Characterizing the gene–environment interaction underlying natural morphological variation in Neurospora crassa conidiophores using high-throughput phenomics and transcriptomics

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Abstract

Neurospora crassa propagates through dissemination of conidia, which develop through specialized structures called conidiophores. Recent work has identified striking variation in conidiophore morphology, using a wild population collection from Louisiana, United States of America to classify 3 distinct phenotypes: Wild-Type, Wrap, and Bulky. Little is known about the impact of these phenotypes on sporulation or germination later in the N. crassa life cycle, or about the genetic variation that underlies them. In this study, we show that conidiophore morphology likely affects colonization capacity of wild N. crassa isolates through both sporulation distance and germination on different carbon sources. We generated and crossed homokaryotic strains belonging to each phenotypic group to more robustly fit a model for and estimate heritability of the complex trait, conidiophore architecture. Our fitted model suggests at least 3 genes and 2 epistatic interactions contribute to conidiophore phenotype, which has an estimated heritability of 0.47. To uncover genes contributing to these phenotypes, we performed RNA-sequencing on mycelia and conidiophores of strains representing each of the 3 phenotypes. Our results show that the Bulky strain had a distinct transcriptional profile from that of Wild-Type and Wrap, exhibiting differential expression patterns in clock-controlled genes (ccgs), the conidiation-specific gene con-6, and genes implicated in metabolism and communication. Combined, these results present novel ecological impacts of and differential gene expression underlying natural conidiophore morphological variation, a complex trait that has not yet been thoroughly explored.

Keywords: Neurospora crassa; natural variation; conidiophore; vegetative development; morphology; sporulation; germination; convolutional neural network; phenomics; transcriptomics

Introduction

The classic filamentous fungus Neurospora crassa reproduces vegetatively through the development of macroconidiophores (conidiophores). These specialized structures serve to generate haploid macroconidia (conidia) that sporulate and germinate to propagate the asexual life cycle throughout an environment. Conidiophore development in N. crassa is well understood following decades of study in the laboratory. Briefly, the conidiation process begins with perpendicular growth of an aerial hypha. The filament undergoes minor and major constriction budding to generate spores that will eventually break off as mature conidia (Springer and Yanofsky 1989). Environmental cues such as desiccation, aeration, nutrient deprivation, and light are required for induction of this process, which is under strict circadian regulation (Sargent and Kaltenborn 1972; Nelson et al. 1975; Loros and Dunlap 2001). Many genes guiding conidiophore development have been identified, and their expression has been characterized throughout the roughly 12-h developmental timeline (Springer and Yanofsky 1989; Greenwald et al. 2010). While these environmental, temporal, and genetic signals of conidiophore development are well understood, little is known about the morphological variation of these structures, particularly in natural populations.

Conidiophore architectural variation has just recently been explored in a wild population collection of N. crassa isolates from Louisiana, United States of America (Krach et al. 2020). This work identified 3 novel and distinct phenotypes, named Wild-Type (WT), Wrap, and Bulky, where WT conidiophores formed linear chains of conidia, Wrap conidiophores wrapped around and/or stuck to hyphal filaments, and Bulky conidiophores formed tight clusters (Krach et al. 2020). These phenotypes were found to be upheld throughout the duration of their development and had a significant impact on resulting “spore shadows,” or distributions of spore dispersal distances. This suggests that conidiophore architectural phenotype may affect colonization capacity of the organism, a hypothesis further explored in this study.
Here, using the wild population collection of 21 strains, we reinforce that conidiophore morphology impacts spore dispersal distances on a larger scale, also affecting the maximum distance a conidium may travel. Furthermore, we show that conidiophore phenotype influences both germination rates and germination times of conidia on growth media containing different carbon sources, particularly in the Bulky strain. Applying methods developed in Krach et al. (2020), we crossed homokaryotic strains for each phenotype and machine-classified conidiophores of the resulting progeny. These phenotypic counts were used to fit a model where at least 3 genes and 2 epistatic interactions contributed to conidiophore morphology, a complex trait with an estimated heritability of 0.47. To identify genes underlying these conidiophore phenotypes, we performed RNA-sequencing (RNA-seq) on both mycelia and conidiophores of a representative strain for each phenotypic group. Our results show that the Bulky strain has a unique transcriptional profile from WT and Wrap, exhibiting the most striking differential expression patterns in clock-controlled genes (ccg-1, ccg-2, ccg-14), metabolic genes responsive to starvation (acu-6, NCU04482), genes involved in communication (doc-1, doc-2, pip-1, pip-2), and the conidiation-specific gene, con-6. To our knowledge, this is the first report of genes implicated in natural morphological variation of N. crassa conidiophores. Together, this work lends insight to phenotypic variation of the conidiophore and its robustness to environmental perturbations, as well as genetic differences underlying this variation in natural populations.

**Materials and methods**

**Strains and media**

Wild Louisiana isolates were obtained from the Fungal Genetics Stock Center (FGSC, Manhattan, KS, USA) and are listed in Table 1. Strains were maintained on 1.8% glucose/1.8% fructose/1.5% agar slants with 1× Vogel’s media and recommended biotin and trace element supplements (hereon referred to as standard VM; Davis and de Serres 1970).

**Sporulation experiment**

Representative strains for each phenotypic class (FGSC8872, FGSC8876, and FGSC3943) were inoculated on standard VM and incubated at 30°C for 30 h. Mycelia were then harvested onto a 60-mm diameter nitrocellulose membrane with 0.45-μm pore size (Whatman Protran BA-85, Maidstone, England), inverted onto a new plate with media as described above, and set under the light for 20 h before being transferred to a new environment for sporulation.

To explore sporulation on a larger scale, this “new environment” was 18 × 18 inch plastic cake platters repurposed as Petri dishes (Fineline Settings, Middletown, NY, USA). The platters and their clear plastic lids were sterilized with ethanol and UV light before use. The nitrocellulose membranes containing isolated conidiophores were placed at the center of each cake platter, where sorbose + fructose + glucose (SFG) media (Davis and de Serres 1970) surrounded a 60-mm diameter blank space now occupied by the membrane. Each platter was covered with a lid through which light could penetrate and placed under the light for 4 days to allow sporulation, germination, and subsequent colonial growth. Pictures of each plate were taken with an iPhone XS, and ImageJ (Schneider et al. 2012) was used to measure the distance from the center of each nitrocellulose membrane to the center of a colony, sampling up to 75 colonies per platter. Three biological replicates of each strain were performed.

**Germination assay**

Representative strains used for each phenotypic group were FGSC8872 for WT, FGSC8876 for Wrap, and FGSC2229 for Bulky. Cultures were grown up on standard VM for 5 days under light (Davis and de Serres 1970). Conidia were then suspended in water, counted using a Cellometer Auto 2000 (Nexcelom, Inc. Lawrence, MA USA), and diluted to roughly 1.00 × 10^3 cells/ml.

| Strain number | FGSC | Perkins | Mat | Strain provenance | Collection site | Substrate/annotation |
|---------------|------|---------|-----|-------------------|----------------|---------------------|
| Wild Strains  |      |         |     |                   |                |                     |
| D110          | 8870 | 4448    | A   | Dettman, J.       | Franklin, LA   | Sugarcane           |
| D111          | 8871 | 4449    | a   | Dettman, J.       | Franklin, LA   | Sugarcane           |
| D112          | 8872 | 4453    | A   | Dettman, J.       | Franklin, LA   | Sugarcane           |
| D114          | 8874 | 4464    | A   | Dettman, J.       | Franklin, LA   | Sugarcane           |
| D116          | 8876 | 4481    | a   | Dettman, J.       | Franklin, LA   | Sugarcane           |
| D118          | 8878 | 4491    | a   | Dettman, J.       | Franklin, LA   | Sugarcane           |
| JW09          | 2229 | 4491    | A   | Welch, J.         | Welsh, LA      | Burned grass        |
| JW10          | 2229 | 4491    | A   | Welch, J.         | Welsh, LA      | Burned grass        |
| JW59          | 3200 |         | a   | Welch, J.         | Coon, LA       | Burned stumps       |
| JW66          | 3211 |         | a   | Welch, J.         | Sugartown, LA  | Pine burn           |
| JW70          | 3199 |         | A   | Welch, J.         | Coon, LA       | Burned stumps       |
| JW75          | 3943 |         | a   | Welch, J.         | Houma, LA      | Sugarcane burn      |
| JW7           | 847  |         | A   | Lein               | Louisiana, LA  | Sugarcane burn      |
| D113          | 8873 | 4454    | a   | Dettman, J.       | Franklin, LA   | Sugarcane           |
| D119          | 8879 | 4500    | a   | Dettman, J.       | Franklin, LA   | Sugarcane           |
| JW20          | 3212 |         | A   | Welch, J.         | Houma, LA      | Sugarcane burn      |
| JW76          | 3943 |         | a   | Welch, J.         | Houma, LA      | Sugarcane burn      |
| JW159         | 2221 |         | a   | Welch, J.         | Iowa, LA       | Grass burn          |
| JW160         | 2222 |         | A   | Welch, J.         | Iowa, LA       | Grass burn          |
| JW162         | 2223 |         | a   | Welch, J.         | Marrero, LA    | Wood burn           |
| JW164         | 2224 |         | a   | Welch, J.         | Roanoke, LA    | Grass burn          |
| JW167         | 2228 |         | a   | Welch, J.         | Marrero, LA    | Unknown             |
| OR74A         | 2489 |         | A   | FGSC               |                |                     |

One of these strains (OR74A) is the genome reference strain (Galagan et al. 2003).
Table 2. Wild isolates selected for first set of crossing to generate F1s.

| Parent (A) | Phenotype  | Parent (a) | Phenotype | Offspring |
|------------|------------|------------|-----------|-----------|
| FGSC8872   | WT 77.78%  | FGSC3943   | Bulky 48.78% | #42, 43   |
| FGSC8872   | WT 77.78%  | FGSC8876   | Wrap 53.85%  | #151, 156 |
| FGSC2229   | Bulky 46.55%  | FGSC8876   | Wrap 53.85%  | #83, 91   |

Some of the Wild isolates in Table 1 were crossed to form an F1 as described in Materials and Methods to make them homokaryotic.

Table 3. F1s selected for crossing to generate F2s.

| Parent (A) | Penetrance* | Parent (a) | Penetrance |
|------------|-------------|------------|-----------|
| WT #151    | 100%        | Bulky #91  | 100%      |
| WT #151    | 100%        | Wrap #43   | 80%       |
| Wrap #156  | 83%         | Bulky #91  | 100%      |

The progeny of these crosses (as described in the Materials and Methods) are shown in Fig. 5. *Fraction of offspring displaying a particular phenotype. The phenotype is defined in the first or fourth column.

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Microscopy and image deconvolution

Nitrocellulose membranes containing conidiophores were visualized on an inverted microscope (Axio Observer A1, Carl Zeiss Microscopy, LLC, Thornwood, NY, USA) at 20× magnification and brightfield images were taken with a charge-coupled device camera (AxioCam HRm, Carl Zeiss Microscopy, LLC, Thornwood, NY, USA). Multiple z-slices were captured and overlaid in ImageJ (Schneider et al. 2012) to convey a complete representation of the 3D conidiophore structure. Augmentation including contrast enhancement and noise and background subtraction was conducted on image stacks to isolate conidiophores from underlying mycelia and/or aerial hyphae.

Classification of conidiophores was carried out based on the pretrained convolutional neural network (Krach et al. 2020). Starting from the classifier published in Krach et al. (2020), we used 327 additional images to fine-tune and adapt the published model for the new F2 dataset. The 3 phenotypes were equally distributed among the image samples and train-validation-test separation was approximately 8:1:1. The model was trained with an additional 100 epochs, and the best model was selected by the best accuracy in the validation set. Other hyperparameters and model evaluations were kept the same as in the original publication (Krach et al. 2020).

RNA extraction

Previous work has shown that growing fungi on solid medium overlaid with a nylon membrane facilitates harvest of the mycelium, produces sufficient biomass, and enhances RNA quality (Schumann et al. 2013). Large Petri dishes (150 × 15 mm) with standard VM were covered with a Hybond XL nylon membrane (Amersham, Buckinghamshire, UK). Membranes were inoculated with conidia from representative wild strains (FGSC8872, FGSC8876, and FGSC2229) and placed at 30°C for 30 h to allow for mycelial growth. Mycelia were then harvested with a sterile razor blade into a microcentrifuge tube and immediately frozen with liquid nitrogen for later RNA extraction. Four biological replicates of each strain were grown and harvested.

Conidiophore cultures were first grown following the protocol above. At the 30-h time point, mycelia were covered with a 0.45-μm pore size nitrocellulose membrane (Whatman Protran BA-85, Maidstone, England) and placed under the light for 20 h. Aerial hyphae penetrated the nitrocellulose membrane, allowing conidiophore isolation from this top layer. Conidiophores were then harvested with a sterile razor blade and immediately frozen with liquid nitrogen for RNA extraction. Four biological replicates of each strain were grown and harvested. The method was adapted from Krach et al. (2020) and Schumann et al. (2013).

Mycelia and conidiophore samples were later ground to fine powder using the Cellcrusher tissue pulverizer (Cellcrusher, Cork, Ireland). The method was adapted from Krach et al. (2020) and Schumann et al. (2013).
Ireland) submerged in liquid nitrogen. Samples were transferred to a new microcentrifuge tube and kept frozen. Total RNA was later isolated and suspended in RNase-free water using the Qiagen RNeasy Plant Mini Kit (QIAGEN, Inc., Valencia, CA, USA) following the protocol outlined by the manufacturer. RNA integrity was assessed using the Agilent Bioanalyzer and RNA concentration was quantified using the fluorometric Qubit analyzer at the Georgia Genomics and Bioinformatics Core.

### RNA library preparation

Libraries were prepared according to the KAPA Stranded RNA-seq Kit. The libraries were then pooled and sequenced on a NextSeq2000 instrument to generate paired-end reads. Library preparation, pooling, and sequencing were conducted at the Georgia Genomics and Bioinformatics Core.

### RNA-seq data analysis

Sequencing reads were demultiplexed by BaseSpace (Illumina). Reads were trimmed from the adaptor sequences using the cutadapt software (Martin 2011) and aligned to the *N. crassa* genome (NC12) following star quantification mRNA-seq pipeline. Gene expression data were analyzed in R version (4.0.3). The differentially expressed genes were measured using Bioconductor: DESeq2 (Love et al. 2014). Differential gene expression was measured between cell type and strain. Power estimation for differential expression analysis was conducted using the Vanderbilt power calculation for RNA-seq experiment Shiny app (Guo et al. 2014). Genes were filtered for an adjusted $P$-value and absolute fold change of $10^{-5}$ and absolute log FC > 3.0, respectively.

Differentially expressed genes were extracted from the results table generated in the DESeq2 pipeline. Expression values for the top 50 most significantly differentially expressed genes were used to hierarchically cluster samples. The DESeq2 package was used to conduct principal component analysis (PCA) and generate PCA plots. Hierarchical clustering was used to discriminate sample phylogeny in the experimental setting. Pretty Heatmaps package (Kolde 2018) was used to visualize the expression patterns of the most significantly differentially expressed genes for each experimental design.

### Results

**Conidiophore architectural phenotype may impact colonization capacity of *N. crassa* by affecting the maximum dispersal distance of released conidia**

Previous work showed that conidiophore architectural phenotype may play a role in colonization capacity of the organism by affecting spore dispersal distance (Krach et al. 2020). Dispersal experiments to characterize population structure have been extensively used in population genetics (Dobzhansky and Wright 1943; Trapnell and Hamrick 2005; Hamrick and Trapnell 2011). For example, neighborhood size is directly related to the dispersal distribution (Dobzhansky and Wright, 1943; Wright, 1968). A variety of studies have been recently carried out on spore dispersal (Roper et al. 2008, 2010; Fritz et al. 2013). A key element to the design of these experiments is the size of the grid on which dispersal is measured (Powell and Dobzhansky 1976). We wanted to scale up our previous spore shadow experiment to explore spore dispersal over a larger area (Krach et al. 2020). Limiting the boundaries of a dispersal experiment to measure neighborhood size introduces biases in specifying the distribution of dispersal distances of the organism (Lemke 1985). To expand these boundaries, we allowed conidiophores of each architectural phenotype to sporulate and germinate on 0.1% SFG medium, reinventing an 18 × 18 inch plastic cake platter as a large Petri dish.

Consistent with the findings in Krach et al. (2020), fewer colonies ($n = 35$) developed on the cake platter (see Materials and Methods) from WT conidiophores at the center of the cake platter compared to that of Wrap and Bulky ($n = 195$ and $n = 104$, respectively; see Supplementary Fig. 1 and histograms in Fig. 1). The distribution of spore dispersal distances by the Wrap conidiophores was significantly different from that of both WT ($P = 0.04926$) and Bulky ($P = 0.004794$) phenotypes, found by 2-sample Kolmogorov-Smirnov tests (Fig. 1; Kendall and Stuart 1979). No significant difference was found in spore dispersal distance distributions between the WT and Bulky groups ($P = 0.2734$). Both findings are consistent with sporulation patterns observed at a smaller scale, indicating that the spore shadows displayed by each conidiophore phenotype are upheld in this new, larger environment (Krach et al. 2020). Taking advantage of this larger controlled environment, we then sought to investigate whether the conidiophore architectural phenotype had an impact on the maximum distance a conidium could travel and subsequently germinate.

Previous work showed that WT spores traveled the least distance following dispersal (Krach et al. 2020). The null hypothesis examined is whether the maximum distance traveled by Wrap and Bulky could be considered to be drawn from the same distribution of the maximum distance squared traveled by WT (Table 4). A Shapiro–Wilk test of normality (Royston 1995) was performed on WT ($SW = 0.9313$, $P = 0.0306$)—departures from normality are not strong. To a first approximation, it is reasonable to suppose that spore dispersal coordinate is normally distributed (Powell and Dobzhansky 1976). The distribution of standardized distance squared is then chi-squared with 1 degree of freedom. The largest distance traveled has a known distribution, and the probability that the largest distances seen in Wrap and Bulky are drawn from the dispersal distribution of WT is extremely unlikely (David 1981).

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**Fig. 1.** Distribution of spore dispersal distances scaled up. Histogram of distances in centimeters to the center of each colony from the center of each nitrocellulose membrane. Total colony counts combining 3 replicates are as follows: WT = 35, Wrap = 195, and Bulky = 104. Strains selected for each phenotypic group were as follows: FGSC8872 for WT, FGSC8876 for Wrap, and FGSC3943 for Bulky. Results of 2-sample Kolmogorov–Smirnov tests for each phenotype pair are as follows: WT-Wrap $D = 0.24982$, $P = 0.04926$; Wrap-Bulky $D = 0.2109$, $P = 0.004794$; WT-Bulky $D = 0.18709$, $P = 0.2734$. The null hypothesis examined is whether the maximum distance traveled by Wrap and Bulky could be considered to be drawn from the same distribution of the maximum distance squared traveled by WT [Table 4]. A Shapiro–Wilk test of normality [Royston 1995] was performed on WT ($SW = 0.9313$, $P = 0.0306$)—departures from normality are not strong. To a first approximation, it is reasonable to suppose that spore dispersal coordinate is normally distributed [Powell and Dobzhansky 1976]. The distribution of standardized distance squared is then chi-squared with 1 degree of freedom. The largest distance traveled has a known distribution, and the probability that the largest distances seen in Wrap and Bulky are drawn from the dispersal distribution of WT is extremely unlikely [David 1981].
Spores from conidiophores with different architectural phenotypes show distinct germination patterns on different carbon sources

We sought to further explore the potential impact of conidiophore phenotype on environmental colonization through the lense of germination. To do this, we plated known amounts of conidia from a WT, Wrap, and Bulky strain (FGSC8872, FGSC8876, and FGSC2229, respectively) onto media containing 1% sorbose, a monosaccharide known to limit the organism to colonial growth. Each colony on a plate represents germination of 1 conidium. We varied the media to assess the germination behavior of each conidiophore phenotype on different carbon sources: Fructose/Glucose, Mannose, and Xylose. In addition to the different carbon sources, we also explored varying concentrations of each carbon source. Germination rates were determined from colonies counted/expected colonies and are depicted below in Fig. 2.

First, we compared spore germination rates on different carbon sources. When compared to different concentrations of the same carbon source, all strains show the highest germination rate with a sugar concentration of 0.1%, with the one exception of WT on 0.01% SFG (Fig. 2). Interestingly, spores from WT conidiophores show a jump in germination from 39% to 91% when fructose and glucose are decreased by an order of magnitude, whereas Wrap and Bulky groups both show lower germination rates. This pattern exhibited by WT indicates a starvation response that is unique to this phenotypic group on SFG medium. In all concentrations of media containing mannose and xylose, Bulky spores consistently show a significantly elevated germination rate compared to the other 2 groups. This is a pattern unique to these 2 carbon sources and is not observed at either concentration of SFG medium.

In addition to germination rate, germination time also varies with conidiophore phenotype under some medium conditions (Fig. 3). All 3 phenotypic groups germinated synchronously on SFG medium, as well as on the positive control of 1.8% Glucose Vogel’s Medium. However, on all concentrations of mannose and xylose, Bulky conidia germinated earlier than the other 2 phenotypic groups, which remained synchronous. Interestingly, this unique temporal behavior exhibited by Bulky conidia is consistent with the uniqueness of higher germination rates observed by Bulky on these same carbon sources.

On all carbon sources, germination of each strain is delayed as concentration of the sugar decreases. This is true in all cases, except for Bulky on 0.1% and 1% Mannose where germination occurs on Day 3 under both conditions. The germination timeline is very predictable on mannose and xylose (Fig. 3). The key points are that Bulky is behaving differently from WT and Wrap with regard to germination rate and timeline on some media (mannose and xylose), and WT is behaving differently on other media (SFG; Figs. 2 and 3).

After characterizing germination on these new carbon sources, we were curious if conidiophore architectural phenotype was upheld once conidiation occurred. We imaged conidiophores of each strain on 0.1% mannose and 0.1% xylose, as

Table 4. Maximum spore dispersal distances traveled by Wrap and Bulky are significantly different from WT.

| Strain (phenotype) | Maximum distance | Tail probability of the WT distribution of the largest rank of distance squared |
|-------------------|-----------------|-----------------------------------------------------------------------------|
| FGSC8872 (WT)    | 23.50 cm        | —                                                                           |
| FGSC8876 (Wrap)  | 25.78 cm        | 3.80 x 10^-64                                                              |
| FGSC3943 (Bulky) | 24.16 cm        | 2.15 x 10^-51                                                              |

Maximum distance in centimeters among 3 biological replicates traveled by a germinating conidium of each phenotype. For all 3 values, the mean movement (–X = 4.729) was subtracted, the sample variance (S^2 = 2.165) was used to standardize, and the resulting z-value (Z = |X–X|/S), squared. The resulting z^2 is then approximately chi-squared in distribution with 1 degree of freedom. The tail probability for the distribution of the largest rank from this chi-squared distribution was then computed as if these z^2 values were all order statistics from the same distribution (David 1981).
Germination rates were highest with this sugar concentration for all 3 strains. On both 0.1% mannose and 0.1% xylose, WT, Wrap, and Bulky architectural phenotypes were upheld in the conidiophores (Fig. 4).

**Crosses between homokaryotic F1s suggest at least 3 genes contribute to conidiophore architectural phenotype**

Previous work has shown that at least 2 genes contribute to the conidiophore architectural phenotype, and the trait has an estimated heritability of 0.23 (Krach et al. 2020). This model was based on progeny phenotype counts from crosses between original Louisiana wild isolates. Because wild strains may possibly be heterokaryotic, this inheritance model would be more robust by crossing homokaryotic F1s, quantifying conidiophore phenotypes of the resulting F2, and fitting the model to those phenotypic ratios.

We selected F1 parents with the greatest penetrance, or fraction of conidiophores displaying a particular conidiophore phenotype, and conducted crosses between groups. We then selected 30 random ascospore progeny from each cross. Conidiophores from these F2 were isolated, imaged, and classified using the automated classification method developed and successfully
applied in Krach et al. (2020). To accommodate the changes of different batches and generations, we fine-tuned the classification model with 327 images and evaluated the classification performance (Supplementary Table 1). The phenotypic ratios of each F1 parent and the resulting F2 progeny are depicted below in Fig. 5 and summed in Table 5.

Fig. 5. Percent of conidiophore phenotypes observed in F1 parents and resulting F2 progeny. Crosses were conducted between majority WT × Bulky (top), Bulky × Wrap (middle), and WT × Wrap (bottom) F1s, whose phenotypic ratios are depicted as the 2 left most bars on each plot. Phenotype ratios of the resulting F2 progeny are plotted as the following 30 bars.
As seen in Fig. 5, there are fewer WT phenotypes in the cross of WT × Wrap than in the other 2 crosses as if Wrap is “standing against” WT. Only 26% of the WT × Wrap offspring are WT as opposed to 50% or 46% WT in the crosses of Wrap × Bulky or WT × Bulky, respectively (Table 5). Conidiophore phenotype counts (Table 5) were used to estimate inheritance models successfully fitted in Krach et al. (2020). The models were fitted by the Method of Maximum Likelihood (ML) using iteratively reweighted least squares (IRLS; Kendall and Stuart 1979; Asmussen et al. 1987). The models incorporate 3 hypothetical genes A, B, and C that may contribute to conidiophore phenotype, and each gene has an allelic effect \( x \), \( \beta \), and \( \gamma \), to give rise to the WT, Wrap, and Bulky phenotypes, respectively. A cross between parents of different genotypes includes 2 dominant genes that interact epistatically to determine the conidiophore phenotype, where interactions are depicted as \( \alpha \beta \), \( \beta \gamma \), and \( \alpha \gamma \). The results for ML fitting the models computed with IRLS to a relative error <10\(^{-8}\) after 9 iterations are summarized in Table 6.

The models fitted in Table 6 represent a hierarchy of hypotheses about the inheritance of conidiophore phenotype varying in complexity from a model with no genetic effects at all to a full epistatic model, in which all 3 loci interact pairwise (Supplementary Fig. 2). The full epistatic model with 3 genes fitted the phenotypic ratios, and the weakest epistatic interaction was between the A (WT) and C (Bulky) genes (Table 7). There is a significant interaction \( \alpha \beta \) between gene A (WT) and gene B (Wrap) that captures the drop in phenotypic count of WT seen in Fig. 5 for the cross A(WT) × B(Wrap). We concluded that at least 3 genes contribute to conidiophore phenotype because none of the allelic effects \( x \), \( \beta \), or \( \gamma \), could be dropped in the hierarchical testing (Supplementary Table 5). Using an additive model and the model without gene effects, we were able to estimate a heritability as the ratio of additive variation to total variation (\( H^2 \)) of 0.47 for this complex trait. This is an increase from our previously estimated \( H^2 \) of 0.23, suggesting that using homokaryotic F2s reduced noise in the data (Krach et al. 2020).

### Table 5. F2 phenotype counts.

| A (WT)   | B (Wrap) | C (Bulky) |
|----------|----------|-----------|
| A × B    | N\(_{11}\) = 103 | N\(_{12}\) = 167 | N\(_{13}\) = 126 |
| B × C    | N\(_{21}\) = 142 | N\(_{22}\) = 86 | N\(_{23}\) = 58 |
| A × C    | N\(_{31}\) = 170 | N\(_{32}\) = 108 | N\(_{33}\) = 88 |

Table 5. In a natural reading of the document, the phenotype counts were generated by classifying the offspring by conidiophore phenotyping by machine learning (see Materials and Methods).

### Table 6. Fitting of hierarchy of inheritance models.

| Model               | \( \chi^2 \)  | df | P      | \( \chi^2_{H0} - \chi^2_{HA} \) | df | P for HA vs. H0 | Notes | Notes |
|---------------------|--------------|----|--------|---------------------------------|----|-----------------|-------|-------|
| Full epistatic      | 0.00         | 1  | —      | —                               | —  | —               | HA = full epistatic |
| \( \alpha \beta = 0 \) | 13.37       | 1  | <0.0001| 13.37-0.00 = 13.37               | 1  | <0.0001         | HA = full epistatic |
| \( \alpha \beta = \alpha = 0 \) | 16.27       | 2  | <0.0001| 16.27-13.37 = 9.12               | 1  | 0.002           | HA = \( \alpha \beta \) = 0 |
| \( \beta \gamma = 0 \) | 43.42       | 1  | <0.0001| 43.42-0.00 = 43.42               | 1  | <0.0001         | HA = full epistatic |
| \( \beta \gamma = \beta = 0 \) | 43.21       | 2  | <0.0001| 43.21-43.42 = 0.00               | 1  | 0.64            | HA = \( \beta \gamma \) = 0 |
| \( \alpha \beta = \beta \gamma = 0 \) | 29.88       | 2  | <0.0001| 29.88-13.37 = 16.51              | 1  | <0.0001         | HA = \( \alpha \beta \gamma \) = 0 |
| \( \alpha \gamma = 0 \) | 11.11       | 1  | 0.0009 | 11.11-0.00 = 11.11               | 1  | 0.0009          | HA = full epistatic |
| \( \alpha \gamma = \alpha \gamma = 0 \) | 126.37      | 1  | <0.0001| 126.37-11.11 = 115.26            | 1  | <0.0001         | HA = \( \alpha \gamma \) = 0 |
| \( \alpha \beta = \beta \gamma = \alpha \gamma = 0 \) additive | 44.61       | 3  | <0.0001| 44.61-13.37 = 31.24              | 3  | <0.0001         | H = \( \alpha \beta \gamma \) = 0 |
| Environmental       | 84.27       | 6  | <0.0001| 84.27-44.61 = 39.66              | 6  | <0.0001         | HA = additive |
| Heritability*       | 84.27       | 6  | <0.0001| 84.27-44.61 = 39.66              | 6  | <0.0001         | H = environmental model |

A nested hierarchy of inheritance models was successfully fitted with at least 3 genes controlling the conidiophore architectural phenotype to the counts of progeny phenotypes from 3 crosses (Table 5), in which the number of offspring from each cross is fixed. Nine iterations were necessary to achieve the desired error tolerance of 10\(^{-8}\) with IRLS. Recommended model was bolded along with its goodness of fit to the counts of phenotypes in crosses. A null hypothesis (H0) is tested against an alternative (HA) using the chi-squared test statistics (\( \chi^2 \)) for goodness of fit with degrees of freedom (df). The models are arranged in complexity from the simplest model with no genetic effects to the most complex, the full epistatic model. In each case, a simpler model (H0) is compared to a more complex alternative model (HA) as the hierarchy is climbed in Supplementary Fig. 2. Goodness of fit was assessed by \( \chi^2 = \sum \sum (N_i - \hat{E})^2 / \hat{E} \), and the difference of such \( \chi^2 \)s, in comparing fit of 2 models.

*Ratio of additive variation to total variation.

### Table 7. ML estimates of allelic effects and epistatic effects in a 3-locus model of inheritance.

| Parameters | \( x_y = 0 \) 3 Genes |
|------------|----------------------|
| \( x \)    | 0.14 ± 0.0130        |
| \( \beta \) | 0.38 ± 0.0182        |
| \( \gamma \) | -0.28 ± 0.0085     |
| \( \alpha \beta \) | -0.24 ± 0.0175   |
| \( \beta \gamma \) | -0.55 ± 0.0175   |
| \( x_y \)  | 0.00                 |

The full epistatic model with 3 genes has 3 allelic effects and 2 epistatic interactions. The standard errors were obtained from the square roots of the diagonal elements of the inverse of the information matrix F2s.
than both WT and Wrap conidiophores. *ccg-2* is allelic with easily wettable (eas) and encodes a hydrophobin critical for maintaining cell wall hydrophobicity in the conidium (Bell-Pedersen et al. 1992). The Bulky samples also show a unique expression pattern of *ccg-1* (Lindgren 1994), where transcription is higher in the mycelium than in the conidiophore. This contrasts with both WT and Wrap, where *ccg-1* expression is higher in the conidiophore than in the mycelium. Expression of *ccg-14* is higher in both Bulky mycelia and conidiophores when compared to the corresponding cell types of WT and Wrap samples.

Another interesting expression pattern is observed with NCU09873 (*acu-6*) and NCU04482 (uncharacterized). Both loci show strikingly lower expression in Bulky mycelia compared to all other strains and cell types.

### Fig. 6. PCA of all samples characterized by RNA-seq. A 2D PCA plot was generated using normalized and variance stabilized transcript expression data (vst transformation, DESeq2) for the top 500 most variant transcripts in the dataset (measured by row variance). The percent of variance explained by each principal component is displayed on each axis. In total, 2229 samples (Hexcode: E69F00), 8872 samples (Hexcode: 68838B), and 8876 (Hexcode CD6839) cluster into 2 distinct groups on PC1 where conidiophores of 8872 and 8876 form a clear cluster away from 2229. Within the mycelium clusters, samples of 8872 and 8876 cluster together on PC2. Seventy-four percent of the total variance can be distinguished by cell type. Samples are assigned a color by strain and shape by cell type. Representative strains used were FGSC8872 for WT, FGSC8876 for Wrap, and FGSC2229 for Bulky. Four biological replicates were measured per condition.

### Fig. 7. Heatmap of normalized counts. Reads were normalized in DESeq2 by the median of ratios method (Love et al. 2014) and are depicted for the 20 most differentially expressed genes for all strains and cell types. Each column represents a biological sample and each row a gene. Representative strains used were FGSC8872 for WT, FGSC8876 for Wrap, and FGSC2229 for Bulky. There are 4 biological replicates per condition. Differential expression model was built using a 2-factor approach estimating the phenotypes effect on cell type in the design formula. The FGSC2229 for Bulky was used as the reference in the design matrix.
for phosphoenolpyruvate carboxykinase (PEPCK), and NCU04482 encodes a hypothetical protein that plays a role in amino acid metabolism (Flavell and Fincham 1968; Schmoll et al. 2012).

To further explore the genes driving separation of the Bulky strain by PCA (Fig. 6), we examined genes differentially expressed when comparing WT to Bulky and Wrap to Bulky (Fig. 8). There were genes significantly up- and downregulated in both strain comparisons. Of these significantly differentially expressed genes, only 5 loci were shared by both strain comparisons, all of which were more highly expressed in Bulky than in WT or Wrap. These shared genes are NCU07191 (doc-1), NCU07192 (doc-2), NCU09244 (plp-1), NCU09245 (plp-2), and NCU05035 (vad-12).

The doc (determinant of communication) genes mediate long-distance kind-recognition, where filaments belonging to the same communication group (CG) are more likely to interact (Heller et al. 2016). Increased expression of doc-1 and doc-2 in the Bulky strain sparked the hypothesis that perhaps there was a relationship between CG and conidiophore phenotype. Previous work has characterized the CGs of 110 wild N. crassa isolates, including the Louisiana collection employed in this work (Heller et al. 2016). While the representative strains selected for RNA-seq do belong to different CGs, this pattern is not upheld when examining the complete population collection (Fig. 9). The plp (palatin-like phospholipase) genes also contribute to N. crassa self-recognition, triggering germling-regulated death following heterokaryon incompatibility (Heller et al. 2018). The final gene more expressed in Bulky than in both WT and Wrap is vad-12, a locus involved in vegetative asexual development that is not well characterized (Carrillo et al. 2017).

To more specifically examine genes contributing to these morphological phenotypes, we compared expression patterns between strains for the conidiophore samples alone. PCA showed clear clustering between conidiophore samples from each of the 3 strains (Fig. 10).

Expression patterns for the 20 genes most differentially expressed in the conidiophore are presented in Fig. 11. Fourteen of these genes are consistent with those presented in Fig. 7, including the previously mentioned ccg-2, ccg-1, ccg-15, acu-6, and NCU04482. However, analyzing the conidiophore samples revealed 2 additional genes displaying striking differential expression patterns: NCU08769 (con-6) and NCU00265 (uncharacterized). The conidiation-specific gene con-6 is also under clock control and is downregulated in Bulky conidiophores. NCU00265 encodes a predicted secreted protein and is upregulated in Bulky conidiophores.

**Discussion**

Conidiation in N. crassa has been thoroughly investigated over the course of decades of study. While much work has been done to characterize the genetic, temporal, environmental, and circadian signals guiding conidiophore development, we lack a complete understanding of the morphological variation of these structures (Sargent and Kaltenborn 1972; Springer and Yanofsky 1989; Loros and Dunlap 2001). Recent work has begun to explore natural conidiophore variation, using a collection of wild isolates to identify 3 architectural phenotypes: WT, Wrap, and Bulky.
We continued this exploration by investigating the impact of these phenotypes on sporulation and germination, developing a more robust model to estimate heritability of the trait, and identifying genes differentially expressed in representative strains for each conidiophore shape. Previous work has demonstrated that conidiophore morphology impacts the distribution of distances traveled by sporulating conidia, or the “spore shadow” (Krach et al. 2020). Similar studies have revealed that dispersal patterns impact gene flow between natural populations of Drosophila and Laelia rubescens, for example (Powell and Dobzhansky 1976; Trapnell and Hamrick 2005). Modeled at a small scale, WT conidia displayed the lowest germination rate and traveled the least maximum distance following their sporulation (Krach et al. 2020). Here, we showed that both patterns were upheld at a larger scale, indicating that WT populations may not have as wide a colonization capacity, and thus as large of gene flow, as Wrap and Bulky (Fig. 1, Table 4). Other features of the spore, such as shape and a natural O-ring, affect spore dispersal (Roper et al. 2010; Fritz et al. 2013). Additional studies could be conducted to explore spore shadows from each
conidiophore phenotype in different environments (Roper et al. 2010), such as on various plant substrates or in different climates.

We also demonstrated that conidiophore phenotype impacted germination rate and germination time of conidia on different carbon sources (Figs. 2 and 3). Germination of conidia from WT conidiophores drastically increased by 52% when fructose and glucose concentration decreased from 1% to 0.1%, presumably a stress response that was not observed in the other 2 phenotypes. On medium containing mannose or xylose as a carbon source, conidia from Bulky conidiophores showed a striking response, consistently germinating earlier and at significantly higher rates than the other 2 phenotypes. This suggests that Bulky is much more responsive than the other 2 strains to the stress of these unfavorable carbon sources.

Implementing tools previously developed in Krach et al. (2020), we used homokaryotic strains to more accurately estimate heritability of conidiophore phenotype. Our model suggested that at least 3 genes contribute to this complex trait with 2 epistatic interactions (Table 6). The fitted model revealed an estimated heritability of 0.47, an increase from the heritability previously estimated using the original, possibly heterokaryotic, wild isolates. To further explore the genes contributing to conidiophore morphology, we conducted RNA-seq on mycelia and conidiophores from each phenotype. The RNA-seq analysis is suggesting more complex genotype–phenotype interactions than the 3 interactions detected in Table 6.

Our RNA-seq results demonstrated that the Bulky (FGSC2229) transcriptional profile is unique from that of WT and Wrap, clearly separating it from the other 2 strains by PCA (Fig. 6). Five of the 20 most differentially expressed genes (Fig. 7) were ccgs, 3 of which showed distinctive expression patterns in the Bulky cell types: ccg-2, ccg-1 (Lindgren 1994), and ccg-14. Notably, both ccg-2 and ccg-14 encode proteins that localize to the conidium cell wall. The CCG-2 protein, a hydrophobin, maintains hydrophobicity of the conidium cell wall (Bell-Pedersen et al. 1992). The protein encoded by ccg-14 (snodprot1) bears sequence similarity to ceratoblatinin, a phytotoxin prevalent in other ascomycetes that has both hydrophobin and expansin properties (Jeong et al. 2007). It is known that the cell wall plays a critical role in fungal morphology, growth rate, and viability, and several mutations affecting N. crassa cell wall components have been associated with a compact phenotype and conidiophore morphology. Among the 14 shared loci are ccg-2, ccg-1, ccg-15, acu-6, and NCU00265 (Fig. 11). Bulky conidiophores showed much lower expression of ccg-14 and higher expression of NCU00265 than WT and Wrap conidiophores. While con-6 has been well characterized as a light-responsive, clock-controlled, conidiation-specific gene, Δcon-6 shows no obvious phenotype and the function of the CON-6 protein remains unknown (White and Yanofsky 1993; Olmedo et al. 2010). Additional work should be conducted to characterize the function of this protein and explore the possible connection of con-6 downregulation and a Bulky phenotype. Further research should also be conducted on NCU00265 to better identify and characterize the predicted secreted protein it encodes. This protein has been detected alongside known cell wall proteins following their secretion by hyphae into growth medium (Maddi and Free 2010). If NCU00265 does encode a cell wall protein that is enriched in Bulky conidiophores, it provides additional evidence, alongside ccg-2 and ccg-14, that the Bulky phenotype may in part be due to an interaction in its cell wall components.

Comparisons between strain pairs revealed 5 loci consistently upregulated in Bulky, 4 of which are involved in communication: doc-1, doc-2, plp-1, and plp-2 (Fig. 8). The doc genes identify CG compatibility prior to hyphal fusion, and the plp genes trigger germing-regulated death following an incompatible fusion (Heller et al. 2016, 2018). We did not find a correlation between CG assignment and conidiophore phenotype (Fig. 9). It is reasonable to suppose that there is more genic interactions taking place between these 4 genes in the Bulky conidiophore, where there is a higher density of cells than in WT or Wrap. As an example, Conidial anastomosis Tubes in N. crassa display density-dependent behavior (Roca et al. 2005). Enrichment of these doc and plp genes could also simply be a result of this increased communication due to density effects.

Alternatively, the Bulky strain may be more heterokaryotic than the WT or Wrap strain, requiring more communication and germing-regulated death following heterokaryon incompatibility. This hypothesis is supported by the fact that this Bulky strain has lower penetrance of the phenotype (46.55%) than the WT (77.78%) and Wrap (53.85%) strains selected for RNA-seq (Krach et al., 2020). To test this hypothesis, RNA-seq could be conducted on homokaryotic F1s representing each conidiophore phenotype.

Identifying genes differentially expressed in just the conidiophore samples revealed many of the same loci differentially expressed according to both cell type and strain (Figs. 7 and 11). Among the 14 shared loci are ccg-2, ccg-1, ccg-15, acu-6, and NCU04482. This provides evidence that these genes were not solely identified by differences between cell types, but showed significant differential expression in the conidiophores and may play a role in the morphological differences observed. Analyzing differential expression patterns in the conidiophore samples did reveal 2 additional loci: con-6 and NCU00265 (Fig. 11). Bulky conidiophores showed much lower expression of con-6 and higher expression of NCU00265 than WT and Wrap conidiophores. While con-6 has been well characterized as a light-responsive, clock-controlled, conidiation-specific gene, Δcon-6 shows no obvious phenotype and the function of the CON-6 protein remains unknown (White and Yanofsky 1993; Olmedo et al. 2010). Additional work should be conducted to characterize the function of this protein and explore the possible connection of con-6 downregulation and a Bulky phenotype. Further research should also be conducted on NCU00265 to better identify and characterize the predicted secreted protein it encodes. This protein has been detected alongside known cell wall proteins following their secretion by hyphae into growth medium (Maddi and Free 2010). If NCU00265 does encode a cell wall protein that is enriched in Bulky conidiophores, it provides additional evidence, alongside ccg-2 and ccg-14, that the Bulky phenotype may in part be due to an interaction in its cell wall components.

Looking deeper into the differentially expressed transcripts revealed a more complex network of gene interactions. Gene set enrichment analysis of the top 500 differentially expressed genes revealed a significant upregulation of biological pathways known...
to be regulated by the clock (Supplementary Fig. 3A). A substantial proportion of ribosomal biogenesis is upregulated in considering the 2-factor differential expression model (Supplementary Fig. 3B). Clock outputs include genes with products heavily involved in multiple key cell processes including DNA metabolism, ribosome biogenesis in RNA metabolism, cell cycle, and protein metabolism (Dong et al. 2008; Al-Omari et al. 2018). While our most significantly differentially expressed genes were highlighted above, the enrichment analysis has revealed a deeper and more complex network of interactions with highly overlapping gene sets forming functional modules in the WT and Wrap phenotype compared to the Bulky phenotype.

High-throughput phenotyping has been recently applied in other fungal systems, such as Saccharomyces cerevisiae, to attribute genes to novel morphological phenotypes (Ohya et al. 2005). This study illustrates how high-throughput methods of phenotyping complex traits using machine learning applied to natural populations can be successfully combined with omics approaches, such as RNA-seq, to implicate the genes and their interactions underlying complex traits. The same approach is now being applied to characterize the morphology of arbuscular mycorrhizal fungi (AMF) in roots of sorghum in field populations to improve the production of biofuel because sorghum accessions grow differently in the presence of AMF (Watts-Williams et al., 2019). High-throughput phenotyping studies not only provide insights into the genetic basis of complex traits in natural and cultivated populations, but they also provide potential insights into the robustness of genetic systems to both environmental and genetic perturbations (Levy and Siegel 2008).

**Data availability**

Transcriptome data are available at NCBI SRA Accession SRP353648. Code used to analyze the RNA-seq data with the RNA-seq data is publicly available on GitHub at https://github.com/michaelSkaro/Neurospora_crassa_transcriptomics.

**Supplemental material** is available at G3 online.

**Acknowledgments**

We acknowledge Mary E. Case for contributing feedback on the sporulation and germination assays and Leidong Mao for providing access to the microscope. We thank Zarin Hossain for a GSEA analysis of the transcriptome data. We thank the Georgia Genomics and Bioinformatics Core for library preparation and sequencing. We thank the Georgia Advanced Computing Resource Center for computing.

**Funding**

This research was funded by the National Science Foundation (MCB-1713746, MCB-2041546), the US Department of Energy (DE-SC0021386), and the Achievement Rewards for College Scientists Foundation (2018-2019-ATL).

**Conflicts of interest**

None declared.

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