

Meta-analysis of transcriptomic responses to biotic and abiotic stress in tomato

Elham Ashrafi-Dehkordi¹, Abbas Alemzadeh¹, Nobukazu Tanaka² and Hooman Razi¹

- Department of Crop Production and Plant Breeding, School of Agriculture, Shiraz University, Shiraz, Iran
- ² Center of Gene Science, Hiroshima University, Kagamiyama, Higashi-Hiroshima, Hiroshima, Japan

ABSTRACT

A wide range of biotic stresses (BS) and abiotic stresses (AS) adversely affect plant growth and productivity worldwide. The study of individual genes cannot be considered as an effective approach for the understanding of tolerance mechanisms, since these stresses are frequent and often in combination with each other, and a large number of genes involved in these mechanisms. The availability of highthroughput genomic data has enabled the discovery of the role of transcription factors (TFs) in regulatory networks. A meta-analysis of BS and AS responses was performed by analyzing a total of 391 microarray samples from 23 different experiments and 2,336 differentially expressed genes (DEGs) involved in multiple stresses were identified. We identified 1,862 genes differentially regulated in response to BS was much greater than that regulated by AS, 835 genes, and found 15.4% or 361 DEGs with the conserved expression between AS and BS. The greatest percent of genes related to the cellular process (>76% genes), metabolic process (>76% genes) and response to stimulus (>50%). About 4.2% of genes involved in BS and AS responses belonged to the TF families. We identified several genes, which encode TFs that play an important role in AS and BS responses. These proteins included Jasmonate Ethylene Response Factor 1 (JERF1), SlGRAS6, MYB48, SlERF4, EIL2, protein LATE ELONGATED HYPOCOTYL (LHY), SIERF1, WRKY 26, basic leucine zipper TF, inducer of CBF expression 1-like, pti6, EIL3 and WRKY 11. Six of these proteins, JERF1, MYB48, protein LHY, EIL3, EIL2 and SlGRAS6, play central roles in these mechanisms. This research promoted a new approach to clarify the expression profiles of various genes under different conditions in plants, detected common genes from differentially regulated in response to these conditions and introduced them as candidate genes for improving plant tolerance through genetic engineering approach.

Submitted 4 September 2017 Accepted 27 March 2018 Published 17 July 2018

Corresponding author Abbas Alemzadeh, alemzadeh@shirazu.ac.ir

Academic editor Yuriy Orlov

Additional Information and Declarations can be found on page 17

DOI 10.7717/peerj.4631

© Copyright 2018 Ashrafi-Dehkordi et al.

Distributed under Creative Commons CC-BY 4.0

OPEN ACCESS

Subjects Bioinformatics, Plant Science **Keywords** Abiotic and biotic stresses, Meta-analysis, Microarray, *Solanum lycopersicum*,
Transcription factors

INTRODUCTION

As sessile organisms, plants constantly encounter a wide range of stresses that negatively affect the yield production, survival and growth of plants worldwide (*Sharma et al.*, 2013). The stresses are divided into two broad categories, abiotic stresses (AS) including a

variety of adverse environmental conditions, (i.e., drought, submergence, salinity, heavy metal contamination, nutrient deficiency, light and extreme temperatures) and biotic stresses (BS) caused by attack of living organisms, (i.e., bacteria, viruses, fungi and nematodes as well as other plants) (*Rastgoo & Alemzadeh, 2011; Liu et al., 2014*). To survive under stress conditions, complex mechanisms have evolved in plants to perceive signals from their environment, transmit through signal transduction pathways and respond to the various stresses that each of which may involve several hundred proteins (*Fujita et al., 2006; Rastgoo, Alemzadeh & Afsharifar, 2011; Atkinson, Lilley & Urwin, 2013; Sharma et al., 2013; Sami & Alemzadeh, 2016*). Functional molecular studies have figured out interactions between BS and AS signaling components with the identification a large number of genes, which are regulated by various types of transcription factors (TFs) (*Kissoudis et al., 2016*). TFs are proteins with critical roles in the control temporal and spatial gene expression during defense response that many of them are induced by more than one stress (*Berrocal-Lobo, Molina & Solano, 2002; Liu et al., 2014; Shaik & Ramakrishna, 2014; Prasad et al., 2016*).

The better understanding of plant stress response mechanisms under various stresses can give us a better view of how to improve the worldwide food production (*Sharma et al.*, 2013). It is necessary to learn more about molecular mechanisms of TFs, MAPKs and ROS signal pathways, and hormone signaling, and the study of genes encoding regulatory molecules, especially TFs, can be more efficient than single-function genes, because TFs enable to regulate various genes to improve plant tolerance against stresses (*Seki et al.*, 2002).

The bioinformatic approach is an emerging and rapidly developing field which allows researchers to analyze biological data (*Khatri & Drăghici*, 2005; *Chang et al.*, 2013; *Zinati*, *Alemzadeh & KayvanJoo*, 2016). Recent advances in genome-wide transcriptome analysis methods, such as RNA sequencing and microarrays, enable researchers to simultaneously study the expression of thousands of genes and their co-expression partners under various stresses (*Khatri & Drăghici*, 2005; *Sharma et al.*, 2013; *Rest et al.*, 2016). Meta-analysis can be used to estimate differential gene expression between normal and stress conditions and find out some genes that their products are key molecules in response to the stress (*Tseng, Ghosh & Feingold, 2012*; *Shaar-Moshe*, *Hübner & Peleg, 2015*).

Today, with recent advances in high-throughput technology, a large number of biological data has become accessible via online resources, and we are now able to obtain results that more reliable by combining information from multiple sources. In this way, statistical methods can be used as useful and powerful tools to identify suitable candidate genes for test under experimental conditions to use in plant breeding programs. The combination of *P*-values is a common method in microarray meta-analysis and recently used in many studies (*Tseng, Ghosh & Feingold, 2012*). A meta-analysis of microarray using 13 experiments querying cytokinin-regulated gene expression changes in *Arabidopsis* combined with empirically defined filtering criteria identified a set of 226 genes differentially regulated by cytokinin (*Bhargava et al., 2013*). In another research project, a meta-analysis of 10 studies identified 4,015 genes showing significant

differential expression in response to water limitation compared to control conditions in *Arabidopsis* (*Rest et al.*, 2016).

Moreover, combining gene expression information across species can improve the ability to identify core gene sets with high evolutionary conservation. For example, a new cross-species meta-analysis of progressive drought stress at the reproductive stage was developed using *Arabidopsis*, rice, wheat and barley. Based on this dataset, important shared genes and metabolic pathways involved in whole plant adaptation to progressive drought stress across-species were identified (*Shaar-Moshe*, *Hübner & Peleg*, 2015). In this study, 225 differentially expressed genes (DEGs) shared across studies and taxa were identified and gene ontology (GO) enrichment and pathway analyses revealed that the shared genes were classified into functional categories involved predominantly in metabolic processes, regulatory function and response to stimulus (*Shaar-Moshe*, *Hübner & Peleg*, 2015).

Tomato (*Solanum lycopersicum*) is one of the most important food crops belongs to Solanaceae family, one of the largest and most important families of flowering plants. In addition, the tomato is a suitable model system to study environmental signal transduction components involved in BS and AS (*Knapp et al.*, 2004; *Arie et al.*, 2007). However, there is no a large-scale transcriptome analysis via meta-analysis of stress response studies in tomato to detect proteins, especially TFs, involved in various stresses.

In this study, for the first time we performed a large-scale comparative transcriptome analysis via meta-analysis of stress response studies in tomato using microarray gene expression data implemented on a single platform (Affymetrix Tomato Genome Array, Santa Clara, CA, USA). The goal of this meta-analysis was to detect DEGs involved in BS and AS and detect DEGs shared by both stresses. In addition, we focused on the identification of TFs have magnitude effects on the plant cells via up- and down-regulation of various genes under stress conditions. Other analysis methods, such as GO and Network analysis, were used to explain the complexity of tomato response to AS and BS. We also discussed the functions of the identified TFs in stress response.

METHODS

Selection microarray studies

In this study, a meta-analysis of stress response was performed in tomato using publicly available microarray gene expression data implemented on a single platform (Affymetrix Tomato Genome Array, Santa Clara, CA, USA). Expression data of tomato plants exposed to BS and AS were combined. The raw expression data of various experiments were obtained from the Array Express from EBI (https://www.ebi.ac.uk/arrayexpress/).

Identification of differentially expressed genes

Microarray expression data from each source study was pre-processed separately as individual datasets. The raw data were normalized using Robust Multi-array Average background correction and quantile normalization. The data were processed in R version 3.2.2 (*R Core Team*, 2015). Meta-analysis of AS and BS was performed separately to identify DEGs involved in stress conditions. Genes that showed significant *P*-value

(false discovery rate (FDR) of 5%), meaning that an estimated five percent of the DEGs are false positive, were considered as DEGs (*Campain & Yang, 2010*; *Chang et al., 2013*). Two methods of the combination of *P*-value have used: Fisher's method and maxP (*Wilkinson, 1951*; *Fisher, 1992*). The Fisher's combined probability test sums the logarithm-transformed *P*-values obtained from individual studies, under the null hypothesis, and follows a chi-squared distribution with 2k degrees of freedom, where *k* is the number of studies being combined (*Rhodes et al., 2002*). A very small *P*-value in just one study can be adequate to make statistical significance, even if the same genes are not significant in any other study. The maxP method takes maximum *P*-values across studies. It follows a beta distribution with parameters *K* (number of transcriptomic studies which are combined for a meta-analysis where each study contains G genes) and 1. This method targets on DEGs that have small *P*-values in all studies. (*Chang et al., 2013*; *Song & Tseng, 2014*).

Gene analysis

The genes selected by meta-analysis were further analyzed and characterized. In the first instance, the results were compared with the plant genes in the Plant Transcription factor & Protein Kinase Identifier and Classifier, iTAK (http://bioinfo.bti.cornell.edu/cgi-bin/itak/index.cgi) and the Plant Transcription Factor, PlantTFDB (http://plntfdb.bio.uni-potsdam.de/v3.0/) databases. The list includes 1,845 genes of tomato which encoding the TFs that directly or indirectly involved in signaling pathway and response to AS and BS. Venn's diagrams (http://bioinfogp.cnb.csic.es/tools/venny/) were used to represent co-occurrence of DEGs. GO enrichment analysis of the DEGs (such as identifying biological processes) was conducted using the AgriGO platform (*Du et al.*, 2010).

Network analysis

The Search Tool for the Retrieval of Interacting Genes/Proteins, STRING 10.5 (http://www.mybiosoftware.com/string-9-0-search-tool-retrieval-interacting-genesproteins. html) (*Szklarczyk et al.*, 2014) database was used to figure out all functional relations between the genes and their occurrence patterns across various genomes. The database makes relations based on several lines of evidences: the empirical evidence from protein-protein interaction assays, co-expression data from the NCBI Gene Expression Omnibus database, the extraction of information from other databases, coexistence of the genes in the same organisms, conserved gene neighborhood in known genomes, gene fusion events, pathway annotation from other resources such as the Kyoto Encyclopedia of Genes and Genomes or GO databases, and automated text-mining tools (*Szklarczyk et al.*, 2014). STRING calculates a confidence value for those interactions according to the pieces of evidences from above 0.4 to 0.9, as the medium to highest score, respectively.

RESULTS AND DISCUSSION

Identification of genes involved in biotic and abiotic stresses

After searching in the database, microarray data from 23 different experiments were selected and analyzed. To obtain a global analysis, 391 microarray samples including 213 AS from different categories: drought, heat, salinity, light, hormone, mineral and

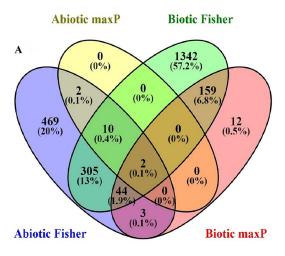
heavy metals deficiencies/toxicities and 178 BS that caused by viruses, nematode, fungi, bacteria and pests in tomato (wild type and transgenic) were chosen (Table S1). We found a total of 1,862 genes differentially regulated in response to BS, was far greater than that induced by AS, 835 genes, with a FDR \leq 0.05 by Fisher's statistical method (Tables S1–S3). We also figured out 15.4% or 361 DEGs with the conserved expression between AS and BS by Fisher's statistical method, suggesting that these genes and their associated cell-signaling pathways are regulated in a similar way in a wide range of stresses (Fig. 1A; Table S5). Using maxP method, a total of 14 genes differentially regulated in response to AS and a total of 220 genes differentially regulated in response to BS were found. We figured out 0.9% or 2 DEGs (LES.1399.1 and LES.3969.1) with the conserved expression between AS and BS by maxP method (Fig. 1A; Tables S2-S4). These methods used in other studies to identify important genes in different physiological processes in other plants. Shaar-Moshe, Hübner & Peleg (2015) identified 225 DEGs across-species by using 17 microarray experiments of drought stress at the reproductive stage of Arabidopsis, rice, wheat and barley. Fisher's and maxP methods are two popular microarray metaanalysis methods used in the literature that have different strength for detecting different types of DEGs (Tseng, Ghosh & Feingold, 2012; Wang et al., 2012). Our results also showed that the two used methods detected different sets of DEGs, suggesting different assumptions behind the methods. Among these methods, Fisher's method is more popular in the microarray meta-analysis studies (Wang et al., 2012); hence, we were more interested in detecting DEGs across all studies through Fisher's statistical method.

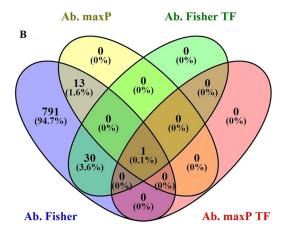
Gene ontology

We focused on the 361 DEGs common to both AS and BS and subjected them to GO and network analysis to explore other possible functions of them. GO enrichment analysis of the DEGs showed that shared DEGs represented genes involved in main biological and cellular processes. Among them, the greatest percentage of genes involved in the cellular process (>76% genes), metabolic process (>76% genes) and response to stimulus (>50%). One of largest functional group of DEGs was associated with metabolic processes (e.g., carbohydrate metabolic process, lipid metabolic process, protein metabolic process, nucleobase, nucleoside, nucleotide and nucleic acid metabolic process, cellular amino acid and derivative metabolic process and regulation of primary metabolic process).

In addition, another large functional group of DEGs was associated with cellular processes (e.g., cellular metabolic process, regulation of cellular process and cellular response to stimulus), the show a significant rearrangement in plant metabolism as part of stress adaptation (*Shaar-Moshe*, *Hübner & Peleg*, 2015).

On the other hand, the lowest percentage of genes involved in positive regulation of biological process (5%) (Fig. 2; Table S6). The most significant GO terms for cellular process were cellular metabolic process (FDR: 1.8E-37), organic acid metabolic process (FDR: 9.2E-29), cellular ketone metabolic process (FDR: 1.2E-28) in metabolic process were primary metabolic process (FDR: 7.8E-39), cellular metabolic process (FDR: 1.8E-37) and organic acid metabolic process (FDR: 9.2E-29). Moreover,





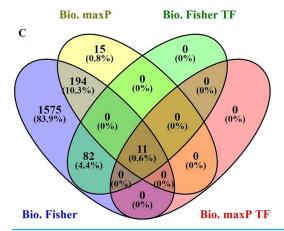


Figure 1 Identification of genes involved in biotic and abiotic stresses. Comparison of differentially expressed genes (DEGs) under abiotic and biotic stress responses. (A) Four-way Venn diagrams showing co-occurrence of DEGs in response to various abiotic and biotic stresses by two meta-analytical approaches: Fisher and maxP methods. (B) Four-way Venn diagrams showing number of transcription factors DEGs in all identified abiotic stresses by two different meta-analytical statistical methods: Fisher and maxP. (C) Four-way Venn diagrams showing number of transcription factors DEGs in all identified biotic stresses by two different meta-analytical statistical methods: Fisher and maxP.

Full-size DOI: 10.7717/peerj.4631/fig-1

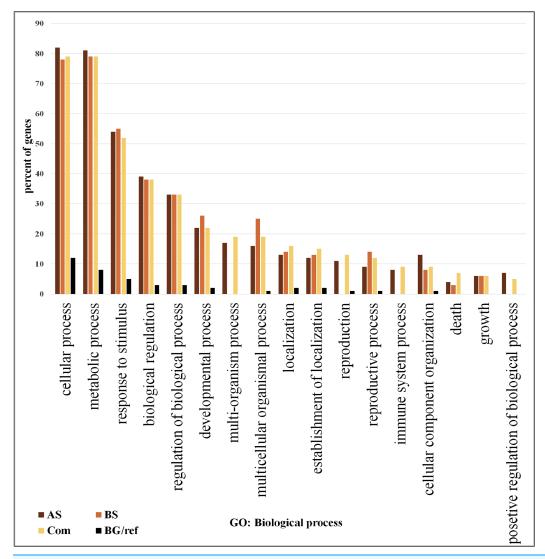


Figure 2 Gene ontology analysis. Frequency of most representative biological process terms. Com: Differentially expressed genes obtained by common genes in biotic and abiotic stresses, AS: Differentially expressed genes obtained by abiotic stresses, BS: Differentially expressed genes obtained by biotic stresses. BG: The frequency of these terms in the reference genes, Tomato Locus set. Gene Ontology analysis made in the AgriGO platform (FDR = 5%). More details in Tables S10 and S11.

Full-size DOI: 10.7717/peerj.4631/fig-2

in response to stimulus were response to abiotic stimulus (FDR: 3.4E-40), response to biotic stimulus (FDR: 9.4E-26), and response to stress (FDR: 9.4E-26). Some of these genes possess important roles in DEGs shared by AS and BS networks (Table S7). A genome-wide transcriptional analysis was performed just for Fe deficiency and was identified 97 DEGs by comparing Fe-deficient and Fe-sufficient in roots of tomato. These transcripts are related to the physiological responses to the nutrient stress resulting in an improved iron uptake, including regulatory aspects, translocation, root morphological modification, and adaptation in primary metabolic pathways, such as glycolysis and TCA cycle (*Zamboni et al.*, 2012). Another research project showed that the largest functional groups of DEGs was associated respectively to metabolic processes

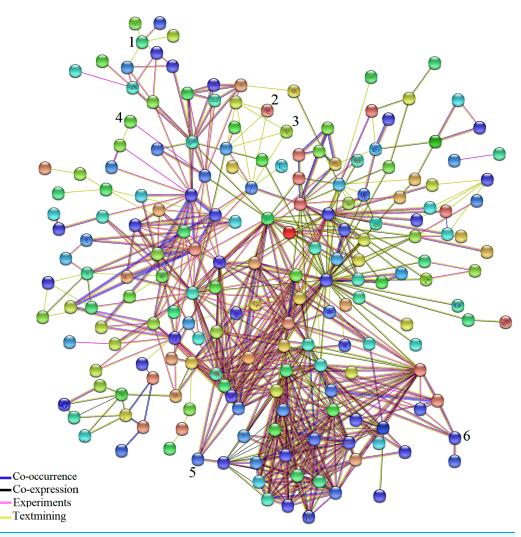


Figure 3 Network analysis of the DEGs identified in biotic and abiotic stresses. Network analysis of the 361 common DEGs identified in biotic and abiotic stresses. Network was generated by STRING (version 10.5) database and represents all connections of those genes with a confidence score > 0.4. The connection colors show the types of evidence for concluding association: co-occurrence of those genes in the same organisms (dark blue), co-expression (black), experimental protein–protein interaction data (pink) and literature text-mining (yellow). A number above hub represent TF. (1) JERF1, (2) MYB48, (3) protein LHY, (4) EIL3, (5) EIL2, (6) WRKY26.

(e.g., amino acid and carbohydrate metabolism), regulator function (e.g., protein degradation and transcription) and response to the stimulus that was identified by the research of *Shaar-Moshe*, *Hübner & Peleg* (2015).

Network analysis

Network analysis using STRING 10.5 identified the connections among DEGs common to AS and BS (Fig. 3; Table S7). In addition, some genes that possess an important role in the three biological pathways; cellular process, metabolic process, response to stimulus were detected. For example, Solyc02g090890.2.1 (*ZE*: plays an important role in resistance to stresses, seed development and dormancy), Solyc04g049350.2.1 (*CS1*: biosynthesis of

aromatic amino acids), Solyc07g052480.2.1 (LOC544276: involved in storage lipid mobilization during the growth of higher plant seedling), Solyc07g049530.2.1 (*ACO1*: involved in ethylene biosynthesis via S-adenosyl-L-methionine), Solyc01g099630.2.1 (*XTH1*: an essential component of the primary cell wall, and thereby participates in cell wall construction of growing tissues), Solyc05g055230.1.1 (*RPS17*: ribosomal small subunit assembly), Solyc12g098350.1.1 (*infA*: no specific function has so far been attributed to this initiation factor; however, it seems to stimulate more or less all the activities of the other two initiation factors, IF-2 and IF-3), Solyc02g082920.2.1 (*CHI3*: defense against chitin containing fungal pathogens), Solyc06g082600.2.1 (*UBC3*: mediates the selective degradation of abnormal and short-lived proteins), Solyc01g096480.2.1 (LOC543657: probable thiol-disulfide oxidoreductase that may play a role in proper chloroplast development and negatively regulates Cf-9/Avr9-mediated cell death and defense responses), Solyc07g047850.2.1 (*CAB4*: it receives and transfers excitation energy to photosystems).

The results of network analysis also showed some of the TFs have connections to other molecules. Based on pieces of literatures reviewing, detected TFs in this study, Jasmonate Ethylene Response Factor 1 (JERF1), EIL3, EIL2, SIGRAS6, Protein LATE ELONGATED HYPOCOTYL (LHY) and MYB 48 affected by a wide range of stresses and possibly play essential and significant roles in multiple stress responses which is in agreement with their functional roles in BS and AS responses. Expression of *JERF1* was induced by salt, ethylene, MeJA and abscisic acid (ABA) treatments in tomato that may be pointed to a central role for JERF1 in different signal transduction pathways (Zhang et al., 2004). It has been also reported that JERF1 activates ABA biosynthesis related genes such as NtSDR (short-chain dehydrogenase/reductase) and enhanced cold and salt stress tolerance in transgenic tobacco (Wu et al., 2007; Ouyang et al., 2007). It showed that the expression of EIL2 and EIL3 was induced by ethylene and salinity treatments (Ying et al., 2004) and SIGRAS6 is one of the most important members of the tomato GRAS family in disease resistance and mechanical stress (Mayrose et al., 2006). Different studies have shown the large family of MYB has essential roles in plant growth, development, primary and secondary metabolism and response to BS and AS (Oh, Park & Han, 2003; Van der Ent et al., 2008; Dou et al., 2016). It was also reported that LHY controls the expression of a large number of ABA-responsive genes by binding directly to their promoters (Huang et al., 2007). These results support that the detected TFs in our study might play an important role in BS and AS responses.

Identification of transcription factors involved in biotic and abiotic stresses

In this research, because of the importance of TFs as a powerful tool for the manipulation of complex metabolic pathways, we have tried to identify genes encoding them. According to the PlantTFDB, 1,845 genes from tomato encoding TFs were identified and classified into 58 families that directly or indirectly involved in signaling and response to AS and BS (Table S8). Based on our results, from the 1,862 genes response to BS selected by Fisher's statistical method, 93 genes (5%) encode TFs (Fig. 1C; Table S9).

TF family	No. of TF in abiotic	No. of TF in biotic	TFBS	
ARF	1	3	TGTCTC auxin response elements (AuxRE)	
BBR-BPC	1	_	(GA/TC)8 and GAGA-binding	
bHLH	3	9	E-box (5'-CANNTG-3')	
C2H2	_	9	A DNA element that contains an AGCT core	
СЗН	_	1	Unknown	
ERF	5	18	GCC box is an 11 bp sequence (TAAGAGCCGCC)	
EIL	3	2	EIL2 BS in ERF1 (TTCAAGGGGGCATGTATCTTGAA)	
FAR1	_	1	FBS for FHY3-FAR1 binding site	
GATA	_	1	WGATAR ($W = T$ or A; $R = G$ or A) motifs	
GRAS	1	7	cis-element AATTT	
HSF	-	3	This consists of a tandem of inverted repeats of the sequence GAA, generating a perfect HSE, TTCnnGAAnnTTC	
LBD	_	1	Core sequence CGGC	
MIKC_MADS	-	1	Keratin-like coiled-coil domain	
MYB and MYB_related	2	8	(CNGTTR), (GKTWGTTR), TAACPy sequence (only one AAC sequence) and (GKTWGGTR; R, A or G; K, G or T; W, A or T) and EE, AGATATTT	
NAC	5	4	NAC recognition site (NACRS), NAC binding element (SNBE) with a longer and variable sequence ([T/A]NN[C/T][T/C/G]TNNNNNNNA[A/C]GN[A/C/T][A/T])	
NF-YB	_	3	CCAAT binding	
NF-YC	1	_	Core nucleotide sequence CCAAT	
PHD	_	1	Unknown	
SPL proteins	_	1	SBP-box, TNCGTACAA	
TALE	2	1	Pbx:Meinox binding site (CTGTCAATCA)	
TCP	_	2	GGNCCCAC sequences and s G(T/C)GGNCCC	
TIFY	_	1	Unknown	
VOZ	1	_	GCGTNx7ACGC	
WRKY	3	9	W box (TTGACY; Y, C or T)	
YABBY	-	1	WATNATW $(W = T \text{ or } A; R = G \text{ or } A)$	
ZIP and HD-ZIP	3	6	Motif AATNATT	

Notes:

Transcription factor and binding site for each, in biotic and abiotic stresses. TF, Transcription factor; TFBS, Transcription factor binding site.

Ethylene-responsive transcription factors (ERF) family has the greatest number of genes (18 genes), and C3H, FAR1, GATA, LBD, MIKC-MADS, PHD, SLP protein, TALE, TIFE and YABBY families have the lowest number of genes (each family has only one gene) (Table 1). From the 835 genes response to AS selected by Fisher's statistical method, 31 of those (3.7%) encode TFs (Fig. 1B; Table S10). ERF and NAC families have the greatest number of genes (five genes), and ARF, BBR-BPC, GRAS, NF-NY and VOZ families have the lowest number of genes (each family has only one gene) (Table 1).

Since TFs are the key regulators of plant growth, development and metabolisms via up- and down-regulation of various genes, identification of the genes encoding these proteins and transferring them into plants to improve their stress tolerance would be

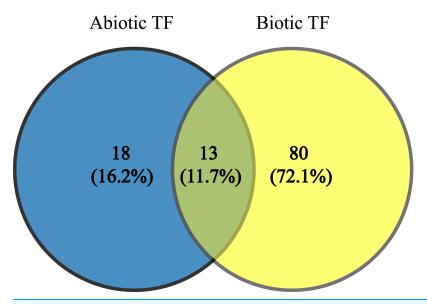


Figure 4 Common transcription factors in biotic and abiotic stresses. Two-way Venn diagram showing the common transcription factors between abiotic and biotic stresses.

Full-size MDOI: 10.7717/peerj.4631/fig-4

interesting. These TFs (JERF1, Pti6, SlERF1, SlERF4, SlGRAS6, MYB48, protein LHY, EIL2, EIL3, WRKY 11, WRKY 26, inducer of CBF expression 1 (ICE1)-like and bZIP TF) can be used as candidate genes in molecular breeding programs because these TFs are common to AS and BS; also, our results indicate they are being affected by a wide range of stresses. In addition, they possibly play essential and significant roles in multiple stress responses, which is in agreement with their functional roles in BS and AS responses (Gu et al., 2002; Ying et al., 2004; Zhang et al., 2004, 2010; Borrás-Hidalgo et al., 2006; Mayrose et al., 2006; Wu et al., 2007; Yáñez et al., 2009; Sharma et al., 2010; Chen et al., 2012; Kim, Stork & Mudgett, 2013; Feng et al., 2013; Li, Luan & Liu, 2015). The results of network analysis also showed some of these proteins have central roles in BS and AS responses (Fig. 3). The two-way Venn diagram indicated that the number of common TFs between BS and AS was 13 while 18 and 80 TFs were uniquely identified in AS and BS, respectively (Fig. 4).

ERF family

The ERF has been shown to play a key role during plant development and stress tolerance mechanisms. Overexpression of ERF family genes in different plant species such as *Arabidopsis*, tobacco, rice and tomato has been shown to increase tolerance to a wide range of BS and AS (*Sharma et al.*, 2010). The tomato has 112 AP2/ERF superfamily genes (*Sharma et al.*, 2010, 2013). Our results showed that four of 13 TFs shared between AS and BS belonged to ERF family, including JERF1, SIERF4, SIERF1 and pti6 (Table 2).

Jasmonate Ethylene Response Factor 1, has critical roles in plant abiotic stress responses, such as salinity, low temperature, dehydration and various BS in different plant species (*Zhang et al.*, 2004, 2010; *Wu et al.*, 2007). It has been shown that the expression of *JERF1* in tomato was induced by salt, ethylene, MeJA and ABA treatments

Table 2 Detected transcription factors.					
Transcription factor family	Transcription factor	Function	References		
ERF	JERF1	Salinity, low temperature, drought and various biotic stresses	Zhang et al. (2004), Wu et al. (2007), Sharma et al. (2010), Zhang et al. (2010)		
	SIERF4	Abiotic stress and pathogen (Xanthomonas campestris)	Sharma et al. (2010), Kim, Stork & Mudgett (2013)		
	Pti6	Pathogen	Gu et al. (2002), Chakravarthy et al. (2003), Sharma et al. (2010)		
	SlERF1	Abiotic stress and pathogen (Plectosphaerella cucumerina and Botrytis cinerea)	Gu et al. (2002), Mayrose et al. (2006)		
GRAS	SlGRAS6	Abiotic stress, mechanical stress and pathogen (<i>Pseudomonas syringae</i> pv. tomato)	Mayrose et al. (2006)		
MYB family and MYB-related family	MYB48	Salt, drought, cold and pathogen	Xiong et al. (2014)		
	LHY	Low temperature and pathogen	Huang et al. (2007), Alves et al. (2013), Grundy, Stoker & Carré (2015), Seo & Mas (2015)		
EIL	EIL2	Salinity and ortholog in tobacco resistance against pathogene (<i>Peronospora hyoscyami</i>)	Ying et al. (2004), Borrás-Hidalgo et al. (2006)		
	EIL3	Salinity and ortholog in tobacco resistance against pathogene (<i>Peronospora hyoscyami</i>)	Ying et al. (2004), Borrás-Hidalgo et al. (2006)		
WRKY	WRKY11	Drought, heat tolerance and pathogen	Wu et al. (2009), Li et al. (2011), Chen et al. (2012), Alves et al. (2013)		
	WRKY26	Heat stress and pathogen	Li et al. (2011), Chen et al. (2012), Alves et al. (2013)		
ICE1-like bHLH	ICE1-like	Cold, chilling, osmotic and salt	Chinnusamy et al. (2003), Feng et al. (2013), Huang et al. (2015)		
bZIP	bZIP11	Abiotic stress and pathogen	Hummel et al. (2009), Alves et al. (2013)		

Detected transcription factors are affected by a diverse set of biotic and abiotic stresses.

that may be pointed to a central role for JERF1 in different signal transduction pathways (Zhang et al., 2004). Overexpression of JERF1 enhanced the tolerance of transgenic tobacco and rice plants to high salinity, low temperature, osmotic stress and pathogen attack by regulating the expression of downstream stress-responsive genes and a number of pathogenesis-related (PR) genes due to binding to DRE/CRT and GCC-box cis-elements (Wu et al., 2007; Zhang et al., 2010). Expressing JERF1 in tobacco induced expression of GCC box-containing genes such as CHN50, Prb-1b, GLA, NtSDR and osmotin that subsequently resulted in enhanced tolerance to salt stress and low temperature (Zhang et al., 2004; Wu et al., 2007). JERF1 up-regulated the expression of genes related to osmolyte synthesis such as OsP5CS and OsSPDS2, which, has increased the production of osmolytes. It also induced the expression of OsABA2 and Os03g0810800 genes (containing DRE element in their promoter regions), which encode two key enzymes involved in ABA biosynthesis and increased the synthesis of ABA in transgenic rice that resulted in elevated tolerance to drought (*Zhang et al.*, 2010). JERF1 may also be located downstream of EIN3 in the ethylene signaling pathway (Zhang et al., 2004). It has been previously reported that SIERF4, an ERF in tomato, is a XopD, a type III secretion effector from Xanthomonas campestris, destabilize SIERF4 and this task has been proposed to promote bacterial growth (Kim, Stork & Mudgett, 2013). SIERF4 reduces ethylene production and, in turn, limits disease symptom development (Kim, Stork & Mudgett, 2013) (Table 2). Pti6 belonged to the ERF family in the plant that binds specifically to the GCC-box *cis*-element present in the promoter regions of many PR genes and induced the expression of a wide range of these genes that have key roles in the plant defense. It was shown that the expression of jasmonic acid- and ethylene-regulated genes, such as PR3, PR4, PDF1.2 and Thi2.1, was being differently affected by Pti4 or other similar genes, Pti5, or Pti6 (Gu et al., 2002; Chakravarthy et al., 2003) (Table 2). The gene encoding SIERF1 was another detected gene that probably has an important role in response to various stresses. It has been shown that the overexpression of SIERF1 in Arabidopsis can confer resistance to necrotrophic fungi like Plectosphaerella cucumerina and Botrytis cinerea (Mayrose et al., 2006) (Table 2).

GRAS family

SIGRAS6 was another identified TF that is one of the most important members of tomato GRAS family in disease resistance and mechanical stress (*Mayrose et al., 2006*). SIGRAS6 is required for tomato disease resistance to the bacterial pathogen *Pseudomonas syringae* pv. tomato. It has been shown that the suppression of *SIGRAS6* gene expression through virus-induced gene silencing reduced resistance of tomato to *Pseudomonas syringae* pv. tomato (*Mayrose et al., 2006*) (Table 2).

MYB family and MYB-related family

The MYB is a large TF family in plants that have essential roles in plant growth, development, primary and secondary metabolism and response to BS and AS (Oh, Park & Han, 2003; Van der Ent et al., 2008; Dou et al., 2016). Most of MYB genes involved in response to various AS, belong to the R2R3-type group. With the comparison of OsMYB48-1 expression with other MYB proteins, a special role for this gene in response to the AS in rice can be suggested. It has been shown that its expression was strongly induced by ABA, H₂O₂, PEG, and dehydration stresses, while slightly induced by salt and cold stresses (Xiong et al., 2014). It was reported that the overexpression of OsMYB48-1 regulated the expression level of some stress responsive-genes, such as RAB21, RAB16D, RAB16C and LEA3, under drought stress conditions (Xiong et al., 2014) (Table 2). The LHY TF plays a critical role in the regulation of low temperature stress response via controlling expression of CBF 1, 2 and 3, COR27 and COL1 genes and contribute to plant defense through the circadian regulation of stomatal aperture (Grundy, Stoker & Carré, 2015; Seo & Mas, 2015). It was also shown that LHY controls the expression of a large number of ABA-responsive genes by binding directly to their promoters (Huang et al., 2007).

EIL family

It has been shown that the expression of *EIL2* and *EIL3*, which encode TFs in ethylene signal transduction, was induced by ethylene and salinity treatments (*Ying et al.*, 2004). It also reported that its ortholog in tobacco is required for resistance against *Peronospora hyoscyami* (*Borrás-Hidalgo et al.*, 2006) (Table 2).

WRKY family

Different studies have shown that the members of WRKY family contribute multiple biological processes such as plant growth and development. They also have important roles in a wide range of BS and AS such as drought, salt, invasion of pathogens, implying that the members of this family may be putative regulators in response to various BS and AS (*Huang et al.*, 2012; *Li*, *Luan & Jin*, 2012). Two of 13 TFs, *WRKY11* and *WRKY26*, which are involved in both AS and BS, belong to the WRKY family (Table 2). In other plant species, it was also shown that the members of this family have a key role in AS and BS responses (*Li et al.*, 2011; *Chen et al.*, 2012). It was reported that *OsWRKY11* involved in drought and heat tolerance in rice (*Wu et al.*, 2009). It was shown that, in *Arabidopsis thaliana*, *WRKY26* positively regulates the cooperation between heat shock proteins and ethylene-activated signaling pathways that mediate responses to heat stress (*Li et al.*, 2011). It has also been shown that the promoters of some heat tolerance-related genes including *Hsp* and *Hsf* genes such as *HsfA2*, *HsfB1*, *Hsp101*, *and MBF1c* contain W-box sequences that are recognized by WRKY proteins (*Li et al.*, 2011).

ICE1-like bHLH family

Inducer of CBF expression 1-like encodes a MYC-type basic helix-loop-helix (bHLH) TF (*Chinnusamy et al.*, 2003). ICE1 is an important inducer of *CBF3/DREB1A* which regulates cold stress tolerance (*Chinnusamy et al.*, 2003; *Huang et al.*, 2015; *Feng et al.*, 2013) (Table 2). In *Arabidopsis*, *ice1* mutants showed decreased cold and chilling tolerance, whereas *ICE1* overexpression in *Arabidopsis* increased the cold tolerance of transgenic plants (*Chinnusamy et al.*, 2003). Furthermore, osmotic and salt tolerance were also much markedly increased in transgenic tobacco carrying *SlICE1* from tomato by increasing the expression level of *DREB1/CBF* and their target genes (*Feng et al.*, 2013) (Table 2).

bZIP family

Transcription factors of the bZIP family regulate various biological processes in plant growth and development, including morphogenesis and seed formation, as well as in AS and BS responses by binding to promoters of specific target genes to up- or down-regulate their expression (*Li et al.*, 2015). WRKY TFs, bZIP domain and MYB TFs involved in plant defense against pathogens (*Alves et al.*, 2013) (Table 2). TF bZIP11 is a member of the S1 class of the bZIP family, and it was shown that *bZIP11* mRNA translation is repressed in response to sucrose (*Hummel et al.*, 2009). It has been shown that sucrose repression of *bZIP11* mRNA translation depends on the presence of the 5'-leader sequence upstream of *bZIP11* gene (*Hummel et al.*, 2009). It has also been shown that sucrose

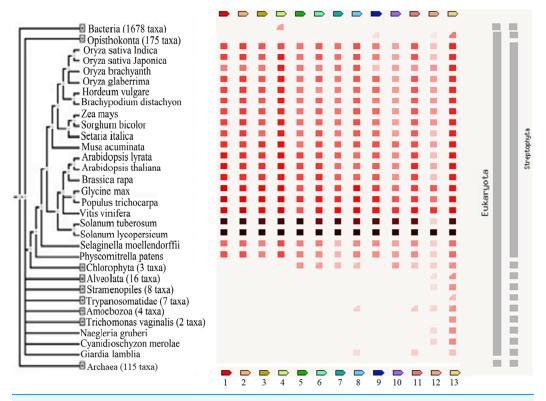


Figure 5 The presence or absence of the transcription factors across various organisms. Transcription factor occurrence patterns across various genomes. (1) EIL2, (2) EIL3, (3) ZIP, (4) GRAS6, (5) SIERF4, (6) SIERF1, (7) JERF1, (8) WRKY26, (9) ICE1-like, (10) pti6, (11) WRKY11, (12) protein LHY, (13) MYB48. The intensity of the color of the red square reflects the amount of conservation of the homologous protein in the species. White color (No similarity detectable) and black/red color (100 similarity detectable).

Full-size DOI: 10.7717/peerj.4631/fig-5

signaling has an important role in the translational control of *bZIP11* in response to stress conditions (*Hummel et al.*, 2009).

These reports implied that these TFs are important regulators of biological processes and may be useful in the breeding of plants to improve their tolerance against adverse stresses. The presence or absence of the TFs across various plant species and other organisms including bacteria, archaea and other eukaryotes is shown in Fig. 5. A high homology was observed for all TFs except protein LHY between *S. lycopersicum* and *S. tuberosum* and the highest percentage of similarity was found between *Vitis vinifera*, *Populus trichocarpa* and *Glycine max*. The results showed that GRAS6 is the only TF present in bacteria with a low similarity to that of tomato GRAS6. Just three of the identified TFs, ICE1-like, protein LHY and MYB48, are present in Opisthokonta that the homology between MYB48 and those of Opisthokonta was the highest. MYB48 is the only protein, which is present in all organisms except bacteria and archaea (Fig. 5). It has been shown that MYB is one of the largest and most diverse TF families, which are the key factors in regulatory networks controlling metabolism, development and responses to BS and AS (*Roy, 2016*). The members of MYB family found in all eukaryotes and molecular studies in *Arabidopsis* has been shown the role of this family in the epigenetic

regulation of stress responses in plants (*Roy, 2016*). The results showed MYB48 as one of the members of MYB family possesses an important role in the network of abiotic and biotic responses (Fig. 3). In addition to MYB48, three other TFs, WRKY11, WRKY26 and protein LHY, were observed in *Amoebozoa* with low similarity to those of tomato (Fig. 5). As well as *Amoebozoa*, the WRKYs were observed in *Giardia lamblia*, also known as *Giardia intestinalis* (Fig. 5). It was previously thought that WRKYs are specific to plants and algae (*Eulgem & Somssich, 2007*) but recent studies have shown their presence in human protozoan parasite *G. lamblia* and *Dictyostelium discoideum* (slime mold) (*Wu et al., 2005*). It has been reported that WRKYs also have important roles in other physiological processes. It was shown that the expression of genes encoding proteins of cell wall increased, when a *WRKY-like* gene overexpressed in *G. lamblia* (*Pan et al., 2009*).

The results showed that protein LHY present in all eukaryotes except *G. lamblia*, *Trichomonas vaginalis* and *Trypanosomatidea* (Fig. 5) and also suggested a central role for this protein in the network of abiotic and biotic responses (Fig. 3). The results of other studies also support an important role for this protein and suggested that this protein involved in circadian clock control (*Lu et al.*, 2009).

CONCLUSION

Transcriptome-wide studies by meta-analysis allowed us to narrow down DEGs to a small number. DEGs represent the genes in different networks and signaling pathways that enable us to figure out which genes play the major role in various stresses and determine the overlap between BS and AS response pathways.

This study was designed to identify genes, which, have important roles in abiotic and biotic tolerance mechanisms in tomato and focused on those shared by both stresses. The novelty of the work presented here is that a large meta-analysis was used to identify important genes in tomato as a model of Solanaceae crops. Meta-analysis has been previously used to figure out important genes that play critical roles in response to BS or AS in other plants such as Arabidopsis (Shaar-Moshe, Hübner & Peleg, 2015; Rest et al., 2016). However, this is the first time that the tomato was selected as the target plant to detect genes involved in a wide range of BS and AS responses. In previous studies, usually one or a few stresses were considered, but here a wide range of stresses was studied simultaneously. We identified 1,862 and 835 genes differentially regulated in response to BS and AS, respectively, and 361 genes identified in this work are to be involved in cross-talk between BS and AS responses. Since TFs regulate the expression of a wide range of genes under stress conditions, we focused on genes which encode TFs and it was revealed that close to 5% of identified genes encoding TFs. Among them, the expression of genes encoding JERF1, GRAS6, MYB48, SIERF4, EIL2, protein LHY, SIERF1, WRKY 26, bZIP TF, ICE1-like, pti6, EIL3 and WRKY 11 are being affected by a diverse set of BS and AS. This study highlighted the importance of detected TFs as regulators of various stress responses, reinforcing the role of these TFs in regulating plant tolerance to a board spectrum of stresses (Borrás-Hidalgo et al., 2006; Yáñez et al., 2009; Sharma et al., 2010; Feng et al., 2013;

Li, Luan & Liu, 2015). The results suggest that detected TFs in this work play important roles in plant tolerance response to various stresses, and the data generated will be helpful in conducting functional genomics studies to figure out their precise role during plant stress response.

The salient features of this study include: (1) a valuable approach to identify the expression profiles of various genes under different conditions in plants; (2) meta-analysis that can be used to characterize candidate genes for both BS and AS tolerance; (3) identification of important genes involved in various BS and AS; and (4) identified genes which are potential candidates for the genetic manipulation of plants to improve their tolerance to BS and AS.

ACKNOWLEDGEMENT

The authors are thankful to Dr. Ahmad Tahmasebi for his invaluable assistance throughout this study.

ADDITIONAL INFORMATION AND DECLARATIONS

Funding

This work was supported by the Iran National Science Foundation (INSF) and Shiraz University. The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

Grant Disclosures

The following grant information was disclosed by the authors: Iran National Science Foundation (INSF) and Shiraz University.

Competing Interests

The authors declare that they have no competing interests.

Author Contributions

- Elham Ashrafi-Dehkordi performed the experiments, analyzed the data, prepared figures and/or tables.
- Abbas Alemzadeh conceived and designed the experiments, authored or reviewed drafts of the paper, approved the final draft.
- Nobukazu Tanaka authored or reviewed drafts of the paper.
- Hooman Razi analyzed the data.

Data Availability

The following information was supplied regarding data availability: The raw data are provided in the Supplemental Files.

Supplemental Information

Supplemental information for this article can be found online at http://dx.doi.org/10.7717/peerj.4631#supplemental-information.

REFERENCES

- Alves MS, Dadalto SP, Gonçalves AB, De Souza GB, Barros VA, Fietto LG. 2013. Plant bZIP transcription factors responsive to pathogens: a review. *International Journal of Molecular Sciences* 14(4):7815–7828 DOI 10.3390/ijms14047815.
- **Arie T, Takahashi H, Kodama M, Teraoka T. 2007.** Tomato as a model plant for plant-pathogen interactions. *Plant Biotechnology* **24**(1):135–147 DOI 10.5511/plantbiotechnology.24.135.
- **Atkinson NJ, Lilley CJ, Urwin PE. 2013.** Identification of genes involved in the response of Arabidopsis to simultaneous biotic and abiotic stresses. *Plant Physiology* **162(4)**:2028–2041 DOI 10.1104/pp.113.222372.
- Berrocal-Lobo M, Molina A, Solano R. 2002. Constitutive expression of ETHYLENE-RESPONSE-FACTOR1 in Arabidopsis confers resistance to several necrotrophic fungi. *Plant Journal* 29(1):23–32 DOI 10.1046/j.1365-313x.2002.01191.x.
- Bhargava A, Clabaugh I, To JP, Maxwell BB, Chiang YH, Schaller EG, Loraine A, Kieber JJ. 2013. Identification of cytokinin-responsive genes using microarray meta-analysis and RNA seq Arabidopsis thaliana. *Plant Physiology* **162**(1):272–294 DOI 10.1104/pp.113.217026.
- Borrás-Hidalgo O, Thomma BPHJ, Collazo C, Chacón O, Borroto CJ, Ayra C, Portieles R, López Y, Pujol M. 2006. EIL2 transcription factor and glutathione synthetase are required for defense of tobacco against tobacco blue mold. *Molecular Plant-Microbe Interactions* 19(4):399–406 DOI 10.1094/MPMI-19-0399.
- **Campain A, Yang YH. 2010.** Comparison study of microarray meta-analysis methods. *BMC Bioinformatics* **11(1)**:408 DOI 10.1186/1471-2105-11-408.
- Chakravarthy S, Tuori RP, D'Ascenzo MD, Fobert PR, Després C, Martin GB. 2003. The tomato transcription factor Pti4 regulates defense-related gene expression via GCC box and non-GCC box cis elements. *Plant Cell* 15(12):3033–3050 DOI 10.1105/tpc.017574.
- Chang LC, Lin HM, Sibille E, Tseng GC. 2013. Meta-analysis methods for combining multiple expression profiles: comparisons, statistical characterization and an application guideline. *BMC Bioinformatics* 14(1):368 DOI 10.1186/1471-2105-14-368.
- Chen L, Song Y, Li S, Zhang L, Zou C, Yu D. 2012. The role of WRKY transcription factors in plant abiotic stresses. *Biochimica et Biophysica Acta (BBA)—Gene Regulatory Mechanisms* 1819(2):120–128 DOI 10.1016/j.bbagrm.2011.09.002.
- Chinnusamy V, Ohta M, Kanrar S, Lee BH, Hong X, Agarwal M, Zhu JK. 2003. ICE1: a regulator of cold-induced transcriptome and freezing tolerance in Arabidopsis. *Genes & Development* 17(8):1043–1054 DOI 10.1101/gad.1077503.
- Dou M, Fan S, Yang S, Huang R, Yu H, Feng X. 2016. Overexpression of *AmRosea1* gene confers drought and salt tolerance in rice. *International Journal of Molecular Sciences* 18(1):2 DOI 10.3390/ijms18010002.
- Du Z, Zhou X, Ling Y, Zhang Z, Su Z. 2010. agriGO: a GO analysis toolkit for the agricultural community. *Nucleic Acids Research* 38(suppl_2):W64–W70 DOI 10.1093/nar/gkq310.
- **Eulgem T, Somssich IE. 2007.** Networks of WRKY transcription factors in defense signaling. *Current Opinion in Plant Biology* **10(4)**:366–371 DOI 10.1016/j.pbi.2007.04.020.
- Feng HL, Ma NN, Meng X, Zhang S, Wang JR, Chai S, Meng QW. 2013. A novel tomato MYC-type ICE1-like transcription factor, SIICE1a, confers cold, osmotic and salt tolerance in transgenic tobacco. *Plant Physiology and Biochemistry* **73**:309–320 DOI 10.1016/j.plaphy.2013.09.014.
- Fisher RA. 1992. Statistical Methods for Research Workers. London: Oliver and Boyd.

- Fujita M, Fujita Y, Noutoshi Y, Takahashi F, Narusaka Y, Yamaguchi-Shinozaki K, Shinozaki K. 2006. Crosstalk between abiotic and biotic stress responses: a current view from the points of convergence in the stress signaling networks. *Current Opinion in Plant Biology* 9(4):436–442 DOI 10.1016/j.pbi.2006.05.014.
- **Grundy J, Stoker C, Carré IA. 2015.** Circadian regulation of abiotic stress tolerance in plants. *Frontiers in Plant Science* **6**:648 DOI 10.3389/fpls.2015.00648.
- Gu YQ, Wildermuth MC, Chakravarthy S, Loh YT, Yang C, He X, Han Y, Martin GB. 2002. Tomato transcription factors Pti4, Pti5, and Pti6 activate defense responses when expressed in Arabidopsis. *Plant Cell* 14(4):817–831 DOI 10.1105/tpc.000794.
- Huang S, Gao Y, Liu J, Peng X, Niu X, Fei Z, Cao S, Liu Y. 2012. Genome-wide analysis of WRKY transcription factors in *Solanum lycopersicum*. *Molecular Genetics and Genomics* 287(6):495–513 DOI 10.1007/s00438-012-0696-6.
- Huang D, Jaradat MR, Wu W, Ambrose SJ, Ross AR, Abrams SR, Cutler AJ. 2007. Structural analogs of ABA reveal novel features of ABA perception and signaling in Arabidopsis. *Plant Journal* 50(3):414–428 DOI 10.1111/j.1365-313X.2007.03056.x.
- Huang XS, Zhang Q, Zhu D, Fu X, Wang M, Zhang Q, Moriguchi T, Liu JH. 2015. *ICE1* of *Poncirus trifoliata* functions in cold tolerance by modulating polyamine levels through interacting with arginine decarboxylase. *Journal of Experimental Botany* **66**(11):3259–3274 DOI 10.1093/jxb/erv138.
- **Hummel M, Rahmani F, Smeekens S, Hanson J. 2009.** Sucrose-mediated translational control. *Annals of Botany* **104**(1):1–7 DOI 10.1093/aob/mcp086.
- **Khatri P, Drăghici S. 2005.** Ontological analysis of gene expression data: current tools, limitations, and open problems. *Bioinformatics* **21(18)**:3587–3595 DOI 10.1093/bioinformatics/bti565.
- Kim JG, Stork W, Mudgett MB. 2013. Xanthomonas type III effector XopD desumoylates tomato transcription factor SIERF4 to suppress ethylene responses and promote pathogen growth. *Cell Host & Microbe* 13(2):143–154 DOI 10.1016/j.chom.2013.01.006.
- Kissoudis C, Sunarti S, Van De Wiel C, Visser RGF, van der Linden CG, Bai Y. 2016.

 Responses to combined abiotic and biotic stress in tomato are governed by stress intensity and resistance mechanism. *Journal of Experimental Botany* 67(17):5119–5132

 DOI 10.1093/jxb/erw285.
- **Knapp S, Bohs L, Nee M, Spooner DM. 2004.** Solanaceae—a model for linking genomics with biodiversity. *Comparative and Functional Genomics* **5(3)**:285–291 DOI 10.1002/cfg.393.
- Li S, Fu Q, Chen L, Huang W, Yu D. 2011. *Arabidopsis thaliana* WRKY25, WRKY26, and WRKY33 coordinate induction of plant thermotolerance. *Planta* 233(6):1237–1252 DOI 10.1007/s00425-011-1375-2.
- Li D, Fu F, Zhang H, Song F. 2015. Genome-wide systematic characterization of the bZIP transcriptional factor family in tomato (*Solanum lycopersicum* L.). *BMC Genomics* 16(1):771 DOI 10.1186/s12864-015-1990-6.
- **Li JB, Luan YS, Jin H. 2012.** The tomato SlWRKY gene plays an important role in the regulation of defense responses in tobacco. *Biochemical and Biophysical Research Communications* **427(3)**:671–676 DOI 10.1016/j.bbrc.2012.09.120.
- Li JB, Luan YS, Liu Z. 2015. SpWRKY1 mediates resistance to Phytophthora infestans and tolerance to salt and drought stress by modulating reactive oxygen species homeostasis and expression of defense-related genes in tomato. Plant Cell, Tissue and Organ Culture (PCTOC) 123(1):67–81 DOI 10.1007/s11240-015-0815-2.
- Liu B, Ouyang Z, Zhang Y, Li X, Hong Y, Huang L, Liu S, Zhang H, Li D, Song F. 2014. Tomato NAC transcription factor SISRN1 positively regulates defense response against biotic stress but

- negatively regulates abiotic stress response. *PLOS ONE* **9(7)**:e102067 DOI 10.1371/journal.pone.0102067.
- Lu SX, Knowles SM, Andronis C, Ong MS, Tobin EM. 2009. CIRCADIAN CLOCK ASSOCIATED1 and LATE ELONGATED HYPOCOTYL function synergistically in the circadian clock of Arabidopsis. *Plant Physiology* 150(2):834–843 DOI 10.1104/pp.108.133272.
- Mayrose M, Ekengren SK, Melech-Bonfil S, Martin GB, Sessa G. 2006. A novel link between tomato GRAS genes, plant disease resistance and mechanical stress response. *Molecular Plant Pathology* **7(6)**:593–604 DOI 10.1111/J.1364-3703.2006.00364.X.
- Oh S, Park S, Han KH. 2003. Transcriptional regulation of secondary growth in *Arabidopsis thaliana*. *Journal of Experimental Botany* 54(393):2709–2722 DOI 10.1093/jxb/erg304.
- Ouyang B, Yang T, Li H, Zhang L, Zhang Y, Zhang J, Fei Z, Ye Z. 2007. Identification of early salt stress response genes in tomato root by suppression subtractive hybridization and microarray analysis. *Journal of Experimental Botany* 58(3):507–520 DOI 10.1093/jxb/erl258.
- Pan YJ, Cho CC, Kao YY, Sun CH. 2009. A novel WRKY-like protein involved in transcriptional activation of cyst wall protein genes in *Giardia lamblia*. *Journal of Biological Chemistry* 284(27):17975–17988 DOI 10.1074/jbc.M109.012047.
- **Prasad KVSK, Abdel-Hameed AAE, Xing D, Reddy ASN. 2016.** Global gene expression analysis using RNA-seq uncovered a new role for SR1/CAMTA3 transcription factor in salt stress. *Scientific Reports* **6(1)**:27021 DOI 10.1038/srep27021.
- **R Core Team. 2015.** *R: A language and environment for statistical computing.* Version 3.2.2. Vienna: R Foundation for Statistical Computing. *Available at https://www.R-project.org/*.
- **Rastgoo L, Alemzadeh A. 2011.** Biochemical Responses of Gouan ('Aeluropus littoralis') to Heavy Metals Stress. Australian Journal of Crop Science **5(4)**:375–383.
- **Rastgoo L, Alemzadeh A, Afsharifar A. 2011.** Isolation of two novel isoforms encoding zinc- and copper-transporting P1B-ATPase from Gouan (*Aeluropus littoralis*). *Plant Omics Journal* **4(7)**:377–383.
- Rest JS, Wilkins O, Yuan W, Purugganan MD, Gurevitch J. 2016. Meta-analysis and meta-regression of transcriptomic responses to water stress in Arabidopsis. *Plant Journal* 85(4):548–560 DOI 10.1111/tpj.13124.
- Rhodes DR, Barrette TR, Rubin MA, Ghosh D, Chinnaiyan AM. 2002. Meta-analysis of microarrays: interstudy validation of gene expression profiles reveals pathway dysregulation in prostate cancer. *Cancer Research* 62(15):4427–4433.
- Roy S. 2016. Function of MYB domain transcription factors in abiotic stress and epigenetic control of stress response in plant genome. *Plant Signaling & Behavior* 11(1):e1117723 DOI 10.1080/15592324.2015.1117723.
- **Sami Z, Alemzadeh A. 2016.** Isolation and molecular characterization of a novel Na⁺/H⁺ antiporter gene, AlNHX2, from *Aeluropus littoralis* and comparison of AlNHX1 and AlNHX2. *Plant Omics* **9(3)**:205–212 DOI 10.21475/poj.16.09.03.pne35.
- Seki M, Narusaka M, Ishida J, Nanjo T, Fujita M, Oono Y, Kamiya A, Nakajima M, Enju A, Sakurai T, Satou M, Akiyama K, Taji T, Yamaguchi-Shinozaki K, Carninci P, Kawai J, Hayashizaki Y, Shinozaki K. 2002. Monitoring the expression profiles of 7000 Arabidopsis genes under drought, cold and high-salinity stresses using a full-length cDNA microarray. *Plant Journal* 31(3):279–292 DOI 10.1046/j.1365-313X.2002.01359.x.
- Seo PJ, Mas P. 2015. STRESSing the role of the plant circadian clock. *Trends in Plant Science* 20(4):230–237 DOI 10.1016/j.tplants.2015.01.001.

- **Shaar-Moshe L, Hübner S, Peleg Z. 2015.** Identification of conserved drought-adaptive genes using a cross-species meta-analysis approach. *BMC Plant Biology* **15**(1):111 DOI 10.1186/s12870-015-0493-6.
- **Shaik R, Ramakrishna W. 2014.** Machine learning approaches distinguish multiple stress conditions using stress-responsive genes and identify candidate genes for broad resistance in rice. *Plant Physiology* **164(1)**:481–495 DOI 10.1104/pp.113.225862.
- Sharma R, De Vleesschauwer D, Sharma MK, Ronald PC. 2013. Recent advances in dissecting stress-regulatory crosstalk in rice. *Molecular Plant* 6(2):250–260 DOI 10.1093/mp/sss147.
- Sharma MK, Kumar R, Solanke AU, Sharma R, Tyagi AK, Sharma AK. 2010. Identification, phylogeny, and transcript profiling of ERF family genes during development and abiotic stress treatments in tomato. *Molecular Genetics and Genomics* 284(6):455–475 DOI 10.1007/s00438-010-0580-1.
- **Song C, Tseng GC. 2014.** Hypothesis setting and order statistic for robust genomic meta-analysis. *Annals of Applied Statistics* **8(2)**:777–800 DOI 10.1214/13-aoas683.
- Szklarczyk D, Franceschini A, Wyder S, Forslund K, Heller D, Huerta-Cepas J, Simonovic M, Roth A, Santos A, Tsafou KP, Kuhn M, Bork P, Jensen LJ, von Mering C. 2014. STRING v10: protein–protein interaction networks, integrated over the tree of life. *Nucleic Acids Research* 43(D1):D447–D452 DOI 10.1093/nar/gku1003.
- Tseng GC, Ghosh D, Feingold E. 2012. Comprehensive literature review and statistical considerations for microarray meta-analysis. *Nucleic Acids Research* 40(9):3785–3799 DOI 10.1093/nar/gkr1265.
- Van der Ent S, Verhagen BWM, Van Doorn R, Bakker D, Verlaan MG, Pel MJC, Joosten RG, Proveniers MCG, Van Loon LC, Ton J, Pieterse CMJ. 2008. MYB72 is required in early signaling steps of rhizobacteria-induced systemic resistance in Arabidopsis. *Plant Physiology* 146(3):1293–1304 DOI 10.1104/pp.107.113829.
- Wang X, Lin Y, Song C, Sibille E, Tseng GC. 2012. Detecting disease-associated genes with confounding variable adjustment and the impact on genomic meta-analysis: with application to major depressive disorder. *BMC Bioinformatics* 13(1):52 DOI 10.1186/1471-2105-13-52.
- **Wilkinson B. 1951.** A statistical consideration in psychological research. *Psychological Bulletin* **48(2)**:156–158 DOI 10.1037/h0059111.
- Wu L, Chen X, Ren H, Zhang Z, Zhang H, Wang J, Wang XC, Huang R. 2007. ERF protein JERF1 that transcriptionally modulates the expression of abscisic acid biosynthesis-related gene enhances the tolerance under salinity and cold in tobacco. *Planta* 226(4):815–825 DOI 10.1007/s00425-007-0528-9.
- Wu KL, Guo ZJ, Wang HH, Li J. 2005. The WRKY family of transcription factors in rice and Arabidopsis and their origins. *DNA Research* 12(1):9–26 DOI 10.1093/dnares/12.1.9.
- Wu X, Shiroto Y, Kishitani S, Ito Y, Toriyama K. 2009. Enhanced heat and drought tolerance in transgenic rice seedlings overexpressing OsWRKY11 under the control of HSP101 promoter. *Plant Cell Reports* 28(1):21–30 DOI 10.1007/s00299-008-0614-x.
- Xiong H, Li J, Liu P, Duan J, Zhao Y, Guo X, Li Y, Zhang H, Ali J, Li Z. 2014. Overexpression of OsMYB48-1, a novel MYB-related transcription factor, enhances drought and salinity tolerance in rice. *PLOS ONE* 9(3):e92913 DOI 10.1371/journal.pone.0092913.
- Ying S, Jing-Yu X, Bian-Yun Y, Ben-Zhong Z, Yun-Bo L. 2004. Influence of salt stress on expression of some genes involved in the ethylene signalling pathway in tomato seedlings. *Chinese Journal of Agricultural Biotechnology* 1(2):115–118.

- Yáñez M, Cáceres S, Orellana S, Bastías A, Verdugo I, Ruiz-Lara S, Casaretto JA. 2009. An abiotic stress-responsive bZIP transcription factor from wild and cultivated tomatoes regulates stress-related genes. *Plant Cell Reports* 28(10):1497–1507 DOI 10.1007/s00299-009-0749-4.
- Zamboni A, Zanin L, Tomasi N, Pezzotti M, Pinton R, Varanini Z, Cesco S. 2012. Genome-wide microarray analysis of tomato roots showed defined responses to iron deficiency. *BMC Genomics* 13(1):101–115 DOI 10.1186/1471-2164-13-101.
- Zhang H, Huang Z, Xie B, Chen Q, Tian X, Zhang X, Zhang H, Lu X, Huang D, Huang R. 2004. The ethylene-, jasmonate-, abscisic acid- and NaCl-responsive tomato transcription factor JERF1 modulates expression of GCC box-containing genes and salt tolerance in tobacco. *Planta* 220(2):262–270 DOI 10.1007/s00425-004-1347-x.
- Zhang Z, Li F, Li D, Zhang H, Huang R. 2010. Expression of ethylene response factor JERF1 in rice improves tolerance to drought. *Planta* 232(3):765–774 DOI 10.1007/s00425-010-1208-8.
- **Zinati Z, Alemzadeh A, KayvanJoo AH. 2016.** Computational approaches for classification and prediction of P-type ATPase substrate specificity in Arabidopsis. *Physiology and Molecular Biology of Plants* **22(1)**:163–174 DOI 10.1007/s12298-016-0351-5.