Title
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Permalink
https://escholarship.org/uc/item/5zc4s499

Journal
mBio, 5(2)

ISSN
2150-7511

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Publication Date
2014-04-01

DOI
10.1128/mBio.01046-13

Peer reviewed
Discovering Functions of Unannotated Genes from a Transcriptome Survey of Wild Fungal Isolates

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ABSTRACT Most fungal genomes are poorly annotated, and many fungal traits of industrial and biomedical relevance are not well suited to classical genetic screens. Assigning genes to phenotypes on a genomic scale thus remains an urgent need in the field. We developed an approach to infer gene function from expression profiles of wild fungal isolates, and we applied our strategy to the filamentous fungus Neurospora crassa. Using transcriptome measurements in 70 strains from two well-defined clades of this microbe, we first identified 2,247 cases in which the expression of an unannotated gene rose and fell across N. crassa strains in parallel with the expression of well-characterized genes. We then used image analysis of hyphal morphologies, quantitative growth assays, and expression profiling to test the functions of four genes predicted from our population analyses. The results revealed two factors that influenced regulation of metabolism of nonpreferred carbon and nitrogen sources, a gene that governed hyphal architecture, and a gene that mediated amino acid starvation resistance. These findings validate the power of our population-transcriptomic approach for inference of novel gene function, and we suggest that this strategy will be of broad utility for genome-scale annotation in many fungal systems.

IMPORTANCE Some fungal species cause deadly infections in humans or crop plants, and other fungi are workhorses of industrial chemistry, including the production of biofuels. Advances in medical and industrial mycology require an understanding of the genes that control fungal traits. We developed a method to infer functions of uncharacterized genes by observing correlated expression of their mRNAs with those of known genes across wild fungal isolates. We applied this strategy to a filamentous fungus and predicted functions for thousands of unknown genes. In four cases, we experimentally validated the predictions from our method, discovering novel genes involved in the metabolism of nutrient sources relevant for biofuel production, as well as colony morphology and starvation resistance. Our strategy is straightforward, inexpensive, and applicable for predicting gene function in many fungal species.
species (13–17). Here, we report on the use of expression as a genome-scale screening tool in fewer than 150 wild individuals, far fewer than the >8,000 mutants of predicted nonessential genes in \textit{N. crassa} that would be screened for phenotype in a library of deletion mutants.

We set out to survey transcriptional variation in wild \textit{N. crassa}, both intrapopulation differences between strains isolated in Louisiana (18) and differences between the Louisiana population and one from the southern Caribbean (11). From these analyses, we formulated hypotheses about genes of unknown function that might mediate growth, metabolic, and regulatory traits in \textit{N. crassa}, and we focused on several as case studies for validation of our methods. Our experiments targeted the roles of these genes with respect to three fundamental aspects of filamentous fungal biology: colony development, regulation of iron acquisition, and global regulation of nitrogen and carbon metabolism. The results led to the discovery of novel phenotypes for four previously uncharacterized genes.

RESULTS
Inferring multigene regulons from expression variation across an \textit{N. crassa} population. To survey variation in \textit{N. crassa} gene expression, we made use of our transcriptional profiles recently generated from wild isolates of this fungus collected in Louisiana (19). Of the 9,733 predicted \textit{N. crassa} genes, 8,876 had mapped reads in at least 24 of the wild isolates, and we considered the latter set of genes to represent the core active transcriptional program of \textit{N. crassa} under the standard growth conditions of our cultured colonies.

To harness regulatory variation across strains to infer gene function, we first used our expression profiles to define coexpressed gene clusters, applying a resampling strategy to assess the significance of cluster sizes (see Materials and Methods). At a cluster size of nine genes, we identified 188 clusters whose gene expression was correlated across wild strains with a correlation coefficient of 0.4 or greater, whereas no such clusters were detected in permuted data sets (see Table S1 in the supplemental material). The majority of clusters (92\%) contained at least three genes that have been annotated in functional categories according to the Functional Catalogue (FunCat) (20). In 72\% of these clusters, we detected functional category enrichment at a \(P\) value of \(\leq0.05\) (Benjamini-Hochberg-corrected hypergeometric test; see Data Set S1), thereby highlighting the potential of our clustering data set as a resource for the inference of function of uncharacterized genes.

Molecular validation of a novel hyphal morphology gene. To investigate, at the molecular level, the inferred function of uncharacterized \textit{N. crassa} genes that underlie growth traits, we first focused on a coexpressed cluster of genes (cluster 48 in Data Set S1) in the supplemental material; see also Table S2) encoding proteins that (i) are localized to septa and cell walls (RHO4, COT2, and ACW11), (ii) act as signaling molecules and transcription factors (RHO1, ADA6, BEK2, and CHM1), and (iii) have suspected roles in hyphal branching and septa formation (RHO4, RHO1, and CHM1). Also among the genes in this cluster was \textit{NCU04826}, which encodes a hypothetical protein lacking any annotated function or protein domains. Sequence searches revealed homologs of this gene in species of filamentous fungi within the Sordariomycetes and Leotiomycetes, with amino acid identity to the \textit{N. crassa} sequence ranging from 28\% (\textit{Glarea lozoyensis}) to 97\% (\textit{Neurospora tetrasperma}). Given the coregulation of \textit{NCU04826} with known cell wall and morphology genes, we hypothesized that this gene would have a role in hyphal morphology. To test this notion, we obtained an \textit{N. crassa} strain of the FGSC (Fungal Genetics Stock Center) 2489 background harboring a deletion in \textit{NCU04826} (21), crossed it to an isogenic wild-type strain, and examined progeny for colony morphology. Strikingly, inheritance of the deletion cassette conferred a distinct hyperbranching phenotype, as predicted from our coexpression analysis (Fig. 1). In assays of the branching phenotypes of strains bearing deletions of the other genes from the \textit{NCU04826} coexpression cluster (see Table S2), the colony and branching morphology of a strain carrying a deletion of \textit{NCU02978} most closely resembled that of the \textit{NCU04826} strain (see Fig. S1). \textit{NCU02978} is predicted to encode a protein similar to Slalp in \textit{S. cerevisiae}, an adaptor protein for endocytosis involved in assembly of the actin cytoskeleton. Consistent with a potential role for \textit{NCU04826} in the cytoskeleton, we reevaluated the predicted protein product of \textit{NCU04826} and detected homology to PFAM “intermediate filament” and “tropomyosin-like” protein families (PF00038 [\(E = 0.055\)] and PF12718 [\(E = 0.0005\)], respectively), both of which comprise cytoskeleton components. We conclude that \textit{NCU04826} represents a previously uncharacterized determinant of colony morphology and hyphal branching, possibly by affecting activity of the
the expression levels of the indicated gene and asi-1

We next set out to use our population genomics approach to investigate the regulatory functions of transcription factors. The majority of annotated transcription factors in N. crassa have no known function or target gene set (6), and we reasoned that our clusters of genes defined by coregulation across wild isolates could be harnessed to infer the pathways in which these transcription factors act. As a positive control for this approach, we first examined the 35 factors act. As a positive control for this approach, we first exam-
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Table 1: Genes coregulated with asi-1 (NCU05257) across wild strains are affected by asi-1 deletion.

| ID          | Annotation                              | DE P value | Coexpression correlation |
|-------------|-----------------------------------------|------------|--------------------------|
| NCU00522    | Cystathionine beta-lyase                 | 2.3104e-05 | 0.78                     |
| NCU00535    | Alanyl-tRNA synthetase                   | 0.0054     | 0.84                     |
| NCU02019    | FAD-dependent oxidoreductase            | 1.5469e-08 | 0.82                     |
| NCU02543    | Aspartate aminotransferase              | 0.0027     | 0.92                     |
| NCU05045    | MFS monocarboxylate transporter         | 1.8375e-09 | 0.90                     |
| NCU05256    | Hypothetical                            | 0.0013     | 0.81                     |
| NCU07126    | Acetyltransferase (GNAT) family         | 0.0503     | 0.83                     |
| NCU111365   | Aminotransferase                        | 0.0533     | 0.82                     |

* Each row shows analysis of one gene whose expression changed significantly upon deletion of asi-1 and was tightly correlated with expression of asi-1 across Louisiana strains of N. crassa. DE P value, Benjamini-Hochberg-corrected significance of differential expression between an engineered asi-1 deletion mutant and an isogenic wild-type strain. The complete set of genes responsive to asi-1 deletion is given in Data Set S2 in the supplemental material. Coexpression correlation, Spearman’s rank correlation coefficient between the expression levels of the indicated gene and asi-1 across Louisiana isolates; the complete set of genes coregulated with asi-1 is given in Table S2. ID, identifier.

cytoskeleton, underscoring the power of our coexpression clustering methods to infer functions of unknown genes. We propose the name hbc-1, hyperbranching and cytoskeleton 1, for this gene.

Molecular validation of a novel amino acid metabolism gene.

We next set out to use our population genomics approach to investigate the regulatory functions of transcription factors. The majority of annotated transcription factors in N. crassa have no known function or target gene set (6), and we reasoned that our clusters of genes defined by coregulation across wild isolates could be harnessed to infer the pathways in which these transcription factors act. As a positive control for this approach, we first examined the 35 N. crassa transcription factors of known function (22) that fall into coexpression clusters enriched for one or more FunCat terms: of these, we identified 15 whose FunCat annotation overlapped with that of genes in their respective clusters (see Data Set S1 in the supplemental material; permutation P < 0.0001). For example, the known carbon catabolite repressor gene cre-1 was annotated in carbohydrate metabolism, as were the genes with which its expression was correlated (see cluster 176 in Data Set S1).

To evaluate this strategy in the context of an unannotated transcription factor, we focused on the gene NCU05257, which encodes a predicted zinc finger and homeobox DNA-binding protein and which fell into an expression cluster containing 58 other genes in our analysis of expression among wild N. crassa strains (cluster 15 in Data Set S1 in the supplemental material). This group was enriched for genes annotated in amino acid metabolism; see Data Set S1 and Table S2), and NCU05257 was previously reported to be a putative target of the N. crassa amino acid biosynthesis regulator CPC1 (23). To test the regulatory impact of NCU05257 directly, we used transcriptome sequencing (RNA-seq) to generate the transcriptional profile of a strain bearing a deletion in NCU05257, and we compared this profile to that of its isogenic wild-type control, identifying 43 genes affected by the NCU05257 mutation at a lenient statistical cutoff (Benjamini-Hochberg-corrected P value < 0.1; see Data Set S2). This expression signature was enriched for genes annotated in amino acid metabolism (see Table S3), a conclusion that was independent of the threshold used to call differential expression (data not shown). Likewise, the NCU05257 deletion signature was enriched for the members of the cluster of genes with which it was coregulated across wild strains (eight genes present in both sets; Table 1; Fisher’s exact test, P = 9.76 × 10^−9), and in a quantitative analysis, this cluster was enriched for dramatic expression change upon NCU05257 deletion (resampling P = 0.007). Interestingly, in the NCU05257 mutant, we also noted upregulation of several iron scavenging genes (the aerobactin siderophore biosynthesis protein IUCB, the fatty acid coenzyme A [CoA] ligase NCU06063, which is involved in siderophore biosynthesis [24], and the siderophore iron transporter NCU06132; see Data Set S2).

To investigate the phenotypic role of NCU05257, we used the histidine biosynthesis inhibitor 3-amino-1,2,4-triazole (3-AT), which induces growth defects in N. crassa amino acid biosynthesis regulator mutants (25, 26). We crossed the NCU05257 deletion strain to an isogenic wild-type strain and measured the growth rate of progeny in the presence of 3-AT. In this cross, strains inheriting ANCU05257 showed significantly compromised growth in 3-AT relative to that of wild-type progeny (Fig. 2), paralleling the behavior of a cpc-1 deletion strain (see Fig. S2 in the supplemental material) and confirming the importance of NCU05257 in the cellular response to amino acid starvation. We conclude that NCU05257 is a novel component of the amino acid metabolic control network in N. crassa with a link to iron scavenging, and we

![FIG 2 Deletion of asi-1 increases sensitivity to the histidine analogue 3-AT.](image-url)

Each column reports the distribution of growth rate among progeny from a cross between the wild-type strain, FGSC 4200, and an isogenic strain bearing a deletion in asi-1 (NCU05257; asi-1::HYG), with symbols as in Fig. 1 except that points >1.5-fold outside the interquartile range are shown as circles. The x axis reports genotype in cross progeny at the asi-1 locus: HYG-, segregants inheriting the wild-type asi-1 gene; HYG+, segregants inheriting the asi-1 deletion. The y axis reports the ratio between growth rate on Vogel’s minimal medium (VMM) containing 6 mM 3-amino-1,2,4-triazole (3-AT) and growth rate on VMM lacking 3-AT. *, the difference between the growth rate on 3-AT in the wild-type and Δasi-1 mutant strains is significant at a Wilcoxon P value of 0.008.

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propose the name asi-1, for amino acid and siderophore regulation, for this gene.

**Regulatory variation between N. crassa populations and the nitrogen metabolite repression pathway.** Having explored the genetic differences among Louisiana strains of *N. crassa*, we next investigated the variation between two populations by incorporating transcriptional profiles previously reported for a southern Caribbean *N. crassa* population (11). To polarize changes with the use of an outgroup, we also added profiles of an *N. crassa* population from Panama (11). Genome-scale comparisons between Caribbean and Louisiana strains detected significant differential expression of 1,539 genes (Benjamini-Hochberg correction, \(P < 0.05\); see Data Set S3 in the supplemental material).

To use expression divergence between populations as a test bed for inference of gene function, we began with the genes most strongly differentially expressed between Caribbean and Louisiana strains. At the top of the list was allantoicase (*alc-1*) (Fig. 3), a purine degradation gene canonically studied as part of the nitrogen metabolite repression (NMR) program, for its repression in the presence of preferred nitrogen sources (27, 28). Ammonium, at the concentrations used in our transcriptional profiling experiments, is sufficient to trigger NMR in *N. crassa* (27). This fact made it all the more striking that *alc-1*, a gene that ought to be under NMR, was highly expressed among wild Louisiana isolates of *N. crassa* (Fig. 3). This finding of unusual expression of an NMR target gene led us to further investigate interpopulation differences in metabolic repression programs. To analyze as many NMR targets as possible, we began with the set of NMR targets previously characterized in *Neurospora*, and we added to it *Neurospora* genes detected by a best-reciprocal-BLAST search of the *N. crassa* genome using known NMR targets in other fungi (see Data Set S4 in the supplemental material). Of these, we examined the 15 genes whose expression differed between the Louisiana and Caribbean populations and found that 9 were upregulated in Louisiana individuals relative to those of the Caribbean and Panamanian populations (Fig. 3, yellow-themed panels). Finding such a large number of genes differentially upregulated in the same population constitutes a directional coherence significantly greater than that expected by chance from the genome as a whole (one-tailed Fisher’s exact test, \(P = 0.005\)). Thus, Louisiana isolates of *N. crassa* exhibited a program of high expression of NMR targets in rich medium that was unique among our study populations.

**PP4 controls expression of nitrogen metabolite repression and carbon catabolite repression targets.** The NMR target genes differentially expressed between the Louisiana and Caribbean *N. crassa* clades also exhibited expression-level variation within the Louisiana population (Fig. 3). To complement our interpopulation comparison, we took a candidate-gene approach in hopes of identifying master regulators of metabolic expression programs, as follows. We identified *N. crassa* orthologs of the set of known NMR regulators in *S. cerevisiae* (29) (see Data Set S4 in the supplemental material) and examined their coding regions for sequence variation among Louisiana *N. crassa* strains. Among these genes, the only case of striking sequence difference between strains was in the putative protein phosphatase PP4 (NCU08301), a regulator of the circadian oscillator FRQ in *N. crassa* (30) whose homolog in *S. cerevisiae* activates the NMR transcription factor Gln3 (31); Louisiana individuals bore two derived amino acid variants in PP4 at high frequency relative to Caribbean strains (Fig. 4). To investigate the regulatory impact of pp4, we therefore transcriptionally profiled a strain harboring a deletion in this gene alongside an isogenic wild-type strain. Comparison between the two expression profiles identified 195 differentially expressed genes (Benjamini-Hochberg-corrected \(P\) value < 0.1; see Data Set S2), of which nearly all (187 genes) showed elevated expression upon pp4 deletion. The response to the deletion of pp4 was significantly more dramatic among NMR targets than expected under a null hypothesis based on genomic resampling (\(P < 0.0001\)), consistent with a role for PP4 in the repression of NMR target genes. In addition, we noted in the pp4 deletion signature a number of genes involved in the metabolism of alternative carbon sources, including the mannose metabolism genes NCU07067, NCU02322, NCU07269, and NCU07318 and the polysaccharide metabolism genes NCU08755, NCU04959, NCU09281, and NCU04431 (see Data Set S2). We thus suspected that in addition to the NMR genes we had originally analyzed (Fig. 3, yellow-themed panels), PP4 also functioned as a repressor of genes subject to carbon catabolite repression (CCR) in glucose medium, which we call CCR targets. As an unbiased test for the role of pp4 in carbon catabolite repression, we considered the 75 genes whose expression increases upon deletion of the *N. crassa* carbon catabolite repressor CreA/CRE1 (32). Eight of these CCR (CRE1) targets were differentially expressed in the pp4 deletion, an overlap beyond that expected by chance (hypergeometric \(P = 0.0001\)). Furthermore, seven of the eight targets shared by CRE1 and PP4 showed elevated expression in the pp4 deletion strain, further supporting a model of PP4 as a repressor of CCR targets. We thus conclude that PP4 functions in the joint regulatory control of nitrogen and carbon metabolism genes in *N. crassa*.

We next hypothesized that if the naturally occurring sequence variants among Louisiana strains in pp4 affected the function of this gene, inheritance at pp4 would be correlated with expression of nitrogen and carbon metabolism pathways across the Louisiana population. To test this notion, we examined both our directly inferred NMR targets and a broader curated set of nitrogen metabolism genes inferred from the nitrogen starvation transcriptional response of *Magnaporthe grisea* (33) (see Data Set S4 in the supplemental material; see also Materials and Methods) for association of their expression with inheritance at pp4 across Louisiana strains. We likewise tested for association between pp4 genotype and expression of CCR targets, again using the CRE1 target gene set as a reflection of the latter (32). Unexpectedly, the results were not consonant with natural variation in the pp4 sequence having a major impact on gene expression: expression of only one gene, NCU00789, showed modest association with the pp4 genotype across Louisiana strains (Benjamini-Hochberg-corrected \(P\) value = 0.12; Fig. 4). Thus, despite the dramatic impact of an engineered deletion in pp4, we concluded that natural polymorphisms in pp4 did not underscore most of the variation in metabolite repression genes within or between wild *N. crassa* populations. However, given that our exploration of pp4 led us to the uncharacterized gene NCU00789, we considered the latter in its own right as another candidate gene for inference of novel function in *N. crassa*.

**Nitrogen and carbon metabolism gene expression associates with *Nc_nmr6*.** GenBank lists the protein encoded by NCU00789 (accession no. AA935520.1) as the *N. crassa* ortholog of the *Hansenula polymorpha* gene NMR6, an unpublished, 12-transmembrane ammonium sensor, and we refer to the *Neurospora* version of this gene as *Nc_nmr6*. The *Nc_nmr6* locus har-
FIG 3  Targets of carbon catabolite and nitrogen metabolite repression are upregulated in the *N. crassa* Louisiana population. Each panel shows expression of one target gene of the nitrogen metabolite repression program (yellow shading; see Data Set S4 in the supplemental material) or of the carbon catabolite repression regulator CRE1 (orange shading [32]), measured in wild strains of *N. crassa* grown on standard rich medium. In a given panel, each column reports the distribution of expression of the indicated gene across the strains of one *N. crassa* population: LA, Louisiana; CARIB, Caribbean; PAN, Panama. Symbols are as in Fig. 2. NMR, nitrogen metabolite repression targets; CCR, carbon catabolite repression targets. Systematic gene identifiers, from top left to bottom right, are NCU01066, NCU01816, NCU02333, NCU03076, NCU03257, NCU05765, NCU08356, NCU10007, NCU00721, NCU00130, NCU01140, NCU01449, NCU02904, NCU03151, NCU04039, NCU04197, NCU04460, NCU04963, NCU08746, NCU10021, and NCU07363.
FIG 4 A derived, high-frequency nonsynonymous variant in pp4 is associated with the expression level of a novel putative ammonium sensor. (A) The top panel shows an amino acid alignment of a region of the putative protein phosphatase PP4, in which the major allele among Louisiana strains of _N. crassa_ (N.crassa.LA) encodes a glycine at residue 177, whereas the major allele among Caribbean isolates (N.crassa_CARIB) and the sequence in other Sordariomycetes encodes an aspartate. The inset shows allele frequencies of the aspartate and glycine alleles in the Louisiana and Caribbean _N. crassa_ populations. (B) Each column reports the expression of _Nc_nmr6_ (NCU00689) across Louisiana _N. crassa_ strains harboring the indicated allele at the PP4 variant in panel A. The _Nc_nmr6_ expression level associated at modest significance with the pp4 genotype across Louisiana strains (nominal _P_ = 0.0001; Benjamini-Hochberg-corrected _P_ = 0.12). Symbols are as in Fig. 2.

bored extensive variation at coding and silent sites across _N. crassa_ isolates (Fig. 5A), which defined two haplotypes. One _Nc_nmr6_ haplotype was borne by most strains of the Caribbean population and a few strains from Louisiana, and a second _Nc_nmr6_ haplotype was apparent in the majority of Louisiana isolates. We thus hypothesized that _Nc_nmr6_ was a previously uncharacterized component of the nitrogen and carbon metabolism gene network in _N. crassa_ and that polymorphisms at _Nc_nmr6_ served to tune the expression of its target genes even in medium containing preferred carbon and nitrogen sources.

To validate our inference of _Nc_nmr6_ as a determinant of metabolic gene expression, we first investigated the relationship between inheritance at _Nc_nmr6_ and NMR target gene expression across Louisiana individuals. The results revealed a robust association signal, with the major haplotype of _Nc_nmr6_ associated with increased NMR gene expression among Louisiana strains (one-tailed paired Wilcoxon _P_ value = 0.006). We next evaluated the effect of the _Nc_nmr6_ genotype on the broader set of nitrogen starvation genes inferred from profiles of _M. grisea_ (33). Again we observed a striking relationship (Fig. 5B and C); among the 62 genes for which the major allele of _Nc_nmr6_ among Louisiana individuals was associated with lower expression, largely components of the translation machinery, 50 were downregulated during nitrogen starvation (Fig. 5B) (binomial _P_ = 9.204e−05). Likewise, of the 37 genes activated by the major allele of _Nc_nmr6_, 31 were upregulated during nitrogen starvation (Fig. 5C; binomial _P_ = 8.197e−07). The latter included genes involved in the metabolism of alternative carbon sources (e.g., the xylanase NCU08189, the gluconate reductase NCU09519, and the rhamnose synthase NCU10683), as well as nitrogen metabolism genes (e.g., the uricase NCU07853 and the pyrimidine catabolism gene hydantoinase NCU00689). We thus suspected that _Nc_NMR6_ had a role in the expression of CCR targets as well as NMR targets. To evaluate the impact of _Nc_nmr6_ on carbon catabolite repression, we examined the association between the _Nc_nmr6_ genotype and expression of CRE1 target genes (32). Of the latter, 18 were differentially expressed between our Louisiana and Caribbean _N. crassa_ populations, with 14 exhibiting a derived program of increased expression in Louisiana individuals relative to outgroups (one-tailed Fisher’s exact _P_ value = 0.005; Fig. 3, orange-shaded panels). As predicted, among Louisiana individuals, the major _Nc_nmr6_ haplotype was associated with high expression of CRE1 targets (one-tailed paired Wilcoxon _P_ value = 0.005). We conclude that the major allele of _Nc_nmr6_ among Louisiana strains is associated with an expression program in rich medium that mirrors the response to nitrogen starvation and the loss of carbon catabolite repression, strongly suggesting that _Nc_NMR6_, like PP4, functions in the regulation of genes that metabolize nonpreferred nutrient sources.

_Nc_nmr6_ is a novel determinant of metabolism gene expression. As a direct test of the regulatory impact of _Nc_nmr6_, we constructed a strain bearing a deletion of this gene (see Materials and Methods), using as the genetic background FGSC 2489, which bears the major allele from the Louisiana population. We tran-
scriptionally profiled this ΔNc_nmr6 strain and its wild-type isogenic progenitor after growth in rich medium, finding 80 genes that were differentially expressed between the strains (Benjamini-Hochberg-corrected \( P \) value \( < 0.1 \); see Data Set S2 in the supplemental material), enriched for annotation in a number of functions, including carbohydrate metabolism (see Table S3). Analysis of the Nc_nmr6 deletion signature bore out our prediction of a role for Nc_nmr6 in nitrogen metabolism regulation: the suite of NMR targets that we had originally ascertained based on upregulation in Louisiana strains relative to levels in other populations (Fig. 3, yellow-shaded panels) were expressed at low levels in the Nc_nmr6 mutant (resampling \( P \) value \( = 0.01 \); Fig. 6). Likewise, again conforming to our expectation, the engineered ΔNc_nmr6 mutant expressed CRE1 targets at lower levels than the wild-type control (Fig. 6). We conclude that Nc_NMR6 functions in the regulation of NMR and CCR gene targets and, in contrast to the behavior of the putative repressor PP4, is required for the high expression of these genes by the Louisiana strain FGSC 2489 in rich medium. Given the impact of both PP4 and Nc_NMR6 on multiple nutrient response pathways, our data thus implicate these proteins as two novel control points for expression of metabolite repression programs in N. crassa.

**DISCUSSION**

Transcripts that are up- and downregulated together across conditions may often encode proteins of similar functions. This idea has motivated studies of coregulation between unknown and well-characterized genes (34, 35), to shed light on novel, species- or condition-specific gene functions. A key roadblock in the field is that such highly specialized genes may be inactive under the experimental conditions used for standard analyses of gene coregulation, even in microbes (35–37). In contrast, when natural ge-
the distribution, across the genes of the indicated set, of the change in expression between a strain harboring an engineered deletion in *Nc_nmr6* (KO) and an isogenic wild-type (WT) strain. All genes, the complete set of genes expressed in the knockout experiment; DE between populations, the set of genes significantly differentially expressed between Louisiana and Caribbean *N. crassa* strains (see Data Set S3 in the supplemental material); NMR genes upregulated in LA, the set of inferred targets of the nitrogen metabolite repression program upregulated in Louisiana strains relative to other *N. crassa* populations (see yellow panels in Fig. 3); CCR genes upregulated in LA, the set of genes repressed by the carbon catabolite repression regulator CRE1 and upregulated in Louisiana strains relative to other *N. crassa* populations (see orange panels in Fig. 3). For the last three columns, the text at the top reports the resampling-based estimate of significance of the difference between the indicated regulon and the set of all genes with respect to the expression effects of *Nc_nmr6* deletion. Symbols are as in Fig. 2, except that for ease of visualization, data points falling outside the box plot whiskers are not shown.

Functional inference methods are in urgent demand for transcription factors, whose biological roles often remain unknown even in exhaustively studied model organisms like budding yeast (37). Our case study for characterization of a novel transcription factor was a coexpression cluster enriched for amino acid metabolism genes that contained NCU05257, a gene encoding a protein with predicted zinc finger and homeobox DNA-binding domains. Of the genes whose expression dropped when NCU05257 was deleted, many overlapped with the original coexpression gene cluster and/or were implicated in amino acid metabolism and iron scavenging via siderophores. Challenging the ΔNCU05257 strain with an amino acid biosynthesis inhibitor demonstrated the importance of this gene in the response to amino acid starvation. We named the NCU05257 gene *asi-1*, for amino acid and siderophore regulation; our results provide a first window onto the potential function of this gene in coordinating the joint regulation of fungal siderophore biosynthesis with amino acid supply (42, 43).

Besides analyzing expression variation across Louisiana strains of *N. crassa*, we also predicted gene function using the differences between two populations of this fungus, isolates from Louisiana and the southern Caribbean. The gene with the strongest interpopulation transcription difference, that encoding allantoicase, is part of the NMR program. This observation led us to a focus on patterns of coordinated expression between nitrogen and carbon metabolism genes and the discovery of a role in these metabolic networks for two putative upstream signaling factors. Our experiments implicated the putative protein phosphatase gene *pp4* in the repression of genes involved in the metabolism of both non-
preferred nitrogen and carbon sources. In *S. cerevisiae*, where nu-
trient sensing networks have been well delineated, no protein
phosphatase is known to participate in both carbon catabolite and
nitrogen metabolite repression (44); it is thus tempting to specu-
late that the division of labor among protein phosphatases in nu-
trient sensing pathways has diverged between yeast and *N. crassa*. 
Likewise, given that PP4 has been well characterized in *N. crassa* as
a regulator of the circadian clock machinery (30), we hypothesize
that this protein may function as a novel component of the re-
cently discovered gene network for circadian control of metabo-
lism in *N. crassa* (45), plausibly linking the circadian clock with
nitrogen and carbon scavenging under nutrient-poor conditions.

Our analysis of the putative transmembrane protein *Nc-
NMR6* identified a haplotype in Louisiana strains associated with
high expression of genes involved in metabolism of nonpreferred
carbon and nitrogen sources, even during growth in glucose- and
ammonium-containing medium. The reduced expression of these
nutrient-scavenging genes that we observed in an *Nc_nmr6* dele-
tion strain is consistent with either of two possible models for the
function of the encoded protein. On the one hand, the ancestral
haplotype of *Nc_NMR6* could transduce a positive signal of the
presence of preferred carbon and nitrogen sources, analogous to
well-characterized glucose and ammonium receptors in many fungi
(46–48). Under this model, the major allele of *Nc_nmr6* in the
Louisiana population would act as a dominant negative loss of
function, failing to signal the availability of preferred nutrient
sources in rich medium and leading to derepression of NMR and
CCR targets. Upon deletion of this allele of *Nc_nmr6* in the Lou-
siana genome, metabolite repression would be reinstated by
CRE1, NMR1, and other NMR and CCR regulators in *N. crassa*
(49, 50). Alternatively, the ancestral form of *Nc_NMR6* could ac-
tively transduce a signal in the absence of preferred nutrient
sources, activating the metabolic machinery for nutrient scaveng-
ing; the major haplotype in the Louisiana population would then
act as a gain of function to elevate expression of NMR and CCR
targets even in rich medium, and deletion of this allele would
abrogate the expression effect. In either case, our results make
clear that *Nc_NMR6*, like PP4, functions as part of a complex
network for the joint control of carbon and nitrogen metabolism.
The regulatory impact of both genes dovetails with that of the
sugar sensor TPS1 in *Magnaporthe oryzae*, which signals to regu-
lators of genes for metabolism of nonpreferred nitrogen and car-
bon sources (51–53). And our de novo inference of the functions
of these previously uncharacterized genes underscores the power of
the comparative transcriptomic approach for gene discovery.

Coexpression analysis can be a powerful approach for the pre-
diction of gene function, though to date, relatively few genome-
scale studies have experimentally validated the *in silico* inference of a
gene’s role at the organismal level (54–59). By validating the
regulatory and organismal impact of four novel genes, our work
makes clear that expression differences between wild isolates can
lead to biologically meaningful inferences of gene function,
whether the loci of interest are involved in gross morphology or
the regulation of metabolism. The coexpression clusters we have
reported here contain 2,243 other unannotated genes well suited
for a similar functional discovery pipeline, in which the deletion
strain of any such gene can be assayed for traits in which its co-
regulated partners are known to play a role. We hope that by
demonstrating the approach, our work will stimulate other re-
searchers to use the collection of Louisiana and Caribbean
*N. crassa* strains that have been transcriptionally profiled to deter-
mine functions for unknown genes in the pathways that interest
them. Given the homology among genes in filamentous Ascomyc-
ota, the *N. crassa* collection will be of immediate use in inference
of gene function in other species. And the strategies we have es-
tablished for population-genomic transcriptional profiling and
analysis in *N. crassa* will be relevant for future work in any fungus
for which sufficient isolates can be collected.

**MATERIALS AND METHODS**

**Strains, sequencing, and gene expression quantification.** The complete
set of wild Louisiana isolates and engineered strains used in this study is
listed in Table S4 in the supplemental material and was obtained from
the Fungal Genetics Stock Center (FGSC) (60). Our analysis of Louisiana
strain transcriptomes used 48 of the RNA-seq libraries from reference 19,
which, for analysis on the same footing as data for other *N. crassa*
populations (see below), we remapped as follows. For each RNA-seq data set,
we mapped Illumina reads to the *Neurospora crassa* OR74A version 10
reference assembly (61) using the software program TopHat (62). We
required reads to map uniquely with a maximum of two mismatches
and minimum and maximum intron lengths of 40 and 200 bp,
respectively, for mapping, coverage, and split-segment searches. We
supplied TopHat with the Broad Institute’s February 2010 release of
the *N. crassa* version 4 annotation (http://www.broadinstitute.org
/annotation/genome/neurospora/) and also allowed the program to
search for novel splice junctions; we then integrated all putative isoforms
into a single expression measure per gene. We used a custom Perl script
and the *N. crassa* annotation to calculate the number of raw reads over-
lapping each gene model and used the third-quartile method from refer-
ence 63 to normalize read counts between lanes. For expression measure-
ments from wild strains, we excluded from further analysis any gene for
which >50% of the individuals analyzed had an expression level of zero.

**Expression clusters.** To generate expression clusters from transcription-
proﬁles of wild Louisiana strains of *N. crassa*, we ﬁrst normalized
read counts for each gene in each RNA-seq library by gene length. We next
used the Popular Statistics::RankCorrelation Perl module to calculate Spearman’s
rank correlation coefficients between all pairwise comparisons of the
8,361 *N. crassa* genes using expression levels from the 48 Louisiana indi-
viduals. We used the absolute value of each correlation coefficient as input
into the hclust package in the R software environment (64) and applied
the method of reference 65 to deﬁne groups of coregulated genes. Briefly,
given a correlation coefﬁcient R, each cluster contains a set of genes
among which any pair exhibit correlated expression across the strains with
a coefﬁcient of at least R. To establish a cutoff for R in Table S1 in the
supplemental material, we permitted expression measurements between
individuals and repeated the clustering analysis; then, given a cluster size s,
in this null set we tabulated the number $n_{perm}$ of clusters containing at
least s genes. Across 10 such permutations, for R = 0.4 and s = 9, $n_{perm}$ was zero (see Table S1); clusters of size 9 or larger in analysis of the real
data are reported in Data Set S1.

**Functional enrichment tests in clusters.** We tested each cluster of
genes subject to coregulation across Louisiana strains for enrichment of
gene functions as follows. We downloaded *N. crassa* gene annotations
from the FunCat database maintained by the Munich Information Center
for Protein Sequences (20). We eliminated from analysis all clusters that
contained <9 genes and those with <2 genes with assigned FunCat an-
notations; the ﬁnal set comprised 143 clusters. We used a custom Perl
script and the R function phyper to assess enrichment of each FunCat
notations; the final set comprised 143 clusters. We used a custom Perl script
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notations; the final set comprised 143 clusters.

**NCU04628 conservation.** Homologs of NCU04628 were identiﬁed
using the Joint Genome Institute MycoCosm fungal genomics web re-

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source (67). Multiple sequence alignment was performed using the software program MUSCLE (68).

**Hyphal morphology assays.** The deletion strain for NCU04826 (FGSC 16805) (6) (mating type a) was crossed to an isogenic wild-type control (mating type A; FGSC 2489) on Westergaard synthetic crosses medium (69), and progeny were screened for hygromycin resistance on sorbose medium (70) with and without 200 μg/ml hygromycin (Sigma). Five hygromycin-resistant and five hygromycin-sensitive progeny were inoculated onto petri dishes containing Vogel’s minimal medium (VMM) (71) and allowed to grow overnight. Strains were observed under a dissecting microscope, and hyphal branching was scored as follows. For each hypha, the number of nodes $n$ was counted, starting from the tip and moving toward the center of the colony, until the first subtending branch that was also branching was reached. Values of $n$ were averaged across 20 randomly selected hyphal tips for each individual.

For imaging of colonies of strains harboring deletions in genes coregulated with NCU04826 in Fig. S1 in the supplemental material, each deletion strain in the FGSC 2489 background (6) was obtained from the Fun- gual Genetics Stock Center and grown on VMM overnight.

**Generation of Nc_nmr6 deletion strain.** The gene knockout procedure for NCU00789 was modified from a previously published method (6) as follows. The strain FGSC 9717 (mus-51; bar a) was grown for 3 days at 30°C in the dark, followed by another 7 days at room temperature, in 100 ml VMM containing 2% (wt/vol) sucrose and 1.5% (wt/vol) agar. Conidia were collected by filtration, washed three times with ice-cold 1 M sorbitol, and resuspended in a final volume of 5 ml. An aliquot of 90 μl (~10^9 conidia) was mixed with 1.0 μg of DNA encoding an NCU00789 deletion cassette (the sequence of the hph hygromycin resistance gene [72] flanked by ~1 kb of the genomic region upstream of NCU00789 in the N. crassa reference genome and ~1 kb of the region downstream of the gene, kindly provided by Carol Ringelberg, Department of Genetics, Dartmouth Medical School). DNA entry into conidia was via electroporation on a Bio-Rad Gene Pulser II with the following settings: 1.5 kV, 600 Ω, and 25 μF. Treated conidia were then mixed with 900 μl ice-cold 1 M sorbitol and 30 ml of a top agar solution containing 20% sucrose, 0.5% fructose, and 0.5% glucose (FGS) warmed to 50°C. This mixture was overlaid on each of three agar plates containing FGS medium and 200 μg/ml hygromycin, and subsequently incubated at 30°C for 3 days. Individual colonies were used to inoculate agar slants (3 ml VMM containing 2% sucrose and 200 μg/ml hygromycin), from which DNA was isolated and screened by PCR with the primers 5’tggaatagtcggctct3’d (hyg) and 5’gcggataaacaatttccacag3’ (NCU00789) using Phire Hot Start polymerase according to the manufacturer’s recommendations (Thermo Scientific; catalog no. F-130).

To reduce the likelihood of background mutations in NCU00789 deletion strains, each PCR-confirmed mutant was backcrossed to a wild type as follows. FGSC 2489 was grown as a female strain on Westergaard medium (69) for 7 days to allow formation of protoperithecia and crossed with conidia from each deletion strain of the opposite mating type. Ascopores from each cross were collected, placed in sterile water, and induced to germinate via incubation at 60°C for 30 min. Treated ascospores were plated on VMM–2% sucrose plates containing 400 μg/ml hygromycin and grown at 30°C for 14 h. Germinated spores were isolated and used to inoculate 3-ml VMM–2% sucrose slants. Cultures were checked for loss of the bar gene by lack of growth on VMM without NH₄NO₃, 0.5% 1-proline, 2% sucrose, and 400 μg/ml Ignite. Progeny from each cross were also checked for integration of the hygromycin cassette by PCR as described above.

**Transcriptional profiling of N. crassa deletion strains.** Following methods described elsewhere (73), we cultured, harvested, and isolated RNA from one replicate of the pp4 deletion strain (FGSC 12454) (6), one of the NCU05257 deletion strain (FGSC 16020) (6), and one of OR74A (FGSC 4200) as a wild-type control, and separately, we isolated RNA from one replicate of the NCU00789 deletion strain and an OR74A wild-type control. Multiplex library construction from each isolate’s cDNA was done according to the manufacturer’s protocol (Truseq v2 LT sample prep kit; Illumina) except that adaptors were diluted 5× before ligation to blunt end cDNA. The RNA-seq libraries were sequenced on an HiSeq2500 instrument (Illumina) as single-end 50-bp reads.

In each RNA-seq data set, we eliminated from analysis all genes with <5 mapped reads. We then assessed the significance of differential expression of each gene between each deletion strain and its respective control using the R software package DEseq (74), normalizing read counts between lanes in each comparison using the estimateSizeFactors function and using expression variance across the pooled samples as described previously (74). Genome-wide expression measurements for each deletion strain are given in Data Set S2 in the supplemental material.

We used a resampling approach to evaluate the expression response of a regulon of interest in the NCU05257 (asi-1), NCU08301 (pp4), or NCU00789 (Nc_nmr6) deletion transcriptional profile. For this purpose, we first tabulated the average of the log2 fold change in expression between the wild type and the deletion strain across the genes of the regulon. We then applied the same procedure to a gene set of the same size as the true regulon, randomly drawn from the pool of expressed genes (excluding those present in the regulon). The latter resampling procedure was repeated 10,000 times. Significance in a one-tailed test of the hypothesis that the true regulon was expressed at lower levels than the null expectation, was assessed as the proportion of resampled gene sets whose average expression ratios were less than or equal to that of the true regulon.

**Amino acid starvation assays.** The deletion strain for NCU05257 (FGSC 16020) was crossed to an isogenic wild-type control (FGSC 4200), and five progeny strains bearing the deletion and five wild-type progeny were recovered, as described above. Agar plugs were taken from each progeny culture and used to inoculate race tubes (75) containing VMM (71) with and without 6 mM 3-AT (Sigma). Each tube was incubated in constant light at 25°C, and the location of the hyphal front was recorded daily until it reached the opposite end of the tube. For comparative purposes, we also measured the growth rate for three replicates of a gpc-1 mutant strain (FGSC 4264) (76) on race tubes containing VMM with and without 6 mM 3-AT.

**Differential expression between N. crassa populations.** We calculated normalized expression measurements for each of 19 Caribbean isolates and 3 Panamanian isolates of N. crassa as described above, using previously published RNA-seq data (11). Given these measures and the expression profiles of Louisiana isolates as described above, we tested each gene in turn for differential expression between the Caribbean and Louisiana populations using the Wilcoxon test; we then corrected these nominal empirical P values using the Benjamini-Hochberg method (66) for multiple hypothesis testing. Using the set of 1,539 genes with differential expression significant at a Benjamini-Hochberg-corrected P value of <0.05, we evaluated enrichment of FunCat annotations as above.

**Curation of nitrogen metabolism gene sets in N. crassa.** To define the nitrogen metabolite repression program in N. crassa in Data Set S4 in the supplemental material, we used the NMR (nitrogen metabolite repression) signature from Saccharomyces cerevisiae (77) to identify 59 homologs in the reference sequence of N. crassa by a best-reciprocal-BLAST search (protein BLAST with E value cutoff of 1e−04). We also included NMR-controlled genes characterized for Neurospora (78–81) for a total of 66 genes.

To define a broader set of nitrogen metabolism genes in N. crassa, we downloaded expression measurements of Magnaporthe grisea under nitrogen-starved and nitrogen-replete conditions from reference 33 and retained for analysis all genes for which the absolute value of the log2 signal intensity ratio between the conditions was 2-fold or greater. We then used the FungiDB database (82) to identify N. crassa orthologs of these genes (see Data Set S4 in the supplemental material). We also added genes annotated as involved in nitrogen metabolism in N. crassa using GO (gene ontology) terms, FunCat terms, and the Broad Institute annotation.
As our set of candidate regulators of NMR in Data Set S4, we compiled a list of *S. cerevisiae* regulators from reference 29 and examined all *Neurospora* homologs of these genes using homology search criteria so as to include both orthologs and paralogs (protein BLAST without best-reciprocal-BLAST requirement; E value cutoff of 1e–10).

**Carbon catabolite repressed genes in *N. crassa***. Our list of carbon catabolite repressed genes was derived from transcriptional profiling of the cre-1 deletion strain performed previously (32). Specifically, we used the 75 genes showing expression levels that were significantly elevated in the cre-1 deletion strain compared to wild-type levels when grown on minimal medium (see Table S2 in reference 32).

Association of gene expression with genotype. To test expression of a given nitrogen metabolism gene for association with inheritance at the DNA level across Louisiana strains, given the normalized expression measurements for a transcript and genotypes at a single-nucleotide variant, we first split the strains into two groups on the basis of inheritance at the variant and then used the Wilcoxon test with Benjamini-Hochberg correction as described above.

**SUPPLEMENTAL MATERIAL**

Supplemental material for this article may be found at http://mbio.asm.org/lookup/suppl/doi:10.1128/mBio.01046-13/-/DCSupplemental.

**ACKNOWLEDGMENTS**

We thank J. Welch for strain handling and microscopy and C. Hall for expression data processing.

This work was supported by NIH R24 GM081597 to J.W.T., N.L.G., and R.B.B. We are pleased to acknowledge use of materials generated by NIH Program Project Grant P01GM068087 (functional analysis of a model filamentous fungus).

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