

1 **Terroir and rootstock effects on leaf shape in California Central Valley**
2 **vineyards**

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29

30 **Summary**

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32 • Embedded in a single leaf shape are the latent signatures of genetic, developmental,
33 and environmental effects. In viticulture, choice of location and rootstock are important
34 decisions that affect the performance and production of the shoot. We hypothesize that
35 these effects influence plant morphology, as reflected in leaf shape.

36 • We sample 1879 leaves arising from scion and rootstock combinations from commercial
37 vineyards in the Central Valley of California. Our design tests 20 pairwise contrasts
38 between Cabernet Sauvignon and Chardonnay scions from San Joaquin, Merced, and
39 Madera counties from vines grafted to Teleki 5C, 1103 Paulsen, and Freedom
40 rootstocks.

41 • Using geometric morphometric approaches, we visualize a morphospace in which, in
42 addition to clear separation of Cabernet Sauvignon and Chardonnay scion leaf shapes,
43 an orthogonal source of shape variation affects both varieties. Comparing the Procrustes
44 distances to within and between group means, the additional source of variance is found
45 to arise from location and rootstock effects.

46 • We describe and visualize a specific shape feature, the angle of the proximal lobe to the
47 midvein that defines the closure of the petiolar sinus, that is attributable to location and
48 rootstock effects and orthogonal to and separate from genetic, developmental, or
49 allometric effects attributable to leaf size.

50

51 **Key words**

52 grapevines; leaf shape; leaf size; morphometrics; Procrustes analysis; rootstocks; terroir; water
53 use efficiency

54

55 **Societal Impact Statement (EN)**

56

57 The innumerable effects of terroir—including climate, soil, microbial environment, biotic
58 interactions, and cultivation practice—collectively alter plant performance and production. A
59 more direct agricultural intervention is grafting, in which genetically distinct shoot and root
60 genotypes are surgically combined to create a chimera that alter shoot performance at a
61 distance. Selection of location and rootstock are intentional decisions in viticulture to positively
62 alter production outcomes. Here, we show that terroir and rootstock alter the shapes of

63 grapevine leaves in commercial vineyards throughout the California Central Valley,
64 documenting the profound effects of these agricultural interventions that alter plant morphology.
65

66 **Introduction**

67

68 Every leaf has only one shape, but that shape is the result of innumerable effects whose
69 signatures are left behind, if only we have the right tools to measure them. Using geometric
70 morphometric methods to quantify shape, these effects can be statistically measured and
71 separated from each other, revealing latent shapes that together comprise leaf morphology
72 (Chitwood and Sinha, 2016). All leaves arising from *Vitis* species are palmate with five lobes,
73 creating a geometric framework in which features are comparable between leaves and species.
74 This framework was leveraged by early ampelographers (after the Greek *ampelos*, ἄμπελος,
75 literally vine; named after the satyr lover of Dionysus that was the personification of grapevines;
76 Nonnus of Panopolis, *Dionysiaca*, Book 12) to distinguish newly introduced North American
77 rootstock species to 19th century France (Goethe, 1876; 1878; Ravaz, 1902) and eventually
78 wine grape varieties in the 20th century (Galet 1979; 1985; 1988; 1990; 2000). The unique
79 geometrical properties of grapevine leaves led to the application of rigorous mathematical
80 approaches to calculate a mean grapevine leaf while preserving intricate details, like the
81 serrations (Martínez and Grenan, 1999). This mathematical framework is the foundation of
82 geometric morphometric methods, in which statistical sampling of high numbers of leaves can
83 resolve underlying genetic (Chitwood et al., 2014; Klein et al., 2017; Demmings et al., 2019;
84 Chitwood, 2021), developmental (Chitwood et al., 2016a; Bryson et al., 2020), and
85 environmental (Chitwood et al., 2016b; Baumgartner et al., 2020; Chitwood et al., 2021)
86 contributions to grapevine leaf shape.

87

88 Yet, even though the field of ampelography was initially created to distinguish shoots of North
89 American rootstock species (Ravaz, 1902), the effects of these roots on grafted scions remains
90 understudied (Migcovsky et al., 2019; Harris et al., 2021). The power of rootstocks lies in the
91 ability for a different genotype than the scion to non-cell autonomously alter the shoot
92 phenotype at a distance (Frank and Chitwood, 2016; Warchefsky et al., 2016; Gaut et al., 2019;
93 Williams et al., 2021). As leaves are a primary constituent of shoot systems, it is reasonable to
94 ask if rootstocks can influence leaf shape. For example, grafting dominant *Me* tomato (*Solanum*
95 *lycopersicum*) mutant roots to non-mutant shoots results in translocation of the associated

96 mutant KNOTTED1-like homeobox transcript and induces leaf shape changes (Kim et al.,
97 2001).

98

99 Here, we apply geometric morphometric approaches to describe the influence of rootstock and
100 location on grapevine leaf shape. We collected 1879 Cabernet Sauvignon and Chardonnay
101 leaves during 2018 and 2019 from commercial vineyards in San Joaquin, Merced, and Madera
102 counties in the Central Valley of California grafted to Teleki 5C, 1103 Paulsen, and Freedom
103 rootstocks. Based on these leaves, we describe a specific shape feature—the angle of the
104 proximal lobe to the midvein that defines the closure of the petiolar sinus—that statistically
105 varies by rootstock and location, and is orthogonal and distinct from genetic, developmental,
106 and allometric effects of leaf size.

107

108 **Materials and Methods**

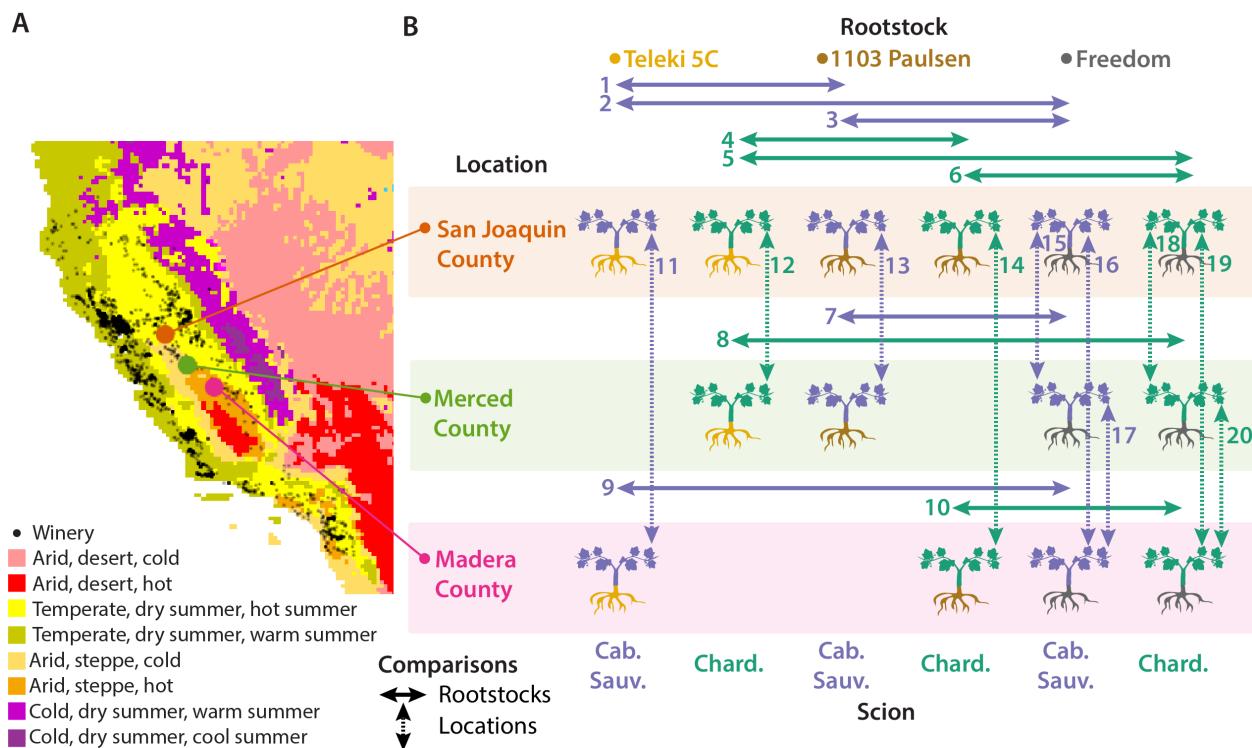
109

110 *Experimental design*

111

112 Prior to sampling in 2018, commercial vineyards plots (each with a unique rootstock by scion
113 combination) were selected in San Joaquin, Merced, and Madera county. The study sites
114 include temperate, dry and hot summer (San Joaquin County) and arid and hot steppe (Merced
115 and Madera Counties) climates according to the Köppen-Geiger classification system (**Figure**
116 **1A**, Chen and Chen, 2013). Vines with Cabernet Sauvignon and Chardonnay scions on Teleki
117 5C, 1103 Paulsen, and Freedom rootstocks were sampled during the 2018 and 2019 growing
118 seasons. In San Joaquin County, all scion and rootstock combinations were present, and all
119 scion and location combinations were sampled for the rootstock Freedom. Only select
120 comparisons could be made for rootstock and scion combinations in the Merced and Madera
121 locations and rootstocks. We chose to analyze contrasting pairs of scion, rootstock, and
122 location combinations, where only one rootstock or location contrast is made at a time. 20 such
123 contrasts are present in this study, each identified by number (**Figure 1B**). As described below,
124 this pairwise contrast design aligns with the morphometric methods we use, in which the overall
125 similarity between two shapes is measured as a Procrustes distance.

126



127 **Figure 1: Experimental design.** **A)** A map of bonded California winery locations (black points)
128 projected onto Köppen-Geiger climate classifications (see legend). **B)** Sampling design of
129 Cabernet Sauvignon (purple) and Chardonnay (dark green) scions across vineyards in San
130 Joaquin (orange), Merced (light green), and Madera (magenta) counties and Teleki 5C (yellow),
131 1103 Paulsen (brown), and Freedom (charcoal) rootstocks. 20 contrasts that evaluate effects of
132 pairs of rootstocks (solid, horizontal arrows) or locations (dotted, vertical arrows) are indicated
133 by number.
134

135

136 *Data collection*

137

138 The vineyard location sampled in San Joaquin is described in detail in Migicovsky et al. (2021;
139 2023). Due to differing vineyard orientations, leaves were collected from either the north- or
140 west-facing side of the vine. Leaves were sampled from each of the three vineyards weekly in
141 2018 for 7 weeks from June 19th to August 9th, and for 6 weeks in 2019 from June 11th to July
142 31st. Vines sampled in 2018 were resampled for the 2019 season, with the exception of those
143 sampled in the last week of 2018 due to the reduction in sampling weeks in 2019. Three vines
144 were sampled for each vineyard block on each sampling date. A LI-6800 Portable
145 Photosynthesis System (LI-Cor Biosciences, Lincoln, NE, USA) was used on two fully expanded
146 mature sunlit leaves on each vine to measure physiological traits between the hours of 10:30am

147 to 2:30pm PST (7:30 to 11:30pm UTC). For LI-6800 measurements, the following parameters
148 were kept constant: flow (600 μmols^{-1}), H_2O (RH_air 50%), CO_2 (CO_2r 400 $\mu\text{mol mol}^{-1}$),
149 temperature (Tleaf 33°C), and light (1800 $\mu\text{mol m}^{-2} \text{s}^{-1}$). At each timepoint, photosynthetic CO_2
150 rate (A , $\mu\text{mol m}^{-2} \text{s}^{-1}$) and transpiration rate (E , $\text{mol m}^{-2} \text{s}^{-1}$) were measured. These measurements
151 were also used to calculate instantaneous Water Use Efficiency (WUEi) as A/E ($\mu\text{mol/mol}$).
152

153 A single undamaged shoot was selected from each vine and three leaves were collected from
154 that shoot: the youngest fully expanded leaf, a leaf roughly in the middle of the shoot, and the
155 oldest intact leaf closest to the shoot base. Leaves were trimmed of petioles, placed in a plastic
156 bag on top of one another in order (young, middle, and oldest), and stored in a cooler. Each leaf
157 had its abaxial surface scanned using either a DS-50,000 (Epson, Suwa, Japan) or CanoScan
158 LiDE 220 (Canon, Ōta, Japan) scanner in color with a white background at 1200 dpi. Resulting
159 images were saved as .jpeg files. 21 landmarks were placed onto each leaf scan following the
160 protocol of Chitwood et al. (2021) using the open source software ImageJ (Schneider et al,
161 2012) or Fiji (Schindelin et al., 2012). Coordinates for each landmark were exported as a CSV
162 file which were merged together with metadata to serve as basis for all subsequent analysis
163 (available here: https://github.com/DanChitwood/terroir_and_rootstock/).
164

165 *Morphometric and statistical analysis*

166
167 The overall similarity between two shapes defined by the same number of points with the same
168 number of dimensions can be measured as a Procrustes distance. Using functions of
169 translation, rotation, scaling, and reflection, the Procrustes distance, calculated between the
170 corresponding points of each shape, is minimized (Goodall, 1991). A population of shapes can
171 be superimposed upon each other through the Generalized Procrustes Analysis (GPA)
172 algorithm (Gower, 1975). Briefly, in GPA an arbitrary shape is chosen as a reference to which
173 all other shapes are aligned. The mean shape is calculated, to which all shapes are again
174 aligned. The algorithm continues until the Procrustes distance between the mean shapes of two
175 cycles falls below an exceedingly low arbitrary threshold. All shapes are then superimposed to
176 the calculated GPA mean shape, allowing corresponding coordinate points to be used
177 comparatively in subsequent statistical analyses.
178

179 In this study, 1879 leaf shapes were analyzed. Each leaf consisted of 21 landmarks with 2
180 coordinates values such that the total dataset size was $1879 \times 21 \times 2 = 78918$. In order to

181 evaluate the overall similarity of two shapes to each other, a Procrustes distance was
182 calculated. To use this measure of similarity to statistically test whether a difference in shape
183 exists between each of the 20 contrasts we evaluated, we used the Kruskal-Wallis one-way
184 analysis of variance to compare within and between group distances to mean leaf shapes. For
185 example, to contrast leaf shapes between groups A and B, we first calculated the GPA means
186 for each group and measured the Procrustes distance of each leaf to its respective mean. We
187 then did the same, but measuring the Procrustes distances of all leaves to the overall common
188 mean. The Kruskal-Wallis one-way analysis of variance was used to determine whether the
189 Procrustes distances of leaves to their respective group means was statistically less than the
190 distance to the overall common mean. A similar method was used to evaluate the physiological
191 data, but instead of using Procrustes distances to a mean shape, the absolute value of residuals
192 to water use efficiency curves modeling photosynthetic rate as a function of transpiration rate
193 were compared between and within groups. We adjusted the reported p values for the contrasts
194 of leaf shape and physiological data using Bonferroni multiple test correction.

195

196 All analyses were performed using Python (version 3.10.9) including the numpy (Harris et al.,
197 2020), pandas (McKinney, 2010), matplotlib (Hunter, 2007), and seaborn (Waskom, 2021)
198 modules. The procrustes and stats modules from scipy (Virtanen et al., 2020) were used for
199 Procrustes analysis, Kruskal-Wallis one-way analysis of variance, and the calculation of
200 Spearman correlation coefficients. The PCA function from sklearn (Pedregosa et al., 2011) was
201 used for Principal Component Analysis (PCA) and the inverse transform to calculate eigenleaf
202 shapes of the morphospaces. statsmodels (Seabold and Perktold, 2010) was used for
203 Bonferroni multiple test adjustment. The curve_fit function from scipy.optimize was used to fit
204 curves for water use efficiency (WUE), modeling photosynthetic rate (A, $\text{umol m}^{-2} \text{s}^{-1}$) as a
205 function of transpiration rate (E, $\text{mol m}^{-2} \text{s}^{-1}$) using the function $A = m * \ln(E) - b$, where m and b
206 are the estimated slope and intercept of a linear function.

207

208 The data and analyses that support the findings of this study are openly available in github at
209 https://github.com/DanChitwood/terroir_and_rootstock.

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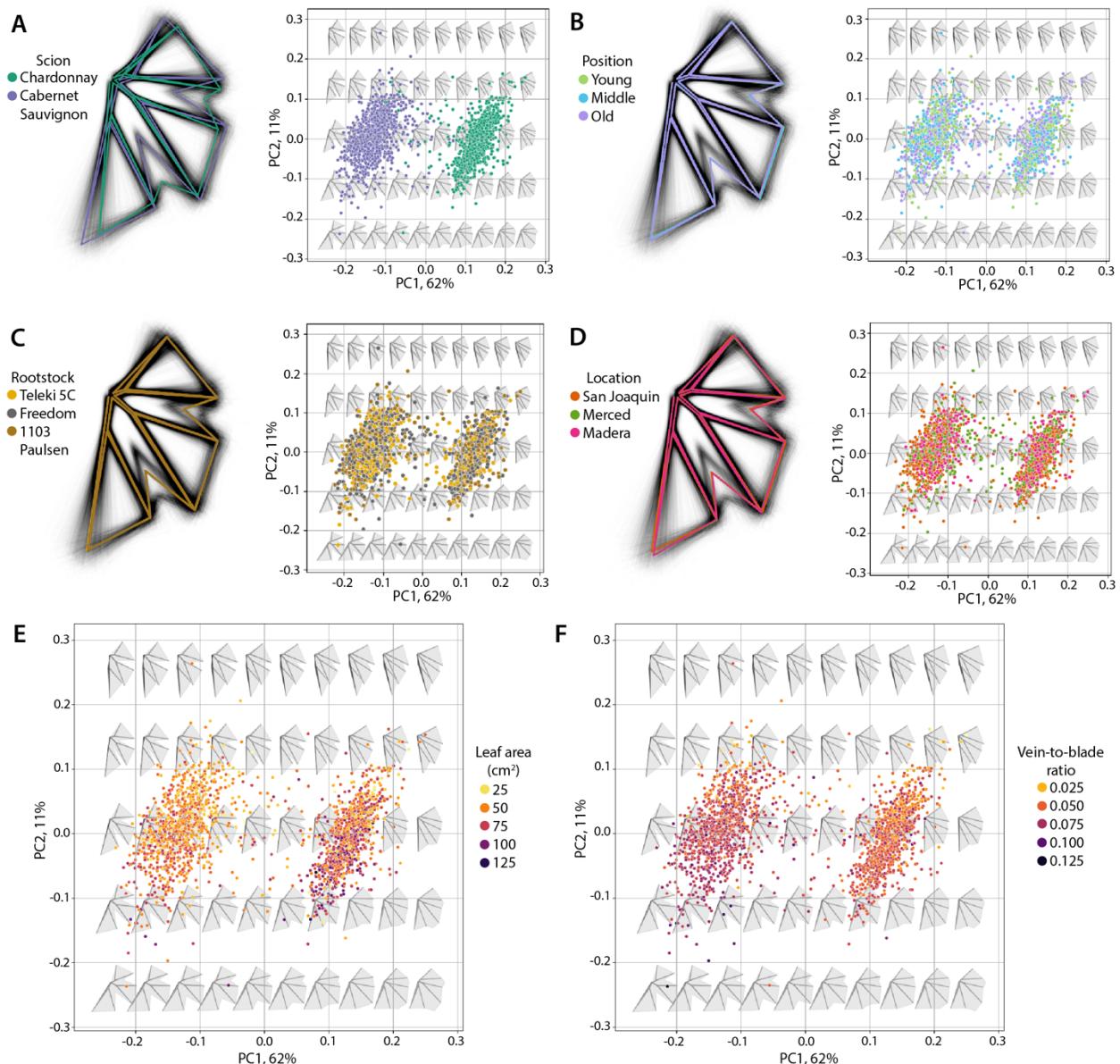
211 **Results**

212

213 In order to visualize the main sources of variance in our data, we performed a Principal
214 Component Analysis (PCA) on the 21 landmarks, each with an x and y coordinate for a total

215 vector length of 42, for all 1879 leaves superimposed using Generalized Procrustes Analysis
216 (GPA) (**Figure 2**). The inverse transform (which given a PCA, provided PC values can return
217 the corresponding leaf shape coordinates) can be used to generate eigenleaves that visualize
218 the variance explained by each PC, or, the morphospace. The first PC explains 62% of the
219 variation in our data and is associated with shape variation corresponding to deeply lobed
220 leaves (low PC1 values) to leaves with less lobing (high PC1 values). Predictably, Cabernet
221 Sauvignon (deeply lobed) and Chardonnay (less lobed) leaves are, with the exception of a few
222 outliers, cleanly separated along this axis and their mean leaf shapes are visibly different from
223 each other, especially with respect to the depth of the sinuses (**Figure 2A**). Despite the clear
224 division of groups based on scion variety, there are a handful of more highly lobed Chardonnay
225 and less lobed Cabernet Sauvignon, as well as several leaves with an intermediate level of
226 lobing that represent a continuum between groups. Given that leaves were sampled at multiple
227 stages of development, and arising from different rootstocks and locations, these seemingly
228 misplaced leaves may represent, for example, a young Cabernet Sauvignon leaf that is less
229 lobed at that stage of development or responding to a unique set of environmental factors.
230 However, shoot position (**Figure 2B**), rootstock (**Figure 2C**) and location (**Figure 2D**), do not
231 correspond in obvious ways to the variance explained by PC1 and PC2 (which combined
232 explain 73% of variance) and the mean leaves of each factor level overlap extensively and can
233 not be differentiated from each other by eye.

234



235

236 **Figure 2: Morphospace.** **A)** Principal Component Analysis (PCA) on Generalized Procrustes
 237 Analysis (GPA)-adjusted landmarks. Left: Superimposed landmarks of all leaves (black) and the
 238 Procrustes mean leaf shapes for Chardonnay (green) and Cabernet Sauvignon (purple). Right:
 239 eigenleaf representations across the PCA morphospace. Points are colored by scion identity.
 240 Similar to (A), panels **(B)**, **(C)**, and **(D)** show superimpositions of Procrustes mean leaves (left)
 241 and projections onto the PCA morphospace (right) for shoot position, rootstock, and location
 242 factors, respectively (see legends). Allometric indicators of E) leaf area (cm^2) and F) vein-to-
 243 blade ratio are projected onto the morphospace. Values are indicated by color (see legends).

244

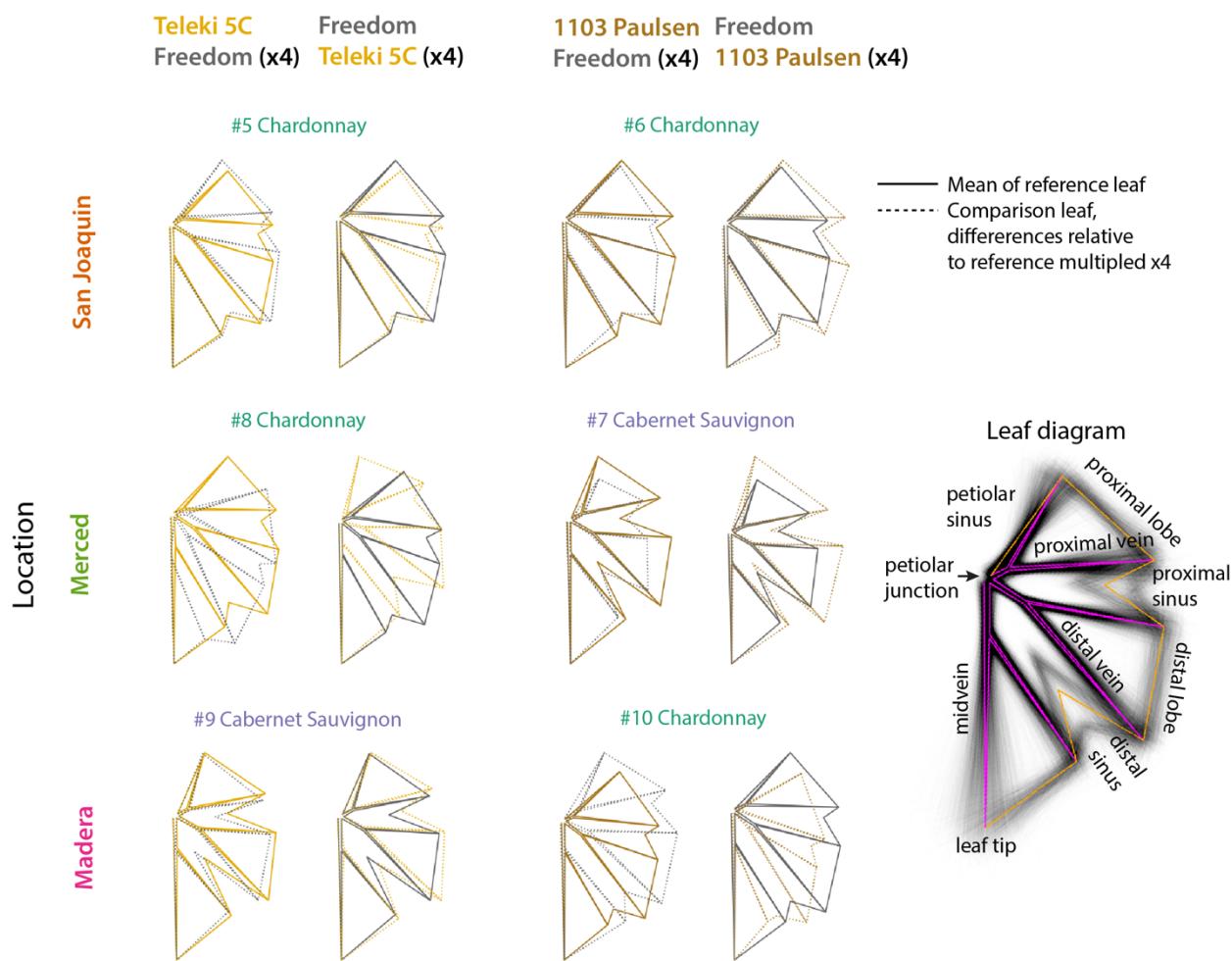
245 While PC1 explains over half of the variation in leaf shape, PC2 explains 11% of the variation.
246 Leaves with low PC2 values have proximal lobes with large angles, sometimes exceeding 180°,
247 away from the midvein, covering a larger proportion of the petiolar sinus region, whereas in
248 leaves with high PC2 values the angle between the proximal lobe and midvein is much smaller,
249 sometimes almost 90°, creating a much flatter leaf base in which the proximal lobes on each
250 side of the leaf do not overlap. Both Cabernet Sauvignon and Chardonnay leaves, although
251 separated along PC1, vary along PC2 in similar ways. The similar distributions within each
252 group are suggestive of a shared effect, which we hypothesize may represent leaf size.
253 Previously we demonstrated that the ratio of vein-to-blade area in grapevine leaves is inversely
254 proportional to leaf area (Chitwood et al., 2021) and is a useful indicator of size in normalized
255 leaves arising from morphometric analysis (as is the case here). Projecting leaf area (**Figure**
256 **2E**) and vein-to-blade ratio (**Figure 2F**) values on our data, some structure is observed.
257 However, the Spearman correlation coefficients for each of these variables with PC2 is
258 marginal. For Cabernet Sauvignon, the correlation coefficient values for leaf area and vein-to-
259 blade ratio are -0.30 (p value = 5.0×10^{-21}) and -0.40 (p value = 5.0×10^{-38}), respectively; for
260 Chardonnay, the correlation coefficient values for leaf area and vein-to-blade ratio are -0.31 (p
261 value = 4.5×10^{-22}) and -0.25 (p value = 4.2×10^{-15}), respectively.
262
263 To determine if rootstock and location significantly contribute to differences in leaf shape, we
264 turned to the unique structure of our experimental design and the power of using Procrustes
265 distance as a measure of the overall similarity between two leaf shapes. Within our
266 experimental design are 20 contrasts, in which for a pair of scion, rootstock, and locations
267 values, samples differ only by rootstock or by location (**Figure 1**). Reducing our analysis to
268 these 20 contrasts allowed us to leverage the ability of the Procrustes distance metric to
269 compare overall similarity between two samples. By comparing the Procrustes distances of
270 each leaf to the respective mean shape of its group to the distances calculated for each leaf to
271 the overall mean shape using a Kruskal-Wallis one-way analysis of variance, we could assign a
272 p value (multiple test adjusted using Bonferroni) that the leaf shapes of the two contrasting
273 groups differ (**Table 1**).
274
275 In San Joaquin County, all rootstock contrasts in both Cabernet Sauvignon and Chardonnay
276 were tested, but none of the Cabernet Sauvignon rootstock contrasts were significant. Across all
277 three locations no comparison of Teleki 5C to 1103 Paulsen was significant for either scion.
278 However, all other rootstock and location contrasts were significant. To visualize these shape

279 differences, for each pair of mean leaves, we magnified the difference of each to the other x4 (to
280 see subtle shape effects) and plotted on top of the other mean leaf. Leaves were rotated and
281 scaled so that their midveins overlapped, allowing relative changes in shape to be more easily
282 discerned. There were no qualitative differences in the types of shape differences between
283 rootstock (**Figure 3**) and location (**Figure 4**) contrasts. For some contrasts, slight differences in
284 sinus depth are observed. However, the strongest observable effect was the angle of the
285 proximal lobe to the midvein, similar to the shape variance associated with PC2 described
286 above (**Figure 2**). Consistent directionality (for example, a particular rootstock or location having
287 a wider proximal lobe angle than another) was not immediately obvious.

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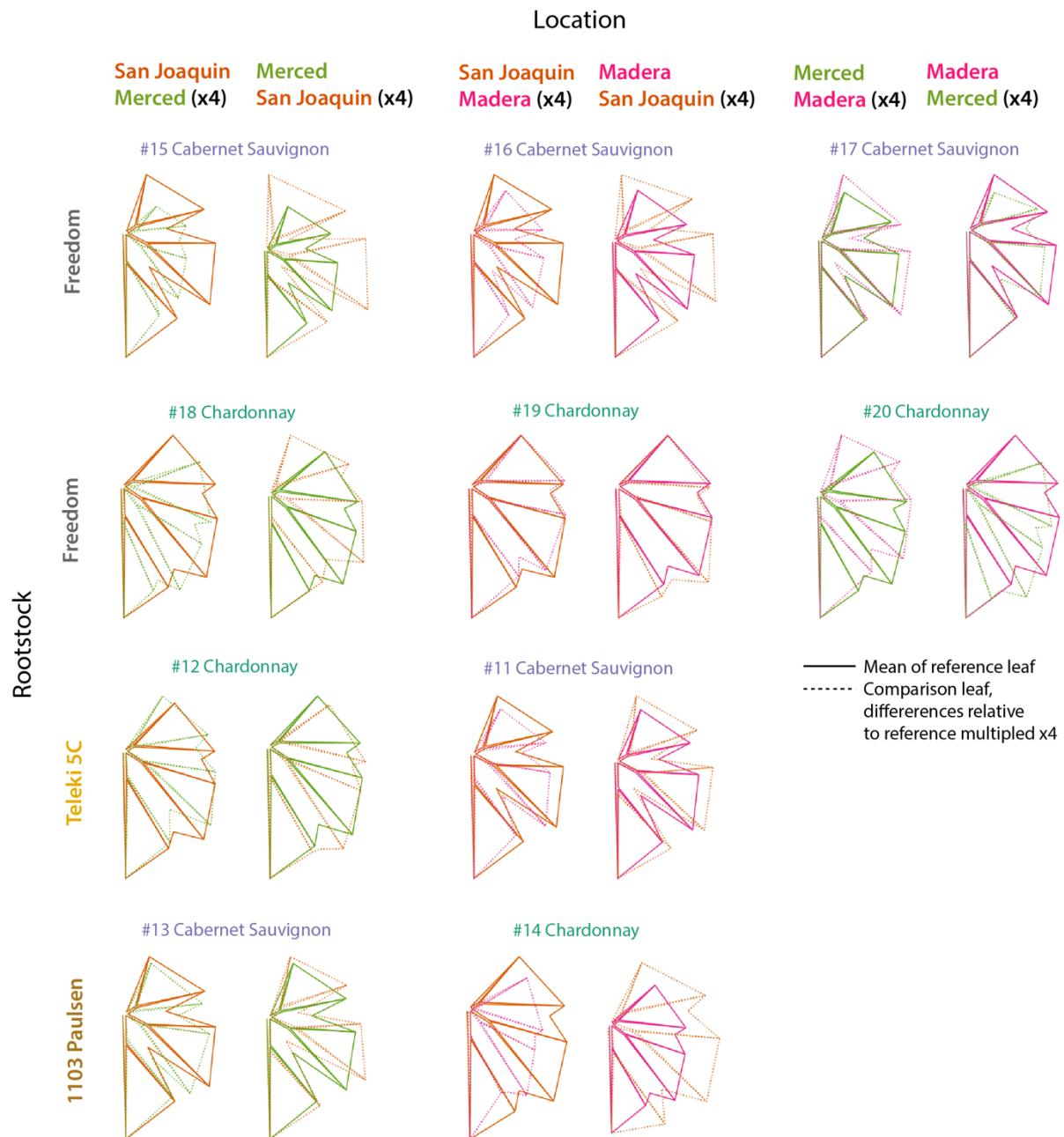
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Rootstock



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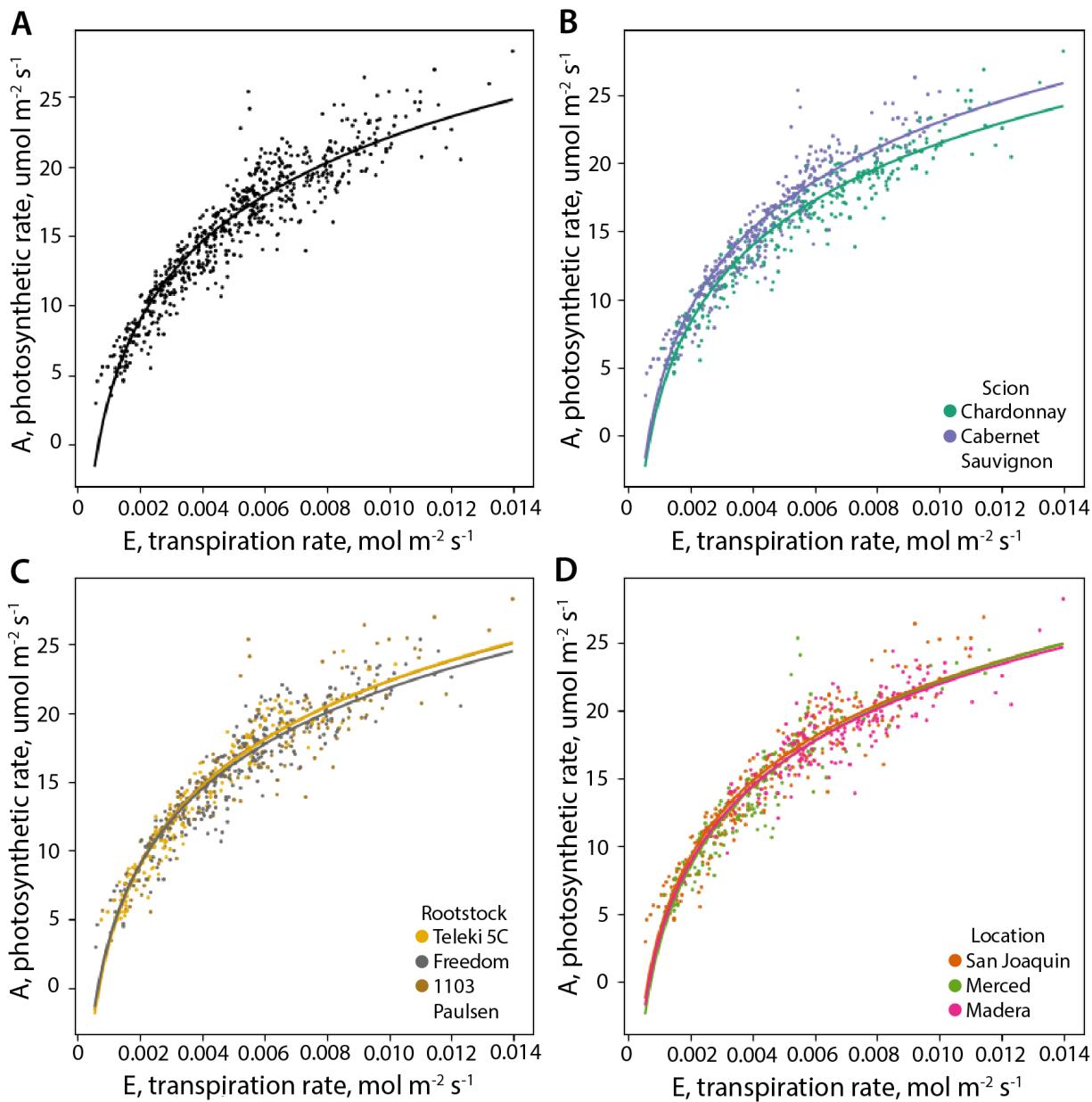
291 **Figure 3: Comparisons of rootstock effects.** For each significant rootstock comparison,
292 visualizations of differences between Procrustes mean leaf shapes are visualized as a
293 reference leaf (solid outline) to a comparison leaf (dotted outline), in which the difference to the
294 reference has been multiplied by x4. The differences of each rootstock to the other are
295 visualized in turn. Rootstock pairs are arranged by column and the locations the samples arise
296 from by row. The identification number of each contrast and the scion that was sampled are
297 indicated. A leaf diagram labels morphological features on top of GPA-adjusted leaf outlines and
298 the overall Procrustes mean leaf (orange and magenta).



299

300 **Figure 4: Comparisons of location effects.** For each significant location comparison,
301 visualizations of differences between Procrustes mean leaf shapes are visualized as a
302 reference leaf (solid outline) to a comparison leaf (dotted outline), in which the difference to the
303 reference has been multiplied by x4. The differences of each location to the other are visualized
304 in turn. Location pairs are arranged by column and the rootstocks the samples arise from by
305 row. The identification number of each contrast and the scion that was sampled are indicated.
306

307 We analyzed the physiological data, collected from the same vines as the leaf shape data, in
308 the same manner as our shape data, so that the results could be directly compared. We first
309 modeled water use efficiency by fitting curves of photosynthetic rate (A, $\text{umol m}^{-2} \text{s}^{-1}$) versus
310 transpiration rate (E, $\text{mol m}^{-2} \text{s}^{-1}$) (**Figure 5A**). Differences in the trajectories of the Cabernet
311 Sauvignon and Chardonnay water use efficiency curves can be seen, with Cabernet Sauvignon
312 assimilating at higher rates for a given transpiration rate than Chardonnay (**Figure 5B**), but
313 differences between rootstocks (**Figure 5C**) and locations (**Figure 5D**) are more subtle. Similar
314 to leaf shape, to see if we could detect differences in any of the 20 contrasts, we compared the
315 absolute value of residuals to the fitted curve of samples to their respective group to the
316 absolute value of residuals of the overall fitted curve using a Kruskal-Wallis one-way analysis of
317 variance. After multiple test adjustment, none of the 20 contrasts was statistically significant
318 (data not shown). Instead, for each sample we calculated instantaneous water use efficiency
319 (WUEi, A/E) and tested if any of the 20 contrasts were statistically significant, again using the
320 Kruskal-Wallis test. A number of the contrasts were statistically significant after multiple test
321 adjustment, but there was no obvious overlap in the contrasts that were significant between leaf
322 shape and WUEi (**Table 1**).
323



324

325 **Figure 5: Water use efficiency (WUE) models. A)** For all samples, photosynthetic rate (A ,
326 $\text{umol m}^{-2} \text{s}^{-1}$) plotted against transpiration rate (E , $\text{mol m}^{-2} \text{s}^{-1}$). A fitted curve modeling
327 photosynthetic rate as a function of transpiration rate, $A = m \ln(E) - b$, is shown. Similar to (A),
328 panels **(B)**, **(C)**, and **(D)** show plots and fitted curves by scion, rootstock, and location factors,
329 respectively (see legends).

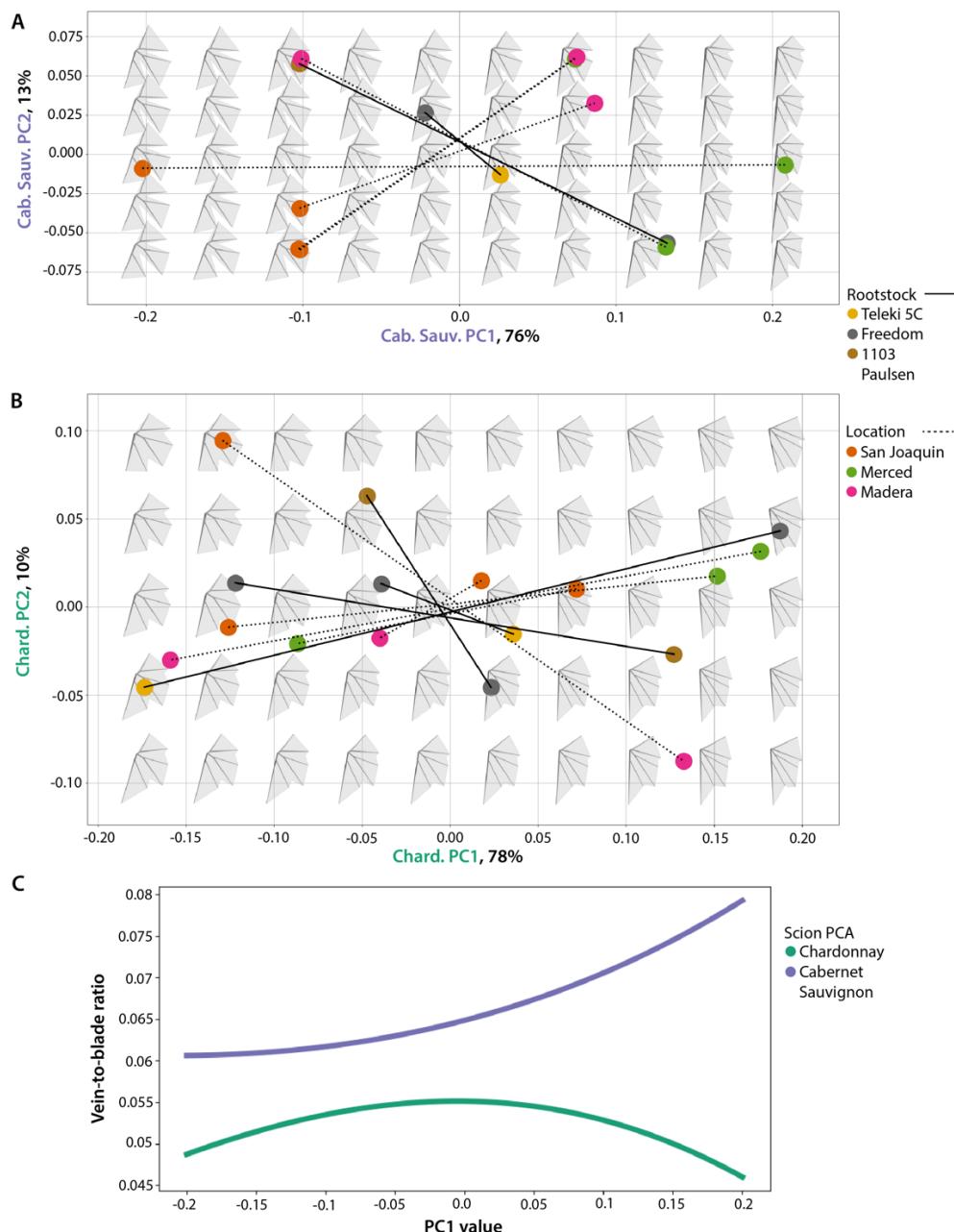
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332

333 Our results demonstrate that there are statistical differences in leaf shape between rootstock
334 and location (**Table 1**) and that this shape variation has qualitative similarities, such as
335 variability in the angle of the proximal lobe to the midvein (**Figures 3 and 4**). To isolate these
336 shape differences more specifically, we created morphospaces for Cabernet Sauvignon (**Figure**
337 **6A**) and Chardonnay (**Figure 6B**) leaves using only differences in leaf shape between the mean
338 leaves for each significant contrast (magnified x4, as shown in **Figures 3 and 4**). The leaf
339 shapes representing each contrast are connected by line segments, and predictably fall on
340 opposite sides of their respective PC1 axes. The overwhelming observed shape variance along
341 each PC1 axis is the angle of the proximal lobe to the midvein, with nearly 180° angles
342 observed in both Cabernet Sauvignon and Chardonnay for the lowest PC1 values that decrease
343 with higher PC1 values. Directionality is not observed, except for the case of location contrasts
344 in Cabernet Sauvignon leaves, in which leaves from San Joaquin County have larger proximal
345 lobe angles than those from Merced or Madera counties.

346



347

348 **Figure 6: Morphospace of rootstock and location effects on leaf shape.** For each
 349 significant contrast, divided by Cabernet Sauvignon (**A**) and Chardonnay (**B**) scions, the
 350 magnified differences (x4) in leaf shape were used to construct a morphospace. Each pair of
 351 contrasted leaf shapes is connected by a line segment indicating the type of comparison, either
 352 rootstock (solid) or location (dotted). Points are colored by identity (see legends). Eigenleaf
 353 representations are provided to visualize the morphospace. **C**) The modeled vein-to-blade ratio
 354 values for eigenleaves across PC1 values for the Cabernet Sauvignon (purple) and Chardonnay
 355 (green) PCA morphospaces.

356

357 Variation in proximal lobe angle in the PCAs representing individual contrasts by scion (**Figure**
358 **6**) reflect the variation observed along PC2 in the overall PCA (**Figure 2**). To determine if this
359 represented an allometric effect related to leaf area, we modeled vein-to-blade ratio (as a proxy
360 of leaf size for normalized leaves) as a function of each PC1 value from **Figure 6A** or **Figure 6B**
361 using eigenleaf representations from the inverse transform (**Figure 6C**). Cabernet Sauvignon
362 values are higher than Chardonnay as expected for a deeply lobed leaf (Migicovsky et al., 2022)
363 and show a marginally positive relationship between PC1 values and vein-to-blade ratio, but
364 Chardonnay does not. The range of vein-to-blade ratios across PC1 values of the PCA of leaf
365 differences is 0.045 to 0.080 (**Figure 6C**), only a fraction of the vein-to-blade ratios observed for
366 actual leaves (0.025 to 0.125, **Figure 2F**). We therefore do not attribute allometric (leaf size)
367 variation to that explaining leaf shape differences by rootstock or location effects.

368

369

370 Discussion

371

372 By measuring leaf shape across three vineyards, two scions, and three rootstocks for a total of
373 13 weeks across two years, we are able to describe the impact of both terroir and rootstock on
374 altering the shapes of grapevine leaves throughout the California Central Valley. At one location
375 in our study, San Joaquin County, all contrasts for the three rootstocks grafted to both Cabernet
376 Sauvignon and Chardonnay are present. The location was a production vineyard, in which 15
377 rootstocks are grafted to both scions in a randomized block design. At this location, we have
378 analyzed historical data showing that rootstock choice can modulate yield and vegetative
379 biomass, and even more strongly Ravaz index (the ratio of yield to vegetative biomass) by
380 almost up to 100% (Migicovsky et al., 2021). At the end of 30 years of production at this site, we
381 analyzed the dendrochronology of scion trunk segments using X-ray Computed Tomography,
382 showing that underlying the effects on Ravaz index, rootstocks had altered secondary
383 patterning of the vasculature, and likely the hydraulic performance, of the scion continuously
384 over the life of the vineyard (Migicovsky et al., 2023). Both the effects of rootstocks on Ravaz
385 index and secondary patterning are strongly additive and robust: that is, regardless of scion
386 properties or environmental effects, rootstocks consistently add or subtract from scion trait
387 values. Although we only measured 3 of the 15 rootstocks present in San Joaquin County that
388 are comparable with the other sites, the lack of directionality in rootstock-induced changes in
389 leaf shape (**Figures 3 and 6**) and associated changes in photosynthetic assimilation or

390 transpiration measured coincident with collecting leaves (**Figure 5, Table 1**), suggests that leaf
391 shape is not a mechanism affecting plant physiology. Nonetheless, we detect clear changes in
392 leaf shape, orthogonal to genetic differences that define varieties, that arise from effects of
393 rootstocks and location (**Figures 2, 3, 4, and 6**). It is possible that such shape changes
394 represent a constraint that results from changes in vascular patterning, hydraulic flux, the
395 canopy, or the ratio of reproductive to vegetative biomass to which it is grafted.

396

397 Although the effects of rootstock and location on leaf shape we are proposing here are new, we
398 previously proposed a framework in which genetic (Chitwood et al., 2014; Klein et al., 2017;
399 Demmings et al., 2019; Chitwood, 2021), developmental (Chitwood et al., 2016a; Bryson et al.,
400 2020), and year-to-year variation in responses to the environment (Chitwood et al., 2016b;
401 Chitwood et al., 2021) are orthogonal and separate to each other (Chitwood and Topp, 2015;
402 Chitwood and Mullins, 2022). Leaf shape variation associated with rootstock and location is
403 strikingly orthogonal to the changes in leaf shape that strongly separate Cabernet Sauvignon
404 and Chardonnay, to the point that it creates very similar distributions in each variety (**Figure 2**).
405 We previously described allometric changes in grapevine leaf shape that are inversely
406 proportional to leaf size across the *Vitis* morphospace, such that the natural log of the ratio of
407 vein area to blade area in a leaf linearly decreases relative to the natural log of leaf area
408 (Chitwood et al., 2021; Chitwood and Mullins, 2022). We reject the hypothesis that the variation
409 orthogonal to the shape differences that define Cabernet Sauvignon and Chardonnay is a
410 random effect due to stochastic sampling of leaf area, since the differences in mean leaf shape,
411 specifically the variation in vein-to-blade ratio (and therefore leaf size) (**Figure 6**), was only a
412 small fraction of the total in the dataset (**Figure 1**). Rather, we believe that the angle of the
413 proximal lobe to the midvein, that defines the petiolar sinus, is a unique leaf morphology trait
414 that varies by rootstock and location arising from physiological effects.

415

416 Conclusion

417

418 We describe an additive effect on leaf shape that varies with rootstock and location in grapevine
419 leaves. The angle of the proximal lobe to the midvein modulates the shape of the petiolar sinus.
420 This shape variation is orthogonal to and qualitatively distinct from variation arising from genetic
421 differences across *Vitis* or allometric variation arising from developmental or year-to-year effects
422 that alter leaf size. The variation does not seem to causally affect leaf photosynthetic rates or

423 transpiration, but likely arises as a developmental constraint from changes in overall plant
424 physiology impacted by terroir and rootstock choice.

425

426 **Acknowledgements**

427

428 All authors were supported by the National Science Foundation Plant Genome Research
429 Program award number 1546869. DHC was additionally supported by National Science
430 Foundation Plant Genome Research Program award numbers IOS-2310355, IOS-2310356, and
431 IOS-2310357.

432

433 We acknowledge Julie Curless (Missouri State University), Mya Ly(Missouri State University),
434 Vy Nguyen (Missouri State University), Dalton Gilig (University of Missouri), and Ilona Natsch
435 (Saint Louis University) for assistance in sampling and landmarking of the leaves. We would
436 also like to acknowledge Laszlo Kovacs (Missouri State University) and Misha Kwasniewski
437 (The Pennsylvania State University) for student supervisory support.

438

439 We thank E & J Gallo Winery and vineyard managers and workers for their generous
440 contributions of time providing safe access and training to sample commercial sites and collect
441 data.

442

443 **Author Contributions**

444

445 Data collection: ZM, JFS, MA, ZH, LLK, LP, KW

446 Student advising and supervision: ZM, JFS, PC, AYF, AJM, DHC

447 Conceptualization: ZM, JFS, PC, AYF, AJM, DHC

448 Data analysis: ZM, JFS, DHC

449 Writing: ZM, JFS, DHC

450 Reading and revising: ZM, JFS, MA, ZH, LLK, LP, KW, PC, AYF, AJM, DHC

451

452 **Data Availability Statement**

453

454 The data that support the findings of this study are openly available in github at
455 https://github.com/DanChitwood/terroir_and_rootstock.

456

457 **Conflict of Interest Statement**

458

459 This study was conducted in collaboration with E & J Gallo Winery, which provided access to
460 commercial vineyards throughout California described in this study and from which data was
461 collected. Peter Cousins, an author of this study, is an employee of E & J Gallo Winery. The
462 remaining authors declare that the research was conducted in the absence of any commercial
463 or financial relationships that could be construed as a potential conflict of interest.

464

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614

615 **Figure and Table Legends**

616

617 **Figure 1: Experimental design.** **A)** A map of bonded California winery locations (black points)
618 projected onto Köppen-Geiger climate classifications (see legend). **B)** Sampling design of
619 Cabernet Sauvignon (purple) and Chardonnay (dark green) scions across vineyards in San
620 Joaquin (orange), Merced (light green), and Madera (magenta) counties and Teleki 5C (yellow),
621 1103 Paulsen (brown), and Freedom (charcoal) rootstocks. 20 contrasts that evaluate effects of

622 pairs of rootstocks (solid, horizontal arrows) or locations (dotted, vertical arrows) are indicated
623 by number.

624

625 **Figure 2: Morphospace.** **A)** Principal Component Analysis (PCA) on Generalized Procrustes
626 Analysis (GPA)-adjusted landmarks. Left: Superimposed landmarks of all leaves (black) and the
627 Procrustes mean leaf shapes for Chardonnay (dark green) and Cabernet Sauvignon (purple).
628 Right: eigenleaf representations across the PCA morphospace. Points are colored by scion
629 identity. Similar to (A), panels **(B)**, **(C)**, and **(D)** show superimpositions of Procrustes mean
630 leaves (left) and projections onto the PCA morphospace (right) for shoot position, rootstock, and
631 location factors, respectively (see legends). Allometric indicators of **E)** leaf area (cm^2) and **F)**
632 vein-to-blade ratio are projected onto the morphospace. Values are indicated by color (see
633 legends).

634

635 **Figure 3: Comparisons of rootstock effects.** For each significant rootstock comparison,
636 visualizations of differences between Procrustes mean leaf shapes are visualized as a
637 reference leaf (solid outline) to a comparison leaf (dotted outline), in which the difference to the
638 reference has been multiplied by x4. The differences of each rootstock to the other are
639 visualized in turn. Rootstock pairs are arranged by column and the locations the samples arise
640 from by row. The identification number of each contrast and the scion that was sampled are
641 indicated. A leaf diagram labels morphological features on top of GPA-adjusted leaf outlines and
642 the overall Procrustes mean leaf (orange and magenta).

643

644 **Figure 4: Comparisons of location effects.** For each significant location comparison,
645 visualizations of differences between Procrustes mean leaf shapes are visualized as a
646 reference leaf (solid outline) to a comparison leaf (dotted outline), in which the difference to the
647 reference has been multiplied by x4. The differences of each location to the other are visualized
648 in turn. Location pairs are arranged by column and the rootstocks the samples arise from by
649 row. The identification number of each contrast and the scion that was sampled are indicated.

650

651 **Figure 5: Water use efficiency (WUE) models.** **A)** For all samples, photosynthetic rate (A,
652 $\text{umol m}^{-2} \text{s}^{-1}$) plotted against transpiration rate (E, $\text{mol m}^{-2} \text{s}^{-1}$). A fitted curve modeling
653 photosynthetic rate as a function of transpiration rate, $A = m * \ln(E) - b$, is shown. Similar to (A),
654 panels **(B)**, **(C)**, and **(D)** show plots and fitted curves by scion, rootstock, and location factors,
655 respectively (see legends).

656

657 **Figure 6: Morphospace of rootstock and location effects on leaf shape.** For each
658 significant contrast, divided by Cabernet Sauvignon (**A**) and Chardonnay (**B**) scions, the
659 magnified differences (x4) in leaf shape were used to construct a morphospace. Each pair of
660 contrasted leaf shapes is connected by a line segment indicating the type of comparison, either
661 rootstock (solid) or location (dotted). Points are colored by identity (see legends). Eigenleaf
662 representations are provided to visualize the morphospace. **C**) The modeled vein-to-blade ratio
663 values for eigenleaves across PC1 values for the Cabernet Sauvignon (purple) and Chardonnay
664 (green) PCA morphospaces.

665

666

667 **Table 1: Contrasts by rootstock and location and associated p values for differences in**
668 **leaf shape and instantaneous water use efficiency (WUEi).**

ID	Comparison	Scion	Constant	Variable 1	Variable 2	Shape p val.	WUEi p val.
1	root	Cab. Sauv.	San Joaquin	Teleki 5C	1103P	1.00E+00	5.35E-01
2	root	Cab. Sauv.	San Joaquin	Teleki 5C	Freedom	1.00E+00	1.00E+00
3	root	Cab. Sauv.	San Joaquin	1103P	Freedom	1.00E+00	2.44E-01
4	root	Chard.	San Joaquin	Teleki 5C	1103P	1.04E-01	1.65E-02
5	root	Chard.	San Joaquin	Teleki 5C	Freedom	3.22E-05	1.00E+00
6	root	Chard.	San Joaquin	1103P	Freedom	4.51E-09	7.44E-02
7	root	Cab. Sauv.	Merced	1103P	Freedom	8.43E-08	2.55E-09
8	root	Chard.	Merced	Teleki 5C	Freedom	4.01E-12	1.63E-09
9	root	Cab. Sauv.	Madera	Teleki 5C	Freedom	7.39E-11	1.00E+00
10	root	Chard.	Madera	1103P	Freedom	3.02E-15	1.00E+00
11	location	Cab. Sauv.	Teleki 5C	San Joaquin	Madera	4.98E-14	1.61E-05
12	location	Chard.	Teleki 5C	San Joaquin	Merced	4.00E-18	1.30E-02
13	location	Cab. Sauv.	1103P	San Joaquin	Merced	4.54E-17	6.77E-02
14	location	Chard.	1103P	San Joaquin	Madera	7.87E-21	6.71E-02
15	location	Cab. Sauv.	Freedom	San Joaquin	Merced	8.10E-20	6.10E-06
16	location	Cab. Sauv.	Freedom	San Joaquin	Madera	6.25E-19	1.97E-05
17	location	Cab. Sauv.	Freedom	Merced	Madera	5.37E-18	1.00E+00
18	location	Chard.	Freedom	San Joaquin	Merced	5.92E-23	7.69E-03
19	location	Chard.	Freedom	San Joaquin	Madera	5.99E-27	3.21E-09
20	location	Chard.	Freedom	Merced	Madera	3.91E-32	2.15E-02

669

670