Molecular adaptation to salinity fluctuation in tropical intertidal environments of a mangrove tree *Sonneratia alba*

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Abstract

**Background:** Mangroves have adapted to intertidal zones - the interface between terrestrial and marine ecosystems. Various studies have shown adaptive evolution in mangroves at physiological, ecological, and genomic levels. However, these studies paid little attention to gene regulation of salt adaptation by transcriptome profiles.

**Results:** We sequenced the transcriptomes of *Sonneratia alba* under low (fresh water), medium (half the seawater salinity), and high salt (seawater salinity) conditions and investigated the underlying transcriptional regulation of salt adaptation. In leaf tissue, 64% potential salinity-related genes were not differentially expressed when salinity increased from freshwater to medium levels, but became up- or down-regulated when salt concentrations further increased to levels found in sea water, indicating that these genes are well adapted to the medium saline condition. We inferred that both maintenance and regulation of cellular environmental homeostasis are important adaptive processes in *S. alba*. i) The sulfur metabolism as well as flavone and flavonol biosynthesis KEGG pathways were significantly enriched among up-regulated genes in leaves. They are both involved in scavenging ROS or synthesis and accumulation of osmosis-related metabolites in plants. ii) There was a significantly increased percentage of transcription factor-encoding genes among up-regulated transcripts. High expressions of salt tolerance-related TF families were found under high salt conditions. iii) Some genes up-regulated in response to salt treatment showed signs of adaptive evolution at the amino acid level and might contribute to adaptation to fluctuating intertidal environments.

**Conclusions:** This study first elucidates the mechanism of high-salt adaptation in mangroves at the whole-transcriptome level by salt gradient experimental treatments. It reveals that several candidate genes (including salt-related genes, TF-encoding genes, and PSGs) and major pathways are involved in adaptation to high-salt environments. Our study also provides a valuable resource for future investigation of adaptive evolution in extreme environments.

**Keywords:** Homeostasis, Mangrove, Salt adaptation, *Sonneratia alba*, Transcription factor, Transcriptome profiles
Background
How species adapt to extreme environments is a fundamental topic of evolutionary biology. Located at the interface between terrestrial and marine ecosystems, intertidal zones impose many harsh or extreme conditions on local organisms, including high salinity, hypoxia, salt fluctuations, strong ultraviolet (UV) light, and high temperature [1]. Nevertheless, mangroves have colonized and became well adapted to these habitats, evolving a series of highly specialized traits including salt tolerance, viviparous embryos, aerial roots, and high tannin content [2–4]. Mangroves are a major group of marine halophytes that thrive and complete their life cycles in high-salinity environments that arise due to fluctuating seawater levels. Thus, they are good model systems for investigating adaptive evolution. Indeed, adaptation in this group has been studied at the phenotypic, ecological, physiological, and genomic levels [5–10]. However, little attention has been paid to transcriptional regulation even though it has been strongly selected for rapid adaptation to changes in environmental conditions and may play a critical role at ecological and evolutionary timescales [11].

Salt adaptation is a long-term and dynamic process brought about by multiple genes involved in many morphological, physiological, molecular, and cellular processes [12–14]. Various salt-responsive genes and important signaling pathways have been identified by global gene profiling in the model plant Arabidopsis thaliana [12, 15], and crop plants rice [16] and soybean [17]. The Salt overly sensitive (SOS) signaling pathway, Ca²⁺-dependent signaling pathways, and ABA signaling pathways contribute to mediating cellular signaling under salt stress and maintain cellular environmental homeostasis [18]. Many molecular mechanisms have been revealed in glycophytes, but the major limiting factor is their inability to survive high and fluctuating salinity. These studies are insufficient to understand molecular aspects of plant salt adaptation.

Sonneratia alba inhabits low intertidal zones of downstream estuarine systems and is one of the most pervasive and salt-tolerant mangrove species. It reaches an optimal growth in 5 to 50% seawater, indicating its capacity to tolerate high salinity and hypoxia [19]. It can synthesize organic solutes to adjust cellular osmotic potential and produce antioxidant enzymes for scavenging reactive oxygen species (ROS) against high salinity [20]. Therefore, it is an attractive ecological model to investigate genetic mechanisms underlying salt adaptation. However, only limited RNA-seq data from Sonneratia have been reported to date [21, 22]. The lack of transcriptomic information for S. alba across salinity conditions is a major obstacle to discerning the molecular bases of adaptation to saline environments. In this study, we designed an experiment to elucidate the transcriptional regulation of high-salt adaptation in S. alba. We sequenced transcriptomes from leaves and roots of S. alba seedlings irrigated with artificial seawater. After genome-guided transcriptome assembly, we characterized the transcriptomes under low, medium, and high salinity conditions for further functional analyses. We wish to gain new insights into the potential genetic, biochemical, and physiological mechanisms of salt adaptation in S. alba. Our study also generates valuable resources to support further research into adaptive evolution of mangroves in extreme environments.

Results
Identification of differentially expressed genes and expression patterns
To generate genomic resources and begin elucidating the mechanisms of salt tolerance in S. alba, we performed RNA-seq experiments on leaves and roots from plants reared in media with various salt levels. The read summary statistics and assembly results are outlined in Table 1. We generated 44.20 to 61.97 million of 100 bp paired-end reads from 12 libraries. After quality filtering (see Methods for details) we retained 41.94 to 59.21 million 99 bp high quality reads and successfully mapped 93.85 to 94.83% of them to the S. alba reference genome. These aligned reads were assembled separately within each sample, yielding 33,930 to 40,937 transcripts with average lengths 3298 to 3530 bp and N50 values of 4214 to 4437 bp (Table 1). The correlation coefficients between biological replicates in each condition were from 0.87 to 0.97 (Additional file 1: Figure S1).

For S. alba, its adaptation to the intertidal zone environment is a long-term and dynamic process, and its growth environment is 5 to 50% seawater concentration, that is, 25 mM to 250 mM [19]. Hence, 500 mM and 0 mM provide the hyper-saline (stressful) and low-saline conditions, respectively. A large number of genes between 0 vs 250 are expected to have similar expression levels. The stress would be between 250 mM and 500 mM. We next performed two pair-wise comparisons of transcript abundance: low (0 mM) vs medium (250 mM), and medium vs high (500 mM) salt concentrations (see Methods for details). In the low to medium NaCl comparison, we found 387 significantly up-regulated and 726 down-regulated genes in leaves, and 169 up- and 295 down-regulated transcripts in roots. We found 1057 up- and 1296 down-regulated genes in leaves, and 247 up- and 133 down-regulated transcripts in roots when comparing the medium and high salt conditions (Fig. 1). There were clearly more differentially expressed genes (DEGs) in leaves than in roots. Moreover, we found more up-regulated genes in the medium to high salt contrast than in the low to medium comparison.

To identify co-expression patterns among the DEGs, we clustered them into eight groups according to their
Table 1 Summary of mapping and assembly statistics of the Sonneratia alba transcriptomes

| Sample | Replicate | Mapping to genome | Multiple mapped | Uniquely mapped | Number of transcripts | Average size (bp) | N50 (bp) |
|--------|-----------|-------------------|-----------------|-----------------|---------------------|------------------|----------|
| L0     | r1        | 52.26 (x 10^6) 94.59 % | 0.44 (x 10^6) 0.84 % | 51.82 (x 10^6) 99.16 % | 36,664 (x 10^6) | 3395 | 4301 |
|        | r2        | 42.27 (x 10^6) 94.12 % | 0.64 (x 10^6) 1.51 % | 41.64 (x 10^6) 98.49 % | 35,970 (x 10^6) | 3378 | 4235 |
| L250   | r1        | 39.85 (x 10^6) 94.02 % | 0.42 (x 10^6) 1.05 % | 39.43 (x 10^6) 98.95 % | 33,930 (x 10^6) | 3318 | 4214 |
|        | r2        | 39.36 (x 10^6) 93.85 % | 0.61 (x 10^6) 1.54 % | 38.75 (x 10^6) 98.46 % | 35,402 (x 10^6) | 3394 | 4273 |
| L500   | r1        | 55.66 (x 10^6) 94.01 % | 0.59 (x 10^6) 1.06 % | 55.08 (x 10^6) 98.94 % | 34,116 (x 10^6) | 3298 | 4221 |
|        | r2        | 43.41 (x 10^6) 93.99 % | 0.87 (x 10^6) 2.01 % | 42.54 (x 10^6) 97.99 % | 34,334 (x 10^6) | 3443 | 4338 |
| R0     | r1        | 45.47 (x 10^6) 93.97 % | 0.49 (x 10^6) 1.09 % | 44.97 (x 10^6) 98.91 % | 40,937 (x 10^6) | 3471 | 4352 |
|        | r2        | 43.01 (x 10^6) 94.43 % | 1.92 (x 10^6) 4.47 % | 41.08 (x 10^6) 95.53 % | 40,005 (x 10^6) | 3469 | 4380 |
| R250   | r1        | 44.20 (x 10^6) 94.48 % | 0.49 (x 10^6) 1.12 % | 43.71 (x 10^6) 98.88 % | 40,696 (x 10^6) | 3425 | 4363 |
|        | r2        | 40.78 (x 10^6) 94.58 % | 1.99 (x 10^6) 4.88 % | 38.79 (x 10^6) 95.12 % | 40,483 (x 10^6) | 3476 | 4383 |
| R500   | r1        | 52.23 (x 10^6) 94.58 % | 0.77 (x 10^6) 1.47 % | 51.46 (x 10^6) 98.53 % | 40,338 (x 10^6) | 3530 | 4437 |
|        | r2        | 40.63 (x 10^6) 94.83 % | 1.13 (x 10^6) 2.79 % | 39.49 (x 10^6) 97.21 % | 39,426 (x 10^6) | 3503 | 4391 |

L and R represent leaf and root tissues; 0, 250 and 500 represent the salinity treatments (mM NaCl); r1 and r2 indicate two biological replicates with three individuals each. For each sample, RNA-seq reads mapping and genome-guided transcriptome assembly were performed individually.

Fig. 1 Differentially expressed genes (DEGs). The number of up- and down-regulated genes in leaves and roots across salinity conditions (0 mM vs 250 mM NaCl and 250 mM vs 500 mM NaCl) is shown. The total number of DEGs is listed on top of each bar.
expression change across salinity conditions and tissues (Additional file 2: Figure S2). Intriguingly, the two largest leaf clusters, LC8 (1045 genes) and LC4 (965 genes), comprised only genes that are significantly down- or up-regulated under high salt treatment (Fig. 2). The same pattern did not hold in the roots, however. To assess whether the LC8 and LC4 clusters are larger than expected by chance, we conducted a permutation test (see Methods for details). We indeed saw that these clusters were larger than expected (Fig. 2). The most genes didn’t differentially express across salinity conditions. The LC4 and LC8 are majority (64%) of all eight clusters. They are only differentially expressed when the salt levels become high (500 mM), but are unchanged in the medium to low NaCl comparison. Thus, only the extremely high levels of salt treatment were able to appreciably change the S. alba transcriptome, as expected given the high salinity tolerance of this species.

**Functional enrichment and transcriptional regulation**

We assigned functional Gene Ontology (GO) categories to the DEGs we identified (Additional file 3: Table S1) and four major co-expression clusters, including LC4, LC8, LC4-RC4 (overlaps between LC4 and RC4), LC8-RC8 (Additional file 4: Table S2). Using a whole-genome annotation as the background, we found several categories that were over-represented among our differentially expressed genes (Fig. 3). Remarkably, many of the genes up-regulated in leaves were involved in the molecular function category “transporter activity.” The biological process category “response to stimulus” was significantly overrepresented in all four comparisons (both leaves and roots). Furthermore, many up-regulated genes were assigned to the “response to stimulus” category in four comparisons (57 in L0 vs L250, 35 in R0 vs R250, 158 in L250 vs L500, 36 in R250 vs R500). Detailed annotations were also performed across up-regulated genes. We note that many genes were annotated with functions involving salt tolerance, especially Sodium/hydrogen exchanger (NHX) encoding gene (SA_23236 in L0 vs L250 and L250 vs L500, SA_23120 in R250 vs R500), Tonoplast intrinsic protein (TIP) encoding gene (SA_14179 in L250 vs L500), Plasma membrane intrinsic protein (PIP) encoding gene (SA_25431 in L0 vs L250 and L250 vs L500, SA_23120 in R250 vs R500), Mangrin (SA_08086, SA_14631 in L250 vs L500), several heat shock protein encoding genes, and main transcription factor encoding genes (described in the following section). Mangrin is partially homologous to the allene oxide cyclase (AOC) encoding gene. Several genes up-regulated under high salt directly adjust cellular osmotic potential or integrate metabolic, hormonal, and environmental signals in stress acclimatization. They include the glutathione peroxidase encoding gene (SA_08946 in roots and SA_18901 in leaves), the monodehydroascorbate reductase encoding gene (SA_20647 in leaves), the aldo-keto reductase encoding gene (SA_02762 in both tissues, SA_02758, SA_02759 in leaves, and SA_15935 in roots), a number of key enzymes in biosynthesis of...
asparagine, cysteine, galactinol, trehalose-6-phosphate, and stachyose. We also found up-regulated transcripts of unknown function. These can be further investigated to gain fresh insights into salt adaptation.

Interestingly, 965 of the transcripts up-regulated in leaves fall into the LC4 cluster, as do 89.24% of the 158 genes that were both up-regulated and assigned to the “response to stimulus” GO category (Additional file 4: Table S2). This suggests that this cluster plays an important role in response to abnormally high salt stress.

The GO terms significantly enriched among down-regulated genes were “extracellular region” and “catalytic activity” categories in leaves and “membrane part” in roots (Fig. 3). Furthermore, 1045 of transcripts down-regulated in leaves fall into the LC8 cluster, were enriched in categories of “carbohydrate metabolic process”, “cell periphery”, “negative regulation of molecular function”, “catalytic activity” and so on (Additional file 4: Table S2). They were closely associated with basic metabolism, plant growth regulation, molecular function regulator, signal transducer activity, and electron transport.

We next investigated metabolic pathways significantly enriched among the DEGs by cross-referencing our list of up-regulated genes with the KEGG annotation of the S. alba reference genome (Additional file 5: Table S3). Of the metabolic pathways overrepresented among transcripts differentially expressed under high salt (four highly and three slightly enriched, Table 2), sulfur metabolism, protein processing in endoplasmic reticulum, and flavone and flavonol biosynthesis were the most interesting because they are known to be involved in S. alba salt adaptation. Flavonoid biosynthesis, nicotinate, and nicotinamide metabolism pathways were also overrepresented among genes up-regulated under high salt treatment in roots (Table 2). No metabolic pathways stood out among genes differentially expressed under the medium compared to low salt conditions.

To further explore the regulation of biological pathways under salt treatment, we visualized the expression profiles of up-regulated genes in transcripts assigned to major pathways (Fig. 4). The highlighted pathways are involved in reactive oxygen species (ROS) scavenging, detoxification, and regulation of cell osmotic pressure. We reason that these pathways play essential roles in salt adaptation. Notably, these functional enrichment analyses of DEGs point to the cellular environment as the possible site of molecular salt adaptation in S. alba. This seems plausible as changes in candidate proteins occur within cells.

Transcription factors involved in salt adaptation

Transcription factors (TFs) are regulatory proteins and therefore have the potential to regulate multiple aspects
ilies were enriched among the up-regulated genes. Recognized families (Additional file 6: Figure S3), we evidenced that these proteins are relevant to salinity resistance. Functional annotation gives us further confidence that these proteins are relevant to salinity resistance.

Positive selection in up-regulated genes
Unusual salt tolerance is a derived characteristic of mangroves. The Ka/Ks values along *S. alba* and its relatives were calculated (Additional file 7: Figure S4). It showed that Ka/Ks ratio was elevated in the *S. alba* relative to relatives. We would therefore expect that some of the genes up-regulated in response to NaCl treatment have acquired their roles recently in evolutionary time. We wondered if their amino acid composition changed along with expression level. To test this idea, we used the PAML package to identify positively selected genes (PSGs) in *S. alba* (see Methods for details). We identified sixteen up-regulated PSGs (Additional file 8: Table S4). The SA_12151 locus has particularly high Ka/Ks ratio (Ka/Ks = 4.42) and has probably undergone positive selection. Functional annotation gives us further confidence that these proteins are relevant to salinity resistance.

Real-time quantitative PCR (qPCR) DEG verification
To assess the reliability of RNA sequencing for DEG identification, transcripts of 16 genes from each cluster were evaluated by qPCR. The expression patterns of selected genes were consistent with the RNA-seq data in eight clusters from both leaves and roots. The results of qPCR are shown in Fig. 6a. The correlation between the log2 fold change of expression level from the RNA-seq experiments and qPCR is shown in Fig. 6b. Correlation between these methods is strong and positive ($R^2 = 0.88$, $P$-value $<1E-10$), suggesting that RNA-seq results are highly reproducible.

Discussion
Comparative genomics has been extensively used to investigate the genetic bases of adaptation to extreme environments [6, 7, 9, 10, 14, 28–31]. In this study, comprehensive transcriptome analyses from leaves and roots of *S. alba* across salinity conditions were performed to probe the genetic mechanisms underlying salt adaptation. Our results yielded candidate genes and mechanistic hypotheses that may deepen our understanding of salt adaptation in halophytes and provide a resource for further studies.

Salt adaptation is both a long-term and dynamic process influenced by multiple genes and involves many morphological, physiological, molecular, and cellular processes [12, 14, 32–34]. The requirement of bulk protein synthesis as part of the global defense strategy of *Arabidopsis* has been extensively shown in previous studies [35]. In *Zostera marina*, which experiences the full marine seawater, Na$^+$ and K$^+$ antiporters were displayed as a typical component. The salt-tolerant H$^+$-ATPase plays an important role in maintaining the low Na$^+$ concentration in the seagrass cells since it is strongly expressed in vegetative tissue [36, 37]. And nitrate assimilation and ion compartmentalization are also required in the process of adaptation [38]. While mangroves induce genes that function in specific salt-responsive pathways and transcription factors that

| Condition | Pathway ID | Pathway | No. of genes in genome | No. of up-regulated genes | P value | Adjusted P value |
|-----------|------------|---------|------------------------|---------------------------|---------|------------------|
| Leaf      | ko00920    | Sulfur metabolism | 121            | 19           | 7.12E-08 | 1.58E-05       |
|           | ko04141    | Protein processing in endoplasmic reticulum | 605          | 46           | 2.41E-06 | 1.79E-04       |
|           | ko04626    | Plant-pathogen interaction | 1253        | 74           | 3.61E-05 | 2.01E-03       |
|           | ko00944    | Flavone and flavonol biosynthesis | 101        | 13           | 7.10E-05 | 3.15E-03       |
|           | ko00592    | alpha-Linolenic acid metabolism | 95         | 10           | 2.28E-03 | 6.32E-02       |
|           | ko00760    | Nicotinate and nicotinamide metabolism | 53         | 7            | 2.84E-03 | 7.01E-02       |
|           | ko00270    | Cysteine and methionine metabolism | 312        | 22           | 3.13E-03 | 6.96E-02       |
| Root      | ko00941    | Flavonoid biosynthesis | 85          | 5            | 7.78E-04 | 9.80E-02       |
|           | ko00760    | Nicotinate and nicotinamide metabolism | 53         | 4            | 1.05E-03 | 6.60E-02       |
influence the cellular environment. Most genes differentially expressed across salt levels in our study were induced or suppressed by the highest salt concentration in leaves. The pattern is different in roots, suggesting that salt adaptation in *S. alba* is tissue-specific.

A major group of up-regulated genes identified in our study were involved in ion transportation and osmotic regulation, including *NHX*, *TIP*, *PIP*, *Mangrin*. These genes have been reported to play major roles in salt tolerance in a variety of plants [39–42]. Particularly, *Mangrin* is partially homologous to the allene oxide cyclase (AOC) encoding gene. It is up-regulated by high salinity and its overexpression enhances salt tolerance in transgenic yeast and tobacco cells [39]. Most transcripts down-regulated under high salt conditions were involved in basic metabolism, plant growth regulation, molecular function regulator, signal transducer activity, and electron transport. This suggests that the long-term high salt treatment might suppress normal metabolic processes and throw off the plant homeostatic balance. Indeed, high salinity is an important agricultural contamination that result in damage to the plant [43]. In *Arabidopsis*, salt stress does not simply decrease the steady-state growth rates. While the growth of several discrete phases was strongly suppressed by salt stress, the others were actually increased [44].

Hypersaline stress always leads to secondary oxidation stress, resulting in reactive oxygen species accumulation. We indeed observed that a number of key enzymes (such as *cysQ*, *ATCYS1*, *cysK*, *GST*) involved in sulfur,
cysteine and methionine, and glutathione metabolism were up-regulated in leaves (Fig. 4). These pathways enhance synthesis of sulfur-rich peptides (such as S-Glutathione) to scavenge ROS [45]. In addition, we found that flavone and flavonol biosynthesis pathway genes were overrepresented in our set of differentially expressed transcripts. Flavonoids are natural antioxidants that can reduce ROS levels and aid in survival and growth of plants [46].

Recycling of nicotinamide and nicotinate for pyridine nucleotide synthesis can be stimulated by salt to protect pyrimidine and purine derivatives for nucleotide synthesis in the mangrove Bruguiera sexangula [47]. We also found that genes from the nicotinate and nicotinamide metabolism pathways were overrepresented among transcripts differentially expressed under high salt treatment in both leaves and roots.

Ethylene signalling also influences the response to salt stress in Arabidopsis. High salinity induces the accumulation and activation of EIN3/EIL1 protein. The activated EIN3 alleviates the accumulation of excess ROS accumulation and therefore increases salt tolerance of the plant [48]. We found that key enzymes in the cysteine and methionine, as well as glutathione, metabolism pathways were up-regulated under salt treatment in leaves but not roots. This suggests that ethylene and glutathione are preferentially synthesized in leaf tissue and play a role in long-term salt adaptation.

Several studies have demonstrated that transcription factors act as regulators of adaptation in plants. The functions of certain TFs (such as AP2/EREBP, NAC) in response to salt tolerance in particular have been well documented [35, 49–51]. We also observed that many TF encoding genes were up-regulated under increasing...
salt concentration. Transcription factors of AP2/EREBP, WRKY, bZIP, NAC, MYB and bHLH regulate downstream factors to involve in the major pathways, such as flavonoid biosynthesis, sulfur and glucosinolate metabolism [52–57]. This result confirms the GO enrichment and pathway enrichment analyses and shows an extensive regulation of transcription during salt adaptation. However, further investigations are required to find the exact regulatory networks of salt response in mangrove trees.

Natural selection is likely the major force driving S. alba salt adaptation to the fluctuating intertidal environment. We lacked comparative transcriptomic data to test how much of the gene expression response to salt treatment is the result of adaptive evolution. However, we did find a number of genes whose evolution at the amino acid level has either accelerated or shows signs of positive selection in the S. alba lineage (Additional file 8: Table S4). Indeed, SA_12151, a member of auxin-responsive family, responds to various abiotic stresses (auxin, high-salinity stresses) in rice [58]. SA_26719 (peroxisomal 3-ketoacyl-CoA thiolase) is a peroxisome-associated gene. Previous studies have reported that salt stress affects peroxisome enzyme activities and protein expression [59]. Salt-induced expression of peroxisome-associated genes requires abscisic acid and other signaling pathways [60]. ABA is a stress hormone that plays a central role in a variety of abiotic stress responses. Finally, SA_14625, encoding a hydrolase, is involved in physiologically important processes in plants. It is up-regulated in response to high salinity stress in salt-tolerant rice [61]. Although our data suggest that these genes are involved in salt tolerance, the mechanisms driving evolution of these up-regulated PSGs must be further confirmed with physiological experiments. It would help us better understand the evolutionary dynamics of salt adaptation.

The most extreme challenge for species living in intertidal zones is the unstable and high salinity due to constant tidal fluctuations. Our analyses of DEGs point to the cellular environment as the possible site of molecular salt adaptation in S. alba. Thus, we suggest that the maintenance and regulation of cellular environmental homoeostasis are important adaptive processes, contributing to evolutionary innovations.

### Conclusions

We performed comprehensive transcriptome analyses of S. alba leaves and roots across salinity conditions. The study demonstrates extensive transcriptional regulation of salt adaptation. Leaves and roots appear to adopt different dynamic gene regulation strategies for salt adaptation in S. alba. Most differentially expressed genes in leaves were well adapted to the medium saline condition. Our analyses report several candidate genes (including salt-related genes, TF-encoding genes, and PSGs) and major pathways including sulfur metabolism, flavone and flavonol biosynthesis, nicotinate and nicotinamide metabolism, and cysteine and methionine metabolism, that are involved in adaptation to high-salt environments. We deduce that the
maintenance and regulation of cellular environmental homeostasis are important adaptive processes. Our study can also serve as a resource for future investigation of adaptive evolution in extreme environments. With additional investigations targeting the methylome and the proteome in the future, the exact mechanisms of salt adaptation in plants can be unveiled.

Methods

Plant materials and growth conditions
Eighteen S. alba seedlings (Additional file 9; Figure S5) were collected from the nursery of Dongzhai Harbor National Nature Reserve in Hainan with permission. All samples have been identified by Mr. Cairong Zhong, an officer of Dongzhai Harbor National Nature Reserve Administration. Then they were cultivated in the greenhouse (23 °C ~ 35 °C, 13.5-h day-length, a light intensity sufficient to allow normal growth) of Sun Yat-Sen University. The seedlings were transferred to sandy soil and grown using 1/2 Hoagland’s solution for 3 weeks to help plants adapt to the new environment. The seedlings were divided into three groups, each group containing two biological replicates. These three experimental groups were irrigated using solutions containing 0, 250, or 500 mM NaCl for 7 days. In natural condition, S. alba is usually found near the estuary. The rainwater and upstream rivers give it little exposure to normal seawater concentrations. S. alba reaches an optimal growth in 5 to 50% seawater. Medium saline (250 mM NaCl), approximately half of the normal seawater salinity, is close to the upper bound for natural S. alba growth [19]. Although 500 mM and 0 mM provide the hyper-saline (stressful) and low-saline conditions, respectively, 0–250 mM is a range well tolerated by mangroves in nature. In contrast, 250–500 mM is genuinely stressful. The treatment duration was based on our previous study [62].

Sample collection, sequencing, and alignment
Healthy young leaves and roots of each seedling were sampled after 7 days and immediately frozen in liquid nitrogen and stored at −80 °C. Three plants from the same group were pooled to minimize variation among individual plants. Total RNA was extracted using the Plant RNA kit (Omega Bio-Tek, Doraville, USA) according to the manufacturer’s instructions. For each sample, at least 20 μg of total RNA was sent to the Novogene Bioinformatics Technology Co., Ltd. for sequencing using the Illumina HiSeq 2000 platform. Raw sequence reads were first purified by trimming adapter sequences. In order to reduce sequencing error, paired-end low-quality reads were removed by customized scripts if they met the following criteria: i) they comprised more than 5% unknown (N) bases; ii) comprised more than 20% bases with quality <= 10; iii) average quality was under 20. The last base of the filtered reads was also trimmed according to the results of FastQC.

Clean reads from each sample were aligned to the S. alba reference genome [10] using TopHat v2.1.1 [63] and Bowtie2 v2.2.9 [64]. We then performed the genome-guided transcriptome assembly and differential transcript expression analyses using Cufflinks v2.2.1 [65]. We reported expression levels as Fragments Per Kilobase of exon model per Million mapped reads (FPKM) values. We then calculated the Pearson’s correlation coefficients between all replicates using gene expression data.

Identification of differentially expressed genes and expression pattern analyses
We identified DEGs from two comparisons: 250 mM NaCl vs 0 mM NaCl and 500 mM NaCl vs 250 mM NaCl. We considered transcripts with q-values (FDR-adjusted P-values) below 5% and fold change higher than 2 in either comparison as differentially expressed. We called genes whose expression increased with salt concentration “up-regulated” and transcripts with change in the opposite direction “down-regulated.” The remaining genes were denoted as “no significant difference”. These DEGs were grouped into eight distinct and representative clusters (LC1 ~ LC8, RC1 ~ RC8) via pairwise comparisons between three expression patterns (up-regulated, down-regulated, or non-significant difference) in leaves and roots. To determine which clusters have statistically significant enrichments in the number of genes assigned, we used a permutation test with 10,000 runs. In each permutation we first shuffled all expression quantities obtained under different conditions randomly to destroy any association between the conditions and expression quantities and assigned them to their original gene groups. We then carried out the cluster analysis as for the real data and recorded whether the number of genes assigned to every cluster in the permutation was no smaller than in the actual observation. After 10,000 permutation runs, we estimated the frequency of assigning no fewer genes to a cluster than what is seen in the real expression data, allowing for significance testing.

Gene annotation and ontology enrichment analyses
We obtained S. alba genome annotations and assigned gene ontology (GO) and KEGG orthology (KO) terms to all genes. GO terms were assigned using the Web Gene Ontology Annotation Plotting (WEGO 2.0) [66] and visualized by Cytoscape_v3.7.2 [67]. Transcription factor annotations for S. alba and its relatives with genome sequences: Sonneratia caseolaris (true mangrove) [10], Punica granatum (non-mangrove) [68], and Eucalyptus grandis (non-mangrove) [69] were performed using iTAK1.7 [70]. We tested for enrichment of TFs among
differentially expressed *S. alba* genes using Fisher’s exact test. KEGG annotation was carried out to assign DEGs to metabolic pathways. We assigned all up-regulated genes in leaves and roots to KEGG pathways and compared these pathways’ gene numbers with the whole genome background. KEGG pathway data were from the R package clusterProfiler [71] which supports downloading the latest KEGG annotations. Pathways with more than three up-regulated genes were processed using Fisher’s exact test. Enrichment was declared significant if Benjamini-Hochberg corrected *P*-values fell below 5%. We identified key salt-responsive genes from *S. alba* genome annotations and BLASTP searches of genes up-regulated in *S. alba* against *Arabidopsis thaliana* genome data (TAIR10, www.arabidopsis.org) with cut-off e-value < 10^-6.

**Selection analyses**

In order to detect up-regulated genes that might undergo positive selection in *S. alba*, we first identified high-quality orthologs within *S. alba* and its relatives, including *Punica granatum*, *Trapa bispinosa*, *Lagerstroemia speciosa*, *Duabanga grandiflora*, and *Eucalyptus grandis*. These species belong to the same order Myrtales (APG IV, www.mobot.org/MOBOT/research/APweb/welcome.html). The whole genome sequences of *Punica granatum*, and *Eucalyptus grandis* were downloaded from phytozome and NCBI [68, 69]. For the more closely related species *Trapa bispinosa*, *Lagerstroemia speciosa*, and *Duabanga grandiflora*, the transcriptomes were sequenced in our previous study [72]. Using the orthologs identified by OrthoMCL [73] as an initial data set, we used the reciprocal BLASTP best-hit method for *S. alba* and five related species. We then individually aligned orthologous proteins using MUSCLE [74] and used the aligned protein sequences to generate codon alignments using PAL2NAL [75]. Alignments shorter than 150 bp after removing sites with ambiguous data were discarded. This process yielded 3936 high-quality orthologs within *S. alba* and its relatives. These codon alignments together with a phylogenetic tree [76] were used for subsequent analyses.

We calculated Ka/Ks ratio for each branch of the phylogenetic tree using the “free-ratio” model in the PAML package [77]. All orthologs were concatenated into a single super gene. We used the branch model in the PAML package to identify positively selected genes, setting the *S. alba* branch as the foreground [78]. We introduced the likelihood comparison between the alternative model allowing the foreground branch to evolve under a different rate and the null model assuming that all branches have been evolving at the same rate for each ortholog set. We also used the branch-site model in PAML by setting model = 2, NSsites = 2 [79]. The comparison of likelihoods between the alternative model assuming that at least one site in the foreground has undergone positive selection and the null model that assumes none was used as the test of positive selection. The likelihood ratio test method was the same as for the branch model. Multiple testing was corrected by applying the Benjamini-Hochberg method implemented in the R function *p.adjust*. We considered the up-regulated genes with FDR below 5% as PSGs. Finally, we manually filtered all PSGs with potential errors in their alignments to minimize the false positive rate.

**RNA-seq data validation with real-time quantitative PCR**

The qRT-PCR was performed to validate the RNA-seq results for 16 transcripts from 16 distinct clusters (SA, 28450 both from LC8 and RC5) in leaves and roots. Gene-specific primers for qRT-PCR were designed using the Primer Premier software and are listed in Additional file 10: Table S5.

First-strand cDNAs were synthesized from 1 μg of total RNA using the PrimeScript™ RT reagent Kit with gDNA Eraser (TaKaRa). The expression of the β-actin gene served as the internal control. qRT-PCR was carried out using the Real-Time PCR system on an ABI 7900HT with the following program: 95 °C for 30 s (95 °C for 5 s, 55 °C for 10 s and 72 °C for 15 s), 40 cycles. Each sample was analysed with three technical replicates. The relative expression profiles of selected genes were calculated using the \(2^{-\Delta\Delta CT}\) method [80]. Correlations between expression profiles of selected genes from RNA-seq and qRT-PCR were estimated using a simple linear regression model using the *lm* function in R.

**Supplementary information**

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**Additional file 1: Figure S1.** Gene expression correlations among all samples. Pearson’s correlation plot visualizes the correlation coefficients. Scale bar represents the range of the value displayed. The correlation coefficients between biological replicates in Leaf 0 mM, Leaf 250 mM, Leaf 500 mM, Root 0 mM, Root 250 mM, and Root 500 mM conditions are 0.92, 0.97, 0.90, 0.97, 0.87, 0.90, respectively.

**Additional file 2: Figure S2.** Cluster analysis and permutation test of differentially expressed genes (DEGs). Expression profiles of DEGs under different salinity contrasts are shown for eight clusters in leaves and roots. DEGs in leaves are significantly enriched in three clusters (LC8, LC4, LC5), whereas DEGs in roots are significantly enriched in RC1 and RC5. In each cluster, the black line represents the expression level (means of log10(FPKM value + 1)) of all enriched genes, the error bars represent standard deviations. A permutation test was used to evaluate the significance of differences. The clusters with statistically significant difference in the number of DEGs are marked with triple (*P*-value < 0.01) asterisks. The observed and expected (only significant clusters) numbers of genes belonging to each cluster are labeled on the top of each cluster.

**Additional file 3: Table S1.** The GO annotation of DEGs. URG represents up-regulated gene; DRG represents down-regulated gene. The
The authors declare that they have no competing interests.

Additional file 4: Table S2. Summary of GO terms enriched among genes in co-expression clusters.

Additional file 5: Table S3. The KEGG annotation of up-regulated genes. URG represents up-regulated gene. The numbers in the brackets are percentages of genes with KO ID and those outside the brackets are gene or pathway numbers.

Additional file 6: Figure S4. Different expression patterns of transcription factor genes in leaves (A) and roots (B) across salinity contrasts (0, 250 and 500 mM NaCl). Various TF families showing differential expression patterns under different conditions are given on the right side of heat map. The middle-upper scale represents FPKM values with z-score normalization.

Additional file 7: Figure S4A. Ka/Ks ratio along each branch of the phylogenetic tree. The mangrove lineage (S. alba) is colored red, set as the foreground.

Additional file 8: Table S4. PSGs Related to Salt Adaptation in Sonneratia alba.

Additional file 9: Figure S5. The seedlings of S. alba.

Additional file 10: Table S5. Primer sequences for real-time quantitative PCR analysis.

Abbreviations
AOC: Allene oxide cyclase; DEGs: Differentially expressed genes; FPKM: Fragments Per Kilobase of exon model per Million mapped reads; GO: Gene Ontology; KEGG: Kyoto Encyclopedia of Genes and Genomes; KOs: KEGG orthology; NHX: Sodium/hydrogen exchanger; PIP: Plasma membrane intrinsic protein; PSGs: Positively selected genes; qPCR: Real-time quantitative PCR; ROS: Reactive oxygen species; SOS: Salt overly sensitive; TFs: Transcription factors; TIP: Tonoplast intrinsic protein; UV: Ultraviolet; WEGO: Web Gene Ontology Annotation Plotting

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Authors’ contributions
S.S. and Z.H. conceived the study; X.F., S.X., J.L. and Q.C. analysed the data; S.X., H.Z. and S.S. wrote the manuscript; all authors read and approved the final manuscript.

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Availability of data and materials
The RNA-seq data have been deposited in the Genome Sequence Archive database (Accession No. PRJCA001412) and the NCBI Sequence Read Archive database (Accession No. PRJNA615770).

Ethics approval and consent to participate
Not applicable.

Consent for publication
Not applicable.

Competing interests
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