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1 **Disparities among crop species in the evolution of growth rates: the**
2 **role of distinct origins and domestication histories**

3

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15 **Summary**

- 16 • Growth rates vary widely among plants with different strategies. For crops,
17 evolution under predictable and high-resource environments might favour rapid
18 resource acquisition and growth, but whether this strategy consistently evolved
19 during domestication and improvement remains unclear.
- 20 • Here, we report a comprehensive study of the evolution of growth rates based on
21 comparisons among wild, landrace, and improved accessions of 19 herbaceous
22 crops grown under common conditions. We also examined the underlying growth
23 components and the influence of crop origin and history on growth evolution.
- 24 • Domestication and improvement did not affect growth consistently, *i.e.* growth rates
25 increased or decreased or remained unchanged in different crops. Crops selected for
26 fruits increased the physiological component of growth (net assimilation rate),
27 whereas leaf and seed crops showed larger domestication effects on morphology
28 (leaf mass ratio and specific leaf area). Moreover, climate and phylogeny
29 contributed to explaining the effects of domestication and changes in growth.
- 30 • Crop-specific responses to domestication and improvement suggest that selection
31 for high yield has not consistently changed growth rates. The trade-offs between
32 morpho-physiological traits and the distinct origins and histories of crops accounted
33 for the variability in growth changes. These findings have far-reaching implications
34 for our understanding of crop performance and adaptation.

35

36 **Keywords:** domestication, functional groups, leaf mass ratio, net assimilation rate,
37 relative growth rate, specific leaf area, *Triticum turgidum* L., wild progenitors

39 **Introduction**

40 Evolution under cultivation involves a diverse range of natural and artificial selection
41 pressures that have changed crop phenotypes for millennia (Evans, 1993; Doebley *et al.*,
42 2006; Purugganan & Fuller, 2009). Our understanding of crop evolution is primarily
43 based on reproductive traits (*e.g.* seed size, flowering time, yield), which have received
44 more attention than vegetative development and growth (Milla *et al.*, 2015; Wood *et al.*,
45 2015; Martin & Isaac, 2018). In resource-rich, predictable systems, growth rates tend to
46 be fast, leading to the assumption that crops may have evolved towards a rapid,
47 acquisitive trait profile (Aerts & Chapin, 1999; Craine, 2009; Milla *et al.*, 2015).
48 Despite the increasing number of studies addressing domestication from an eco-
49 evolutionary perspective or a trait-based approach (*e.g.* Blesh, 2018; Martin *et al.*, 2018;
50 Roucou *et al.*, 2018; Chacón-Labela *et al.*, 2019; Preece *et al.*, 2021), there is a lack of
51 comparative work assessing the evolution of growth dynamics in cultivation.

52 Crops are generally larger than their wild progenitors (Preece *et al.*, 2016; Milla
53 & Matesanz, 2017) and invest less in chemical and physical defences (Meyer *et al.*,
54 2012; Chen *et al.*, 2015; Simpson *et al.*, 2017). Increased resource allocation to
55 harvestable organs and earlier and more synchronous flowering and maturation
56 phenologies are typical of crops (Meyer & Purugganan, 2013). In addition, some
57 herbaceous crops have higher photosynthetic rates and leaf nitrogen concentrations than
58 their wild progenitors (Delgado-Baquerizo *et al.*, 2016; Roucou *et al.*, 2017; Nadal &
59 Flexas, 2018). However, the effects of domestication on growth rates appear to be
60 inconsistent or variable across crops. For example, modern cereals and other crop
61 species show no increase in growth rates during domestication (Gifford & Evans, 1981).
62 These results have recently been supported by other studies on a number of cereal and
63 legume species, which found no overall effect of domestication on growth rates (Preece
64 *et al.*, 2016; Simpson *et al.*, 2017).

65 Why previous work has reported idiosyncratic growth responses to
66 domestication may be due in part to the properties of the most common metric of
67 growth, relative growth rate (RGR), and the methods used to measure it. RGR, defined
68 as the rate of biomass increase relative to the biomass of the plant at the beginning of a
69 given time interval, is the product of a physiological (net assimilation rate, NAR), a
70 biomass allocation (leaf mass ratio, LMR), and a morphological component (specific
71 leaf area, SLA; Poorter, 1990). Given the mathematical relationships among these traits,

72 changes in RGR depend not only on variation in its components but also on how they
73 co-vary with each other (see Supporting Information Table S1 for a list of abbreviations
74 and a diagram of the mathematical relationships among growth traits). For example, a
75 change in NAR will result in a change in RGR unless NAR co-varies negatively with
76 LMR and/or SLA. Empirical studies of plant domestication often report changes in
77 physiology, biomass allocation, and leaf morphology in opposite directions and in
78 inconsistent ways. For example, SLA is lower in wild progenitors of several crops,
79 whereas leaf/stem fraction is higher compared to domesticates (Milla & Matesanz,
80 2017). Alternatively, leaf photosynthetic rate (*i.e.* an instantaneous proxy for NAR) is
81 higher in modern soybean, while SLA is lower than in its wild progenitors (Togashi &
82 Oikawa, 2021). Therefore, RGR might not differ between crops and their progenitors
83 because domestication has exerted opposite effects on its underlying components.

84 Another confounding effect may arise from the fact that RGR tends to decrease
85 as plants grow larger through increased investment in structural components, self-
86 shading and tissue turnover (Evans, 1972; Grime & Hunt, 1975). The larger size of
87 domesticated crops compared to their wild progenitors could therefore mask a faster
88 growth rate at a given size and have compromised the accuracy of previous work
89 (Turnbull *et al.*, 2008; Rose *et al.*, 2009). In addition, the methods used to measure
90 growth and the experimental settings differ between studies. Growth can be compared
91 between different experimental conditions, standardized by plant size or age, measured
92 once or over the entire plant ontogeny, and samples can be collected destructively or
93 non-destructively (Pommerening & Muszta, 2016). These diverse approaches to
94 measuring, calculating, and standardizing growth could contribute to the idiosyncratic
95 and crop-specific responses of growth to domestication.

96 The differential effects of domestication on plant growth could also be explained
97 by the heterogeneity of domestication processes (Purugganan & Fuller, 2009). Crops
98 with diverse origins and histories may have evolved in response to different
99 environmental pressures, human selection purposes, and over different time periods
100 (Hufford *et al.*, 2019). For example, latitude and temperature at the geographic origin of
101 each crop influence the response of leaf C, N, and P concentrations and ratios to
102 domestication (Delgado-Baquerizo *et al.*, 2016). In addition, the effects of
103 domestication on herbivore resistance vary depending on human selection, such that
104 crops selected for seed and fruit production show greater changes in herbivore
105 resistance and damage compared to leaf crops (Whitehead *et al.*, 2016). Finally, some of

106 the differences among crops in the effects of domestication on RGR could also be
107 explained by phylogenetic relationships among species, as RGR and its components
108 show phylogenetic signal (Kempel *et al.*, 2011; Atkinson *et al.*, 2016).

109 Our major crops were domesticated over the last c. 10,000 years, and modern
110 varieties are the product of the last c. 100 years of intensive breeding for high-yielding
111 crops. Here, we explore the extent to which domestication and modern plant breeding
112 have impacted RGR and its components in a wide range of herbaceous crops. We
113 conducted two experiments: an intensive one, in which the domestication history of
114 durum wheat was addressed in detail, and an extensive one, in which 18 crop species
115 were investigated more broadly. In both experiments, we grew multiple accessions of
116 wild progenitors, landraces, and improved cultivars of each crop under common
117 conditions and non-destructively measured their growth dynamics using a size-
118 standardized approach (Rees *et al.*, 2010). By comparing landraces with their wild
119 progenitors and with improved cultivars, we addressed the effects of domestication and
120 modern breeding, respectively. To investigate differences among taxa, we also collected
121 data on the origin and domestication history of each crop. Specifically, we asked: i)
122 How have domestication (wild progenitors *vs.* landraces) and modern plant breeding
123 (landraces *vs.* improved cultivars) impacted crop growth rates?; ii) Which components
124 of RGR have changed the most during crop evolution?; and iii) Can changes in growth
125 rates be explained by phylogeny, organ under selection, time in cultivation, and climate
126 at crop origin?

127

128 **Materials and Methods**

129 Two experiments were carried out to investigate how growth rates evolved after
130 domestication and modern plant breeding. The first experiment, called the *intensive*
131 *experiment*, examined in detail the variation in growth rate during the evolution of
132 durum wheat (*Triticum turgidum* L.). The second experiment, the *extensive experiment*,
133 explored growth rate changes after domestication and further improvement in a diverse
134 set of 18 crops. In both experiments, we estimated total mass, leaf mass, and leaf area at
135 different times during the vegetative growth period on individual plants. Using non-
136 linear growth models, we obtained the relative growth rate and its components at a
137 common size. Finally, we computed the magnitudes and directions of domestication and
138 improvement effects for all 19 crops and tested whether they varied as a function of the
139 origin and history of domestication and phylogenetic relationships among species.

140 *Study system*

141 Over the course of crop domestication and subsequent improvement, three main
142 domestication statuses can be distinguished: wild progenitors (W), the closest wild
143 relatives contributing to the gene pool of the crop; landraces (L), domesticated
144 genotypes that have not undergone intensive breeding in the last century and therefore
145 most closely represent early domesticates; and improved cultivars (I), genotypes from
146 more recent breeding programs (Abbo *et al.*, 2014). The identity of the putative wild
147 progenitor of each crop was taken from the Crop Origins database (Milla, 2020;
148 accessed 16 March 2021). Note that most crops are attributed a single wild progenitor,
149 but some have several wild progenitor taxa, either due to knowledge gaps, taxonomic
150 uncertainties, or hybrid origins. In addition, wild progenitors are thought to represent
151 the closest extant wild taxa, rather than the original ancestral populations of the
152 domesticated gene pool.

153 In both experiments, we grew several accessions belonging to the three
154 domestication statuses and covering a wide range of geographical origins (Fig. 1a). For
155 the *intensive experiment*, 32 accessions summarizing the domestication history of
156 durum wheat were selected. In particular, eight accessions of wild emmer wheat (*T.*
157 *turgidum* L. ssp. *dicoccoides* (Asch. & Graebn.) Thell.), eight accessions of early
158 landraces (domesticated emmer originating c. 10,000 years ago; *T. turgidum* L. ssp.
159 *dicoccum* (Schrank ex Schübl.) Thell.), eight accessions of late landraces (domesticated
160 durum originating c. 7,000 years ago; *T. turgidum* L. ssp. *durum* (Desf.) Husn.), and
161 eight accessions of modern wheat (*T. turgidum* L. ssp. *durum* (Desf.) Husn.) (Matsuoka,
162 2011; Roucou *et al.*, 2017). For the *extensive experiment*, we selected 18
163 phylogenetically diverse herbaceous species, mostly annuals, belonging to different
164 functional groups (Table 1). About 26% of them were cereals, 26% legumes, and 48%
165 forbs (*i.e.* herbaceous flowering plants that are neither graminoids nor legumes). These
166 species have C₃ photosynthesis, except for *Amaranthus*, *Pennisetum*, and *Sorghum*,
167 which have C₄ photosynthesis. For each species, we selected three wild accessions, two
168 landrace accessions, and two improved accessions, for a total of 126 accessions (see
169 Supporting Information Table S2 and Table S3 for accessions identifiers and seed
170 donors).

171 *Experimental procedures*

172 The *intensive* and *extensive experiments* were conducted in spring 2018 and 2019,
173 respectively. In both experiments, 12–35 seeds per accession were randomly selected
174 and individually sown on peat-filled flats. Those with thick and/or hard testas (mostly
175 legumes) were first scarified with a wire cutter to facilitate seed imbibition. About two
176 weeks after sowing, seedlings were transplanted into 3.6-l square pots (15 x 15 x 20 cm)
177 containing washed sand and slow-release fertilizer (5 g l⁻¹ Basacote Plus 6M; Compo,
178 Barcelona, Spain). The amount of fertilizer was set according to the manufacturer's
179 recommended dose for high nutrient availability conditions. Pot size was chosen to
180 allow unrestricted growth for the largest species following the recommendations of
181 Poorter *et al.* (2012). All pots were randomly placed on two contiguous benches in the
182 CULTIVE glasshouse of the Universidad Rey Juan Carlos (Madrid, Spain) and received
183 full sun (mean photosynthetically active radiation during light hours (10:00–20:00 h),
184 PAR ± SD = 892 ± 204 μmol m⁻² s⁻¹). Pots were watered regularly to ensure adequate
185 water supply, and air temperature (T) and relative humidity (RH) in the glasshouse were
186 recorded hourly (*intensive experiment*, mean T ± SD = 16.1 ± 8.1 °C, mean RH ± SD =
187 68 ± 22.6%; *extensive experiment*, mean T ± SD = 23.9 ± 5.2 °C, mean RH ± SD = 57.2
188 ± 15.5%).

189 Each experiment was divided into two groups: the focal and calibration plants.
190 In the focal plants, we measured several traits (see below) non-destructively at regular
191 intervals during the vegetative growth period. In the calibration plants, we measured the
192 same traits but also harvested individuals at regular intervals to obtain the dry mass of
193 leaves and the whole plant, and total leaf area. Calibration plants were used to develop
194 statistical models predicting the dry mass of leaves and plants and total leaf area from
195 the non-destructively measured traits. These models were then used to estimate the
196 masses and areas of focal plants at each monitoring date. Below we describe the
197 experimental procedures used, while the mathematical methods to estimate biomass
198 from the non-destructive traits are described in the Mass Estimations subsection of Data
199 Analyses.

200 For focal plants, six and three plants per accession were used in the *intensive* (N
201 = 192 focal plants) and *extensive* (N = 378 focal plants) *experiments*, respectively. Each
202 plant was monitored individually every three to ten days (8–12 times in total); more
203 frequently during early growth. During monitoring, the following non-destructive traits
204 were measured: plant height, canopy diameter, number of branches, number of leaves,
205 and length of the longest leaf. Basal stem diameter was also measured using a digital

206 calliper (0.01 mm resolution), but only in the *extensive experiment*, as wheat showed
207 little variation in this trait.

208 For calibration plants, six to nine destructive harvests were conducted during the
209 vegetative growth period. At each harvest, one plant per accession (*intensive*
210 *experiment*) or one plant per species and domestication status (either wild or
211 domesticate; *extensive experiment*) was harvested after measuring the non-destructive
212 traits. Harvested plants were washed and divided into stems, leaves, roots, leaf litter and
213 reproductive fraction (buds, flowers and fruits). Petioles and rachises were included in
214 the stem fraction. We scanned all leaf laminae at a 400-dpi resolution and measured the
215 total leaf area per plant using Photoshop software (CS6; Adobe Systems, Inc., San Jose,
216 CA, USA). Each plant fraction was dried at 60 °C for three days and weighed to the
217 nearest mg. Total mass (g) per plant was computed as the sum of all mass fractions at
218 each harvest date.

219 *Data compilation on phylogeny, origin and history of crops*

220 We built a phylogeny with our set of 19 crops (Fig. 1b). This phylogenetic tree was
221 pruned from the most comprehensive tree to date for angiosperms (Qian & Jin, 2016)
222 using the *drop.tip* function of the 'phytools' R package (Revell, 2012). *Abelmoschus*
223 *esculentus* was not in the reference tree, so its placement was taken as that of a sister
224 Malvaceae (*Hibiscus sabdariffa*), included in the reference tree. We also collected data
225 on time in cultivation (*i.e.* earliest record of exploitation in cultivation (ya)) and organ
226 under artificial selection (either fruits, leaves, or seeds) (Fig. 1c) from the Crop Origins
227 database (Milla, 2020; accessed 16 March 2021). The geographic location (latitude and
228 longitude) of each accession was also searched on the website of the corresponding
229 germplasm bank (Fig. 1d, Supporting Information Table S2 and Table S3). For each
230 location, past climatic data on temperature and precipitation regimes (Fig. 1e) were
231 obtained as follows. Considering the large climatic variability during the Holocene, time
232 in cultivation was divided into three periods according to available global paleoclimatic
233 models: early-Holocene (11,700–8,300 years BP), mid-Holocene (8,300–4,200 years
234 BP), and late-Holocene (4,200 years BP to present). Then, for crops originating in the
235 late-, mid-, or early-Holocene, we used their respective paleoclimatic model from the
236 PaleoClim database at ~5 km resolution (www.paleoclim.org; Brown *et al.*, 2018).
237 Models were read into R using the *raster* function of the 'raster' R package (Hijmans,
238 2021). Of the 19 bioclimatic variables provided, six were selected for the primary

239 analyses, including mean annual temperature, total annual precipitation, temperature
240 seasonality, precipitation seasonality, temperature of the coldest quarter, and
241 precipitation of the driest quarter. This selection aimed to cover annual trends,
242 seasonality, and extreme conditions. We calculated the arithmetic mean of the
243 bioclimatic variables for each location and species as a proxy for the climate at the
244 geographic origin of each crop.

245 *Data analyses*

246 Prior to data analysis, four dead individuals from the *intensive experiment* were
247 excluded from the data set, as was one individual from the *extensive experiment* that
248 was a clear outlier. All analyses were performed separately for each experiment in R
249 v.4.1.1. (R Core Team, 2021).

250

251 MASS ESTIMATIONS. Linear regressions were performed to obtain prediction
252 equations for total mass, leaf mass, and leaf area using data from the calibration plants.
253 Trait, mass, and area variables were \log_e -transformed. We fitted linear mixed-effects
254 models (LMM) to account for the factorial design of the experiments. Models were run
255 with the response variable (*i.e.* total plant mass, leaf mass, or leaf area), the non-
256 destructive trait measurements as fixed-effects predictors, and harvest date as a
257 covariate. The random effects structure varied between experiments. In the *intensive*
258 *experiment*, accession identity was included as a random effect over the intercept,
259 whereas in the *extensive experiment*, a combined variable between crop identity and
260 domestication status (either wild or domesticate) was used. To allow the relationship
261 between the response variable and predictors to vary across accessions in the *intensive*
262 *experiment* and between species and domestication status (combined variable) in the
263 *extensive experiment*, we included a random slope effect over the non-destructive trait
264 measurements.

265 For model selection, we looked for the optimal fixed structure by fitting models
266 with all combinations of fixed-effects predictors. The inclusion/exclusion of random
267 effects over the slopes depended on the presence/absence of certain predictors. Model
268 selection was based on the minimum AIC value. Selected models explained a great
269 proportion of the variation in the response variable (*intensive experiment*, mean $R^2_m \pm$
270 $SD = 0.98 \pm 0.004$, mean $R^2_c \pm SD = 0.99 \pm 0.004$; *extensive experiment*, mean $R^2_m \pm$
271 $SD = 0.86 \pm 0.040$, mean $R^2_c \pm SD = 0.99 \pm 0.002$) and were used to predict total mass,

272 leaf mass, and leaf area of focal plants (see Supporting Information Methods S1 for
273 more details). All models were run with the *lmer* function of the ‘lme4’ R package
274 (Bates *et al.*, 2015) with maximum likelihood (ML) estimation.

275

276 CURVE FITTING. We fitted logistic functions to the increase in mass of focal plants
277 over the vegetative growth period. Logistic functions are commonly used to describe
278 biological growth patterns and are appropriate when the data span the entire vegetative
279 lifespan (Paine *et al.*, 2012). Specifically, the three- and four-parameter logistic models
280 were tested and implemented with the *SSlogis* and *SSfpl* functions, respectively, in the
281 ‘nlme’ R package (Pinheiro *et al.*, 2021). We modelled $\log_e(\text{total mass})$ as a function of
282 time, adding plant identity as a random factor to all curve parameters (*i.e.* curve
283 parameters were allowed to vary among individuals). For both experiments, the most
284 parsimonious model based on minimizing AIC was the four-parameter logistic model
285 (Supporting Information Fig. S1) which modelled the variation of $\log_e(\text{total mass})$
286 ($\log_e M$) over time (t) as follows:

$$287 \quad \log_e M = A + \frac{B-A}{1 + e^{(xmid-t)/scal}} \quad (\text{Eqn 1})$$

288 where A , B , $xmid$, and $scal$ are the free parameters. Parameters A and B are the
289 minimum and maximum asymptotic $\log_e(\text{mass})$, respectively; $xmid$ is the time at which
290 $\log_e(\text{mass})$ is midway between the minimum and maximum asymptotes, and $1/scal$ is
291 the slope at the inflection point (Richards, 1959; R function *SSfpl* in Pinheiro *et al.*
292 (2020)). A separate curve was fitted for $\log_e(\text{leaf mass})$ and $\log_e(\text{leaf area})$ following the
293 same steps, and again the four-parameter logistic function provided the best fit.

294

295 RGR CALCULATION. To compare relative growth rates between plants at a common
296 size, we extracted the curve parameters from the fitted model and calculated a size-
297 standardized relative growth rate (sRGR) as:

$$298 \quad sRGR = \frac{(1/scal)(A - \log_e M_C)(B - \log_e M_C)}{(A - B)} \quad (\text{Eqn 2})$$

299 where $\log_e M_C$ is the common $\log_e(\text{mass})$ (Rees *et al.*, 2010). We used the median of the
300 mass distribution across all focal plants as the common size because all species occurred
301 at this size (0.555 g in the *intensive experiment* and 0.383 g in the *extensive experiment*).
302 Plant mass in the data set ranged from 0.006 g to 17.910 g in the *intensive experiment*
303 and from 0.001 g to 66.836 g in the *extensive experiment*. Because our size-standardized
304 metric focused on small plants, we supplemented it with metrics based on ontogenetic

305 criteria. In particular, we calculated the time-standardized RGR (tRGR) at two
306 ontogenetic stages: seedling and adult. Because the correlations among the three RGR
307 metrics were very high (Supporting Information Fig. S2), we used the common size
308 criteria for the analyses shown in the body of the paper to control for the widely
309 reported effects of plant size on RGR (Evans, 1972; Grime & Hunt, 1975; Rees *et al.*,
310 2010).

311

312 COMPONENTS OF RGR. Size-standardized RGR components were calculated from
313 sRGR following Rees *et al.* (2010). On logarithmic scales, sRGR can be expressed as
314 the sum of its components:

$$315 \quad \log_e(\text{sRGR}) = \log_e(\text{sNAR}) + \log_e(\text{sLMR}) + \log_e(\text{sSLA}) \quad (\text{Eqn 3})$$

316 These components are functions of total mass (M), leaf mass (ML), and leaf area (AL) as
317 follows:

$$318 \quad \log_e(\text{sRGR}) = \log_e\left(\frac{1}{AL_C} \frac{M_C - M_0}{t_C - t_0}\right) + \log_e\left(\frac{ML_C}{M_C}\right) + \log_e\left(\frac{AL_C}{ML_C}\right) \quad (\text{Eqn 4})$$

319 To calculate the contribution of each growth component to sRGR, we first
320 calculated the time (t_C) at which each focal plant reached the common mass (M_C) using
321 the four-parameter logistic equation (Eqn 1). This allowed us to calculate the
322 corresponding values of leaf mass (ML_C) and leaf area (AL_C) reached at that time from
323 their respective fitted curve. We used the estimates of ML_C and AL_C to calculate size-
324 standardized LMR (sLMR) and SLA (sSLA) applying equation 4. The value of NAR at
325 the common mass (sNAR) was then estimated as the ratio between sRGR and the
326 product of sLMR and sSLA (Eqn 3). For a detailed description of the calculation of
327 growth traits, see Supporting Information Methods S2.

328

329 RELATIVE IMPORTANCE OF RGR COMPONENTS. We decomposed the variation
330 in sRGR into its three components, following the protocol described by Rees *et al.*
331 (2010). Briefly, the variance of $\log_e(\text{sRGR})$ was equated to the sum of the variances and
332 covariances of the three \log_e -transformed sRGR components. The relative importance of
333 each component to sRGR variation was then calculated as the sum of the absolute
334 values of the component's variance and covariances divided by the sum of the absolute
335 values of all variances and covariances.

336

337 DOMESTICATION AND BREEDING EFFECT SIZE CALCULATIONS. Hedges' G
338 statistic was computed to measure the magnitude and direction of domestication and
339 improvement effects on sRGR and its components. For domestication, this was
340 calculated as the difference in means between landraces and wild progenitors of each
341 crop divided by the pooled and weighted standard deviation of the two groups (Hedges
342 *et al.*, 1999). In the *intensive experiment*, early and late landraces were considered
343 together to make the two experiments comparable. Effect sizes of modern breeding on
344 sRGR and its components were computed in the same way, but using improved
345 cultivars and landraces as reference groups. Hedges' G and its 95% confidence interval
346 were calculated using the *cohen.d* function of the 'effsize' R package (Torchiano,
347 2020).

348

349 STATISTICAL ANALYSES. To assess the impact of domestication and improvement
350 on sRGR, we ran linear mixed-effects models (LMMs) using the *lme* function in the
351 'nlme' R package (Pinheiro *et al.*, 2021). The models included sRGR as a response
352 variable and domestication status (with functional group and their interaction in the
353 *extensive experiment*) as fixed effects. Accession identity (nested within species in the
354 *extensive experiment*) was included as a random factor over the intercept. Loge-
355 transformations were used to meet the assumptions of the models. In the presence of
356 heteroscedasticity (checked with Levene's and Bartlett's test), the variance structure of
357 the data was modelled, with the best variance structure determined by comparing AIC
358 and standardized residual plots (Zuur *et al.*, 2009). Specifically, the variance structure of
359 the data was modelled using the *weights* option (VarIdent command) within the *lme*
360 function. The significance of the fixed factors of the models was estimated using the
361 *anova.lme* function with marginal (type III) sums of squares in the 'nlme' R package
362 (Pinheiro *et al.*, 2021). The amount of variance explained by the models was quantified
363 by calculating the marginal and conditional pseudo- R^2 with the *r.squaredGLMM*
364 function from the 'MuMIn' R package (Barton, 2020). Multiple comparison tests
365 among all levels and interactions of the fixed-effect factors were applied with false
366 discovery rate control, using the *glht* function in the 'multcomp' R package (Hothorn *et*
367 *al.*, 2008).

368 We investigated whether the effect sizes of domestication and modern breeding
369 on growth traits could be explained by phylogenetic relationships. We calculated the
370 phylogenetic signal in the effect sizes (Hedges' G) on growth traits (*i.e.* sRGR, sNAR,

371 sLMR, and sSLA) using Blomberg's K statistic (Blomberg *et al.*, 2003). K values near
372 zero indicate a lack of phylogenetic dependence, and values near one mean that closely
373 related species tend to have more similar values than species drawn randomly from the
374 tree. The significance of K values was tested using randomization tests with 1,000
375 permutations. To calculate K statistics and their significance we used the *phylosig*
376 function of the 'picante' R package (Kembel *et al.*, 2010).

377 We performed phylogenetic generalized least squares models (PGLSs) to assess
378 whether the effect sizes of domestication and modern breeding on sRGR and its
379 components were explained by the origin and history of crops. PGLSs incorporate
380 phylogenetic correlation structure in model residuals to account for phylogenetic
381 non-independence of species (Symonds & Blomberg, 2014). Domestication and
382 improvement effects on sRGR and its components were included as response variables,
383 while organ under artificial selection, time in cultivation and bioclimatic variables as
384 predictors. Models were run separately for each response and predictor variable.
385 Because C₃ and C₄ species differ in their climate optima, the models for climate effects
386 included the two-way interaction with photosynthetic pathway (Yamori *et al.*, 2014).
387 Prior to analyses, precipitation-related variables were log-transformed. PGLSs were
388 implemented using the *gls* function of the 'nlme' R package (Pinheiro *et al.*, 2021). To
389 account for heteroscedasticity, the variance structure of the data was modelled using the
390 weights option (VarIdent command) within the *gls* function. The significance of fixed
391 factors was estimated using the *anova* function with marginal (type III) sums of squares
392 in the 'nlme' R package (Pinheiro *et al.*, 2021). In models for bioclimatic variables,
393 significance levels were adjusted for false-discovery rates with the *p.adjust* function of
394 the 'stats' R package (R Core Team, 2021).

395

396 **Results**

397 *Evolution of RGR under cultivation*

398 sRGR varied considerably among crops, ranging from 0.10 for peanut to 0.27 g g⁻¹ d⁻¹
399 for amaranth (global mean \pm SD = 0.17 \pm 0.06). We found no consistent change in
400 sRGR after domestication and subsequent plant breeding in any of the experiments
401 (Table 2 and Table 3). The directions and effect sizes of domestication and
402 improvement varied among crops (Fig. 2). The magnitudes of domestication effects on
403 sRGR were significantly greater than those of subsequent plant breeding ($F_{1,95} = 15.95$,
404 $P < 0.001$; Fig. 2).

405 In the *extensive experiment*, sRGR did not consistently differ with
406 domestication status, but it differed significantly among functional groups (Fig. 3, Table
407 2). C₄ cereals had the highest and legumes the lowest average growth rates (0.24 and
408 0.11 g g⁻¹ d⁻¹, respectively). In the *intensive experiment*, sRGR increased in
409 domesticated plants when the entire domestication process was considered (*i.e.* wilds vs.
410 all landraces; $F_{1,22} = 7.08$, $P = 0.014$), but when the domestication process was split, we
411 found no effect of early or late domestication on sRGR in durum wheat (Fig. 4, Table
412 3). In both experiments, neither domestication nor modern breeding had consistent
413 effects on growth curve parameters ($P > 0.05$ for each of the four fitted parameters;
414 Supporting Information Fig. S3 and Fig. S4).

415 *Responses of RGR components to domestication and breeding*

416 None of the components of sRGR evolved consistently across species after
417 domestication and modern breeding, with the exception of sSLA, which increased in
418 improved cultivars (Table 2). Moreover, the high proportion of variance explained by
419 the random structure in the *intensive experiment* indicated high variability in responses
420 to domestication and improvement among the 32 durum wheat accessions (Table 3).

421 C₄ cereals and forbs had the highest sNAR and sLMR, respectively (Fig. 3,
422 Table 2). Moreover, the effect of domestication varied among functional groups for
423 sRGR and sLMR (interaction domestication status × functional group, Table 2). In the
424 *intensive experiment*, sNAR increased and sLMR decreased when the entire
425 domestication process was considered (*i.e.* wilds vs. all landraces; sNAR: $F_{1,22} = 6.81$, P
426 = 0.016, and sLMR: $F_{1,22} = 6.40$, $P = 0.019$; Fig. 4); however, when considered
427 separately, we found no effect of early and late domestication on any of the growth
428 traits of durum wheat (Fig. 4, Table 3).

429 sRGR was positively correlated with sNAR ($F_{1,394} = 118.6$, $P < 0.001$;
430 Supporting Information Fig. S5) and sSLA ($F_{1,394} = 8.9$, $P < 0.001$; Supporting
431 Information Fig. S5), whereas there was no relationship with sLMR ($F_{1,394} = 1.6$, $P =$
432 0.204). sNAR was by far the main driver of variation in sRGR in both experiments
433 (relative importance of NAR ± SD = 0.52 ± 0.02), followed by sLMR and sSLA
434 (relative importance of sLMR ± SD = 0.28 ± 0.15; and of sSLA ± SD = 0.20 ± 0.14;
435 Supporting Information Fig. S6).

436 *Factors influencing domestication and improvement effects*

437 Differences among crops in the effect sizes of domestication and improvement on
438 sRGR, sNAR, sLMR, and sSLA were partially explained by the organ under artificial
439 selection (Table 4). In crops selected for fruits, sNAR tended to increase after
440 domestication, whereas in those selected for leaves and seeds, sLMR and sSLA
441 increased (Fig. 5a, Table 4a). Only the increase in sLMR in leaf crops continued after
442 improvement, leading to an increase in sRGR (Table 4b).

443 The relationships between climate at crop origin and effect sizes of
444 domestication on growth traits were modulated by the photosynthetic pathway. For
445 mean annual temperature and temperature of the coldest quarter, C₃ species showed an
446 increase in sRGR and sNAR, a decrease in sLMR, and no effect on sSLA, while C₄
447 species showed the inverse relationships (Fig. 5b, Table 4a). Temperature seasonality
448 showed the opposite patterns for the same traits (Table 4a). Precipitation-related
449 variables hardly explained the effect sizes of domestication on sRGR components
450 (Table 4a, Supporting Information Table S4). Variation in effect sizes of modern
451 breeding among crops was statistically explained by some bioclimatic variables, such as
452 temperature seasonality, in the same direction as domestication effects on C₃ species
453 (Table 4b, Supporting Information Table S5).

454 Time in cultivation did not significantly explain the variation in effect sizes of
455 domestication and improvement on sRGR and its components (Table 4). Effect sizes on
456 sSLA showed a significant phylogenetic signal, suggesting that changes in sSLA during
457 domestication tended to be similar in magnitude and direction in phylogenetically
458 related species (Table 4a). The size and magnitude of modern breeding effects did not
459 show phylogenetic signals (Table 4b).

460

461 **Discussion**

462 In this study, we examined the evolution of RGR and its components during
463 domestication and modern plant breeding in a wide range of herbaceous crops. We
464 found that crops responded differently to domestication, suggesting that high yields,
465 typical of agricultural plants, were not consistently accompanied by an increase in
466 growth rates. These differential responses of RGR and its components to domestication
467 and further plant breeding were dependent on the phylogeny, organ under selection, and
468 climate at the geographic origin of each crop. Moreover, domestication affected RGR
469 components in opposite directions, resulting in no or smaller net effects on RGR. Thus,

470 the evolution of RGR was also constrained by trade-offs between its underlying
471 components.

472 *Evolution of growth rates under cultivation*

473 We found that size-standardized RGR changed from wild progenitors to landraces to
474 improved cultivars in idiosyncratic ways, *i.e.* the direction and magnitude of the effects
475 of domestication and modern breeding differed among crops. Of the 19 crops studied,
476 six had a negative effect size, four had a positive effect size, and nine showed no effect
477 (based on 95% CIs, Fig. 3). This species-specific response of RGR is consistent with
478 previous studies that focused on individual crops. For example, RGR increased with
479 domestication in tomato (Conesa *et al.*, 2017), decreased in rice (Cook & Evans, 1983)
480 and barley (Chapin *et al.*, 1989), but showed no effect in wheat (Evans & Dunstone,
481 1970), maize (Duncan & Hesketh, 1968) and millet (Evans & Bush, 1985). These
482 studies were conducted under dissimilar conditions and with different methodologies.
483 However, even when comparisons are made between plants of the same size and under
484 the same conditions, the effects of domestication and improvement on growth rates vary
485 widely among crops (Preece *et al.*, 2015; Simpson *et al.*, 2017). Our extensive
486 screening, together with previous case studies, therefore supports the scenario of an
487 inconsistent pattern of growth rate evolution during domestication and modern plant
488 breeding.

489 The idiosyncratic changes in growth rates across crops contrast with the widely
490 reported decline in defence investment during domestication and subsequent plant
491 breeding (Rosenthal & Dirzo, 1997; Gepts, 2004; Meyer *et al.*, 2012; Chen *et al.*, 2015;
492 but see Simpson *et al.*, 2017; Whitehead *et al.*, 2017). Plant defence theory predicts a
493 trade-off between growth and defence because secondary metabolism and physical plant
494 structures are physiologically costly (Coley *et al.*, 1985). Trade-offs between growth
495 and defence have been particularly well studied in natural ecosystems (Endara & Coley,
496 2011; Lind *et al.*, 2013), but have not been consistently supported in crops (Kempel *et al.*
497 *et al.*, 2011; Turcotte *et al.*, 2014; Simpson *et al.*, 2017; Moreira *et al.*, 2018). In wheat,
498 barley, and maize, for example, silicon-based defences decreased after domestication,
499 but growth rates did not (Simpson *et al.*, 2017). We speculate that reduced defence traits
500 in crops are the result of early and direct selection for palatable and fast-growing wild
501 progenitors and early domesticates, rather than the result of later selection through
502 trade-offs with growth. Our results therefore raise the question of whether wild

503 progenitors have faster growth rates and lower defensive traits than other wild species
504 that have not been selected for agricultural purposes.

505 In this study, sNAR was the main driver of variation in sRGR, which is
506 consistent with previous work (Shipley, 2006; Cunniff *et al.*, 2014; Atkinson *et al.*,
507 2016; but see Lambers & Poorter, 1992 and Wilson *et al.*, 1999 for contrasting results).
508 However, the magnitude of change in sNAR during crop evolution was less than in
509 sSLA and sLMR. Previous literature suggests that selection for higher yields has not
510 altered crop physiology as much as allocation patterns and morphology (Gifford &
511 Evans, 1981; Gifford *et al.*, 1984; Richards, 2000; Driever *et al.*, 2014; Sinclair *et al.*,
512 2019). For example, traits such as high harvest index (*i.e.* the ratio of yield to
513 aboveground mass), lower allocation to chaff and pods, lower root mass fraction, or
514 larger leaves and stems are more often claimed to drive yield (Evans & Dunstone, 1970;
515 Donald & Hamblin, 1976; Sinclair, 1998; Waines & Ehdaie, 2007). In addition, other
516 traits typically associated with the domestication syndrome, such as large initial and
517 final body size, earlier reproduction, and lower branching have also contributed to
518 higher yields (Preece *et al.*, 2015; Holland *et al.*, 2019; Houshmandfar *et al.*, 2020). In
519 our study, the strong physiological basis of sRGR supports the notion that physiology
520 has not consistently changed over the course of evolution under cultivation and is
521 therefore not a major driver of variation in crop yield.

522 It is noteworthy that the changes in growth traits were greater after
523 domestication than in later plant breeding. In fact, the magnitude of domestication
524 effects was *c.* 74% greater than that of further breeding. This is consistent with other
525 studies. For example, wild progenitors and landraces of wheat and maize show higher
526 phenotypic diversity than modern cultivars for root or kernel traits (Flint-Garcia *et al.*,
527 2009; Roucou *et al.*, 2017). One explanation for these results is that the domestication
528 process, when broadly defined, *i.e.* from the initial domestication of wild progenitors to
529 their spreading and diversification into landraces, spanned longer periods of time,
530 whereas modern breeding practises began about a century ago (Faris, 2014). Moreover,
531 the current study compared landraces with wild progenitors from diverse geographical
532 regions, where natural selection pressures might be different. On the other hand, modern
533 cultivars are derived from a limited number of landraces and intensive artificial
534 selection for specific traits, which in turn has reduced phenotypic and genetic diversity
535 (Tanksley & McCouch, 1997; Meyer & Purugganan, 2013). Therefore, wild progenitors

536 and/or landraces harbour a greater diversity in growth traits compared to modern
537 cultivars, which could lead to stronger effect sizes in the domestication process.

538 *Factors explaining variation in domestication effects*

539 Interestingly, the effect sizes of domestication on sRGR components were partially
540 explained by the organ under selection. Specifically, fruit crops showed the highest
541 domestication effects on sNAR, whereas leaf and seed crops showed larger effects on
542 sSLA and sLMR. We are unaware of any previous studies reporting differential growth
543 responses to domestication depending on which organ was primarily selected.
544 Investment in fleshy fruits can be physiologically more costly than in leaves and seeds
545 because they are typically photosynthetic sinks that require substantial amounts of
546 carbon, nutrients, and water (Coombe, 1976). As a result, yields of fruit crops are often
547 more limited by source strength (*i.e.* photosynthesis) rather than sink capacity (Li *et al.*,
548 2015), in contrast to what occurs in seed crops such as wheat, maize and soybean
549 (Borrás *et al.*, 2004). Other physiological traits such as photosynthetic rate, stomatal
550 conductance, and water and nutrient use efficiency may have accompanied the increase
551 in sNAR during domestication of fruit crops; however, more evidence is needed to test
552 this hypothesis. Furthermore, these results are in line with the idea that if sRGR does
553 not differ between crops and their progenitors, this could be because domestication had
554 opposite effects on the underlying components of RGR.

555 When C₃ and C₄ species were looked at separately, we found significant growth
556 differences between crops from different geographic origins. After domestication,
557 sRGR and sNAR tended to decrease with temperature and increase with seasonality in
558 wild C₃ progenitors, whereas the opposite trend was observed in C₄ species (Supporting
559 Information Fig. S7). For C₃ species, variation in growth rates with temperature is
560 congruent with adaptation to the length of the growing season (T-plant physiology
561 hypothesis; Reich & Oleksyn, 2004). Thus, previous studies showed faster growth rates
562 in populations from regions with shorter growing seasons (either at high altitudes or
563 high latitudes), both in crop progenitors (Alexander, 2010) and wild species (Weber &
564 Schmid, 1998; Ryser & Aeschlimann, 1999; Milla *et al.*, 2009; but see Li *et al.*, 1998).
565 In contrast, for C₄ species, the positive relationship between sRGR and sNAR with
566 temperature is likely a result of the adaptive advantage that C₄ photosynthesis provides
567 in regions with higher photorespiration and potential evapotranspiration losses
568 (Watcharamongkol *et al.*, 2018). In our study, despite the low number of C₄ crops, we

569 found that climate adaptations of wild progenitors modulated the growth response to
570 domestication. The effect of domestication (*i.e.* landraces *vs.* progenitors) tended to be
571 positive when wild C₃ progenitors came from regions with higher temperatures or lower
572 seasonality, whereas C₄ showed the opposite trend. Similarly, Delgado-Baquerizo *et al.*
573 (2016) found significant relationships between temperature at crop origin and changes
574 during domestication in other growth-related traits such as leaf N, C, and P
575 concentrations. Therefore, we speculate that wild C₃ and C₄ progenitors from regions
576 with low and high temperatures (or high and low seasonal variation), respectively,
577 already grew fast enough to meet agricultural needs or had reached their physiological
578 limits and thus experienced little or even negative changes in plant growth during
579 domestication. Exploring the specific adaptations of wild progenitors to climate could
580 have important implications for our understanding of current crop performance and for
581 future breeding and conservation programmes.

582 Variation in domestication effect sizes among crops was phylogenetically
583 constrained only for sSLA, suggesting that phylogeny can partially explain the diversity
584 of growth responses. Despite the fact that most growth traits showed significant effects
585 of functional group (*i.e.* a factor largely related to phylogeny), common selection
586 pressures during domestication and improvement may have favoured convergence in the
587 direction and magnitude of growth traits changes among species in distant clades
588 (Pickersgill, 2018). Finally, time in cultivation did not explain the differences in effect
589 sizes of domestication and modern plant breeding on sRGR and its components. This
590 result was also found for root traits in a number of crops (Martín-Robles *et al.*, 2018). It
591 has been suggested that evolutionary rates are similar to those measured for wild species
592 (Purugganan & Fuller, 2011), or that they vary over time, both accelerating and
593 decelerating depending on the prevailing selective force (Abbo & Gopher, 2020). For
594 example, the spreading to new environments and intense directional selection have far
595 greater potential for rapid evolutionary change than mutation or unconscious selection
596 (Zeder, 2017). Therefore, time in cultivation may not be as relevant as other factors in
597 explaining evolutionary changes in crop growth.

598 In conclusion, our comprehensive survey suggests that growth rates have not
599 responded consistently to domestication and modern plant breeding, in line with
600 previous case studies. Crop-specific responses of growth to domestication and
601 improvement depended on artificial selection purposes and climate at crop origin, and
602 were constrained by correlations between traits rather than phylogenetic position. Thus,

603 in fruit crops, artificial selection changed the physiological component of growth,
604 whereas in leaf and seed crops it changed the components related to allocation and leaf
605 morphology. The specific adaptations of wild progenitors to the climate at their origins
606 further modulated the evolution of growth rates. Overall, our study sheds light on the
607 factors underlying the diversity of crop responses to evolution under cultivation.
608 Research in this area should further explore the causes and consequences of this
609 diversity, given the importance of growth rates to crop performance.

610

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621

622 **Author Contribution**

623 AG-F and RM designed the study. AG-F, CI, GG, and JP collected the data. AG-F, MR,
624 and RM analysed the data. AG-F wrote a first draft of the paper, and CPO, JP, MR, and
625 RM contributed to further revisions. All authors read and approved the final version.

626

627 **Data Availability**

628 The raw data and R codes supporting the findings of this study are available at
629 <https://doi.org/10.6084/m9.figshare.16818046.v2>.

630

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637 **References**

- 638 **Abbo S, Gopher A. 2020.** Plant domestication in the Neolithic Near East: the humans-
639 plants liaison. *Quaternary Science Reviews* **242**: 106412.
- 640 **Abbo S, Van-Oss RP, Gopher A, Saranga Y, Ofner I, Peleg Z. 2014.** Plant
641 domestication versus crop evolution: a conceptual framework for cereals and grain
642 legumes. *Trends in Plant Science* **19**: 351–360.
- 643 **Aerts R, Chapin FS. 1999.** The mineral nutrition of wild plants revisited: a re-
644 evaluation. *Advances in Ecological Research* **30**: 1–67.
- 645 **Alexander JM. 2010.** Genetic differences in the elevational limits of native and
646 introduced *Lactuca serriola* populations. *Journal of Biogeography* **37**: 1951–1961.
- 647 **Atkinson RRL, Mockford EJ, Bennett C, Christin PA, Spriggs EL, Freckleton RP,
648 Thompson K, Rees M, Osborne CP. 2016.** C₄ photosynthesis boosts growth by
649 altering physiology, allocation and size. *Nature Plants* **2**: 1–5.
- 650 **Barton K. 2020.** Mu-MIn: multi-model inference. [https://cran.r-](https://cran.r-project.org/package=MuMIn)
651 [project.org/package=MuMIn](https://cran.r-project.org/package=MuMIn).
- 652 **Bates D, Mächler M, Bolker BM, Walker SC. 2015.** Fitting linear mixed-effects
653 models using lme4. *Journal of Statistical Software* **67**: 1–48.
- 654 **Blesh J. 2017.** Functional traits in cover crop mixtures: biological nitrogen fixation and
655 multifunctionality. *Journal of Applied Ecology* **55**: 38–48.
- 656 **Blomberg SP, Garland T, Ives AR. 2003.** Testing for phylogenetic signal in
657 comparative data: behavioral traits are more labile. *Evolution* **57**: 717–745.
- 658 **Borrás L, Slafer GA, Otegui ME. 2004.** Seed dry weight response to source-sink
659 manipulations in wheat, maize and soybean: a quantitative reappraisal. *Field Crops
660 Research* **86**: 131–146.
- 661 **Brown JL, Hill DJ, Dolan AM, Carnaval AC, Haywood AM. 2018.** PaleoClim, high
662 spatial resolution paleoclimate surfaces for global land areas. *Scientific Data* **5**: 1–9.
- 663 **Chacón-Labela J, García Palacios P, Matesanz S, Schöb C, Milla R. 2019.** Plant
664 domestication disrupts biodiversity effects across major crop types. *Ecology Letters* **22**:
665 1472–1482.
- 666 **Chapin FS, Groves RH, Evans LT. 1989.** Physiological determinants of growth rate in
667 response to phosphorus supply in wild and cultivated *Hordeum* species. *Oecologia* **79**:
668 96–105.
- 669 **Chen YH, Gols R, Benrey B. 2015.** Crop domestication and its impact on naturally
670 selected trophic interactions. *Annual Review of Entomology* **60**: 35–58.
- 671 **Coley PD, Bryant JP, Chapin FS. 1985.** Resource availability and plant antiherbivore
672 defense. *Plant Ecology* **230**: 895–899.
- 673 **Conesa M, Muir CD, Roldán EJ, Molins A, Perdomo JA, Galmés J. 2017.** Growth
674 capacity in wild tomatoes and relatives correlates with original climate in arid and semi-

- 675 arid species. *Environmental and Experimental Botany* **141**: 181–190.
- 676 **Cook MG, Evans LT. 1983.** Some physiological aspects of the domestication and
677 improvement of rice (*Oryza* spp.). *Field Crops Research* **6**: 219–238.
- 678 **Coombe BG. 1976.** The development of fleshy fruits. *Annual Review of Plant*
679 *Physiology* **27**: 207–228.
- 680 **Craine JM. 2009.** *Resource strategies of wild plants*. New Jersey, USA: Princeton
681 University Press.
- 682 **Cunniff J, Wilkinson S, Charles M, Jones G, Rees M, Osborne CP. 2014.**
683 Functional traits differ between cereal crop progenitors and other wild grasses gathered
684 in the neolithic Fertile Crescent. *PLoS ONE* **9**: e87586.
- 685 **Delgado-Baquerizo M, Reich PB, García-Palacios P, Milla R. 2016.** Biogeographic
686 bases for a shift in crop C:N:P stoichiometries during domestication. *Ecology Letters*
687 **19**: 564–575.
- 688 **Doebley JF, Gaut BS, Smith BD. 2006.** The molecular genetics of crop domestication.
689 *Cell* **127**: 1309–1321.
- 690 **Donald CM, Hamblin J. 1976.** The biological yield and harvest index of cereals as
691 agronomic and plant breeding criteria. *Advances in Agronomy* **28**: 361–405.
- 692 **Driever SM, Lawson T, Andralojc PJ, Raines CA, Parry MAJ. 2014.** Natural
693 variation in photosynthetic capacity, growth, and yield in 64 field-grown wheat
694 genotypes. *Journal of Experimental Botany* **65**: 4959–4973.
- 695 **Duncan WG, Hesketh JD. 1968.** Net photosynthetic rates, relative leaf growth rates,
696 and leaf numbers of 22 races of maize grown at eight temperatures. *Crop Science* **8**:
697 670–674.
- 698 **Endara MJ, Coley PD. 2011.** The resource availability hypothesis revisited: a meta-
699 analysis. *Functional Ecology* **25**: 389–398.
- 700 **Evans GC. 1972.** *The quantitative analysis of plant growth*. California, USA:
701 University of California Press.
- 702 **Evans LT. 1993.** *Crop evolution, adaptation and yield*. Cambridge, UK: Cambridge
703 University Press.
- 704 **Evans LT, Bush MG. 1985.** Growth and development of channel millet (*Echinochloa*
705 *turneriana*) in relation to its potential as a crop plant and compared with other
706 *Echinochloa* millets, rice and wheat. *Field Crops Research* **12**: 295–317.
- 707 **Evans LT, Dunstone RL. 1970.** Some physiological aspects of evolution in wheat.
708 *Australian Journal of Biological Sciences* **23**: 725–741.
- 709 **Faris JD. 2014.** Wheat domestication: key to agricultural revolutions past and future.
710 In: Tuberosa R, Graner A, Frison E, eds. *Genomics of plant genetic resources*.
711 Dordrech, The Netherlands: Springer, 439–464.
- 712 **Flint-Garcia SA, Bodnar AL, Scott MP. 2009.** Wide variability in kernel
713 composition, seed characteristics, and zein profiles among diverse maize inbreds,

- 714 landraces, and teosinte. *Theoretical and Applied Genetics* **119**: 1129–1142.
- 715 **Gepts P. 2004.** Crop domestication as a long-term selection experiment. *Plant Breeding*
716 *Reviews* **24**: 1–44.
- 717 **Gifford RM, Evans LT. 1981.** Photosynthesis, carbon partitioning, and yield. *Annual*
718 *Review of Plant Physiology* **32**: 485–509.
- 719 **Gifford RM, Thorne JH, Hitz WD, Giaquinta RT. 1984.** Crop productivity and
720 photoassimilate partitioning. *Science* **225**: 801–808.
- 721 **Grime JP, Hunt R. 1975.** Relative growth-rate: its range and adaptive significance in a
722 local flora. *Journal of Ecology* **63**: 393–422.
- 723 **Hedges L V., Gurevitch J, Curtis PS. 1999.** The meta-analysis of response ratios in
724 experimental ecology. *Ecology* **80**: 1150–1156.
- 725 **Hijmans RJ. 2021.** raster: geographic data analysis and modeling. [https://cran.r-](https://cran.r-project.org/package=raster)
726 [project.org/package=raster](https://cran.r-project.org/package=raster).
- 727 **Holland BL, Monk NAM, Clayton RH, Osborne CP. 2019.** A theoretical analysis of
728 how plant growth is limited by carbon allocation strategies and respiration. *In Silico*
729 *Plants* **1**: diz004.
- 730 **Hothorn T, Bretz F, Westfall P. 2008.** Simultaneous inference in general parametric
731 models. *Biometrical Journal* **50**: 346–363.
- 732 **Houshmandfar A, Ota N, O’Leary GJ, Zheng B, Chen Y, Tausz-Posch S,**
733 **Fitzgerald GJ, Richards R, Rebetzke GJ, Tausz M. 2020.** A reduced-tillering trait
734 shows small but important yield gains in dryland wheat production. *Global Change*
735 *Biology* **26**: 4056–4067.
- 736 **Hufford MB, Berny Mier y Teran JC, Gepts P. 2019.** Crop biodiversity: an
737 unfinished magnum opus of nature. *Annual Review of Plant Biology* **70**: 727–751.
- 738 **Kembel S, Cowan P, Helmus M, Cornwell W, Morlon H, Ackerly D, Blomberg S,**
739 **Webb C. 2010.** Picante: R tools for integrating phylogenies and ecology.
740 *Bioinformatics* **26**: 1463–1464.
- 741 **Kempel A, Schädler M, Chrobock T, Fischer M, Van Kleunen M. 2011.** Trade-offs
742 associated with constitutive and induced plant resistance against herbivory. *Proceedings*
743 *of the National Academy of Sciences of the United States of America* **108**: 5685–5689.
- 744 **Lambers H, Poorter H. 1992.** Inherent variation in growth rate between higher plants:
745 a search for physiological causes and ecological consequences. *Advances in Ecological*
746 *Research* **23**: 187–261.
- 747 **Li T, Heuvelink E, Marcelis LFM. 2015.** Quantifying the source-sink balance and
748 carbohydrate content in three tomato cultivars. *Frontiers in Plant Science* **6**: 416.
- 749 **Li B, Suzuki JI, Hara T. 1998.** Latitudinal variation in plant size and relative growth
750 rate in *Arabidopsis thaliana*. *Oecologia* **115**: 293–301.
- 751 **Lind EM, Borer E, Seabloom E, Adler P, Bakker JD, Blumenthal DM, Crawley M,**
752 **Davies K, Firn J, Gruner DS, et al. 2013.** Life-history constraints in grassland plant

- 753 species: a growth-defence trade-off is the norm. *Ecology Letters* **16**: 513–521.
- 754 **Martín-Robles N, Morente-López J, Freschet GT, Poorter H, Roumet C, Milla R.**
755 **2018.** Root traits of herbaceous crops: pre-adaptation to cultivation or evolution under
756 domestication? *Functional Ecology* **33**: 273–285.
- 757 **Martin AR, Hale CE, Cerabolini BEL, Cornelissen JHC, Craine J, Gough WA,**
758 **Kattge J, Tirona CKF. 2018.** Inter- and intraspecific variation in leaf economic traits
759 in wheat and maize. *AoB PLANTS* **10**: ply006.
- 760 **Martin AR, Isaac ME. 2018.** Functional traits in agroecology: advancing description
761 and prediction in agroecosystems. *Journal of Applied Ecology* **55**: 5–11.
- 762 **Matsuoka Y. 2011.** Evolution of polyploid *Triticum* wheats under cultivation: the role
763 of domestication, natural hybridization and allopolyploid speciation in their
764 diversification. *Plant and Cell Physiology* **52**: 750–764.
- 765 **Meyer RS, DuVal AE, Jensen HR. 2012.** Patterns and processes in crop
766 domestication: an historical review and quantitative analysis of 203 global food crops.
767 *New Phytologist* **196**: 29–48.
- 768 **Meyer RS, Purugganan MD. 2013.** Evolution of crop species: genetics of
769 domestication and diversification. *Nature Reviews Genetics* **14**: 840–852.
- 770 **Milla R. 2020.** Crop Origins and Phylo Food: a database and a phylogenetic tree to
771 stimulate comparative analyses on the origins of food crops. *Global Ecology and*
772 *Biogeography* **29**: 606–614.
- 773 **Milla R, Giménez-Benavides L, Escudero A, Reich PB. 2009.** Intra- and interspecific
774 performance in growth and reproduction increase with altitude: a case study with two
775 *Saxifraga* species from northern Spain. *Functional Ecology* **23**: 111–118.
- 776 **Milla R, Matesanz S. 2017.** Growing larger with domestication: a matter of
777 physiology, morphology or allocation? *Plant Biology* **19**: 475–483.
- 778 **Milla R, Osborne CP, Turcotte MM, Violle C. 2015.** Plant domestication through an
779 ecological lens. *Trends in Ecology and Evolution* **30**: 463–469.
- 780 **Moreira X, Abdala-Roberts L, Gols R, Francisco M. 2018.** Plant domestication
781 decreases both constitutive and induced chemical defences by direct selection against
782 defensive traits. *Scientific Reports* **8**: 12678.
- 783 **Nadal M, Flexas J. 2018.** Variation in photosynthetic characteristics with growth form
784 in a water-limited scenario: implications for assimilation rates and water use efficiency
785 in crops. *Agricultural Water Management* **216**: 457–472.
- 786 **Paine CET, Marthews TR, Vogt DR, Purves D, Rees M, Hector A, Turnbull LA.**
787 **2012.** How to fit nonlinear plant growth models and calculate growth rates: an update
788 for ecologists. *Methods in Ecology and Evolution* **3**: 245–256.
- 789 **Pickersgill B. 2018.** Parallel vs. convergent evolution in domestication and
790 diversification of crops in the Americas. *Frontiers in Ecology and Evolution* **6**: 1–15.
- 791 **Pinheiro J, Bates D, DebRoy S, Sarkar D, Team RC. 2021.** nlme: linear and

- 792 nonlinear mixed effects models. <https://cran.r-project.org/package=nlme>.
- 793 **Pommerening A, Muszta A. 2016.** Relative plant growth revisited: towards a
794 mathematical standardisation of separate approaches. *Ecological Modelling* **320**: 383–
795 392.
- 796 **Poorter H. 1990.** Interspecific variation in relative growth rate: on ecological causes
797 and physiological consequences. In: Lambers H, ed. Causes and consequences of
798 variation on growth rate and productivity of higher plants. The Hague, The Netherlands:
799 SPB Academic Publishing, 45–68.
- 800 **Poorter H, Bühler J, Van Dusschoten D, Climent J, Postma JA. 2012.** Pot size
801 matters: a meta-analysis of the effects of rooting volume on plant growth. *Functional*
802 *Plant Biology* **39**: 839–850.
- 803 **Preece C, Jones G, Rees M, Osborne CP. 2021.** Fertile Crescent crop progenitors
804 gained a competitive advantage from large seedlings. *Ecology and Evolution* **11**: 3300–
805 3312.
- 806 **Preece C, Livarda A, Christin PA, Wallace M, Martin G, Charles M, Jones G,**
807 **Rees M, Osborne CP. 2016.** How did the domestication of Fertile Crescent grain crops
808 increase their yields? *Functional Ecology* **31**: 387–397.
- 809 **Preece C, Livarda A, Wallace M, Martin G, Charles M, Christin PA, Jones G,**
810 **Rees M, Osborne CP. 2015.** Were Fertile Crescent crop progenitors higher yielding
811 than other wild species that were never domesticated? *New Phytologist* **207**: 905–913.
- 812 **Purugganan MD, Fuller DQ. 2009.** The nature of selection during plant
813 domestication. *Nature* **457**: 843–848.
- 814 **Purugganan MD, Fuller DQ. 2011.** Archaeological data reveal slow rates of evolution
815 during plant domestication. *Evolution* **65**: 171–183.
- 816 **Qian H, Jin Y. 2016.** An updated megaphylogeny of plants, a tool for generating plant
817 phylogenies and an analysis of phylogenetic community structure. *Journal of Plant*
818 *Ecology* **9**: 233–239.
- 819 **R Core Team. 2021.** R: a language and environment for statistical computing.
820 <https://www.r-project.org/>.
- 821 **Rees M, Osborne CP, Woodward FI, Hulme SP, Turnbull LA, Taylor SH. 2010.**
822 Partitioning the components of relative growth rate: how important is plant size
823 variation? *The American Naturalist* **176**: E152–E161.
- 824 **Reich PB, Oleksyn J. 2004.** Global patterns of plant leaf N and P in relation to
825 temperature and latitude. *Proceedings of the National Academy of Sciences of the*
826 *United States of America* **101**: 11001–11006.
- 827 **Revell LJ. 2012.** phytools: an R package for phylogenetic comparative biology (and
828 other things). *Methods in Ecology and Evolution* **3**: 217–223.
- 829 **Richards FJ. 1959.** A flexible growth function for empirical use. *Journal of*
830 *Experimental Botany* **10**: 290–301.

- 831 **Richards RA. 2000.** Selectable traits to increase crop photosynthesis and yield of grain
832 crops. *Journal of Experimental Botany* **51**: 447–458.
- 833 **Rose KE, Atkinson RL, Turnbull LA, Rees M. 2009.** The costs and benefits of fast
834 living. *Ecology Letters* **12**: 1379–1384.
- 835 **Rosenthal JP, Dirzo R. 1997.** Effects of life history, domestication and agronomic
836 selection on plant defence against insects: evidence from maizes and wild relatives.
837 *Evolutionary Ecology* **11**: 337–355.
- 838 **Roucou A, Violle C, Fort F, Roumet P, Ecarnot M, Vile D. 2017.** Shifts in plant
839 functional strategies over the course of wheat domestication. *Journal of Applied*
840 *Ecology* **55**: 25–37.
- 841 **Ryser P, Aeschlimann U. 1999.** Proportional dry-mass content as an underlying trait
842 for the variation in relative growth rate among 22 Eurasian populations of *Dactylis*
843 *glomerata* s.l. *Functional Ecology* **13**: 473–482.
- 844 **Shipley B. 2006.** Net assimilation rate, specific leaf area and leaf mass ratio: which is
845 most closely correlated with relative growth rate? A meta-analysis. *Functional Ecology*
846 **20**: 565–574.
- 847 **Simpson KJ, Wade RN, Rees M, Osborne CP, Hartley SE. 2017.** Still armed after
848 domestication? Impacts of domestication and agronomic selection on silicon defences in
849 cereals. *Functional Ecology* **31**: 2108–2117.
- 850 **Sinclair TR. 1998.** Historical changes in harvest index and crop nitrogen accumulation.
851 *Crop Science* **38**: 638–643.
- 852 **Sinclair TR, Rufty TW, Lewis RS. 2019.** Increasing photosynthesis: unlikely solution
853 for world food problem. *Trends in Plant Science* **24**: 1032–1039.
- 854 **Symonds MR, Blomberg SP. 2014.** A primer on phylogenetic generalised least
855 squares. In: Garamszegi LZ, ed. *Modern phylogenetic comparative methods and their*
856 *application in evolutionary biology*. Heidelberg, Germany: Springer, 105–130.
- 857 **Tanksley SD, McCouch SR. 1997.** Seed banks and molecular maps: unlocking genetic
858 potential from the wild. *Science* **277**: 1063–1066.
- 859 **Togashi A, Oikawa S. 2021.** Leaf productivity and persistence have been improved
860 during soybean (*Glycine max*) domestication and evolution. *Journal of Plant Research*
861 **134**: 223–233.
- 862 **Torchiano M. 2020.** effsize: efficient effect size computation. [https://cran.r-](https://cran.r-project.org/package=effsize)
863 [project.org/package=effsize](https://cran.r-project.org/package=effsize).
- 864 **Turcotte MM, Turley NE, Johnson MTJ. 2014.** The impact of domestication on
865 resistance to two generalist herbivores across 29 independent domestication events. *New*
866 *Phytologist* **204**: 671–681.
- 867 **Turnbull LA, Paul-Victor C, Schmid B, Purves DW. 2008.** Growth rates, seed size,
868 and physiology: do small-seeded species really grow faster? *Ecology* **89**: 1352–1363.
- 869 **Waines JG, Ehdaie B. 2007.** Domestication and crop physiology: roots of green-

- 870 revolution wheat. *Annals of Botany* **100**: 991–998.
- 871 **Watcharamongkol T, Christin PA, Osborne CP. 2018.** C₄ photosynthesis evolved in
872 warm climates but promoted migration to cooler ones. *Ecology Letters* **21**: 376–383.
- 873 **Weber E, Schmid B. 1998.** Latitudinal population differentiation in two species of
874 *Solidago* (Asteraceae) introduced into Europe. *American Journal of Botany* **85**: 1110–
875 1121.
- 876 **Whitehead SR, Turcotte MM, Poveda K. 2016.** Domestication impacts on plant-
877 herbivore interactions: a meta-analysis. *Philosophical Transactions of the Royal Society*
878 *B: Biological Sciences* **372**: 20160034.
- 879 **Wilson PJ, Thompson K, Hodgson JG. 1999.** Specific leaf area and leaf dry matter
880 content as alternative predictors of plant strategies. *New Phytologist* **143**: 155–162.
- 881 **Wood SA, Karp DS, DeClerck F, Kremen C, Naeem S, Palm CA. 2015.** Functional
882 traits in agriculture: agrobiodiversity and ecosystem services. *Trends in Ecology and*
883 *Evolution* **30**: 531–539.
- 884 **Yamori W, Hikosaka K, Way DA. 2014.** Temperature response of photosynthesis in
885 C₃, C₄, and CAM plants: temperature acclimation and temperature adaptation.
886 *Photosynthesis Research* **119**: 101–117.
- 887 **Zeder MA. 2017.** Domestication as a model system for the extended evolutionary
888 synthesis. *Interface Focus* **7**: 20160133.
- 889 **Zuur AF, Ieno EN, Walker NJ, Saveliev AA, Smith GM. 2009.** *Mixed effects models*
890 *and extensions in ecology with R*. New York, USA: Springer.

891 **Figure legends**

892 **Fig. 1** Description of the study system. (a) Evolution under cultivation of durum wheat
893 (included in the *intensive experiment*) and lettuce (included in the *extensive experiment*),
894 from wild progenitors to landraces (domestication process) and from landraces to
895 improved cultivars (improvement process). (b) Phylogeny of the 19 crop species studied
896 and histogram of time in cultivation (*i.e.* earliest record of exploitation in cultivation)
897 indicating photosynthetic pathway (C₃ vs. C₄) and major organ under artificial selection
898 (either fruit, leaf, or seed) for each crop. (c) Geographical distribution of wild and
899 landrace accessions. The distribution of wild progenitors was used to infer the
900 geographic origins of each crop. (d) Climate distribution at the origin of C₃ and C₄
901 accessions for mean annual temperature and total annual precipitation. Drawings are
902 based on observations from this study and previous descriptions in the literature (see
903 *e.g.* Roucou *et al.* (2017) for wheat).

904 **Fig. 2** Changes in growth traits during (a) domestication and (b) improvement of the 19
905 crops studied. The dots are the effect sizes estimated by Hedges' *G*, and the bars are the
906 95% confidence intervals. Negative scores of Hedges' *G* indicate negative effects of
907 domestication or improvement on size-specific relative growth rate (sRGR), net
908 assimilation rate (sNAR), leaf mass ratio (sLMR), and specific leaf area (sSLA), and
909 vice versa for positive scores. Colours indicate functional group affiliation: C₃ cereals
910 (yellow), C₄ cereals (blue), forbs (pink), and legumes (red). The *intensive experiment*
911 was included in the plot (Wheat*).

912 **Fig. 3** Size-specific (a) relative growth rate (sRGR), (b) net assimilation rate (sNAR),
913 (c) leaf mass ratio (sLMR), and (d) specific leaf area (sSLA) in the *extensive experiment*
914 – 18 crop species – plotted separately by functional group: C₃ cereals, C₄ cereals, forbs,
915 and legumes, and by domestication status: wild (W), landrace (L), and improved (I)
916 accessions. Boxplots show the median and 25th and 75th percentiles of the data, with
917 whiskers extending to 1.5 times the interquartile range. Different letters denote
918 significant differences at $P < 0.05$ after Tukey's post hoc test and false discovery rate
919 correction.

920 **Fig. 4** Size-specific (a) relative growth rate (sRGR), (b) net assimilation rate (sNAR),
921 (c) leaf mass ratio (sLMR), and (d) specific leaf area (sSLA) in the *intensive experiment*
922 – durum wheat – plotted separately by domestication status: wild (W), early landrace

923 (EL), late landrace (LL), and improved (I) accessions. Boxplots show the median and
924 25th and 75th percentiles of the data, with whiskers extending to 1.5 times the
925 interquartile range. Different letters denote significant differences at $P < 0.05$ after
926 Tukey's post hoc test and false discovery rate correction.

927 **Fig. 5** Effect sizes of domestication (Hedges' G_{L-w}) on the size-specific relative growth
928 rate (sRGR), net assimilation rate (sNAR), leaf mass ratio (sLMR), and specific leaf
929 area (sSLA) of 19 crop species plotted against (a) the organ under artificial selection
930 and (b) the mean annual temperature (MAT) at the geographic origin of each crop.
931 Boxplots show the median and 25th and 75th percentiles of the data, with whiskers
932 extending to 1.5 times the interquartile range. Different letters indicate significant
933 differences at $P < 0.05$, after Tukey's post hoc test and false discovery rate correction.
934 Solid lines represent the fitted phylogenetic generalized least squares models. Symbols
935 indicate the photosynthetic pathway: C₃ (circles) and C₄ (triangles).

936 **Supporting Information**

937 **Fig. S1** Comparison of three alternative approaches to calculating RGR.

938 **Fig. S2** Comparison of size- and time-standardized RGR.

939 **Fig. S3** Comparison of growth curve parameters between functional groups and
940 domestication statuses in the *extensive experiment*.

941 **Fig. S4** Comparison of growth curve parameters between domestication statuses in the
942 *intensive experiment*.

943 **Fig. S5** Pairwise correlation between sRGR and its components.

944 **Fig. S6** Relative importance of the three components of growth on the variation of
945 sRGR.

946 **Fig. S7** Average sRGR as a function of mean annual temperature at crop origin.

947 **Table S1** List of abbreviations, definitions, formulae, and units of the growth traits
948 studied in the experiments and a diagram showing the relationships between them.

949 **Table S2** List of accessions used in the *extensive experiment*, including accession
950 identifier, functional group, domestication status, seed donor, country of origin, and
951 geographic coordinates of the collection site.

952 **Table S3** List of accessions used in the *intensive experiment*, including accession
953 identifier, domestication status, seed donor, country of origin, and geographic
954 coordinates of the collection site.

955 **Table S4** ANOVA results on the influence of 19 bioclimatic variables on changes in
956 growth traits during domestication.

957 **Table S5** ANOVA results on the influence of 19 bioclimatic variables on changes in
958 growth traits during improvement.

959 **Methods S1** Details on the estimation of total mass, leaf mass, and leaf area.

960 **Methods S2** Details on the calculation of growth traits.

961 **Tables**

962 **Table 1** Common and botanical names of the crop species used in the two experiments,
 963 as well as their domestication status (W = wild progenitor; D = domesticate) and
 964 functional group affiliations. In the *extensive experiment*, domesticate status refers to
 965 accessions belonging to both landraces and improved cultivars.

Common name	Botanical name	Domestication status	Functional group
<i>Intensive experiment</i>			
Emmer wheat	<i>Triticum dicoccoides</i> (Asch. & Graebn.) Schweinf.	W	C ₃ cereal
	<i>Triticum dicoccum</i> (Schrank ex Schübl.)	D (early landrace)	
Durum wheat	<i>Triticum durum</i> Desf.	D (late landrace)	C ₃ cereal
	<i>Triticum durum</i> Desf.	D (improved)	
<i>Extensive experiment</i>			
Barley	<i>Hordeum spontaneum</i> K.Koch	W	C ₃ cereal
	<i>Hordeum vulgare</i> L.	D	
Oat	<i>Avena sterilis</i> L.	W	C ₃ cereal
	<i>Avena sativa</i> L.	D	
Pearl millet	<i>Pennisetum glaucum</i> (L.) R.Br.	W	C ₄ cereal
	<i>Pennisetum glaucum</i> (L.) R.Br.	D	
Sorghum	<i>Sorghum arundinaceum</i> (Desv.) Stapf	W	C ₄ cereal
	<i>Sorghum bicolor</i> (L.) Moench	D	
Amaranth	<i>Amaranthus hybridus</i> L.	W	Forb
	<i>Amaranthus cruentus</i> L.	D	
Lettuce	<i>Lactuca serriola</i> L.	W	Forb
	<i>Lactuca sativa</i> L.	D	
Borage	<i>Borago officinalis</i> L.	W	Forb
	<i>Borago officinalis</i> L.	D	
Cabbage	<i>Brassica oleracea</i> L.	W	Forb
	<i>Brassica oleracea</i> L.	D	
Flax	<i>Linum usitatissimum</i> L.	W	Forb
	<i>Linum usitatissimum</i> L.	D	
Okra	<i>Abelmoschus tuberculatus</i> Pal & Singh	W	Forb
	<i>Abelmoschus esculentus</i> (L.) Moench	D	
Sesame	<i>Sesamum indicum</i> L.	W	Forb
	<i>Sesamum indicum</i> L.	D	

Chili pepper	<i>Capsicum baccatum</i> L.	W	Forb
	<i>Capsicum baccatum</i> L.	D	
Tomato	<i>Solanum pimpinellifolium</i> L.	W	Forb
	<i>Solanum lycopersicum</i> L.	D	
Faba bean	<i>Vicia narbonensis</i> L.	W	Legume
	<i>Vicia faba</i> L.	D	
Lentil	<i>Lens culinaris</i> ssp. <i>orientalis</i> (Boiss.) Ponert	W	Legume
	<i>Lens culinaris</i> Medik.	D	
Peanut	<i>Arachis monticola</i> Krapov. & Rigoni	W	Legume
	<i>Arachis hypogaea</i> L.	D	
Vetch	<i>Lathyrus cicera</i> L.	W	Legume
	<i>Lathyrus sativus</i> L.	D	
White clover	<i>Trifolium repens</i> L.	W	Legume
	<i>Trifolium repens</i> L.	D	

966

967 **Table 2** Effects of domestication and improvement on size-specific relative growth rate
 968 (sRGR), net assimilation rate (sNAR), leaf mass ratio (sLMR), and specific leaf area
 969 (sSLA) in the *extensive experiment*. All models included a two-way interaction (‘×’) *vs.*
 970 between domestication status (either Dom –wild *vs.* landrace– or Imp –landrace *vs.*
 971 improved–) and functional group (FG). Species nested within accession were
 972 considered as random factors. The table shows the $F_{d.f.}$ score and significance of
 973 predictor variables. Significant values ($P < 0.05$) are highlighted in bold. The variance
 974 of the models explained by the fixed effects is indicated by the marginal pseudo- R^2
 975 (R^2m), and the variance explained by both the fixed and random effects is indicated by
 976 the conditional pseudo- R^2 (R^2c).

	Domestication (Wild – Landrace)					Improvement (Landrace – Improved)				
	Dom	FG	Dom × FG	R^2m	R^2c	Imp	FG	Imp × FG	R^2m	R^2c
	$F_{1,68}$	$F_{3,14}$	$F_{3,68}$			$F_{1,50}$	$F_{3,14}$	$F_{3,50}$		
sRGR	1.15	9.06	3.17	0.59	0.91	0.18	10.3	1.50	0.61	0.87
sNAR	0.04	11.4	0.40	0.68	0.95	2.05	11.7	1.45	0.74	0.98
sLMR	0.02	24.8	4.25	0.77	0.96	0.62	22.7	0.80	0.80	0.99
sSLA	1.57	2.13	0.74	0.22	0.92	5.45	1.90	2.70	0.21	0.96

977 **Table 3** Effects of early domestication (earlyDom), late domestication (lateDom), and
 978 improvement (Imp) on size-specific relative growth rate (sRGR), net assimilation rate
 979 (sNAR), leaf mass ratio (sLMR), and specific leaf area (sSLA) in the *intensive*
 980 *experiment*. Accession was considered as a random factor. The table shows the $F_{d.f.}$
 981 score and significance of predictor variables. Significant values ($P < 0.05$) are
 982 highlighted in bold. The variance of the models explained by the fixed effects is
 983 indicated by the marginal pseudo- R^2 (R^2m), and the variance explained by both the fixed
 984 and random effects is indicated by the conditional pseudo- R^2 (R^2c).

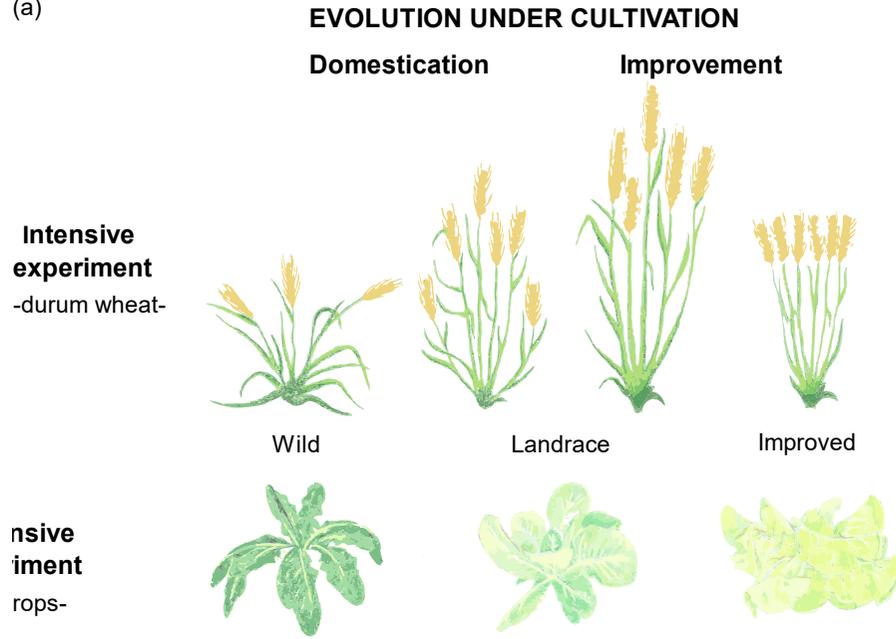
	Early domestication (Wild – Early Landrace)			Late domestication (Early landrace – Late landrace)			Improvement (Late landrace – Improved)		
	earlyDom	R^2m	R^2c	lateDom	R^2m	R^2c	Imp	R^2m	R^2c
	$F_{1,14}$			$F_{1,14}$			$F_{1,14}$		
sRGR	2.67	0.12	0.72	1.62	0.08	0.72	0.97	0.05	0.82
sNAR	2.15	0.09	0.56	2.11	0.08	0.52	0.61	0.03	0.64
sLMR	2.71	0.13	0.82	0.32	0.02	0.88	1.24	0.06	0.80
sSLA	2.42	0.11	0.77	0.04	0.001	0.47	0.49	0.02	0.40

985 **Table 4** Phylogenetic signal and the effects of organ under selection (Organ), time in cultivation (Time) and some bioclimatic variables –mean
986 annual temperature (MAT), temperature seasonality (TS), temperature of the coldest quarter (TCQ), total annual precipitation (TAP),
987 precipitation seasonality (PS), and precipitation of the driest quarter (PDQ) at the geographic origin of each crop– on changes in size-specific
988 relative growth rate (sRGR), net assimilation rate (sNAR), leaf mass ratio (sLMR), and specific leaf area (sSLA) during (a) domestication
989 (Hedges’ G_{L-W}) and (b) improvement (Hedges’ G_{I-L}). The table shows the Blomberg’s K statistic for growth trait changes as well as the $F_{d.f.}$ score
990 and significance of predictor variables. Significant values ($P < 0.05$) are highlighted in bold. Models for the bioclimatic variables included the
991 two-way interaction (‘×’) with photosynthetic pathway (Photo; C_3 vs. C_4) and their P -values were corrected for multiple testing using false
992 discovery rate. Results for the remaining bioclimatic variables can be found in Supporting Information Table S4 and Table S5.

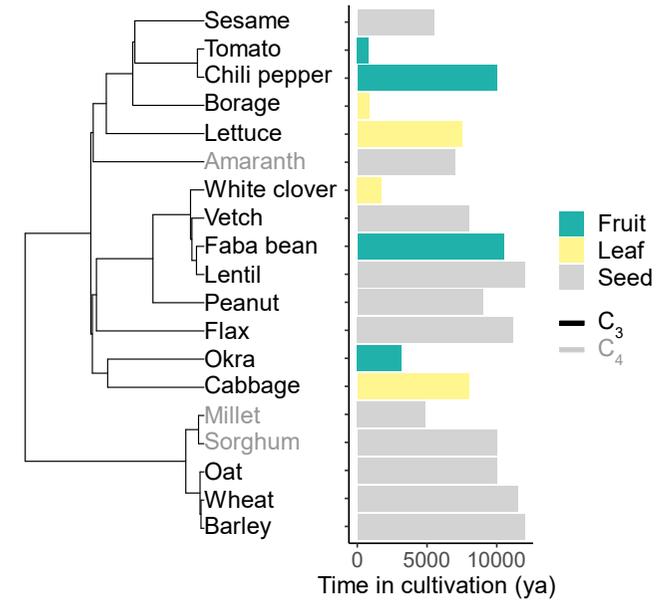
Effect size	Phylogenetic signal	Phylogenetic generalized least squares models																				
	Blomberg’s K	Model A	Model B	Model C			Model D			Model E			Model F			Model G			Model H			
		Organ	Time	MAT	Photo	MAT × Photo	TS	Photo	TS × Photo	TCQ	Photo	TCQ × Photo	TAP	Photo	TAP × Photo	PS	Photo	PS × Photo	PDQ	Photo	PDQ × Photo	
$F_{1,16}$	$F_{1,17}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$		
(a) G_{L-W}																						
sRGR	0.14	0.52	0.77	1.42	8.76	12.1	0.04	0.29	17.2	0.15	2.06	7.95	0.25	0.04	0.03	1.30	25.4	25.2	0.20	1.42	18.0	
sNAR	0.17	4.90	0.46	2.50	4.62	6.92	6.58	8.04	9.98	8.92	2.08	14.0	0.83	0.83	0.87	0.03	3.06	3.19	0.84	2.27	2.86	
sLMR	0.09	5.85	2.89	3.40	5.76	7.98	3.70	5.21	49.6	8.46	3.79	34.8	2.93	0.79	0.80	1.17	2.54	2.47	1.90	1.85	2.82	
sSLA	0.30	19.1	1.28	0.21	0.75	0.64	1.02	0.04	0.07	0.27	0.38	0.07	0.55	0.02	0.03	0.55	4.91	5.20	0.13	0.25	7.27	
(b) G_{I-L}																						
sRGR	0.11	7.81	1.39	0.80	0.15	0.20	10.2	5.30	0.78	8.10	1.60	2.07	1.77	10.6	13.3	2.29	0.50	0.29	5.07	0.48	0.23	
sNAR	0.06	0.91	2.13	0.67	0.44	2.17	29.1	2.10	1.12	3.52	0.14	1.22	1.09	0.00	0.00	3.86	7.52	11.7	6.18	0.78	1.10	
sLMR	0.08	3.23	0.55	0.07	1.94	2.60	9.89	3.33	5.13	5.37	1.39	5.09	2.54	0.01	0.01	3.40	0.38	0.26	5.02	0.11	7.51	
sSLA	0.04	0.15	0.85	0.00	1.66	2.09	33.6	2.15	0.02	6.95	0.87	0.13	0.83	0.10	0.08	1.86	2.27	2.21	5.28	0.92	2.70	

Fig. 1

(a)



(b)



(c)



(d)

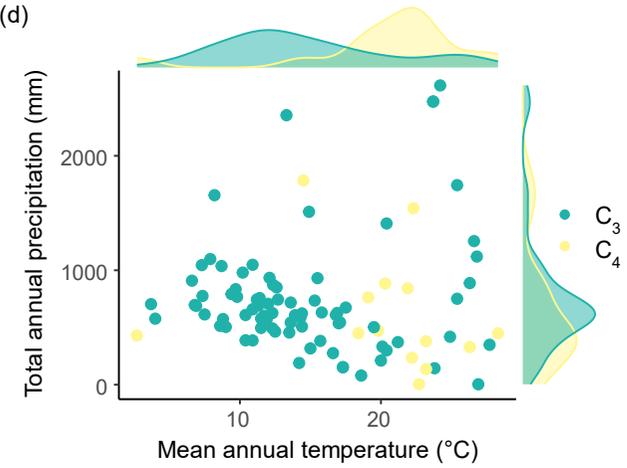
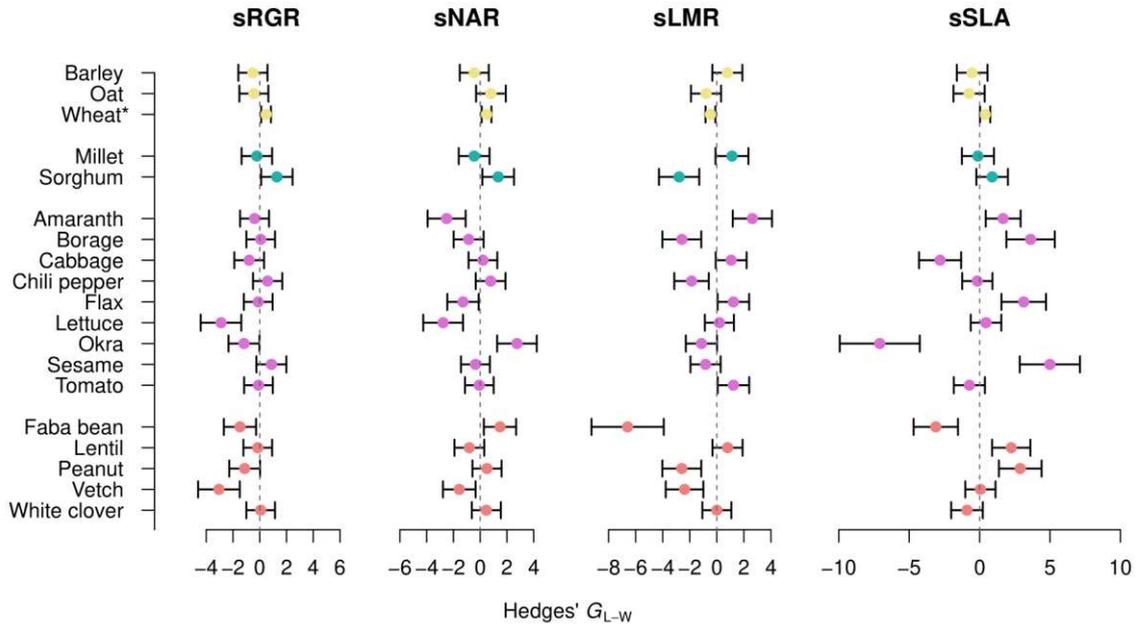


Fig.2

(a)

Domestication



(b)

Improvement

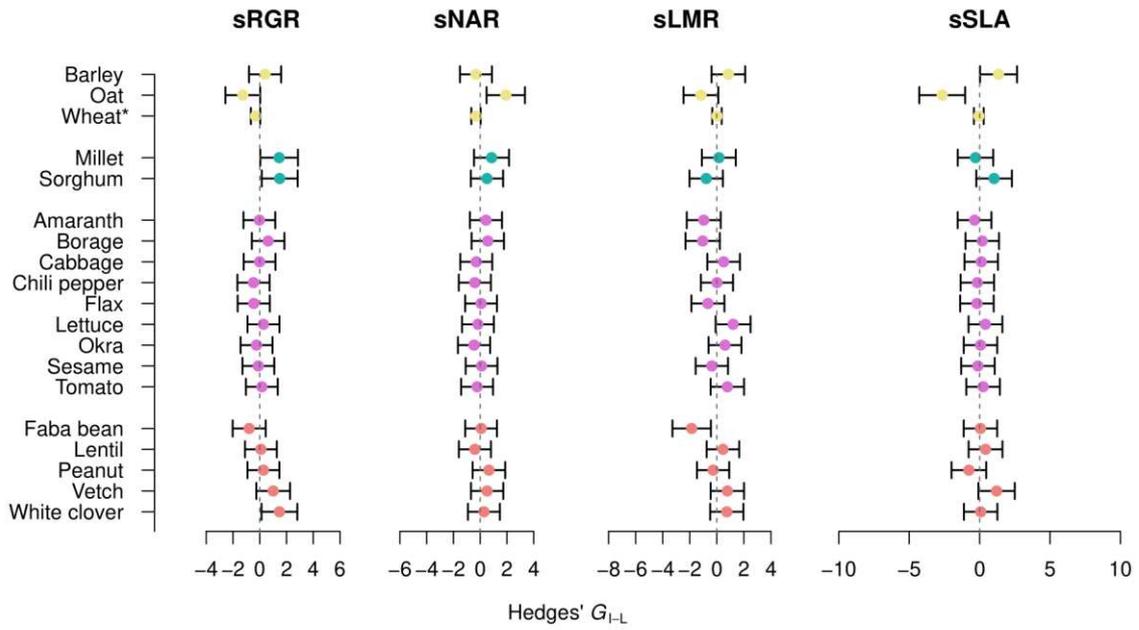


Fig. 3

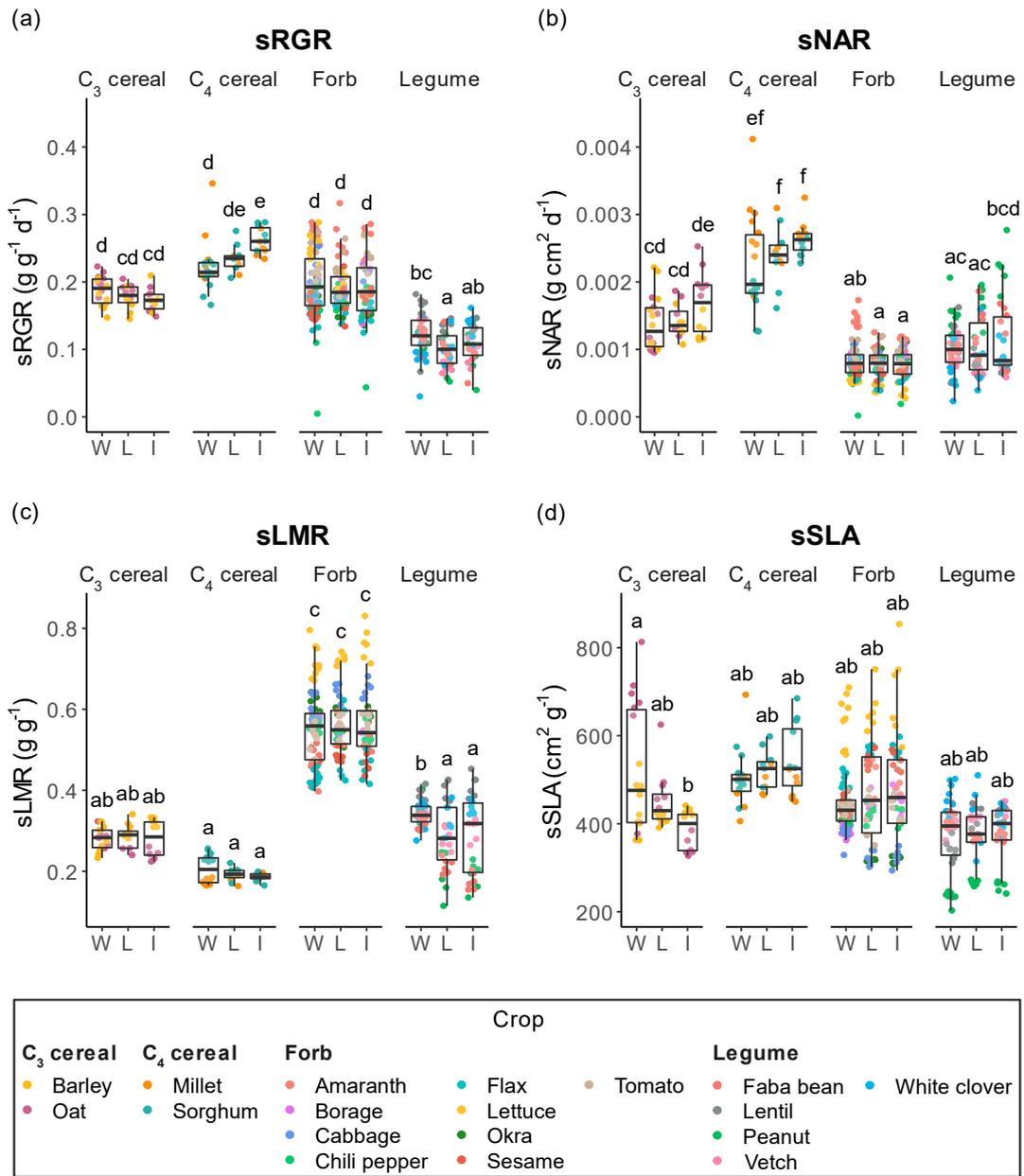
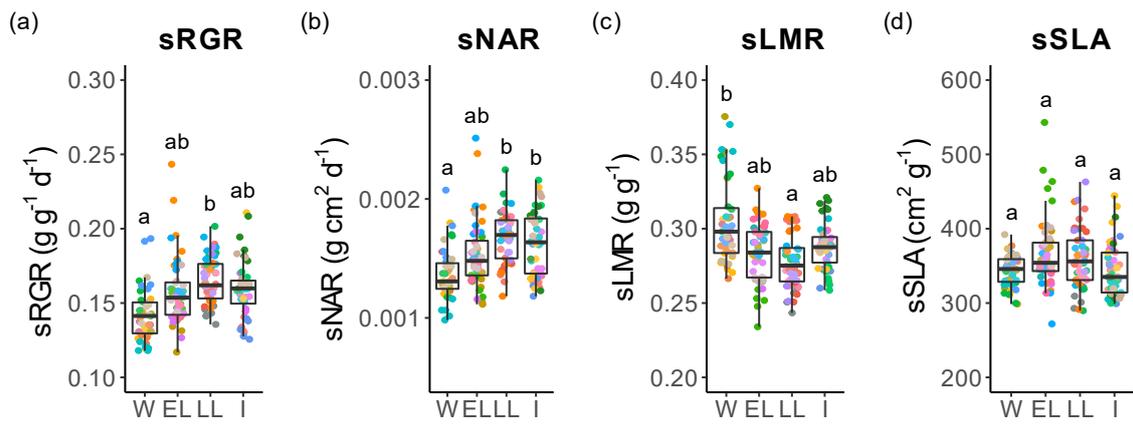
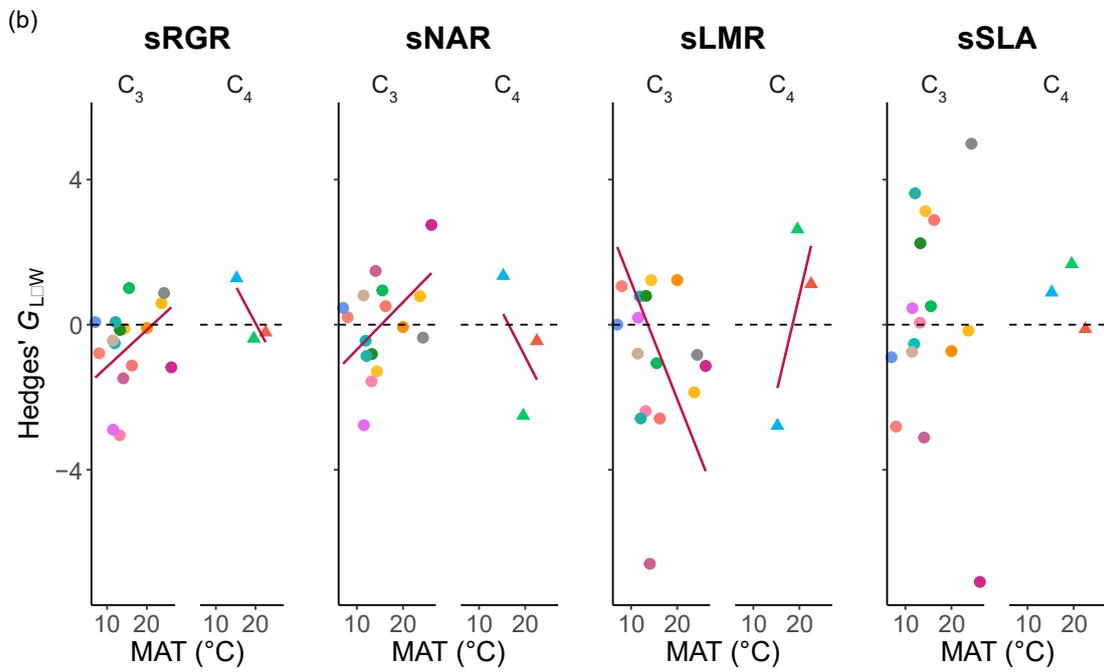
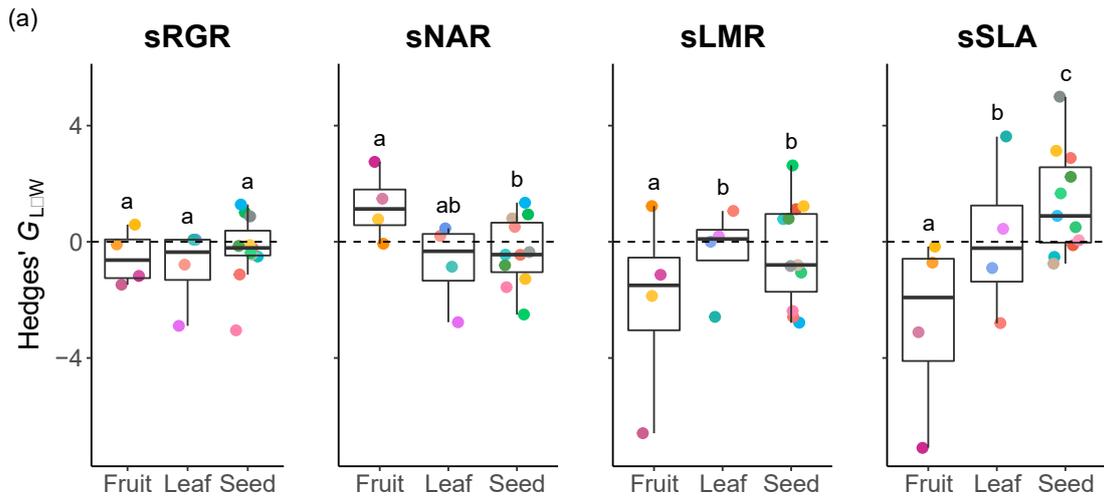


Fig. 4



Accession ID									
Wild	Early landrace				Late landrace				Improved
● 27004	● 27024	● 26894	● 33760	● 26899	● 26974	● 14060	● 30727		
● 27020	● 27025	● 33756	● 33761	● 26931	● 26982	● 14063	● 31269		
● 27021	● 33774	● 33757	● 33762	● 26966	● 33799	● 27246	● 33801		
● 27023	● 33776	● 33759	● 33764	● 26970	● 33800	● 27288	● 33802		

Fig. 5



Crop					
Fruit		Leaf		Seed	
● Chili pepper	● Borage	▲ Amaranth	▲ Millet	▲ Sorghum	
● Faba bean	● Cabbage	● Barley	● Oat	● Vetch	
● Okra	● Lettuce	● Flax	● Peanut	● Wheat	
● Tomato	● White clover	● Lentil	● Sesame		

***New Phytologist* Supporting Information**

Article title: **Disparities among crop species in the evolution of growth rates: the role of distinct origins and domestication histories.**

Authors: *Alicia Gómez-Fernández, Colin Osborne, Mark Rees, Javier Palomino, Carlos Ingala, Guillermo Gómez, and Rubén Milla.*

Article acceptance date: 24 October 2021.

The following Supporting Information is available for this article:

Fig. S1 Comparison of three alternative approaches to calculating RGR.

Fig. S2 Comparison of size- and time-standardized RGR.

Fig. S3 Comparison of growth curve parameters between functional groups and domestication statuses in the *extensive experiment*.

Fig. S4 Comparison of growth curve parameters between domestication statuses in the *intensive experiment*.

Fig. S5 Pairwise correlation between sRGR and its components.

Fig. S6 Relative importance of the three components of growth on the variation of sRGR.

Fig. S7 Average sRGR as a function of mean annual temperature at crop origin.

Table S1 List of abbreviations, definitions, formulae, and units of the growth traits studied in the experiments and a diagram showing the relationships between them.

Table S2 List of accessions used in the *extensive experiment*, including accession identifier, functional group, domestication status, seed donor, country of origin, and geographic coordinates of the collection site.

Table S3 List of accessions used in the *intensive experiment*, including accession identifier, domestication status, seed donor, country of origin, and geographic coordinates of the collection site.

Table S4 ANOVA results on the influence of 19 bioclimatic variables on changes in growth traits during domestication.

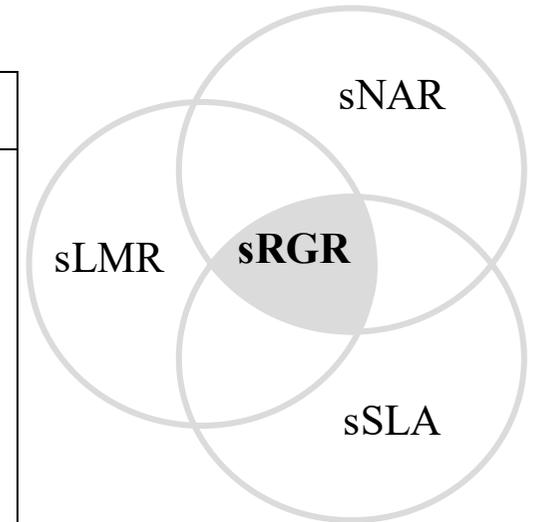
Table S5 ANOVA results on the influence of 19 bioclimatic variables on changes in growth traits during improvement.

Methods S1 Details on the estimation of total mass, leaf mass, and leaf area.

Methods S2 Details on the calculation of growth traits.

Table S1 List of abbreviations, definitions, formulae, and units for the growth traits studied in the experiments, and a diagram showing the relationships among them.

Trait	Abbr.	Definition	Formula	Unit
Size-specific relative growth rate	sRGR	The rate of dry mass accumulation at a specific plant size per unit of existing dry mass	$\frac{1}{M} \frac{dM}{dt}$	$\text{g g}^{-1} \text{d}^{-1}$
Size-specific net assimilation rate	sNAR	The rate of total dry mass increase at a specific plant size per leaf area and time	$\frac{1}{AL} \frac{dM}{dt}$	$\text{g cm}^2 \text{d}^{-1}$
Size-specific leaf mass ratio	sLMR	The ratio of total dry mass allocation to the leaves at a specific plant size	$\frac{ML}{M}$	g g^{-1}
Size-specific specific leaf area	sSLA	The ratio of total leaf area to leaf dry mass at a specific plant size	$\frac{AL}{M}$	$\text{cm}^2 \text{g}^{-1}$



$$\text{sRGR} = \text{sNAR} \times \text{sLMR} \times \text{sSLA}$$

Table S2 Common and botanical names, family, functional group, domestication status, and seed origin information (country and geographic coordinates) for each accession used in the *extensive experiment*. Accession identifier refers to the code assigned by each seed donor, except for commercial companies (N.A. = not applicable). The country and coordinates (latitude and longitude) where seeds were originally collected are indicated (N.A. = not available). Seed donor (BGVCU: Banco de Germoplasma Vegetal de Cuenca, Spain; CGN: Center for Genetic Resources, The Netherlands; CITA: Centro de Investigación y Transferencia Agroalimentaria de Aragón, Spain; COMAV: Instituto Universitario de Conservación y Mejora de la Agrodiversidad Valenciana, Spain; CRF: Centro Nacional de Recursos Fitogenéticos-INIA, Spain; ICARDA: International Center for Agricultural Research in Dry Areas, Lebanon; IPK: Germplasm Bank of the Leibniz Institute of Plant Genetics and Crop Plant Research, Germany; NPGS: National Plant Germplasm System-USDA, U.S.A.; *: commercial company).

Common name	Functional group	Family	Botanical name	Domestication status	Accession identifier	Accession country	Latitude	Longitude	Seed donor
Barley	C ₃ cereal	Poaceae	<i>Hordeum spontaneum</i> K.Koch	Wild	BGE025385	Morocco	N.A.	N.A.	CRF
					PI 662181	Turkey	37.746	39.661	NPGS
					BGE025389	Morocco	N.A.	N.A.	CRF
			<i>Hordeum vulgare</i> L.	Landrace	BGE011162	Morocco	35.574	-5.375	CRF
					BGE024314	Greece	38.537	22.622	CRF
				Improved	N.A.	N.A.	N.A.	N.A.	Battle*
					BGE000214	Spain	N.A.	N.A.	CRF
Oat	C ₃ cereal	Poaceae	<i>Avena sterilis</i> L.	Wild	BGE049076	Spain	38.786	-0.263	CRF
					BGE049079	Spain	42.841	-1.676	CRF
					IG 100379 IFMI 3096	Turkey	N.A.	N.A.	ICARDA
			<i>Avena sativa</i> L.	Landrace	BGE008136	Spain	41.983	2.825	CRF
					BGE008166	Spain	42.483	-3.199	CRF
				Improved	N.A.	N.A.	N.A.	N.A.	Battle*
					BGE024681	Spain	N.A.	N.A.	CRF
Millet	C ₄ cereal	Poaceae	<i>Cenchrus americanus</i> (L.) Morrone	Wild	PI 537068	Niger	17.767	8.950	NPGS
					PEN 1028	Yemen	14.083	44.167	IPK
					PEN 1048	Yemen	16.07	43.300	IPK

				Landrace	PEN 837	Tunisia	36.803	10.172	IPK
					PEN 687	Libya	26.633	13.633	IPK
				Improved	PI 586660	Burkina Faso	N.A.	N.A.	NPGS
					PEN 1257	Soviet Union	N.A.	N.A.	IPK
Sorghum	C ₄ cereal	Poaceae	<i>Sorghum arundinaceum</i> (Desv.) Stapf	Wild	PI 524718	Sudan	12.723	29.804	NPGS
					PI 482605	Zimbabwe	-20.383	30.667	NPGS
					PI 539066	Soviet Union	52.453	56.224	NPGS
			<i>Sorghum bicolor</i> (L.) Moench	Landrace	PI 532206	Oman	17.333	54.000	NPGS
					PI 535999	Cameroon	12.117	14.750	NPGS
				Improved	PI 563327	Sudan	N.A.	N.A.	NPGS
					PI 563437	Chad	N.A.	N.A.	NPGS
Amaranthus	Forb	Amaranthaceae	<i>Amaranthus hybridus</i> L.	Wild	Ames 2072	Nepal	27.701	85.300	NPGS
					PI 500234	Zambia	-15.300	23.150	NPGS
					PI 652417	Brazil	-16.217	-47.917	NPGS
			<i>Amaranthus cruentus</i> L.	Landrace	Ames 2001	Ghana	N.A.	N.A.	NPGS
					PI 643050	Mexico	18.717	-98.750	NPGS
				Improved	AMA 169	Nepal	N.A.	N.A.	IPK
					Ames 15197	Argentina	N.A.	N.A.	NPGS
Lettuce	Forb	Asteraceae	<i>Lactuca serriola</i> L.	Wild	BGV009232	Spain	43.094	-6.253	COMAV
					BGE034705	Spain	40.517	-3.283	CRF
					LAC 1079	Italy	45.427	12.178	IPK
			<i>Lactuca sativa</i> L.	Landrace	BGV003526	Spain	42.601	-6.724	COMAV
					BGV001094	Spain	37.692	-4.480	COMAV
				Improved	N.A.	N.A.	N.A.	N.A.	Battle*
					BGV005752	Spain	N.A.	N.A.	COMAV
Borago	Forb	Boraginaceae	<i>Borago officinalis</i> L.	Wild	BGHZ5329	Spain	40.978	-0.055	CITA
					BGHZ2103	Spain	42.173	-0.029	CITA
					BGHZ4294	Spain	42.279	-5.100	CITA
				Landrace	BGHZ0363	Spain	40.976	-0.443	CITA
					BGHZ2340	Spain	42.388	-0.717	CITA
				Improved	N.A.	N.A.	N.A.	N.A.	Battle*

Cabbage	Forb	Brassicaceae	<i>Brassica oleracea</i> L.	Wild	N.A.	N.A.	N.A.	N.A.	Rocalba*	
					CGN06903	France	50.180	1.483	CGN	
					CGN18947	Germany	54.200	7.867	CGN	
					CGN25455	Netherlands	53.310	5.622	CGN	
					Landrace	CGN14079	Belgium	40.976	-0.443	CGN
						CGN15773	Portugal	42.388	-0.717	CGN
					Improved	N.A.	N.A.	N.A.	N.A.	Rocalba*
N.A.	N.A.	N.A.	N.A.	N.A.	Battle*					
Flax	Forb	Linaceae	<i>Linum usitatissimum</i> L.	Wild	Ames 29165	Georgia	41.660	43.053	NPGS	
					PI 231945	Belgium	N.A.	N.A.	NPGS	
					PI 253972	Irak	35.479	43.419	NPGS	
					Landrace	LIN 2020	Yemen	14.633	43.633	IPK
						LIN 2288	Colombia	N.A.	N.A.	IPK
					Improved	BGE030455	Spain	N.A.	N.A.	CRF
						PI 598151	Nepal	N.A.	N.A.	NPGS
Okra	Forb	Malvaceae	<i>Abelmoschus tuberculatus</i> Pal & Singh	Wild	Grif 12671	India	24.483	72.783	NPGS	
					PI 639676	Sri Lanka	6.275	81.157	NPGS	
					PI 639681	India	21.537	78.803	NPGS	
					Landrace	PI 489782	Ivory Coast	5.667	-4.167	NPGS
						PI 505564	Zambia	-27.417	17.167	NPGS
					Improved	N.A.	N.A.	N.A.	N.A.	Battle*
						PI 548700	India	N.A.	N.A.	NPGS
Sesamum	Forb	Pedaliaceae	<i>Sesamum indicum</i> L.	Wild	SESA 17	Yemen	15.333	43.000	IPK	
					SESA 20	Yemen	15.210	43.340	IPK	
					SESA 22	Yemen	16.339	43.704	IPK	
					Landrace	SESA 4	North Korea	38.949	125.765	IPK
						SESA 5	Irak	33.354	43.779	IPK
					Improved	N.A.	N.A.	N.A.	N.A.	Rocalba*
						SESA 14	N.A.	N.A.	N.A.	IPK
Chili pepper	Forb	Solanaceae	<i>Capsicum baccatum</i> L.	Wild	CGN21515	N.A.	N.A.	N.A.	CGN	
					CGN16973	Bolivia	-16.800	64.400	CGN	

					CGN17025	Bolivia	-16.800	64.400	CGN
				Landrace	CGN16972	India	19.000	85.000	CGN
					CGN23260	Bolivia	-16.800	-64.400	CGN
				Improved	CGN21470	Chile	N.A.	N.A.	CGN
					CGN22181	Peru	N.A.	N.A.	CGN
Tomato	Forb	Solanaceae	<i>Solanum pimpinellifolium</i> L.	Wild	BGV007948	Peru	-7.200	-79.050	COMAV
					LYC 1	N.A.	N.A.	N.A.	IPK
					LYC 2671	N.A.	N.A.	N.A.	IPK
			<i>Solanum lycopersicum</i> L.	Landrace	LYC 15	Switzerland	47.148	8.526	IPK
					LYC 1014	Guatemala	14.835	-91.518	IPK
				Improved	N.A.	N.A.	N.A.	N.A.	Battle*
					N.A.	N.A.	N.A.	N.A.	Clause*
Faba bean	Legume	Fabaceae	<i>Vicia narbonensis</i> L.	Wild	IG 111590 IFVI5266	Tunisia	37.284	9.836	ICARDA
					BGE031092	Spain	40.817	-3.617	CRF
					BGE031093	Spain	38.100	-3.083	CRF
			<i>Vicia faba</i> L.	Landrace	BGE022388	Spain	42.850	-1.767	CRF
					BGE031076	Spain	40.573	-5.060	CRF
				Improved	N.A.	N.A.	N.A.	N.A.	Rocalba*
					N.A.	N.A.	N.A.	N.A.	Battle*
Lens	Legume	Fabaceae	<i>Lens culinaris</i> ssp. <i>orientalis</i> (Boiss.) Ponert	Wild	PI 572374	Iran	31.067	56.350	NPGS
					PI 572399	Turkey	37.167	29.579	NPGS
					BCU001423	Turkey	N.A.	N.A.	BGVCU
			<i>Lens culinaris</i> Medik.	Landrace	PI 297287	Argentina	N.A.	N.A.	NPGS
					PI 298022	Turkey	39.996	32.867	NPGS
				Improved	N.A.	N.A.	N.A.	N.A.	Battle*
					PI 379368	Serbia	N.A.	N.A.	NPGS
Peanut	Legume	Fabaceae	<i>Arachis monticola</i> Krapov. & Rigoni	Wild	PI 263393	Brazil	-22.870	-47.077	NPGS
					PI 468196	Argentina	-24.117	-65.383	NPGS
					PI 497261	Argentina	-24.133	-65.383	NPGS
			<i>Arachis hypogaea</i> L.	Landrace	PI 602352	Brazil	N.A.	N.A.	NPGS
					Grif 373	Sudan	N.A.	N.A.	NPGS

				Improved	PI 538758	Burkina Faso	N.A.	N.A.	NPGS
					PI 550688	China	N.A.	N.A.	NPGS
Vetch	Legume	Fabaceae	<i>Lathyrus cicera</i> L.	Wild	BGE019570	Spain	40.200	-2.267	CRF
					BGE016953	Spain	39.917	-5.167	CRF
					BGE016954	Spain	39.550	-5.400	CRF
			<i>Lathyrus sativus</i> L.	Landrace	BGE014724	Spain	40.003	3.839	CRF
					BGE046719	Spain	42.803	-8.898	CRF
				Improved	LAT 440	India	N.A.	N.A.	IPK
					LAT 466	Soviet Union	N.A.	N.A.	IPK
White clover	Legume	Fabaceae	<i>Trifolium repens</i> L.	Wild	CGN22512	Uzbekistan	41.150	70.417	CGN
					CGN22513	Kyrgyzstan	40.980	73.183	CGN
					CGN22516	Kyrgyzstan	41.230	73.367	CGN
				Landrace	CGN21763	France	45.700	2.900	CGN
					CGN22506	Netherlands	53.500	6.267	CGN
				Improved	N.A.	N.A.	N.A.	N.A.	Intersemillas*
					CGN23145	Denmark	N.A.	N.A.	CGN

Table S3 Botanical name, domestication status and seed origin information (country and geographic coordinates) for each accession used in the *intensive experiment*. Accession identifier refers to the code assigned by each seed donor, except for commercial companies. The country and coordinates (latitude and longitude) where seeds were originally collected are indicated (N.A. = not available). All seeds come from INRA - CRB: Small grain cereals Biological Resources Centre, France. Durum wheat belongs to the functional group of C₃ cereals.

Botanical name	Domestication status	Accession identifier	Accession country	Latitude (°)	Longitude (°)
<i>Triticum dicoccoides</i> (Asch. & Graebn.) Schweinf.	Wild	27004	Israel	N.A.	N.A.
		27020	Israel	N.A.	N.A.
		27021	Israel	N.A.	N.A.
		27023	Syria	32.783	36.200
		27024	Iraq	N.A.	N.A.
		27025	Iraq	N.A.	N.A.
		33774	Turkey	37.920	40.55
		33776	Israel	32.867	35.533
<i>Triticum dicoccum</i> (Schränk) Schübl	Early landrace	26894	Algeria	34.800	3.117
		33756	Turkey	39.000	35.000
		33757	Iraq	32.000	53.000
		33759	Iran	32.000	53.000
		33760	Italy	41.283	15.100
		33761	Russia	57.600	39.867
		33762	Slovakia	48.731	17.406
		33764	Germany	51.500	7.000
<i>Triticum durum</i> Desf.	Late landrace	26899	Algeria	N.A.	N.A.
		26931	Pakistan	N.A.	N.A.
		26966	Egypt	24.091	32.899
		26970	Palestine	32.500	35.500
		26974	Russia	34.717	33.083
		26982	Spain	37.167	-3.600

		33799	Turkey	37.420	31.850
		33800	Turkey	38.750	34.850
<i>Triticum durum</i> Desf.	Improved	14060	France	N.A.	N.A.
		14063	France	N.A.	N.A.
		27246	France	N.A.	N.A.
		27288	France	N.A.	N.A.
		30727	France	N.A.	N.A.
		31269	France	N.A.	N.A.
		33801	France	N.A.	N.A.
		33802	France	N.A.	N.A.

Table S4 Effects of the 19 bioclimatic variables at the geographic origin of each crop on the effect size of domestication (Hedges' $G_{L,W}$) on size-specific relative growth rate (sRGR), net assimilation rate (sNAR), leaf mass ratio (sLMR), and specific leaf area (sSLA). Models included the two-way interaction ('×') with photosynthetic pathway (Photo; C_3 vs. C_4). The table shows the $F_{d.f.}$ score and the significances of the predictor variables. Significant P -values ($P < 0.05$) are highlighted in bold after false discovery rate correction. Models were tested with phylogenetic generalized least squares. Abbreviations: BIO1, annual mean temperature; BIO2, mean diurnal range; BIO3, isothermality; BIO4, temperature seasonality; BIO5, maximum temperature of warmest month; BIO6, minimum temperature of coldest month; BIO7, temperature annual range; BIO8, mean temperature of wettest quarter; BIO9, mean temperature of driest quarter; BIO10, mean temperature of warmest quarter; BIO11, mean temperature of coldest quarter; BIO12, annual precipitation; BIO13, precipitation of wettest month; BIO14, precipitation of driest month; BIO15, precipitation seasonality; BIO16, precipitation of wettest quarter; BIO17, precipitation of driest quarter; BIO18, precipitation of warmest quarter; BIO19, precipitation of coldest quarter.

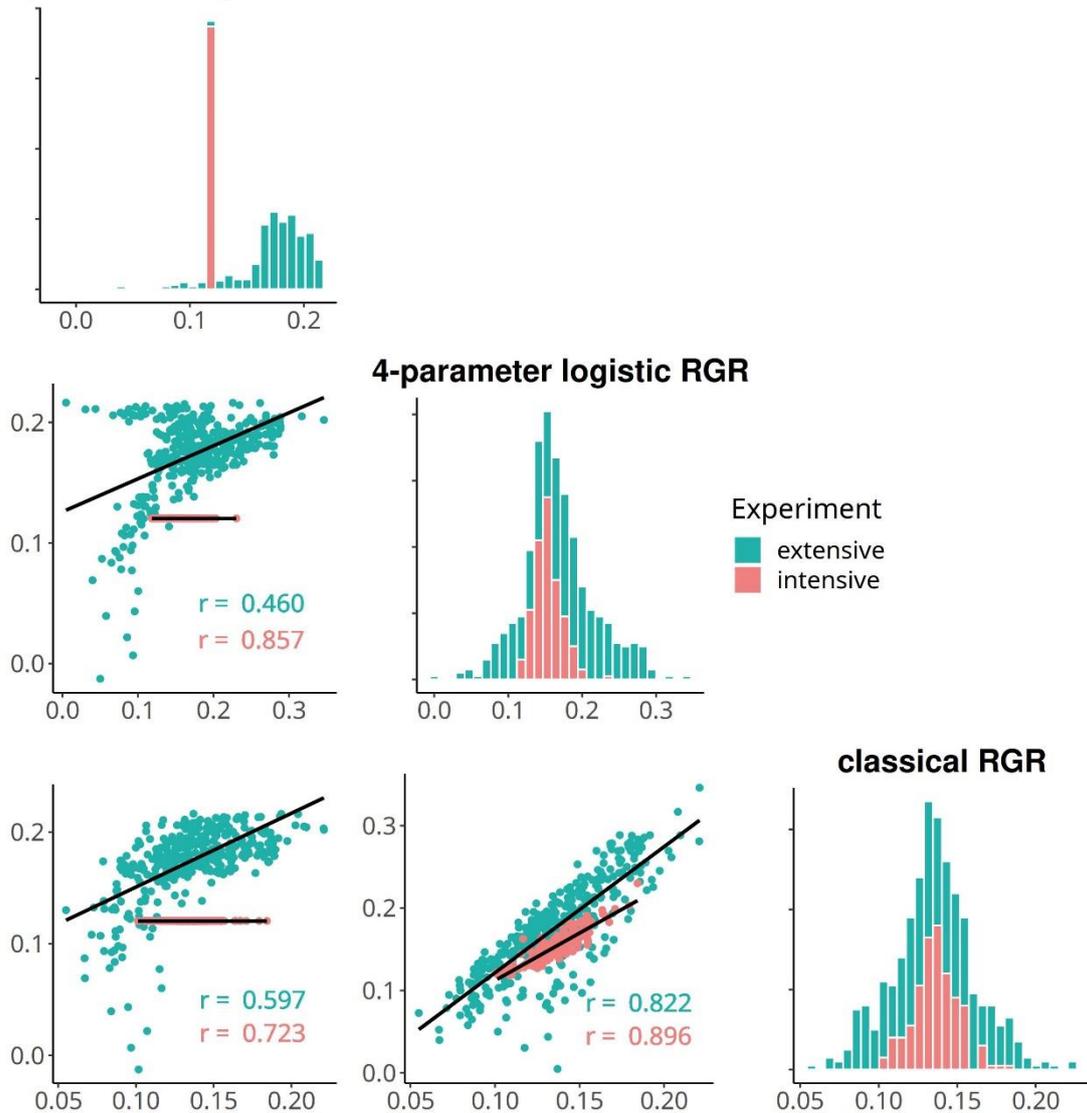
Response Hedges' G_{L-w}	Predictors	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19
		$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$
sRGR	BIO	1.40	1.36	0.00	0.03	0.23	0.00	0.07	0.21	0.42	0.60	0.14	0.29	0.01	0.10	1.51	0.00	0.21	2.86	0.18
	Photo	8.78	1.84	26.6	0.30	1.73	0.74	1.20	5.26	3.26	3.26	2.07	0.04	1.26	0.93	24.8	0.92	1.44	2.09	0.56
	BIO × Photo	12.1	2.07	55.4	17.6	1.66	6.60	20.8	5.54	10.3	3.12	7.96	0.03	1.16	17.6	24.4	0.86	18.1	1.73	18.5
sNAR	BIO	2.50	5.52	0.01	6.97	1.91	8.85	7.66	10.13	0.47	0.79	8.76	0.75	2.32	0.29	0.04	1.73	0.79	0.25	0.89
	Photo	4.62	56.1	3.60	8.36	1.11	0.01	9.78	6.25	0.28	2.59	2.05	0.80	4.51	2.17	2.95	3.33	2.27	3.53	1.78
	BIO × Photo	6.92	83.0	3.93	10.3	1.17	13.6	10.5	15.7	1.75	2.76	13.8	0.84	4.80	2.94	3.08	3.49	2.87	3.86	2.86
sLMR	BIO	3.41	6.10	0.40	3.79	0.38	11.66	4.70	7.43	0.06	0.03	8.42	2.79	3.26	0.56	1.56	3.51	1.84	0.46	1.92
	Photo	5.77	67.7	3.43	5.37	1.38	0.11	11.6	12.8	1.15	3.03	3.75	0.74	4.36	1.73	2.32	3.37	1.85	3.18	1.34
	BIO × Photo	7.99	127.1	3.47	49.9	1.40	16.3	61.6	24.8	6.32	3.13	34.7	0.75	4.54	2.87	2.24	3.46	2.82	3.40	2.81
sSLA	BIO	0.21	4.68	0.25	0.95	0.99	0.55	1.27	1.00	0.68	1.07	0.32	0.61	0.70	0.00	0.45	0.35	0.14	1.15	0.15
	Photo	0.74	0.04	2.25	0.04	8.87	0.33	0.00	0.14	1.10	10.19	0.21	0.02	0.03	0.13	4.38	0.03	0.26	0.21	0.12
	BIO × Photo	0.63	0.06	2.88	0.07	9.70	0.03	0.07	0.05	2.87	18.24	0.12	0.03	0.02	4.52	4.55	0.02	7.39	0.13	7.14

Table S5 Effects the 19 bioclimatic variables at the geographic origin of each crop on the effect size of improvement (Hedges' G_{1-L}) on size-specific relative growth rate (sRGR), net assimilation rate (sNAR), leaf mass ratio (sLMR) and specific leaf area (sSLA). Models included the two-way interaction ('×') with photosynthetic pathway (Photo; C_3 vs. C_4). The table shows the $F_{d.f.}$ score and the significances of the predictor variables. Significant P -values ($P < 0.05$) are highlighted in bold after false discovery rate correction. Models were tested with phylogenetic generalized least squares. Abbreviations: BIO1, annual mean temperature; BIO2, mean diurnal range; BIO3, isothermality; BIO4, temperature seasonality; BIO5, maximum temperature of warmest month; BIO6, minimum temperature of coldest month; BIO7, temperature annual range; BIO8, mean temperature of wettest quarter; BIO9, mean temperature of driest quarter; BIO10, mean temperature of warmest quarter; BIO11, mean temperature of coldest quarter; BIO12, annual precipitation; BIO13, precipitation of wettest month; BIO14, precipitation of driest month; BIO15, precipitation seasonality; BIO16, precipitation of wettest quarter; BIO17, precipitation of driest quarter; BIO18, precipitation of warmest quarter; BIO19, precipitation of coldest quarter.

Response Hedges' G_{i-L}	Predictors	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19
		$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$	$F_{1,15}$
sRGR	BIO	0.81	18.4	0.01	10.4	5.48	12.0	15.7	4.35	2.20	2.19	8.04	1.75	1.95	3.62	2.88	1.68	4.96	6.29	5.06
	Photo	0.15	0.52	0.17	5.46	0.05	4.58	4.73	0.26	3.40	0.15	1.61	10.27	0.25	0.55	0.65	0.37	0.47	0.66	0.79
	BIO × Photo	0.20	0.16	0.02	0.81	0.12	2.95	1.03	1.61	0.83	0.03	2.04	13.3	0.08	0.24	0.40	0.20	0.22	0.19	0.24
sNAR	BIO	0.71	39.5	0.41	29.2	58.2	5.38	43.1	2.25	22.2	35.73	3.38	1.09	1.19	5.19	4.96	0.59	6.15	18.4	6.62
	Photo	0.46	0.14	0.53	2.19	1.05	0.06	1.94	0.00	10.23	2.86	0.13	0.00	0.00	0.74	8.80	0.05	0.79	0.11	0.71
	BIO × Photo	2.23	1.21	2.38	1.15	1.06	2.14	1.36	0.33	10.1	3.07	1.14	0.00	0.01	1.16	13.0	0.05	1.08	0.03	1.16
sLMR	BIO	0.05	32.5	0.41	10.2	8.12	8.63	16.2	4.28	3.79	3.76	5.14	2.48	2.35	3.05	4.68	1.94	4.95	6.55	5.55
	Photo	1.88	5.97	5.67	3.46	1.18	0.24	5.31	2.52	0.14	11.33	1.37	0.01	0.82	0.07	0.25	0.43	0.10	0.76	0.03
	BIO × Photo	2.49	7.02	7.40	5.26	1.14	6.97	6.80	4.92	0.01	24.1	4.92	0.00	0.79	5.02	0.15	0.40	7.55	0.89	0.50
sSLA	BIO	0.00	22.5	0.16	34.4	25.7	9.32	50.7	4.35	11.1	14.3	6.93	0.81	1.60	4.42	2.29	0.91	5.14	7.82	5.68
	Photo	1.65	4.50	9.05	2.31	1.89	1.19	1.12	0.56	9.83	4.06	0.88	0.11	0.51	0.77	2.53	0.52	0.91	1.56	0.60
	BIO × Photo	2.07	6.61	22.90	0.02	1.75	0.22	0.01	0.17	8.96	3.85	0.12	0.09	0.42	2.74	2.48	0.45	2.66	1.11	2.83

Fig. S1 Comparison of alternative approaches to modelling RGR. Relationships between the different RGR measures (below the main diagonal, all $g\ g^{-1}\ d^{-1}$), histograms of RGR calculated using each method (diagonal), and the R^2 for relationships between RGR values calculated by alternative methods. Classical RGR was calculated as mass increase per unit of initial mass and per unit of time [$RGR = (\ln M_1 - \ln M_2) / (t_2 - t_1)$, where M_1 and M_2 are plant mass at the beginning (t_1) and end (t_2) of the vegetative growth period, respectively]. Details on the calculation of three- and four-parameter logistic RGRs can be found in Paine *et al.* (2012)¹.

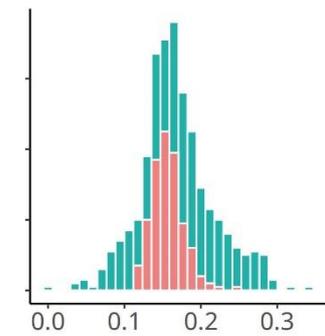
3-parameter logistic RGR



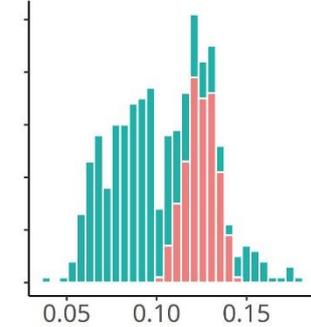
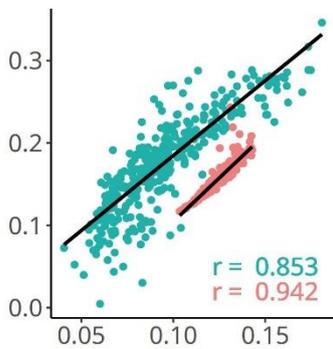
¹ Paine CET, Marthews TR, Vogt DR, Purves D, Rees M, Hector A, Turnbull LA. 2012. How to fit nonlinear plant growth models and calculate growth rates: an update for ecologists. *Methods in Ecology and Evolution* 3: 245–256.

Fig. S2 Comparison of RGRs calculated at different reference sizes. Relationships between the different RGRs (below the main diagonal, all $g\ g^{-1}\ d^{-1}$), histograms of RGRs calculated using each reference size (diagonal), and the R^2 for relationships between RGR values calculated using alternative reference size criteria. As a common size, we used the median of the $\log_e(\text{mass})$ distribution across all focal plants, since all plants occurred at this size. As ontogenetic stages, we used the $\log_e(\text{mass})$ reached at both the inflection point (adult stage) and mid-inflection point (seedling stage) of each focal plant.

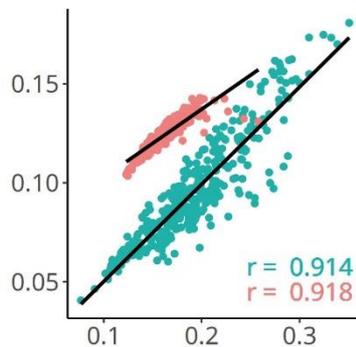
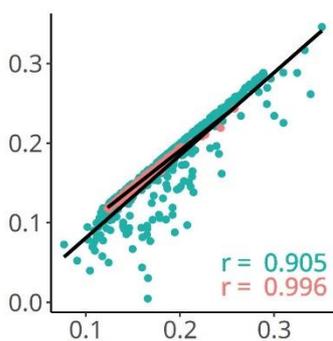
RGR at a common size



RGR at an early stage



Experiment
■ extensive
■ intensive



RGR at an adult stage

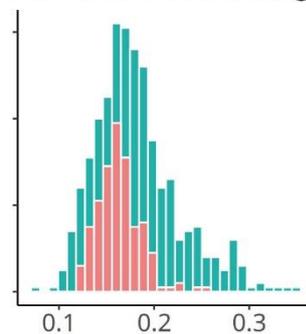


Fig. S3 Comparison of growth curve parameters in the *extensive experiment*, plotted separately by functional group: C₃ cereals, C₄ cereals, forbs and legumes, and by domestication status: wild (W), landrace (L), and improved (I) accessions. The parameters are: (a) minimum asymptote (*i.e.* the lower horizontal asymptote), (b) maximum asymptote (*i.e.* the upper horizontal asymptote), (c) steepest slope (*i.e.* the absolute increase in mass per unit time at the inflection point), and (d) arc length (*i.e.* time when the plant mass is midway between the minimum and maximum asymptotes). Boxplots show the median and 25th and 75th percentiles of the data, with whiskers extending to 1.5 times the interquartile range. Different letters denote significant differences at $P < 0.05$ after Tukey's post hoc test and false discovery rate correction.

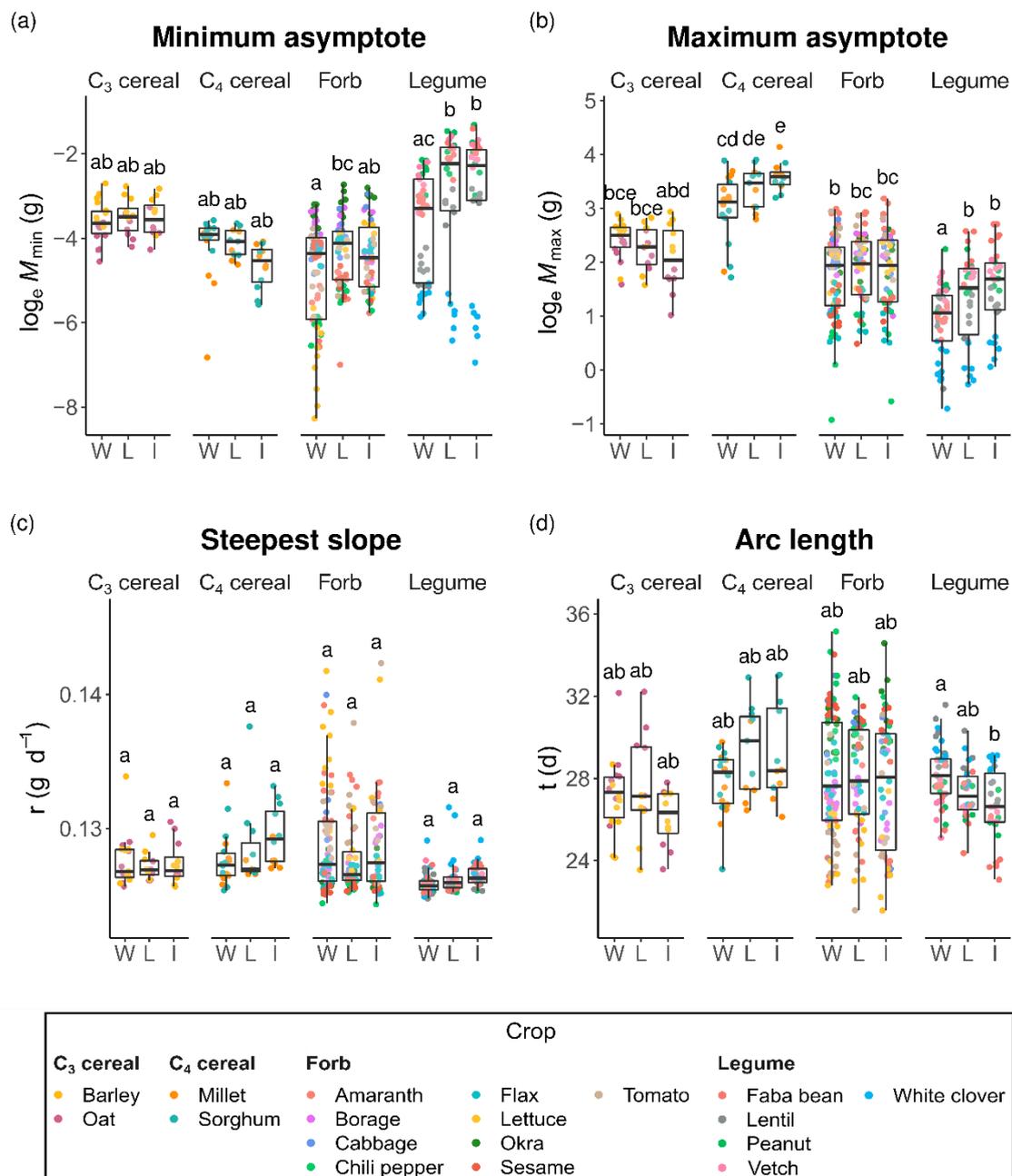


Fig. S4 Comparison of growth curve parameters in the *intensive experiment*, plotted separately by domestication status: wild (W), early landrace (EL), late landrace (LL) and improved (I) accessions. The parameters are: (a) minimum asymptote (*i.e.* the lower horizontal asymptote), (b) maximum asymptote (*i.e.* the upper horizontal asymptote), (c) steepest slope (*i.e.* the absolute increase in mass per unit time at the inflection point), and (d) arc length (*i.e.* time when the plant mass is midway between the minimum and maximum asymptotes). Boxplots show the median and 25th and 75th percentiles of the data, with whiskers extending to 1.5 times the interquartile range. Different letters denote significant differences at $P < 0.05$ after Tukey's post hoc test and false discovery rate correction.

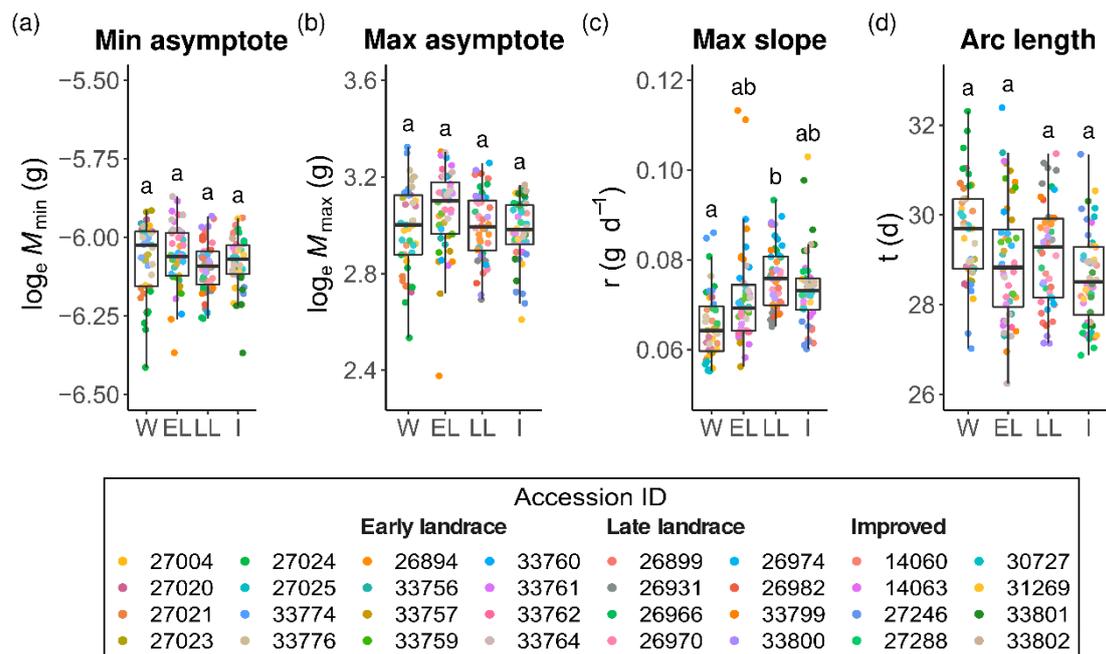
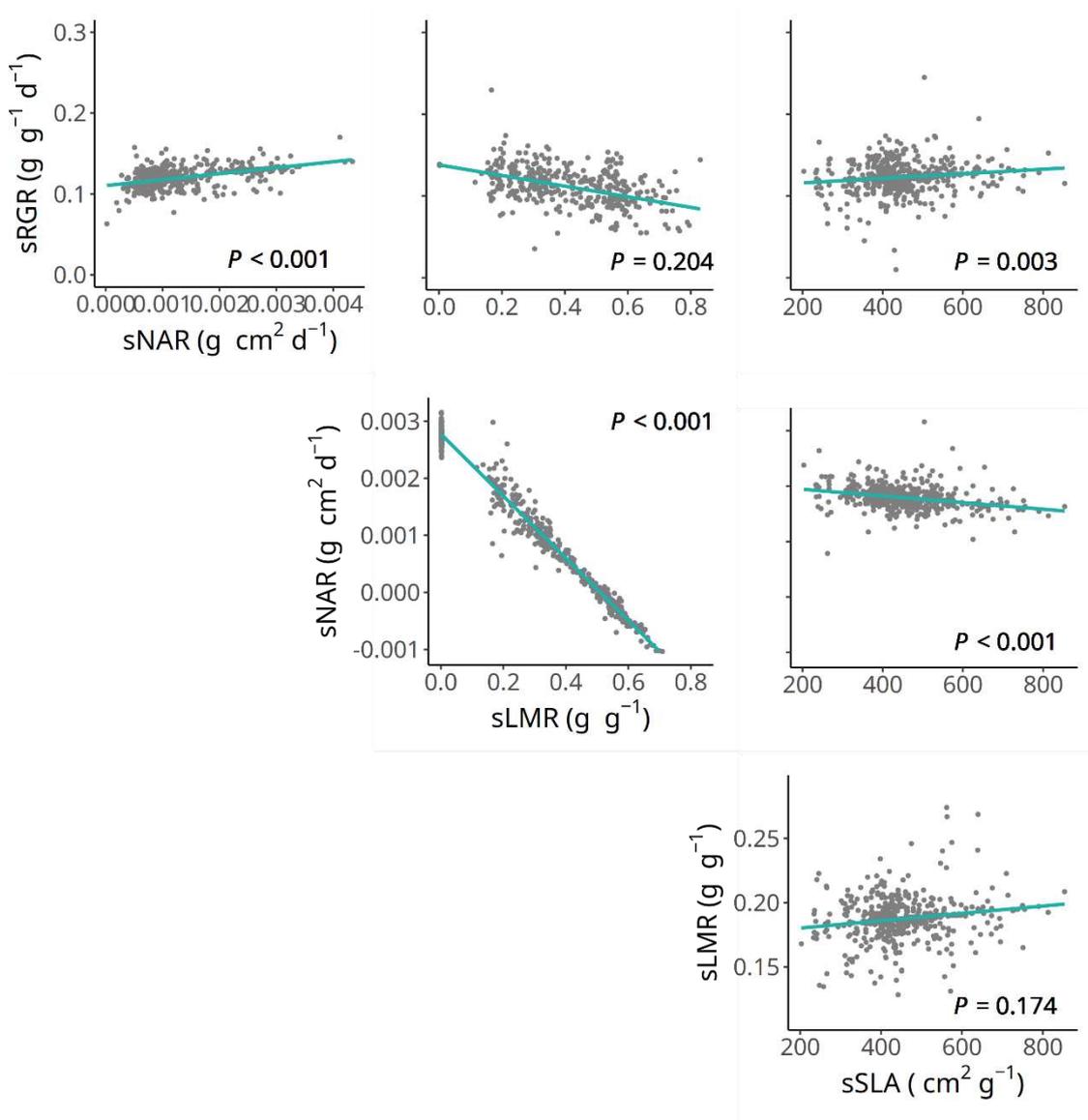


Fig. S5 Partial residuals and prediction line of the linear mixed-effects model showing the relationship between size-specific relative growth rate (sRGR), net assimilation rate (sNAR), leaf mass ratio (sLMR), and specific leaf are (sSLA). For sRGR, linear mixed-effects models included the interaction between one sRGR component, domestication status and functional group as fixed effects, and accession identity (nested within species) as random effects over the intercept. This model structure was repeated for the sRGR components as response variables. The plot was generated using the *visreg* function of the ‘visreg’ R package (Breheny & Burchett, 2017²).



² Breheny P, Burchett W. 2017. Visualization of regression models using visreg. *The R Journal* 9: 56–71.

Fig. S6 Importance of interspecific variation in size-specific net assimilation rate (sNAR), leaf mass ratio (sLMR), and specific leaf area (sSLA) to variation in size-specific relative growth rate (sRGR). Percentage variation is shown for (a) functional group: C₃ cereals, C₄ cereals, forbs, and legumes; and (b) domestication status: wild, landraces and improved cultivars, for both experiments across all percentile plant sizes.

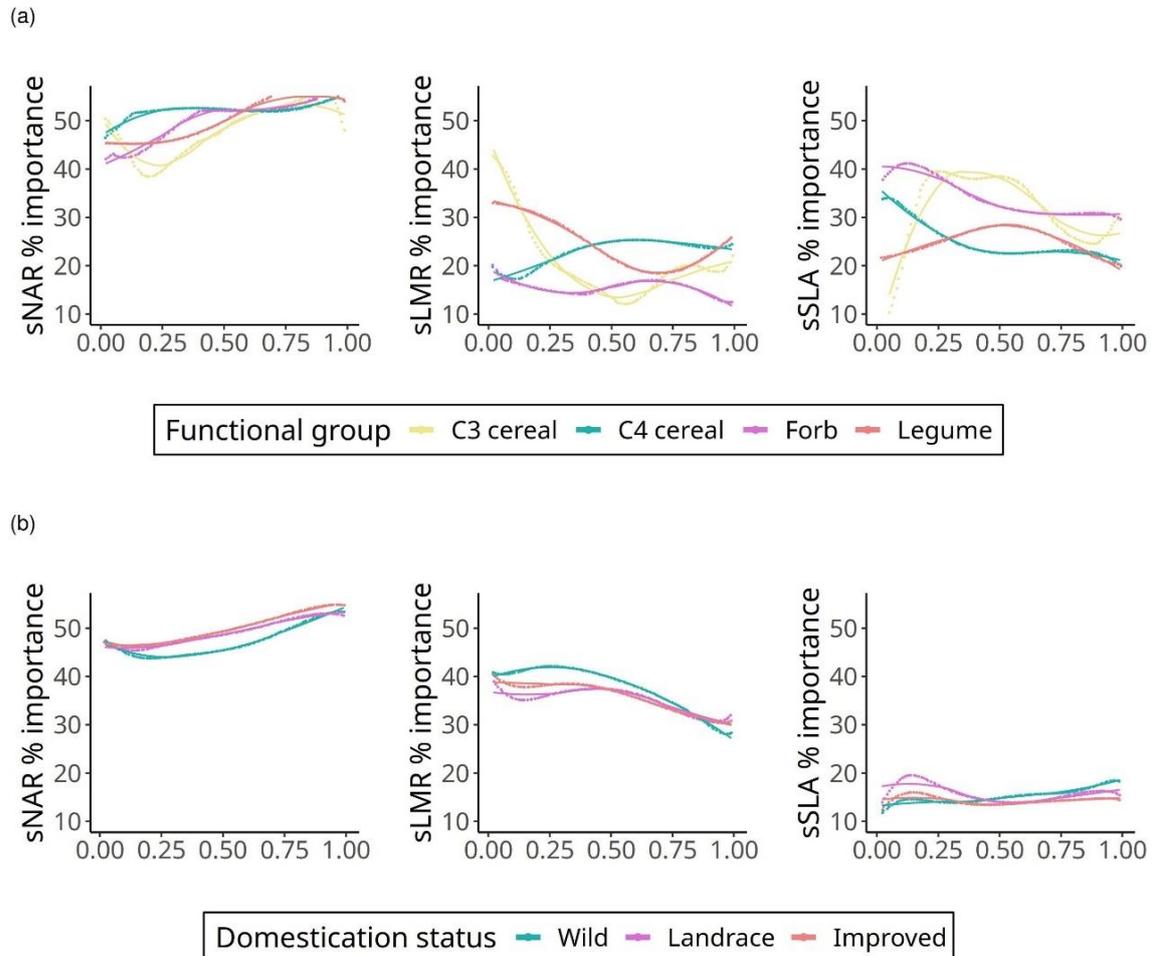
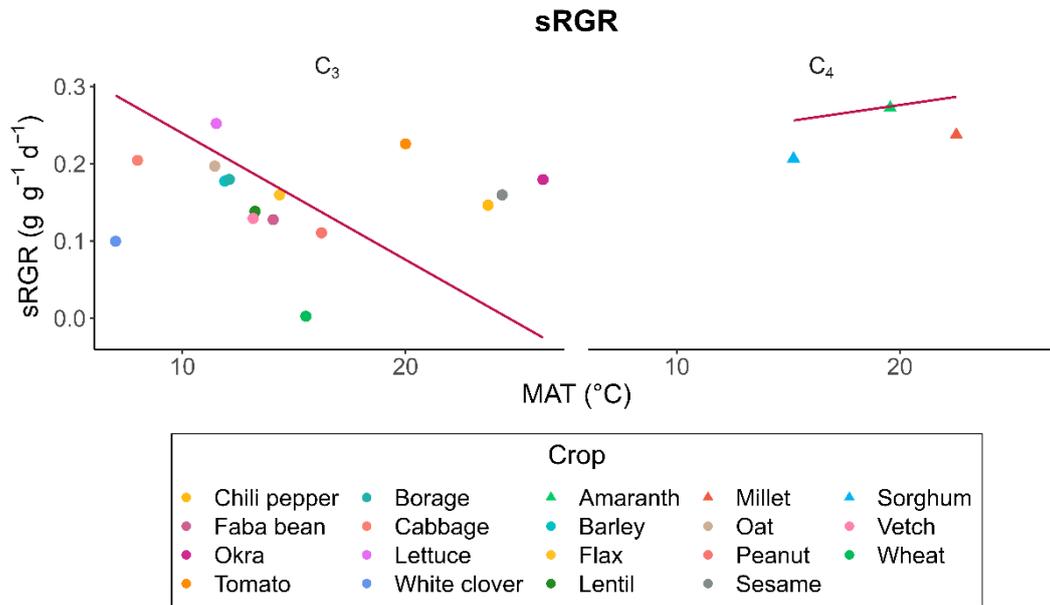


Fig. S7 Mean size-specific relative growth rate (sRGR) as a function of mean annual temperature (MAT) at crop origin and photosynthetic pathway (C_3 vs. C_4). Solid lines represent the fitted phylogenetic generalized least squares model (PGLS). Symbols represent the photosynthetic pathway: C_3 (circles) and C_4 (triangles).



Methods S1 Supplementary details on the estimation of total mass, leaf mass, and leaf area.

Linear regressions were performed to obtain prediction equations for total mass (IntotalM), leaf mass (InleafM), and leaf area (InleafA) using data from calibration plants (harvest_IN and harvest_EX for the *intensive* and *extensive experiments*, respectively). The final models for each experiment and response variable were:

INTENSIVE EXPERIMENT

1. Total mass calibration

$\text{lmer}(\text{IntotalM} \sim \text{Inheight} + \text{Incanopyd} + \text{Inleafn} + \text{time} + (1 + \text{Inheight} + \text{Incanopyd} + \text{Inleafn} | \text{acc_number}), \text{data} = \text{harvest_IN})$

2. Leaf mass calibration

$\text{lmer}(\text{InleafM} \sim \text{Inheight} + \text{Incanopyd} + \text{Inleafn} + \text{time} + (1 + \text{Inheight} + \text{Incanopyd} + \text{Inleafn} | \text{acc_number}), \text{data} = \text{harvest_IN})$

3. Leaf area calibration

$\text{lmer}(\text{InleafA} \sim \text{Intillern} + \text{Inleafn} + \text{Inleafl} + \text{time} + (1 + \text{Intillern} + \text{Inleafn} + \text{Inleafl} | \text{acc_number}), \text{data} = \text{harvest_IN})$

EXTENSIVE EXPERIMENT

1. Total mass calibration

$\text{lmer}(\text{IntotalM} \sim \text{Inheight} + \text{Incanopyd} + \text{Inleafn} + \text{Inleafl} + \text{Inbasald} + \text{time} + (1 + \text{Inheight} + \text{Incanopyd} + \text{Inleafn} + \text{Inleafl} + \text{Inbasald} | \text{sps_dom}), \text{data} = \text{harvest_EX})$

2. Leaf mass calibration

$\text{lmer}(\text{InleafM} \sim \text{Inheight} + \text{Incanopyd} + \text{Inleafn} + \text{Inleafl} + \text{Inbasald} + \text{time} + (1 + \text{Inheight} + \text{Incanopyd} + \text{Inleafn} + \text{Inleafl} + \text{Inbasald} | \text{sps_dom}), \text{data} = \text{harvest_EX})$

3. Leaf area calibration

$\text{lmer}(\text{InleafA} \sim \text{Incanopyd} + \text{Intillern} + \text{Inleafn} + \text{Inleafl} + \text{Inbasald} + \text{time} + (1 + \text{Incanopyd} + \text{Intillern} + \text{Inleafn} + \text{Inleafl} + \text{Inbasald} | \text{sps_dom}), \text{data} = \text{harvest_EX})$

where \ln_{height} is plant height (cm), \ln_{canopyd} is canopy diameter (cm), \ln_{tillern} is the number of branches, \ln_{leafn} is the number of leaves, \ln_{leafl} is the length of the largest leaf, \ln_{basald} is the diameter of the basal stem, and time is the number of days from sowing to harvest. Note that ‘ln’ stands for \log_e -transformed variables. In the *intensive experiment*, accession identity (acc_number) was considered as random effects, whereas in the *extensive experiment*, a combined variable between crop identity and domestication status (sps_dom) was used. All models were run with the `lmer` function of the ‘lme4’ R package (Bates *et al.*, 2015)³ with maximum likelihood (ML) estimation. Each of the final models was checked by plotting predicted values against observed values from the calibration plant data and calculating Pearson correlation.

³ **Bates D, Mächler M, Bolker BM, Walker SC.** 2015. Fitting linear mixed-effects models using lme4. *Journal of Statistical Software* 67: 1–48.

Methods S2 Details on the calculation of growth traits.

Note that for calculating RGR and its components, it is more convenient to work on a logarithmic scale. Therefore, we use lowercase letters to indicate \log_e -transformed variables (*e.g.* $\log_e(AL) = al$, $\log_e(RGR) = rgr$).

CALCULATION OF sRGR.

We calculated the size-specific RGR (sRGR) from the four-parameter logistic function using the 50th percentile of the total mass distribution (m) as the common size. For this function, the sRGR for a given individual can be written as follows:

$$sRGR_i = \frac{1/scal (m_{min} - m_c)(m_{max} - m_c)}{(m_{min} - m_{max})} \quad (\text{Eqn 1})$$

where m_{min} , m_{max} , and $scal$ are the free parameters of the function, and m_c is the common reference size. The parameters m_{min} and m_{max} are the minimum and maximum asymptotic m , respectively, and $1/scal$ is the slope at the inflection point of the curve (R function *SSfpl* in Pinheiro *et al.* (2020)⁴).

CALCULATION OF THE COMPONENTS OF sRGR

size-standardized RGR components were calculated from sRGR according to Rees *et al.* (2010)⁵. On logarithmic scales, sgr can be expressed as the sum of its components:

$$sgr = snar + slmr + sla \quad (\text{Eqn 2})$$

These components are functions of total mass (m), leaf mass (ml), and leaf area (al) as follows:

$$sgr = \log_e \left(\frac{1}{AL_C} \frac{dM}{dt} \right) + (ml_C - m_C) + (al_C - ml_C) \quad (\text{Eqn 3})$$

To calculate the contribution of each growth component to sgr , we first calculated the time (t_c) at which each focal plant reached the common reference mass (m_c) using the four-parameter logistic equation as follows:

⁴ Pinheiro J, Bates D, DebRoy S, Sarkar D, Team RC. 2021. nlme: linear and nonlinear mixed effects models. R package version 3.1-152.

⁵ Rees M, Osborne CP, Woodward FI, Hulme SP, Turnbull LA, Taylor SH. 2010. Partitioning the components of relative growth rate: how important is plant size variation? *The American Naturalist* 176: E152–E161.

$$t_c = xmid - \frac{1}{scal} \log_e \left(-\frac{m_{max}-m_c}{m_{min}-m_c} \right) \quad (\text{Eqn 4})$$

where m_{min} , m_{max} , $xmid$ and $scal$ are the free parameters of the curve and m_c is the common reference size. The parameters m_{min} and m_{max} are the minimum and maximum asymptotic m , respectively, $xmid$ is the time at which m is midway between the minimum and maximum asymptotes, and $1/scal$ is the slope at the inflection point.

Second, we estimated leaf mass (ml_c) and leaf area (al_c) at the common reference size by fitting the four-parameter logistic model to ml and al . For ml_c , the logistic model is given by:

$$ml_c = ml_{min} + \frac{ml_{max}-ml_{min}}{1+e^{(xmid-t_c)/scal}} \quad (\text{Eqn 5})$$

where ml_{min} , ml_{max} , $xmid$ and $scal$ are the free parameters of the curve and t_c is the time at the common reference size. The parameters ml_{min} and ml_{max} are the minimum and maximum asymptotic ml , respectively, $xmid$ is the time at which ml is midway between the minimum and maximum asymptotes, and $1/scal$ is the slope at the inflection point of the curve. For al_c , the logistic model is given by:

$$al_c = al_{min} + \frac{al_{max}-al_{min}}{1+e^{(xmid-t_c)/scal}} \quad (\text{Eqn 6})$$

where al_{min} , al_{max} , $xmid$ and $scal$ are the free parameters of the curve, and t_c is the time at the common reference size. The parameters al_{min} and al_{max} are the minimum and maximum asymptotic al , respectively, $xmid$ is the time at which al is midway between the minimum and maximum asymptotes, and $1/scal$ is the slope at the inflection point of the curve.

Finally, we used the estimates of ml_c and al_c to calculate the size-standardized lmr ($s\text{lmr}$) and sla ($s\text{sla}$) using equation 3. The value of nar at the common mass ($s\text{nar}$) was then estimated as $s\text{rgr} - s\text{lmr} - s\text{sla}$.