

**Network analysis of publicly available RNA-seq data provides insights into the molecular mechanisms of plant defense against multiple fungal pathogens in *Arabidopsis thaliana***

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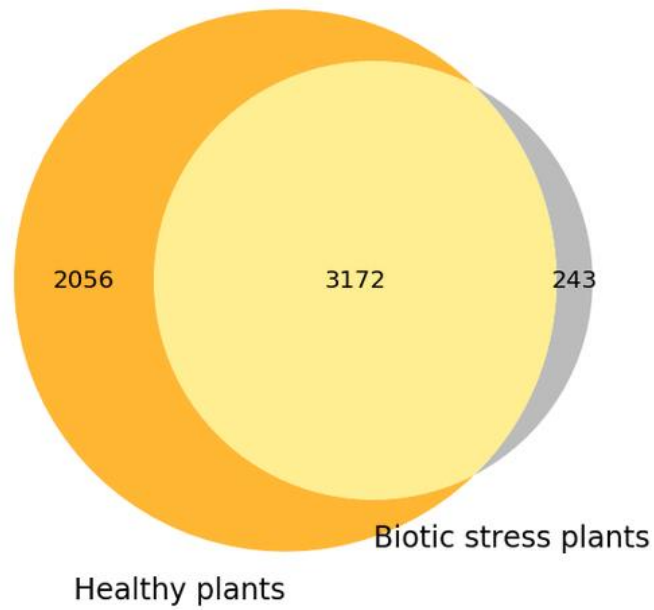
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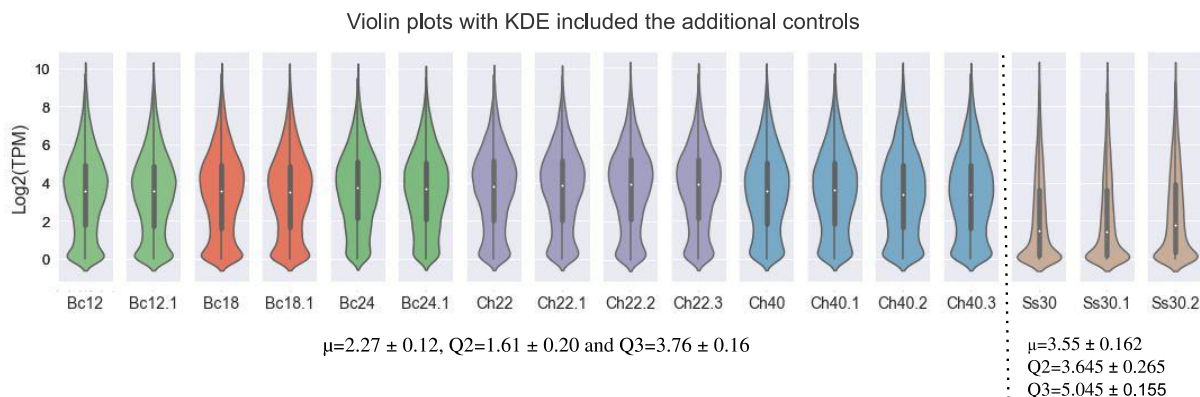
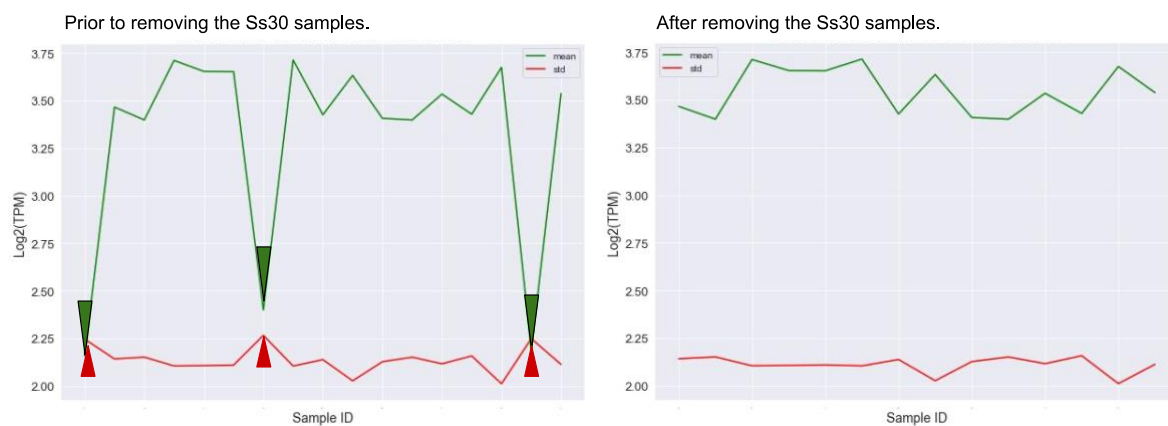
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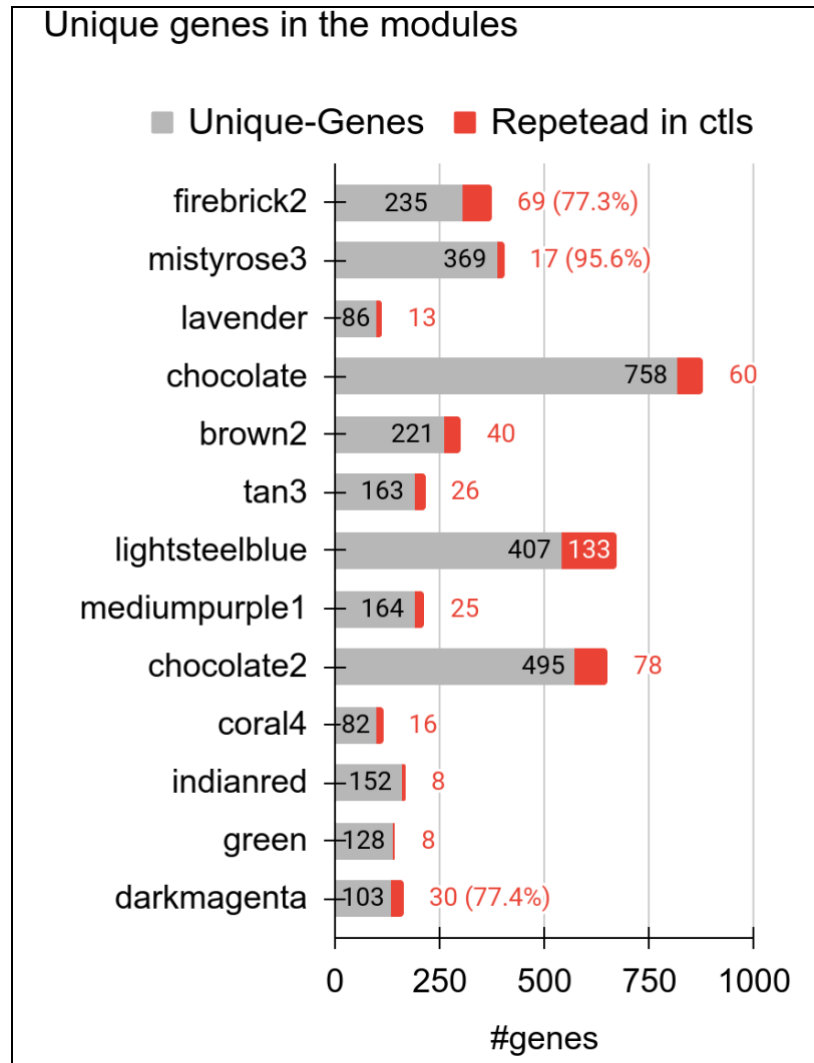
## Supplemental Material



**Figure S1. Venn diagrams for non-expressed genes.** Expression values equal to zero were removed from both control sample and stress-treated sample datasets.

**(A)****(B)****(C)**

**Figure S2. Distributions of atypical samples in stress-treated gene networks.** (A) Distribution of stress-treated samples without including the Ss30s samples, (B) Distribution of the atypical samples, (C) Mean (green line) and the standard deviation (red line) before and after removing the Ss30, Ss30.1 and Ss30.2 samples from the stress-treated dataset.



**Figure S3. Unique genes identified in infected-plants gene modules.** The gray bars show the number of genes that were unique in each gene module. The red bar show the number of repeated genes found in the healthy-control network when comparing both gene co-expression networks.

**Table S1.** Sample information obtained from NCBI SRA.

Sample	Treatment	HPI	Sample ID	BioProject	Reference
<i>Arabidopsis thaliana</i> (Col-0) leaves	<i>Colletotrichum higginsianum</i>	22	Ch22	PRJNA148307	[1,2]
		40	Ch40		
	<i>Sclerotinia sclerotiorum</i>	30	Ss30	PRJNA418121	[3]
	<i>Botrytis cinerea</i>	12	Bc12	PRJNA315516 PRJNA593073	Utrecht University (2016) Beijing normal University (2019)
		18	Bc18		
		24	Bc24		
	Control	12	Healthy12	PRJNA315516 PRJNA418121	[1]
		18	Healthy18		
		24	Healthy24		
		30	Healthy30		

HPI: hours post-inoculation

**Table S2.** Databases, bioinformatic tools and source code.

Databases to extract RNA-Seq data and complementary data		
DB OR FILE NAME	DESCRIPTION	WEBSITE
SRA-NCBI DB	Bulk RNASeq data	https://www.ncbi.nlm.nih.gov/sra
TAIR10 Genome Fasta File	TAIR10 genome release 2010	https://www.arabidopsis.org/download/index-auto.jsp?dir=/download_files/Sequences
Araport11 Genome Annotation	Genome Annotation Release 2016	https://www.arabidopsis.org/download/index-auto.jsp?dir=/download_files/Sequences
Bioinformatic tools to estimate the RNA-Seq raw-counts		
TOOL NAME	DESCRIPTION	WEBSITE
SRA-Toolkit	To download SRA accessions	Toolkithttps://hpc.nih.gov/apps/sratoolkit.html
FastQC	To perform quality analysis	https://www.bioinformatics.babraham.ac.uk
Trimmomatic	To preprocessing fq files	http://www.usadellab.org/cms/?page=trimmomat
STAR	To perform read alignments	https://github.com/alexdobin/STAR
HT-Seq-qa HT-Seq-count	To test SAM files quality To estimate raw-counts	https://htseq.readthedocs.io/en/master/
4-Step methodology for RNA-Seq integration. Python (v3.8.10) <a href="https://github.com/cyntsc/RNA-Seq-raw-integration">https://github.com/cyntsc/RNA-Seq-raw-integration</a> . DOI 10.5281/zenodo.7076416		
STEP 1: Raw-count integration	1_Step1_integrating_raw_counts (integrate and clean datasets) Venn_diagram_genes_in_ceros.ipynb (compare TT vs CL repressed genes)	
STEP 2: TPM normalization	2_Step2_TPM_normalization.ipynb (normalize datasets) *(prepare a gene length file)	
STEP 3: Data standardization	3_Step3_TPM_standardization.ipynb (cut off extreme and underrepresented values)	
STEP 4: Data log transformation and atypical identification	4_Step4_Log2_scale.ipynb (reduce scale representation and identify source of noise)	
Script to extract gene lengths for step 2.	Gene_length_extraction_from_GTF.ipynb (extract gene lengths for normalization)	
Script to extract gene modules for annotation tasks	6_modules_percentual_differentiation.ipynb	
Network implementation and gene module identification. R (v4.1.0) <a href="https://github.com/cyntsc/RNA-Seq-raw-integration">https://github.com/cyntsc/RNA-Seq-raw-integration</a> . DOI 10.5281/zenodo.7076416		
01_Healthy_SignedNtw_D.R 01_Infected_SignedNtw_E.R 02_Healthy_GS_MM_24hpi.R 02_Infected_GS_MM_24hpi.R	Network implementation for a signed-ntw for the healthy plants Network implementation for a signed-ntw for the infected plants Gene-Significance and Module-Membership for the healthy network Gene-Significance and Module-Membership for the infected network	
Gene Ontology annotation		
Panther v17.0 (Released 20221013)	http://pantherdb.org/webservices/go/overrep.jsp	

**Table S3.** Significant gene modules identified in the healthy-control and fungal-infected samples.

Condition	Module Name	Cluster Size	R <sup>2</sup>
Control samples gene networks	coral3 *	2024	+0.98
	blue3 *	79	+0.96
	navajowhite3 *	492	+0.79
	magenta2	204	-0.98
	darkolivegreen4	1792	-0.97
	antiquewhite	764	-0.88
Stress-treated gene networks	chocolate *	818	+0.98
	dodgerblue1 *	72	+0.96
	green3 *	472	+0.95
	chocolate2 *	573	+0.91
	tomato2	218	-0.94
	palevioletred1	368	-0.93
	green	236	-0.90
	deepskyblue	39	-0.80

\* Gene modules with positive R<sup>2</sup> were selected to perform the Gene Ontology overrepresentation gene test.

**Table S4.** Gene Ontology overrepresentation test summary.

	Module name	#Genes	GO Category	Mapped genes	Classified	Unclassified
Control samples gene networks	Coral3	2024	MF	1991	726	1265
	navajowhite3	492	BP	489	168	321
	Blue3	79	BP	79	76	3
Stress-treated gene networks	Chocolate	818	MF	805	247	558
	Chocolate2	573	MF	566	180	386
	Green3	472	MF	456	118	338
	Dodgerblue1	72	None	70	NA	NA

MF: Molecular function, BP: Biological Process

**Table S5.** Genes functional groups related to multiple fungal infections identified by DAVID

Group Description	Number of genes (including duplicates)	EASE Score* (>0.8)
Cell wall / extracellular region	28	0.42835
Developmental protein	23	0.59716
Glycosyltransferase	34	<b>1.34897</b>
Methyltransferase	13	<b>0.87954</b>
Microtubule binding	11	0.77645
Protein kinase activity	85	0.44928
Protein transport	23	<b>1.11126</b>
Regulation of transcription	75	0.31824
Sterol metabolism	22	<b>1.35113</b>
WD40 repeat / G-protein	28	<b>1.55271</b>
Zinc finger	17	<b>1.01733</b>
mRNA binding	14	0.62084
LRR Receptors	15	0.52845

\*Scores highlighted in bold show the groups selected in this study.

## References

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2. Robin, G.P.; Kleemann, J.; Neumann, U.; Cabre, L.; Dallery, J.-F.; Lapalu, N.; O'Connell, R.J. Subcellular Localization Screening of *Colletotrichum Higginsianum* Effector Candidates Identifies Fungal Proteins Targeted to Plant Peroxisomes, Golgi Bodies, and Microtubules. *Front. Plant Sci.* **2018**, *9*, 562, doi:10.3389/fpls.2018.00562.
3. Badet, T.; Voisin, D.; Mbengue, M.; Barascud, M.; Sucher, J.; Sadon, P.; Balagué, C.; Roby, D.; Raffaele, S. Parallel Evolution of the POQR Prolyl Oligo Peptidase Gene Conferring Plant Quantitative Disease Resistance. *PLOS Genetics* **2017**, *13*, e1007143, doi:10.1371/journal.pgen.1007143.