The genome-wide transcriptional response to varying RpoS levels in *Escherichia*

coli* K-12*.

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Running title: Effect of RpoS concentration on *E. coli* transcriptome

Abstract

The alternative sigma factor RpoS is a central regulator of a many stress responses in

Escherichia coli. The level of functional RpoS differs depending on the stress. The effect

of these differing concentrations of RpoS on global transcriptional responses remains

unclear. We investigated the effect of RpoS concentration on the transcriptome during

stationary phase in rich media. We show that 23% of genes in the *E. coli* genome are

regulated by RpoS level, and we identify many RpoS-transcribed genes and promoters.

We observe three distinct classes of response to RpoS by genes in the regulon: genes

whose expression changes linearly with increasing RpoS level, genes whose

expression changes dramatically with the production of only a little RpoS ("sensitive"

genes), and genes whose expression changes very little with the production of a little

RpoS ("insensitive"). We show that sequences outside the core promoter region

determine whether a RpoS-regulated gene in sensitive or insensitive. Moreover, we

show that sensitive and insensitive genes are enriched for specific functional classes,

and that the (in)sensitivity of a gene to RpoS corresponds to the timing of induction as

cells enter stationary phase. Thus, promoter sensitivity to RpoS is a mechanism to

coordinate specific cellular processes with growth phase, and may also contribute to the

diversity of stress responses directed by RpoS.

Importance

The sigma factor RpoS is a global regulator that controls the response to many

stresses in *Escherichia coli.* Different stresses result in different levels of RpoS

production, but the consequences of this variation are unknown. We describe how

changing the level of RpoS does not influence all genes equally. The cause of this

variation is likely the action of transcription factors that bind the promoters of the genes.

We show that the sensitivity of a gene to RpoS levels explains the timing of expression

as cells enter stationary phase, and that genes with different RpoS sensitivities are

enriched for specific functional groups. Thus, promoter sensitivity to RpoS is a

mechanism to coordinate specific cellular processes in response to stresses.

Introduction
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Genome-wide measurements of RNA levels have revolutionized our understanding

of how cells organize their patterns of transcription. These studies have given us

snapshots of how patterns of gene expression change in response to changes in the

external environment. They have also allowed us to define the regulons controlled by

specific transcription factors. A major weakness of the vast majority of these studies is

that they explore the function of a regulatory protein only by comparing expression of

target genes in a wild-type strain to a gene knock-out, or to a mutant with a single

diminished or increased level of activity. While some genetic regulatory networks are

certainly switch-like (1, 2) and can be fully characterized by only two levels of activity,

many other regulatory proteins vary continuously in their abundance and/or activity.

How regulons respond to a range of regulator levels is a largely unstudied question (3).

Should we expect all genes in a regulon to increase or decrease expression by the

same relative amount following a change in abundance of a regulatory protein? Or

should we expect genes to respond in different ways? These questions motivate this

study.

A paradigmatic example of a bacterial regulatory protein whose abundance and

activity vary continuously in response to different conditions is the alternative sigma

factor RpoS in *Escherichia coli*. Transcription by RNA polymerase containing RpoS is

responsible for the general stress response (4–6). Under conditions of optimal growth in

the laboratory (such as exponential phase growth in rich media at 37°C), RpoS levels

are nearly undetectable in the model *E. coli* strain K-12. As conditions become poorer

for growth, either because cells begin to starve for various nutrients, or because they

face physical challenges such as low temperature or elevated osmolarity, RpoS levels

rise (4–6). RpoS coordinates the transcription of genes that are critical for the response

to these stresses.

RpoS expression is not an all-or-none phenomenon. For example, RpoS levels rise

continuously as cells transition from exponential growth to stationary phase (7).

Moreover, starvation for different nutrients upregulates RpoS to differing levels (8, 9).

This level of control over RpoS levels is accomplished by regulating transcription,

translation, and degradation (4, 7, 10), allowing for careful control over protein levels. In

addition to regulation of protein abundance, RpoS activity can also be directly

modulated by a number of factors, such as Crl (11).

Not only do RpoS levels vary across conditions for a single strain, but different

strains of *E. coli* also differ in their patterns of expression of RpoS. For example,

naturally occurring strains can differ in the amount of RpoS they produce during

exponential phase (12) or stationary phase (13). All studies that have measured RpoS

levels in naturally occurring strains of *E. coli* have detected variation (13–17), though the

extent and cause of this variation in RpoS between strains is still a matter of some

controversy (16, 17).

Microarray studies (18–20) have shown that RpoS controls the expression of at least

500 genes (over 10% of the genome) either directly or indirectly, but the set of RpoS-

regulated genes differs across environmental conditions. For most RpoS-controlled

genes, it is not known whether the gene is directly transcribed by RpoS or regulated

indirectly, as a consequence of RpoS transcribing other genes. Previous studies have

not investigated the impact of changes in RpoS levels on the RpoS regulon, or whether

quantitative differences in RpoS levels between environmental conditions influence the

observed differences in what genes are RpoS-regulated. It is clear that *E. coli* has a

complicated regulatory network to fine-tune RpoS levels to different conditions, but we

do not yet know the consequences of this regulation.

In this study, we tested the hypothesis that members of the RpoS regulon vary in

their response to RpoS levels. Using a combination of ChIP-seq and RNA-seq, we

identified RpoS-regulated genes, and we showed that genes vary in their sensitivity to

RpoS levels in a manner dependent on sequences outside of the core promoter region.

Sensitivity of genes to RpoS levels corresponds to the order in which genes are induced

during the transition into stationary phase, and genes with different levels of sensitivity

are enriched for specific functional groups. Thus, the level of sensitivity of genes to

RpoS controls the physiological response to different stress conditions.

Materials and Methods
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Culture conditions

Cells were grown in 5mL of LB broth (1% tryptone, 0.5% yeast extract, 1% NaCl) in

150 x 18 mm tubes shaking at 225 rpm at 37°C in a water bath unless otherwise

specified in the text. When required, antibiotics were used at the following final

concentrations: ampicillin at 100 μg/ml (for plasmids) or 25 μg/ml (for chromosomal

integration); chloramphenicol at 20 μg/ml; kanamycin at 50 μg/ml.

Strains and genetic manipulations

Strains and plasmids used in this study are listed in Table S1. The wild-type genetic

background for this all experiments except for ChIP-seq is BW27786, a strain designed

to give a graded transcriptional response to increasing arabinose concentration (21). To

create a strain of this background lacking *rpoS*, the *ΔrpoS746::kan* allele of JW5437

(22) was moved by P1 transduction into BW27786, creating strain DMS2545.

The arabinose-inducible RpoS strain was created by PCR amplifying the *kan* gene

and the P$_{araB}$ promoter of plasmid pAH150 (23) using primers ParaBRpoSRecomb-F

and ParaBRpoSRecomb-R (Table S2). This PCR product was then integrated into the

nlpD gene (i.e. 5' of *rpoS*) in a MG1655 background using plasmid pKD46 (24), and P1

transduced into BW27786, creating strain DMS2564. This strain thus lacks both the

native transcriptional and translational control of RpoS.

For flow cytometry experiments, plasmid pDMS123 (25), which contains a *otsB-gfp*

transcriptional fusion, was transformed into DMS2564 by the method of Groth et al. (26).

ChIP-seq experiments used strain RPB104, an unmarked derivative of MG1655 that

expresses a C-terminally SPA-tagged derivative of RpoS from its native locus. This

strain was constructed by P1 transduction of kan[R]-linked *rpoS*-SPA from a previously

described strain (27). The kan[R] cassette was removed using the pCP20 plasmid that

encodes Flp recombinase (24).

Construction and chromosomal integration of lacZ fusions

The *gadB* and *astC* plasmids were built by standard cloning methods. The *gadC*

and *astA* promoter regions with transcription factor binding sites were PCR-amplified

with primers gadCpromoter +/- and astApromoter +/- (Table S2), which included *Kpn*I

and *Eco*RI restriction sites for cloning. Cloning of core promoter regions was performed

by annealing oligonucleotides designed to contain the whole RpoS binding region, as

predicted by Fraley *et al.* (28) and Castanie-Cornet and Foster (29). Oligonucleotides

were annealed by heating 1 μM of forward and reverse primer for one minute at 100°C

with 5 mM MgCl$_2$ and 7 mM Tris-HCl and then cooling slowly to room temperature.

Inserts and plasmid (pLFX) were digested with *Eco*RI-HF and *Kpn*I-HF (NEB), ligated

with T7 ligase (NEB) and cloned into strain BW23473. Transformants were miniprepped

and inserts were verified by Sanger sequencing.

The *gadA* and *hdeA* plasmids were built using Gibson assembly (30) with the

NEBuilder HiFi Assembly kit (New England Biolabs). The *gadA* and *hdeA* promoter

regions were PCR amplified with primers hdeAHiFi+/- and gadAHiFi+/-. The core

promoter was cloned with the single long oligo hdeAcoreHiFi or gadAcoreHiFi, as

predicted by Arnqvist et al (31) and De Biase et al (32). PCR products or

oligonucleotides were mixed with pLFX digested with *Kpn*I-HF and *Eco*RI-HF, and

assembled according to the manufacturers' instructions. Mixtures were cloned into

strain BW23473, transformants were miniprepped, and inserts were verified by Sanger

sequencing.

lacZ fusion plasmids were integrated into strain DMS2564 with helper plasmid

pPFINT (33). Single-copy integrants were confirmed using the PCR assay of Haldimann

and Wanner (23).

Quantitative Western Blotting

Quantitative western blotting was used to measure RpoS levels. Cells were

inoculated from frozen cultures into 5 mL of LB, and grown overnight at 37 °C, shaken

at 225 rpm. 5 μL of this overnight culture was diluted into 5 mL of LB with the

appropriate concentration of arabinose, and grown for 20 h. 100 μL of overnight culture

to be assayed was centrifuged and resuspended in 1X Laemmli sample buffer (Sigma-

Aldrich) and boiled for 5 minutes. Samples were diluted 1:10 in 1X Laemlli, and 10 μL

was electrophoresed on a 10% polyacrylamide gel (Bio-Rad) in tris-glycine running

buffer (25 mM Tris base, 250 mM glycine, 0.5% SDS) at 100 V for 90 min at room

temperature. Proteins were transferred to an Immobilon-FL PVDF membrane by

electrophoresis at 100 V for 45 min at 4 °C in transfer buffer (48 mM Tris base, 39 mM

glycine, 20% methanol, and 0.0375% SDS). Membranes were blocked by overnight

incubation in Odyssey blocking buffer (Li-Cor) at 4 °C.

The blocked membrane was probed with affinity-purified monoclonal antibodies

to RpoS (clone 1RS1) and RpoD (clone 2G10) (NeoClone) at a final concentration of 0.4

μg / mL in Odyssey blocking buffer plus 0.2% Tween 20 for 1 hour at room temperature.

The membrane was washed four times for five minutes each with 15 mL of 1X TBST. A

fluorescent secondary antibody (IRDye 800CW goat anti-mouse, Li-Cor) was diluted

1:10,000 in a solution of Odyssey Blocking Buffer plus 0.2% Tween 20 and 0.01% SDS

and incubated in the dark for 1 hour at room temperature. The membrane was washed

as before, dried for 2 hours between sheets of Whatmann 3MM blotting paper, and

imaged on a LiCor Clx fluorescent imager.

Band intensity was estimated using Image Studio 2.1 (LiCor). RpoS levels were

divided by RpoD levels to normalize for differences in total protein levels. The ratio of

RpoS to RpoD is biologically meaningful because this ratio, rather than RpoS level

alone, dictates levels of transcription from RpoS-dependent promoters due to sigma

factor competition (34).

RNA-seq experiments and analysis

Cells were inoculated from frozen cultures into 5 mL of LB, and grown overnight. 5

μL of this overnight culture was diluted into 5 mL of LB and grown for 20 hours. (For the

26% RpoS condition, a final concentration of 10^{-4}% arabinose was also added.) RNA

was purified from 200 μL of overnight culture by pelleting and resuspending in 500 μL of

Trizol at 65 °C, followed by purification on a column (Direct-Zol, Zymo Research).

Samples received two 30-minute DNase treatments using TURBO DNA-free (Ambion)

following the manufacturer's instructions. RNA samples were then purified on a column

(RNA Clean & Concentrator, Zymo Research). Samples were stored at -80 °C until

used. Three samples were prepped from each culture and pooled to generate sufficient

RNA. Two biological replicates were prepared for each strain or condition of interest.

rRNA depletion, cDNA synthesis, library preparation, and sequencing were performed

by a commercial provider (Otogenetics, Norcross, GA). Paired-end, 100 bp sequences

were generated for 7-15 million reads per sample.

Before reads were mapped, the first ten base pairs of each read were trimmed using

FASTX-Toolkit 0.0.13. Reads were mapped to the NCBI K12 reference genome

(NC_000913.2 Escherichia coli str. K-12 substr. MG1655) using BWA v0.7.5a (35).The

number of read pairs mapped to each gene was counted with HTSeq 0.6.1 (36).

Differential expression analysis was performed with DESeq v2.13 (37). All p-values

were first FDR adjusted using the procedure of Benjamini and Hochberg (38). p-values

were then further Bonferroni-adjusted for the three comparisons between pairs of RpoS

levels. All differential expression p-values reported in this paper reflect both the FDR

and Bonferroni-adjustments.

To determine if a gene differed significantly from the null expectation of linearity, we

calculated the probability of the observed read count value at 26% RpoS if the true

expression level was given by the linear prediction between the endpoints. Note that to

calculate both the expected read count and the probability of the observed value under

the null (i.e. the p-value) our model required estimating both the DEseq size factor (for

scaling) and the dispersion (a variance factor) for each of the samples. The negative

binomial probability model (routinely used to measure count data, (e.g. (37)) was used,

with the size factors and dispersion estimated from DESeq (37) to calculate the

probability of the observed read count at 26% RpoS.

GO term analysis was performed using the topGO package (39) together with the

org.EcK12.eg.db annotation package (40) in R 3.1.0. Enrichment was assessed using

the weight01 algorithm (39) together with the Fisher's-exact test. The GO hierarchy was

pruned to include only nodes with at least five associated genes, as significance tests

can be unstable for GO terms with fewer genes (39).

A Venn diagram of the number of significant genes in each condition was prepared

with EulerAPE 3.0.0 (41).

ChIP-seq analysis

E. coli strain RPB104 (MG1655 with C-terminally SPA-tagged *rpoS*) was grown

overnight in M9 minimal medium with 0.4% glycerol at 30 °C and then subcultured 1:100

in the same medium, and grown for 60 hours to saturation (OD_{600} of ~3). ChIP-seq

using the M2 monoclonal anti-FLAG antibody was performed as described previously

(42). Regions of enrichment ("peaks") were identified as described previously (43).

Relative enrichment was reported as a "Fold Above Threshold" (FAT) score.

We used MEME-ChIP (Version 4.11.2; default parameters) to analyze enriched

regions identified by ChIP-seq (44) Regions within 100 bp were merged. The reported

sequence motif was identified by MEME (Version 4.11.2) (45), run within the MEME-

ChIP environment.

We used MEME (45) within the MEME-ChIP environment (Version 4.11.2; default

parameters except that only the given strand was analyzed) to identify enriched

sequence motifs in regions surrounding TSSs associated with RpoS ChIP-seq peaks.

We analyzed sequences from -45 to +5 relative to each TSS.

We identified directly RpoS-transcribed genes by requiring that (i) the gene start is

within 300 bp of a ChIP-seq peak, (ii) the gene is positively regulated by RpoS, as

determined by RNA-seq, (iii) no other positively regulated gene starts within 300 bp of

the ChIP-seq peak, (iv) there is no associated sequence motif or TSS (identified in (46))

that would be consistent with transcription in the opposite orientation. For regulated

genes, we determined whether other genes in the same operon were also regulated,

using a published operon list (47) for *E. coli.* Peaks were associated with transcription

start sites from a previous study (46) by identifying transcription start sites within 20 bp

of a peak.

QPCR

RNA was isolated as for RNA-seq, except that samples underwent three 30-

minute DNase treatments using TURBO DNA-free (Ambion). cDNA was made from the

RNA samples using SuperScript VILO master mix (Invitrogen) and stored at -20 °C for

use in quantitative real-type PCR (qPCR) reactions.

qPCR was performed using Power SYBR Green master mix (Invitrogen), 2 μL of

cDNA, and primers at 300 nM. Cycling was performed at 95 °C for 10 min; and 40

cycles of 95 °C for 15 sec, 58 °C for 15 sec, and 68 °C for 30 sec. Three control genes

(*ftsZ, pgm,* and *hemL*) were used in addition genes of interest. These genes were

selected using the approach of Vandesompele et al. (48). Details of this selection

process, including the other seven genes tested, are available in Supplementary Text 1,

Table S3, and Figure S1.

For each gene, a standard curve was made using the following amounts of

genomic DNA: 200 ng, 20 ng, 2 ng, 200 pg, 20 pg, and 2 pg. Genomic DNA was

extracted from overnight cultures using a Puregene Kit (Qiagen), following the

manufacturer's instructions. Expression levels for each gene were interpolated from the

standard curve. Expression levels of experimental genes were divided by the geometric

mean of the three control genes.

Statistical assessment of sensitivity was performed by a bootstrapping approach.

Bootstrapping, rather than a parametric approach, was appropriate because the data

did not conform to parametric assumptions. In this approach, the unit of resampling was

the RNA isolated from the 0%, 26% and 100% RpoS conditions on an individual day.

For each resampled data set, the median expression at 0% and 100% RpoS was

calculated. The linear fit from the 0% RpoS median and the 100% RpoS median was

then used to predict the level of expression at the intermediate level of RpoS. Repeating

this re-sampling of the data 10,000 times yielded a 95% confidence interval for 26%

RpoS. The observed median for 26% was then compared to the confidence interval to

test for significance.

Beta-galactosidase activity

Beta-galactosidase activity was measured using the method of Miller (49). With

lacZ fusion strains, the level of sensitivity was quantified, rather than only categorized as

sensitive, linear, or insensitive. To quantify the level of sensitivity, for each replicate we

calculated the distance between the observed expression at the intermediate RpoS

concentration and the expected level based on a linear pattern, standardized by the

difference in expression between high and low RpoS conditions. Testing if sensitivities

were different from zero used a one-sample t-test, with p-values adjusted for multiple

comparisons using Holm's sequential adjustment method (50). Testing if two

sensitivities were different used a two-sample t-test with the same method of adjustment

for multiple comparisons.

Analysis of published RNA-seq data

We analyzed published RNA-seq data for wild-type and Δ*rpoS E. coli* over a

time-course of growth into stationary phase (47). Using normalized genome coverage

information extracted from wiggle files, we calculated relative abundance for all genes at

each of the four stationary phase time-points in the growth curves (the last four time-

points for each strain). We arbitrarily selected a threshold coverage value of 500 and

excluded any genes scoring below this threshold at all four time-points in wild-type cells.

This reduced variability associated with low expression levels. We excluded any genes

for which coverage at the final time-point was 0 since this would have prevented

normalization. We also excluded any gene for which the first stationary phase time-point

had the highest expression value of the four time-points for wild-type cells, since RpoS-

dependent expression of these genes is likely to be masked by other factors. We then

selected genes whose expression we had found to be induced by RpoS, and separated

these genes into insensitive, linear, and sensitive classes. We calculated expression

levels for each of these genes relative to expression at the final time-point.

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Results

The RpoS regulon in late stationary phase.

To understand the role of the RpoS protein in late stationary phase, we used RNA-

seq to compare the transcriptome of wild-type and Δ*rpoS* cells. We observed differential

expression of 1044 genes (23% of genes) between these two conditions ($p < 0.05$). Of

the 1044 genes whose expression is influenced by RpoS, 605 are upregulated, and 439

are downregulated (Table S4).

Influencing transcription of 23% of the genome could have many potential

phenotypic effects. To better understand the function of these genes, we examined

which kinds of gene functions, as described by the gene ontology (GO), are more

abundant in the regulon than expected by chance. GO enrichment analysis indicates

that the RpoS regulon includes many genes involved in metabolic processes (Table 1);

17 of 18 significantly enriched GO terms are metabolic terms. This metabolic

reorganization includes the upregulation of genes encoding glycolytic enzymes and

pathways for metabolism of L-arginine to glutamine, or from L-arginine to putrescine and

then into succinate. RpoS also drives the downregulation of genes involved in the TCA

cycle. These patterns of metabolic regulation are very similar to those identified in late

stationary phase in *Salmonella enterica* (51)*.* The only significant GO term not explicitly

linked to metabolism (GO:0006970, "response to osmotic stress") also includes

metabolic genes, such as *otsA* and *otsB,* that are involved in trehalose biosynthesis.

Central metabolism is not the only phenotype similarly regulated by RpoS in both *S.*

enterica and *E. coli*. Other similarities include transcription of genes involved in

antioxidant activities, iron regulation, and Fe-S cluster assembly, the upregulation of

proteases, and down regulation of porins. As in *S. enterica,* RpoS in *E. coli* influences

the expression of many genes encoding other regulatory proteins, including *csrA, arcA,*

cra, fur, ihfA, hupA and *hupB.* Not all of the regulation is identical between *E. coli* and *S.*

enterica, however. For example, Lévi-Meyrueis *et al.* (51) noted that RpoS appears to

direct switching between many pairs of isozymes. While some pairs show a similar

pattern in *E. coli* (such as *tktA/B* and *acnA/B*), others (such as *fumA/fumC*) do not show

this pattern.

Genome-wide binding profile of RpoS

While RNA-seq identifies which genes are regulated by RpoS, it cannot distinguish

between direct and indirect effects. To determine sites where RpoS binds (and hence

likely plays a direct role in transcription), we used ChIP-seq to map the association of

RpoS across the *E. coli* chromosome during stationary phase growth in minimal medium.

To facilitate ChIP, RpoS was C-terminally SPA-tagged at its native locus. We reasoned

that RpoS would only be identified with promoter regions since it is likely released from

elongating RNAP complexes. We identified 284 peaks of RpoS ChIP-seq signal

covering 260 genomic regions (peaks within 100 bp of each other were merged). 217 of

the RpoS-bound regions are intergenic, and 67 are located within genes. We reasoned

that annotated genes that are transcribed by RpoS would be positioned close to an

RpoS-bound region. Consistent with this, 213 RpoS-bound regions are ≤300 bp

upstream of an annotated gene start. These 213 regions include 27 that are intragenic.

In 79 cases, we observed an RpoS-bound region ≤300 bp from the starts of two

divergently transcribed genes.

We used MEME to search for enriched sequence motifs within the RpoS-bound

regions. We detected a highly enriched motif in 107 regions; this motif closely

resembles the known -10 hexamer recognized by RpoS (Figure 1A). Moreover,

occurrences of the motif are positionally enriched with respect to the ChIP-seq peak

center (Figure 1B), indicating that the data have high spatial resolution. Note that the

motifs tend to be located just upstream of the peak centers (Figure 1B), as we have

observed previously for the *E. coli* flagellar Sigma factor, FliA (43). This presumably

reflects the fact that the footprint of initiating RNA polymerase associated with RpoS is

not centered on the -10 hexamer.

The identification of RpoS binding sites allowed us to better understand the role of

RpoS in both positive and negative regulation. While similar proportions of the RpoS

regulon are positively and negatively regulated by RpoS, it is not clear if this is true at

the level of direct regulation. Only 19 genes within 300 bp of a binding site are

negatively regulated by RpoS. 19 of the 286 such genes is *fewer* than we would expect

by chance if binding sites were randomly distributed around the genome, given than 439

of 4513 genes in the genome are negatively regulated. On the other hand, 111 of 286

genes within 300 bp are positively regulated, a highly significant effect (Fisher's exact

test, $p < 10^{-16}$). Thus, the binding profile of RpoS is consistent with direct positive

regulation of many genes, but provides no evidence of direct negative regulation.

Identification of RpoS-transcribed genes and promoters

We combined the ChIP-seq and RNA-seq data to identify genes that are directly

transcribed by RNA polymerase containing RpoS. Thus, we identified 123 RpoS-

transcribed genes in 99 transcripts (Table S5), and compared them to other published

analyses (18–20, 46, 52–55). In some cases, we identified RpoS-bound regions

upstream of genes that were not detected as being RpoS-regulated by RNA-seq. These

genes may have promoters that bind transcriptionally inactive RNA polymerase (56, 57).

Alternatively, the disparity between RpoS binding and regulation could be explained by

differences in growth conditions between the ChIP-seq and RNA-seq experiments, or by

the possibility that the C-terminal SPA tag affects the function of the C-terminus of the

protein in the response to transcription activators (56, 58).

The high spatial resolution of sigma factor ChIP-seq can facilitate the

identification of specific promoters (43, 59) when combined with nucleotide resolution

transcription start site (TSS) maps. Using published TSS data from *E. coli* under

stationary phase conditions similar to those used for the ChIP-seq experiment (46), we

determined all pairwise distances between RpoS ChIP-seq peaks and stationary phase

TSSs. We observed a strong enrichment for peak-TSS distances ≤20 bp (Figure S2).

We infer that these 112 TSSs are RpoS-transcribed. Consistent with this, the putative

RpoS-transcribed TSSs are associated with -10 hexamers that have features expected

of RpoS promoters (Figure 1C). In some cases, RpoS promoters were identified ≤300

bp upstream of genes that were not RpoS-regulated in the RNA-seq data set. We

presume that these are RpoS-transcribed genes that otherwise escaped detection

because of differences in the growth conditions used for ChIP-seq and RNA-seq.

The transcriptome at three different RpoS levels

This first view of the RpoS regulon considers RpoS as either present or absent.

RpoS levels vary continuously across environmental stresses (7), so we sought to better

understand how the level of RpoS in the cell influences transcription of target genes. To

do this, we placed the *rpoS* gene under the control of the arabinose-inducible promoter

P$_{araB}$. This promoter was integrated just upstream of the native *rpoS* gene, placing

transcription under the control of arabinose concentration and removing the 5' region

that regulates translation of the native mRNA (4).

To measure the resulting arabinose-induced expression of RpoS, we employed

quantitative western blotting. RpoS levels increased with increasing arabinose

concentration, from undetectable RpoS levels, to levels similar to those in wild-type cells

(Fig 2a). To confirm that expression was graded and not all-or-none in this system (21),

we used flow cytometry to measure expression in individual cells. We transformed the

arabinose-inducible RpoS strain with the plasmid pDMS123 (25), which contains the

RpoS-dependent *otsBA* promoter fused to *gfp*. As expected, *gfp* expression increased

with increasing arabinose concentrations, and at each expression level the population

was unimodal (Fig 2b).

We measured the transcriptome in cells with 26% of wild-type levels of RpoS,

achieved with the addition of 10^{-4} % arabinose to cells with our arabinose-inducible *rpoS*

strain. Of the genes that are differentially expressed between 26% RpoS and either 0%

(ΔrpoS) or 100% (wild-type), 95% are also differentially expressed between 0% and

100% (Fig 3) ($p < 0.05$).

Nearly all genes that are significantly differentially expressed have monotonically

increasing or decreasing patterns of expression between three levels of RpoS. Only

two genes (*ytfR* and *ytfT*) have an expression level at 26% rpoS that is significantly

higher than the expression at both 100% and 0% RpoS. The only two genes with

expression lower in the 26% RpoS condition than either 100% or 0% are *nlpD* and *pcm*.

These genes lie immediately upstream of *rpoS*. *nlpD* was removed from the genome

during the construction the arabinose inducible RpoS strain. The *pcm* gene is still

present, but the level of transcription was lowered by the genetic modification.

Classifying quantitative responses to RpoS levels

To explore how genes respond to changing levels of RpoS, we developed a new

metric, *sensitivity.* Our null expectation was that gene expression would increase

linearly with increasing RpoS concentration. We observed many genes in our RNA-seq

data set (such as *osmY*) whose expression at intermediate RpoS levels falls on, or

close, to a line drawn between the 0% and 100% RpoS conditions (Fig 4a). We refer to

these genes as *linear* in their response to increasing RpoS levels. Other genes (such as

astA) are transcribed more at 26% than would be expected based on their expression

levels at 0% and 100% (Fig 4b). We refer to these genes as *sensitive*, because only a

small amount of RpoS results in relatively high levels of transcription. In contrast, some

genes (like *gadC*) are expressed at intermediate RpoS levels less than expected based

on expression at 0% and 100% RpoS; such genes are referred to as *insensitive* (Fig 4c).

We identified 910 linear, 102 sensitive, and 32 insensitive genes. 96% of sensitive

genes and 88% of insensitive genes are positively regulated by RpoS. In contrast, only

53% of linear genes are positively regulated by RpoS, a significant difference (chi-

square test, $p < 0.001$).

To determine whether sensitive or insensitive (hereafter *(in)sensitive*) genes are

associated with specific physiological responses to increasing RpoS levels, we again

used GO enrichment. We tested the null hypothesis that the functions of these genes

are a random sample from the entire RpoS regulon (not the whole genome). The GO

terms significantly enriched in the sensitive class are response to osmotic stress,

cellular amino acid catabolic process, and fatty acid oxidation (Table 2). Several genes

encoding regulators are among the sensitive genes, including *arcA*, which encodes a

global regulator of respiratory metabolism, and *rssB,* which encodes the adaptor protein

required for degradation of RpoS by ClpXP.

GO enrichment is less useful for understanding the possible function of the

insensitive gene set. Three GO terms are enriched (Table 3), but only a few genes with

these annotations are present in the insensitive gene set, and their enrichment probably

reflects the relatively small number of insensitive genes. More strikingly, the insensitive

genes include nearly all of the genes required for acid resistance system 2: the

structural genes *gadA, gadB,* and, *gadC*, and the regulator of this system *gadE* (60, 61).

In addition, the genes *yhiM, yhiD*, *hdeA, hdeB , hdeD, mdtE, mdtF,* all of which have

been described as having roles in acid resistance (60, 61), are insensitive.

We used reverse transcription coupled to qPCR to confirm the expression

patterns of two insensitive genes (*gadC* and *gadE*), and three sensitive genes (*prpR,*

prpD, and *astA*). All genes were positively regulated by RpoS (Fig 5), and the median

expression at 26% RpoS is consistent with RNA-seq expression patterns for all genes.

We used a bootstrapping approach to assess if expression at 26% was significantly

above or below the linear expectation at 26% RpoS. The expression of *gadC* was

significantly insensitive ($p < 10^{-4}$), and *prpR* expression was significantly sensitive ($p <$

10^{-4}). *astA* expression was marginally significantly sensitive ($p = 0.06$ for sensitivity),

while *gadE* and *prpD* were not significantly different from the linear expectation ($p =$

0.10 and $p = 0.56$ respectively).

Control of (in)sensitivity of expression

What makes one promoter sensitive to RpoS levels and another promoter insensitive

to RpoS levels? We hypothesized three possible mechanisms. First, chromosomal

location could determine the response to RpoS levels, as is known to occur in the

context of total transcription levels (62). Second, it is possible that the DNA sequence of

the core promoter drives the response. Finally, it is possible that the binding of

transcription factors upstream of the core promoter influences the response to RpoS

levels. To test these hypotheses, we cloned the promoters (including all upstream

transcription factor binding sites annotated in EcoCyc (63)) of four operons into the *lacZ*

fusion plasmid pLFX (33). The four promoters were the sensitive *astCADBE*, and the

insensitive *gadA, gadBC,* and *hdeAB-yhiD.* Plasmid pLFX recombines into the lambda-

attachment site, placing the fusion in a novel genomic context. While we did not detect

binding of RpoS upstream of *astC, gadA, gadB,* or *hdeA* by ChIP-seq, this is likely due

to the difference in growth conditions, since these genes have been previously shown to

be directly transcribed by RpoS (31, 64, 65).

The pattern of transcription of all four fusions was the same as observed for the

respective genes in the RNA-seq data (Fig 6 a-d). *astC* transcription was sensitive to

RpoS levels (one sample t-test, p = 0.04), while *gadA, gadB*, and *hdeA* transcription

were all insensitive (one sample t-test, $p = 2 \times 10^{-6}$, $p = 10^{-5}$, p = 0.04 respectively).

Since all reporters were placed at the same genomic locus, this result suggests that

genomic location is not the determinant of response to RpoS levels.

A second potential mechanism to explain the difference between sensitive and

insensitive genes is interactions between RpoS and the core promoter sequence. For

example, specific nucleotides (or combinations of nucleotides) might tend to confer

(in)sensitive patterns of transcription. The majority of both sensitive and insensitive

genes were not associated with RpoS-bound regions in the ChIP-seq experiment,

suggesting that they are indirectly regulated by RpoS (Table 4). The fact that most

sensitive and insensitive genes are not bound by RpoS argues against the hypothesis of

direct RpoS-DNA interactions driving (in)sensitivity. To see if specific sequence motifs

were consistently associated with (in)sensitivity, we used the discriminative motif search

feature of DREME (66) to search for motifs that differed between sensitive and linear, or

insensitive and linear regulatory sequences. There are 28 ChIP-seq peaks associated

with an operon with at least one sensitive gene, and 4 ChIP-seq peaks associated with

an operon with at least one insensitive gene. We found no motifs that distinguished

these sets of sequences, although the small number of sequences would have

restricted the power of such a test.

To directly test the hypothesis that the core promoter region of RpoS-transcribed

genes is responsible for determining the sensitivity to RpoS levels, we cloned this short

region from the *astC, gadA, gadB, hdeA* promoters into pLFX and recombined the

plasmids into the chromosome. These core promoters had absolute levels of

transcription much lower than the entire promoter that included upstream transcription

factor binding sites (Fig 6e - h). The core promoter sequence for *astC* alone was

somewhat less sensitive to RpoS levels than the full-length construct (two sample t-test,

p = 0.08), although not significantly so, probably due to the variability of expression from

the full-length *astC* reporter. The *gadA* and *gadB* core promoters differed significantly

from their full-length promoters (two sample t-test, p= 0.01 and $p < 10^{-6}$, respectively).

The *hdeA* core promoter (Fig 6h) is not RpoS-dependent, showing a decline in

expression of approximately 10% in the presence of RpoS. This is consistent with the

previous finding that the ability of RpoD (but not RpoS) to transcribe *hdeA* is repressed

by H-NS protein bound upstream of the promoter (67). The core promoter construct

lacks the native H-NS binding sites upstream, and so the selectivity is apparently lost.

Thus, the core promoters do not replicate the RpoS sensitivity of their whole promoter

sequences. In addition, the three RpoS-dependent core promoters (Fig 6 e - g) do not

differ from each other in sensitivity (ANOVA, p = 0.32). We conclude that the core

promoter is not responsible for RpoS sensitivity.

Sensitivity to RpoS determines the timing of induction during entry into stationary

phase

We hypothesized that the degree of sensitivity to RpoS could impact the timing of

expression under conditions when levels of active RpoS increase, such as during entry

into stationary phase. A previous study used RNA-seq to monitor the transcriptome over

a time-course of growth, including four time-points in stationary phase (47). We

analyzed these data to determine whether insensitive, linear, and sensitive genes show

differences in the timing of induction. We selected only those RpoS-induced genes

whose transcription increased upon entry into stationary phase. (It total, there were 250

such linear genes, 90 sensitive, and 19 insensitive.) We then determined the pattern of

expression for each such gene over four time-points beginning at the onset of stationary

phase. Although there is considerable variability in the expression patterns of genes, as

a group the three classes show clear differences in the timing of induction (Figure 7a).

Specifically, sensitive genes are induced most rapidly, followed closely by linear genes.

In both cases, expression peaked early in stationary phase and fell between 30 and 180

minutes after entering stationary phase. In contrast, insensitive genes showed relatively

little change in expression until the final time-point, 180 minutes into stationary phase.

To determine the importance of RpoS on the patterns of gene expression, we repeated

the above analysis using data generated from a Δ*rpoS* strain. As expected, the large

difference in timing between the groups of genes was greatly diminished (Figure 7b).

We conclude that the (in)sensitivity of a gene is associated with the timing of expression

during stationary phase in an RpoS-dependent manner.

Discussion

Expanding the known set of RpoS-transcribed genes and promoters

The *E. coli* RpoS regulon has been widely investigated using targeted and genome-

scale approaches. Most genome-scale studies have focused on genes whose

expression is altered in the absence of RpoS (18–20, 53, 54). Hence, these studies

cannot distinguish between genes that are transcribed by RpoS, and those that are

indirectly regulated. ChIP-seq affords a high resolution view of RpoS binding. By

combining ChIP-seq with RNA-seq, we have identified 123 RpoS-transcribed genes with

high confidence, considerably expanding the known RpoS regulon. Previous studies

have suggested a role for RpoS in direct repression of some target genes (46, 56, 68).

While we observed negative regulation of 439 genes by RpoS, there were fewer of

these repressed genes associated with ChIP-seq peaks than expected by chance,

suggesting that direct negative regulation by RpoS is rare.

Only two previous studies have used ChIP methods to map RpoS binding genome-

wide in *E. coli*. The first used ChIP-chip to identify 868 RpoS-bound regions (46), many

more identified in our study but with considerably lower resolution (median peak length

of 324 bp for RpoS ChIP-chip). The second used ChIP-seq, but identified relatively few

RpoS-bound regions (52). Of the 63 RpoS-bound regions identified in that study, 41 are

shared with those from our study.

The high resolution of ChIP-seq allowed us to identify specific promoter sequences

recognized by RpoS. By combining ChIP-seq data with a TSS map, we identified many

high-confidence RpoS promoters. These promoters are strongly enriched for the

presence of a -10 hexamer, with sequence preferences consistent with several of the

previously described features of RpoS promoters (69). Specifically, we observed a

preference for a C at position -8 (within the -10 hexamer), a C at position -13

(immediately upstream of the -10 hexamer), and a TAA a positions -6 to -4 (immediately

downstream of the -10 hexamer). Previous studies have suggested that RpoS

promoters often contain a -35 hexamer (70), although the spacing relative to the -10

hexamer is considerably more variable than for σ^{70} promoters. However, we did not

detect enrichment of a -35 hexamer-like sequence among RpoS promoters, suggesting

that the requirement for this element is weak.

Sensitivity to RpoS affects timing of expression, and groups functionally related

genes

As is true for many transcription factors, RpoS levels vary continuously across a

wide range of conditions. Our data show that genes differ in the sensitivity of their

response to RpoS levels. Moreover, whether a gene is (in)sensitive to RpoS levels is

associated with its function, suggesting a physiological rationale for sensitivity. For

example, insensitive genes include many of those involved in the glutamate-dependent

acid resistance 2 system (AR2). These genes are of particular interest because AR2

allows *E. coli* to survive a pH of 2, an important trait for its ability to pass through the

stomach and colonize the gastrointestinal tract (71, 72). To some extent, the shared

sensitivity of functionally related genes can be explained by operon structure, i.e. co-

transcription of multiple functionally related genes from a single promoter. However, the

phenomenon of shared sensitivity for functionally related genes extends beyond

operons. For example, the insensitive genes involved in AR2 are transcribed from at

least five different promoters (63).

The fact that functionally related genes often have similar patterns of sensitivity to

RpoS suggests that sensitivity can serve as a mechanism to control the timing of gene

expression, and hence to coordinate specific cellular processes as part of a response to

environmental stresses. Consistent with this idea, we have shown that the sensitivity of

genes to RpoS levels correlates with the timing of their expression. RpoS sensitivity

may drive similar patterns of expression in response to other stresses. Different

environmental stresses are known to upregulate RpoS to varying levels (8), suggesting

that some insensitive genes may only be expressed under certain stresses. In addition

to the effects on gene expression, sensitivity to RpoS may also impact the effects of

mutations in *rpoS,* which have been seen to evolve in the lab (73, 74). We expect

mutations attenuating RpoS to have the strongest effect on insensitive genes, and the

weakest effect on sensitive genes.

Possible mechanisms of RpoS sensitivity

While the connection between RpoS sensitivity and the timing of gene expression is

clear, the molecular basis of sensitivity is less so. Our data indicate that the genomic

location of these operons does not determine the expression pattern. Moreover, several

lines of evidence suggest that direct interactions between RpoS and the core promoter

are also not responsible for determining sensitivity. First, analysis of the ChIP-seq data

for the sensitive and insensitive genes finds no motif that distinguishes between them.

Second, core promoters from both sensitive and insensitive genes do not replicate the

pattern of expression of the full-length promoters. Third, the core promoters of a

sensitive operon (*astC*) and two insensitive operons (*gadA* and *gadB*) have

indistinguishable patterns of sensitivity, suggesting that what was excluded from those

constructs (i.e., binding sites of regulatory proteins) determines the shape of the

relationship.

Given our finding that core promoter sequences cannot explain the difference in

(in)sensitivity between promoters, we suggest that sensitivity is largely due to the action

of specific regulatory proteins bound upstream. If this hypothesis is correct, it could also

explain the physiological coherence of these groups. For example, many insensitive

genes are involved in the AR2 phenotype and are also regulated by GadX, GadW, and

GadE (75–77). If one or more of these three regulators was directly responsible for the

insensitive pattern of expression, then this could help to explain the physiological

coherence of the insensitive group. The sensitive genes, being a larger group, have no

obvious single regulator, although relatively little is known about regulators that function

in stationary phase.

RpoS responds to a wide variety of environmental cues, and regulates genes

responsible for many different kinds of responses. This work has demonstrated that one

facet of that response, the level of RpoS produced, has varying effects across the

entirety of the regulon. The level of RpoS produced in a stress response, together with

the action of other transcription factors, may help to tune the RpoS-dependent stress

response in ways appropriate for individual stresses.

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900 **Table 1** Biological processes enriched in the RpoS regulon.

GO ID	GO Term	Genes in genome	Observed in regulon	Expected in regulon	p-value
GO:0009063	cellular amino acid catabolic process	69	38	18.5	3.5e-07
GO:0006099	tricarboxylic acid cycle	20	15	5.35	8.3e-06
GO:0006096	glycolytic process	18	14	4.81	8.3e-06
GO:0006094	gluconeogenesis	13	9	3.48	0.0016
GO:0009441	glycolate metabolic process	6	5	1.6	0.0063
GO:0006090	pyruvate metabolic process	21	17	5.61	0.0180
GO:0006970	response to osmotic stress	28	13	7.49	0.0190
GO:0006542	glutamine biosynthetic process	10	6	2.67	0.0270
GO:0009060	aerobic respiration	65	33	17.38	0.0285
GO:0006006	glucose metabolic process	36	20	9.62	0.0317
GO:0016052	carbohydrate catabolic process	276	94	73.79	0.0326
GO:0009101	glycoprotein biosynthetic process	13	7	3.48	0.0342
GO:0019395	fatty acid oxidation	30	13	8.02	0.0354
GO:0000162	tryptophan biosynthetic process	11	6	2.94	0.0463
GO:0009246	enterobacterial common antigen biosynthesis	11	6	2.94	0.0463
GO:0042398	cellular modified amino acid biosynthesis	17	7	4.55	0.0472
GO:0009239	enterobactin biosynthetic process	6	4	1.6	0.0472
GO:0000270	peptidoglycan metabolic process	51	14	13.64	0.0475

903 **Table 2** Biological processes enriched in the sensitive genes.

GO ID	GO Term	Genes in genome	Observed in sensitive set	Expected in sensitive set	p-value
GO:0006970	response to osmotic stress	13	4	0.79	0.0055
GO:0009063	cellular amino acid catabolic process	38	6	2.31	0.0210
GO:0019395	fatty acid oxidation	13	3	0.79	0.0385

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910 **Table 3** Biological processes enriched in the insensitive genes.

GO ID	GO Term	Genes in genome	Observed in insensitive set	Expected in insensitive set	p-value
GO:0006412	translation	21	3	0.6	0.018
GO:0006164	purine nucleotide biosynthetic process	10	2	0.28	0.030
GO:0006805	xenobiotic metabolic process	13	2	0.37	0.049

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923   **Table 4** Contingency table of the (in)sensitivity of a gene and if is promoter is or is
924   not bound by RpoS.
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	Sensitive	Linear	Insensitive
Bound by RpoS	31	87	5
Not bound by RpoS	71	823	27

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Figure 1. Analysis of RpoS ChIP-seq data. (A) Sequence logo (78) of the enriched
sequence motif associated with RpoS-bound regions (MEME E-value 2.0e-40). The
sequence logo was generated using WebLogo (79) (B) Histogram showing the
frequency distribution of distances between identified motifs and ChIP-seq peak
centers. The black, vertical, dashed line indicates the peak center position. (C)
Sequence logo (78) of the enriched sequence motif associated with TSSs located
within 20 bp of RpoS ChIP-seq peak centers (MEME E-value 2.6e-56). Note that
only positions 21-29 of the motif are shown. The sequence logo was generated
using WebLogo (79).
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945 **Figure 2. Arabinose control of RpoS protein levels in strain DMS2564.** (a)

946 Increasing arabinose results in more RpoS expression in strain DMS2564, as measured

947 by western blotting. n = 4 to 5; error bars represent SEM. (b) Flow cytometric

948 measurement of an *otsB-gfp* fusion under 0%, 10^{-4}%, and 10^{-1}% arabinose. (Dotted line,

949 dashed line, and solid line, respectively.)

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957 **Figure 3**: Area proportional Venn diagram showing the number of genes
958 differentially expressed between each of the three conditions.

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974   **Figure 4:** Examples of three classes of patterns of RpoS response. *osmY*
975   expression is linear with RpoS level, *astA* is sensitive to RpoS level, and *gadC* is
976   insensitive to RpoS level. Dots represent normalized expression levels from individual
977   RNA-seq samples. The line is drawn between the mean at 0% and the mean at 100%.
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Figure 5. Testing of RNA-seq expression patterns using qPCR. Expression patterns were measured for (a,b) *gadC* and *gadE*, which were insensitive in the RNA-seq data, and (c-e) *astA*, *prpR,* and *prpD*, which were sensitive in the RNA-seq data. In all cases, the level of transcription at 26% RpoS was on the side of the line predicted by the RNA-seq data, though not always significantly so. *gadC* was significantly insensitive (p < 10^{-4}), *gadE* was not significantly different from linear (p = 0.10), *astA* was (barely) not significantly sensitive (p = 0.06), *prpR* was significant sensitive (p < 10^{-4}), and *prpD* was not significant (p = 0.56). n = 6, error bars = SEM.

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Figure 6. **Expression patterns of whole promoters and core promoters only.**
Expression patterns were measured for the (a - d) whole upstream regulatory region
and the (e - h) core promoter only of *astC, gadA, gadB,* and *hdeA.* As expected by
RNA-seq data, the whole *astC* fusion was sensitive (p = 0.04, t-test), and the *gadA,*
gadB, and *hdeA* fusions were all insensitive (p = 2 x 10^{-6}, p = 10^{-5}, p = 0.04, all by one
sample t-test). The p-values are adjusted for multiple comparisons using Holm's
sequential adjustment method (50). The core promoters (e - h) have much lower levels
of maximal transcription than the full length promoters. All four have much altered
patterns of sensitivity, although only *gadA* and *gadB* was significantly different from the
whole promoter (p = 0.08 (*astC*), p = 0.01 (*gadA*), p = 10^{-6} (*gadB*), p = 0.21 (*hdeA*), all
by two-sample t-test with Holm's correction.) n = 6 to 8, error bars = SEM.
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Figure 7. Expression profiles of genes when cells enter stationary phase. (A)
Relative expression levels of RpoS-regulated genes were taken from a published
study (47) for four time-points following entry into stationary phase. All expression
values were normalized to those from the final time-point for each gene. The graph
shows the range of relative expression values for insensitive (white boxes), linear
(light gray boxes), and sensitive (dark gray boxes) genes for each time-point. Boxes
represent the 25th-75th percentile range, and horizontal lines indicate the median
value. **(B)** As above, but for a Δ*rpoS* strain of *E. coli*.