



1 Beyond the laboratory: performance and agreement of rapid 2 methodologies for soil health assessment

3 Lur Epelde¹, Mikel Anza¹, Jasmin Fetzer², Katy Jo Stanton³, Josiah Judson³, Nerea Mandaluniz¹

4 ¹NEIKER-Basque Institute for Agricultural Research and Development, Basque Research and Technology Alliance (BRTA),
5 Parque Científico y Tecnológico de Bizkaia, P812, 48160 Derio, Spain

6 ²Digit Soil AG, Adliswil, 8134, Switzerland

7 ³Soil Association Limited, 51 Victoria Street, Bristol, BS1 6AD, UK

8 *Correspondence to:* Lur Epelde (lepelde@neiker.eus)

9 **Abstract.** Soil health assessment increasingly relies on biological indicators because of their sensitivity and direct links to
10 ecosystem functioning. However, conventional laboratory methods are time-consuming, require specialized infrastructure, and
11 are often incompatible with rapid decision-making in applied contexts. Several rapid or field-deployable tools have recently
12 been developed to address this limitation, but their comparability with standard laboratory methods remains insufficiently
13 evaluated. Here, we compared four rapid approaches with their corresponding laboratory reference methods in a long-term
14 grassland experiment: aggregate stability (SLAKES), soil respiration (portable CO₂ analyzer), microbial biomass carbon and
15 fungal-to-bacterial ratio (microBIOMETER®), and enzyme activities (Soil Enzymatic Activity Reader, SEAR). Agreement
16 between methods was assessed using Spearman correlations, redundancy analyses and Procrustes analysis. Aggregate stability
17 showed strong correspondence between rapid and laboratory measurements ($R = 0.64$), whereas soil respiration exhibited weak
18 agreement, likely reflecting that in situ and laboratory approaches capture different aspects of respiratory activity. Microbial
19 biomass carbon displayed moderate comparability between methods ($R = 0.51$), while fungal-to-bacterial ratios did not.
20 Enzyme activities measured with SEAR were generally consistent with laboratory assays. Multivariate analyses indicated that
21 overall, rapid methods captured ecological patterns similar to those revealed by laboratory protocols. These findings support
22 the use of selected rapid tools as complementary or alternative options when laboratory facilities are unavailable or timely soil
23 health information is required to inform management decisions.

24 1 Introduction

25 Soil health is fundamental to the maintenance of ecosystem functions, including primary productivity, nutrient cycling, water
26 regulation, and climate mitigation (Diaz de Otalora et al., 2021; Vogel et al., 2019). Despite its importance, there is ongoing
27 debate regarding how to operationalize soil health and which indicators most appropriately capture it, reflecting its inherently
28 multidimensional nature (Lehmann et al., 2020). Among the different types of indicators, biological indicators are increasingly



29 recognized as particularly informative due to their high sensitivity to management and disturbance and their direct links to soil
30 multifunctionality (Creamer et al., 2022; Garbisu et al., 2011).

31 In many applied contexts, however, soil health assessment is needed rapidly to support management decisions. Traditional
32 laboratory-based analyses often require specialized facilities and can involve substantial delays, ranging from several days to
33 weeks between sampling and results. This limits their practical utility for on-farm decision-making or adaptive management.
34 Consequently, there is growing interest in the development of low-cost, easy-to-use, non-invasive and rapid assessment tools
35 that can be applied directly in the field (Ameer et al., 2024), even if this comes at the expense of analytical precision.

36 Soil biological measurements have lacked such rapid, field-deployable approaches, but recent years have seen the emergence
37 of several tools addressing this gap. Aggregate stability, while sometimes classified as a physical property, can be considered
38 a hybrid indicator because the persistence of aggregates largely depends on biologically derived binding agents, such as
39 glomalin, microbial exudates and mucilages, fungal hyphal networks, plant roots, and the activity of soil macrofauna (Six et
40 al., 2004). Recent tools such as SLAKES allow for rapid field-based assessment of aggregate stability (Fajardo et al., 2016;
41 Fajardo et al., 2025). Similarly, soil respiration can be measured in situ using portable CO₂ analyzers (e.g., Biau et al., 2013),
42 providing an immediate proxy for overall biological activity. More recently, commercial kits such as microBIOMETER®
43 (mB) (2026) have been developed to estimate microbial biomass and the fungal-to-bacterial ratio, the latter being an
44 ecologically relevant indicator linked to sustainable soil systems (Fierer et al., 2021). Finally, portable systems such as the Soil
45 Enzymatic Activity Reader (SEAR) enable standardized measurements of soil enzyme activities (Fetzer et al., 2025), which
46 are closely associated with organic matter turnover and nutrient acquisition.

47 Despite their increasing availability and potential utility, the performance of these rapid assessment tools relative to established
48 laboratory methods remains insufficiently evaluated. Assessing their comparability is essential to determine whether they can
49 reliably inform soil health assessments and management decisions. The objective of this study was therefore to compare four
50 rapid, field-based biological soil health measurements (i.e., aggregate stability measured with SLAKES, soil respiration
51 measured with EGM-5, microbial biomass and fungal-to-bacterial ratio measured with mB, and enzyme activities measured
52 with SEAR) with their corresponding laboratory-based reference methods, to evaluate their agreement and to identify potential
53 sources of divergence where discrepancies occur.

54 **2 Materials and methods**

55 **2.1 Experimental design, soil sampling and characterization**

56 Soil sampling was conducted in the experimental multispecies pastures of NEIKER in Arkaute (42°51'11.41" N, 2°37'27.20"
57 W; Basque Country, Spain) from April to November 2024, with sampling performed every three weeks, resulting in a total of
58 11 sampling campaigns. During this period, the mean air temperature was 15.2 °C (mean maximum 21.4 °C and mean
59 minimum 9.1 °C), and cumulative precipitation amounted to 492 mm. The experimental fields host a long-term experiment
60 established in 2013, ten years prior to the present sampling. The experiment follows a completely randomized block design.



61 The grazing area comprises 4.5 ha of semipermanent pasture divided into three plots, which are further subdivided into two
62 grazing management treatments: rotational and conventional grazing (for details on the experimental design, see Diaz de
63 Otálora et al., 2021).

64 Within each of the six resulting subplots, a random sampling point was selected and repeatedly sampled at soil depths of 0–20
65 cm and 20–50 cm following the standardized protocol of the LUCAS network. Briefly, five subsamples (one collected at the
66 central point and four additional subsamples collected 2 m away in the cardinal directions) were composited to obtain a single
67 representative sample (Ballabio et al., 2016). In total, 132 soil samples were collected. One representative fraction of each
68 sample was kept unsieved for aggregate stability determination, a second fraction was sieved fresh to <2 mm for the analysis
69 of soil biological properties, and a third fraction was air-dried to constant weight for physicochemical characterization. Soil
70 pH was measured following MAPA et al. (1994). Soil organic matter content was estimated from organic carbon content,
71 determined according to DIN 19539 (2016) using a SoliTOC® Cube analyzer (Elementar, Langenselbold, DE), and converted
72 using a factor of 1.72. Particle size distribution was determined by laser diffraction (Mastersizer 3000 Hydro LV, Malvern
73 Panalytical, Malvern, UK) in accordance with ISO 13320 (2009). On average (\pm standard deviation), soils exhibited a pH of
74 8.17 ± 0.18 , an organic matter content of $4.42 \pm 1.69\%$, and a clay loam texture.

75 **2.2 Aggregate stability**

76 A low-cost assessment of aggregate stability was performed using SLAKES, a free smartphone application developed by the
77 University of Sydney (Australia) and promoted by the Soil Health Institute (USA). The methodology underpinning the
78 SLAKES software is described in Fajardo et al. (2016). All measurements were conducted following the step-by-step
79 instructions provided within the application.

80 In the laboratory, soil aggregate stability was determined using the wet sieving apparatus of Eijkelkamp (Giesbeek, The
81 Netherlands). A known mass of air-dried soil aggregates (comparable to the amount used for SLAKES) was subjected to
82 vertical oscillation in a water column through a 2 mm sieve. After a 3 min immersion period, the mass of soil remaining on
83 the sieve was recorded and used to calculate the percentage of water-stable aggregates.

84 **2.3 Respiration**

85 In situ soil respiration was measured using a portable CO₂ infrared gas analyzer (IRGA) (EGM-5, PP Systems, Amesbury,
86 MA, USA) connected to a cylindrical soil respiration chamber (SRC-2, PP Systems), following the procedure described by
87 Biau et al. (2013). Laboratory soil respiration measurements were conducted in accordance with ISO 16072 (2002). For this
88 analysis, the comparison between in situ and laboratory measurements was restricted to the 0–20 cm soil layer, comprising a
89 total of 66 samples.



90 **2.4 Microbial biomass carbon and fungal to bacteria ratio**

91 Microbial biomass carbon (MBC) and fungal:bacterial (F:B) ratio were assessed using the commercial mB system (Everett,
92 WA, USA) following the procedure provided by the manufacturer. In parallel, laboratory-based MBC was determined using
93 the chloroform fumigation–extraction method described by Vance et al. (1987). Soil DNA was extracted from 0.25 g dry-
94 weight soil using the DNeasy PowerSoil™ Pro Kit (Qiagen, Venlo, NL). Quantification of bacterial (16S rRNA gene) and
95 fungal (18S rRNA gene) abundances was performed by qPCR using the primers, reaction mixtures, and cycling conditions
96 described in Epelde et al. (2014).

97 **2.5 Enzyme activities**

98 Finally, β -glucosidase (GLU), phosphomonoesterase (PHO), L-leucine aminopeptidase (LEU), and β -xylosidase (XYL)
99 activities were measured on site using the laboratory-independent SEAR (Fetzer et al., 2025). Corresponding laboratory
100 measurements of these enzyme activities were conducted following ISO 22939 (2010) using fluorogenic substrates in 96-well
101 microplates, as described by Anza et al. (2019).

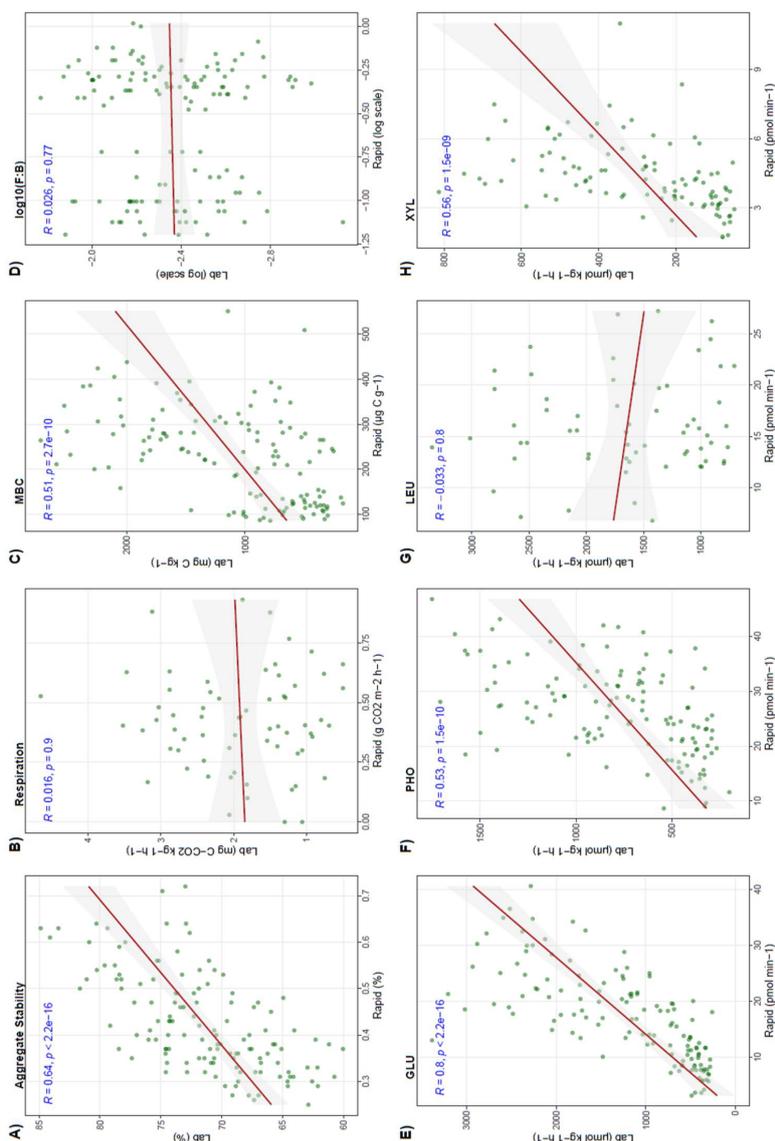
102 **2.6 Statistical analysis**

103 Spearman's rank correlation was employed as the primary metric to assess the relationship between methods. Before this
104 analysis, data for the F:B ratio were log-transformed to reduce heteroscedasticity and improve data symmetry. All correlation
105 analyses were performed in R.

106 Redundancy analyses (RDA) were applied separately to the datasets derived from the rapid assessment tools and from the
107 standard laboratory protocols to evaluate whether both approaches detected the effects of the experimental factors.
108 Measurements of soil respiration were excluded from the RDA due to the lack of data for the 20–50 cm depth in the rapid
109 method. For enzyme activity values below the limit of quantification, the lowest quantified value was substituted prior to
110 analysis. Redundancy analyses were performed using Canoco 5 (ter Braak and Šmilauer, 2012). Complementarily, to evaluate
111 multivariate consistency between the rapid field-based suite of measurements and the laboratory reference methods, a
112 Procrustes analysis was performed using the *protest* function of the *vegan* package (Oksanen et al., 2025).

113 **3 Results and discussion**

114 Figure 1 summarizes the Spearman rank correlations between rapid field-based measurements and their corresponding
115 laboratory reference methods, and all raw data are provided in Supplement A. Overall, the strength and significance of the
116 relationships varied markedly among indicators. Aggregate stability showed a strong and significant correlation between
117 methods (Fig. 1a), supporting the validity of SLAKES as a rapid assessment tool. This result is consistent with previous
118 validation studies of the application (Flynn et al., 2020) and further supports its use in the growing body of studies that have
119 applied this approach (see examples in Fajardo et al., 2025). In contrast, the relationship between in situ and laboratory



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Figure 1: The relationship between rapid metrics (x axis) and standard laboratory protocols (y axis) for: (A) aggregate stability, (B) soil respiration, (C) microbial biomass carbon (MBC), (D) fungal to bacterial (F:B) ratio, and (E-H) activities of β-glucosidase (GLU), phosphomonoesterase (PHO), L-leucine aminopeptidase (LEU), and β-xylosidase (XYL). Solid red lines represent the linear regression fit, with light gray shaded areas indicating the 95% confidence interval. Blue text denotes Spearman's rank correlation coefficient and associated p-values.



126 measurements of soil respiration was weak and not statistically significant (Fig. 1b), indicating limited agreement between the
127 two approaches. For the mB outputs, MBC exhibited a moderate and significant correlation with the laboratory-based method
128 (Fig. 1c), whereas the F:B ratio showed a weak relationship (Fig. 1d), suggesting lower comparability between methods for
129 this parameter. Similarly, Laine et al. (2025) reported that mB did not match PLFA-based estimates when determining the F:B
130 ratio. The performance of the SEAR varied among enzymes. Correlations were very strong and significant for GLU (Fig. 1e),
131 moderate and significant for PHO (Fig. 1f) and XYL (Fig. 1h), and weak and non-significant for LEU (Fig. 1g). For LEU, the
132 device prototype tested was not optimized, as this enzyme requires a different wavelength; this limitation will be addressed in
133 future versions of the device, which will incorporate additional wavelengths.

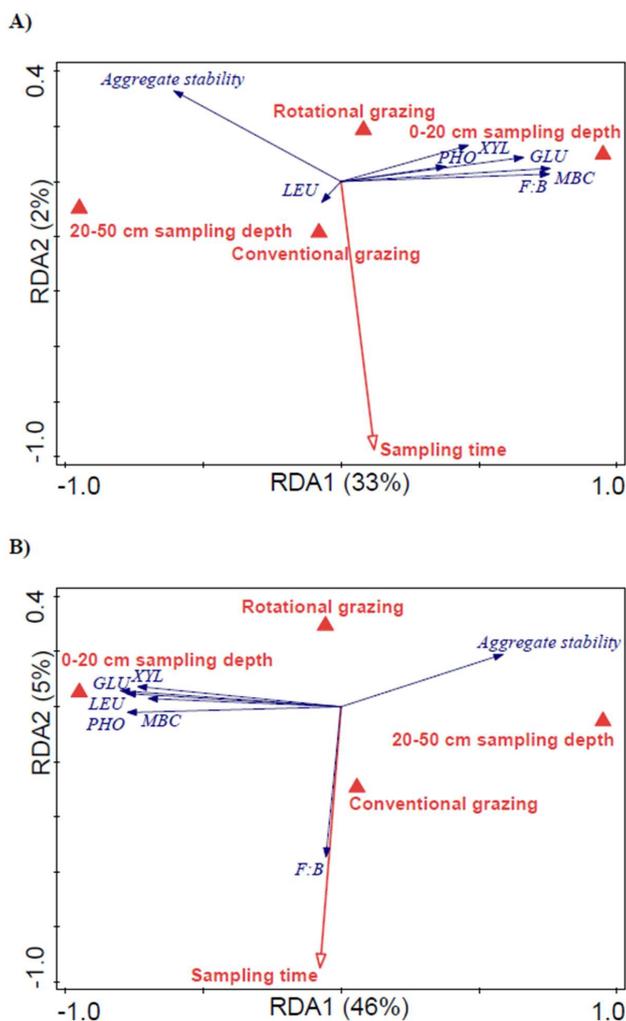
134 These results must be interpreted in light of methodological differences between rapid and laboratory measurements. In
135 particular, soil respiration and enzyme activities measured with rapid tools reflect the temperature and moisture conditions at
136 the time of sampling, whereas laboratory assays are performed under standardized and near-optimal conditions of temperature,
137 moisture, and, in the case of enzyme activities, pH. This interpretation is supported by the high correlation reported by Haney
138 et al. (2008) between respiration measured by titration and by an IRGA when both methods were applied under laboratory
139 conditions, suggesting that the lack of correlation observed in this study may be related to the use of this equipment under field
140 conditions. The relative value of each approach therefore depends on the study objective: rapid measurements more faithfully
141 capture the in situ functional status of the soil at a given moment, while laboratory measurements estimate potential activity
142 (Nannipieri et al., 2018) and offer greater comparability across samples and studies.

143 The Procrustes analysis revealed a statistically significant concordance between the rapid assessment tools and the laboratory
144 standard methods ($R = 0.473$; $P = 0.011$), indicating a moderate but meaningful similarity in the multivariate configuration of
145 samples. Consistent with this finding, redundancy analyses showed that the suite of rapid indicators captured the same overall
146 ecological patterns as the laboratory standards (Fig. 2). Sampling depth emerged as the dominant factor structuring the data
147 along the first ordination axis (RDA1), explaining 33% of the total variance in the rapid assessment dataset (Fig. 2a) and 46%
148 in the laboratory dataset (Fig. 2b). Higher aggregate stability was associated with deeper soil layers (20–50 cm), whereas
149 higher values of microbial biomass and activity were predominantly linked to surface soils (0–20 cm), consistent with
150 established vertical patterns in soil biological functioning (Taylor et al., 2002). The second ordination axis (RDA2) captured
151 the effects of sampling time and grazing management, although these factors explained a substantially smaller proportion of
152 the total variance, accounting for 2% in the rapid assessment data and 5% in the laboratory data. Despite this lower explanatory
153 power, both datasets showed qualitatively similar trends along this axis.

154 Examination of correlations among response variables within the RDAs further supported the patterns observed in the
155 Spearman analyses. Rapid indicators that showed weak agreement with laboratory methods displayed distinct ordination
156 behavior. In particular, LEU activity measured with SEAR showed a poor representation and low correlation with the
157 explanatory variables in the selected dimensions. Also, the F:B ratio derived from the rapid method was strongly correlated
158 with the MBC value obtained from the same kit. Meanwhile, the laboratory-based F:B ratio was primarily associated with



159 sampling time, suggesting a strong temporal signal that may reflect batch effects or a higher sensitivity to short-term
 160 environmental variability.



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 162 **Figure 2: Biplots of the redundancy analyses (RDA) performed using (A) rapid assessment metrics (Pseudo-F=23.4, P=0.002) and**
 163 **(B) standard laboratory protocol data (Pseudo-F=45.2, P=0.002) as response variables (blue), with experimental factors and**
 164 **sampling time included as explanatory variables (red). F:B: fungal to bacterial ratio; GLU: β -glucosidase; LEU: L-leucine**
 165 **aminopeptidase; MBC: microbial biomass carbon; PHO: phosphomonoesterase; XYL: β -xylosidase.**



166 **4 Conclusions**

167 Overall, under the conditions tested, the performance of the rapid methods varied among indicators. SLAKES provided results
168 that closely matched laboratory measurements, supporting its use for aggregate stability assessment, whereas soil respiration
169 measured in situ showed poor agreement with laboratory assays, likely reflecting that the two approaches capture different
170 aspects of respiratory activity. The mB yielded comparable estimates of MBC but did not reliably reproduce laboratory-based
171 F:B ratios. Enzyme activities measured with SEAR were generally consistent with laboratory methods. Despite these
172 differences at the individual indicator level, multivariate analyses demonstrated that the suite of rapid measurements captured
173 ecological patterns similar to those identified by laboratory protocols. Overall, these results indicate that selected rapid methods
174 can provide meaningful and timely insights into soil biological functioning and represent a valuable complement, or alternative,
175 to laboratory analyses when rapid decision-making or limited analytical infrastructure is a constraint.

176 **Author contributions**

177 LE: Conceptualization, funding acquisition, writing - original draft. MA: Formal analysis, writing - review & editing. JF:
178 Writing - review & editing. KJS: Project administration, writing - review & editing. JH: Project administration, writing -
179 review & editing. NM: Funding acquisition, writing - review & editing.

180 **Competing interests**

181 Author JF was employed by Digit Soil. The remaining authors declare that they have no conflict of interest.

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