Ancestral class-promiscuity as a driver of functional diversity in the BAHD acyltransferase family in plants

Lars H. Kruse[1], Austin T. Weigle[3], Jesús Martínez-Gómez[1,2], Jason D. Chobirko[1,5], Jason E. Schaffer[6], Alexandra A. Bennett[1,7], Chelsea D. Specht[1,2], Joseph M. Jez[6], Diwakar Shukla[4], Gaurav D. Moghe[1*]

Footnotes:

[1] Plant Biology Section, School of Integrative Plant Sciences, Cornell University, Ithaca, NY, 14853, USA

[2] L.H. Bailey Hortorium, Cornell University, Ithaca, NY, 14853, USA

[3] Department of Chemistry, University of Illinois at Urbana-Champaign, Urbana, IL, 61801, USA

[4] Department of Chemical and Biomolecular Engineering, University of Illinois at Urbana-Champaign, Urbana, IL, 61801, USA

[5] Present address: Department of Molecular Biology and Genetics, Cornell University, Ithaca, NY, 14853, USA

[6] Department of Biology, Washington University in St. Louis, St. Louis, MO, 63130, USA

[7] Present address: Institute of Analytical Chemistry, Universität für Bodenkultur Wien, Vienna, 1190, Austria

* Corresponding author: gdm67@cornell.edu

ABSTRACT

Gene duplication-divergence and enzyme promiscuity drive metabolic diversification in plants, but how they contribute to functional innovation in enzyme families is not clearly understood. In this study, we addressed this question using the large BAHD acyltransferase family as a model. This fast-evolving family, which uses diverse substrates, expanded drastically during land plant evolution. *In vitro* characterization of 11 BAHDs against a substrate panel and phylogenetic analyses revealed that the ancestral enzymes prior to origin of land plants were likely capable of promiscuously utilizing most of the substrate classes used by current, largely specialized enzymes. Motif enrichment analysis in anthocyanin/flavonoid-acylating BAHDs helped identify two motifs that potentially contributed to specialization of the ancestral anthocyanin-acylation capability. Molecular dynamic simulations and enzyme kinetics further resolved the potential roles of these motifs in the path towards specialization. Our results illuminate how promiscuity in robust and evolvable enzymes contributes to functional diversity in enzyme families.

KEY WORDS

Evolutionary biochemistry, enzyme family, comparative genomics, gene duplication, promiscuity, protein structure analysis, BAHD acyltransferase

INTRODUCTION

Enzymes involved in plant specialized metabolism often belong to enzyme
families, some of whom (e.g. cytochrome P450s, lipases, acyltransferases,
dioxygenases) have several hundred members in angiosperm genomes. Such enzyme
families are characterized by frequent gene duplication, functional divergence, and
promiscuity, all of which contribute to metabolic diversification. For duplication, various
models explaining duplicate gene evolution have been proposed, such as neo-
functionalization, sub-functionalization, escape from adaptive conflict, dosage balance,
and pseudogenization (reviewed in Panchy et al., 2016). Due to advances in sequencing
technologies, much is known today about how duplicates evolve at the genomic,
epigenetic and transcriptomic levels (Ganko et al., 2007; Zou et al., 2009; Schnable et
al., 2011; Moghe et al., 2014; J. Wang et al., 2014); however, our understanding of how
substrate preference evolves in duplicate enzymes, especially in large enzyme families
is lacking. Specifically, while it is common knowledge that different members of large
enzyme families use substrates containing very different chemical/structural scaffolds,
the extent to which this "family multi-functionality" or functional diversity is due to ancestral
promiscuity vs. neo-functionalization after duplication is not clear.

Promiscuity refers to the ability of an enzyme to catalyze multiple reactions, either
by using different substrates (substrate promiscuity), producing multiple products from
the same substrate (product promiscuity), or performing secondary reactions that cause
different chemical transformations (catalytic promiscuity) (Copley, 2015). Here, we do not
consider the physiological relevance of the secondary products but only study an
enzyme's ability to use multiple substrates – a definition of promiscuity typically used by
molecular/structural biologists (Copley, 2015; Kreis and Munkert, 2019). We also define
a special type of substrate promiscuity called "class-promiscuity", referring to the ability
of an enzyme to use substrates containing very different structural scaffolds e.g. aliphatic
alcohol vs. anthocyanin. In contrast, the term "multi-functionality" is used here in the
context of the collective enzyme family using multiple substrates e.g. multi-functionality
of the BAHD family. Even if the product at first is irrelevant in a physiological context, the
promiscuous reaction may still continue to occur and may get selected upon if the product
directly or indirectly increases organismal fitness. Existence of such promiscuity-driven

"underground metabolism" can occur via drift and contributes to the standing natural

variation of metabolites (Notebaart et al., 2014).

In this study, we address the question of how gene duplication and promiscuity

contribute to plant specialized metabolic diversity using the large BAHD acyltransferase

family (referred to as BAHDs hereafter) as a model. The ease of heterologous protein

expression in *Escherichia coli*, intronless nature of many BAHD genes, their ability to use

structurally diverse substrates, and availability of functional data from multiple species

make this an attractive family to address the above question. Named after the four first

discovered enzymes of this family – benzyl alcohol O-acetyltransferase (BEAT)

(Dudareva et al., 1998), anthocyanin O-hydroxycinnamoyltransferase (AHCT) (Fujiwara

et al., 1997; H. Fujiwara et al., 1998; Hiroyuki Fujiwara et al., 1998), N-

hydroxycinnamoyl/benzoyltransferase (HCBT) (Yang et al., 1997), and deacetylvindoline

4-O-acetyltransferase (DAT) – members of this large family (referred to as BAHDs

hereafter) catalyze the transfer of an acyl group from a coenzyme A (CoA) conjugated

donor to a –OH or –NH_2 group on an acceptor (D'Auria, 2006). BAHDs play important

roles in the biosynthesis of several phenylpropanoids, amides, volatile esters, terpenoids,

alkaloids, anthocyanins, flavonoids, and acylsugars (D'Auria, 2006; Tuominen et al.,

2011). Although >150 members of this family have been experimentally characterized

across the plant kingdom, transfer of known functions using sequence similarity to these

characterized enzymes is difficult owing to their rapid sequence divergence, substrate

promiscuity and functional divergence. For example, the 4-5 acylsugar acyltransferases

involved in acylsugar biosynthesis in Solanaceae trichomes are BAHDs (Moghe et al.,

2017) but are only 40-50% identical, and yet all of them use sucrose/acylated sucrose as

substrates. In contrast, a single amino acid change is sufficient to convert a BAHD from

preferentially using phenylpropanoid substrates to using phenolic amine substrates

(Levsh et al., 2016). Compared to the scale of BAHD acceptor diversity, there is an

incomplete understanding of BAHD sequence-function and structure-function

relationships, which is representative of a similar lack of knowledge in other large enzyme

families generated via gene duplication.

Previous studies on BAHDs have revealed existence of substrate promiscuity

(Aharoni et al., 2000; Aymerick Eudes et al., 2016; Levsh et al., 2016; Moghe et al., 2017;

Chiang et al., 2018a), but compared to the known number of BAHD substrates, the extent

of our knowledge about BAHD class-promiscuity is still limited. Furthermore, it is unclear

whether BAHD multi-functionality is a result of multiple rounds of neo-functionalization –

where new activities emerged completely afresh in the BAHD family (also referred to

below as "innovation") – vs. specialization of activities that were already possible in the

common ancestor.

The BAHD family is speculated to have arisen from carnitine acyltransferases

involved in fatty acid metabolism (St Pierre and De Luca, 2000; D'Auria, 2006), however,

their evolution across plants has not been studied. We were interested in characterizing

evolution of the *capability*/*potential* of BAHDs to use different substrate classes rather

than their *actual in vivo* substrates, since the inherent capability of an enzyme can be a

starting point for selection to act and fix diversified enzyme activities in different *in vivo*

contexts. Although it may be argued that BAHDs might use any substrates with hydroxyl

or amine groups, previous results provide clear evidence of specialization (D'Auria, 2006),

and it is unclear how these specializations emerged. We first characterized the known

substrate space of extant characterized BAHDs and used it as a template in the context

of BAHD phylogeny to delineate the putative ancestral substrate space. Prediction of the

ancestral state helped us differentiate between neo-functionalization vs. ancestral

promiscuity, and the sequence and structural features that enabled specialization of

ancestrally accessible functions. Overall, this study provides a template to assess

functional evolution after duplication in large enzyme families, and generates resources

foundational for rational prediction of BAHD function in plant genomes.

RESULTS

The BAHD enzyme family occupies a wide substrate space

128   BAHDs have been experimentally characterized across land plants (Sander and

129   Petersen, 2011; Aymerick Eudes et al., 2016; Levsh et al., 2016; Moghe et al., 2017;

130   Chiang et al., 2018a). To get a complete picture of the range of known BAHD substrates,

131   we first compiled a database of 136 biochemically characterized BAHDs that used a total

132   of 187 acceptor substrates and ~30 acyl donor substrates from 64 species across the

green plants **(File S2)**. BAHDs characterized solely using other types of experimental evidence such as gene knock-out and knock-down, or gene expression analysis were excluded from this study, because of the remaining uncertainty about the actual substrate the enzyme is acting on. These substrates were empirically classified into thirteen

Figure 1. BAHD substrate space. (A) Substrate similarity network of BAHD acceptor substrates (**File S2B**), shown as the "Prefuse Force Directed" layout in Cytoscape. An MCS-Tanimoto similarity cutoff of 0.5 used to draw edges between two substrate nodes. Substrates are colored based on the substrate type they belong to. **(B)** BAHD substrates in the context of a larger network of plant compounds. Plant compounds with KNApSAcK ID were gathered from the ChEBI database. The network was visualized as in part A and BAHD substrates are highlighted by colors corresponding to their substrate class. For each labeled cluster/region in the network, five compounds were chosen randomly and the compound class for each compound was determined based on the ChEBI Ontology (see **File S2**). Representative classes are shown for each analyzed region. Names in bold are the classes closest to the known BAHD substrates.

different types based on similarity of their chemical scaffolds and functional groups **(Table**

S1; Fig. S1A) (Wang et al., 2013), and organized into a network for visualization of the

already characterized BAHD substrate space **(Fig. 1A; Fig. S1B)**.

Distinct clusters obtained in this visualization suggested that BAHDs have evolved

to accept at least eight different structural scaffolds **(classes A-G-Mix; Fig. 1A**). A

majority (109, 80%) of the characterized enzymes use substrates in four classes – class

A comprising of aromatic alcohol and amines (69 enzymes, 51%), class B containing

aliphatic alcohols and amines (38 enzymes, 28%), class C that includes monoterpenoids

and a sesquiterpenoid (7 enzymes, 5%), and class D comprising flavonoids,

anthocyanins, and phenolic glycosides (38 enzymes, 28%) **(Fig. 1A; File S2B)** – roughly

indicative of the degree of research attention on lignins, cuticular lipids, floral volatiles and

pigments, respectively. BAHDs can also transform a wider array of terpenoids including

monoterpenoid (e.g. geraniol), diterpenoid (e.g. taxol intermediates), and triterpenoid

alcohols (e.g. thalianol) in classes E and F. Some polyamines (spermine, spermidine)

were more distant from aliphatic alcohols primarily due to different functional groups (-OH

vs. $-NH_2$), but still were classified into the same class B due to overall scaffold relatedness

(Fig. S1A; Table S1). Other substrates such as alkaloids (class Mix) and sugar

derivatives (class G), represented smaller and independent classes in the network.

Alkaloids themselves are a loosely defined compound type, and hence class Mix

comprises of alkaloids that could actually be assigned to other classes.

Although large, the characterized BAHD substrate space still only represents a

small proportion of the larger phytochemical space. As described in a later section, only

~50% of all multi-species BAHD orthologous groups (OGs) have characterized activities,

which suggests that many substrate classes remain undiscovered. We thus used the

substrate network to obtain a bird's eye view of known BAHD substrates in the context of

the wider phytochemical space. Using 7128 additional plant-specific compounds, we

mapped a broader phytochemical similarity network **(Fig. 1B).** Since BAHDs can only use

substrates containing amine or hydroxyl groups, only ~70% of the compounds from this

pool are actually available to them.

Previous studies recognize the role of substrate ambiguity and structural motifs in

generating secondary reactions (Bar-Even et al., 2011), which, under appropriate

conditions, can be selected upon. An investigation of the global "underground

metabolism" found that promiscuously used substrates of *E. coli* enzymes tend to be

structurally similar (Notebaart et al., 2014). Based on these observations, we divided

compounds in the above phytochemical network **(Fig. 1B)** into three groups: (1) known

BAHD substrate classes; these classes are spread out over a large region of the

phytochemical space **(colored nodes, Fig. 1B)**, which potentially allows for transitions to

new substrate classes through mutational changes, (2) compound classes that are

theoretically accessible – due to similarity to known substrates – but not yet known as

BAHD substrates, such as small organic acids in central metabolism, some amino acids,

nitrogen bases/nucleosides, many alkaloids, sesquiterpenoids, peptides, polyphenols,

and oligosaccharides **(Fig. 1B; File S3)**, and (3) compound classes whose utilization

cannot be inferred due to absence of prior data or are unlikely to be BAHD substrates. As

far as we could determine, no biochemical or genetic studies have identified BAHDs using

sulfur/phosphorus containing compounds (e.g. glucosinolates, nucleotides) and long-

chain lipid types (e.g. sulfolipids, sphingolipids, carotenoids/tetraterpenoids) as acyl

acceptor substrates – despite some members having hydroxyl groups – and hence, these

substrate classes were categorized into group 3. From groups 2 and 3, some substrate

types may be inaccessible to BAHDs due to their lipophilic nature or their absence in

cytoplasmic environments where most BAHDs are known to exist. Nonetheless, such

visualization of the larger network shows that several metabolite classes are structurally

very similar to existing BAHD substrates, and represent the latent catalytic potential of

BAHDs that could be selected upon after duplication-divergence or could be used for *in*

vitro enzyme engineering. These activities may still lie undetected in uncharacterized

enzymes or as secondary activities of characterized enzymes.

To address our questions about class-promiscuity, we further assessed enzymes

accepting class A, B and D substrates, which are structurally very distinct **(Fig. 1A)**. Three

enzymes utilizing class A substrates (AtHCT, SmHCT, PsHCT2) were previously shown

in one study to use naringenin, a flavonoid (Chiang et al., 2018a). No class D substrate

utilizing enzyme was shown to use class A/B substrates **(File S2A)**. While this difference

could represent true functional differentiation, it may likely be a result of experimenter-

bias (i.e. researchers studying pigmentation may not assess lignin or volatile esters (class

A/B substrates), and vice versa). To minimize the effect of such a bias on our later
evolutionary inferences, we performed extensive characterization of 11 BAHDs that use
class A, B and D substrates and obtained better insights into promiscuity of BAHD
enzymes.

Determining the class-promiscuity of characterized BAHDs

We tested a total of 11 novel and previously characterized BAHD enzymes against
an acceptor substrate array **(Fig. S2)**, using optimal conditions described for those
enzymes. For donors, we selected the most preferred donor of each tested enzyme,
which, for most enzymes was coumaroyl-CoA or malonyl-CoA. Cocaine synthase from
Erythroxylum coca (EcCS) was described to use benzoyl-CoA and cinnamoyl-CoA to
produce cocaine from methylecgonine (Schmidt et al., 2015; A. Eudes et al., 2016), but
in our enzyme assays, we discovered that it is also able to use coumaroyl-CoA (**Fig. S2**).
For the hydroxyacid/alcohol hydroxycinnamoyl transferase of the liverwort *Marchantia*
emarginata (MeHFT), we used its preferred donor feruloyl-CoA. Based on the substrate
networks **(Fig. 1A; Fig. S1),** we selected 11 substrates from six substrate types in classes
A, B and D as acceptors for initial analysis, and based on these results, further assayed
some enzymes with additional anthocyanin and terpenoid substrates **(Fig. S2B,C)** to
confirm additional hypotheses.

Of the eleven enzymes, two used only one tested substrate, six showed substrate
promiscuity within the same class (which we also refer to as "specialized" below), while
three showed class-promiscuity (**Fig. 2A; Fig. S2**) According to our assays and gathered
literature information **(Fig. 4)**, 75% (103 enzymes) of BAHDs analyzed in this study can
use more than one acceptor substrate, 27% (37 enzymes) use $\geq$5 substrates, and 9%
(12 enzymes) can use $\geq$10 substrates, highlighting the considerable substrate promiscuity
in the family. One of the two enzymes using only one substrate was a previously
uncharacterized enzyme from the outgroup species selected for this study *Chara braunii*
– representing charophytic algae, the most closely related sister lineage to land plants
(Cheng et al., 2019; Donoghue and Paps, 2020; Vries and Rensing, 2020). Under the
testing conditions, this enzyme exclusively accepted quinate (an aromatic alcohol) **(Fig.**
2A,C; Fig. S2A), leading us to rename this enzyme as CbHQT-like**.** This observation,

 coupled with the evolutionary analysis performed below, suggests that activities essential

 for monolignol biosynthesis in land plants (Weng and Chapple, 2010; Renault et al., 2019)

Figure 2. Enzyme activities of selected enzyme representatives. (A) Matrix of tested enzymes and substrates. Class A/B/C/F substrates are pink, Class D substrates are blue. Larger red circle corresponds to major enzyme activity and smaller orange circles indicate moderate activity. See **Fig. S2** for more detailed calculations. n.t. = not tested. **(B-E)** extracted ion chromatograms of the quantifier ions of different enzymatic products of **(B)** MeHFT with its preferred donor feruloyl-CoA **(C)** CbHQT-like using quinate and coumaroyl-CoA **(D)** Dm3MAT3 using malonyl-CoA, and **(E)** SlHCT using coumaroyl-CoA. Each product was measured using product-specific PRM methods (see **Table S4**) and the most abundant fragment ion was used for quantification. This ion is noted in the upper right corner of each chromatogram. Structures represent the best-inference based on previously reported structures and the observed fragmentation patterns. Species photographs are from following sources: *Chara braunii* (Picture by Rob Palmer released under the CC BY-NC-SA license), *Marchantia emarginata* (Picture by Boon-Chuan Ho), *Dendranthema × morifolium* (Wikimedia Commons released into the public domain), and *Solanum lycopersicum* (no license attached).

existed before the last common ancestor (LCA) of land plants. Given the timescale of

hundreds of millions of years representing the split between charophytes and land plants,

it is possible CbHQT-like specialized for utilizing quinate after the split from land plants.

Alternatively, CbHQT-like might natively use substrates not tested in this study. Further

in vivo experiments coupled with untargeted metabolomics could provide further insights

into the functional capacity of algal BAHDs.

Three out of eleven tested BAHDs were class-promiscuous. The previously

uncharacterized *Solanum lycopersicum* hydroxycinnamoyl CoA transferase (SlHCT) was

able to acylate shikimate and quinate (aromatic alcohols, class A), dopamine and

tyramine (aromatic amines, class A), geraniol (monoterpenoids, class C), curcubitacin E

(triterpenoids, class F) as well as malvidin 3-O glucoside, cyanidin-3,5-O diglucoside,

cyanidin 3-O glucoside, and cyanidin 3-O rutinoside (anthocyanins, class D) **(Fig. 2A,E;**

Fig. S2). Aromatic alcohol, amine and flavonoid use has been described for HCT-type

enzymes before (Sander and Petersen, 2011; Aymerick Eudes et al., 2016; Levsh et al.,

2016; Peng et al., 2016; Chiang et al., 2018a), but not anthocyanin acylation. No activity

was found with free sucrose or glucose for any of the 11 enzymes, despite evidence of

acylation on the glycoside of the anthocyanin. The liverwort enzyme MeHFT showed a

similarly diverse substrate utilization pattern **(Fig. 2A,E; Fig. S2)**. We note that

Cucurbitacin E is not known as a substrate for any known BAHD, but still, both SlHCT

and MeHFT showed significant activity with it. Another enzyme, EcCS which has only

been tested with its native substrate methylecgonine (alkaloid, mix class) (Schmidt et al.,

2015; A. Eudes et al., 2016), showed high specific activities with shikimate and quinate

(class A) – two substrates that are structurally not very similar to methylecgonine (MCS-

Tanimoto = 0.47, **File S2C**)**.** Methylecgonine is most similar to aliphatic alcohols (class B)

– and thus, EcCS can also be considered a class-promiscuous enzyme. These findings

show that some BAHDs have specialized *in vivo* substrate profiles, but have the capability

to explore a large region of the phytochemical space through secondary activities.

As opposed to the relatively promiscuous HCT/HQT-type enzymes, the AnAT-type

enzymes show a more specialized substrate usage. All four AnATs, exclusively used

other flavonoid and anthocyanin substrates under the testing conditions **(Fig. 2A,D; Fig.**

S2A,C). A previously characterized BAHD (*Dendranthema morifolium* 3-O

malonyltransferase, Dm3MAT3) known to malonylate cyanidin 3-O glucoside acylated other anthocyanins and a flavonoid containing a 3-O glycosylation, and to a lesser extent, 3,5-O glycosylation. At5MAT (5-O malonyltransferase from *Arabidopsis thaliana*) and Pc3MAT (3-O malonyltransferase from *Pericallis cruenta*) showed the same specificity with 5-O glycosylated and 3-O glycosylated anthocyanins, respectively. These results suggest that enzymes that have specialized for class D substrate acylation may have undergone adaptations constraining them from using class A/B substrates. Since we tested AnATs using malonyl-CoA donor and class A/B-utilizing enzymes with coumaroyl- and feruloyl-CoA, we also tested whether these enzymes can acylate aromatic alcohols and anthocyanins respectively, using non-native donors. We incubated Dm3MAT3 with coumaroyl-CoA and shikimate or quinate. For comparison, SlHQT and SlACT were incubated with malonyl-CoA using five different anthocyanin substrates. In both cases, no product formation was detected **(Fig. S2A,B)**. These results suggest that enzymes transforming substrates in classes A, B and D form two separate groups. While some class A/B enzymes are able to acylate class D substrates, no evidence was obtained for the reverse being true. This finding was robust to changes in donor CoAs, which may be expected considering AnAT-BAHDs are postulated to carry out an ordered bi-bi type of reaction, with the donor binding first (Shaw and Leslie, 1991; Tanner et al., 1999; Suzuki et al., 2003).

These observations of promiscuity raise questions about how substrate specificity – especially of broad substrate range enzymes – is maintained *in vivo*. One mechanism is restricting gene expression – EcCS is specifically expressed in palisade parenchyma of *E. coca,* which coincides with the highest concentrations of its product cocaine in the plant (Schmidt et al., 2015). Highly specific expression patterns have been shown before for BAHDs and other enzymes (e.g. St-Pierre et al., 1999; Kruse et al., 2017). Another HCT from *Plectranthus scutellarioides* (PsHCT2, previously *Coleus blumei*) maintains sufficient substrate specificity *in vivo* by using a conserved Arg residue near the active site as a handle that regulates substrates entry into the active site (Levsh et al., 2016) through inducing conformational change (Chiang et al., 2018a). Authors also found evidence for alternative substrate binding sites that increased BAHDs substrate permissiveness through diffusion of the non-native substrate towards the active site, while

maintaining its specialized native function (Chiang et al., 2018a). Other BAHDs may show

similar mechanistic adaptations to maintain *in vivo* substrate specificity.

Through these enzyme assays, we found existence of alternate, previously

undocumented scale of BAHD class-promiscuity, which ranged from strong (EcCS using

shikimate) to moderate (SlHCT using malvidin-3-O-glucoside) to weak (MeHFT using

cyanidin-3-O-glucoside) **(Fig. S2)**. Such latent capacity of BAHDs, like all enzyme

families, is critical for emergence of new reactions and new metabolites in cellular

environments. We only assayed BAHD promiscuity among the known substrate classes;

however, additional promiscuity existing for structurally related classes **(Fig. 1B)** or under

other conditions may influence BAHD incorporation into new pathways. Enzyme assays

of *C. braunii* HQT-like and MeHFT, coupled with the knowledge of lignin and cutin/suberin

as ancestral polymers in land plants (Renault et al., 2019; Philippe et al., 2020), also

suggested that enzymes utilizing substrates in class A/B were members of an ancestral

clade. To address this hypothesis and determine origins of other activities, we next

studied BAHD family evolution.

Functional innovation and expansion in the BAHD acyltransferase family in plants

Through an evolutionary analysis, we asked two questions. First, what was the

ancestral state of the BAHD family in plants? Second, how did the BAHD family expand

and diversify in the plant kingdom? We first identified BAHDs from 49 sequenced plant

genomes using the acyltransferase domain model (PF02458). BAHDs were detected in

1-5 copies in multiple green algal genomes, however, angiosperm and gymnosperm

genomes contain dozens to hundreds of copies, with non-seed plants showing

intermediate BAHD counts **(Fig. 3A,C)**. Ancestral state reconstruction of normalized

BAHD counts using a Bounded Brownian Motion model (Boucher and Démery, 2016)

revealed that the relative BAHD gene content began to increase upon origin of the land

plants **(Fig. 3A,B)**. The modal BAHD count per thousand genes rose from <1 in algae to

~1.3 in the ancestor of land plants, increasing to ~3 in the ancestor of seed plants. While

the values for vascular plants and euphyllophytes were intermediate, there was also less

confidence in the specific modal value at these internal nodes, shown by the broad spread

of their distributions. These patterns suggest a gradual increase of relative BAHD count

from the origin of land plants to the origin of seed plants, after which the relative counts

generally stabilized in the genome with some lineage-specific exceptions **(Fig. 3).** We

thus sought to determine the ancestral state of BAHD activities in the last common

ancestor of land plants, prior to their expansion.

Figure 3. Number of BAHD acyltransferases in different across the plant phylogeny. (A) Likelihood of the number of BAHDs per 1000 genes in the genome of different phylogenetic groups, as per the bound Brownian motion model. The corresponding nodes are indicated in **Fig. 3B**. **(B)** Species tree of the analyzed species with sequenced genomes. Color scale of the branches indicates the normalized BAHD count. Maximum extent of OG conservation for each substrate class is highlighted by a solid black line, while tentative assignment of emergence of activities is indicated by the dashed line. See **Fig. 4** and **Fig. 8** for more details. **(C)** Bar graph showing BAHD count per analyzed genome.

Ancestral state is typically obtained in two different ways. Performing activity
assays after ancestral sequence resurrection is one approach (Huang et al., 2016).
However, BAHDs are fast evolving enzymes that maintain activity and structural folds
despite undergoing large-scale sequence evolution. Due to such rapid sequence change,
there is very little confidence in amino acid state prediction in deep nodes **(**example: **File**
S6), and thus, ancestral sequence reconstruction or resurrection for deep nodes is not
possible. The second approach – predicting ancestral state based on analyses of
activities of extant enzymes – is also restricted due to poor confidence in many internal
nodes of the BAHD phylogeny and absence of accurate knowledge about extant
character states i.e. absence of knowledge about a substrate's utilization by an enzyme
could be truly an inability to use the substrate or simply an untested interaction. Thus, we
used a different approach, where we first constructed 765 BAHD OGs across the 49
sequenced plant genomes, of which 132 comprised of >2 members and 89 comprised of
members from >2 species (multi-species OGs). We then used sequence similarity to
assign biochemically characterized enzymes to OGs, thus roughly predicting the OGs
biochemical function at the substrate class utilization level. Depending on the breadth of
conservation of a specific OG, we inferred the deepest internal node in the species tree
likely housing the OG's associated function. The 136 BAHDs were mapped to only 47
OGs, suggesting that ~50% of the multi-species OGs still have unidentified functions.
While this method enables assignment of discrete states to specific ancestral nodes,
there are no probability estimates associated with the predictions, which is a caveat of
this approach. However, most internal node functional inferences were supported by
multiple characterized BAHDs with the same function **(File S4)**, thus providing confidence
in assigning the phylogenetic extent of each clade in the BAHD gene tree **(Fig. 4)**.

The characterized enzymes can be divided into seven clades with high bootstrap
support, of which four (clades 1-4) are the same as defined previously (D'Auria, 2006).
Clade V in D'Auria, 2006 was divided into three separate clades 5-7 **(Fig. S3)**. Three
clades containing HCT/HQT enzymes (clade 5a), alcohol acyltransferases (clade 7a) and
polyamine acyltransferases (clade 4a) were the most widely conserved, with orthologs
extending from angiosperms to liverworts **(Fig. 4)**. Clade 5a, most of whose members are
involved in phenylpropanoid pathway and lignin biosynthesis, is – based on branch

lengths – under the most purifying selection of all BAHDs **(Fig. S3)**, suggesting that
sequence similarity-based, substrate class-level functional prediction of unknown BAHDs
mapping to this clade will likely be accurate. Accurate class-level predictions may also be
possible for clades 1a/b (all but one anthocyanin/flavonoid acylating despite long branch
lengths) and 7a/7b (cuticular wax biosynthesis and slow-evolving), however, other clades
appear to have diverged rapidly at the sequence level as well as for substrate utilization
(Fig. S3). A deeper analysis of the sub-clades will be needed to determine their predictive
potential.

Combined with substrate preference patterns from our enzyme assays and
previous studies **(Figs. 2,4)**, these results suggest that aliphatic and aromatic alcohol and
amine acylating activities (clades 4a, 5a, 7a) were already established in the ancestor of
land plants **(Fig. 3)**. At ~1.3 BAHDs per 1000 genes in this ancestor, and given the
predicted gene content of *Chara*, mosses and liverworts, it is possible that ~15-30 BAHDs
may have existed in this ancestor, housing the three clades described above (with
additional uncharacterized/lineage-specific activities). Of these, only the aromatic alcohol
acylating activity was detected in the outgroup species *C. braunii*, providing more support
to the inference that at least this activity was present in the ancestor of all land plants and
perhaps in the LCA of charophytes-land plants. Given the possibility of specialization of
the enzyme in the lineage leading to *C. braunii*, occurrence of other activities (aromatic
amine, aliphatic alcohol and amine acylation) in this ancestor cannot be ruled out.

The evolution of the AnAT and terpenoid acylation activities is slightly more
complex. First, the AnAT- and terpenoid-specialized clades have OGs only containing
angiosperm species, but these activities exist in multiple distantly related BAHD clades
(clades 1,3,5,7 and 3,5,6,7, respectively). Second, these activities also exist in MeHFT
and SIHCT clades, whose OGs extend up to liverworts **(Fig. 2B,E, Fig. S2A,B)**. Both
observations combined together suggest that AnAT and terpenoid acylation activities may
have been accessible to BAHDs for a long time before their specialization in angiosperms.

Figure 4. Defining BAHD clades. The tree was rooted using the algal enzyme clade (Clade 0a). Maroon circles on branches refer to clades with bootstrap values > 70. Clades were first defined based on deepest, high-confidence monophyletic clades. Clades 1-4 are same as D'Auria et al, 2006 definitions **(Fig. S3)**, while Clade V from that study is divided into Clades 5-7 here, based on the above criterion. Sub-clades were further defined within each clade based on the extent of conservation of OGs across the plant phylogeny **(File S3)**. These clade/sub-clade definitions can be further expanded as new, uncharacterized BAHDs are characterized in the future. Solid circles in the concentric circles around the tree represent activities characterized in this and previous studies, with "gen" referring to genetically characterized BAHDs whose substrates were not considered. * AtEPS1 showes unusual reaction mechanisms for BAHDs and has an isochorismoyl-glutamate A pyruvoyl-glutamate lyase activity that produces salicylic acid.

Evolution of specialized BAHD activities

Orthology-based ancestral state reconstruction suggested that BAHDs accessed
new substrate classes or specialized in ancestrally accessible classes over their
expansion through gene duplication. Comparisons of OG sequences could thus help
identify specific residues that contributed to functional specialization for a given substrate
class. We first compared the HCT/HQT-like enzymes (clade 5a/b) to amine-acylating
enzymes (clade 4a), both clades being conserved across land plants. We identified 35
residues that were present in >70% of the tested sequences in clade 5a/b but had
completely switched to a different residue in >70% of the tested sequences in clade 4a
(File S5). Of these, only 2 – AtHCT F303L and R356D **(Fig. 5A)** – were close to the active
site, with R356 present in ~90% of clade 5a/b but 0% of clade 4a, being replaced by Asp
or Glu in a majority of clade 4a sequences. The effect of R356D switch was previously
described (Levsh et al., 2016; Chiang et al., 2018b) in several HCTs, which use shikimate
as its primary substrate but do not show activity with aromatic amine substrates. It was
found that this switch was able to convert this enzyme into using amine-containing
substrates. While this promiscuous activity exists in at least some *O*-acylating enzymes
(Fig. 2, Fig. S2), the ubiquity of the R356D mutation in *N*-acylating enzymes suggests
that this residue was critical for specialization towards positively charged substrates
despite alternative binding sites in the protein contributing to increased promiscuity
(Chiang et al., 2018a). For F303L, we found that while 95% of the HCT/HQT-like enzymes
have the Phe residue, this is completely reversed, with 95% of the amine-acylating
enzymes having a Leu residue. Although this residue was not experimentally tested, we
hypothesize that this position also plays an important role in specialization towards the
respective substrates.

Figure 5. Conserved residues in different clades and ancestral state reconstruction. (A) Alignment of clade 5a and clade 4a OG sequences. Highlighted in red is the Arg residue conserved in ~90% of clade 5a BAHDs that predominantly use aromatic alcohols. In green, the corresponding Asp, Glu or Asn residues in amine acylating BAHDs are shown, of which Asp is present in ~90% of clade 4a sequences. **(B)** Conserved TFFDxxW and YFGNC motif region in different clades of the BAHD phylogeny. Groups 1-5 are defined in **Figure S4B**. A 20 amino acid broad window is shown. **(C)** Ancestral sequence reconstruction of the conserved TFFDxxW (purple) and YFGNC (green) region. Large letters are residues with a posterior probability >80%. Smaller letters represent residues with a posterior probability <80%. Additional information can be found in **File S6**.

Using OG comparisons, we also looked at sequence differences between
HCT/HQT-like enzymes (clade 5a/b) and AnAT-like enzymes (clades 1a/b) by
supplementing the single-residue analysis with enrichment analysis of larger motifs given
the larger structural difference between the substrates. We focused on AnAT-like
enzymes, since they are the most widespread and well-characterized clades among
plants **(Fig. 4)**. The OGs of clades 1a/1b extend farthest back to angiosperms, which
suggests that the fixation of this activity occurred only in this lineage. This spread
corroborates with previous knowledge about evolution of the core anthocyanidin pathway
in seed plants (Davies et al., 2020; Piatkowski et al., 2020), which is further extended via

glycosylation, methylation and acylation. Anthocyanin acylation can improve the stability of the molecule to heat, higher pH, UV light, also providing an evolutionary advantage to its fixation in flowering plants.

Motif enrichment analysis revealed among others, two large, over-represented motifs in these clades (TFFDxxW: E-value=1.9e-248; YFGNC: E-value=4.5e-221**, Fig. 5B, Fig. S4A)**. Single-residue analysis also confirmed that two residues Trp36 and Cys320 contained in the conserved motifs, were highly conserved AnAT enzymes in comparison to other biochemically characterized BAHDs (100% vs 0% and 95% vs 4%, respectively) **(Fig. S4; Fig. S5A,B)**. Both these residues were closer to the catalytic His than other identified residues in the crystallized structure of Dm3MAT3 in complex with malonyl-CoA (PDB: 2E1T) (Unno et al., 2007). Ancestral sequence reconstruction was performed to determine when these residues appeared in the BAHD phylogeny. Over 80% of residues in the ancestor of all AnAT-type enzymes could not be predicted confidently (posterior probability<0.8), owing to the rapid sequence evolution of BAHDs **(File S6)**; however, both Trp36 and Cys320 were confidently placed in the ancestral node of clade 1a/b **(Fig. 5C**; posterior probability>0.95**)**. Emergence of these two residues was most likely preceded by a Tyr (<80% posterior probability) and an Ala, respectively, in the prior ancestral node **(Fig. 5C)**. These results suggest that the acquisition of Trp36 and Cys320 was important for the angiosperm-restricted specialization of the ancestrally accessible AnAT activity. We next performed molecular dynamic (MD) simulations to determine the role of these residues in the AnAT activity. MD simulations were performed using the wild-type Dm3MAT3 enzyme and by replacing the two residues with Ala, a catalytically neutral residue.

The role of Trp36 and Cys320 in anthocyanin malonyltransferase catalysis

The first step of the acyltransferase reaction involves proton abstraction from the cyanidin 3-*O*-glucoside (C3G) 6"-hydroxyl by the deprotonated, basic nitrogen of the His170 imidazole (**Fig. 6B**) (Unno et al., 2007). For successful intermolecular proton transfer to occur, the distance between these two atoms should be less than 4 Å to account for the longest possible hydrogen forming (Harris and Mildvan, 1999). Simulations revealed that the distance for C3G 6"-hydroxyl proton abstraction fell within

a 4 Å threshold for WT Dm3MaT3 for the entire 1 μs of production runs. After maintaining

a catalytically competent distance for the first 100 ns of simulation, the distance for proton

abstraction in the C320A mutant exceeded the 4 Å threshold. The W36A mutant never

achieved a distance satisfactory for catalysis to proceed **(Fig. 6A)**. This result suggested

that the W36A mutant would have a lower catalytic efficiency than wild-type and the

C320A mutant for both C3G and malonyl-CoA.

Figure 6. Distance calculations between the cyanidin 3-O glucosides 6"-OH hydrogen and the deprotonated, basic nitrogen of the His170 imidazole. (A) Distance calculation for WT (purple), W36A (red) and C320A (green) over 1000 ns of simulation time. **(B)** Illustration of the distance measured during the simulation. Active site organization is altered upon replacement of Trp36 and Cys320. Active site organization in the **(C)** WT protein, **(D)** the W36A variant, and **(E)** C320A variant during catalysis. Ala substitutions are colored and labeled as goldenrod in the appropriate systems.

We next assessed how C3G is repositioned by each of the Dm3MaT3 variants

(Fig. 6C-E). In wild-type Dm3MaT3, the glucose moiety of C3G maintained a tight

proximity to His170 within the active site, as it was flanked by bulky residues Trp36 and

Trp383 **(Fig. 6C)**. The benzene diol moiety of C3G stacked with Trp383 and the

chromenylium moiety stacked with Trp36. Incorporation of these bulky residues in the

proximal active site imposed a restriction on the movement of C3G, encouraging catalysis

by keeping it close to the reactive His170. Upon W36A substitution, the bulky Trp residue

was no longer present, which allowed C3G to sample other conformations within a more

open active site, while still remaining bound to the enzyme. Trp383 initially stacked with

aromatic portions of C3G in the WT, as seen in the Dm3MaT3 WT simulation. However,

in W36A, the glucose moiety of C3G was flipped from its starting pose, where backbone

hydrogens on the hexose ring demonstrate C-H⋯pi interactions with Trp383, positioned

the 6"-hydoxyl far from His170 **(Fig. 6D)**. Thus, the W36A mutation directly altered active

site organization, providing a straightforward justification for this Dm3MaT3 variant's

reduced k_{cat}/K_m.

While the C320A mutant initially maintained a native active site organization due

to the interactions between C3G and Trp36, C3G completely left and was excluded from

the active site by the end of the simulation **(Fig. 6A,E)**. Ala substitution altered the native

C-H⋯pi interactions between the backbone of residue 320 and the aromatic portion of

Phe32. In *apo* simulations, the distance between the backbone hydrogen of Ala320 and

the pi system of Phe32 was consistently less than 3 Å, but it experienced fluctuations

ranging between 2.5 and 6 Å during *holo* simulations **(Fig. S7A,B)**. For WT and W36A

variants, these distances were constant and unperturbed by the introduction of ligand,

suggesting that the basis of reduced activity in the C320A mutant is distinct from that

seen in the W36A mutant **(Fig. 6D)**.

Snapshots from simulation reveal that upon experiencing fluctuations in the

distance between the backbone alpha hydrogen of Ala320 and the aromatic portion of

Phe32, a series of stacking rearrangements occurred along the helix containing the

TFFDxxW motif **(Fig. S6C,G)**. First, Phe32 became destabilized due to inconsistent

backbone C-H⋯pi interactions with Ala320, resulting in increased edge-to-face stacking

between Phe31 and Phe32 **(Fig. S6A)**. Alternation of face-to-face and edge-to-face

stacking between Phe31 and Phe32 was seen throughout each of the *apo* and *holo*

systems simulated **(Fig. S7)**, but the extent of fluctuation between the two modes of

stacking, and the sampling of continuous no stacking interactions, was greatest for *holo*

C320A.

To recover stability within the C320A variant's TFFDxxW motif, Phe35 broke its

edge-to-face stacking with Trp36 and began to stack with Phe31 and Phe32 **(Fig. S6D,**

Fig. S7). While the W36A substitution prevented Phe35 from edge-to-face stacking with an active site tryptophan, Phe35 never stacked with neither Phe31 nor Phe32 **(Fig. S8)**. WT Dm3MaT3 also demonstrated virtually no Phe31-Phe35 nor Phe32-Phe35 stacking **(Fig. S9)**, suggesting these interactions were developed in response to instabilities caused by C320A substitution. Thus, as Phe35 sampled new stacking interactions with Phe31 and Phe32, Phe35-Trp36 stacking interactions were momentarily lost **(Fig. S6D)**. With the network of stacking interactions between Phe35, Trp36, and His170 disrupted, the active site lost the tight organization which is requisite for maintaining a catalytically-competent distance between His170 and the C3G 6"-hydroxyl (**Fig. S6E,F**). C3G then left the immediate vicinity of His170 and the TFFDxxW motif to bind elsewhere within the enzyme.

In contrast to the Ala substituted variants, wild-type Dm3MaT3 maintained an ordered active site which better maintained proximity to C3G and overall acyltransferase activity **(Fig. 6C)**. The stability and organization of the wild-type enzyme is underscored by consistent Phe31-Phe32 face-to-face stacking and Phe35-Trp36 edge-to-face stacking, neither of which were disrupted by Phe31-Phe35 or Phe32-Phe35 stacking at any point during the *holo* wild-type simulation **(Fig. S9)**.

The MD simulation results thus suggested that that the W36A variant would have inferior reaction kinetics in comparison to the wildtype due to absence of the bulky Trp36 that is important for keeping the C3G proximal to His170. Furthermore, we were interested in examining how severely the C320A variant activity would be affected, because the MD simulations demonstrated that C320A may position the acyl acceptor in a catalytically competent position, suggesting the possibility for some functionality despite mutation. In addition, for ~10% of the total simulation time (about 100 ns), the C320A mutant maintained C3G 6"-hydroxyl and the His170 imidazole within a distance of 4 Å (**Fig. 6A**). Therefore, we tested the effects of Ala replacement of Trp36 and Cys320 using site specific mutagenesis and subsequent *in vivo* enzyme assays.

The catalytic importance of the two residues in anthocyanin acyltransferase activity

We heterologously expressed and purified mutagenized Dm3MAT3 variants from

E. coli. These proteins showed similar folding behavior based on similar gel migration and

retention time in size-exclusion chromatography compared to the wild-type protein **(Fig.**

S11A,B), suggesting that the mutation did not affect protein folding; however, both

mutants substantially affected enzyme reaction kinetics when comparing specific

activities (**Fig. 7A,B,C**). Comparing the pseudo-first-order reaction kinetics of the wild-

type Dm3MAT3 enzyme with the W36A mutant revealed that the mutation did not

influence the acceptor K_m value (**Fig. 7D**) but drastically reduced turnover number (k_{cat})

and catalytic rate (k_{cat}/K_m) by ~97% and 97.5%, respectively (**Fig. 7E,F**). All three kinetic

estimates were significantly reduced for the acyl donor malonyl-CoA (**Fig. 7D-F**).

Figure 7. Enzyme activity and enzyme kinetics of Dm3MAT3 wildtype and mutants. (A) Specific activity of Dm3MAT3 wild-type, W36A and C320A mutant. The activity of the wild-type (WT) is set to 100 %. The error bar represents the standard deviation of three expression replicates which were each assayed in technical triplicates. **(B)** Michaelis-Menten curve for Dm3MAT3 WT (WT), Dm3MAT3 W36A (W36A), and Dm3MAT3 C320A (C320A) in dependence of cyanidin 3-O glucoside (C3G). **(C)** Michaelis-Menten curve for the malonyl-CoA donor. **(D)** Km values of WT, W36A and C320. **(E)** k_{cat} values. **(F)** k_{cat}/K_m values.

These changes in enzyme activity suggest that Trp36 is catalytically important for

the AnAT activity in clades 1a/b. The relatively similar acceptor K_m between wild-type and

W36A mutant suggested that the mutation did not affect acceptor substrate binding. Thus,

while the MD simulations suggested a quick C3G departure from the active site, the

molecule may still remain in the binding cavity due to interactions with other residues, and

exit normally. The kinetic assays – specifically the reduced k_{cat} and k_{cat}/K_m values for both

acceptor and donor – support MD simulation predictions that catalysis would be
drastically affected due to C3G departure from a catalytically competent distance with
His170. The K_m of the W36A mutant towards malonyl-CoA was also significantly reduced.
Considering AnAT-like acyltransferases are postulated to operate through a two-
substrate ordered bi-bi reaction mechanism with malonyl-CoA binding first (Suzuki et al.,
2003; Unno et al., 2007), we hypothesize that the reduced catalysis in W36A results in
the malonyl-CoA continually bound to the donor binding site, further resulting in a
saturation of the available donor sites at lower substrate concentration.

In the case of the C320A mutant, the observed effects were less severe than for
the W36A mutant (**Fig. 7A**), and the C320A still exhibited a k_{cat} ~47-74% of the WT (**Fig.**
7E). Compared to the wild-type, the C320A mutant showed an improved K_m towards C3G
(**Fig. 7D**). This lower K_m and the only slightly reduced k_{cat} resulted in a 50% improved
catalytic rate than WT (**Fig. 7F**). The C320A mutant showed an unchanged K_m and a
lower k_{cat} for malonyl-CoA. This resulted in a slightly less efficient enzyme with respect to
the donor (**Fig. 7F**).

These results support the simulation prediction that Cys320 plays a role in
optimizing the enzyme's activity rather than catalysis. While MD simulations indicated that
stability of the C3G bound form is reduced, this results in only a slightly decreased k_{cat}.
However, we postulate that the acceptor substrate remains in the substrate binding site
without catalysis, reducing the K_m to a greater extent than reduction in k_{cat}, thereby
mathematically increasing the catalytic efficiency. Combined together, these results
suggested that the presence of both Trp36 and Cys320 is necessary for optimal
anthocyanin malonyltransferase activity in clades 1a/b, explaining their conservation in
AnATs spread out over 150 million years of angiosperm evolution.

DISCUSSION

Evolution of functional diversity in large enzyme families that contribute to the
metabolic diversity in plants is still incompletely understood. In this study, we extensively
characterized the BAHD family to determine how gene duplication and promiscuity
contributed to the diversity of BAHD functions. Nine out of eleven BAHDs assayed in this

study were substrate-promiscuous **(Fig. 2; Fig. S2; File S2)**, and three (MeHFT, SlHCT,

EcCS) were class-promiscuous under the testing conditions **(Fig. 2A, Fig. 1)**. We only

tested nine substrate types (including glucose/sucrose), and the specialized enzymes

may still exhibit class-promiscuity with other, untested classes under different conditions.

More enzymes that cross class boundaries are known from previous studies **(Fig. 4)**.

Nonetheless, these observations suggest that while most BAHDs are not able to

discriminate between structurally related compounds (e.g. shikimate, quinate for SlHQT

or putrescine, agmatine for SlACT), they are able to differentiate between very different

chemical scaffolds. For individual enzymes, these adaptations for specialization may

constrain them and their duplicates from traversing large distances in the phytochemical

space. On the other hand, the class-promiscuity of multiple enzymes such as SlHCT,

AtHCT, SmHCT1, MeHFT, EcCS, alcohol acyltransferases provides a foundation on

which duplicates may traverse larger distances in the phytochemical space and "plug into"

emerging metabolic pathways. Since such class-promiscuity is not typically assayed, it

may be much more prevalent than currently known and may form an important basis for

metabolic diversification. Significant presence of class-promiscuity can also confound

evolutionary inferences and thus, functional annotation for enzyme family members.

Thus, more studies to determine if there are any rules for class-promiscuity are needed.

Such studies may also reveal new insights about the nature of selection on protein

structural features that enable promiscuity.

The integrative analysis of the AnAT activity can be interpreted following the

previously proposed potentiation-actualization-refinement model of emergence of new

functions (Blount et al., 2008). Our results suggest that the ability to acylate anthocyanins

was already actualized (first appeared/already existed) in ancestral BAHDs prior to land

plant evolution, or that the ancestral enzymes were potentiated to evolve the AnAT activity

through different routes. The refinement of the AnAT activity in clades 1a/b required

fixation of two residues, one of which (Trp36) is critical for positioning anthocyanins in the

active site and the second (Cys320) for positioning the first residue. Both these residues

together optimize the AnAT activity by affecting acceptor binding, however, they are likely

not sufficient to confer the AnAT activity. The median clade 1a/b AnAT identity is itself

~50%, hence identifying all residues necessary to transform a class A/B utilizing enzyme

to class D substrate utilizing enzyme is challenging. Comparative sequence and structural

analysis can help further identify regions in the protein that could be tested by

mutagenesis.

An important question we asked was whether the structural diversity of substrates

used across BAHDs is related to an ancestral ability to use these substrates or

emergence of new activities through multiple rounds of duplication-neo-functionalization.

Our results can be discussed in the context of two models of BAHD evolution **(Fig. 8)**. In

both models, the most likely ancestral activity in the root node prior to BAHD expansion

is aromatic alcohol acylation. Based on the fact that many HCT/HQT enzymes also

acylate aromatic amines **(Fig. 4)**, class A substrates as a whole are the likely ancestral

BAHD substrate space **(Fig. 8A,B)**. For other activities, strong evidence for their

existence exists only in the ancestor of all land plants. For example, in our enzyme assays

(Fig. 2A) or the accumulated database **(Fig. 4)**, we do not see any enzyme from any

orange/yellow (extending to LCA of dicot-liverwort and dicot-mosses, respectively) clades

acylating both aliphatic and aromatic alcohols, but these activities are seen together in

seed plant clades 1c, 3a, 6a, 5b, 7c/d. Also, many aliphatic alcohol acyltransferases can

also acylate monoterpenoids. Thus, while it is possible that the 1-5 root node BAHDs

acylated aliphatic alcohols and thus, monoterpenoids, we cannot be certain about this

inference. Finally, the presence of anthocyanin acylation is seen in MeHFT, SlHCT and

multiple distantly related clades in the BAHD phylogeny **(Fig. 4)**. While this could

represent multiple instances of convergent evolution, a simpler explanation is existence

of this activity at least among the dozen-odd BAHDs in the land plant ancestor as a

promiscuous activity. No evidence exists for the assignment of this activity to the root

node, but given presence of class-promiscuous enzymes, this possibility cannot be

completely ruled out **(Fig. 8B)**. These inferences reveal that most of the currently known

BAHD acyltransferase substrate space was already accessible by the time BAHDs

started drastically expanding, either as fixed or unfixed, promiscuous activities. Compared

to simpler prior models of individual enzyme evolution such as the "patchwork model"

(Jensen, 1976; Matsumura and Ellington, 2001) and the "specialization-generalization-

specialization" model (Aharoni et al., 2005), our study of enzyme family reveals a complex

picture where ancestral promiscuity played a central role in incorporation of member

Figure 8. Two models of evolution of BAHD multi-functionality. (A) Conservative model: Only aromatic alcohols and amines are acylated in the root node. First steps of expansion in the branch towards the land plants ancestor leads to emergence of new activities that are fixed in the ancestor (big circles) as well as other promiscuous activities that remain unfixed (smaller circles) **(B)** Relaxed model: The 1-5 BAHDs in the LCA of charophytes and land plants are capable of utilizing most of the substrates in the current known substrate space. Specializations evolve in lineage-specific duplications. The aliphatic alcohol acyltransferase enzyme is likely a different one than the aromatic acyltransferase enzyme in the root node ancestor. The first detected monoterpenoid activities appear in the green angiosperm-gymnosperm ancestral clade **(Fig. 4)**, however, strong secondary activities were also detected in MeHFT and SlHCT.

enzymes into new pathways. Both of these models likely apply to evolution of individual

BAHD enzymes.

Our inferences are based on ~140 BAHDs, and activities of many BAHDs are

unknown. For example, to the best of our knowledge in tomato, only 10 out of 100 BAHDs

have an experimentally determined function (**Table S2**) (Niggeweg et al., 2004; Goulet et

al., 2015; Fan et al., 2016) (including this study), and >80% of BAHDs have a generic

domain annotation. It is possible that uncharacterized BAHDs may reveal new primary

activities *in vitro* and *in vivo*. However, our meta-analysis suggests that – despite the

dangers of class-promiscuity – some clade relationships **(Fig. 4)** can provide a strong

functional signal. Combined with other enzyme attributes, these clade relationships can

be useful in predicting putative BAHD substrate class utilizations with greater accuracy

than simply a similarity-based approach. e.g. using machine learning (Mahood et al.,

2020). At the same time, among the 42 multi-species OGs not represented among

characterized BAHDs, many show broad conservation across angiosperms and vascular

plants. These uncharacterized OGs may indeed reveal novel BAHD activities. As new

substrate classes are identified, it would be informative to test – using methods similar to

those employed in this study – whether the ability to use those classes also exists across

the BAHD family.

Findings from this study also provide insights into the role of enzyme families in

metabolic diversification in plants. BAHDs are fast-evolving enzymes; even in the well-

conserved clade 5a, where all known enzymes are associated with substrates related to

lignin biosynthesis (**Fig. 4**), the median percent identity is only ~60%. The ability to use

aromatic alcohols such as shikimate and quinate exists in BAHDs that are only 30-40%

identical to the clade 5a BAHDs. In other words, these enzymes are robust in retaining

their aromatic alcohol acyltransferase activity despite changes to ~70% of their sequence.

Similar behavior is observed for acylsugar acyltransferases (Moghe et al., 2017) and

aliphatic alcohol acyltransferases **(Fig. 4)**. While being robust, the class-promiscuity of

BAHDs and thus, the ability to specialize in one of the classes via duplication, would also

make them evolvable. The paradox of a biological system being both robust and

evolvable at the same time has previously been addressed in detail (Wagner, 2005, 2008;

Bloom et al., 2006; Pigliucci, 2008; Tokuriki and Tawfik, 2009; Payne and Wagner, 2019);

these properties may enable enzyme families to become involved in newly emerging

metabolic pathways or detoxify harmful metabolites without compromising their core

activities. The examples of aromatic alcohol, anthocyanin and terpenoid acylating

enzymes – whose OGs have a much narrower phylogenetic spread than the predicted

spread of the biochemical activity – also highlight a common theme in metabolism, where

enzyme activities may exist promiscuously for millions of years prior to the actual

genome-level signatures of their fixation. While presence of substrate promiscuity is

helpful in this regard, having class-promiscuous enzymes may also offer an added

benefit. Presence of robust evolvable enzymes is likely an important feature of metabolic
networks and needs to be studied in greater detail in the context of biochemical evolution.

The present study describes an integrative analysis of enzyme evolution,
biochemistry and structure-function relationships that captures potentially emergent
behaviors of enzyme families in the form of robustness and evolvability. The described
analysis can serve as a template for characterizing class-promiscuity in enzyme families,
although more high-throughput means of analysis are needed. Our results also identify
patterns in duplication-divergence of BAHDs that can be explored in other large enzyme
families to determine their involvement in metabolic diversification in plants.

<u>MATERIALS AND METHODS</u>

Creation of a database of biochemically characterized enzyme activities

Biochemically characterized BAHD enzymes were gathered through an extensive
literature search. Only sequences belonging to the BAHD acyltransferase protein fold
(PFAM domain: PF02458) were considered. To ensure a high level of confidence for *in*
vitro activities and the resulting substrate-enzyme pairs, only enzymes that were subject
to *in vitro* biochemical assays were considered. For each enzyme, all tested acceptor and
donor substrates with their associated PubChem CID, the associated chemical structure,
cDNA and protein sequence as well as the species name from which the gene was
isolated were compiled **(File S2)**.

Generation of substrate similarity networks

Structures of known BAHD substrates were downloaded from the PubChem
database using the PubChem ID as structure-data file (sdf) format. Substrates not found
in the PubChem database were created using ChemDraw, exported in the MOL data
format and manually brought into sdf format. To calculate substrate similarity based on
the maximum common substructure, overlap coefficient (MCS-Overlap), the R packages
ChemmineR v3.34.1 and fmcsR v1.24.0 were used (Cao et al., 2008; Wang et al., 2013)
with default values, except for the time threshold for the comparison of two molecules set

to 12s. The similarity network was visualized using Cytoscape v3.8.0 (Shannon et al.,

2003). Additional plant specific compounds were downloaded from the Chemical Entities

of Biological Interest (ChEBI) database (Hastings et al., 2013) and only compounds were

chosen that were also represented in the plant-centric KNApSAcK database (Afendi et

al., 2012),

Identification of BAHD acyltransferases and OG assignment

For identification of BAHD acyltransferases from the analyzed genomes, we used

hmmsearch v3.1b2 (Potter et al., 2018) using the BAHD PFAM domain (PF02458) with

all default parameters except *cut_ga* as the hit significance threshold. OGs were

constructed using OrthoFinder v2.3.3 (Emms and Kelly, 2019) with default parameters

except an inflation parameter of 1.5 to make larger OGs. After defining OGs between

these species, we used blastp (Camacho et al., 2009) to assign biochemically

characterized enzymes to OGs. For each enzyme, its individual phylogenetic

conservation was determined based on the phylogenetic spread of its assigned OG **(File**

S4). An internal node in the species tree was assigned an activity if multiple characterized

enzymes shared their conservation up to that node.

Creating a time-calibrated species phylogeny for ancestral state reconstruction

While recent attempts to infer a well-sampled land plant phylogeny (Gitzendanner

et al. 2018) and estimate divergence times (Nie et al. 2019) of green plants exist, a time

calibrated phylogeny encompassing the breath of taxa included in our dataset does not.

In order to obtain one, we first obtained a time calibrated phylogeny from TimeTree, a

database of synthesized time calibrated studies (Kumar et al. 2017). Because not all taxa

were present in TimeTree we used 'proxy taxa' (closest relative represented in TimeTree)

to assign a position within the phylogenetic tree. **(File S7A)**. Using this approach, we

accounted for all taxa in our database with the exception of *Klebsormidium nitens*

(charophytic algae), which was manually added to the most likely position according to

(Leliaert et al., 2012). To obtain divergence time for this group, we recalibrated the

phylogeny using a penalized likelihood under a relaxed clock model (Sanderson, 2002),

a lambda=0 and 95% CI intervals for major clades, inferred from previous studies
(Magallón et al., 2015; Nie et al., 2020) **(File S7B)**, with the function chronos() in R
package ape v.5.4 (Sanderson, 2002; Paradis and Schliep, 2019).

BAHD family size evolution

In order to determine the expansion dynamics of the BAHD gene family, we
modeled the evolution of normalized BAHD gene counts. Counts were normalized against
predicted total number of coding sequences in each genome with a methionine start
residue that were longer than 100bp. If >30% of the gene models did not fit these criteria,
these genomes (9 gymnosperm, 2 red algae) were eliminated from further analyses. We
fitted normalized counts and our phylogeny using Evolutionary Brownian Motion (BM) and
Bounded Brownian Motion (BBM) type models and performed model selection using the
R package BBMV v2.1 with default parameters except those specified below (Boucher et
al. 2016). BBM is a special case of Brownian motion (BM) where values are constrained
between a minimum and a maximum value (Boucher et al. 2016). We used a minimum
bound of 0, as negative gene copy is biologically nonsensical, and a maximum bound of
0.02914. It was suggested that the maximum bound should be set to the largest value in
the dataset (in our data set *Petunia axillaris*: 0.00679) (Boucher et al. 2016). However,
this value is value may not realistically represent the upper bound of the gene count for
an existing gene family. Therefore, we used *Arabidopsis thaliana*, the best annotated
plant genome, to identify the largest gene family present, which was the EAR repressome
family with 403 members (https://www.arabidopsis.org/browse/genefamily/index.jsp;
accessed August 20, 2020). We normalized this value against the total number of gene
in the Arabidopsis genome (https://www.arabidopsis.org/browse/genefamily/index.jsp;
accessed August 20, 2020) and set the upper bound of the model at 2x the normalized
gene number in this family (upper normalized limit = 0.02914). We acknowledge the
potential existence of gene families in nature that are larger but assume that this limit
should properly account for a majority of potential gene families.

Plant material, RNA extraction and cDNA synthesis

Plant material from the cultivated tomato variety M82 was used for cloning tomato

BAHDs. Plants were grown in a growth chamber under constant light/dark (16 h/8 h)

regime at 24 ºC for 8 weeks until first flowers appeared. For RNA extraction, young leaves

were cut from a single plant using clean tweezers and scalpel, immediately flash frozen

using liquid nitrogen (liq. N2) and stored at -80 ºC until further use. Total RNA was

extracted using the E.Z.N.A Plant RNA Kit (Omega Bio-tek, Norcross, GA) as per

manufacturers protocol with on-column DNAse digestion. cDNA was synthesized using

the Protoscript II Reverse Transcriptase (New England Biolabs, Ipswich, MA [NEB]) with

Oligo-dT$_{17}$ primer **(Table S3)** at 45 ºC for one hour. After heat inactivation for 20 minutes

at 80 ºC, the reaction mix was diluted with four volumes of nuclease-free water and stored

at -20 ºC until further use.

774　**Amplification of candidate enzymes, cloning of expression constructs, and site-**
775　**specific mutagenesis**

For amplifying cDNA sequences of candidate BAHD enzymes for cloning, the Q5

Hot Start High Fidelity DNA Polymerase (NEB) with gene specific primers was used

(Table S3). To allow fast and easy cloning using the Gibson assembly, the primers

contained matching overlaps to the pET28b vector, with successful insertion resulting in

N-terminal fusion of the candidate enzyme with a 6x His tag. After successful

amplification, the PCR product was purified using the E.Z.N.A Cycle Pure Kit (Omega

Bio-tek, Norcross, GA) and eluted in 30 µl nuclease-free water. First, the pET28b vector

was linearized using BamHI and XhoI restriction enzymes (NEB) and purified using the

E.Z.N.A Cycle Pure Kit. Second, 50 ng of linearized vector was mixed with 100 ng of

purified PCR product and incubated with HiFi DNA Assembly Master Mix (NEB) according

the manufacturers protocol. After assembly, 3 µl of the reaction mix were used for

transformation of *E. coli* 10-beta (NEB) cells. The transformation mix was plated on Luria-

Bertani (LB) medium containing kanamycin (50 µg/ml) and streptomycin (50 µg/ml).

Grown colonies were screened using colony PCR with construct-specific primers. Positive

clones were confirmed using Sanger sequencing at the Cornell Institute of Biotechnology.

In case of Dm3MAT3, the full-length gene was codon-optimized, synthesized and cloned

into pET28b by Gene Universal (Newark, DE). For cloning the C. braunii BAHD, the *C.*

braunii genome sequence (Nishiyama et al., 2018) was used. Previous studies postulated

presence of a BAHD orthologous to HCT potentially involved in a progenitor of lignin

biosynthesis (de Vries et al., 2017; Renault et al., 2019) however, no experimentally

characterized functions are available. We chose two out of four *C. braunii* BAHD enzymes

(PFAM: PF02458) that contained an intact catalytic HxxxD motif. The gene was

synthesized in two overlapping (42 bp overlap) parts (gBLOCKS) (Integrated DNA

Technologies, Coralville, IA [IDT]) and assembled into the pET28b vector as described

above. All constructs were sequenced before transformation into *E. coli* Rosetta2 cells

(EMD-Millipore-Sigma, Burlington, MA) for heterologous expression. Site-specific

mutagenesis of Dm3MAT3 was conducted using the Q5 site-directed mutagenesis kit

(NEB) **(Table S3)**.

Heterologous expression* and *protein purification

806   Protein expression was induced in mid-log phase growing *E. coli* cells containing

the respective expression construct after addition of 0.5 mM isopropyl-β-D-

thiogalactopyranoside (IPTG). The cultures (300 ml) were incubated for 16 hours at 25°C

in LB medium supplemented with 50 µg/ml kanamycin and 50 µg/ml chloramphenicol

while shaking at 250 rpm. For protein extraction, the overnight grown cells were pelleted

by centrifugation for 15 minutes at 5000g and re-suspended in lysis buffer (50 mM

NaH_2PO_4, 300 mM NaCl, 10 mM imidazole, pH 8.0), to which lysozyme was added

(~1 mg/ml) and incubated for 1 hour at 4°C. Subsequent sonification (intensity 90%,

Lawson Scientific, Harrogate, United Kingdom) for 10 minutes was conducted to shear

remaining intact cells. After centrifugation for 30 min at 21,000 g to remove cell debris

and insoluble proteins the cell-free lysate was incubated with Ni-NTA-Agarose matrix

(Qiagen, Germantown, MD [Qiagen]) for 30 min. After removing the lysate, the matrix was

washed with 10 column volumes (cv) wash buffer (50 mM NaH_2PO_4, 300 mM NaCl,

20 mM imidazole, pH 8.0) and then added to a 2 ml spin column (BioRad, Hercules, CA)

equipped with a frit. The matrix was washed again with 5 cv wash buffer after which

proteins were eluted three to four times using one cv elution buffer (50 mM NaH_2PO_4,

50 mM NaCl, 250 mM imidazole, pH 8.0). Fractions with similar protein content were

combined and the protein concentration was determined in triplicate using the Bradford
method (Bradford, 1976).

Size exclusion chromatography

Protein was expressed from *E. coli* BL21 grown at 37°C with 50 µg/mL kanamycin
to select for desired cells. When A_{600} values reached approximately 0.4, cells were
induced with 0.1 M IPTG, cooled to 15°C, and incubated with shaking for 18 hours. Cells
were then pelleted with centrifugation, supernatant discarded, and cell pellets frozen at -
80°C until ready for next steps. Cell pellets were thawed on ice, then resuspended in lysis
buffer (50 mM tris pH 8, 20 mM imidazole, 500 mM NaCl, 10% glycerol, 1% Tween-20),
and lysed by sonication. Lysate was centrifuged to pellet debris, and the supernatant was
passed through 2 mL of Ni-NTA resin. Resin was washed with buffer (50 mM Tris pH 8,
20 mM imidazole, 500 mM NaCl, 10% glycerol), then protein eluted in a single fraction
(50 mM Tris pH 8, 250 mM imidazole, 500 mM NaCl, 10% glycerol). To the fraction
containing desired protein, 30 U of thrombin was added, the entire solution was
transferred to dialysis tubing, and dialyzed overnight in 1 L of wash buffer. The resulting
solution was removed from dialysis tubing and passed through a column containing 300
µL Ni-NTA resin and 200 µL Benzamidine-sepharose resin. The flowthrough was
collected and further purified by FPLC size-exclusion chromatography (Hi-Load 26/600
S-200 column, buffer contained 100 mM NaCl, 25 mM HEPES, pH 7.5).

Biochemical synthesis of p-coumaroyl-CoA and feruloyl-CoA

For enzyme assays using p-coumaroyl-CoA and feruloyl-CoA as acyl donor, the
donor was synthesized biochemically using 4-coumaroyl-CoA ligase (4CL) from tomato
as described previously (Beuerle and Pichersky, 2002). The cDNA of tomato 4CL was
amplified using gene specific primers **(Table S3)**, cloned into pET28b, heterologously
expressed, and purified as described above. To generate the CoAs, 5 mM MgCl2, 2 mM
ATP, 2 mM p-coumaric acid (TCI America, Portland, OR) or ferulic acid (MP Biomedicals,
Solon, OH), and 3 mM coenzyme A (MP Biomedicals) in 50 mM sodium phosphate buffer
(pH 8.0) were incubated for 24 hours at room temperature (Beuerle and Pichersky, 2002).
The reaction was stopped by heating the reaction mix to 70 ºC for 20 min. After

centrifugation at 21,000 g for 30 min to remove precipitated protein the supernatant was

transferred into a new reaction tube and stored at -20 ºC until further use. The final

concentration of activated donor was estimated to be 1.6 mM based on the initial

substrate concentrations and an assumed conversion rate of 80% (Beuerle and

Pichersky, 2002).

Enzyme assays

In general, enzyme assays used for screening the substrate promiscuity of

selected BAHD enzymes were performed in 50 µl reactions containing 50 mM sodium

phosphate buffer (pH 7.2), supplemented with 100 µM acceptor substrate, and 300 µM

donor substrate. The assay was started by the addition of 20 µg purified enzyme.

Preliminary enzyme assays with lower substrate concentrations did not show any

indications for substrate inhibition at such substrate concentrations. For each enzyme,

three replicates per substrate were performed and incubated for 1 hour. Reactions were

stopped by adding 100 µl of a mixture of isopropanol, acetonitrile, and water (ratio of 2:2:1

+ 0.1% (v/v) formic acid) containing 15 µM of the internal standard telmisartan. The

reaction mix was centrifuged for 10 min at 21,000 g to remove precipitated proteins,

transferred into LC vials, and stored at -20 ºC until LC-MS analysis. Mock controls using

purified protein extracts from empty vector transformed *E. coli* were run to exclude *E. coli*

background activity, for each of the used acceptor and donor combinations.

To test whether SlACT, CbHCT, and MeHFT can use additional substrates from

the aromatic alcohol, aliphatic and aromatic amine class not captured using 100 uM of

substrate, a modified version with higher substrate concentrations were used (assay

version 2). A 50 µl reaction containing 50 mM sodium phosphate buffer (pH 7.2), 1 mM

acceptor substrate, 300 µM coumaroyl-CoA was started using 20 µg of the respective

enzyme, incubated for 1 hour, stopped, and subsequently run on LC-MS as described

above.

Enzyme assay reactions for screening for enzyme activities were measured

individually on the LC-MS and each product was detected with specific PRM parameters

(see below). In order to determine more detailed enzyme kinetics for anthocyanin

acylating enzymes, a modified version of the above described enzyme assays was used

(assay version 3). Varied concentrations of cyanidin-3-O-glucoside (25, 50, 100, 200,

300 µM) and malonyl-CoA (25, 50, 100, 200, 300 µM) were used to determine enzyme

kinetics. The counter substrate was kept at a saturating concentration of 300 µM. Two

biological replicates (independent enzyme expression and purification) were used and

Km, Kcat and Vmax were determined using non-linear curve fitting in Prism 8 (GraphPad,

San Diego, CA). All enzyme reactions were measured under initial rate conditions and

stopped as described above. The wildtype enzyme reactions were incubated for 10

minutes and mutant reactions for 20 and 30 minutes for C320A and W36A variants,

respectively.

LC-MS measurements

LC-MS analysis was performed on a ThermoScientific Dionex Ultimate 3000 HPLC

equipped with an autosampler coupled to a ThermoScientific Q-Exactive HF Orbitrap

mass spectrometer using solvent A (water + formic acid (0.1% v/v)) and solvent B

(acetonitrile + formic acid (0.1%)) at a flow rate of 0.6 ml/min. Products of enzyme

reactions were detected with specific PRM methods using their predicted parent ion mass

in positive or negative ionization mode with an isolation window of 2 m/z **(Table S4)**.

Additional details of chromatographic and mass spectrometric methods are described in

Table S4 and S5. LC-MS data was analyzed with the ThermoScientific Dionex

Chromeleon 7 Chromatography Data System v7.2 software. Peaks were selected using

their specific masses (**Table S4**) and default peak detection parameters.

Phylogenetic analysis

Protein sequence alignment was generated using MAFFT v.7.453-with-extensions

(Katoh et al., 2002) using following parameters: *--maxiterate 1000 --genafpair --thread*

70. The alignment was inspected manually to ensure proper alignment e.g. by inspecting

that known motifs like the HxxxD and DFGWG motif are aligned properly. Afterwards, IQ-

Tree v.1.6.10 (Nguyen et al., 2015) was used to infer a phylogenetic tree using following

parameters: *-st AA -nt AUTO -ntmax 70 -b 1000 -m TEST* after model selection

(LG+F+G4) using ModelFinder (Kalyaanamoorthy et al., 2017). Tree visualization was
created using iTol v.5.6.2 (Letunic and Bork, 2019).

Identification of enriched motifs in anthocyanin acylating enzyme

For identifying enriched motifs in a specific clade of anthocyanin acylating
enzymes, we used MEME v.5.0.5 (Bailey et al., 2009) in discriminative mode using default
parameters but the maximum number of motifs to find set to 5. Positive and negative
examples were set as described in **Fig. S5**. The TFFDxxW and YFGNC sub-motifs were
selected for further analysis due to their high degree of conservation and their proximity
to the active site within the Dm3MAT3 protein. Clade-wise single residues were identified
using custom Python scripts.

Prediction of ancestral sequence of AnATs

Ancestral sequence reconstruction was performed using IQ-TREE v1.6.10
(Nguyen et al., 2015). Twenty randomly selected sequences from each of the three
orthologous groups representing clades 1a-c and 5a (outgroup) were combined together
with previously characterized BAHDs from these clades. All protein sequences were
aligned using MAFFT v7.453 and provided as input to IQ-TREE, which was run with
model selection and ancestral state reconstruction with 1000 standard bootstraps. The
optimal tree was obtained using the JTT+I+G4 model. Per-site posterior probabilities of
the ancestral state prediction were filtered using a threshold of 0.95, and the resultant
FASTA sequence at each ancestral node was extracted using a custom Python script.

Preparation of PDB structures for docking and molecular simulations

The *holo* structure of Dm3MaT3 bound to malonyl-CoA (MLC), the acyl-donor, was
retrieved from the Protein Data Bank (PDB: 2E1T) (Berman et al., 2000; Unno et al.,
2007).After treatment with the PROPKA-plugin in VMD to verify residue charge states at
pH 7.0, (Humphrey et al., 1996; Rostkowski et al., 2011) *holo*-Dm3MaT3 was then
submitted to the Solution Builder Input Generator in CHARMM-GUI (Jo et al., 2008).

Because 2E1T was crystallized as a dimer, only segment "A" of the protein was input to

Solution Builder. The N- and C-terminal residues were modeled and patched using the

ACE and CT3 terminal groupings, respectively. For the W36A and C320A Dm3MaT3

mutants, the respective point mutations were also made during pdb structure

manipulation. Resulting Dm3MaT3 models were solvated with 150 mM NaCl, neutralized,

and output in Amber format. PyMol 2.3.3 was then used to perform structural alignment

between the Dm3MaT3 models and the original 2E1T structure containing MLC so that

the donor molecule could be introduced in the bound pose to each Dm3MaT3 model

(Schrödinger, LLC, 2015). *Apo* simulations were prepared in a similar fashion but from

the 2E1U PDB file (Unno et al., 2007).

The MLC structure file was taken directly from the 2E1T PDB file and edited in the

Maestro 2017-3 Release for improved structural resolution by adding hydrogens not seen

in the crystal structure and revising charge states at pH 7.0 (Schrödinger, LLC, 2020).

Cyanidin 3-*O*-glucoside (C3G), the acyl-acceptor, was also edited in Maestro to reflect an

accurate charge state at pH 7.0 (Schrödinger, LLC, 2020). Both ligands were introduced

to antechamber and parameterized with the AM1-BCC charge model (Jakalian et al.,

2002). At pH 7.0, MLC bears a net charge of -5 while C3G has a net charge of 0, although

it is zwitterionic due to its oxonium moiety.

Docking of acyl acceptor into donor-containing Dm3MaT3 structures

Holo-Dm3MaT3 was converted into a receptor pdbqt file for docking via AutoDock4

command-line (Morris et al., 1998). The box for docking C3G into each *holo* Dm3MaT3

variant (wild-type, WT; and mutants W36A and C320A) was determined in VMD

(Humphrey et al., 1996). Autodock Vina v1.1.2 was then used for docking. A docking

procedure was then performed using the previously determined box size and an

exhaustiveness score of 8 (Trott and Olson, 2010). A low exhaustiveness score was used

to obtain greater diversity of bound poses. This procedure was repeated 30 times in order

to generate a greater number of different starting seeds for C3G, where bound poses for

each iteration were evaluated based on the proximity of the reactive cyanidin 3-*O*-

glucoside 6"-hydroxyl to be ≤ 4.0 Å from the thioate moiety of MLC (Unno et al., 2007),

as proximity between the two ligands would serve as a proxy for an intermediate step
occurring within the reaction.

From this docking procedure, 268 C3G poses were generated for WT Dm3MaT3,
270 poses for W36A Dm3MaT3, and 261 poses for C320A Dm3MaT3. Regardless of the
orientation of the docked C3G, the pose selected for MD system assembly was that which
had the most favorable energy among those with the shortest distance (less than or equal
to 4 Å) from the MLC thioate. The C3G poses selected for simulation had docking scores
of -5.8 kcal/mol (WT), -7.6 kcal/mol (W36A), and -6.4 kcal/mol (C320A) **(Fig. S10)**.

Molecular simulation setup

Each Dm3MaT3 system (WT, W36A, C320A) was then reassembled using the
holo-aligned protein and MLC structures, the AutoDock Vina output, 150 mM NaCl, and
~18,500 water molecules in Packmol 18.169 (Martínez et al., 2009). Parameterization
and periodic box conditions were then applied using tleap in Amber18 (Case et al., 2018).
All ligands were parameterized with GAFF2 (Wang et al., 2004), Dm3MaT3 variants with
the AMBER-FB15 (Wang et al., 2017), and water with the TIP3PFB forcefields (L.-P.
Wang et al., 2014), respectively. Ions were parameterized using the 2008 parameter set
developed Joung and Cheatham (Joung and Cheatham, 2008). Hydrogen mass
repartitioning (Hopkins et al., 2015) was then applied to the resulting files in ParmEd 3.2.0
(Case et al., 2018). *Apo* systems were prepared in an identical fashion.

The following conditions were then applied to all Dm3MaT3 systems during
initialization stages. Minimization was performed for 50000 cycles, where steepest
descent was used for the first 5000 and then conjugate gradient for the remaining 45000
cycles. Each system was heated from 0 to 300 K in the NVT ensemble for 5 ns. The
systems were then held at 300 K for 5 ns at NPT. A Berendsen thermostat and barostat
were used throughout heating and equilibration stages for efficiency (Braun et al., 2018),
where the pressure was maintained at 1 bar using isotropic scaling and temperature at
300 K (Berendsen et al., 1984). Each Dm3MaT3 variant and the bound ligands were
restrained during the heating simulation. Restraints were removed during the 50 ns
equilibration. The SHAKE algorithm was applied to all stages of initialization except for
minimization (Krautler et al., 2001), while the Particle Mesh Ewald method used for

treating long-range electrostatics at a 10 Å cutoff (Darden et al., 1993). A Langevin
thermostat was implemented for temperature maintenance and a Monte Carlo barostat
was implemented for pressure maintenance in the production runs (Loncharich et al.,
1992; Åqvist et al., 2004). Simulations were performed for a total of 1 µs for each of the
six systems (*holo* and *apo* versions of Dm3MaT3 WT, W36A, and C320A).

1009

Data analysis

Trajectories were visualized using VMD 1.9.3 whereas images were rendered
using Chimera 1.14 (Humphrey et al., 1996; Pettersen et al., 2004). Distance
measurements between atoms were conducted using MDTraj 1.9.3 (McGibbon et al.,
2015). Residue-residue face-to-face and edge-to-face stacking interactions were
determined using a script which was adapted from a previously reported analysis
(Ferreira de Freitas and Schapira, 2017). Briefly, the plane of an aromatic residue's pi
system was determined using the same atoms as in GetContacts (*GetContacts*, 2020).
The normal vector was then determined for each aromatic plane under inspection, where
the intersection between the two vectors was solved for. Solving for the angle between
the center of mass of one aromatic ring, the center of mass of another aromatic, and the
intersection of the rings' normal vectors, allowed for the supplementary angle, θ, to be
determined. Angle θ and distance cutoffs were then applied to determine the extent of
aromatic stacking (Ferreira de Freitas and Schapira, 2017; "Schrödinger – Knowledge
Base.," 2020), where exact stacking classifications are detailed in Supplementary
Information. All graphs were generated using Matplotlib 3.2.0 (Hunter, 2007).

1026

ACKNOWLEDGEMENTS

We thank Dr. John D'Auria, Dr. Toru Nakayama and Dr. Sangeeta Dhaubhadel for
providing BAHD constructs, Dr. Bennett Fox and Dr. Frank Schroeder for use of and
assistance with the LC-MS experiments. We thank Dr. Thomas Stegemann for support in
LC-MS method establishment, Elizabeth Mahood and Dr. Nicholas Santantonio for
support in establishing the substrate similarity pipeline, Dr. Hening Lin for helpful advice
regarding our biochemical experiments, Dr. Elisabeth Kaltenegger for helpful comments
on the manuscript, and Anna-Lena Sprick and Dr. Kai Fan for support in the laboratory.
We thank Dr. Florian Boucher for assistance with interpreting BBMV R package output.

1036

COMPETING INTEREST

1038   The authors declare that the research was conducted in the absence of any commercial
1039   or financial relationship that could be construed as a potential conflict of interest.

1040

FUNDING INFORMATION

This work was supported by the following funding sources:
• LK/GM: Deutsche Forschungsgemeinschaft award #411255989 (LK); Cornell startup
funds (GM)
• JMG/CS: NSF award #DGE-1650441
• JC: NSF REU award #DBI-1850796 to the Boyce Thompson Institute/Georg Jander
• AW/DS: Blue Waters sustained-petascale computing project (NSF awards OCI-
0725070 and ACI-1238993, the State of Illinois, and the National Geospatial-
Intelligence Agency).

1050

REFERENCES

Afendi FM, Okada T, Yamazaki M, Hirai-Morita A, Nakamura Y, Nakamura K, Ikeda S, Takahashi H, Altaf-Ul-Amin M, Darusman LK, Saito K, Kanaya S. 2012. KNApSAcK Family Databases: Integrated Metabolite–Plant Species Databases for Multifaceted Plant Research. *Plant Cell Physiol* **53**:e1–e1. doi:10.1093/pcp/pcr165

Aharoni A, Gaidukov L, Khersonsky O, Gould SM, Roodveldt C, Tawfik DS. 2005. The "evolvability" of promiscuous protein functions. *Nat Genet* **37**:73–76. doi:10.1038/ng1482

Aharoni A, Keizer LC, Bouwmeester HJ, Sun Z, Alvarez-Huerta M, Verhoeven HA, Blaas J, van Houwelingen AM, De Vos RC, van der Voet H. 2000. Identification of the SAAT gene involved in strawberry flavor biogenesis by use of DNA microarrays. *Plant Cell* **12**:647–661.

Åqvist J, Wennerström P, Nervall M, Bjelic S, Brandsdal BO. 2004. Molecular dynamics simulations of water and biomolecules with a Monte Carlo constant pressure algorithm. *Chem Phys Lett* **384**:288–294. doi:10.1016/j.cplett.2003.12.039

Bailey TL, Boden M, Buske FA, Frith M, Grant CE, Clementi L, Ren J, Li WW, Noble WS. 2009. MEME Suite: tools for motif discovery and searching. *Nucleic Acids Res* **37**:W202–W208. doi:10.1093/nar/gkp335

Bar-Even A, Noor E, Savir Y, Liebermeister W, Davidi D, Tawfik DS, Milo R. 2011. The Moderately Efficient Enzyme: Evolutionary and Physicochemical Trends Shaping Enzyme Parameters. *Biochemistry* **50**:4402–4410. doi:10.1021/bi2002289

Berendsen HJC, Postma JPM, van Gunsteren WF, DiNola A, Haak JR. 1984. Molecular dynamics with coupling to an external bath. *J Chem Phys* **81**:3684–3690. doi:10.1063/1.448118

Berman HM, Westbrook J, Feng Z, Gilliland G, Bhat TN, Weissig H, Shindyalov IN, Bourne PE. 2000. The Protein Data Bank. *Nucleic Acids Res* **28**:235–242. doi:10.1093/nar/28.1.235

Beuerle T, Pichersky E. 2002. Enzymatic Synthesis and Purification of Aromatic Coenzyme A Esters. *Anal Biochem* **302**:305–312. doi:10.1006/abio.2001.5574

Bloom JD, Labthavikul ST, Otey CR, Arnold FH. 2006. Protein stability promotes evolvability. *Proc Natl Acad Sci* **103**:5869–5874. doi:10.1073/pnas.0510098103

Blount ZD, Borland CZ, Lenski RE. 2008. Historical contingency and the evolution of a key innovation in an experimental population of *Escherichia coli*. *Proc Natl Acad Sci* **105**:7899–7906. doi:10.1073/pnas.0803151105

Boucher FC, Démery V. 2016. Inferring Bounded Evolution in Phenotypic Characters from Phylogenetic Comparative Data. *Syst Biol* **65**:651–661. doi:10.1093/sysbio/syw015

Bradford MM. 1976. A rapid and sensitive method for the quantitation of microgram
quantities of protein utilizing the principle of protein-dye binding. *Anal Biochem*
72:248–254. doi:10.1016/0003-2697(76)90527-3

Braun E, Gilmer J, Mayes HB, Mobley DL, Monroe JI, Prasad S, Zuckerman DM. 2018.
Best Practices for Foundations in Molecular Simulations [Article v1.0]. *Living J*
Comput Mol Sci **1**:5957. doi:10.33011/livecoms.1.1.5957

Camacho C, Coulouris G, Avagyan V, Ma N, Papadopoulos J, Bealer K, Madden TL.
2009. BLAST+: architecture and applications. *BMC Bioinformatics* **10**:421.
doi:10.1186/1471-2105-10-421

Cao Y, Charisi A, Cheng L-C, Jiang T, Girke T. 2008. ChemmineR: a compound mining
framework for R. *Bioinformatics* **24**:1733–1734. doi:10.1093/bioinformatics/btn307

Case DA, Ben-Shalom IY, Brozell SR, Cerutti DS, Cheatham TEI, Cruziero VWD, Darden
TA, Duke RE, Ghoreishi D, Gilson MK, Gohlke H, Goetz AW, Greene D, Harris R,
Homeyer N, Izadi S, Kovalenko A, Kurtzman T, Lee T-S, LeGrand S, Li P, Lin C,
Liu J, Luchko T, Luo R, Mermelstein DJ, Merz KM, Miao Y, Monard G, Nguyen C,
Nguyen H, Omelyan I, Onufriev A, Pan F, Qi R, Roe DR, Roitberg A, Sagui C,
Schott-Verdugo S, Shen J, Simmerling CL, Smith J, Salomon-Ferrer R, Swails J,
Walker RC, Wang J, Wei H, Wolf RM, Wu X, Xiao L, York DM, Kollman PA. 2018.
AMBER. San Francisco, CA: University of California, San Francisco.

Cheng S, Xian W, Fu Y, Marin B, Keller J, Wu T, Sun W, Li X, Xu Y, Zhang Y, Wittek S,
Reder T, Günther G, Gontcharov A, Wang S, Li L, Liu X, Wang J, Yang H, Xu X,
Delaux P-M, Melkonian B, Wong GK-S, Melkonian M. 2019. Genomes of Subaerial
Zygnematophyceae Provide Insights into Land Plant Evolution. *Cell* **179**:1057-
1067.e14. doi:10.1016/j.cell.2019.10.019

Chiang Y-C, Levsh O, Kei Lam C, Weng J-K, Wang Y. 2018a. Active Site Dynamics and
Substrate Permissiveness of Hydroxylcinnamoyltransferase (HCT). *Biophys J*
114:583a–584a. doi:10.1016/j.bpj.2017.11.3193

Chiang Y-C, Levsh O, Lam CK, Weng J-K, Wang Y. 2018b. Structural and dynamic basis
of substrate permissiveness in hydroxycinnamoyltransferase (HCT). *PLOS*
Comput Biol **14**:e1006511. doi:10.1371/journal.pcbi.1006511

Copley SD. 2015. An evolutionary biochemist's perspective on promiscuity. *Trends*
Biochem Sci **40**:72–78. doi:10.1016/j.tibs.2014.12.004

Darden T, York D, Pedersen L. 1993. Particle mesh Ewald: An N·log(N) method for Ewald
sums in large systems. *J Chem Phys* **98**:10089–10092. doi:10.1063/1.464397

D'Auria JC. 2006. Acyltransferases in plants: a good time to be BAHD. *Curr Opin Plant*
Biol **9**:331–340. doi:10.1016/j.pbi.2006.03.016

Davies KM, Jibran R, Zhou Y, Albert NW, Brummell DA, Jordan BR, Bowman JL, Schwinn
KE. 2020. The Evolution of Flavonoid Biosynthesis: A Bryophyte Perspective.
Front Plant Sci **11**:7. doi:10.3389/fpls.2020.00007

de Vries J, de Vries S, Slamovits CH, Rose LE, Archibald JM. 2017. How Embryophytic
is the Biosynthesis of Phenylpropanoids and their Derivatives in Streptophyte
Algae? *Plant Cell Physiol* **58**:934–945. doi:10.1093/pcp/pcx037

Donoghue P, Paps J. 2020. Plant Evolution: Assembling Land Plants. *Curr Biol* **30**:R81–
R83. doi:10.1016/j.cub.2019.11.084

Dudareva N, D'Auria JC, Nam KH, Raguso RA, Pichersky E. 1998. Acetyl-
CoA:benzylalcohol acetyltransferase – an enzyme involved in floral scent
production in *Clarkia breweri*. *Plant J* **14**:297–304. doi:10.1046/j.1365-
313X.1998.00121.x

Emms DM, Kelly S. 2019. OrthoFinder: phylogenetic orthology inference for comparative
genomics. *bioRxiv* 466201. doi:10.1101/466201

Eudes A., Mouille M, Robinson DS, Benites VT, Wang G, Roux L, Tsai YL, Baidoo EE,
Chiu TY, Heazlewood JL, Scheller HV, Mukhopadhyay A, Keasling JD, Deutsch
S, Loque D. 2016. Exploiting members of the BAHD acyltransferase family to
synthesize multiple hydroxycinnamate and benzoate conjugates in yeast. *Microb*
Cell Fact **15**:198. doi:10.1186/s12934-016-0593-5

Eudes Aymerick, Pereira JH, Yogiswara S, Wang G, Teixeira Benites V, Baidoo EEK,
Lee TS, Adams PD, Keasling JD, Loqué D. 2016. Exploiting the Substrate
Promiscuity of Hydroxycinnamoyl-CoA:Shikimate Hydroxycinnamoyl Transferase
to Reduce Lignin. *Plant Cell Physiol* **57**:568–579. doi:10.1093/pcp/pcw016

Fan P, Miller AM, Schilmiller AL, Liu X, Ofner I, Jones AD, Zamir D, Last RL. 2016. In
vitro reconstruction and analysis of evolutionary variation of the tomato
acylsucrose metabolic network. *Proc Natl Acad Sci* **113**:E239–E248.
doi:10.1073/pnas.1517930113

Ferreira de Freitas R, Schapira M. 2017. A systematic analysis of atomic protein–ligand
interactions in the PDB. *MedChemComm* **8**:1970–1981.
doi:10.1039/C7MD00381A

Fujiwara Hiroyuki, Tanaka Y, Fukui Y, Ashikari T, Yamaguchi M, Kusumi T. 1998.
Purification and characterization of anthocyanin 3-aromatic acyltransferase from
Perilla frutescens. *Plant Sci* **137**:87–94. doi:10.1016/S0168-9452(98)00119-8

Fujiwara H, Tanaka Y, Fukui Y, Nakao M, Ashikari T, Kusumi T. 1997. Anthocyanin 5-
aromatic acyltransferase from *Gentiana triflora*: purification, characterization and
its role in anthocyanin biosynthesis. *Eur J Biochem* **249**:45–51.

Fujiwara H., Tanaka Y, Yonekura-Sakakibara K, Fukuchi-Mizutani M, Nakao M, Fukui Y,
Yamaguchi M, Ashikari T, Kusumi T. 1998. cDNA cloning, gene expression and
subcellular localization of anthocyanin 5-aromatic acyltransferase from *Gentiana*
triflora. *Plant J* **16**:421–31. doi:10.1046/j.1365-313x.1998.00312.x

Ganko EW, Meyers BC, Vision TJ. 2007. Divergence in Expression between Duplicated
Genes in Arabidopsis. *Mol Biol Evol* **24**:2298–2309. doi:10.1093/molbev/msm158

GetContacts. 2020.

Goulet C, Kamiyoshihara Y, Lam NB, Richard T, Taylor MG, Tieman DM, Klee HJ. 2015.
Divergence in the Enzymatic Activities of a Tomato and *Solanum pennellii* Alcohol
Acyltransferase Impacts Fruit Volatile Ester Composition. *Mol Plant*, Plant
Metabolism and Synthetic Biology **8**:153–162. doi:10.1016/j.molp.2014.11.007

Harris TK, Mildvan AS. 1999. High-Precision Measurement of Hydrogen Bond Lengths in
Proteins by Nuclear Magnetic Resonance Methods. *Proteins Struct Funct*
Bioinforma **35**:275–282. doi:https://doi.org/10.1002/(SICI)1097-
0134(19990515)35:3<275::AID-PROT1>3.0.CO;2-V

Hastings J, de Matos P, Dekker A, Ennis M, Harsha B, Kale N, Muthukrishnan V, Owen
G, Turner S, Williams M, Steinbeck C. 2013. The ChEBI reference database and
ontology for biologically relevant chemistry: enhancements for 2013. *Nucleic Acids*
Res **41**:D456–D463. doi:10.1093/nar/gks1146

Hopkins CW, Le Grand S, Walker RC, Roitberg AE. 2015. Long-Time-Step Molecular
Dynamics through Hydrogen Mass Repartitioning. *J Chem Theory Comput*
11:1864–1874. doi:10.1021/ct5010406

Huang R, O'Donnell AJ, Barboline JJ, Barkman TJ. 2016. Convergent evolution of
caffeine in plants by co-option of exapted ancestral enzymes. *Proc Natl Acad Sci*
113:10613–10618. doi:10.1073/pnas.1602575113

Humphrey W, Dalke A, Schulten K. 1996. VMD: Visual molecular dynamics. *J Mol Graph*
14:33–38. doi:10.1016/0263-7855(96)00018-5

Hunter JD. 2007. Matplotlib: A 2D Graphics Environment. *Comput Sci Eng* **9**:90–95.
doi:10.1109/MCSE.2007.55

Jakalian A, Jack DB, Bayly CI. 2002. Fast, efficient generation of high-quality atomic
charges. AM1-BCC model: II. Parameterization and validation. *J Comput Chem*
23:1623–1641. doi:10.1002/jcc.10128

Jensen RA. 1976. Enzyme recruitment in evolution of new function. *Annu Rev Microbiol*
30:409–425. doi:10.1146/annurev.mi.30.100176.002205

Jo S, Kim T, Iyer VG, Im W. 2008. CHARMM-GUI: A web-based graphical user interface
for CHARMM. *J Comput Chem* **29**:1859–1865. doi:10.1002/jcc.20945

Joung IS, Cheatham TE. 2008. Determination of Alkali and Halide Monovalent Ion
Parameters for Use in Explicitly Solvated Biomolecular Simulations. *J Phys Chem*
B **112**:9020–9041. doi:10.1021/jp8001614

Kalyaanamoorthy S, Minh BQ, Wong TKF, von Haeseler A, Jermiin LS. 2017.
ModelFinder: fast model selection for accurate phylogenetic estimates. *Nat*
Methods **14**:587–589. doi:10.1038/nmeth.4285

Katoh K, Misawa K, Kuma K, Miyata T. 2002. MAFFT: a novel method for rapid multiple
sequence alignment based on fast Fourier transform. *Nucleic Acids Res* **30**:3059–
3066.

Krautler V, van Gunsteren WF, Hunenberger PH. 2001. A fast SHAKE algorithm to solve
distance constraint equations for small molecules in molecular dynamics
simulations. *J Comput Chem* **22**:501–508.

Kreis W, Munkert J. 2019. Exploiting enzyme promiscuity to shape plant specialized
metabolism. *J Exp Bot* **70**:1435–1445. doi:10.1093/jxb/erz025

Kruse LH, Stegemann T, Sievert C, Ober D. 2017. Identification of a Second Site of
Pyrrolizidine Alkaloid Biosynthesis in Comfrey to Boost Plant Defense in Floral
Stage. *Plant Physiol* **174**:47–55. doi:10.1104/pp.17.00265

Leliaert F, Smith DR, Moreau H, Herron MD, Verbruggen H, Delwiche CF, Clerck OD.
2012. Phylogeny and Molecular Evolution of the Green Algae. *Crit Rev Plant Sci*
31:1–46. doi:10.1080/07352689.2011.615705

Letunic I, Bork P. 2019. Interactive Tree Of Life (iTOL) v4: recent updates and new
developments. *Nucleic Acids Res* **47**:W256–W259. doi:10.1093/nar/gkz239

Levsh O, Chiang Y-C, Tung CF, Noel JP, Wang Y, Weng J-K. 2016. Dynamic
conformational states dictate selectivity toward the native substrate in a substrate-
permissive acyltransferase. *Biochemistry* **55**:6314–6326.

Loncharich RJ, Brooks BR, Pastor RW. 1992. Langevin dynamics of peptides: the
frictional dependence of isomerization rates of N-acetylalanyl-N'-methylamide.
Biopolymers **32**:523–535. doi:10.1002/bip.360320508

Magallón S, Gómez-Acevedo S, Sánchez-Reyes LL, Hernández-Hernández T. 2015. A
metacalibrated time-tree documents the early rise of flowering plant phylogenetic
diversity. *New Phytol* **207**:437–453. doi:10.1111/nph.13264

Mahood E, Kruse L, Moghe G. 2020. Machine learning: A powerful tool for gene function
prediction in plants. *Appl Plant Sci* **8**:e11376–e11376. doi:10.1002/aps3.11376

Martínez L, Andrade R, Birgin EG, Martínez JM. 2009. PACKMOL: A package for building
initial configurations for molecular dynamics simulations. *J Comput Chem*
30:2157–2164. doi:10.1002/jcc.21224

Matsumura I, Ellington AD. 2001. *In vitro* evolution of beta-glucuronidase into a beta-
galactosidase proceeds through non-specific intermediates. *J Mol Biol* **305**:331–
339. doi:10.1006/jmbi.2000.4259

McGibbon RT, Beauchamp KA, Harrigan MP, Klein C, Swails JM, Hernández CX,
Schwantes CR, Wang L-P, Lane TJ, Pande VS. 2015. MDTraj: A Modern Open
Library for the Analysis of Molecular Dynamics Trajectories. *Biophys J* **109**:1528–
1532. doi:10.1016/j.bpj.2015.08.015

Moghe GD, Hufnagel DE, Tang H, Xiao Y, Dworkin I, Town CD, Conner JK, Shiu S-H.
2014. Consequences of Whole-Genome Triplication as Revealed by Comparative
Genomic Analyses of the Wild Radish *Raphanus raphanistrum* and Three Other
Brassicaceae Species. *Plant Cell Online* **26**:1925–1937.
doi:10.1105/tpc.114.124297

Moghe GD, Leong BJ, Hurney SM, Daniel Jones A, Last RL. 2017. Evolutionary routes
to biochemical innovation revealed by integrative analysis of a plant-defense
related specialized metabolic pathway. *eLife* **6**:e28468. doi:10.7554/eLife.28468

Morris GM, Goodsell DS, Halliday RS, Huey R, Hart WE, Belew RK, Olson AJ. 1998.
Automated docking using a Lamarckian genetic algorithm and an empirical binding

free energy function. *J Comput Chem* **19**:1639–1662. doi:http://doi:
10.1002/(SICI)1096-987X(19981115)19:14<1639::AID-JCC10>3.0.CO;2-B

Nguyen L-T, Schmidt HA, von Haeseler A, Minh BQ. 2015. IQ-TREE: A Fast and Effective
Stochastic Algorithm for Estimating Maximum-Likelihood Phylogenies. *Mol Biol*
Evol **32**:268–274. doi:10.1093/molbev/msu300

Nie Y, Foster CSP, Zhu T, Yao R, Duchêne DA, Ho SYW, Zhong B. 2020. Accounting for
Uncertainty in the Evolutionary Timescale of Green Plants Through Clock-
Partitioning and Fossil Calibration Strategies. *Syst Biol* **69**:1–16.
doi:10.1093/sysbio/syz032

Niggeweg R, Michael AJ, Martin C. 2004. Engineering plants with increased levels of the
antioxidant chlorogenic acid. *Nat Biotechnol* **22**:746–754. doi:10.1038/nbt966

Nishiyama T, Sakayama H, de Vries J, Buschmann H, Saint-Marcoux D, Ullrich KK, Haas
FB, Vanderstraeten L, Becker D, Lang D, Vosolsobě S, Rombauts S, Wilhelmsson
PKI, Janitza P, Kern R, Heyl A, Rümpler F, Villalobos LIAC, Clay JM, Skokan R,
Toyoda A, Suzuki Y, Kagoshima H, Schijlen E, Tajeshwar N, Catarino B,
Hetherington AJ, Saltykova A, Bonnot C, Breuninger H, Symeonidi A,
Radhakrishnan GV, Van Nieuwerburgh F, Deforce D, Chang C, Karol KG, Hedrich
R, Ulvskov P, Glöckner G, Delwiche CF, Petrášek J, Van de Peer Y, Friml J, Beilby
M, Dolan L, Kohara Y, Sugano S, Fujiyama A, Delaux P-M, Quint M, Theißen G,
Hagemann M, Harholt J, Dunand C, Zachgo S, Langdale J, Maumus F, Van Der
Straeten D, Gould SB, Rensing SA. 2018. The *Chara* Genome: Secondary
Complexity and Implications for Plant Terrestrialization. *Cell* **174**:448-464.e24.
doi:10.1016/j.cell.2018.06.033

Notebaart RA, Szappanos B, Kintses B, Pál F, Györkei Á, Bogos B, Lázár V, Spohn R,
Csörgő B, Wagner A, Ruppin E, Pál C, Papp B. 2014. Network-level architecture
and the evolutionary potential of underground metabolism. *Proc Natl Acad Sci*
111:11762–11767. doi:10.1073/pnas.1406102111

Panchy N, Lehti-Shiu M, Shiu S-H. 2016. Evolution of gene duplication in plants. *Plant*
Physiol **171**:2294–2316. doi:10.1104/pp.16.00523

Paradis E, Schliep K. 2019. ape 5.0: an environment for modern phylogenetics and
evolutionary analyses in R. *Bioinformatics* **35**:526–528.
doi:10.1093/bioinformatics/bty633

Payne JL, Wagner A. 2019. The causes of evolvability and their evolution. *Nat Rev Genet*
20:24–38. doi:10.1038/s41576-018-0069-z

Peng M, Gao Y, Chen W, Wang W, Shen S, Shi J, Wang C, Zhang Y, Zou L, Wang S,
Wan J, Liu X, Gong L, Luo J. 2016. Evolutionarily Distinct BAHD N-
Acyltransferases Are Responsible for Natural Variation of Aromatic Amine
Conjugates in Rice. *Plant Cell* **28**:1533–50. doi:10.1105/tpc.16.00265

Pettersen EF, Goddard TD, Huang CC, Couch GS, Greenblatt DM, Meng EC, Ferrin TE.
2004. UCSF Chimera--a visualization system for exploratory research and
analysis. *J Comput Chem* **25**:1605–1612. doi:10.1002/jcc.20084

Philippe G, Sørensen I, Jiao C, Sun X, Fei Z, Domozych DS, Rose JK. 2020. Cutin and
suberin: assembly and origins of specialized lipidic cell wall scaffolds. *Curr Opin*
Plant Biol, Physiology and Metabolism **55**:11–20. doi:10.1016/j.pbi.2020.01.008

Piatkowski BT, Imwattana K, Tripp EA, Weston DJ, Healey A, Schmutz J, Shaw AJ. 2020.
Phylogenomics reveals convergent evolution of red-violet coloration in land plants
and the origins of the anthocyanin biosynthetic pathway. *Mol Phylogenet Evol*
151:106904. doi:10.1016/j.ympev.2020.106904

Pigliucci M. 2008. Is evolvability evolvable? *Nat Rev Genet* **9**:75–82. doi:10.1038/nrg2278

Potter SC, Luciani A, Eddy SR, Park Y, Lopez R, Finn RD. 2018. HMMER web server:
2018 update. *Nucleic Acids Res* **46**:W200–W204. doi:10.1093/nar/gky448

Renault H, Werck-Reichhart D, Weng J-K. 2019. Harnessing lignin evolution for
biotechnological applications. *Curr Opin Biotechnol*, Food Biotechnology • Plant
Biotechnology **56**:105–111. doi:10.1016/j.copbio.2018.10.011

Rostkowski M, Olsson MH, Søndergaard CR, Jensen JH. 2011. Graphical analysis of pH-
dependent properties of proteins predicted using PROPKA. *BMC Struct Biol* **11**:6.
doi:10.1186/1472-6807-11-6

Sander M, Petersen M. 2011. Distinct substrate specificities and unusual substrate
flexibilities of two hydroxycinnamoyltransferases, rosmarinic acid synthase and
hydroxycinnamoyl-CoA:shikimate hydroxycinnamoyl-transferase, from *Coleus*
blumei Benth. *Planta* **233**:1157–71. doi:10.1007/s00425-011-1367-2

Sanderson MJ. 2002. Estimating Absolute Rates of Molecular Evolution and Divergence
Times: A Penalized Likelihood Approach. *Mol Biol Evol* **19**:101–109.
doi:10.1093/oxfordjournals.molbev.a003974

Schmidt GW, Jirschitzka J, Porta T, Reichelt M, Luck K, Torre JCP, Dolke F, Varesio E,
Hopfgartner G, Gershenzon J, D'Auria JC. 2015. The Last Step in Cocaine
Biosynthesis Is Catalyzed by a BAHD Acyltransferase. *Plant Physiol* **167**:89–101.
doi:10.1104/pp.114.248187

Schnable JC, Springer NM, Freeling M. 2011. Differentiation of the maize subgenomes
by genome dominance and both ancient and ongoing gene loss. *Proc Natl Acad*
Sci **108**:4069–4074. doi:10.1073/pnas.1101368108

Schrödinger – Knowledge Base. 2020.

Schrödinger, LLC. 2020. Schrödinger Release 2020-1: Maestro.

Schrödinger, LLC. 2015. The AxPyMOL Molecular Graphics Plugin for Microsoft
PowerPoint, Version 1.8.

Shannon P, Markiel A, Ozier O, Baliga NS, Wang JT, Ramage D, Amin N, Schwikowski
B, Ideker T. 2003. Cytoscape: A Software Environment for Integrated Models of
Biomolecular Interaction Networks. *Genome Res* **13**:2498–2504.
doi:10.1101/gr.1239303

Shaw WV, Leslie AGW. 1991. Chloramphenicol Acetyltransferase. *Annu Rev Biophys*
Biophys Chem **20**:363–386. doi:10.1146/annurev.bb.20.060191.002051

St Pierre B, De Luca V. 2000. Evolution of acyltransferase genes: Origin and
diversification of the BAHD superfamily of acyltransferases involved in secondary
metabolism In: Romeo JT, Ibrahim R, Varin L, De Luca V, editors. Recent
Advances in Phytochemistry, Evolution of Metabolic Pathways. Elsevier. pp. 285–
315. doi:10.1016/S0079-9920(00)80010-6

St-Pierre B, Vazquez-Flota FA, Luca VD. 1999. Multicellular compartmention of
Catharanthus roseus alkaloid biosynthesis predicts intercellular translocation of a
pathway intermediate. *Plant Cell* **11**:887–900. doi:10.1105/tpc.11.5.887

Suzuki H, Nakayama T, Nishino T. 2003. Proposed mechanism and functional amino acid
residues of malonyl-CoA : anthocyanin 5-O-glucoside-6 ' ' '-O-malonyltransferase
from flowers of *Salvia splendens*, a member of the versatile plant acyltransferase
family. *Biochemistry* **42**:1764–1771. doi:10.1021/bi020618g

Tanner KG, Trievel RC, Kuo M-H, Howard RM, Berger SL, Allis CD, Marmorstein R, Denu
JM. 1999. Catalytic Mechanism and Function of Invariant Glutamic Acid 173 from
the Histone Acetyltransferase GCN5 Transcriptional Coactivator. *J Biol Chem*
274:18157–18160. doi:10.1074/jbc.274.26.18157

Tokuriki N, Tawfik DS. 2009. Protein Dynamism and Evolvability. *Science* **324**:203–207.
doi:10.1126/science.1169375

Trott O, Olson AJ. 2010. AutoDock Vina: improving the speed and accuracy of docking
with a new scoring function, efficient optimization and multithreading. *J Comput*
Chem **31**:455–461. doi:10.1002/jcc.21334

Tuominen LK, Johnson VE, Tsai C-J. 2011. Differential phylogenetic expansions in BAHD
acyltransferases across five angiosperm taxa and evidence of divergent
expression among *Populus* paralogues. *BMC Genomics* **12**:236.
doi:10.1186/1471-2164-12-236

Unno H, Ichimaida F, Suzuki H, Takahashi S, Tanaka Y, Saito A, Nishino T, Kusunoki M,
Nakayama T. 2007. Structural and Mutational Studies of Anthocyanin
Malonyltransferases Establish the Features of BAHD Enzyme Catalysis. *J Biol*
Chem **282**:15812–15822. doi:10.1074/jbc.M700638200

Vries J de, Rensing SA. 2020. Gene gains paved the path to land. *Nat Plants* **6**:7–8.
doi:10.1038/s41477-019-0579-5

Wagner A. 2008. Robustness and evolvability: a paradox resolved. *Proc R Soc B Biol Sci*
275:91–100. doi:10.1098/rspb.2007.1137

Wagner A. 2005. Robustness, evolvability, and neutrality. *FEBS Lett*, Systems Biology
579:1772–1778. doi:10.1016/j.febslet.2005.01.063

Wang J, Marowsky NC, Fan C. 2014. Divergence of Gene Body DNA Methylation and
Evolution of Plant Duplicate Genes. *PLOS ONE* **9**:e110357.
doi:10.1371/journal.pone.0110357

Wang J, Wolf RM, Caldwell JW, Kollman PA, Case DA. 2004. Development and testing
of a general amber force field. *J Comput Chem* **25**:1157–1174.
doi:10.1002/jcc.20035

Wang L-P, Martinez TJ, Pande VS. 2014. Building Force Fields: An Automatic,
1373   Systematic, and Reproducible Approach. *J Phys Chem Lett* **5**:1885–1891.
1374   doi:10.1021/jz500737m

1375   Wang L-P, McKiernan KA, Gomes J, Beauchamp KA, Head-Gordon T, Rice JE, Swope
1376   WC, Martínez TJ, Pande VS. 2017. Building a More Predictive Protein Force Field:
1377   A Systematic and Reproducible Route to AMBER-FB15. *J Phys Chem B*
1378   **121**:4023–4039. doi:10.1021/acs.jpcb.7b02320

1379   Wang Y, Backman TWH, Horan K, Girke T. 2013. fmcsR: mismatch tolerant maximum
1380   common substructure searching in R. *Bioinformatics* **29**:2792–2794.
1381   doi:10.1093/bioinformatics/btt475

1382   Weng J-K, Chapple C. 2010. The origin and evolution of lignin biosynthesis. *New Phytol*
1383   **187**:273–285. doi:https://doi.org/10.1111/j.1469-8137.2010.03327.x

1384   Yang Q, Reinhard K, Schiltz E, Matern U. 1997. Characterization and heterologous
1385   expression of hydroxycinnamoyl/benzoyl-CoA:anthranilate N-
1386   hydroxycinnamoyl/benzoyltransferase from elicited cell cultures of carnation,
1387   *Dianthus caryophyllus* L. *Plant Mol Biol* **35**:777–789.
1388   doi:10.1023/A:1005878622437

1389   Zou C, Lehti-Shiu MD, Thomashow M, Shiu S-H. 2009. Evolution of Stress-Regulated
1390   Gene Expression in Duplicate Genes of Arabidopsis thaliana. *PLOS Genet*
1391   **5**:e1000581. doi:10.1371/journal.pgen.1000581

1392

1393

SUPPLEMENTARY TABLES

Table S1: Substrate class definitions. Substrates were grouped into different classes based on their chemical properties and their membership in known compound classes. Also see **Fig. S1A**.

Scaffold	Class	Definition	Exemplary structures
Aromatic alcohols	A	Substrates were classified as aromatic alcohols when they contained a planar ring of carbon atoms and a hydroxy group attached to a sidechain. Compounds containing additional hydroxy groups directly attached to the ring were also classified into this category. All substrates in this class except 2-naphthaleneethanol had one ring.	Benzoyl alcohol, Coniferyl alcohol, Shikimate, Quinate
Anthocyanins	D	Substrates were grouped into the anthocyanins category if they fulfill the classical definition of an anthocyanin: contain an anthocyanidin (flavylium cation) with attached sugar group(s).	Cyanidin 3-O glucoside, Cyanidin 3-5-O glucoside, Pelargonidin 3-O rutinoside
Flavonoids	D	Substrates being a flavonoid or isoflavonoid were grouped into this category. Flavonoids a generally defined as having a 15-carbon skeleton which contains two benzene rings that are connected with a 3-carbon linking chain. Also, flavonoid glucosides were grouped into this class.	Naringin, Genistin, Quercetin 3-O- sophoroside
Aliphatic amines	B	Aliphatic amines were defined as amines not containing any aromatic rings but were allowed to contain more than one amine group and or have additional hydroxy groups.	Agmatine, Putrescine, Spermidine
Aliphatic alcohols	B	Aliphatic alcohols do not contain any aromatic rings but can range from short chain to very long chains. We also grouped fatty acids into this category because of their high chemical similarity.	1-Butanol, 1-Decanol, 16-Hydroxypalmitic acid
Monoterpenoids	C	Compounds formally containing of isoprene that do not contain aromatic rings or had branched chains were classified as terpenoids.	Geraniol, Nerol, Linalool
Sesquiterpenoids	C	Compounds build from three isoprene units.	Farnesol
Aromatic amines	A	Substrates containing an aromatic ring and an amine group were classified as aromatic amines. However, substrates containing an additional hydroxy group were also categorized as aromatic amine because their chemistry was expected to be more similar to amines than to aromatic alcohols, due to presence of the positively charged -NH2 group.	Tyramine, Dopamine, 3-Aminobenzoate
Alkaloids	Mix	Substrates were classified as alkaloids if they contained a heterocyclic bound nitrogen atom.	Methylecgonine, Tryptamine, Serotonin
Phenolic glycosides	D	We classified substrates as phenolic glycosides if they contained a sugar group bound via a glycosidic bond to a aromatic molecule that could not be classified as either flavonoids or anthocyanins.	Scopolin, 1-Naphtol glucoside, Umbelliferone glucoside
Sugar derivatives	G	Substrates classified as sugar derivatives are either mono- and disaccharides and their acylated variants.	Sucrose, Glucose, S1:5(5)
Diterpenoids	E	Substrates involved in the biosynthesis of taxenes were grouped into the diterpenoid class	Baccatin III, 10-Deacetylbaccatin III, Taxadien-5a-ol
Triterpenoids	F	Compounds that consisted of six isoprene units were grouped into the triterpenoid class.	Thalianol, Arabidiol, Tirucalla-7,24-dien-3beta-ol

Table S2. BAHD acyltransferases in the *Solanum lycopersicum* (tomato) genome. For characterized BAHDs their name and the respective reference is given.

Protein ID	PFAM domain name	Pfam ID	Name/ Annotation	Reference
Solyc08g005770.2.1	Transferase	PF02458.15	SlAAT1	(Goulet et al., 2015)
Solyc08g005760.1.1	Transferase	PF02458.15	SlAAT2	(Goulet et al., 2015)
Solyc11g071470.1.1	Transferase	PF02458.15	SlACT1	**this study**
Solyc11g071480.1.1	Transferase	PF02458.15	SlACT2	**this study**
Solyc12g006330.1.1	Transferase	PF02458.15	SlASAT1	(Fan et al., 2016)
Solyc04g012020.1.1	Transferase	PF02458.15	SlASAT2	(Fan et al., 2016)
Solyc11g067270.1.1	Transferase	PF02458.15	SlASAT3	(Fan et al., 2016)
Solyc01g105580.1.1	Transferase	PF02458.15	SlASAT4	(Fan et al., 2016)
Solyc03g117600.2.1	Transferase	PF02458.15	SlHCT	**this study**
Solyc07g005760.2.1	Transferase	PF02458.15	SlHQT	(Niggeweg et al., 2004)
Solyc05g052670.1.1	Transferase	PF02458.15		
Solyc05g052680.1.1	Transferase	PF02458.15		
Solyc11g008630.1.1	Transferase	PF02458.15		
Solyc07g015960.1.1	Transferase	PF02458.15		
Solyc09g014280.1.1	Transferase	PF02458.15		
Solyc03g097500.2.1	Transferase	PF02458.15		
Solyc02g079490.2.1	Transferase	PF02458.15		
Solyc07g014580.2.1	Transferase	PF02458.15		
Solyc05g014330.1.1	Transferase	PF02458.15		
Solyc07g049660.2.1	Transferase	PF02458.15		
Solyc08g005890.2.1	Transferase	PF02458.15		
Solyc07g049670.2.1	Transferase	PF02458.15		
Solyc03g025320.2.1	Transferase	PF02458.15		
Solyc04g078660.1.1	Transferase	PF02458.15		
Solyc05g015800.2.1	Transferase	PF02458.15		
Solyc02g093180.2.1	Transferase	PF02458.15		
Solyc04g079720.2.1	Transferase	PF02458.15		
Solyc06g074710.1.1	Transferase	PF02458.15		
Solyc07g006680.1.1	Transferase	PF02458.15		
Solyc12g096250.1.1	Transferase	PF02458.15		
Solyc06g051320.2.1	Transferase	PF02458.15		
Solyc04g082350.1.1	Transferase	PF02458.15		
Solyc01g105550.1.1	Transferase	PF02458.15		
Solyc01g107080.2.1	Transferase	PF02458.15		
Solyc02g062710.1.1	Transferase	PF02458.15		

Solyc11g066640.1.1	Transferase	PF02458.15
Solyc08g013830.1.1	Transferase	PF02458.15
Solyc07g006670.1.1	Transferase	PF02458.15
Solyc04g080720.2.1	Transferase	PF02458.15
Solyc02g081740.1.1	Transferase	PF02458.15
Solyc01g105590.2.1	Transferase	PF02458.15
Solyc06g051130.1.1	Transferase	PF02458.15
Solyc08g014490.1.1	Transferase	PF02458.15
Solyc07g043700.1.1	Transferase	PF02458.15
Solyc12g088170.1.1	Transferase	PF02458.15
Solyc07g043710.2.1	Transferase	PF02458.15
Solyc02g081760.1.1	Transferase	PF02458.15
Solyc12g096770.1.1	Transferase	PF02458.15
Solyc00g040390.1.1	Transferase	PF02458.15
Solyc07g043670.1.1	Transferase	PF02458.15
Solyc02g081800.1.1	Transferase	PF02458.15
Solyc12g010980.1.1	Transferase	PF02458.15
Solyc02g081750.1.1	Transferase	PF02458.15
Solyc08g075210.1.1	Transferase	PF02458.15
Solyc11g069680.1.1	Transferase	PF02458.15
Solyc12g096790.1.1	Transferase	PF02458.15
Solyc11g067290.1.1	Transferase	PF02458.15
Solyc12g096800.1.1	Transferase	PF02458.15
Solyc01g068140.2.1	Transferase	PF02458.15
Solyc11g067340.1.1	Transferase	PF02458.15
Solyc07g008380.1.1	Transferase	PF02458.15
Solyc07g008390.1.1	Transferase	PF02458.15
Solyc12g005430.1.1	Transferase	PF02458.15
Solyc11g067330.1.1	Transferase	PF02458.15
Solyc02g081770.1.1	Transferase	PF02458.15
Solyc12g005440.1.1	Transferase	PF02458.15
Solyc05g039950.1.1	Transferase	PF02458.15
Solyc00g040290.1.1	Transferase	PF02458.15
Solyc07g017320.1.1	Transferase	PF02458.15
Solyc01g008300.1.1	Transferase	PF02458.15
Solyc11g020640.1.1	Transferase	PF02458.15
Solyc10g079570.1.1	Transferase	PF02458.15
Solyc10g008680.1.1	Transferase	PF02458.15
Solyc04g009680.1.1	Transferase	PF02458.15

Solyc07g052060.2.1	Transferase	PF02458.15
Solyc12g087980.1.1	Transferase	PF02458.15
Solyc09g092270.2.1	Transferase	PF02458.15
Solyc08g075180.1.1	Transferase	PF02458.15
Solyc05g052650.2.1	Transferase	PF02458.15
Solyc10g055730.1.1	Transferase	PF02458.15
Solyc00g135260.1.1	Transferase	PF02458.15
Solyc01g107070.2.1	Transferase	PF02458.15
Solyc05g050760.1.1	Transferase	PF02458.15
Solyc01g005900.2.1	Transferase	PF02458.15
Solyc00g134620.1.1	Transferase	PF02458.15
Solyc01g107050.2.1	Transferase	PF02458.15
Solyc08g007210.2.1	Transferase	PF02458.15
Solyc08g078030.2.1	Transferase	PF02458.15
Solyc08g075200.1.1	Transferase	PF02458.15
Solyc07g026890.1.1	Transferase	PF02458.15
Solyc11g067350.1.1	Transferase	PF02458.15
Solyc05g015810.1.1	Transferase	PF02458.15
Solyc06g071940.1.1	Transferase	PF02458.15
Solyc03g078130.1.1	Transferase	PF02458.15
Solyc10g008670.2.1	Transferase	PF02458.15
Solyc04g009670.1.1	Transferase	PF02458.15
Solyc08g036440.1.1	Transferase	PF02458.15
Solyc04g078350.1.1	Transferase	PF02458.15
Solyc12g044660.1.1	Transferase	PF02458.15

1400
1401

1402 **Table S3. Primer sequences.** Sequences of Primers used to clone enzymes analyzed in this study.

Name	Sequence (5' to 3' end)	Used for
Oligo DT$_{17}$	GTCGACTCGAGAATTCTTTTTTTTTTTTTTTTT	cDNA synthesis
Gibson_N_6xHis4CL_F2	AGCATGACTGGTGGACAGCAAATGGGTCGGATGCCGAT GGATACCGAAACA	Cloning of Sl4CL
Gibson_4CL_R2	GCCGGATCTCAGTGGTGGTGGTGGTGGTGCATGCCGAT GGATACCGAAACA	Cloning of Sl4CL
Gibson_N_6xHis_HCT_F1	AGCATGACTGGTGGACAGCAAATGGGTCGGATGAAGAT CGAGGTGAAAAACTCA	Cloning of SlHCT
Gibson_HCT_R1	GCCGGATCTCAGTGGTGGTGGTGGTGGTGCTCAAATGT CATACAAGAACTTCTCAA	Cloning of SlHCT
SlHQT_Gibson_N_6xHis_F1	AGCATGACTGGTGGACAGCAAATGGGTCGGATGGGAAG TGAAAAAATGATGAAAATTAATATC	Cloning of SlHQT
SlHQT_Gibson_R1	GCCGGATCTCAGTGGTGGTGGTGGTGGTGCTCATAATT CATATAAATATTTTTCAAATAGTG	Cloning of SlHQT
Gibson_N_6xHis_ACT2_F2	AGCATGACTGGTGGACAGCAAATGGGTCGGATGAATGT GAAAATTGAGAGTTCAAAAATCATCAAGCCATTG	Cloning of SlACT
Gibson_ACT2_R2	GCCGGATCTCAGTGGTGGTGGTGGTGGTGCTCACTTTG CTTTCAAATCTAGAGAGTAACAAATTTTC	Cloning of SlACT
Dm3MAT3_W70A_F2	CCTTTTTCGATTTCTTTGCGCTGCGCAGTCCGCC	Mutation of Trp36 to Ala
Dm3MAT3_W70A_R2	GGCGGACTGCGCAGCGCAAAGAAATCGAAAAAGG	Mutation of Trp36 to Ala
Dm3MAT3_C322A_F	GCATATTTTGGCAATGCCGTGGGTGGTTGCGCC	Mutation of Cys320 to Ala
Dm3MAT3_C322A_R	GGCGCAACCACCCACGGCATTGCCAAAATATGC	Mutation of Cys320 to Ala

1403
1404 **Table S4: Masses of detected products.** Parent ion mass (bold) for each donor was used for isolation
1405 with product specific PRM method (see methods section) and the quantifier ion to quantify this product (if
1406 product was detected). For each product (if detected) the retention time is given in brackets. n.t = not tested;
1407 n.d. = not detected. LC methods are described in **Tab. S5**.

Substrate	+ Coumaroyl Product (m/z)	+ Feruloyl Product (m/z)	+ Malonyl Product (m/z)	Ionization mode	LC method
Shikimate	**319.0817** 163.0398 (1.6)	**349.0924** n.d.	**259.0454** n.d.	negative	4-min method
Quinate	**337.0927** 191.0555 (1.4)	**367.1034** n.d.	**277.0565** 191.0558 (1.4)	negative	4-min method
Tyramine	**284.1279** 147.0443 (1.85)	**314.1386** 177.0544 (1.66)	**224.0917** n.d.	positive	4-min method
Dopamine	**300.1229** 147.0443 (1.65)	**330.1336** 177.0545 (1.6)	**284.0764** n.d.	positive	4-min method
Agmatine	**277.1667** 147.0440 (1.1)	**307.1774** 177.0546 (1.3)	**217.1295** 131.1291 (1.1)	positive	4-min method
Putrescine	**235.1445** 147.0440 (0.6)	**265.1552** 177.0546 (1.0)	**175.1077** n.d.	positive	4-min method
1-Decanol	**303.1964** n.d.	**333.2071** 177.0194 (5.5)	**243.1601** n.d.	negative	8-min method
Naringin	**725.2091** 163.0401 (1.7)	**755.2198** 271.0603 (2.1)	**665.1723** 271.0608 (1.8)	negative	4-min method
Genistin	**577.1351** 269.0434 (1.65)	**607.1458** 269.0445 (2.1)	**517.0987** 269.0443 (1.8)	negative	4-min method
Cyanidin 3,5-O glucoside	**757.1974** 287.0546	**787.208** **n.d**	**697.1610** 287.0546	positive	4-min method

	(1.45)		(1.25)		
Malvidin 3-O glucoside	**639.1713** 331.0835 (1.65)	**669.1820** 331.0829 (1.6)	**579.1350** 331.0835 (1.45)	positive	4-min method
Pelargonidin 3-O rutinoside	**725.2080** n.d.	**755.2270** 271.0604 (1.5)	**665.1717** n.t.	positive	4-min method
Cyanidin 3-O glucoside	**595.1451** 287.0546 (1.6)	**625.1558** 287.0541 (1.5)	**535.1088** 287.0544 (2.5 (8-min method))	positive	4-min method 8-min method
Cyanidin 3-O rutinoside	**741.2029** 287.0546 (1.25)	**771.2136** 287.0545 (1.5)	**681.1666** n.t.	positive	4-min method
Linalool	**299.1725** n.d.	**329.183165** n.d.	**239.1362** n.d.	negative	8-min method
Geraniol	**299.1654** 163.0393 (4.9)	**329.1754** 329.1754 (4.9)	**239.1362** n.d.	negative	8-min method
Cucurbitacin E	**701.3380** 175.0393 (4.3)	**731.3502** 175.0393 (4.4)	**641.2967** n.d.	negative	8-min method

Table S5. LC methods used for analyzing enzyme assay reactions. All solvents and reagents used
were LC-MS grade quality. For the 150 mm column, a 2 mm C18 guard column (AJ0-8782, SecurityGuard
Ultra, 2.1 mm, Phenomenex, Torrance, CA) was used. The autosampler was kept at 20 ºC and the column
compartment was heated to 40 ºC. For negative ion mode, the mass spectrometer parameters were as
follows: spray voltage 3800V, capillary temperature 380C, sheath gas pressure 60 psi, auxiliary gas 20 psi,
spare gas 2 psi, max. spray current 100eV, probe 400C, RF lens 50V, and a collision energy ramped from
20 to 40 eV. For positive ion mode MS parameters were the same except for the spray voltage that was
set to 3500V.

4min-C18

	Flow (mL/min)	%A	%B	Curve	Solvents	Column
					A: Water + 0.01% (v/v) Formic acid B: Acetonitrile + 0.01% (v/v) Formic acid	Kinetex PS C18 (00B-4780-AN, 2.6µm particle size, 100Å pore size, 50 x 2.10 mm, Phenomenex (Torrance, CA))
Initial	0.6	95	5	5		
0.5	0.6	95	5	5		
3.0	0.6	30	70	5		
3.3	0.6	95	5	5		
3.6	0.6	5	95	5		
4.0	0.6	5	95	5		

8min-C18

	Flow (mL/min)	%A	%B	Curve	Solvents	Column
					A: Water+0.01% (v/v) Formic acid B: Acetonitrile + 0.01% (v/v) Formic acid	Kinetex C18 (00F-4462-AN, 2.6µm particle size, 100Å pore size, 150 x 2.10 mm, Phenomenex (Torrance, CA))
Initial	0.6	95	5	5		
1	0.6	95	5	5		
5.6	0.6	5	95	5		
6.6	0.6	5	95	5		
6.61	0.6	5	95	5		
7.5	0.6	95	5	5		
8.0	0.6	95	5	5		

1416

Figure S1

A

B

Figure S1: Substrate classification and BAHD substrates represented using an alternative layout. (A) Decision tree to group BAHD substrates into different substrate types. Further information about substrate class definitions can be found in **Table S1**. **(B)** The network was created in the same way and is colored the same way as **Fig. 1A**. The "Edge-weighted Spring Embedded" Layout using MCS-Tanimoto coefficient was used to create an edge-weighted network.

Figure S2

A

Enzyme	Donor	acceptor concentration	aromatic alcohols		aromatic amines		aliphatic amines		aliphatic alcohols	flavonoids		anthocyanins	
			Shikimate	Quinate	Tyramine	Dopamine	Putrescine	Agmatine	Decanol	Naringin	Genistin	Cyanidin 3,5-O diglucoside	Malvidin 3-O glucoside
CbHQT-like	Coumaroyl-CoA	100 µM	0.037 ±0.008	4.763 ±2.812	0.023 ±0.003	0.021 ±0.001	0.016 ±0.006	0.024 ±0.003	0.015 ±0.001	0.028 ±0.006	0.434 ±0.237	0.001 ±0.000	0.001 ±0.001
CbHQT-like*	Coumaroyl-CoA	1 mM	6.956 ±0.476	605.346 ±59.314	6.322 ±0.835	16.998 ±3.392	9.289 ±2.779	4.469 ±1.269	0.603 ±0.056	n.t.	n.t.	n.t.	n.t.
MeHFT	Coumaroyl-CoA	100 µM	0.052 ±0.008	0.003 ±0.001	0.083 ±0.026	0.134 ±0.015	0.037 ±0.006	0.025 ±0.007	0.000 ±0.000	0.067 ±0.013	0.193 ±0.028	0.009 ±0.001	0.180 ±0.014
MeHFT	Feruloyl-CoA	100 µM	n.t.	n.t.	n.t.	n.t.	n.t.	n.t.	8.332 ±0.000	0.041 ±0.004	0.336 ±0.040	0.191 ±0.173	2.588 ±0.083
MeHFT*	Feruloyl-CoA	1 mM	n.d.	n.d.	2.014 ±0.066	0.967 ±0.101	0.296 ±0.032	0.286 ±0.007	n.t.	n.t.	n.t.	n.t.	n.t.
SlACT	Coumaroyl-CoA	100 µM	0.006 ±0.001	0.003 ±0.001	0.050 ±0.007	0.054 ±0.002	22.873 ±3.184	6.476 ±0.455	0.018 ±0.003	0.022 ±0.006	0.065 ±0.113	0.002 ±0.001	0.008 ±0.002
SlACT*	Coumaroyl-CoA	1 mM	1.396 ±0.433	1.854 ±0.679	10.576 ±1.632	17.718 ±7.195	3,914.042 ±1,218.783	1,076.305 ±141.838	0.715 ±0.139	n.t.	n.t.	n.t.	n.t.
SlHCT	Coumaroyl-CoA	100 µM	65.373 ±14.024	70.896 ±17.802	2.238 ±0.292	1.093 ±0.136	0.057 ±0.014	0.027 ±0.007	n.t.	0.078 ±0.022	0.228 ±0.079	0.210 ±0.052	6.497 ±1.347
SlHQT	Coumaroyl-CoA	100 µM	21.606 ±1.636	160.623 ±27.189	0.041 ±0.009	0.093 ±0.033	0.056 ±0.006	0.030 ±0.004	n.t.	0.102 ±0.008	0.169 ±0.009	0.012 ±0.002	0.015 ±0.011
EcHQT	Coumaroyl-CoA	100 µM	9.695 ±1.284	132.568 ±7.880	0.034 ±0.004	0.055 ±0.008	0.111 ±0.091	0.035 ±0.015	n.t.	0.040 ±0.004	0.557 ±0.629	0.001 ±0.000	0.002 ±0.001
EcCS	Coumaroyl-CoA	100 µM	25.696 ±3.045	94.004 ±8.606	0.025 ±0.003	0.033 ±0.005	0.021 ±0.001	0.025 ±0.002	n.t.	0.031 ±0.021	0.188 ±0.049	0.001 ±0.000	0.002 ±0.000
At5MAT	Malonyl-CoA	100 µM	n.d.	n.d.	n.d.	0.005 ±0.005	n.d.	n.d.	n.t.	0.042 ±0.015	n.d.	1.876 ±0.317	0.943 ±0.147
GmIF7MAT	Malonyl-CoA	100 µM	n.d.	0.039 ±0.050	n.d.	n.d.	n.d.	0.017 ±0.015	n.t.	0.043 ±0.011	0.638 ±0.049	0.004 ±0.001	0.029 ±0.020
Dm3MAT3	Malonyl-CoA	100 µM	n.d.	n.d.	n.d.	0.002 ±0.003	n.d.	0.195 ±0.041	n.t.	0.000 ±0.000	1.229 ±0.116	2.683 ±0.163	13.098 ±4.598
Dm3MAT3	Coumaroyl-CoA	100 µM	n.d.	n.d.	n.d.	n.t.	n.t.	n.t.	n.t.	n.t.	n.t.	n.t.	n.t.
Pc3MAT3	Malonyl-CoA	100 µM	n.d.	0.016 ±0.007	n.d.	n.d.	n.d.	n.d.	n.t.	0.057 ±0.014	0.030 ±0.006	3.751 ±0.888	100.680 ±3.219
Mock	Coumaroyl-CoA	100 µM	0.597 ±0.049	0.940 ±0.367	0.047 ±0.001	0.093 ±0.026	0.034 ±0.002	0.023 ±0.003	0.001 ±0.002	0.071 ±0.011	0.296 ±0.200	0.017 ±0.004	0.018 ±0.007
Mock*	Coumaroyl-CoA	1 mM	22.757 ±3.350	0.212 ±0.084	6.018 ±0.530	14.886 ±4.448	3.156 ±2.660	4.539 ±0.643	0.111 ±0.000	n.t.	n.t.	n.t.	n.t.
Mock	Malonyl-CoA	100 µM	n.d.	n.d.	n.d.	n.d.	n.d.	n.d.	n.t.	0.039 ±0.005	0.000 ±0.000	0.003 ±0.000	0.006 ±0.001
MeHFT	Feruloyl-CoA	100 µM	n.d.	n.d.	0.045 ±0.004	n.d.	0.189 ±0.005	0.124 ±0.070	n.d.	n.t.	n.t.	n.t.	n.t.
Mock*	Feruloyl-CoA	1 mM	n.t.	n.t.	n.t.	n.t.	n.t.	n.t.	0.006 ±0.001	0.014 ±0.004	0.064 ±0.040	0.057 ±0.007	

B

Enzyme	Donor	acceptor concentration	anthocyanins				
			Cyanidin 3,5-O diglucoside	Malvidin 3-O glucoside	Pelargonidin 3-O rutinoside	Cyanidin 3-O glucoside	Cyanidin 3-O rutinoside
CbHCT	Coumaroyl-CoA	100 µM	0.063 ±0.015	0.111 ±0.024	1.337 ±0.112	1.341 ±0.354	0.133 ±0.028
MeHFT	Feruloyl-CoA	100 µM	0.191 ±0.173	2.588 ±0.083	54.127 ±8.146	3.243 ±0.468	0.431 ±0.101
SlACT	Coumaroyl-CoA	100 µM	0.059 ±0.014	0.105 ±0.085	1.143 ±0.256	0.549 ±0.228	0.053 ±0.052
SlACT	Malonyl-CoA	100 µM	0.002 ±0.001	0.000 ±0.000	0.001 ±0.001	0.007 ±0.003	0.011 ±0.000
SlHCT	Coumaroyl-CoA	100 µM	1.127 ±0.332	38.458 ±19.705	0.908 ±0.313	2.445 ±0.955	1.550 ±2.472
SlHQT	Malonyl-CoA	100 µM	0.001 ±0.000	0.000 ±0.000	0.001 ±0.000	0.033 ±0.047	0.007 ±0.001
EcHQT	Coumaroyl-CoA	100 µM	0.037 ±0.064	0.060 ±0.103	0.443 ±0.768	0.726 ±1.258	0.080 ±0.070
Mock	Coumaroyl-CoA	100 µM	0.063 ±0.057	0.166 ±0.007	0.987 ±0.069	0.770 ±0.169	0.121 ±0.048
Mock	Feruloyl-CoA	100 µM	0.064 ±0.040	0.057 ±0.007	0.052 ±0.025	0.915 ±0.049	0.000 ±0.000
Mock	Malonyl-CoA	100 µM	0.000 ±0.000	0.000 ±0.000	0.000 ±0.000	0.000 ±0.000	0.000 ±0.000

C

Enzyme	Donor	acceptor concentration	Monoterpenoids		Triterpenoids
			Linalool	Geraniol	Curcubitacine E
CbHQT-like	Coumaroyl-CoA	100 µM	3.383 ±0.358	0.000 ±0.000	0.436 ±0.042
MeHFT	Feruloyl-CoA	100 µM	0.034 ±0.011	45.255 ±20.153	10.589 ±0.421
SlHCT	Coumaroyl-CoA	100 µM	2.350 ±0.034	1.386 ±0.307	10.975 ±0.466
SlHQT	Coumaroyl-CoA	100 µM	2.893 ±0.163	0.000 ±0.000	0.529 ±0.029
Dm3MAT3	Malonyl-CoA	100 µM	3.952 ±0.643	3.900 ±0.334	0.000 ±0.000
Mock	Coumaroyl-CoA	100 µM	2.742 ±0.399	0.000 ±0.000	0.651 ±0.193
Mock	Feruloyl-CoA	100 µM	0.095 ±0.031	0.101 ±0.020	1.236 ±0.081
Mock	Malonyl-CoA	100 µM	4.217 ±0.580	4.497 ±0.689	0.000 ±0.000

Figure S2. Specific activities of selected enzymes against representative substrates. Enzymes are ordered by the phylogenetic relationship to each other, starting with the most basal at the top. The average enzyme activity and standard deviation of three technical replicates is given as nmol/mg/min. Each enzyme was tested with its preferred donor substrate (300 uM) and 100 uM of the acceptor substrate. Main activities for each enzyme are colored dark red, medium activities in orange, and low/trace activities in light red. Each activity considered had to be three times higher than the respective Mock control (empty vector). **(A)** Enzyme activities of selected enzymes against a large panel of substrates. In case of EcCS, coumaroyl-CoA was used instead of benzoyl-CoA which had been shown to be used previously. MeHFT was assayed with coumaroyl-CoA and feruloyl-CoA. n.d. = not detected n.t. = not tested. * = higher acceptor substrate concentration than other assays. In addition to the shown activities, all enzymes were tested with glucose and sucrose as acceptor substrates but no activities were observed. **(B)** Enzyme assays testing the ability of selected enzymes to acylate anthocyanins with their preferred donor. To exclude that such enzymes can acylate anthocyanins with a different donor, SlHQT and SlACT were also tested with malonyl-CoA, a donor not previously shown to be used by those enzymes. As expected, changing the donor did not allow those enzymes to acylate anthocyanins. **(C)** Enzyme assays testing the ability to use terpenoids. Two monoterpenoids and one triterpenoid were tested using different enzymes with their preferred donor. Mock controls for each donor were performed to exclude the unspecific formation of the analyzed reaction products.

Figure S3

Figure S3. Defining BAHD clades. The tree was rooted using the algal enzyme clade (Clade 0a). Maroon circles on branches refer to clades with bootstrap values > 70. Clades were defined based on deepest, high-confidence mono-phyletic clades. Clades 1-4 are same as D'Auria et al, 2006 definitions, while Clade V from that study is divided into Clades 5-7 here, based on the above criterion. Branch colors indicate the different clades. Branch lengths indicate number of amino acid substitutions per site.

Figure S4

A

		E-value
1.		1.9e-248
2.		1.2e-402
3.		4.2e-248
4.		4.5e-221
5.		9.9e-192

B

Figure S4. Sequences used for motif enrichment analysis. (A) Five top most enriched motifs in anthocyanin/flavonoid-acylating BAHD acyltransferases. **(B)** Groups of sequences that were further analyzed with respect to the TFFDxxW (1. motif) and YFGNC (4. motif). The same tree as in **Fig. 4** is shown. Black circles on branches refer to clades with bootstrap values > 70. Clades not analyzed in more detail are displayed in light gray and bootstrap values are not indicated. Groups of sequences correspond to **Fig. 4**. Group 5, highlighted in purple, contains the sequences of anthocyanin/flavonoid-acylating BAHDs that were analyzed in more detail.

Figure S5. Alignment of HCT/HQT-like and anthocyanin/flavonoid-acylating BAHDs. (A) Alignment of the TFFDxxW region of sequences belonging to clade 1a/b (anthocyanin/flavonoid) and 5a/b (HCT/HQT). **(B)** Alignment of the same sequences for the YFGNC region. Conserved (95%) residues are highlighted in colors. The positions of the highly conserved Trp and Cys residues are highlighted with a box.

Figure S6. Analysis of distance fluctuations in Dm3MAT3 variants. **(A)** Distance calculations between backbone hydrogen of C/A320 and the center of mass of the aromatic ring of F32 for the apo simulation. WT is represented as purple, W36A is represented as orange-red, and C320A is represented as green. **(B)** Distance calculations for the holo simulation **(C)** Stacking rearrangements of phenylalanine aromatic sidechains within the TFFDXXW motif of C320A Dm3MaT3. Amino acid coloring: F31, cornflower blue; F32, hot pink; F35, spring green; W36, H170, A320, and C3G are colored as in previous images. Malo-nyl-CoA and W383 are hidden for simplicity. Distance fluctuations between A320 alpha hydrogen and F32 aromatic system cause aromatic stacking reorganization in the TFFDXXW motif. A catalytically competent distance between C3G-6"-OH and H170 is maintained. **(D)** C3G can still undergo reaction as F35 breaks its t-stacking with W36 to stabilize F31 and F32. **(E)** F31 and F32 have been stabilized. F32 moves closer to A320 while F35 returns to t-stack with W36, although C3G has distanced itself from H170. **(F)** Stacking interactions between residues in the active site and across the TFFDXXW motif have been reset, but now C3G is excluded from re-entering the active site due to steric and non-conventional bonding interactions with peripher-al residues. **(G)** Schematic demonstrating θ calculations for discriminating between face-to-face and edge-to-face stacking interactions along the TFFDXXW motif. Figure and methodology as described in the Data Analysis subsection from the Main Text. Rings A and B represent the aromatic systems of two different residues. The normal vectors for each ring were then calcu-lated so that the point of intersection between Rings A and B could be determined. Solving for the angle between the center of mass of Ring A, the center of mass of Ring B, and the intersection of the rings' normal vectors, allowed for the supplementary angle, θ, to be determined. When $θ ≤ 30°$ or $150° ≥ θ$ and the distance between the aromatic systems' centers of mass is ≤ 4.4 Å, face-to-face stacking is occurring. When $60° ≤ θ ≤ 120°$ and the distance between the aromatic systems' centers of mass is ≤ 5.5 Å, edge-to-face stacking is occurring. When $30° < θ < 60°$ or $120° < θ < 150°$ or the distance requirements for either form of stacking are unmet, no stacking interaction is occurring.

Figure S7

A aC320 stackings

<table>
<tr><td>face-to-face stacking</td><td>edge-to-face stacking</td><td>no stacking (can include)</td></tr>
<tr><td>θ ≤ 30° or 150° ≥ θ</td><td>60° ≤ θ ≤ 120°</td><td>30° < θ < 60° or 120° < θ < 150°</td></tr>
<tr><td>distance ≤ 4.4 Å</td><td>distance ≤ 5.5 Å</td><td>distance criteria for stacking unmet</td></tr>
</table>

B hC320A stackings

<table>
<tr><td>face-to-face stacking</td><td>edge-to-face stacking</td><td>no stacking (can include)</td></tr>
<tr><td>θ ≤ 30° or 150° ≥ θ</td><td>60° ≤ θ ≤ 120°</td><td>30° < θ < 60° or 120° < θ < 150°</td></tr>
<tr><td>distance ≤ 4.4 Å</td><td>distance ≤ 5.5 Å</td><td>distance criteria for stacking unmet</td></tr>
</table>

Figure S7. Stacking interactions for the C320A mutant. (A) Apo C320A (aC320A) stacking interactions categorized as face-to-face stacking, edge-to-face stacking, or no stacking interactions. **(B)** Holo C320A (hC320A) stacking interactions. Face-to-face stacking is represented as blue, edge-to-face stacking as orange, and no stacking as gray. Distance measurements were taken between the centers of mass of each residue listed on the y-axis of each panel.

Figure S8

Figure S8. Stacking interactions for the W36A mutant. (A) Apo W36A (aW36A) stacking interactions categorized as face-to-face stacking, edge-to-face stacking, or no stacking interactions. **(B)** Holo W36A (hW36A) stacking interactions. Face-to-face stacking is represented as blue, edge-to-face stacking as orange, and no stacking as gray. Distance measurements were taken between the centers of mass of each residue listed on the y-axis of each panel. Only three panels are provided because W36A substitution prevents F35-W36 stacking from occurring.

A aWT stackings

Figure S9. Stacking interactions for the WT Dm3MAT3. (A) Apo WT (aWT) stacking interactions categorized as face-to-face stacking, edge-to-face stacking, or no stacking interactions. **(B)** Holo WT (hWT) stacking interactions. Face-to-face stacking is represented as blue, edge-to-face stacking as orange, and no stacking as gray. Distance measurements were taken between the centers of mass of each residue listed on the y-axis of each panel.

Figure S10

Figure S10. Substrate docking. Cyanidin 3-O-glucoside (C3G) docking into the Dm3MaT3 acceptor site for **(A)** WT, **(B)** W36A, and **(C)** C320A variants. WT is represented as purple, W36A is represented as orange-red, and C320A is represented as green.

A

B

Figure S11. SDS-PAGE and Size Exclusion Chromatography of Dm3MAT3 variants. (A) Coomassie-stained SDS-PAGE of Dm3MAT3 variants purified to homogeneity using Ni-NTA-agarose. 20 and 40 ug protein were separated on a 12% gel. Similar migration pattern and purity for all variants was observed. **(B)** Size exclusion chromatography traces of the different Dm3MAT3 variants show that mutant proteins retain to similar elution volumes as wild-type, indicating proper size and shape of the protein.