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A workflow for genome-wide mapping of archaeal transcription factors with ChIP-seq

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ABSTRACT

Deciphering the structure of gene regulatory networks across the tree of life remains one of the major challenges in postgenomic biology. We present a novel ChIP-seq workflow for the archaea using the model organism Halobacterium salinarum sp. NRC-1 and demonstrate its application for mapping the genome-wide binding sites of natively expressed transcription factors. This end-to-end pipeline is the first protocol for ChIP-seq in archaea, with methods and tools for each stage from gene tagging to data analysis and biological discovery. Genome-wide binding sites for transcription factors with many binding sites (TfbD) are identified with sensitivity, while retaining specificity in the identification the smaller regulons (bacteriorhodopsin-activator protein). Chromosomal tagging of target proteins with a compact epitope facilitates a standardized and cost-effective workflow that is compatible with high-throughput immunoprecipitation of natively expressed transcription factors. The Pique package, an opensource bioinformatics method, is presented for identification of binding events. Relative to ChIP-Chip and qPCR, this workflow offers a robust catalog of protein-DNA binding events with improved spatial resolution and significantly decreased cost. While this study focuses on the application of ChIP-seq in H. salinarum sp. NRC-1, our workflow can also be adapted for use in other archaea and bacteria with basic genetic tools.

INTRODUCTION

The dynamic modulation of gene expression is an important mechanism that allows organisms to sense and respond to changes in their environment. These changes in expression profiles are mediated by dynamic associations of transcription factors and their cognate regulatory regions, collectively known as gene-regulatory networks (GRNs) (1). Regulatory networks integrate complex cellular and environmental cues, orchestrating intricate phenotypes essential for physiology and development. The evolutionary rewiring of these regulatory circuits is thought to be an important driver of speciation (2). Elucidating the structure and function of GRNs is therefore a major research initiative in functional genomics and systems biology (3–8).

The characterization of GRN architecture has been driven by advances in experimental and computational methods for identifying genome-wide protein-DNA interactions (9-13). One such approach is chromatin immunoprecipitation (IP) coupled with high-throughput sequencing (ChIP-seq), a method that provides quantitative genome-wide mapping of target protein-binding events. ChIP-seq identifies protein-binding sites with improved spatial resolution and decreased cost relative to previous microarray-based ChIP-chip technologies (10). While ChIP-seq has become a widely used tool in eukaryotic systems, this method has been applied only once in a bacterial system (14) and there exist no instances of such work in archaea. The small size of bacterial and archaeal genomes makes this high-throughput sequence technology particularly attractive, as sample multiplexing can be used to dramatically reduce costs relative to microarray-based platforms.

Developing a ChIP-seq protocol for archaea would stimulate high-throughput characterization of GRNs,

The authors wish it to be known that, in their opinion, the first two authors should be regarded as joint First Authors.

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which are a nascent area of study relative to work in the other two domains of life. Archaea are essential drivers of global biogeochemical cycling, integral players in industrial applications and biomedically important organisms. Furthermore, the transcriptional apparatus of archaea exhibits properties of both eukaryotic and bacterial systems, making it an intriguing target for investigating basic principles of regulatory mechanisms across the tree of life (15). Improved understanding of archaeal information processing and transcriptional regulation has widespread applicability.

We present a novel ChIP-seq workflow for the archaea using the model organism Halobacterium salinarum sp. NRC-1 (Hb. NRC-1) and demonstrate its application for mapping the genome-wide binding sites of natively expressed transcription factors. Previous bacterial and archaeal ChIP methods have taken different approaches involving either costly protein-specific antibodies against native proteins (14) or a standard antibody against epitope-tagged target proteins that are constitutively overexpressed from a heterologous plasmid (16,17). This protocol combines these methods by employing a single, commercially available antihemagglutinin (HA) antibody against natively expressed recombinant target proteins. This ChIP-seq method maintains sensitivity and specificity with as little as $\sim 1 \text{ ml}$ of the typical bacterial or archaeal culture, making it suitable for high-throughput analyses. Multiplexing of samples during sequencing significantly decreases experimental costs relative to previous ChIP-chip methods, without diminishing sensitivity and specificity.

A complimentary bioinformatics method is presented for user-friendly identification of binding events using the Pique python package. Integration with the Gaggle toolkit streamlines the exploration and analysis of putative protein-binding sites (18,19). This end-to-end workflow for ChIP-seq of natively expressed proteins provides a suitable platform for large-scale studies of the structure and dynamic remodeling of GRNs. The first protocol of its kind for archaea, this method can be adapted for work in all bacteria and archaea with suitable genetic tools.

MATERIALS AND METHODS

Construction of the pRSK01 ligation-independent cloning vector for *Halobacterium salinarum* NRC-1

The plasmid pNBK07 (obtained from N. Baliga, Institute for Systems Biology, Seattle, WA) has been previously used to create targeted gene knockouts (17,20–22) in the *Hb. NRC-1* $\Delta pyrF$ uracil auxotroph strain (*Hb. NRC-1* $\Delta pyrF$). For this study, pNBK07 was modified to facilitate Gateway ligation-independent cloning of segments of DNA suitable for chromosomal modification by homologous recombination. Plasmid pNBK07 (sequence and maps in Supplementary Information) was digested with StuI (New England Biolabs cat. #R0187S, Ipswitch, MA), a blunt-end cutting restriction endonuclease. The digested vector was subsequently dephosphorylated with calf intestinal alkaline phosphates (New England Biolabs cat. #M0290S) and purified via agarose gel extraction. PCR primers m13F and m13R (Supplementary Table S1) were used to amplify a fragment of the pDONR221 vector containing an *att*P1 recombination site, the *ccdB* gene, a chloramphenicol resistance marker and an *att*P2 recombination site. This PCR product was ligated into the StuI-digested pNBK07 backbone to create the Gateway (Invitrogen, Carlsbad, CA) compatible pRSK01 vector. The pRSK01 vector was sequenced using the pNBK07F, pNBK07R, ccdB500F and ccdB500B primers. Sequences (plasmid and oligonucleotide) and plasmid maps are provided in Supplementary Information.

Construction of chromosomally tagged transcription factors

Chromosomally epitope-tagged transcription factors were made by way of site-specific homologous recombination in *Hb.* NRC-1 $\Delta pvrF$. This method is analogous to that used for making in-frame gene knockouts using the pNBK07 vector as previously described (22). Two different approaches were utilized to make the epitope-tagging constructs for this study. In the first approach, PCR-mediated splicing by overlap extension (SOEing) (23) was used to join two PCR products used to add a sequence encoding a region \sim 500 bp upstream of the bacteriorhodopsinactivator protein (bat) stop codon, an HA epitope coding sequence, a new stop codon and \sim 500 bp downstream of the bat genomic stop codon. PCR primers are listed in Supplementary Table S1. This PCR product was cloned into the StuI site of plasmid pNBK07, which was subsequently transformed into strain Hb. NRC-1 $\Delta pyrF$. A two-step, double-crossover process was followed to chromosomally insert the HA epitope. First crossover recombinants were selected by plating on 2% (w/v) complete media (CM) agar plates containing 20 µg/ml mevinolin. Second crossover recombinants were enriched by selecting on 2% CM agar plates containing 300 µg/ml 5-fluoroorotic acid (5-FOA). The absence of a functional pyrF gene is required for survival on 5-FOA, indicating loss of plasmid.

The second method for chromosomal epitope tagging was used to tag the general transcription factor tfbD. This method takes advantage of commercial DNA synthesis technologies and the new Gateway cloning compatible pRSK01 vector. In this case, a construct consisting of an attB1 recombination site, ~ 500 bp upstream of the tfbD stop codon, the sequence encoding an HA epitope tag, a stop codon, \sim 500 bp downstream of the *tfbD* chromosomal stop codon, and an attB2 recombination site were directly synthesized by Geneart (Invitrogen, Carlsbad, CA) and delivered, cloned, in a pANY backbone vector also encoding an ampicillin-resistance marker. This vector was used directly in an *in vitro* recombination reaction (Gateway cloning, Invitrogen, Carlsbad, CA) with the pRSK01 vector according to manufacturer's protocols to move the synthetic construct into pRSK01. Once the synthetic construct is inserted into pRSK01, the rest of the tagging procedure is identical to that used for pNBK07based tagging. We have also used a combination of SOEing and Gateway recombination to directly clone

PCR products, flanked by appropriate *attB* recombination sites, directly into pRSK01 (data not shown).

Verification of chromosomal tagging

The insertion of HA epitopes at the C-terminal ends of the chromosomally encoded *bat* and *tfbD* genes was verified both by PCR and DNA sequencing. The following PCR reactions summarized in Supplementary Figure S2 were conducted to verify insertion of the HA epitope.

The initial PCR screen (Reaction 1) verified the presence of the C-terminally tagged gene of interest in the cell using a forward primer (ct_gene_of_interest_a_F) located in the gene and a reverse primer complimentary to the HA epitope tag's sequence (HA_epiope_R). PCR products of the expected size indicate either successful chromosomal tagging or the presence of residual tagging vector in the cell.

Recombinant strains were also screened for the presence of chromosomally encoded *pyrF* using primers flanking the chromosomally encoded gene (k_*vng167g3*_e_F and k_*vng1673g*_d_R, Reaction 2). The presence of chromosomally encoded *pyrF* yields a 2050 bp PCR product, while the disrupted *pyrF* in the *Hb*. *NRC-1* $\Delta pyrF$ strain yields a PCR product of 712 bp. Reaction 2 was performed to verify that the *pyrF* gene from the plasmid has not reintegrated into the chromosome of the *Hb*. *NRC-1* $\Delta pyrF$ strain.

A second *pyrF* PCR screening (Reaction 3) was carried out to confirm that the plasmid carrying the *pyrF* had been cured and that *pyrF* had not recombined into the chromosome of the *Hb*. *NRC-1* $\Delta pyrF$ strain. Reaction 3 amplifies a region from 465 bp upstream of the *pyrF* stop codon to 70 bp downstream of the *pyrF* stop codon (primers k_*vng1673g_g_F* and k_*vng1673g_h_R*). This final reaction yields no product in the *Hb*. *NRC-1* $\Delta pyrF$ strain and its derivatives. In strains that do carry the *pyrF* gene, such as wild type *Hb*. *NRC-1*, or strains transformed with either the pNBK07 or pRSK01 plasmids, a 535 bp product is formed.

Finally, we screen specifically for plasmid-encoded copies of the *pyrF* gene using primers that amplify a segment of the *pyrF* encoded by the pNBK07 or pRSK01 vectors (*k_vng1673g_g_F* and o_pNBK07_a_R, Reaction 4). The absence of product confirms that the plasmid has been cured, when the reaction is run in conjunction with plasmid-containing positive control.

PCR products derived from PCR reaction using primers ct_gene_of_interest_a_F and ct_gene_of_interest_d_R on strains meeting all the criteria established by the verification Reactions 1–4 above were sequenced via standard Sanger sequencing to verify the integration of the HA epitope. Sequences are provided in the Supplementary Information. Tag integration was further verified by analyzing the genome re-sequencing data for each strain that was generated in the process of the ChIP-seq experiment.

Culture preparation

All cultures were grown in the standard CM for *H. salinarum* (250 g/l NaCl, 20 g/l MgSO₄, 2 g/l KCl,

3 g/L Na–Citrate, 10 g/l Oxoid peptone (Oxoid cat# LP0034) (Oxoid, Basingstoke, UK) supplemented with 50 mg/l uracil and filled to volume with distilled water. Cultures were revived from -80° C freezer stocks and were streaked on agar plates. Starter cultures were inoculated from individual colonies and allowed to reach an optical density at 600 nm of 0.7 before inoculating a culture at a starting optical density at 600 nm of 0.03. Cells were grown under ambient light conditions in unbaffled flasks in a volume equal to 25% of the flask's maximum volume. All cultures were grown at 37 °C and shaken at 150 rpm on a New Brunswick G-53 orbital shaker (New Brunswick, Edison, NJ).

Immunoprecipitation

Biological replicates were conducted as inoculations in separate but identical volumes on the same orbital shaker. Cells were harvested at an OD 600 between 0.9 and 1.0, which corresponds to early stationary phase for *Hb.* NRC-1. Cells were immediately fixed with 1% (v/v) formaldehyde for 10 min. Fixing was stopped through the addition of glycine to a final concentration of 125 mM. Batches of 1.75×10^{10} cells were removed and pelleted at $5000 \times g$. Cell pellets were washed twice with citrate-free basal salts, after which the pellets were frozen at -80° C. Though this was our standard input for the IP reaction, we also examined the effect of decreasing the number of input cells for the IP reaction using the *tfbD*::HA strain. In addition to 1.75×10^{10} cells, these scaling experiments also used 8.75×10^9 , 3.50×10^9 , 1.75×10^9 and 3.50×10^8 cells as input material for IP. The rest of the method is described in detail for 1.75×10^{10} cells. Appropriate volumes and quantities of reagents for the scaled-down experiments are reported in Supplementary Table S3.

Cell pellets were resuspended in 1.6 ml of lysis buffer (50 mM HEPES, 140 mM NaCl, 1 mM EDTA, 1% (v/v) Triton X-100, 0.1% (w/v) sodium deoxycholate, pH 7.5) containing protease inhibitors (Roche cat# 04693159001). Resuspended pellets were sonicated using a Bioruptor (Diagenode, Denville, NJ) until DNA fragment size reached an average of ~500 bp (2–7.5-min cycles, 30 s on/30 s off, high power setting).

Cell lysate was combined with 1 µg of anti-HA antibody (Abcam cat# ab9110) (Abcam, Cambridge, MA) and protein A-conjugated Dynabeads (Invitrogen cat. # 100.2D) preblocked with 5 mg/ml BSA in phosphatebuffered saline and incubated overnight at 4°C. Dynabeads were washed two times with the lysis buffer, two times with 1 ml of the lysis buffer supplemented with 500 mM NaCl, two times with 1 ml wash buffer (10 mM Tris, 250 mM LiCl, 0.5% NP-40 (v/v), 0.5% Na-deoxycholate (w/y), 1 mM EDTA, pH 8.0) and one time with 1 ml TE buffer. Enriched ChIP DNA/transcription factor complexes were eluted by the addition of $50 \,\mu$ l elution buffer (50 mM Tris, 10 mM EDTA, 1% SDS (w/v), pH 8.0) and incubation at 65° C for 10 min. Cross-links were reversed by incubating in TE/SDS (10 mM Tris, 1 mM EDTA, 1% SDS) overnight at 65°C. RNA was digested and DNA sample was subsequently prepared for Illumina single read sequencing.

ChIP-seq library preparation

Individual ChIP samples were blunt ended with T4 DNA polymerase (NEB cat. # M0203L), Klenow large fragment (NEB cat. # M0210L) and T4 polynucleotide kinase (NEB cat. # M0201L) at 20°C for 30 min. Blunt-ended DNA was 3' A tailed with 3' > 5' exo- Klenow fragment (NEB cat. # M0212L) for 30 min at 37°C. Adapters containing 6 bp barcodes were ligated to the prepared ChIP DNA samples for 15 min at room temperature with T4 DNA ligase (Enzymatics cat. # L603-HC-L). Barcode sequences are provided Supplementary Table S4. A background control of whole cell extract genomic DNA from each sample was prepared as mentioned above. Samples were then used as template for an 18-cycle PCR amplification. PCR products were quantified and visualized with a high-sensitivity DNA chip (Agilent cat. # 5067-4626) on a bioanalyzer (Agilent, Santa Clara, CA). ChIP and background libraries were pooled in equimolar concentrations and loaded onto a single Illumina lane.

Western blotting

Transcription factors were immunoprecipitated under the same conditions as ChIP methods mentioned above. IP samples were run in one dimension on 4–12%, 1.5 mm polyacrylamide gel (Invitrogen cat. # NP0335) in MOPS buffer (Invitrogen cat. # NP0001). Protein was then blotted onto a 0.2- μ m pore size PVDF membrane (Invitrogen cat. # LC2002) at 30 V for one hour in transfer buffer (25 mM Tris, 192 mM glycine, 10% (v/v) methanol, pH 8.4). PVDF was blocked in 0.5% (w/v) casein overnight and subsequently probed with HRP-conjugated anti-HA antibody (Abcam cat. # ab1265). The blot was incubated with GE ECL plus reagents (Amersham cat. # RPN2132) according to the manufacturer's suggestions and exposed to light-sensitive film.

qPCR verification

qPCR was performed on a Bio-Rad Chromo 4 Real-Time Detector (Bio-Rad, Hercules, CA) using KAPA SYBR[®] FAST Universal 2× qPCR master mix (Kapa Biosystems cat.# KK4601) (Kapa Biosystems, Woburn, MA) according to the supplied protocol. Primer sets for enriched regions and negative regions were designed using known enriched sites and unenriched sites, respectively, from previous ChIP-seq and ChIP-chip data (see Supplementary Table S1 for primer sequences). Fold enrichment above background was calculated as 2 to the power of cycle threshold difference between a non-enriched region and an expected enriched site. WCE extract, ChIP samples and amplified libraries were all used as template for a qPCR reaction. These were all confirmed by comparing to a set of ChIP-control reactions on the Hb *NRC-1* $\Delta pyrF$ strain.

Sequencing, processing and ChIP-seq peak calling

Multiplexed samples were sequenced to 40 bp on the Illumina GA-II. Sequences were barcode sorted and quality trimmed (minimum Phred quality 20, minimum length 25 bp) using the FASTX-Toolkit (http://

hannonlab.cshl.edu/fastx_toolkit/) (Gordon,A and Hannon,G.J., unpublished results). Sequencing primer and adapter contamination were filtered using the TagDust package (24). Quality-filtered reads were mapped using Bowtie (25) to the *Hb. NRC-1* reference genome with repeat sequences masked, and SAM format sequence files were converted to sorted BAM files using the samtools package (26).

Putative protein-DNA binding events were detected using Pique, a novel microbially focused and freely available peak calling application (available at https:// github.com/ryneches/pique, version tag: halo egw) (Neches, R.Y., Wilbanks, E.G. and Facciotti, M.T., in preparation). Pique is written in Python and makes use of the SciPy signal-processing subroutines (27). Pique is able to operate on systems that have genomic complexities such as IS elements, gene dosage polymorphisms and accessory genomes that cause coverage variations unrelated to ChIP, or in cases where the organism under study is not identical to the reference genome. The resulting enrichment 'pedestals' and 'holes' can be problematic for accurately detecting binding events and calculating enrichment levels. Hb. NRC-1 has several IS elements and two plasmids that exhibit dosage variations, and so a segmented analysis was performed by providing a genomic map of these features in the reference genome.

ChIP-seq coverage data and candidate peaks were visualized using the Gaggle Genome Browser (18). Shared peaks were assessed using a combination of BEDTools (28) and custom R scripts. To assess the required sequencing depth for accurate and sensitive binding site identification, random subsampling of a 6 million read TfbD ChIP and WCE control runs were performed.

RESULTS

Epitope-tagged strain construction

We developed a protocol for rapidly engineering the strains of the *Hb. NRC-1* with epitope-tagged target proteins under the control of their native promoters. Using different classes of transcription factors, we demonstrate this approach's application and utility. In bacteria and archaea, different transcription factors may bind a wide span of target sites, ranging from one to hundreds. We chose to collect localization data from two extreme cases. The general transcription factor TfbD is known to bind hundreds of promoters (16). As an example of a specific regulator of a smaller regulon, we examined the Bat transcription factor. This transcription factor predicted to bind up to four potential sites and is one of the few haloarchaeal transcription factors with a well-described binding motif (29).

Our epitope-tagging protocol for *Hb. NRC-1* employed a homologous recombination method analogous to the gene deletion strategy developed by Peck *et al.* (30). This epitope knock-in strategy was described by Zhang *et al.* for ChIP-chip applications in human somatic cell lines (31). For *tfbD*::HA strain construction, the novel vector pRSK01, which is compatible with Gateway ligation-independent cloning (Invitrogen, Carlsbad, CA), was created to facilitate and further standardize strain construction. This new vector allowed the use of either commercial DNA synthesis technology or PCR-mediated SOEing (23) with Gateway compatible primers to rapidly construct the vectors used for strain construction. While in most instances PCR SOEing is still less expensive, the decreasing cost of DNA synthesis should make this approach a more attractive alternative for future large-scale strain construction projects.

Hb. NRC-1 $\triangle pvrF$ was transformed with a recombinant plasmid that contained the terminal 500 bp of the target gene, a hemagglutinin (HA) tag, stop codon and 500 bp downstream sequence (Figure 1). Homologous recombination between the chromosomal target gene and the recombinant plasmid sequence introduced the HA tag to the chromosomal sequence. Successful first recombinants were determined by PCR screening of Mev^R colonies (Supplementary Figure S1). The plasmid was subsequently resolved using 5-FOA counter-selection as previously described (22,30). Strains with C-terminally HA-tagged target proteins were further verified by PCR and Sanger (Supplementary Figures sequencing S1–S2 and Supplementary Information). For the rapid construction of epitope-tagged transcription factor strains, this general strategy of utilizing DNA synthesis and homologous recombination-based chromosomal modification can be readily extended to any organisms with a system for targeted genetic knockouts.

Western blotting with anti-HA antibody confirmed the specificity of the ChIP assay in these chromosomally (Supplementary Figure S3). tagged strains The chromosomally integrated, epitope-tagged proteins remain under the control of their native promoters, as observed in the differential expression of the TfbD-HA protein over the course of growth in the recombinant strain $\Delta pyrF$ tfbD::HA (Supplementary Figure S4). Increase in the abundance of TfbD during stationary phase is consistent with previous reports of tfbD transcriptional abundance (20). This approach can be used to monitor dynamic changes in the GRN that occur under different physiological conditions.

Identifying target protein DNA-binding sites

We used ChIP-seq to analyze with two different classes of target proteins: the general transcription factor TfbD and the Bat using recombinant strains that natively express the target protein ($\Delta pyrF$ bat::HA and $\Delta pyrF$ tfbD::HA). From the ChIP-seq datasets of 1.2 million reads for each factor, we identified 380 binding sites for TfbD and two binding sites for Bat (the *brp* and *crtB1* promoters).

These punctate target protein DNA-binding events produce a distinctive bimodal, strand-specific enrichment pattern in sequence coverage (Figure 2). This enrichment pattern was leveraged to identify binding sites using our open source software package Pique, which reports the candidate binding site's enrichment as the ratio of sequence coverage in the IP data relative to a background control (Neches, R.Y., Wilbanks, E.G. and Facciotti, M.T., in preparation). Pique is implemented from a user-friendly





Figure 1. Epitope tag-in approach for Hb. NRC-1. The Hb. NRC-1 $\Delta pyrF$ strain is transformed with a plasmid containing the mevinolin-resistance determinant (Mev^R; dark gray box) and the pyrFgene (black box) that confers 5-FOA sensitivity. The plasmid carries an engineered sequence containing the HA epitope sequence (white box) flanked by the last 500 bp of the target gene and the 500 bp downstream of the target gene (light gray boxes). Plasmid sequence is shown as solid line, chromosomal sequence is shown as solid, wavy lines. Cross-over can occur between target gene (light gray box) and flanking sequence (gray wavy line) in the chromosome and the homologous regions in the plasmid sequence, at either position 1 or 2 (position 1 example shown). PCR screening of mevinolin-resistant colonies is used to determine successful first recombinants. Subsequent plating on 5-FOA selects for second recombinants (via counter-selection with the pyrF gene). In this example, a second cross-over at site 2 produces the desired chromosomally integrated recombinant target_gene::HA fusion. PCR screening of these colonies is required to distinguish this desired second recombinant from a second recombinant occurring at position 1. Drawing is not to scale.

graphical user interface and exports predicted binding sites to the Gaggle Genome Browser (18). From the Gaggle genome browser, users can explore and curate the data before proceeding with analysis via other downstream Gaggle tools (Figure 3).

To confirm the specificity of the HA antibody for the ChIP assay on the recombinant HA-tagged strains, we conducted ChIP-seq on the $\Delta pyrF$ parent strain where no HA tags were expressed. The data contained a single peak, at chromosomal position 166589, likely because of the presence of a similar epitope in a native protein. This peak was present in all datasets examined and was subsequently filtered from all downstream analysis.



Figure 2. The 5' to 3' sequencing requirement and short read length produce stranded bias in sequence coverage. The shaded blue oval represents the protein of interest bound to DNA (solid black lines). Wavy lines represent either sense (blue) or antisense (red) DNA fragments from ChIP enrichment. The thicker portion of the line indicates regions sequenced by short read sequencing technologies. Sequenced tags are aligned to a reference genome and shown below is the strand-specific sequence coverage at each position in the genome. Punctate-binding events (e.g. transcription factors) are characterized by well-defined strand-specific bimodality in sequence coverage.

Independent biological replicates of the TfbD ChIP-seq good experiment showed reproducibility (82%) overlapping binding sites). Differences between the two replicates are due to the smallest peaks in each dataset, as indicated by examining increasingly stringent enrichment thresholds (Supplementary Figure S5). Both binding sites in the Bat dataset and two example sites in the TfbD datasets (VNG906H and *atp* p promoters) were confirmed with a ChIP-qPCR assay. The relationship between the quantification of ChIP enrichment at the binding sites determined by qPCR and sequencing was found to be well correlated across several experiments (Figure 4). The TfbD ChIP-seq binding sites agreed well with previously reported sites from TfbD ChIP-chip experiments (16). For both the ChIP-seq and ChIP-chip methods, 80% of consensus binding sites from biological replicates were identified by at least one of the replicates from the other method.

Spatial resolution

The spatial resolution of the ChIP-seq binding site prediction was assessed for the Bat and TfbD datasets. As the Bat transcription factor has a well-characterized predicted binding motif (29), we used the distance from our predicted binding site to the motif center as an estimate of the spatial resolution. The binding sites at the *brp* and *crtB1* promoters were found, respectively, at 20 and 27 bp upstream of the predicted Bat-binding motif center (3' displaced).

As there exists no well-defined binding motif for the general transcription factor TfbD, we used proximity to the nearest predicted transcript start site (TSS) as a measure of spatial resolution for this factor. Archaeal TFB proteins, homologs of the eukaryotic factor TFIIB, canonically bind at B-recognition elements \sim 30–50 bp upstream of the TSS, in association with a TATA-binding protein (32). Eighty-six percent of the 312 consensus binding sites from the TfbD ChIP-seq biological replicates were found within 250 bp of a predicted TSS (in either 3' or 5' direction). We measured the distance from each of these predicted binding sites to the nearest predicted TSS. These values were compared to the distance to TSS from previously reported TfbD ChIP-chip sites, determined with both 500-bp contiguous and 12-bp overlapped tiling microarray (Figure 5) (16). The distance to TSS for the ChIP-seq binding sites (median = 32, mean = 51) is in agreement with the expected binding pattern for this general transcription factor, and is significantly smaller than that predicted by both resolutions of ChIP-chip microarray (Mann Whitney U-test, p value < 0.005). The variance in these distance measurements provides an estimate of the precision with which each method maps binding sites. The ChIP-seq dataset has significantly decreased variance relative to both ChIP-chip datasets (significance assessed by Bartlett test, p value < 0.005; samples were tested for normality by two-sided Kolmogorov-Smirnov p value < 0.0005). Taken together, these data indicate that the ChIP-seq assay offers improved spatial resolution in mapping the target protein-binding site relative to prior ChIP-chip assays.

ChIP assay cell number and sequencing depth

We investigated the number of cells required for ChIP to produce sufficient enrichment at target protein-binding sites. Decreasing the number of cells in the ChIP reaction lowered the observed enrichment at target protein-binding sites; however, enrichment (>5x) was detectable for highly enriched TfbD-binding sites when as few as 3.50×10^8 cells were used for ChIP, equivalent to 1 ml of a typical culture (Figure 6A). Because of the overall decrease in enrichment, smaller cell number ChIPs were less sensitive in binding site detection but maintained specificity (Figure 6B). The relatively small volume required for this ChIP assay should enable the high-throughput application of this method in the context of dynamic binding studies by allowing for the repetitive sampling of numerous strains with minimal perturbation.

One of the main advantages to the ChIP-seq platform for small microbial genomes is the ability to decrease the experimental cost by multiplexing many samples in a single sequencing lane. We carried out an *in silico* analysis to determine the depth of sequencing necessary to achieve sensitive and accurate detection of binding sites. Sequence reads were randomly subsampled to decreasing coverage levels from 1.2 M reads



Figure 3. The Pique software package processes ChIP-seq coverage data to predict protein-binding sites. Strand-specific coverage data are output as tracks for the Gaggle Genome Browser, and putative-binding sites (peaks) are output as 'bookmark files'. (A) Screenshot of data browsing in the Gaggle Genome Browser. Green box outlines the navigation window for clicking through bookmarks of predicted binding sites. Details of each site can be displayed (inset). The Gaggle toolbar (shown with black arrow) can be used to broadcast selected data to other 'geese' in the gaggle package, programs such as R, cytoscape, BLAST or KEGG. (B) Schematic overview of bioinformatics workflow.



Figure 4. ChIP enrichment of binding sites determined by qPCR and sequencing show a linear relationship. Data shown are drawn from multiple ChIP experiments: the Bat ChIP (P_{brp} and P_{crtB1} closed and open circles) and the TfbD ChIP and the reduced cell number TfbD ChIPs ($P_{VNG906H}$ and P_{atp_p} ; closed and open triangles). Differences in enrichment at the TfbD-bound promoters corresponded to changes produced by decreasing the number of cells in the ChIP reaction (see Figure 5 for further details).

(15.4 x coverage) to 10 000 reads (0.13 x coverage). For the TfbD dataset, the number of peaks identified remained stable down to 500K reads (5 x coverage), after which the sensitivity began to decrease (Figure 7). The specificity of binding site identification remained excellent below 500 K reads, even though the sensitivity decreased (Figure 7).

For the Bat dataset, the more strongly enriched binding site at the *brp* promoter could be detected in datasets as small as $50\,000$ reads (0.64 x coverage), whereas the



Figure 5. Distance from predicted TfbD-binding site for ChIP-seq (consensus between biological replicates), 500-bp tiling microarray ChIP-Chip (consensus between biological replicates) and 12-bp tiling microarray ChIP-Chip experiments. The observed difference in means was statistically significant (Mann Whitney U-test, p value < 0.005), as is the observed difference in variance (Bartlett test, p value < 0.005).

weaker binding at the crtB1 promoter was undetectable below 150 000 reads (1.9 x coverage). No false positives were detected in these lower coverage datasets, with the exception of a single site in the 200 000 read dataset. Examining the effect of decreased coverage on the spatial resolution, we found that the automated prediction of the binding site remained accurate, until just before the site became undetectable (Figure 8).

DISCUSSION

We report here a workflow for the genome-wide mapping of archaeal natively expressed transcription factors using a standardized, cost-effective, highthroughput ChIP-seq platform. This is the first example of a ChIP-seq protocol for archaea. The development of



Figure 6. TfbD ChIP with decreasing numbers of cells. (A) ChIP-qPCR determined fold enrichment at two test promoters (VNG906H, dark bars and *atp_p* light bars) decreases with lower numbers of cells; however, strong (>5 x) enrichment is still observed with 3.5×10^8 cells. (B) For decreasing cell volume ChIP-seq experiments, fewer peaks could be identified (number of peaks identified, squares and lines), resulting in a significant loss in sensitivity. However, the percentage of identified peaks that were true positives stayed high (% true positives, triangles). True positives were defined as binding sites that were shared with at least one of the optimized 1.75×10^{10} ChIP TfbD biological replicates.

this high-throughput method for mapping protein–DNA binding events in archaea should catalyze the investigation of the GRNs in this third domain of life.

While major advances have been made in mapping the GRNs of bacteria and eukaryotes, the GRNs of archaea represent a nascent area of exploration, with only a handful of genome-wide experimental studies (8,16,17,33,34). Understanding regulatory mechanisms in archaea will greatly inform our understanding of the basic biology of this important domain, relevant to diverse fields of study from biogeochemistry to biotechnology. The development of improved methods for surveying archaeal GRNs is timely, coinciding with a new wave of archaeal genome sequencing that has suggested many conserved archaeal regulatory mechanisms (34,35).

Archaea also possess an intriguing mosaic transcriptional apparatus that exhibits properties of both eukaryotic and bacterial systems. While the basal transcriptional machinery of archaea is homologous to that



Figure 7. Subsampling sequence coverage. Sequences were randomly sampled from a TfbD ChIPseq dataset of 6M reads to create subsampled datasets of decreasing coverage. The number of peaks that could be identified in each subset (filled circles) is shown as a function of the number of sequence reads in the dataset. The specificity of these smaller lists is assessed as the percentage of the identified peaks which overlapped the larger 1.2M read dataset (solid line).



Figure 8. Spatial resolution of binding sites calculated from randomly subsampled Bat ChIP-seq datasets of decreasing coverage. The distance from ChIPseq predicted binding site to nearest Bat-binding motif was calculated for both the crtB1 (dark gray bars) and brp (light gray bars) promoter regions. Binding sites for in the crtB1 promoter could not be detected for datasets with 100 000 reads and below (no bars in plot).

of eukaryotes, archaeal transcriptional regulators are more similar to those of bacteria. Studies of archaeal transcription have provided insight into both the mechanisms and evolution of information processing in the three domains of life (32,36). Likewise, deciphering archaeal GRNs holds a great potential for advancing our understanding of fundamental principles employed by GRNs across the tree of life.

The workflow presented here is cost efficient and amenable to high-throughput scaling. To increase throughput and minimize costs, we relied on recombinant strains with low-profile HA epitope-tagged target proteins and a standard anti-HA antibody for the ChIP assay. The HA epitope tag in conjunction with the well-characterized commercial antibody proved efficient for IP while minimally perturbing the target protein. The choice of the sterically slim HA epitope tag can prove quite important; we have found that the larger repeated myc epitope tag (37) can render some DNA-binding proteins nonfunctional. The HA tag does not disrupt Bat function, as seen in its ability to complement a bat knockout mutation and induce bacteriorhodopsin production (Supplementary Figure S6).

A transcription factor's occupancy of possible binding sites depends, in part, on its concentration within the cell. As our ultimate goal is to map the dynamics of regulatory network rearrangement, native expression of the target transcription factor, rather than constitutive expression from a plasmid construct, was an important feature of our approach. To accomplish this, we used recombinant target proteins that were chromosomally integrated under the control of the wild-type promoter. The construction of these recombinant archaeal strains required the development of a method for generating chromosomally integrated recombinant proteins in *Hb. NRC-1*, thereby expanding the genetic toolbox available for this model archaeon.

Previous ChIP-chip studies in Hb. NRC-1 used target proteins that were constitutively expressed at nonnative levels from a heterologous plasmid (16,33,38). The resultant protein-DNA associations are, therefore, perhaps best viewed as lists of all possible interactions rather than a snapshot of protein–DNA association network under physiological conditions. While this approach is appropriate for some applications and can offer technical advantages, such as improving ChIP efficiency for proteins present in low abundance, expression of target proteins at nonnative levels can produce artifacts in the list of protein-DNA binding sites. In the simplest case, constitutive overexpression can drive transcription factor association to weak or nonspecific sites without significantly perturbing expression. Ambiguities concerning which binding sites are physiologically relevant can sometimes be resolved by incorporating data such as transcriptomes and regulatory motifs in the analysis. However, the perturbation of transcription factor expression can also have more serious consequences that cannot be easily resolved, such as unintended protein-protein interactions and changes in the cellular phenotype. Lastly, constitutive expression precludes investigating nonnative the dynamics of transcription factor association, a fundamental aspect in understanding the relationship between the GRN structure and function.

We demonstrated the application of our ChIP-seq protocol on two different classes of archaeal transcription factor: a general transcription factor with many binding sites (TfbD) and a more specific transcriptional activator (Bat). ChIP-seq data were analyzed with the user-friendly, open-source Pique package, designed for identifying protein–DNA binding events in small bacterial and archaeal genomes. Our bioinformatics pipeline integrates with the Gaggle toolkit to facilitate downstream data visualization, curation and analysis. The predicted binding sites were consistent between biological replicates and with previously published ChIP-chip results for TfbD. We observed few significant trends in the gene classes bound by TfbD, with the exception of gene functionally associated with GTPase activity (p value = 1×10^{-4}). The lack of obvious functional partitioning of TfbD target genes is unsurprising, given this factor's broad role in global transcription initiation. Dynamic ChIP-seq experiments under different physiological conditions would likely be an appropriate future method for determining the potential regulatory roles carried out by TfbD and other archaeal general transcription factors.

The two Bat-binding sites discovered in the *brp* and *crtB1* promoters (P_{brp} , and P_{crtB1}) were verified by qPCR and contain two of the four previously reported occurrences of the Bat regulatory motif (P_{bop} and P_{blp} were not bound) (29). It seems initially surprising that Bat binding was not detected upstream of the bacteriorhodopsin apoprotein (*bop*), a gene it regulates (29,39–41). However, recent research has shown that Bat regulation of *bop* expression is complex and may work cooperatively with accessory proteins Brz and Brb (42,43).

Interestingly, the Bat-binding motif at P_{brp} , and P_{crtBI} share a single nucleotide insertion relative to the two unbound motif sites at P_{bop} and P_{blp} . Furthermore, the spacer sequence between the Bat motif and the TATA-box is shorter at P_{brp} and P_{crtBI} sites (2 and 3 bp spacer, respectively), relative to the unbound P_{bop} and P_{blp} motif occurrences (5 bp spacer) (29). We note that these small differences in the Bat motif, in concert with our binding data, may provide some preliminary evidence to suggest a way by which Bat (and potential coregulators) distinguishes between the four predicted binding sites in a condition-specific manner.

ChIP-seq identifies protein-binding sites with fine spatial resolution and provides accurate estimates of binding site enrichment. The quantification of enrichment found at protein binding sites calculated by ChIP-seq was very similar to that determined by ChIP-qPCR (Figure 5). Unlike ChIP-chip enrichment values, which become saturated at high levels of enrichment, ChIP-seq has excellent dynamic range, and thus provides an accurate metric for the level of enrichment at target protein-binding sites. A narrow size selection of chromatin immunoprecipitated DNA fragments enhances the enrichment in sequence coverage at target-binding sites. The use of automated size-selection instruments, such as the Pippin Prep[®] (Sage Science, Beverly, MA), in the preparation of ChIP-seq libraries may improve data quality.

The number of cells required for the ChIP assay was investigated to determine an optimal protocol that balances sensitivity of binding site detection with throughput and ease of sample handling. Decreasing the number of cells for the ChIP assay was found to decrease the enrichment level at target protein-binding sites, and thus the sensitivity of the assay. However, the more strongly enriched sites could still be accurately detected with 3.50×10^8 cells, equivalent to ~1 ml of a typical culture. The false-positive rate remains very low in the lower cell number ChIP-seq experiments. The ability to use low cell counts as input makes this approach tractable for high-throughput assessment of the more prominent binding sites in the genome, though precludes development of an exhaustive list of possible protein–DNA interactions.

By randomly subsampling deeply sequenced datasets, we determined that the required sequence coverage for the sensitive detection of binding sites corresponds to $\sim 6.5 \,\mathrm{x}$ coverage of the complete genome (approximately 500 K reads in Hb. NRC-1). We estimate that for the average sequencing run on the Illumina HiSeq (80 million reads) and the typical bacterial and archaeal genome $(\sim 3 \text{ Mb})$, 130 samples can be multiplexed per lane. With this level of multiplexing, the ChIP-seq assay would cost roughly \$15 per sample. The per-sample cost is expected to drop even further with continuing improvements in the output of sequencing technologies. Relative to ChIP-chip, this ChIP-seq workflow greatly reduces the experimental cost of defining the genome-wide binding sites of target transcription factors while also improving spatial resolution. From gene tagging to data analysis, this workflow provides an excellent model for conducting large-scale, dynamic mapping of bacterial and archaeal gene regulator networks.

ACCESSION NUMBERS

All high-throughput sequencing data generated in this work are available via BioTorrents (http://www .biotorrents.net/details.php?id=259) or from our lab website (www.bme.ucdavis.edu/facciotti/resources_data/ data/). The open access Pique package and source code can be obtained via github at https://github.com/ryneches/ pique. Primer and plasmid sequences used in this study are available in Supplementary Information.

SUPPLEMENTARY DATA

Supplementary Data are available at NAR Online: Supplementary Tables 1–4, Supplementary Figures 1–6, Supplementary Information and Supplementary Reference (44).

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