CAM evolution is associated with gene family expansion in an explosive bromeliad radiation

Short title: CAM is associated with gene family expansion

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1.   Abstract

The subgenus *Tillandsia* (Bromeliaceae) belongs to one of the fastest radiating clades in the plant kingdom and is characterised by the repeated evolution of Crassulacean Acid Metabolism (CAM). Despite its complex genetic basis, this water-conserving trait has evolved independently across many plant families and is regarded as a key innovation trait and driver of ecological diversification in Bromeliaceae. By producing high quality genome assemblies of a *Tillandsia* species pair displaying divergent photosynthetic phenotypes, and combining genome-wide investigations of synteny, TE dynamics, sequence evolution, gene family evolution and temporal differential expression, we were able to pinpoint the genomic drivers of CAM evolution in *Tillandsia*. Several large-scale rearrangements associated with karyotype changes between the two genomes and a highly dynamic TE landscape shaped the genomes of *Tillandsia*. However, our analyses show that rewiring of photosynthetic metabolism is mainly obtained through regulatory evolution rather than coding sequence evolution, as CAM-related genes are differentially expressed across a 24-hour cycle between the two species, but are no candidates of positive selection. Gene orthology analyses reveal that CAM-related gene families manifesting differential expression underwent accelerated gene family expansion in the constitutive CAM species, further supporting the view of gene family evolution as a driver of CAM evolution.

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2. Introduction

Crassulacean Acid Metabolism (CAM) is a photosynthetic phenotype playing a major role in plant adaptation to arid environments and the epiphytic lifeform (Winter and Smith, 2012; Cushman, 2001; Silvera et al., 2010), and has been described as a key innovation trait driving plant diversification and speciation in several plant lineages (Quezada and Gianoli, 2011; Silvera et al., 2009; Ogburn and Edwards, 2009). CAM functions as a carbon concentrating mechanism by assimilating CO_2 overnight and storing it as malate in the vacuole, which greatly enhances the efficiency of Rubisco, the first enzyme of the Calvin cycle (Osmond, 1978). This also has the secondary effect of improving the plant's overall water use efficiency by reducing evapotranspiration, as stomata can remain closed during the day (Borland et al., 2014). Though often presented as a discrete trait, CAM actually encompasses a large spectrum of photosynthetic phenotypes including intermediate and facultative forms (Edwards, 2023). Phenotypes from this CAM continuum have evolved repeatedly in at least 37 plant families (Winter et al., 2021), yet the underlying evolutionary mechanisms allowing this complex and diverse trait to emerge multiple times throughout plant history are not fully understood.

Due to the sparse availability of CAM plant genomes, the majority of studies on CAM evolution have focussed on transcription levels and sequence evolution to understand its underlying genetic drivers. However, novel variation can be generated by other mechanisms which have not been investigated thoroughly in the context of CAM evolution. For example, several studies have pointed at the potential importance of gene family expansion as a driver of CAM evolution (Cai et al., 2015; Silvera et al., 2014). In C4 plants, duplicated gene copies tend to be more often retained compared to closely related C3 lineages (Hoang et al., 2023). Gene duplication occurs at higher rates than point mutation in many lineages (Katju and Bergthorsson, 2013) and can lead to novel functional variation through dosage effects, neo-functionalization, or subfunctionalization (Ohno, 1970), as observed in teleost fish (Arnegard et al., 2010; Moriyama et al., 2016) and orchids (Mondragón-Palomino and Theissen, 2009). Another form of structural variation that can contribute to the evolution of complex traits is TE insertion in and around genes, which has been shown to play a role in local adaptation, for example in *Arabidopsis* (Baduel et al., 2019). Large-scale rearrangements such as chromosomal fusions, inversions or translocations can increase linkage between co-adapted alleles and generate reproductive barriers (Luo et al., 2018; Lowry and Willis, 2010).

The adaptive radiation of *Tillandsia* subgenus *Tillandsia* (Bromeliaceae) is part of one of

the fastest diversifying clades known in the plant kingdom (Tillandsioideae) (Givnish et al., 2014)

and is characterised by a number of key innovation traits such as epiphytism, absorptive trichomes,

water-impounding tanks and photosynthetic metabolism driving extraordinary diversity both on

the taxonomic and ecological level (Barfuss et al., 2016). The group displays a broad range of

phenotypes of the CAM continuum, resulting in repeated evolution of constitutive CAM (Crayn

et al., 2015; De La Harpe et al., 2020). CAM evolution has been described as an ecological driver

of diversification in the subgenus *Tillandsia* (Barfuss et al., 2016; Crayn et al., 2004), and across

Bromeliaceae in general (Givnish et al., 2014; Crayn et al., 2004; Benzing and Bennett, 2000).

This renders the radiation a fascinating system both for studies on speciation and rapid adaptation

generally, and for studies on CAM evolution specifically, as comparative investigations between

recently diverged species with contrasting phenotypes prevent the overestimation of evolutionary

changes needed to evolve adaptations such as CAM (Heyduk et al., 2019a).

While Bromeliaceae is generally regarded as a homoploid radiation with conserved

chromosome counts and little genome size variation (Gitaí et al., 2014), more recent work has

pointed at a high "genomic potential" of the subgenus *Tillandsia*, notably from elevated gene loss

and duplication rates (De La Harpe et al., 2020), providing an exemplary system to study the role

of genome evolution and structural variation in CAM evolution. Not only are adaptive radiations

like *Tillandsia* characterised by repeated evolution of key innovation traits, the short timescales at

which novel variation arises in these systems challenges classical views of adaptive evolution,

stimulating a range of studies pointing at the potential importance of genome evolution as a

genomic driver of diversification (McGee et al., 2020; Brawand et al., 2015; Cicconardi et al.,

2021).

In this study, we comparatively investigated *de novo* assembled genomes of two

ecologically divergent members of the subgenus *Tillandsia* to further our understanding of genome

evolution in this recent radiation and its link to CAM evolution as a key innovation trait. *Tillandsia*

fasciculata (Fig. 1a) displays a set of phenotypes typically described as "grey" or "atmospheric"

Tillandsia (Benzing and Bennett, 2000): a dense layer of absorptive trichomes, CAM

photosynthesis and occurrence in arid places with high solar incidence and low rainfall. On the

other hand, *T. leiboldiana* (Fig. 1b) represents a typical "green" *Tillandsia* displaying tank

formation, C3-like leaf morphology, a lack of absorptive trichomes and occurrences in cooler,

wetter regions. While not sister species, *T. fasciculata* and *T. leiboldiana* belong to sister clades

displaying a shift in photosynthetic metabolism (Fig. 1c), and represent phenotypic extremes

within subgenus *Tillandsia*. Their photosynthetic metabolisms have been described with carbon

isotope ratios as strong CAM for *T. fasciculata* (carbon isotope ratio: $\delta^{13}C$ = -11.9 / -16.1, (Crayn

et al., 2015; De La Harpe et al., 2020)) and C3 for *T. leiboldiana* ($\delta^{13}C$ = -28.0 / -31.3; (Crayn et

al., 2015; De La Harpe et al., 2020), and are some of the most distinct values reported for the

subgenus on the CAM continuum.

However, due to the limited ability of carbon isotope measurements in capturing

intermediate CAM phenotypes and therefore representing the full CAM continuum (Messerschmid

et al., 2021; Pierce et al., 2002), the exact photosynthetic phenotypes of *T. leiboldiana* and *T.*

fasciculata need to be corroborated to fully understand what range of the CAM continuum is truly

encompassed in the subgenus. By characterising the photosynthetic metabolisms of *T. leiboldiana*

and *T. fasciculata* and investigating genomic variation between these two species on multiple

levels, from karyotype, chromosomal rearrangements, to molecular evolution, gene family

evolution and temporal differential gene expression, we thoroughly explore the degree of genomic

divergence found within this radiation and the link between this variation and the evolution of a

key innovation trait. We ascertain that the photosynthetic metabolisms of these species are clearly

distinct, with *T. fasciculata* at the late stages of CAM evolution (i.e., constitutive, strong CAM),

and *T. leiboldiana* likely at the very early stages (i.e., no night-time malate accumulation, but

CAM-like expression profiles of certain enzymes). We further document karyotype differences,

multiple chromosomal rearrangements, distinct TE landscapes and gene family evolution rates

between the two species. We further find evidence that molecular variation underlying the

difference in phenotype is largely found at the transcriptomic level. We also found a clear

association between CAM-related temporal gene expression differences and both gene family

expansion in the constitutive CAM plant and pre-existing gene duplications shared between both

species.

Figure 1: a) Tillandsia fasciculata*, a "grey" or "atmospheric"* Tillandsia *with a dense layer of trichomes, carbon isotope values within the CAM photosynthesis range, a lack of water-impounding tank and roots adapted to the epiphytic lifestyle.* **b)** Tillandsia leiboldiana*, a green* Tillandsia *with C3-like leaf morphology and carbon isotope values, an impounding tank and reduced trichome layer.* **c)** *Schematic representation of the evolutionary relationship between the two investigated species of* Tillandsia *within the subgenus. Colours indicate reported carbon isotope values* (De La Harpe et al., 2020; Crayn et al., 2015)*. The average was taken when multiple values have been reported for the same species. Pie charts at internal nodes show the ancestral state of photosynthetic metabolism as reported in De La Harpe et al. 2020. WHZ stands for Winter-Holtum Zone* (Males, 2018) *and represents intermediate forms of the* CAM *continuum.*

3. Results

3.1. Photosynthetic phenotypes of *T. fasciculata* and *T. leiboldiana*

To better understand the difference in photosynthetic metabolism between *T. fasciculata* and *T. leiboldiana*, we measured metabolite abundances with GC-MS for six samples per species at six timepoints across a 24-hour cycle. The overall composition of 49 metabolite abundances separate samples of both species within the first two principal components (Fig. 2a), which combined explain 41.5 % of variance. This suggests a pronounced general metabolic differentiation between *T. fasciculata* and *T. leiboldiana*. Besides amino acids, the organic acids malate, citrate and gluconic acid contribute most to this differentiation along PC1 (Fig. S1), which is a common pattern for species with diverging photosynthetic metabolism (De La Harpe et al., 2020; Benzing and Bennett, 2000; Popp et al., 2003). Sugars appear large contributors to differentiation along PC2 (Fig. S1).

Overnight malate accumulation is a core feature of CAM photosynthesis and therefore an indicator of the respective photosynthetic phenotypes of *T. fasciculata* and *T. leiboldiana*. Malate abundances in the leaf fluctuated strongly in *T. fasciculata* over 24 hours, with highest median abundances around midday (D+5) and lowest abundances in the early night (N+1), representing a 3.8-fold difference (Fig. 2b). In comparison, malate abundances were overall lower for *T. leiboldiana* and fluctuated less. The highest median abundance in the latter species was found at N+1, while the lowest was found at D+5, representing a 2.2-fold difference. Interestingly, the accumulation times of malate seem reversed in the two species, with the highest abundances in *T. fasciculata* found at the time of lowest abundances in *T. leiboldiana* and vice versa. The reversed timing of malate accumulation has been described as a key difference between C3 and CAM metabolisms ((Winter and Smith, 2022), but see also (Bräutigam et al., 2017)). The median accumulation in malate within 24-hours differs significantly between species (Wilcoxon rank-sum test p-value = 8.6xE-3, Fig. 2c), and is 3.2-fold higher in *T. fasciculata* than in *T. leiboldiana*.

Overall, the malate accumulation curves suggest distinct photosynthetic phenotypes for *T. fasciculata* and *T. leiboldiana*. *T. fasciculata* appears to behave as a constitutive CAM plant in normal conditions, accumulating malate from the early night until the early day, while *T. leiboldiana*'s flux is more C3-like, without a clear accumulation overnight.

Figure 2: *Metabolomic analyses of* T. fasciculata *and* T. leiboldiana *leaf material throughout a 24-hour cycle. Abundances of individual metabolites were measured with GC-MS and normalised against the Main Total Ion Count (MTIC).* **a)** *Principal component analysis of metabolic composition of 72 leaf samples based on 77 metabolic compounds including soluble sugars, amino acids and organic acids. The first and second principal components are displayed. Blue arrows show the loadings of a subset of metabolites relevant for photosynthetic metabolism.* **b)** *Malate abundance in leaf material of* T. fasciculata *and* T. leiboldiana *at six timepoints across a 24-hour cycle. Dots represent individual observations across timepoints. Timepoints are noted as hours into the day (D) or into the night (N).* **c)** *Distribution of per-individual accumulation of malate per species over a 24-hour cycle. Accumulations were obtained by taking the difference in malate abundance between the highest and lowest reported abundances across time for each accession.*

3.2. Genome assembly and annotation

We constructed *de novo* haploid genome assemblies for both species (Table S1) using a combination of long-read (PacBio), short read (Illumina) and chromosome conformation capture (Hi-C) data. This resulted in assemblies of 838 Mb and 1,198 Mb with an N50 of 23.6 and 43.3 Mb in *T. fasciculata* and *T. leiboldiana* respectively. The assembly sizes closely match the estimated genome size of each species based on flow cytometry and k-mer analysis (Table S2, SI Note 1,2). The 25 and respectively 26 longest scaffolds (hereafter referred to as 'main scaffolds') contain 72 % and 75.5 % of the full assembly, after which scaffold sizes steeply decline (SI Note 3, Fig. S2). The number of main scaffolds corresponds with the species karyotype in *T. fasciculata*, but deviates from the *T. leiboldiana* karyotype (SI Note 1), suggesting that a few fragmented chromosome sequences remain in the latter assembly.

Structural gene annotation resulted in a total of 34,886 and 38,180 gene models in *T. fasciculata* and *T. leiboldiana* respectively, of which 92.6 % and 71.9 % are considered robust based on additional curation (Methods, Section 5). Annotation completeness was evaluated with BUSCO using the Liliopsida dataset resulting in a score of 89.7 % complete genes in *T. fasciculata* and 85.3 % in *T. leiboldiana* (Table S2).

3.3. Genic, repetitive and GC content

TE annotation performed with EDTA (Ou et al., 2019) revealed a total repetitive content of 65.5 % and 77.1 % in *T. fasciculata* and *T. leiboldiana* respectively. This closely matches estimates derived from k-mer analyses (66 % and 75 %, SI Note 2). Compared to *T. fasciculata*, the repetitive content in *T. leiboldiana* is enriched for Gypsy LTR retrotransposon and *Mutator* DNA transposon content, with a 1.7-fold and 4.2-fold increase in total covered genomic length, respectively (Table S3).

Repetitive content per scaffold is negatively correlated with gene count in both assemblies (Kendall's correlation coefficient: -0.79 in *T. fasciculata*, -0.82 in *T. leiboldiana,* p-values < 2.2E-16), with gene-rich regions in distal positions (Fig. 3a, green track) and repetitive regions primarily in median positions (Fig. 3a, yellow track). This pattern is accentuated in *T. leiboldiana*: on average, the repetitive-to-exonic content per scaffold is 1.6 times larger compared to *T. fasciculata* (Mann Whitney U, p-value = 4.3E-04). The genome size difference between the two assemblies is

therefore mostly explained by differential accumulation of TE content, mostly in heterochromatic regions.

Figure 3: a) *Circular overview of the main scaffolds of the* T. fasciculata *(right) and* T. leiboldiana *(left) genome assemblies. Scaffolds 25 and 26 of* T. leiboldiana *are not shown due to their reduced size. Going inwards, the tracks show: (1, green) gene count; (2, yellow) proportion of repetitive content; (3, red), and GC content per 1-Mb windows.* ***b)*** *TE and GC content, GC content exclusively in TEs, and genic content in a triplet of syntenic scaffolds between* Ananas comosus *(LG3, black),* T. fasciculata *(scaffold 4, grey) and* T. leiboldiana *(scaffold 1, green; see Fig. S3 for other syntenic chromosomes).* ***c)*** *Syntenic plot linking blocks of orthologous genes between* A. comosus, T. fasciculata *and* T. leiboldiana. *The size of each scaffold on the y-axis is proportional to genic content and therefore does not represent the true scaffold size. Colour-filled boxes indicate scaffolds with reversed coordinates as compared to the sequences in* A. comosus.

Contrary to the general expectation, GC content is negatively correlated with gene content

in both species (Kendall's correlation coefficient: -0.68 in *T. fasciculata*, -0.71 in *T. leiboldiana,*

p-values < 2.2E-16, red track in Fig. 3a, detailed in Fig. 3b). By visualising GC and TE content

across a syntenic chromosome triplet of *A. comosus*, *T. fasciculata* and *T. leiboldiana,* we show

that this unusual relationship can be mostly explained by elevated GC content in repetitive regions

(Fig. 3b). TE-rich regions indeed exhibit a much higher GC content than TE-poor regions, a pattern

which is exacerbated as the overall TE content per species increases (Fig. 3b, SI Note 4).

3.4.   Synteny and chromosomal evolution

Cytogenetic karyotyping (SI Note 1) revealed a difference of six chromosome pairs

between *T. fasciculata* ($2n = 50$) and *T. leiboldiana* ($2n = 38$), which is atypical in this largely

homoploid clade with generally constant karyotype (Gitaí et al., 2014; Brown and Gilmartin,

1989). To investigate orthology and synteny, we inferred orthogroups between protein sequences

of *Ananas comosus* (Ming et al., 2015) (pineapple), *T. fasciculata* and *T. leiboldiana* using

Orthofinder (Emms and Kelly, 2019). This resulted in 21,045 (78 %), 26,325 (87.5 %) and 23,584

(75 %) gene models assigned to orthogroups respectively, of which 10,021 were single-copy

orthologs between all three species (Table S4).

Syntenic blocks were then defined across all three assemblies using Genespace (Lovell et

al., 2022) (Fig. 3c). The observed karyotype difference between species could arise from

chromosomal loss in *T. leiboldiana*. However, our Genespace analysis reveals highly conserved

synteny between the two assemblies without major orphan regions in *T. leiboldiana*, consistent

with a scenario of chromosomal fusion, rather than loss. We found clear evidence of such a fusion

on scaffold 14 in *T. leiboldiana* (Fig. 3c, Fig. S4a), which was confirmed with in-depth analyses

of potential breakpoints (SI Note 5). However, chromosomal rearrangements are not limited to

fusions, since we also detected two major reciprocal translocations (Fig. 3c, hereafter referred as

Translocations 1 and 2, Fig. S4b-c).

3.5.   Gene family evolution

6,261 genes in *T. fasciculata* and 4,693 genes in *T. leiboldiana* were assigned to non-unique

gene families with multiple gene copies in at least one species, after correcting gene family sizes

(Table S4, SI Note 6). On average, the multicopy gene family size is 1.3x larger in *T. fasciculata*
than in *T. leiboldiana* (Mann Whitney U, p-value: 8.8E-16, Fig. 4a).

To investigate the role of expanded gene families in CAM evolution, we combined gene
ontology (GO) enrichment tests on multicopy orthogroups (SI Note 7) with a targeted search of
known genes involved in the CAM pathway. This highlighted 25 multi-copy gene families with
functions putatively related to CAM (Table S5), of which 17 have expanded in *T. fasciculata* and
eight in *T. leiboldiana.* Families expanded in *T. fasciculata* included a malate dehydrogenase
(MDH) and a β-carbonic anhydrase (CA), which are putatively involved in the carbon fixation
module of CAM photosynthesis, and subunits of the two vacuolar pumps (V-ATPase and V-
PPiase) known to energise the night-time transport of malate in pineapple (McRae et al., 2002)
(Table S5, Fig. 6). Additionally, two enolase families (members of the glycolysis pathway), a
vacuolar acid invertase putatively involved in day-time soluble sugar accumulation in the vacuole
(McRae et al., 2002; Holtum et al., 2005), and a pyrophosphate-dependent phosphofructokinase
associated with night-time conversion of soluble sugars through the glycolysis to PEP (Carnal and
Black, 1989) were expanded in *T. fasciculata*. Two subunits of succinate dehydrogenase, a
member of the tricarboxylic acid cycle and the electron transport chain which also plays a role in
stomatal opening regulation, a relevant aspect of CAM photosynthesis (Araújo et al., 2011), were
also expanded in *T. fasciculata* (Fig. 6). XAP5 CIRCADIAN TIMEKEEPER (XCT), a regulator
of circadian rhythm and disease resistance (Liu et al., 2022) which was previously identified as
undergoing rapid gene family evolution in *Tillandsia* (De La Harpe et al., 2020) is also expanded
in *T. fasciculata*.

Gene families expanded in *T. leiboldiana* contained three members of the glycolysis, two
aquaporins, a member of the tricarboxylic acid cycle and a regulator of stomatal opening (Table
S5).

c

Orthogroup	dN/dS	adj-P	Function
OG0002972	∞	0.00162	jacalin-related lectin 3-like
OG0005000	∞	0.00628	cucumber peeling cupredoxin-like
OG0006253	∞	0.00066	mitochondrial prohibitin-3
OG0009278	∞	0.00155	chloroplastic Peroxiredoxin-2E-2
OG0012770	∞	0.01374	metallo-hydrolase/oxidoreductase superfamily protein
OG0011786	9.3	0.00008	U-box_domain-containing_protein
OG0015603	8.1	0.01575	anaphase-promoting complex subunit CDC27
OG0009004	3.1	0.01148	Hydroquinone glycosyltransferase
OG0008977	2.6	0.00512	glutamate receptor 2.8-like
OG0010014	2.0	0.00603	Glycerophosphodiester phosphodiesterase GDPDL7

Figure 4: a) Scatterplot: composition of per-species gene counts among orthogroups. Upper histogram: distribution of per-orthogroup gene count in T. fasciculata. *Lower histogram: distribution of per-orthogroup gene count in* T. leiboldiana. *b) Distribution of d_N/d_S values of one-to-one orthologs across non-rearranged scaffolds (grey profile) and scaffold 14 in* T. leiboldiana *(blue profile), which is the result of a fusion. c) Single-copy orthogroups with significant d_N/d_S values and their functions. Three uncharacterized genes that are excluded here are detailed in Table S6. Infinite d_N/d_S values correspond to genes with $d_S=0$ (no synonymous substitutions), an expected situation considering the low divergence of the two species. Further explanation about the biological significance of these functions can be found in SI Note 8.*

3.6.   Adaptive sequence evolution

Adaptive sequence evolution was evaluated in 9,077 orthologous gene pairs using the non-synonymous to synonymous substitution ratio ($\omega = d_N/d_S$). Little among-scaffold variation in d_N/d_S was observed, with per-scaffold median d_N/d_S values ranging from 0.32 to 0.39 in *T. fasciculata* and 0.31 to 0.4 in *T. leiboldiana* (Fig. S5a). Regions of large chromosomal rearrangement such as the fused scaffold 14 in *T. leiboldiana* do not exhibit strong signatures of fast coding sequence evolution (Fig. 4b), though for Translocation 1, d_N/d_S values are slightly, yet significantly, lower for scaffold 13 in *T. fasciculata* and scaffold 19 in *T. leiboldiana* (p-value < 0.05, Fig. S5b, SI Note 5).

Among the 9,077 orthologous gene pairs, 13 candidates (0.21%) exhibit a significant d_N/d_S > 1 (adjusted p-value < 0.05, Fig. 4c, Table S6, SI Note 8). Notably, we recover a significant signal in a type B glycerophosphodiester phosphodiesterase (GDPDL-7). GDPDL's are involved in cell wall cellulose accumulation and pectin linking, and play a role in trichome development (Hayashi et al., 2008), a main trait differentiating the two species and more broadly, green and grey *Tillandsia*. Additionally, GDPDL-7 may be involved in response to drought and salt stress (Cheng et al., 2011).

A glutamate receptor (GLR) 2.8-like also exhibits a significant d_N/d_S > 1. By mediating Ca2+ fluxes, GLRs act as signalling proteins and mediate a number of physiological and developmental processes in plants (Weiland et al., 2015), including stomatal movement (Kong et al., 2016). Although it is associated with drought-stress response in *Medicago trunculata* (Philippe et al., 2019), the specific function of GLR2.8 still remains unknown.

3.7.   Gene expression analyses

To study gene expression differences linked to distinct photosynthetic phenotypes, we performed a time-series RNA-seq experiment using six plants of each species (Table S1, SI Note 9), sampled every four hours in a 24-hour period. We recovered 907 genes with a differential temporal expression (DE) profile between *T. fasciculata* and *T. leiboldiana*. Among them are 46 known CAM-related genes and 22 genes associated with starch metabolism and glycolysis/gluconeogenesis. GO term enrichment of the 907 DE genes revealed many CAM-related functions such as malate and oxaloacetate transport, circadian rhythm, light response, water and proton pumps, sucrose and maltose transport and starch metabolism (Table S7; Fig. 5a). While

none of the candidate genes for adaptive sequence evolution recovered in this study were
differentially expressed, nine of 22 genes reported by De La Harpe et al. 2020 as candidates for
adaptive sequence evolution during transitions to constitutive CAM in the wider context of the
genus *Tillandsia* were also differentially expressed in this study (Table S7).

Core CAM enzymes phosphoenolpyruvate carboxylase (PEPC) and phosphoenolpyruvate
carboxylase kinase (PEPC kinase, PPCK) display clear temporal expression cycling in *T.*
fasciculata (Fig. 5c, S5). PPCK also shows a night-time increase in expression in *T. leiboldiana*
(Fig. S6), albeit with a milder temporal effect, a phenomenon that has been documented before in
C3-assigned *Tillandsia* (De La Harpe et al., 2020) and also in other C3-like species belonging to
CAM- and C4-evolving lineages (Heyduk et al., 2019b, 2019a). Clustering analysis distributed DE
genes across seven clusters with sizes ranging from 209 to 38 genes (Table S7). CAM-related
genes were distributed across six of seven clusters, highlighting the diversity of expression profiles
associated with CAM (Fig. S7). While core CAM genes (see Fig. 6) are mainly found in cluster 5,
we find malate transporters in cluster 1, circadian regulators in clusters 2 and 3, sugar transport in
clusters 3 and 6, and vacuolar transport regulators in clusters 2, 4 and 6. Cluster 7, though not
containing any core CAM candidate genes, was enriched for salt and heat stress response and
contains a mitochondrial isocitrate dehydrogenase, which has been proposed as an alternative
carbon fixator in CAM plants (Tay et al., 2021; Töpfer et al., 2020).

The expression curves of the respective clusters (Fig. S7) demonstrate a complex web of
expression changes between photosynthetic phenotypes. The most common expression change
pattern among CAM-related genes is an overall increase in expression in the strong CAM plant
(*T. fasciculata*), paired with increased diel cycling peaking in the early night (clusters 2, 5 and 6).
This involved members of the night-time carbon fixating module of CAM such as PEPCK, PPCK
and MDH, enzymes involved in malate transport as V-ATPase and several glycolysis enzymes
such as glucose-6-phosphate isomerase, aldolase, Ppi-dependent phosphofructokinase (PFK) and
enolase (Fig. 6, S6). Enzymes of both soluble sugar transport (SUT2, ERD6, cluster 6) and starch
metabolism (Starch synthase I, α- and β-amylase and glucose-6-phosphate/phosphate translocator
(GPT), cluster 5 and 6) are overall up-regulated and show cycling expression curves in *T.*
fasciculata compared to *T. leiboldiana,* all showing highest activity in the late day. While the
increased night-time expression of Ppi-dependent PFK suggests a primary role of soluble sugars
as night-time source for PEP (Carnal and Black, 1989) in CAM *Tillandsia*, the simultaneously

cycling expression patterns in starch metabolic enzymes point also at transitory starch as a
potential source. A handful of CAM-related genes show increased expression in *T. leiboldiana*:
aluminium-activated malate transporter (ALMT), secondary vacuolar proton pump AVP1, which
also displays a phase shift peaking later in the night in *T. fasciculata*, and three circadian clock
regulators (LHY, GI and RVE1, cluster 1), which all show similar but reduced cycling patterns in
T. fasciculata compared to *T. leiboldiana*. We also see a phase shift in succinate dehydrogenase
peaking earlier in the night in *T fasciculata* versus *T. leiboldiana*'s early morning peak.

Overall, the majority of CAM-related DE gene expression profiles align with the view that
T. fasciculata is a constitutive CAM plant while *T. leiboldiana* performs a C3-like metabolism in
normal conditions, though showing signs of very early CAM evolution. The difference in
metabolism between both plants seems to be largely attained through regulatory rewiring of
functional enzymes.

3.8.   Genomic features of DE and CAM-related genes

3.8.1.   Differentially expressed genes have more TE insertions

To investigate whether transposable element (TE) activity and differential gene expression
are associated in *Tillandsia*, we tested whether intronic TE insertions are significantly enriched in
DE genes in both species. Both the presence of one or more TE insertions in a gene, as well as the
average number of TE insertions per gene is higher across all genes of *T. leiboldiana* compared to
the *T. fasciculata* gene set, which was expected given its larger proportion of repetitive content
(See Results section 3.2.).

While in both genomes, the proportion of DE genes with one or more TE insertions in
intronic regions is not significantly different to that of the full gene set, the average number of TE
insertions per gene is significantly higher in DE compared to non-DE genes (Table 1).

On the other hand, DE genes related to CAM, starch metabolism and glycolysis /
gluconeogenesis display on average reduced, but non-significant TE insertion rates in both
genomes (Table S8). However, the proportion of CAM-related DE genes with a TE insertion is
larger in *T. fasciculata* (41.2 %) than in *T. leiboldiana* (36.2 %), despite *T. leiboldiana*'s generally
elevated intronic TE insertion rate.

Table 1: Statistical test results on TE insertions in DE versus non-DE genes, and in DE genes previously described as underlying CAM, glycolysis or starch metabolism, versus all other genes.

Total number of genes with one or more TE insertions			
	All genes	DE genes	Chi-square p-value
In *T. fasciculata*	15844 (50 %)	473 (52 %)	0.2324
In *T. leiboldiana*	18251 (54.6 %)	472 (54.6 %)	0.9645
	All genes	DE CAM genes	Chi-square p-value
In *T. fasciculata*	15844 (50 %)	28 (41.2 %)	0.178
In *T. leiboldiana*	18251 (54.6 %)	21 (36.2 %)	0.008*
Average and (median) TE insertion counts per gene			
	in DE genes	in non-DE genes	Mann-Whitney U p-value
In *T. fasciculata*	3.66 (1)	2.88 (1)	0.0179
In *T. leiboldiana*	4.2 (1)	3.27 (1)	0.0159
	in CAM DE genes	in all other genes	Mann-Whitney U p-value
In *T. fasciculata*	2.618 (0)	2.906 (1)	0.375
In *T. leiboldiana*	2.862 (0)	3.291 (1)	0.083

Nine DE genes related to CAM, starch and gluconeogenesis display more than twice the
number of TE insertions as the genome-wide average in *T. fasciculata*. This includes a V-type
proton ATPase subunit H (vacuolar transport and acidification), an aluminium-activated malate
transporter, a ABC transporter C family member 4 (stomatal opening and circadian rhythm) and a
mitochondrial isocitrate dehydrogenase subunit (Table S9), which all had more TE insertions than
their orthologs in *T. leiboldiana*. On the other hand, six DE genes of interest had elevated TE
insertion rates in *T. leiboldiana,* including a vacuolar acid invertase (sugar metabolism) and a sugar
transporter ERD6-like. Four of these genes showed high amounts of TE insertions in both
genomes, such as a glucose-1-phosphate adenylyltransferase subunit (starch synthesis), a V-type
proton ATPase subunit C, and circadian clock regulator GIGANTEA (Dalchau et al., 2011), which
also plays a role in stomatal opening (Ando et al., 2013).

3.8.2. Differentially expressed genes belong more often to multi-copy orthogroups

To investigate the consequences of gene family evolution on gene expression, we tested whether the proportion of multi-copy orthogroups underlying DE genes was significantly elevated to that of the whole-genome set of orthogroups in both species (Fig. 5b, SI Note 10). The 907 DE genes in *T. fasciculata* are found in 738 orthogroups (hereafter called DE orthogroups) containing a total of 2,141 and 910 genes in *T. fasciculata* and *T. leiboldian*a, respectively. Genes from multi-copy orthogroups are more likely to be differentially expressed: while multi-copy orthogroups account for 24 % of all orthogroups in the genome, they represent 31 % of DE orthogroups. This difference is primarily explained by a 3.2x larger proportion of multi-copy orthogroups with a larger family size in *T. fasciculata* than in *T. leiboldiana* in the DE subset compared to the whole genome (Chi-square P = $1.59e^{-66}$).

Reciprocally, the DE analysis on the *T. leiboldiana* genome (SI Note 10) resulted in 836 DE genes belonging to 714 orthogroups, of which 489 overlap with the DE orthogroups resulting from the analysis on the *T. fasciculata* genome. As in the analysis on the *T. fasciculata* genome, we find that orthogroups with a larger family size in *T. leiboldiana* are enriched among DE genes. Additionally, both analyses point at a significant enrichment for multi-copy orthogroups with equal family sizes in both species, suggesting that also older duplications preceding the split of *T. fasciculata* and *T. leiboldiana* play a role in day-night regulatory evolution. This highlights the importance not only of novel, but also ancient variation in fuelling trait evolution in *Tillandsia*.

Figure 5: a) *CAM-related enriched GO terms among differentially expressed (DE) genes between* T. fasciculata *and* T. leiboldiana*. The genome in which a GO term has been found to be enriched among DE genes is shown by [F] and [L] for* T. fascuculata *and* T. leiboldiana*, respectively. The family size difference for the underlying orthogroups is represented as a Z-score: a negative score indicates a tendency towards gene families with larger size in* T. leiboldiana *than in* T. fasciculata*, and vice versa. The p-value displayed represents the significance of the GO-term enrichment among DE genes in* T. fasciculata*, unless the term was only enriched in* T. leiboldiana*. The number of DE genes underlying each function is shown next to the GO-term name.* **b)** *Composition of gene families by relative size between T. fasciculata (F) and T. leiboldiana (L) for three subsets of gene families (whole genome, DE and CAM-related DE). Species-specific orthogroups are not included in this analysis. A chi-square test of independence was applied to test the significance of composition changes in 2x2 contingency tables for each category when testing the entire DE gene subset. For CAM DE genes, the Fisher's exact test was applied. Significant p-values of both tests are reported as:* *0.05-0.01, **0.01-0.0001, ***0.0001-0. **c)** *Expression profiles in a 24-hour period of exemplary CAM-related gene families (PEPC, MDH and XCT) displayed at the*

orthogroup level. For each gene copy and timepoint, the average read count (in Counts Per
Million), and the standard deviation across accessions are displayed. Read counts of each
ortholog are obtained by mapping conspecific accessions to their conspecific reference genome.
We show two families with older duplications preceding the split of T. fasciculata *and* T.
leiboldiana *(PEPC and MDH) and one gene family with a recent duplication in* T. fasciculata
(XCT).

Multi-copy gene families are also enriched in a restricted subset of DE orthogroups related
to CAM, starch metabolism and gluconeogenesis (68 genes in 67 orthogroups), especially gene
families with equal copy number and with copy number expansion in *T. fasciculata* (Fig. 5b, Table
S5). Importantly, expanded gene families in *T. leiboldiana* are not significantly enriched in this
functional subset, showing that while the full set of DE orthogroups exhibit increased gene family
dynamics in both lineages, CAM-related gene family expansion is only associated with *T.*
fasciculata (constitutive CAM). This pattern is also reflected on the GO-term level, where enriched
CAM-related biological functions appear disproportionally associated with gene family expansion
in *T. fasciculata* (nine functions) than in *T. leiboldiana* (one function). Functions associated with
V-ATPase proton pumps especially tend to have larger gene family size in *T. fasciculata* than in
T. leiboldiana (ATPase binding, proton-transporting ATPase activity).
CAM-related expanded gene families often show one highly expressed copy that performs
diel cycling, while the other copies are lowly or not expressed (e.g. SDH, glyceraldehyde-3-
phosphate dehydrogenase (GAPDH), V-ATPase subunit H, SI Note 11), however, several gene
families show diel and/or elevated expression in *T. fasciculata* in two or more gene copies (starch
synthase, Ppi-dependent PFK and enolase, SI Note 11). Both copies of XCT are expressed in *T.*
fasciculata, though showing no diel cycling or increased expression compared to *T. leiboldiana*
(Fig. 5c), making its role in CAM photosynthesis unclear. V-ATPase subunit H has eight copies
in *T. fasciculata* and three in *T. leiboldiana*. While in both species only one copy is highly
expressed (with diel cycling peaking at night in *T. fasciculata*), the copies implemented for
elevated expression in either species are not each other's homologs (SI Note 11), suggesting that
different copies are recruited for the distinct photosynthetic phenotypes of these species. The other
copies are lowly or not expressed. A candidate gene family with larger gene family size in *T.*
leiboldiana is a probable aquaporin PIP2-6 (OG0005047, Fig. S8), which is involved in water

regulation and follows a diel pattern in pineapple (*A. comosus*) (Zhu and Ming, 2019). While lowly

expressed in *T. fasciculata*, one of the two gene copies shows strong diel expression in *T.*

leiboldiana, with highest expression in the early night. This is another indication that an early,

latent CAM cycle may be present in *T. leiboldiana*.

CAM-related gene families with duplications preceding the divergence of *T. leiboldiana*

and *T. fasciculata* include a malate dehydrogenase (MDH) with two copies in both species, where

only one copy is highly expressed and cycling in *T. fasciculata*, and the core CAM enzyme

Phosphoenolpyruvate carboxylase (PEPC), which shares an ancient duplication among monocots

(Deng et al., 2016) (Fig. 5c). The widely varying expression patterns of multicopy DE CAM-

related families suggest a variety of mechanisms possibly contributing to CAM regulatory

evolution: dosage changes ("more of the same"), subfunctionalization and neofunctionalization.

4.   Discussion

The sources of variation fuelling trait evolution in rapid radiations have been a long-

standing topic in evolutionary biology (Simpson, 1953), and our understanding of how complex

traits such as CAM evolve repeatedly is still incomplete. By showcasing a broad range of

photosynthetic phenotypes and repeated evolution of constitutive CAM, the subgenus *Tillandsia*

provides an excellent opportunity to study CAM evolution. The recent divergence between

members of *Tillandsia* allows to pinpoint the necessary evolutionary changes to evolve a

constitutive CAM phenotype. By integrating comparative genomics using *de novo* assemblies and

in-depth gene expression analyses of two closely related *Tillandsia* species representing one of the

most distinct photosynthetic phenotypes within the clade, we found support for regulatory

evolution and gene family expansion as major features of CAM evolution (Fig. 6).

Our metabolic analyses of night-time malate accumulation and gene expression analyses

provide a much more detailed understanding of the photosynthetic phenotypes present in the

subgenus than the previously reported carbon isotope measurements, which do not accurately

reflect all stages of the CAM continuum. For example, weak CAM phenotypes have been reported

in Bromeliaceae for species with carbon isotope ratios falling in the C3 range (-26.5) (Pierce et al.,

2002), indicating that the photosynthetic metabolism of *T. leiboldiana* could be different from C3

sensu stricto as its carbon isotope measurements suggest. Our timewise malate measurements

show distinct fluxes for *T. fasciculata* and *T. leiboldiana*, with constitutive accumulation of night-

time malate in *T. fasciculata* which is absent in *T. leiboldiana*. On the other hand, *T. leiboldiana*
does display CAM-like temporal expression profiles for certain enzymes, such as PEPC kinase
and Aquaporin PIP-6. It has been suggested that repeated evolution of CAM (similar to C4) may
be facilitated in lineages where C3 species already display increased or CAM-like expression of
core genes (Heyduk et al., 2019a, 2019b; Kajala et al., 2012). However, while a CAM cycle is
seemingly not being expressed in *T. leiboldiana* under normal circumstances, we can not exclude
that a latent CAM cycle could become activated under certain conditions, for example under
drought stress. In that case, *T. leiboldiana* would rather be at the very early stages of CAM
evolution than a "pre-adapted" C3 plant (De La Harpe et al., 2020). We hope that future studies
will study the drought response of *T. leiboldiana* to better understand its exact position in the CAM
continuum.

On the other hand, even if *T. leiboldiana* and potentially all subgenus *Tillandsia* species
previously labelled as C3 represent in fact the early stages of the CAM continuum, our analyses
show widely distinct photosynthetic phenotypes within the radiation which required divergent
evolution. Therefore, while this study may underestimate the total number of evolutionary changes
needed to establish constitutive CAM from C3 *sensu stricto*, it highlights the evolutionary drivers
underlying the least understood section of the CAM continuum: from early CAM to constitutive
CAM.

Differences between the two genomes related to CAM evolution can be primarily found
on the regulatory level, with CAM-related genes showing temporal differential expression between
species across a 24-hour period. These reveal a complex web of underlying expression changes,
as they are distributed over all inferred co-expression clusters (Fig. S7), emphasising the lack of a
master regulator and a clear overall direction of expression changes (Wickell et al., 2021; Heyduk
et al., 2022).

Gene family expansion has been previously observed in CAM lineages (Cai et al., 2015;
Silvera et al., 2010) and suggested as a driver of CAM evolution (Silvera et al., 2014). Strikingly,
both the total subset of differentially expressed genes and a more stringent group of CAM-related
DE genes was significantly enriched for gene families that have expanded in *T. fasciculata*
(constitutive CAM). CAM-related functions that were enriched in DE genes show a
disproportionate bias towards gene family expansion in *T. fasciculata* (circadian rhythm, vacuolar
ATPase activity, tricarboxylic acid cycle and starch metabolism) compared to *T. leiboldiana*

(glycolysis) (Fig. 5a). Gene duplications, preceding the split of *T. fasciculata* and *T. leiboldiana*
are also significantly associated with day-night expression differences in CAM-related genes (Fig.
5b), suggesting that older, already existing gene duplications may also be recruited in CAM
evolution, alongside novel duplications.

The expression curves of DE multicopy gene families with a potential link to CAM reveal
a multitude of expression behaviours (e.g. Fig. 5c), which supports that complex regulatory
evolution on the transcriptional level underlies CAM evolution. Our findings suggest that gene
family evolution played a significant role in modulating regulatory changes underlying the
evolution towards constitutive CAM in *Tillandsia*. As gene family expansion leads to increased
redundancy, selection on individual gene copies and their expression relaxes, facilitating the
assimilation of a constitutively expressed CAM expression profile (Ohno, 1970).

Another potential driver of trait evolution is TE insertion, though its role in CAM evolution
in *Tillandsia* remains unclear. TE insertions are overall less common in CAM-related DE genes
compared to all genes in both genomes, suggesting a selection pressure against TE insertions
around these genes. However, the proportion of CAM-related DE genes with TE insertions is
greater in *T. fasciculata* than in *T. leiboldiana,* despite the overall higher genic TE insertion rate
in *T. leiboldiana*. This suggests that the pressure to maintain CAM-related genes TE-free is
reduced in the constitutive CAM lineage relative to *T. leiboldiana*. We detect nine and six CAM-
related DE genes with more than twice the whole-genome average TE insertion count in *T.*
fasciculata and *T. leiboldiana* respectively, of which four are single-copy orthologs shared
between species. The high degree of sharedness of TE-rich DE genes between species rather
suggests that TE insertions are not a major driver of CAM-related gene expression changes. In
fact, genes with exceptionally high insertion rates in *T. fasciculata* tend to show reduced
expression (ALMT, V-ATPase subunit H copy Tfasc_v1.24696, ABCC4, SI Note 11). Instead, the
larger proportion of CAM-related DE genes with one or more TE insertions in *T. fasciculata* may
be a consequence of higher rates of gene family expansion and eventual pseudogenization of
redundant copies.

Figure 6: *Pathway of Crassulacean Acid Metabolism (CAM), highlighting underlying genes detected in this study as differentially expressed, with gene family expansion, with signature of adaptive sequence evolution or elevated TE insertion counts. The colour of the process indicates the species showing diel or elevated expression, increased gene family size or increased TE insertion rate. Enzymes are shown in coloured boxes, while pathway products are shown in bold*

*outside boxes. Enzymatic members of CAM metabolism pathways are shown in orange boxes, while stomatal and circadian regulators are highlighted in yellow. Stomatal regulators are shown on the left outside the cell and circadian regulators on the right. **a)** CO_2 is absorbed at night and first converted to HCO_3^- by carbonic anhydrase (CA). Then, it is converted to malate by carboxylating phosphoenol pyruvate (PEP), a key component of the glycolysis. In a first step, PEP carboxylase (PEPC) converts PEP to oxaloacetate, after being activated by PEPC kinase (PPCK). In a second step, malate dehydrogenase (MDH) converts oxaloacetate to malate. Malate is then transported into the vacuole by two possible transporters, either a tonoplast dicarboxylate transporter or an aluminium-activated malate transporter, which are assisted by V-ATPase proton pumps. During the day, the accumulated malate becomes the main source of CO_2 for photosynthesis. This allows the stomata to remain closed, which greatly enhances the water use efficiency of the plant. Malate is again transported out of the vacuole and reconverted to oxaloacetate by MDH, and then decarboxylated to PEP and CO_2 by PEP carboxykinase (PEPCK). The CO_2 will cycle through the Calvin cycle and generate sugars. Abbreviations: GLR2.8 – Glutamate receptor 2.8, ABCC4 - ABC transporter C 4, SDH – Succinate Dehydrogenase, XCT – XAP5 CIRCADIAN TIMEKEEPER, GI – protein GIGANTEA, RVE1 – REVEILLE 1. **b)** Glycolysis, transitory starch and sugar metabolism are tightly linked with the core CAM pathway as providers of starting materials such as PEP. During the day, CAM plants can store starch in the chloroplast and hexoses in the vacuole. In Bromeliaceae, the relative importance of soluble sugars versus starch as a source for PEP is variable across species[56]. At night, the stored starch and/or sugars are transported into the cytoplasm, converted to glucose or fructose and broken down via the glycolysis to PEP. Abbreviations: G6P - glucose-6-phosphate; PGM – phosphoglucomutase; GPT - glucose-6-phosphate/phosphate translocator; PFK – phosphofructokinase; GAPDH - Glyceraldehyde-3-phosphate dehydrogenase; F6P – fructose-6-phosphate; F1,6BP – fructose-1,6-biphosphate; SSI – starch synthase I; 1,3BPG - 1,3-Bisphosphoglyceric acid; PEP – Phosphoenol pyruvate; SUT2 – Sucrose transporter 2; ERD6 - EARLY RESPONSE TO DEHYDRATION 6. For a detailed description and accompanying per-gene expression profiles, see SI Note 11.*

Candidate genes under positive selection underlie a broad array of functions, but had no immediate link to CAM photosynthesis. The lack of overlap between regulatory and adaptive

sequence evolution is in line with previously proposed mechanisms of CAM evolution largely

relying on regulatory changes in other systems (Deng et al., 2016). A small number of cases of

convergent and adaptive sequence evolution between distantly related CAM and C3 species have

been described (Yang et al., 2017), though no overlap was found between convergence in

expression and sequence evolution. Our study suggests that while on larger evolutionary scales

adaptive sequence evolution may play an important role, distinct photosynthetic phenotypes

between closely related species may be achieved primarily with gene expression changes (but see

SI Note 14), or may be especially relevant between the transition of C3 *sensu stricto* to the CAM

continuum.

Though we observe a karyotype difference of six chromosome pairs between *T. fasciculata*

and *T. leiboldiana* and we identified one fusion in the *T. leiboldiana* assembly, along with two

reciprocal translocations, we did not find detectable consequences of large-scale rearrangements

for either functional diversification or adaptation in *Tillandsia*, unlike other studies (Davey et al.,

2016; Cicconardi et al., 2021) (Fig. 4b, S4, S6, SI Note 12, but see SI Note 5 and 14). The role of

these large-scale rearrangements in the evolution of species barriers and/or the evolution of other

key innovation traits in *Tillandsia* will be an exciting area of study for future research.

The two *de novo* assemblies presented in this study are the first tillandsioid and fourth

bromeliad genomes published so far. Despite both genomes exhibiting one of the highest TE

contents reported to date for a non-polyploid plant species (Pedro et al., 2021), the joint use of

long-read sequencing and chromatin conformation capture successfully led to highly contiguous

assemblies with high-quality gene sets (SI Note 13). Along with other recently developed

resources for Bromeliaceae (Yardeni et al., 2021; Liu et al., 2021), these genomes will be crucial

in future investigations of this highly diverse and species-rich plant family, and in further studies

of CAM evolution.

Our analyses reveal genomic changes of all scales between two members of an adaptive

radiation representing a recent shift to constitutive CAM. However, in this recent shift between

closely related species, differences in photosynthetic metabolism are brought about largely by

temporal expression changes enabled by both existing and *de novo* gene duplication, rather than

adaptive sequence evolution of existing gene copies, which may play a role at later stages of

divergence (but see SI Note 14). Large scale rearrangements observed so far seem unlinked from

functional divergence, more likely affecting reproductive isolation (de Vos et al., 2020; Faria and

Navarro, 2010), and need further study. Our findings support an important role for gene family
expansion in generating novel variation that fuels the evolution of the CAM continuum.
641
642 **5.   Materials & Methods**
643
644   **5.1.   Flow cytometry and cytogenetic experiments**
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5.1.1.   Genome size measurements
Approximately 25 mg of fresh leaf material was co-chopped according to the chopping method of
(Galbraith et al., 1983) together with an appropriate reference standard (*Solanum pseudocapsicum*,
1.295 pg/1C) (Temsch, 2010; Temsch et al., 2022) in Otto´s I buffer (Otto et al., 1981). After
filtration through a 30 µm nylon mesh (Saatilene Hitech, Sericol GmbH, Germany) and incubation
with RNase A (0.15mg/ml, Sigma-Aldrich, USA) at 37°C, Otto´s II buffer (Otto et al., 1981)
including propidium iodide (PI, 50mg/L, AppliChem, Germany) was added. Staining took place
in the refrigerator for at least one hour or up to over-night. Measurement was conducted on a
CyFlow ML or a CyFlow Space flow cytometer (Partec/Sysmex, Germany) both equipped with a
green laser (532nm, 100mW, Cobolt AB, Sweden). The fluorescence intensity (FI) of 10,000
particles were measured per preparation and the 1C-value calculation for each sample followed
the equation: $1C_{Obj} = (\text{FI peak mean}_{G1\,Obj} / \text{FI peak mean}_{G1\,Std}) \times 1C_{Std}$
658
5.1.2.   Karyotyping
Actively growing root meristems of genome assembly accessions (see Table S1) were harvested
and pretreated with 8-hydroxyquinoline for 2 hrs at room temperature and 2 hrs at 4°C. The roots
were then fixed in Carnoy's fixative (3 : 1 ethanol : glacial acetic acid) for 24 hours at room
temperature and stored –20 °C until use. Chromosome preparations were made after enzymatic
digestion of fixed root meristems as described in (Jang and Weiss-Schneeweiss, 2015).
Chromosomes and nuclei were stained with 2 ng/µl DAPI (4',6-diamidino-2-2phenylindole) in
Vectashield antifade medium (Vector Laboratories, Burlingame, CA, USA). Preparations were
analyzed with an AxioImager M2 epifluorescent microscope (Carl Zeiss) and images were
captured with a CCD camera using AxioVision 4.8 software (Carl Zeiss). Chromosome number

was established based on analyses of several preparations and at least five intact chromosome

spreads. Selected images were contrasted using Corel PhotoPaint X8 with only those functions

that applied equally to all pixels of the image and were then used to prepare karyotypes.

5.2.   Genome Assembly

5.2.1.   Plant material selection and sequencing

Genome assemblies were constructed from the plant material of one accession per species

(see Table S1). The accessions were placed in a dark room for a week to minimise chloroplast

activity and recruitment, after which the youngest leaves were collected and flash frozen with

liquid nitrogen. High molecular weight extraction for ultra-long reads, SMRTbell library

preparation and PacBio Sequel sequencing was performed by Dovetail Genomics™ (now Cantata

Bio). Dovetail Genomics™ also prepared Chicago (Putnam et al., 2016) and Hi-C (Lieberman-

Aiden et al., 2009) libraries which were sequenced as paired-end 150bp reads on an Illumina HiSeq

X instrument. Additional DNA libraries were prepared for polishing purposes using Illumina's

TruSeq PCR-free kit, which were sequenced on a HiSeq2500 as paired-end 125 bp reads at the

Vienna BioCenter Core Facilities (VBCF), Austria.

RNA-seq data of *T. fasciculata* used for gene annotation was sampled, sequenced and

analyzed in De La Harpe et al. 2020 under SRA BioProject PRJNA649109. For gene annotation

of *T. leiboldiana*, we made use of RNA-seq data obtained during a similar experiment, where

plants were kept under greenhouse conditions and sampled every 12 hours in a 24-hour cycle.

Importantly, while the *T. fasciculata* RNA-seq dataset contained three different genotypes, only

clonal accessions were used in the *T. leiboldiana* experiment. For *T. leiboldiana*, total RNA was

extracted using a QIAGEN RNeasy® Mini Kit, and poly-A capture was performed at the Vienna

Biocenter Core Facilities (VBCF) using a NEBNext kit to produce a stranded mRNA library. This

library was sequenced on a NovaSeq SP as 150 bp paired end reads.

For both species, sequencing data from different time points and accessions were merged

into one file for the purpose of gene annotation. Before mapping, the data was quality-trimmed

using AdapterRemoval (Schubert et al., 2016) with default options (--trimns, --trimqualities). We

allowed for overlapping pairs to be collapsed into longer reads.

5.2.2. First draft assembly and polishing

We constructed a draft assembly using long-read PacBio data with CANU v1.8 (Koren et al., 2017) for both species. To mitigate the effects of a relatively low average PacBio coverage (33x), we ran two rounds of read error correction with high sensitivity settings (corMhapSensitivity=high corMinCoverage=0 corOutCoverage=200) for *T. fasciculata.* Additionally, we applied high heterozygosity (correctedErrorRate=0.105) settings, since K-mer based analyses pointed at an elevated heterozygosity in this species (See SI Note 2), and memory optimization settings (corMhapFilterThreshold=0.0000000002 corMhapOptions=" --repeat-idf-scale 50" mhapMemory=60g mhapBlockSize=500).

Given that the coverage of *T. leiboldiana* PacBio averaged 40x, we limited error correction for this species to only one round. CANU was run with additional settings accommodating for high frequency repeats (ovlMerThreshold=500) and high sensitivity settings as mentioned above.

To minimise the retention of heterozygous sequences as haplotigs in *T. fasciculata* (see SI Note 2), we reassigned allelic contigs using the pipeline Purge Haplotigs (Roach et al., 2018). Raw PacBio data was mapped to the draft assembly produced in the previous step with minimap2 (Li, 2018), before using the Purge Haplotigs pipeline.

Since the size of the *T. leiboldiana* draft assembly indicates, together with previous analyses, that this species is largely homozygous (SI Note 2), we did not include a purge_haplotigs step. However, we did make use of the higher average coverage of the *T. leiboldiana* PacBio data to polish the assembly with two rounds of PBMM v.1.0 and Arrow v2.3.3 (Pacific Biosciences).

5.2.3. Scaffolding and final polishing

Scaffolding of both assemblies was performed in-house by Dovetail GenomicsTM using Chicago and Hi-C data and the HiRise scaffolding pipeline (Putnam et al., 2016). To increase base quality and correct indel errors, we ran additional rounds of polishing with high-coverage Illumina data (See above, section 2.1.) using Pilon v1.22 (Walker et al., 2014). The Illumina data was aligned to the scaffolded assembly using BWA-MEM (Li, 2013), and then Pilon was run on these alignments. We evaluated the result of each round using BUSCO v.3 (Waterhouse et al., 2018) with the Liliopsida odb9 library and proceeded with the best version. For *T. fasciculata*, polishing was performed twice, fixing SNPs and indels. We did not fix small structural variation in this genome due to the relatively low coverage (35x) of Illumina data. For *T. leiboldiana*, one round

of polishing on all fixes (SNPs, indels and small structural variants) resulted in the highest BUSCO

scores.

5.3.   Annotation

5.3.1.   TE annotation and repeat masking

De novo TE annotation of both genome assemblies was performed with EDTA v.1.8.5 (Ou

et al., 2019) with option –sensitive. To filter out genes that have been wrongly assigned as TEs, *A.*

comosus (pineapple) coding sequences (Ming et al., 2015) were used in the final steps of EDTA.

Using the species-specific TE library obtained from EDTA, we masked both genomes

using RepeatMasker v.4.0.7 (Smit et al., 2013-2015). Importantly, we excluded all TE annotations

marked as "unknown" for masking to prevent potentially genic regions flagged as TEs to be

masked during annotation. The search engine was set to NCBI (-e ncbi) and simple and low-

complexity repeats were left unmasked (-nolow). We produced both hard-masked and soft-masked

(--xsmall) genomes.

5.3.2.   Transcriptome assembly

We constructed transcriptome assemblies for both species using Trinity *de novo* assembler

v.2.4.8. (Grabherr et al., 2011) using default parameters starting from the raw mRNA-seq data.

These were evaluated with BUSCO. Additionally, before feeding the transcriptome assemblies to

the gene annotation pipeline, we ran a round of masking of interspersed repeats to avoid an

overestimation of gene models due to the presence of active transposases in the RNA-seq data.

5.3.3.   Gene prediction and functional annotation

Gene models were constructed using a combination of BRAKER v.2.1.5 (Hoff et al., 2019)

and MAKER2 v.2.31.11 (Campbell et al., 2014). Starting with BRAKER, we obtained splicing

information from RNA-seq alignments to the masked genome as extrinsic evidence using the

bam2hints script of AUGUSTUS v.3.3.3 (Stanke et al., 2008). A second source of extrinsic

evidence for BRAKER were single-copy protein sequences predicted by BUSCO when run on the

masked genomes in genome mode with option --long. Predictions made by BRAKER were

evaluated with BUSCO and with RNA-seq alignments.

Subsequently, we built our final gene predictions using MAKER2. As evidence, we used

(1) the gene models predicted by BRAKER, (2) a transcriptome assembly of each respective

species (see above section 3.2.), (3) a protein sequence database containing proteins of *Ananas*

comosus comosus (F153) (Ming et al., 2015) and *Ananas comosus bracteatus* (CB5) (Chen et al.,

2019) and manually curated swissprot proteins from monocot species (64,748 sequences in total)

and (4) a GFF file of complex repeats obtained from the masked genome (see above section 3.1.)

and an extended repeat library containing both the EDTA-produced *Tillandsia*-specific repeats and

the monocot repeat library from RepBase (7,857 sequences in total). By only providing masking

information of complex repeats and setting the model organism to "simple" in the repeat masking

options, hardmaksing in MAKER2 was limited to complex repeats while simple repeats were soft-

masked, which makes these available for gene prediction. MAKER2 predicts genes both *ab initio*

and based on the given evidence using AUGUSTUS.

We evaluated the resulting set of predicted gene models by mapping the RNA-seq data

(section 2.1.) back to both the transcript and full gene model sequences and running BUSCO in

transcriptome mode. We also calculated the proportion of masked content in these gene models to

ascertain that MAKER2 hadn't predicted TEs as genes. A second run of MAKER, which included

training AUGUSTUS based on the predicted models from the first round, resulted in lower

BUSCO scores and was not further used. We functionally annotated the final set of gene models

in Blast2Go v.5.2.5 (Götz et al., 2008) using the Viridiplantae database.

5.4.   Inferring gene orthology

Orthology between gene models of *T. fasciculata*, *T. leiboldiana* and *Ananas comosus* was

inferred using Orthofinder v.2.4.0 (Emms and Kelly, 2019). Protein sequences produced by

MAKER2 of inferred gene models were used for *T. fasciculata* and *T. leiboldiana*. For *A. comosus*,

the publicly available gene models of F153 were used. The full Orthofinder pipeline was run

without additional settings. Counts per orthogroup and the individual genes belonging to each

orthogroup were extracted from the output file Phylogenetic_Hierarchical_Orthogroups/N0.tsv.

Orthofinder was run a second time on gene models present only on main contigs (See

Results). For each gene model, the longest isoform was selected, and gene models with protein

sequences shorter than 40 amino acids were removed. This resulted in 27,024, 30,091 and 31,194

input sequences for *A.* comosus, *T. fasciculata* and *T. leiboldiana* respectively. Then, the steps
mentioned above were repeated.

5.5. Gene model assessment and curation

Gene model sets were assessed and curated using several criteria. Gene models with
annotations indicating a repetitive nature (transposons and viral sequences) together with all their
orthologs were marked with "NO_ORTHOLOGY" in the GFF file and excluded from downstream
analyses. Using the per-exon expression data obtained in our mRNA-seq experiment (see below,
section 10.) and information gathered on the length of the CDS and the presence / absence of a
start and stop codon, we further classified our gene models into ROBUST and NOT-ROBUST
categories. A gene model was considered ROBUST (i) if all exons are expressed or, (ii) if both
start and stop codons are present and the CDS has a minimum length of 50 amino-acids.

5.6. Analyzing TE class abundances

By rerunning EDTA with step --anno, we obtained TE abundances and detailed annotation
of repetitive content for the whole assembly. Per-contig abundances of each class were calculated
with a custom python script (available at https://github.com/cgrootcrego/). Using this curated TE
library, the assemblies were masked again with RepeatMasker for downstream analyses. The
resulting TE class abundances reported by RepeatMasker were then compared between species
and reported.

5.7. Spatial distribution of repetitive, genic and GC content

The spatial distribution of genes, transposable elements and GC content as shown in Fig.
3a, was analysed on a per-window basis, using windows of 1 Mb. Gene counts were quantified as
the number of genes starting in every window, based on genes with assigned orthology, including
both single and multicopy gene models. Repetitive content was measured as the proportion of
masked bases in each window, stemming from the hard-masked assembly using the curated TE
library. Per-window gene counts and proportion of repetitive bases was then visualized using the
R package circlize (Gu et al., 2014). GC content was calculated as the proportion of G and C bases
per 1 Mb windows. Correlation between genic, repetitive and GC content was calculated and tested

for significance using the Kendall Rank Correlation Coefficient, after testing for normality using
the Shapiro-Wilk test.

Repetitive, GC and gene content as shown in Fig. 3b was estimated directly from the soft-
masked reference genomes using 100 kb non-overlapping sliding windows as described in (Leroy
et al., 2021). TE content corresponds to the proportion of soft-masked positions per window, using
the curated TE library (see above, section 6.) as a basis for soft-masking in RepeatMasker. As
compared to the version of Leroy et al. 2021, this script was modified to estimate GC content in
repetitive regions only. In addition to this, we estimated the genic fraction by considering the total
number of genomic positions falling in genes based on the GFF files (feature = "gene") divided
by the size of the window (100 kb). This estimate was derived for the same window boundaries as
used for GC and TE content to be able to compare all statistics. The relative per-window proportion
of genic bases corresponding to non-robust genes (see above, section 5) was also estimated by
dividing the number of non-robust gene positions with the total number of gene positions.

5.8.   Synteny between *T. fasciculata* and *T. leiboldiana*

Synteny was inferred with GENESPACE (Lovell et al., 2022), using orthology information
obtained with Orthofinder of the gene models from *A. comosus*, *T. fasciculata* and *T. leiboldiana*.
This provided a first, visual graphical to detect large-scale rearrangements. We used GENESPACE
with default parameters, except that we generated the syntenic map (riparian plot) using
minGenes2plot=200. Other methods have also been used to confirm the chromosomal
rearrangements and to identify the genomic breakpoints more precisely (see SI Note 5).

5.9.   Gene family evolution

Gene family counts were corrected for multi-copy orthogroups due to unusual coverage
distribution, especially in *T. fasciculata* (see SI Note 6). The distribution of gene counts per
multicopy orthogroup was compared between *T. fasciculata* and *T. leiboldiana* with a non-
parametric test (Mann-Whitney U). Using the log-ratio of per-species gene count, we investigated
which gene families experienced large changes in gene count compared to the background (SI
Note 7).

Functional characterization of multicopy families was done with a GO term enrichment
analysis of the underlying genes using the Fisher's exact test in TopGo (Alexa and Rahnenführer,

2009). Enrichment analyses were done on all genes belonging to multicopy orthogroups, on a
subset of genes belonging to families that are larger in *T. fasciculata* and on a subset of genes
belonging to families that are larger in *T. leiboldiana*. The top 100 significantly enriched GO terms
were then evaluated. GO terms putatively associated with key innovation traits were used to list
multicopy gene families of interest.

Additionally, we searched for specific genes that are known to underlie CAM evolution in
these multi-copy gene families. The IDs of candidate pineapple genes for CAM were obtained
from (Yardeni et al., 2021), who compiled extensive lists of genes from a diverse set of studies.
For CAM, we considered all genes listed in Table S1 in this study under the categories
"Differentially expressed in CAM / C3 experiment" (186 genes) (De La Harpe et al., 2020),
"Positive selection in CAM / C3 shifts" (22) (De La Harpe et al., 2020), Gene families associated
with CAM/C3" (79) (De La Harpe et al., 2020), "CAM-related *A. comosus*" (29) (Ming et al.,
2015), "stomatal function" (48) (Christin et al., 2014), "aquaporin regulation" (24) (Vera-Estrella
et al., 2012), "drought resistance" (61) (Xiao et al., 2007), "circadian metabolism" (47) (Wai et al.,
2017), "malate transferase" (28) (Cosentino et al., 2013) and "circadian clock" (3) (McClung,
2006), resulting in a total of 527 genes. A separate list was made for gluconeogenesis and starch
metabolism genes (288 genes) (Cushman et al., 2008). After obtaining these lists of pineapple gene
IDs, we searched for their orthologs in *T. fasciculata* and *T. leiboldiana*, and investigated their
presence in multi-copy gene families.

872
873 **5.10. d_N/d_S analysis**
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5.10.1.   On single-copy orthologous pairs

One-to-one orthologous genes were subjected to a test of positive selection using the non-
synonymous to synonymous substitution ratio ($\omega = d_N/d_S$). Gene pairs where both genes were
incomplete (missing start and/or stop codon) or where the difference in total length was more than
20 % of the length of either gene were removed to avoid misalignments. We performed codon-
aware alignments using the alignSequences program from MACSE v.2.05 (Ranwez et al., 2018)
with options -local_realign_init 1 -local_realign_dec 1 for optimization. Pairwise d_N/d_S ratios were
estimated with the codeML function of PAML v.4.9. (Yang, 2007). Using a single-ratio model
across sites and branches (Nssites = 0, model = 0), we tested for a fixed $\omega = 1$ as null hypothesis,

against an unfixed ω as the alternative hypothesis. Automization of codeML was achieved with a

modified script from AlignmentProcessor

(https://github.com/WilsonSayresLab/AlignmentProcessor/). The results of codeML under both

the null and alternative model were compiled and significance of the result was calculated with the

likelihood-ratio test (Wong et al., 2004). Multiple-testing correction was applied with the

Benjamini-Hochberg method and an FDR threshold of 0.05. Orthologous gene pairs with a d_N/d_S

ratio larger than one and an adjusted p-value under 0.05 were considered candidate genes under

divergent selection.

The d_N/d_S values of all orthologous gene pairs with five or more variant sites in the MACSE

alignment were used to obtain per-scaffold distributions of d_N/d_S values in both genomes. We

visualized d_N/d_S distributions of all main scaffolds in both assemblies with boxplots and used

density plots to visualize the d_N/d_S distribution in rearranged chromosomes compared to all non-

rearranged chromosomes. To test whether these distributions were significantly different, we ran

a non-parametric test (Mann-Whitney U) between the distribution of each single rearranged

chromosome and that of all non-rearranged chromosomes in each assembly.

5.10.2.   On duplicated orthogroups

We also performed tests of selection using d_N/d_S on all orthogroups that were consisted of

a single gene in *A. comosus* and a duplicated gene in either T. *leiboldiana* (1:1:2), or *T. fasciculata*

(1:2:1). Only orthogroups that maintained this conformation after size correction (SI Note 6) were

used in this analysis. Pairwise alignments were performed between the ortholog of one species and

either paralog of the other species using MACSE. Then, ω was estimated in the same way as

mentioned above.

5.11.   RNA-seq experiment capturing photosynthetic phenotypes and expression

5.11.1.   Experiment set-up and sampling

To capture gene expression patterns related to CAM, we designed an RNA-seq experiment

where individuals of *T. fasciculata* and *T. leiboldiana* were sampled at six time points throughout

a 24-hour cycle. Six plants of each species were placed in a PERCIVAL climatic cabinet at 22 °C

and a relative humidity (rH) of 68 % for 4 weeks, with a 12-hour light cycle. Light was provided

by fluorescent lamps with a spectrum ranging from 400 to 700 nm. The light intensity was set at

124 umol/m2s. The plants acclimated to these conditions for 4 weeks prior to sampling, during

which they were watered every second day.

Leaf material from each plant was sampled every 4 hours in a 24-hour cycle starting one

hour after lights went off. One leaf was pulled out of the base at each time-point without cutting.

The base and tip of the leaf were then removed, and the middle of the leaf immediately placed in

liquid nitrogen, then stored at -80 °C.

5.11.2.   Targeted metabolite analyses

To corroborate the photosynthetic phenotypes of *T. fasciculata* and *T. leiboldiana*, we

measured malate abundances in the leaf throughout a 24-hour cycle. An approximate amount of

20 mg of frozen leaf material collected at six timepoints during the above-mentioned experiment

was collected and ground to a powder with a TissueLyser and metal beads. Subsequent steps were

performed at the Vienna Metabolomics Center (VIME, Department of Ecogenomics and Systems

Biology, Vienna, Austria).

Polar metabolites were extracted by modifying the procedure of Weckwerth et al., 2004. A

weighed amount of deep frozen and ground plant tissues was combined with 750 µL of ice-cold

extraction solvent, consisting of methanol (LC-MS grade, Merck), chloroform (anhydrous >99 %,

Sigma Aldrich), and water (MilliQ) in a ratio of 2.5:1:0.5 (v/v). Additionally, 7 µL of a solution

of 10 mmol of pentaerythritol (PE) and 10 mM phenyl-β-d-glucopyranoside (PGP) respectively in

water (MilliQ) were added as an internal standard mix. After ultrasonication at 4 °C for 20 minutes

and centrifugation (4 min, 4 °C, 14,000 g) the supernatant was transferred to a new 1.5 mL tube

(polypropylene). Another 250 µL of extraction solvent was added to the remaining pellet and after

another cycle of ultrasonication and centrifugation as described before, the supernatant was

combined with the previous. To induce phase separation 350 µL of water (MilliQ) was added.

After thorough mixing and consecutive centrifugation (4 min, 4 °C, 14,000 g) 900 µL of the upper

phase were transferred to a new 1.5 mL tube. Approximately 100 µL of the remaining polar phase

of all samples were combined. The 900 µL aliquotes of this mixed sample were used as quality

control during measurements. The polar phases and the aliquotes of the sample mix were dried in

a vacuum centrifuge for 5 hours at 30 °C and 0.1 mbar.

The dried extracts were derivatized as described earlier (Doerfler et al., 2013) by dissolving the metabolite pellet carefully in 20 µL of 40 mg of methoxyamine hydrochloride (Sigma Aldrich) in 1 mL pyridine (anhydrous >99,8 %, Sigma Aldrich). After incubation at 30 °C and 700 rpm for 1.5 hours on a thermoshaker, 80 µL of *N*-methyl-*N*-trimethylsilyl-trifluoroacetamid (Macherey-Nagel) were added. The samples were incubated for 30 min at 37 °C and 750 rpm and consecutively centrifuged for 4 min at room temperature and 14,000 g.

Metabolite analysis was performed on an Agilent 7890B gas chromatograph equipped with a LECO Pegasus® BT-TOF mass spectrometer (LECO Corporation). Derivatized metabolites were injected through a Split/Splitless inlet equipped with an ultra inert single tapered glass liner with deactivated glass wool (5910-2293, Agilent Technologies), a split ratio of 1:25 was used and the temperature was set to 230 °C. Components were separated with helium as carrier gas on a Restek Rxi-5Sil MS column (length: 30 m, diameter: 0.25 mm, thickness of film: 0.25 µm). The initial oven temperature was set to 70 °C held for 1 minute and ramped with a rate of 9 °C per minute until reaching 340 °C help for 10 minutes. Collection of spectra started after an acquisition delay of 280 seconds with a detector voltage of 1692.5 V, a rate of 15 spectra per second and a mass range of 50−500 *m/z*. Retention indices were calculated based on the retention times of the alkane mixture C_{10}-C_{40} run within each of the 2 batches. Samples were measured in randomized order and randomly distributed across the batches. Within each batch, a mixture of standard metabolites was measured for MSI level I identification of metabolites. Deconvolution, annotation, and processing of chromatograms was performed according to (Zhang et al., 2023) using ChromaTOF® (Version 5.55.29.0.1187, LECO Cooperation) and MS-DIAL, version 4.7 (Tsugawa et al., 2015). Areas of derivatization products of single metabolite were summed and normalized by the main targeted ion content of each sample.

5.11.3.   RNA extraction and sequencing

Using the same sampled leaf material as for targeted metabolite analyses, total RNA was extracted using the QIAGEN RNeasy® Mini Kit in an RNAse free laboratory. Samples were digested using the kit's RLT buffer with 1 % Beta-mercaptoethanol. Elution was done in two steps. The purity and concentration of the extractions was measured using Nanodrop, and RIN and fragmentation profiles were obtained with a Fragment Analyzer™ system. RNA libraries were

prepared by the Vienna Biocenter Core Facilities (VBCF) using a NEBNext stranded mRNA kit before sequencing 150-bp paired-end reads on one lane of Illumina NovaSeq S4.

5.11.4. RNA-seq data processing

The raw RNA-seq data was evaluated with FastQC (https://www.bioinformatics.babraham.ac.uk/projects/fastqc/) and MultiQC (Ewels et al., 2016), then quality trimmed using AdapterRemoval v.2.3.1 (Schubert et al., 2016) with settings --trimns --trimqualities --minquality 20 --trimwindows 12 --minlength 36. The trimmed data was then aligned to both the *T. fasciculata* and *T. leiboldiana* genomes using STAR v.2.7.9 (Dobin et al., 2013) using GFF files to specify exonic regions. Because mapping bias was lowest when mapping to *T. fasciculata* (see SI Note 9, Fig. S9, S10), our main analyses have been performed on the reads mapped to this genome. However, the alignments to *T. leiboldiana* were used for verification or expansion of the main analysis (SI Note 10).

5.11.5. Differential Gene Expression analysis

We quantified read counts per exon using FeatureCounts from the Subread package v.2.0.3. (Liao et al., 2014) for paired-end and reversely stranded reads (-p -s 2). The counts were then summed up across exons per gene to obtain gene-level counts. The composition of the count data was investigated with PCA in EdgeR (Robinson et al., 2009). Then, counts were normalised using the TMM method in EdgeR, and every gene with a mean cpm < 1 was removed. We ran a differential gene expression (DE) analysis in maSigPro (Conesa et al., 2006), which detects genes with differential diurnal expression profiles between species using a regression approach. *Tillandsia leiboldiana* was used as the baseline in this analysis. Significant DE genes were then clustered using the hclust algorithm into modules, with the number of modules being determined with the K-means algorithm. Expression curves were plotted by taking the average expression across all replicates per species at each time point with a custom R script. Expression curves for entire clusters (Fig. S7) were plotted by median-centering the log(CPM) of each gene and time point against the median of all genes at each time point, while expression curves for individual genes or gene families (Fig. 5c, S5, S7) report average CPM.

GO term enrichments were performed for each cluster using the R package TopGO (Alexa and Rahnenführer, 2009). Separately, known candidate genes underlying CAM and starch metabolism (See Methods, Section 5.9.) were searched among differentially expressed genes.

5.12. Intersecting findings of gene family evolution, TE insertion ad differential gene expression

5.12.1. Transposable element insertions and differential gene expression

Intronic TE insertions were obtained using *bedtools intersect* on the GFF files of the TE and gene annotations of both species. We used the full transcript length of a gene (feature = "mRNA" in GFF file) for this analysis, and only applied "known" TE annotations. This resulted in a dataset reporting the number of TE insertions per gene. We then performed two tests on the resulting TE counts per gene: (1) whether the proportion of genes with one or more TE insertions is elevated in DE genes compared to the full gene set (chi-square test), and (2) whether the rate of TE insertions per gene measured, as the total count of intersections for each gene annotation with a TE annotation, is elevated in DE genes compared to non-DE genes (Mann-Whitney U test).

The same test was also applied to a restricted set of DE genes previously described as CAM-related, or involved in starch metabolism and gluconeogenesis. Then, genes of interest with a TE insertion rate higher than twice the genome-wide average were selected and the difference in number of TE insertions between orthologs of *T. leiboldiana* and *T. fasciculata* was taken in case of a one-to-one relationship.

5.12.2. Gene family evolution and differential gene expression

Orthogroups were split based on relative family size in *T. fasciculata* (F) versus *T. leiboldiana* (L) in the following categories: Single-copy orthogroups (F = 1 : L = 1), orthogroups with family size larger in *T. fasciculata* (F > L), orthogroups with family size smaller in *T. fasciculata* (F < L), orthogroups with equal family sizes that are larger than 1 (F = L). Orthogroups unique to one species (F:0 or 0:L) were not considered in this analysis. We counted the number of orthogroups belonging to each category for the full orthogroup set, for the subset of orthogroups containing DE genes (DE orthogroups) and for the subset of orthogroups containing DE genes that

have been previously described as CAM-related (CAM DE orthogroups). We then tested whether

counts in each orthogroup category were enriched in DE orthogroups and DE CAM orthogroups

compared to all orthogroups. For comparisons of all orthogroups versus DE orthogroups, we used

the chi-square test of independence in R. For the comparison of CAM DE orthogroups versus all

orthogroups, Fisher's exact test on 2x2 contingency tables of each orthogroup category versus all

other categories, and DE genes versus non-DE genes was used due to small sample sizes. To study

the effect of the reference genome used on our findings on gene family evolution in DE genes, we

performed the same analysis on read counts obtained from mapping to *T. leiboldiana* (SI Note 10).

1043

6.   Accession Numbers

The genome assemblies and raw data used in this study are available at NCBI-SRA under

BioProject [PRJNA927306](#). Specifically, the *T.fasciculata* genome assembly TFas_v1 can be

downloaded   here: https://ncbi.nlm.nih.gov/datasets/genome/GCA_029168755.1/.   The   *T.*

leiboldiana   genome   assembly   TLei_v1   can   be   found   at:

https://www.ncbi.nlm.nih.gov/datasets/genome/GCA_029204045.1/. The annotation of both

genomes   is   available   on   the   github   repository   at:

https://github.com/cgrootcrego/Tillandsia_Genomes, together with the list of orthogroups, counts

table used for RNA-seq analyses, full GO term enrichment results, and all scripts written for this

manuscript. The *A. comosus* sequences used in this study stem from BioProject PRJNA371634

(F153): https://www.ncbi.nlm.nih.gov/bioproject/PRJNA371634/ and BioProject PRJNA747096

(CB5): https://www.ncbi.nlm.nih.gov/bioproject/747096. Information on all accessions used in

this study can be found in Supplemental Table 1.

1057

7.   Supplemental data

Supplemental Figure S1. Per-metabolite loadings of PC1 and PC2

Supplemental Figure S2. Distribution of scaffold sizes and orthologous gene counts for the top

100 largest scaffolds of both assemblies.

Supplemental Figure S3. TE, GC and gene content at three examples of syntenic chromosome

triplets.

Supplemental Figure S4. In-depth visualisation of large-scale rearrangements between *T.*

fasciculata and *T. leiboldiana.*

Supplemental Figure S5. Genome-wide distribution of d_N/d_S values between single-copy
orthologous genes.

Supplemental Figure S6. Average expression curve of PEPC kinase (PPCK) in *T. fasciculata* and
T. leiboldiana.

Supplemental Figure S7. Per-gene expression curves of all differentially expressed genes.

Supplemental Figure S8. Average expression curve of Aquaporin 2-6 in *T. fasciculata* and *T.*
leiboldiana.

Supplemental Figure S9. Unique mapping rates of RNA-seq reads to three different reference
genomes.

Supplemental Figure S10. Multi-mapping rates of RNA-seq reads to three different reference
genomes.

Supplemental Figure S11. Genome size measurement histograms.

Supplemental Figure S12. Mitotic metaphase chromosomes and karyotypes of *Tillandsia*
fasciculata and *Tillandsia leiboldiana.*

Supplemental Figure S13. Heterozygosity and genome size estimation with a k-mer based
approach.

Supplemental Figure S14. Distribution of heterozygous sites per 1000 mappable variants.

Supplemental Figure S15. Mean per-gene coverage distribution across different gene family
categories.

Supplemental Figure S16. Distribution of DE genes across the genome.

Supplemental Table S1. List of accessions used in this study.

Supplemental Table S2. Assembly statistics.

Supplemental Table S3. TE Abundances.

Supplemental Table S4. Orthology statistics.

Supplemental Table S5. List of CAM-related expanded orthogroups.

Supplemental Table S6. Full list of candidate genes under positive selection.

Supplemental Table S7. Full list of co-expression modules.

Supplemental Table S8. List of DE genes related to CAM, starch metabolism and gluconeogenesis
with a high TE insertion rate.

Supplemental File 1. Supplementary Notes.

1096

8.   Acknowledgments

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9.   Author Contributions

This study was conceived by CL, JH, OP, TL and CGC. Sampling was conducted by MHJB, WT, GY and MDH. Laboratory work was conducted by MHJB, SS, LACS and CGC. Cytogenetic work was performed by HWS and EMT. The RNA-Seq experiment and DE analysis was conducted under the guidance of KH and OP. Analyses were performed by CGC, JH, GY, TL and FB. The manuscript was primarily written by CGC and amended following the dedicated reading and feedback of all co-authors, especially KH, TL and OP.

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1415