Cover Page

Title: *GIGANTEA* Is A Negative Regulator Of Abscisic Acid Transcriptional Responses And Sensitivity In Arabidopsis

Running head: *GIGANTEA* suppresses abscisic acid responses

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Abstract

Transcriptional reprogramming plays a key role in drought stress responses, preceding the onset of morphological and physiological acclimation. The best-characterised signal regulating gene expression in response to drought is the phytohormone abscisic acid (ABA). ABA-regulated gene expression, biosynthesis and signalling are highly organised in a diurnal cycle, so that ABA-regulated physiological traits occur at the appropriate time of the day. The mechanisms that underpin such diel oscillations in ABA signals are poorly characterised. Here we uncover *GIGANTEA* (*GI*) as a key gatekeeper of ABA-regulated transcriptional and physiological responses. Time-resolved gene expression profiling by RNA sequencing under different irrigation scenarios indicates that *gi* mutants produce an exaggerated ABA response, despite accumulating wild-type levels of ABA. Comparisons with ABA-deficient mutants confirm the role of GI in controlling ABA-regulated genes and the analysis of leaf temperature, a read-out for transpiration, supports a role for GI in the control of ABA-regulated physiological processes. Promoter regions of GI/ABA-regulated transcripts are directly targeted by different classes of transcription factors, especially PHYTOCHROME-INTERACTING FACTORs, and (ABRE)-BINDING FACTOR, together with GI itself. We propose a model whereby diel changes in GI control oscillations in ABA responses. Peak GI accumulation at midday contributes to establishing a phase of reduced ABA sensitivity and related physiological responses, by gating DNA binding or function of different classes of transcription factors that cooperate or compete with GI at target regions.

Keywords

1   Introduction

2   Several drought responses rely on short term transcriptional reprogramming of
3   physiological and metabolic-related genes to enable long term morphological
4   adjustments. These gene regulatory events respond to combinations of signals to allow
5   precise spatial/temporal organization. As these signals converge to chromatin regions,
6   one key question is to understand what mechanisms enable their transduction and
7   integration onto regulatory DNA sequences, as this would allow a better understanding
8   of the evolution of adaptive strategies in response to water deficit.

9   The best characterised messenger of water deficit conditions is the phytohormone
10   abscisic acid (ABA). Cellular ABA levels are detected by a class of soluble ABA
11   receptors of the PYRABACTIN RESISTANCE/PYRABACTIN RESISTANT-
12   LIKE/REGULATORY COMPONENT OF ABA RECEPTOR (PYR/PYL/RCAR)
13   protein family (hereafter referred to as PYLs). ABA-bound PYLs interact with
14   PROTEIN PHOSPHATASES 2C (PP2Cs) that act as negative regulators of ABA
15   signalling (Ma et al., 2009; Miyazono et al., 2009; Park et al., 2009; Rubio et al., 2009;
16   Santiago et al., 2009). The repressive role of PP2Cs is exerted through their binding to
17   SUCROSE-NON-FERMENTATION KINASE SUBFAMILY 2 (SnRK2s) proteins,
18   causing repression of their kinase activity (Umezawa et al., 2009; Vlad et al., 2009).
19   Thus, ABA-stimulated PYLs inhibit PP2Cs so that the SnRK2s can initiate ABA
20   responses. Activated SnRK2s quickly phosphorylate and activate multiple target
21   proteins, including transcription factors (TFs) that control ABA-responsive genes
22   (Fujii et al., 2009; Furihata et al., 2006; Minkoff et al., 2015; Umezawa et al., 2013;
23   Wang et al., 2013). ABA-responsive element (ABRE)-BINDING FACTORS (ABFs),
24   belonging to the basic leucine zipper (bZIP) family are master regulators of ABA-
25   dependent transcriptional reprogramming (Fujita et al., 2011, 2005; Yoshida et al.,
26   2015). However, integration of RNA-sequencing and Chromatin immunoprecipitation
27   (ChIP)-seq studies revealed additional levels of complexity associated with ABA
28   transcriptional responses, orchestrated by a wide network of TFs with different binding
29   dynamics and combinatorial interactions at target genes promoters (Song et al., 2016).
30   Coordination of these transcriptional events may dictate the regulation of ABA
31   production and sensitivity in space (tissue types) and at different timescales,
32   appropriate with the different water stress conditions. While ABA is the major

33   regulator of plant gas exchange, the molecular mechanisms that control the diel
34   organization of its responses are only beginning to emerge (Dubois et al., 2017; Endo
35   et al., 2008; Fukushima et al., 2009). One key mechanism relies on the circadian clock
36   that affords coordination of physiological and metabolic processes according to
37   transpiration demands. Stomatal opening, photosynthetic rates, carbon metabolism and
38   assimilation impact lifetime biomass accumulation and also drought tolerance (Dodd
39   et al., 2005; Nakamichi et al., 2016; Simon et al., 2020). Downstream of the core clock
40   oscillator, several mechanisms may restrict specific transcriptional programs at
41   specific times. *GIGANTEA* (*GI*), encoding a plant-specific gene, has been implicated
42   in multiple signalling cascades, including the regulation of circadian rhythms and
43   several plant environmental responses (Fowler et al., 1999; Huq et al., 2000; Martin-
44   Tryon et al., 2007; Mizoguchi et al., 2005; Park et al., 1999). *GI* transcript and protein
45   follow a similar diel accumulation pattern, with peaks occurring at approximately
46   midday in a typical long day photocycle of 16h (Fowler et al., 1999; Park et al., 1999).
47   This oscillatory pattern of GI is associated with the regulation of different sets of genes
48   within the circadian cycle, thereby controlling numerous phenotypic traits (Mishra and
49   Panigrahi, 2015). The activation of photoperiodic flowering is one of the best-studied
50   modes of action of GI with respect to transcriptional regulation. GI binds to the blue
51   light photoreceptor FLAVIN-BINDING, KELCH REPEAT, F-BOX 1 (FKF1) to
52   promote degradation of CYCLING DOF FACTORs (CDFs) that repress the floral
53   activator *CONSTANS* (*CO*) and *FLOWERING LOCUS* (*FT*) (Fornara et al., 2009;
54   Imaizumi et al., 2005; Sawa et al., 2007). Interestingly, the GI-CDFs module regulates
55   freezing tolerance in *Arabidopsis,* pointing to a wider contribution of this
56   transcriptional mechanism to gene regulation (Fornara et al., 2015; Kim et al., 2012).
57   GI has been described as a scaffold for direct interaction with the blue light
58   photoreceptor (and FKF1 homologue) ZEITLUPE (ZTL). GI promotes ZTL
59   accumulation and, in turn, ZTL proteolytic activity against TIMING OF CAB
60   EXPRESSION 1 (TOC1), a transcription factor that controls clock function (Ito et al.,
61   2012; Kim et al., 2007). Thus, one clear mode of GI influence on transcriptional events
62   is through FKF1 and ZTL, stimulating targeted protein degradation of their specific
63   targets.

GI can be found in different protein complexes (Ito et al., 2012; Krahmer et al., 2019), including nuclear complexes that control its own stability and localization to regulate chromatin accessibility (Y. Kim et al., 2013; Yu et al., 2008). As GI protein lacks a recognized DNA binding domain, its association with DNA regulatory elements may be indirect, possibly mediated by a variety of TFs (Baek et al., 2020; Kubota et al., 2017; Nohales et al., 2019; Sawa and Kay, 2011). The association between GI and different TFs may explain the high level of phenotypic pleiotropy described in *gi* mutants, including enhanced resistance to oxidative stress (Kurepa et al., 1998) and salt stress (W. Y. Kim et al., 2013) which cannot be ascribed to the above well-characterised GI interactors. Despite the importance of GI in gene-environment regulation across plant species (Izawa et al., 2011), little is known about its role in regulating transcriptional responses elicited by water deficit. *gi* mutants present signatures of ABA deregulated gene expression in the absence of external stress suggesting a role in mediating water deficit signals (Fornara et al., 2015; Kim et al., 2012) and GI function is sensitive to ABA signalling (Riboni et al., 2016). Physiological data support a role for GI in promoting stomatal opening and thus water loss (Ando et al., 2013), and in activating ABA biosynthetic genes (Baek et al., 2020). In this work we describe GI as a key component of ABA-regulated transcriptional outputs. The observed patterns of gene expression and mutant analyses are consistent with a model where GI establishes diurnal oscillations in ABA sensitivity so that peak GI accumulation leads to a phase of minimal ABA sensitivity and maximal transpiration.

Results and discussion

GI represses ABA responses in a phase specific manner

We determined the contribution of *GI* in controlling ABA signalling and responses by analysing differentially expressed genes (DEGs) in *gi-2* mutant plants subject to water deficit. DEGs were also compared to strong ABA deficient mutant plants (*aba1-6*), and wild-type Columbia (Col-0) plants undergoing the same conditions. Samples derived from plants grown for 3-week-old under short day photoperiods (SDs) and undergoing either water deficit or well-watered conditions before shifting to long days (LDs), while maintaining the same irrigation scheme (well-watered and water deficit,

96　maintained by gravimetric measurements) (Figure 1A). This experimental design
97　enabled us to synchronise plants at the vegetative stage and to control for potential
98　effects on gene expression caused by the different developmental stages of the
99　genotypes. It also allowed us to evaluate the impact of GI on ABA/water deficit -
100　regulated traits before and after photo stimulation of the photoperiodic pathway (i.e.,
101　when the role of GI is best described) as we sampled tissues at different times in the
102　day including ZT1, (Zeitgeber Time 1, i.e., within 1 hour after dawn of the last SD)
103　and the remaining time points capturing the end of the last SD (ZT8) and the photo-
104　extension to LD (ZT12 and ZT16) (Figure 1A).

105　In the wild type, water deficit resulted in a statistically significant deregulation of
106　genes at all the time points considered. A slight over representation of repressed genes
107　compared with normal irrigation was observed (Figure 1B). At every time point,
108　pairwise comparisons of sets of DEGs revealed both common and time-of-the-day-
109　specific patterns of expressions (Figure 1C). The observation that water deficit could
110　impose strong and specific gene deregulations already in the morning timepoint (ZT1)
111　may support a model of anticipation of expression of drought tolerance-related genes
112　to prepare for higher transpiration demand later in the day (Mizuno and Yamashino,
113　2008). We recovered a total of 4780 distinct individual DEGs (FDR < 0.05) by
114　comparing mutants to wild type plants, at any given time-point and irrigation condition
115　(Table S1). Principal component analysis (PCA) of log normalised DEGs expression
116　levels revealed that the most extreme difference in gene expression (PC1, 39%) was
117　observed between time of the day ZT1 and the remaining time points (ZT8, ZT12 and
118　ZT16) across all genotypes (Figure 1D). PC2 broadly reflected the contribution of
119　genotype and treatment but explained smaller proportions of the variation in gene
120　expression (19% of the variability of the DEGs). Interestingly, the *gi* mutant displayed
121　less separation compared with ABA deficient mutants and the wild type on both the
122　first (time points) and second (genotype/treatment) components of the PCA,
123　suggesting a deregulation in the diel responses to water deficit stimuli in *gi* mutant
124　plants compared to the other genetic backgrounds analysed in this study.

125　We could detect a remarkable separation of gene expression patterns between morning
126　DEGs detected at ZT1 and the remaining time points. Based on the consideration that
127　ZT8, ZT12 and ZT16 clustered together in our PCA, and that ZT12 could coincide

with light-dependent stimulation of the photoperiodic cascade (Sawa et al., 2007), we selected the ZT1 and ZT12 time points to perform more detailed analyses. GI function had a large impact on gene expression at ZT12 compared with ZT1, with a clear prevalence of upregulated genes at ZT12 under well-watered and water deficit conditions when the number of DEGs increased up to 2.4-fold compared to control conditions (1817 vs 730 DEGs, respectively) (Figure 2A,B). A similar over representation of upregulated genes in *gi* mutants occurred at ZT8 and ZT16 compared with the wild type and similar observations were made in direct comparisons with *aba1* mutants (Figure S1), further supporting a general repressive role of GI on water deficit-regulated gene expression.

The analysis of *GI* transcript accumulation across different ZTs revealed a peak in accumulation at ZT8 with negligible variations observed in response to water deficit or ABA deficiency (Figure S2). Based on prior reports (Sawa et al., 2007), GI protein accumulation broadly reflects its transcript levels, indicating that the large increase in DEGs detected at ZT12 does not coincide with the peak of GI accumulation. This could point to additional molecular mechanisms mediated by GI that are dependent on the extended light period, namely the light-dependent interaction with blue light photoreceptors. Despite this observation, a large proportion of DEGs in *gi* mutants were common between ZT1 and ZT12 (Figure 2C) and intersections of sets of DEGs were statistically significant between any time point (hypergeometric distribution p-values all <= 1E-33). Moreover, a sizable fraction of GI-DEGs at ZT1 and ZT12 (39% and 66%, respectively) observed under normal irrigation conditions were also deregulated under water deficit at matched time points. Thus, even in the absence of water deficit stress *gi* plants already present alterations in the water deficit-regulated transcriptome. As water deficit caused an increased representation of upregulated genes in *gi*, we tested the possibility that this could depend on an interaction between the genotype and water deficit (i.e., an amplification of water deficit-mediated gene upregulation in the absence of GI function). We compared fold change of expression of GI-DEGs under water deficit at ZT12 (*gi* vs. wild type) to changes in gene expression observed under water deficit vs. control in the wild type and *gi* mutants, respectively (Figure 2D). Strikingly, water deficit caused a global downregulation of these genes in the wild type, and no consistent upregulation could be observed in *gi*

plants under water deficit compared with control conditions. These observations suggest that GI contributes to repress expression of these genes and water deficit inputs have a limited contribution in driving further gene upregulation in *gi* mutants. Genes that were upregulated in the wild type in response to water deficit had similar levels of expression in *gi*, independent of water deficit conditions. However, these genes displayed a higher level of variability in their expression patterns compared with genes that were downregulated in the same condition (coefficient of variation of the logFC distribution = 55.3 and 3.4, respectively for the up- and down- regulated genes), suggesting that GI-activated genes are more significantly influenced by water deficit conditions in the absence of GI.

In the ABA deficient mutant background *aba1* a higher number of DEGs was recovered at ZT1 compared with ZT12 (Figure S3). Water deficit conditions strongly amplified this time-of-the-day dependency on ABA-DEGs (547 DEGs under normal irrigation vs 1610 under water deficit), when we observed a mild over-representation of down regulated genes. These patterns of gene deregulation may reflect diel oscillations of ABA production, which peaks at dusk (Adams et al., 2018; Fukushima et al., 2009). In this view and since severe ABA deficient mutants like *aba1-6* can synthesize small quantities of ABA (Rock and Zeevaart, 1991), *aba1-6* plants may have the lowest point of ABA accumulation in the morning with a gradual recovery at later time points (Figure S3). Despite the different diel contribution of GI signalling to gene regulation under water deficit conditions we could detect highly significant overlaps between ABA and GI DEGs both at ZT1 and ZT12 time points, under both well-watered conditions (53 and 66 common DEGs at ZT1 and ZT12, respectively) or water deficit (419 and 400 common DEGs, at ZT1 and ZT12, respectively) (Figure S3). Our data suggest that diel changes in GI accumulation cause different phases of ABA sensitivity whereby high levels of GI signalling at ZT12 contribute to repress ABA responses.

GI binding overlaps with different classes of TFs and is associated with the repression of target genes

The increased number of shared DEGs in *aba1* and *gi* plants under water deficit points to a convergence of GI and ABA signalling to gene regulation. Cis motif analysis at

192   the promoter regions (1000 bps upstream of the transcription start site - TSS) of ABA

193   and GI DEGs predicted significant enrichments for several TFs binding sites. We

194   computed scores for binding sites enrichment at DEGs detected under different

195   combinations of genotype comparison/treatment/time point (Table S2). Focusing on

196   the binding motifs detected in DEGs common among ABA and GI DEGs under water

197   deficit, we found an over-representation for bZIPs transcription factors binding

198   (including ABA INSENSITIVE 5, p value = 2.23E-25 at ZT1 and related ABF2, p

199   value = 1.76E-22 at ZT1) which are key in coordinating ABA-dependent

200   transcriptional responses and PHYTOCHROME-INTERACTING FACTORs (PIFs)

201   (PIF4, p value = 6.84E-18 at ZT1), acting as central components of plant

202   photomorphogenesis (Leivar and Monte, 2014) (Figure 3A). While there are no

203   obvious indications about the role of PIFs in response to drought stress in *Arabidopsis*,

204   recent data indicate that their binding at target chromatin regions is negatively

205   regulated by GI (Nohales et al., 2019). We also detected a significant enrichment of

206   CDF binding sites (CDF2, p value = 2.80E-03 at ZT1) at the promoter of ABA and GI

207   DEGs, consistent with their role in conferring plant survival upon freezing temperature

208   (Fornara et al., 2015) and drought (Corrales et al., 2017).

209   A meta-analysis of a selection of ChIP-seq datasets of known ABA-related

210   transcription factors (Song et al., 2016) and PIF proteins (Pfeiffer et al., 2014)

211   confirmed the strong enrichment for the binding of ABFs and PIFs at the promoters of

212   ABA and GI DEGs (Table S3). PIFs and ABFs direct targets were more represented

213   under water deficit conditions, irrespective of the time point, and were not associated

214   with a clear direction of regulation (Figure 3B). Most DEGs in *gi* or *aba1* mutants

215   were also targets of both ABFs and PIFs, suggesting cooperation between these

216   families of TFs at target genes promoters and pointing to contribution of PIFs in

217   coordinating drought stress responses.

218   Many other ABA-regulated families of TFs were similarly and significantly enriched

219   in the promoters of our lists of DEGs, potentially indicating a pervasive role of GI in

220   the regulation of ABA transcriptional responses (Table S3). GI ChIP-seq peaks were

221   found to be significantly associated with the promoters of both *gi* and *aba1* mutants

222   DEGs (Table S3). GI binding was also significantly more associated with upregulation

223   of the target genes detected in *gi* mutants (Fisher's exact test p value = 1.09E-02),

224   confirming the predominantly repressive role of GI with respect to gene expression

225   (Figure 3C). The comparison of independent ChIP datasets allowed us to uncover a

226   general and significant overlap between ABA-regulated TFs and GI binding peaks

227   (Table S4). As these comparisons included different families of transcriptional

228   regulators, these results could point to a direct role for GI at the regulatory chromatin

229   of ABA-responsive genes, in cooperation or competition with different TF families.

230   Such a general association between GI and several families of TFs offers intriguing

231   insights into the complexity of the molecular interplay between different TFs and their

232   target genes upon water deficit conditions. In this scenario GI might alter the stability,

233   activation, or occupancy potential of TFs in a phase specific manner.

234

235   **GI regulates ABA signalling genes but not ABA accumulation**

236   Similarities in gene deregulation patterns between *gi* and *aba1* do not derive from

237   altered ABA accumulation in *gi* mutants. Unlike recent reports, we found no

238   significant deregulation in the levels of the rate limiting ABA metabolic genes *9-CIS-*

239   *EPOXYCAROTENOID DIOXYGENASE 3* (*NCED3*) under well-watered conditions

240   (Baek et al., 2020) nor other ABA metabolic or catabolic genes in *gi* mutants (Figure

241   S2, Table S1). A slight, but significant, decrease in *NCED3* transcript accumulation

242   was observed under water deficit conditions in *gi* mutants at ZT12 and ZT1 (Figure

243   4A, Figure S2 and S4). To further verify the possible role of GI in promoting ABA

244   accumulation under water deficit conditions, we conducted an independent experiment

245   to measure total ABA accumulation. Samples derived from plants grown under water

246   deficit conditions or well-watered conditions in a continuous LD photoperiod. We

247   harvested tissue at ZT10, which coincides with high GI-stimulated photoperiodic

248   signal stimulation (Sawa et al., 2007). To control for the different developmental

249   stages of *gi* and wild-type plants under LDs we also analysed *co* and *fkf1* mutants,

250   which are phenotypically comparable to *gi* with respect to the duration of the

251   vegetative phase and growth but are defective at different steps of the photoperiodic

252   cascade. Basal levels of ABA accumulation under normal irrigation conditions were

253   similar in all the genotypes considered (Figure S5). Cellular ABA accumulation

254   increased in wild-type plants undergoing water deficit, but similarly so in *gi*, *fkf1* and

co mutants. Thus, our data are more consistent with a model where GI regulates ABA-responses or signalling via modulation of transcriptional processes.

Gene Ontology analysis revealed both common and unique functions associated with GI-DEGs at ZT1 and 12 under normal irrigation conditions (Figure S6). DEGs at ZT1 and ZT12 were particularly enriched in carbohydrate metabolic processes and photosynthesis-related functions respectively, which could be linked to the known role of GI in carbon metabolism and sugar hold-release signalling (Dalchau et al., 2011; Eimert et al., 1995; Mugford et al., 2014). Notably, DEGs related to "response to abiotic stimulus" and "water deprivation" were significantly enriched under control conditions at ZT12 and ZT1, respectively. As expected, most gene responses observed under water deficit conditions were associated with water deprivation terms, irrespective of the time point analysed.

Despite the known interplay between the circadian clock and ABA-related responses (Mizuno and Yamashino, 2008; Seung et al., 2012), we found limited examples of deregulation of core circadian clock genes in response to the water deficit conditions used in this study, which extends similar observations under mild drought scenarios (Dubois et al., 2017). A notable exception to this pattern was *TOC1* which was similarly and significantly upregulated in *gi* and *aba1* plants under water deficit at ZT1 (Figure S4). Such an increase in *TOC1* transcript levels may translate into elevated TOC1 protein abundance in *gi* mutant plants, as GI mediates the proteolytic degradation of TOC1 in association with blue light stimulated F-box protein ZTL at dawn (Kim et al., 2007). Furthermore, *TOC1* over-expression causes reduced ABA sensitivity of guard cells, constitutive stomata opening and decreased plant survival under drought conditions (Legnaioli et al., 2009). By comparing changes in DEGs previously assigned to circadian clock regulation (Covington et al., 2008), we confirmed an over-representation for circadian clock-controlled genes among the *GI* DEGs at time points representative of morning and dusk, particularly under water deficit (Figure S7). A similar high representation for circadian genes was observed for ABA DEGs, confirming the enrichment for ABA-regulated processes among circadian-related genes (Covington et al., 2008).

Previous studies also defined a set of ABA-responsive genes under circadian controls including *EARLY RESPONSE TO DEHYDRATION* (*ERD*) *7*, *COLD-REGULATED*

287 (*COR*) *15A* and *B* (Mizuno and Yamashino, 2008). Besides verifying that the

288 accumulation of these genes was ABA-dependent we also found a general and

289 significant pattern of increased accumulation in *gi* mutants compared with the wild

290 type (Figure 4A,B, Figure S2 and S4). This pattern of deregulation was previously

291 associated with the increased freezing tolerance of *gi* mutant plants (Fornara et al.,

292 2015; Kim et al., 2012).

293 We set up an independent experiment to test if overexpression of *GI* could reduce the

294 accumulation of these ABA-regulated markers. We compared *35S::HA-GI* plants with

295 the wild type, upon exogenous ABA applications or in control conditions after two

296 weeks of growth on soil under a long day photocycle and measured transcript levels

297 by quantitative real time PCR. Under control conditions, *COR15A* and *ERD7*

298 transcripts accumulation followed a strong daytime-dependent increase in the wild

299 type, with higher levels of accumulation detected at ZT8 compared to ZT1. At ZT8,

300 we observed a significant decrease in these transcripts levels in *35S::HA-GI* plants

301 compared with the wild type (factorial ANOVA analysis p = 7.90E-03 and 4.97E-02

302 for *COR15A* and *ERD7*, respectively) (Figure S8). Upregulation of *COR15A* in

303 response to ABA (i.e., the slope) was significantly stronger in *35S::HA-GI* plants,

304 while overall ABA-dependent accumulation of these markers in *35S::HA-GI* plants

305 reached similar levels of the wild type. Thus, under physiological, non-stressed

306 conditions *GI* can reduce the accumulation of these genes, but an acute increase in

307 cellular ABA concentration (as in this condition) may ultimately overcome GI-

308 repressive function despite the 5-6 fold excess of *GI* accumulation detected at ZT8 in

309 *35S::HA-GI* plants compared with the wild type (Figure S8).

310 RNAseq analysis of other ABA responsive target genes including *MITOGEN-*

311 *ACTIVATED PROTEIN KINASE KINASE KINASE 18* (*MAPKKK18*), *Rab-related*

312 *gene 18* (Lång and Palva, 1992; Mitula et al., 2015) did not reveal significant variations

313 between the genotypes analysed under water deficit conditions (Figure 4A,B Figure

314 S2 and S4). This could be due to transcriptional desensitization of ABA downstream

315 targets upon prolonged water deficit conditions (Asensi-Fabado et al., 2017). The

316 ABA-regulated *MYB96* transcription factor (Lee et al., 2016), controlling ABA

317 sensitivity in the evening was slightly (but not significantly) upregulated in *gi* mutants,

318   which could contribute to increase ABA responses. However, no further upregulation

319   was observed under water deficit condition (Figure S2).

320   Inspection of genes associated with ABA signalling in *gi* plants at ZT12 revealed a

321   general and significant downregulation of clade A *PP2C*-encoding genes *ABI1*, *ABI2*

322   and *HIGHLY ABA-INDUCED* (*HAI*) *1* and *2*, the proteostasis – related *ABI FIVE*

323   *BINDING PROTEIN* (*AFP*) *1* and *3* genes, which function as negative regulators of

324   ABA signalling (Bhaskara et al., 2012; Leung et al., 1997; Lopez-Molina, 2003) and

325   a slight upregulation of the ABA receptor *PYL4* (Figure 4A). The reduced levels of

326   *PP2Cs ABI1*, *ABI2* transcript levels in *gi* plants undergoing water deficit was further

327   verified by quantitative real-time PCR analysis on the same time points analysed by

328   RNAseq and across additional time points encompassing the previous short day and

329   the subsequent long day (Figure S9). This pattern of accumulation of ABA signalling

330   and responsive genes echoed that observed in *aba1* mutants (Figure 4B and Figure S9)

331   which would be expected as results of impaired ABA transcriptional responses (Wang

332   et al., 2019). Because we found downregulation of clade A *PP2Cs* and upregulation

333   of *PYL4*, our data support a role for GI in regulating the early steps of the ABA

334   signalling cascade, which could further impact ABA-regulated gene expression.

335   Previously we detected an over representation for ABA-regulated bZIPs at the *GI*-

336   regulated promoters under water deficit (Figure 3B). Transcript levels of *ABF3* and

337   *ABF4* did not change in *gi* mutants compared with the wild type at ZT12, pointing to

338   post-transcriptional effects (Figure 4A). As loss in *PP2Cs* function results in increased

339   ABA responsiveness (Bhaskara et al., 2012; Rubio et al., 2009), GI may normally act

340   to repress ABA signalling at this time of the day at multiple levels, by de-sensitising

341   the core ABA signalling cascade, and by interfering with TFs (e.g., the ABFs)

342   function.

343

344   **GI regulates phase specific ABA sensitivity to control transpiration**

345   We next examined the role of GI in regulating ABA-specific physiological traits. We

346   measured leaf surface temperatures, which is highly related to transpiration through

347   stomata (Yang et al., 2016). Stomatal movement is regulated via ABA through rapid

348   post-transcriptional activation of ion channels localised at the plasma membrane

349   (Munemasa et al., 2015). Thus, changes in ABA signalling and response in *gi* guard

cells should cause alterations in water loss compared with the wild type, which can be monitored using an infrared imaging approach. To control for the different development of wild-type and *gi* plants we analysed plants undergoing a shift from short to long days, with data collected on the second long day (Figure 5A). We evaluated the impact of *gi* on leaf temperature in response to ABA treatment at three time periods (ZT1, ZT8, and ZT16) using a factorial linear model. We tested for the main effect of genotype (*gi* vs. wild type), treatment (mock vs. ABA applications), or their interaction across each time period. Here, a significant interaction indicates that the effect of *gi* mutation on leaf temperature differed from wild type in response to ABA treatment. At ZT1 and ZT16 we discovered simple main effects of Genotype and Treatment. At both of these ZTs, ABA treatment slightly increased leaf temperature (T; $P<0.05$). At ZT1, the wild type had higher leaf temperature compared to *gi* and the reverse was true at ZT16 (G; in both cases $P<0.05$). However, genotype responses to ABA were similar (GxT $p>0.05$) at these time points. In contrast, we detected a significant Genotype x Treatment interaction at ZT8. Here, the *gi* mutant showed a stronger increase in leaf temperature in response to ABA treatment compared to the wt (GxT; $P=0.008$) (Figure 5B). The detected patterns of warmer leaf temperature observed in *gi* mutants at ZT8 and ZT16 may reflect the consequences of time-of-the day changes in GI function as a repressor of ABA-regulated processes. Notably, impairing GI function caused increased sensitivity to ABA compared with the wild type, but only at ZT8. This indicates that other layers of ABA responsiveness are regulated independent of GI function.

In summary, our study provides a framework for defining how GI exerts multilevel influence on ABA-regulated gene expression, ABA signalling sensitivity, and the phenotypic traits that depend on these molecular processes. Our data indicate that GI acts as a general hub for the ABA transcriptional network, in conjunction with multiple TFs families. GI was recently shown to prevent PIFs binding to chromatin via direct interaction (followed by PIFs degradation) and competition at chromatin region (Nohales et al., 2019). Here we extend this model to suggest that GI may exert similar regulatory roles on many other TFs to gate ABA responses according to diurnal cycles of GI accumulations (Figure 5C).

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387   **Materials and Methods**

388

389   **Plant material and growth conditions**

390   In this study we used wild–type *Arabidopsis* plants, ecotype Columbia (Col–0),

391   transgenic lines *35S::HA-GI* (David et al., 2006) and mutant lines *aba1-6* (Niyogi et

392   al., 1998), *gi-2* (Fowler et al., 1999), *gi-100* (Huq et al., 2000), *fkf1* line SALK_059480

393   (Riboni et al., 2013), *co-10* (Laubinger et al., 2006). Seeds were germinated and plants

394   grown in a controlled environment at a temperature of 21-23 °C, 65% relative humidity,

395   either under long day (16 h light / 8 h dark) or short day (8 h light / 16 h dark)

396   photocycles. Light was cool white fluorescent tubes (Osram, Sylvania) at a fluency of

397   120–150 μmol m^{-2} s^{-1} (Photosynthetically active radiation). Water deficit conditions

398   were imposed two days after germination, so that under normal irrigation conditions

399   plants grew under a Relative Soil Water Content (RSWC) of 80 – 90%, and 30%

400   RSWC under water deficit (Riboni et al., 2013). RSWC was kept constant throughout

401   the experiment (i.e., during the short day and long day part) by daily weighing of pots

402   and applications of water to maintain the desired values. Samples used for RNA-seq

403   analysis (or real time PCR) derived from an experiment previously described (Riboni

404   et al., 2013). For each time point / treatment / genotype combination, we analysed two

405   biological replicates, each one consisting of approximately 50 seedlings pooled from

406   three different Arabasket pots.

407   ABA quantification derived from an independent experiment. Stratified seeds (20–50)

408   were sown in Arabasket pots and grown under water deficit irrigation (or control) for

409   20 days in a growth chamber set under long day photocycle. Plants were harvested at

410   ZT10 in three biological replicates, and each replicate consisting of 100 mg of pooled

411   seedlings derived from 2-3 independent Arabasket pots. To avoid soil carryover, we

412   harvested only the aerial part of plants (i.e., above the hypocotyl).

413 To quantify gene expression in response to exogenous ABA, plants were grown in
414 Arabasket pots (at a density of approx. 20 seedlings in each pot) under normal
415 irrigations for two weeks (after germination) in a long-day growth chamber (as above).
416 On the evening of the 15th day plants were sprayed with 10 µm ABA or a mock
417 solution. Sampling occurred on the following day at ZT1 and ZT8 and four replicates
418 for each genotype/timepoint/treatment combination were harvested from independent
419 pots.

420

421 **RNAseq and expression analysis by real time PCR**
422 RNA was extracted using the Trizol reagent (Invitrogen). For RNA sequencing, RNA
423 Quality Control was performed with an electrophoretic run on a Bioanalyzer
424 instrument using the RNA 6000 Nano Kit (Agilent). RNA Integrity Number was
425 determined, and all the samples were considered suitable for processing (RIN > 8).
426 RNA concentration was estimated through a spectrophotometric measurement using a
427 Nanoquant Infinite M200 instrument (Tecan). Sequencing libraries were prepared
428 using the TruSeq™ RNA Sample Preparation Kit (Illumina). Polyadenylated
429 transcripts were purified using poly-T oligo-attached magnetic beads. PolyA RNA was
430 fragmented at 94 °C for 8 min and retrotranscribed using random hexamers. Multiple
431 indexing adapters were ligated to the ends of the cDNA and the amount of DNA in the
432 library was amplified by PCR. Final libraries were validated and quantified with the
433 DNA1000 kit on the Agilent Bioanalyzer Instrument. Pooled libraries were sequenced
434 on the Illumina Genome Analyzer IIx producing 72nt paired-end reads.
435 Reads were mapped on the reference assembly of *Arabidopsis thaliana* genome TAIR
436 vs.10 as available from ftp://ftp.arabidopsis.org/home/tair) using the bowtie2 program
437 (Langmead and Salzberg, 2012). Estimation of gene expression levels was performed
438 using RSEM (Li and Dewey, 2011) and the TAIR10 annotation of gene models.
439 Summary statistics concerning total number of reads, total number of mapped reads
440 and number of unambiguously mapped reads are reported in Supplementary Table S5.
441 Identification of differentially expressed genes was performed by the quasi-likelihood
442 F-test as implemented by edgeR (Robinson et al., 2009). A False Discovery Rate
443 (FDR) cut-off value of 0.05 was applied for the identification of significantly
444 differentially expressed genes. Functional enrichment analysis of sets of differentially

445   expressed genes were performed by means of ShinyGO, terms from the "biological

446   process" domain of the Gene Ontology (GO) were set as the "pathway database" (Ge

447   et al., 2020). Only the top 10 terms with most statistically significant enrichment were

448   included in the graphical representation of the results enclosed in Supplementary

449   Figure S6. Quantitative real-time PCR, changes in gene expression were calculated

450   relative to *ACTIN2*. Values were either expressed as fold change variations relative to

451   the wild type ($\Delta\Delta Ct$) or expressed as -ΔCt values (Castelletti et al., 2020) and analysed

452   by fitting a factorial ANOVA model (ZT, genotype, treatment, genotype x treatment).

453   Quantitative real-time PCR primers are provided in Supplemental Table S6.

454

455   **ABA quantification**

456   ABA was quantified according to the method described by (Salem et al., 2020) with

457   minor modifications. Briefly, ABA was extracted from freeze-dried leaves (10 mg)

458   after grinding to a powder using 1 mL of pre-cooled (-20°C) extraction solvent

459   (methyl-tert-butyl-ether:methanol, MTBE:MeOH, 3:1, v:v). The extracted samples

460   were vortexed and incubated on an orbital shaker at 4 °C for 30 min. Liquid-liquid

461   phase separation was induced by adding a volume of 0.5 ml of acidified water (0.1%

462   HCl) followed by 30 min incubation on an orbital shaker at 4 °C. The samples were

463   centrifuged at 4 °C for 5 min at 13.000 x g and 1 mL from the MTBE layer was then

464   evaporated in a vacuum concentrator before re-dissolving the residue in 100 µL of

465   methanol: water (1:1, v/v). ABA was analysed by ultra-performance liquid

466   chromatography coupled to mass spectrometry (UPLC-MS/MS) analysis. The LC-MS

467   analysis was performed on a quadruple linear ion trap mass spectrometer (4000

468   QTRAP MS/MS System, SCIEX, Redwood City, U.S.A.) connected to an Acquity

469   ultra performance liquid chromatography (UPLC, Waters, Milford, MA, USA). The

470   UPLC was equipped with a reversed-phase HSS T3 C18 column (100 mm×2.1 mm×

471   1.7 µm particles, Waters). ABA was identified and quantified using a multiple reaction

472   monitoring (MRM) method (Salem et al., 2020).

473

474   **Thermal Imaging**

475   For leaf temperature analysis, plants were grown in 2-inch pots randomized in 32 cell

476   trays. Trays were also randomly cycled between the top and bottom shelves in a short-

477 day growth chamber (8 h light period, 22°C, a light intensity of 90 to 110-μmol m^{-2} s^{-1}

478). Pots were bottom watered with hyponex nutrient solution (~1 g L21) as needed. On

479 the 8th day, seedlings were thinned to leave one plant per pot. On the 29th day (when

480 plants were ~4-week-old), replicates were split into treatment and sprayed with 10 µm

481 ABA or a mock solution. These were shifted to long days (16 h light). The following

482 day, thermal imaging (FLIRA325sc) was carried out on whole plant trays at three time

483 points (ZT1, ZT8, and ZT16). Leaf temperature was measured using FLIR ResearchIR

484 Max4 software. We used the freehand ROI (Region of Interest) tool to trace the entire

485 rosette carefully to avoid background and obtained the mean temperatures for

486 individual rosettes.

487

488 **Transcription factor binding analyses**

489 The pscan software (Zambelli et al., 2009), in conjunction with the JASPAR_fam

490 matrices set (Vlieghe et al., 2006), was used to calculate transcription factor binding

491 site (TFBS) family score enrichment profiles for promoters (1000 bp upstream of

492 TSS).

493 Publicly available ChIP-seq peaks for a selection of Transcription Factors were

494 obtained from the GEO repository, under the following accessions GSE129865 (GI),

495 GSE35059 (PIF5), GSE35315 (PIF4 etiolated seedling), GSE39215 (PIF3),

496 GSE43283 (PIF1), GSE43284 (PIF 4 seedling 3 days), GSE68193 (PIF4 and PIF 5,

497 seedling, 5 days), GSE80564 (Song et al., 2016). Genomic coordinates of promoter

498 sequences, defined as -1000 bp upstream and +100 bp downstream of an annotated

499 TSS, based on the TAIR10 annotation of the reference *A. thaliana* genome were

500 obtained by a custom Perl script.

501 Intersection of promoters and ChIP-seq peaks coordinates were performed by means

502 of the bedtools intersect utility. Total number of overlaps were recorded by the means

503 of a custom Perl script. A statistical test based on the hypergeometric distribution was

504 applied to infer statistical significance. The total number peaks was used as the "total

505 number of successes" in the population (k), while the size of the population was set to

506 the total number of promoters in the genome as defined by the criteria outlined above.

507 P-values were corrected by applying the Benjamini-Hochberg procedure for the

508 control of False Discovery Rate.

509

Graphical representation and statistical analyses

Graphical representations of the data and comparisons between the raw data and published data sets were prepared with R software (R Core Team, 2020, https://www.R-project.org/). The Base R was used in combination with packages ggplot2 3.3.3 (Wickham, 2016), pheatmap 1.0.12, rstatix 0.6.0. For the PCA calculation the R base method using the singular value decomposition was applied (prcomp, R base). The input for PCA was a matrix of log2 scaled transcript abundance values for each gene in different conditions. For the final PCA displays the values referring to genes occurring as significantly altered in any given comparison were used. Heatmaps for motif analysis were generated with the pheatmap package (Raivo, 2012). Leaf temperature data were analysed by three-way repeated measures ANOVA with rstatix R package (Kassambara, 2021) considering genotype (*gi* mutant vs. wild type), treatment (ABA addition versus mock) and their interaction as fixed factors at each time point (ZT1, ZT8, ZT1) separately. The outliers in the data were identified based boxplot method and eliminated from analysis. Values above Q3+1.5xIQR (Inter Quantile Range) or below Q1-1.5xIQR were considered as outliers. Q1 and Q3 are the first and third quartile, respectively. IQR is the interquartile range (IQR = Q3 - Q1). The data was tested for normal distribution by Shapiro-Wilk's test of normality and the homogeneity of variance was assessed by Leven's test of homogeneity of variance. The significant two-way interactions were further analysed as simple main effects to investigate the effect of genotype on leaf temperatures at each level of treatment. The R script used for the leaf temperature data analysis and visualization can be found here https://github.com/BhaskaraGB/ABA_photoperiod_GI.

Data availability

The data discussed in this publication have been deposited in NCBI's Gene Expression Omnibus (Edgar et al., 2002) and are accessible through GEO Series accession number GSE181083 (https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE181083).

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551 **Author contributions**

552 BS, MC, BGB, MR, FD, DB, MAAS, investigation with help from DM and SC who

553 contributed with supervision and data curation. CT, MG, PG, TEJ and LC

554 conceptualisation and methodology. BS, MC, BGB Visualisation and formal analysis.

555 BS, MC and LC Writing – original draft. CT, MG, PG, TEJ and LC Writing – review

556 & editing.

557

558 **Conflicts of interest**

559 No conflicts of interest declared

560

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Legends to Figures

Figure 1. Strong contribution of time of the day in the regulation of drought responses.

(A) Schematic representation of the experimental design. Samples for RNA-seq were

harvested at the indicated time points (ZT1 to ZT16) in the light phase encompassing

the transition from the last short day and the first long day (photo-extension).

(B) Pyramid plots showing numbers of up-regulated and down-regulated DEGs in the

comparison of water deficit vs. well-watered conditions (dubbed as treatment and

control, respectively) in the wild type at matched time-points (ZT1 to ZT16). Right

(blue) and left (red) bars represent upregulated or downregulated genes, respectively.

(C) Heatmap illustrates the total number of DEGs in response to water deficit in the

wild type from ZT1 to ZT16. Colours indicate the overlap (%) for each pairwise

comparison according to the colour scale positioned on the right. The horizontal bar

chart on the right represents the total number of DEGs detected at each ZT/condition

combination (grey) and those which are in common with any the other ZTs/conditions

(black).

848　　(D) PCA analyses of DEGs (n = 4780). The same plot is shown in three panels with

849　　different colour codes to highlight the genotype, time points and treatment (left to

850　　right). Each dot represents one biological replicate.

851

852　　Figure 2. RNA-seq analysis of GI and ABA regulated genes reveals a phase of ABA

853　　insensitivity regulated by GI.

854　　(A, B) Pyramid plots showing numbers of up-regulated and down-regulated DEGs in

855　　the comparison of *gi-2* vs. wild type at matched time-points (ZT1 to ZT16) under

856　　control (A) or water deficit conditions (B). Right (blue) and left (red) bars represent

857　　upregulated or downregulated genes, respectively.

858　　(C) Heatmap displaying the total number of DEGs in *gi-2* compared with the wild type

859　　at ZT1 and ZT12 under different irrigation schemes. Colours indicate the overlap (%)

860　　for each pairwise comparison according to the colour scale positioned on the right. The

861　　horizontal bar chart on the right represents the total number of DEGs detected at each

862　　ZT/condition combination (grey) and those which are in common with any the other

863　　ZTs/conditions (black).

864　　(D) Boxplot of the distribution of log2(Fold Change) of DEGs between *gi-2* and wild-

865　　type plants under water deficit, at ZT12 in (from left to right); wild type water deficit

866　　vs. control condition; *gi-2* water deficit vs. control condition; *gi-2* vs. wild type under

867　　water deficit. Genes upregulated in the comparison between *gi-2* and the wild type

868　　under water deficit are represented in blue, downregulated genes are represented in

869　　red.

870

871　　Figure 3. Enrichment of PIFs, ABFs and GI binding at GI/ABA DEGs

872　　(A) Heatmap of the statistically significant (FDR < = 1e-2 in at least one comparison)

873　　enriched transcription factor binding sites at the ABA and GI DEGs at ZT1 and ZT12

874　　time points under control or water deficit conditions. See Table S2 for a complete list.

875　　(B) Pyramid plot showing number of up-regulated and downregulated DEGs in each

876　　comparison and their overlap with PIFs and ABFs ChIP-seq peaks. The following

877　　datasets - PIF4, GEO: GSE43284, GSE68193, GSE35315, PIF3, GEO: GSE39215,

878　　PIF1, GEO: GSE43283, PIF5, GEO: GSE35059, GSE68193 - were pooled to identify

879　　potential target of the PIF family of transcription factors. Similarly, candidate target

genes of ABFs were obtained by pooling GBF3, GBF2, ABF1, ABF4 (GEO: GSE80564) datasets, with no regard to experiment conditions (ABA treatment, EtOH treatment). Colour code represents binding to individual or both classes of transcription factors.

(C) Pyramid plot showing the overlap between GI ChIP-seq peaks (SD ZT8) and DEGs obtained in this study. *gi-2* vs. WT (top), *aba1-6* vs. WT (bottom). Asterisks in (A) and (B) are used to indicate a statistically significant over-representation of up regulated genes based on Fisher's exact test.

Figure 4. Deregulation of ABA signalling genes in *gi-2* mutants

(A) and (B) Volcano plots showing a selection of ABA/Cold and circadian-related DEGs in *gi-2* (top) or *aba1-6* mutants (bottom panel) at ZT12 under water deficit. LogFC is reported on the X-axis. FDR on the Y axis. Differentially expressed genes (FDR for differential expression $<= 0.05$) are represented in purple, non-differentially expressed genes (FDR > 0.05) in green. Selected genes are displayed in yellow. See also Fig. S4 for ZT1 analysis.

Figure 5. GI controls daytime variations in transpiration.

(A) Thermogram of representative plants grown under SDs under well-watered conditions and then sprayed with ABA (or mock) before shifting to LDs. Plants were grown in separate pots and photographed for thermal imaging on the second LD at the indicated time points.

(B) Leaf temperature data extracted from images shown in (A). n = 5-7 biological replicates per genotype. Values for each genotype / condition combination represent the mean and associated SD. Asterisks indicate statistically significant effect (*$p<0.05$, **$P<0.01$, ***$P<0.001$ and **** $p<0.0001$) for the genotype (Geno). A detected statistically significant Genotype X Treatment effect (G X T) is show for ZT8.

(C) Simplified model of the GI regulatory role of water deficit responses. GI protein levels increase during the day (blue dotted line) reaching a peak at approx. ZT8. Peak GI expression and function contribute to reducing ABA-depended responses, perhaps to counteract increasing ABA levels (red dotted line). GI may interfere with TFs

911 functions (e.g., the ABFs and the PIFs) for the activation of ABA signalling and

912 optimise plant growth performances under water deficit.