Genome-wide transcriptional and physiological responses to drought stress in leaves and roots of two willow genotypes

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Abstract

Background: Drought is a major environmental stress that can have severe impacts on plant productivity and survival. Understanding molecular mechanisms of drought responses is crucial in order to breed for drought adapted plant cultivars. The aim of the present study was to investigate phenotypic and transcriptional drought responses in two willow genotypes (520 and 592) originating from an experimental cross between S. viminalis × (S. viminalis × S. schwerinii). Willows are woody perennials in the Salicaceae plant family that are grown as bioenergy crops worldwide.

Methods: An experiment was conducted where plants were exposed to drought and different eco-physiological parameters were assessed. RNA-seq data was furthermore generated with the Illumina technology from root tips and leaves from plants grown in drought and well-watered (WW) conditions. The RNA-seq data was assembled de novo with the Trinity assembler to create a reference gene set to which the reads were mapped in order to obtain differentially expressed genes (DEGs) between the drought and WW conditions. To investigate molecular mechanisms involved in the drought response, GO enrichment analyses were conducted. Candidate genes with a putative function in the drought response were also identified.

Results: A total of 52,599 gene models were obtained and after filtering on gene expression (FPKM ≥ 1), 35,733 gene models remained, of which 24,421 contained open reading frames. A total of 5,112 unique DEGs were identified between drought and WW conditions, of which the majority were found in the root tips. Phenotypically, genotype 592 displayed less growth reduction in response to drought compared to genotype 520. At the transcriptional level, genotype 520 displayed a greater response in the leaves as more DEGs were found in genotype 520 compared to genotype 592. In contrast, the transcriptional responses in the root tips were rather similar between the two genotypes. A core set of candidate genes encoding proteins with a putative function in drought response was identified, for example MYBs and bZIPs as well as chlorophyll a/b binding proteins.

Discussion: We found substantial differences in drought responses between the genotypes, both at the phenotypic and transcriptional levels. In addition to the genotypic variation in several traits, we also found indications for genotypic variation in trait plasticity, which could play a role in drought adaptation. Furthermore, the two genotypes displayed overall similar transcriptional responses in the root tips, but more variation in the leaves. It is thus possible that the observed phenotypic differences could be a result of transcriptional differences mostly at the leaf level.

Conclusions: This study has contributed to a better general understanding of drought responses in woody plants, specifically in willows, and has implications for breeding research towards more drought adapted plants.

Keywords: Drought stress, Response to drought, RNA-seq, De novo assembly, Salix, Differentially expressed genes, Candidate genes

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Background

Drought is a major environmental stress that can have severe impacts on plant productivity and survival. Upon drought, plants will perform an array of complex responses, involving molecular, biochemical, physiological and morphological changes [1, 2]. As a result of adaptation to environmental conditions, for example water availability and drought, various drought avoidance and tolerance mechanisms have evolved. These mechanisms include whole-plant changes such as shoot-root allocation, growth rate, leaf morphology, leaf abscission, stomatal conductance and photosynthetic rate, as well as molecular changes underpinned by remodelling of the transcriptome that may include upregulation of stress signalling, transcription factors and defence processes [1, 3]. Due to the availability of the *Populus trichocarpa* genome sequence [4], transcriptional responses to abiotic stresses in woody plants have primarily been studied in various poplar species. Comparisons of gene expression levels between samples from different tissues, time points and treatments has been performed using microarrays [5–7] or more recently massively parallel sequencing of total RNA, i.e. RNA-seq [8–10]. As a result, a wealth of information on genes and processes involved in the molecular mechanisms of drought responses in different poplar species is available. These studies have revealed large variation in transcriptional drought responses between different species. For example, in the leaves of *P. trichocarpa*, 5689 genes were differentially expressed between drought and control conditions [10], while in the leaves of *P. balsamifera* only 98 probe sets displayed increased transcript abundance under drought [11]. Other studies have revealed variation in drought responses between populations [12–14] and even between genotypes of the same species [11]. These observed differences could originate from genetic diversification as a result of adaptation to different drought conditions [15]. Such diversifying selection should be visible as genetic variation in important genes involved in drought response. Thus, identifying genetic variation associated with variation in phenotypic responses, could provide valuable insight into the genetic basis of drought adaptation. The ultimate goal would be to pinpoint genetic variants that are associated with phenotypic responses to drought that could then be used in marker assisted selection to produce cultivars better adapted to drought conditions. In order to do so, there is an urgent need for information on phenotypic and transcriptional responses from species other than model-species. In this work we have examined and compared drought responses in willow genotypes (genus *Salix*, the sister genus to *Populus*) that, like poplars, are woody perennials in the Salicaceae family. Species from both genera have for a long time been grown as bioenergy crops worldwide and there are active breeding programs for developing new high performing varieties [16, 17]. Willows are particularly suitable for cultivations in regions with good water supply such as most of Northern Europe [18]. However, considering that willows appear to be prone to drought stress [19, 20], drought tolerant varieties are desired for cultivations in southern Europe [21]. For the purpose of bioenergy, *Salix viminalis*, *S. dascyclados* and *S. schwerinii* and their hybrids are the mostly used species in Europe as they display rapid growth and high biomass yields [16, 22]. There is relatively high genetic diversity in natural populations of willows [23], which have proven useful for the identification of quantitative trait loci (QTLs) of different traits such as frost and rust resistance and phenology [24–28], growth, water-use efficiency and drought tolerance [19, 29, 30].

The aim of the present study was to investigate phenotypic and transcriptional drought responses in two willow genotypes originating from an experimental cross between *S. viminalis* × (*S. viminalis* × *S. schwerinii*) [31]. In order to achieve this, we performed a controlled experiment in a phytotrone where a number of phenotypic measurements were assessed. Since different organs show very different responses to drought [32–34] we estimated the effects of drought separately in mature leaves and root tips. We performed massive parallel sequencing of RNA (RNA-seq) on the Illumina platform, assembled the sequencing reads de novo into gene models and then mapped the reads back to the gene models in order to quantify the level of gene expression per gene model among the different samples. We then performed GO enrichment analyses on the set of differentially expressed genes to identify important functional categories involved in the response to drought in the two tissues and genotypes.

Results

Phenotypic responses to drought

Plants grown in WW conditions were both taller (mean ± standard deviation (SD)) 135.5 ± 6.4 cm), had more sylleptic shoots and greater total root, shoot and leaf biomasses compared to the drought stressed plants (height mean 84.3 ± 10.5 cm; Fig. 1a, d, e, i, m; Table 1, main effect treatment). These differences were established already shortly after the onset of the experimental treatments, and the same overall pattern remained across the whole experimental period (Fig. 2a for plant height). Leaf chlorophyll concentration, here assessed in terms of SPAD i.e. leaf nitrogen (N) per leaf area [35], was significantly higher in the drought treatment compared to the WW condition (Table 1), and also that pattern was similar across the entire experimental period (Fig. 2b). In contrast to the leaf chlorophyll (or leaf N) concentration, the total accumulation of N was reduced by drought. Thus, at the end of the experiment, total number of leaves and total N accumulation in leaves,
Fig. 1 Means (± SD) for various eco-physiological traits in genotype 592 and 520. The genotypes were grown in a growth chamber (three blocks) and exposed to two experimental conditions (well-watered (white) and drought (black)). DW = dry weight (biomass), RGR = relative growth rate, LAR = leaf area ratio, LAP = leaf area productivity, SLA = specific leaf area, RWC = relative water content, SPAD values indicate leaf chlorophyll content.
Table 1 ANOVA for various eco-physiological traits measured in genotype 592 and 520 grown in a growth chamber (three blocks) and exposed to two experimental conditions: well-watered or drought.

| Traits                      | Source of variation | Block (df. = 2) | Genotype (G) (df. = 1) | Treatment (T) (df. = 1) | G x T (df. = 1) | Error (df. = 52) |
|-----------------------------|---------------------|-----------------|------------------------|-------------------------|-----------------|------------------|
| Total plant DW* (g)         |                     | 0.048           | 0.055                  | 20.355                  | 0.315           | 0.059            |
|                            | p-value             | 0.452           | 0.338                  | <0.001                  | 0.025           |                  |
| Total leaf N pool (g)       |                     | 0.020           | 0.024                  | 0.972                   | 0.034           | 0.005            |
|                            | p-value             | 0.030           | 0.038                  | <0.001                  | 0.014           |                  |
| RGR* (g g⁻¹ wk⁻¹)           |                     | 0.003           | 0.006                  | 1.188                   | 0.018           | 0.003            |
|                            | p-value             | 0.456           | 0.184                  | <0.001                  | 0.025           |                  |
| Number of syll. shoots      |                     | 50              | 2269                   | 1050                    | 874             | 46               |
|                            | p-value             | 0.342           | <0.001                 | <0.001                  | <0.001          |                  |
| Leaf DW* (g)                |                     | 0.343           | 0.857                  | 31.523                  | 0.057           | 0.302            |
|                            | p-value             | 0.328           | 0.098                  | <0.001                  | 0.665           |                  |
| L/W/W (%)                   |                     | 108             | 2640                   | 399                     | 1465            | 220              |
|                            | p-value             | 0.614           | <0.001                 | 0.183                   | 0.013           |                  |
| LAR* (m² g⁻¹)               |                     | 2.87            | 97.30                  | 13.57                   | 1.39            | 0.92             |
|                            | p-value             | 0.052           | <0.001                 | <0.001                  | 0.223           |                  |
| SLA* (m² kg⁻¹)              |                     | 58.0            | 42.2                   | 390.8                   | 10.7            | 12.7             |
|                            | p-value             | 0.015           | 0.074                  | <0.001                  | 0.363           |                  |
| Shoot DW* (g)               |                     | 0.018           | 0.077                  | 22.310                  | 0.042           | 0.067            |
|                            | p-value             | 0.768           | 0.289                  | <0.001                  | 0.434           |                  |
| SW/W (%)                    |                     | 441             | 1410                   | 15                      | 2190            | 223              |
|                            | p-value             | 0.149           | 0.015                  | 0.787                   | 0.003           |                  |
| LAP* (g m⁻² wk⁻¹)           |                     | 68              | 728                    | 8080                    | 5               | 40               |
|                            | p-value             | 0.191           | <0.001                 | <0.001                  | 0.729           |                  |
| SPAD* (rel. units)          |                     | 3.6             | 844.1                  | 46.9                    | 2.3             | 8.1              |
|                            | p-value             | 0.645           | <0.001                 | 0.020                   | 0.596           |                  |
| Root DW* (g)                |                     | 0.476           | 0.321                  | 15.384                  | 0.405           | 0.126            |
|                            | p-value             | 0.029           | 0.116                  | ≤0.001                  | 0.079           |                  |
| RW/W (%)                    |                     | 2490            | 31                     | 4997                    | 643             | 114              |
|                            | p-value             | <0.001          | 0.606                  | <0.001                  | 0.021           |                  |
| No. of leaves               |                     | 546             | 2640                   | 36803                   | 1995            | 231              |
|                            | p-value             | 0.103           | 0.001                  | <0.001                  | 0.005           |                  |
| Leaf temperature (°C)       |                     | 9.56            | 0.19                   | 13.56                   | 1.26            | 0.70             |
|                            | p-value             | <0.001          | 0.607                  | <0.001                  | 0.185           |                  |
| RWC* (%)                    |                     | 105.1           | 86.7                   | 818.2                   | 132.0           | 199.3            |
|                            | p-value             | 0.595           | 0.513                  | 0.048                   | 0.420           |                  |
| Leaf Δ¹³C (%)               |                     | 1.7             | 40.6                   | 138.8                   | 14.5            | 1.1              |
|                            | p-value             | 0.224           | <0.001                 | <0.001                  | 0.001           |                  |
| Leaf Δ¹⁸O (%)               |                     | 0.16            | 0.12                   | 59.12                   | 5.94            | 0.27             |
|                            | p-value             | 0.555           | 0.503                  | <0.001                  | <0.001          |                  |
which is a main driver for growth, were greatly reduced by drought (Fig. 1b; Table 1, main effect treatment). The drought-exposed plants also had greater relative biomass allocation to roots compared to the WW plants (Fig. 1n; Table 1, main effect treatment). In contrast to roots, relative shoot and leaf allocation (SW/W, LW/W) was not significantly affected by drought stress (Fig. 1f, k; Table 1).

Many traits differed between the two genotypes (Table 1, main effect genotype). At the end of the experiment, mean height growth was significantly higher for genotype 592 compared to 520, both in the WW and in the drought conditions (Fig. 2a; ANOVA: main effect genotype, p-value (p) = 0.007, not shown). Genotype 592 also produced many more sylleptic shoots than genotype 520, but had fewer leaves and lower total leaf N pool (Fig. 1b, d, o; Table 1, main effect genotype). Mean total biomass at the start of the experimental treatments was similar in the two treatments (ANOVA: main effect genotype = 0.231). The biomass allocation to roots (RW/W) was similar between the genotypes, whilst leaf and shoot allocation varied significantly between the genotypes (Fig. 1f, j, n; Table 1 main effects genotype). There were also several significant interaction effects between genotype and treatment, indicating differential drought responses for the two genotypes. For example, genotype 592 displayed greater drought-induced increase in root biomass allocation (RW/W) than 520; leaf biomass allocation decreased (i.e. 520) and increased (i.e. 592) in response to drought; and shoot biomass allocation increased in response to drought only in genotype 520 (Fig. 1f, j, n; Table 1 Genotype x Treatment interaction (G x T) effects). In contrast, genotype 520 was characterised by generally higher leaf area ratio (LAR) and lower leaf area productivity (LAP) compared to 592 (eqn 1; Fig. 1g, k; Table 1, G x T effects).

Relative growth rate (RGR) was greatly reduced by drought condition (Table 1, main effect genotype). The end of the experiment, mean height growth was significantly higher for genotype 592 compared to 520, both in the WW and in the drought conditions (Fig. 2a; ANOVA: main effect genotype, p-value (p) = 0.007, not shown). Genotype 592 also produced many more sylleptic shoots than genotype 520, but had fewer leaves and lower total leaf N pool (Fig. 1b, d, o; Table 1, main effect genotype). Mean total biomass at the start of the experimental treatments was similar in the two treatments (ANOVA: main effect genotype = 0.231). The biomass allocation to roots (RW/W) was similar between the genotypes, whilst leaf and shoot allocation varied significantly between the genotypes (Fig. 1f, j, n; Table 1 main effects genotype). There were also several significant interaction effects between genotype and treatment, indicating differential drought responses for the two genotypes. For example, genotype 592 displayed greater drought-induced increase in root biomass allocation (RW/W) than 520; leaf biomass allocation decreased (i.e. 520) and increased (i.e. 592) in response to drought; and shoot biomass allocation increased in response to drought only in genotype 520 (Fig. 1f, j, n; Table 1 Genotype x Treatment interaction (G x T) effects). In contrast, genotype 520 was characterised by generally higher leaf area ratio (LAR) and lower leaf area productivity (LAP) compared to 592 (eqn 1; Fig. 1g, k; Table 1, G x T effects).
main effect genotype). Following eqn. 2, the greater LAR of 520 was mostly a consequence of greater leaf biomass allocation (LW/W; Table 1, main effect genotype; Fig. 1f). In contrast, the higher LAP of 592 was a consequence of greater leaf chlorophyll content (SPAD index; Fig. 1; Table 1, main effect genotype) in that genotype. Ultimately, the RGR of the high-SPAD genotype (i.e. 592), was less affected by drought compared to the high-LAR genotype (i.e. 520).

Leaf temperature and relative water content (RWC) varied between growth conditions, but were similar between genotypes (Fig. 1p, q; Table 1, main effect treatment and genotype). In drought condition, RWC decreased, leaf temperature increased, stomata were more closed (higher leaf Δ18O indicates lower stomatal conductance, g), intrinsic water use efficiency was higher (lower leaf Δ13C indicates higher intrinsic water use efficiency), and also carboxylation efficiency was higher (lower ratio of Δ13C and Δ18O indicates higher carboxylation efficiency) (Fig. 1p-t, Table 1, main effect treatment). Genotype affected isotope ratios, and the genotype 592 had lower intrinsic water use efficiency (i.e. higher leaf Δ13C) and lower carboxylation efficiency (i.e. higher ratio of Δ13C and Δ18O) than genotype 520 (Table 1, main effect genotype). In general, leaf gas exchange as reflected by the isotope data indicates small genotype differences under WW conditions along with great genotype differences in drought conditions. This pattern was statistically supported by highly significant interaction effects in all isotope measures (Fig. 1r-t, Table 1). More specifically, the drought-induced decrease in stomatal conductance (i.e. increase in leaf Δ18O) was less pronounced in 592 compared to 520. Also drought-induced increase in intrinsic water use efficiency (i.e. decrease in leaf Δ13C) was less pronounced in 592 than in 520 (Table 1, G x T effects).

De novo assembly and identification of DEGs

Illumina sequencing of 36 libraries prepared from RNA extracted from leaf and root tip tissues from two willow genotypes generated 7.91 × 10^8 paired-end sequencing read pairs that corresponds to 158 × 10^9 base pairs (bp) (Table 2). Sequencing read pair numbers per library varied between 18.1 and 33.3 million and did not differ significantly between tissues (ANOVA: p = 0.60), genotypes (ANOVA: p = 0.55) or treatment (ANOVA: p = 0.39). When all sequencing reads were combined and assembled de novo using the Trinity assembler, 91,701 contigs were obtained, which represented 52,599 gene models. After filtering on gene expression (fragments per kilobase of transcript per million mapped fragments (FPKM) ≥ 1), 35,733 gene models remained of which 24,421 contained open reading frames (Additional file 1). The gene models in this high confidence gene set ranged in size from 301 to 16,559 bp with a mean length of 1485 bp and a median length of 1273 bp (Additional file 2). 40.6 % of all reads mapped to the high confidence gene set with the parameters used for expression analysis. The majority of them (99.7 %) mapped to one unique position. More detailed mapping statistics can be found in Table 2. The program edgeR was used with the “calcNormFactors()” function to identify DEGs between drought and WW conditions in: 1) leaves of genotype 592 (five replicates), 2) leaves of genotype 520 (five replicates), 3) root tips of genotype 592 (four replicates) and 4) root tips of genotype 520 (four replicates). Normalized read counts are listed in Additional file 3 and normalization factors used in the edgeR analyses are listed in Additional file 4. Genes were defined as upregulated DEGs if false discovery rate (FDR) was ≤ 0.05 and log2 fold change (FC) ≥ 1 and as downregulated DEGs if FDR was ≤ 0.05 and log2 FC ≤ -1. A total of 6935 DEGs were identified across genotypes and tissues, representing 5112 unique gene models. 3082 DEGs were upregulated (of which 2175 were unique) and 3853 were downregulated (of which 2974 were unique) (Table 3, Additional file 5). Much fewer DEGs were found in the leaves than in the root tips (Table 3, Additional file 5), a pattern that was consistent across both genotypes. This result manifests the small effects of drought on the overall transcription in the young mature leaves in these two genotypes and shows that root tips instead have a much stronger

| Table 2 Summary of Illumina sequencing and mapping of the sequencing read to the genes in the high confidence gene set |
|-------------------------------------------------|-----------------|-----------------|----------------|-----------------|
| Total reads                               | Total mapped reads | % mapped reads | # uniquely mapped reads | % uniquely mapped reads |
|-------------------------------------------------|-----------------|----------------|----------------|-----------------|
| 520 leaves WW                               | 208,238,566     | 85,078,408     | 40.86          | 84,848,522      | 99.73            |
| 520 roots WW                                | 177,073,828     | 70,448,906     | 39.79          | 70,245,780      | 99.71            |
| 520 leaves drought                          | 223,031,526     | 111,012,324     | 49.77          | 110,732,434     | 99.75            |
| 520 roots drought                           | 170,916,478     | 70,687,714     | 41.36          | 70,463,550      | 99.68            |
| 592 leaves WW                               | 202,996,992     | 96,790,050     | 47.68          | 96,548,610      | 99.75            |
| 592 roots WW                                | 185,878,236     | 90,739,420     | 48.82          | 90,492,450      | 99.73            |
| 592 leaves drought                          | 234,211,410     | 63,601,116     | 27.16          | 63,220,510      | 99.40            |
| 592 roots drought                           | 178,292,458     | 52,870,112     | 29.55          | 52,564,102      | 99.42            |
| Total                                      | 1,581,276,494   | 641,228,050   | 40.55          | 639,115,958     | 99.67            |

*a* Relative to the total number of reads

*b* Relative to the number of mapped reads
transcriptional response to drought than leaves do. When comparing the two genotypes, more DEGs were found in the leaves of genotype 520 compared to genotype 592, while the number of DEGs in the root tips was more similar between the genotypes (this was particularly evident for the upregulated DEGs) (Table 3). Twenty-eight genes (Additional file 5) displayed significant G x T effects when using a generalized linear model and the gene ontology (GO) term ADP-binding was significantly enriched among these genes. When comparing the drought responses in the leaves and the root tips of the two genotypes for these genes, 25 displayed greater drought responses in leaves than in the root tips. The hypothesis that there was an equal chance that the effect is stronger in either tissue could thus be rejected ($p = 2.74 \times 10^{-5}$, binomial test see Additional file 6). This result further strengthens the finding of greater genotypic differences in the drought responses in the leaves than in the roots.

**Functional annotation and GO enrichment analysis**

The Blast2GO software tool was used to functionally annotate the de novo assembled gene products and for GO enrichment analyses in order to identify functions and genes involved in drought stress responses in the leaves and root tips. Of the 24,421 genes in the high confidence gene set, 15,980 were annotated with a GO term.

The GO enrichment analysis was done for up and downregulated annotated DEGs in leaves and root tips for each genotype separately using the annotated high confidence gene set as reference. All enriched GO terms in every comparison are presented in Additional file 7. Overall, few GO terms were enriched for the DEGs in the leaves, in fact none was significantly enriched for the DEGs in genotype 592, suggesting that drought had little impact on the function of the leaves in this genotype. However, genotype 520 showed a greater functional drought response as several GO terms were significantly enriched, particularly for the upregulated DEGs. For example at the molecular function (F) ontology level, “peptidyl-prolyl cis-trans isomerase activity” was one of the most significantly overrepresented terms and at the biological process (P) ontology level, different response and regulation terms were significantly overrepresented (Additional file 7). At the cellular component (C) ontology level, several terms related to the thylakoid and the cell wall were overrepresented (Additional file 7). For the downregulated DEGs at the F ontology level, “oxygen binding” was the most significantly overrepresented term (Additional file 7). At the P ontology level, “arachidonic acid metabolic process” and “epoxygenase P450 pathway” were overrepresented. No term was significantly overrepresented at the C ontology level.

Many more GO terms were enriched for the DEGs in the root tips compared to the leaves, which demonstrates the greater drought responses in this tissue compared to the leaves. In Fig. 3 we present overrepresented GO terms (FDR < 0.0001) in the P ontology level associated with upregulated (Fig. 3a) and downregulated (Fig. 3b) DEGs for both genotypes. Many GO terms were significantly or nearly significantly enriched in both genotypes, which means that the genotypes displayed overall similar responses to drought in the root tips. For the upregulated DEGs, the most affected functions were associated with biosynthetic and metabolic processes (Fig. 3a).

A core set of candidate genes involved in the drought response

To identify candidate genes involved in drought stress responses in the two willow genotypes, we extracted the upregulated DEGs in the root tips of both genotypes annotated with any of the 15 “response” GO terms listed in Table 4. This resulted in a list with 115 genes for genotype 592 and 141 genes for genotype 520 (Additional file 8). Based on annotations and reported functions in stress responses, a core set of 28 candidate genes were identified (Table 5) that could function as targets for detailed functional studies of drought responses at the molecular level in willows and related species. Several of the genes were homologous to genes encoding transcription factors, e.g. MYBs, WRKYs, bZIPs and heat stress transcription factors with known functions in stress responses. Other genes were homologous to genes encoding dehydrin and chlorophyll a/b binding proteins.

**Discussion**

**Genotype specific physiological responses to drought**

This study accommodated a growth experiment in which cuttings of two genotypes were grown in an irrigation contrast for sufficiently long time to develop considerable variation both at the transcriptional and phenotypic levels. We demonstrated that the two genotypes displayed contrasting phenotypic responses to drought. Genotype 592 was overall less affected by drought as it displayed a weaker growth

| Genotype | 592 | 520 | 592 | 520 | Total no. of DEGs |
|----------|-----|-----|-----|-----|------------------|
| Tissue   |     |     |     |     |                  |
| Leaves   |     |     |     |     |                  |
| Upregulated DEGs FDR ≤ 0.05, log2 FC ≥ 1 | 1   | 126  | 1457 | 1498 | 3082            |
| Downregulated DEGs FDR ≤ 0.05, log2 FC ≤ −1 | 34  | 123  | 2163 | 1533 | 3853            |
| Total no. of DEGs | 35  | 249  | 3620 | 3031 | 6935            |

**Table 3 Number of DEGs between drought and well-watered conditions for each genotype and each tissue**
Fig. 3 (See legend on next page.)
reduction. The weaker growth reduction of 592 was associated with greater increase in root biomass allocation (RW/W) in response to drought, enhancing the capacity to explore available water and nutrient resources. Interestingly, genotype 592 also displayed generally higher foliar SPAD values, which is in agreement with the hypothesis that higher area-based leaf N content is an acclimation to drought [36]. In contrast to 592, genotype 520 displayed generally higher evapotranspiration area (e.g. LAR) and responded to drought with strong leaf area reduction, greatly decreased stomatal conductance and increased intrinsic water use efficiency. Thus, in addition to the observed genotypic variation in mean traits (e.g. LAR, LAP, SPAD), we also found indications for genotype differences in trait variability in a drought contrast, especially for the stomatal physiology traits (leaf $\Delta^{13}C$ and $\Delta^{18}O$). Genotypic differences in trait variability, observed in a drought contrast, may be indicative of genotypic variation in trait plasticity in relation to drought. Large genotypic variation in trait plasticity of leaf traits similar to the ones in our study was found among Eucalyptus provenances originating from a rainfall gradient in Australia, and interpreted as important features in the climate adaptation of these trees [37].

**Generation of the high confidence gene set**

De novo assembly of all sequencing reads generated 52,599 gene models and after filtering out lowly expressed gene models, 35,733 remained. In a next filtering step, gene models that did not contain an open reading frame were filtered out, leaving 24,421 gene models in the high confidence gene set. Out of these, 18,128 were functionally annotated with a GO term in Blast2Go. Although it is problematic to compare these numbers with those from other studies as different tissues and also filtering criteria are often used, de novo assembly of RNA-seq data in P. pruinosa resulted in 48,653 gene models [38] and in S. matsudana 48,817 unigenes were retrieved [39], showing that the figures were overall rather similar across the different species. The number of annotated genes in this high confidence set was considerably higher than the number of gene models annotated with a GO term in P. pruinosa and P. euphratica where 11,587 [40] and 9296 in [9] were reported respectively, possibly reflecting improved annotations tools.

**Variation in the number of DEGs and enriched GO terms between the two genotypes and tissues**

When comparing the responses to drought between the two tissues we found that drought affected gene

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**Table 4** GO terms associated with upregulated DEGs in the root tips that were enriched (FDR $\leq 0.05$) in at least one of the genotypes and used to select candidate genes with putative functions in drought responses in willows

| GO-ID          | Term                                      | FDR Genotype 592/520 | No of genes in test set Genotype 592/520 | No of genes in ref set Genotype 592/520 |
|---------------|-------------------------------------------|----------------------|------------------------------------------|-----------------------------------------|
| GO:0009628    | Response to abiotic stimulus              | $7.9 \times 10^{-2}$ | 106/114                                  | 1004/996                                |
| GO:0009644    | Response to high light intensity          | $2.4 \times 10^{-2}$ | 15/15                                    | 73/73                                   |
| GO:0009642    | Response to light intensity               | $4.9 \times 10^{-3}$ | 19/19                                    | 90/90                                   |
| GO:0009636    | Response to toxic substance               | $3.5 \times 10^{-2}$ | 5/-                                      | 8/-                                     |
| GO:0009415    | Response to water                         | $4.9 \times 10^{-2}$ | 17/-                                     | 96/-                                    |
| GO:0042538    | Hyperosmotic salinity response            | $3.2 \times 10^{-2}$ | 9/-                                      | 73/73                                   |
| GO:0098542    | Defense response to other organism        | $9.3 \times 10^{-3}$ | 47/-                                     | 34/2                                    |
| GO:009637     | Response to blue light                    | $1.8 \times 10^{-5}$ | 19/-                                     | 51/-                                    |
| GO:0010218    | Response to far red light                | $1.1 \times 10^{-4}$ | 15/-                                     | 37/-                                    |
| GO:009416     | Response to light stimulus                | $1.4 \times 10^{-2}$ | 55/-                                     | 432/-                                   |
| GO:009314     | Response to radiation                     | $2.6 \times 10^{-2}$ | 57/-                                     | 466/-                                   |
| GO:0010114    | Response to red light                     | $5.1 \times 10^{-5}$ | 14/-                                     | 29/-                                    |
| GO:0051788    | Response to misfolded protein             | $3.4 \times 10^{-2}$ | 0/-                                      | 93/-                                    |
| GO:0002679    | Respiratory burst involved in defense response | $2.5 \times 10^{-2}$ | 7/-                                      | 19/-                                    |
| GO:006974     | Cellular response to DNA damage stimulus  | $4.9 \times 10^{-2}$ | 6/6                                      | 256/256                                 |
### Table 5: A core set of candidate genes with reported or putative functions in stress responses

| Sequence ID       | Description                                                                 | BLAST E-value | Protein description at TAIR obtained by BLASTX with the Salix sequence                                                                 |
|-------------------|------------------------------------------------------------------------------|---------------|----------------------------------------------------------------------------------------------------------------------------------------|
| c34790_g1_i6      | ap2-like ethylene-responsive transcription factor ails                        | $5.2 \times 10^{-75}$ | Encodes a member of the AP2 family of transcriptional regulators.                                                                        |
| c19050_g1_i3      | bzip transcription factor 60                                                 | $4.2 \times 10^{-144}$ | Consists of a bZIP DNA binding domain followed by a putative transmembrane domain.                                                    |
| c25859_g1_i2      | Chlorophyll a-b binding protein chloroplastic                                 | $6.5 \times 10^{-166}$ | Photosystem II encoding the light-harvesting chlorophyll a/b binding protein CP26 of the antenna system of the photosynthetic apparatus. |
| c16933_g1_i1      | Chlorophyll a-b binding chloroplastic                                         | $2.5 \times 10^{-175}$ | PSII type II chlorophyll a/b-binding protein (Lhca2*) mRNA.                                                                         |
| c3652_g1_i1       | Chlorophyll a-b binding protein chloroplastic                                 | $5.1 \times 10^{-157}$ | Encodes a component of the light harvesting complex associated with photosystem I.                                                  |
| c28579_g1_i2      | Chloroplast stem-loop binding protein of 41 kda chloroplastic                | $0$           | Encodes CHLOROPLAST RNA BINDING (CRB), a putative RNA-binding protein.                                                               |
| c22127_g1_i1      | Dehydrin                                                                     | $5.0 \times 10^{-93}$ | Encodes a gene induced by low temperature and dehydration.                                                                         |
| c16392_g1_i1      | Ethylene-responsive transcription factor 4-like                               | $4.1 \times 10^{-62}$ | Encodes a member of the ERF (ethylene response factor) subfamily B-1 of ERF/AP2 transcription factor family (ATERF-4).               |
| c20465_g2_i1      | Heat shock transcription factor a2 isoform 1                                 | $1.3 \times 10^{-57}$ | Member of Heat Stress Transcription Factor (Hsf) family. Expression is regulated by DREB2A.                                           |
| c23076_g1_i1      | Heat stress transcription factor a-3                                         | $0$           | Member of Heat Stress Transcription Factor (Hsf) family.                                                                           |
| c33073_g3_i1      | Heat stress transcription factor a-6b                                         | $2.9 \times 10^{-98}$ | Member of Heat Stress Transcription Factor (Hsf) family.                                                                           |
| c14799_g1_i1      | Heat stress transcription factor a-6b-like                                    | $5.7 \times 10^{-127}$ | Member of Heat Stress Transcription Factor (Hsf) family.                                                                           |
| c32554_g10_i1     | Heat stress transcription factor b-2-b                                       | $5.1 \times 10^{-96}$ | Member of Heat Stress Transcription Factor (Hsf) family.                                                                           |
| c29038_g2_i1      | Heat stress transcription factor c-1                                         | $5.8 \times 10^{-166}$ | Member of Heat Stress Transcription Factor (Hsf) family.                                                                           |
| c31828_g1_i3      | Mitochondrial transcription termination factor family protein isoform 1      | $0$           | Encodes a putative mitochondrial transcription termination factor.                                                                     |
| c6068_g1_i1       | myb-like transcription factor partial                                         | $1.2 \times 10^{-103}$ | Putative transcription factor MYB108 (MYB108) mRNA.                                                                                 |
| c21976_g1_i1      | myb transcription factor family protein                                       | $2.9 \times 10^{-179}$ | Encodes a MYB transcription factor involved in wounding and osmotic stress response. Member of the R2R3 factor gene family.          |
| c35418_g2_i1      | Photosystem ii 10 kda chloroplastic                                         | $8.9 \times 10^{-75}$ | Encodes for the 10 kDa PsbR subunit of photosystem II (PSII).                                                                       |
| c26360_g1_i3      | Photosystem ii stability assembly factor chloroplastic                        | $0$           | Encodes a stability and/or assembly factor of photosystem II.                                                                      |
| c34906_g2_i3      | Probable wrky transcription factor 33                                         | $0$           | Member of the plant WRKY transcription factor family. Involved in response to various abiotic stresses.                               |
| c31974_g2_i2      | Probable wrky transcription factor 40                                         | $1.7 \times 10^{-179}$ | Pathogen-induced transcription factor.                                                                                               |
| c26874_g1_i3      | Probable wrky transcription factor 48                                         | $4.9 \times 10^{-151}$ | Encodes WRKY48, a member of the WRKY Transcription Factor. WRKY48 is a stress- and pathogen-induced transcriptional activator that represses plant basal defense. |
| c27657_g4_i1      | Protein odorant1-like                                                         | $0$           | Encodes a MYB transcription factor involved in wounding and osmotic stress response. Member of the R2R3 factor gene family.          |
| c3736_g1_i1       | Protochlorophyllide chloroplastic                                            | $1.1 \times 10^{-82}$ | Encodes for a protein with protochlorophyllide oxidoreductase activity. The enzyme is NADPH+ and light-dependent.                     |
| c24362_g1_i1      | Protochlorophyllide reductase-like                                           | $0$           | Encodes for a protein with protochlorophyllide oxidoreductase activity. The enzyme is NADPH+ and light-dependent.                     |
| c21511_g1_i2      | Stress enhanced protein chloroplastic                                       | $1.7 \times 10^{-80}$ | Encodes a stress enhanced protein that localizes to the thylakoid membrane and whose mRNA is upregulated in response to high light intensity. It may be involved in chlorophyll binding. |
| c34938_g1_i3      | BZIP17, transcription factor                                                 | $0$           | bZIP17 appears to regulate transcription as part of a salt and osmotic stress response.                                              |
| c21535_g1_i1      | Transcription factor myb108-like                                            | $7.1 \times 10^{-150}$ | Member of the R2R3 factor gene family.                                                                                               |

*E-value between the sequence and the best hit in the Viridiplantae subset of the NCBI nr database*
expression of many more genes in the root tips compared to the leaves, a result that was consistent across both genotypes. A similar pattern was previously observed in hybrid poplars (P. deltoides × P. nigra) [7] where about twice as many probe sets displayed a significant change in response to drought in the root tips compared to the mature leaves. The authors proposed that this difference could in part be explained by the higher sensitivity to water deprivation in an actively growing tissue, i.e. the root tips. Judging from the number of DEGs in the root tips, the two genotypes displayed rather similar responses. This similarity was also evident in the enriched GO terms associated with these DEGs that were present in both genotypes. A somehow more unexpected result was the numerous enriched terms in the cellular component level that was related to the chloroplast. The overall lack of response of drought on gene expression in the mature leaves was also unexpected as several previous studies have reported a greater response in leaves. For instance, in P. trichocarpa 5689 DEGs were identified in leaves [10], in P. euphratica, where 5083 genes were differentially expressed in leaves [8] and in hybrid poplar (P. deltoides × P. nigra) 2120 were affected in leaves [7]. Results from the present study instead resemble the results in P. balsamifera where 98 probe sets displayed increased transcript abundance to drought [11]. Less responses were also seen in shoots of P. balsamifera [6]. In contrast to the root tips, the genotypes displayed a greater difference in the responses in the leaves with much fewer DEGs and enriched GO terms in genotype 592 than in genotype 520. Strong genotypic effects were previously found in leaves of hybrid poplars [7] and P. balsamifera (six genotypes) [11]. This was also seen in shoots of hybrid poplar clones (P. deltoides × P. nigra and P. nigra × P. maximowiczii) [6]. Interestingly, genotype 592 also displayed less physiological responses at the leaf level compared to genotype 520 supporting a correlation between transcriptional and physiological responses. For example, the drought-induced shifts in stomatal physiology (as reflected by Δ^{13}C and Δ^{18}O) were more pronounced in the genotype 520, which coincides with the overrepresented GO terms “thylakoid” and cell wall in the cellular component ontology level. Among the 28 genes that displayed significant G x T effects, several encoded disease resistant proteins and leucine-rich repeat proteins. An exciting hypothesis is that the two genotypes carry different alleles at these genes rendering different drought responses.

A core set of candidate genes involved in drought stress responses
We have produced a core set of candidate genes that we find particularly interesting and that should be further examined for their role in the molecular stress responses and we discuss some of them here. Among our candidate genes were several transcription factors i.e. MYBs, bZIPs and WRKYs, with known functions in ABA-dependent stress responses [41]. They among other things regulate stress responses via modulation of transcription of downstream genes. Furthermore, several candidate genes encoded chlorophyll a/b binding proteins (i.e. CABs). Typical for this protein family is the CAB domain, containing the amino acid residues involved in pigment binding. There are some indications that they function in high-light protection in a broad sense and that they encode homologs to light-harvesting-like proteins (LILs). All organisms performing oxygenic photosynthesis also contain LILs [42], which are highly homologous to the higher plant light-harvesting antenna and contain the CAB domain. The LILs are regulated opposite to the light-harvesting complex proteins (LHCs); under high light condition - when the expression of LHCs is repressed - the LILs are upregulated. While in cyanobacteria LILs have been found to respond to different stresses, their function in plant was thought to be restricted to light stress [43]. It is thus possible that these genes and gene products play an important role in the stress response – even in non-photosynthetic tissue. Also in Arabidopsis [44] and rice [45] enhanced expression of such genes was observed in roots exposed to drought stress, indicating a general importance of these genes independent of the species.

Conclusions
In this study we report the first large transcriptome study in willows where we have compared physiological and transcriptional responses to drought between two genotypes and two tissues. A de novo assembly and subsequent filtering of root tip and leaf transcriptomes across the two genotypes and two conditions, generated a set with 24,421 genes. A total of 5112 unique DEGs were identified between the drought and WW conditions, of which the majority were found in the root tips. We also found substantial differences in drought responses between the genotypes, both at the phenotypic and transcriptional level. Phenotypically, genotype 592 displayed less growth reduction in response to drought compared to genotype 520, which suggests that this genotype is more drought adapted than genotype 520. In addition to the genotypic variation in several mean traits, we also found considerable genotypic variation in trait plasticity (especially leaf traits), which may play a role in drought adaptation. At the transcriptional level, genotype 520 displayed a greater response in the leaves as many more DEGs were found compared to genotype 592. As the two genotypes displayed overall similar transcriptional responses in the root tips, but more variation in the leaves, it is possible
that the observed phenotypic differences between the genotypes are due to transcriptional differences in their leaves. We also identified a core set of candidate genes encoding proteins with a putative function in drought response, for example MYBs, bZIPs and WRKYs and chlorophyll a/b binding proteins. Knowledge from this study increases our understanding of the physiological and molecular basis of drought responses in willows, however these results are applicable to all woody plants.

Methods

Plant material

Two Salix genotypes (520 and 592) from the S₁ pedigree (N = 463) [31] were used in this study. S₁ is a cross between the female clone 78183, a diploid S. viminalis and Björn, a diploid Salix viminalis L × S. schwerinii E Wolf hybrid male. Results from previous greenhouse and phytotron experiments showed that both genotype 520 and 592 were high producing clones, both in WW and in drought conditions with repeated periods of drought [30]. Interestingly, they displayed different responses to drought, as an example, genotype 592 lost many more leaves than 520 [30].

The S₁ population is planted and conserved in a plant archive owned by the Swedish University of Agricultural Sciences, near Uppsala (59°49’ N 17°40’ E, central Sweden) where dormant shoots from the two genotypes were collected the 2nd of February 2011 and stored in −4 ºC. On the 17th of February, fifty, six cm long cuttings were prepared for each genotype and planted in 3-L pots filled with commercial soil (“Weibull’s Krukväxtjord Lera & Kisel”, organic matter 95 %; pH 5.5-6.5; 182 g/m³ N, 91 g/m³ P, 195 g/m³ K, maximum 50 g/m³ micro nutrients). The relatively small pot size and the resulting high biomass to substrate volume ratio at final harvest has probably decreased growth in all genotypes and treatments compared to field conditions [46], but visible observation revealed that roots never occupied more than approximately half of the substrate volume per pot at final harvest. The experiment was conducted in one walk-in growth chamber where the plants were organized in three groups. Each group consisted of equal number of plants per genotype, thus each group represented one block or replicate. The plants were grown for 38 days under 20 ºC constant temperature, 70 % relative humidity and 16 h photoperiod (300 µmol PAR m⁻² s⁻¹) with regular watering. The positions of the plants within the blocks as well as the positions of the blocks in the chamber were changed every day and the plants were pruned so that only the main shoots were preserved. For the drought treatments starting the 28th of March, 30 plants at the same growth stage of each genotype were selected, of which 15 were assigned as controls with regular watering (well-watered or WW condition) and 15 subjected to drought stress (drought condition) for 31 days. The aim was to keep the drought stressed plants just above their wilting point throughout the treatment period. The wilting point or the minimal amount of soil moisture the plants required not to wilt was therefore studied in six plants for each genotype prior to the drought stress treatment was initiated. The plants were weighed daily until they started to wilt (no water was supplied). To apply drought stress, the onset of visible wilting signs (“wilting point”) of individual plants was used as a physiological drought stress indicator. Thus, in the drought treatment, any water supply to individual pots was initially withheld until the plants had reached their individual wilting points, at which pot weights were determined. At a daily basis throughout the experimental period in April, all pots were weighed and water was added in the necessary amounts to keep the substrate at a more or less constant minimum water supply level. Applying this procedure, all plants in the drought stress condition were kept close to their wilting points throughout the whole treatment period. The WW plants were regularly watered to field capacity of the soil throughout the treatment period and weighed every day. To account for increased consumption as the plants grew, watering quantities were slightly increased throughout the treatment period. When the drought stressed plants were kept near the wilting point, they received approximately 50 % of the water supplied to the WW plants.

Eco-physiological measurements

Two harvests were carried out during the experiment, one initial harvest just before the start of the treatment period (28th of March) and a final harvest at the end of the treatment period (27th – 28th of April). The harvested plants (five plants per block, treatment and genotype) were separated into leaves, stems, the original cutting, roots, sylleptic shoots and sylleptic leaves and all parts were oven dried for 48 h at 70 °C. Excised leaves were not included. Leaf areas were determined on fresh leaves with an area meter (LI-3100, LiCor Inc., Lincoln, NE, USA). Both for the initial and the final harvest, the dry weights of each plant compartment as well as leaf areas of all plants were assessed. The total dry weight (DW) per plant was estimated as the sum of all plant compartments excluding the cutting. Height was assessed non-destructively throughout the treatment period. Height was defined as length from the base of the shoot to the thickest part of the apical bud. Leaf chlorophyll content (SPAD index) of one leaf (per plant) in the upper middle canopy was measured non-destructively at the same occasions as plant height by using a portable chlorophyll meter (SPAD-502, Konica Minolta Sensing Inc. Japan).
In *Salix* spp, the SPAD index has previously been shown to be closely related to leaf nitrogen (N) content [35]. Also, leaf temperature was measured on the same leaf using a portable infrared thermometer (IR400, Extech Instruments). An average of five measurements per leaf was used for both assessments. In general, relative water content (RWC) is a measure of drought resistance [47] and was measured on leaves sampled from all plants at the day of the final harvest. For determination of RWC, two fully developed leaves per plant were sampled. A simplified measure of RWC was determined as described by Weih and Nordh [48] and indicates here the rate of water loss from the leaf tissue. Leaves were weighted directly after sampling (wt0) and placed on sheets of filter paper in room temperature. Leaves were weighed again after 4 h (wt4) and after they had oven dried for 24 h at 70 °C (wt24). RWC was calculated according to the following formula: 

\[
\text{RWC} (\%) = \frac{\text{wt4} - \text{wt24}}{\text{wt0} - \text{wt24}} \times 100
\]

The sampled leaves were used for total nitrogen (N); carbon and oxygen isotope analysis performed by Iso-Analytical Limited, Crewe, UK. All analyses were conducted by Elemental Analyser Isotope Ratio Mass Spectrometry (EA-IRMS, Europa Scientific 20-20). Leaf N concentrations were multiplied with total leaf DW of all leaves per plant to obtain total leaf N pool. Leaf carbon and oxygen isotope measurements reflect important characteristics of gas exchange at leaf level and integrate over the lifetime of the leaf. Thus, leaf Δ¹³C is correlated with intercellular CO₂ concentration Ci and higher Ci indicates lower intrinsic water use efficiency [49]. Leaf Δ¹⁸O is a measure of stomatal conductance (gs), and increased leaf Δ¹⁸O is correlated with decreased gs [49, 50]. Finally, the ratio of Δ¹³C and Δ¹⁸O (i.e. Ci/gs) represents an estimate of carboxylation efficiency, and higher Ci/gs is correlated with lower carboxylation efficiency [51]. The reference material used for δ¹³C analysis of the samples was IA-R001 (wheat flour, δ¹³C_V-PDB = −26.43 ‰). Carbon isotope ratio of leaf samples (δ¹³C_sample) is expressed as: 

\[
\delta^{13}C_{\text{sample}} (\%) = \left( \frac{R_{\text{sample}}}{R_{\text{PDB}} - 1} \right) \times 1000
\]

where Rsample and RPDB are the ¹³C/¹²C molar abundance ratio of the leaf material and PeeDee Belemnite standard. Carbon isotope discrimination (Δ¹³C, ‰) was calculated assuming the isotopic composition of atmospheric air (δ¹³C_air) to be −8 ‰ [51] and computed as: 

\[
\Delta^{13}C = (\delta^{13}C_{\text{air}} - \delta^{13}C_{\text{sample}}) / (1 + \delta^{13}C_{\text{sample}}/1000)
\]

The reference material used for δ¹⁸O analysis was IAEA-CH-6 (sucrose, δ¹⁸O_V-SMOW = 36.4 ‰). The δ¹⁸O of the irrigation water (CO₂/H₂O equilibrium technique) was −13.8 ‰. The δ¹⁸O enrichment over the irrigation water (Δ¹⁸Osample) was computed as: 

\[
\Delta^{18}O_{\text{sample}} = \delta^{18}O_{\text{irrigation water}} - \delta^{18}O_{\text{sample}}
\]

The isotope ratio was calculated as \( \Delta^{13}C / \Delta^{18}O_{\text{sample}} \) [51].

Changes in plant growth over time were analyzed using the methods of classical growth analysis [52] and based on biomass and leaf area changes between the two consecutive harvests. Thus possible differences in relative growth rate (RGR, final plant biomass per initial plant biomass and week, g g⁻¹ wk⁻¹) between genotypes were related to differences in leaf area ratio (LAR, leaf area per total plant biomass, m² g⁻¹); leaf area productivity (LAP, total biomass growth per leaf area and week, g m⁻² wk⁻¹); specific leaf area (SLA, leaf area per leaf biomass, m² kg⁻¹); and leaf biomass fraction (LW/W, leaf biomass per total plant biomass, %). The following relationships among growth traits were utilized to analyze differences among the treatments and clones [53]:

\[
\text{RGR} = \text{LAR} \times \text{LAP}
\]

\[
\text{LAR} = \text{SLA} \times \text{LW} / \text{W}
\]

**RNA extractions and sequencing**

The day before the final harvest, two fully developed young leaves and about one centimetre of several root tips were collected from each plant that were immediately after harvesting snap frozen in liquid nitrogen and stored in −80 °C awaiting RNA extractions. Total RNA was extracted from 40 plants representing five biological replicates from leaves and root tips from the two genotypes in drought and WW conditions. Approximately 100 mg of leaves and 30 mg of root tips were taken for RNA extractions using Spectrum Plant Total RNA Kit (Sigma-Aldrich) with the On-Column DNase I digestion set (Sigma-Aldrich). After quality checking on a 2100 Bioanalyzer (Agilent), four root tip samples were discarded, leaving a total of 36 RNA samples for sequencing. One sequencing library was prepared for each of the 36 samples. The RNA samples were first treated with DNase, then one library per sample was prepared using Illumina’s TruSeq RNA Sample Prep Kit v1, where polyA-fragments were selected, followed by cDNA synthesis and ligation of amplification and sequencing adapters. Sequencing libraries were individually barcoded and then pooled with nine libraries per lane on an Illumina HiSeq 2000 instrument. All samples were sequenced as paired-end, and for 27 samples the read lengths were 100 bp and for nine samples the read lengths were 107 bp for read 1 and 144 bp for read 2. Library preparations and sequencing was performed by the SNP&SEQ Technology Platform in Uppsala.

**De novo assembly and filtering**

Sequencing reads from all 36 libraries were combined and assembled de novo using the Trinity software (version 20140717) [54] with the read trimming and digital normalization options. Default parameters were implemented except for maximum coverage of 50 in the normalization step and a minimum k-mer coverage of 20
as well as a minimum contig length of 300 in the assembly step. The contigs generated in Trinity are by default clustered in “trinity components” or gene models, where each gene model represents a putative gene. The assembly was filtered based on gene expression levels where gene models that had no contig with FPKM (fragments (read pairs) per kilobase of transcript per million mapped fragments) above or equal to 1 were filtered out. For every gene model, the contig with the highest FPKM value was retained. FPKM values were obtained by mapping the sequencing reads (from all libraries) to the contigs with the program RNA-Seq by Expectation-Maximization (RSEM) v. 1.2.12 [55]. In a second filtering step, only gene models that contained and open reading frame was retained. TransDecoder version 20140704 [56] was used with default options to predict open reading frames in the gene models from the filtered de novo assembly. This assembly is named the high confidence gene set, which was the dataset used for gene expression analyses.

Differential gene expression analysis
In order to identify genes with a putative function in the drought response, differentially expressed genes between drought and WW conditions were identified. Gene specific read counts for the genes in the high confidence gene set were obtained using the RSEM v. 1.2.12 [55], that internally uses bowtie v. 1.0.0 [57]. All sequencing libraries were mapped separately to this assembly. The applied mapping parameters allowed for up to two mismatches to reflect the allelic variation. Using the read counts obtained by RSEM, the R/Bioconductor package edgeR [58] was used to detect differentially expressed genes (DEGs) among the genes in the high confidence gene set between the drought and WW condition in leaves of genotype 592, leaves of genotype 520, root tips of genotype 592 and root tips of genotype 520. The four pairwise comparisons were made using a classic Fisher exact test. For all the analyses in edgeR, the read counts were normalized for RNA composition using the “calcNormFactors()” function. Genes that showed significant G x T effects were identified using a generalized linear model (GLM) in edgeR with an interaction model (Treatment + Tissue + Genotype + (Treatment × Tissue) + (Genotype × Treatment) + (Tissue × Genotype) + (Treatment × Tissue × Genotype)). Genotype × Treatment interaction terms were retained. In order to quantify whether the genotypes differed more in their drought responses (drought vs. WW conditions) in the leaves compared to the root tips, a statistic test was constructed and implemented to the genes with significant G x T effects (the test is described in Additional file 6). For all tests, genes were defined as differentially expressed if false discovery rate (FDR) was ≤ 0.05 and log₂ fold change (FC) ≥ 1.

Functional annotation and GO enrichment analysis
Genes in the high confidence gene set were functionally annotated with GO terms in Blast2GO (version 3.1) [59]. The blast step was performed locally against the Viridiplantae subset of the NCBI nr database downloaded 2015-08-06. All up- and downregulated DEGs for leaves, root tips in the two genotypes were tested for GO enrichment using the Fisher’s exact test, which uses the Gossip software integrated in the Blast2GO software. The rationale was to test for over- or underrepresentation of GO terms among the DEGs compared to all annotated genes in the high confidence gene set, using a cut-off threshold of FDR ≤ 0.05. The analyses were performed for the three ontology levels, biological processes (P), molecular function (F) and cellular component (C).

Availability of supporting data
The raw sequencing reads are available in the European Nucleotide Archive (ENA) with the reference number PRJEB10883 (http://www.ebi.ac.uk/ena/data/view/PRJEB10883).

Additional files

| Additional file 1: Sequences of all gene models in the high confidence gene set. (TXT 40356 kb) |
| Additional file 2: Length distribution of transcripts in filtered and high confidence gene set. (EPS 33 kb) |
| Additional file 3: Normalized read counts used in egdeR. (XLSX 12012 kb) |
| Additional file 4: Normalization factors used in egdeR. (XLSX 58 kb) |
| Additional file 5: Differentially expressed gene models (DEGs). (XLSX 1313 kb) |
| Additional file 6: Description of statistical test. (DOCX 89 kb) |
| Additional file 7: Enriched GO terms. (XLSX 203 kb) |
| Additional file 8: Genes associated with response GO terms for upregulated DEGs in 592-roots and 520-roots. (XLSX 72 kb) |

Abbreviations
ANOVA: Analysis of variance; bp: base pair; C ontology: Cellular component ontology; cm: Centimetre; DEG: Differentially expressed gene; df: Degrees of freedom; DW: Dry weight; F ontology: Molecular function ontology; FC: Fold change; FDR: False discovery rate; FPKM: fragments (read pairs) per kilobase of transcript per million mapped fragments; G x T: Genotype x Treatment interaction; GO: Gene ontology; h: hours; LAP: Leaf area productivity; LAIR: Leaf area ratio; LW/W: Leaf biomass per total plant biomass; N: Nitrogen; p: p-value; P ontology: Biological process ontology; QTL: Quantitative trait locus; RGR: Relative growth rate; RNA-seq: RNA sequencing; RW/W: Root biomass per total plant biomass; RWC: Relative water content; SD: Standard deviation; SLA: Specific leaf area; SPAD: Leaf chlorophyll content; SW/W: Shoot biomass per total plant biomass; WW: Well-watered; Δ¹³C: Carbon isotope ratio ¹³C/¹²C; Δ¹⁸O: Oxygen isotope ratio ¹⁸O/¹⁶O.

Competing interests
The authors declare that they have no competing interests.

Authors’ contributions
PP and PS carried out the bioinformatic analyses and drafted parts of the manuscript. MW analysed physiological data and drafted parts of the manuscript. ACRW carried out experiments, acquired funding and drafted
parts of the manuscript. SB conceived and designed the study, performed the experiments, acquired funding and had the main responsibility for writing the manuscript. All authors read and approved the final manuscript.

Authors’ information
Not applicable.

Availability of data and materials
Not applicable.

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