# 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 *SUPPLEMENTARY MATERIALS and METHODS*  Bacterial Strains and Growth Conditions Serotype M1 GAS strains commonly cause pharyngitis and invasive infection<sup>1</sup>. Strain MGAS5005 (ATCC # BAA-947), a wild-type (WT) clinical strain (serotype M1), and its isogenic ∆*covR* derivative strain (JRS950) have been described<sup>2-4</sup>. Bacteria were cultured statically on Trypticase soy agar containing 5% sheep blood agar (Becton Dickinson, Cockeysville, Md.), or in Todd-Hewitt (TH) broth (Becton Dickinson) containing 0.2% (wt/vol) yeast extract (THY; Difco Laboratories, Detroit, MI), at  $37^{\circ}$ C in an atmosphere containing 5% CO<sub>2</sub>. Mouse Soft-Tissue Infection Model This model of GAS soft-tissue infection has been used extensively to study bacterial-host interactions<sup>2-5</sup>. Our experimental protocol was approved by the Institutional Animal Care and Use Committee, National Institute of Allergy and Infectious Diseases (NIAID). Bacterial strains MGAS5005 and JRS950 were grown to late exponential (LE) phase (OD<sub>600</sub> $\sim$ 0.75) in THY broth, washed twice in pyrogen-free Dulbecco's PBS (DPBS), and resuspended as inocula in pyrogenfree PBS. Cells were harvested at LE phase to limit infectivity differences associated with upregulated capsule biosynthesis, which is maximal in the early-to-mid exponential growth phases. Immediately before inoculation, the animals were weighed and anesthetized with isoflurane (Aerrane; Ohmeda Caribe, Guayama, P.R.) inhalation. Five-week-old (20- to 25-g) outbred,



## 18 Experimental Design

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20 21 22 23 A one-factor experimental design with two treatment levels was used for array experiments in this study. Mice (*n* = 54) were assigned at random to one of the two treatment groups (WT or ∆*covR* mutant GAS strain, each  $n = 27$ ). As data quality is of paramount importance, we took care during the design of our custom oligonucleotide Affymetrix GeneChip® (designated RMLChip herein) to

1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 exclude cross-hybridizing sequences and to use randomized blocking in our experimental design. Samples were randomized before all preparation procedures and we ensured that batches of sample preparation, hybridizations, and post-hybridization washes were not confounded with treatment. For example, samples were randomly allocated to positions on a single 96-well plate in a 6 x 9 rectangle near the center of the plate to protect against positional biases such as thermal edge effects and other unknown positional effects. In order to protect against the possibility of random "unlucky" positional allocation, 10,000 random assignments were generated and then each assignment was tested for right-to-left, top-to-bottom, and edge-to-center imbalances in allocation of wild type and *covR*-minus. These tests were performed by first computing the distance of each sample to the top, left, and middle of the plate. For each random assignment, three *t*-tests were performed to test for a difference between wild type and *covR*-minus in the three distances and the minimum *P-*value recorded. Finally, the results of the 10,000 random allocations were sorted by the minimum of the three *P-*values. The randomization with the largest minimum *P-*value was selected as the actual assignment of samples to positions. To minimize confounding variables, samples were placed in the same position for each 96-well plate used during sample preparation (e.g., RNA Extraction II, cDNA synthesis, cDNA cleanup, *in-vitro* transcription, clean-up, concentration adjustments, and target fragmentation).

18 19 20 21 22 23 Batches for RNA extraction I were performed with 12 samples per batch. For all steps including hybridization and wash batches, samples were run sequentially in such a way that each batch was balanced to include equal numbers in each batch of both WT- and ∆*covR-* infected samples. An alternating RMLChip scanning order (WT / ∆*covR* / WT / etc.) also was used to balance scanner effects over time evenly across treatment groups. All samples were run on the same GeneChip<sup>®</sup> lot to avoid unnecessary noise in the data.

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# 2 RNA isolation

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### 1 cDNA Labeling

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# 18 GeneChip® Design

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20 21 22 23 A custom 18-micron feature size, anti-sense oligonucleotide array representing approximately 249,690 25-mer probe-pairs (16 probe pairs, each consisting of one perfect match probe (PM) and a mismatch probe (MM), per probe set) manufactured by Affymetrix Inc. was used as described<sup>7-9</sup>. Each probe-set is used to detect the presence of a single transcript. To facilitate analysis of GAS



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# 18 GeneChip® Hybridization

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20 21 22 23 24 Target hybridizations, washing, staining, and scanning were performed by the NIAID Affymetrix core facility (SAIC-Frederick, MD) using a GeneChip® hybridization oven and the *Pseudomonas aeruginosa* hybridization protocol (Affymetrix, Santa Clara, CA). The hybridization solution volume used was 200 µl because the RMLChip is a standard size array. Each array was scanned at 570 nm at 3-µm resolution with a GeneArray® scanner. Scanned DAT-image files were 1 2 3 analyzed with Affymetrix<sup>®</sup> Microarray Suite (MAS) 5.0 software (Affymetrix, Santa Clara, CA). The raw CEL-files have been submitted to Gene Expression Omnibus (www.ncbi.nlm.nih.gov/geo/).

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## 6 Microarray Data Quality Assessment and Statistical Analysis

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21 22 23 24 Randomization and blocking enabled us to assess the respective error variance contributed by unavoidable (technical) experimental variables. Technical variables included in the analysis included sample preparation, sample hybridization batches, and post-hybridization wash batches. For example, during array washing the fluidic station could process only a maximum of 8 slides at

1 2 3 4 5 6 7 8 9 a time and wash batch was found to contribute variance (Supplementary Figure 1). Downstream data analysis was performed on the composite array data with MGAS5005-specific probe sets using Partek Pro<sup>TM</sup> (Partek Inc., St. Louis, MO). Image files were converted to text format, and normalized according to a symmetric square root transformation. To evaluate expression rankings, the normalized absolute square root expression estimates were integer-ranked such that the most abundant transcript reported as transcript rank "1". Integer rank assignments were increased correspondingly with ever decreasing transcript detection. To investigate expression correlations between genes, standard Pearson correlation coefficients were determined for select genes *versus* all other genes.

10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 Lastly, to investigate the strain effect, the resultant normalized expression estimates were analyzed by Analysis of Variance (ANOVA) with treatment (WT versus ∆*covR* strain) as a fixed effect. Final results were subjected to multiple testing correction using  $Q \leq 0.05$  false discovery rate (FDR) cutoff values<sup>13</sup>. Using rigorous permutation-based statistics, we also performed significance analysis of function and expression (termed  $SAFE$ )<sup>14</sup> to assess the significance of multiple gene categories in GAS *in vivo* transcriptional responses across strains. All GAS probe sets were assigned to one of 17 functional categories (including "Unknown") as described<sup>7</sup>. For each gene, the Student's *t*-statistic was calculated as local statistic measuring the association between expression and treatment (strain). The Wilcoxon rank sum was then computed as global statistic to assess how the distribution of local statistics within a functional category differed from local statistics outside the category. To account for multiple testing of functional categories, empirical *P*-values for each functional category were obtained by recomputing Wilcoxon statistics across 10,000 permutations of the array assignments in which treatment assignments were randomized. Unknown correlation among genes was conserved across permutations using this approach and allowed computation of permutation-based estimates for the  $FDR<sup>14</sup>$  using the

1 2 3 4 5 6 7 Benjamini-Yekutieli method<sup>15</sup>. In the resultant SAFE plots,  $P$ -values would trace the identity line (solid diagonal) representing the null hypothesis if no differential expression were present. When small *P*-values occur with greater frequency than expected by chance, then the cumulative distribution function for a gene category diverges from the diagonal, and responsible genes are designated as differentially regulated. Low-ranking, negative *t*-statistics correlate to upregulated expression in the mutant strain (left side of the SAFE plot); conversely positive *t*-statistics correlate to downregulated expression in the mutant.

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10 Quantitative Real-Time PCR Analysis

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12 13 14 15 16 17 18 19 20 Real-time reverse transcription-PCR (RT-PCR) assays were conducted to validate a subset of the microarray data<sup>7</sup>. Eight oligonucleotide primer pairs and 6FAM-labeled probe sets (specific for *cfa, dppA, emm1, sceD*, *sclA*, *sic*, *slo*, and *speA*2) were used to perform target amplification and detection from cDNA templates in 20  $\mu$ L multiplex two-step RT-PCR reactions as described<sup>7</sup>. Targets were selected to encompass the full range of expression signal values identified by array transcriptome analysis. Target abundance was normalized to JOE-labeled internal reference transcript *proS*, which is transcribed at constant levels throughout the GAS growth cycle *in vitro* and not affected by  $\text{cov}R$  inactivation<sup>3</sup>. Differences in median values were evaluated for statistical significance with the Mann-Whitney Rank Sum Test at the  $P \le 0.001$  level.







# 1 *REFERENCES*





