### SUPPORTING INFORMATION FOR

## Diversification, disparification, and hybridization in the desert shrubs Encelia

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## Methods S1

#### Sampling: Common Garden

Collected seeds were planted in a common garden for use in all genetic and phenotypic analyses. The common garden was first planted in 2011 and then expanded in 2016. By using a common garden to collect both genetic and phenotypic data, we ensured that our phenotypic data were matched to a genotype and that phenotypic differences among individuals were not due to environmental variation. Seeds were germinated in greenhouse trays using an equal mix of pumice, sand, and vermiculite. *Encelia ravenii* and *E. resinifera* were not included in the garden because their seeds failed to germinate. When seedlings were 5 cm tall or had four true leaves, they were transplanted into the common garden at the University of California Agricultural Operations Station in Riverside, CA. Seedlings were randomized across rows and then irrigated weekly. Plants were phenotyped from 2015 to 2016, and in October 2016, adult leaves were collected in silica gel for use in genetic analysis. For phenotypic measurements of *E. ravenii* and *E. resinifera*, we used field-collected adult leaves.

#### Genetic data collection

Because ddRAD data have not yet been collected for *Encelia*, we first conducted a series of *in silico* experiments to determine the best restriction enzymes and target size to use. The closest available reference genome to *Encelia* is *Helianthus annuus*, another species of the family Asteraceae estimated to be 7.5 million years diverged from *Encelia* (Smith & Brown, 2018). Many of the species in Asteraceae, including *Encelia*, have extensive repetitive content (Badouin *et al.*, 2017). To avoid sequencing this repeat content, we focused on testing restriction enzymes that are sensitive to the DNA methylation common in repeat content (Pegadaraju *et al.*, 2013). In total, we tested 14 different restriction enzymes in all possible

pairwise combinations. Ultimately, to give the desired yield of 20,000 loci, we used *Pstl* and *Mspl*, size-selecting between 250 - 700 bp.

To collect ddRAD data, we first extracted DNA from silica-dried adult leaves using a QIAGEN DNeasy Plant Kit (cat. no.: 69104) following manufacturer's instructions. We determined DNA quality by running extracts on a 2% agarose gel and DNA concentration by using a NanoDrop DNA spectrometer. The commercial provider RTL Genomics (Lubbock, TX, USA) then prepared doubly-barcoded ddRAD libraries (Peterson *et al.*, 2012). Equimolar amounts of samples were combined to generate pools of 100 individuals; each pool was sequenced across one lane of 100 PE sequencing on the Illumina HiSeq 4000 Sequencing Platform.

Per individual, we first cleaned all reads, trimming adaptors and low quality sequence using Trimmomatic v36 (Bolger *et al.*, 2014). We then merged any overlapping reads using PEAR v0.9.8 (Zhang *et al.*, 2014). Due to extensive adaptor contamination, many of our paired reads became orphaned through the cleaning process. We then assembled cleaned and merged reads using Velvet v1.2.10 (Zerbino & Birney, 2008) across multiple k-mer values. Although Velvet is not often used in ddRAD assembly, other commonly used tools like Rainbow or STACKS are not designed to assemble mixed sets of single-end and paired-end reads. We then retained all assembled loci with coverage ≥2×, resulting in an average of 140K loci per individual.

### **Phylogenetic Inference**

Although all sampled individuals were identified to nominal species, species boundaries have not been well-tested in *Encelia*. Thus, our first set of analyses were designed to determine lineage identity per individual. We used VSEARCH v2.9.1 (Rognes *et al.*, 2016) to identify homologous loci using a 70% identity cutoff. After aligning homologous loci using mafft v7.9.34 (Katoh *et al.*, 2009), we then concatenated homologous loci across individuals, removing any individual sequenced for <5% of loci or any loci sequenced for <30% of individuals. We used RAxML v8.2.11 to generate a phylogeny from the 11.6K locus, 1.5 Mb concatenated alignment (Stamatakis, 2014). By comparing clade identity to nominal species identity, we determined likely lineage identities for each individual (Table S1). We then inferred individual-level phylogenies using both concatenated and coalescent-based approaches. First, we used the lineage-level pseudo-reference genomes and the variant call set to determine loci sequences per individual. We removed any sequences with >30% of missing sites and any loci sampled for <60% of individuals. We then used RAxML to infer a phylogeny and 100 bootstraps from the 31K locus, 3.9 Mb concatenated alignment. For a coalescent-based approach, we used SVDquartets (Chifman & Kubatko, 2014). We randomly sampled one SNP per locus, retaining SNPs with <20% missing and  $\geq$ 2 minor allele count. We used the resulting set of 2.8K SNPs to infer the phylogeny and to conduct 100 bootstrap analyses. Additionally, we calculated gene concordance factors (gCF) and site concordance factors (sCF) using IQ-TREE v1.6.4 (Minh *et al.*, 2020), to assess conflict across loci and sites.

Then, we used two different coalescent-based approaches - one based on gene trees and the other based on SNPs - to infer a lineage-level phylogeny. We used ASTRAL-III (Zhang et al., 2018), which, when given an individual-to-lineage mapping and a set of gene trees inferred across individuals, can generate a lineage-level phylogeny. Here, we filtered loci to include only loci inferred for >60% of individuals, inferred gene trees using RAxML, collapsed all randomly-bifurcated nodes using 'dimulti' in the R package ape (Paradis et al., 2004), and then inferred the species tree from these gene trees with ASTRAL-III. Second, using the same SNP dataset used for the individual-level SVDquartets analysis, we additionally inferred a lineage-level phylogeny. Neither ASTRAL-III nor SVDguartets provides terminal branch lengths. Accordingly, to infer an ultrametric phylogeny for use in comparative analyses, we generated a lineage-level concatenated alignment by randomly selecting one sequence per locus per lineage. We then used RAXML to estimate branch lengths, constraining the topology to each of the lineage-level phylogenies. To infer an ultrametric tree for this group, we used a calibration from a comprehensive angiosperm phylogeny that estimated the crown age of Encelia as 1.36 million years (Myr) (Magallón et al., 2015; Smith & Brown, 2018). This aligns with previous divergence dating based on population genomic data that inferred the root of Encelia as 1.05 Myr (Singhal, unpublished). We used this root age to infer a chronogram using the 'chronos' function in the R package 'ape' under a strict clock model with a lambda of 0. We selected our model and lambda value by comparing  $\Phi IC$  across all possible models and a range of lambda values from 1e-6 to 0.1 (Fig. S1; Paradis, 2013). We used this time-calibrated phylogeny in all comparative analyses.

## **MicroCT** imaging

High-resolution, three-dimensional (3D) images of stem and leaf structure were obtained by performing hard X-ray microcomputed tomography (microCT) at the Advanced Light Source, Lawrence Berkeley National Laboratory (LBNL), Beamline 8.3.2 (Brodersen, 2013; Brodersen & Roddy, 2016). These images were used to visualize fine-scale morphology and anatomy of these leaves and to calculate trichome density and stem xylem vessel diameter and area. Leaf samples were collected from plants growing in cultivation at the Agricultural Operations Station, University of California, Riverside in June 2015 and September 2016 and transported to LBNL, keeping the leaves hydrated until imaging. Samples were placed on a rotating stage in the 24 keV synchrotron X-ray beam and continuously rotated from 0 to 180 degrees. As the sample rotated continuously through the beam, 1025 two-dimensional projections were recorded using a 10x objective, yielding a final pixel resolution of 1.25  $\mu$ m. Each scan was completed in ~15 min. These raw tomographic projections were reconstructed using TomoPy, a Python-based framework for reconstructing tomographic data (Gürsoy et al., 2014). Images were reconstructed using gridrec (Dowd et al., 1999) and phase retrieval (Davis et al., 1995) reconstruction methods. Image stacks were rotated, aligned, and traits measured in FIJI (Schindelin et al., 2012). Three-dimensional stack visualization was performed in Avizo v. 9.2.0 (ThermoFisher Scientific).

### Trait data collection & analysis

### Leaf area, shape, and color

To analyze leaf area, shape, and color, we collected three to five adult leaves per individual growing in the common garden. Using a digital camera, we photographed these leaves with a black-and-white reference scale bar. Leaves were then oven-dried at 70 C for at least 72 hours and weighed for dry mass using a Sartorius Practum (resolution of 0.0001 mg, Sartorius AG, Goettingen, Germany). We analyzed leaf images using ImageJ (Abràmoff *et al.*, 2004), first calibrating the white balance and size of each image using the scale bar as a reference (Mascalchi, 2017). For each leaf, we set the threshold across the leaf, excluding the petiole, and then collected data on area, perimeter, fit ellipse, and shape using the Analyze Particles function. To measure color, we used the white-balanced leaf images in Adobe Photoshop and

measured the arithmetic mean of all pixels of the largest circumscribed rectangle possible within the center of the leaf. We calculated leaf mass per area (LMA) using measurements of leaf area and dry mass. We conducted a principal component analysis of all leaf size and shape measurements after centering and scaling them. The first two axes summarize 42% and 24% of the variation respectively; PC1 largely loads on size-related measurements and PC2 largely reflects the roundness of the leaf.

## Canopy ramification

We estimated the degree of canopy ramification as the number of terminal branch tips per stem cross-sectional area (BTSA; Roddy *et al.*, 2019). For each plant in the common garden, we took two perpendicular measurements of the diameter of the stem base using manual calipers and averaged these two measurements to estimate the area of the stem base. We then counted all branch tips descendant from this base, regardless of their size.

## Trichome density

Trichomes were characterized from 3D microCT images of fresh leaves. Subregions of each leaf surface were cropped and rotated to align the leaf surface with the plane of view. Digital slices parallel to the leaf surface were taken through the trichomes, allowing trichomes to be counted per unit projected leaf surface area.

## Vessel dimensions

Stem xylem vessel dimensions were measured from 2D cross sections of microCT image stacks obtained from dried stems. Images were thresholded and binarized to distinguish vessel lumens from plant tissue, and the area of each of these lumens measured using the Analyze Particles function in ImageJ.

## Wood density

Wood density was measured using Archimedes' principle. A small segment of stem was excised, its bark removed, and its volume measured as the mass of water displaced when it was submerged in a beaker of water sitting on a balance. The segment was then oven-dried at 70 C for at least one week, after which its dry mass was measured.

#### Stem hydraulic conductance

Whole-shoot hydraulic conductance was measured using a low pressure flow meter (Kolb *et al.*, 1996), which enables measuring the entire shoot regardless of branch ramification and can be applied to morphologically diverse structures (Roddy *et al.*, 2016, 2019). Mature plants growing under well-watered conditions at the Agricultural Operations Station, University of California, Riverside, were sampled in May 2013. Healthy shoots were selected from each plant.

In the late evening, entire shoots were cut at their base, and immediately recut underwater while the shoot was enclosed in black plastic bags and transported to the lab. Shoots were allowed to rehydrate for at least two hours before hydraulic measurements, and measurements were completed within 24 hours of sampling. Branching shoots were defoliated underwater, the leaves scanned for leaf area, and the cut shoot base was recut under clean water with a fresh razor blade. While keeping the cut stem surface covered in water, the shoot base was guickly enclosed in a compression fitting that was connected to a hard-sided tube, which was filled with and plumbed back to a bottle of 0.01 M KCl that sat on a microbalance with resolution to 0.0001 g (Sartorius CPA225D, Sartorius AG, Goettingen, Germany). The shoot was enclosed in an acrylic vacuum chamber. The vacuum chamber was connected to a pressure gauge and a vacuum pump, which created a series of partial vacuums (60, 50, 40, 30, 20, 15 kPa below ambient) on the stem that pulled KCI from the balance and into the stem. Every fifteen seconds the mass of KCI on the balance was logged by a computer interfaced with the balance (WinWedge, TAL Technologies Inc., Philadelphia, PA, USA). The partial vacuum was maintained for at least three minutes at each pressure and until the coefficient of variation of the previous ten measurements was below 0.05. The average of these last ten measurements was taken as the flow rate at the given pressure. Hydraulic conductance was calculated as the slope of the regression of flow rate versus pressure, and no more than one of the six measurements for each shoot was removed to improve linearity. Because shoots differed in size and ramification, hydraulic conductance was normalized by leaf area of the shoot, which is taken as a metric of hydraulic efficiency (Roddy et al., 2019).

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# Supplementary Tables

**Table S1:** Information on the 72 individuals for which we collected genetic data, including their species identification, the locality from which they were collected, and whether they are in our common garden. For those nominal species that contain multiple putative lineages, we indicate lineage identity in parentheses. This table also outlines the efficacy of our genetic data collection, including trimmed sequencing read yield in megabases (Mb), number of assembled homologous loci, and number of high-quality, high coverage variant and invariant sites.

sample	species	locality	In common garden	read yield (Mb)	loci count (K)	sites (Mb)	SRA
ACT-9M-1	Encelia actoni	9 Mile Canyon Road, CA, USA	*	2055.59	68.9	12.6	SAMN17310799
ACT-9M-2	E. actoni	9 Mile Canyon Road, CA, USA	*	223.83	37.8	5.3	SAMN17310800
ACT-AB-1	E. actoni	Anza Borrego, CA, USA	*	176.34	33.3	4.8	SAMN17310801
ACT-AB-2	E. actoni	Anza Borrego, CA, USA	*	102.6	22.6	3.2	SAMN17310802
ASP-CD-1	E. asperifolia	Central Desert, BC, Mexico	*	72.55	19.2	2.3	SAMN17310803
ASP-CD-2	E. asperifolia	Central Desert, BC, Mexico	*	308.75	38	5.5	SAMN17310804
ASP0129	E. asperifolia	10 km south of El Rosarito, BC, Mexico		87.05	20.9	2.5	SAMN17310805
CAL-CR-1	E. californica (1)	north of El Rosario, BC, Mexico	*	79.68	17.5	2.2	SAMN17310806
CAL-CR-2	E. californica (1)	north of El Rosario, BC, Mexico	*	64.54	13.3	1.7	SAMN17310807
CAL-SC-1	E. californica (1)	San Clemente, CA, USA	*	177.14	24	3.4	SAMN17310808
CAL-SC-2	E. californica (1)	San Clemente, CA, USA	*	214.8	27.7	4.3	SAMN17310809
CAL-ELR-1	E. californica (2)	pass south of El Rosario, BC, Mexico	*	72.35	14	1.9	SAMN17310810
CAL-ELR-2	E. californica (2)	pass south of El Rosario, BC, Mexico	*	78.58	17.6	2.2	SAMN17310811
CAN-COP-5	E. canescens	Atacama Desert, Chile		384.14	40.6	6.5	SAMN17310812
CAN-DOM-1	E. canescens	Atacama Desert, Chile		483.94	13.4	1.1	SAMN17310813
CAN-HUA-4	E. canescens	Atacama Desert, Chile		494.51	42.5	7.3	SAMN17310814
CAN-HUA-6	E. canescens	Atacama Desert, Chile		325.6	27.9	4.8	SAMN17310815

CAN-ING-1	E. canescens	Atacama Desert, Chile		122.41	14.3	2.3	SAMN17310816
CAN-OBI-7	E. canescens	Atacama Desert, Chile		330.76	32	5.3	SAMN17310817
CAN-OBI-9	E. canescens	Atacama Desert, Chile		85.97	18.3	2.3	SAMN17310818
CAN-PERU-2	E. canescens	Arequipa 2, Peru	*	4091.65	49.9	11.8	SAMN17310819
CAN-PERU-3	E. canescens	Mollebaya, Peru	*	1162.26	67.1	9	SAMN17310820
CAN-VAL-9	E. canescens	Atacama Desert, Chile		769.06	62.6	8	SAMN17310821
DEN-12-A	E. densifolia	Sierra Santa Clara, BCS, Mexico	*	513.66	55.4	6.7	SAMN17310822
DEN-13-A	E. densifolia	Sierra Santa Clara, BCS, Mexico	*	323.22	52	6.2	SAMN17310823
DEN-16	E. densifolia	Sierra Santa Clara, BCS, Mexico	*	966.03	77.7	10.2	SAMN17310824
DEN-17	E. densifolia	Sierra Santa Clara, BCS, Mexico	*	279.31	50.3	6	SAMN17310825
DEN-18-A	E. densifolia	Sierra Santa Clara, BCS, Mexico	*	365.55	53	6.5	SAMN17310826
DEN-21	E. densifolia	Sierra Santa Clara, BCS, Mexico	*	42.47	16.4	1.4	SAMN17310827
FAR-FAR-AB-1	E. farinosa farinosa	Anza Borrego, CA, USA	*	1186.43	90.4	12	SAMN17310828
FAR-FAR-AB-2	E. farinosa farinosa	Anza Borrego, CA, USA	*	1905.66	96.8	13.3	SAMN17310829
FAR-FAR-DVHZ- 1	E. farinosa farinosa	Death Valley, CA, USA	*	815.27	79.3	10	SAMN17310830
FAR-FAR-DVHZ- 2	E. farinosa farinosa	Death Valley, CA, USA	*	1381.38	87.7	11.8	SAMN17310831
FAR-FAR-DVSC- 1	E. farinosa farinosa	Surprise Canyon, Death Valley, CA, USA	*	514.03	70.9	8.9	SAMN17310832
FAR-FAR-HH-1	E. farinosa farinosa	Hidden Hills, Mojave National Preserve, CA, USA	*	419.46	60.8	7.1	SAMN17310833
FAR-FAR-HH-2	E. farinosa farinosa	Hidden Hills, Mojave National Preserve, CA, USA	*	1324.88	88.7	11.9	SAMN17310834
FAR-FAR-RV-1	E. farinosa farinosa	Riverside, CA, USA	*	356.7	65	7.9	SAMN17310835
FAR-FAR-RV-2	E. farinosa farinosa	Riverside, CA, USA	*	1258.34	88	11.6	SAMN17310836
FAR-PHE-52	E. farinosa phenicodonta	Central Desert, BC, Mexico	*	675.51	77.5	9.9	SAMN17310837
FAR-PHE-60-2	E. farinosa phenicodonta	Sea of Cortez, BC, Mexico	*	892.79	77.3	9.9	SAMN17310838
FAR-PHE-71-1	E. farinosa phenicodonta	north of San Felipe, BC, Mexico	*	865.67	84.5	11.1	SAMN17310839
FAR-PHE-73-1	E. farinosa phenicodonta	north of San Felipe, BC, Mexico	*	696.14	72	9.2	SAMN17310840

FAR-SON	E. farinosa phenicodonta	Sonora, Mexico	*	548.82	70.4	8.6	SAMN17310841
FRU-FRU-DV-1	E. frutescens frutescens	Death Valley, CA, USA	*	416.43	61.4	7.5	SAMN17310842
FRU-FRU-DV-2	E. frutescens frutescens	Death Valley, CA, USA	*	456.33	59.7	7.1	SAMN17310843
FRU-FRU-DVHZ- 1	E. frutescens frutescens	Death Valley, CA, USA	*	1019.08	76.2	9.7	SAMN17310844
FRU-FRU-DVHZ- 2	E. frutescens frutescens	Death Valley, CA, USA	*	683.58	70.2	8.8	SAMN17310845
FRU-FRU-MNP-1	E. frutescens frutescens	Mojave National Preserve, CA, USA	*	814.49	77.8	9.8	SAMN17310846
FRU-FRU-MNP-2	E. frutescens frutescens	Mojave National Preserve, CA, USA	*	1115.06	85.7	11	SAMN17310847
FRUGLA1	E. frutescens glandulosa	25 km north of Condensadora, BC, Mexico		1570.74	94.1	12.6	SAMN17310848
FRUGLA2	E. frutescens glandulosa	25 km north of Condensadora, BC, Mexico		462.31	69.3	8.2	SAMN17310849
PAL-BA-1	E. palmeri	Bahia Asuncion, BCS, Mexico	*	59.69	20.4	1.8	SAMN17310850
PAL-BA-2	E. palmeri	Bahia Asuncion, BCS, Mexico	*	1305.15	90.3	12.1	SAMN17310851
PAL-JESUS-01	E. palmeri	Villa Jesus Maria, BC, Mexico		903.89	79.7	10.4	SAMN17310852
PAL648	E. palmeri	El Marasal, BC, Mexico		1143.91	87.8	11.7	SAMN17310853
RAV-1	E. ravenii	San Felipe Desert, BC, Mexico		1763.21	85.8	11.5	SAMN17310854
RAV-10	E. ravenii	San Felipe Desert, BC, Mexico		1971.25	92.5	12.6	SAMN17310855
RAV-11	E. ravenii	San Felipe Desert, BC, Mexico		1224.63	72.4	9.9	SAMN17310856
RAV-13	E. ravenii	San Felipe Desert, BC, Mexico		1577.5	87.2	11.7	SAMN17310857
RAV-2	E. ravenii	San Felipe Desert, BC, Mexico		1231.7	84	11.1	SAMN17310858
RES-1	E. resinifera	Zion National Park, Washington Co., UT, USA		919.69	92	12.1	SAMN17310859
RES-2	E. resinifera	Zion National Park, Washington Co., UT, USA		1322.5	98.6	13.1	SAMN17310860
VEN-BA-1	E. ventorum	Bahia Asuncion, BCS, Mexico	*	1452.92	89.5	11.7	SAMN17310861
VEN-BA-2	E. ventorum	Bahia Asuncion, BCS, Mexico	*	2350.32	97.5	13.3	SAMN17310862
VIR-GMDRC	E. virginensis (1)	Granite Mountains Desert Research Center, CA, USA	*	1575.73	96.1	12.9	SAMN17310863
VIR-MNP-1	E. virginensis (1)	Foshay Pass, Mojave National Preserve, CA, USA	*	871.66	84.5	11.3	SAMN17310864
VIR-MNP-2	E. virginensis	Foshay Pass, Mojave National	*	1792.58	98.6	13.7	SAMN17310865

	(1)	Preserve, CA, USA					
VIR-UT-1	E. virginensis (2)	Highway 91, Washington Co., UT, USA		1565.36	108.4	14.6	SAMN17310866
VIR-UT-2	E. virginensis (2)	Highway 91, Washington Co., UT, USA		637.33	71.7	9.2	SAMN17310867
ENC-1	Enceliopsis covillei	Surprise Canyon, Death Valley, CA, USA	*	1066.09	48.7	6	SAMN17310868
ENC-2	Enceliopsis covillei	Surprise Canyon, Death Valley, CA, USA	*	1378.21	49	6.3	SAMN17310869
XYL	Xylorhiza tortifolia	Mojave National Preserve, CA, USA	*	2756.57	10.3	0.9	SAMN17310870

nominal species	BTSA	stem hydraulic conductance	leaf area	leaf color	LMA	leaf roundness	trichome density	vessel diameter	wood density	phylogenetic sampling
Encelia actoni	16	6	6	6	6	6	2	1	0	4
E. asperifolia	7	8	3	3	3	3	1	1	8	3
E. californica	26	8	7	7	7	7	1	1	0	6
E. canescens	11	3	3	3	3	3	1	1	0	10
E. densifolia	11	0	3	3	3	3	1	1	8	6
E. farinosa farinosa	31	6	22	22	22	22	4	1	8	7
E. farinosa phenicodonta	18	0	36	36	36	36	3	1	0	5
E. frutescens frutescens	27	6	15	15	15	15	1	1	0	6
E. palmeri	7	6	3	3	3	3	1	1	7	4
E. ravenii	0	0	0	0	0	0	2	0	0	5
E. resinifera	0	0	0	0	0	0	1	0	0	2
E. ventorum	7	4	6	6	6	6	2	1	8	2
E. virginensis	10	0	6	6	6	6	2	0	0	5
Enceliopsis covillei	2	0	0	0	0	0	2	1	0	2
Xylorhiza tortifolia	6	0	0	0	0	0	1	0	0	1

**Table S2:** Sample sizes for morphological traits and genetic sampling for *Encelia* nominalspecies and outgroups. Abbreviations for morphological traits follow Table S3.

**Table S3:** The nine morphological and physiological traits measured in this study.

trait	description	# of lineages measured
canopy ramification	measured as the number of branch tips per stem cross-sectional area (BTSA)	15
trichome density	top and bottom trichome density was measured on microCT leaf images; top and bottom trichome density were highly correlated ( <i>r</i> = 0.99, p-val < 5e-14)	15
stem hydraulic conductance	slope of flow rate versus pressure for stems, normalized by leaf area of shoot	8
leaf color	leaf color was calculated on color-balanced leaves as the arithmetic mean of pixels	12
leaf roundness	leaf roundness is highly correlated with leaf PC2 ( $r =$ -0.98) which captures 24% of the variation	12
leaf area	leaf area is highly correlated with leaf PC1 ( $r = -0.97$ , p-val < 2e-16) which captures 42.5% of the variation	12
leaf mass area (LMA)	leaf dry mass divided by leaf area	12
wood density	woody stem (stripped of bark) dry mass divided by its volume	5
vessel diameter	stem xylem vessel diameter	11

**Table S4:** Results of significant D-statistic tests. D-statistics were calculated using the allele frequency approach, with all samples assigned to a lineage included in the calculation. In all tests, *Enceliopsis covillei* was used as the outgroup species. Shown are D-statistic analyses that significantly deviate from 0, as measured by their Z-score. The two species inferred to have exchanged genes, as determined by the sign of the D-statistic, are listed in the first two columns. The third species in the triad is listed in the third column.

species 1	species 2	Species 3	D-stat	Z-score	p-val
E. actoni	E. virginensis 1	E. virginensis 2	0.19	34.66	3.12E-263
E. actoni	E. frutescens frutescens	E. frutescens glandulosa	0.06	8.52	1.62E-17
E. virginensis 1	E. frutescens frutescens	E. actoni	0.04	6.5	7.96E-11
E. californica 1	E. californica 2	E. asperifolia	0.1	5.13	2.86E-07
E. palmeri	E. farinosa farinosa	E. californica 1	0.11	7.07	1.55E-12
E. palmeri	E. farinosa phenicodonta	E. californica 1	0.03	6.17	6.79E-10
E. palmeri	E. ravenii	E. californica 2	0.03	4.62	3.83E-06
E. ravenii	E. frutescens frutescens	E. actoni	0.03	5.44	5.40E-08
E. frutescens	E virginensis 2	E actoni	0.04	6 97	3.06E-12
frutescens	E. Virginensis E	E. dolom	0.04	0.07	0.002 12
E. frutescens	E. virainensis 2	E. actoni	0.07	10.73	7.73E-27
glandulosa	gonoio 2	<u> </u>	0.07		1.1.92 21

**Table S5:** Global correlations between traits and the environment as inferred from a phylogenetic canonical correlation analysis. This approach requires complete data; accordingly, we selected those variables and species for which we have the most complete data. For species, we included all in-group species (n = 11) but *Encelia resinifera* and *E. ravenii*. For environmental data, we included all four bioclimatic variables measuring extreme climates: maximum temperature of the warmest month (BIO5), minimum temperature of the coldest month (BIO6), precipitation of wettest month (BIO13), and precipitation of driest month (BIO14). For morphological data, we included six of our nine measured traits: leaf area, leaf roundness, canopy ramification, leaf color, trichome density, and leaf mass area. All variables were scaled and centered prior to analysis. We recovered no significant correlations between trait and environmental variables.

dimension	correlation	$\chi^2$	p-value
1	0.98	29.6	0.19
2	0.96	14.7	0.48
3	0.58	2.24	0.97
4	0.26	0.33	0.95

**Table S6:** Correlations between traits and the environment. Shown are the expected correlations between trait and the mean of the environmental variable (temperature, precipitation) and citations to studies that guided this expectation. For some comparisons, we are unaware of a citation from the literature to support our prediction; we mark the 'citation' column for these comparisons with an asterisk ('\*'). We report the actual correlation to extreme values of each environmental variable and p-value from PGLS. Trait abbreviations follow Table 1. For leaf color, higher values indicate lighter color. P-values have not been corrected for multiple-testing. Although many trait-environment correlations trend in the direction expected, almost all correlations are weak and only one correlation (LMA vs. warmest month; in bold) is significant.

trait	environmental variable	predicted correlation	citation	extreme	correlation	p-value
	temperature	+	(Schenk <i>et al.</i> ,	warmest month	0.17	0.64
BTSA	tomporataro		2008)	coldest month	0.09	0.8
Dion	precipitation	_	(Schenk <i>et al.</i> ,	wettest month	-0.42	0.23
	procipitation		2008)	driest month	0.06	0.87
	temperature	_	(Lambers <i>et al.</i> ,	warmest month	-0.06	0.88
leaf area			2008)	coldest month	0.28	0.44
	precipitation	C	(Moles <i>et al.</i> ,	wettest month	-0.09	0.8
	precipitation		2014)	driest month	0.32	0.37
	leaf color	warmest month	-0.17	0.63		
leaf color				coldest month	-0.49	0.15
	precipitation	-	*	wettest month	0.12	0.74
				driest month	-0.05	0.89
	temperature		*	warmest month	0.4	0.25
leaf roundness	temperature			coldest month	-0.07	0.84
	precipitation		*	wettest month	-0.5	0.14
	procipitation			driest month	0.19	0.6
	temperature	+	(Moles <i>et al.</i> ,	warmest month	0.67	0.04
I MA			2014)	coldest month	-0.02	0.95
	precipitation	_	(Moles <i>et al.</i> ,	wettest month	-0.15	0.67
	procipitation		2014)	driest month	0.32	0.37
	temperature	+	(He et al. 2020)	warmest month	0.05	0.9
stem hydraulic			(110 01 01., 2020)	coldest month	-0.22	0.6
conductance	precipitation	+	(He <i>et al.</i> , 2020)	wettest month	0.69	0.06
			( , )	driest month	0.62	0.1

trichomes	temperature	+	*	warmest month	0.31	0.33
	tomporataro			coldest month	-0.03	0.92
	precipitation	_	*	wettest month	0.09	0.77
	P P			driest month	-0.08	0.81
vessel diameter	temperature	+	(Morris <i>et al.</i> , 2018)	warmest month	-0.62	0.07
				coldest month	-0.03	0.94
	precipitation	+	(Morris <i>et al.</i> , 2018)	wettest month	0.44	0.23
				driest month	0.22	0.56
	temperature	+	(Martínez-Cabrer a <i>et al.</i> , 2009)	warmest month	-0.68	0.21
wood density	temperature			coldest month	0.04	0.95
	precipitation	-	(Martínez-Cabrer	wettest month	-0.36	0.55
	precipitation		a <i>et al.</i> , 2009)	driest month	-0.47	0.42

**Table S7:** Instances of hybridization and introgression in the *Encelia* genus. Shown are the involved species, what type of data was used to infer the hybridization or introgression event, and notes, including the reference for the data. Data types "dstat" and "SNaQ" come from the present study.

species 1	species 2	data type	notes
E. actoni	E. frutescens	natural hybrids	no explicit data,(Clark, 1998)
E. actoni	E. frutescens frutescens	dstat	conservative test
E. actoni	E. frutescens frutescens	hybrid species	result in <i>E. virginensis</i> ; genetic analysis from (Allan <i>et al.</i> , 1997)
E. actoni	E. frutescens glandulosa	hybrid species	result in <i>E. resinifera</i> ; no formal test but included in (Allan <i>et al.</i> , 1997)
E. actoni	E. virginensis 1	dstat	conservative test
E. asperifolia	E. californica	natural hybrids	field data from DiVittorio et al., unpublished
E. asperifolia	E. californica 2	SNaQ	approximately 50% admixture
E. asperifolia	E. canescens	dstat	conservative test
E. asperifolia	E. farinosa	natural hybrids	no explicit data, (Clark, 1998)
E. asperifolia	E. palmeri	natural hybrids	field data from DiVittorio et al., unpublished
E. asperifolia	E. ventorum	natural hybrids	(Kyhos, 1967; Clark, 1998; DiVittorio <i>et</i> <i>al.</i> , 2020)
E. californica	E. farinosa	natural hybrids	http://nathistoc.bio.uci.edu/plants/Asterace ae/Encelia%20hybrids/Encelia%20hybrids .htm
E. californica	E. frutescens	hybrid species	result in <i>E. asperifolia</i> ; no formal test but included in (Allan <i>et al.</i> , 1997)
E. californica 1	E. californica 2	dstat	conservative test

E. californica 2	E. farinosa phenicodonta	dstat	conservative test
E. densifolia	E. farinosa	natural hybrids	field data from DiVittorio et al., unpublished
E. farinosa	E. frutescens	natural hybrids	field data from DiVittorio et al., unpublished; (Clark, 1998)
E. farinosa	E. palmeri	hybrid species	result in <i>E. canescens</i> ; no formal test but included in (Allan <i>et al.</i> , 1997)
E. farinosa	E. palmeri	natural hybrids	no explicit data, (Clark, 1998)
E. farinosa farinosa	E. palmeri	dstat	conservative test
E. farinosa phenicodonta	E. palmeri	dstat	conservative test
E. frutescens	E. virginensis	natural hybrids	no explicit data, (Clark, 1998)
E. frutescens frutescens	E. ravenii	dstat	conservative test
E. frutescens frutescens	E. virginensis 1	dstat	conservative test
E. frutescens frutescens	E. virginensis 2	dstat	conservative test
E. frutescens glandulosa	E. virginensis 2	dstat	conservative test
E. palmeri	E. ravenii	dstat	conservative test
E. palmeri	E. ventorum	natural hybrids	(Kyhos, 1967; Clark, 1998; DiVittorio <i>et</i> <i>al</i> ., 2020)
E. palmeri	E. virginensis 2	dstat	conservative test

# **Supplementary Figures**



**Figure S1:** Effects of model choice and lambda values ( $\lambda$ ) on  $\Phi$ IC (model fit) on chronogram fit to the *Encelia* phylogeny. We tested a range of lambda values across three models for substitution rate variation implemented in chronos in APE: "correlated", "discrete" under a strict clock, and "relaxed". The best model is the one with lowest  $\Phi$ IC, which here is  $\lambda = 0$  under a strict clock model.



**Figure S2:** Ancestral range reconstruction in *Encelia* based on the DEC model implemented in BioGeoBEARS. Top left: phylogeny for the genus; pie charts at tips show the biogeographic ranges to which species were assigned and pie charts at nodes represent the ancestral range reconstruction with the highest probability at that node. Top right: the map shows the biogeographic regions. Bottom bar graphs depict the distribution of ancestral range reconstructions for nodes A and B labeled in the phylogeny above. All possible reconstructions with >5% marginal probability are shown; for node A, these seven possible reconstructions total 88% and, for node B, these four possible reconstructions total 99%.



**Figure S3:** Individual-level phylogenies for *Encelia* inferred using (L) a concatenated maximum-likelihood approach implemented in RAxML and (R) a single nucleotide polymorphism (SNP) coalescent-based approach implemented in SVDquartets. The SVDquartets phylogeny (n = 56) contains a subset of the tips in the RAxML phylogeny (n = 69) because individuals with high levels of missing SNP data were removed. Phylogenies rooted by outgroups (not shown). Nodes labeled in red differ between the two topologies. In both phylogenies, all lineage groupings encompass monophyletic groups. Most of the nodes that differ are among individuals within a species. However, the placement of *E. actoni* and the clade consisting of *E. asperifolia* and *E. californica* differ between the two trees.



**Figure S4:** Individual-level phylogenies for *Encelia* inferred using a concatenated maximum-likelihood approach implemented in RAxML. Phylogeny rooted by outgroups (not shown). Nodal values indicate bootstrap values; all nodes with >95% bootstrap are marked with a solid circle. The individual-level phylogeny exhibits high bootstrap support for the monophyly of all lineages but *E. palmeri* (bootstrap: 84%) and *E. asperifolia* (bootstrap: 84%). Plots show (top) gene concordance factors (gCF) relative to site concordance factors (sCF) and (bottom) sCF relative to branch length. sCF and gCF values are relatively low throughout the topology. These low sCF and gCF values are likely both because ddRAD loci are short and therefore contain few phylogenetically informative sites and because of extensive incomplete lineage sorting, as is common in recent and rapid radiations like *Encelia*.



**Figure S5:** Lineage-level phylogenies inferred using coalescent-based approaches (L) using gene trees implemented in ASTRAL and (R) single nucleotide polymorphisms implemented in SVDquartets. Nodes that differ between the two topologies shown in red. The sole node that differs between the two topologies has low local posterior probability (0.93; Fig. 2).



**Figure S6:** Effects of missing data on phylogenetic inference in *Encelia*. We filtered gene trees to retain only those that were >60%, >70%, and >80% complete; number of gene trees retained were 29K, 14K, and 3.9K respectively. We then inferred a species tree for each dataset using ASTRAL-III. Outgroups not shown. Nodes with >95% local posterior probability support are marked with a solid black dot; all other nodes are shown in light blue. The only topological difference across trees is in the placement of *E. actoni* in the >80% complete tree, but the discrepant node has low support (0.67).



**Figure S7:** Reconstruction of *Encelia*'s evolutionary history as a network using SNaQ. The best fit model inferred one reticulation event; *E. asperifolia* is a hybrid with ~50% ancestry of *E. californica 2. Encelia asperifolia* has been hypothesized to be a hybrid species resulting from hybridization between *E. californica* and *E. frutescens glandulosa* (Table S7), although this hypothesis has not been formally tested previously.



**Figure S8:** Introgression across *Encelia* as measured by the D-statistic. Plotted are the p-values of measured D-statistics (Table S4). The D-statistic was calculated across all possible triads in *Encelia*; for a given species pair, shown is the greatest p-value. Insignificant tests are shown in yellow; all significant tests are shown in shades of green & blue. Some species pairs are shown in white. For these species pairs, we either could not evaluate the D-statistic because they are sister species or because, in all comparisons involving these two species, introgression occurred with the third species. Even using this conservative approach, several species-pairs show strong evidence for introgression, including some species known to hybridize in nature (Fig. 5).



**Figure S9:** Phenotypic variation in *Encelia*, depicted as phenograms. The y-axis indicates phenotypic spread across (top two rows) morphological and physiological traits and (bottom row) environmental variation. Trait abbreviations follow Table 1. Branches are colored by clade identity as shown in Fig. 2, and all species names are abbreviated to the first three characters. Closely-related species in *Encelia* often exhibit dramatically different phenotypes.



**Figure S10:** The environmental space occupied by *Encelia* species, as summarized by four bioclimatic variables: maximum temperature of the warmest month (BIO5), minimum temperature of the coldest month (BIO6), precipitation of wettest month (BIO13), and precipitation of driest month (BIO14). Species are broken up by the major clades in which they occur (Fig. 2). The species in *Encelia* occupy a diversity of climatic environments, although many of them overlap in climatic space.



**Figure S11:** Pairwise correlations among all measured traits in *Encelia* (Table S3). Wood density was excluded due to low sample size. Shown in the upper diagonal are estimates of phylogenetic correlations and p-values between In-transformed traits. Very few traits are correlated strongly |r > 0.6| and even fewer are significantly correlated. One of the few significant correlations is between leaf color and trichome density; this is expected as a high density of trichomes can result in a leaf looking more white.



**Figure S12:** Correlations between (L) trait and (R) environmental variables. Trait abbreviations follow Table 1. Width of lines indicate absolute strength of phylogenetic correlation between In-transformed traits and environment. Red lines indicate significant correlations (p < 0.05), and dotted lines indicate negative correlations. Very few traits are correlated significantly with environmental variation.