

1 This is a post-peer-review, pre-copyedit version of an article published in *Journal of*
2 *Applied Ecology*. The final authenticated version is available online at:
3 <https://besjournals.onlinelibrary.wiley.com/doi/10.1111/1365-2664.13027>

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8 **Farmer perception and utilization of leaf functional traits in managing**
9 **agroecosystems**

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27 **Running title:** Farmer perceptions of leaf functional traits

28 **Keywords:** Agroecology, agroforestry, coffee, functional traits, intraspecific trait
29 variation, Leaf Economics Spectrum, local ecological knowledge, shade trees, trait-based
30 ecology, visual elicitation tool

31

32 **Abstract**

33

34 1. Using knowledge of leaf functional traits, such as those forming the Leaf Economics
35 Spectrum (LES), to understand plant responses to environmental change is well
36 established and now being more widely applied to agroecosystems. Yet, little is known
37 about how farm managers invoke leaf functional traits to inform management decisions.

38

39 2. The objectives of this research were to i) evaluate if farmers use knowledge of
40 intraspecific trait variation (ITV) in LES traits (or trait proxies) of target crops as
41 response indicators of management conditions; ii) determine if LES trait values are
42 ranked consistently among multiple farmers along a “Farmer Leaf Economics Spectrum”
43 (FES); iii) evaluate how a FES corresponds to the LES; and iv) identify the farmer- and
44 farm attributes that best predict the agreement between the FES and the LES.

45

46 3. We collaborated with coffee (*Coffea arabica*) farmers in the Turrialba Valley, Costa
47 Rica. We used a visual elicitation tool of fresh leaves along an intraspecific spectrum of
48 leaf size, leaf thickness, and leaf colour (as a proxy for leaf nutrients); respondents were
49 asked to rank leaves in response to shade and nutrient scenarios as well as yield potential.
50 On-farm biophysical data, management practices, and socio-economic attributes were
51 also collected.

52

53 4. The majority of farmers demonstrated a developed system of utilizing coffee leaf- and
54 whole-plant ITV as indicators of management practices. Farmers managing smaller farms
55 tended to more commonly acknowledge ITV in LES chemical-morphological traits, as
56 compared to those managing large farms. The agreement between a respondent-identified
57 ranking of leaf thickness ITV as a function of light environment, and an empirically-
58 defined thickness-to-light ranking, was partially explained by farmers’ physical
59 engagement with plants.

60

61 5. *Synthesis and implications.* In scientific literature, analyses of crop intraspecific trait
62 variation have provided important insights into the mechanistic bases of multiple key

63 agroecological processes. We demonstrate that farmers use crop leaf variation as an
64 indicator to both evaluate management prescriptions, and to initiate management actions
65 including shade-tree species selection and abundance, crop- and shade-tree pruning
66 regimes, and fertilization treatments. These findings signify that functional traits
67 represent a key nexus between scientific and local knowledge.

68

Accepted

69 **Introduction**

70 **Plant functional traits in agroecosystems**

71 Diversified agroecosystems are increasingly being recognized as an
72 environmentally and economically sustainable alternative to biologically simplified
73 conventional agricultural systems (Tschardt *et al.* 2005; Tomich *et al.* 2011).
74 Commonly, greater diversity in agroecosystems stimulates nutrient capture and cycling
75 processes leading to a reduced reliance on fertilizers (Drinkwater & Snapp 2007; Isaac &
76 Kimaro 2011), while simultaneously enhancing the provisioning of other ecosystem
77 services such as carbon (C) sequestration, improved hydrological cycling, or enhanced
78 maintenance of biodiversity across multiple trophic levels (Jose 2009). In agroecological
79 research and management, on-farm diversity has traditionally been described based on
80 taxonomy (i.e. number of crop species or varieties) or the range of plant functional types
81 (e.g. number of annuals vs. perennials; herbaceous vs. woody crops; N₂-fixing legumes
82 vs. non-N₂-fixing species). However, in diverse agroecosystems, theory and observations
83 suggest that describing agricultural diversity based on plant functional trait variation is
84 likely to offer mechanistic insights into agroecological processes, including (but not
85 limited to) plant responses to environmental change, determinants of soil nutrient cycles,
86 and rates of species interactions such as pollination (Garnier & Navas 2012; Martin &
87 Isaac 2015; Wood *et al.* 2015).

88 There has been considerable work identifying the morphological, physiological,
89 and phenological traits of plants (and crops) that drive differences in plant performance
90 across environmental gradients (e.g. Violle *et al.* 2007). Leaf functional traits in
91 particular have been shown to be critical predictors of aboveground biomass (e.g. Diaz *et*
92 *al.* 2007), litter decomposition (e.g. Santiago 2007), and ultimately yield (Gagliardi *et al.*
93 2015); as a result, certain leaf traits are included, in some form, into most regional or
94 global assessments of agroecosystem structure and function (e.g. Box 7-1 in Porter *et al.*
95 2014).

96 In the functional ecology literature, the concept of a global functional leaf trait
97 spectrum – a continuum that describes universal patterns of leaf morphological,
98 physiological, and chemical trait co-variation among all plants worldwide – was arguably
99 most widely popularized by publication of the “Leaf Economics Spectrum” (LES)

100 (Wright *et al.* 2004). The LES describes co-variation between certain leaf functional traits
101 (namely, leaf mass per area (LMA), leaf nitrogen (N) and phosphorus (P) concentrations,
102 maximum photosynthetic capacity, leaf respiration, and leaf lifespan) that occurs along a
103 single axis, or spectrum, which in turn describes trade-offs among resource acquiring
104 traits and resource conservation traits (Westoby *et al.* 2002; Wright *et al.* 2004; Chave *et*
105 *al.* 2009; Reich 2014; Diaz *et al.* 2016). While foundational papers on the LES and other
106 trait spectra (e.g. stems, roots, and whole-plant functional traits) have been based on trait
107 comparisons across hundreds of species (e.g. Wright *et al.* 2004), recent work has shown
108 that certain spectra also describe patterns of intraspecific trait variation (ITV) among
109 plants within a species (e.g. Hajek, Hertel & Leuschner 2013; Gagliardi *et al.*, 2015;
110 Niinemets, 2015; Isaac *et al.* 2017). For example, recent studies on ITV in coffee (*Coffea*
111 *arabica*) – the focus of our study here – suggest that the LES also describes functional
112 variation in crops of the same variety (Martin *et al.* 2017).

113 **Plant functional traits in action**

114 Functional trait-based approaches have not been widely applied to crops and
115 agroecosystems. This is despite the fact that trait-based research, particularly research
116 that focuses on ITV in crops, could make significant contributions to understanding the
117 effects of management and changes in the farm environment on productivity and general
118 plant performance (Martin & Isaac, 2015). Certain traits do factor heavily into applied
119 research on plant breeding, which largely focuses on how long-term trends in crop
120 domestication are due to selection for suites of “domestication traits”, including yield,
121 nutritional quality, harvest factors (such as shattering in grains), or propagation (Doebley,
122 Brandon & Smith 2006; Meyer, DuVal & Jensen 2012). However, farmers are also
123 known to both evaluate the impacts of management and make management decisions
124 based on other observable and tactile traits, such as leaf colour, size, and thickness, or
125 phenology (Gibson *et al.* 2008; Mwangi *et al.* 2011) that are not necessarily part of a
126 domestication syndrome. For example, in coffee agroforestry systems in Central America,
127 farmers commonly select shade trees through an understanding of traits such as leaf
128 texture and size, foliage density, and rooting patterns (Cerdán *et al.* 2012), while in cocoa
129 agroforestry systems in West Africa, farmers select shade trees based on observable leaf
130 traits that are critical for ecosystem processes including organic matter accumulation

131 (Isaac, Dawoe & Sieciechowicz 2009). Similarly, in rice production systems, studies have
132 found that farmers select varieties based on traits associated with plant phenology, which
133 has greater implications for the timing of harvests in response to market prices, as
134 opposed to strictly yield maximization (Dalton 2004).

135 Research further suggests that the degree to which farmers make decisions based
136 on functional traits that are not strictly related to yield, is contingent on management
137 practices. For example, Isaacs *et al.* (2016) showed that farmers placed plant architectural
138 traits as the second most important factor in breeding programs, just below yield
139 potential; the authors also showed that the use of plant traits in crop selection other than
140 yield alone, was of two-fold-greater importance for farmers managing diverse
141 intercropping systems as compared to those managing monocultures. There is reason to
142 expect farmers who manage for greater on-farm diversity, are also likely to manage for
143 multiple simultaneous agroecosystem functions such as C storage and nutrient cycling;
144 ecosystem functions that require the selection and management of plant characteristics
145 beyond just yield.

146 **Plant functional traits in perception**

147 Despite a large and growing number of independent analyses on i) ITV in the
148 ecological literature, and ii) local knowledge of functional traits for agricultural
149 management and decision-making, there are few studies to date that have integrated these
150 two fields (Martin & Isaac 2015, Wood *et al.* 2015). Applied understandings of
151 functional traits may not be precisely analogous to the ecological literature, but farmers
152 may employ knowledge of certain plant characteristics that approximate the traits
153 commonly evaluated in ecological literature such as the LES. For example, from a strict
154 empirical perspective, leaf N concentration determinations demand some type of
155 elemental analysis (Pérez-Harguindeguy *et al.* 2013), but farmers may make decisions
156 based on this trait as approximated by leaf colour (Friedman, Hunt Jr. & Mutters 2016).
157 Similarly, there is evidence that farmers evaluate plant response based on leaf texture, or
158 how “soft” or “hard” a leaf feels to the touch (Cerdán *et al.* 2012). While texture may not
159 be a correlate of the LES *per se*, from a functional perspective it may approximate LMA
160 or other ecologically important measures of leaf morphology such as thickness
161 (Niinemets 2001) or toughness (Westbrook *et al.* 2011). There is therefore reason to

162 expect that certain traits ecologists have focused on when exploring plant community
163 structure and function – namely LES traits – also factor prominently into the development
164 and deployment of local knowledge surrounding farm management decision making
165 (Clinquart 2010; Humphries *et al.* 2015; Isaacs *et al.* 2016).

166 To address this expectation, we collaborated with coffee farmers in the Turrialba
167 Valley, Costa Rica. Our research objectives were to i) evaluate if farmers use target crop
168 intraspecific trait variation (ITV) in LES traits (or trait proxies) as response indicators of
169 management conditions; ii) determine if LES traits are ranked consistently among
170 multiple farmers along a “Farmer leaf Economics Spectrum” (FES); iii) evaluate how a
171 FES corresponds to the LES; and iv) identify the farmer and farm attributes that best
172 predict the agreement between the FES and the LES. We hypothesize that ITV in coffee
173 leaf traits are perceived and utilized by farmers as response indicators of management
174 decisions in coffee agroecosystems. We also expect that the agreement on leaf trait
175 expression under various light and nutrient environments between a FES and LES, is
176 explained by key management-related characteristics including the size of the farm, the
177 level of on-farm biodiversity, the degree of input intensification, and the degree to which
178 farmers report touching leaves of coffee plants.

179 We wish to emphasize that our study is not designed as a verification or test of the
180 “accuracy” of local knowledge, but rather to investigate an interchange between i)
181 established general ecological theory, and ii) the highly nuanced and lived local
182 knowledge. While these two bodies of knowledge vary both spatially and temporally
183 (Brook & McLachlan 2005), and impressing scientific boundaries on this exchange
184 creates inherent power dynamics, we impose the leaf as the unit of analysis within this
185 study, with the full recognition that we are confining our analysis based on the establish
186 LES. In doing so, it is our explicit intention to locate complementarity among knowledge
187 systems (Altieri 2004; Richards 1985) in order to inform our understanding of
188 agroecosystem management and function.

189 **Materials and Methods**

190 **Study region**

191 This study was conducted in the Cantón Turrialba, Cartago Province of Costa
192 Rica (9.7960° N, 83.5497° W). Within this region, coffee is produced at both low and

193 high altitudes (range 690-1100 m a.s.l.) and across a range of management practices from
194 monocultures, to highly diverse shade coffee agroforestry systems. Long-term
195 consequences of on-farm nutrient depletion and unpredictable rainfall patterns are also
196 prevalent in the region (Cerdán *et al.* 2012), and farmers in the region are currently
197 contending with the devastating effects of coffee leaf rust caused by the fungus (*Hemileia*
198 *vastatrix*): a widespread leaf fungal epidemic throughout Central and South America that
199 has occurred within the last decade (Avelino *et al.* 2015). Amid low soil fertility and
200 constraints to, or undesirability of, fertilizer use, farmers have developed techniques to
201 promote soil and crop nutrition, primarily through the maintenance of shade tree
202 intercropping and predominantly with the commonly occurring shade tree *Erythrina*
203 *poeppigiana* (Fabaceae).

204 The effects of shade trees on coffee have been fairly well described with respect
205 to bulk soil nutrient pools and aboveground coffee growth, yield, and physiology (Beer *et*
206 *al.* 1998; DaMatta, 2004; Harmand *et al.* 2007; Frank & Vaast 2009; Munroe *et al.* 2015).
207 In the region, shade trees moderate adverse microclimatic conditions, stabilize year-to-
208 year coffee yields, boost yields under poor edaphic and climatic conditions, extend coffee
209 plant longevity, and improve nutrient availability for coffee plants, particularly in low or
210 moderate fertilizer systems (Beer *et al.* 1998; Harmand *et al.* 2007; Haggard *et al.* 2011;
211 Gagliardi *et al.* 2015). The incorporation of dinitrogen (N₂)- fixing shade trees in the
212 region results in significant inputs of N (246-340 kg N ha⁻¹ year⁻¹ (Haggard *et al.* 2011;
213 Beer *et al.* 1998)) with additional contributions from pruned shade tree litter (Munroe &
214 Isaac 2014). Ultimately though, coffee farms in the region are variable in shade tree
215 density and composition.

216 **Respondent and biophysical attributes**

217 To identify farmer participants, we worked with the farmer network of the
218 “Ecosystem-based Adaptation for Smallholder Subsistence and Coffee Farming
219 Communities in Central America” (CASCADE) project, led by Conservation
220 International and the Tropical Agricultural Research and Higher Education Center
221 (CATIE) (Cerdeira *et al.* 2017). Specifically, our study identified 25 randomly selected
222 participants from the broader CASCADE group of farmers. Potential respondents in this
223 region were selected randomly at two levels, for the original CASCADE project, and

224 from the CASCADE network for our case. The 25 respondents in this study received no
225 substantial or exclusive training in comparison to other farmers in the region of study,
226 though the general proximity to CATIE may enhance the agricultural knowledge base of
227 these farmers in comparison to other regions of Costa Rica. This research received ethics
228 approval from the Social Sciences, Humanities and Education Research Ethics Board,
229 University of Toronto for research involving human participants. Informed consent was
230 secured in advance of every interview.

231 At each farm semi-structured interviews were conducted. We collected data from
232 each respondent on the number of years farming (Y), the age of coffee (A), the number of
233 coffee varieties planted (V), and level of inputs, which included both “fertilizers”
234 (applications per year; F) and “total practices” (the sum of all practices including F, as
235 well as pesticide inputs, herbicide inputs, weeding and pruning of coffee and shade trees;
236 P). We also operationalized an “engagement with leaves” (E) variable, coded as a binary
237 response to whether or not respondents reported touching coffee leaves on their farms,
238 when inspecting the farm and making management decisions and prescriptions. Field
239 surveys of each coffee farm were also conducted to collect data on farm size (L; ha),
240 coffee planting density (D; plants ha⁻¹), and shade tree composition which were converted
241 into two diversity indices; Shannon Index (*H*) and Simpson Index (*D*) as well as a
242 categorical shade class (S) which entailed classes categorized into i) full sun (no shade
243 trees) ii) coffee intercropped only with *Erythrina poeppigiana*, or iii) coffee intercropped
244 with diverse shade trees (up to seven species per farm). Of these continuous variables *H*,
245 *D*, P, A, and V met assumptions of normality (Shapiro-Wilk test *p*-value > 0.05), Y was
246 marginally non-normal and not transformed prior to analysis (Shapiro-Wilk *p* = 0.04), F
247 could not be transformed to meet normality but did visually approximate a normal
248 distribution and was not transformed (Shapiro-Wilk test *p* = 0.01), while L and V were
249 log-transformed prior to analysis to improve normality (Shapiro-Wilk test *p* = 0.4 and
250 0.02, respectively on log-transformed data).

251 **Semi-structured interviews: setting the leaf spectra**

252 Following completion of questions on farming characteristics and practices, we
253 used a visual elicitation tool based on empirically framed researcher-created visual data
254 (Prosser & Loxley 2008). The visual elicitation tool, or the “leaf book”, was based on a

255 collection of fresh *C. arabica* leaves displaying a range of ITV (Fig. 1). We used four leaf
256 spectra that closely link with ecological studies on the importance of leaf traits, including:
257 1) coffee leaf size, 2) coffee leaf thickness, 3) a coffee leaf colour trait (as an
258 approximation of leaf nutrition), and 4) multiple coffee leaf traits in relation to yield
259 potential. We used *C. arabica* var. Caturra, as it is the most dominant variety of coffee in
260 Central America (McCook & Vandermeer 2015). For each interview day, a new leaf
261 book was made (based on data described below), to ensure that all respondents in the
262 study were presented with directly comparable leaves. Leaves in the leaf book were not
263 adjusted to individual farms, in order to maintain comparability of responses across all
264 respondents.

265 In order to bound an appropriate coffee leaf size spectrum, our leaf size values
266 were based on a collection of 216 fully expanded coffee leaves at the same location in the
267 coffee canopy (i.e. two-thirds the height of the plant) under monoculture and shade
268 treatments, in order to collect maximum ITV of leaf size within the study region. We
269 measured leaf area (cm²) using digital photos analyzed with Image J software (Abramoff,
270 Magalhaes & Ram 2004). Our leaf area ranged 4-fold from 14.6 to 64.7 cm², with a mean
271 of 39.0 cm². We selected six leaves covering this range (minimum and maximum values
272 as well as the 20th, 40th, 60th and 80th percentiles) in leaf area, which were in
273 turn associated with decreasing light availability from full sun monoculture to high shade
274 agroforestry. This is supported by Gagliardi *et al.* (2015) in a study from the same
275 species, where the authors report a ~4 fold increase in coffee leaf area under shade
276 canopies versus leaf area under full sun monoculture.

277 The leaf thickness spectrum was based on the same 216 leaves. A low-force
278 micrometer (No. 227-101, Mitutoyo Co., Japan) was used to measure leaf thickness
279 across a given leaf while avoiding major veins. Leaf thickness ranged from 0.23-0.33
280 mm, which corresponded to light environments ranging from high shade agroforestry to
281 full sun monoculture. We selected six leaves covering the minimum and maximum
282 values as well as 20th, 40th, 60th and 80th percentiles in leaf thickness in order to capture
283 the full potential range of leaf thickness associated with increasing light availability. Both
284 Morais *et al.* (2004) and Fahl *et al.* (1994) report such an increase in coffee leaf thickness
285 under sun versus under shade (due to increase size of palisade and spongy parenchyma

286 tissue under sun conditions). Similarly, Martins *et al.* (2014) report an increase of 15% in
287 thickness of coffee leaves under sun versus under shade.

288 Coffee leaf colour was not based on the collection of 216 leaves, but was rather
289 determined by locally developed colour patterns which were based on an established
290 coffee nutrient deficiency chart for N (published by the Instituto del Cafe de Costa Rica),
291 as well as for P and zinc (Zn), two other common deficiencies in coffee plantations. We
292 displayed a spectrum of leaves ranging from deficient to sufficient status. We selected
293 five leaves ranging from green to yellowing veins (indicative of N deficiency), or
294 exhibiting red coloration (indicative of P deficiency) or inter-vein fading (indicative of
295 Zn deficiency).

296 The leaf-based yield potential spectrum was based on multiple studies. Gagliardi
297 *et al.* (2015) showed that multivariate suites of leaf traits were significantly related to
298 plant-level reproductive output, but in general along the LES, coffee plants associated
299 with higher LMA were associated with greater reproductive output as compared to those
300 with low LMA. Therefore, our high leaf-based yield potential end of the spectrum
301 depicted mid-range to larger green leaves with average thickness. DaMatta (2004)
302 showed that smaller leaf area, particularly under drought, corresponds to lower yield due
303 to modified C partitioning to plant tissue in pot-grown *C. arabica*. Therefore, the lower
304 end of the leaf spectrum associated with leaf-based yield potential was populated with
305 smaller leaves exhibiting yellowing in their venation.

306 **Semi-structured interviews: respondent rankings**

307 We asked a set of questions with regards to ranking the importance of leaf and
308 plant characteristics for farm decision-making. Specifically, respondents were asked to
309 rank a suite of coffee leaf traits (i.e. leaf size, colour, and thickness) and a suite of whole-
310 plant traits (i.e. coffee plant height, sprout diameter, and crown radius), in order of
311 importance for making decision on plant and farm management.

312 Following a pre-tested semi-structured interview guide, respondents were also
313 asked to rank leaves in relation to a specific environmental scenario. For leaf size, we
314 described various light environments (a six-point qualitative gradient ranging from “high
315 shade” to “no shade”) and respondents were asked to rank leaves in relation to shade
316 growing conditions. Similarly for leaf thickness, we operationalized light environments

317 (high shade to no shade gradient) and respondents were asked to rank the six leaves in the
318 leaf thickness spectrum that correspond to high shade growing conditions to low shade
319 growing conditions. For leaf nutrients, we operationalized soil fertility (a five-point
320 qualitative gradient ranging from “high soil fertility” to “low soil fertility”) and
321 respondents were asked to rank the five leaves in the leaf nutrient spectrum that
322 correspond to sufficient to deficient conditions. For yield potential, respondents were
323 asked to rank the six leaves in the leaf-based yield potential spectrum that correspond to a
324 six-point qualitative gradient ranging from “high yielding” to “low yielding” plants.
325 These scenarios were used to delimit the spectra in order to achieve a ranking from
326 respondents and therefore, define a FES. Following this ranking, questions and prompts
327 were used to gather information on reasons for the selection.

328 **Statistical Analysis**

329 We use the concept of “agreement” in order to evaluate the agreement between
330 respondent-identified leaf rankings and empirical leaf rankings. For the size, thickness
331 and nutrient leaf spectra, we used a cumulative proportion agreement where the
332 proportion of agreement between i) the empirically-defined leaf trait values and ii) the
333 respondent-identified leaf rank in response to described environmental scenarios per
334 respondent, were recorded. For yield potential, we conducted a Spearman rank
335 correlation between i) empirically-supported rankings of leaves to yield potential, and ii)
336 respondent-identified ranking of leaves to yield potential. All response data was normally
337 distributed.

338 We then used a backwards-stepwise linear modelling procedure to identify the
339 strongest farmer- and farm-level variables predictors of the agreement between the
340 respondent rank and the established rank. For this analysis, the full model was of the
341 form:

$$342 \rho_i = \beta_0 + \beta_1 \log L_i + \beta_2 A_i + \beta_3 D_i + \beta_4 Y_i + \beta_5 \log V_i + \beta_6 E_i + \beta_7 F_i + \beta_8 P_i + \beta_9 H_i + \beta_{10} D_i + \beta_{11} S_i$$

343 (Equation 1)

344 where ρ_i represents the agreement between how the i^{th} respondent ranked leaves, and the
345 empirical rankings of leaf traits based on empirical analysis (quantified as either a
346 Spearman rank correlation coefficient (for the yield potential spectrum) or a proportion of
347 agreement (for size, thickness, and colour spectra)), β_0 represents an overall model

348 intercept, and β_1 through β_{11} represent the parameter estimates for L, A, D, Y, V, E, F, P,
349 *H*, *D* and *S*, respectively, all of which were measured on the i^{th} farm. Collinearity among
350 all pair-wise sets of predictor variables was initially assessed with correlation analysis; of
351 the 55 possible pairwise relationships only six were statistically significant (Spearman
352 rank correlation $p \leq 0.05$; Table S1) so all variables were included in our full model.
353 Models were then compared using Akaike's Information Criteria (AIC), with the lowest
354 AIC score indicating the most parsimonious model fit. Significance of predictor variables
355 in each AIC-selected model was then assessed using multiple regression.

356 We then used a principal component analysis (PCA) the 'vegan' R package
357 (Oksanen *et al.*, 2013), to evaluate responses in multivariate space. Specifically, for each
358 respondent, there were four different values of agreement (i.e. ρ_i in Equation 1), which
359 were evaluated in multivariate space. We then used a one-tailed *t*-test to determine if
360 PCA axis one scores differed significantly across farms of different sizes (i.e. between
361 small (< 1 ha) vs. large (≥ 1 ha) farms). All statistical analyses were performed using R v.
362 3.0.2 (R Foundation for Statistical Computing, Vienna, Austria).

363 **Results**

364 **Respondent attributes**

365 The majority of respondents maintained on average $2.8 (\pm 2.1$ s.d.) coffee
366 varieties, on farms ranging in size from very small farms (0.1 ha) to moderate sized farms
367 (4 ha) (Table 1). On-farm shade tree communities were composed primarily of *E.*
368 *poeppigiana*, and tended toward higher diversity rather than monocultures (Table 1).
369 Respondents experience with farming ranged from a low of 4 years, to a maximum of 86
370 years (Table 1). Over half of respondents (60%) reported evaluating leaf traits as a
371 regular process during farm inspection (Table 1).

372 **Important leaf- and plant traits**

373 Respondents indicated that the most important leaf- and whole-plant traits when
374 prescribing management practices were the leaf colour and the coffee canopy radius,
375 respectively (Table 2). Surprisingly, plant height was less important than plant canopy
376 radius in decision-making, and sprout diameter was considered the least important trait to
377 farmers when making management decisions (Table 2).

378 Certain traits ranked consistently among multiple farmers along an FES (Table 3).
379 Well over half of respondents (64%) ranked the largest leaf in the leaf size spectrum as
380 corresponding to a low light/ high shade environment, while 56% of respondents agreed
381 that the smallest leaf in the size spectrum corresponded to a high light/ no shade
382 environment (Table 3). The thickest leaf was identified by 68% of respondents as
383 associating with a low light/ high shade environment. In terms of leaf nutrition traits,
384 80% of respondents identified a large green leaf as an indication of sufficient nutrient
385 supply, while 68% identified small yellow leaves as belonging to nutrient deficient
386 conditions (Table 3). Over 60% of respondents agreed that the largest green leaf in the
387 spectrum was associated with the highest yield potential (Table 3).

388 **Selecting leaf traits based on management scenarios**

389 *Leaf size*

390 The AIC-selected model that best predicted the agreement between i) the
391 empirically-defined leaf trait values and ii) the respondent-identified leaf rank in response
392 to described leaf size-light environment relationships, included only farm size (L) ($r^2 =$
393 0.146; Table 4). With a negative coefficient, this indicates that farmers managing larger
394 farms tend to have a lower agreement between the ranks of leaf size-light environment
395 relationships (Table 4).

396 *Leaf thickness*

397 The agreement between i) the empirically-defined leaf trait values and ii) the
398 respondent-identified leaf rank in response to described leaf thickness-light environment
399 relationships, was most parsimoniously predicted as a function of eight variables: the age
400 of coffee plants (A), the coffee density (D), the number of coffee varieties (V), whether a
401 respondent was physically engaged with the coffee leaves (E), the level of fertilizer use
402 (F), both shade tree diversity indices (H and D) as well as shade tree class ($r^2 = 0.213$;
403 Table 4). Whether or not a respondent physically touched leaves when inspecting plants
404 had a significantly positive effect on the likelihood of agreement between the ranks of
405 leaf thickness-light environment relationships (Table 4). Similarly, respondents with
406 older coffee plants and higher shade tree diversity tended to have a more positive
407 likelihood of agreement between the ranks, as compared to farmers with younger plants.

408 In contrast, higher reported fertilizer use was associated with lower agreement between
409 the respondent-identified and empirically-defined ranks (Table 4).

410 *Leaf colour*

411 Farmer and farm-level attributes that best predict the agreement between a
412 respondent-identified and empirical-defined rank of leaf colour to soil nutrient status,
413 were farm size (L) and the sum of all practices (P), based on the most parsimonious AIC
414 selected model ($r^2 = 0.170$; Table 4). Farm size (L) was significantly negatively
415 associated with the agreement between ranks, such that the smaller the farm, the higher
416 likelihood a farmer will match observed leaf nutrition requirements to documented leaf
417 nutritional deficiencies. The number of practices was positively related to the proportion
418 of agreement between the ranks (Table 4).

419 *Yield potential*

420 The correlation between a respondent-identified ranking of leaves in relation to
421 yield potential, and an established ranking of leaves based on empirically assessed
422 relationships between leaf traits and yield, was most parsimoniously predicted by the
423 number of varieties (V), the number of years farming (Y), the sum of practices (P) and
424 the shade class ($r^2 = 0.263$; Table 4). There was a significantly negative association
425 between overall number of practices and the agreement of the two rankings; the less
426 practices a farmer reported using, the stronger the correlation (Table 4).

427 *Effects of farm size*

428 The first two PCA axes explained 79.3% of the variation in the four response
429 variables (Fig. 2). PCA axis 1 explained the majority of this variation (46.5%), while
430 PCA axis 2 explained a further 32.8% of the variation in the four response variables. PCA
431 axis 1 scores were significantly related to three of four responses including the agreement
432 between farmer rankings and empirical rankings of i) leaf area, ii) leaf thickness, and iii)
433 leaf colour ($r^2 = 0.80, 0.59, \text{ and } 0.41$, respectively, and $p < 0.001$ in all cases, as per linear
434 regression analyses performed between these variables and PCA axis 1 scores). PCA axis
435 2 was significantly related to the agreement between farmers and empirical rankings of
436 yield potential ($r^2 = 0.80, p < 0.001$; Fig. 2). PCA 1 axis scores differed significantly
437 across small and large farms (one-tailed $t = 1.92, p = 0.035$), indicating that respondents
438 with smaller farms generally showed higher agreement between their ranking of leaves

439 and the empirical ranking of leaves (Fig. 2).

440 **Discussion**

441 **The breadth and depth of farmer knowledge on ITV**

442 There is increasing interest in using plant functional traits to better understand the
443 causes and consequences of environmental change on plants in agroecosystems (Garnier
444 & Navas 2012; Martin & Isaac 2015; Wood *et al.* 2015). We contribute to this growing
445 literature with an examination of farmer use of plant traits, namely, the use of ITV in LES
446 traits for on-farm evaluation and decision-making. While farmer selection of shade trees
447 traits in coffee agroforestry systems is well documented (Cerdán *et al.* 2012; Valencia *et*
448 *al.* 2015), we show that leaf ITV in coffee is also recognized by farmers, and employed in
449 farm management. In our study, the majority of farmers demonstrated a developed
450 system of utilizing coffee leaf and whole-plant ITV as response indicators of
451 management practices.

452 To be relevant for management, LES traits or their proxies must be visible and
453 distinguishable by a manager, insofar as they are physical, mutable and tractable. The
454 “soft” traits that comprise part of the LES (cf. Wright *et al.* 2010), namely LMA and
455 (arguably) leaf N, are not immediately analogous with those visible traits described as
456 important by farmers, but LMA does represent certain proxies of observable leaf traits
457 such as leaf size and thickness trade-offs. However, leaf colour was identified as the most
458 important leaf ITV characteristic used as an indicator of crop response to environmental
459 and management conditions (Table 2). This is not unexpected as farmers typically use
460 leaf colour to identify both systemic issues (e.g. nutrient deficiencies) and localized
461 issues (e.g. disease).

462 **Agreement between the LES and the FES**

463 The range of agreement between the LES and the FES (Table 3) highlights the
464 discrete nature of the perception and use of leaf traits in farm management. Our
465 predictive models provide evidence that differences among farmers in how they evaluate
466 leaf traits in response to environmental and management conditions, is at least partially
467 explained by systematic differences in farmer- and farm attributes (Table 4). Though the
468 size of the farm, the sum of practices, and the number of coffee varieties appear as

469 significant predictors in several predictive models, no one farmer- or farm attribute was
470 consistently retained in all AIC-selected models across all traits (Table 4).

471 Given the established axis of resource acquisition to resource conservation
472 hypothesized by the LES, leaf traits that are observable by farmers may be used to
473 diagnose the relationship with agroecological processes. While a farmer-defined FES was
474 evident, for instance relatively larger leaves were widely associated with a high shade
475 environment (Table 3), our study demonstrates that an understanding of how plants
476 respond to environmental conditions in scientific literature does not necessarily
477 correspond to how farmers or managers interpret ITV or plant ecological strategies. For
478 instance, respondents associated thick leaves with high shade environments (Table 3).
479 However, peer-reviewed studies suggest coffee leaf thickness decreases with increasing
480 shade due to increase size of palisade and spongy parenchyma tissue under sun conditions
481 (Martins *et al.* 2014; Fahl *et al.* 1994). Therefore, though 68% of respondents allocated
482 thick leaves to shade conditions, thus defining an FES for leaf thickness, it is not in
483 accordance with empirically derived relationships between leaf thickness and light
484 availability in, specifically, coffee (Martins *et al.* 2014).

485 The size of farm appears as significant in multiple models and was consistently
486 associated with the agreement between the rankings: respondents with smaller farms (<1
487 ha) showed higher coordination between the LES and their FES, as compared to
488 respondents on larger farms (≥ 1 ha) (Table 4; Fig. 2). Notably, this PCA axis 1 was
489 significantly related to three of the four response variables and was linked with farm size
490 (Fig. 2). Whether this is a function of i) experience – those with smaller farms have more
491 opportunity to engage with coffee plants, – or ii) labour – those with larger farms are
492 more likely to hire labour, or are burdened with administrative management, and
493 therefore reduce their engagement with coffee plants (Laffourcade 2012; Hugøy 2016) –
494 demands further exploration. In most cases, when farmers are less engaged on larger
495 farms, standardized management protocols were used. Specifically, the agriculture
496 ministry in the Turrialba region deploys messages that recommend broadly applicable
497 spraying and fertilizing schedules.

498 As expected, physical engagement with coffee plants was retained as a predictor
499 in terms of how farmers ranked leaf thickness in relation to light environment (Table 4).

500 Unlike leaf size, which is visually observable, leaf thickness requires a physical
501 engagement with the plant. However, despite potential benefits of doing so, not all
502 farmers reported physically engaging with plants to assess leaf attributes. Farmers who
503 report touching their leaves during farm inspection also tended to rank leaf thickness in
504 agreement with the empirically-defined rank for coffee leaf thickness. This finding
505 supports Ingold's (1992) assertion that knowledge production of the environment is
506 formed through action. Farmer perception of leaf traits is presumably formed through
507 action, seeing, or being engaged in the experience with plants, as described by Ingold
508 (1992), Okely (2001), and Hugøy (2016). Farmers have clear and coherent ways of
509 understanding the diversity, services, and yield potential on their coffee farms (Cerdán *et*
510 *al.* 2012). Based on our results here, this knowledge is at least partly derived from a
511 physical engagement with the plant.

512 **The role of plant characteristics in management actions**

513 We make the case here that farmers use crop ITV in leaves as “response traits”
514 that are indicators of how the coffee plant is responding to the agroecological
515 environment. The accumulation of farm management (here represented as fertilizers
516 inputs, pesticide and herbicide application, weeding, and pruning of coffee and shade
517 trees) synchronizes farmers along an axis of agricultural intensification. This
518 characteristic is a significant negative predictor of the correlation of leaf ranking for yield
519 potential (Table 4); in other words, farmers with less intensified farms matched the LES
520 and the FES for yield potential, as compared to those with more intensified farms. Such
521 low input agricultural systems are repeatedly linked with higher diversity (Tscharrntke *et*
522 *al.* 2005), given that species interactions often replace external inputs. For instance, in our
523 study region, Cerda *et al.* (2017) found that diversified coffee agroforestry systems
524 maintain higher soil potassium content than either simple coffee agroforestry systems (e.g.
525 one shade tree species only) and coffee monocultures, and some of these shade trees are
526 leguminous species, contributing to soil fertility. Here, we show that farmers in this
527 region with such highly diverse farms use a well-structured system of leaf ITV as a
528 response indicator to management.

529 Employing ITV as indicators of management may result in action. For instance, in
530 the shade tree stratum, farmers typically reduce shade cover via pruning of shade trees, at

531 the beginning of coffee flowering and at the beginning of harvest; however, farmers also
532 reduce shade when coffee plants exhibit signs of small leaves or otherwise stunted growth.
533 Consequently, under heterogeneous light environments, coffee leaf ITV is recognized as
534 changes in leaf size and is used to assess shade tree success in the provisioning of optimal
535 conditions for yield. In this study, we show that farmers deem coffee plant canopy radius
536 as a more important indicator for management decision-making than coffee plant height
537 (Table 3). DaMatta (2004) identified coffee crown architecture as a leading whole-plant
538 trait in the selection of coffee cultivars, particularly in water limited environments.
539 Though coffee planting density is typically a fixed variable, the space between crowns is
540 managed with pruning practices, which in turn can affect canopy “lushness”, or the
541 amount of leaves on a given plant. All reported explanations for canopy radius as the
542 most important indicator of a coffee plant success were related to plant requirements – a
543 larger canopy radius requires less fertilizer, and simultaneously produces greater yield –
544 and a farmers’ ability to change the canopy radius and lushness via pruning.

545 **Trait trade-offs and trait typologies**

546 The LES is constructed from leaf-level trade-offs among traits (Wright *et al.*
547 2004), for instance, leaf mass per area (LMA) trade-offs with mass-based photosynthesis
548 and leaf N concentration. Yet, there remains uncertainty as to whether these strong global
549 correlations among and between LES traits also function at the local scale (Wright &
550 Sutton-Grier 2012). For instance, trait expression and the strength of trade-offs are
551 affected by soil fertility (Ordoñez *et al.* 2009), climate (Sandel *et al.* 2010), and in crops,
552 a history of domestication (Milla *et al.* 2014; Martin *et al.* 2017), and therefore may
553 follow other drivers that influence the degree, or even presence, of trade-offs.

554 From an applied perspective, what remains unresolved is if farmers also perceive,
555 and make decisions based on, trade-offs among traits, which may consciously result in
556 certain ecosystem functions being optimized at the expense of others. For instance,
557 Mason & Donovan (2015) report that allocation to leaf defense may stimulate
558 conservative leaf strategies. Similarly, in this study the ranking of leaf size or thickness to
559 light environments often resulted in farmer references to leaf disease. A thick, dark, small
560 leaf, for example, may be identified as situated under an excess of shade, but discussions
561 of such trait syndromes often triggered supplementary comments by farmers on leaf

562 susceptible to disease, most notably, coffee rust. Alternatively, in several cases the
563 inverse was stated: the presence of some specific diseases or fungus indicated an excess
564 of shade. Several respondents identified the second largest leaf as the most productive in
565 the yield potential ranking, given that the largest leaf “...needs more fertilizer to
566 maintain”: an identified management trade-off between nutrient acquisition and leaf area.

567 Respondents often defined leaf- and whole-plant characteristics as indicative of
568 plant health, and occasionally in terms of human health, supporting previous work that
569 identified the tendency of farmers to compare coffee plant health to human health when
570 managing farms (Hugøy 2016). Leaves appeared “poisoned” (“intoxicado”) due to an
571 excess of fertilizer inputs or “wilted/sad” (“sodayadita”) due to insufficient shade or an
572 excess of light. Based on leaf size and colour, one respondent identified leaves as being
573 on “...a branch who has no future”. Trait trade-offs in the strictest ecological sense were
574 not at the forefront of respondent ranking of leaves, however, upon further exploration of
575 multiple leaf properties or trait syndromes, respondents clearly perceived trade-offs
576 between leaf size, thickness, nutrition, and disease resistance. This has been previously
577 reported as the concept ‘hielo’, or the tendency to group a variety of symptoms related to
578 plant disease, water imbalances, and nutritional deficiencies together (Bentley 1990).

579 **Implications**

580 The application of ecological theory to agricultural systems, for instance the use
581 of trait-based approaches in agroecology, needs to consciously take into account existing
582 on-farm practices. Just as the LES offers a framework to understand the traits
583 underpinning variation in plant ecological strategies, an FES can provide structure to
584 understanding farmer knowledge on plant trait expression and ITV, and the use of traits
585 that are not strictly related to yield in management decision-making. Our findings show
586 both commonality in knowledge of leaf traits between the scientific literature and farmers,
587 as well as apparent disconnects. It should be noted that given proximity to an agricultural
588 training research station, respondents in this study might have a distinctive agricultural
589 knowledge base than those in the surrounding regions. Regardless though, our findings
590 open an opportunity for mutually beneficial exchange on the role of functional traits in
591 agroecosystems that may advance management prescriptions that are consistent with
592 contemporary scientific knowledge, and practical understandings of agroecological

593 systems. Specifically, traits can be used to assess the success of a range of management
594 practices including shade-tree species selection and abundance, crop- and shade-tree
595 pruning regimes, and fertilization treatments. Furthermore, several emerging areas of
596 agroecological management, including the implementation of new crop genotypes,
597 control of novel pests and pathogens, and climate change mitigation strategies, demand a
598 highly robust and diverse array of ecological knowledge. Given environmental change in
599 many agriculture regions, coordination between scientific- and local knowledge systems
600 is increasingly required to effectively adapt management practices to changing conditions.

601

602 **Acknowledgements:** The authors would like to sincerely thank the participants of this
603 study for their knowledge and generosity. The authors would like to thank J. Baumgartner
604 and L. Hethcote for invaluable logistical support and assistance in the field. We thank
605 three journal reviewers for their insightful and constructive comments on earlier versions
606 of the manuscript. Thanks to the Agroforestry Systems with Perennial Crops Scientific
607 Partnership Platform (PCP AFS-PC) for access to its network. This work was conducted
608 with the support of the CASCADE project (“Ecosystem-based Adaptation for
609 Smallholder Subsistence and Coffee Farming Communities in Central America”), which
610 was funded by the International Climate Initiative (ICI) of the German Federal Ministry
611 for the Environment, Nature Conservation, Building and Nuclear Safety (BMUB). This
612 research was undertaken thanks to funding from the Canada Research Chairs program
613 and a Natural Sciences and Engineering Research Council of Canada Discovery Grant to
614 M.E.I.

615

616 **Authors’ contributions:** M.E.I. designed and coordinated the study, conducted interview
617 data collection and analysis, and drafted the manuscript. R.C. participated in study design
618 and interview data collection, conducted farm data collection, provided logistical support,
619 and assisted in manuscript writing. B.R. participated in study design and interview data
620 collection, provided logistical support and assisted in manuscript writing. A.R.M.
621 participated in statistical analyses and manuscript writing. A.K.D. participated in
622 interview data collection and manuscript writing. N.S. provided assistance in study

623 design, interview data collection and manuscript writing. All authors gave final approval
624 for publication.

625

626 **Data accessibility:** Based on confidentiality protection outlined in our approved ethics
627 protocols for research with human participants, data from this research will not be
628 archived.

629

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Accepted

909 Table 1. Mean (\pm s.d.) and range of farm-level variables attributes. In all cases $n = 25$
910 farms, and acronyms in brackets correspond to how they appear in Equation 1.
911

Predictor variables	Mean (\pm s.d.)	Range
Land under coffee (L)	0.89 ± 0.92	0.1-4.0
Coffee Age (A)	15.76 ± 10.04	2-35
Coffee Density (D)	5094.6 ± 1189.9	3464-8069
Years farming (Y)	36.1 ± 16.7	4-86
Variety number (V)	2.8 ± 2.1	0-11
Engagement with plants (E)	0.6 ± 0.5	0-1
Fertilizer use (F)	1.52 ± 0.92	0-3
Sum of practices (P)	17.08 ± 4.95	1-23
Shannon Index (H)	0.84 ± 0.55	0-1.611
Simpson Index (D)	0.38 ± 0.29	0-1
Shade class (S)	2.52 ± 0.71	1-3

912

913 Table 2. Ranking of leaf (colour, size, and thickness) and whole-plant (canopy radius,
914 coffee plant height, and coffee sprout diameter) trait importance for on-farm decision-
915 making. Values following the name of each trait denote the average ranking of that trait
916 (with 1 representing the highest importance, and three representing the least important),
917 based on interviews with $n = 25$ respondents.

918

Ranking	Leaf traits	Whole-plant traits
1 st rank	Colour = 1.4 ± 0.71	Canopy radius = 1.16 ± 0.37
2 nd rank	Size = 2.04 ± 0.68	Coffee height = 2.28 ± 0.61
3 rd rank	Thickness = 2.52 ± 0.65	Sprout diameter = 2.52 ± 0.71

919

920 Table 3. Mean (\pm s.d.) proportion of agreement for response variables, as well as the most
 921 frequently identified trait under three management scenarios: light environment (high and
 922 low light), nutrient status (sufficient and deficient nutrition) and yielding potential (high
 923 yield and low yield). All values are based on $n = 25$ responses.
 924

Response variables	Mean proportion (\pm s.d.)	Most frequently identified trait (% of respondents)	
		High light	Low light
Proportion of agreement for leaf size	0.55 \pm 0.32	Smallest leaf (56%)	Largest leaf (64%)
Proportion of agreement for leaf thickness	0.39 \pm 0.26	Thinnest leaf (36%)	Thickest leaf (68%)
		Sufficient nutrients	Deficient nutrients
Proportion of agreement for leaf colour	0.66 \pm 0.34	Large green leaf (80%)	Small yellow leaf (68%)
		High yield	Low yield
Correlation for yield potential	0.47 \pm 0.27	Largest green leaf (64%)	Small yellow leaf (32%)

925

926 Table 4. Step-wise and multiple regression model analysis evaluating which farm-level characteristics best predict the agreement
 927 between i) a farmer's ranking of leaf traits to light environments (size and thickness), soil fertility (colour) and reproduction (yield
 928 potential), and ii) an empirical ranking of leaf traits. AIC values for both the full model and the most parsimonious model are
 929 presented, with Δ AIC values representing the difference between the two. Parameter estimates (and p -values in brackets) are shown
 930 only for those parameters retained in the AIC-selected model. Parameters highlighted in bold are those that were significant ($p < 0.05$)
 931 in a multiple regression analysis, and acronyms and associated β values (in brackets) correspond to Table 1 and Equation 1. Also
 932 shown for each AIC-model are is explained variance, where $n = 25$ for each model.
 933

Model	AIC-retained parameters (β values in Equation 1)	Coefficient (p -value)	Full AIC	AIC	Delta AIC	Model r^2 (p -value)
Leaf size	Intercept (β_0)	0.47 (<0.001)	-44.77	-58.97	14.2	0.146 (0.033)
	log-L (β_1)	-0.149 (0.03)	-	-	-	-
Leaf thickness	Intercept (β_0)	0.934 (0.031)	-50.92	-54.97	4.05	0.213 (0.148)
	A (β_2)	0.009 (0.0187)	-	-	-	-
	D (β_3)	-0.0001 (0.091)	-	-	-	-
	log-V (β_5)	-0.145 (0.255)	-	-	-	-
	E (β_6)	0.302 (0.058)	-	-	-	-
	F (β_7)	-0.142 (0.056)	-	-	-	-
	H (β_9)	-0.398 (0.165)	-	-	-	-
	D (β_{10})	-0.581 (0.091)	-	-	-	-

	S (β_{11})	0.272 (0.195)	-	-	-	-
Leaf colour	Intercept (β_0)	0.24 (0.323)	-45.7	-56.5	10.8	0.170 (0.049)
	log-L (β_1)	-1.66 (0.026)	-	-	-	-
	P (β_8)	0.02 (0.141)	-	-	-	-
Yield potential	Intercept (β_0)	1.231 (<0.001)	-61.6	-69.0	7.45	0.263 (0.037)
	log-V (β_5)	0.327 (0.007)	-	-	-	-
	Y (β_4)	-0.004 (0.206)	-	-	-	-
	P (β_8)	-0.034 (0.006)	-	-	-	-
	S (β_{11})	-0.119 (0.139)	-	-	-	-

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938 Figure 1. The “leaf book”, a visual elicitation tool based on empirically framed
939 researcher-created visual data for ranges of coffee leaf traits. See Materials and Methods
940 for description of the collection and presentation of leaves in the book.

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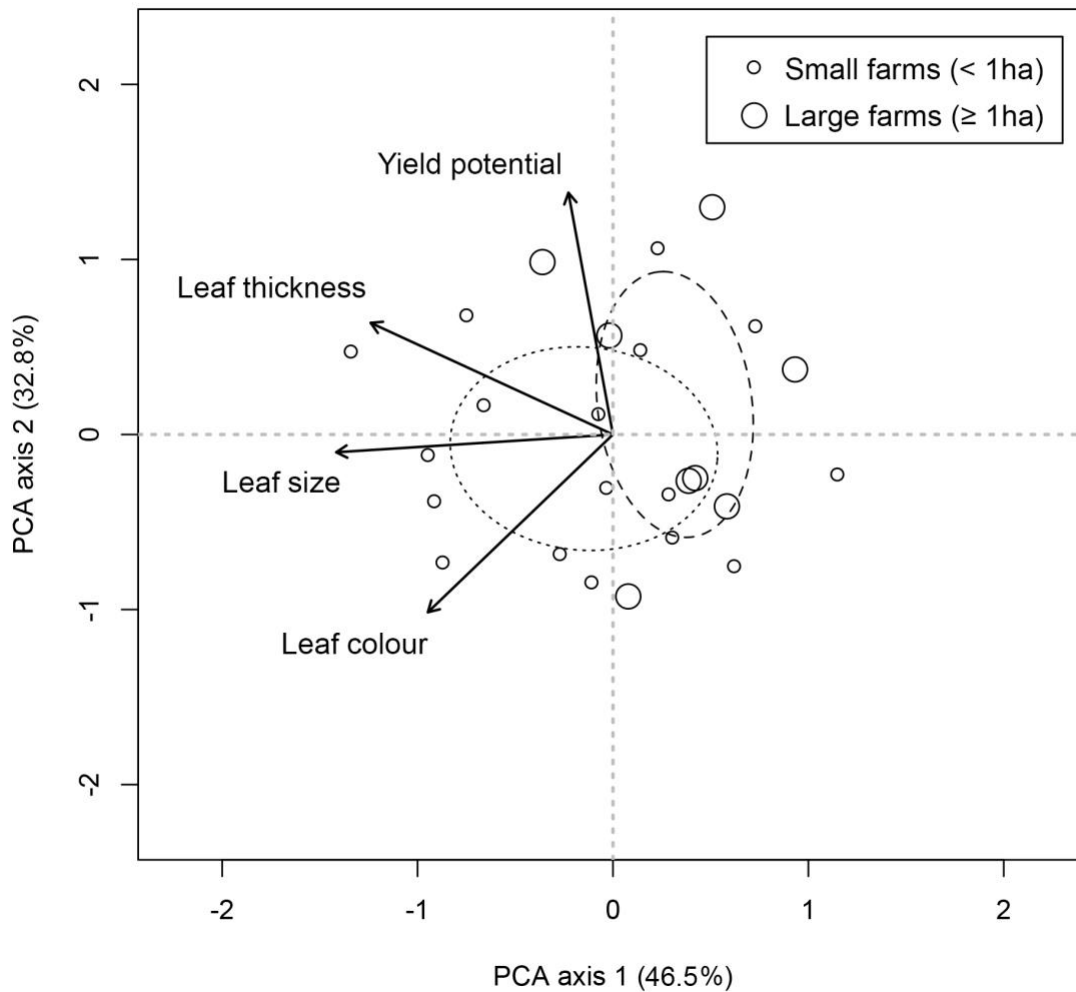
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960 Figure 2. Principal components analysis assessing relationships among the proportion of
 961 agreement between respondent-identified ranking of leaves and an empirically assessed
 962 ranking of leaves for i) leaf size, ii) leaf thickness, iii) leaf colour, and iv) yield potential).

963 Small circles represent small farms <1 ha in size, while large circles represent farms ≥ 1
 964 ha in size. Dotted and dashed circles represent 95% confidence ellipses surrounding small
 965 and large farms, respectively, which differed significantly across PCA axis 1 (one-tailed
 966 *t*-test $p = 0.035$).

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