

What can the cold-induced transcriptomes of Arctic Brassicaceae tell us about the evolution of cold tolerance?

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Abstract

- By studying the molecular basis of cold response in plants adapted to some of the world's coldest biomes, we can gain insight into the evolution of cold tolerance - an important factor in determining plant distributions worldwide.
- Although cold tolerance in temperate plants have been extensively studied, little is known about the evolutionary changes needed to transition from temperate to the more extreme polar zones.
- Here, we conducted a time series experiment to examine the transcriptional responses of three Arctic Brassicaceae to low temperatures. RNA was sampled before onset of treatment, and after 3h, 6h, and 24h with 2 °C. We identified sets of genes that were differentially expressed in response to cold and compared them between species, as well as to published data from the temperate *Arabidopsis thaliana*.
- We found that the cold response is highly species-specific. Among thousands of differentially expressed genes, ~200 genes were shared among the three Arctic species and *A. thaliana*, and only ~100 genes were specific to the three

Arctic species alone. This pattern was also reflected in the functional comparison.

- Our results show that the cold response of Arctic plant species has mainly evolved independently, although it likely builds on a conserved basis found across Brassicaceae. The findings also confirm that highly polygenic traits, such as cold tolerance, may show less repeatable patterns of adaptation than traits involving only a few genes.

1. Introduction

Temperature is one of the most important factors determining plant distributions across the world, and only a few species have been able to occupy the cold biomes found towards the poles (Wiens and Donoghue 2004; Billings and Mooney 1968). Evolutionary history seems to play an important role in determining how plant cold tolerance is distributed globally, with tropical-to-temperate transitions being key events in plant evolution (Lancaster and Humphreys 2020). However, little is known about the evolutionary changes required for the transition from temperate to the more extreme polar zones. It is still unclear if the cold response of polar plant species is distinct from that of temperate relatives, and whether it may have evolved in a similar or convergent fashion because of the same extreme selection pressures. Polar environments exhibit lower year-round temperatures and shorter growing seasons (with up to 24 hours of daylight) than temperate environments, although some temperate-alpine environments share similar characteristics (Billings 1974). In the Arctic, the average temperature of the warmest summer month is not more than 10 °C (Elvebakk 1999), and the growing season can be as short as one month in the coldest areas (e.g. Jónsdóttir 2005). By studying how Arctic plants cope with low temperatures, we can gain insights into how plants acquire cold tolerance, and if there are general trends in plant adaptation to extreme polar environments.

Plants face many challenges upon the transition to colder environments. Low temperatures can affect nearly all aspects of plant cell biochemistry, and influence protein properties, photosynthesis reactions, and cell membrane fluidity (Shi, Ding, and Yang 2018). Ice formation comes with its own set of challenges and is in general deadly if ice forms within the cell (Körner 2003). Some plants can tolerate freezing of the apoplast (the space between the cells), but this may draw water out of the cell and lead to severe dehydration, as well as increase the level of salts and toxic solutes

(Wisniewski et al. 2004; Steponkus 1984; Körner 2003; Wisniewski and Fuller 1999). In temperate environments (from where Arctic plants most likely are derived; Abbott and Brochmann 2003), plants prepare for predictable cold periods via cold acclimation, i.e. an increase in freezing tolerance in response to low non-freezing temperatures (Thomashow 2010). Exposing temperate plants to low temperatures typically results in complete reorganization of the transcriptome, ultimately leading to increased freezing tolerance (Kreps et al. 2002; Thomashow 2010; Kilian et al. 2007). The CBF transcription factors (C-repeat-Binding Factors) are among the main “regulatory hubs” of this cold response, and have been isolated in many different plant species (Shi, Ding, and Yang 2018; Park et al. 2015; Thomashow 2010). The CBFs are induced shortly after exposure to cold stress, and control the expression of >100 cold-regulated (COR) genes downstream (the CBF pathway; Park et al. 2015). *Arabidopsis thaliana* exhibits three of these cold-induced CBFs (*CBF1*, *CBF2* and *CBF3*, also called *DREB1b*, *DREB1c* and *DREB1a*; Jia et al. 2016), but the same genes are not always found in other species (Zhao et al. 2012). It is also becoming increasingly clear that the CBF regulon involves extensive co-regulation by other lesser-known transcription factors, and that the low temperature regulatory network is highly complex (Park et al. 2015). One could envision that polar plant species are in less need of a cold acclimation period as temperatures are low year-round (e.g., their transcriptomes could be less responsive to a drop in temperature), or that their cold response is somehow more complex (e.g., involving more genes) or more effective (e.g., faster, or involving fewer genes). There are surprisingly few in-depth studies on the cold-induced transcriptomes of polar plants, but Archambault and Strömvik (2011) studied Arctic *Oxytropis*, Wang et al. (2017) the temperate-subarctic *Eutrema* (*Thellungiella*) *salsugineum*, and Lee et al. (2013) the Antarctic *Dechampsia antarctica*. Although these studies give valuable species-specific information on cold response, they give limited insight into how polar plants differ from temperate relatives. In this study, we therefore perform a whole-transcriptome investigation of cold response in three plant species adapted to the high Arctic, and then compare their response to that of temperate relatives.

The focal species of this study, *Cardamine bellidifolia*, *Cochlearia groenlandica* and *Draba nivalis*, have independent Arctic origins (Carlsen et al. 2009; Jordon-Thaden et al. 2010; Koch 2012) and represent three of the main clades of Brassicaceae (clade A, B, and C; divergence time ~30 Mya; Huang et al. 2016).

They are ideal model species for studying cold response of Arctic plants for three main reasons: 1) all have their main distribution above the Arctic Circle, 2) they are found in all Arctic bioclimatic subzones (even in polar deserts; Elven et al. 2011), and 3) they belong to the plant family in which cold response has been most extensively studied, as it includes the model species *A. thaliana* and various economically important crop species (Kilian et al. 2007; Park et al. 2015; Shi, Ding, and Yang 2018). Furthermore, we recently found evidence of positive selection in genes associated with cold stress in all three species (Birkeland et al. 2020). Different genes seem to be under positive selection in each species, suggesting that they have independently adapted to the Arctic by modifying different components of similar stress response pathways (Birkeland et al. 2020). However, our previous study was limited to protein coding regions, and theory predicts that there could be a higher chance of convergence in their expression profiles (e.g. Stern 2013; Sackton et al. 2019). The reason is that mutations in *cis*-regulatory regions should have fewer pleiotropic effects than mutations in coding regions, as protein function may be affected only in a subset of the full expression domain (Gompel and Prud'homme 2009; Stern 2013). This is tied to the fact that a gene can have several *cis*-domains and that one *cis*-domain may bind several different transcription factors. To evaluate the degree of similarity in Arctic Brassicaceae expression profiles and to describe how Arctic cold-induced transcriptomes differ from those of temperate relatives, we therefore subjected these three species to a simulated summer cold shock and identified differentially expressed genes after 3h, 6h and 24h with cold treatment. We aimed to i) characterize the cold-induced transcriptomes of *C. bellidifolia*, *C. groenlandica* and *D. nivalis*, ii) describe how their cold response differ from that of *A. thaliana*, and iii) identify potential convergent expression patterns in the three Arctic species.

2. Material and methods

2.1 Plant material

For each species, we sowed seeds from a single selfed parent derived from wild Arctic populations from Alaska (Supplementary table 1). Seeds were sown in six pots per species (several seeds per pot), of which four pots were used as biological replicates at each time point (see below). Because the plants were siblings derived from selfed

parents, and because selfing is assumed to be the predominant mode of reproduction of these species in the wild (Brochmann and Steen 1999), we expected the replicates to be close to genetically identical. The plants were grown in a natural daylight room in the phytotron at the University of Oslo from 22nd of March to 18th of May 2018 with day temperature at 22 °C and night temperature at 18 °C. Supplementary artificial light was given from 08:00-24:00 to mimic Arctic light conditions (400 W high-intensity discharge lamps), and moisture was at ~65 % RH.

2.2 Cold shock treatment

Eight weeks after sowing, the plants were given a 24-hour cold shock to simulate a sudden drop in temperature during a typical Arctic summer. At 13:00 p.m. (to minimize correlation with circadian change), we transferred the pots from the 22 °C daylight room to a 2 °C cooling room with artificial light from 250 W high-intensity discharge lamps (140-160 $\mu\text{mol m}^{-2} \text{s}^{-1}$ measured at plant height). Leaf tissue was sampled at four time points; just before they were transferred (0h; control), and after 3h, 6h, and 24h. We sampled all six pots for RNA extractions at each time point, but only used the four best RNA extracts per time point for sequencing (i.e., in terms of RNA quality and quantity). These constituted our four biological replicates per time point per species. RNA was immediately extracted from fresh, fully expanded leaves as described below. During the 24h cold shock treatment, the plants were given supplementary artificial light to replicate the conditions in which they were sprouted (see above).

2.3 RNA extraction and sequencing

For extraction of total RNA, we used the Ambion RNAqueous Kit (Thermo Fisher Scientific, Waltham, USA), following the manufacturer's protocol for fresh plant tissue: ~50 mg leaf tissue was immediately ground in Lysis/Binding Solution together with 1 volume of Plant RNA Isolation Aid and consecutively extracted. The RNA quantity was measured with Broad Range RNA Kit on a Qubit v.2.0 fluorometer (Life Technologies, Carlsbad, USA); RNA quality with an Experion Automated Electrophoresis System Station (Bio-Rad Laboratories, Hercules, USA) and a Nandrop One spectrophotometer (Thermo Fisher Scientific, Waltham, USA). The Norwegian Sequencing Centre (www.sequencing.uio.no) prepared the libraries using the TruSeq protocol for stranded mRNA (Illumina, San Diego, USA) and performed

the sequencing. Samples were indexed, pooled, and run on three lanes (16 samples/lane) on an Illumina HiSeq 3000 (Illumina, San Diego, USA), producing paired end reads with a default insert size of 350 bp and read lengths of 150 bp. The raw reads were quality-checked with FastQC v.0.11.8 (Andrews 2010), and a single FastQC report per species was generated with MultiQC v.1.7 (Ewels et al. 2016).

2.4 Transcriptome assembly and annotation

As there were no available genome assemblies at the start of this study, we assembled a reference transcriptome *de novo* for each species using Trinity v.2.8.5 (Grabherr et al. 2011) based on all acquired RNA samples. Trinity was run with the integrated Trimmomatic option (Bolger, Lohse, and Usadel 2014), strand-specificity, and a minimum assembled contig length of 300 bp. The transcriptomes were filtered and annotated with EnTAP (Eukaryotic Non-Model Transcriptome Annotation Pipeline; Hart et al. 2020) in two rounds: first to apply the EnTAP filtering option on the raw transcriptome (in order to reduce inflated transcript estimates), and a second time to annotate the highest expressed isoform and filter out contaminants (used for the annotation of DEGs; see below). For expression filtering, an alignment file was generated with bowtie2 (Langmead and Salzberg 2012) in combination with RSEM (B. Li and Dewey 2011) using default options in the “align_and_estimate_abundance.pl” script provided with the Trinity software suite. Numbers of complete and fragmented BUSCOs (Benchmarking Universal Single-Copy Orthologs) in the filtered transcriptomes were estimated with BUSCO v4.0.6 (Simão et al. 2015). The filtered transcriptomes were used as the final reference in the differential expression analyses (see below).

2.5 Differential expression analyses

The Trimmomatic filtered reads were mapped to the reference transcriptomes using the alignment free mapper Salmon with a GC content bias correction (Patro et al. 2017). Genes that were differentially expressed after 3h, 6h and 24h of cold treatment were identified with DESeq2 v.1.22.1 (Love, Huber, and Anders 2014), using a design formula controlling for the effect of pot number (design = ~ pot number + time). This means that we tested for the effect of time with 2 °C treatment, while controlling for the individual effects of the sampled pots. A generalized linear model was fitted to each gene and a Wald test (Love, Huber, and Anders 2014) applied to test if the 3h,

6h and 24h model coefficients differed significantly from zero when contrasted to the 0h model coefficient. A gene was considered as differentially expressed if the transcript level exhibited \geq twofold change in response to the cold treatment at the different time points (\log_2 fold change threshold = 1). We used a Benjamini-Hochberg adjusted p-value (Benjamini and Hochberg 1995) to evaluate the significance of each differentially expressed gene (False Discovery Rate cutoff of $\alpha = 0.05$). Heatmaps of the top 30 differentially expressed genes with the lowest false discovery rate were generated with the pheatmap package in R (Kolde 2019) using the regularized log function (rld) on original count data. The mean expression value of a gene was subtracted from each observation prior to heatmap generation.

2.6 Comparison of DEG sets among Arctic species and *A. thaliana*

To enable the comparison of DEGs among the Arctic species, and among the Arctic species and *A. thaliana*, we used already published data on differentially expressed genes in wild type *A. thaliana* in response to 24h chilling treatment (Table S1 in Park et al. 2015). In this experiment, *A. thaliana* wild type plants were grown at 22 °C and constant illumination, then exposed to a 4 °C chilling treatment for 24 h (Ws-2; see Park et al. 2015 for details). Many of our analyses thus focus on the 24h DEG sets. We used two different approaches to compare the 24h DEG sets among species. First, we ran OrthoFinder v.2.3.12 (Emms and Kelly 2019) to identify sets of genes putatively descended from a single gene in the last common ancestor of all four species (orthogroups), using the assembled transcriptomes (filtered based on highest expressed isoform) and the Araport11 peptide file of *A. thaliana* downloaded from www.arabidopsis.org. This enabled us to compare orthogroup identity among DEG sets. Second, we used the BLASTP (protein-protein) search of BLAST+ v.2.9.0 (Camacho et al. 2009) to identify putative *A. thaliana* homologs in the three Arctic species, using the Araport11 peptide file as database and each of the Arctic transcriptome files as query (with e-value < 0.01 and max target seqs = 1). This second approach enabled us to compare gene identity and function more closely among species, but with the caveat that the *A. thaliana* homolog might not always be the true homolog or represent the same gene function in all species. The significance of the overlaps among differentially expressed orthogroups at 24h were evaluated using the supertest function in SuperExactTest v.1.0.7 (M. Wang, Zhao, and Zhang 2015). We also visualized potential unique overlaps among differentially expressed

orthogroups at 24h and among differentially expressed *A. thaliana* homologs at 24h, using UpSetR v.1.4.0 (Conway, Lex, and Gehlenborg 2017). To compare transcription factor composition, we annotated the *A. thaliana* 24h DEG set with EnTAP (as above).

2.7 Gene Set Enrichment Analyses

To characterize sets of upregulated and downregulated genes further functionally, we performed gene ontology (GO) enrichment analyses within the Biological Process (BP), Cellular Component (CC) and Molecular Function (MF) domains for each species and time point. We used the Fisher's exact test in combination with the elim algorithm implemented in topGO 2.34.0 of Bioconductor to test for overrepresented GO-terms in each set of significantly upregulated or downregulated genes (Gentleman et al. 2004; Alexa, Rahnenfuhrer, and Lengauer 2006). The elim algorithm works by traversing the GO-graph bottom-up and discarding genes that already have been annotated to significant child terms, and is the recommended algorithm by the creators of topGO due to its simplicity (Alexa, Rahnenfuhrer, and Lengauer 2006). For the enrichment analyses, we used the gene annotations of the transcriptomes as background gene sets in each test (using the GO-annotations acquired with EnTAP). For *A. thaliana*, we used the org.At.tair.db R package v.3.7.0 to annotate the 24h DEG set of Park et al. (2015), and for creating a background gene set used in the enrichment tests (Carlson 2018). A GO-term was considered significantly enriched if $p < 0.05$. We did not correct for multiple testing as the enrichment-tests were not independent. Overlaps among enriched GO-terms in similar DEG sets (i.e., upregulated, and downregulated genes at similar time points) were compared among species using UpSetR as above.

2.8 Comparison of DEGs with data set on positively selected genes

Because we previously have identified convergent substitutions and tested for positive selection in the three focal species (see Birkeland et al. 2020 for details), we were able to check for potential overlaps between positively selected genes/convergent genes and the 24h DEG sets. We blasted the newly assembled transcriptomes against the transcriptomes of the previous study using BLASTP with an e-value cutoff of < 0.01 and max target seqs = 1.

2.9 Gene co-expression network analyses

To identify co-expressed gene modules, we performed weighted correlation network analysis for each species using the R package WGCNA (Langfelder and Horvath 2008). The gene expression matrix was prepared by first filtering out genes with consistent low counts (row sums ≤ 10), and then applying a variance stabilizing transformation within DESeq2. We also filtered out genes with low expression variance by only maintaining genes with variance ranked above the 25 percentile in each data set. A gene adjacency matrix was constructed using default settings and raised to a soft thresholding power of 18 (signed network type). The soft thresholding power was chosen based on recommendations in the WGCNA FAQs, since a scale-free topology fit was reached only at very high values. This is not uncommon when a subset of the samples is globally different from the rest (for instance cold treated versus non-cold treated samples), which causes high correlation among large groups of genes. The adjacency matrix was translated into a Topological Overlap Matrix (TOM) and the corresponding dissimilarity measure. Co-expressed gene modules were then identified by using the resulting dissimilarity matrix as input in an average linkage hierarchical clustering analysis and pruning the dendrogram into modules using the Dynamic Tree-Cut algorithm (minimum module size of 30, deepSplit of 2, and pamRespectsDendro as FALSE; Langfelder et al. 2008). Close modules were merged based on the correlation of their eigengenes (correlation threshold of 0.75). We identified hub genes within each module as the genes with the top 10 % eigengene-based connectivity (module membership). We also calculated the correlation between each module and a binary measure of cold (0/1) as well as a time-based measure of cold (0, 3, 6, and 24 hours), which was visualized in a heatmap. Gene set enrichment analyses of each module were performed with topGO as described above.

3. Results

3.1 Transcriptome assemblies and Differentially Expressed Genes (DEGs)

The three *de novo* assemblies contained ~22,000-24,000 (Trinity) genes and were highly complete in terms of BUSCOs (>90 % complete; Table 1). We identified a

gradual increase in the number of DEGs with time at 2 °C. About 700-1000 DEGs were identified after 3h, with varying rates of increase at 6h (+33 DEGs in *C. bellidifolia*, +283 DEGs in *C. groenlandica*, and +895 DEGs in *D. nivalis*), and with ~2500-3000 DEGs identified at 24h (Table 2; Supplementary tables S2-4). Initially, most DEGs were upregulated, but after 24h we found similar numbers of downregulated and upregulated genes in all species (Table 2). Most 3h and 6h DEGs were also found in the 24h set, but many DEGs were also unique for each time point (Supplementary tables S5-7).

Based on orthogroup identity, most 24h DEGs seemed to be species-specific, but we found 212 differentially expressed orthogroups shared by the three Arctic species and *A. thaliana*, and 106 differentially expressed orthogroups shared by Arctic species but not by *A. thaliana* (Figure 1). These unique orthogroup intersections corresponded to overlaps of 195 and 119 genes based on *A. thaliana* homologs, respectively (Supplementary Fig. 1). Note that several genes from the same species occasionally can have the same orthogroup identity. We also found that all species (i.e. all species combinations) shared significant orthogroup overlaps in the 24h DEG sets (all $p < 0.01$ based on the supertest; Supplementary table 8). Thus, our main findings were that the shared portion of the cold response was bigger than expected by chance, and that more genes were shared by the three Arctic species and *A. thaliana*, than by the Arctic species alone.

Table 1. Statistics for de novo transcriptome assemblies

Species (source)	Total no. read pairs	No. of “genes” in final assembly ^b [raw assembly]	No. of isoforms in final assembly [raw assembly]	% complete BUSCOs in final assembly
<i>Cardamine bellidifolia</i>	403,256,653 (16 samples ^a)	21,818 [42,151]	42,646 [98,419]	93.8%
<i>Cochlearia groenlandica</i>	389,172,001 (16 samples ^a)	22,396 [49,768]	40,639 [102,855]	93.4%
<i>Draba nivalis</i>	368,925,923 (16 samples ^a)	23,871 [52,096]	46,282 [109,658]	92.6%

^aFour replicates at four time points (0h, 3h, 6h, 24h), ^bCorresponding to Trinity genes (or transcript clusters)

Table 2. Number [percentage] of differentially expressed genes (DEGs) after 3h, 6h and 24h with 2 °C. (bold = total number of DEGs, ↑/↓ = upregulated/downregulated DEGs)

	3h *	6h *	24h *
<i>Cardamine bellidifolia</i>	1012 857↑, 155↓ [85% ↑, 15% ↓]	1045 877↑, 168↓ [84% ↑, 16% ↓]	2520 (1301↑, 1219↓) [52% ↑, 48% ↓]
<i>Cochlearia groenlandica</i>	733 521↑, 212↓ [71% ↑, 29% ↓]	1016 (636↑, 380↓) [63% ↑, 37% ↓]	3010 (1534↑, 1476↓) [51% ↑, 49% ↓]
<i>Draba nivalis</i>	688 (505↑, 183↓) [73% ↑, 27% ↓]	1583 (998↑, 585↓) [63% ↑, 37% ↓]	2839 (1484↑, 1355↓) [52% ↑, 48% ↓]

3.2 Comparison of DEGs among species

3.2.1 Transcription factors in Arctic species

Transcription factors, the genes potentially orchestrating the cold response, made up 9-14% of all Arctic DEGs at all times (Table 3, Supplementary tables 2-4). AP2/ERF (Figure 2) was the most common family of transcription factors based on the 24h DEG set; known to include important candidates for cold regulation such as CBFs/DREBs and RAVs.

Among the transcription factors in the AP2/ERF family were *CBF1* (*DREB1B*) and *CBF4* (*DREB1D*), which were upregulated upon cold treatment in all species. Other upregulated AP2/ERFs included *DREB2A* (at all time points in *C. groenlandica*, and at 24h in *C. bellidifolia*), *DREB2C* (at all time points in *C. bellidifolia* and *D. nivalis*), *DREB3* (at all time points in *C. groenlandica* and *D. nivalis*), and *RAV1* (found to be upregulated at 6h and 24h in *C. bellidifolia*).

Other common transcription factors in all three species were those containing SANT/Myb domains, MYC-type basic helix-loop-helix (bHLH) domains, basic-leucine zipper domains, and NAC domains (Figure 2). Many of these transcription factors were among the DEGs shared only by Arctic species based on the *A. thaliana* homolog (for instance *REVEILLE 2*, *RAP2.10*, *RAP2.2*, *PCL1*, and *HY5*; Supplementary table 9).

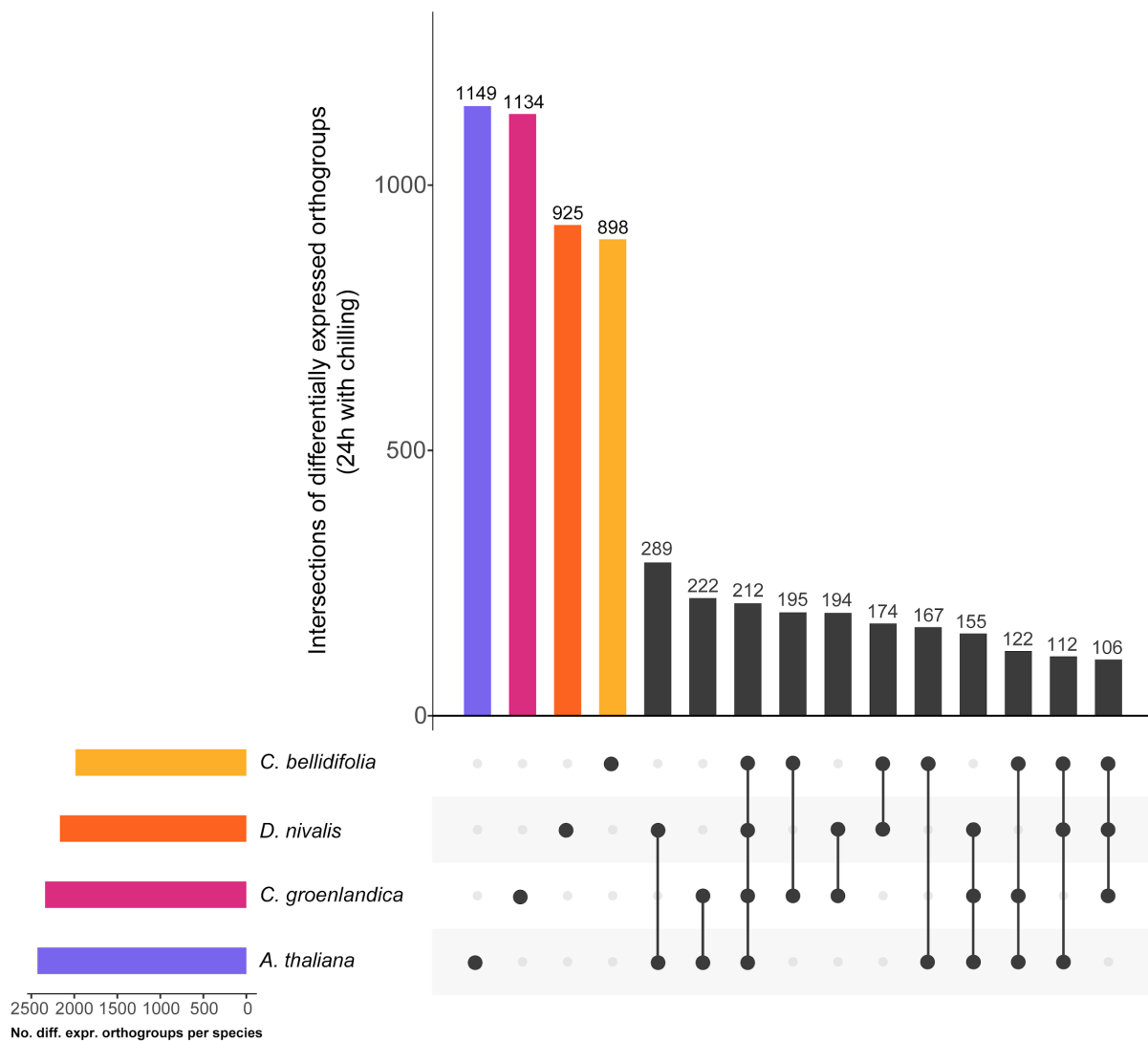


Figure 1. UpSet plot of differentially expressed orthogroups. The plot in the left corner shows total numbers of differentially expressed orthogroups, and the main plot shows the number of unique differentially expressed orthogroups, followed by orthogroups intersections/overlaps between species (connected dots).

Table 3. Number of differentially expressed genes (DEGs) annotated with “regulation of transcription” (GO:0006355) after 3h, 6h and 24h with 2 °C. The percentage of the total DEG set is given in parentheses.

	3h *	6h *	24h *
<i>Cardamine bellidifolia</i>	123 (12%)	139 (13%)	266 (11%)
<i>Cochlearia groenlandica</i>	106 (14%)	143 (14%)	344 (11%)
<i>Draba nivalis</i>	92 (13%)	180 (11%)	260 (9%)

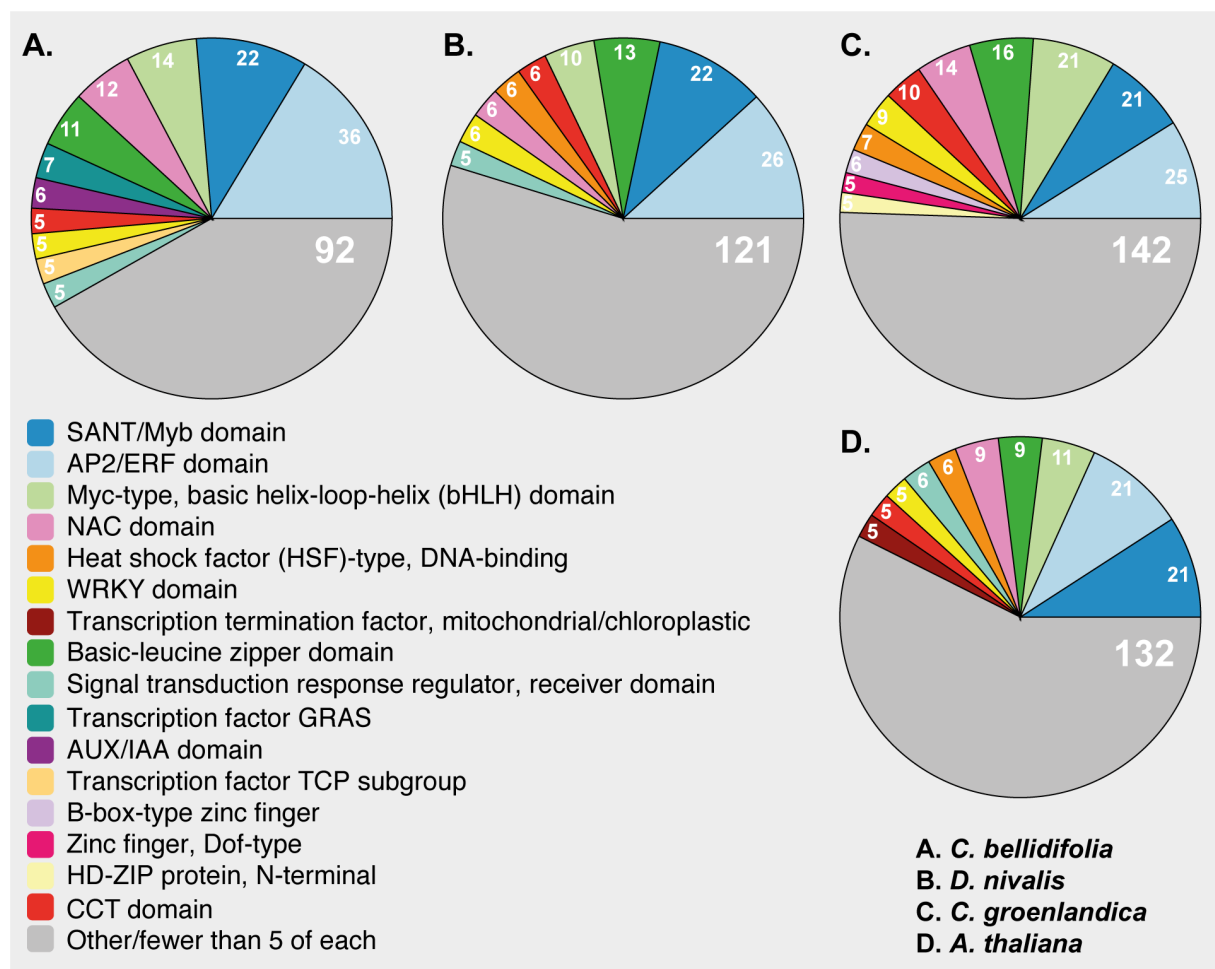


Figure 2. InterPro domains in 24h DEGs annotated with “regulation of transcription” (GO:0006355). Transcription factors that did not have InterPro domain information are not included.

3.2.2 Other shared cold regulated genes in Arctic species

The gene overlaps based on *A. thaliana* homologs were used to further investigate cold regulated genes that were common in the cold response of all Arctic species at 24h (Supplementary tables 9-13). Among the 119 uniquely shared Arctic DEGs, 109 shared similar expression patterns in all species (74 upregulated and 35 downregulated; Supplementary figure 2). The upregulated DEGs might be especially important in acquiring freezing tolerance, and included many transcription factors (15, cf. above), but also genes annotated with e.g., response to karrikin (7), circadian rhythm (6), flavonoid biosynthetic process (5), and proline transport (4; Supplementary table 9). Other upregulated DEGs shared by the Arctic species included e.g., *MAPK7*, *MAPKKK14*, *MAPKKK18*, *SUCROSE SYNTHASE 3*, *RAB18*, and a Late Embryogenesis Abundant gene (LEA; Supplementary table 9). Several of

the uniquely shared DEGs could be found in the heatmaps of the top 30 differentially expressed genes with the lowest false discovery rate (Figure 3).

3.2.3 Shared cold regulated genes among Arctic species and *A. thaliana*

Among the 195 DEGs shared by Arctic species and *A. thaliana*, 188 shared similar expression patterns in all species (Supplementary table 10). Among the 122 shared upregulated DEGs (Supplementary figure 2) were several genes annotated with e.g., response to cold (23), response to salt stress (20), response to abscisic acid (16), response to water deprivation (14), and flavonoid biosynthetic process (13; (Supplementary table 10). The shared gene list included many known cold induced genes like *LEA14*, *COR78*, *COR15B*, *TCF1*, *COR27* and *COR28* (Supplementary table 10). There were also several shared genes related to Absciscic acid and ethylene regulated pathways, like *ABF1*, *AITR5*, *ERT2*, and *ERF43* (Supplementary table 10). Several of the DEGs shared between *A. thaliana* and the three Arctic species could also be found in the heatmaps of the top 30 differentially expressed genes with the lowest false discovery rate (Figure 3).

3.3 Functional characterization of DEG sets (GO-enrichment)

Most significantly enriched GO-terms within the BP, CC, and MF domains were species-specific (Table 4, Supplementary tables 14-22). However, as many as 27 of the same GO-terms (BP and MF) were significantly enriched in the Arctic upregulated DEG sets at 24h (Supplementary table 23). Most of these (22) were also found in *A. thaliana*. Among the GO-terms shared by *A. thaliana* and the Arctic species were many BP terms like “response to cold”, and “response to heat” (Figure 4-5). Other shared BP terms included those associated with cold and freezing like “response to salt stress”, “response to water deprivation”, “response to oxidative stress”, and “response to cadmium ion” (Figure 5). We also observed a shared overrepresentation of genes associated with the hormones abscisic acid, and ethylene, and the substance karrikin, as well as genes associated with possible cryoprotectants like flavonoids and sucrose (BP, Figure 5). Among the GO-terms uniquely found in Arctic 24h upregulated DEG sets were e.g., “spermidine biosynthetic process” (BP) and “arginine decarboxylase activity” (MF). The 24h downregulated DEGs were enriched for genes associated with growth related GO-terms such as “phototropism” (BP; all species) and “auxin-activated signaling pathway” (BP; only Arctic species)

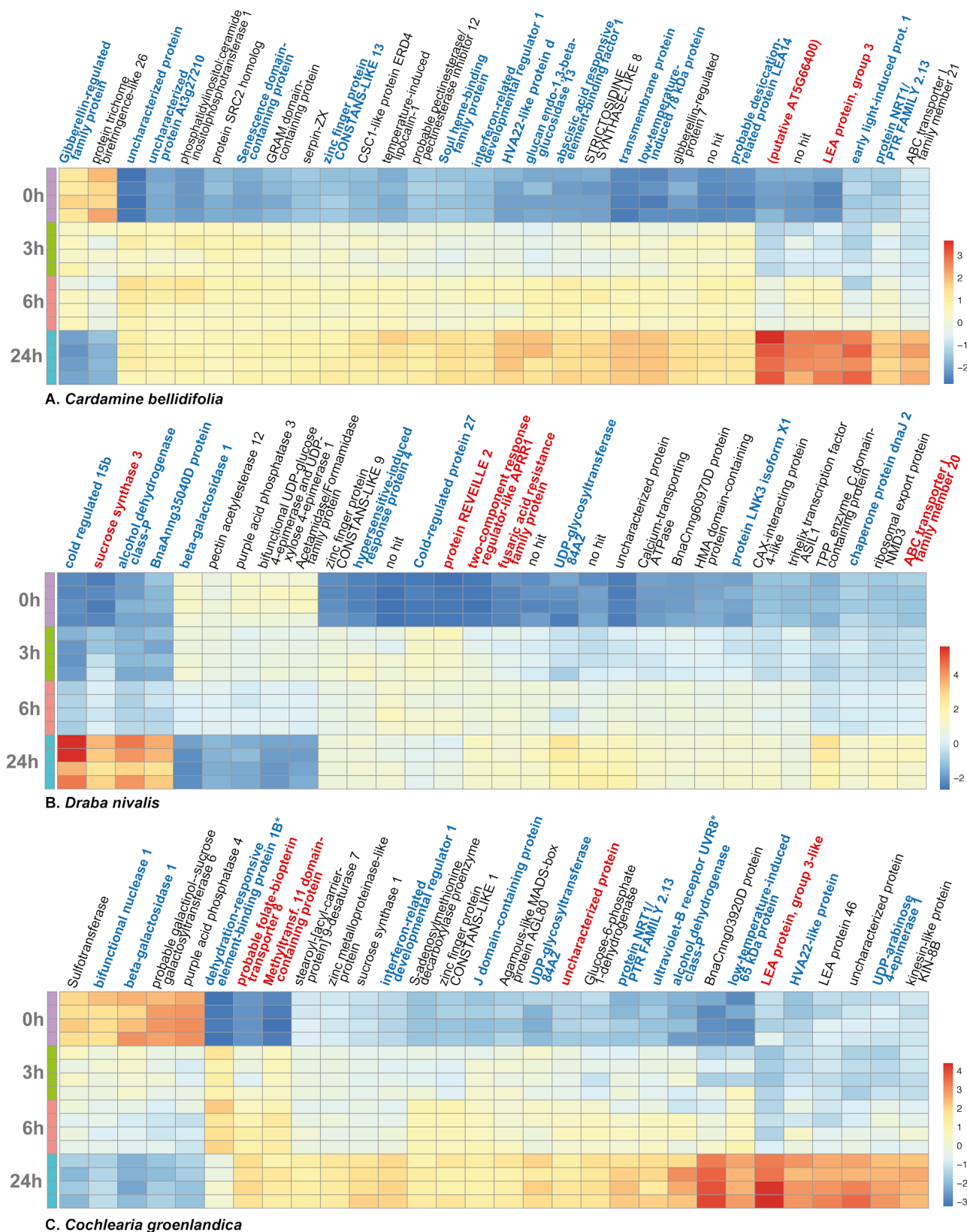


Figure 3. Heatmaps of the most significantly differentially expressed genes (DEGs): A) *Cardamine bellidifolia*, B) *Draba nivalis*, and C) *Cochlearia groenlandica*. Color scale = log₂ transformed counts. Each row corresponds to a replicate, and there are four replicates at each time point (0h, 3h, 6h, 24h). Gene names in **bold and blue** = Found as 24h DEG in *Arabidopsis thaliana* and all Arctic species (based on *A. thaliana* homologs), gene names in **bold and red** = Found as 24h DEG only in Arctic species (based on *A. thaliana* homologs). *dehydration-responsive element-binding protein 1B = DREB1b/CBF1, ultraviolet-B receptor UVR8 = TCF1.

Table 4. Numbers of significantly enriched GO-terms ($p < 0.05$) in differentially expressed genes (DEGs) after 3h, 6h and 24h at 2 °C when applying the elim algorithm.

Abbreviations: ↑ = Upregulated DEG set, ↓ = Downregulated DEG set, Biological Process = BP, Cellular Component = CC, Molecular Function = MF domains.

		<i>C. bellidifolia</i>	<i>C. groenlandica</i>	<i>D. nivalis</i>	Shared*
3h ✱	↑	109 BP, 12 CC, 40 MF	101 BP, 26 CC, 42 MF	102 BP, 4 CC, 38 MF	11 BP, 1 CC, 3 MF
	↓	44 BP, 7 CC, 26 MF	61 BP, 14 CC, 34 MF	101 BP, 1 CC, 44 MF	0 BP, 0 CC, 0 MF
6h ✱	↑	127 BP, 13 CC, 33 MF	108 BP, 5 CC, 37 MF	120 BP, 9 CC, 44 MF	16 BP, 1 CC, 4 MF
	↓	51 BP, 4 CC, 25 MF	73 BP, 8 CC, 37 MF	121 BP, 8 CC, 45 MF	1 BP, 0 CC, 0 MF
24h ✱	↑	156 BP, 22 CC, 62 MF	172 BP, 30 CC, 76 MF	139 BP, 24 CC, 82 MF	22 BP, 0 CC, 5 MF
	↓	138 BP, 10 CC, 78 MF	108 BP, 11 CC, 60 MF	136 BP, 25 CC, 62 MF	7 BP, 4 CC, 0 MF

*Also shared with *A. thaliana* after 24h: ↑20 BP, 0 CC, 2 MF, and ↓ 3 BP, 3 CC, 0 MF.

3.4 Note on positively selected/convergent cold-responsive genes

We identified several positively selected and convergent genes in the 24h DEG sets of the Arctic species (Table 5-6; Supplementary tables 24-25; (Birkeland et al. 2020).

Upregulated positively selected genes (PSGs) included genes such as *COR15B* and *CSDP1* in *D. nivalis*, *LEA4-5* in *C. groenlandica*, and a highly upregulated transmembrane protein (putative homolog of *A. thaliana* AT1G16850) in *C. bellidifolia*. We found that a gene with convergent substitutions in all Arctic species, *EMB2742*, was upregulated in all species (and in *A. thaliana*). Some convergent genes showed different expression patterns depending on the species. For instance, *CAT2* was downregulated in *D. nivalis* and upregulated in *C. groenlandica*. This gene has previously been found to contain convergent substitutions in *D. nivalis* and *C. groenlandica*, and to be under positive selection in *D. nivalis*. The low temperature responsive transcription factor *RAV1* has previously been found to be under positive selection in *C. groenlandica*, but was not differentially expressed in this species. We also note that *MAPKKK14*, (previously found to contain convergent substitutions in

the Arctic species) had blast hits in several different MAPKKs in this study. Its status as a convergent gene is therefore uncertain.

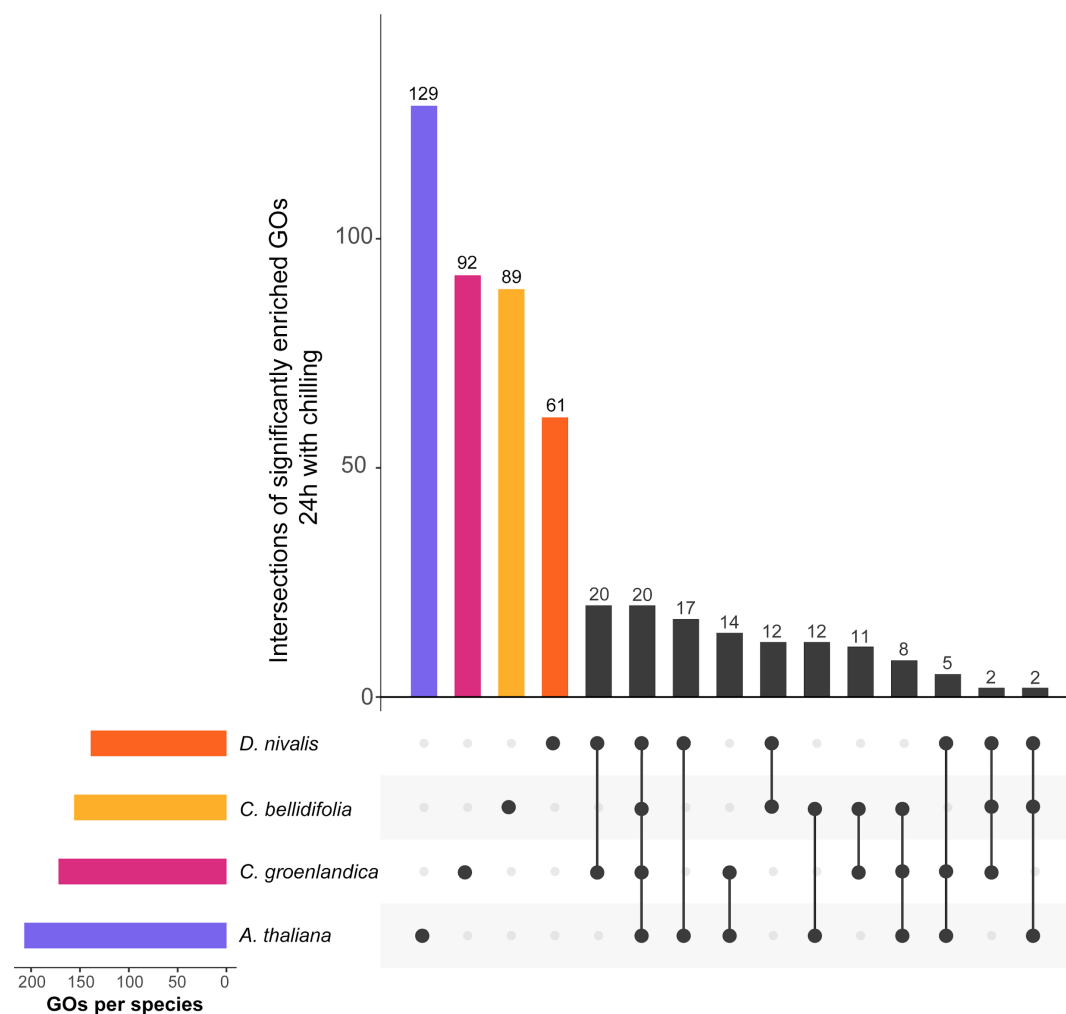


Figure 4. UpSet plot of overrepresented GO-terms in the 24h upregulated gene sets (Biological Process domain only). The plot in the left corner shows total numbers of significantly enriched GO-terms, and the main plot shows the number of unique significantly enriched GO-terms, followed by GO-term intersections/overlaps between species (connected dots).

Table 5. 24h DEGs under positive selection based on Birkeland et al. (2020).

PSG = Positively selected gene

	Total number of PSGs Birkeland et al. (2020)	PSGs among all DEGs, 24h	PSGs among upregulated DEGs, 24h	PSGs among downregulated DEGs, 24h
<i>C. bellidifolia</i>	201	26	14	12
<i>C. groenlandica</i>	159	23	12	11
<i>D. nivalis</i>	360	46	24	22

Table 6. 24h DEGs with convergent substitutions based on Birkeland et al. (2020).

Cb = *Cardamine bellidifolia* 24h DEG set, Cg = *Cochlearia groenlandica* 24h DEG set,
Dn = *Draba nivalis* 24h DEG set

	Total number of convergent genes Birkeland et al. (2020)	Convergent genes among all DEGs, 24h	Convergent genes among upregulated DEGs, 24h	Convergent genes among downregulated DEGs, 24h
<i>C. bellidifolia</i> , <i>C. groenlandica</i>	58	17 Cb 14 Cg	9 Cb 12 Cg	8 Cb 2 Cg
<i>C. groenlandica</i> , <i>D. nivalis</i>	33	8 Cg 8 Dn	5 Cg 2 Dn	3 Cg 6 Dn
<i>D. nivalis</i> , <i>C. bellidifolia</i>	126	19 Dn 20 Cb	13 Dn 17 Cb	6 Dn 3 Cb

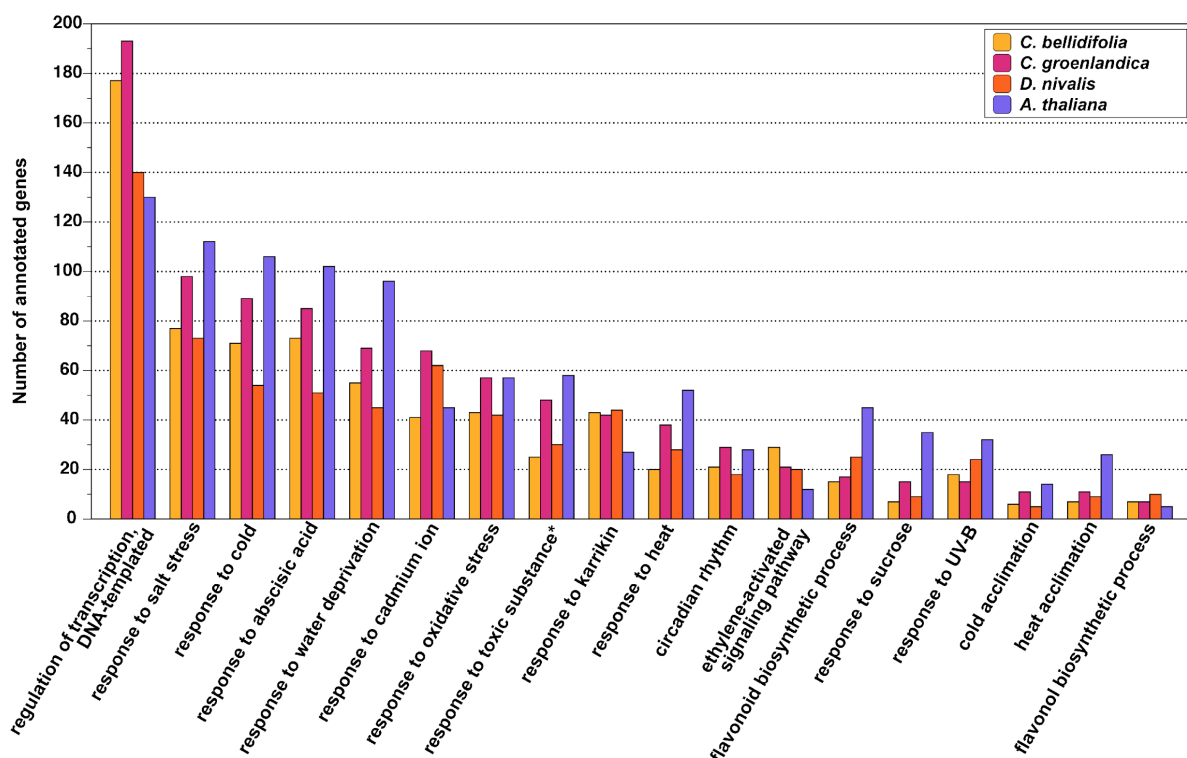


Figure 5. Barchart of genes annotated with Gene Ontology (GO) terms found to be significantly enriched in the 24h upregulated gene set of the three Arctic species. There were 22 overlapping GO-terms among the Arctic 24h upregulated DEG sets, and 20 of these overlapped with *A. thaliana*. Only terms with at least 10 annotated genes are shown. *Not significantly enriched in *A. thaliana*.

3.5 Gene co-expression modules

The gene co-expression network analyses resulted in 13 co-expressed modules in *C. bellidifolia*, 23 co-expressed modules in *C. groenlandica* and 14 co-expressed modules in *D. nivalis* (after module merging; Figure 6; Supplementary figure 3). At least one module in each species stood out as highly positively correlated with the binary measure of cold temperature (i.e., Pearson correlation coefficient > 0.90 and $p < 0.001$), but often several modules in each species were positively correlated with this measure, and with 3h, 6h and 24h of cold.

Focusing on the highest positively correlated module in relation to the binary measure of cold in each species (lightyellow in *C. bellidifolia*, lightcyan in *C. groenlandica*, and darkorange2 in *D. nivalis*, Figure 6), we found that 16 significantly enriched GO-terms overlapped between the modules in question (domain BP, CC, MF; Table 7). However, most GO-term enrichment results were module or species-specific (Supplementary tables 26-34). The three cold correlated modules had slightly different hub-genes (genes with high kME), but some general patterns were found (Supplementary tables 35-37): *MAPK7* (a gene found only in the 24h DEG sets of all Arctic species) was among the hub-genes of *C. bellidifolia* and *D. nivalis*, *CONSTANS-like 4* was among the hub genes of *C. groenlandica* and *D. nivalis*, and *CIPK4* had high kME in all species and was a hub gene in *D. nivalis*. In addition, several related genes like *REVEILLE1*, 2 and 6, and *CONSTANS-like 4*, 9, 10 and 13 had high kME in all species, but the exact genes differed from species to species. We also noticed several well-known cold regulated genes among the hub genes in these modules like, for instance, *COR78* and *COR47* in the lightyellow module of *C. bellidifolia*, *CBF1* in the lightcyan module of *C. groenlandica* and *COR27* in the darkorange2 module of *D. nivalis*. (Note that the genes orchestrating each module may have different expression patterns and that it is possible that they do not cluster with the module they are regulating).

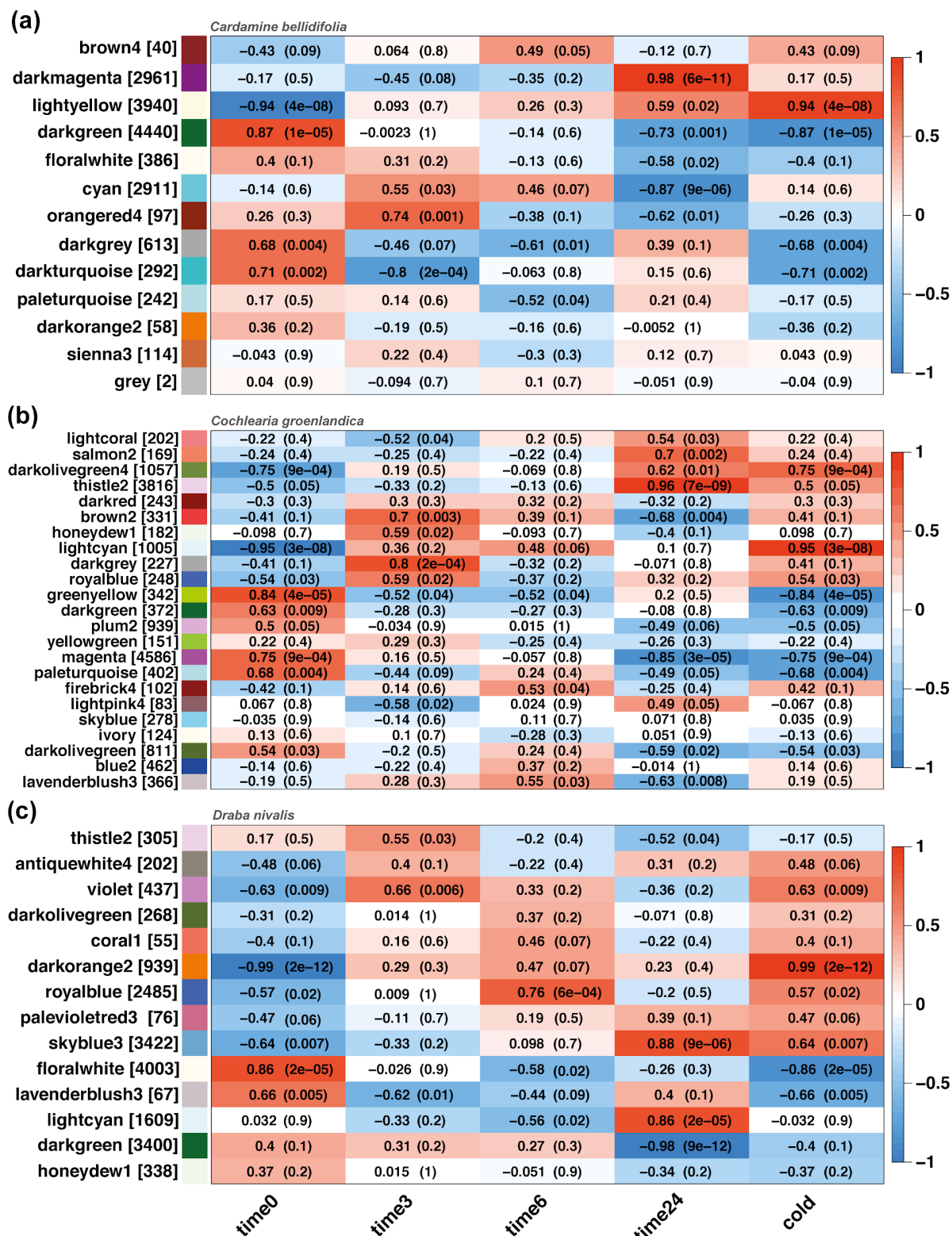


Figure 6. Heatmaps showing Pearson correlation between each co-expression module (module eigengenes) and temperature: 0h, 3h, 6h, and 24h with cold treatment, and a binary measure of cold (cold treatment / no cold treatment). Each row corresponds to a co-expression module and each column to a temperature trait. The number of genes in each module is indicated in brackets after the module name. Cells show the correlation coefficient and corresponding p-value (in parentheses). From the top: a) *Cardamine bellidifolia*, b) *Cochlearia groenlandica* and c) *Draba nivalis*.

Table 7. Gene Ontology terms that were overrepresented in the lightyellow (*C. bellidifolia*) lightcyan (*C. groenlandica*) and darkorange2 (*D. nivalis*) co-expression modules. These modules were the most positively correlated with the binary measure of cold in each species.

GO Identifier	GO Term Name	Domain
GO:0080167	response to karrikin	BP
GO:0007623	circadian rhythm	BP
GO:0009737	response to abscisic acid	BP
GO:0006355	regulation of transcription, DNA-templated	BP
GO:0009409	response to cold	BP
GO:0009719	response to endogenous stimulus	BP
GO:0009753	response to jasmonic acid	BP
GO:0097305	response to alcohol	BP
GO:0010017	red or far-red light signaling pathway	BP
GO:0009414	response to water deprivation	BP
GO:0009651	response to salt stress	BP
GO:0005634	nucleus	CC
GO:0003677	DNA binding	MF
GO:0005509	calcium ion binding	MF
GO:0043565	sequence-specific DNA binding	MF
GO:0003700	DNA-binding transcription factor activity	MF

4. Discussion

4.1 The cold response of Arctic Brassicaceae is highly species-specific

Our main finding was that the cold response of the three Arctic species, *C. bellidifolia*, *C. groenlandica*, and *D. nivalis*, is highly species-specific. Not only were most DEGs unique for each species, but the number of DEGs shared by the three Arctic species and *A. thaliana* were higher than the number shared by the three Arctic species alone. This suggests that evolution of cold response in Arctic Brassicaceae followed independent genetic trajectories, but with some conserved components. These results are in concordance with our previous study of protein sequence evolution in the same three species, where we found very little overlap in positively selected genes among species, and also very few genes with convergent substitutions (Birkeland et al. 2020). Although we expected a higher degree of convergence in the cold induced expression profiles, the results agree well with those of a similar study of cold acclimation in the temperate grass subfamily Pooideae (Schubert et al. 2019). Schubert et al. (2019) found that phylogenetically diverse species of grasses showed widespread species-specific transcriptomic responses to low temperatures, but with some conserved aspects.

The independent evolutionary trajectories of cold response might be tied to the polygenic nature of the trait, involving thousands of genes. Highly polygenic traits have high genetic redundancy and should show less repeatable patterns of adaptation compared to traits involving only a few genes (Yeaman 2015; Yeaman et al. 2016; Barghi et al. 2019). Low levels of repeatability have also been found in other polygenic traits, such as drought tolerance. Marín-de la Rosa et al. (2019) recently showed that divergent strategies underlie drought resistance in closely related Brassicaceae species. However, there are also studies demonstrating a high degree of convergence in polygenic traits, like Yeaman et al. (2016) showing convergent local adaptation to climate in two distantly related conifers, and Yang et al. (2018) showing convergent evolution of flowering time in *Capsella rubella* due to independent deletions at the same locus (flowering time is controlled by >60 genes in *A. thaliana*; Anderson, Willis, and Mitchell-Olds 2011). Although it is not known what caused evolutionary repeatability in these exact examples, adaptation may end up taking the same routes in the presence of pleiotropic constraints or restricted available standing

genetic variation (as also discussed in Gould and Stinchcombe 2017). Since, we find that cold tolerance mainly evolved independently in the three species, we can assume that such constraints have been of less importance. Following, there seem to be many ways to evolve cold tolerance, or at least many ways to build upon cold tolerance once a basis for cold tolerance has evolved.

4.2 Conserved aspects of the Arctic Brassicaceae cold response

Another major finding in our data was that the Arctic species seem to have more in common with the temperate relative *A. thaliana*, than they have with each other. This shared aspect of the cold response may represent conserved parts of the CBF pathway, which also have been found in other plant lineages (Jaglo et al. 2001; Shi, Ding, and Yang 2018). Shared genes included, for instance, those that were both tied to circadian regulation and regulation of freezing tolerance (i.e. COR27 and COR28; X. Li et al. 2016). This fits well with the CBF pathway being gated by the circadian clock (Dong, Farré, and Thomashow 2011; Fowler, Cook, and Thomashow 2005). The Arctic cold response also seemed to follow similar trends as in *A. thaliana*, with a continuous increase in the number of DEGs in response to cold temperatures (Kilian et al. 2007; Calixto et al. 2018). This indicates that Arctic plants respond to low temperatures in a similar way as temperate plants, and that they are not completely hard-wired to tolerate low temperatures.

The transcriptomic changes also triggered sets of biological processes similar to the cold response of *A. thaliana*, and that of other plant species within and outside the Brassicaceae family (e.g. Zhao et al. 2012; Lee et al. 2013; Buti et al. 2018). First and foremost, this involved stress responses associated with low temperatures and freezing. The upregulation of genes associated with salt stress, water deprivation, and cadmium ion (a heavy metal highly toxic to plants; di Toppi and Gabbrielli 1999) could be linked to cold and freezing stress in two different ways. First, ice formation in the apoplast will draw water out of the cells and increase the concentration of salts and toxic solutes (Wisniewski and Fuller 1999; Körner 2003). This may lead to severe cell dehydration (Shi, Ding, and Yang 2018). Second, stress response pathways associated with cold have been shown to be partially homologous with those of dehydration and salt tolerance (Bartels and Souer 2003; Shamustakimova et al. 2017; Ingram and Bartels 1996). We also found that genes annotated with “response to oxidative stress” were overrepresented among the upregulated genes. This fits well

with oxidative stress accompanying other abiotic stresses like cold, and especially high light intensity in combination with low temperatures (Heino and Palva 2004; Lütz 2010; Kilian et al. 2007).

Genes associated with the hormones abscisic acid (ABA) and ethylene were upregulated in response to cold in all species. ABA is an important hormone in plant stress signalling and has previously been shown to increase in abundance during cold acclimation (Heino and Palva 2004; Tuteja 2007). Exogenous application of ABA will even increase freezing tolerance in *A. thaliana* and other plants (Thomashow 1999). The hormone ethylene is similarly reported to influence freezing tolerance and regulates the CBF pathway itself (Kazan 2015). In most plant species, this entails a positive regulation of freezing tolerance (like e.g. in tomato and tobacco; Zhang and Huang 2010). However, in *A. thaliana* ethylene appears to also negatively regulate freezing tolerance depending on the growing conditions (Kazan 2015). This hormone's role in the Brassicaceae cold response is therefore still uncertain.

One way that plants increase their freezing tolerance is by accumulating compounds that in various ways hinder ice nucleation, or alleviate the effects of ice formation by protecting plant tissues against freezing damage (i.e. cryoprotectants; Ruelland et al. 2009). We observed that genes associated with possible cryoprotectants like flavonoids (especially flavonol) and sucrose were upregulated in response to cold in all species. Sucrose has been reported to lower the freezing point in *A. thaliana* (i.e. leading to supercooling and avoidance of freezing; Reyes-Díaz et al. 2006), to diminish the water potential between the apoplastic space and the cell in the face of ice formation (osmotic adjustment; Ruelland et al. 2009), and even to regulate cold acclimation itself (Rekarte-Cowie et al. 2008). Similarly, accumulation of flavonoids is associated with cold acclimation and higher freezing tolerance in *A. thaliana* (Schulz et al. 2016), and flavonol might have a role in protecting cell membranes during freezing stress (Korn et al. 2008). Finally, we also observed another compound that seems to be important in acquiring cold/freezing tolerance in all species. Karrikin is a compound that has received little attention in relation to cold stress, but a recent study suggests that karrikin inhibits germination under unfavourable conditions, and also improves plant vigour in the face of abiotic stress through regulation of redox homeostasis (Shah et al. 2020). Genes responsive to karrikin have also been documented to be important in the cold response of *Chorispora bungeana* (a subnival Brassicaceae; Zhao et al. 2012). The fact that

flavonoids, sucrose and karrikin seem to have important roles in the cold response of four distantly related Brassicaceae species, highlights their importance in achieving cold tolerance.

These shared trends indicate that cold response in Brassicaceae is built upon a similar scaffold and support the claim that cold tolerance is a complex trait that is difficult to evolve (Donoghue 2008). It is highly likely that the last common ancestor of the three Arctic species had some kind of cold tolerance machinery as the major clades of the Brassicaceae are thought to have radiated in response to a colder and drier climate (~33.9 Ma in the Eocene-Oligocene transition; Huang et al. 2016). Perhaps the basis of the Arctic cold response could also have contributed to the Brassicaceae family's success in cold and dry environments. However, the highly species-specific cold responses found in *C. bellidifolia*, *C. groenlandica*, and *D. nivalis* also indicate that there is great evolutionary flexibility in cold coping strategies once there is a basis to build upon.

4.3 The Arctic cold response - is there such a thing?

The low degree of overlap in the cold response of Arctic Brassicaceae provokes the question if there exists anything like a true Arctic cold response. There are, however, a few shared genes and characteristics that stand out as potentially unique for the three Arctic species. For instance, among the ~100 genes shared only by the three Arctic species, there are particularly many transcription factors that potentially could have large effects. Yeaman et al. (2016) similarly found that convergent genes in adaptation to climate in distantly related conifers were enriched for transcription factors. Although we can only speculate about their function in this study, several are related to transcription factors that have important roles in cold tolerance in other plant species. Examples included e.g. *CBF4* (mainly activated by drought in *A. thaliana*; Haake et al. 2002), *RAP2.10*, and *RAP2.2*. Considering the important role, the related *CBF1-3* has in regulating the cold response, *CBF4* stands out as a potential important candidate for the regulation of an Arctic cold response. Furthermore, this gene has also been found to be cold responsive in other species (Cai et al. 2019). Two other examples, *RAP2.10*, and *RAP2.2*, are related to *DEAR1*; a transcription factor known to mediate freezing stress responses (Tsutsui et al. 2009). In addition, several transcription factors related to circadian rhythm like e.g., *REVEILLE 2*, *PCL1* and *HY5* are also among the DEGs shared by all Arctic species. One possibility is that the

extreme Arctic light regime might have triggered the evolution of links between the CBF pathway and the circadian clock that are not found in temperate environments. We cannot completely rule out that certain aspects of the experimental light regime may have provoked the expression of these genes, but the interplay between the circadian clock and the CBF pathway is important for balancing freezing tolerance and plant growth (e.g. Dong, Farré, and Thomashow 2011; Shi, Ding, and Yang 2018).

Another interesting finding among the shared genes is the potential traces of a mitogen activated protein kinase (MAPK) cascade. Such cascades are known to be important in regulating the CBF pathway (Teige et al. 2004; Shi, Ding, and Yang 2018). Intriguingly, we found one MAPK (*MAPK7*), as well as two MAPKKs (*MAPKK14* and *MAPKK18*) that are uniquely upregulated in the three Arctic species. In *A. thaliana*, it has been shown that the *MKK2* pathway regulates cold stress signaling (Teige et al. 2004), and that *MAPK6* is involved in releasing inhibitory effects on CBF gene expression (Kim et al. 2017). Accordingly, the shared MAPK and MAPKKs could potentially have important roles in a putative Arctic cold response.

Functionally, “spermidine biosynthetic process” stands out as having a special status in the Arctic species, although the significant overrepresentation was caused by only a few cold induced genes. One of these genes was arginine decarboxylase gene 1 (*ADC1*), which was tied to both the shared significant enrichment of “spermidine biosynthetic process” and “arginine decarboxylase activity” in Arctic species. This gene has previously been found to be involved in acquiring freezing tolerance in wild potato (Kou et al. 2018). It has been shown that cucumber plants pretreated with spermidine (a polyamine) show less decline in photosynthesis rates during chilling than non-treated plants (He, Nada, and Tachibana 2002). Maintaining photosynthesis rates during low temperatures should be especially important for Arctic plants as temperatures are low year-round, and accordingly, they also have optimum photosynthesis rates at lower temperatures than other plants (Chapin 1983). In addition, there was an overrepresentation of shared genes related to proline transport (an amino acid with a similar role in freezing tolerance; Ruelland et al. 2009), as well as some of the abovementioned cryoprotectants, like flavonoid biosynthesis and sucrose.

Finally, we also found other potential traces of adaptation to extreme Arctic temperatures. Some of the genes that we previously had found to be under positive

selection (Birkeland et al. 2020), are upregulated in the same lineages in response to cold. This includes some of the top significantly differentially expressed genes, like e.g., *COR15B* in *D. nivalis*, a transmembrane protein in *C. bellidifolia*, and *LEA46* in *C. groenlandica*. Furthermore, *EMB2742*, which has convergent substitutions in all three species (Birkeland et al. 2020), is significantly upregulated after 24h of cold, possibly indicating that it has a unique Arctic function. As the findings in Birkeland et al. (2020) were not based on cold induced transcriptomes (just “untreated” transcriptomes), there could also be other positively selected genes among the cold induced DEGs that have not been detected. That important cold regulated genes are under positive selection, shows that the Arctic climate may have imposed strong selection pressure on the cold response of Arctic plants.

Despite a few potentially important overlaps in cold response, it is apparent that there probably is not a single, but many ways Arctic plants respond to low temperatures. Considering the polygenic nature of the trait, and that the three species likely descend from different temperate relatives, it would also be surprising if their cold response had converged into something uniquely Arctic. To truly understand what has made each of the three Arctic species successful in the extreme Arctic climate, a better understanding of gene functions in cold response in these species is needed. A good place to start could be to study the effects of shared transcription factors on cold and freezing tolerance.

4.4 Limitations and future perspectives

We note that there might be methodological differences between our study and the study on *A. thaliana* (Park et al. 2015) that may affect the number of DEGs considered as significant in each study. This may potentially have a small effect on the number of shared genes, both between *A. thaliana* and the Arctic species, and between the Arctic species alone. An important difference is, for instance, that our study is based on *de novo* assembled transcriptomes, which may have small inaccuracies in the delimitation of genes (e.g., isoforms of the same gene mistakenly being delineated as different genes). However, we also used a stringed filtering scheme to reduce inflated transcript numbers, and such differences should not affect the overall results.

This experiment represents one of the first snapshots of Arctic cold-induced transcriptomes, but more studies are needed to understand how Arctic plant species

can cope with the low summer temperatures at higher latitudes. Future studies on Arctic cold tolerance could delve deeper into performance under long-term cold stress, or performance under cold stress in combination with high light (typical of Arctic environments).

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