1	Exploring the genetic basis of gene transcript abundance and
2	metabolite level in loblolly pine (Pinus taeda L.) using
3	association mapping and network construction
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22	

23 Abstract

24 Gene transcripts and metabolites are important regulatory checkpoints between 25 genetic variation and complex biological processes such as wood development and 26 drought response in conifers. Loblolly pine (Pinus taeda L.) is one of the most 27 commonly planted forest tree species in the southern U.S. In this study, we tested for 28 associations between 2.8 million exome-derived SNPs and the transcript abundance of 29 110 wood development genes, 88 disease or drought related genes as well as levels of 30 82 known metabolites. We identified 1841 SNPs associated with 191 gene expression 31 phenotypes and 524 SNPs associated with 53 metabolite level phenotypes. The 32 identified SNPs reside in genes with a wide variety of functions. We further integrated 33 the identified SNPs and their associated expressed genes and metabolites into 34 networks. We described the SNP-SNP interactions that significantly impacted the 35 gene transcript abundance and metabolite level in the networks. The key loci and 36 genes in the wood development and drought response networks were identified and 37 analyzed. This work provides candidate genes for research on the genetic basis of 38 gene expression and metabolism linked to wood development and drought response in 39 loblolly pine, and highlights the efficiency of using association-mapping-based 40 networks to discover candidate genes with important roles in complex biological 41 processes. 42 43 Keywords: gene expression, metabolism, epistasis, stress response, wood development, **SNP** 44

46 Introduction

Understanding the genetic basis of complex traits in the important forest tree species, 47 48 loblolly pine (Pinus taeda L.), can contribute to the improvement of its growth and 49 quality. The majority of previous genetic studies have focused on the dissection of 50 adaptive or commercially important traits like growth, wood properties, or drought 51 tolerance (Neale and Savolainen 2004; González-Martínez et al. 2007; Cumbie et al. 52 2011; Westbrook et al. 2013), while only a few studies have sought to characterize 53 phenotypes in depth by surveying the levels of transcripts and metabolites associated 54 with such traits of interest. Palle et al. (2011) analyzed expression of genes involved 55 in loblolly pine wood development and reported key regulatory genes. A total of 33 56 wood development gene expression phenotypes were associated with 80 single 57 nucleotide polymorphisms (SNPs). Seeve (2010) detected the expression of 88 genes 58 related to disease or drought responses in loblolly pine and found that 27 expression 59 phenotypes were associated with 94 SNPs. Eckert et al. (2012) detected multiple 60 SNP-metabolite associations in loblolly pine. 61 Gene transcript abundance and metabolite levels are complex intermediate 62 phenotypes that link genetic variations to whole-plant phenotypes. Each is regulated 63 by genetic and environmental cues, and perturbations in these intermediate 64 phenotypes may be manifested as changes in higher-order traits (Schadt et al. 2008). 65 Thus, studies linking gene expression or metabolite phenotypes to genetic variations may enhance our understanding of the molecular mechanisms that underlie broader 66

67	whole-plant phenotypes. For example, Bossu et al. (2016) found secondary
68	metabolites influence wood properties. Obata et al. (2015) demonstrated that
69	metabolite levels in maize respond to stress conditions and can be used to predict the
70	grain yield under drought. Furthermore, integrating SNPs and their associated
71	gene expression and metabolite level phenotypes into networks aids in
72	connecting the two phenotypes, and in identifying key genes in regulatory
73	networks that contribute to adaptive traits (Wentzell et al. 2007; Burkhardt et al.
74	2015).
75	To gain insights into the regulatory mechanism underlying wood development and
76	disease and drought responses, we tested for associations between 2.8 million SNPs
77	derived from exome target sequencing and gene transcript abundance and metabolite
78	levels. The expression data includes 110 wood development genes and 88 disease or
79	drought related genes. The metabolite data includes 82 metabolites with known names.
80	We constructed networks to analyze the loci associated with multiple phenotypes.
81	Since epistatic interaction between loci is another factor that may further influence
82	phenotypes in loblolly pine (Lu et al. 2017), the SNP-SNP interactions were also
83	detected among the identified loci. The identified genes are valuable resources to
84	study the genetic basis of gene expression and metabolite level phenotypes linked to
85	complex biological processes in loblolly pine.
86	

88 Materials and methods

89 Plant material and genotypic data

90 The loblolly pine population used in this study was originally established for the 91 Allele Discovery of Economic Pine Traits 2 (ADEPT2) project and included trees 92 with parents from a wide range across the southeastern U.S. (Eckert et al. 2010a; 93 Cumbie et al. 2011). Genotypic data were obtained for 375 trees in this population 94 (Lu et al. 2016). The NimbleGen SeqCap EZ system (Roche NimbleGen, Inc., 95 Madison, WI) was used to capture and enrich the exome of each tree. The detailed 96 procedures of probe design, raw SNP detection and genotyping were described in Lu 97 et al. (2016). The raw SNPs were filtered, accepting only bi-allelic sites with at least 98 5X sequencing depth for all of the individuals without missing data and a minor allele 99 frequency (MAF) ≥ 0.01 . A total of 2,822,609 SNPs were retained, and a total of 100 94,478 haplotype blocks were detected for this population (Lu et al. 2017). 101 Additionally, 23 simple sequence repeat (SSR) markers have been used to genotype 102 ADEPT2 trees (Eckert et al. 2010a). SSR genotype data were used for estimating 103 covariates to adjust for the selectively neutral population structure.

104

105 Phenotypic data

106 Abundance of functional gene transcripts and levels of metabolites were analyzed in

107 this study. Relative transcript abundance was measured using reverse transcription

108 quantitative polymerase chain reaction (RT-qPCR). Palle et al. (2011) analyzed the

109	expression of 111 genes with probable roles in xylem/wood development in woody
110	tissue collected from 475 trees. Seeve (2010) detected the expression of 88 disease or
111	drought responsive genes in woody tissue collected from 354 trees. However, only
112	278 trees with gene expression data were genotyped for this study. Therefore, 278
113	trees were used for association tests with expression data for 199 genes. The gene
114	expression phenotypes from the two data sets were organized into seven functional
115	groups based on the biological processes which they were involved: genes related to
116	reactive oxygen species (ROS) biosynthesis and signaling, terpenoid biosynthesis,
117	programmed cell death (PCD), phenylpropanoid pathway, wood-related,
118	disease-related, and drought-related genes. The genes in each group were further
119	assigned to sub-groups (see Table S1 available as Supplementary Data at Tree
120	Physiology Online). Metabolite data were obtained from the study of Eckert et al.
121	(2012). They measured the concentration of 292 metabolites in woody tissue of
122	ADEPT2 trees. In this study, we only used data of the 82 metabolites with known
123	names. Only 212 of the trees with metabolite data were genotyped for this study.
124	Therefore, 212 trees were used for association tests with concentration data for 82
125	metabolites.
126	

127 Association analyses

Association analyses for the individual SNPs and phenotypes were conducted using
TASSEL 5.0 (Bradbury et al. 2007). The SSR genotype data were used for estimating

130	covariates to adjust for the selectively neutral population structure. The SSR
131	genotypes were available for 195 of the trees used for the gene expression analysis
132	and 196 of the trees used for the metabolite concentration analysis. We used this
133	group of trees (named as the str population) for a population structure analysis.
134	Population structure within this group was mainly due to the Mississippi River
135	discontinuity (Lu et al. 2016). We named the trees from east of the Mississippi River,
136	namely 223 trees used for gene expression analysis and 184 trees used for metabolite
137	concentration analysis, as the <i>east</i> population. Therefore, three populations: <i>total</i> (N =
138	278), east (N = 223) and str (N = 195) populations, were used to perform association
139	analyses for the gene expression data. Three populations, <i>total</i> ($N = 212$), <i>east</i> ($N =$
140	184) and str (N = 196), were used to perform association analyses for the metabolite
141	concentration data. For the total and east populations, the simple general linear model
142	(GLM) method (S model) and the mixed linear model (MLM) method incorporating a
143	kinship matrix (K model) were applied. For the str population, in addition to the S and
144	K models, the GLM incorporating the covariate to adjust for population structure (Q
145	model) and the MLM incorporating both the kinship matrix and population structure
146	covariate (QK model) were applied. The population structure covariate was estimated
147	using the software STRUCTURE (Pritchard et al. 2000; Hubisz et al. 2009) and 23
148	SSR markers. A kinship matrix for each population was estimated by TASSEL 5.0
149	(Bradbury et al. 2007) using the 2.8 million SNP markers. The kinship relatedness is
150	low in this population with an average range between -0.03 and 0.10 (excluding the

151	self-relatedness). Quantile-quantile plots were generated for observed against
152	expected $-\log_{10}P$ to examine the model fitness, where observed <i>P</i> -values were
153	obtained from association mapping and expected P-values from the assumption that
154	no association occurred between marker and trait. Significance of associations
155	between loci and traits were determined by the P-values. A corrected Bonferroni
156	threshold 0.05/94,478=5.29E-7, where 94,478 was the estimated number of haplotype
157	blocks, was applied to screen for significant loci. The squared correlation coefficient
158	(R^2) between genotypes on the same scaffold was used as an LD measure and
159	calculated using the "geno-r2" function in the VCFtools software (Danecek et al.
160	2011). The triangular heatmaps were produced using the R package "LDheatmap"
161	(Shin et al. 2006; R Core Team 2017).
162	

163 Annotation of genes that contained SNPs associated with traits

- 164 The VCF tools software (Danecek et al. 2011) was used to calculate the minor allele
- 165 frequencies (MAFs) and perform Hardy-Weinberg Equilibrium (HWE) tests for the
- 166 identified SNPs. Annotation of the genes containing the identified SNPs was obtained
- 167 from loblolly pine Gene Annotation v3.0
- 168 (<u>https://treegenesdb.org/FTP/Genomes/Pita/v1.01/annotation</u>) (Wegrzyn et al. 2014).
- 169 Very few regulatory sequences such as promoters, enhancers and silencers have been
- 170 identified in the loblolly pine reference genome. SNPs within 5000 bp downstream or
- 171 upstream of a gene were considered to be within a putative regulatory sequence of the

172	gene. If a SNP	was located in a	region without	annotation. th	e flanking sec	juence 1500
	0					

- bp upstream and downstream of the SNP was used as a query to do a blastx search
- against the entire National Center for Biotechnology Information (NCBI)
- 175 non-redundant (nr) protein database (<u>http://blast.ncbi.nlm.nih.gov/Blast.cgi</u>). The
- 176 NCBI GI numbers of candidate genes were uploaded to the "Gene List Analysis" tool
- 177 in the PANTHER Classification System (<u>http://www.pantherdb.org</u>) (Mi et al. 2013;
- 178 Mi et al. 2016). The genes were mapped to the PANTHER databases and analyzed for
- their classification according to their molecular functions and protein classes.
- 180

181 Network plots and SNP-SNP interaction analyses

- 182 To visualize the relationships between SNPs and their associated phenotypes, R
- 183 package "igraph" was used to plot the networks (Csardi and Nepusz 2006; R Core
- 184 Team 2017). Blue, yellow and pink nodes represent SNPs, gene expression
- 185 phenotypes and metabolite level phenotypes, respectively. Red and gray edges
- 186 represent the significant SNP-metabolite-level and SNP-gene-expression associations.
- 187 In addition, for the SNPs in the networks, the epistatic SNP-SNP interaction test was
- 188 implemented using PLINK 1.9 (Purcell et al. 2007). The Bonferroni correction was
- applied to screen for significant SNP-SNP interactions. In the networks, purple edges
- 190 represent the significant SNP-SNP interactions.
- 191
- 192

193 **Results**

194 Significant associations between SNPs and phenotypes

195 Association analyses of 2.8 million SNPs with 199 gene expression phenotypes and

196 82 metabolite level phenotypes were conducted. After summarizing the results from *S*,

197 K, Q and QK models, a total of 2,562 associations between 1,841 SNPs and 191 gene

- expression phenotypes and 524 associations between 524 SNPs and 53 metabolite
- 199 concentration phenotypes were identified (see Tables S2 & S3 available as
- 200 Supplementary Data at *Tree Physiology* Online). A total of 40 % and 23 % of the
- 201 SNPs associated with gene expression and metabolite concentration phenotypes,

respectively, had a MAF \geq 0.05. The MAFs of other SNPs were between 0.01 and

- 203 0.05. There were 9 % of the SNPs associated with gene expression and 6 % of the
- 204 SNPs associated with metabolite concentrations that departed from HWE. Among the
- 205 2562 gene expression associations, 1195 (47 %) were related to expression of wood

development genes, 661 (26 %) to drought-related genes, 232 (9 %) to terpenoid

- biosynthesis genes, 162 (6 %) to PCD genes, 110 (4 %) to ROS genes, 104 (4 %) to
- 208 phenylpropanoid pathway genes and 98 (4 %) to disease related genes. Expression of
- 209 the CYPB gene (involved in terpenoid biosynthesis) was associated with the largest
- 210 number of SNPs (181 SNPs). It was followed by the genes encoding ATAF-1 [a
- drought-responsive transcription factor (TF), 138 SNPs], RAP2.1 (a
- drought-responsive TF, 133 SNPs), CS-5828 (cellulose synthase-like, 128 SNPs),
- 213 CslA1 (cell wall- related, 117 SNPs), PtEMB4 (a late embryogenesis abundant protein,

214	114 SNPs), <i>αtub1</i> (α-tubulin, 105 SNPs), <i>ANR</i> (involved in phenylpropanoid pathway,
215	76 SNPs), PtMLO2 (involved in PCD, 75 SNPs), NCED (related to drought signaling,
216	73 SNPs), PtMLO1 (involved in PCD, 74 SNPs), CesA2 (a cellulose and callose
217	synthase, 66 SNPs), RP-L2 (a wood development protein, 62 SNPs), and CaS3 (a
218	cellulose and callose synthase, 57 SNPs). For levels of the metabolites glucose and
219	melezitose, each were associated with 30 SNPs. They were followed by
220	3,4-dihydroxybenzoic acid (24 SNPs), glycerol-3-galactoside (22 SNPs), glycine (21
221	SNPs), and raffinose (21 SNPs). Complete lists of the identified SNPs and their
222	associated phenotypes were presented in Tables S4 & S5 (available as Supplementary
223	Data at Tree Physiology Online).
224	The SNP-trait r^2 values in the association outputs represent the proportion of
225	phenotypic variation that is explained by the corresponding markers. The median of r^2
226	values was 0.15 for both gene expression and metabolite level associations. However,
227	the r^2 values of gene expression associations had a wide range, from 0.09 to 0.85,
228	while the r^2 values of metabolite level associations ranged from 0.11 to 0.22 (see
229	Figure S1 available as Supplementary Data at Tree Physiology Online). We examined
230	the 323 gene expression associations with high r^2 values (> 0.40). A total of 181 were
231	associated with the CYPB gene involved in terpenoid biosynthesis, 133 with the
232	RAP2.1 gene encoding a drought-responsive TF, 4 with the PtMLO1 gene involved in
233	PCD, 2 with the <i>PtGPX3</i> gene (a peroxidase), 2 with the <i>CesA2</i> gene (a cellulose and
234	callose synthase), and 1 with the CaS3 gene (a cellulose and callose synthase).

235	In the previous association studies on the ADEPT2 population, nearly 4000
236	EST-derived SNPs were associated with metabolite level and gene expression
237	phenotypes (Seeve 2010; Eckert et al. 2012; Palle et al. 2013). To cross-reference
238	associated SNPs identified in the current study with associated SNPs in the three prior
239	studies, we mapped the sequences with previously identified SNPs to loblolly pine
240	reference assembly v1.01 (https://treegenesdb.org/FTP/Genomes/Pita/v1.01) using the
241	GMAP software (Wu and Watanabe 2005). We found that the SNPs
242	scaffold596656_40783 and tscaffold2197_12732 discovered in the current study
243	reside also in sequences identified in the prior study. The SNP scaffold596656_40783
244	was associated with expression of the CAD1 gene (encoding cinnamyl-alcohol
245	dehydrogenase involved in a lignin biosynthesis). This SNP resides in a gene
246	encoding cystathionine gamma-synthase. The SNP tscaffold2197_12732 was
247	associated with expression of the CesA2 gene (encoding a cellulose and callose
248	synthase). This SNP resides in a gene encoding E3 ubiquitin-protein ligase. These two
249	associations are consistent with the associations reported by Palle et al. (2013). Other
250	identified SNPs in this study could not be mapped to the sequences identified in prior
251	studies. Nonetheless, SNPs and genes identified in the current study provide valuable
252	clues to understand the genetic basis of gene transcript abundance and metabolite
253	level in loblolly pine.
254	Annotation of the genes containing identified SNPs

255 We obtained the annotation for the genes containing the identified SNPs from loblolly

256	pine Gene Annotation v3.0 or blastx alignment. The SNPs that were associated with
257	gene expression phenotypes reside in 1635 different annotated genes. Of this total, 57 $\%$
258	reside in coding sequences (CDS), 2 % in 5' untranslated sequences (5'UTR), 3 % in 3'
259	untranslated sequences (3'UTR), 23 % in introns, 7 % in putative 3' regulatory
260	sequences (P3'RS) and 8 % in putative 5' regulatory sequences (P5'RS). The SNPs that
261	were associated with metabolite level phenotypes reside in 374 different annotated
262	genes. Of these, 58 % reside in CDS, 2 % in 5'UTR, 2 % in 3'UTR, 25 % in introns, 6 %
263	in P3'RS, and 7 % in P5'RS. The SNP-containing genes encode proteins with
264	functions of nucleic acid binding, transporter, oxidoreductase, transferase, hydrolase,
265	receptor, enzyme modulator, ligase, cytoskeletal protein, TF, membrane traffic protein,
266	and signaling molecule chaperone. The major molecular functions of SNP-containing
267	genes include: catalytic activity, DNA binding, transporter activity, receptor activity
268	and structural molecule activity.
269	Among the identified associations, some gene expression phenotypes were
270	associated with a large number of SNPs. For example, expression of the CYPB gene,
271	which encodes a terpenoid biosynthesis enzyme cytochrome P450 monooxygenase,
272	was associated with 181 SNPs. The SNPs associated with CYPB gene expression
273	mainly reside in the genes involved in secondary metabolites biosynthesis and defense
274	resistance, including genes encoding beta-glucosidase, phosphofructokinase,
275	polygalacturonase, shikimate O-hydroxycinnamoyltransferase-like, cytochrome P450
276	78A3, glucosinolate transporter-2, TIR-NBS-LRR protein, serine/threonine protein

277	kinase, and lipase. The expression phenotypes of genes encoding drought-responsive
278	TFs, RAP2.1 and ATAF-1, were also associated with a large number of SNPs, 133 and
279	138 SNPs, respectively. The associated SNPs mainly reside in drought responsive
280	genes or TF genes that confer drought tolerance to plants including genes encoding
281	cysteine-rich receptor-like protein, glucan endo-1,3-beta-glucosidase, COBRA-like
282	protein, cinnamoyl-CoA reductase, root phototropism protein, putative
283	TIR-NBS-LRR protein, laccase, cellulose synthase, UDP-glucuronyltransferase-like
284	protein, and TFs of ethylene-responsive, bHLH, MADS-box and MYBs. Table 1
285	presents a partial list of the genes containing SNPs associated with gene expression
286	and metabolite level phenotypes. More details are presented in Tables S2 & S3
287	(available as Supplementary Data at Tree Physiology Online).
288	TFs regulate gene expression in response to a variety of endogenous and
289	environmental cues. The SNP-containing genes that encode TFs were assigned to
290	plant TF families according to the Plant Transcription Factor Database v4.0
291	(http://planttfdb.cbi.pku.edu.cn/index.php). A total of 12 TF families were associated
292	with gene expression and metabolite level phenotypes (Figure 1). Twelve
293	SNP-containing TF family genes belong to MYB family, associating with expressed
294	genes encoding wood development protein (1CAB-3A), cellulose and callose synthase
295	(<i>CesA</i>), cell wall protein (<i>CslA</i>), α -tubulin (α tub1), lignin biosynthesis enzyme
296	(TC4H), drought-responsive TF (RAP2.1), phenylpropanoid pathway enzyme (ANR)
297	and metabolites 4-hydroxybenzoate, aspartic acid, maltose and melezitose. Details of

the TFs annotations, SNPs and their associated phenotypes are listed in Table S6

299 (available as Supplementary Data at *Tree Physiology* Online).

300

301 LD among identified SNPs that reside in the same scaffolds

- 302 Among the identified SNPs, we found that even though some loci are more than 10
- 303 kbp apart along the same scaffolds, they were associated with the same gene
- 304 expression phenotypes with similar r^2 values. To examine whether these loci are in
- 305 linkage disequilibrium (LD), we calculated their pairwise zygotic LD (squared

306 correlation coefficient R^2) values. From the results, we identified 10 scaffolds

307 containing correlated SNPs. For example, the SNPs tscaffold2867_628232,

308 tscaffold2867_651263, and tscaffold2867_755157 span 128 kbp on tscaffold2867

309 (Figure 2). They all were associated with expression of the *ATAF-1* gene

- 310 (drought-responsive TF) with $r^2 = 0.31$. High pairwise LD values (> 0.89) were
- 311 detected between these SNPs.

To further inspect for haplotype blocks on these scaffolds, we plotted LD heatmaps

- 313 for SNPs in these regions with SNPs with high correlation values. Figure 2 illustrates
- all LD values between SNP pairs around the investigated regions on tscaffold2867.
- 315 Other LD heatmaps are presented in Figures S2-S10 (available as Supplementary
- 316 Data at *Tree Physiology* Online). We did not observe long LD blocks along the

317 investigated regions in the LD heatmaps.

319 Network plots

320 Among the identified SNPs, some are associated with multiple gene expression and 321 metabolite level phenotypes. Plotting these SNPs with the gene expression and 322 metabolite level phenotypes with which they are associated in networks can provide 323 insight into the complex regulatory mechanisms underlying biological processes and 324 help us recognize key genes in the pathways. The network graphs were based on the 325 functional groups we assigned. The effects of SNP-SNP interactions were also 326 demonstrated in the networks. 327 The wood development and drought response networks (Figures 3 & 4, respectively) contain the largest number of SNPs. In the wood development network 328 329 (Figure 3), a total of 52 SNPs (each represented as a number in a blue node) are 330 connected to 56 gene expression phenotypes (yellow nodes, grey edges) and 8 331 metabolite level phenotypes (pink nodes, red edges). In the drought response network 332 (Figure 4), a total of 80 SNPs (each represented as a number in a blue node) are 333 connected to 10 gene expression phenotypes (yellow nodes, grey edges) and 4 334 metabolite level phenotypes (pink nodes, red edges). In both networks, purple edges 335 denote SNP-SNP interactions that significantly impact the phenotypes. The putative 336 functions of the SNP-containing genes included in these networks were determined 337 from loblolly pine gene annotation 338 (https://treegenesdb.org/FTP/Genomes/Pita/v1.01/annotation) or blastx alignment 339 (Tables 2 & 3 and Tables S7 & S8 available as Supplementary Data at Tree

340	Physiology Online). SNP #33 in the wood development network (Figure 3) and SNPs
341	#13, #20, #57, #70 and #78 in the drought response network reside in TF genes
342	(Figure 4).
343	Fewer associations between SNPs and gene expression phenotypes belonging to
344	the other functional groups were identified. Therefore, limited connections are shown
345	in the ROS response and disease response networks (see Figure S11, Table S9
346	available as Supplementary Data at Tree Physiology Online). No networks could be
347	plotted for gene expression phenotypes related to terpenoid biosynthesis, PCD or the
348	phenylpropanoid pathway.
349	Modules of genes with similar functionality can be recognized from the networks.
350	Gene-module level analysis can help us understand developmental and stress
351	resistance phenotypes in the context of biological network design and system
352	behavior rather than as a product of individual genes (Wang et al. 2008). A large gene
353	module related to wood development can be recognized in Figure 3. It contains 33
354	SNPs, 4 metabolites and 28 expressed genes that encode cellulose and callose
355	synthases, lignin biosynthetic enzymes, wood development enzymes, and tubulins.
356	Figure 4 includes two gene modules linked to drought responsive processes. One
357	module is composed of 24 SNPs, 2 metabolites and 4 expressed genes that encode
358	drought responsive TFs, drought signaling molecules and phenylpropanoid pathway
359	enzymes. The other module contains 52 SNPs and two expressed genes that encode a
360	drought responsive TF and a late embryogenesis abundant protein. These modules

361 supplement current regulatory and biosynthetic pathways for wood development and362 drought response.

363

364 **Discussion**

365 Genetic variations do not lead to changes in whole-plant traits directly, but instead act 366 on intermediate, molecular phenotypes, which in turn induce changes in higher-order 367 traits (Schadt et al. 2008). Therefore, identification of the genetic variants that 368 associate with intermediate phenotypes and description of molecular networks that 369 genes operate are important to understand the genetic basis underlying complex traits. 370 In this study, we explored the genetic regulation of gene transcript abundance and 371 metabolite level linked to important whole-plant traits, wood development and stress 372 responses by constructing networks comprised of the SNPs and their associated gene 373 expression and metabolite level phenotypes. The SNP-SNP interactions were also 374 described in the networks. These results provide valuable sources to bridge 375 connections between genetic variation, intermediate molecules produced in the 376 biological pathways, and whole-plant traits. 377 We identified 1841 SNPs associated with 191 gene expression phenotypes and 524 378 SNPs associated with 53 metabolite level phenotypes. Compared to a wide range of r^2 379 values of gene expression associations (0.09 to 0.85), we did not find strong 380 association signals for metabolite level associations (0.11 to 0.22), probably because SNP effects for metabolite level are generally low and the genetic basis underlying 381

382 metabolism involves more complex factors.

383	Among the SNP-gene expression associations, we detected 181 associations with
384	CYPB gene expression and 133 associations with RAP2.1 gene expression that have
385	remarkably high r^2 values, ranging from 0.40 to 0.85. The <i>CYPB</i> gene encodes a
386	cytochrome P450 monooxygenase enzyme involved in the synthesis of diverse
387	oleoresin terpenoids important for constitutive and induced defenses against pests and
388	pathogens (Ro et al. 2005), while the RAP2.1 gene encodes a
389	dehydration-responsive-element binding (DREB) protein type transcriptional
390	repressor. We also detected SNPs in strong associations with other gene expression
391	phenotypes, including the gene encoding abiotic stress responsive TF ATAF-1, and the
392	gene encoding phenylpropanoid pathway enzyme ANR. High r^2 values indicate that
393	the corresponding markers can explain a large proportion of the variation in
394	expression of these genes, and that the associated SNPs offer potential to discover
395	genes that regulate these biosynthetic pathways and stress responses. The SNPs
396	highly associated with CYPB and RAP2.1 gene expression are found in diverse genes.
397	SNPs associated with CYPB gene expression were discovered in genes involved in
398	secondary metabolite biosynthesis and defense pathways, including genes encoding
399	NBS-LRR type disease resistance protein and genes encoding MADS-box TF. SNPs
400	associated with RAP2.1 gene expression were discovered in drought responsive genes
401	or TF genes that contribute to drought tolerance, such as genes encoding MYB, which
402	plays a great role in controlling responses to biotic and abiotic stresses (Ambawat et

403	al. 2013). Although the effects of genes containing the identified SNPs on the
404	expressed genes need to be confirmed by the evidence from forward genetics
405	experiments, association studies are an efficient method to discover clusters of
406	candidate genes in biosynthetic pathways.
407	The pattern and extent of LD in the genome is important for association mapping
408	studies (Yu et al. 2008). In this study, we detected loci located more than 10kbp apart
409	along the same scaffolds that were associated with the same gene expression
410	phenotypes and had similar r^2 values. This observation raised the possibility that these
411	SNPs are in LD with each other or are even found within LD blocks. Although
412	outcrossing conifer trees are thought to have a rapid decline of LD, the rate of LD
413	decay may vary from gene to gene (Brown et al. 2004; Pavy et al. 2012). Furthermore,
414	if loci associated with the same phenotypes are in LD, it may suggest epistatic
415	interaction between these loci due to natural selection. In the current study, we
416	detected ten scaffolds that contained identified SNPs in strong LD with each other.
417	However, no LD blocks were observed in LD heatmap plots for the regions
418	surrounding the correlated SNPs (Figure 2 & Figures S2-S10 available as
419	Supplementary Data at Tree Physiology Online). These results diminish the potential
420	of interaction among investigated loci due to natural selection since large blocks of
421	LD should be maintained, if the interacted loci are under selection (Gabriel et al. 2002;
422	Slatkin 2008). The occasional LD observed here probably rise from mixing of
423	individuals from subpopulations. The population used in this study was comprised of

424 individuals with parents from a wide range across the southeastern U.S. Differences in
425 allele frequencies among subpopulations can create resemblance of LD (Slatkin
426 2008).

427	Gene networks demonstrate the potential interactions among genes and help us						
428	prioritize the candidate genes (Li et al. 2015). In the wood development network						
429	(Figure 3), SNP#33 resides in a TF GAMYB gene. It has been identified as an						
430	activator of gibberellin (GA)-regulated genes in plant growth (Woodger et al. 2003).						
431	SNP#33 was found to be associated with expressed genes encoding wood						
432	development enzyme and lignin biosynthetic enzyme, indicating that the GAMYB						
433	gene may influence lignin biosynthesis and wood formation through its regulatory						
434	interactions with a large number of genes. SNP #17 resides in a gene encoding						
435	arabinosyltransferase ARAD1. It is responsible for the polymerization of arabinose						
436	into the arabinan of arabinogalactan (Belanger et al. 1996). Arabinogalactan protein						
437	have been found functional during secondary wall formation in loblolly pine (Zhang						
438	et al. 2003). SNP#17 is associated with seven gene expression phenotypes all related						
439	to lignin biosynthesis. Lignin biosynthesis can be induced when cell wall is damaged						
440	(Denness et al. 2011). The associations between SNP#17 and lignin biosynthesis gene						
441	expression phenotypes imply a link between arabinogalactan protein and lignin						
442	biosynthesis for cell wall formation. SNP#31 resides in an aspartokinase gene.						
443	Aspartokinase is an enzyme that catalyzes the phosphorylation of aspartic acid. Data						
444	from bacteria has shown that decreasing aspartokinase activity results in blockage of						

445	cell wall growth (Rosenberg et al. 1973). The SNP#31 is associated with multiple						
446	lignin biosynthesis and wood development gene expression phenotypes, suggesting						
447	aspartokinase-mediated amino acid metabolism is involved in cell wood development						
448	and lignin biosynthesis. Laccase provides the oxidative capacity during lignification.						
449	The large number of gene family members makes it difficult to study (Piscitelli et al.						
450	2010). From the network in the Figure 3, we can identify a series of candidate genes						
451	that may function in the laccase synthesis pathway. Lac3 gene expression is						
452	associated with SNPs that reside in genes encoding cytochrome, disease resistance						
453	protein, calcium dependent protein kinase, LRR receptor-like and aspartokinase. Lac6						
454	gene expression is associated with SNPs that reside in genes encoding						
455	transmembrance protein, 1-phosphatidylinositol 3-phosphate, arabinosyltransferase						
456	and CBL-interacting protein kinase. These associations provide clues to understand						
457	the laccase oxidation process.						
458	Additive or epistatic interaction between loci is another factor that may further						
459	influence phenotypes (Phillips 2008). Lu et al. (2017) reported 11 SNP-SNP						
460	interactions in loblolly pine that in some cases, contributed more to the clonal and						
461	phenotypic variance of the quantitative traits than the identified additive loci. Thus by						
462	integrating SNP-SNP epistatic relationships into the network, we can acquire a more						
463	complete understanding of gene interactions. In the wood development network						
464	(Figure 3), <i>RP-L2</i> (ribosomal protein L2) gene expression is impacted by interactions						
465	of multiple SNP-SNP pairs. RP-L2 together with the 23S RNA are the main						

467 2000). The SNP-SNP interactions suggest genes encoding dormancy/auxin associated 468 protein, pentatricopeptide repeat-containing protein and histone H2A interact to affect 469 the formation of ribosomal protein. Additionally, interaction between an aspartokinase 470 gene and a disease resistance gene significantly influence CCoAMT gene expression, 471 but the mechanism remains unclear. 472 We also discovered important loci and phenotypes from the drought response network (Figure 4). Four gene expression phenotypes stand in the center of a series of 473 474 SNP associations. NCED is a key enzyme in abscisic acid (ABA) biosynthesis, which is induced by drought stress. ANR functions in the phenylpropanoid pathway. 475 476 Expression of NCED and ANR genes are widely associated with the same set of SNPs, 477 which reside in genes mainly encoding drought responsive products. This result 478 indicates ANR and NCED genes play key roles in the drought response pathway. 479 *PtEMB4* is a Late Embryogenesis Abundant protein. *ATAF-1* gene belongs to the *NAC* 480 (No Apical Meristem) family genes, which encode plant-specific TFs involved in

candidates for catalyzing peptide bond formation on the 50S subunit (Diedrich et al.

466

481 diverse biological processes (Wu et al. 2009). We found the expression of *ATAF-1* and

482 *PtEMB4* genes were associated with the same 52 SNPs, which reside in genes

483 encoding proteins such as wall-associated receptor kinase-like, heat stress TF. The

- 484 above results suggest the *PtEMB4* and *ATAF-1* genes as well as the *NCED* and *ANR*
- 485 genes may perform redundant functions during drought response processes.

486 Alternatively, there could be a synergetic mechanism for these genes to function487 together during drought response processes.

488	Metabolic changes in response to drought conditions play a key role for drought
489	adaptation in plants (Silvente et al. 2012). In the drought response network (Figure 4),
490	we found some SNPs were associated with both drought-related gene expression
491	phenotypes and metabolite level phenotypes. The genes containing the SNPs and the
492	expressed genes provide candidates to analyze the genetic basis of metabolic changes
493	in response to drought. Drought stress increases stearic acid (Júnior et al. 2008).
494	SNP#56 resides in a gene encoding a cytochrome P450. It is associated with stearic
495	acid concentration and NAC1 (a drought-responsive TF) gene expression. Melezitose
496	is found in the manna of many pine trees. During droughts, bees that collect manna
497	from these trees produce honey containing elevated concentrations of melezitose
498	(Purich 2017). SNPs #54 and #70 are associated with melezitose concentration and
499	RAP2.1 (a drought-responsive TF) gene expression. SNPs #54 and #70 reside in the
500	genes encoding a cytochrome P450 and a MYB domain protein, respectively. It is
501	probable that biosynthesis of melezitose in response to drought is under regulation of
502	drought responsive genes.
503	This study is an attempt to compose networks for exploring the genetic basis of
504	gene expression and metabolite level involved in complex biological processes. A

506 investigated genes and metabolites are too limited to cover all the genes related to the

total of 2.8 million SNPs were used to do association mapping, yet the numbers of

505

507 biosynthetic pathways. Numbers of genes related to ROS, PCD, terpenoid 508 biosynthesis and phenylpropanoid pathway are too few to compose networks. In 509 addition, gene expression and metabolite level were measured in different populations. 510 If these data were to be measured with the same samples collected at the same time, 511 the correlations between gene expression and metabolite level could be used to enrich 512 the current networks. In the future, we wish to take advantage of the active 513 development of transcriptome and metabolome profile technologies to improve the 514 quantification of gene transcripts and metabolites.

515

516 **Conclusion**

517 We have identified a total of 1841 SNPs associated with 191 gene expression

518 phenotypes and 524 SNPs associated with 53 metabolite level phenotypes. The

519 identified SNPs reside in genes with a wide variety of functions. We constructed

520 wood development and drought response networks and discovered key loci and genes

521 that contribute to the biological processes. This work provides candidate genes to

522 study the genetic basis of gene expression and metabolism involved in complex

- 523 biological processes, and highlights the efficiency of using
- association-mapping-based networks to discover candidate genes involved in complex
- 525 biological processes.

526

528 Supplementary Data

529 Supplementary Data for this article are available at *Tree Physiology* Online.

530 **Conflict of interest**

- 531 The authors declare that the research was conducted in the absence of any commercial
- 532 or financial relationships that could be construed as a potential conflict of interest.

533 Author contributions

- 534 ML performed the sample collection and measurement, data analysis, and wrote the
- 535 manuscript. KVK and CAL conceived and designed the study, coordinated the
- research and participated in the drafting of the manuscript. CMS helped with
- 537 expression data analysis, interpretation and manuscript editing. All authors read and
- 538 approved the final manuscript.
- 539

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544

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554

556

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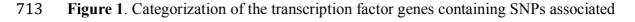
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711 Figure legends

712



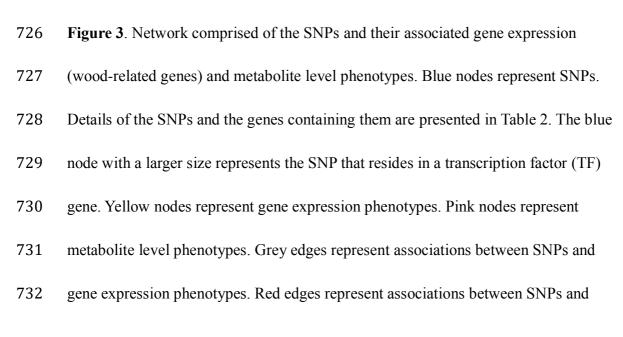
- with gene expression and metabolite level phenotypes. The gene expression
- phenotypes were classified into different functional groups: wood-related,
- 716 disease-related, drought-related, reactive oxygen species (ROS)-related, terpenoid
- biosynthesis, programmed cell death (PCD), and phenylpropanoid pathway. The
- numbers above each bar represent the numbers of the identified SNPs associated with
- 719 gene expression or metabolite level phenotypes.
- 720

Figure 2. Pairwise linkage disequilibrium (LD) values for SNPs in the scaffold

tscaffold2867. The bottom vertex of each red triangle highlights the high LD values

for SNPs tscaffold2867_628232, tscaffold2867_651263 and tscaffold2867_755157

724 $(R^2 > 0.89)$ located in the scaffold tscaffold 2867.



- 733 metabolite level phenotypes. Purple edges represent SNP-SNP interactions that
- significantly impact the phenotypes. Expressed genes in the network include:
- arabinogalactan-protein and cell wall protein genes: *AGP1-6*; cell expansion genes:
- 736 *COB*, *KORRI*; cell wall related (resistance related) genes: *CslA1*; cellulose and callose
- synthase genes: *CesA3*, *CslA2*, *CS-1343*; lignin biosynthesis enzyme genes: *4CL1*,
- 738 *C3H*, *CAD1*, *CCoAMT*, *COMT*, *Lac1-8*, *PAL1*, *TC4H*; α-tubulin gene: *αtub2*; wood
- development enzyme genes: BKACPS, BQR, Cellulase, EndChi, Importin, LP6,
- 740 PCBER, PLR, prxC2, SAH7, SPL, XET1; wood development protein genes: 1CAB-3A,
- 741 NH-10, NH-9, RP-L2; wood development TF genes: SND1, AIP, APL, eIF-4A, FRA2,
- 742 KNAT4, KNAT7, LZP, MYB1, MYB4, MYB85.
- 743
- Figure 4. Network comprised of the SNPs and their associated gene expression
- 745 (drought-related genes) and metabolite level phenotypes. Blue nodes represent SNPs.
- 746 Details of the SNPs and the genes containing them are presented in Table 3. The blue
- nodes with a larger size represent the SNPs that reside in transcription factor (TF)
- 748 genes. Pink nodes represent metabolite level phenotypes. Grey edges represent
- associations between SNPs and gene expression phenotypes. Red edges represent
- associations between SNPs and metabolite level phenotypes. Purple edges represent
- 751 SNP-SNP interactions that significantly impact the phenotypes. Expressed genes in
- the network include: drought signaling genes: *ABI1*, *NCED*, *RPK1*;
- 753 drought-responsive TF genes: NAC1, RAP2.1, RAP2.4, ATAF-1; late embryogenesis

abundant protein genes: *PtEMB3-4*; phenylpropanoid pathway gene: *ANR*.