Short title: Coregulation of Cannabinoid and Terpenoid Pathways

Gene Networks Underlying Cannabinoid and Terpenoid Accumulation in Cannabis¹

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¹ This study was supported by gifts from private individuals, with no association with the cannabis industry. All work with raw materials was conducted by A.S. at a facility accredited to National Environmental Laboratory Accreditation Program standards and licensed by the Oregon Liquor Control Commission. Work of employees of Washington State University (J.J.Z, I.L., and B.M.L.) was performed in accordance with the OR/ORSO Guideline of July 2017.

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ONE-SENTENCE SUMMARY

Metabolite and transcriptome profiling of cannabis glandular trichomes differentiates strains and provides evidence for co-regulation of cannabinoid and terpenoid volatile biosynthesis.

AUTHOR CONTRIBUTIONS

Author contributions: J.J.Z., A.S. and B.M.L. designed the experiments; A.S. harvested and extracted plant materials; A.S. performed metabolite analyses; J.J.Z. and I.L. cloned terpene synthase genes and performed functional assays; J.J.Z., A.S. and B.M.L. analyzed the data; J.J.Z and B.M.L. wrote the manuscript, with input from all authors.

Key words: cannabinoid, cannabis, coexpression network, functional characterization, glandular trichome, metabolic regulation, terpene synthase, terpenoid.

Abbreviations and Acronyms:

CBD, cannabidiol; CBDA, cannabidiolic acid; BN, cannabinol; HCA, hierarchical clustering analysis; MEP, 2C-methyl-D-erythritol 4-phosphate; MVA, mevalonic acid; OPLS-DA, orthogonal projections to latent structures discriminant analysis; PCA, principal component analysis; RNAseq, ribonucleic acid sequencing; SCC, Spearman correlation coefficient; THC, tetrahydrocannabinol; THCA, tetrahydrocannabinolic acid; TPM, transcripts per kilobase million; TPS, terpene synthase; WGCNA, weighted gene correlation network analysis.

DISCLOSURE DECLARATION

JJZ and BML are members of Dewey Scientific LLC, a biotechnology company based in Pullman, WA, USA. A.S. serves as Chief Scientific Officer for Evio Labs, a cannabis analytical testing company based on Central Point, OR, USA.

- 1 Abstract
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3 Glandular trichomes are specialized anatomical structures that accumulate secretions with important biological roles in plant-environment interactions. These secretions also have 4 5 commercial uses in the flavor, fragrance, and pharmaceutical industries. The capitate-stalked 6 glandular trichomes of *Cannabis sativa* (cannabis), situated on the surfaces of the bracts of the female flowers, are the primary site for the biosynthesis and storage of resins rich in 7 cannabinoids and terpenoids. In this study, we profiled nine commercial cannabis strains with 8 9 purportedly different attributes, such as taste, color, smell and genetic origin. Glandular 10 trichomes were isolated from each of these strains and cell type-specific transcriptome data sets were acquired. Cannabinoids and terpenoids were quantified in flower buds. Statistical 11 12 analyses indicated that these data sets enable the high-resolution differentiation of strains by providing complementary information. Integrative analyses revealed a coexpression network of 13 genes involved in the biosynthesis of both cannabinoids and terpenoids from imported 14 precursors. Terpene synthase genes involved in the biosynthesis of the major mono- and 15 sesquiterpenes routinely assayed by cannabis testing laboratories were identified and 16 17 functionally evaluated. In addition to cloning variants of previously characterized genes, specifically CsTPS14CT ((-)-limonene synthase) and CsTPS15CT (β-myrcene synthase) we 18 19 functionally evaluated genes that encode enzymes with activities not previously described in cannabis, namely CsTPS18VF and CsTPS19BL (nerolidol/linalool synthases); CsTPS16CC 20 (germacrene B synthase); and CsTPS20CT (hedycaryol synthase). This study lays the groundwork 21 for developing a better understanding of the complex chemistry and biochemistry underlying 22 23 resin accumulation across commercial cannabis strains.

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27 INTRODUCTION

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29 Cannabis sativa (cannabis) was originally discovered in Central Asia and has likely been cultivated for tens of thousands of years by human civilizations, with the first mention about 30 31 5,000 years ago in Chinese texts (Unschuld, 1986). Whereas the initial utility was primarily as a 32 source of grain and fiber, strains with medicinal properties were already in use in northwest China some 2,700 years ago, as evidenced by the detection of the psychoactive cannabinoid, 33 (-)-trans- Δ^9 -tetrahydrocannabinol (THC), in plant residues recovered from an ancient grave 34 35 (Russo et al., 2008). Cannabis strains containing less THC but more of the non-psychoactive 36 cannabidiol (CBD), commonly referred to as hemp, were grown in Roman Britain for grain and fiber, but later found additional uses as a medicine during the Anglo-Saxon period (Grattan and 37 38 Singer, 1952). The 1925 Geneva International Opium Convention required signatories to control 39 the trade of certain drugs (including cannabis), which was followed by increasingly restrictive resolutions by the League of Nations and later United Nations (United Nations, 1966). Until very 40 41 recently, cannabis was considered an illicit substance of abuse by many governments, and 42 could only be researched by selected, authorized scientists in tightly supervised laboratories. 43 Despite these restrictions, evidence for the medicinal potential was sufficiently convincing that, by the mid-1980s, the synthetic cannabinoids nabilone and dronabinol had been granted 44 45 approval by the U.S. Food and Drug Administration to suppress nausea during chemotherapy (Abuhasira et al., 2018). The discovery of the existence of a high-affinity cannabinoid receptor 46 in the rat brain during the late 1980s (Devane et al., 1988) prompted further research to 47 identify the endogenous ligands. This resulted in the characterization, beginning in the early 48 49 1990s, of several lipid-based retrograde neurotransmitters (endocannabinoids) and multiple 50 enzymes involved in their biosynthesis, trafficking and perception (the endocannabinoid system), which were subsequently demonstrated to regulate a multitude of physiological and 51 cognitive processes in humans and other animals (Devane et al., 1992). With receptor targets in 52 hand, follow-up research and clinical trials brought several additional cannabis-related products 53 to the pharmaceutical marketplace, including nabiximols (marketed as Sativex[®] in Canada since 54 55 2005), a cannabis extract used to treat symptoms of multiple sclerosis, and a formulation of

highly-purified, plant-sourced CBD (marketed as Epidiolex[®] in the USA since early 2018) to treat 56 certain forms of epilepsy. In the meanwhile, several jurisdictions and even entire countries 57 changed their policies on cannabis, endorsing laws that allow its therapeutic use, and 58 decriminalizing or even legalizing it for recreational purposes (Abuhasira et al., 2018). 59 Legislation has not been able to keep up with these recent developments, and specific labeling 60 regulations with regard to the composition of active ingredients, serving sizes and 61 62 recommended doses are woefully lacking (Subritzky et al., 2016). This situation is exacerbated by an inadequate understanding of how the chemistry (cannabinoids and other specialized 63 metabolites) of cannabis extracts and formulations relates to their biological effects. 64

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Since the original structural elucidation, during the early 1960s, of THC as a psychoactive 66 principle in cannabis (Gaoni and Mechoulam, 1964), the structures of more than 90 biogenic 67 68 cannabinoids have been reported to occur in members of the genus *Cannabis* (Andre et al... 2016), with a handful of constituents being the most prominent across strains (Fig. 1). These 69 cannabinoids accumulate primarily in capitate-stalked glandular trichomes of female plants at 70 the flowering stage (Mahlberg and Kim, 2003). A second class of metabolites with high 71 72 abundance and even greater chemical diversity in cannabis glandular trichomes are monoterpenes and sesquiterpenes (Brenneisen, 2007) (Fig. 1). These volatile terpenoids are 73 responsible for the distinctive aromas of different cannabis strains. The popular press and 74 trade magazines liberally use the term "entourage effect" to suggest that synergism among 75 cannabinoids or between cannabinoids and other constituents (in particular terpenoids) may 76 contribute to different psychological perceptions of cannabis varieties by users. In support of 77 this view, β -caryophyllene, a sesquiterpene with almost ubiquitous occurrence in plant oils and 78 79 resins, was demonstrated to bind with high affinity to the CB2 cannabinoid receptor, and has 80 therefore been referred to as a dietary cannabinoid (Gertsch et al., 2008). However, there is only limited clinical evidence for entourage effects of terpenoids in cannabis formulations 81 (Gertsch et al., 2010; Russo, 2011). Irrespective of these considerations, the chemical 82 composition of each cannabis strain is unique and acquiring a "metabolic fingerprint" is an 83 excellent first step in building a more robust scientific foundation for assessing the correlation 84

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85 between the composition of plant material and the perception by users (Fischedick et al., $\frac{6}{6}$

86 2010).

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Most of the cannabis products traded licitly or illicitly today are sourced from strains for 88 89 which minimal documentation is available in the public domain, and for which the primary goal 90 was clearly to breed high-THC strains (Fidelia et al., 2012). In other words, the genetics underlying chemical diversity in commercial cannabis strains is currently poorly understood 91 92 (Welling et al., 2016). In this context, it is interesting to note that cannabinoids and terpenoids share a common biosynthetic origin. The biosynthesis of the prominent cannabinoids involves 93 two direct precursor pathways. The polyketide pathway gives rise to olivetolic acid from a 94 95 short-chain fatty acid intermediate (hexanoyl-CoA), whereas the methylerythritol 4-phosphate (MEP) pathway provides geranyl diphosphate (GPP) (Fellermeier et al., 2001; Taura et al., 2009; 96 Gagne et al., 2012; Stout et al., 2012; Page and Stout, 2013) (Fig. 1). An aromatic 97 prenyltransferase catalyzes the formation of cannabigerolic acid from oilyetolic acid and GPP 98 (Fellermeier and Zenk, 1998; Page and Boubakir, 2012). The pathway then branches again 99 toward different cyclized products, such as tetrahydrocannabinolic acid (THCA), cannabidiolic 100 acid (CBDA), and cannabichromanic acid (Marimoto et al., 1998; Sirikantaramas et al., 2005; 101 102 Taura et al., 2007) (Fig. 1). Reduced metabolic products of these acids are formed nonenzymatically by exposure to heat (Degenhardt et al., 2017). Plant monoterpenes are mostly 103 derived from the plastid-localized MEP pathway, whereas the cytosolic/peroxisomal 104 mevalonate (MVA) pathway is a common source of precursors for sesquiterpenes, although 105 106 crosstalk between both pathways has also been reported (Hemmerlin et al., 2012) (Fig. 1). Terpene synthases catalyze the first committed step in the biosynthesis of a specific terpenoid 107 from a prenyl diphosphate precursor of the appropriate chain length. To date, monoterpene 108 109 synthases (accepting a C10 precursor) and sesquiterpene synthases (acting on a C15 precursor) 110 that are responsible for the production of about half a dozen terpenoids in cannabis have been reported (Günnewich et al., 2008; Booth et al., 2017) (Fig. 1), with many more awaiting 111 functional characterization. In the current manuscript, we report the chemical profiles and 112 corresponding gene networks across several cannabis strains, thereby building the foundation 113 for a better understanding of their chemical and biochemical diversity. 114

- 117 **RESULTS**
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119 Strategic Considerations for Logistics, Strain Selection, and Experimental Design

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One of the goals of this pilot study was to test the utility of combining metabolic and 121 122 transcriptomic data to differentiate cannabis strains with regard to the most relevant traits. To ensure the consistency of data sets, all plant materials were sourced from the same facility, 123 124 where they had been maintained under comparable growth conditions (Shadowbox Farms in 125 Williams, OR, USA). Plant harvest was performed when the appearance of glandular trichome 126 content had changed from a turbid white to clear, and before another change to an amber-like color occurred. For most strains, the pistils had changed color from white to a yellow or orange. 127 128 These are the visual cues used by experienced growers to indicate optimal harvest time. All 129 further processing was performed with fresh (uncured) material to avoid the previously reported loss of terpenoid volatiles during drying (Ross and El Sohly, 1996). Cannabinoids and 130 131 terpenoids were extracted and quantified at a testing facility licensed according to the National Environmental Laboratory Accreditation Program's TNI 2009 Standard (Evio Labs, Central Point, 132 133 OR, USA). At this facility, fractions highly enriched in glandular trichomes were obtained and RNA was isolated, with minor modifications, using previously established protocols (Lange et 134 al., 2000). Glandular trichome-specific RNA sequencing (RNA-seq) data were then acquired by a 135 commercial service provider (Quick Biology Inc., Pasadena, CA). Metabolite and transcriptome 136 137 data were acquired for three biological replicates per strain.

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This study involved a selection of strains with *Cannabis sativa* L. ancestry, whereas *Cannabis indica* Lam. (formally classified as *Cannabis sativa* forma *indica*) was dominant in others (Fig. 2). Strains of *C. sativa* provenance are generally characterized by fairly thin and narrow leaves, comparatively longer flowering cycles, and a relatively tall stature. A typical example in the current study is 'Mama Thai', which is generally considered a landrace of *C. sativa*. In contrast, *C. indica* strains ordinarily have large and thick leaves, a rather short flowering cycle (6-8 weeks), and a proportionately short habitus (Fig. 2A). Our pilot study featured 'Blackberry Kush'



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Strain	Origin	Aromatic Description	Lineage
Blackberry Kush'	Indica Dominant Hybrid	Hashy, jet fuel aroma with a sweet berry taste	Afghani x Blackberry
Black Lime'	Sativa Dominant Hybrid	Pine, lemon, and black pepper	Northern Lights x Purple Kush
Canna Tsu'	Sativa Dominant Hybrid	Sweet, earthy, citrus	Cannatonic x Sour Tsunami
Mama Thai'	Sativa	Fruity, citrus	Likely a landrace
Valley Fire'	Sativa Dominant Hybrid	Earthy, pine, citrus, flowery	The White x Fire OG
Cherry Chem'	Indica Dominant Hybrid	Sweet cherry, grassy, spicy	Cherry Pie x Chemdawg
Terple'	Unclear	Earthy, woody, tobacco, spicy, slightly fruity	Unknown
Sour Diesel'	Sativa Dominant Hybrid	Pungent diesel like aroma, earthy	Chemdawg 91 x Super Skunk
White Cookies'	Sativa Dominant Hybrid	Sweet, earthy	White Widow x Girl Scout Cookies

146 as a *C. indica* dominant strain. The remaining strains were hybrids of mixed *C. sativa* and *C.*

indica lineage, plus one strain ('Terple') with poorly documented origin (Fig. 2B).

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To address our goal of assessing the utility of our data for classifying strains, RNA-seq and chemical data (cannabinoid and terpenoid profiles) were subjected to multivariate statistical analyses. We then tested the hypothesis that cannabinoid and terpenoid pathways are coregulated by performing gene coexpression network analyses. A combination of gene network and phylogenetic analyses was subsequently used to identify candidate genes for hitherto uncharacterized terpene synthases that contribute significantly to the cannabis volatile bouquet.

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157 Strain Differentiation Based on RNA-seq Data

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159 High-quality libraries reflecting transcripts expressed in isolated glandular trichomes were subjected to RNA-seg analysis (nine strains, three biological replicates each, 27 samples total) 160 on the Illumina HiSeq 4000 platform. A *de novo* consensus transcriptome assembly was 161 generated using the Trinity suite (Haas et al., 2013) (assembly statistics in Supplemental Table 162 163 S1). The reads were assembled into contigs covering a total of 305 Mbp of sequence with a GC 164 content of 40.4%. The resulting assembly produced an N50 value of 833 bp, containing 514,208 contigs of at least 201 bp in length. The assembled transcriptome data set was searched against 165 the NCBI non-redundant protein database, which resulted in the annotation of 82,523 166 sequences at e-values < 1e-5. Read counts for each transcript in each sample were then 167 processed with the RSEM software package (Li and Dewey, 2011) to calculate normalized 168 expression levels as Transcripts Per Kilobase Million (TPM). Transcripts with TPM values lower 169 170 than 5.0 across all varieties were removed from subsequent analysis, resulting in 46,559 171 predicted genes with significant expression (Supplemental Table S2).

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As a first step to investigate the utility of RNA-seq for strain categorization, transcriptome data sets were subjected to <u>Principal Component Analysis</u> (PCA), a statistical procedure that reduces attribute space from a larger number of variables to a smaller number of so-called



principal components, thereby decreasing the dimensionality of the original data. The first 176 three principal components accounted for 83% of the variability in the data set (Fig. 3A). The 177 replicates for each strain clustered together in a three-dimensional PCA plot, whereas the 178 component scores for each strain were separated from those of all other strains, indicating that 179 the overall transcriptome of each strain is unique (Fig. 3A). Processing of RNA-seq data by 180 181 Hierarchical Clustering Analysis (HCA), which builds a cluster hierarchy that is commonly displayed as a dendrogram, grouped strains into two major clades (Fig. 3B). The first clade 182 contained 'Blackberry Kush', 'Cherry Chem' and 'Terple', whereas the second consisted of 183 'Mama Thai', 'White Cookies', 'Valley Fire', 'Black Lime', 'Canna Tsu' and 'Sour Diesel', indicating 184 a clear separation of strains by heritage (C. indica for clade 1 and C. sativa for clade 2). 185

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187 Strain Differentiation Based on Metabolite Profiling Data

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The highly robust analytical platforms that served as the basis for the analysis of six 189 190 cannabinoids and 24 terpenoids were described in a previous report (Fischedick et al., 2010) 191 and used here with minor modifications. Cannabinoid concentration was highest in 'White Cookies' (28.4% of flower bud dry weight), with relatively high contents also occurring in 192 'Cherry Chem' (17.7%), 'Black Lime' (17.5%), 'Backberry Kush' (15.8%), 'Valley Fire' (15.7%), 193 'Terple' (15.6%), 'Sour Diesel' (12.4%), and 'Canna Tsu' (12.2%) (Table 1). Significantly lower 194 concentrations were detected in 'Mama Thai' (6.4%). In eight of the nine strains investigated, 195 THCA was the major cannabinoid, ranging from 26.3% of the flower bud dry weight in 'White 196 Cookies' to 5.9% in 'Mama Thai' (Table 1). The only exception was the 'Canna Tsu' strain, in 197 198 which CBDA (7.8% of flower bud dry weight) dominated over THCA (3.2%), whereas CBDA in all other strains remained at 1% or less. Two additional cannabinoids of fairly high abundance 199 200 were cannabinol (CBN), which accumulated to 0.2-1.7% of flower bud dry weight, and 201 tetrahydrocannabinol, which amounted to 0.2–1.6% (Table 1) (for structures see Fig. 1). Cannabichromene was not detected in any of the sampled varieties. 202

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Terpenoid content was highest in 'Black Lime' (8.8% of flower bud dry weight), with fairly 204 205 high contents also occurring in 'White Cookies' (4.8%), 'Terple' (4.8%), 'Valley Fire' (3.8%), 'Cherry Chem' (3.5%), 'Blackberry Kush' (3.4%), and 'Canna Tsu' (3.3%) (Table 1). Significantly 206 lower concentrations were detected in 'Sour Diesel' (1.8%) and 'Mama Thai' (0.7%). The 207 monoterpene (C10) to sesquiterpene (C15) ratio was generally very high (> 10), with only three 208 strains in which the ratio was below 3 ('Cherry Chem', 'Mama Thai', and 'Sour Diesel') (Table 1). 209 It should be noted that this ratio only applies to the terpenoids we were able to quantify based 210 on the availability of authentic standards. β -Myrcene was the most abundant monoterpene in 211 212 most strains (up to 4.3% of flower bud dry weight in 'Black Lime'). The only exceptions were 213 'Mama Thai' (generally low terpenoid contents, with terpinolene as most abundant monoterpene at 0.1%) and 'White Cookies' (with limonene at 1.5%) (Table 1). Limonene 214 content was also high in 'Black Lime' (0.9%) and 'Valley Fire' (0.7%). α -Pinene and β -pinene 215 216 amounts were quite high in 'Black Lime' (2.0% and 0.5%, respectively). 1,8-Cineole was 217 particularly abundant in 'Canna Tsu' and 'Cherry Chem' (0.5% in both) (Table 1). All other

Table 1. Constituents of Cannabi	is Female Flower Buds (Me	etabolite Content in Ni	ne Strains Expressed a	as Percent of Dry Weig	ht).				
Abbreviations: n.d., not detectab	ble.								
Metabolite	Black Berry Kush	Black Lime	Canna Tsu	Cherry Chem	Valley Fire	Mamma Thai	Sour Diesel	Terple	White Cookies
Cannabinoids									
Tetrabydrocannahinolic acid	13 56 ± 0 90	15.02 + 1.10	3 19 + 0 20	16 55 ± 0.81	13 89 + 1 33	5 91 + 0 60	11 31 + 1 04	13 72 + 1 36	26 33 + 0 54
Tetrahydrocannabinol	0.31 + 0.02	1.62 + 0.19	0.55 ± 0.055	0.15 + 0.008	0.41 + 0.049	0.14 ± 0.02	0.22 ± 0.027	1.15 + 0.12	0.86 ± 0.091
Cannabidiolic acid	0.45 ± 0.02	0.12 ± 0.012	7.76 ± 0.63	0.079 ± 0.007	0.037 ± 0.001	0.016 ± 0.003	0.032 ± 0.002	0.067 ± 0.002	0.088 ± 0.004
Cannabidiol	0.95 ± 0.07	0.139 ± 0.016	0.085 ± 0.013	0.079 ± 0.008	0.12 ± 0.004	0.047 ± 0.005	0.086 ± 0.006	0.11 ± 0.008	0.098 ± 0.013
Cannabigerol	0.12 ± 0.015	0.086 ± 0.008	0.093 ± 0.008	0.051 ± 0.005	0.15 ± 0.027	0.016 ± 0.001	0.052 ± 0.004	0.093 ± 0.002	0.25 ± 0.005
Cannabinol	1.74 ± 0.20	0.55 ± 0.019	0.53 ± 0.051	0.83 ± 0.019	1.12 ± 0.14	0.29 ± 0.028	0.68 ± 0.033	0.502 ± 0.007	0.78 ± 0.025
Cannabichromene	n.d.	n.d.	n.d.	n.d.	n.d.	n.d.	n.d.	n.d.	n.d.
Total Cannabinoids	15.87 ± 1.13	17.53 ± 1.25	12.20 ± 0.85	17.74 ± 0.82	15.70 ± 1.53	6.41 ± 0.65	12.387 ± 1.05	15.64 ± 1.47	28.40 ± 0.54
Monoterpenes									
β-Myrcene	2.35 ± 0.2	4.34 ± 0.36	1.70 ± 0.15	1.61 ± 0.049	2.24 ± 0.28	0.11 ± 0.009	0.70 ± 0.046	2.96 ± 0.25	1.14 ± 0.17
(-)-Limonene	0.29 ± 0.02	0.89 ± 0.08	0.16 ± 0.021	0.23 ± 0.015	0.65 ± 0.098	0.03 ± 0.003	0.17 ± 0.011	0.23 ± 0.019	1.53 ± 0.24
α-Pinene	0.015 ± 0.001	1.99 ± 0.12	0.38 ± 0.039	0.016 ± 0.001	0.044 ± 0.008	0.007 ± 0.001	0.004 ± 0	0.82 ± 0.051	0.20 ± 0.032
β -Pinene	0.086 ± 0.005	0.50 ± 0.034	0.18 ± 0.025	0.056 ± 0.003	0.11 ± 0.013	0.026 ± 0.002	0.039 ± 0.002	0.31 ± 0.022	0.04 ± 0.007
1,8-Cineole	0.26 ± 0.02	0.38 ± 0.038	0.52 ± 0.075	0.464 ± 0.012	0.22 ± 0.028	0.057 ± 0.007	0.11 ± 0.011	0.00 ± 0	0.31 ± 0.037
Linalool	0.082 ± 0.005	0.079 ± 0.004	0.052 ± 0.005	0.13 ± 0.003	0.16 ± 0.027	0.023 ± 0.002	0.074 ± 0.005	0.067 ± 0.006	0.57 ± 0.072
Terpinolene	0.019 ± 0.001	0.034 ± 0.003	0.019 ± 0.002	0.019 ± 0.001	0.02 ± 0.003	0.13 ± 0.016	0.017 ± 0.001	0.02 ± 0.002	0.041 ± 0.006
Borneol	0.039 ± 0.002	0.041 ± 0.003	n.d.	0.032 ± 0.002	0.033 ± 0.005	0.021 ± 0.002	0.026 ± 0.002	0.036 ± 0.002	0.048 ± 0.008
β-Ocimene	n.d.	0.039 ± 0.003	n.d.	n.d.	0.006 ± 0.001	0.13 ± 0.014	n.d.	0.086 ± 0.007	0.015 ± 0.002
Camphene	n.d.	0.089 ± 0.008	0.055 ± 0.007	n.d.	0.004 ± 0.001	n.d.	n.d.	0.019 ± 0.002	0.07 ± 0.012
δ-3-Carene	0.029 ± 0.002	0.052 ± 0.006	0.003 ± 0	0.008 ± 0.001	0.022 ± 0.003	0.003 ± 0.001	n.d.	0.027 ± 0.002	0.016 ± 0.002
Camphor	0.044 ± 0.003	0.006 ± 0.001	n.d.	n.d.	n.d.	n.d.	n.d.	n.d.	0.101 ± 0.013
(+)-Terpinene	0.001 ± 0.001	n.d.	n.d.	n.d.	n.d.	0.005 ± 0.001	n.d.	n.d.	0.002 ± 0
Total monoterpenes	3.23 ± 0.26	8.43 ± 0.66	3.07 ± 0.32	2.56 ± 0.085	3.52 ± 0.47	0.54 ± 0.057	1.14 ± 0.078	4.57 ± 0.36	4.09 ± 0.60
Sesquiterpenes									
β -Caryophyllene	0.13 ± 0.01	0.24 ± 0.023	0.21 ± 0.022	0.74 ± 0.012	0.23 ± 0.034	0.12 ± 0.013	0.45 ± 0.026	0.15 ± 0.009	0.60 ± 0.068
α-Humulene	0.03 ± 0.002	0.06 ± 0.005	0.051 ± 0.005	0.20 ± 0.011	0.087 ± 0.014	0.068 ± 0.008	0.19 ± 0.009	0.058 ± 0.003	0.15 ± 0.018
Nerolidol	n.d.	0.06 ± 0.004	n.d.	n.d.	n.d.	n.d.	n.d.	n.d.	n.d.
Total sesquiterpenes	0.16 ± 0.015	0.361 ± 0.032	0.26 ± 0.027	0.93 ± 0.019	0.32 ± 0.048	0.19 ± 0.021	0.64 ± 0.035	0.21 ± 0.012	0.75 ± 0.086
Total terpenoids	3.39 ± 0.27	8.79 ± 0.69	3.33 ± 0.35	3.49 ± 0.10	3.84 ± 0.51	0.73 ± 0.078	1.78 ± 0.11	4.78 ± 0.38	4.83 ± 0.69

218 monoterpenes had concentrations below 0.2%. All strains contained sesquiterpenes, of which 219 β -caryophyllene was consistently the most abundant (0.1–0.7% of flower bud dry weight). α -220 Humulene was also detectable in all strains (< 0.2%), whereas 'Black Lime' was the only strain in 221 which the nerolidol concentration rose above the limit of quantitation (< 0.1%) (Table 1).

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Processing of the metabolite data (cannabinoids and terpenoid profiles) by PCA resulted in 223 a clear separation of the strains, with individual biological replicates clustering closely together 224 225 (Fig. 4A). Remarkably, 99% of the data variation across genotypes was captured by the first 226 three principal components. Application of Orthogonal Projections to Latent Structures Discriminant Analysis (OPLS-DA), a statistical modeling tool used commonly in metabolomics 227 research (Worley, 2013), indicated a separation of strains into two groups based on our 228 metabolite profiling data, one representing the *C. indica*-dominant strains, whereas another 229 constituted the C. sativa-dominant strains (Fig. 4B). Biological replicates for each strain once 230 again clustered together, whereas significant separation was observed across strains. In 231 summary, glandular trichome-specific gene expression and metabolite data were consistent in 232 233 differentiating cannabis strains.

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Evidence for Co-Expression of Cannabinoid and Terpenoid Pathways
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237 Our glandular trichome RNA-seq data sets were filtered to eliminate genes with consistently 238 low expression levels (below 50 TPM), thereby retaining roughly 16,000 expressed genes with significant expression levels in at least one strain. Gene abundance across strains was then 239 evaluated using the Weighted Gene Correlation Network Analysis (WGCNA) package in R 240 (Langfelder and Horvath, 2008), which resulted in the binning of genes (only those with 241 Spearman Correlation Coefficients (SCCs) of ≥ 0.8 were considered) into seven co-expression 242 modules (Supplemental Table S3). Further analysis using the 'moduleEigengenes' function 243 244 indicated that the accumulation of CBDA, the signature cannabinoid of the 'Canna Tsu' strain, was highly correlated (SCC of 0.97, P-value of 2e-17) with one of the co-expression modules 245 (indicated by brown color in Fig. 5A). Interestingly, this module contained the gene coding for 246 CBDA synthase, the enzyme responsible for the conversion of cannabigerolic acid to CBDA 247 (Table 2). An analogous analysis for THCA or THC (which correlated with a module indicated by 248 yellow color in Fig. 5A) and THCA synthase was not possible, because single nucleotide 249 polymorphisms in this gene (and not lack of expression) result in an inactive enzyme in strains 250 251 that accumulate primarily CBDA (Kojoma et al., 2006; Laverty et al., 2018) (Table 2).



252 Interestingly, the THCA synthase sequences were essentially identical, with the exception of that of the 'Canna Tsu' strain, the only CBDA accumulator in our pilot study (Supplemental Fig. 253 S1). Consequently, a full-length CBDA synthase gene was expressed only in the 'Canna Tsu' 254 strain (Supplemental Fig S2), which is novel information that furthers our understanding of the 255 mechanisms underlying CBDA accumulation. Finally, the yellow-colored module (which as 256 257 mentioned above contained THCA synthase) also comprised cannabigerolic acid synthase (Table 258 2), the gene preceding tetrahydrocannabinolic acid synthase in the cannabinoid pathway (Fig. 259 1), thereby providing additional evidence for gene-to-metabolite correlation in the cannabinoid 260 pathway.

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We then asked if similar gene-to-metabolite correlations occurred in the terpenoid pathway. Interestingly, two co-expression modules (indicated by black and yellow color in Fig.

Table 2	 Transcript 	Abundance fo	or Genes	Involved	in the E	liosynthesis	s of Canna	binoids and	d Terpenoids i	n Cannabis St
Abbrev	iations: n.d	., not detectab	ole; MEP,	2C-methy	yl-D- er	ythritol 4-p	hosphate;	MVA, mey	alonate.	

Gene Annotation	UniProt Identifier	Transcript Abundance [Transcripts Per Kilobase								
		Million]	Die ekstimen	Come Tour	Charge Charge	Manua Thai	Cours Discol	Terrie	Volley Fire	White Cookies
Cannabing id nathway		Black Berry Kush	Black Lime	Canna Tsu	Cherry Chem	wama Inai	Sour Diesei	repie	valley Fire	White Cookies
Acyl activating enzyme 1	HOA1V3 CANSA	80.63	160.44	316.06	840.92	377 29	307 00	93 59	188 84	229.65
Olivetol synthese		3946.85	9454.00	10400.03	14619.66	17955.05	4984.60	9706.06	11374 75	12373 11
Geranyl dinbosphate: olivetolate	CsPT1	422.42	222.42	189.43	407.76	649 37	263.62	246.13	175.87	115 21
geranyltransferase	00/11	422.42	222.92	105.45	407.70	045.57	200.02	240.10	1/5.0/	110.21
CBDA synthase	CBDAS CANSA	n.d.	n.d.	1282.46	n.d.	n.d.	18.39	n.d.	n.d.	n.d.
THCA synthase	THCAS CANSA	885.17	423.29	1203.31	2321.64	2317.68	1557.54	619.22	309.23	524.08
MEP pathway										
1-Deoxy-D-xylulose-5-phosphate synthase	A0A1V0QSH6_CANSA	221.85	284.41	412.74	1627.02	319.76	1751.70	533.57	288.69	16.32
1-Deoxy-D-xylulose 5-phosphate	A0A1V0QSG8_CANSA	172.63	228.15	185.07	667.96	304.62	117.92	176.79	256.25	16.01
reductoisomerase	_									
2-C-Methyl-D-erythritol 4-phosphate	A0A1V0QSI6_CANSA	36.77	95.99	96.25	168.24	160.40	146.38	46.96	75.40	64.73
cytidylyltransferase										
4-(Cytidine 5'-diphospho)-2-C-methyl-D-	A0A1V0QSI2_CANSA	35.20	3.70	67.94	211.85	212.43	109.88	57.60	104.05	80.23
erythritol kinase										
2-C-Methyl-D-erythritol 2,4,-	G9C075_HUMLU	67.75	118.23	315.86	338.21	184.98	419.84	69.75	171.17	207.15
cyclodiphosphate synthase										
(E)-4-Hydroxy-3-methylbut-2-enyl-	A0A1V0QSG3_CANSA	107.65	287.57	794.25	744.09	444.09	596.36	349.56	297.07	317.55
diphosphate synthase										
(E)-4-Hydroxy-3-methylbut-2-enyl-	A0A1V0Q5H9_CANSA	1485.98	561.96	3447.50	3468.57	3090.49	3024.22	1889.37	1031.90	4544.35
diphosphate reductase										
Isopentenyldiphosphate isomerase	A0A1V0QSG5_CANSA	165.10	272.72	433.46	1836.07	306.03	347.85	476.86	509.70	9.96
MVA pathway		20.25	11.00	252.20	202.50	212.00		252.40	54.25	240.42
Acetoacetyi-CoA thiolase	AUA1VUQSH3_CANSA	38.35	11.90	253.38	302.58	313.99	134./1	252.40	54.35	248.13
2 Hudsons 2 mothulalutand soonnume A	AGAINOOSH2 CANEA	12.44	22.08	20.91	21.60	37.91	24.22	0.24	10.22	01.24
sunthase	AUAIVUQ5H5_CANSA	13.44	22.90	20.81	21.00	27.01	34.33	9.24	19.52	91.24
3-Hydroxy-3-methylalutapil-coenzyme A	ADA1VOOSES CANSA	26.69	56.02	21.92	42.41	29.05	107 71	19.75	69 30	48.26
reductase	NONITOODID_CHINDH	20.05	50.55	21.92	43.41	25.05	107.71	19.75	09.50	40.20
Mevalonate kinase	A0A1VOOSIO CANSA	1.63	1449.32	3.63	3.41	5.81	4.75	2.45	5.93	5.05
Phosphomevalonate kinase	A0A1V0OSH8 CANSA	3.68	7.58	7.99	6.63	8.09	6.03	3.81	7.40	305.27
Mevalonate diphosphate decarboxylase	A0A1V0OSG4 CANSA	5.00	11.89	10.21	14.89	21.24	19.39	9.67	9.64	9.96
		2.00	22100					2.07	2.701	2.00

1

rains

5A) correlated with β -myrcene accumulation (Fig. 5B). This metabolite is formed by a 264 monoterpene synthase encoded by the CsTPS3FN gene (Booth et al., 2017), which was 265 contained in one of these modules (yellow color in Fig. 5A) (Table 3). Analogous gene-to-266 267 metabolite correlations were observed for limonene and CsTPS1FN, α -pinene and CsTPS2FN, β ocimene and CsTPS6FN, and β -carvophyllene/ α -humulene and CsTPS9FN (color of modules in 268 Fig. 5A: black, yellow, yellow, turquoise, respectively; terpene synthase annotation based on 269 Günnewich et al., 2007 and Booth et al., 2017) (Fig. 5B). Transcripts corresponding to CsTPS5FN 270 $(\beta$ -myrcene/ α -pinene synthase), CsTPS4FN (alloaromadendrene synthase), CsTPS8FN (γ -271 eudesmol/valencene synthase) and CsTPS13PK (a second β -ocimene synthase) (Booth et al., 272 2017) remained below the threshold expression level in our data sets. The corresponding 273 274 terpenoids were not detected in the strains investigated, indicating that the expressed gene complement was generally sufficient to account for the presence of the major terpenoids 275

 Table 3. Transcript Abundance for Terpene Synthases Across Cannabis Strains.

 Symbols: *, functionally characterized as part of the present study; #, from Booth et al., 2017

Gene	GenBank	CsTPS	Transcript Abundance (TPM)								
	Accession	Identifier	Black Berry Kush	Black Lime	Canna Tsu	Cherry Chem	Valley Fire	Mamma Thai	Sour Diesel	Terple	White Cookies
Monoterpene synthases (TPS-b clade)											
(-)-Limonene synthase *	MK801766	CsTPS14CT	646.24	898.94	612.37	651.86	2272.48	751.48	201.94	2.46	895.86
(+)-α-Pinene synthase #	KY014565	CsTPS2FN	217.36	2041.33	1554.77	101.32	96.90	n.d.	n.d.	1298.95	49.52
β-Myrcene synthase*	MK801765	CsTPS15CT	183.29	597.88	325.85	272.65	157.78	254.10	183.29	436.63	n.d.
β -Myrcene/(-)- α -pinene synthase #	KY014560	CsTPS5FN	217.59	640.97	483.09	547.24	157.78	445.85	125.94	472.33	50.51
(E)-β-Ocimene synthase #	KY014563	CsTPS6FN	n.d.	n.d.	n.d.	n.d.	n.d.	103.41	n.d.	191.65	n.d.
(Z)-β-Ocimene synthase #	KY014558	CsTPS13PK	n.d.	n.d.	n.d.	n.d.	n.d.	n.d.	n.d.	n.d.	n.d.
Acyclic terpene synthases (TPS-g clade)											
(E)-Nerolidol/(+)-linalool synthase *	MK801764	CsTPS18VF	2.82	9.41	2.62	16.21	16.39	2.51	4.80	16.77	8.76
(E)-Nerolidol/linalool synthase *	MK801763	CSTPS19BL	56.78	81.13	27.22	80.23	249.23	62.53	47.73	90.86	66.47
Sesquiterpene synthases (TPS-a clade)											
Alloaromadendrene synthase #	KY014564	CsTPS4FN	n.d.	108.92	n.d.	639.56	n.d.	329.87	148.17	n.d.	323.36
γ-Eudesmol/valencene synthase #	KY014556	CsTPS8FN	n.d.	n.d.	n.d.	n.d.	n.d.	n.d.	n.d.	n.d.	n.d.
(putative)											
δ-Selinene synthase # (putative)	KY014554	CsTPS7FN	356.34	n.d.	367.47	n.d.	316.74	210.58	n.d.	n.d.	268.50
β-Caryophyllene/α-humulene synthase #	KY014555	CsTPS9FN	764.18	794.46	435.11	3241.85	1090.94	738.74	555.25	495.72	591.86
Germacrene B synthase *	MK131289	CsTPS16CC	16.14	19.44	9.13	156.08	20.60	40.36	20.22	7.19	22.72
Hedycaryol synthase *	MK801762	CSTPS20CT	310.43	27.00	498.70	98.21	19.35	11.98	17.67	0.00	17.02

1

(Table 3). Linalool and nerolidol were exceptions for which the corresponding terpene synthases had hitherto not been identified from cannabis. Notably, genes involved in the formation of these terpenoids (and others) were cloned and functionally characterized as part of the current study (details below), which contributes significantly to a better understanding of the genetic underpinnings of terpenoid diversity.

281

The yellow module featured prominently in our gene-to-metabolite correlation analysis for the cannabinoid and terpenoid pathways. Interestingly, a Gene Ontology (GO) analysis implied a substantial enrichment of genes involved in terpenoid biosynthesis in the yellow module (*P*value of 1.4e-05) (Supplemental Table S3) (note that GO terms for cannabinoid biosynthesis as a biological process have not yet been released). Interestingly, a total of 22 genes involved in the conversion of precursor metabolites into cannabinoid and terpenoid end products were coexpressed with THCA synthase (Fig. 5C). Specifically, these genes code for enzymes involved in glycolysis (conversion of an imported carbon source into triose phosphates and pyruvic acid), the MEP pathway toward GPP and ultimately monoterpenes, the production of sesquiterpenes, the formation of olivetolic acid from fatty acid precursors, and the incorporation of olivetolic acid and GPP into cannabinoids (Fig. 5D).

293

294 Target Gene Identification and Characterization

295

Building on our terpenoid profiling and glandular trichome-specific transcriptome data sets, we 296 297 embarked on gene discovery efforts aimed at characterizing terpene synthases associated with the biosynthesis of major mono- and sesquiterpenes routinely quantified in commercial 298 299 cannabis testing, as well as other terpenoids that are not assayed routinely. The analytical 300 chemistry data were employed to assess which genes would be expected to be expressed to 301 support the observed terpenoid profiles. We then performed BlastX searches with previously characterized terpene synthases to identify contigs with high sequence identity in our 302 transcriptome data sets. We then asked which of the putative cannabis terpene synthases 303 304 were expressed at appreciable levels in particular cannabis strains. Sequences of selected contigs were then chosen to perform a sequence relatedness analysis with previously 305 characterized terpene synthases, thereby enabling their categorization by class. cDNAs of 306 307 putative terpene synthases were cloned into appropriate vectors, expressed heterologously in 308 Escherichia coli, the corresponding recombinant proteins purified, and assays performed with appropriate prenyl diphosphate substrates. Expression for genes putatively encoding geranyl 309 diphosphate synthase and trans, trans-farnesyl diphosphate synthase was readily detectable in 310 transcriptome data sets of all strains; in contrast, no putative orthologs of neryl diphosphate 311 synthase and *cis,cis*-farnesyl diphosphate synthase were recognizable based on sequence 312 identity (Supplemental Table S1 and S2). Nevertheless, terpene synthase assays were 313 performed with GPP, neryl diphosphate (NPP), 2-trans, 6-trans farnesyl diphosphate (tFPP) and 314 315 2-cis, 6-cis-farnesyl diphosphate (cFPP).

316

317 β -Myrcene and (-)-limonene were principal monoterpenes in all strains (Table 1) and, expectedly, contigs with high sequence identity to the previously characterized β -myrcene and 318 (-)-limonene synthases of cannabis (Günnewich et al., 2007; Booth et al., 2017), which belong to 319 the TPS-b clade of terpene synthases (Fig. 6 and Supplemental Table S4), were expressed at 320 321 high levels across most strains investigated in the present study (Table 2). Cloning was successful for the corresponding cDNAs from the 'Canna Tsu' strain (CsTPS14CT and 322 CsTPS15CT), and a functional evaluation confirmed the annotation ((-)-limonene synthase and 323 β -myrcene, respectively) (Fig. 7A, B). The translated peptide sequences of β -myrcene synthases 324 (CsTPS3FN and CsTPS15CT; excluding plastidial targeting sequence) had thirteen mismatches 325 326 (Supplemental Fig. S3) but identical specificity (100% β -myrcene as product with GPP as substrate). The sequence of the (-)-limonene synthase characterized as part of the present 327 study (CsTPS14CT; excluding plastidial targeting sequence) had two mismatches when 328 329 compared to CsTPS1SK and nine mismatches when compared to CsTPS1FN (Supplemental Fig. 330 S3). As described for CsTPS1SK, CsTPS14CT generated several other products, and we report the stereochemistry of those (Fig. 7A). 331

332

333 The monoterpene linalool was accumulated to fairly high amounts in the 'Valley Fire' and 334 'White Cookies' strains, whereas the sesquiterpene nerolidol was guantifiable only in the 'Black Lime' strain (Table 1). Contigs with moderate sequence identity (slightly above 50%) to 335 bifunctional nerolidol/linalool synthases (strawberry: Aharoni et al., 2004; snapdragon: 336 Nagegowda et al., 2008) and considerable expression in glandular trichomes were identified in 337 our transcriptome data sets (Table 3), and corresponding cDNAs were cloned from the 'Valley 338 Fire' (CsTPS18VF) and 'Black Lime' (CsTPS19BL) strains. These sequences belong to the TPS-g 339 340 clade of terpene synthases (Fig. 6 and Supplemental Table S4). Heterologous expression and 341 functional characterization confirmed that the corresponding recombinant proteins were able to catalyze the formation of (E)-nerolidol from tFPP and linalool from GPP, but no activity was 342 detected with NPP or cFPP (Fig. 8A, B). Interestingly, follow-up chiral separation of products 343 from assays performed with GPP as substrate indicated that CsTPS18VF generated almost 344 exclusively (+)-linalool, whereas CsTPS19BL produced a mixture of (-)-linalool and (+)-linalool 345



(Fig. 7C, D). Sequence differences across sesquiterpene synthases with different product
 profiles included residues with potential roles in catalysis (Fig. 9), and the implications are
 evaluated in the Discussion section.

349



To further investigate the genetic potential for generating terpenoid chemical diversity, two representatives of the TPS-b clade of terpene synthases (CsTPS16CC and CsTPS20CT) were selected for functional characterization. Cs*TPS16CC* had very high expression levels in the

Α

CsTPS18VF	ISDAWKCLNKECILR	NPAFPPPFLKASLNLARL	VPLMYNYDH-NQRLPH	LEEHIKSLL	554
CsTPS19BL	ISDAWKCLNKECILR	NPAFPPPFLKASLNLARL	VPLMYNYDH-NQRLPH	LEEHIKSLL	554
AmNES/LIS12	ISSEWKLLNKECFNLM	NHVSTSSIKKAALNTARM	VPLMYSYDE-N <mark>Q</mark> GLPI	LEEYVKIMLFD	563
AmNES/LIS2	ISSEWKLLNKECFSLN	NHVSTSSLKKAALNTAKI	VPLMYSYDE-N <mark>Q</mark> RLPI	LEEYVKIMLFD	563
FvNES1	ISDEWKKLNRELLSPN	I-PFPATFTSASLNLARM	IPLMYSYDG-NQSLPS	LKEYMKLMLYE	573
FaNES2	ISDEWKKLNRELLSPN	I-PFPATITLASLNLARM	IPLMYSYDG-NQCLPS	LKEYMKLMLYE	571
FaNES1	ISDEWKKLNRELLSPN	I-PFPASFTLASLNLARM	IPLMYSYDG-NQCLPS.	LKEYMKLMLYE	513
CsTPS16CC CsTPS20CT	VVNLWKEINQEFLR-I VDTHWKEINEDFIR-I	PTSMPSSILVRILNFTKV PAVVPFPILVRVLNFTKI	LDIIYKEGD-G Y THVG VDLLYKEGDDQ Y TNVG	KLVKDSVAALL KVLKESIAALL	565 545
B	✓ →			PS20CT	\bigcirc
0 PP		∕ <u>⊳</u> @@	⊕]	но-Т	
Farnesyl diphosphate		Nerolidyl diphosphate		Hedycar	ryol
(FPP)	CsTPS16CC	CsTPS18VF CsTPS19BL		ļ	Heat
Heat		С		HO	
-Flomono	Germacrene B	(E)-Nerolidol		Elemo	bl

353 'Cherry Chem' strain (Table 3). The sequence was most similar to that of the previously
354 characterized alloaromadendrene synthase (Booth et al., 2017) (Fig. 6 and Supplemental Table
355 S4). In our assays, the recombinant protein generated germacrene B from tFPP (Fig. 8C), with γ-



elemene being detected as a thermal breakdown product (de Kraker et al., 1998). Other prenyl
diphosphate substrates were not accepted as substrates with appreciable conversion rates (Fig.
8). The 'Canna Tsu' strain had a particularly high expression level of Cs*TPS20CT* (Table 3). Its

closest neighbor in the sequence relatedness tree was a putative δ -selinene synthase from 359 360 cannabis (Booth et al., 2017) (Fig. 6 and Supplemental Table S4). Functional assays with the 361 purified, recombinant protein indicated a conversion of tFPP to elemol, a thermal breakdown product of the sesquiterpene hedycaryol (Koo and Gang, 2012; Hattan et al., 2016), but there 362 was little or no activity with other prenyl diphosphate substrates (Fig. 8D). In summary, we 363 demonstrate that the resources and approaches described here can be employed to identify 364 candidates and subsequently characterize functions of terpene synthase genes that belong to 365 three different clades, thereby contributing to a better understanding of the genetic 366 determinants of terpenoid chemical diversity in cannabis. 367

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369

- 370 **DISCUSSION**
- 371

372 Utility of Transcript Profiling for Strain Differentiation

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Competition in decriminalized retail markets for cannabis has put pressure on breeders to 374 375 differentiate their product from that of their competitors. This has led to branding with a plethora of distinct and memorable names, which has caused both confusion and controversy 376 (Small, 2015). Chemical profiling can be employed as a powerful tool in strain differentiation 377 378 but adding genotyping information further increases the resolution of the analysis. The 379 differentiation of drug-type and fiber-type cannabis strains can be achieved with standard genotyping analyses (Piluzza et al., 2013). However, a differentiation of genetically related 380 381 strains has been much more challenging (Sawler et al., 2015; Punja et al., 2017). Traditional genotyping approaches benefit significantly from high-quality reference genome sequences 382 (Scheben et al., 2017) but, unfortunately, only fairly low-quality genome sequences have been 383 published for two cannabis strains (van Bakel et al., 2011). We employed RNA-seg as an 384 385 alternative approach for genotyping (Haseneyer et al., 2011), which does not depend on prior 386 sequence data (Wang et al., 2009). We used RNA-seq to obtain the transcriptome of glandular trichome cells of nine selected cannabis strains (with three biological replicates each). 387 Importantly, statistical analyses of these data sets allowed the differentiation of strains into 388 broader clades (descendants of landraces of C. sativa or C. indica), but also resulted in the full 389 separation of all individual strains (with biological replicates clustering closely together) (Fig. 3). 390 We fully recognize that RNA-seq is not a viable option for routine genotyping but it can be used 391 392 to develop Single Nucleotide Polymorphism (SNP)-based genotyping platforms. This approach 393 has been employed successfully for a number of crops, including alfalfa (Medicago sativa; Yang et al., 2011), maize (Zea mays; Hansey et al., 2012), and wheat (Triticum aestivum; Ramirez-394 Gonzalez et al., 2015). Our data sets are therefore highly valuable for building resources for 395 follow-up research with cannabis. As an added benefit, RNA-seq data can be used for gene 396 expression analysis, thereby providing a functional context, which is discussed in more detail 397 398 below.

399

400 Utility of Metabolite Profiling for Strain Differentiation

401

We assessed the utility of cannabinoid and terpenoid profiling, in addition to strain 402 403 differentiation by genotyping as discussed above, to demarcate nine commercial cannabis strains. Two independent statistical approaches, PCA and OPLS-DA, grouped biological 404 405 replicates closely together, while still separating individual strains and classes of strains (those of C. sativa or C. indica heritage) (Fig. 4). Several authors have advocated the profiling of both 406 cannabinoids and terpenoids in recent publications (Fischedick et al., 2010; Elzinga et al., 2015; 407 408 Aizpurua-Olaizola et al., 2016; Hazekamp et al., 2016; Fischedick, 2017; Lewis et al., 2018; Orser et al., 2018; Richins et al., 2018; Sexton et al., 2018). The key advantage of this approach over 409 merely profiling cannabinoids lies in the enormous diversity of terpenoids accumulated in 410 411 cannabis (and in other plants as well), which significantly increases the power of statistical analyses. It also reflects the fact that many users select cannabis strains based on both the 412 reported THC content and aroma (which is largely imparted by terpenoids) (Gilbert and DiVerdi, 413 2018). A comprehensive analysis of cannabis strains recently indicated the presence of close to 414 415 200 detectable volatiles, which were tentatively identified based on searches against various 416 spectral databases (Rice and Koziel, 2015). A notable challenge with terpenoid profiling pertains to the limitation that authentic standards are often very costly or unavailable from commercial 417 sources, which is particularly true for sesquiterpenes (dozens detected by Rice and Koziel, 418 2015). Commercial cannabis testing laboratories therefore rarely offer services that comprise 419 more than 20 terpenoids. While such analyses may detect the most abundant terpenoids for 420 popular strains, it is not unlikely that important aroma volatiles with a low odor detection 421 422 threshold could be missed (Chin and Marriott, 2015). Another reason why a comprehensive 423 profiling of terpenoids would be desirable relates to testing the validity of the "entourage effect", the proposed synergism between cannabinoids and other constituents (in particular 424 terpenoids) that might affect the experience of the user (Gertsch et al., 2010; Russo, 2011). 425 426 Should such effects be substantiated by empirical evidence, it would be advisable to reconsider 427 the current laws and rules for formulations containing cannabis extracts, which are based solely 428 on THC. An improved understanding of terpenoid phytochemistry in cannabis would be an 429 important first step in this direction (Booth and Bohlmann, 2019).

430

431 Co-Regulation of Metabolic Pathways in Cannabis is Consistent with Gene Expression 432 Patterns Commonly Observed in Glandular Trichomes

433

Our statistical analyses using the WGCNA package indicated a tight correlation of biosynthetic 434 genes with cannabinoid and terpenoid end products (Fig. 5). We recently performed a meta-435 analysis of gene expression patterns in glandular trichomes across various species (Zager and 436 437 Lange, 2018). One of the conclusions, consistent with the data presented here, was that gene 438 expression patterns correlate well with the metabolic specialization in these anatomical structures. Co-regulation has been observed for genes across multiple pathways of specialized 439 metabolism, such as cannabinoids and terpenoids (this study), monoterpenes and diterpenes 440 (Salvia pomifera; Trikka et al., 2015), flavonoids and acyl sugars (Salpiglossis sinuata and 441 Solanum quitoense; Moghe et al., 2017), and bitter acids and prenylflavonoids (Humulus 442 lupulus; Kavalier et al., 2011; Clark et al. 2013). These tight gene-to-metabolite correlations 443 444 were also reflective of predicted fluxes through the relevant pathways (Zager and Lange, 2018). 445 In contrast, gene expression patterns appear to be less predictive of fluxes through central carbon metabolism, where regulation at the protein level plays a more significant role (Paul and 446 Pellny, 2003; Koch, 2004; Gibon et al., 2006; Rocca et al., 2014; Schwender et al., 2014). This 447 does not mean that feedback regulation of specialized metabolism is negligible in glandular 448 trichomes; there is just a particularly strong overall gene-to-metabolite correlation, and 449 unraveling the details will be an exciting topic for future research. 450

451

Functional Characterization of Terpene Synthases Contributes to an Improved Understanding the Genetic Determinants of Terpenoid Diversity

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The observed gene-to-metabolite correlations in cannabis glandular trichomes provided opportunities for gene discovery efforts. Booth et al. (2017) analyzed transcriptome data sets obtained with the 'Finola' and 'Purple Kush' strains to obtain candidate genes for terpene 458 synthases that were subsequently characterized to encode enzymes for the production of 14 459 mono- and sesquiterpenes. Those that contribute to the formation of some of the common 460 mono- and sesquiterpenes (e.g., β -myrcene, (-)-limonene, α -pinene, β -caryophyllene, and α -461 humulene) were found to be expressed at fairly high levels across the strains included in the present analysis, whereas those that generate less common products (e.g., (Z)- β -ocimene, γ -462 eudesmol, alloaromadendrene, δ -selinene, and valencene) were found to be expressed only in 463 464 a limited number of strains or not at all (Table 3). To assess sequence variation among these 465 genes, we cloned genes with high sequence identity to the previously characterized β -myrcene and (-)-limonene synthases. 466

467

Prior to the present study, a notable gap existed with regard to the terpene synthases 468 underlying the formation of the monoterpene linalool and the sesquiterpene nerolidol, which 469 470 are both common constituents in cannabis resin. We now identified a gene coding for an enzyme (CsTPS19BL) that generates a mixture of (+)-linalool and (-)-linalool from GPP and (E)-471 nerolidol from tFPP in the 'Black Lime' strain. We also cloned a putative ortholog from the 472 'Valley Fire' strain to evaluate the effects of sequence variation. Interestingly, the encoded 473 474 enzyme (CsTPS18VF) had the same specificity as CsTPS19BL with regard to the tFPP substrate ((E)-nerolidol as product); however, with GPP as substrate, (+)-linalool was detected as the 475 essentially exclusive product. This difference in specificity is surprising given that the peptide 476 sequences have only three mismatches (Supplemental Fig. S3). 477

478

Finally, we cloned genes that, based on sequence relatedness, were expected to code for 479 enzymes that generate sesquiterpene products not previously detected in assays with cannabis 480 481 terpene synthases. Indeed, CsTPS16CC was demonstrated to produce germacrene B and 482 CsTPS20CT formed hedycaryol as primary product. In assays with CsTPS16CC, y-elemene was also detected, but this is a well-known product of thermal degradation in the GC inlet (de 483 Kraker et al., 1998). Elemol was the sole product of assays with CsTPS20 CT, which is also a 484 thermal degradation product, in this case of hedycaryol (Koo and Gang, 2012; Hattan et al., 485 486 2016). Consequently, the enzyme activities are referred to as germacrene B synthase and 487 hedycaryol synthase, respectively. To the best of our knowledge, the sesquiterpenes generated 488 by these terpene synthases (germacrene B and hedycaryol) have not been identified in 489 cannabis samples yet, indicating the need for a more comprehensive coverage of terpenoids to better understand strain-specific aroma profiles. It should also be noted that several recent 490 491 studies reporting on comprehensive chemical and sensory analyses of volatiles emitted from cannabis found that non-terpenoid alcohols and aldehydes have potent odor impacts (Rice and 492 Koziel, 2015; Wiebelhaus et al., 2016; Calvi et al., 2018). These considerations indicate that 493 more emphasis needs to be placed on comprehensive metabolite profiling, including 494 cannabinoids and terpenoids but also extending to other volatiles, for future efforts focused on 495 496 strain characterization.

497

With a larger number of functionally characterized genes in cannabis, sequence 498 comparisons are now allowing us to ask questions about some of the determinants of 499 specificity. The overall sequence identity of the sesquiterpene synthases characterized here is 500 fairly low (< 70% at the amino acid level) but there are striking differences in the nature of a 501 conserved aromatic residue (Y527) that had previously been hypothesized to stabilize the 502 503 positive charge of the carbocation occurring during the formation of a germacrene 504 intermediate in the *epi*-aristolochene synthase catalytic sequence (Starks et al., 1997). The equivalent residues in sesquiterpene synthases that catalyze the formation of cyclic products 505 (CsTPS16CC and CsTPS20CT) are also tyrosines (Fig. 9). In contrast, glutamine residues occupy 506 this position in CsTPS18VF, CsTPS19BL and other characterized enzymes of the TPS-g clade 507 (Aharoni et al., 2004; Nagegowda et al., 2008) (Fig. 9A) which, possibly because of insufficient 508 carbocation stabilization, generate (E)-nerolidol as a non-cyclic product (Fig. 9). Testing this 509 510 hypothesis will be an important future goal for follow-up research.

511

512 MATERIALS AND METHODS

513

514 Plant Materials and Chemicals

515

Clonal plant cuttings of nine strains ('Sour Diesel', 'Canna Tsu', 'Black Lime', 'Valley Fire', 'White 516 517 Cookies', 'Mama Thai', 'Terple', 'Cherry Chem', and 'Blackberry Kush') were placed in 250 l pots 518 and grown in hoop-style, light-deprivation greenhouses (Shadowbox Farms, Williams, OR, USA) 519 under a 18-h light/6-h dark regime (natural light) to stimulate vegetative growth, before shifting 520 to a 12-h light/12-h dark cycle to induce flowering. The length of these time periods varied from strain to strain and was adjusted based on phenotypic evaluations. All aspects of plant growth, 521 harvest and transport were performed in accordance with the laws and rules under Chapter 522 475B, as released by the Oregon Liquor Control Commission. Plant harvest was performed 523 when the consistency of glandular trichome content had changed from a turbid white to clear, 524 525 and before another change to an amber-like color occurred. For most strains the pistils had 526 changed color from white to a yellow or orange. Buds were harvested, parts with low glandular trichome content removed using scissors, and the remainder placed on ice until further 527 processing (always within 3 h). Monoterpene and sesquiterpene reference standards were 528 purchased from Restek (Bellefort, PA, USA). Cannabinoid reference standards were obtained 529 from Sigma-Aldrich (St. Louis, MO, USA). Solvents for extraction were procured from Sigma-530 Aldrich (St. Louis, MO, USA). Solvents and chemicals for chromatography were sourced from 531 532 Burdick & Jackson (Morris, Plains, NJ, USA). Substrates for enzyme assays (GPP, NPP and E,E-533 FPP) were prepared synthetically (Davisson et al., 1986) or obtained from a commercial source (Z,Z-FPP, Echelon Biosciences, Salt Lake City, UT, USA). The sources of standards for enzyme 534 assays were as follows: germacrene B, isolated as a side product from assays with germacrene C 535 synthase (Colby et al., 1998)); y-elemene, obtained by heating germacrene B under argon (de 536 Kraker et al., 1998); elemol, institutional chemical repository (originally purchased from 537 Parchem, New Rochelle, NY, USA); hedycaryol, institutional chemical repository (source 538 539 unknown); (S)-(+)-linalool, isolated from coriander oil; (-)-limonene, (+)-limonene, (R)-(-)-540 linalool, β -myrcene, (E)-nerolidol, (-)- α -pinene, (-)- β -pinene, α -terpinolene, all purchased from Sigma-Aldrich, St. Louis, MO, USA). 541

542

543 Metabolite Extraction and Analysis

544

545 Cannabinoids and terpenoids were extracted and quantified according to Fischedick et al. (2010), with modifications, at a testing facility with accreditation by ISO/IEC 17025 and licensed 546 547 through the National Environmental Laboratory Accreditation Program (Evio Labs, Central Point, OR, USA). Briefly, roughly 2.0 g of fresh bud tissue was crushed in a falcon tube, 548 549 suspended in 10 ml methyl tert-butyl ether (containing 1-octanol as internal standard) with gentle shaking for 15 min, followed by centrifugation at 2,000 x q for 5 min. The supernatant 550 551 was transferred to a new vial and the plant material extracted two more times as above (no 552 addition of internal standard to solvent). The combined supernatants were filtered through a polytetrafluoroethylene syringe filter (0.45 µm pore size, 25 mm diameter) and an aliquot 553 transferred to a screw-cap glass vial, which was stored at -20°C until further analysis. Following 554 extraction, the remaining plant material was dried in an oven (50°C) and weighed to determine 555 556 dry weights for each sample.

557

Cannabinoids were separated via high performance liquid chromatography (model LC-2030C, 558 559 Shimadzu, Columbia, MD, USA) using a Kinetex C18 reversed phase column (50 x 4.6 mm, 2.6 μ m particle size; Phenomenex, Torrance, CA) and a binary gradient of solvent A (water 560 containing 0.1% (v/v) formic acid and 10 mM ammonium formate) and solvent B (methanol 561 containing 0.05% (v/v) formic acid) with the following settings: $0-9 \min$, 68-78% B; $9-11.9 \min$, 562 563 78–100% B; 11.9–13.5 min, hold at 100% B. Analytes were monitored at 228 nm in a diode array detector. Peak identification was achieved based on comparisons of retention times and 564 spectral characteristics with those of authentic cannabinoid reference standards. Analytes were 565 quantified based on calibration curves acquired with authentic standards. The validation of the 566 analytical method was performed according to Fischedick et al. (2010). 567

568

Terpenoids were separated via gas chromatography (model 6890, Agilent Technologies, Santa Clara, CA, USA) using a DB5 column (30 m x 25 mm, 25 μ m film thickness; Agilent Technologies, Santa Clara, CA, USA) and detected with a flame ionization detector. The conditions for separation were as follows: injector at 250°C, 20 : 1 split injection mode (1 μ L injected); detector at 250°C (H₂ flow at 30 ml/min, airflow at 400 ml/min, makeup flow (He) at 25 574 ml/min); oven heating from 40°C to 120°C at 2°C/min, then ramped to 200°C at 50°C/min, with 575 a final hold at 200°C for 2 min. GC peaks were identified based on comparisons of retention 576 times of authentic standards (purchased from Sigma-Aldrich, St. Louis, MO, USA). Analytes were 577 quantified based on calibration curves acquired with authentic standards. The validation of the 578 analytical method was performed according to Fischedick et al. (2010).

579

580 RNA Isolation from Glandular Trichomes and cDNA Library Preparation

581

Secretory cells of glandular trichomes were removed from 10–15 g of bud tissue by surface abrasion and then collected by filtering through a series of nylon meshes (Lange et al., 2000). Total RNA was isolated from secretory cells using the RNeasy Plant kit (Qiagen, Germantown, MD, USA) according to the manufacturer's instructions. RNA integrity was determined using a BioAnalyzer 2100 (Agilent Technologies, Santa Clara, CA, USA). cDNA libraries from 1–2 μg of total RNA were generated using the SuperScript III Reverse Transcriptase kit (Invitrogen, Carlsbad, CA, USA) according to the manufacturer's instructions.

589

590 **RNA Sequencing and Transcriptome Assembly**

591

592 RNA sequencing libraries were prepared from 250 ng total glandular trichome RNA with the 593 Stranded mRNA-Seq Poly(A) Selection kit (KAPA Biosystems, Wilmington, MA, USA). The quality 594 and quantity of the sequencing library was assessed using a Bioanalyzer 2100 and a Qubit 3.0 Fluorometer (Agilent Technologies; Life Technologies, Carlsbad, CA, USA). Sequencing of 150-bp 595 paired end reads was performed on a HiSeg 4000 instrument (Illumina, San Diego, CA, USA). 596 Sequenced reads were trimmed of adapter sequences with Trimmomatic (Bolger et al., 2014) 597 598 and sequence quality was checked with FastQC (Andrews, 2010). Trimmed sequences were 599 merged and assembled using the Trinity de novo assembler and downstream functional 600 annotation of the assembly was performed with Trinotate (Haas et al., 2013). The resulting 601 transcriptome assembly contained 514,208 contigs, with a mean contig length of 875 bp and an N50 value of 1529 bp. Transcript abundance in each RNA-seq data set (3 biological replicates 602 603 per strain) was determined with RSEM (Li and Dewey, 2011).

604

605 Analysis of Global Gene Expression Patterns and Gene Ontology Enrichment

606

607 Testing for differential gene expression across strains was performed using the Bioconductor 608 package DESeq2 (version 1.18.1) (Love et al., 2014). P-values were adjusted using the 609 Benjamini-Hochberg procedure (Benjamini and Hochberg, 1995). An adjusted P-value (false discovery rate) \leq 1.0e-10 and log₂ ratio \geq 3.0 were set as thresholds. A Cluster analysis of gene 610 expression patterns between strains was performed within the Trinity suite (Haas et al., 2013) 611 by partitioning genes into clusters by cutting the hierarchically clustered gene tree at 60% 612 613 height of the tree. A Gene Ontology (GO) enrichment analysis of differentially expressed genes was performed using the GOseq package in R (Young et al., 2010). GO terms with an adjusted P-614 value < 0.01 were considered significantly enriched. 615

616

617 Gene Co-Expression Network Analysis

618

A gene co-expression network was built using the WGCNA package in *R* (Langfelder and Horvath, 2008). Transcriptome datasets were filtered to remove genes with an average expression value of 50 TPM or smaller. Co-expression modules were identified using the function blockwiseModules with the following settings: power at 7, mergeCutHeight at 0.55, and minModuleSize at 30. Eigengene values were determined for each co-expression module to test for association significance. Modules with similar eigengene values were merged to obtain the final co-expression modules.

626

627 Phylogenetic Analysis of TPS Candidates

628

The identification of TPS candidate genes was accomplished by searching the translated transcriptome consensus assembly against a manually curated protein database specific to characterized plant TPSs using the Blastx algorithm. A reciprocal search (tBlastn) was performed with sequences of 114 characterized angiosperm TPSs against the assembly for each individual strain. Predicted TPS sequences were then analyzed for gene expression values across strains.
Translated amino acid sequences of these and reference TPSs (from *Cannabis sativa* and *Humulus lupulus*) were aligned using the MUSCLE algorithm. Alignments were analyzed with maximum likelihood analysis using a Jones-Taylor-Thorton model with Gamma distribution for rates among amino acid sites. One thousand bootstrap replicates were then used to construct a phylogeny using MEGA7 (Jones et al., 1992; Kumar et al., 2015).

639

640 Cloning of TPS cDNAs

641

642 First-strand cDNA was prepared from RNA with the SuperScript III First Strand Synthesis kit (Invitrogen) with random hexamer oligonucleotides. Open reading frames for TPSs were 643 amplified using gene-specific primers (Supplemental Table S5) (amplicons for full-length cDNAs 644 645 were generated for putative sesquiterpene synthases, whereas cDNAs devoid of the plastidial targeting sequence were amplified for putative monoterpene synthases). Amplicons were 646 ligated into the pGEM-T Easy vector (Promega, Fitchburg, WI, USA) and sequence-verified. For 647 expression in *E. coli*, full-length or truncated genes were subcloned into the pSBET expression 648 649 vector (predigested with Ndel and BamHI) (Steinbiss et al., 1995). Several terpene synthase cDNAs (CsTPS18VF, CsTPS19BL and CsTPS20CT) were purchased as synthetic products (in the 650 pET28B expression vector) from GenScript (Piscataway, NJ, USA). 651

652

653 In Vitro Functional Assays for Recombinant TPSs

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Plasmids were transformed into chemically competent cells of several *E. coli* strains (BL21 (DE3), C41 (DE3), C43 (DE3), C43 (DE3) pLysS and ArcticExpress (DE3)), which were then grown in 25 ml of liquid LB medium at 37°C with shaking to an OD₆₀₀ of 0.8. Expression of TPS genes was induced with 0.1 or 0.5 mM Isopropyl β -D-1-thiogalactopyranoside (Goldbio, St. Louis, MO, USA) and cells grown for another 24 h at three different temperatures (16°C, 10 °C and 4 °C). Bacterial cells were harvested by centrifugation at 5,000 x *g* and resuspended in 300 ul MOPSO buffer, pH 7.0, supplemented with 1 mM dithiothreitol (DTT; Goldbio, St. Louis, MO, USA). Cells 662 were lysed using a model 475 sonicator (VirTis, Gardiner, NY, USA), with three 15 s bursts and cooling on ice between bursts. The resulting homogenate was centrifuged at $15,000 \times q$ for 30 663 min at 4°C. and the clear supernatant mixed with ceramic hydroxyapatite (Biorad, Hercules, CA, 664 665 USA). The purification of recombinant protein was performed as described in Srividya et al. (2016) for constructs in the pSBET expression vector, whereas those in the pET28B expression 666 vector were purified over Ni²⁺ affinity columns according to the manufacturer's instructions 667 (Novagen-EMD Millipore, Burlington, MA, USA). In vitro assays were performed in 2-ml glass 668 vials containing 200 µg purified enzyme in MOPSO buffer containing DTT and MgCl₂ (total 669 volume 100 µl). A prenyl diphosphate substrate (GPP, NPP, tFPP or cFPP) was added to a final 670 671 concentration of 0.5 mM. The assay mixtures were overlaid with 100 µl n-hexane (Avantor. Center Valley, PA, USA) and incubated at 30°C for 16 h on a multi-tube rotator (Labguake, 672 Barnstead Thermolyne, Ramsey, MN, USA). The enzymatic reaction was stopped by vigorous 673 mixing of the contents of the tubes, followed by 30 min at -80°C for phase separation. The 674 organic phase was removed and transferred to glass vial inserts and stored in GC vials at -20°C 675 until further analysis. 676

677

Enzymatically formed products were analyzed on a 6890N gas chromatograph coupled to a 5973 mass selective detector (Agilent, Santa Clara, CA, USA). Analyte separation was achieved under the conditions developed by Adams (2007), which includes a comprehensive resource for spectral comparisons of volatiles. The chiral separation of monoterpenes was achieved as described in Turner et al. (2019). Enzymatically generated products were identified based on retention times and mass spectral properties when compared to those of authentic standards.

684

685 Statistical Analyses

686

For metabolite analyses, statistical analyses were performed in *R* using the MetaboAnalystR package (Chong and Xia, 2018). Quantitative terpenoid and cannabinoid data were scaled by dividing mean centered values by the standard deviation of each variable to generate principal component (PC) loadings. Principal components were then plotted in three dimensions within the *R* environment. OPLS-DA analysis was also performed in the same way using the MetaboAnalystR package. Differential gene 692 expression patterns were assessed using the Bioconductor package DESeg2 (version 1.18.1) (Love et al., 693 2014), with the P-value for the Benjamini-Hochberg false discovery threshold being adjusted to \leq 1.0e-694 10 and the \log_2 fold-change ratio to \geq 3.0. Cluster analysis of differential gene expression was performed 695 within the Trinity suite (Haas et al., 2013) by cutting the clustered gene tree at 60% tree height and 696 differentially expressed genes subjected to further analysis within GOSeq as described above (Young et 697 al., 2010). TPS candidates were identified based on sequence identity with functionally characterized 698 TPSs in tBLASTn searches. Candidates with e-values > 0.001 and bitscores < 250 were removed from further consideration. 699

700

701 Accession Numbers

702

The raw transcriptome sequence data for cannabis strains are available at the NCBI Sequence Read Archive, project number PRJNA498707. Nucleotide sequences for genes characterized as part of this study were deposited in GenBank and received the accession numbers MK131289 (Cs*TPS16CC*), MK801762 (Cs*TPS20CT*), MK801763 (Cs*TPS19BL*), MK801764 (Cs*TPS18VF*), MK801765 (Cs*TPS15CT*), and MK801766 (Cs*TPS14CT*).

708

Supplemental Data

Supplemental Figure S1. Alignment of translated peptide sequences, based on RNA-seq data, of tetrahydrocannabinolic acid synthase across cannabis strains.

Supplemental Figure S2. Nucleotide and translated peptide sequence, based on RNA-seq data, of cannabidiolic acid synthase from the cannabis strain 'Canna Tsu'.

Supplemental Figure S3. Alignment of terpene synthase sequences.

Supplemental Table S1. Statistics of *de novo* assemblies performed based on cannabis glandular trichome-specific RNA-seq data sets.

Supplemental Table S2. Annotation of transcripts represented in cannabis glandular trichomespecific RNA-seq data sets.

Supplemental Table S3. Clustering of genes into coexpression modules obtained by Weighted Gene Correlation Network Analysis of cannabis glandular trichome-specific RNA-seq data sets.

Supplemental Table S4. Accession numbers and sequences of terpene synthases considered for phylogenetic analysis.

Supplemental Table S5. Primers used to clone cannabis cDNAs for functional characterization.

709

710 **ACKNOWLEDGEMENTS**

This study was supported by gifts from private individuals and we are grateful for their generosity. We would also like to thank Shadowbox Farms for allowing A.S. to harvest plant materials.

- 714
- 715 **TABLES**
- 716

717 FIGURE LEGENDS

718

Figure 1. Shared origin of the cannabinoid and terpenoid biosynthetic pathways. A circled "P"
denotes phosphate moieties.

721

Figure 2. Characteristics of cannabis strains. A, Floral phenotype. B, Origin and aroma
 description (according to https://www.leafly.com).

724

Figure 3. Cannabis strain differentiation based on glandular trichome-specific RNA-seq data. A, Three-dimensional plot representing outcomes of a Principal Component Analysis. B, Heatmap of a two-way Hierarchical Clustering Analysis. The numerical values and red-white-blue color code indicate the log₂ fold-change compared to the average gene expression value across all strains. Abbreviations at the bottom of panel B: BB, 'Black Berry Kush'; BL; 'Black Lime'; CC, 'Cherry Chem'; CT, 'Canna Tsu'; MT, 'Mama Thai'; SD, 'Sour Diesel'; T, 'Terple'; VF, 'Valley Fire'; WC, 'White Cookies'.

732

Figure 4. Cannabis strain differentiation based on cannabinoid and terpenoid profiles. A, Three dimensional plot representing outcomes of a Principal Component Analysis. B, Two-dimensional
 plot of the outcomes of an Orthogonal Projections to Latent Structures Discriminant Analysis.

736

Figure 5. Co-expression of genes involved in cannabinoid and terpenoid biosynthesis. A, 737 Weighted Gene Correlation Network Analysis (WGCNA) of glandular trichome-specific RNA-seq 738 data categorizes transcripts into eight color-coded modules (for gene lists see Supplemental 739 740 Table S3). B, Correlation of WGCAN modules with metabolites. A color code is used to visualize 741 the Spearman Correlation Coefficients (SCCs) for each module-metabolite pair, with red color representing positive and blue color indicating negative SCCs. C, Genes involved in cannabinoid 742 and terpenoid biosynthesis are enriched in the yellow co-expression module obtained by 743 744 WGCNA. Color code for pathways: light blue, hexanoate formation; dark green, precursors for monoterpenes; light green, monoterpene synthases; orange, sesquiterpenes; dark blue, 745

746 cannabinoids; cyan, remaining genes. D, Functional context of genes highlighted in (C) in 747 simplified metabolic pathway scheme. Abbreviations: ACC1, acetyl-CoA carboxylase; Ac-CoA, 748 acetyl coenzyme A; AAE1, acyl activating enzyme for short-chain fatty acids: 749 CsTPS1FN/CsTPS14CT, (-)-limonene synthase; CsTPS2SK. $(+)-\alpha$ -pinene svnthase: CsTPS3FN/CsTPS15CT, β-myrcene synthase; CsTPS16CC, germacrene B synthase; PT1, 750 cannabigerolic acid synthase; DHAP, dihydroxyacetone phosphate; DXS, 1-deoxy-D-xylulose-5-751 phosphate synthase; ENO, enolase; FNR-Root, ferredoxin-NADP+ reductase (isoform of roots 752 and glandular trichomes); FPPS, farnesyl diphosphate synthase; GAP, glyceraldehyde-3-753 phosphate; GAPDH, glyceraldehyde-3-phosphate dehydrogenase; GPP, geranyl diphosphate; 754 755 GPPS, geranyl diphosphate synthase; KR, β -ketoacyl reductase (fatty acid synthase complex); OA, olivetolic acid; PDH, pyruvate dehydrogenase; PFK, phosphofructokinase; PGI, 756 757 phosphoglucoisomerase; *PGM*, phosphoglucomutase; *PK*, pyruvate kinase; *Pyr*, pyruvate; 758 THCA, tetrahydrocannabinolic acid; THCAS, tetrahydrocannabinolic acid synthase; and TPI, 759 triose phosphate isomerase.

760

Figure 6. Maximum likelihood phylogenetic tree of selected, functionally characterized terpene 761 762 synthases. The tree is rooted with the ancestral *ent*-kaurene synthase of *Physcomitrella patens* 763 (PpCPS/KS). A color code is used to indicate different clades (vellow, TPS-a; green, TPS-b; and purple, TPS-g). Abbreviations: BL, 'Black Lime'; CC, 'Cherry Chem'; CT, 'Canna Tsu'; Cs, Cannabis 764 sativa; FN, 'Finola'; FRAAN, Fragaria x ananassa; FRAVE, Fragaria vesca; HUMLU, Humulus 765 lupulus; OCIBA, Ocimum basilicum; ROSRU, Rosa rugosa; SALOF, Salvia officinalis; and VF, 766 'Valley Fire'; and VITVI, Vitis vinifera. The accession numbers and sequences of the terpene 767 synthases are provided in Supplemental Table S4. 768

769

Figure 7. Functional characterization of cannabis terpene synthases that act on GPP as substrate. A to D, Left panel: chiral GC chromatogram. Center panel: mass spectra of primary products. Right panel: product distribution. A, (-)-limonene synthase (CsTPS14CT). B, β-myrcene synthase (CsTPS15CT). C, (E)-nerolidol/(+)-linalool synthase (CsTPS18VF). D, (E)-nerolidol/(+)linalool synthase (CsTPS19BL). 775

Figure 8. Functional characterization of cannabis terpene synthases that act on tFPP as
substrate. A to D, Left panel, GC-MS chromatogram. Center panel, mass spectra of primary
products. Right panel, product distribution. A, (E)-nerolidol/(+)-linalool synthase (CsTPS18VF).
B, (E)-nerolidol/(+)-linalool synthase (CsTPS19BL). C, germacrene B synthase (CsTPS16CC). D,
hedycaryol synthase (CsTPS20CT).

781

Figure 9. Variation of the residue putatively stabilizing carbocation intermediates correlates with outcome of catalysis in cannabis sesquiterpene synthases. A, Sequence alignment of sesquiterpene synthases (with carbocation-stabilizing residue highlighted). B, Proposed cyclization reactions catalyzed by sesquiterpene synthases. Identifiers for sequences from the literature (Aharoni et al., 2004; Nagegowda et al., 2008): AmNES/LIS1, EF433761; AmNES/LIS2, EF433762; FvNES1, AX529002; FaNES2, AX529067; FaNES1, KX450224 (species abbreviations: Am, Antirrhinum majus; Fa, Fragaria x ananassa; Fv, Fragaria vesca).

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