hypothetical proteins SMU.276c; M2: putative hydrolase SMU.488; M3: putative thioesterase SMU.633; M4: transporter SMU.1149; M5: hypothetical protein SMU.688. **(b)** Five mRNAs were downregulated at OD₆₀₀ = 1.5. **(c)** M1 of *S. mutans* UA159 in pH 5.5 and pH 7.5 TYE medium at 0.5, 1.5, 2.5, 3.5, and 4.5 hours. **(d)** M2 of *S. mutans* UA159 in pH 5.5 and pH 7.5 TYE medium at 0.5, 1.5, 2.5, 3.5, and 4.5 hours. **(e)** M3 of *S. mutans* UA159 in pH 5.5 and pH 7.5 TYE medium at 0.5, 1.5, 2.5, 3.5, and 4.5 hours. **(f)** M4 of *S. mutans* UA159 in pH 5.5 and pH 7.5 TYE medium at 0.5, 1.5, 2.5, 3.5, and 4.5 hours. **(g)** M5 of *S. mutans* UA159 in pH 5.5 and pH 7.5 TYE medium at 0.5, 1.5, 2.5, 3.5, and 4.5 hours. The levels target mRNAs changed differently in pH 5.5 and pH 7.5 from 0.5 to 2.5 hours but were consistently higher in pH 5.5 than in pH 7.5, at 3.5 and 4.5 hours.

conserved intergenic sequences, and each locus is annotated for several features, including conservation in other species, association with one of several transcription factor binding sites and homology to any of over 300 previously identified sRNAs and *cis*-regulatory RNA elements (Livny, 2007; Livny et al., 2008). sRNASVM is a model for the prediction of small non-coding RNAs in *E. coli* using support vector machines (Saha and Raghava, 2006). These programs are set up using the features of sRNAs that have been found, and so it is hard to detect novel sRNAs that are completely different from known sRNAs. Some research has used genomic tiling microarrays to detect sRNAs in bacteria (Kumar et al., 2010). The advantage of this method is that it will find all transcribed sRNAs, but sRNA is related to the environment, so more specific experiments are required to find stress-related sRNAs. Furthermore, tiling microarrays require copious amounts of computational analysis of the detected sequence of sRNA. We know from research in *S. pyogenes* that only 7 of 75 candidate sRNAs were identified by both tiling microarrays and bioinformatic methods (Perez et al., 2009). Re-

gardless of whether bioinformatic prediction or tiling microarrays are used to analyze sRNA, they need to be verified experimentally.

In this study, we first verified the existence of sRNA in *S. mutans* using experimental approaches. The sequence of sRNA we studied was annotated as L10-leader (RF00557) family in Rfam website. L10-Leader family is a putative ribosomal protein leader auto-regulatory structure found in *Bacillus subtilis* and other low-guanine-cytosine content Gram-positive bacteria. It has also been identified in *L. monocytogenes* (Toledo-Arana et al., 2009). It is located in the 5'-UTR of mRNAs encoding ribosomal proteins L10 and L12 (*rplJ-rplL*). A Rho-independent transcription terminator structure that is probably involved in regulation is included at the 3' end (Zengel and Lindahl, 1994). The lowest-energy secondary-structure prediction for the L10-Leader sRNA was identified for *S. mutans*, and like most sRNAs, it has a loop structure. L10-Leader was also detected in clinical strains of *S. mutans*. The sequence was conserved among *Streptococci*, such as *Streptococcus dysgalactiae* subsp. *Equisimilis*, *Streptococcus*

equi subsp. *Zooepidemicus*, *S. pyogenes*, and *Streptococcus parauberis*. However, in clinical strains, the levels of L10-Leader are different. There were also some sRNAs expressed differently in different strains, which is similar to results found in other studies. In *S. aureus*, strains of RN6390, Newman, and COL, transcriptional analyses showed different transcription profiles of 11 sRNAs between the different strains (Geissmann et al., 2009). In *S. pyogenes*, sRNAs PEL, FASX, and SR195750 are variably expressed in different strains. In contrast, 5 candidate sRNAs do not show obviously change in *S. pyogenes* (Perez et al., 2009). sRNAs regulate the virulence of bacteria and different strains have different features, so different levels of some sRNAs may suggest they are related to virulence, while the others may not be.

The levels of L10-Leader were gradually downregulated from logarithmic growth phase to stationary growth phase, regardless of whether it was in *S. mutans* UA159 or the clinical strains. Numerous sRNAs in Gram-positive pathogens are growth phase dependent. In *L. monocytogenes*, there are 3 growth phase dependent sRNAs, which vary in response to growth media (Christiansen et al., 2004). Expression of most of the 11 sRNAs in *S. aureus* was dependent on the growth phase. In *S. pyogenes*, several sRNAs changed in different growth phases (Perez et al., 2009). It is well known that there are drastic morphological and physiological properties changes during different bacterial growth stages, from the exponential to the stationary phase. In exponentially growing cells, translation is the most energetically expensive process; its efficiency is under important selective pressure. Under some physiological conditions, a small set of genes regulate the large majority of transcription and translation taking place in the cell (Andersson and Kurland, 1990). These sets include genes related to translation, transcription, and energy metabolism, and are under strong selective pressure for translation efficiency. In *S. mutans*, *htrA* and *vicRXX* genes were shown to be related to growth phase (Ahn et al., 2005; Tremblay et al., 2009).

Our predicted target mRNAs of L10-Leader included putative hydrolase (SMU.488), putative thioesterase (SMU.633), and transporter (SMU.1149). Hydrolase, thioesterase, and transporters are related to the metabolic activity of bacteria. L10-Leader, in combination with these mRNAs, may be involved in the regulation of metabolism, cell growth, and division. For example, hydrolase may be involved in the adenosine triphosphate (ATP) hydrolysis reaction, releasing energy. Thioesterase catalyzes fatty chain termination, releasing free fatty acids, which break down to produce a lot of energy (Wakil, 1989). These reactions are able to supply sufficient energy for physical activity. Transporters are specialized transmembrane proteins, which recognize the substance and allow it to cross the membrane. The proteins involved are usually pumps, and normally use the chemical energy of ATP and transfer particles.

The levels of L10-Leader in *S. mutans* UA159 varied in response to low pH growth medium. The ability of *S. mutans* to tolerate low pH has received considerable attention because it is this property that contributes substantially to the virulence of this cariogenic organism. In one study (Fozo and Quivey, 2004a), it was demonstrated that *S. mutans* must make major changes to its membrane to survive at low pH, and the action of an enzyme named FabM is required. In fact, changes of cell membrane fluidity by adjustment of the

membrane fatty acid composition play an important role in the acid resistance phenotype of *S. mutans*, which has been shown to influence the virulence of this organism (Fozo and Quivey, 2004b; Fozo et al., 2007). More specifically, studies have shown that at low pH, changes in the membrane fatty acid composition involve a higher proportion of long-chain, monounsaturated fatty acids. This results in a membrane that is less permeable to protons, thereby facilitating survival in highly acidic environments (Fozo et al., 2004). In a study that examined the fatty acid profiles of *S. mutans*, *Streptococcus gordonii*, *Streptococcus salivarius*, and *Lactobacillus casei* in response to environmental acidification, it was shown that aciduric oral bacteria were able to increase their levels of long-chain, monounsaturated membrane fatty acids at low pH (Fozo and Quivey, 2004b). In *S. mutans*, the loss of *VicK* significantly enhanced its survival in an acid tolerance response assay (Senadheera et al., 2009). These observations suggest that membrane fatty acid alteration is a common mechanism utilized by bacteria to withstand environmental stress.

Alterations in the expression of mRNA in *S. mutans* in acidic environments were analyzed by microarray analysis. Results showed that the expression of most stress response and DNA repair related genes increased (Fozo et al., 2007). Bioinformatic predictions of L10-Leader target mRNAs show related genes, such as the putative manganese-type superoxide dismutase Fe/Mn-SOD (sod SMU.629) and putative DNA mismatch repair protein MutS₂ (SMU.1870). An anaerobic bacterium, *Propionibacterium shermanii*, synthesized either Fe- or Mn-SOD with an identical protein moiety, depending on the metal supplied (Meier et al., 1982). The presence of Fe-containing Mn-SOD in *E. coli* grown in tryptic soy yeast extract medium suggested the possibility that Fe-substituted Mn-SOD could have a physiological function (Beyer and Fridovich, 1991). Even if a Mn-SOD had been classified as having highly metal-specific enzyme activity, it may also exhibit at least some activity with iron at an acidic pH. The increased expression of L10-Leader in the acidic environment may suggest it combines with the target mRNA to increase its stability and promote the translation of target mRNA into protein. This allows bacteria to adapt to the acidic environment and ensures the correct matching of DNA under stress conditions. However, the role of the L10-Leader of *S. mutans* under acid conditions remains to be investigated.

We choose 5 possible target mRNAs of L10-Leader according to bioinformatics prediction web services sTarPicker (Ying et al., 2011) (<http://ccb.bmi.ac.cn/starpicker/index.php>) and RNAPredator (Eggenhofer et al., 2011) (http://rna.tbi.univie.ac.at/RNAPredator2/target_search.cgi). The 5 target mRNAs encoded hypothetical proteins (SMU.276c), putative hydrolase (SMU.488), putative thioesterase (SMU.633), transporter (SMU.1149), and hypothetical protein (SMU.688). *S. mutans* UA159 was grown to logarithmic growth phase and stationary growth phase. The results of qRT-PCR showed the transcript level of mRNAs was lower in stationary growth phase than in logarithmic growth phase. These changes are similar to that seen for L10-Leader. L10-Leader may combine with the target mRNAs to increase their stability, promoting mRNA translation. The proteins could provide particles and energy for the growth of bacteria. For acid shock assays, the levels of 5 target mRNAs changed differently in pH 5.5 than in pH 7.5, from 0.5 hours

to 2.5 hours. But interestingly, the levels of 5 target mRNAs were consistently higher in pH 5.5 than in pH 7.5, at 3.5 hours and 4.5 hours. The changes at 3.5 hours and 4.5 hours are similar to L10-Leader. The changes of mRNAs are smaller and slower than L10-Leader. As we know, sRNA can respond quickly to a variable environment and then regulate mRNAs. An mRNA may be regulated by more than 1 sRNA. So the final changes of mRNAs are very complicated. According to our predictions, many of the potential mRNA targets of L10-Leader are related to bacterial growth and metabolism, such as 30S ribosomal protein S6 s6 – rpsF (SMU.1860), putative citrate lyase alpha subunit cilA (SMU.1021), peptide chain release factor 1 prfA (SMU.1085), and permease protein psaB (SMU.1928). Therefore, we suggest that L10-Leader and its predicted mRNA targets may play important roles in the regulation of growth phase and acid adaptation in *S. mutans*.

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