RECTA: Regulon Identification Based on Comparative Genomics and Transcriptomics Analysis

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ABSTRACT

Regulons, which serve as co-regulated gene groups contributing to the transcriptional regulation of microbial genomes, have the potential to aid in understanding of underlying regulatory mechanisms. In this study, we designed a novel computational pipeline, RECTA, for regulon prediction related to the gene regulatory network under certain conditions. To demonstrate the effectiveness of this tool, we implemented RECTA on *Lactococcus lactis* MG1363 data to elucidate acid-response regulons. *Lactococcus lactis* is one of the most important Gram-positive lactic acid-producing bacteria, widely used in food industry and has been proved to have advantages in oral delivery of drug and vaccine. The pipeline carries out differential gene expression, gene co-expression analysis, *cis*-regulatory motif finding, and comparative genomics to predict and validate regulons related to acid stress response. A total of 51 regulons

were identified, 14 of which have computational-verified significance. Among these 14 regulons, five of them were computationally predicted to be connected with acid stress response with (i) known transcriptional factors in MEME suite database successfully mapped in *Lactococcus lactis* MG1363; and (ii) differentially expressed genes between pH values of 6.5 (control) and 5.1 (treatment). Validated by 36 literature confirmed acid stress response related proteins and genes, 33 genes in *Lactococcus lactis* MG1363 were found having orthologous genes using BLAST, associated to six regulons. An acid response related regulatory network was constructed, involving two trans-membrane proteins, eight regulons (*llrA*, *llrC*, *hllA*, *ccpA*, NHP6A, *rcfB*, regulons #8 and #39), nine functional modules, and 33 genes with orthologous genes known to be associated to acid stress. Our RECTA pipeline provides an effective way to construct a reliable gene regulatory network through regulon elucidation. The predicted response pathways could serve as promising candidates for better acid tolerance engineering in *Lactococcus lactis*. RECTA has strong application power and can be effectively applied to other bacterial genomes where the elucidation of the transcriptional regulation network is needed.

KEYWORDS

Lactococcus lactis MG1363; Acid stress response; Differentially expressed gene; Gene co-expression; *Cis*-regulatory motif finding; Regulon; Gene regulatory network.

INTRODUCTION

Genomic and transcriptomic analyses have been widely used for elucidating gene regulatory network (GRN) hierarchies and offering insight into the coordination of response capabilities in microorganisms (Arnoldini et al. 2012, Carvalho et al. 2013, Levine et al. 2013, Locke et al. 2011). One way to study the mechanism of transcriptional regulation in microbe genomics is regulon prediction. A regulon is a group of co-regulated operons, which contains single or multiple consecutive genes along the genome (Cao, Ma, et al. 2017, Mao et al. 2015, Zhou et al. 2014). Genes in the same operon are controlled by the same promoter and are co-regulated by one or a set of TFs (Jacob et al. 1960). The elucidation of regulons can improve the identification of transcriptional genes, and thus, reliably predict the gene transcription regulation networks (Liu, Zhou, et al. 2016).

There are three ways for regulon prediction: (i) Predicting new operons for a known regulon (Kumka and Bauer 2015, Tan et al. 2001). This method combines motif profiling with a comparative genomic strategy to search for related regulon members and carries out systematical gene regulation study. (ii) Integrating *cis*-regulatory motif (*motif* for short) comparison and clustering to find significantly enriched motif candidates (Gupta et al. 2007, Ma et al. 2013). The candidate motifs are then

assembled into regulons. (iii) Performing *ab initio* novel regulon inference using *de novo* motif finding strategy (Novichkov et al. 2010). This approach uses a phylogenetic footprinting technique which mostly relies on reference verification (Blanchette et al. 2002, Katara et al. 2012, Liu, Zhang, et al. 2016) and can perform a horizontal sequential comparison to predict regulons in target organisms by searching known functionally-related regulons or TFs from other relevant species. One algorithm for phylogenetic footprinting analysis called Motif Prediction by Phylogenetic Footprinting (MP3) has been used for regulon prediction in *E. coli* (Liu, Zhang, Zhou, Li, Fennell, Wang, Kang, Liu and Ma 2016). MP3 was then integrated into the DMINDA web server along with other algorithms, such as the Database of Prokaryotic Operons 2.0 (DOOR2) (Cao, Ma, Chen and Xu 2017, Mao et al. 2014), BOttleneck BROken (BoBro) (Li, Ma, et al. 2011) and BoBro-based motif Comparison (BBC) (Ma, Liu, Zhou, Yin, Li and Xu 2013), to construct a complete pipeline for regulon prediction. In the latest research, a newly developed pipeline called Single-cell Regulatory Network Inference and Clustering (SCENIC) combines motif finding from co-expression gene modules (CEMs) with regulon prediction for single-cell clustering and analysis (Aibar et al. 2017). Such method builds up a way of regulon application in single-cell and metagenomic research. Nevertheless, without a suitable regulon database, researchers need to build up the library first through operon identification, CEM analysis, motif prediction and comparison (Jensen et al. 2005). Here, we reported an integrated computational framework of <u>re</u>gulon identification based on <u>c</u>omparative genomics and <u>t</u>ranscriptomics <u>a</u>nalysis (**RECTA**) to elucidate the GRN responses in microbes under specific conditions. To better elucidate the methodology of RECTA, we built a regulatory network responding to the acid stress in *Lactococcus lactis* (*L. lactis*) species.

Lactococcus lactis (*L. lactis*) is one of the mesophilic Gram-positive lactic acid-producing bacteria. It is has been widely applied in dairy fermentations, such as cheese and milk product (Wegmann et al. 2007). Several studies have provided evidence of its essential roles in wrapping and delivering proteins or vaccinations for immune treatment, such as diabetes (Ma, Liu, et al. 2014), malaria (Ramasamy et al. 2006), tumors (Bermudez-Humaran et al. 2005, Zhang et al. 2016) and infections (Hanniffy et al. 2007). Holding the advantage of higher acid tolerance to protect vectors from resolving during delivery inside of the animal body, *L. lactis* has more potential and safety in oral drug development (Hols et al. 1999). Moreover, it has been found that *L. lactis*, along with some *Lactobacillus*, *Bifidobacterium* and other gut microbiota, were associated with obesity (Million et al. 2012). Such studies lead to the possibility and availability of *L. lactis* in metagenomic studies to investigate the effect of microbial interaction between *L. lactis* and other species in the human body. It is now well established that *Lactococcus* have evolved stress-sensing systems, which enable them to tolerate harsh environmental conditions (Carvalho, Turner,

Fonseca, Solopova, Catarino, Kuipers, Voit, Neves and Santos 2013, Hutkins and
Nannen 1993, van de Guchte et al. 2002).

Among the harsh environmental conditions that microorganisms confront, acid stress
is known to change the level of the alarmones (guanosine tetraphosphate and
guanosine pentaphosphate, collectively referred to as (p)ppGpp) (Hauryliuk et al.
2015) and leads to a stringent response to cellular regulation (Rallu et al. 2000). The
reason that bacteria maintain the protection mechanism against acid stress is to
withstand the deleterious effects caused by the harmful high level of protons in the
exposed environment. Many mechanisms or genes related to the acid stress response
(ASR) have been identified. Proton–pumping activity, the direct regulator to acid
stress response, controls the intracellular pH level by pumping extra protons out of the
cell (Koebmann et al. 2000, Lund et al. 2014), and the increase of alkaline compound
levels also counters the acidification found in streptococci (Shabayek and Spellerberg
2017). Acid damage repair of cells by chaperones or proteases, such as GroES,
GroEL, GrpE, HrcA, DnaK, DnaJ and Clp (Frees et al. 2003, Jayaraman et al. 1997),
and hdeA/B and Hsp31 in *Escherichia coli* (*E. coli*) (Kern et al. 2007, Mujacic and
Baneyx 2007), the arginine deiminase system (ADI) (Budin-Verneuil et al. 2006,
Ryan et al. 2009, Sun et al. 2012, Zuniga et al. 2002) and glutamate decarboxylases
(GAD) pathways, etc. (Hoskins et al. 2001, Nomura et al. 1999, Sanders et al. 1998)
have been proven to be associated with the acid response. Additionally,
transcriptional regulators, sigma factors, and two-component signal transduction
system (TCSs) have also been demonstrated to be responsible for ASR by modifying
gene expression (Cotter and Hill 2003). These genes or pathways suggest low pH has
widespread adverse effects on cell functions and inflicts response at genomic,
metabolic, and macromolecular levels. To better understand the mechanism that
controls the acid tolerance and response to the acid stress in *L. lactis*, we considered
MG1363, a strain extensively studied for acid resistance, to carry out computational
analyses (Carvalho, Turner, Fonseca, Solopova, Catarino, Kuipers, Voit, Neves and
Santos 2013, Hartke et al. 1996, Linares et al. 2010, Sanders et al. 1999).
Nevertheless, to adequately describe the transcriptional state and gene regulation
responsible for ASR in *L. lactis*, a GRN integrating all individual pathways is needed.

The experiment was conducted by six steps and the general framework is showcased
in Figure 1. (i) MG1363 co-expression gene modules (CEMs) and differentially
expressed genes (DEG) were generated from microarray data by hcluster package
(Antoine Lucas 2006) and Wilcoxon test (Bauer 1972) in R, respectively. MG1363
operons were predicted from genome sequence using DOOR2 webserver and assigned
into each CEM; (ii) For each CEM, the 300bp upstream to the promoter was extracted
and the sequences were used to find motifs using DMINDA2.0; (iii) The top five
significant motifs in each CEM were reassembled by their similarity comparison and
clustering to predict regulons; (iv) The motifs were compared to known transcription
factor binding sites (TFBSs) in the MEME suite, and the TFs corresponding to these

TFBSs were mapped to MG1363 using BLAST. Only regulons with DEGs and

mapped TF were kept as ASR-related regulons; (v) Experimentally identified ASR-

related genes in other organisms were mapped to MG1363 using BLAST and

allocated to corresponding regulons for further verification; and (vi) The relationship

between regulons and functional gene modules were established to elucidate the

overall ASR mechanism in MG1363.

Figure 1 should be placed here.

As a result, 14 regulons are identified, literature verified or putative, to be connected

to ASR. Eight regulons, related to nine functional modules and 33 associated genes,

are considered as the essential elements in acid resistance in MG1363. This proposed

computational pipeline and the above results significantly expand the current

understanding of the ASR system, providing a new method to predict systematic

regulatory network based on regulon clustering.

METHODS

Data Acquisition

The *L. lactic* MG1363 genome sequence was downloaded from NCBI (GenBank

accession number: AM406671). The microarray dataset contains eight samples under

different acid stress conditions for MG1363 was downloaded from the Gene

Expression Omnibus (GEO) Database (Series number: GSE47012). The data has been

treated with LOWESS normalization by the provider. The details on cell culture

preparation and data processing can be found in the previous study (Carvalho, Turner,

Fonseca, Solopova, Catarino, Kuipers, Voit, Neves and Santos 2013). This dataset has

all bacteria grown in basic conditions: a 2-liter fermenter in chemically defined

medium containing 1% (w/v) glucose at 30°C. The control and treatment samples

were grown at a pH of 6.5 and 5.1, respectively.

Several TFBS databases integrated in the MEME suite, including DPINTERACT (*E.*

coli) (http://arep.med.harvard.edu/dpinteract/), JASPAR (Anthony Mathelier 2016),

RegTransBase (Prokaryotes) (Kazakov et al. 2007), Prodoric Release (Prokaryotes)

(Munch et al. 2003), and Yeastract (Yeast) (Teixeira et al. 2014), were utilized for

regulon filtering in known TF templates to find homologous TFs and corresponding

genes in MG1363 using BLAST with default parameters. In the literature validation

part, all ASR-related transporters and genes were collected from published articles,
and their sequences were obtained from NCBI and UniProt databases.

Operon identification

The genome-scale operons of MG1363 were identified by DOOR2. It is a one-stop
operon-centered resource including operons, alternative transcriptional units, motifs,
terminators, and conserved operons information across multiple species (Mao, Ma,
Zhou, Chen, Zhang, Yang, Mao, Lai and Xu 2014). Operons were predicted by the
back-end prediction algorithm with a prediction accuracy of 90-95%, based on the
features of intergenic distance, neighborhood conservation, short DNA motifs, length
ratio between gene pairs, and newly developed transcriptomic features trained from
the strand-specific RNA-seq dataset (Chou et al. 2015, Li, Liu, et al. 2011).

Gene differential expression analysis and co-expression analysis

201 DEGs were identified based on the Wilcoxon signed-rank test (Bauer 1972) between
202 the control and treatment, which was performed in R. The gene co-expression analysis
203 was performed using a hierarchical clustering method ('hcluster' package in R)
204 (Antoine Lucas 2006) to detect the CEMs under the acid stress in MG1363.

Motif finding and regulon prediction

Genes from each CEM were first mapped to the identified operons to retrieve the
basic transcription units. Next, 300 bps in the upstream of the translation starting sites
for each operon were extracted, in which motif finding was carried out using the web
server DMINDA (Ma, Zhang, et al. 2014, Yang et al. 2017). DMINDA is a dominant
motif prediction tool, embraced five analytical algorithms to find, scan, and compare
motifs (Li, Liu, Ma and Xu 2011, Liu et al. 2017, Ma, Liu, Zhou, Yin, Li and Xu
2013), including a phylogenetic footprint framework to elucidate the mechanism of
transcriptional regulation at a system level in prokaryotic genomes (Li, Ma, Mao, Yin,
Zhu and Xu 2011, Liu, Zhang, Zhou, Li, Fennell, Wang, Kang, Liu and Ma 2016, Liu,
Zhou, Li, Zhang, Zeng, Liu and Ma 2016). All sequences were uploaded to the server
and default parameters were used to find the top five significant motifs (p-value <
0.05) in each cluster. The identified motifs were subjected to motif comparison and
grouped into regulons based a sequence similarity cutoff using the BBC program in
DMINDA (Ma, Liu, Zhou, Yin, Li and Xu 2013).

Regulon validation based on TF BLAST and DEG filtering.

Each highly conserved motif was considered to contain the same TFBS among species. Therefore, a comparison study was performed using TOMTOM in the MEME Suite (Bailey et al. 2009) between identified motif and public-domain TFBS databases, including DPINTERACT, JASPAR, RegTransBase, Prodoric Release and Yeastract, to find TFBSs and corresponding TFs with significant p-values in other prokaryotic species. Those TFs were then mapped to MG1363 using BLAST by default parameters to predict the connection between regulons and TFs in MG1363. On the other hand, since genes without differential expression were supposed not to react to pH changes, and thus, irrelevant to ASR, regulons without DEGs were not involved in the GRN, and thus, excluded from the following steps.

Regulon validation based on known ASR proteins from the literature.

To validate the performance of the above computational pipeline for regulon prediction, a literature-based validation was performed. Thirty-six ASR-related proteins and genes in other organisms including *L. lactis, E. coli*, *Streptococci*, *etc.* were first manually collected from literature, and their sequence was retrieved from the NCBI and UniProt databases. They were used to examine the existing known mechanisms in response to pH changes in MG1363 using the BLAST program by default parameters on NCBI. Such literature-based validation can either confirm the putative regulons when known ASR-related genes can be found in the significant regulons or expand our results to some insufficiently significant regulons, which indicate both false positive and true negative rate to evaluate the computational pipeline.

RESULTS

Predicted operons and CEM generation

A total of 1,565 operons with 2,439 coding genes of MG1363 (Dataset S1) were retrieved from the DOOR2 database. Through co-expression analysis, the 1,565 operons were grouped into 124 co-expressed clusters. Among these clusters, two large groupings contain more than 200 operons. Each of which was removed from the subsequent analyses as larger clusters may have higher chances to induce false positive operons which were connected with true operons by co-expression analysis. For the remaining 122 clusters covering 2,122 genes, 26 (21%) contain no more than 10 operons; the smallest cluster had two operons, and most of the clusters (90%) contained between 10 and 50 operons (Dataset S2 and Figure S1).

Predicted regulons based on motif finding and clustering.

Using BoBro in the DMINDA web server, multiple motif sequences were identified from the 300 bps in the upstream of the translation start sites for each operon. Only the top five significant motifs (p-value < 0.05) were selected in each cluster, giving rise to a total of 610 (122×5) identified. The motif comparison-and-clustering analysis was then performed on the 610 motifs, and 51 motif clusters were identified with a motif similarity 0.8 as a cutoff. Intuitively, the operons sharing highly similar motifs in each motif cluster are supposed to be regulated by the same TF and tend to be in the same regulon. Hence, these 51 motif clusters correspond to 51 regulons (Dataset S3).

Computationally-verified regulon based on TF BLAST and differential gene expression (DGE) analysis.

Among the above 51 regulons, 14 were found containing motifs significantly (E-value < 0.05) matched to known TFBSs using TOMTOM in the MEME suite, representatively. The motif logos are shown in Figure S2, and more details can be found in Dataset S4. The 14 TFBS-corresponding TFs were then mapped to MG1363 using BLAST to identify the real TFs/genes regulating each regulon. As a result, eight known TFs, spo0A, lhfB, GAL80, CovR, c4494, ihfA, CovR, and RHE_PF00288, were successfully mapped to MG1363 resulting in eight TFs with duplicates. *llrA* (llmg_0908) regulates regulons #12 and #37, *ccpA* (llmg_0775) regulated regulons #15 and #47, *hllA* (llmg_0496) regulates regulons #7 and #31 (Table 1). *CcpA* (Abranches et al. 2008, Zomer et al. 2007), *llrA* (O'Connell-Motherway et al. 2000), *llrC* (O'Connell-Motherway, van Sinderen, Morel-Deville, Fitzgerald, Ehrlich and Morel 2000), and *hllA* (Bolotin et al. 1999) were known to be ASR-related genes in *L. lactis*; gene (llmg_0271), without any related known TF, was found similar to template TF GAL80 in yeast, which has not been associated with any ASR regulation pathways yet. For all 14 significant regulons, regulons #3, #4, #20, #28, #40, and #44 are potential candidates as, currently, no related TFs in *L. lactis* have been found.

Table 1 should be placed here.

Additionally, 86 down-regulated genes and 55 up-regulated genes (Dataset S5), resulting from DGE analysis were integrated into the regulons. Regulons #10, #37,

#44 and #47 were found lacking DEGs. Thus, gene llmg_0271, related to regulon
#10, was not likely to respond to acid stress in MG1363 even though it has been
successfully mapped to MG1363, and was then grouped into the potential candidate.
On the contrary, *ccpA* and *llrA* were still retained due to their involvements in
regulons #15 and #12 with DEGs, respectively.

By the end of the computational pipeline, we predicted that regulons #2, #7, #12, #15
and #31 were related to GRN in MG1363. Merging regulon #7 and #31 as one, we
referred their TF names (*ccpA, llrA, llrC,* and *hllA*) to represent the five regulons for
convenience.

Verified regulons based on literature verification

Altogether, 36 literature-supported ASR-related transporters were successfully
mapped to MG1363 using blast with an E-value cutoff as 1e-10 and resulted in a total
of 33 mapped genes. All the 36 transporters were categorized into nine modules based
on their biological functions or regulated pathways, including L-lactate
dehydrogenase (LDH), GAD, ADI, urea degradation, F1/F0ATPase, acid stress,
protein repair and protease, envelope alterations and DNA repair. The 33 mapped
genes generate 22 operons and six regulons: *llrA*, *llrC*, *hllA*, NHP6A, regulon #8 and
#39, which were subjected, one or more, to each functional module (Table 2).

Regulon *llrA*, *llrC*, and *hllA* have already been computationally identified in Table 1
and supported again by literature verification results. NHP6A which interestingly has
a homologous TF in human and fungi but not in *L. lactis* (Kolodrubetz and Burgum
1990, Stillman 2010), yet failed to map in MG1363. Here, we are using NHP6A to
represent regulon #20, as their relationship has been predicted computationally in
Table 1. Regulon #39 was identified to be regulated by *llrD*, one of the six two-
component regulatory systems in MG1363 (O'Connell-Motherway, van Sinderen,
Morel-Deville, Fitzgerald, Ehrlich and Morel 2000). Regulons #8 (llmg_1803) and
#39 (*llrD*) were not included in the 14 significant regulons in Table 1. For NHP6A,
regulons #8 and #39 were enriched by literature validation as it expanded regulon
results of the RECTA pipeline. Among the nine functional modules, *llrA* was found
connected to five of them, and NHP6A related to three. On the other hand, the GAD
and urea degradation functional modules failed to connect to any previous regulons.

Table 2 should be placed here.

Compared to the regulon verification based on TF BLAST and DGE, the literature
verification identified two more regulons (#8 and #39) that lay in the insignificant
group, however, with no sign of *ccpA* regulon. Thus, such a result indicates a possible
false positive rate of 1/5 and a true negative rate of 2/37 of our computational
pipeline, indicating a reliability and feasibility of using RECTA to predict the ASR-
related regulons. In Figure 2, we show the processes and results for both literature
verification and computational pipeline in detail. The final eight regulons predicted
from both parts were then compared to construct a GRN response to acid stress,
integrated with other information found in the literature.

Figure 2 should be placed here.

A model of regulatory network in response to pH change

According to the results outlined above, we are presenting a working model of the
transcriptional regulatory network for acid stress response in MG1363 (Figure 3). The
network consists of two transmembrane proteins (Dataset S6), eight regulons, nine
functional modules, and 33 orthologous genes known for ASR in other bacteria that
are also contributing in MG1363.

Figure 3 should be placed here.

The network is subjected to respond to the change of intracellular proton level. The
signal is captured by H+ sensor and regulons are initiated to be regulated. Although
significance was not shown for *rcfB* in our computational results, it has been reported
to recognize and regulate promoter P170 (Madsen et al. 1999), P1 and P3 (Hindre et
al. 2004, Rince et al. 1994), which are activated by the ACiD box and essential to acid
response (Madsen et al. 2005). With the ACiD box, operons like *groESL*, *lacticin 481*
and *lacZ* have been proved to be regulated by *rcfB*, while *als*, *aldB*, etc. have not
(Madsen, Hindre, Le Pennec, Israelsen and Dufour 2005). The homologous
comparative study also predicted the existence of the ACiD box in *llrA* (Akyol et al.
2008, O'Connell-Motherway, van Sinderen, Morel-Deville, Fitzgerald, Ehrlich and
Morel 2000). With such evidence, we separated *rcfB* from regulon #39 and predicted

that *rcfB* is first triggered by H+ sensor and acts as the global initiator that controls

the other seven regulons. It is reasonable that *rcfB*-related regulon #39 failed to show

significant TF matching results after CEM treatment in the operon clustering step.

rcfB worked as a trustworthy global factor; its differential expression should be less

significant than regulons directly responding to acid stress, thus leading to the failure

of being predicted by the RECTA pipeline. Nevertheless, the low number of

microarray data sets (eight) also limited the real performance to the ASR. However,

the mechanism of how H+ sensor is activating and regulating the GRN and *rcfB*

remains unclear. In the seven regulons, three--*llrA*, *llrC* and *hllA*--were verified

through literature to be related to ASR; regulons #8 and #39 showed less significant in

regulon prediction; NHP6A was considered as putative regulon due to its failure to

map in MG1363; and *ccpA* was another putative regulon without literature support.

The six downstream regulons (*llrA*, *llrC*, *hllA*, NHP6A, regulon #39, and regulon #8)

other than *ccpA*, interact with each other to regulate six ASR-related functional

modules, including ADI system, DNA repair, LDH, protein repair, envelope

alterations, and F0/F1ATPase. The ADI pathway, which generates ATP and protects

cells from acid stress (Zuniga, Perez and Gonzalez-Candelas 2002), is under the

regulation of NHP6A, *llrC*, *llrA*, and *hllA*. Another important pathway is the LDH

(EC 1.1.1.27) under the regulation of NHP6A and *llrA*, which converts pyruvate and

H+ to lactate which is exported outside of cells (Dennis and Kaplan 1960). Chaperons

which take part in macromolecule protection and repairing are subjected to regulon

llrA. Chaperons have functions that includes providing protection to against

environmental stress, helping protein folding, and repairing damaged proteins, and

have been demonstrated to show clear linkage with acid stress in numerous Gram-

positive bacteria (Frees, Vogensen and Ingmer 2003, Jayaraman, Penders and Burne

1997, Kern, Malki, Abdallah, Tagourti and Richarme 2007, Mujacic and Baneyx

2007). F0/F1ATPase, controlled by *llrA* and regulon #8, also plays an important role

in maintaining normal cellular pH, which pumps H+ out of cells at the expense of

ATP (Amachi et al. 1998, Koebmann, Nilsson, Kuipers and Jensen 2000, Lund,

Tramonti and De Biase 2014, O'Sullivan and Condon 1999). The GAD (Nomura,

Nakajima, Fujita, Kobayashi, Kimoto, Suzuki and Aso 1999, Sanders, Leenhouts,

Burghoorn, Brands, Venema and Kok 1998) and urea degradation (Cotter and Hill

2003) functional modules are missing reliable associations with the regulons in

MG1363 while maintaining functions in ASR mechanism in other species.

DISCUSSION AND CONCLUSIONS

Implementation of the novel computational pipeline RECTA resulted in the construction of an eight-regulons enrolled ASR regulatory network. The framework provides a useful tool and will be a starting point toward a more systems-level understanding of the question (Cao, Wei, et al. 2017). The identified motifs and regulons suggest acid resistance is a coordinated response regarding regulons, although most of which have not been identified or experimentally verified. From the three well-identified regulons—*llrA*, *llrC*, and *hllA*, it appears the gene regulation is also complex, as these regulons also interact with other proteins and TFs. F0/F1ATPase is directly involved in the concentration regulations of the intracellular proton. Other pathways are responsible for repairing the damage caused by acid stress, such as DNA repair, protein repair, and cell envelops alterations. However, there were also several reported ASR-related genes or transporters such as *htrA* in *Clostridium spp.* (Alsaker et al. 2010), CovS/CovR acid response regulator in *Streptococcus* (Cumley et al. 2012), cyclopropane fatty acid (cfa) synthase for cell-membrane modification (Budin-Verneuil et al. 2005), and oxidative damage protectant genes like *sodA*, *nox-1* and *nox-2* (Santi et al. 2009) that failed to map to MG1363. Using more gene expression datasets for CEM and DGE analyses could be a way to strengthen the result of our computational pipeline, which might cover more significant regulons to construct a more solid and complete regulatory network.

Homology mapping at the genomic level showed very a long evolutionary distance between MG1363 and currently well-annotated model species. Hence, the functional analysis for MG1363 is limited, and it is hard to apply gene functional enrichment to verify our prediction results. With more expression datasets and experiments about protein-protein interactions, the ASR mechanism can be largely improved in *L. lactic* MG1363.

In summary, through the implementation of RECTA, we found that the ASR at the transcriptome level in MG1363 is an orchestrated complex network. Functional annotation shows these regulons are involved in many levels of biological processes, including but not limited to DNA expression, transcription, and metabolism. Our method builds a TF-regulons-GRN relationship so that the new ASR-related genes can be predicted. Besides, the low false positive and true negative rate indicate the RECTA pipeline as sensitive and reasonable. In fact, considering the high accuracy, we regarded *ccpA* as the putative regulon, though not connected to any related functional modules, while more robust methods are required. Such results expand current pathways to those that can corroborate cell structures—cell wall, cell

membranes, etc.—and related functions. Our findings suggest that acid has profound adverse effects and inflict systems-level response. Such predicted response pathways can inform better resistance design.

Looking forward to the acid tolerance advantage of *L. lactis*, which makes its prospective application in drug and vaccine delivery, the effects on anti-obesity research, and metagenomic studies, the ASR-related GRN in *L. lactis* shows an excellent research value. Fully understanding of its theory may contribute to the development of *Lactococcus* therapy and can even expand to other close species by genetic modification. Furthermore, our computational pipeline provides an effective method to construct reliable GRN based on regulon prediction, integrating CEMs, DGE analysis, motif finding, and comparative genomics study. It has a durable application power and can be effectively applied to other bacterial genomes, where the elucidation of the transcriptional regulation network is needed.

In this study, we designed a computational framework, RECTA, for acid-response regulon elucidation. This tool integrates differential gene expression, co-expression analysis, *cis*-regulatory motif identification, and comparative genomics to predict and validate regulons associated with acid response. In demonstrating the efficacy of this tool, we analyzed *Lactococcus lactis* MG1363. This implementation resulted in the expanded understanding of the acid-response regulon network for this one strain of *L. lactis* and provides an applicable method for acid-response regulon elucidation of further species. Through utilization of the RECTA pipeline, researchers can readily evaluate acid-response mechanisms for numerous bacterial species, while simultaneously validating the results of their study.

LIST OF ABBREVIATIONS

ADI: arginine deiminase

BoBro-based motif Comparison (BBC)

BOttleneck BROken (BoBro)

CEM: co-expression (gene) module

DEG: differentially expressed gene

DGE: differential gene expression

E. coli: Escherichia coli

GAD: glutamate decarboxylases

GRN: gene regulatory network

LDH: lactate dehydrogenase

L. lactis: *Lactococcus lactis*

MG1363: *Lactococcus lactis* MG1363

Motif: cis-regulatory motif

RECTA: regulon identification based on comparative genomics and transcriptomics
analysis
TF: transcription factors
TFBS: transcriptional factor binding site

DECLARATIONS

Ethics Approval and Consent to Participate

Not applicable

Conflict of Interest Statement

The authors declare that they have no competing interests

Data Availability Statements

All datasets analyzed and generated for this study are included in the manuscript and
the supplementary files.

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Authors' Contributions

XC and QM carried out the framework of the paper and are responsible for the
regulon prediction work. HZ carried out the operon prediction and motif finding
work. CL was responsible for DGE analysis. HC, A. McDermaid, and A. Ma drafted
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FIGURE LEGEND

Figure 1: The flowchart of constructing global ASR transcriptional network in
MG1363. Step 1: microarray data was used to generate co-expressed gene clusters

and DEGs, and MG1363 genome sequence was used to find operons. Step 2: a motif
finding progress was carried out to identify all statistically significant motifs in each
of the CEMs. Step 3: a regulon finding procedure was designed to identify all the
possible regulon candidates encoded in the genome based on motif comparison and
clustering. Step 4: the motifs of each of these regulons were compared to known
TFBSs, and DGE analysis between low pH condition and normal condition was used
to figure out the ASR-related regulons. Step 5: regulon validation based on literature
information verified the significant putative regulons and expanded the results to
some insufficiently significant regulons. Step 6: the ASR-related GRN in MG1363
was predicted and described with eight regulons, nine functional modules, and 33
genes. The combination of the above information forms a genome-scale regulatory
network constructed for ASR.

Figure 2.   Regulon prediction using RECTA pipeline (red) and validation and
enrichment using literature information and gene blast (blue). All processes were
shown in rectangles and results were highlighted with corresponding background
colors. In the computational pipeline, 51 regulons with assigned motifs and operons
were analyzed sequentially through significant TFBS pairing, DEG conformation, and
TF blast. Only regulons contained DEGs (ten) and had related mapped TF (eight)
were believed to be the final predicted ASR-related regulons (five). These five
regulons were then merged into four and using the corresponding TFs to represent
their names. In the literature validation process, known ASR-related transporters were
first mapped to the MG1363 genome and resulted in 33 genes. Those genes were then
searched in 51 regulons and determine six related regulons. All regulons resulted from
both computational pipeline and literature validation were combined, along with the
information of functional modules, to determine the GRN.

Figure 3. A working model of the transcriptional gene regulatory network in response
to pH change in *L. lactis*. The mechanism is activated by the change of proton signal
in a cell. *RcfB* is assumed to be the overall activator for the rest seven regulons and
controls the ASR functional module solely. Three kinds of literature verified and
significant ASR-related regulons, *llrA*, *llrC* and *hllA*, two insufficient significant
regulon *llrD* (regulon #39) and regulon #8 (llmg_1803) predicted via our workflow
but results under 0.8 motif similarity cutoff or could not find hit, and one putative
significant regulon NHP6A control the seven functional modules which are
experimentally verified in close species MG1363. The other significant regulon *ccpA*
failed to be confirmed by any literature proved genes or transporters. Two extra
functional modules, GAD, and urea degradation show no direct connection to all
seven the regulons. One or more homology genes are found in MG1363 for all the
nine modules using BLAST. The solid arrows indicate regulation between
regulons/TFs and functional modules/genes, and the dashed arrows indicate uncertain
control processes. Besides, two ovals indicate two trans-membrane proteins, one is

confirmed as F0/F1ATPase and the other one with dash line that we still not find its
related information in the public-domain literature.

TABLE LEGEND

Table 1. 14 significant regulons that are verified and mapped to known TFs**.**
According to analyses, operon numbers and DEG determination (yes or no), matched
template TFs and mapped TFs were assigned for each significant regulon,
respectively, and were aligned based on regulon ID number. Five regulons containing
DEGs and having the corresponding TF at the same time were bolded, as which were
being computationally verified regulons to be responsible for acid stress in MG1363.

Regulon ID	No. of operons	DEG	TF template	TF(gene) blast in MG1363
Regulon2	**82**	**Y**	**spo0A**	***llrC*(llmg_0414)**
Regulon3	32	Y	FoxQ1	N/A
Regulon4	20	Y	SPT2	N/A
Regulon7	**49**	**Y**	**lhfB**	***hllA*(llmg_0496)**
Regulon10	5	N	GAL80	-(llmg_0271)
Regulon12	**259**	**Y**	**CovR**	***llrA*(llmg_0908)**
Regulon15	**19**	**Y**	**c4494**	***ccpA*(llmg_0775)**
Regulon20	79	Y	NHP6A	N/A
Regulon28	5	Y	1Z916	N/A
Regulon31	**65**	**Y**	**ihfA**	***hllA*(llmg_0496)**
Regulon37	10	N	CovR	*llrA*(llmg_0908)
Regulon40	7	Y	Awh	N/A
Regulon44	12	N	YBR182C	N/A
Regulon47	5	N	RHE_PF00288	*ccpA*(llmg_0775)

Table 2. Known ASR-related gene mapping from literature in response to pH
change. Literature-supported ASR-related genes found in close species or other *L.*
lactis strains. The template transporters and genes were first identified in published
studies from the NCBI and UniProt databases. *L. lactis Il1403* was used as the
organism which is very close to MG1363 if template gene existed. Only 36 templates
that successfully mapped to the MG1363 genome were listed, which resulted in 33
genes. All mapped genes and corresponding templated were organized by their
regulated pathways which were further used as functional modules. Mapped genes
were searched in 51 regulons to build the connections between functional modules
and regulons.

Template organisms	MG1363

Organisms	Transporters	Functions/ Pathways	Mapped genes (locus tag)	Regulons
L. lactis	*ldh* *ldhB* *ldhX*	LDH	*ldh* (llmg_1120) *ldhB* (llmg_0392, llmg_0475) *ldhX* (llmg_1429)	NHP6A, *llrA*
L. lactis	*gadB* *gadC*	GAD	*gadB* (llmg_1179) *gadC* (llmg_1178)	N/A
L. lactis	*arcA* *arcB* *arcC1* *arcC2* *argF*	ADI pathway	*arcA* (llmg_2313) *arcB* (llmg_2312) *arcC1* (llmg_2310) *arcC2* (llmg_2309) *argF* (llmg_1754)	NHP6A, *llrA*, *llrC*, *hllA*
Bacteria	ureA/B/C[$]	Urea degradation	*pyrC* (llmg_1508)	N/A
L. lactis	atpEBFHAGDC[$$]	F0/F1ATPase	llmg_1952, llmg_1951, llmg_1950, llmg_1949, llmg_1948, llmg_1947, llmg_1946, llmg_1945	*llrA*, (Regulon8, llmg_1803)[$$$]
L. lactis	rcfB	Acid response	rcfB (llmg_2512)	(Regulon39, *llrD*)[$$$]
L. lactis, *E. coli* K12	*dnak* *groEL* *groES* *grpE* *clpE* *clpC* *clpP*	Chaperone, Protein repair and protease	*dnaK* (llmg_1574) *groEL2* (llmg_0411) *groES* (llmg_0410) *grpE* (llmg_1575) *clpE* (llmg_0528) *clpC* (llmg_0615) *clpP* (llmg_0638)	*llrA*
L. lactis, *Bacillus subtilis*	*dltC, agK, SGP, ffh*	Envelope alterations	llmg_0878	NHP6A, *llrA*
L. lactis	*recA, uvr, smn*	DNA repair	llmg_0374, llmg_0534, llmg_1718 llmg_1221	(Regulon39, *llrD*)[$$$]

[$]Three subunits of urease enzymes coded by ureABC operon found preserved in

multiple bacteria.

[$$]Altogether eight genes.

[$$$]The homolog prediction or motif research results with low homolog similarity but

have meaningful biological relevance.

SUPPLEMENTAL MATERIALS

Supplement Number 1: Figure S1 operon_distribution.pptx;

The distribution of operon numbers among 122 clusters.

Supplement Number 2: Figure S2 motif_logos_for_14_significant_regulons.pptx;

The motif logos substracted from 14 significant regulons, used for TFBS comparison.

Supplement Number 3: Dataset S1 NC_009004_operons_and_genes.xlsx;

The operon prediction results for MG1363.

Supplement Number 4: Dataset S2 122_operons_to_clusters.xlsx;

The gene clustering results for microarray "GSE47012" containing operons.

Supplement Number 5: Dataset S3 51_motif-regulons.xlsx;

The summarized motif prediction results for regulon prediction.

Supplement Number 6: Dataset S4 14_verified_regulons.xlsx;

Computational validation results for predicted regulons.

Supplement Number 7: Dataset S5 differential_expression_genes.xlsx;

86 up-regulated genes and 55 down-regulated genes.

Supplement Number 8: Dataset S6 trans-memberane.xlsx;

The prediction results for trans-membrane proteins.