

Differences in heat tolerance, water use efficiency and growth among Douglas-fir families and varieties evidenced by GWAS and common garden studies

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ABSTRACT

Severe and frequent heat and drought events challenge the survival and development of long-generation trees. In this study, we investigated the genomic basis of heat tolerance, water use efficiency, and growth by performing genome-wide association studies in coastal Douglas-fir (*Pseudotsuga menziesii*) and intervarietal (*menziesii* x *glauca*) hybrid seedlings. GWAS results identified 32 candidate genes involved in primary and secondary metabolism, abiotic stress and signaling, among other functions. Water use efficiency (inferred from carbon isotope discrimination), photosynthetic capacity (inferred from %N), height, and heat tolerance (inferred from electrolyte leakage in a heat stress experiment) were significantly different among Douglas-fir families and varieties. High elevation seed sources had increased water use efficiency, which could be a result of higher photosynthetic capacity. Similarly, families with greater heat tolerance also had higher water use efficiency and slower growth, suggesting a conservative growth strategy. Intervarietal hybrids showed increased heat tolerance (lower electrolyte leakage at 50 °C and 55 °C) and higher water use efficiency compared to coastal families, suggesting hybridization might be a source of pre-adapted alleles to warming climates and should be considered for large-scale reforestation projects under increasingly arid conditions.

Keywords: Water use efficiency, carbon isotope discrimination, heat tolerance, electrolytic leakage, Douglas-fir, GWAS, hybridization

INTRODUCTION

High temperatures are an emerging, climate-driven stress that can cause numerous biochemical, physiological, and morpho-anatomical changes in plants leading to stunted growth and development (Kotak *et al.* 2007; Wahid *et al.* 2007). Drought, caused by low soil water content and/or high evaporative demand, often accompanies high temperature and is one of the most important climatic stressors affecting growth, performance, and survival of plant species (McDowell *et al.* 2008; Estravis-Barcala *et al.* 2020). When working in tandem, drought and heat stress have the potential to cause permanent physiological harm to plants (Shah and Paulsen 2003; Wang and Huang 2004). Climate change is predicted to cause more extreme temperatures and alter precipitation regimes, causing an increase in the severity and duration of heat stress and drought leading to reduced growth and increased tree mortality (Loarie *et al.* 2009; Trumbore *et al.* 2015). While coniferous gymnosperms are the dominant trees of global arid forests and are generally considered more heat and drought resistant than angiosperms, investigation of the genetics of heat and drought stress in gymnosperms has focused on only a few species and lags studies in angiosperms, especially crops (Moran *et al.* 2017). Understanding the genetics and physiology of heat and drought tolerance can help us better predict the effects of climate change on gymnosperms and develop management strategies that target the hardiest genotypes for tree improvement and reforestation.

The physiological and genomic basis of drought tolerance in conifers have been investigated via transcriptomic, genomic and/or common garden studies (reviewed in Moran *et al.* 2017). Common garden studies of commercial conifer species have provided information on the differential response to drought among populations or seed sources (Montwe *et al.* 2015; Bansal *et al.* 2016; Dixit *et al.* 2022). The recent development of sequencing technologies has facilitated investigation of genome-wide drought responses via gene expression analyses, thereby enabling the identification of candidate genes involved in stress responses (Behringer *et al.* 2015; Hess *et al.* 2016; Du *et al.* 2018; De Maria *et al.* 2020; Haas *et al.* 2021). Candidate genes identified by transcriptomic studies or other methods can be associated with specific drought-related traits through genome-wide association studies (Neale and Savolainen 2004; Cumbie *et al.* 2011). However, these methods have received less attention in conifers due to the difficulties in identifying large numbers of genes with limited number of reference genomes and incomplete annotations, and due to the extensive time required to obtain phenotypic data.

Douglas-fir (*Pseudotsuga menziesii*, Pinaceae family) is a long generation, mostly outcrossing, evergreen conifer species of great economic and ecological importance in North America (St. Clair *et al.* 2005). The species is categorized into three varieties based on their geographic location along the Pacific coast (coastal, var. *menziesii*), across the mountainous regions from the Rocky Mountains to Arizona (interior, var. *glauca*), or in the Sierra Madre Mountain ranges of Mexico (Mexican, var. *lindleyana*) (St. Clair *et al.* 2005). The coastal and interior varieties naturally hybridize in British Columbia (Canada), and the Washington Cascades (U.S.). Douglas-fir natural hybrids have been understudied, and their adaptation and stress response are unknown. Douglas-firs can grow under a variety of conditions, but the climate in which they are most found consists of warm, dry summers, and cool, wet winters. Their natural species' distribution range in North America spans from 19°N to 55°N degrees latitude and from sea level to over 3000 m of elevation (Hardin *et al.* 2001; Gould *et al.* 2011). Drought is present throughout a large part of the Douglas-fir's geographic range, and climate change brings warmer temperatures on a global scale; as such, being able to identify which families and varieties of Douglas-fir are the most heat and drought tolerant is of key importance to foresters (Bansal *et al.* 2016).

Morphological, physiological and plastic responses to drought have been widely studied in Douglas-fir (Bansal *et al.* 2015; Bansal *et al.* 2016; Marias *et al.* 2016; Müller *et al.* 2012; Hess *et al.* 2016). In contrast, the genomics of heat tolerance and traits associated with drought tolerance have not been investigated in any variety or hybrids of the species. In addition to evaluating heat tolerance by cellular electrolyte leakage (Ruter 1993; Agarie *et al.* 1995), we measured seedling growth and the trait of carbon isotope discrimination, a time-integrated measure of photosynthetic water use efficiency (Farquhar *et al.* 1982). The lower carbon isotope discrimination of woody plants of dry habitats compared with plants of wet habitats (Diefendorf *et al.* 2010) suggest that water use efficiency is an important drought adaptation. These traits are relevant to conifer adaptation to arid conditions, as trade-offs between growth rate and water use efficiency have been reported in other conifers (Kerr *et al.* 2015; Dixit *et al.* 2022). The objectives of this study were to (a) investigate differences in water use efficiency, cellular heat tolerance and growth among families and varieties of Douglas-fir; and (b) investigate the genomic basis (candidate genes, gene families and pathways) of the measured traits.

MATERIALS AND METHODS

Greenhouse experimental design

A total of 4,025 seeds (35 for each one of the 109 coastal and 6 hybrid families) were selected from parents throughout the species' natural distribution in western Oregon and Washington, ensuring a wide variation of temperature and moisture in the sample collection (Figure 1). In December 2018, seeds were cold stratified for 1 month at 4°C before planting and later sown in SC10 containers (8.25 x 1.5", Ray Leach Cone-tainers-SC10 Super, Stuewe & Sons, Inc., Tangent, OR, USA) containing a soil mixture made up of 1-part sphagnum peat moss, 1-part coarse vermiculite, and 1-part horticultural perlite. Seedling containers were placed in racks (30.5 cm x 37.6 cm each), and groups of racks were on adjacent benches (4.5 m² each) in a completely randomized design in the same greenhouse room at the Northern Arizona University (NAU) greenhouse facility (Flagstaff, Arizona). Seeds were watered three times a week until germination. After germination, seedlings were watered daily until secondary needles appeared, for about 4 weeks. Seedlings were fertilized once weekly during the months of April through July of 2019, using a balanced water-soluble fertilizer at a concentration of 60 ppm. The greenhouse temperatures were maintained between 15-23° C during growing season, from April to September. To follow the natural process of dormancy in the species, we mimicked dormancy by placing trees in a cool greenhouse from October (2019) to March (2020). During this period, the greenhouse temperatures varied between 7-15° C. All seedlings were then transplanted into D-40 pots (0.65 L) in the summer of 2020.

Survival and growth measurements

Germination and seedling survival were evaluated in March 2020. Five coastal families with less than 5% germination rate were excluded for posterior experiments (all other families had an average germination rate of 80% with an average of 22 individuals per family). Heights for all remaining seedlings (2,474 total) growing in the greenhouse were measured in October 2020 when seedlings were 21 months old. The height of the uppermost bud, measured from the stem base, was recorded for each seedling. All measurements were taken in millimeters using a standard meter stick.

Heat stress experiment

Needles were subjected to a range of treatment temperatures to simulate heat stress, upon which conductivity of electrolyte leakage was measured in microsiemens (μS) and compared to the same measurement after 100% damage (Ruter 1993; Agarie *et al.* 1995). This approach is similar to the use of electrolyte leakage to study freeze tolerance (Sutinen *et al.* 1992; Lu *et al.* 2007), but in our study heat is used instead of below-freezing temperatures. A pilot study was conducted to develop the protocol and assess the viability of the experiment. In this pilot study, 2 individuals from each of 3 different families were subjected to 8 different treatment temperatures in increments of 5°C (30°C-65°C) (data not shown). Upon review, the 30°C treatment was excluded from subsequent experiments due its similarity with the 35°C treatment.

The morning of each treatment, 4 needles were picked from 1 individual of each of the 100 selected families. Needles were placed in separate white paper envelopes and labeled accordingly. Each of the 4 needles per individual was then cut into 1cm segments, with both ends having been cut, and placed into a corresponding labeled test tube. Next, 1ml of laboratory-grade deionized water was added to each test tube and capped with a square of aluminum foil to prevent dehydration during treatments. Each treatment group of 100 individuals was then submerged into a Precision GP10 Thermo Scientific water bath (Thermo Fisher Scientific, Waltham, MA, USA) at the desired treatment temperature (35°C, 40°C, 45°C, 50°C, 55°C, 60°C, 65°C) for 60 minutes each. After removing the treatment group(s) from the bath, 4ml of additional deionized water was added to each test tube and the aluminum foil caps were replaced. Each treatment group was then placed on a KJ-201BD Orbital Shaker at 150 RPM for 40 minutes. Following the shaking, each treatment group was placed in a room temperature (25°C) bath for 30 minutes. Using a Fisherbrand conductivity meter (Thermo Fisher Scientific, Waltham, MA, USA), conductivity of the solution in each test tube was measured and recorded. To determine the total number of electrolytes for each sample, each treatment group was placed in a Fisher Scientific SterilElite24 autoclave at 121°C, for the full 50-minute cycle. After one final manual inversion of each tube, final conductivity of the solution in each test tube was then measured and recorded to represent 100% leaf damage. The following formula was used to calculate the percentage of electrolyte leakage for each individual temperature treatment after taking the conductivity after 100% damage into account:

$$\% \text{ electrolytic leakage} = (\text{conductivity at selected Temp} / \text{conductivity at } 121^{\circ}\text{C}) \times 100$$

This process was repeated for 5 individuals from each of the 100 families, in all 7 treatment temperatures. Percent electrolyte leakage was then averaged among the 5 individuals from each family.

Carbon isotope and leaf nitrogen analysis

Carbon isotope discrimination ($\Delta^{13}\text{C}$) and nitrogen concentrations (%N) were measured in newly developed needles in 22-month-old seedlings from 105 families, with a total of 525 individual seedlings (same families used for the heat stress experiment plus 5 additional ones). These traits were measured using elemental analyzer-continuous flow isotope ratio mass spectrometry with an Erba NC 2100 EA (Carlo) interfaced to a Finnigan Delta Plus XL IRMS (Thermoquest, San Jose, Calif.) at the Colorado Plateau Stable Isotope Laboratory located at NAU.

At least 6 needles were collected from 5 individuals per each of the 105 families. Needles were placed into labeled paper coin envelopes and placed in a SHEL LAB Forced Air Oven at 65°C for 48 hours to dry. Needles were then placed in 2mL tubes with 1 grinding ball (4mm) bearing and placed in a SPEX SamplePrep 1600 MiniG mill for 2 minutes. Following an additional 24-hour drying period, ground needles were then rolled into tin capsules (4x6mm) with a target sample weight of 2.00mg +/-0.100mg. Upon receiving the data from the carbon isotope analysis, $\delta^{13}\text{C}$ data was then used to calculate carbon isotope discrimination ($\Delta^{13}\text{C}$) using the following equation (Farquhar *et al.* 1989):

$$\Delta^{13}\text{C} \text{ ‰} = \frac{-0.008 + \left[\frac{\delta^{13}\text{C}}{-1000} \right]}{1 - \left[\frac{\delta^{13}\text{C}}{-1000} \right]} \times 1000$$

Here, $\delta^{13}\text{C}$ is the ratio of the stable carbon isotope in Douglas-fir needle tissues, and -0.008 is the approximate $\delta^{13}\text{C}$ of atmospheric CO_2 established in the Pee Dee Belemnite reference standard.

Statistical analyses of phenotypic traits

A one-way analysis of variance (ANOVA) was used to test for significant differences among families for growth, heat tolerance for each of the seven temperature treatments (35°C, 40°C, 45°C, 50°C, 55°C, 60°C, 65°C), carbon isotope discrimination, and leaf nitrogen. The analysis was repeated three

times, including only coastal, only hybrids, and coastal and hybrid families. We used mixed-model ANOVA to test for differences in measured traits among the two varieties with variety as a fixed effect and family as a random effect using the lme4 package. The residuals of the models approximated a normal distribution. A Tukey's Honestly Significant Difference (Tukey's HSD) post-hoc test was used for pairwise comparisons of all families. Relationships among family trait means were evaluated using correlation and regression analyses. All data analyses were carried out using R studio version 1.1.442 (packages Hmisc and ggcorrplot) and JMP PRO 15.2.

Correlations among geographic, environmental variables and phenotypic traits

Seed source elevation data was obtained using coordinates and a digital elevation model (DEM) in GIS. Climate data was obtained using ClimateNA software that downscales PRISM monthly climate data from 1962-1990 to scale-free point data, allowing for a more accurate representation of maternal tree climate variables (Wang *et al.* 2016). From the available data, 23 annual climate variables and 3 geographic variables were obtained and included in the final dataset alongside the phenotypic traits that were measured in the greenhouse and the lab (Supporting Information Tables S1 and S2). Correlations among geographic, environmental, and phenotypic variables were analyzed in R (version 4.0.4).

DNA extraction and genotyping

Douglas-fir seeds were soaked in a 7:3 solution of water and 3% hydrogen peroxide for 12 hours. Megagametophyte haploid tissues were dissected from ten half-sib individuals for each family and pooled together to infer the maternal genotype. DNA was extracted using the Qiagen DNeasy mini-prep Plant kit and an Eppendorf automated pipetting workstation, with a lab protocol that included one day of tissue lysis and incubation at 96°C, followed by several steps of precipitation and filtering. The quality and concentration of DNA were assessed using a NanoDrop Spectrophotometer and a Qubit 2.0 Fluorometer. Genotyping for all samples was done with 20,397 Single Nucleotide Polymorphism (SNPs) markers obtained from a custom-designed gene-based Illumina Infinium SNP array (Weiss *et al.* 2020; De La Torre *et al.* 2021). A total of 14,980 SNPs was retained for posterior analyses. This array was developed to represent genome-wide variation by containing both coding and non-coding regions of the genome. To assure species-wide levels of genetic diversity, whole-genome resequencing data from individuals across the species' geographic range were used as input for SNP array construction (Weiss *et al.* 2020). GenomeStudio Genotyping Module v.2.0 (Illumina

2016) was used to call genotypes, filter, and generate genotyping statistics for all samples and SNPs. SNPs with a call frequency ≤ 0.65 and a call rate ≤ 0.8 were not included. Further, markers were filtered out based on minor allele frequency (>0.01) to remove all monomorphic and low-quality SNPs. SNP functional annotations were obtained from aligning against the full NCBI non-redundant protein sequences database (nr) using BLASTP (e value $< e^{-10}$), and by using the Douglas-fir's version Psme.1.0 genome annotations (https://treegenesdb.org/FTP/Genomes/Psme/v1.0/annotation/Psme.1_0.entap_results.tsv).

Population Structure

Population structure in the dataset was evaluated using a Principal Component Analysis (PCA) in TASSEL v.5 (Bradbury *et al.* 2007). In addition, the Python 2.X FastSTRUCTURE (Raj *et al.* 2014) algorithm was used for posterior inference of the number of clusters (K) that better explain the population genetic structure of the dataset. Models in fastSTRUCTURE were replicated ten times for each K value from 1 to 10 with default prior parameters and random seed numbers for each run. All runs were combined using CLUMPP (Jakobsson *et al.* 2007) and ancestry barplots were generated in R.

Univariate and multivariate genome-wide association studies

General and mixed linear models (GLM and MLM) were implemented in the univariate GWAS analyses in TASSEL v.5 (Bradbury *et al.* 2007). The first two principal components of a PCA were used as co-variables to control for population structure in the GLM model, and a kinship matrix was used to account for relatedness in the MLM model (Bradbury *et al.* 2007). Associations between SNP markers and heat, water use efficiency, and growth-related traits were tested using univariate linear mixed models (uLMM) and multivariate linear mixed models (mvLMM) in GEMMA v0.98.3 (Zhou *et al.* 2013). In contrast to the uLMM method, mvLMM associates multiple phenotypic traits with all markers simultaneously, while controlling for population structure and relatedness. To run GEMMA, PLINK binary ped format was generated using PLINK v.1.9 software for association analysis. Bonferroni and false discovery rate (FDR) (<0.05) were applied for correction for multiple testing to identify significant SNPs.

Functional gene annotations

Genomic positions of significant SNPs were investigated to identify the annotated genes by scanning the genomic VCF files of Douglas-fir. Subsequently, the identified significant SNPs were annotated using annotation files downloaded from TreeGenes (<https://treegenesdb.org/TripalContactProfile/588450>). The annotation was confirmed using some other approaches such as pfam (Finn *et al.* 2014), blastp (Johnson *et al.* 2008), and BlastKOALA (Kanehisa *et al.* 2016b). Pfam was run using HMMER (Finn *et al.* 2011) at default parameters with e-value 1.0 to search proteins families. The blastp was run at expected threshold-0.05; matrix-BLOSUM 62; database- non-redundant protein sequence (nr) to search the similar hits. The BlastKOALA at KEGG (Kanehisa *et al.* 2016a) was performed for protein pathways and annotations. Identical matching genes were chosen to identify annotations and KEGG pathways.

RESULTS

Growth measurements

A total of 2,474 seedlings were evaluated for height, and significant differences were found among families and varieties. Family-averaged height (mm) varied from 61.3 (family 8049) to 188.75 (family 3027). Height measurements significantly differed among the coastal ($F = 14.5$, $p\text{-value} = <0.0001$) and hybrid Douglas-fir families ($F = 25.73$, $p\text{-value} = <0.0001$; Table 1); with coastal families growing faster and taller (128.11 ± 29.17) than families from the hybrid variety (86.41 ± 27.93) ($F=30.211$, $p\text{-value} < 0.0001$; Table 1). No significant correlations were found between seedling height and geographic variables (latitude, longitude) over all seedlings.

Heat stress experiment

The evaluation of Douglas-fir families suggested significant genetic variation in heat tolerance at different temperatures across geographical origin of coastal and hybrid Douglas-fir families (Figures 2 and 3; Table 1). Significant differences in electrolyte leakage were found among coastal families at temperatures 35°C, 40°C, 45°C and 50°C ($p\text{-value} < 0.05$, Table 1). In hybrid families, differences in electrolyte leakage were only significant at 40°C ($F = 3.497$, $p\text{-value} = 0.0401$, Table 1). Across all families, extensive damage of Douglas-fir foliage occurred between 50°C and 60°C (Figure 2). Hybrid

families had significantly lower electrolyte leakage than coastal families at 55°C (p-value <0.05; Table 1; Figure 4).

All correlations among environmental variables and between environmental variables and traits can be found in Figure 4 and Supporting Information Figures S1 and S2. On average, families from higher elevation, more inland locations with drier and colder weathers had less heat induced electrolyte leakage than families from lower elevation, coastal locations with warmer and more humid climates.

Carbon isotope and leaf nitrogen analyses

Significant differences in $\Delta^{13}\text{C}$ ($F = 2.691$, p-value <0.0001), %N ($F = 4.53$, p-value <0.0001), and C:N ($F = 5.161$, p-value = <0.0001) were found among coastal Douglas-fir families (Table 1). Hybrid families also had significant differences in all above traits (p-value <0.05, Table 1). Family-averaged $\Delta^{13}\text{C}$ varied from 22.94 ‰ (family 6089) to 25.74 ‰ (family 5097) (mean = 24.47 ± 0.84) among coastal Douglas-fir families. $\Delta^{13}\text{C}$ was positively correlated with electrolytic leakage across all temperatures at which extensive cell damage occurs (Figure 2). In addition, elevation was negatively correlated with $\Delta^{13}\text{C}$ and C:N (p-values of 0.0037, 0.0032 respectively) and positively correlated with %N ($r=0.3$, p-values = 0.002). $\Delta^{13}\text{C}$, %N, and C:N ratio were significantly different between coastal and hybrid families (p-values <0.05; Table 1).

Across all families, height was positively correlated with electrolyte leakage at 55°C with an r of 0.36 and a p-value of 0.000197 (Figure 2), and with electrolyte leakage at 60°C ($r=0.25$, p-value = 0.0133; Figure 2). In addition, seedling height was negatively correlated with %N and positively correlated with C:N (Figure 2).

Genome-wide association studies

A principal component analysis of Douglas-fir families suggested a low population structure in the dataset with the coastal individuals divided between one main cluster and one smaller cluster (Oregon's southern distribution); and the hybrid individuals located in eastside Washington's Cascades (Supporting Information Figure S3). The fastSTRUCTURE ancestry analysis suggested the

mother trees of these hybrid families have mixed ancestry from the coastal and interior varieties (Supporting Information Figure S4). Univariate and multivariate GWAS analyses were repeated twice, the first time for all dataset (coastal and hybrids), and a second time only for coastal families. Population structure in the datasets was accounted for in the GWAS analyses to avoid the presence of false positives. Univariate and multivariate GWAS investigated the presence of significant associations among 11 heat, drought, and growth-related traits and 14,980 genetic markers. Results of the univariate and multivariate GWAS of coastal and hybrids identified 35 significant associations between 6 traits (electrolytic leakage at 35°C, 40°C, and 45°C and 50°C; C:N ratio and height) and 31 genes (Table 2 and Supporting Information Table S3). Most genes were associated with only one trait, except for transcription factor VOZ1 PSME_15499 gene, which was associated with electrolytic leakage at 35°C and 50°C; and uncharacterized gene PSME_42829, which was associated with electrolytic leakage at 40°C and 45°C. The proportion of phenotypic variance explained by each significant SNP marker (based on the GLM model) varied from 22 to 26.4% (Supporting Information Table S3).

The GLM analyses of coastal families identified 12 significant SNPs ($p\text{-value} < 3.00 \times 10^{-6}$), 10 of them were associated with electrolytic leakage at 35°C, and 2 with C:N ratio. Those significant SNPs matched 7 genes in 7 different scaffolds in the genome of Douglas-fir (Supporting Information Table S4; Table 3). The proportion of phenotypic variance explained by each significant SNP marker varied from 21.7 to 27.3% (Supporting Information Table S4). Subsequently, univariate linear mixed model (uLMM) and multivariate linear mixed model (mvLMM) approaches were performed in GEMMA to identify significant SNPs. MvLMM identified 10 significant SNPs ($p\text{-value} < 3.00 \times 10^{-6}$) (Supporting Information Table S5); and uLMM identified 3 SNPs ($p\text{-value} < 3.00 \times 10^{-6}$) associated with electrolytic leakage at 35°C, 40°C, and 45°C (Supporting Information Table S6). The GWAS results from both datasets (coastal plus hybrids, and only coastal) identified 7 common genes, identified as F-box proteins, reticulon-like protein, inactive purple acid phosphatase and a gene involved in the inositol phosphate metabolic pathway (Tables 2 and 3). Other genes were involved in the biosynthesis of primary and secondary metabolites, Ubiquitin mediated proteolysis, transport, transcription, and ribosome biogenesis (Tables 2 and 3). Ten of the genes associated with height in this study (Table 2) were found to be associated with growth, cold hardiness, phenology and environmental variables (continentality and/or days below zero degrees) in a previous study in Douglas-fir (De La Torre et al. 2021). Seeds from some of the mother trees in the 2021 study were grown and analyzed in the present study.

DISCUSSION

Differences in heat tolerance among Douglas-fir families and varieties

Our heat stress experiment included foliar temperatures (50 to 55 °C) similar to those reported in forests of Douglas-fir in the Pacific Northwest region during the unprecedented “heat dome” event in summer of 2021 (Philip *et al.* 2021). Thus, our results are relevant to emerging extreme heat events that are expected to increase in the future with climate warming. Exposure to high temperatures caused substantial tissue damage across all families. We found differences among families at moderate exposure (40 °C and 45 °C) but not at temperatures above 55°C, suggesting that exposure to such extreme temperatures causes irreversible damage across all families due to denaturing of proteins. At such high temperatures, seedling survival might depend on maintaining high transpiration rates for heat dissipation (Kolb and Robberecht 1996). However, this might be challenging under more frequent and intense droughts predicted with climate change. Differences in electrolyte leakage at moderate temperatures among Douglas-fir families were attributed to an adaptation to source climate in an earlier study (Marias *et al.* 2016). Lower electrolyte leakage in hybrids at 55°C compared with the coastal variety suggest a higher tolerance to these temperatures for hybrids and should be explored further to test if results of electrolyte leakage translate to enhanced performance under field conditions. In addition, future studies should test electrolyte leakage at different temperatures under high and low water conditions to determine tree response to an interaction of heat and water stress.

Higher elevation Douglas-fir families have higher water use efficiency

Carbon isotope discrimination significantly varied among coastal families, among hybrid families, and between coastal and hybrid varieties. This result is consistent with earlier studies that found differences among Douglas-fir varieties in carbon isotope discrimination (Zhang *et al.* 1993; Anekonda *et al.* 2004). Our results suggest an increase in water use efficiency and photosynthetic capacity with an increase in seed source elevation. Lower carbon isotope discrimination (higher water use efficiency) in families from higher elevations could be a result of higher photosynthetic capacity, as suggested by a significant positive correlation between source elevation and %N. Our results of a decrease in carbon isotope discrimination and an increase in foliar nitrogen with elevation are consistent with an earlier study on riparian tree species in southern Utah (Sparks and Ehleringer 1996). Future studies should focus on direct measurements of leaf level gas exchange to

gather information about differences in net carbon assimilation and stomatal conductance among Douglas-fir families.

Correlations between heat stress, water use efficiency, and growth

Interestingly, families with higher water use efficiency also had higher tolerance to heat stress. We found a positive correlation between electrolyte leakage after exposure to high temperatures and carbon isotope discrimination (under well-watered conditions) suggesting that families with high tolerance to extreme temperatures were also better able to control water loss by stomatal closure. In addition, faster growing families were more susceptible to heat damage as evidenced by a positive correlation between tree height and electrolyte leakage after exposure to temperatures between 55 to 65°C (Figure 2); this suggests a trade-off between growth and heat tolerance. Families with the highest electrolyte leakage under extreme heat stress were from locations with a higher number of frost-free days, less precipitation as snow, fewer days below 0°C, and warmer average temperatures, both annually and during the coldest and warmest months (Supporting Information Figure S2). Our results suggest that slow-growing families from colder climates are more tolerant to heat stress than fast-growing families from warmer climates. Previous research shows trade-offs between tree growth and cold hardiness in Douglas-fir, however trade-offs between tree growth and heat/drought resistance have not been clear (Darychuk *et al.* 2012). We used tree height to represent overall growth, however radial growth is also an important metric of growth which is responsive to drought (Sergent *et al.* 2014). Overall, the results of our study indicate that intervarietal hybrid families with slower growth and higher water use efficiency had greater tolerance of extreme heat, which suggests they might be better adapted to future water conditions than the coastal variety given that high temperatures are an emerging environmental stress and threat to global forests (Marias *et al.* 2016, Phillip *et al.* 2021).

Previous studies of Douglas-fir found a trade-off between drought tolerance and growth (St. Clair and Howe 2007) and cold hardiness and growth (De La Torre *et al.* 2021). Our results suggest that fast-growing families are less tolerant to heat stress as well (Figure 2), so it appears that all three stress-tolerance traits (heat, drought, and cold tolerance) are inversely correlated to height in Douglas-fir. In a previous greenhouse study by Bansal *et al.* (2016), cold hardiness and drought tolerance were found to be linked in coastal Douglas-fir. Trees from regions with cold winters had higher cold and drought tolerance likely due to the overlapping adaptations and their ability to deal

with winter desiccation (Bansal *et al.* 2016). The results of our experiment support this finding. In addition, our results indicate that the colder the seed source environment (whether it be in terms of the number of frost-free days or mean coldest month temperature), the more resistant foliar cells are to extreme heat, which indicates greater heat tolerance (Supporting information Figure S2).

Role of hybridization in heat tolerance and water use efficiency in Douglas-fir

Our results indicate that hybrid families had higher heat tolerance and water use efficiency compared to coastal families (Figure 4). As inferred by the proportions of interior and coastal ancestries in the mother trees, seedlings in our study are not first-generation hybrids but most likely are the product of several generations of hybridization and introgression among the interior and coastal varieties (Supporting Information Figure S4). Strong differences in heat tolerance and water use efficiency were evident in families in which the mother trees had a higher ancestry from the interior variety (all hybrid families except for family 5029 which had a higher coastal ancestry, Supporting Information Figure S4). Therefore, we suggest that adaptive introgression from the interior variety might have resulted in natural intervarietal hybrids with higher water use efficiency and tolerance to heat. However, conclusive evidence requires further drought and heat tolerance studies including all species' varieties (coastal, interior, Mexican) and hybrid classes (F1s, F2s, backcrosses). It is also important to consider that our study was restricted to the Washington Cascades, but natural hybrids are also reported in British Columbia. Therefore, our results might not represent all hybrid individuals naturally growing across the species' geographic range. Indirect evidence from a common garden study growing the same families in Oregon (St. Clair *et al.* 2005) suggests these hybrids might also have higher cold hardiness. Examples of adaptive introgression between widely distributed, mostly outcrossing, closely related, forest tree species or varieties have been reported in *Populus tremula* (Rendon-Anaya *et al.* 2021), *Quercus robur* x *Q. petraea* (Leroy *et al.* 2020), *Populus balsamifera* x *P. trichocarpa* (Suarez-Gonzalez *et al.* 2016), *Picea glauca* x *P. engelmannii* (De La Torre *et al.* 2014a; De La Torre *et al.* 2014b), among others. The results of our study suggest hybridization might be a source of pre-adapted alleles to warming climatic conditions and should be considered for planting under increasingly arid conditions.

Functional annotation of candidate genes

Trees respond to abiotic stress in numerous ways, including physiological, genetic, cellular, and morphological changes (Kijowska-Oberc *et al.* 2020). In this study, we identified 32 candidate genes associated with water use efficiency, heat tolerance, and growth. These genes were involved in

many important biological processes such as primary and secondary metabolism (inositol phosphate metabolism, ubiquinone and other terpenoid-quinone biosynthesis, pyrimidine metabolism, alanine, aspartate and glutamate metabolism); abiotic, biotic stress, and signaling; DNA replication and repair; transport; among other processes (Tables 2 and 3).

From the genes associated with heat stress, we identified PSME_48755 and PSME_04475, F-box/kelch-repeat candidate genes associated with heat tolerance at 35°C. F-box proteins play critical roles in plant responses to biotic/abiotic stresses and plant developmental process (Baute *et al.* 2017; Wei *et al.* 2021). F-box gene transcripts in *Arabidopsis* highly accumulated in roots and altered their response under drought stress conditions. In addition, F-box mutant plants displayed better growth under drought stress conditions compared to the wild type with a reduced accumulation of H₂O₂ and malondialdehyde (MDA) (Rao and Virupapuram 2021). F-box proteins (FBX92) had been linked to reduced leaf growth, since overexpression of *AtFBX92* resulted in smaller leaves than the wild type in *Arabidopsis thaliana* (Baute *et al.* 2017).

Our results also identified an inositol-tetrakisphosphate 1-kinase 3-like candidate gene (PSME_39629) associated with electrolytic leakage at 35°C. Inositol signaling plays a crucial role in various aspects of plant growth and adaptation (Jia *et al.* 2019). Inositol phosphate kinases (IPK1) play important roles in diverse cellular processes by functioning as structural and functional cofactors, regulators, and second messengers (Shears *et al.* 2012). In the study of Zhang *et al.* (2020), transgenic expression of *ThIPK2* (an inositol polyphosphate kinase gene) in wheat led to improved drought tolerance. Compared to the wild type (WT) plants, the authors identified that transgenic plants showed higher seed germination rates, better developed root systems, a higher relative water content (RWC), total soluble sugar content, and less cell membrane damage under drought stress conditions (Zhang *et al.* 2020).

Candidate gene PSME_47800, identified as inactive purple acid phosphatase (PAPs) protein was associated with electrolytic leakage at 35°C and height. In a previous study by our group, we found that this gene was associated with growth, phenology, degree days below 0°C and continentality in Douglas-fir (De La Torre *et al.* 2021). PAPs proteins play important roles in phosphate (Pi) acquisition, utilization, and developmental processes. It has been reported that differential regulation of CaPAPs under different nutrient deficiencies revealed their roles under multiple nutrient stresses, including

Pi deficiency. Most of the CaPAPs have prominently expressed in flower and seed development in Chickpea (Bhadouria *et al.* 2017). Finally, Transcription factor VOZ1 candidate gene PSME_15499, associated with heat tolerance in our study, was also associated with heat, drought, and salt stress in *Arabidopsis* (Koguchi *et al.* 2017; Song *et al.* 2018).

CONCLUSIONS

Warmer climates bring significant challenges to the survival of Douglas fir populations. While coastal varieties are widely planted due to their fast growth, their low water use efficiency (a trait of mesic-adapted plants) and low tolerance to heat might predispose them to maladaptation to future climate conditions, as suggested by trade-offs between growth, water use efficiency and heat tolerance found in this study. Our study indicates significant genetic variation in water use efficiency (inferred from carbon isotope discrimination), photosynthetic capacity (inferred from %N), growth, and tolerance of leaves to heat stress among Douglas-fir families and varieties. High elevation families had higher water-use efficiency and photosynthetic capacity than low elevation families. In addition, families with greater heat tolerance had slower growth and higher water-use efficiency. Intervarietal hybrids, with mixed ancestry from coastal and interior varieties, had higher water use efficiency and higher heat tolerance than coastal families, suggesting hybridization might be a source of pre-adapted alleles to extreme heat events and drought in increasingly warming climate conditions. Due to the long-generation nature of the species, these results should be considered for reforestation and the selection of seeds sources for commercial plantations under arid conditions.

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AUTHOR CONTRIBUTIONS

A.D.L.T, S.C, A.D and T.K designed the research; A.D.L.T and S.C perform and supervised lab work; S.C and C.S performed all physiological tests; A.D.L.T, S.C, A.D, M.K, and C.S performed all genomic and physiological data analyses; A.D.L.T, S.C, M.K, A.D, T.K, and C.S wrote the manuscript; all authors reviewed and approved the final manuscript.

DATA AVAILABILITY

The data available for this study is included in the Supporting Information section.

REFERENCES

- Agarie S, Hanaoka N, Kubota F, Agata W, Kaufman PB. 1995. Measurement of cell membrane stability evaluated by electrolyte leakage as a drought and heat tolerance test in rice (*Oryza sativa* L.). *J. Fat. Agr., Kyushu Univ* 40(1): 233-240. doi: 10.5109/2410
- Anekonda T, Jones C, Smith BN, Hansen LD. 2004. Differences in physiology and growth between coastal and inland varieties of Douglas fir seedlings in a common garden. *Thermochimica Acta* 422(1-2):75-79.
- Bansal S, Harrington CA, St. Clair JB. 2016. Tolerance to multiple climate stressors: a case study of Douglas-fir drought and cold hardiness. *Ecology and Evolution* 6(7): 2074-2083.
- Bansal S, Harrington CA, Gould PJ, St. Clair JB. 2015. Climate-related genetic variation in drought-resistance of Douglas-fir (*Pseudotsuga menziesii*). *Global Change Biology* 21:947-958.
- Baute J, Polyn S, De Block J, Blomme J, Van Lijsebettens M, and Inze, D. 2017. F-Box Protein FBX92 Affects Leaf Size in *Arabidopsis thaliana*. *Plant Cell Physiol*, 58: 962-975.
- Behringer D, Zimmermann H, Ziegenhagen B, Liepelt S. 2015. Differential Gene Expression Reveals Candidate Genes for Drought Stress Response in *Abies alba* (Pinaceae). *PLOS ONE* 10(4). <https://doi.org/10.1371/journal.pone.0124564>
- Bhadouria J, Singh AP, Mehra P, Verma L, Srivastawa R, Parida S K, Giri J. 2017. Identification of purple acid phosphatases in chickpea and potential roles of *CaPAP7* in seed phytate accumulation. *Sci Rep* 7: 11012.
- Bradbury PJ, Zhang Z, Kroon DE, Casstevens TM, Ramdoss Y, Buckler, ES. 2007. TASSEL: software for association mapping of complex traits in diverse samples. *Bioinformatics* 23:2633–2635.
- Brodribb TJ, McAdam SAM, Jordan GJ, Martins SCV. 2014. Conifer species adapt to low-rainfall climates by following one of two divergent pathways. *Proceedings of the National Academy of Sciences* 111:14489–14493.

Cornejo-Oviedo EH, Voelker SL, Mainwaring DB, Maguire DA, Meinzer FC, Brooks JR. 2017. Basal area growth, carbon isotope discrimination, and intrinsic water use efficiency after fertilization of Douglas-fir in the Oregon Coast range. *Forest Ecology and Management* 389: 285–295.

Cumbie W, Eckert A, Wegrzyn J. *et al.* 2011. Association genetics of carbon isotope discrimination, height and foliar nitrogen in a natural population of *Pinus taeda* L. *Heredity* 107: 105–114. <https://doi.org/10.1038/hdy.2010.16>

Darychuk N, Hawkins BJ, Stoehr M. 2012. Trade-offs between growth and cold and drought hardiness in subarctic Douglas-fir. *Canadian Journal of Forest Research* 42(8): 1530-1541.

De La Torre AR, Roberts DR, Aitken SN. 2014a. Genome-wide admixture and ecological niche modelling reveal the maintenance of species boundaries despite long history of interspecific gene flow. *Molecular Ecology* 23(8), 2046–2059. <https://doi.org/10.1111/mec.12710>

De La Torre AR, Wang T, Jaquish B, Aitken SN. 2014b. Adaptation and exogenous selection in the *Picea glauca* x *P. engelmannii* hybrid zone and its implications for forest management under climate change. *New Phytologist* 201(2): 687-699.

De La Torre AR, Wilhite B, Puiu D, St. Clair JB, Crepeau MW, Salzberg SL, Langley CH, Allen B, Neale DB. 2021. Dissecting the polygenic basis of cold adaptation using genome-wide association of traits and environmental data in Douglas-fir. *Genes* 12(1): 110.

De María N, Guevara MÁ, Perdiguero P, *et al.* 2020. Molecular study of drought response in the Mediterranean conifer *Pinus pinaster* Ait.: Differential transcriptomic profiling reveals constitutive water deficit-independent drought tolerance mechanisms. *Ecol Evol* 10:9788–9807. <https://doi.org/10.1002/ece3.6613>

Diefendorf AF, Mueller KE, Wing SL, Koch PL, Freeman KH. 2010. Global patterns in leaf $\delta^{13}\text{C}$ discrimination and implications for studies of past and future climate. *Proceedings of the National Academy of Sciences* 107 (13): 5738-5743.

Dixit A, Kolb TE. 2020. Variation in seedling budburst phenology and structural traits among southwestern ponderosa pine provenances. *Can. J. For. Res.* 50(9):dx.doi.org/10.1139/cjfr-2019-0333

Dixit A, Kolb T, Burney O, Mock K, Grady K. 2021. Provenance variation in early survival, growth, and carbon isotope discrimination of southwestern ponderosa pine growing in three common gardens across an elevational gradient. *Forests* 12(11):1561.

Dixit A, Kolb T, Burney O. 2022. Trade-off between growth rate and water use efficiency in southwestern ponderosa pine provenances. *Forest Ecology and Management* 515:120239.

Du M, Guijie D, Qiong C. 2018. The Transcriptomic Responses of *Pinus massoniana* to Drought Stress. *Forests* 6:326. <https://doi.org/10.3390/f9060326>

Estravis-Barcala M, Mattera MG, Soliani C, Bellora N, Opgenoorth L, Heer K, Arana MV. 2020. Molecular bases of responses to abiotic stress in trees. *Journal of Experimental Botany* 71(13): 3765–3779, <https://doi.org/10.1093/jxb/erz532>

Farquhar GD, O'Leary MH, Berry JA. 1982. On the relationship between carbon isotope discrimination and the intercellular carbon dioxide concentration in leaves. *Functional Plant Biology* 9(2):121-137.

Farquhar GD, Ehleringer JR, Hubick KT. 1989. Carbon isotope discrimination and photosynthesis. *Annual review of plant biology* 40(1): 503-537.

Finn RD, Bateman A, Clements J, Coghill P, Eberhardt RY, Eddy SR, Heger A, Hetherington K, Holm L, Mistry J, Sonnhammer EL, Tate J, Punta M. 2014. Pfam: the protein families database. *Nucleic Acids Res*, 42 (Database issue), D222-30, doi: 10.1093/nar/gkt1223.

Finn RD, Clements J, Eddy SR. 2011. HMMER web server: interactive sequence similarity searching. *Nucleic Acids Res*, 39 (Web Server issue), W29-37, doi: 10.1093/nar/gkr367.

Gould PJ, Harrington CA, St. Clair JB. 2011. Incorporating genetic variation into a model of Budburst phenology of Coast Douglas-fir (*Pseudotsuga menziesii* var. *menziesii*). *Canadian Journal of Forest Research* 41(1):139–150. <https://doi.org/10.1139/x10-191>

Haas JC, Vergara A, Serrano AR, Mishra S, Hurry V, Street NR. 2021. Candidate regulators and target genes of drought stress in needles and roots of Norway spruce. *Tree Physiology* 41(7): 1230–1246, <https://doi.org/10.1093/treephys/tpaa178>

Hess M, Wildhagen H, Junker LV, Ensminger I. 2016. Transcriptome responses to temperature, water availability and photoperiod are conserved among mature trees of two divergent Douglas-fir provenances from a coastal and an interior habitat. *BMC Genomics* 17(1). <https://doi.org/10.1186/s12864-016-3022-6>

Illumina 2016. GenomeStudio Genotyping Module v2.0 Software guide. Document 11319113 v01. Illumina, California. https://support.illumina.com/content/dam/illumina-support/documents/documentation/software_documentation/genomestudio/genomestudio-2-0/genomestudio-genotyping-module-v2-user-guide-11319113-01.pdf

Jakobsson M, Rosenberg NA. 2007. CLUMPP: A cluster matching and permutation program for dealing with label switching and multimodality in analysis of population structure. *Bioinformatics* 23:1801–1806.

Jia Q, Kong D, Li Q, Sun S, Song J, Zhu Y, Liang K, Ke Q, Lin W, and Huang J. 2019. The function of inositol phosphatases in plant tolerance to abiotic stress. *Int J Mol Sci* 20: 3999.

Johnson M, Zaretskaya I, Raytselis Y, Merezuk Y, McGinnis S, Madden TL. 2008. NCBI BLAST: a better web interface. *Nucleic Acids Res*, 36 (Web Server issue), W5-9, doi: 10.1093/nar/gkn201.

Kanehisa M, Sato Y, Morishima K. 2016a. BlastKOALA and GhostKOALA: KEGG tools for functional characterization of genome and metagenome sequences. *J Mol Biol*, 428, (4):726-731, doi: 10.1016/j.jmb.2015.11.006.

Kanehisa M, Sato Y, Kawashima M, Furumichi M, Tanabe M. 2016b. KEGG as a reference resource for gene and protein annotation. *Nucleic Acids Research*, 44 (D1), D457-D462, doi: 10.1093/nar/gkv1070.

Kerr KL, Meinzer FC, McCulloh KA, Woodruff D, Marias D. 2015. Expression of functional traits during seedling establishment in two populations of *Pinus ponderosa* from contrasting climates. *Tree Physiology* 35(5): 535–548. <https://doi.org/10.1093/treephys/tpv034>

Kijowska-Oberc J, Staszak AM, Kamiński J, Ratajczak E. 2020. Adaptation of forest trees to rapidly changing climate. *Forests* 11(2):123. <https://doi.org/10.3390/f11020123>

Koguchi M, Yamasaki K, Hirano T, Sato MH. 2017. Vascular plant one-zinc-finger protein 2 is localized both to the nucleus and stress granules under heat stress in *Arabidopsis*. *Plant Signal Behav* 12: e1295907.

Kolb PF, Robberecht R. 1996. High temperature and drought stress effects on survival of *Pinus ponderosa* seedlings, *Tree Physiology* 16(8): 665–672, <https://doi.org/10.1093/treephys/16.8.665>

Kotak S, Larkindale J, Lee U, Von Koskull-Döring P, Vierling E, & Scharf K-D. 2007. Complexity of the heat stress response in plants. *Current Opinion in Plant Biology* 10(3):310–316. <https://doi.org/10.1016/j.pbi.2007.04.011>

Loarie S, Duffy P, Hamilton H. *et al.* 2009. The velocity of climate change. *Nature* 462, 1052–1055. <https://doi.org/10.1038/nature08649>

Leroy T, Louvet JM, Lalanne C, Le Provost G, Labadie K, Aury JM, Delzon S, Plomion C, Kremer A. 2020. Adaptive introgression as a driver of local adaptation to climate in European white oaks. *New Phytologist* 226:1171– 1182.

Lu P, Colombo SJ, Sinclair RW. 2007. Cold hardiness of interspecific hybrids between *Pinus strobus* and *P. wallichiana* measured by post-freezing needle electrolyte leakage. *Tree physiology* 27(2):243-250.

Marias DE, Meinzer FC, Woodruff, DR, McCulloh, KA. 2016. Thermotolerance and heat stress responses of Douglas-fir and ponderosa pine seedling populations from contrasting climates. *Tree Physiology* 37(3):301–315. <https://doi.org/10.1093/treephys/tpw117>

McDowell NG, Pockman WT, Allen CD, Breshears DD, Cobb N, Kolb TE, Plaut J, Sperry J, West A, Williams DG, Yezzer EA. 2008. Mechanisms of plant survival and mortality during drought: why do some plants survive while others succumb to drought?. *New phytologist* 178(4): 719-739.

Montwe D, Spiecker H, Hamann. 2015. Five decades of growth in a genetic field trial of Douglas-fir reveal trade-offs between productivity and drought tolerance. *Tree Genetics & Genomes* 11:29.

Moran E, Lauder J, Musser C, Stathos A, Shu M. 2017. The genetics of drought tolerance in conifers. *The New Phytologist*. 10.1111/nph.14774

Müller T, Ensminger I, Schmid KJ. 2012. A catalogue of putative unique transcripts from Douglas-fir (*Pseudotsuga menziesii*) based on 454 transcriptome sequencing of genetically diverse, drought stressed seedlings. *BMC Genomics* 13(673). <https://doi.org/10.1186/1471-2164-13-673>

Neale DB, Savolainen O. 2004. Association genetics of complex traits in conifers. *Trends in Plant Science* 9(7):325-330.

Philip SY, Kew SF, van Oldenborgh GJ, Anslow FS, Seneviratne SI, Vautard R, et al (2021). Rapid attribution analysis of the extraordinary heatwave on the Pacific Coast of the US and Canada June 2021. *Earth System Dynamics Discussions*, 1-34.

Raj A, Stephens M, Pritchard JK. 2014. fastSTRUCTURE: Variational Inference of Population Structure in Large SNP Data Sets. *Genetics* 197: 573–589.

Rao V, Virupapuram V. 2021. Arabidopsis F-box protein At1g08710 interacts with transcriptional protein ADA2b and imparts drought stress tolerance by negatively regulating seedling growth. *Biochem Biophys Res Commun* 536: 45-51.

Rendón-Anaya M, Wilson J, Sveinsson S, Fedorkov A, Cottrell J, Bailey MES, Runģis D, Lexer C, Jansson S, Robinson KM, Street NR, Ingvarsson PK. 2021. Adaptive Introgression Facilitates Adaptation to High Latitudes in European Aspen (*Populus tremula* L.). *Molecular Biology and Evolution* 38(11): 5034–5050. <https://doi.org/10.1093/molbev/msab229>

Ruter JM. 1993. High-temperature-induced electrolyte leakage from excised leaves and roots of three hollies. *Hortscience* 28(9):927-928.

Sergent AS, Bréda N, Sanchez L, Bastein JC, Rozenberg P. 2014. Coastal and interior Douglas-fir provenances differ in growth performance and response to drought episodes at adult age. *Annals of Forest Science* 71(6):709–720. <https://doi.org/10.1007/s13595-014-0393-1>

Shah N, Paulsen G. 2003. Interaction of drought and high temperature on photosynthesis and grain-filling of wheat. *Plant and Soil* 257:219–226.

Shears SB, Ganapathi SB, Gokhale NA, Schenk TM, Wang H, Weaver JD, Zaremba A, and Zhou Y. 2012. Defining signal transduction by inositol phosphates. *Subcell Biochem* 59: 389-412

Song XP, Hansen MC, Stehman SV, Potapov PV, Tyukavina A, Vermote EF, Townshend JR: Global land change from 1982 to 2016. 2018. *Nature* 560: 639-643.

Sparks, J.P. and Ehleringer, J.R., 1997. Leaf carbon isotope discrimination and nitrogen content for riparian trees along elevational transects. *Oecologia*, 109(3), pp.362-367.

St. Clair JB, Howe GT. 2007. Genetic maladaptation of coastal Douglas-fir seedlings to future climates. *Global Change Biology* 13(7):1441–1454. <https://doi.org/10.1111/j.1365-2486.2007.01385.x>

St. Clair JB, Mandel NL, Vance-Borland KW. 2005. Genecology of Douglas Fir in Western Oregon and Washington. *Annals of Botany* 96(7):1199–1214. <https://doi.org/10.1093/aob/mci278>

Suarez-Gonzalez A, Hefer CA, Christe C, Corea O, Lexer C, Cronk QCB, Douglas CJ. 2016.

Genomic and functional approaches reveal a case of adaptive introgression from *Populus balsamifera* (balsam poplar) in *P. trichocarpa* (black cottonwood). *Mol Ecol* 25:2427–2442.

Sutinen ML, Palta JP, Reich PB.1992. Seasonal differences in freezing stress resistance of needles of *Pinus nigra* and *Pinus resinosa*: evaluation of the electrolyte leakage method. *Tree Physiology* 11(3):241-254.

Valliere JM. 2019. Tradeoffs between growth rate and water-use efficiency in seedlings of native perennials but not invasive annuals. *Plant Ecology* 220(3):361-369.

Wahid A, Gelani S, Ashraf M, Foolad, MR. 2007. Heat tolerance in plants: An overview. *Environmental and Experimental Botany* 61(3):199–223. <https://doi.org/10.1016/j.envexpbot.2007.05.011>

Wang T, Hamann A, Spittlehouse D, Carroll C. Locally Downscaled and Spatially Customizable Climate Data for Historical and Future Periods for North America. 2016. *PLoS ONE* 11. <https://doi.org/10.1371/journal.pone.0156720>

Wang Z, Huang B. 2004. Physiological Recovery of Kentucky Bluegrass from Simultaneous Drought and Heat Stress. *Crop Science* 44(5):1729–1736. <https://doi.org/10.2135/cropsci2004.1729>

Wei C, Zhao W, Fan R, Meng Y, Yang Y, Wang X, Foroud NA, Liu D, and Yu X. 2021. Genome-wide survey of the F-box/Kelch (FBK) members and molecular identification of a novel FBK gene TaAFR in wheat. *PLoS One* 16: e0250479.

Weiss M, Snieszko RA, Puiu D, Crepeau MW, Stevens K, Salzberg SL, Langley CH, Neale DB, & De La Torre, AR. 2020. Genomic basis of white pine blister rust quantitative disease resistance and its relationship with qualitative resistance. *The Plant Journal* 104(2):365–376. <https://doi.org/10.1111/tpj.14928>

Zhang J, Marshall JD, Jaquish BC. 1993. Genetic differentiation in carbon isotope discrimination and gas exchange in *Pseudotsuga menziesii*. *Oecologia* 93(1): 80-87.

Zhang SJ, Li Y L, Song GQ, Gao J, Zhang RZ, Li W, Chen ML, Li GY. 2020. Heterologous expression of the ThIPK2 gene enhances drought resistance of common wheat. *Journal of Integrative Agriculture* 19: 941-952.

Zhou X, Carbonetto P, Stephens M. 2013. Polygenic modeling with Bayesian Sparse Linear Mixed Models. *PLoS Genet* 9, e1003264, doi:10.1371/journal.pgen.10032

FIGURE LEGENDS

Figure 1. Geographic distribution of coastal and intervarietal hybrid Douglas-fir families included in this study.

Figure 2. Results of heat stress experiments and correlations among carbon isotope discrimination (an index of water use efficiency) and heat traits measured in this study. (A) Average foliar % electrolytic leakage of 100 Douglas-fir families (blue lines) at 7 temperature exposures; (B) Heatmap showing correlations among height, heat tolerance at 7 temperatures (T35C-T65C), %N (N), C:N ratio (CN), carbon isotope discrimination (X13C). Color legend indicates the Pearson's correlation value (r). Crosses indicate non-significant correlations.

Figure 3. Variation in heat tolerance (measured through % electrolytic leakage) across geographical origin of coastal and hybrid Douglas-fir families in Oregon and Washington. (A) Geographic variation in % electrolytic leakage at 55°C (T55C) and 45°C (T45C); (B) Geographic variation in mean annual temperature (MAT) and % electrolytic leakage at 40°C (T40C).

Figure 4. Correlations among heat tolerance, carbon isotope discrimination (an index of water use efficiency), growth and environmental variables in Douglas-fir families. Coastal families are shown in green and intervarietal hybrids in blue. Heat tolerance was measured by % electrolytic leakage at 7 temperatures (T35C-T65C). Bottom boxplots show differences between coastal and hybrid families in % nitrogen, electrolytic leakage at 55°C (T55C) and carbon isotope discrimination. A higher percentage of electrolytic leakage indicates a lower tolerance to heat. All correlations are significant with a P-value<0.001. Environmental variables included Mean Annual Temperature (MAT), Mean Annual Radiation (MAR), Precipitation as snow (PAS), Extreme Minimum Temperature (EMT), Continentality (Temperature difference between mean warmest and mean coldest month) and Degree Days below 0° Celsius.

Table 1- Differences in height, heat tolerance (electrolyte leakage from 35°C to 65°C), carbon isotope discrimination ($\Delta^{13}\text{C}$), and foliar nitrogen concentration (%N) and C:N ratio among coastal families, among hybrid families, and between coastal and hybrid varieties based on ANOVA results. Electrolytic leakage was measured at 35, 40, 45, 50, 55, 60 and 65 Celsius degrees.

Trait	Coastal Families		Hybrid Families		Coastal and Hybrids	
	F value	P value	F value	P value	F value	P value
Height	14.50	<0.0001	25.73	<0.0001	30.211	<.0001
T35C	1.511	0.003	1.218	0.335	0.601	0.439
T40C	1.728	0.000	3.497	0.040	0.479	0.490
T45C	1.588	0.001	2.605	0.087	1.358	0.246
T50C	1.396	0.015	2.161	0.132	3.362	0.069
T55C	0.907	0.713	0.793	0.515	6.150	0.013
T60C	1.210	0.108	0.321	0.809	1.442	0.232
T65C	1.241	0.081	0.930	0.449	0.134	0.714
$\Delta^{13}\text{C}$	2.691	<0.0001	11.356	<0.0001	5.079	0.026
%N	4.530	<0.0001	5.517	0.018	8.695	0.003
C:N	5.161	<0.0001	6.873	0.0005	15.589	<.0001

Table 2. Functional annotation of the genes of significant SNPs identified by univariate linear mixed models (ulmm), multivariate linear mixed models (mvmm) at GEMMA, and general linear model (GLM) in TASSEL for coastal and hybrid Douglas-fir. The “Trait” column identifies the heat tolerance at different temperatures, carbon isotope discrimination, nitrogen content, or growth trait; “Gene” column identifies the gene associated with the trait; “Similar search ID” column indicates the genbank or uniprot ID identifying similar BLAST hits; “SNP” column indicates the number of associated SNPs; P-value is the significance level; and Annotation and KEGG provide the functional annotation for candidate genes. Genes in bold were identified as associated with growth, cold hardiness, phenology and environmental variables in a previous Douglas-fir study (De La Torre et al. 2021).

Trait	Gene	Similar Search ID	SNPs	P-value	Annotation	KEGG Pathway/KEGG-Brite/Process
T35C	PSME_20450 ⁹²	KAF7835953.1	1	6.26 ⁻⁰⁷	Ecotropic viral integration site protein	-
T35C	PSME_48755 ⁹²	XP_031386554.1	1	3.16 ⁻⁰⁶	F-box/FBD/LRR-repeat protein	-
T35C	PSME_04475 ⁹²	XP_024384147.1	1	8.06 ⁻⁰⁷	F-box/Kelch-repeat protein	-
T35C	PSME_39629 ⁹²	XP_031480439.1	1	1.36 ⁻⁰⁶	Inositol-tetrakisphosphate 1-kinase 3-like isoform	Inositol phosphate metabolism, phosphatidylinositol signaling system pathways
T35C	psme_t025434 m.78800 ⁹²	-	1	1.86 ⁻⁰⁶	-	-
T35C	PSME_15499 ⁹²	XP_010260370.1	1	2.89 ⁻⁰⁶	transcription factor VOZ1	Abiotic and biotic stress

T40C	PSME_42829 ⁹²	CAN68428.1	1	3.21 ⁻⁰⁶	uncharacterized protein	-
T45C	PSME_42829 ⁹²	CAN68428.1	1	2.11 ⁻⁰⁶	uncharacterized protein	-
T50C	PSME_15499 ⁹³	XP_010260370.1	1	2.69 ⁻⁰⁶	transcription factor VOZ1	Abiotic and biotic stress
CN	PSME_31599 ⁹²	XP_031503996.1	1	1.19 ⁻⁰⁶	reticulon-like protein B1	Membrane trafficking
Height	PSME_27219⁹³	XP_039125026.1	1	1.14 ⁻⁰⁸	cyclin-dependent kinase C-2 isoform	Cell cycle control
Height	PSME_27198 ⁹³	XP_024396643.1	1	2.40 ⁻⁰⁸	nucleolin-like	Ribosomal biogenesis
Height	PSME_47504⁹³	ADB97926.1	1	1.45 ⁻⁰⁷	thaumatin-like protein L2	-
Height	PSME_29636 ⁹³	XP_023884839.1	1	1.74 ⁻⁰⁷	uncharacterized protein	-
Height	PSME_48306 ⁹³	-	1	1.74 ⁻⁰⁷	-	-
Height	PSME_35875⁹³	XP_030930392.1	1	1.79 ⁻⁰⁷	NAD(P)H dehydrogenase	Ubiquinone and other terpenoid-quinone biosynthesis pathway
Height	PSME_47942⁹³	XP_038710372.1	1	2.35 ⁻⁰⁷	cysteine-rich repeat secretory protein	Salt stress

Height	PSME_47800 ^h	XP_014505538.1	1	2.56 ⁻⁰⁷	inactive purple acid phosphatase	Growth and development
Height	PSME_52709 ^h	-	2	2.78 ⁻⁰⁷	-	-
Height	PSME_42675 ^h	ABR16422.1	1	3.71 ⁻⁰⁷	unknown	-
Height	PSME_45982 ^h	XP_024973942.1	1	4.38 ⁻⁰⁷	UPF0481 protein At3g47200-like	-
Height	PSME_01087 ^h	XP_017217789.1	1	5.03 ⁻⁰⁷	AT-hook motif nuclear-localized protein	DNA replication and repair
Height	PSME_02019 ^h	XP_009346468.1	1	6.00 ⁻⁰⁷	uridine-cytidine kinase C-like	Pyrimidine metabolism, Metabolic pathways
Height	PSME_33611 ^h	XP_010920821.1	1	8.23 ⁻⁰⁷	protein PMR5	Biotic stress
Height	PSME_00292 ^h	NP_013268.1	1	1.00 ⁻⁰⁶	ubiquitin-ribosomal 40S subunit protein	Ubiquitin mediated proteolysis
Height	PSME_28922 ^h	QBI90547.1	1	1.29 ⁻⁰⁶	flavanone-3-hydroxylase	Flavonoid biosynthesis, Biosynthesis of secondary metabolites pathways
Height	PSME_01405 ^h	XP_038985456.1	1	1.40 ⁻⁰⁶	putative disease resistance protein	Plant-pathogen interaction
Height	PSME_32617 ^h	XP_011077233.1	1	1.48 ⁻⁰⁶	Phosphoenolpyruvate carboxylase kinase 2	Calcium signaling pathway, biotic stress
Height	PSME_42381 ^h	XP_021276483.1	4	2.11 ⁻⁰⁶	uncharacterized protein	-
Height	PSME_04717 ^h	ABF58895.1	1	2.46 ⁻⁰⁶	GABA aminotransferase	Alanine, aspartate and glutamate metabolism pathway
All	PSME_03650 ^h	NP_012017.1	1	2.31 ⁻⁰⁶	Mitochondrial 54S ribosomal protein	-
All	PSME_42829 ^h	CAN68428.1	1	7.25 ⁻⁰⁶	Uncharacterized protein	-

All	PSME_15499	XP_010260370.1	1	2.75 ⁻⁰⁶	transcription factor VOZ1	Abiotic and biotic stress
All	PSME_15952	XP_024168655.1	1	2.87 ⁻⁰⁶	mRNA turnover protein 4	Ribosome biogenesis pathway

⁰ univariate linear mixed models (ulmm) at GEMMA; ⁰⁰ general linear model (GLM) analysis at TASSEL; ⁰⁰⁰ multivariate linear mixed models (mvmm) at GEMMA, All = phenotypes including Height, 35°C, 40°C, 45°C, 50°C, 55°C, 60°C, 65°C, N, C: N, $\Delta^{13}\text{C}$

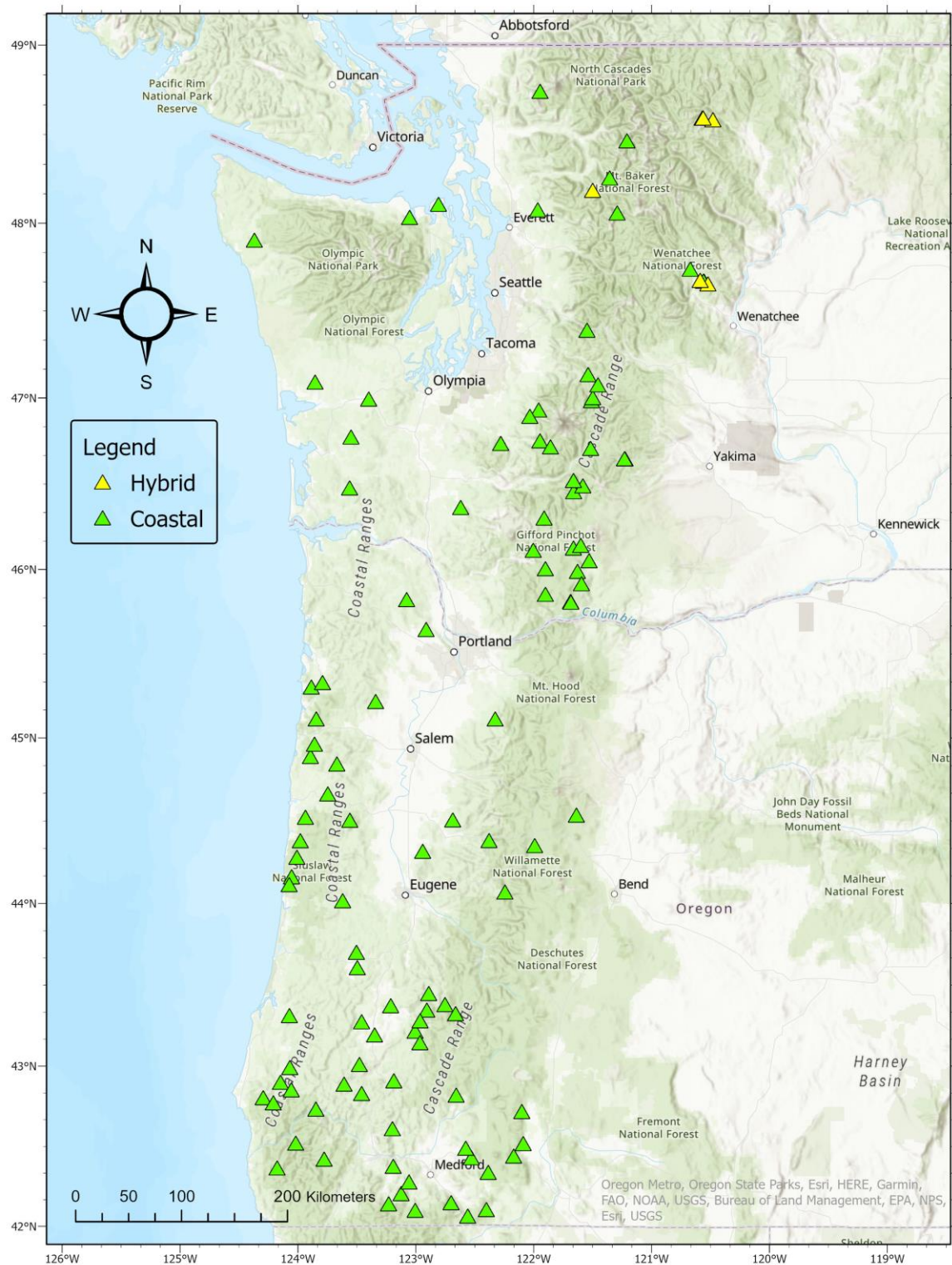
Table 3. Functional annotation of the genes of significant SNPs identified by univariate linear mixed models (ulmm), multivariate linear mixed models (mvmm) at GEMMA, and general linear model (GLM) in TASSEL for coastal Douglas-fir (*Pseudotsuga menziesii*). The “Trait” column identifies the associated heat tolerance at different temperatures, carbon isotope discrimination, nitrogen content, or growth trait; “Gene” column identifies the gene associated with the trait; “Similar search ID” column indicates the genbank or uniprot ID identifying similar BLAST hits; “SNP” column indicates the number of associated SNPs; P-value is the significance level and “Annotation” provides the functional annotation for candidate genes.

Trait	Gene	Similar search (ID)	SNP	P-value	Annotation
T35C	PSME_39629 ²²	XP_031480439.1	1	1.73 ⁻⁰⁶	Inositol-tetrakisphosphate 1-kinase 3-like isoform
T35C	PSME_20450 ²²	KAF7835953.1	1	3.51 ⁻⁰⁷	Ecotropic viral integration site protein
T35C	PSME_48755 ²²	XP_031386554.1	2	9.79 ⁻⁰⁷	F-box/FBD/LRR-repeat protein
T35C	PSME_04475 ²²	XP_024384147.1	1	1.13 ⁻⁰⁶	F-box/Kelch-repeat protein
T35C	PSME_00998 ^{22*}	XP_031497232.1	1	1.73 ⁻⁰⁶	Pentatricopeptide repeat-containing protein
T35C	PSME_47800 ²²	XP_014505538.1	1	1.73 ⁻⁰⁶	Probable inactive purple acid phosphatase
T40C	PSME_42829 ²¹	CAN68428.1	1	1.97 ⁻⁰⁷	Uncharacterized protein
T45C	PSME_42829 ²¹	CAN68428.1	1	1.14 ⁻⁰⁷	Uncharacterized protein

CN	PSME_31599 ²¹²	XP_031503996.1	1	1.92 ⁻⁰⁶	Reticulon-like protein B1
All	PSME_03650 ²¹²	NP_012017.1	1	1.74 ⁻⁰⁷	Mitochondrial 54S ribosomal protein
All	PSME_36695 ²¹² *	XP_024401260.1	1	5.39 ⁻⁰⁷	Protein EXECUTER2, chloroplastic-like isoform
All	PSME_42829 ²¹²	CAN68428.1	1	1.25 ⁻⁰⁷	Uncharacterized protein
All	PSME_15952 ²¹²	XP_024168655.1	1	2.68 ⁻⁰⁷	mRNA turnover protein

²¹² univariate linear mixed models (ulmm) at GEMMA; ²¹³ general linear model (GLM) analysis at TASSEL; ²¹⁴ multivariate linear mixed models (mvmm) at GEMMA; All = phenotypes including Height, 35°C, 40°C, 45°C, 50°C, 55°C, 60°C, 65°C, N, C: N, $\Delta^{13}\text{C}$; *not found as candidate genes in Table 2.

Figure 1



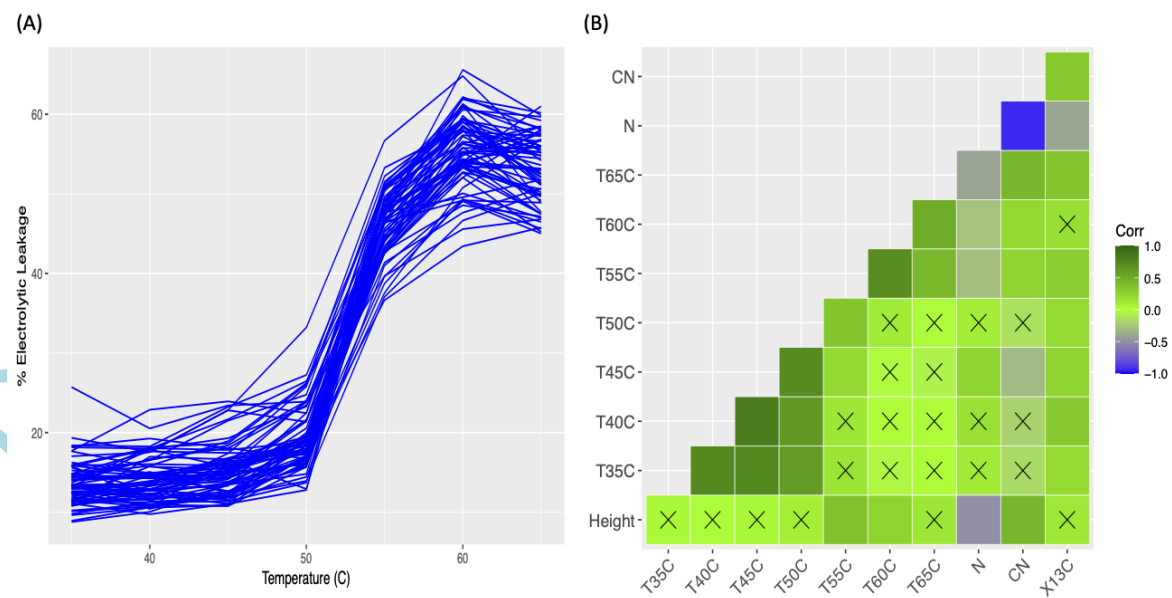


Figure 3

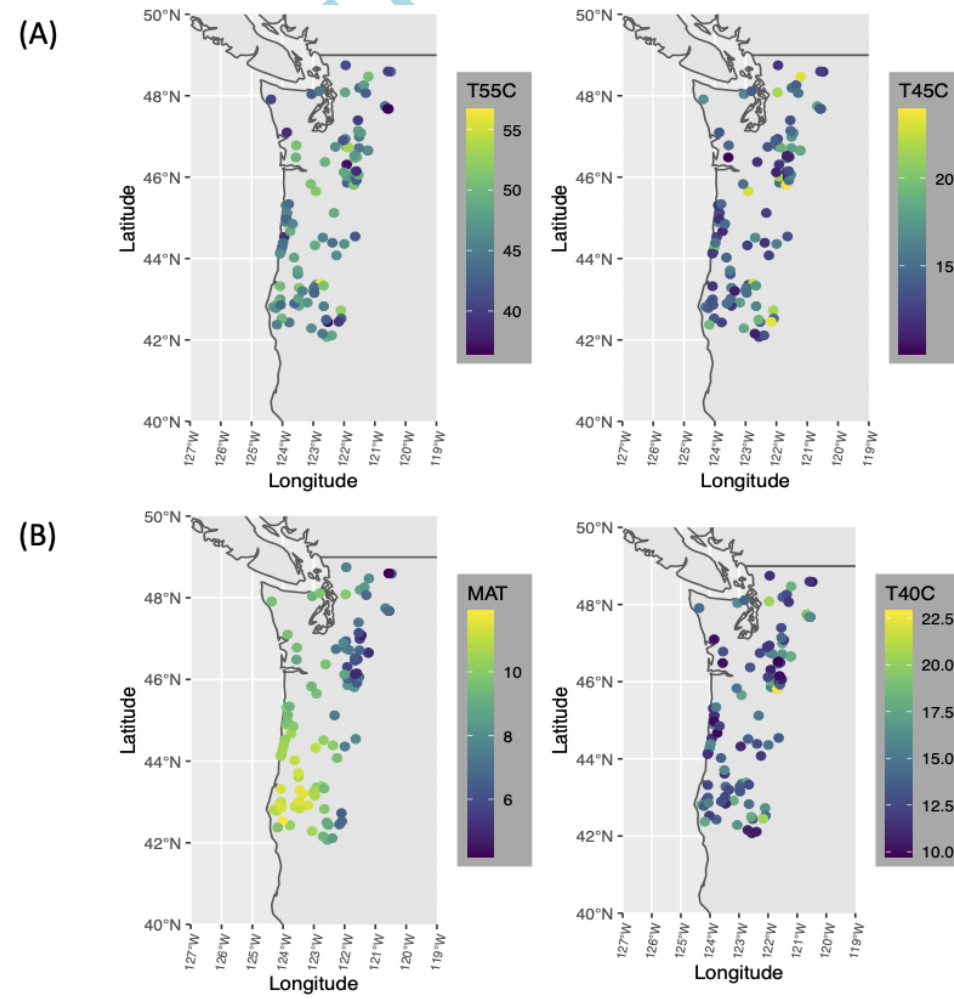


Figure 4

