Plant-Necrotroph Co-transcriptome Networks Illuminate a Metabolic Battlefield

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Abstract

A central goal of studying host-pathogen interaction is to understand how host and pathogen

manipulate each other to promote their own fitness in a pathosystem. Co-transcriptomic

approaches can simultaneously analyze dual transcriptomes during infection and provide a

systematic map of the cross-kingdom communication between two species. Here we used the

Arabidopsis-*B. cinerea* pathosystem to test how plant host and fungal pathogen interact at the

transcriptomic level. We assessed the impact of genetic diversity in pathogen and host by

utilization of a collection of 96 isolates infection on Arabidopsis wild-type and two mutants with

jasmonate or salicylic acid compromised immunities. We identified ten *B. cinerea* gene co-

expression networks (GCNs) that encode known or novel virulence mechanisms. Construction of

a dual interaction network by combining four host- and ten pathogen-GCNsrevealed potential

connections between the fungal and plant GCNs. These co-transcriptome data shed lights on the

potential mechanisms underlying host-pathogen interaction.

INTRODUCTION

How a host and pathogen manipulate each other within a pathosystem to facilitate their own fitness remains a long-standing question. The difference between the pathogen's ability to infect and the host's ability to resist generates the resulting disease symptomology. This interaction forces host-pathogen dynamics to shape the genomes of the two species via adaptive responses to each other (Dangl and Jones, 2001; Bergelson et al., 2001; Benton, 2009; Kanzaki et al., 2012; Karasov et al., 2014). Plants have evolved a sophisticated set of constitutive and inducible immune responses to cope with constant selective pressures from antagonistic microbes (Jones and Dangl, 2006). Reciprocally, plant pathogens have also evolved a variety of different invasion and virulence strategies to disarm or circumvent plant defense strategies (Glazebrook, 2005; Toruno et al., 2016). This has resulted in complex relations between plant hosts and fungal pathogens for survival and fitness.

The plant innate immune system includes several functional layers with overlapping functions to detect and defend against phytopathogens. This multi-layer immune system can be categorized as a signal monitor system to detect invasion, local and systemic signal transduction components to elicit and coordinate responses, and defensive response proteins and metabolites focused on combatting the invading pathogen (Tsuda and Katagiri, 2010; Corwin and Kliebenstein, 2017). These functional layers, as well as the components within them, are highly interconnected and tightly regulated by the host plant to respond appropriately to various phytopathogens (Couto and Zipfel, 2016; Tang et al., 2017). For instance, Arabidopsis utilizes a complex signaling network to regulate the production of indole-derived secondary metabolites, such as camalexin and indole glucosinolates, that contribute to resistance against pathogens (Kliebenstein et al., 2005; Clay et al., 2009; Bednarek et al., 2009; Frerigmann et al., 2016; Xu, J. et al., 2016;

Mine et al., 2018). This layered immune system provides pathogens with numerous targets in the

plant immune system that the pathogen can utilize, evade or attack. Most biotrophic pathogens,

evolved from commensal microbes, attempt to dismantle the plant immune system by injecting

effector proteins into host cells or the inter-cellular space (Dangl and Jones, 2001; Buttner and

He, 2009; Stergiopoulos and de Wit, 2009). For example, the biotrophic bacterial pathogen

Pseudomonas syringae can utilize the jasmonic acid (JA) signaling pathway through the

production of a JA-mimic, coronatine, to enhance its fitness (Mittal and Davis, 1995; Brooks et

al., 2005; Cui et al., 2018). Alternatively, necrotrophic pathogens, which often evolved from

environmental saprophytic microbes, can utilize toxic secondary metabolites, small secreted

proteins, and small RNAs to aggressively attack host defenses while also defending against host-

derived toxins (Choquer et al., 2007; Arbelet et al., 2010; Mengiste, 2012; Weiberg et al.,

2013; Kubicek et al., 2014; Macheleidt et al., 2016). In addition, pathogens can directly resist

downstream defenses as is done by *B. cinerea*, where it has an ATP-binding cassette (ABC)

transporter BcatrB that provides resistance by exporting camalexin from the pathogen cell

(Stefanato et al., 2009). This high level of interactivity between the immune system and

pathogen virulence mechanisms generates the final level of disease severity. However, a

functional description of this combative cross-kingdom communication between a plant host and

necrotrophic pathogen remains elusive.

Co-transcriptomic approaches whereby the host and pathogen transcriptomes are simultaneously

analyzed provide the ability to systematically map the cross-kingdom communication between

plant hosts and their pathogens, both for individual genes and gene co-expression network

(GCN) levels (Stuart et al., 2003; Musungu et al., 2016; Zhang, W. et al., 2017; Lanver et al.,

2018; McClure et al., 2018). Recent advances have enabled the measurement of pathogen *in*

planta transcriptome. For example, *in planta* measurements of the pathogens' transcriptome

within the biotrophic Arabidopsis-*Pseudomonas syringae* pathosystem has enabled the

investigation of early effects on Arabidopsis host immunity and the consequent effects on

bacterial growth (Nobori et al., 2018). This enabled the identification of a bacterial iron

acquisition pathway that is suppressed by multiple plant immune pathways (Nobori et al., 2018).

This shows the potential for new hypothesis to be generated by a co-transcriptome approach

(Swierzy et al., 2017; Westermann et al., 2017; Lee et al., 2018).

The Arabidopsis-*B. cinerea* pathosystem is well suited for exploring plant-pathogen interaction

to understand host defenses and necrotrophic virulence in ecological and agricultural settings. *B.*

cinerea is a necrotrophic generalist pathogen that attacks a broad range of diverse plant hosts,

including dicots, gymnosperms, and even bryophytes (Williamson et al., 2007). This

necrotrophic pathogen is endemic throughout the world and can cause severe pre- and post-

harvest losses in many crops. A high level of standing natural genetic variation within *B. cinerea*

population is hypothesized to facilitate the extreme host range of *B. cinerea*. This genetic

variation affects nearly all known *B. cinerea* virulence strategies, including penetration and

establishment, evading detection, and combatting/coping with plant immune responses (Atwell

et al., 2015; Walker et al., 2015; Corwin, Subedy et al., 2016). For example, a key virulence

mechanism is the secretion of phytotoxic secondary metabolites, including the sesquiterpene

botrydial (BOT) and the polyketide botcinic acid (BOA) that trigger plant chlorosis and host cell

collapse (Deighton et al., 2001; Colmenares et al., 2002; Wang et al., 2009; Rossi et al., 2011;

Ascari et al., 2013; Porquier, A. et al., 2016). These metabolites are linked to virulence, but

some pathogenic field isolates fail to produce either compounds pointing to additional

pathogenic strategies. The combination of a high level of genetic diversity and extensive

recombination means that a population of *B. cinerea* is a mixed collection of virulence strategies

that can be used to interrogate by the co-transcriptome.

In the present study, we used the Arabidopsis-*B. cinerea* pathosystem to test how the

transcriptomes of the two species interact during infection and assess how natural genetic

variation in the pathogen impacts disease development. Isolates were inoculated on Arabidopsis

Col-0 wild-type (WT) in conjunction with immune-deficient hormone mutants *coi1-1* (jasmonate

defense signaling) and *npr1-1* (salicylic acid defense signaling). For the pathogen, we utilized a

collection of 96 isolates of *B. cinerea* that harbor a wide scope of natural genetic variation within

the species (Atwell et al., 2015; Corwin et al., 2016; Zhang et al., 2016; Corwin, Copeland et

al., 2016; Zhang et al., 2017; Soltis et al., 2018; Fordyce et al., 2018). From individual

infected leaves, we simultaneously measured both the Arabidopsis and *B. cinerea* transcripts at

16 hours post-infection (HPI). We have previously analyzed the Arabidopsis transcripts to

identify four host-derived GCNs that are sensitive to natural genetic variation in *B. cinerea*

(Zhang et al., 2017). In present analysis, we identified ten fungal pathogen-derived GCNs that

encode either known or novel virulence mechanisms within the species. Some of these *B. cinerea*

GCNs responsible for BOT production, exocytosis regulation and copper transport are highly

linked with the host's defense phytohormone pathways. By combining the plant host- and

pathogen-GCNs into a single network, we constructed a dual-transcriptomic network that

identifies potential interactions between the components of plant host innate immune system and

fungal pathogen virulence. These connections highlight potential targets for fungal pathogen

phytotoxins and prevailing counter-responses from plant host. Collectively, co-transcriptomic

analysis shed lights on the potential mechanisms underlying how the host and pathogen combat

each other during infection and illustrate the continued need for advancements of *in planta*

analysis of dual-species interaction.

RESULTS

Genetic Variation in Pathogen and Hosts Influence *B. cinerea* Transcriptome

To investigate how genetic variation within a pathogen differentially interacts with plant host

immunity at the transcriptomic level, we profiled the *in planta* transcriptomes of 96 *B. cinerea*

isolates infection across three host genotypes, the Arabidopsis accession Col-0 WT and two

immune-signaling mutants *coi1-1* and *npr1-1* that are respectively compromised in JA or

salicylic acid (SA) driven immunity. This previously described collection of 96 isolates

represents a broad geographical distribution and contains considerable natural genetic variation

that affects a diversity of virulence strategies within *B. cinerea* (Denby et al., 2004; Rowe and

Kliebenstein, 2007; Atwell et al., 2015; Corwin et al., 2016; Zhang et al., 2016). Four

independent biological replicates across two separate experiments per isolate/genotype pair were

harvested at 16HPI for transcriptome analysis. A total of 1,152 independent RNA samples were

generated for library preparation and sequenced on Illumina HiSeq platform (NCBI accession

number SRP149815). These libraries were previously used to study Arabidopsis transcriptional

responses to natural genetic variation in *B. cinerea* (Zhang et al., 2017). Mapping the dual-

transcriptome reads against the *B. cinerea* reference genome (B05.10), we identified 9,284

predicted gene models with a minimum of either 30 gene counts in one isolate or 300 gene

counts across 96 isolates. The total of identified genes corresponds to ~79% of the 11,701

predicted encoding genes in B05.10 reference genome (Supplemental Data Set 1 to 3) (Van Kan

153 et al., 2017). The two different thresholds allowed the identification of pathogen transcripts that

154 express only in a specific isolate.

155 Measuring the abundance of individual pathogen transcripts in relation to the host transcripts can

156 be used as a molecular method to estimate fungal biomass (Blanco-Ulate et al., 2014). Given

157 this, we hypothesized that the fraction of total reads that map to *B. cinerea* might be a

158 biologically relevant indicator of pathogen virulence (Supplemental Data Set 4). Comparing *B.*

159 *cinerea* transcript abundance at 16HPI to lesion development at 72HPI revealed a significant

160 partial correlation in the WT Col-0 ($R^2 = 0.1101$, *P*-value = 0.0016, Figure 1). In contrast to WT,

161 the early transcriptomic activities of most *B. cinerea* isolates were more vigorous in the two

162 Arabidopsis mutants, resulting in a significant curvilinear relationship between total fraction of

163 *B. cinerea* reads and final lesion area (*P*-value = 3.914e-07, *P*-value = 0.0001, respectively,

164 Figure 1). Interestingly, the total reads fraction was better correlated with final lesion area in

165 *coi1-1* ($R^2 = 0.2562$) than either WT ($R^2 = 0.1101$) or *npr1-1* ($R^2 = 0.161$). This suggests that

166 early transcriptomic activity from the pathogen can be a partial indicator of pathogen virulence,

167 but also depends on the respective resistance from the plant host.

Figure 1. Correlation between earlier estimated *B. cinerea* biomass and later lesion area. Model-corrected lesion area means were estimated using the linear model on the six replicates data from three Arabidopsis genotypes at 72-hours post-infection with 96 *B. cinerea* isolates. Estimated biomass of *B. cinerea* was calculated using the linear model-corrected fraction of *B. cinerea* mapped reads against total mapped reads to Arabidopsis and *B. cinerea* reference genomes. RNA-Seq analysis was conducted at 16-hours post-infection for each pathosystem. Three Arabidopsis genotypes are wild-type Col-0 (purple dot), jasmonate insensitive mutant *coi1-1* (green triangle), and salicylic acid insensitive mutant *npr1-1* (orange diamond). The 90% confidence ellipse intervals are plotted for each Arabidopsis genotype for references. Quadratic regression lines are: Col-0: y = -0.00059x² + 0.729x + 10.037, *P* = 0.0016, adjusted R² = 0.1101; *coi1-1*: y = -0.117x² + 4.44x – 0.1585, *P* = 3.914e-07, adjusted R² = 0.2562; *npr1-1*: y = -0.0579x² + 2.26x + 1.673, *P* = 0.0001, adjusted R² = 0.161.

Plant defense phytohormone networks, like SA and JA, help shape the immune responses of a

plant host while also shape the virulence gene expression within bacterial pathogens, such as

Pseudomonas syringae (Nobori et al., 2018). To test how variation in host SA/JA-signaling

influences the fungal pathogen transcriptome, we applied a generalized linear model linked with

negative-binomial function (nbGLM) to each *B. cinerea* transcript across the experiment. This

analysis allowed us to estimate the relative broad-sense heritability (H^2) of genetic variation from

the pathogen, plant host, or their interaction contributing to each transcript. Of the 9,284

detectable *B. cinerea* transcripts, 8,603 and 5,244 transcripts were significantly influenced by

genetic variation in pathogen and host, respectively (74% and 45% of predicted *B. cinerea* gene

models, respectively) (Figure 2A, Supplemental Data Set 3 and 5). While this result shows that

the plant phytohormone pathways influence *B. cinerea* gene expression, the variation in host

defense responses (average $H^2_{Host} = 0.010$) has far less influence on *B. cinerea* gene expression

than that of the pathogens' own natural genetic variation (average $H^2_{Isolate} = 0.152$). The host

defense hormones also affected *B. cinerea* gene expression in a genotype-by-genotype dependent

manner on 4,541 genes (39% of *B. cinerea* predicted gene models, average $H^2_{Isolate \times Host} = 0.116$)

(Figure 2B-2I). Illustrating this potential for host x pathogen interactions on pathogen gene

expression are the two genes encoding the well-studied polygalacturonase 1 (*Bcpg1*) and

oxaloacetate acetyl hydrolase. *Bcpg1* has dramatic expression variation across 96 isolates in

different host backgrounds (Figure 3, Figure 3-figure supplemental 1, and Supplemental Data Set

1). Extending this to 500 genes showing the strongest host x pathogen effect showed that there is

a wide range of patterns that differs in the host *coi1-1* or *npr1-1* background with diverse

pathogen strain specific patterns (Figure 4). One potential complication of this analysis is for

sequence variation between the reference B05.10 genome and the diverse strains to create

artificially low expression estimates. However, very few genes showed consistently low

expression within a strain and instead when a gene showed no expression in one host genotype, it

was expressed in a different host genotype (Figure 3 and supplemental Table 1). This

conditionality argues against a sequencing error as the sequence has not altered. The genes that

did show a loss of expression across all host genotypes within a strain (i.e. BOT and BOA genes)

were frequently linked to whole gene deletions that abolished their expression (Soltis et al.,

2019). Thus, while there are likely some sequence variation associated expression errors, they are

not a dominant signature in the data. Thus, within the Arabidopsis/*B. cinerea* pathosystem the

Figure 2. Transcriptomic responses of *B. cinerea* on Arabidopsis are controlled by genetic variation in pathogen population, host genotypes, and their interaction. (A) Distribution of broad-sense heritability (*H²*) of *B. cinerea* transcripts contributed by genetic variation in the *B. cinerea*, Arabidopsis genotypes, and the interaction between pathogen and host. Violin plots illustrating the distribution of *H²* for transcripts from 96 *B. cinerea* isolates infecting on Arabidopsis genotypes. Heritability is partitioned across the different sources, 96 pathogen genotypes = "Isolate", plant genotypes Col-0, *coi1-1* and *npr1-1* plant genotypes = "Host", and the corresponding interaction. The transcriptomic analysis was conducted by sequencing mRNA extracted from *B. cinerea* infected Arabidopsis leaves at 16-hours post-infection. Red lines indicate the average broad-sense heritability values of lesion area caused by isolates, Arabidopsis genotypes, and their interaction. (B) to (E) Expression profiles of *B. cinerea* transcripts significantly influenced by host genotypes. The model-corrected means (log₂) for *B. cinerea* transcript were used for plotting. The Arabidopsis genotypes, wild-type Col-0 (purple), jasmonate insensitive mutant *coi1-1* (green), and salicylic acid mutant *npr1-1* (orange), are shown on the x axis. *B. cinerea* transcripts are: (B) *Bcin01g03790.1*, class IV chitin synthase; (C) *Bcin02g02980.1*, Monooxygenase; (D*) Bcin10g02260.1*, NADH oxidoreductase; (E) *Bcin11g05200.1*, caffeine-induced death Cid2; (F) to (I) Expression profiles of *B. cinerea* transcripts significantly influenced by the interaction between pathogen and host genotypes. (F) *Bcin02g03810.1*, GTP cyclohydrolase; (G) *Bcin09g01190.1*, Tripeptidyl-peptidase 1; (H) *Bcin10g01130.1*, in secretory pathway; (I) *Bcin14g05090.1*, a transcription factor.

Figure 3. Expression profiles of an endopolygalacturonase gene *Bcpg1* from diverse *B. cinerea* isolates across Arabidopsis genotypes. Rank plot shows the relationship of *Bcpg1* expression from 32 diverse *B. cinerea* isolates (right) across three Arabidopsis genotypes (x axis). Three Arabidopsis genotypes are wild-type Col-0 (purple dot), jasmonate insensitive mutant *coi1-1* (green triangle), and salicylic acid mutant *npr1-1* (orange diamond). The model-corrected means (log₂) for the transcript of *Bcpg1* (*Bcin14g00850.1*) encoding an endopolygalacturonase gene are utilized for plotting. The transcript expression levels from the same isolate across three Arabidopsis genotypes are connected with a colored line. The names of 32 isolates are represented with the same colored lines as induced *Bcpg1* expression levels. Black lines indicate the expression levels of *Bcpg1* are higher in *coi1-1* and *npr1-1* than in Col-0. Red lines indicate the higher expression levels of *Bcpg1* in *coi1-1* but lower in *npr1-1*. Blue lines indicate the highest expression levels of *Bcpg1* are in Col-0. Dark green lines indicate the higher expression levels of *Bcpg1* in *npr1-1* but lower in *coi1-1*.

Figure 4. Interaction effects of host genotypes and pathogen isolates on *B. cinerea* transcriptome. Hierarchical clustering of relative expression of 500 genes from 96 *B. cinerea* isolates infection on Arabidopsis mutants *coi1-1* (A) or *npr1-1* (B) are plotted based on pairwise comparison of pathogen gene expression under Col-0. The 500 *B. cinerea* genes with highest broad-sense heritability (H^2) of host X pathogen were used for analysis. Lesion area induced by 96 isolates are compared under *coi1-1* (orange bar plot) and *npr1-1* (green bar plot).

pathogens transcriptional responses are influenced by a blend of the pathogens' natural variation

and its interaction with the host, while there is less evidence for the host's defense responses to

unilaterally affect *B. cinerea*. Future work will hopefully assess how this extends to other host-

pathogen systems.

Identification of Virulence Factors Among the Early *B. cinerea* Transcripts

This data set also allows us to test for specific *B. cinerea* transcripts whose early expression is

associated with later lesion development. These genes can serve as potential biomarkers of

overall pathogen virulence and may elucidate the functional mechanisms driving early virulence

in the interaction. To find individual pathogen transcripts link with lesion development, we

conducted a genome-wide false discovery rate-corrected Spearman's rank correlation analysis

between 72HPI lesion area and individual *B. cinerea* transcripts accumulation at 16HPI. We

identified 2,521 genes (22% of *B. cinerea* predicted gene models) with significant positive

correlations and 114 genes (1% of *B. cinerea* predicted gene models) with significant negative

correlations to lesion area across three Arabidopsis genotypes, respectively (*P*-value < 0.01,

Supplemental Data Set 6). The top 20 positively correlated *B. cinerea* genes contained all seven

genes involved in BOT biosynthesis (Deighton et al., 2001; Colmenares et al., 2002; Wang et

al., 2009; Rossi et al., 2011; Ascari et al., 2013; Porquier et al., 2016). In addition to

phytotoxins, more than 30 genes of the top 100 lesion-correlated genes encode plant cell wall

degrading enzymes, i.e., glucosyl hydrolases, carbohydrate esterases, cellobiose dehydrogenases

and *Bcpg1* (Figure 3 and Supplemental Data Set 6) (Gerbi et al., 1996; Zamocky et al., 2006;

Cantarel et al., 2009; Van Vu et al., 2012; Igarashi et al., 2014; Ingo Morgenstern et al., 2014;

Blanco-Ulate et al., 2014; Tien-chye Tan et al., 2015; Courtade et al., 2016; Nelson et al.,

2017; Pérez-Izquierdo et al., 2017). Additionally, 10 of the top 100 lesion-correlated genes were

annotated as putative peptidase activities, which are critical for fungal virulence (Movahedi et

al., 1991; Poussereau, Creton et al., 2001; Poussereau, Gente et al., 2001; ten Have et al.,

2004; ten Have et al., 2010). A final classical virulence gene in the top 100 gene list is *Bcoah*

(*Bcin12g01020*) encoding oxaloacetate acetyl hydrolase, which is a key enzyme in oxalic acid

biosynthesis that positively contributes to virulence (Figure 3-figure supplemental 1 and

Supplemental Data Set 6) (Greenberg et al., 1994; Williamson et al., 2007; Walz et al., 2008;

Schumacher et al., 2012; Schumacher et al., 2015; Tayal et al., 2017). In addition, this method

identified 37 of the top 100 lesion-correlated genes with no gene ontology (GO) terms, which

likely represent unknown virulence mechanisms (Supplemental Data Set 6). Thus, this approach

readily creates new hypothesis about known and novel pathogen virulence functions.

In Planta* Virulence Gene Co-expression Networks (GCNs) in *B. cinerea

To develop a systemic view of fungal pathogen *in planta* gene expression, we used a co-

expression approach to identify *B. cinerea* networks that associated with growth and virulence *in*

planta. Using solely *B. cinerea* transcriptome at 16HPI from Arabidopsis Col-0 WT infected

leaves, we calculated Spearman's rank correlations of gene counts across all *B. cinerea* isolates,

filtered gene pairs with correlation greater than 0.8. We then used the filtered gene pairs as input

to construct GCNs. We identified ten distinct GCNs containing more than five *B. cinerea* genes

(Figure 5, Supplemental Table 1, Figure 5-figure supplemental 1 and Supplemental Data Set 7).

The largest GCN with 242 genes contains members responsible for phospholipid synthesis,

eiosome function, and membrane-associated stress signaling pathways (Figure 5-

Vesicle/virulence). The biological function of this GCN suggests its role in fungal membrane-

and vesicle-localized processes, which are normally involved with general hyphae growth,

fungal cell wall deposition, and exudation of fungal toxins to the intercellular space

Figure 5. Gene co-expression networks identified from *B. cinerea* transcriptomic responses to Arabidopsis wild-type Col-0 immunity. Ten gene co-expression networks (GCNs) with more than five nodes were identified from 96 *B. cinerea* isolates infecting on Arabidopsis wild-type Col-0. The similarity matrix is computed using Spearman's rank correlation coefficient. Nodes with different colors represent *B. cinerea* genes condensed in GCNs with different biological functions. Edges represent the Spearman's rank correlation coefficients between gene pairs. Trans- and cis-GCNs means GCNs are regulated by trans- and cis-regulatory elements, respectively. GCNs were named after their biological functions, which were determined by hub and bottleneck genes within each network. GCNs are: vesicle/virulence (red), translation/growth (green), exocytosis regulation (pink), cyclic peptide (yellow), peptidase (gray), isopentenyl pyrophosphate (IPP, turquoise), polyketide (violet), botcinic acid (BOA, blue), copper transport (slate blue), botrydial (BOT, purple).

(Supplemental Data Set 7). The second largest network contains 128 genes that were entirely

associated with translation and protein synthesis (Figure 5-Translation/growth and Supplemental

Data Set 7). Of the smaller GCNs identified (5-20 genes), five networks were identified with

genes distributed across *B. cinerea* 16 chromosomes, suggesting that these GCNs arise from

coordinated trans-regulation (Figure 5-Trans-networks and Figure 5-figure supplemental 1D, 1F,

1H-1G). These networks are associated with diverse array of virulence functions, including the

regulation of exocytosis, copper transport, the production of peptidases and isoprenoid

precursors (IPP), and polyketide secretion.

In contrast to the whole-genome distributed GCNs, three of the smaller GCNs were

predominantly comprised of genes tandemly clustered within a single chromosome with no or a

few genes on other chromosomes (Figure 5-BOA, -Cyclic Peptide, -BOT, Figure 5-figure

supplemental 1C, 1E, and 1G). A functional analysis showed that all of the genes within these

networks encoded known or putative biosynthetic enzymes for specialized metabolic pathways.

For example, seven genes responsible for BOT biosynthesis cluster on chromosome 12 and form

a small GCN with a Zn(II)2Cys6 transcription factor that is specific to the pathway (Figure 6A,

6B, Figure 5-figure supplemental 1G and Supplemental Data Set 7) (Siewers et al., 2005;

Pinedo et al., 2008; Urlacher and Girhard, 2012; Moraga et al., 2016). Similarly, all 13 genes

involved in BOA biosynthesis cluster in Chromosome 1 and form a highly connected GCN

(Figure 5-BOA, Figure 6E, Figure 5-figure supplemental 1C and Supplemental Data Set 7)

(Dalmais et al., 2011; Porquier, Antoine et al., 2019). In addition to previously characterized

secondary metabolic pathways, we identified an uncharacterized set of ten genes that cluster on

Chromosome 1 (Figure 5-Cyclic Peptide, Figure 6F, Figure 5-figure supplemental 1E and

Supplemental Data Set 7). These genes share considerable homology with enzymes related to

Figure 6. Variation of transcripts accumulation for secondary metabolites production across diverse *B. cinerea* isolates. Expression profiles of genes responsible for botrydial, botcinic acid, cyclic peptide production across 96 isolates under Arabidopsis wild-type Col-0 are shown. **(A)** Schematic shows the genomic locus of seven botrydial (BOT) biosynthesis genes clustered together. Exons are represented by gray boxes. Introns and intergenic regions are represented by the grey line. Seven BOT genes are: *BcBOT1, BcBOT3* and *BcBOT4,* encoding a cytochrome P450 monooxygenase, respectively; *BcBOT2* encoding a sesquiterpene cyclase; *BcBOT5* encoding an acetyl transferase; *BcBOT6* encoding a Zn(II)2Cys6 transcription factor, *BcBOT7* encoding a dehydrogenase reductase. (B) BOT biosynthesis pathway in *B. cinerea.* (C) Bar plots compare expression variation of *BcBOT*2 across 96 *B. cinerea* isolates in responding to Arabidopsis wild-type Col-0 immunity. The model-corrected means ($\log_2$) of transcripts were used for plotting. (D) Scatter plot illustrates the positive correlations between lesion area and accumulation of *BcBOT2* transcript across the 96 isolates in response to varied Arabidopsis immunities. Model-corrected lesion area means were estimated for three Arabidopsis genotypes at 72-hours post-infection with 96 *B. cinerea* isolates. The three Arabidopsis genotypes are labeled next to the confidence ellipse curves: wild-type Col-0 (purple dot), jasmonate insensitive mutant *coi1-1* (green triangle), and salicylic acid mutant *npr1-1* (orange diamond). The 90% confidence ellipse intervals are plotted for each Arabidopsis genotype for reference. Linear regression lines: Col-0: y = 3.2532x + 4.4323, *P* = 1.008e-10, Adjusted R^2 = 3.3537; *coi1-1*: y = 7.4802x + 10.3289, *P* = 7.895e-15, adjusted R^2 = 0.4700; *npr1-1*: y = 3.7086x + 7.3487, *P* = 2.425e-11, adjusted R^2 = 0.3726. (E) and (F) Bar plots compare expression variation of *BcBOA6* in botcinic acid (BOA) pathway and *Bcin01g11460.* in cyclic peptide pathway across 96 *B. cinerea* isolates in response to Arabidopsis wild-type Col-0 immunity. (G) Venn diagram illustrates the number of *B. cinerea* isolates with the ability to induce BOT, BOA, and cyclic peptide.

cyclic peptide biosynthesis and may represent a novel secondary metabolic pathway in *B.*

cinerea (Supplemental Data Set 7). The expression of these pathways *in planta* was extremely

variable among the isolates and included some apparent natural knockouts in the expression of

the entire biosynthetic pathway (Figure 6G and Supplemental Data Set 1). Isolate 94.4 was the

sole genotype lacking the entire BOT pathway, while 19 isolates and 24 isolates did not

transcribe respectively the BOA and the putative cyclic peptide pathways (Figure 6E to 6G and

Supplemental Data Set 1). We decomposed the expression of these pathways into expression

vectors, referred to as eigengenes, using a principle component analysis and used a linear mixed

model to test for a relationship between early expression of secondary metabolic pathways and

later lesion area. This showed a significant relationship between the expression of BOT and

BOA pathways and lesion area measured at 72HPI (Supplemental Table 2). In contrast, the

putative cyclic peptide pathway was only associated with lesion development in a BOT-

dependent manner, suggesting that it may have a synergism to BOT (Supplemental Table 2).

Thus, *in planta* analysis of the fungal transcriptome can identify known and novel potential

virulence mechanisms and associate them with the resulting virulence.

Covariation of Fungal Virulence Networks Under Differing Plant Immune Responses

The *B. cinerea* GCNs measured within Arabidopsis WT provide a reference to investigate how

phytohormone-signaling in host innate immunity may shape the pathogen's transcriptional

responses during infection. Comparing the *B. cinerea* GCN membership and structure across the

three Arabidopsis genotypes (WT, *coi1-1*, and *npr1-1*) showed that the core membership within

networks was largely maintained but the specific linkages within and between GCNs were often

variable (Figure 7, Supplemental Table 1, Figure 7-figure supplemental 1, 2, 3, and

Supplemental Data Set 7). For example, the two largest *B. cinerea* GCNs in WT developed

multiple co-expression connections during infection in the JA-compromised *coi1-1* host (Figure

7 and Supplemental Table 1). In contrast, some GCNs have a highly robust structure across three

host genotypes, including three GCNs associated with BOT, BOA and cyclic peptide production,

and GCNs associated with exocytosis regulation, copper transport, and peptidase activity

(Supplemental Table 1, Figure 7-figure supplemental 1, 2, 3, and Supplemental Data Set 7). In

addition, we also identified additional small GCNs that demonstrated host specificity in *coi1-1*

(Supplemental Figure 4 and Supplemental Data Set 7). In particular, there were four small GCNs

that are associated with plant cell wall degradation, siderophores, glycolysis, ROS, and S-

adenosylmethionine biosynthesis (Figure 7-figure supplemental 3). Thus, the coordinated

transcriptional responses of *B. cinerea* GCNs are at least partially dependent on variation in the

host immune response.

Host immunities showed different impacts on expression profiles of genes condensed in

individual *B. cinerea* GCNs (Figure 7-figure supplemental 4). Compared with WT, expression

profiles of genes within the largest membrane/vesicle virulence GCN were elevated in the SA-

and JA-compromised Arabidopsis mutants on average (Figure 7-figure supplemental 4A).

Fungal genes associated with copper transport and polyketide production were upregulated under

SA-compromised host immunity (Figure 7-figure supplemental 4F and 4J). Whereas, members

of GCNs responsible for plant cell wall degradation and siderophore biosynthesis were

upregulated under JA-compromised host immunity (Figure 7-figure supplemental 4K and 4L).

Finally, GCNs associated with BOT and exocytosis regulation showed robust gene expression

profiles across all three Arabidopsis genotypes (Figure 7-figure supplemental 4D and 4G). The

above observation indicates host immunity influences the *B. cinerea* transcriptional response of

B. cinerea and suggests that *B. cinerea* isolates have varied abilities to tailor virulence strategy in

response to host immunity.

Figure 7. Comparison of plasticity of *B. cinerea* gene co-expression network under vaired host immunity. *B. cinerea* gene co-expression networks (GCNs) of vesicle/virulence (red) and translation/growth (green) identified under three Arabidopsis genotypes are compared. Three Arabidopsis genotypes are wild-type Col-0, jasmonate insensitive mutant *coi1-1*, and salicylic acid mutant *npr1-1*. Nodes marked with red and green colors represent *B. cinerea* genes condensed in GCNs with different biological functions. The same node condensed in GCNs across three Arabidopsis genotypes was marked with same color. Nodes specificaly condensed in GCNs under two mutants *coi1-1* and *npr1-1* background are marked with orange color. Edges represent the Spearman's rank correlation coefficients between gene pairs.

Cross-kingdom Co-transcriptomic Networks Revealed Direct Gene-for-gene Interaction

To test the interaction between individual genes from two organisms, we generated Arabidopsis-*B. cinerea* GCNs using co-transcriptome data under each host genotype. We calculated Spearman's rank correlation coefficients among 23,898 Arabidopsis transcripts and 9,284 *B. cinerea* transcripts. This approach identified three cross-kingdom GCNs (CKGCNs) under Arabidopsis WT and JA- or SA-compromised two mutants (Figure 8, Supplemental Table 5, and Supplemental Data Set 8). Under Arabidopsis WT, a total of 54 hub genes were identified, half from *B. cinerea* and half from Arabidopsis. Furthermore, CKGCNs contain a majority of genes in the BOT GCN and a small proportion of genes in the vesicle/virulence GCN (Figure 8-figure supplemental 1C, 1G, 1F and 1H). For plants, CKGCNs contain a majority of genes from Arabidopsis Defense/camalexin GCN (Figure 8-figure supplemental 1B). These CKGCNs also contain genes associated with extensive host defense responses, i.e., genes encoding membrane-localized leucine-rich repeat receptor kinases (LRR-RKs), stress signal sensing and transduction, tryptophan-derived phytoalexin production, regulation of cell death, cell wall integrity, nutrition transporters, etc. (Supplemental Data Set 8). The topological structure and gene content of the CKGCNs shifted across the three Arabidopsis genotypes (Figure 8). These changes illustrate how the host genotype can influence the intercommunication in the host-pathogen interaction.

A   Col-0

B   *coi1-1*

C   *npr1-1*

A Dual Interaction Network Reveals Fungal Virulence Components Targeting Host

Immunity

To begin assessing how two species influence each other's gene expression during infection, we

constructed a co-transcriptome network using both the host- and pathogen-derived GCNs (Figure

9 and Figure 9-figure supplemental 1). We converted the ten *B. cinerea* GCNs and the four

Arabidopsis GCNs into eigengene vectors that capture the variation of the general expression of

all genes within a GCN into a single value (Zhang et al., 2017). The Arabidopsis GCNs were

defined in response to this same transcriptome but by using solely the host transcripts. Of these

four Arabidopsis GCNs, one is largely comprised of genes in Defense/Camalexin signaling, two

are linked to different aspects of photosynthesis and the fourth is largely comprised of host genes

in cell division. We calculated Spearman's rank coefficients among each GCN eigengene pairs

without regard for the species. In this dual transcriptome network, the Arabidopsis/*B. cinerea*

GCN eigengenes are displayed as nodes and positive/negative correlations between the GCNs as

edges (Figure 9 and Figure 9-figure supplemental 1). Of the host-derived GCNs, the Arabidopsis

Defense/camalexin and Photosystem I (PSI) GCNs have a higher degree of centrality within the

dual interaction network, suggesting that they have the most interactions with *B. cinerea* GCNs.

The fungal GCNs highest degrees were associated with the exocytosis regulation, BOT, and IPP,

which further illustrating the importance of specialized metabolism exudation in pathogen

virulence. Interestingly, fungal GCNs (copper transport, exocytosis regulation, BOT and IPP

biosynthesis) that were positively correlated with the host Defense/camalexin GCN showed

negative correlations with PSI eigengene. The opposite relationships between pathogen- and

host-GCNs suggest that these fungal GCNs may target plant PSI while the plant counters this

attack using the Defense/camalexin GCN.

To test if these connections were dependent upon the host immunity, we used the eigengene

values derived from fungal GCNs to conduct mixed linear modelling of how they were linked to

variation in the host genotype and/or host GCNs (Supplemental Table 3 and 4). Some *B. cinerea*

GCNs (Vesicle/virulence and TSL/growth, etc.) were affected by variation in the host genotypes

but others showed host-independent manner (BOT, copper transport, etc.). Critically, the

connections between pathogen- and host-GCNs were largely independent of the host genotypes,

suggesting that the dual interaction network is a relatively robust structure (Supplemental Table

3 and 4). Collectively, pathogen virulence and host immunity GCNs showed complex

connections within dual interaction network identified from co-transcriptome data, suggesting

functional relationships between host defense and pathogen virulence mechanisms for future

experimentation.

Figure 9. A dual interaction network reveals links between Arabidopsis immunity and *B. cinerea* virulence. A dual interaction network was constructed using gene co-expression networks (GCNs) from Arabidopsis and *B. cinerea* co-transcriptome. The first eigenvectors were derived from individual GCNs and used as input to calculate Spearman's rank correlation coefficiency between GCN pairs. Green dots and orange triangles represent Arabidopsis immune- and *B. cinerea* virulence-GCNs, respectively. Blue and red lines (edges) represent the positive and negative Spearman's rank correlation coefficients between GCN pairs, respectively. The thickness of line signifies the correlational strength.

Germinations Influence on the Co-Transcriptome

One potential complicating factor that may influence the co-transcriptome is variation in germination of the spores between *B. cinerea* strains. The lack of universal genomic patterns for host x pathogen interactions in the co-transcriptome argues that germination is not causing global effects on the co-transcriptome (Figures 3 and 4). To begin examining how variation in *B. cinerea* spore germination may influence the co-transcriptome and our identified link to

virulence, we investigated the germination of 19 isolates. This showed that there was some

variation in germination with all but a few isolates germinating within the 6-7 hours' time frame

at room temperature (Figure 9-figure supplemental 2). To extend this to an *in planta* analysis, we

utilized an existing microarray study on *B. cinerea* germination to develop an eigengene that

estimates the relative germination between the strains using the *in planta* transcriptomic data

(Leroch et al., 2013). We then used this *in planta* estimation of germination to test if our

previously identified co-transcriptome to virulence links were altered by controlling for

germination. Using linear models, we ran the same test whereby the major *B. cinerea* GCNs

were tested for a link to virulence although this time, we included the germination eigengene as a

co-variate. This analysis showed that the *in planta* estimation of germination significantly

associated with virulence. Critically, even with germination taken into account, all the *B. cinerea*

networks remained significantly associated to lesion area. Some GCNs link, i.e. BOT and BOA,

were largely unaffected by the germination estimates (Supplemental Table 6), showing that some

aspects of virulence are independent of spore germination. In contrast other GCNs like the

vesicle linked GCN had their link to virulence decreased but not abolished by including the

germination co-variate. Thus, while spore germination plays a role in our measurement of the

plant-pathogen interaction, it is only one of multiple factors influencing the co-transcriptome and

is not imparting a dominant global influence on the observed patterns.

DISCUSSION

In recent decades, improvements in the understanding of the molecular basis of plant-pathogen

dynamics have facilitated breeding strategies for disease resistance in a variety of crop species.

However, breeding for disease resistance remains difficult for crops susceptible to pathogens that

harbor diverse polygenic virulence strategies targeting multiple layers and components of the

plant innate immune system (Corwin and Kliebenstein, 2017). In this study, we investigated the

transcriptome profiles at an early infection stage both within *B. cinerea* and across the interaction

within the Arabidopsis-*B. cinerea* pathosystem. This showed that the transcriptional virulence

strategy employed by *B. cinerea* is dependent both on fungal genotype and the functional

response of the host plant's immune system. We identified a batch of early *B. cinerea* transcripts

associated with later lesion development and several pathogen GCNs responsible for mediating

virulence in *B. cinerea*, including a potential specialized metabolic pathway of cyclic peptide

virulence factor. Lastly, we constructed a co-transcriptome network that revealed known and

novel fungal virulence components coordinated expressed with plant host GCNs during

infection.

There are some potential limitations on the utility of cross-species GCNs. Predominantly, they

are a correlational approach where links are made between host and pathogen transcriptional

changes. While this leads to the development of new hypothesis, it will equally require future

validation efforts to assess if these are direct or indirect relationships. Additionally, the cross-

species GCN approach as implemented in this work does not distinguish between host/pathogen

cells that are directly interacting versus those that are having long-distance responses. An

important future avenue will be to integrate cell-specific RNA sequencing approaches to better

delineate what are the responses within host/pathogen cells that are directly interacting versus the

long-distance responses. This would greatly increase the power of elucidating direct versus

indirect effects in this system.

426   **Secondary Metabolites May Mediate Plant and Fungus Transcriptomic Interactions**

427   **During Infection**

428   Necrotrophic pathogen *B. cinerea* has evolved an arsenal of virulence strategies to establish

429   colonization and enhance infection within the plant host, including production of secondary

430   metabolites. The co-transcriptome approach shows that the expression of fungal specialized

431   pathways early in infection correlates with later lesion development (Supplemental Table 3).

432   Three secondary metabolite GCNs are clustered within the fungal genome and two of them

433   identified with pathway-specific transcription factors (Figure 6, Figure 7-figure supplemental 2,

434   3, and Supplemental Data Set 7). Further, the expression of these pathways displayed a large

435   range of phenotypic variation across the isolates (Figure 6G and Supplemental Data Set 1).

436   However, the topology and memberships of GCNs for the three pathways are largely insensitive

437   to variation in host immunity. Robustness to host immunity suggests that these GCNs are

438   somehow insulated from the host's immune response, possibly to protect toxin production from a

439   host counter-attack. The co-transcriptome approach showed the ability to identify known and

440   novel secondary metabolic pathways that mediate plant host and fungal pathogen interaction.

441   Importantly, the dual interaction networks provide hypothesis for pathogen-GCNs responsible

442   for fungal secondary metabolites production link to specific plant host-GCNs (Figure 9 and

443   Figure 9-figure supplemental 1). Specifically, the co-transcriptome approach revealed that *B.*

444   *cinerea* GCNs responsible for secondary metabolite production are positively associated with

445   plant immune responses and negatively associated with primary plant metabolism (Figure 9,

446   Supplemental Table 3 and 4). For example, the BOT GCN shows a strong positive correlation

447   with the Arabidopsis defense/camalexin GCN, suggesting that BOT production may directly

induce the host's defense system. Concurrently, the BOT GCN is negatively linked to the plant's

PSI GCN, suggesting that BOT may repress the plant's photosynthetic potential. Further work is

needed to test the direction of causality within the dual interaction network as it is possible that

the host's defense/camalexin network is responding to the PSI alteration, or vice versa.

Collectively, these results strongly implicate the ability of secondary metabolites biosynthesis to

mediate the interactions between pathogen virulence and plant host immunity at the

transcriptomic level. The co-transcriptome approach showed the potential to enable us to form

new hypotheses about how this linkage may occur.

Fungal Virulence Components Correlated with Plant Immune Response

In addition to secondary metabolite biosynthesis, the co-transcriptome identified a number of key

virulence mechanisms that could be mapped to the two species interaction. One key GCN is

enriched for genes involved in exocytosis associated regulation (Figure 5-Exocytosis regulation

and Supplemental Data Set 7). The exocytosis complex is responsible for delivery of secondary

metabolites and proteins to the extra-cellular space and plasma membrane in fungi (Colombo et

al., 2014; Rodrigues et al., 2015). Additionally, we found many *B. cinerea* genes associated

with secretory vesicles within the membrane/vesicle virulence GCN that likely serve a similar

function during infection (Figure 5-Vesicle/virulence and Supplemental Data Set 7). These

GCNs also provide support for the role of exocytosis-based spatial segregation of different

materials during fungal hyphae growth *in planta* (Samuel et al., 2015). The dual interaction

network suggests that the exocytosis regulation and membrane/vesicle virulence GCNs are

differentially linked to the Arabidopsis defense/camalexin GCN, indicating varied connections

between fungal secretory pathways and plant immune responses (Figure 9 and Supplemental

Table 3 and 4). Another conserved GCN in the *B. cinerea* species is associated with copper

uptake and transport (Figure 5-Copper transport, Figure 7-figure supplemental 1, 2, 3, and

Supplemental Data Set 7). Although copper is essential for *B. cinerea* penetration and redox

status regulation within plant tissues, further work is required to decipher the precise molecular

mechanism involved in acquisition and detoxification of copper. Thus, the co-transcriptome

approach can identify both known and unknown mechanisms and links within the host-pathogen

interaction.

Fungal Virulence Transcriptomic Responses Are Partly Shaped by Host Immunity

It is largely unknown how plant host immunity contributes to the transcriptomic behavior of the

fungus during infection. Even less is known about the role of genetic variation in the pathogen in

responding to, or coping with, the inputs coming from the host immune system. In the current

study, we found that the host immune system's effect on pathogen transcripts and GCNs was

largely via an interaction with the pathogens genotype (Figure 2, Figure 7, Figure 7-figure

supplemental 4, and Supplemental Data 5). For example, fungal GCNs associated with

membrane/vesicle virulence and fungal growth shifted drastically between the WT and *coi1-1* or

npr1-1 Arabidopsis genotypes (Figure 7). In addition, some GCNs only appeared in specific

backgrounds. For example, those linked to siderophores and a polyketide production were only

identified during infection of the JA-compromised Arabidopsis mutant (Figure 7-figure

supplemental 4J and 4L). However, other fungal GCNs, like those involved in secondary

metabolism, were largely insensitive to variation in the host immunity (Figure 7-figure

supplemental 1, 2, 3, Supplemental Table 1, and Supplemental Data Set 7). Critically, the gene

membership of these GCNs is largely stable across the collection of pathogen isolates, even

while their expression level across the *B. cinerea* isolates is highly polymorphic (Supplemental

Data Set 7 and Figure 7-figure supplemental 4). This suggests that natural variation in the host

immunity and pathogen shapes how the co-transcriptome responds to host's immune system.

Further, the natural variation in the pathogen may be focused around these functional GCNs.

Plant Disease Development Can Be Predicted by Early Transcriptome Data

Plant disease development is an abstract phenomenon that is the result of a wide set of spatio-

temporal biological processes encoded by two interplaying species under a specific environment.

In current study, we used late stage lesion area as a quantitative indicator of *B. cinerea* virulence.

We have previously shown that early Arabidopsis transcriptomic response could be linked to

later lesion development (Zhang et al., 2017). Here, our findings suggest that the late-stage

disease development of a *B. cinerea* infection is determined during the first few hours of

infection by the interaction of plant immune and fungal virulence responses. It was possible to

create a link between early transcripts' accumulation and late disease development using solely

the *B. cinerea* transcriptome (Figure 1 and Supplemental Data Set 6). This could be done using

either individual pathogen genes, GCNs, or more simply the total fraction of transcripts from the

pathogen. As the transcriptomic data were from plant leaf tissue only 16HPI, there is not a

significant amount of pathogen biomass and this is more likely an indicator of transcriptional

activity in the pathogen during infection. It is possible to develop these methods as possible

biomarkers for likely fungal pathogen caused disease progression.

CONCLUSION

By analyzing an early stage co-transcriptome of a *B. cinerea* population infection on

Arabidopsis, we identified a number of *B. cinerea* GCNs that contained a variety of virulence-

associated gene modules with different biological functions. The characterization of these GCNs

simultaneously identified mechanisms known to enhance *B. cinerea* virulence and implicated

several novel mechanisms not previously described in the Arabidopsis-*B. cinerea* pathosystem.

In addition, the plant-fungus co-transcriptome network revealed the potential interaction between

fungal pathogen- and plant host-GCNs. Construction of GCNs and dual networks shed lights on

the biological mechanisms driving quantitative pathogen virulence in *B. cinerea* and their

potential targets in the plant innate immune system.

METHODS

Collection and Maintenance *B. cinerea* Isolates

A collection of 96 *B. cinerea* isolates were selected in this study based on their phenotypic and

genotypic diversity (Denby et al., 2004; Rowe and Kliebenstein, 2007; Corwin et al., 2016;

Zhang et al., 2016; Zhang et al., 2017). This *B. cinerea* collection was sampled from a large

variety of different host origins and contained a set of international isolates obtained from labs

across the world, including the well-studied B05.10 isolate. A majority of isolates are natural

isolates that isolated from California and can infect a wide range of crops. Isolates are

maintained in -80°C freezer stocks as spores in 20% glycerol and were grown on fresh potato

dextrose agar (PDA) 10 days prior to infection.

Plant Materials and Growth Conditions

The Arabidopsis accession Columbia-0 (Col-0) was the wildtype background of all Arabidopsis

mutants used in this study. The three Arabidopsis genotypes used in this study included the WT

and two well-characterized immunodeficient mutants, *coi1-1* and *npr1-1*, that abolish the major

JA- or SA-defense perception pathways, respectively (Cao et al., 1997; Xie et al., 1998; Xu, L.

et al., 2002; Pieterse and Van Loon, 2004). All plants were grown as described previously

(Zhang et al., 2017). Two independent randomized complete block-designed experiments were

conducted and a total of 90 plants per genotype were grown in 30 flats for each experiment.

Approximately 5 to 6 fully developed leaves were harvested from the five-week old plants and

placed on 1% phytoagar in large plastic flats prior to *B. cinerea* infection.

Inoculation and Sampling

We infected all 96 isolates onto each of the three Arabidopsis genotypes in a random design with

6-fold replication across the two independent experiments. A total of twelve infected leaves per

isolate/genotype pair were generated. For inoculation, all *B. cinerea* isolates were cultured and

inoculated on three Arabidopsi*s* genotypes as described previously (Denby et al., 2004; Corwin

et al., 2016; Zhang et al., 2017). Briefly, frozen glycerol stocks of isolate spores were first used

for inoculation on a few slices of canned peaches in petri plates. Spores were collected from one-

week-old sporulating peach slices. The spore solution was filterred and the spore pellet was re-

suspended in sterilized 0.5x organic grape juice (Santa Cruz Organics, Pescadero, CA). Spore

concentrations were determined using a hemacytometer and suspensions were diluted to

10spores/μL. Detached leaf assays were used for a high-throughput analysis of *B. cinerea*

infection, which has been shown to be consistent with whole plant assay (Govrin and Levine,

2000; Mengiste et al., 2003; Denby et al., 2004; Sharma et al., 2005; Windram et al., 2012).

Five-week old leaves were inoculated with 4μL of the spore solution. The infected leaf tissues

were incubated on 1% phytoagar flats with a humidity dome at room temperature. The

inoculation was conducted in a randomized complete block design across the six planting blocks.

All inoculations were conducted within one hour of dawn and the light period of the leaves was

maintained. Two blocks were harvest at 16HPI for RNA-Seq analysis. The remaining four

blocks were incubated at room temperature until 72HPI when they were digitally imaged for

lesion size and harvested for chemical analysis as described previously (Zhang et al., 2017).

RNA-Seq Library Preparation, Sequencing, Mapping and Statistical Analysis

Two *B. cinerea* infected leaf tissues of the six blocks were sampled at 16HPI for transcriptome

analysis, which resulted in a total of 1,052 mRNA libraries for Illumina HiSeq sequencing.

RNA-Seq libraries were prepared according to a previous method (Kumar et al., 2012) with

minor modifications (Zhang et al., 2017). Briefly, infected leaves were immediately frozen in

liquid nitrogen and stored at -80°C until processing. RNA extraction was conducted by re-

freezing samples in liquid nitrogen and homogenizing by rapid agitation in a bead beater

followed by direct mRNA isolation using the Dynabeads oligo-dT kit. First and second strand

cDNA was produced from the mRNA using an Invitrogen Superscript III kit. The resulting

cDNA was fragmented, end-repaired, A-tailed and barcoded as previously described. Adapter-

ligated fragments were enriched by PCR and size-selected for a mean of 300 base pair (bp) prior

to sequencing. Barcoded libraries were pooled in batches of 96 and submitted for a single-end,

50bp sequencing on a single lane per pool using the Illumina HiSeq 2500 platform at the UC

Davis Genome Center (DNA Technologies Core, Davis, CA).

Transcriptomic Data Analysis

Fastq files from individual HiSeq lanes were separated by adapter index into individual RNA-

Seq library samples. The quality of individual libraries was estimated for overall read quality and

over-represented sequences using FastQC software (Version 0.11.3,

www.bioinformatics.babraham.ac.uk/projects/). We conducted downstream bioinformatic

analysis, like reads mapping, normalization and nbGLM model analysis, using a custom script

from the *Octopus* R package (https://github.com/WeiZhang317/octopus). The mapping of

processed reads against Arabidopsis and *B. cinerea* reference genomes was conducted by Bowtie

1 (V.1.1.2, http://sourceforge.net/projects/bowtie-bio/files/bowtie/1.1.2/) using minimum

phred33 quality scores (Langmead et al., 2009). The first 10bp of reads was trimmed to remove

low quality bases using the fastx toolkit

(http://hannonlab.cshl.edu/fastx_toolkit/commandline.html). Total reads for each library were

firstly mapped against the Arabidopsis TAIR10.25 cDNA reference genome. The remaining un-

mapped reads were then aligned against *B. cinerea* B05.10 isolate cDNA reference genome

(Lamesch et al., 2010; Lamesch et al., 2012; Krishnakumar et al., 2015; Van Kan et al., 2017)

and the gene counts for both species were pulled from the resulting SAM files (Li et al., 2009).

For pathogen gene expression analysis, we first filtered genes with either more than 30 gene

counts in one isolate or 300 gene counts across 96 isolates. We normalized *B. cinerea* gene

counts data set using the trimmed mean of M-values method (TMM) from the *EdgeR* package

(V3.12) (Robinson and Smyth, 2008; Bullard et al., 2010; Robinson and Oshlack, 2010). We

then ran the following generalized linear model (GLM) with a negative binomial link function

from the MASS package for all transcripts using the following equation (Venables and Ripley,

2002):

$Y_{egai} = E_e + E_e(Gf_g) + E_e(Gf_g(Af_a)) + I_i + H_h + H_h* I_i$

where the main categorical effects E, I, and H are denoted as experiment, isolate genotype, and

plant host genotype, respectively. Nested effects of the growing flat (Gf) within the experimental

replicates and agar flat (Af) nested within growing flat are also accounted for within the model.

Model corrected means and standard errors for each transcript were determined for each

isolate/plant genotype pair using the lsmeans package (Russell V. Lenth, 2016). Raw *P*-values

for F- and Chi Square-test were determined using Type II sums of squares using *car* package

(Fox and Weisberg, 2011). *P*-values were corrected for multiple testing using a false discovery

rate correction (Benjamini Yoav and Yekutieli Daniel, 2001). Broad-sense heritability (H^2) of

individual transcripts was estimated as the proportion of variance attributed to *B. cinerea*

genotype, Arabidopsi*s* genotype, or their interaction effects.

Gene Ontology Analysis

GO analysis was conducted for several *B. cinerea* gene sets that were identified with high heritability,

correlated with lesion size, and condensed in network analysis. We first converted sequences of these *B.*

cinerea genes into fasta files using *Biostrings* and *seqRFLP* packages in R (Ding Qiong and Zhang

Jinlong, 2012; Pages et al., 2017). The functional annotation of genes was obtained by blasting the

sequences against the NCBI database using Blast2GO to obtain putative GO annotations (Conesa et al.,

2005; Gotz et al., 2008). The GO terms were compared to the official GO annotation from the *B.*

cinerea database (http://fungi.ensembl.org/Botrytis_cinerea/Info/Index) and those obtained by Blast2GO

analysis. The official gene annotations for host genes was retrieved from TAIR10.25

(https://apps.araport.org/thalemine/bag.do?subtab=upload).

B. cinerea **Gene Co-expression Network Construction**

To obtain a representative subset of *B. cinerea* genes co-expressed under *in planta* conditions,

we generated gene co-expression networks (GCNs) among genes in the *B. cinerea* transcriptome.

GCNs were generated using the model-corrected means of 9,284 *B. cinerea* transcripts from

individual isolate infection across three Arabidopsis genotypes. Only genes with average or

medium expression greater than zero across all samples were considered. This preselection

process kept 6,372 genes and those with negative expression values were adjusted to set

expression at zero before network construction. Spearman's rank correlation coefficients for each

gene pair was calculated using the *cor* function in R. Three gene-for-gene correlation similarity

matrixes were generated independently for each of the three Arabidopsis genotypes. Considering

the cutoff for gene-pair correlation usually generates biases of GCN structure and the candidate

gene hit, we utilized several cutoff threshold values at 0.75, 0.8, 0.85, and 0.9 to filter the gene

set. Comparing the structure and content of GCNs among those GCN sets using filtered gene set

as input, we selected the correlation threshold at 0.8. A total of 600, 700 and 494 *B. cinerea*

candidate genes passed the criterion under Arabidopsis WT, mutants *coi1-1* and *npr1-1*,

respectively. To obtain a representative subset of *B. cinerea* gene candidates across three host

genotypes, we selected gene candidates that presented across the above three gene subsets. This

process generated a gene set with 323 *B. cinerea* candidate genes that were common to each of

the plant genotype backgrounds and had at least 0.8 significant correlations. Using this gene set

as kernel, we extended gene candidate sets under each Arabidopsis genotype. The expanded *B.*

cinerea gene candidate set under individual Arabidopsis genotypes was further used as input for

gene co-expression network construction.

GCNs were visualized using Cytoscape V3.2.1 (Java version:1.8.0_60) (Shannon et al., 2003).

The nodes and edges within each network represent the *B. cinerea* genes and the Spearman's

rank correlations between each gene pair. The importance of a given node within each network

was determined by common network analysis indices, such as connectivity (degree) and

betweenness. Nodes with higher connectivity and betweenness were considered as hub and

bottleneck genes, respectively, and the biological functions of each network were determined by

the GO terms of hub and bottle neck genes using Blast2GO.

Cross-kingdom Arabidopsis-*B. cinerea* Gene Co-expression Network Construction

We used model-corrected means of transcripts from three Arabidopsis host genotypes and 96 *B.*

cinerea isolates to construct the cross-kingdom Arabidopsis-*B. cinerea* GCNs. Model-corrected

means of 23,959 Arabidopsis transcripts and 6,372 *B. cinerea* transcripts derived from two

negative binomial linked generalized linear models were served as input data sets (Zhang et al.,

2017). Spearman's rank correlation coefficient was calculated between genes from Arabidopsis

and *B. cinerea* data sets. The gene pairs with positive correlations greater than 0.74 under each

Arabidopsis genotype were considered to construct cross-kingdom GCNs.

Dual Interaction Network Construction

To construct a cross-kingdom, dual interaction network of plant-pathogen GCNs, we performed

principle component analysis on individual GCNs within each species to obtain eigengene

vectors describing the expression of the entire gene network as previously described (Zhang and

Horvath, 2005; Langfelder and Horvath, 2008; Okada et al., 2016). From these eigengene

vectors, we calculated the Spearman's rank correlation coefficient between the first eigengene

vectors for each network. The resulting similarity matrices were used as input to construct the

interaction network and Cytoscape was used to visualize the resulting network.

Statistical Analysis of Network Components

All the analyses were conducted using R V3.2.1 statistical environment (R Core Team, 2014). To

investigate how secondary metabolite induction in *B. cinerea* contributes to disease

development, we conducted a multi-factor ANOVA on *B. cinerea* three secondary metabolic

pathways upon impacts on host genotypes. The three secondary metabolic pathways included the

biosynthetic pathways of two well-known secondary metabolites, BOT and BOA, and a cyclic

peptide biosynthetic pathway predicted in this study. We calculated the z-scores for all

transcripts involved in BOT pathway, the BOA, and the putative cyclic peptide pathway for each

isolate/plant genotype pair. The multi-factor ANOVA model for lesion area was:

$y_{Lesion} = \mu + T * A * C * G_h + \varepsilon$

where T, A, C, and G_h stand for BOT, BOA, Cyclic peptide, and host genotype, respectively.

In addition, we used multi-factor ANOVA models to investigate interactions between GCNs

within species for impacts upon host genotypes. The ANOVA models contain all GCNs within a

species. The first eigengene vector derived from principal component analysis on each network

was used in ANOVA models. The ANOVA model for individual *B. cinerea* GCNs was:

$y_{BcNeti} = \mu + D * P * C * PSI * G_h + \varepsilon$

where D, P, C, PSI, and G_h stand for Arabidopsis Defense/Camalexin GCN, Arabidopsis Plastid

GCN, Arabidopsis Cell/Division GCN, Arabidopsis PSI GCN, and Host genotypes, respectively.

G_h stands for HostGenotype, respectively. BcNeti represents one of the ten *B. cinerea* GCNs

identified in this study. The ANOVA model for individual Arabidopsis GCNs was:

$y_{AtNeti} = \mu + \sum BcNeti + G_h + \varepsilon$

where $\sum$BcNeti represents the summation of each of the ten *B. cinerea* GCNs identified in this

study: BcVesicle/Viru GCN, BcTSL/Growth GCN, BcBOA GCN, BcExocytoRegu GCN,

BcCycPep GCN, BcCopperTran GCN, BcBOT GCN, BcPeptidase GCN, BcIPP GCN,

BcPolyketide GCN, while G_h stands for Host genotypes. Interactions among the terms were not

tested to avoid the potential for overfitting. AtNeti stands for one of the four Arabidopsis GCNs

(e.g. AtDefense/Camalexin GCN, AtPlastid GCN, AtCell/Division GCN, AtPSI GCN). All

multi-factor ANOVA models were optimized by trimming to just the terms with a significant *P*-

value (*P*-value < 0.05).

Germination Assay

To assess the potential for natural variation in germination time in the isolate collection, 19 *B.*

cinerea isolates were investigated by germination assay. The isolates were grown on PDA.

Mature spores were collected in water, filtered and resuspended in 50% grape juice, as

previously described, and further diluted to 1000spores/μL. To prevent germination before the

beginning of the assay, spores were continuously kept on ice or in the fridge at 4°C. During the

germination assay, the spores were maintained at 21°C in 1.5mL tubes. Every hour, the tubes

were mixed by manual inversion and sampled for 25μL that were transferred to microscope

slides. The spores within the drops were let to set down shortly. Without using slide covers, the

spores were observed within the drops at two locations, used as technical replicates. The spores

were categorized and counted based on the picture of every microscope observations taken every

hour from 2 to 11 hours. Germination was defined as the hyphae emerged out of the spore.

To assess the contribution of germination to the observed *B. cinerea* transcriptomic networks

involved in lesion development, we generated germination estimates based on gene expression

by extracting the first principal component of a publicly available time series microarray data

including 101 gemination-associated genes (Leroch et al., 2013). Based on this principal

component, we predicted the level of expression of germination-associated genes for the 96

isolates on the three Arabidopsis genotypes at 16HPI. Theses germination predictions for

individual isolates were used in a linear ANOVA model to estimate the co-linearity of the

germination eigengene vector to virulence. Using linear ANOVA models with and without this

germination eigengene vector, we compared how germination influences the 10 *B. cinerea*

transcriptomic networks involved in lesion area in the three host genotypes. The ANOVA

models with and without germination eigengene vector were:

$Y_{Lesion} = \mu + Germination + \sum BcNeti + G_h + \varepsilon$

$Y_{Lesion} = \mu + \sum BcNeti + G_h + \varepsilon$

where Germination represents the scores of first principal components on expressions of

germination associated genes from *B. cinerea* transcriptomic data in this study, $\sum BcNeti$

represents the summation of each of the ten *B. cinerea* GCNs identified in this study:

BcVesicle/Viru GCN, BcTSL/Growth GCN, BcBOA GCN, BcExocytoRegu GCN, BcCycPep

GCN, BcCopperTran GCN, BcBOT GCN, BcPeptidase GCN, BcIPP GCN, BcPolyketide GCN,

while G_h stands for Host genotypes. Interactions among the terms were not tested to avoid the

potential for overfitting.

Data Availability

The datasets in this study are available in the following database: Bioproject PRJNA473829

(https://www.ncbi.nlm.nih.gov/bioproject/?term=PRJNA473829). The computer scripts used in

this study are available in GitHub (https://github.com/WeiZhang317/octopus).

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Figure legends

Figure 1. Correlation between earlier estimated *B. cinerea* biomass and later lesion area.

Model-corrected lesion area means were estimated using the linear model on the six replicates

data from three Arabidopsis genotypes at 72-hours post-infection with 96 *B. cinerea* isolates.

Estimated biomass of *B. cinerea* was calculated using the linear model-corrected fraction of *B.*

cinerea mapped reads against total mapped reads to Arabidopsis and *B. cinerea* reference

genomes. RNA-Seq analysis was conducted at 16-hours post-infection for each pathosystem.

Three Arabidopsis genotypes are wild-type Col-0 (purple dot), jasmonate insensitive mutant

coi1-1 (green triangle), and salicylic acid insensitive mutant *npr1-1* (orange diamond). The 90%

confidence ellipse intervals are plotted for each Arabidopsis genotype for references. Quadratic

regression lines are: Col-0: $y = -0.00059x^2 + 0.729x + 10.037$, $P = 0.0016$, adjusted $R^2 = 0.1101$;

coi1-1: $y = -0.117x^2 + 4.44x – 0.1585$, $P = 3.914e-07$, adjusted $R^2 = 0.2562$; *npr1-1*: $y = -$

$0.0579x^2 + 2.26x + 1.673$, $P = 0.0001$, adjusted $R^2 = 0.161$.

Figure 2. Transcriptomic responses of *B. cinerea* on Arabidopsis are controlled by genetic

variation in pathogen population, host genotypes, and their interaction. (A) Distribution of

broad-sense heritability (H^2) of *B. cinerea* transcripts contributed by genetic variation in the *B.*

cinerea, Arabidopsis genotypes, and the interaction between pathogen and host. Violin plots

illustrating the distribution of H^2 for transcripts from 96 *B. cinerea* isolates infecting on

Arabidopsis genotypes. Heritability is partitioned across the different sources, 96 pathogen

genotypes = "Isolate", plant genotypes Col-0, *coi1-1* and *npr1-1* plant genotypes = "Host", and

the corresponding interaction. The transcriptomic analysis was conducted by sequencing mRNA

extracted from *B. cinerea* infected Arabidopsis leaves at 16-hours post-infection. Red lines

indicate the average broad-sense heritability values of lesion area caused by isolates, Arabidopsis

genotypes, and their interaction. (B) to (E) Expression profiles of *B. cinerea* transcripts

significantly influenced by host genotypes. The model-corrected means ($\log_2$) for *B. cinerea*

transcript were used for plotting. The Arabidopsis genotypes, wild-type Col-0 (purple),

jasmonate insensitive mutant *coi1-1* (green), and salicylic acid mutant *npr1-1* (orange), are

shown on the x axis. *B. cinerea* transcripts are: (B) *Bcin01g03790.1*, class IV chitin synthase; (C)

Bcin02g02980.1, Monooxygenase; (D*) Bcin10g02260.1*, NADH oxidoreductase; (E)

Bcin11g05200.1, caffeine-induced death Cid2; (F) to (I) Expression profiles of *B. cinerea*

transcripts significantly influenced by the interaction between pathogen and host genotypes. (F)

Bcin02g03810.1, GTP cyclohydrolase; (G) *Bcin09g01190.1*, Tripeptidyl-peptidase 1; (H)

Bcin10g01130.1, in secretory pathway; (I) *Bcin14g05090.1*, a transcription factor.

Figure 3. Expression profiles of an endopolygalacturonase gene *Bcpg1* from diverse *B.

cinerea isolates across Arabidopsis genotypes.** Rank plot shows the relationship of *Bcpg1*

expression from 32 diverse *B. cinerea* isolates (right) across three Arabidopsis genotypes (x

axis). Three Arabidopsis genotypes are wild-type Col-0 (purple dot), jasmonate insensitive

mutant *coi1-1* (green triangle), and salicylic acid mutant *npr1-1* (orange diamond). The model-

corrected means ($\log_2$) for the transcript of *Bcpg1* (*Bcin14g00850.1*) encoding an

endopolygalacturonase gene are utilized for plotting. The transcript expression levels from the

same isolate across three Arabidopsis genotypes are connected with a colored line. The names of

32 isolates are represented with the same colored lines as induced *Bcpg1* expression levels.

Black lines indicate the expression levels of *Bcpg1* are higher in *coi1-1* and *npr1-1* than in Col-0.

Red lines indicate the higher expression levels of *Bcpg1* in *coi1-1* but lower in *npr1-1*. Blue lines

indicate the highest expression levels of *Bcpg1* are in Col-0. Dark green lines indicate the higher

expression levels of *Bcpg1* in *npr1-1* but lower in *coi1-1*.

Figure 4. Interaction effects of host genotypes and pathogen isolates on *B. cinerea*

transcriptome. Hierarchical clustering of relative expression of 500 genes from 96 *B. cinerea*

isolates infection on Arabidopsis mutants *coi1-1* (A) or *npr1-1* (B) are plotted based on pairwise

comparison of pathogen gene expression under Col-0. The 500 *B. cinerea* genes with highest

broad-sense heritability (H^2) of host X pathogen were used for analysis. Lesion area induced by

96 isolates are compared under *coi1-1* (orange bar plot) and npr1-1 (green bar plot).

Figure 5. Gene co-expression networks identified from *B. cinerea* transcriptomic responses

to Arabidopsis wild-type Col-0 immunity. Ten gene co-expression networks (GCNs) with

more than five nodes were identified from 96 *B. cinerea* isolates infecting on Arabidopsis wild-

type Col-0. The similarity matrix is computed using Spearman's rank correlation coefficient.

Nodes with different colors represent *B. cinerea* genes condensed in GCNs with different

biological functions. Edges represent the Spearman's rank correlation coefficients between gene

pairs. Trans- and cis-GCNs means GCNs are regulated by trans- and cis-regulatory elements,

respectively. GCNs were named after their biological functions, which were determined by hub

and bottleneck genes within each network. GCNs are: vesicle/virulence (red), translation/growth

(green), exocytosis regulation (pink), cyclic peptide (yellow), peptidase (gray), isopentenyl

pyrophosphate (IPP, turquoise), polyketide (violet), botcinic acid (BOA, blue), copper transport

(slate blue), botrydial (BOT, purple).

Figure 6. Variation of transcripts accumulation for secondary metabolites production

across diverse *B. cinerea* isolates. Expression profiles of genes responsible for botrydial,

botcinic acid, cyclic peptide production across 96 isolates under Arabidopsis wild-type Col-0 are

shown. **(**A) Schematic shows the genomic locus of seven botrydial (BOT) biosynthesis genes

clustered together. Exons are represented by gray boxes. Introns and intergenic regions are

represented by the grey line. Seven BOT genes are: *BcBOT1, BcBOT3* and *BcBOT4,* encoding a

cytochrome P450 monooxygenase, respectively; *BcBOT2* encoding a sesquiterpene cyclase;

BcBOT5 encoding an acetyl transferase; *BcBOT6* encoding a Zn(II)2Cys6 transcription factor,

BcBOT7 encoding a dehydrogenase reductase. (B) BOT biosynthesis pathway in *B. cinerea*. (C)

Bar plots compare expression variation of *BcBOT*2 across 96 *B. cinerea* isolates in responding to

Arabidopsis wild-type Col-0 immunity. The model-corrected means (log_2) of transcripts were

used for plotting. (D) Scatter plot illustrates the positive correlations between lesion area and

accumulation of *BcBOT2* transcript across the 96 isolates in response to varied Arabidopsis

immunities. Model-corrected lesion area means were estimated for three Arabidopsis genotypes

at 72-hours post-infection with 96 *B. cinerea* isolates. The three Arabidopsis genotypes are

labeled next to the confidence ellipse curves: wild-type Col-0 (purple dot), jasmonate insensitive

mutant *coi1-1* (green triangle), and salicylic acid mutant *npr1-1* (orange diamond). The 90%

confidence ellipse intervals are plotted for each Arabidopsis genotype for reference. Linear

regression lines: Col-0: y = 3.2532x + 4.4323, *P* = 1.008e-10, Adjusted R^2 = 3.3537; *coi1-1*: y =

7.4802x + 10.3289, *P* = 7.895e-15, adjusted R^2 = 0.4700; *npr1-1*: y = 3.7086x + 7.3487, *P* =

2.425e-11, adjusted R^2 = 0.3726. (E) and (F) Bar plots compare expression variation of *BcBOA6*

in botcinic acid (BOA) pathway and *Bcin01g11460.* in cyclic peptide pathway across 96 *B.*

cinerea isolates in response to Arabidopsis wild-type Col-0 immunity. (G) Venn diagram

illustrates the number of *B. cinerea* isolates with the ability to induce BOT, BOA, and cyclic

peptide.

Figure 7. Comparison of plasticity of *B. cinerea* gene co-expression network under vaired

host immunity. *B. cinerea* gene co-expression networks (GCNs) of vesicle/virulence (red) and

translation/growth (green) identified under three Arabidopsis genotypes are compared. Three

Arabidopsis genotypes are wild-type Col-0, jasmonate insensitive mutant *coi1-1*, and salicylic

acid mutant *npr1-1*. Nodes marked with red and green colors represent *B. cinerea* genes

condensed in GCNs with different biological functions. The same node condensed in GCNs

across three Arabidopsis genotypes was marked with same color. Nodes specificaly condensed in

GCNs under two mutants *coi1-1* and *npr1-1* background are marked with orange color. Edges

represent the Spearman's rank correlation coefficients between gene pairs.

Figure 8. Cross-kingdom Arabidopsis-*B. cinerea* gene co-expression networks. Networks

showing the co-expression connectivity between Arabidopsis and *B. cinerea* genes within three

Arabidopsis genotypes are shown. (A) shows connectivity within Arabidopsis wild-type Col-0,

(B) shows connectivity within the Arabidopsis jasmonate insensitive mutant *coi1-1*, and (C)

shows connectivity within the Arabidopsis salicylic acid insensitive mutant *npr1-1*. Within each

connectivity plot, orange and green nodes show transcripts from *B. cinerea* and Arabidopsis,

respectively. Nodes with red and violet colors represent the *B. cinerea* transcripts that were

found to be members of the *B. cinerea* membrane/vesicle virulence network and BOT network,

respectively. Node size shows the number of interactions with a specific gene. The connectivity

between the nodes was derived using Spearman's rank correlation analysis.

Figure 9. A dual interaction network reveals links between Arabidopsis immunity and *B.*

***cinerea* virulence.** A dual interaction network was constructed using gene co-expression

networks (GCNs) from Arabidopsis and *B. cinerea* co-transcriptome. The first eigenvectors were

derived from individual GCNs and used as input to calculate Spearman's rank correlation

coefficiency between GCN pairs. Green dots and orange triangles represent Arabidopsis

immune- and *B. cinerea* virulence-GCNs, respectively. Blue and red lines (edges) represent the

positive and negative Spearman's rank correlation coefficients between GCN pairs, respectively.

The thickness of line signifies the correlational strength.

Supplemental Data

Figure 3-figure supplemental 1. **Expression profiles of an oxaloacetate hydrolase gene**

***Bcoah* from diverse *B. cinerea* isolates across Arabidopsis genotypes.** Rank plot shows the

relationship o *Bcoah* expression from 32 diverse *B. cinerea* isolates (right) across three

Arabidopsis genotypes (x axis). Three Arabidopsis genotypes are wild-type Col-0 (purple dot),

jasmonate insensitive mutant *coi1-1* (green triangle), and salicylic acid mutant *npr1-1* (orange

diamond). The model-corrected means ($\log_2$) for the transcript of *Bcoah* (*Bcin12g01020.1*)

encoding an oxaloacetate hydrolase gene are utilized for plotting. The transcript expression

levels from the same isolate across three Arabidopsis genotypes are connected with a colored

line. The names of 32 isolates are represented with the same colored lines as induced *Bcoah*

expression levels. Black lines indicate the expression levels of *Bcoah* are higher in *coi1-1* and

npr1-1 than in Col-0. Red lines indicate the higher expression levels of *Bcoah* in *coi1-1* but

lower in *npr1-1*. Blue lines indicate the highest expression levels of *Bcoah* are in Col-0. Dark

green lines indicate the higher expression levels of *Bcoah* in *npr1-1* but lower in *coi1-1*.

Figure 5-figure supplemental 1. Genomic location of *B. cinerea* gene co-expression

networks. The circle diagrams showed the genome-wide distribution of gene pairs identified by

B. cinerea gene co-expression networks (GCNs) under Arabidopsis wild-type Col-0. (A) to (J)

Genomic locations of gene pairs identified by GCNs: (A) Vesicle/virulence, (B)

translation/growth, (C) botcinic acid (BOA), (D) exocytosis regulation, (E) cyclic peptide, (F)

copper transport, (G) botrydial (BOT), (H) peptidase (gray), (I) isopentenyl pyrophosphate (IPP),

(G) polyketide. The rings show 18 *B. cinerea* chromosomes on a Mb scale. Genomic locations of

co-expressed gene pairs are connected by the colored lines.

Figure 7-figure supplemental 1. Gene co-expression networks identified from *B. cinerea*

transcriptomic responses to Arabidopsis wild-type Col-0 immunity. *B. cinerea* gene co-

expression networks (GCNs) were identified from 96 *B. cinerea* isolates infecting on

Arabidopsis wild-type Col-0. The similarity matrix is computed using Spearman's rank

correlation coefficient. All co-expressed gene pairs with correlation greater than 0.85 were

shown. Nodes with different colors represent *B. cinerea* genes condensed in GCNs with different

biological functions. Edges represent the Spearman's rank correlation coefficients between gene

pairs. GCNs were ordered as number of nodes within each network. GCNs were named after

their biological functions, which were determined by hub and bottleneck genes within each

network: vesicle/virulence (red), translation/growth (green), botcinic acid (BOA, blue),

exocytosis regulation (pink), cyclic peptide (yellow), copper transport (slate blue), botrydial

(BOT, purple), peptidase (gray), isopentenyl pyrophosphate (IPP, turquoise), polyketide (violet).

Figure 7-figure supplemental 2. Gene co-expression networks identified from _B. cinerea_

transcriptomic responses to Arabidopsis jasmonate-compromised immunity. _B. cinerea_

gene co-expression networks (GCNs) were identified from 96 _B. cinerea_ isolates infecting on

Arabidopsis jasmonate insensitive mutant _coi1-1_. The similarity matrix is computed using

Spearman's rank correlation coefficient. All co-expressed gene pairs with correlation greater

than 0.8 were shown. Nodes with different colors represent _B. cinerea_ genes condensed in GCNs

with different biological functions. Nodes were marked with same color as under Arabidopsis

wild-type Col-0 background. Nodes specificaly condenced in GCNs under Arabidopsis mutant

coi1-1 background are marked with orange color. Edges represent the Spearman's rank

correlation coefficients between gene pairs. GCNs were ordered as number of nodes within each

network. GCNs were named after their biological functions, which were determined by hub and

bottleneck genes within each network: vesicle/virulence (red/orange), translation/growth

(green/orange), botrydial/ isopentenyl pyrophosphate (BOT/IPP, blue/turquoise/orange), botcinic

acid (BOA, blue), exocytosis regulation (pink), peptidase (gray/orange), copper transport (slate

blue/orange), plant cell wall degradation (orange), cyclic peptide (yellow), peptidase II

(red/orange), siderophores (orange), 26S proteasome regulated protein degradation (red/orange),

sugar, ROS/NO stress (orange), ATP (orange), polyketide (violet/orange).

Figure 7-figure supplemental 3. Gene co-expression networks identified from _B. cinerea_

transcriptomic responses to Arabidopsis salicylic acid-compromised immunity. _B. cinerea_

gene co-expression networks (GCNs) were identified from 96 _B. cinerea_ isolates infecting on

Arabidopsis salicylic acid insensitive mutant _npr1-1_. The similarity matrix is computed using

Spearman's rank correlation coefficient. All co-expressed gene pairs with correlation greater

than 0.8 were shown. Nodes with different colors represent _B. cinerea_ genes condensed in GCNs

with different biological functions. Nodes were marked with same color as under Arabidopsis

wild-type Col-0 background. Nodes specificaly condensed in GCNs under Arabidopsis mutant

npr1-1 background are marked with orange color. Edges represent the Spearman's rank

correlation coefficients between gene pairs. GCNs were ordered as number of nodes within each

network. GCNs were named after their biological functions, which were determined by hub and

bottleneck genes within each network: translation/growth (green/orange), vesicle/virulence

(red/orange), peptidase (gray/orange), polyketide (violet/orange), botrydial/isopentenyl

pyrophosphate (BOT/IPP, blue/turquoise/orange), botcinic acid (BOA, blue), exocytosis

regulation (pink), copper transport (slate blue/orange), cyclic peptide (yellow), unknown (red),

sugar (orange), peptidase II (red/orange).

Figure 7-figure supplemental 4. Plasticity in expression profiles of genes identified by *B.*

***cinerea* gene co-expression networks under varied Arabidopsis immunities.** Violin plots of

(A) to (O) show the expression profiles of *B. cinerea* genes in response to variation of

Arabidopsis immunity. Genes shown are condensed in *B. cinerea* gene co-expression networks

(GCNs). The model-corrected means ($\log_2$) for *B. cinerea* transcript were used for plotting. The

Arabidopsis genotypes, wild-type Col-0 (purple), jasmonate insensitive mutant *coi1-1* (green),

and salicylic acid mutant *npr1-1* (orange), are shown on the x axis.

Figure 8-figure supplemental 1. Associations between gene co-expression networks

identified from co- and single-transcriptome. Venn diagrams highlights: (A) the overlap of

plant (left) and pathogen (right) genes condensed in Arabidopsis-*B. cinerea* gene co-expression

networks (GCNs) across three Arabidopsis genotypes, (B) the overlap of plant genes in

Arabidopsis-B. cinerea GCNs and Arabidopsis GCNs across three genotypes, (C) the overlap of

pathogen genes in Arabidopsis-*B. cinerea* GCNs and *B. cinerea* GCNs across three genotypes.

Figure 9-figure supplemental 1. Dual networks reveal links between Arabidopsis immunity

and *B. cinerea* virulence under *coi1-1* and *npr1-1*. Dual interaction networks were constructed

using gene co-expression networks (GCNs) from Arabidopsis and *B. cinerea* co-transcriptome

under JA-insensitive mutant *coi1-1* and SA-insensitive mutant *npr1-1*, respectively. The first

eigenvectors were derived from individual GCNs and used as input to calculate Spearman's rank

correlation coefficiency between GCN pairs. Green dots and orange triangles (nodes) represent

Arabidopsis immune- and *B. cinerea* virulence-GCNs, respectively. Blue and red lines (edges)

represent the positive and negative Spearman's rank correlation coefficients between GCN pairs,

respectively. The thickness of line signifies the correlational strength.

Figure 9-figure supplemental 2. Variation of *B. cinerea* spore germination time under *in*

***vitro* condition.**

Supplemental Table 1. Topology traits of *B. cinerea in planta* gene co-expression networks.

Supplemental Table 2. ANOVA table of lesion area and *B. cinerea* pathways. A mixed linear

model was fitted to test lesion area and *B. cinerea* pathways responsible for botrydial (BOT),

botcinic acid (BOA), and cyclic peptide (CycPep) produced under three Arabidopsis genotypes.

The lesion area data used in the model were GLM corrected least square means induced by 96 *B.*

cinerea isolates. Model-corrected means of transcripts from 96 *B. cinerea* isolates were z-scaled

and used in ANOVA. Df is the degrees of freedom for a term within the model. SS is the Sum of

Squares variation. MS is the Mean of Squared variation. F value is derived from the F statistic

and *P*-value indicates the statistical significance for a given term within the model. Significance

of differences are shown as $P < 0.001$ '***', 0.01 '**' and 0.05 '*'.

Supplemental Table 3. ANOVA tables of *B. cinerea* gene co-expression networks. Mixed

linear models were fitted to individual *B. cinerea* (Bc) gene co-expression networks (GCNs) and

variation of host genotypes and Arabidopsis (At) GCNs. Variation was estimated among host

genotypes and first eigenvectors from four individual Arabidopsis GCNs. Df is the degrees of

freedom for a term within the model. SS is the Sum of Squares variation. MS is the Mean of

Squared variation. F value is derived from the F statistic and *P*-value indicates the statistical

significance for a given term within the model. Significance of difference are shown as $P < 0.001$

'***', 0.01 '**' and 0.05 '*'.

Supplemental Table 4. **ANOVA tables of Arabidopsis gene co-expression networks.** Linear

mixed models were fitted to individual Arabidopsis (At) gene co-expression networks (GCNs)

and variation of host genotypes and ten *B. cinerea* (Bc) GCNs. Variation was estimated among

host genotypes and first eigenvectors from individual *B. cinerea* GCNs. Df is the degrees of

freedom for a term within the model. SS is the Sum of Squares variation. MS is the Mean of

Squared variation. F value is derived from the F statistic and *P*-value indicates the statistical

significance for a given term within the model. Significance of difference are shown as $P < 0.001$

'***', 0.01 '**' and 0.05 '*'.

Supplemental Table 5. Topology traits of cross-kingdom Arabidopsis-*B. cinerea* gene co-

expression networks. Three Arabidopsis genotypes are wild-type Col-0, jasmonate insensitive

mutant *coi1-1*, and salicylic acid insensitive mutant *npr1-1*.

Supplemental Table 6. Analysis of potential impact of germination variation. To test if germination may influence the observed network to lesion connections, we estimated germination using the first principal component of genes linked to germination in Leroch et al., 2013. We then estimated the value of this principal component in the isolates grown on the three host genotypes and conducted a linear model to compare how this eigengene links to virulence. Using a linear model with and without this germination eigengene, we compared the link of the 10 *B. cinerea* transcript networks to lesion size with and without the germination eigengene vector in the model.

Supplemental Data Set 1. Model-corrected means of *B. cinerea* transcripts. A table of model-corrected least-square means of *B. cinerea* transcripts from 96 isolates infection on Arabidopsis wild-type Col-0, jasmonate insensitive mutant *coi1-1*, and salicylic acid insensitive mutant *npr1-1*.

Supplemental Data Set 2. Standard errors of *B. cinerea* transcripts. A table of model-corrected standard errors of *B. cinerea* transcripts infection on Arabidopsis wild-type Col-0, jasmonate insensitive mutant *coi1-1*, and salicylic acid insensitive mutant *npr1-1*.

Supplemental Data Set 3. GLM deviance tables and broad-sense heritability of *B. cinerea* transcripts. Summary of deviance tables derived from generalized linear model and estimated broad-sense heritability (H^2) for *B. cinerea* transcripts. All significance values are corrected by false discovery rate.

Supplemental Data Set 4. Model-corrected means of estimated *B. cinerea* biomass. *B. cinerea* biomass of 96 isolates infection on Arabidopsis genotypes was estimated using the

fraction of uniquely mapped reads against B05.10 reference genome. Three Arabidopsis

genotypes are wild-type Col-0, jasmonate insensitive mutant *coi1-1*, and salicylic acid

insensitive mutant *npr1-1*.

Supplemental Data Set 5. Top 100 heritability of *B. cinerea* transcripts. Broad-sense

heritability (H^2) of individual *B. cinerea* transcript contributed by pathogen, host and their

interaction were estimated. Three Arabidopsis genotypes are wild-type Col-0, jasmonate

insensitive mutant *coi1-1*, and salicylic acid insensitive mutant *npr1-1*.

Supplemental Data Set 6. Spearman's rank correlation between lesion area and *B. cinerea*

transcripts abundance. A table of spearman's rank correlation coefficiency between lesion area

and *B. cinerea* transcripts accumulation across three Arabidopsis genotypes or under individual

genotypes. Three Arabidopsis genotypes are wild-type Col-0, jasmonate insensitive mutant *coi1-*

1, and salicylic acid insensitive mutant *npr1-1*.

Supplemental Data Set 7. Gene list of *B. cinerea* gene co-expression networks. Tables of *B.*

cinerea genes identified by *B. cinerea* gene co-expression networks (GCNs) during 96 isolates

infection on Arabidopsis wild-type Col-0, jasmonate insensitive mutant *coi1-1*, and salicylic acid

insensitive mutant *npr1-1*.

Supplemental Data Set 8. Gene list of cross-kingdom Arabidopsis-*B. cinerea* gene co-

expression networks. Tables of Arabidopsis and *B. cinerea* genes identified by co-transcriptome

gene co-expression networks (GCNs) during 96 isolates infection on Arabidopsis wild-type Col-

0, jasmonate insensitive mutant *coi1-1*, and salicylic acid insensitive mutant *npr1-1*.

References

Arbelet, D., Malfatti, P., Simond-Cote, E., Fontaine, T., Desquilbet, L., Expert, D., Kunz,

C., and Soulie, M.C. (2010). Disruption of the Bcchs3a chitin synthase gene in Botrytis cinerea

is responsible for altered adhesion and overstimulation of host plant immunity. Mol. Plant

Microbe Interact. **23,** 1324-1334, doi/10.1094/MPMI-02-10-0046.

Ascari, J., Boaventura, M.A., Takahashi, J.A., Duran-Patron, R., Hernandez-Galan, R.,

Macias-Sanchez, A.J., and Collado, I.G. (2013). Phytotoxic activity and metabolism of

Botrytis cinerea and structure-activity relationships of isocaryolane derivatives. J. Nat. Prod. **76,**

1016-1024, doi/10.1021/np3009013.

Atwell, S., Corwin, J.A., Soltis, N.E., Subedy, A., Denby, K.J., and Kliebenstein, D.J.

(2015). Whole genome resequencing of Botrytis cinerea isolates identifies high levels of

standing diversity. Front. Microbiol. **6,** 996, doi/10.3389/fmicb.2015.00996.

Bednarek, P., Pislewska-Bednarek, M., Svatos, A., Schneider, B., Doubsky, J., Mansurova,

M., Humphry, M., Consonni, C., Panstruga, R., Sanchez-Vallet, A., Molina, A., and

Schulze-Lefert, P. (2009). A glucosinolate metabolism pathway in living plant cells mediates

broad-spectrum antifungal defense. Science **323,** 101-106, doi/10.1126/science.1163732.

Benjamini Yoav and Yekutieli Daniel (2001). The Control of the False Discovery Rate in

Multiple Testing under Dependency. **29,** 1165-1188, doi/10.1214/aos/1013699998.

Benton, M.J. (2009). The Red Queen and the Court Jester: species diversity and the role of

biotic and abiotic factors through time. Science **323,** 728-732, doi/10.1126/science.1157719.

Bergelson, J., Kreitman, M., Stahl, E.A., and Tian, D. (2001). Evolutionary dynamics of plant

R-genes. Science **292,** 2281-2285, doi/10.1126/science.1061337.

Blanco-Ulate, B., Morales-Cruz, A., Amrine, K.C., Labavitch, J.M., Powell, A.L., and

Cantu, D. (2014). Genome-wide transcriptional profiling of Botrytis cinerea genes targeting

plant cell walls during infections of different hosts. Front. Plant. Sci. **5,** 435,

doi/10.3389/fpls.2014.00435.

Brooks, D.M., Bender, C.L., and Kunkel, B.N. (2005). The Pseudomonas syringae phytotoxin

coronatine promotes virulence by overcoming salicylic acid-dependent defences in Arabidopsis

thaliana. Mol. Plant. Pathol. **6,** 629-639, doi/10.1111/j.1364-3703.2005.00311.x.

Bullard, J.H., Purdom, E., Hansen, K.D., and Dudoit, S. (2010). Evaluation of statistical

methods for normalization and differential expression in mRNA-Seq experiments. BMC

Bioinformatics **11,** 94, doi/10.1186/1471-2105-11-94.

Buttner, D. and He, S.Y. (2009). Type III protein secretion in plant pathogenic bacteria. Plant

Physiol. **150,** 1656-1664, doi/10.1104/pp.109.139089.

Cantarel, B.L., Coutinho, P.M., Rancurel, C., Bernard, T., Lombard, V., and Henrissat, B.

(2009). The Carbohydrate-Active EnZymes database (CAZy): an expert resource for

Glycogenomics. **37,** D233–D238, doi/10.1093/nar/gkn663.

Cao, H., Glazebrook, J., Clarke, J.D., Volko, S., and Dong, X. (1997). The Arabidopsis NPR1

gene that controls systemic acquired resistance encodes a novel protein containing ankyrin

repeats. Cell **88,** 57-63, doi/10.1016/S0092-8674(00)81858-9.

Choquer, M., Fournier, E., Kunz, C., Levis, C., Pradier, J.M., Simon, A., and Viaud, M.

(2007). Botrytis cinerea virulence factors: new insights into a necrotrophic and polyphageous

pathogen. FEMS Microbiol. Lett. **277,** 1-10, doi/10.1111/j.1574-6968.2007.00930.x.

Clay, N.K., Adio, A.M., Denoux, C., Jander, G., and Ausubel, F.M. (2009). Glucosinolate

metabolites required for an Arabidopsis innate immune response. Science **323,** 95-101,

doi/10.1126/science.1164627.

Colmenares, A.J., Aleu, J., Duran-Patron, R., Collado, I.G., and Hernandez-Galan, R.

(2002). The putative role of botrydial and related metabolites in the infection mechanism of

Botrytis cinerea. J. Chem. Ecol. **28,** 997-1005, doi/1015209817830.

Colombo, M., Raposo, G., and Thery, C. (2014). Biogenesis, secretion, and intercellular

interactions of exosomes and other extracellular vesicles. Annu. Rev. Cell Dev. Biol. **30,** 255-

289, doi/10.1146/annurev-cellbio-101512-122326.

Conesa, A., Gotz, S., Garcia-Gomez, J.M., Terol, J., Talon, M., and Robles, M. (2005).

Blast2GO: a universal tool for annotation, visualization and analysis in functional genomics

research. Bioinformatics **21,** 3674-3676, doi/10.1093/bioinformatics/bti610.

Corwin, J.A., Copeland, D., Feusier, J., Subedy, A., Eshbaugh, R., Palmer, C., Maloof, J.,

and Kliebenstein, D.J. (2016). The Quantitative Basis of the Arabidopsis Innate Immune

System to Endemic Pathogens Depends on Pathogen Genetics. PLoS Genet. **12,** e1005789,

doi/10.1371/journal.pgen.1005789.

Corwin, J.A., Subedy, A., Eshbaugh, R., and Kliebenstein, D.J. (2016). Expansive

Phenotypic Landscape of Botrytis cinerea Shows Differential Contribution of Genetic Diversity

and Plasticity. Mol. Plant Microbe Interact. **29,** 287-298, doi/10.1094/MPMI-09-15-0196-R.

Corwin, J.A. and Kliebenstein, D.J. (2017). Quantitative Resistance: More Than Just

Perception of a Pathogen. Plant Cell **29,** 655-665, doi/10.1105/tpc.16.00915.

Courtade, G., Wimmer, R., Røhr, ÅK., Preims, M., Felice, A.K.G., Dimarogona, M., Vaaje-

Kolstad, G., Sørlie, M., Sandgren, M., Ludwig, R., Eijsink, V.G.H., and Aachmann, F.L.

(2016). Interactions of a fungal lytic polysaccharide monooxygenase with β-glucan substrates

and cellobiose dehydrogenase. **113,** 5922, doi/10.1073/pnas.1602566113.

Couto, D. and Zipfel, C. (2016). Regulation of pattern recognition receptor signalling in plants.

Nat. Rev. Immunol. **16,** 537-552, doi/10.1038/nri.2016.77.

Cui, H., Qiu, J., Zhou, Y., Bhandari, D.D., Zhao, C., Bautor, J., and Parker, J.E. (2018).

Antagonism of Transcription Factor MYC2 by EDS1/PAD4 Complexes Bolsters Salicylic Acid

Defense in Arabidopsis Effector-Triggered Immunity. Mol. Plant. **11,** 1053-1066,

doi/10.1016/j.molp.2018.05.007.

Dalmais, B., Schumacher, J., Moraga, J., LE Pêcheur, P., Tudzynski, B., Collado, I.G., and

Viaud, M. (2011). The Botrytis cinerea phytotoxin botcinic acid requires two polyketide

synthases for production and has a redundant role in virulence with botrydial. **12,** 564,

doi/10.1111/j.1364-3703.2010.00692.x.

Dangl, J.L. and Jones, J.D. (2001). Plant pathogens and integrated defence responses to

infection. Nature **411,** 826-833, doi/10.1038/35081161.

Deighton, N., Muckenschnabel, I., Colmenares, A.J., Collado, I.G., and Williamson, B.

(2001). Botrydial is produced in plant tissues infected by Botrytis cinerea. Phytochemistry **57,**

689-692, doi/10.1016/S0031-9422(01)00088-7.

Denby, K.J., Kumar, P., and Kliebenstein, D.J. (2004). Identification of Botrytis cinerea

susceptibility loci in Arabidopsis thaliana. Plant J. **38,** 473-486, doi/10.1111/j.0960-

7412.2004.02059.x.

Ding Qiong and Zhang Jinlong (2012). seqRFLP: Simulation and visualization of restriction

enzyme cutting pattern from DNA sequences. **R package version 1.0.1.,** .

Fordyce, R.F., Soltis, N.E., Caseys, C., Gwinner, R., Corwin, J.A., Atwell, S., Copeland, D.,

Feusier, J., Subedy, A., Eshbaugh, R., and Kliebenstein, D.J. (2018). Digital Imaging

Combined with Genome-Wide Association Mapping Links Loci to Plant-Pathogen Interaction

Traits. Plant Physiol., doi/10.1104/pp.18.00851.

Fox, J. and Weisberg, S. (2011). An R Companion to Applied Regression, Second Edition ed.

(Sage, CA: Thousand Oaks).

Frerigmann, H., Pislewska-Bednarek, M., Sanchez-Vallet, A., Molina, A., Glawischnig, E.,

Gigolashvili, T., and Bednarek, P. (2016). Regulation of Pathogen-Triggered Tryptophan

Metabolism in Arabidopsis thaliana by MYB Transcription Factors and Indole Glucosinolate

Conversion Products. Mol. Plant. **9,** 682-695, doi/10.1016/j.molp.2016.01.006.

Gerbi, C., Bata, J., Breton, A., and Prensier, G. (1996). Glycoside and polysaccharide

hydrolase activity of the rumen anaerobic fungus Caecomyces communis (Sphaeromonas

communis SENSU ORPIN) at early and final stages of the developmental cycle. Curr. Microbiol.

32, 256-259, doi/10.1007/s002849900046.

Glazebrook, J. (2005). Contrasting mechanisms of defense against biotrophic and necrotrophic

pathogens. Annu. Rev. Phytopathol. **43,** 205-227, doi/10.1146/annurev.phyto.43.040204.135923.

Gotz, S., Garcia-Gomez, J.M., Terol, J., Williams, T.D., Nagaraj, S.H., Nueda, M.J.,

Robles, M., Talon, M., Dopazo, J., and Conesa, A. (2008). High-throughput functional

annotation and data mining with the Blast2GO suite. Nucleic Acids Res. **36,** 3420-3435,

doi/10.1093/nar/gkn176.

Govrin, E.M. and Levine, A. (2000). The hypersensitive response facilitates plant infection by

the necrotrophic pathogen Botrytis cinerea. Curr. Biol. **10,** 751-757, doi/10.1016/S0960-

9822(00)00560-1.

Greenberg, J.T., Guo, A., Klessig, D.F., and Ausubel, F.M. (1994). Programmed cell death in

plants: a pathogen-triggered response activated coordinately with multiple defense functions.

Cell **77,** 551-563, doi/10.1016/0092-8674(94)90217-8.

Igarashi, K., Uchihashi, T., Uchiyama, T., Sugimoto, H., Wada, M., Suzuki, K., Sakuda, S.,

Ando, T., Watanabe, T., and Samejima, M. (2014). Two-way traffic of glycoside hydrolase

family 18 processive chitinases on crystalline chitin. Nat. Commun. **5,** 3975,

doi/10.1038/ncomms4975.

Ingo Morgenstern, Justin Powlowski, and Adrian Tsang (2014). Fungal cellulose degradation

by oxidative enzymes: from dysfunctional GH61 family to powerful lytic polysaccharide

monooxygenase family. **13,** 471, doi/10.1093/bfgp/elu032.

Jones, J.D. and Dangl, J.L. (2006). The plant immune system. Nature **444,** 323-329,

doi/10.1038/nature05286.

Kanzaki, H., Yoshida, K., Saitoh, H., Fujisaki, K., Hirabuchi, A., Alaux, L., Fournier, E.,

Tharreau, D., and Terauchi, R. (2012). Arms race co-evolution of Magnaporthe oryzae AVR-

Pik and rice Pik genes driven by their physical interactions. Plant J. **72,** 894-907,

doi/10.1111/j.1365-313X.2012.05110.x.

Karasov, T.L., Kniskern, J.M., Gao, L., DeYoung, B.J., Ding, J., Dubiella, U., Lastra, R.O.,

Nallu, S., Roux, F., Innes, R.W., Barrett, L.G., Hudson, R.R., and Bergelson, J. (2014). The

long-term maintenance of a resistance polymorphism through diffuse interactions. Nature **512,**

436-440, doi/10.1038/nature13439.

Kliebenstein, D.J., Rowe, H.C., and Denby, K.J. (2005). Secondary metabolites influence

Arabidopsis/Botrytis interactions: variation in host production and pathogen sensitivity. Plant J.

44, 25-36, doi/10.1111/j.1365-313X.2005.02508.x.

Krishnakumar, V., Hanlon, M.R., Contrino, S., Ferlanti, E.S., Karamycheva, S., Kim, M.,

Rosen, B.D., Cheng, C.Y., Moreira, W., Mock, S.A., Stubbs, J., Sullivan, J.M., Krampis, K.,

Miller, J.R., Micklem, G., Vaughn, M., and Town, C.D. (2015). Araport: the Arabidopsis

information portal. Nucleic Acids Res. **43,** 1003, doi/10.1093/nar/gku1200.

Kubicek, C.P., Starr, T.L., and Glass, N.L. (2014). Plant cell wall-degrading enzymes and

their secretion in plant-pathogenic fungi. Annu. Rev. Phytopathol. **52,** 427-451,

doi/10.1146/annurev-phyto-102313-045831.

Kumar, R., Ichihashi, Y., Kimura, S., Chitwood, D.H., Headland, L.R., Peng, J., Maloof,

J.N., and Sinha, N.R. (2012). A High-Throughput Method for Illumina RNA-Seq Library

Preparation. Front. Plant. Sci. **3,** 202, doi/10.3389/fpls.2012.00202.

Lamesch, P., Dreher, K., Swarbreck, D., Sasidharan, R., Reiser, L., and Huala, E. (2010).

Using the Arabidopsis information resource (TAIR) to find information about Arabidopsis genes.

Curr. Protoc. Bioinformatics **Chapter 1,** Unit1.11, doi/10.1002/cpbi.36.

Lamesch, P., Berardini, T.Z., Li, D., Swarbreck, D., Wilks, C., Sasidharan, R., Muller, R.,

Dreher, K., Alexander, D.L., Garcia-Hernandez, M., Karthikeyan, A.S., Lee, C.H., Nelson,

W.D., Ploetz, L., Singh, S., Wensel, A., and Huala, E. (2012). The Arabidopsis Information

Resource (TAIR): improved gene annotation and new tools. Nucleic Acids Res. **40,** 1202,

doi/10.1093/nar/gkr1090.

Langfelder, P. and Horvath, S. (2008). WGCNA: an R package for weighted correlation

network analysis. BMC Bioinformatics **9,** 559, doi/10.1186/1471-2105-9-559.

Langmead, B., Trapnell, C., Pop, M., and Salzberg, S.L. (2009). Ultrafast and memory-

efficient alignment of short DNA sequences to the human genome. Genome Biol. **10,** r25. Epub

2009 Mar 4, doi/10.1186/gb-2009-10-3-r25.

Lanver, D., Muller, A.N., Happel, P., Schweizer, G., Haas, F.B., Franitza, M., Pellegrin, C.,

Reissmann, S., Altmuller, J., Rensing, S.A., and Kahmann, R. (2018). The Biotrophic

Development of Ustilago maydis Studied by RNA-Seq Analysis. Plant Cell **30,** 300-323,

doi/10.1105/tpc.17.00764.

Lee, H.J., Georgiadou, A., Walther, M., Nwakanma, D., Stewart, L.B., Levin, M., Otto,

T.D., Conway, D.J., Coin, L.J., and Cunnington, A.J. (2018). Integrated pathogen load and

dual transcriptome analysis of systemic host-pathogen interactions in severe malaria. Sci. Transl.

Med. **10,** 10.1126/scitranslmed.aar3619, doi/10.1126/scitranslmed.aar3619.

Leroch, M., Kleber, A., Silva, E., Coenen, T., Koppenhofer, D., Shmaryahu, A., Valenzuela,

P.D., and Hahn, M. (2013). Transcriptome profiling of Botrytis cinerea conidial germination

reveals upregulation of infection-related genes during the prepenetration stage. Eukaryot. Cell.

12, 614-626, doi/10.1128/EC.00295-12 [doi].

Li, H., Handsaker, B., Wysoker, A., Fennell, T., Ruan, J., Homer, N., Marth, G., Abecasis,

G., Durbin, R., and 1000 Genome Project Data Processing Subgroup (2009). The Sequence

Alignment/Map format and SAMtools. Bioinformatics **25,** 2078-2079,

doi/10.1093/bioinformatics/btp352.

Macheleidt, J., Mattern, D.J., Fischer, J., Netzker, T., Weber, J., Schroeckh, V., Valiante,

V., and Brakhage, A.A. (2016). Regulation and Role of Fungal Secondary Metabolites. Annu.

Rev. Genet. **50,** 371-392, doi/10.1146/annurev-genet-120215-035203.

McClure, R.S., Overall, C.C., Hill, E.A., Song, H.S., Charania, M., Bernstein, H.C.,

McDermott, J.E., and Beliaev, A.S. (2018). Species-specific transcriptomic network inference

of interspecies interactions. ISME J. **12,** 2011-2023, doi/10.1038/s41396-018-0145-6.

Mengiste, T., Chen, X., Salmeron, J., and Dietrich, R. (2003). The BOTRYTIS

SUSCEPTIBLE1 gene encodes an R2R3MYB transcription factor protein that is required for

biotic and abiotic stress responses in Arabidopsis. Plant Cell **15,** 2551-2565,

doi/10.1105/tpc.014167.

Mengiste, T. (2012). Plant immunity to necrotrophs. Annu. Rev. Phytopathol. **50,** 267-294,

doi/10.1146/annurev-phyto-081211-172955.

Mine, A., Seyfferth, C., Kracher, B., Berens, M.L., Becker, D., and Tsuda, K. (2018). The

Defense Phytohormone Signaling Network Enables Rapid, High-Amplitude Transcriptional

Reprogramming during Effector-Triggered Immunity. Plant Cell **30,** 1199-1219,

doi/10.1105/tpc.17.00970.

Mittal, S. and Davis, K.R. (1995). Role of the phytotoxin coronatine in the infection of

Arabidopsis thaliana by Pseudomonas syringae pv. tomato. Mol. Plant Microbe Interact. **8,** 165-

171.

Moraga, J., Dalmais, B., Izquierdo-Bueno, I., Aleu, J., Hanson, J.R., Hernandez-Galan, R.,

Viaud, M., and Collado, I.G. (2016). Genetic and Molecular Basis of Botrydial Biosynthesis:

Connecting Cytochrome P450-Encoding Genes to Biosynthetic Intermediates. ACS Chem. Biol.

11, 2838-2846, doi/10.1021/acschembio.6b00581.

Movahedi, S., Norey, C.G., Kay, J., and Heale, J.B. (1991). Infection and pathogenesis of cash

crops by Botrytis cinerea: primary role of an aspartic proteinase. **306,** 213.

Musungu, B.M., Bhatnagar, D., Brown, R.L., Payne, G.A., OBrian, G., Fakhoury, A.M.,

and Geisler, M. (2016). A Network Approach of Gene Co-expression in the Zea

mays/Aspergillus flavus Pathosystem to Map Host/Pathogen Interaction Pathways. Front. Genet.

7, 206, doi/10.3389/fgene.2016.00206.

Nelson, C.E., Attia, M.A., Rogowski, A., Morland, C., Brumer, H., and Gardner, J.G.

(2017). Comprehensive functional characterization of the glycoside hydrolase family 3 enzymes

from Cellvibrio japonicus reveals unique metabolic roles in biomass saccharification. Environ.

Microbiol. **19,** 5025-5039, doi/10.1111/1462-2920.13959.

Nobori, T., Velasquez, A.C., Wu, J., Kvitko, B.H., Kremer, J.M., Wang, Y., He, S.Y., and

Tsuda, K. (2018). Transcriptome landscape of a bacterial pathogen under plant immunity. Proc.

Natl. Acad. Sci. U. S. A. **115,** E3064, doi/10.1073/pnas.1800529115.

Okada, H., Ebhardt, H.A., Vonesch, S.C., Aebersold, R., and Hafen, E. (2016). Proteome-

wide association studies identify biochemical modules associated with a wing-size phenotype in

Drosophila melanogaster. Nat. Commun. **7,** 12649, doi/10.1038/ncomms12649.

Pages, H., Aboyoun, P., Gentleman, R., and DebRoy, S. (2017). Biostrings: String objects

representing biological sequences, and matching algorithms. **R package version 2.44.2.,** .

Pérez-Izquierdo, L., Morin, E., Maurice, J.P., Martin, F., Rincón, A., and Buée, M. (2017).

A new promising phylogenetic marker to study the diversity of fungal communities: The

Glycoside Hydrolase 63 gene. **17,** e11, doi/10.1111/1755-0998.12678.

Pieterse, C.M. and Van Loon, L.C. (2004). NPR1: the spider in the web of induced resistance

signaling pathways. Curr. Opin. Plant Biol. **7,** 456-464, doi/10.1016/j.pbi.2004.05.006.

Pinedo, C., Wang, C., Pradier, J., Dalmais, B., Choquer, M., Le Pêcheur, P., Morgant, G.,

Collado, I.G., Cane, D.E., and Viaud, M. (2008). Sesquiterpene synthase from the botrydial

biosynthetic gene cluster of the phytopathogen Botrytis cinerea. **3,** 791, doi/10.1021/cb800225v.

Porquier, A., Morgant, G., Moraga, J., Dalmais, B., Luyten, I., Simon, A., Pradier, J.M.,

Amselem, J., Collado, I.G., and Viaud, M. (2016). The botrydial biosynthetic gene cluster of

Botrytis cinerea displays a bipartite genomic structure and is positively regulated by the putative

Zn(II)2Cys6 transcription factor BcBot6. Fungal Genet. Biol. **96,** 33-46,

doi/10.1016/j.fgb.2016.10.003.

Porquier, A., Moraga, J., Morgant, G., Dalmais, B., Simon, A., Sghyer, H., Collado, I., and

Viaud, M. (2019). Botcinic acid biosynthesis in Botrytis cinerea relies on a subtelomeric gene

cluster surrounded by relics of transposons and is regulated by the Zn2Cys6 transcription factor

BcBoa13(), doi/10.1007/s00294-019-00952-4.

Poussereau, N., Creton, S., Billon-Grand, G., Rascle, C., and Fevre, M. (2001). Regulation of

acp1, encoding a non-aspartyl acid protease expressed during pathogenesis of Sclerotinia

sclerotiorum. Microbiology **147,** 717-726, doi/10.1099/00221287-147-3-717.

Poussereau, N., Gente, S., Rascle, C., Billon-Grand, G., and Fevre, M. (2001). aspS encoding

an unusual aspartyl protease from Sclerotinia sclerotiorum is expressed during

phytopathogenesis. FEMS Microbiol. Lett. **194,** 27-32, doi/S0378-1097(00)00500-0 [pii].

R Core Team (2014). R: A Language and Environment for Statistical Computing(Vienna,

Austria).

Robinson, M.D. and Smyth, G.K. (2008). Small-sample estimation of negative binomial

dispersion, with applications to SAGE data. Biostatistics **9,** 321-332,

doi/10.1093/biostatistics/kxm030.

Robinson, M.D. and Oshlack, A. (2010). A scaling normalization method for differential

expression analysis of RNA-seq data. Genome Biol. **11,** r25. Epub 2010 Mar 2, doi/10.1186/gb-

2010-11-3-r25.

Rodrigues, M.L., Godinho, R.M., Zamith-Miranda, D., and Nimrichter, L. (2015). Traveling

into Outer Space: Unanswered Questions about Fungal Extracellular Vesicles. PLoS Pathog. **11,**

e1005240, doi/10.1371/journal.ppat.1005240.

Rossi, F.R., Garriz, A., Marina, M., Romero, F.M., Gonzalez, M.E., Collado, I.G., and

Pieckenstain, F.L. (2011). The sesquiterpene botrydial produced by Botrytis cinerea induces the

hypersensitive response on plant tissues and its action is modulated by salicylic acid and

jasmonic acid signaling. Mol. Plant Microbe Interact. **24,** 888-896, doi/10.1094/MPMI-10-10-

0248.

Rowe, H.C. and Kliebenstein, D.J. (2007). Elevated genetic variation within virulence-

associated Botrytis cinerea polygalacturonase loci. Mol. Plant Microbe Interact. **20,** 1126-1137,

doi/10.1094/MPMI-20-9-1126.

Russell V. Lenth (2016). Least-Squares Means: The R Package lsmeans. **69,** 1-33,

doi/10.18637/jss.v069.i01.

Samuel, M., Bleackley, M., Anderson, M., and Mathivanan, S. (2015). Extracellular vesicles

including exosomes in cross kingdom regulation: a viewpoint from plant-fungal interactions.

Front. Plant. Sci. **6,** 766, doi/10.3389/fpls.2015.00766.

Schumacher, J., Pradier, J.M., Simon, A., Traeger, S., Moraga, J., Collado, I.G., Viaud, M.,

and Tudzynski, B. (2012). Natural variation in the VELVET gene bcvel1 affects virulence and

light-dependent differentiation in Botrytis cinerea. PLoS One **7,** e47840,

doi/10.1371/journal.pone.0047840.

Schumacher, J., Simon, A., Cohrs, K.C., Traeger, S., Porquier, A., Dalmais, B., Viaud, M.,

and Tudzynski, B. (2015). The VELVET Complex in the Gray Mold Fungus Botrytis cinerea:

Impact of BcLAE1 on Differentiation, Secondary Metabolism, and Virulence. Mol. Plant

Microbe Interact. **28,** 659-674, doi/10.1094/MPMI-12-14-0411-R.

Shannon, P., Markiel, A., Ozier, O., Baliga, N.S., Wang, J.T., Ramage, D., Amin, N.,

Schwikowski, B., and Ideker, T. (2003). Cytoscape: a software environment for integrated

models of biomolecular interaction networks. Genome Res. **13,** 2498-2504,

doi/10.1101/gr.1239303.

Sharma, H.C., Pampapathy, G., Dhillon, M.K., and Ridsdill-Smith, J.T. (2005). Detached

leaf assay to screen for host plant resistance to Helicoverpa armigera. J. Econ. Entomol. **98,** 568-

576, doi/10.1603/0022-0493-98.2.568.

Siewers, V., Viaud, M., Jimenez-Teja, D., Collado, I.G., Gronover, C.S., Pradier, J.,

Tudzynski, B., and Tudzynski, P. (2005). Functional analysis of the cytochrome P450

monooxygenase gene bcbot1 of Botrytis cinerea indicates that botrydial is a strain-specific

virulence factor. **18,** 602, doi/10.1094/MPMI-18-0602.

Soltis, N.E., Atwell, S., Shi, G., Fordyce, R., Gwinner, R., Gao, D., Shafi, A., and

Kliebenstein, D. (2018). Crop domestication and pathogen virulence: Interactions of tomato and

Botrytis genetic diversity. , doi/10.1101/255992.

Soltis, N.E., Zhang, W., Corwin, J.A., Atwell, S., and Kliebenstein, D.J. (2019). Pathogen

genetic control of transcriptome variation in the *Arabidopsis thaliana – Botrytis cinerea*

pathosystem. 577585, doi/10.1101/577585.

Stefanato, F.L., Abou-Mansour, E., Buchala, A., Kretschmer, M., Mosbach, A., Hahn, M.,

Bochet, C.G., Metraux, J.P., and Schoonbeek, H.J. (2009). The ABC transporter BcatrB from

Botrytis cinerea exports camalexin and is a virulence factor on Arabidopsis thaliana. Plant J. **58,**

499-510, doi/10.1111/j.1365-313X.2009.03794.x.

Stergiopoulos, I. and de Wit, P.J. (2009). Fungal effector proteins. Annu. Rev. Phytopathol. **47,**

233-263, doi/10.1146/annurev.phyto.112408.132637.

Stuart, J.M., Segal, E., Koller, D., and Kim, S.K. (2003). A gene-coexpression network for

global discovery of conserved genetic modules. Science **302,** 249-255,

doi/10.1126/science.1087447.

Swierzy, I.J., Handel, U., Kaever, A., Jarek, M., Scharfe, M., Schluter, D., and Luder,

C.G.K. (2017). Divergent co-transcriptomes of different host cells infected with Toxoplasma

gondii reveal cell type-specific host-parasite interactions. Sci. Rep. **7,** w, doi/10.1038/s41598-

017-07838-w.

Tang, D., Wang, G., and Zhou, J.M. (2017). Receptor Kinases in Plant-Pathogen Interactions:

More Than Pattern Recognition. Plant Cell **29,** 618-637, doi/10.1105/tpc.16.00891.

Tayal, P., Raj, S., Sharma, E., Kumar, M., Dayaman, V., Verma, N., Jogawat, A., Dua, M.,

Kapoor, R., and Johri, A.K. (2017). A Botrytis cinerea KLP-7 Kinesin acts as a Virulence

Determinant during Plant Infection. Sci. Rep. **7,** 5, doi/10.1038/s41598-017-09409-5.

ten Have, A., Dekkers, E., Kay, J., Phylip, L.H., and van Kan, J.A. (2004). An aspartic

proteinase gene family in the filamentous fungus Botrytis cinerea contains members with novel

features. Microbiology **150,** 2475-2489, doi/10.1099/mic.0.27058-0.

ten Have, A., Espino, J.J., Dekkers, E., Van Sluyter, S.C., Brito, N., Kay, J., Gonzalez, C.,

and van Kan, J.A. (2010). The Botrytis cinerea aspartic proteinase family. Fungal Genet. Biol.

47, 53-65, doi/10.1016/j.fgb.2009.10.008.

Tien-chye Tan, Daniel Kracher, Rosaria Gandini, Christoph Sygmund, Roman Kittl,

Dietmar Haltrich, B Martin Hällberg, Roland Ludwig, and Christina Divne (2015).

Structural basis for cellobiose dehydrogenase action during oxidative cellulose degradation. **6,**

7542, doi/10.1038/ncomms8542.

Toruno, T.Y., Stergiopoulos, I., and Coaker, G. (2016). Plant-Pathogen Effectors: Cellular

Probes Interfering with Plant Defenses in Spatial and Temporal Manners. Annu. Rev.

Phytopathol. **54,** 419-441, doi/10.1146/annurev-phyto-080615-100204.

Tsuda, K. and Katagiri, F. (2010). Comparing signaling mechanisms engaged in pattern-

triggered and effector-triggered immunity. Curr. Opin. Plant Biol. **13,** 459-465,

doi/10.1016/j.pbi.2010.04.006.

Urlacher, V.B. and Girhard, M. (2012). Cytochrome P450 monooxygenases: an update on

perspectives for synthetic application. Trends Biotechnol. **30,** 26-36,

doi/10.1016/j.tibtech.2011.06.012.

Van Kan, J.A., Stassen, J.H., Mosbach, A., Van Der Lee, T A, Faino, L., Farmer, A.D.,

Papasotiriou, D.G., Zhou, S., Seidl, M.F., Cottam, E., Edel, D., Hahn, M., Schwartz, D.C.,

Dietrich, R.A., Widdison, S., and Scalliet, G. (2017). A gapless genome sequence of the

fungus Botrytis cinerea. Mol. Plant. Pathol. **18,** 75-89, doi/10.1111/mpp.12384.

Van Vu, B., Itoh, K., Nguyen, Q.B., Tosa, Y., and Nakayashiki, H. (2012). Cellulases

belonging to glycoside hydrolase families 6 and 7 contribute to the virulence of Magnaporthe

oryzae. Mol. Plant Microbe Interact. **25,** 1135-1141, doi/10.1094/MPMI-02-12-0043-R.

Venables, W.N. and Ripley, B.D. (2002). Modern Applied Statistics with S, Fourth Edition ed.

(New York: Springer).

Walker, A.S., Gladieux, P., Decognet, V., Fermaud, M., Confais, J., Roudet, J., Bardin, M.,

Bout, A., Nicot, P.C., Poncet, C., and Fournier, E. (2015). Population structure and temporal

maintenance of the multihost fungal pathogen Botrytis cinerea: causes and implications for

disease management. Environ. Microbiol. **17,** 1261-1274, doi/10.1111/1462-2920.12563.

Walz, A., Zingen-Sell, I., Loeffler, M., and Sauer, M. (2008). Expression of an oxalate oxidase

gene in tomato and severity of disease caused by Botrytis cinerea and Sclerotinia sclerotiorum.

57, 453-458, doi/10.1111/j.1365-3059.2007.01815.x.

Wang, C.M., Hopson, R., Lin, X., and Cane, D.E. (2009). Biosynthesis of the sesquiterpene

botrydial in Botrytis cinerea. Mechanism and stereochemistry of the enzymatic formation of

presilphiperfolan-8beta-ol. J. Am. Chem. Soc. **131,** 8360-8361, doi/10.1021/ja9021649.

Weiberg, A., Wang, M., Lin, F.M., Zhao, H., Zhang, Z., Kaloshian, I., Huang, H.D., and

Jin, H. (2013). Fungal small RNAs suppress plant immunity by hijacking host RNA interference

pathways. Science **342,** 118-123, doi/10.1126/science.1239705.

Westermann, A.J., Barquist, L., and Vogel, J. (2017). Resolving host-pathogen interactions by

dual RNA-seq. PLoS Pathog. **13,** e1006033, doi/10.1371/journal.ppat.1006033.

Williamson, B., Tudzynski, B., Tudzynski, P., and van Kan, J.A. (2007). Botrytis cinerea: the

cause of grey mould disease. Mol. Plant. Pathol. **8,** 561-580, doi/10.1111/j.1364-

3703.2007.00417.x.

Windram, O., Madhou, P., McHattie, S., Hill, C., Hickman, R., Cooke, E., Jenkins, D.J.,

Penfold, C.A., Baxter, L., Breeze, E., Kiddle, S.J., Rhodes, J., Atwell, S., Kliebenstein, D.J.,

Kim, Y.S., Stegle, O., Borgwardt, K., Zhang, C., Tabrett, A., Legaie, R., Moore, J.,

Finkenstadt, B., Wild, D.L., Mead, A., Rand, D., Beynon, J., Ott, S., Buchanan-Wollaston,

V., and Denby, K.J. (2012). Arabidopsis defense against Botrytis cinerea: chronology and

regulation deciphered by high-resolution temporal transcriptomic analysis. Plant Cell **24,** 3530-

3557, doi/10.1105/tpc.112.102046.

Xie, D.X., Feys, B.F., James, S., Nieto-Rostro, M., and Turner, J.G. (1998). COI1: an

Arabidopsis gene required for jasmonate-regulated defense and fertility. Science **280,** 1091-

1094, doi/10.1126/science.280.5366.1091.

Xu, J., Meng, J., Meng, X., Zhao, Y., Liu, J., Sun, T., Liu, Y., Wang, Q., and Zhang, S.

(2016). Pathogen-Responsive MPK3 and MPK6 Reprogram the Biosynthesis of Indole

Glucosinolates and Their Derivatives in Arabidopsis Immunity. Plant Cell **28,** 1144-1162,

doi/10.1105/tpc.15.00871.

Xu, L., Liu, F., Lechner, E., Genschik, P., Crosby, W.L., Ma, H., Peng, W., Huang, D., and

Xie, D. (2002). The SCF(COI1) ubiquitin-ligase complexes are required for jasmonate response

in Arabidopsis. Plant Cell **14,** 1919-1935, doi/10.1105/tpc.003368.

Zamocky, M., Ludwig, R., Peterbauer, C., Hallberg, B.M., Divne, C., Nicholls, P., and

Haltrich, D. (2006). Cellobiose dehydrogenase--a flavocytochrome from wood-degrading,

phytopathogenic and saprotropic fungi. Curr. Protein Pept. Sci. **7,** 255-280,

doi/10.2174/138920306777452367.

Zhang, B. and Horvath, S. (2005). A general framework for weighted gene co-expression

network analysis. Stat. Appl. Genet. Mol. Biol. **4,** 6115.1128. Epub 2005 Aug 12,

doi/10.2202/1544-6115.1128.

Zhang, W., Kwon, S.T., Chen, F., and Kliebenstein, D.J. (2016). Isolate Dependency of

Brassica rapa Resistance QTLs to Botrytis cinerea. Front. Plant. Sci. **7,** 161,

doi/10.3389/fpls.2016.00161.

Zhang, W., Corwin, J.A., Copeland, D., Feusier, J., Eshbaugh, R., Chen, F., Atwell, S., and

Kliebenstein, D.J. (2017). Plastic Transcriptomes Stabilize Immunity to Pathogen Diversity:

The Jasmonic Acid and Salicylic Acid Networks within the Arabidopsis/Botrytis Pathosystem.

Plant Cell **29,** 2727-2752, doi/10.1105/tpc.17.00348.