

GENERATING AND TESTING HYPOTHESES ABOUT THE FOSSIL RECORD OF INSECT HERBIVORY WITH A THEORETICAL ECOSPACE

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Abstract

A typical fossil flora examined for insect herbivory contains a few hundred leaves and a dozen or two insect damage types. Paleontologists employ a wide variety of metrics to assess differences in herbivory among assemblages: damage type diversity, intensity (the proportion of leaves, or of leaf surface area, with insect damage), the evenness of diversity, and comparisons of the evenness and diversity of the flora to the evenness and diversity of damage types. Although the number of metrics calculated is quite large, given the amount of data that is usually available, the study of insect herbivory in the fossil record still lacks a quantitative framework that can be used to distinguish among different causes of increased insect herbivory and to generate null hypotheses of the magnitude of changes in insect herbivory over time. Moreover, estimates of damage type diversity, the most common metric, are generated with inconsistent sampling standardization routines. Here we demonstrate that coverage-based rarefaction yields valid, reliable estimates of damage type diversity that are robust to differences among floral assemblages in the number of leaves examined, average leaf surface area, and the inclusion of plant organs other than leaves such as seeds and axes. We outline the potential of a theoretical ecospace that combines various metrics to distinguish between potential causes of increased herbivory. We close with a discussion of the most appropriate uses of a theoretical ecospace for insect herbivory, with the overlapping damage type diversities of Paleozoic gymnosperms and Cenozoic angiosperms as a brief case study.

1 INTRODUCTION

In recent years, the number of fossil plant assemblages examined for insect herbivory has increased markedly. The wealth of available data has already been used to inform a variety of biotic and abiotic phenomena

31 (Smith, 2008; Carvalho et al., 2014; Labandeira and Currano, 2013), but raises the question of how to
32 compare the patterns of insect herbivory observed on different host plants or in different assemblages.

33 An increase in herbivory in deep time can occur in response to various environmental and evolutionary
34 phenomena, demonstrating the need for analytical techniques that can be used to distinguish among them.
35 Two explanations that are commonly invoked as causes of increased insect herbivory are insect and plant
36 evolution, which lead to an expanded suite of feeding behaviors (Labandeira, 2006; Martinez et al., 2019;
37 Wagner et al., 2015), and the nutrient dilution hypothesis, in which a sudden increase in atmospheric $p\text{CO}_2$
38 increases the carbon-to-nitrogen ratio in many plant tissues, increasing the amount of leaf area that each
39 insect must consume in order to ingest a given amount of protein (Bazzaz, 1990). The techniques currently
40 used in paleontological studies do not distinguish among these disparate causes of increased herbivory.

41 A standardized method for comparing insect herbivory would allow the use of published data to generate
42 null expectations for findings at new localities. In addition, a standardized method would facilitate the
43 differentiation of these and other causes of increased insect herbivory in cases where the distinction is not so
44 clear.

45 1.1 THE END-TRIASSIC AS A HYPOTHETICAL CASE STUDY

46 The end-Triassic extinction event exemplifies the potential utility of statistical methods with the capacity
47 to generate null expectations and disentangle the various potential causes of fluctuations in herbivory. Few
48 latest Triassic floras have been examined for insect herbivory (Ghosh et al., 2015) and, of the geologic
49 periods that contain more than five described insect fossils, the Jurassic is the least studied in this context
50 (McLoughlin et al., 2015; Ding et al., 2015; Pinheiro et al., 2016; Na et al., 2018; Santos et al., 2021).

51 One could generate any number of predictions about changes in insect herbivory across the end-Triassic
52 event. Patterns of insect herbivory may have remained constant because it is widely agreed that insects
53 did not suffer major losses at the Triassic/Jurassic boundary (Dmitriev and Zherikin, 1988; Labandeira
54 and Sepkoski, 1993; Jarzembowski and Ross, 1996). Insect herbivory may have decreased because plant
55 communities do appear to have endured noticeable turnover and losses across this extinction event (Belcher
56 et al., 2010; Li et al., 2020; Lucas, 2021; McElwain and Punyasena, 2007). The $p\text{CO}_2$ spike associated with
57 the end-Triassic event (Knobbe and Schaller, 2017) complicates matters further. Whether or not an increase
58 in $p\text{CO}_2$ led to an increase in plant biomass and a corresponding dilution of nutrients such as nitrogen
59 (Mattson, 1980) depends greatly on interacting environmental parameters (Shaw et al., 2002; McMurtrie
60 et al., 2008; Reich et al., 2014). Nutrient dilution has very rarely been directly examined in the plant clades
61 that were present in Triassic and Jurassic ecosystems; this phenomenon has been studied almost exclusively

62 in angiosperms, which had not yet evolved at the time of the end-Triassic event (Bazzaz, 1990; Boyce and
63 Zwieniecki, 2012; Ramírez-Barahona et al., 2020).

64 If data were available, a comparison of insect herbivory levels immediately before and after the end-
65 Triassic event would be hampered by the lack of available statistical techniques. The first obstacle would be
66 the lack of a null, or baseline, prediction of the magnitude of change in insect herbivory that would occur
67 from the Late Triassic to Early Jurassic in the absence of a major environmental or evolutionary event. How
68 much variation in insect herbivory is best attributed to statistical noise? How much is best attributed to the
69 passage of time rather than an external trigger? After these sources of variation are taken into account, how
70 much variation remains? The fern- and gymnosperm-dominated Permian, Triassic, and Cretaceous floras
71 that have already been examined for insect herbivory provide an opportunity to generate a null prediction
72 and quantify the uncertainty surrounding it. What is needed is a comparative framework to generate this
73 null prediction.

74 The second obstacle would be the lack of a comparative framework for disentangling the biotic and
75 abiotic causes of fluctuations in insect herbivory. The environmental perturbation most thoroughly
76 examined in existing studies of insect herbivory, the Paleocene–Eocene Thermal Maximum, or PETM (Wilf
77 and Labandeira, 1999; Currano et al., 2008, 2016), began and ended far too quickly for much plant or
78 insect evolution to have occurred (Zeebe and Lourens, 2019). Many other environmental perturbations,
79 such as the increase in $p\text{CO}_2$ at the end-Triassic, which occurred in multiple pulses (Ruhl and Kürschner,
80 2011), span a long enough interval that abiotic and biotic factors can be confounded.

81 1.2 THEORETICAL ECOSPACES IN PALEONTOLOGY

82 Morphospaces are a useful tool for quantifying change over time. The axes of empirical morphospaces,
83 constructed with techniques such as principal component analysis, change with the addition of new data; in
84 contrast, the axes of theoretical morphospaces remain unchanged as new data are added (McGhee, 2006).
85 Morphospaces can be multidimensional (Raup, 1967; Lohman et al., 2017), can consist of various two-
86 dimensional comparisons (Wilson and Knoll, 2010), or, with sufficiently clear and specific hypotheses, require
87 only two dimensions (Raup, 1967; Gerber, 2017; Balisi and Van Valkenburgh, 2020).

88 Ecospace extend the logic of empirical and theoretical morphospaces to ecological data. The canonical
89 use of ecospace in paleontology applies to the marine realm (Valentine, 1969; Bambach, 1983), with an
90 updated version now forming the foundation of many quantitative studies (Bush et al., 2007; Wiedl et al.,
91 2013; Knope et al., 2015; Mondal and Harries, 2016; Frey et al., 2018; Laing et al., 2019). The three-
92 dimensional ecospace as updated by Bush et al. (2007) has also been modified for the sedimentary ichnological

93 study of terrestrialization (Minter et al., 2017) and for the study of terrestrial vertebrates (Chen et al., 2019).
94 The primary difference between the two ecospace formulations currently used in studies of marine animals is
95 the number of characters and character states (Novack-Gottshall, 2007; Bush and Novack-Gottshall, 2012).
96 Of note, both theoretical ecospace for marine animals use qualitative character states (Novack-Gottshall,
97 2007; Bush et al., 2007).

98 2 A THEORETICAL ECOSPACE FOR INSECT HERBIVORY IN THE 99 FOSSIL RECORD

100 Studies of insect herbivory typically use at least one of three common metrics. The diversity of insect damage
101 types (Labandeira et al., 2007) measures the richness of herbivory. The percentage of leaf area removed by
102 herbivores (known as the herbivory index) and the percentage of leaf specimens with feeding damage both
103 measure the intensity of herbivory; however, the latter is highly sensitive to leaf size, behaves differently from
104 the herbivory index (Smith, 2008), and thus is not recommended as an alternative to the former (Schachat
105 et al., 2018). Although damage type diversity and the herbivory index are often discussed interchangeably,
106 with an increase in either referred to as “more herbivory,” they measure fundamentally different aspects of
107 herbivory (Figure 1).

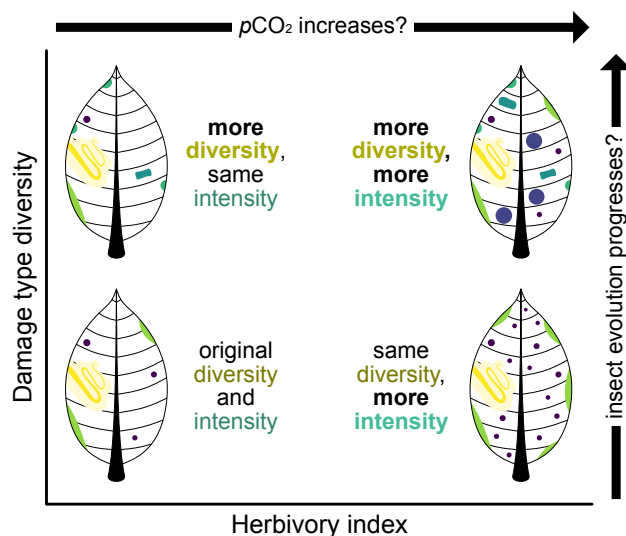


Figure 1: The difference between the diversity and intensity of insect herbivory.

108 The simplest theoretical ecospace for insect herbivory consists of damage type diversity plotted against
109 the herbivory index. Although this ecospace has not been illustrated or analyzed in previous studies, it holds
110 the potential to disentangle the various causes of increased herbivory.

2.1 THE UTILITY OF AN ECOSPACE

111
112 The dilution of nitrogen in leaves, caused by increased plant biomass, in turn caused by increased $p\text{CO}_2$, will
113 increase the amount of leaf biomass that an herbivore will need to consume in order to ingest enough protein
114 to satisfy its metabolic demands. In other words, an increase in $p\text{CO}_2$ is predicted to cause an increase in
115 the intensity of insect herbivory. The increased intensity of herbivory may well permit an increase in the
116 observed diversity of herbivory because the more leaf mass is consumed, the more opportunities there are
117 for the preservation of additional damage types that were present at low frequencies. However, this need
118 not necessarily be the case. When the intensity of herbivory increases, the diversity of herbivory can remain
119 constant or can even decrease.

120 In contrast, the expanding suite of insect feeding behaviors that developed from the Carboniferous through
121 Permian increased the diversity of herbivory more measurably and consistently than it increased the intensity
122 of herbivory. Whereas arthropod herbivory, probably caused by mites, has been found on Devonian liverworts
123 (Labandeira et al., 2013), the oldest known insect herbivory in the fossil record consists only of external
124 foliage feeding, with no accompanying piercing and sucking or specialized behaviors such as galling and
125 mining (Iannuzzi and Labandeira, 2008). The size of the archetypal herbivorous arthropod and its feeding
126 traces increased, but exhibited no more sophistication or diversity of feeding behavior than already seen with
127 probable mite herbivory on Devonian liverworts. As another example, definitive evidence of leaf mining only
128 appears in the earliest Mesozoic—tens of millions of years after insects began eating plants (Labandeira et al.,
129 1994; Ding et al., 2014)—at a time when the diversity of gall-inducing arthropods also increased (Labandeira,
130 2021). An increase in the diversity of insect herbivory does not necessitate a corresponding increase in the
131 intensity of herbivory: for example, some Permian gymnosperms have herbivory indices between 3.08%
132 (Schachat et al., 2014) and 3.95% (Beck and Labandeira, 1998). These represent the twenty-sixth and
133 thirty-second percentiles, respectively, of angiosperm leaf area removed per year by insect herbivores in
134 extant ecosystems—slightly less than the mean value (Turcotte et al., 2014).

135 Thus, a bivariate theoretical ecospace can be used to differentiate among the expected impacts of nitrogen
136 dilution versus insect and plant evolution as potential causes of increased insect herbivory. An increase in
137 the diversity of herbivory that outpaces the intensity of herbivory is consistent with insect evolution but not
138 with nitrogen dilution, and vice-versa.

139 The cyclical Holocene range expansions and contractions of eastern hemlock, *Tsuga canadensis* L.,
140 illustrate why the distinction between the diversity and intensity of herbivory matters (Labandeira, 2012).
141 The decline of hemlock is believed to have been caused by either climate change or outbreaks of insect
142 herbivores (Foster et al., 2006). Increased herbivory on subfossil hemlock leaves would support the

143 hypothesis that insects drove eastern hemlock into decline, but only if an increase in the intensity of
144 herbivory exceeds any increase in the diversity of herbivory. The intensity of herbivory would need to
145 increase more strongly than the diversity of herbivory in order for the insect-outbreak hypothesis to be
146 supported because, of the myriad insects known to feed on eastern hemlock (Buck et al., 2005; Dilling
147 et al., 2007), only three—two moths and one conifer aphid—have been identified as possible culprits in the
148 Holocene decline of this species (Filion et al., 2006; Simard et al., 2002; Oswald, 2016; Orwig and Foster,
149 1998; Labandeira, 2012). Moreover, due to indirect interspecific competition among insect herbivores
150 (Janzen, 1973; Price et al., 1980; van Veen et al., 2006; Kaplan and Denno, 2007), an outbreak of a single
151 herbivore species would quite possibly decrease the total diversity of insect herbivory through suppression
152 of other herbivores. Therefore, range contractions of eastern hemlock caused by insect herbivores would
153 require an increase in the intensity, but not the diversity, of herbivory.

154 2.2 ESTIMATING DAMAGE TYPE DIVERSITY FOR AN ECOSPACE

155 Because differences in sampling completeness threaten to bias estimates of damage type diversity (Schachat
156 et al., 2018), the act of comparing damage type diversity to the herbivory index is no simple task. Estimates
157 of the herbivory index are largely robust to differences in sample size: as sampling intensity increases,
158 confidence intervals become narrower but point estimates of the herbivory index do not change perceptibly
159 because a mean is an unbiased estimator (Schachat et al., 2018). For damage type diversity, on the other
160 hand, point estimates and the ranges of confidence intervals vary substantially with sampling intensity
161 because a tally—in this case, a tally of the damage types observed—is a biased estimator that continues
162 to increase as sampling progresses. This is illustrated by the Colwell Creek Pond and Mitchell Creek Flats
163 assemblages from the Early Permian of Texas (Schachat et al., 2014, 2015) which overlap considerably in
164 composition of their floral communities but vary in sampling completeness; the former contains over fifteen
165 times as much broadleaf surface area and over five times as many broadleaf specimens as the latter (Figure
166 S1). As detailed in the supplemental material, we found that the Chao1 estimator is unable to overcome the
167 lack of sampling completeness in insect herbivory datasets. This estimator returns inaccurate and imprecise
168 estimates of damage type diversity whether used alone (Chao, 1987) or in conjunction with rarefaction (Chao
169 et al., 2014).

170 Coverage-based rarefaction provides an alternative to traditional rarefaction (Chao and Jost, 2012). In
171 traditional, or size-based, rarefaction, curves are scaled by the number of leaves or the amount of leaf area
172 examined. In coverage-based rarefaction, samples are standardized by sampling completeness as indicated
173 by the slope of the rarefaction curve (Good, 1953; Jost, 2010). At the base of a rarefaction curve, sampling

174 is incomplete; the slope of the rarefaction curve equals 1 and coverage equals 0. As sampling approaches
175 completeness, the slope of the rarefaction curve reaches an asymptote—*i.e.*, a slope of 0—and coverage
176 equals 1 (Figure 2). Thus, with coverage-based rarefaction, damage type datasets are rarefied not to a
177 particular number of leaves but to a particular slope of the rarefaction curve. If a high proportion of damage
178 types are observed on only one specimen, this indicates that coverage is relatively low. In contrast, if a low
179 proportion of damage types are observed on only one specimen, coverage is relatively high. Coverage-based
180 rarefaction is performed by removing the rarest damage types from the dataset until coverage is reduced to
181 a predetermined threshold such as 0.8 (Figure 3).

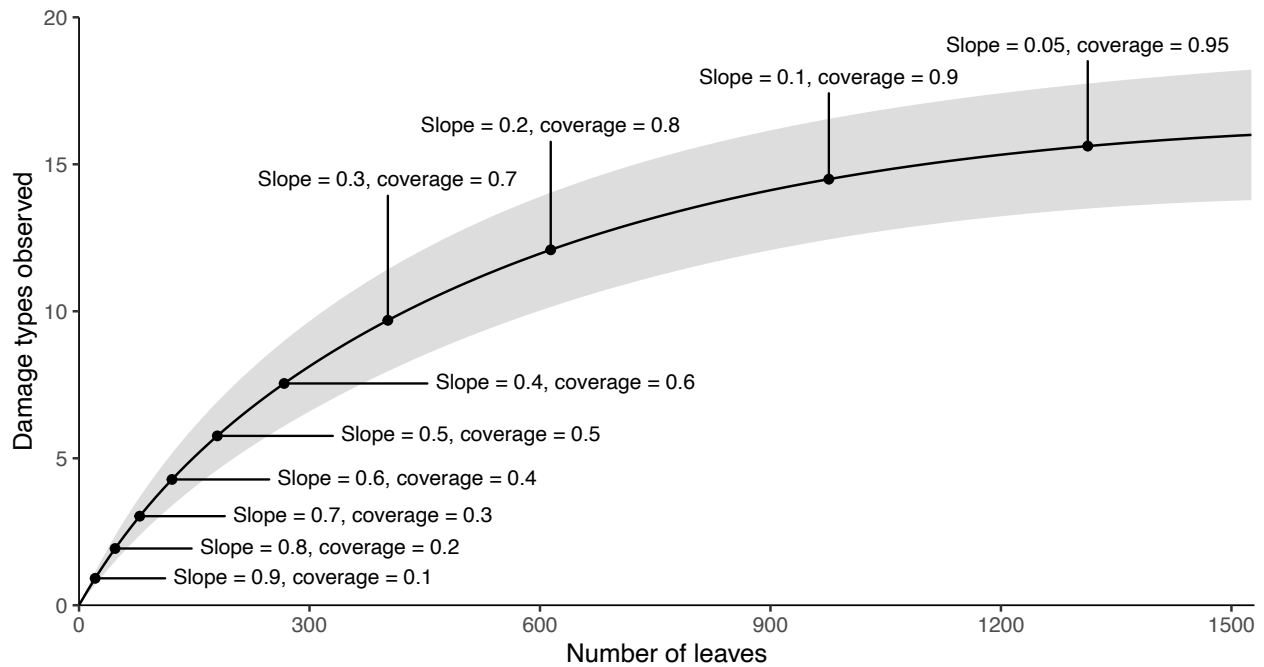
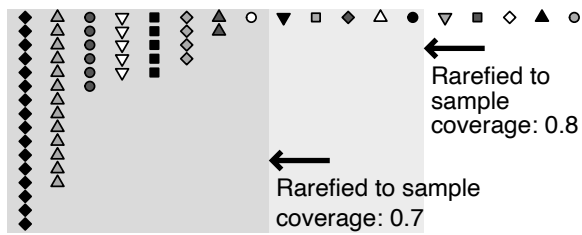
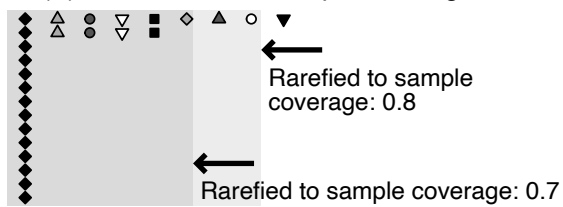


Figure 2: An illustration of the relationship between size-based and coverage-based rarefaction, showing that sample coverage is determined by subtracting the slope of the rarefaction curve from 1. These data are from the Hlatimbe Valley 213 assemblage (Labandeira et al., 2018), which was selected for inclusion here because it is also featured in Figure 3.

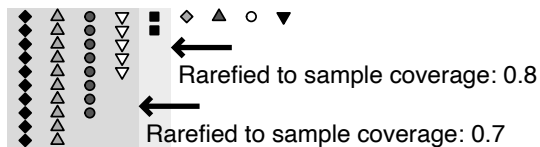
(A) Nuwejaarspruit 111B, sample coverage = 0.82



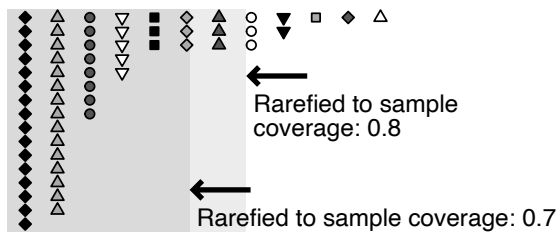
(B) Sani Pass 111, sample coverage = 0.85



(C) Klein Hoek 111B, sample coverage = 0.90



(D) Greenvale 121, sample coverage = 0.95



(E) Hlatimbe Valley 213, sample coverage = 0.97

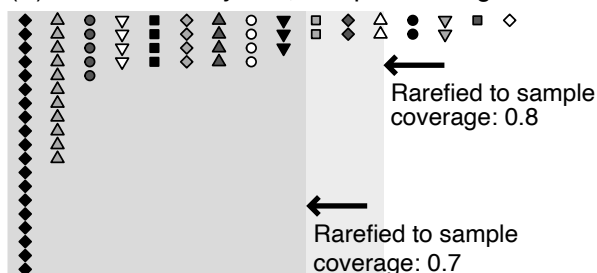


Figure 3: An illustration of the concepts of sample coverage and of rarefaction by sample coverage. The five assemblages illustrated here were examined by [Labandeira et al. \(2018\)](#). For each assemblage, the sample coverage for the complete dataset is listed in the title. The lightest gray boxes denote the damage type diversity at each assemblage when rarefied to a sample coverage of 0.8, and the darker gray boxes denote damage type diversity rarefied to a sample coverage of 0.7. The assemblages are organized from the lowest to highest levels of sample coverage. Each column of symbols represents a damage type, and the number of times each symbol is illustrated represents the number of specimens on which the damage type was observed. (A) has relatively low sample coverage because the majority of damage types are observed on only one specimen. In contrast, (E) has relatively high sample coverage because so few damage types are observed on only one specimen.

182 Because coverage-based rarefaction follows the replication principle (Chao and Jost, 2012) it provides
183 an unbiased and consistent estimator of damage type diversity that is robust to sample size, leaf size, and
184 fragmentation. In statistical parlance, an “unbiased estimator” is an estimator whose average expected value
185 for a sample is equal to the true value in the population from which the sample was drawn. In other words,
186 whether the dataset for a given fossil assemblage reaches a sample coverage of 0.85 or 0.99, the estimated
187 damage type diversity will, on average, be the same when the dataset is rarefied down to a sample coverage of
188 0.8. A “consistent estimator” converges on the true population value as sample size becomes large. In other
189 words, if one fossil assemblage reaches a sample coverage of 0.85 and another reaches a sample coverage of
190 0.99, the uncertainty surrounding damage type diversity rarefied to a sample coverage of 0.8 will be smaller
191 for the latter, more completely-sampled assemblage. The assemblages listed in Tables 1 and 2 with over
192 7,000 leaves examined were randomly subsampled down to 1,000 and 2,000 leaves in a procedure that was
193 iterated 1,000 times (Figure 4). Whereas accuracy and precision typically suffer when these assemblages are
194 subsampled down to 1,000 or even 2,000 leaves before performing traditional, size-based rarefaction (Figure
195 S3), the loss of precision and accuracy are minimal with coverage-based rarefaction (Figure 4). The loss of
196 precision with coverage-based rarefaction is far less severe than with size-based rarefaction (Figure S3) or
197 with the Chao1 estimator (Figure S2) and still allows the differentiation of damage type diversity among
198 many assemblages. With coverage-based rarefaction, the most notable lack of overlap among the confidence
199 intervals for the different sample sizes occurs for Clouston Farm (Figure 4). The confidence intervals overlap
200 to a reassuring extent for half of the pre-angiosperm and all of the angiosperm assemblages (Figure 4).

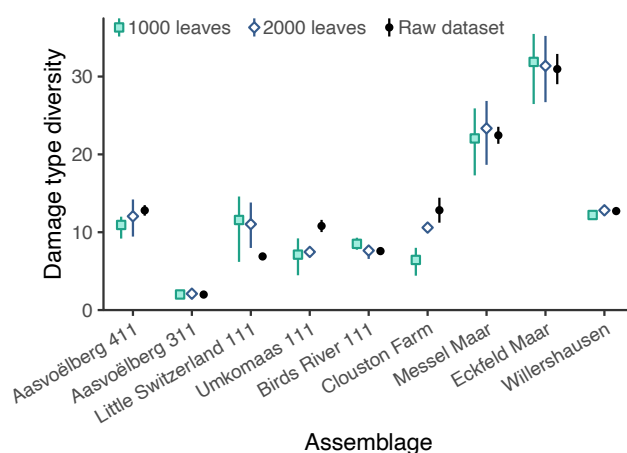


Figure 4: Precision and accuracy of estimates of damage type diversity generated with coverage-based rarefaction.

201 An additional benefit of coverage-based rarefaction is that, because samples are standardized by
202 damage type completeness rather than number of leaves examined, this method is relatively robust to the

203 variable inclusion of non-broadleaf specimens, from pine needles to axes to seeds, in the datasets for
204 different assemblages. Some published comparisons of insect herbivory among assemblages include
205 broadleaf specimens only, whereas others include all plant specimens. Some floral assemblages do not
206 contain any seeds and other non-foliar specimens, and of those that do, the non-foliar specimens may or
207 may not be included in published datasets. Within angiosperm-dominated floras, ferns, gymnosperms, and
208 monocots may (Robledo et al., 2018; Giraldo et al., 2021) or may not (Azevedo Schmidt et al., 2019;
209 Currano et al., 2019) be examined. Criteria for inclusion in comparative analyses can even vary within the
210 same research group, with some workers using a wider definition of foliage that includes needles, liverworts,
211 phyllids, photosynthetic wings of seeds, and even flattened horsetail axes (Prevec et al., 2009; Labandeira
212 et al., 2018), and others using a narrower definition restricted to multi-veined broad leaves or leaves with a
213 defined midvein (Schachat et al., 2014, 2015, 2018, 2020). (For additional information, see “criteria for
214 inclusion of leaves and damage types” in the supplemental information.)

215 Because non-broadleaf specimens typically contain little or no evidence of insect herbivory, they contribute
216 little or nothing to metrics of damage type sampling completeness and will therefore have minimal or no
217 impact on estimates of damage type diversity at an assemblage. And, because some herbivory datasets code
218 plants as morphotypes only, with no information on whether the morphotype represents a dicot or a conifer
219 (Maccracken, 2020), these datasets cannot be analyzed with traditional size-based rarefaction. However,
220 these datasets can indeed be analyzed with coverage-based rarefaction. Plant organs that contain little or no
221 insect damage, such as axes and pine needles, contribute little or nothing to estimates of sample coverage,
222 and thus their inclusion in coverage-based damage diversity estimates does not bias the estimates downward.

223 A remaining question is the confidence level that should be used in confidence intervals for rarefied
224 damage type diversity. Schachat et al. (2018) used 84% confidence intervals, rather than 95%, because the
225 comparison of two curves with 84% confidence intervals yields a Type I error rate below 5%. However, the
226 confidence intervals generated with the iNEXT package rely in part on extrapolated diversity, which is shown
227 above to be biased by sample size in the case of damage type diversity datasets. As a result, the confidence
228 intervals generated by iNEXT are narrower than those generated with the iterative procedure of Schachat
229 et al. (2018). Therefore, we recommend generating 95% confidence intervals with the iNEXT package for use
230 in the ecospace.

231 2.2.1 CONSIDERATIONS FOR ASSEMBLAGES WITH INSUFFICIENT COVERAGE

232 Just as size-based rarefied estimates of damage type diversity are conducted at a predetermined level of
233 sampling, such as 400 leaves, coverage-based rarefied estimates of damage type diversity require a
234 predetermined level of coverage, or sampling completeness. Here we rarefy to sample coverage 0.8.

235 Coverage increases throughout the sampling process as a function of sample size, the evenness of damage
236 types, the richness of damage types, and the frequency with which damage types occur. Accordingly, there
237 is no predetermined number of leaves or amount of surface area at which an assemblage is assured to have
238 sufficient coverage.

239 The impossibility of assessing all assemblages that have been sampled to a certain size, such as 400 leaves
240 or 1000 cm² of surface area, is perhaps the only disadvantage of using coverage-based rarefaction instead of
241 size-based rarefaction. Five of the assemblages in Tables 1 and 2 have insufficient coverage for rarefaction
242 to sampling completeness of 0.8: the Permian Wuda (Feng et al., 2020) and Bletterbach (Bernardi et al.,
243 2017) assemblages, and the Triassic Qachasnek 111, Kraai River 111, and Konings Kroon 222 assemblages
244 (Labandeira et al., 2018). (The Triassic Makoaneng 111 and Konings Kroon 111A assemblages both have a
245 total sample coverage of 0.79. We extrapolated the damage type diversity at a sample coverage of 0.8 for
246 these two assemblages, but do not recommend extrapolating diversity estimates for assemblages with coverage
247 below 0.79, as discussed in the supplemental material.) All of these contain little damage. Of note, the
248 Boesmanshkoek 112 assemblage from the same Triassic study contains only two damage types, but these
249 occur frequently enough that the assemblage has sufficient coverage. Thirty-nine of the 44 assemblages
250 with fewer than 1,000 leaves listed in Table 3 have sufficient coverage as well. Of the ten assemblages with
251 coverage below 0.79, eight have values between 0.45 and 0.69, and the remaining two assemblages have values
252 of 0.70 and 0.72. Rarefying to a sample coverage of 0.7 would decrease the power to detect differences among
253 assemblages while providing a negligible increase in the number of assemblages included.

254 The best way to handle assemblages with insufficient coverage depends on the number of leaves examined
255 and the temporal scope of the study. In the case of assemblages with only a few hundred leaves, estimates
256 of damage type diversity comparable to those generated with coverage-based simply cannot be calculated.
257 For studies such as that of Labandeira et al. (2002b) with a limited temporal scope and a large quantity of
258 assemblages, many of which contain small numbers of leaves, the exclusion of assemblages that do not reach
259 sufficient sample coverage can decrease the power of the study to detect changes in herbivory through time
260 by excluding assemblages whose low sample coverage may be due to insufficient sampling or a true scarcity
261 of damage types. In cases such as this, size-based rarefaction curves scaled by the amount of surface area
262 sampled rather than the number of leaves sampled (Schachat et al., 2018) are the most appropriate choice.
263 These size-based estimates of damage type diversity cannot be incorporated into meta-analyses that use
264 coverage-based rarefaction. We conducted simulations to evaluate extrapolated estimates of damage type
265 diversity in cases when the raw dataset does not reach a sample coverage of 0.8, and found that extrapolated
266 estimates are not reliable and are often invalid (Figure S4).

2.3 EVALUATING OTHER POTENTIAL DIMENSIONS OF AN ECOSPACE FOR HERBIVORY

Whereas the intensity and diversity of insect damage are the two most commonly studied aspects of insect herbivory and are the two most obvious choices for dimensions of an ecospace, additional dimensions bear consideration. Examples discussed here include the proportion of feeding occurrences belonging to external foliage feeding and to piercing and sucking; damage type evenness; damage type evenness compared to floral evenness; floral diversity; damage type diversity compared to floral diversity; and the prevalence of each functional feeding group, best quantified through the amount of herbivorized surface area. Feeding occurrences are defined here as the number of times that a damage type occurs on an individual plant specimen; these data are rarely collected in studies of fossil herbivory (Robledo et al., 2018; Ma et al., 2020).

The first additional dimension is the prevalence of external foliage feeding, a functional feeding group containing generalist modes of herbivory in which an insect typically chews on a leaf. The proportion of feeding occurrences belonging to external foliage feeding—as opposed to piercing and sucking or specialized herbivory such as mining and galling—is perhaps the metric with the greatest potential to distinguish between insect evolution and nutrient dilution, the two most common explanations of increased herbivory invoked by paleontologists.

When nutrient dilution occurs, the prevalence of piercing-and-sucking feeding damage—whether measured as number of damage types, percentage of leaf area damaged, or number of feeding occurrences—will likely remain constant. This is because piercing-and-sucking insects often feed by puncturing individual phloem cells (Will et al., 2013). Whereas the number of phloem cells may increase in response to nutrient dilution, the content of each cell and the pressure within each cell will most likely remain the same. Therefore, piercing-and-sucking insects that feed on phloem will not need to feed more in order to meet their nutritional requirements. Similarly, when nutrient dilution occurs, the number of specialized mining and galling damage types is unlikely to increase, as is the number of occurrences of these damage types. Gall-inducing insects hijack their host plant’s metabolism; their control over the tissues that form galls may well be sufficient to shield them from the effects of nutrient dilution. However, the amount of surface area lost to mining may increase if larvae need to make longer mines in order to fulfill their nutritional requirements. External foliage feeding is the only broad category of herbivory for which one would expect to see the number of feeding event occurrences increase with nutrient dilution. External foliage feeders, like leaf miners, will need to consume more leaf surface area if nutrient dilution reduces the nutritional quality of the leaf blade. And unlike leaf miners, external foliage feeders have the opportunity to partition the additional leaf surface area they consume among multiple feeding event occurrences.

299 Whereas “damage type diversity” as discussed here is a measure of the richness of damage types, the
300 evenness of herbivore damage also holds potential to inform differences in herbivory among assemblages.
301 Gunkel and Wappler (2015) extended the insights of Olszewski (2004) regarding the relationship between
302 rarefaction and evenness to the study of insect herbivory. A number of other authors have also quantified
303 evenness—of the floral community, of damage types, or both—when evaluating insect herbivory in the fossil
304 record (Wappler et al., 2015; Currano et al., 2019; Azevedo Schmidt et al., 2019). These too can be used
305 as additional dimensions of the herbivory ecospace. Of note, the frequencies of each plant host and damage
306 type can be replaced with their respective surface areas when calculating evenness metrics.

307 Another metric, related to evenness, is the offset between the prevalence of each plant host and the
308 prevalence of insect damage on it. For example, if a locality contains two plant hosts, one that accounts for
309 70% of total leaf surface area and another that accounts for 30%, a null expectation would be that 70% of
310 herbivorized surface area would belong to the first plant host and 30% would belong to the second. The
311 extent to which this null expectation is violated can be quantified by summing the absolute values of the
312 differences between each plant host’s prevalence in the assemblage and the proportion of herbivorized leaf
313 area that it contains. In this example, if the first host plant contains 65% of herbivorized leaf area and
314 the second contains 35%, the offset between total and herbivorized leaf area summed across the locality is
315 $|0.7 - 0.65| + |0.3 - 0.35| = 0.1$. If, on the other hand, the first host plant contains only 25% of herbivorized
316 leaf area and the second contains 75%, the offset between total and herbivorized leaf area summed across
317 the locality is $|0.7 - 0.25| + |0.3 - 0.75| = 0.9$. This evenness-offset score ranges from 0 (no offset) to 2
318 (maximum offset). This metric can be leveraged as an additional dimension for an herbivory ecospace. An
319 84% confidence interval for this evenness-offset score can be generated by resampling plants and damage type
320 occurrences, with replacement, to the amount of surface area observed in the original dataset, calculating
321 an evenness-offset score for the resampled datasets, and iterating this process 5,000 times. (For discussion
322 of the number of iterations needed, see the supplemental material.)

323 Floral diversity, and the relationship of floral diversity to damage type diversity, can comprise additional
324 dimensions of the herbivory ecospace. A damage-diversity-to-floral-diversity ratio, in which rarefied damage
325 type diversity is divided by rarefied floral diversity, holds the potential to reveal the extent to which floral
326 diversity underlies damage type diversity. A confidence interval for this ratio can be calculated by randomly
327 selecting a value within the confidence interval for rarefied damage type diversity, randomly selecting a value
328 within the confidence interval for rarefied floral diversity, dividing the former by the latter, and iterating this
329 process 1,000 times. The number of specimens required to rarefy floral diversity to sample coverage of 0.8
330 may differ from the number of specimens required to rarefy damage type diversity to sample coverage of 0.8.

331 Lastly, the prevalences of each functional feeding group within an assemblage can comprise yet more

332 dimensions of the ecospace. Ideally, the prevalence of each functional feeding group would be quantified by
333 the absolute and relative amounts of herbivorized surface area that it represents. However, as noted above,
334 surface area measurements are unavailable for nearly all herbivory datasets. Thus, the proportion of damage
335 type occurrences attributable to each functional feeding group can instead be used as a proxy. Confidence
336 intervals for the proportions of either herbivorized area or damage type occurrences can be calculated by
337 iteratively resampling with replacement as discussed above.

338 Additional measures, such as bipartite network metrics (*sensu* Blüthgen et al., 2008) and host specificity,
339 require so much data that it may not be possible to precisely and accurately estimate them with a typical
340 fossil herbivory dataset. This is further discussed in the supplemental material (Figure S6; “evaluation of
341 other metrics”).

342 2.4 THE OPTIMAL NUMBER OF DIMENSIONS IN THE ECOSPACE

343 When traditional multivariate statistics are used in ecological studies, scree plots allow researchers to
344 determine the number of axes that warrant interpretation (McGarigal et al., 2013). We are not aware of
345 any analysis comparable to a scree plot that would facilitate determination of whether all of the above
346 dimensions of an ecospace warrant consideration. However, we also do not see any *a priori* reason to
347 expect that all of these dimensions will be informative in all studies of insect herbivory. We have listed all
348 of these dimensions because they are already used in studies of insect herbivory and we aim to standardize
349 statistical practices and to encourage use of the most robust metrics. It is entirely reasonable to assume at
350 the beginning of any study that damage type diversity and the herbivory index are the only necessary
351 dimensions of this ecospace unless another dimension can directly address any hypotheses under
352 consideration. If two or more dimensions of an ecospace are highly colinear, or if one or more dimensions
353 contribute only statistical noise, the most appropriate course of action is to eschew those dimensions.

354 3 THE AMOUNT OF SAMPLING REQUIRED

355 The ability to determine *a priori* how much sampling is needed for a given assemblage—or whether an
356 assemblage contains sufficient material for robust comparisons with other assemblages—would allow
357 investigators to allocate their efforts most efficiently. Unfortunately, our results show that neither the
358 number of leaves nor the amount of surface area in an assemblage are reliable predictors of sufficient
359 sampling for coverage-based rarefaction. The frequency of damage on leaves is a far more reliable predictor,
360 but is much more difficult to estimate prior to data collection.

361 One advantage of coverage-based rarefaction is that it can be integrated with a “stopping rule”. In the case

362 of the sample coverage threshold of 0.8 used here, sample coverage can be re-calculated whenever new data
363 are added to a dataset and sampling can be considered sufficient when the threshold of 0.8 has been reached.
364 Although stoppage of data collection at this threshold would allow a maximum number of assemblages to be
365 examined and compared, it is worth noting that the confidence intervals surrounding damage type diversity
366 and the herbivory index will narrow if sampling continues beyond a sample coverage of 0.8.

367 As discussed in a recent contribution, estimates of damage type diversity require more sampling than
368 estimates of the herbivory index do (Schachat et al., 2020). As shown in the supplemental material
369 (“evaluation of other metrics”; Figure S6), even more sampling would be needed to estimate host-specificity
370 of individual damage types and to estimate more complex metrics that match particular damage types
371 with particular host plants; the extreme sensitivity of such metrics to sample size has already been
372 established in the neontological literature (Blüthgen, 2010).

373 In fact, it has yet to be demonstrated that any realistic amount of sampling completeness in studies
374 of fossil insect herbivory is adequate for reliable calculation of complex metrics such as bipartite network
375 properties. The sampling completeness necessary for reliably calculating bipartite network properties could
376 be established by iteratively subsampling an assemblage to half of its original leaf surface area and calculating
377 network properties for each subsample. The results of this subsampling routine could be used to generate a
378 confidence interval for each network property. If, for each network property, the confidence interval is narrow
379 enough that it does not overlap with the confidence intervals generated for other assemblages subsampled
380 to the same amount of surface area, the amount of surface area available is sufficient to reliably calculate
381 network properties. Of note, however, the requisite amount of surface area will vary with the dominance–
382 diversity structure of the plant host and damage type communities at each assemblage. It is not yet clear
383 whether it is worth pursuing sufficient sampling for the implementation of bipartite network analyses.

384 4 GENERATION OF NULL HYPOTHESES

385 The study of insect herbivory in the fossil record has become vastly more popular over the past few decades.
386 Early studies that examined patterns of insect herbivory caused by a major event, such as a mass extinction,
387 had only one available option: to quantify the magnitude of the differences in insect herbivory before and
388 after the event in question. A question that could not be addressed at the time those early studies were
389 conducted, but can be addressed now, is whether the amount of change in insect herbivory after a major
390 event exceeds the amount of change that one would expect if the event had not happened.

391 Returning to the end-Triassic example mentioned in the Introduction, a number of factors preclude the
392 attribution of changes in herbivory across a boundary to the events that occurred at the boundary. First, the

393 difficulty of finding floral assemblages deposited immediately before or after the Triassic/Jurassic boundary,
394 and establishing their proximity to the boundary, necessitates that this event be examined with Triassic and
395 Jurassic floras that are separated by millions of years—a long enough time for plant and insect communities
396 to change without any extraordinary abiotic events. Second, even within a single basin, plant communities,
397 soils, precipitation, temperature, and perhaps also predators and parasitoids change with major abiotic
398 events, introducing various potential causes of changes in insect herbivory that may not be directly related
399 to the abiotic event in question (Currano et al., 2019). On a related note, although nutrient dilution is
400 expected to cause an increase in the herbivory index, several of these alternative phenomena could also cause
401 such an increase. Third, extant ecosystems demonstrate that coeval plant communities in close geographic
402 proximity can vary tremendously due to microclimate and other factors such as soil type and frequency of
403 disturbance (Tamme et al., 2010); the difficulty of identifying a characteristic flora for a particular place and
404 time makes it more difficult still to compare characteristic patterns of insect herbivory for a particular place
405 across different time slices.

406 To overcome these obstacles, many studies of insect herbivory have examined multiple plant assemblages
407 from each time interval of interest (Labandeira et al., 2002a; Currano et al., 2008; Wappler, 2010; Donovan
408 et al., 2018). If the assemblages from each time interval form distinct clusters, in an NMDS plot for example,
409 this provides far stronger evidence of change over time than can be gleaned from only two assemblages. The
410 outstanding question is how much change over time is to be expected in the absence of an event such as
411 an environmental disturbance (the amount of change expected as a null hypothesis), and how much change
412 can be attributed to the event under consideration (the amount of change needed to support the alternative
413 hypothesis that the event in question caused significantly more change in patterns of insect herbivory than
414 can be explained only by the passage of time).

415 The theoretical ecospace outlined here holds the potential to go beyond an examination of the number of
416 damage types observed per number of leaves, by quantifying the following. First is the amount of variation
417 in insect herbivory to be expected among coeval floras that occur in close proximity and have slightly
418 different plants, soils, depositional conditions, and microclimates. Second is the amount of variation in
419 insect herbivory to be expected on the time scales examined in relevant studies (10 Kyr, 100 Kyr, 1 Myr,
420 10 Myr) in the absence of any major environmental changes. Third is the amount of, and the nature of,
421 changes in insect herbivory associated with environmental changes, such as the Cretaceous/Paleogene Event
422 (Labandeira et al., 2002b; Donovan et al., 2018) and the Paleocene/Eocene Thermal Maximum (Currano
423 et al., 2008). Fourth is the amount of, and the nature of, changes in insect herbivory associated with insect
424 evolution over tens of millions of years. Nearly any comparison of floral assemblages will yield evidence
425 of different patterns of insect herbivory; the key question is how much change needs to occur in order to

426 rise above background levels. In the future, as data become available to include increasing numbers of
427 assemblages in this theoretical ecospace, confidence intervals can be established in each dimension for these
428 four types of changes in insect herbivory.

429 Traditional morphospaces have already been leveraged to generate null hypotheses and to disentangle
430 the results of directional evolution from changes through time that are best attributed to a “random walk”
431 model (Pie and Weitz, 2005; Puttick et al., 2020). The same logic can be extended to the study of insect
432 herbivory in the fossil record.

433 5 PRELIMINARY RESULTS

434 5.1 COMPARABLE HERBIVORY DURING THE PALEOZOIC, MESOZOIC, AND 435 CENOZOIC

436 When damage type diversity from the assemblages in Tables 1, 2, and 3 is plotted against time, the most
437 striking result is the overlap of damage type diversity among Permian, Triassic, and angiosperm-dominated
438 assemblages (Figure 5). The dataset analyzed here contains five times as many angiosperm-dominated
439 assemblages as it does Paleozoic assemblages, and over three times as many angiosperm-dominated
440 assemblages as Triassic assemblages. And yet the Laguna Polina assemblage from the Permian of
441 Patagonia (Cariglino, 2018), the Kühwiesenkopf assemblage from the Triassic of Italy (Labandeira et al.,
442 2016), and Cyphergat 111A assemblage from the Triassic of South Africa (Labandeira et al., 2018) all
443 contain damage type diversities within the highest 12% of values for angiosperm-dominated assemblages.
444 The supplemental material includes a sensitivity analysis and a discussion of the role of rare damage types
445 in diversity estimates.

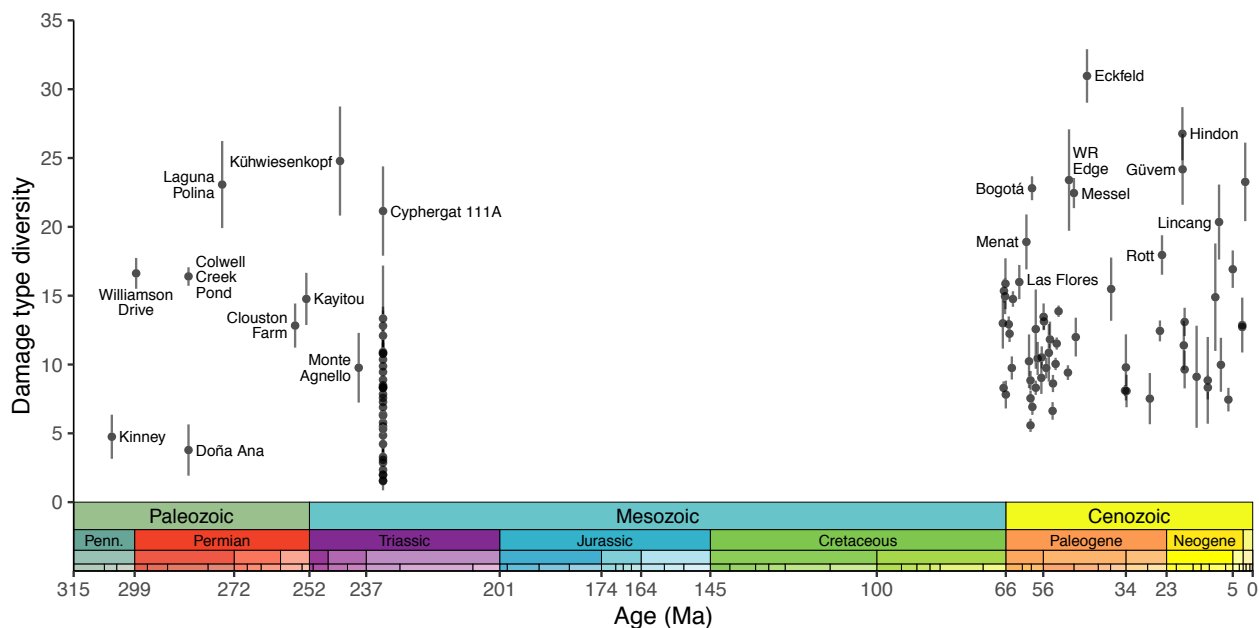


Figure 5: Damage type diversity estimated with coverage-based rarefaction, plotted against time.

446 The minimal increase in the upper bound of damage type diversity from the Paleozoic through Cenozoic
 447 is further supported by the two-dimensional ecospace which we advocate for here (Figure 6). The available
 448 data are sufficient to include only eight assemblages in this ecospace: Williamson Drive (Xu et al., 2018)
 449 and Colwell Creek Pond (Schachat et al., 2015) from the late Paleozoic; and Daiye Spa, Hubble Bubble,
 450 PN, Fifteenmile Creek, Republic, and Bonanza from the interval surrounding the Paleocene/Eocene Thermal
 451 Maximum (Wilf et al., 2001; Labandeira, 2002; Currano et al., 2008, 2010, 2016). This ecospace also supports
 452 the prediction outlined earlier in this contribution that an abiotic event—in this case, exemplified by the
 453 Hubble Bubble locality during the PETM—can indeed lead to a change in damage type diversity, but most
 454 clearly manifests with a change in the herbivory index.

455 This finding of similar damage type diversities in Paleozoic, Triassic, and angiosperm-dominated
 456 assemblages may be somewhat surprising given the angiosperm potential for uniquely high productivity
 457 (Boyce and Zwieniecki, 2012) and their tremendous diversity (Soltis and Soltis, 2004)—and especially their
 458 codiversification with insects (Pellmyr, 1992). Leaf mining, for example, is a highly specialized form of
 459 herbivory that accounts for a wide array of described damage types (Eiseman, 2019) and occurs
 460 overwhelmingly on angiosperm leaves (Scott et al., 1992). One might expect the origin of angiosperms to
 461 have caused a diversification of leaf-mining insects, which in turn would have caused an increase in damage
 462 type diversity. However, the data do not support an overall increase in damage type diversity in
 463 angiosperm-dominated assemblages.

464 The difficulty of quantifying the role of angiosperm diversity and biomass in shaping extant ecosystems

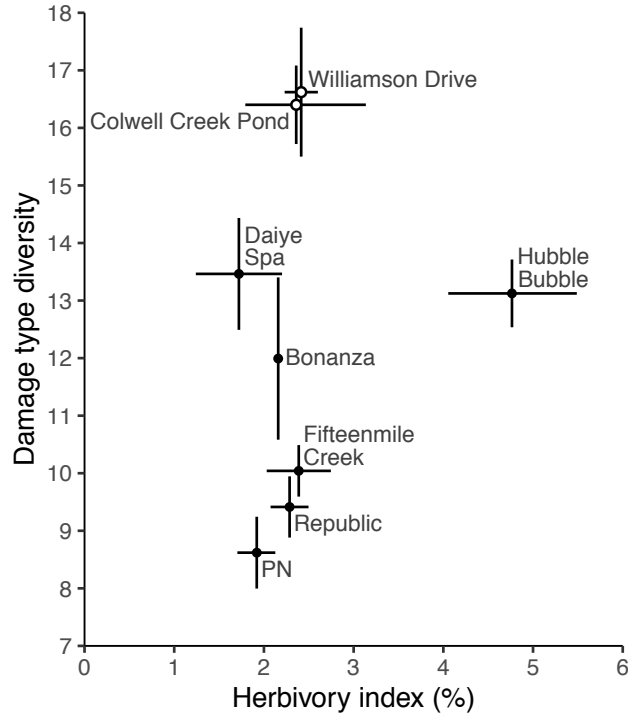


Figure 6: Damage type diversity plotted against the herbivory index in a two-dimensional ecospace. Paleozoic localities are denoted with white circles and angiosperm localities are denoted with dark circles. The confidence intervals for the herbivory index were derived with the method of [Schachat et al. \(2018\)](#) for Williamson Drive and Colwell Creek Pond, and with the method of [Currano et al. \(2016\)](#) for all other assemblages.

465 stems largely from their overwhelming dominance. If angiosperms accounted for 70% of the diversity and
466 biomass of land plants, any disproportionate ecological role they may fulfill—for example, if they hosted
467 95% of leaf mine damage types—would be easier to establish. But, because angiosperms account for
468 approximately 96% of described vascular plant species ([Christenhusz and Byng, 2016](#)), any
469 disproportionate ecological role they may fulfill—for example, if they hosted 98% of leaf mine damage
470 types—may be statistically indistinguishable from the null expectation of this ecological role (in the case of
471 this example, that angiosperms would host 96% of leaf mine damage types). To use another hypothetical
472 example, if it were known with absolute certainty that angiosperms host only 93% of leaf mine damage
473 types, one could still say that the overwhelming majority of leaf mine damage types occur on angiosperms,
474 such that leaf mine diversity and thus overall damage type diversity likely increased alongside the radiation
475 of flowering plants.

476 5.2 THE ROLE OF $p\text{CO}_2$ DEPENDS ON THE RATE OF ATMOSPHERIC CHANGE

477 The changes in damage type diversity in Figure 5 show no general relationship with $p\text{CO}_2$. For example, the
478 Kayitou assemblage dates to the late Lopingian, an interval for which many geochemical models reconstruct
479 high $p\text{CO}_2$ (Mills et al., 2019), but Kayitou does not have particularly prevalent insect damage (Liu et al.,
480 2020) nor is its damage type diversity unusually high (Figure 5).

481 Rising $p\text{CO}_2$ may drive nutrient dilution and, thus, increased insect herbivory, but stable $p\text{CO}_2$ —whether
482 high or low—may favor instead compensatory physiological changes that allow for stable rates of carbon
483 fixation under different atmospheres. The carbon cycle perturbation of the PETM may have constituted
484 an atmospheric change too rapid for plants to adapt to; it is under this circumstance that nutrient dilution
485 may be most likely to occur. For this reason, the increased intensity of herbivory seen at the PETM is not
486 a template for the changes in herbivory that would occur if $p\text{CO}_2$ rose slowly enough for plants to remain
487 well adapted to the atmospheric composition.

488 6 CONSIDERATIONS FOR FUTURE STUDIES

489 Overlapping damage type diversities through time are far easier to demonstrate than changing diversities.
490 Labandeira et al. (2002b) evaluated damage type diversity before and after the end-Cretaceous event. Their
491 study included 17 beds with 200 or more dicot leaves: eleven Cretaceous and six Paleogene. Coverage-
492 based rarefaction supports the authors' conclusion that damage type diversity was higher during the latest
493 Cretaceous than during the earliest Paleogene. However, damage type diversities vary widely within each of
494 these two intervals. These fossil beds all occur within the same basin, and the beds from each interval are
495 dominated by a single type of depositional environment. Aside from the first-order finding of less herbivory
496 in Paleogene beds, variation in damage type diversity does not occur in a predictable manner through time.

497 This variability observed within 1 Myr in a single basin highlights the difficulty of establishing a
498 representative amount of damage type diversity for a particular interval—even for a single depositional
499 environment within a single basin during a very small amount of geologic time. The low signal-to-noise
500 ratio among the damage type diversities for a pooled set of assemblages highlights the difficulty of
501 identifying temporal trends in insect herbivory.

502 Additional factors further conceal temporal trends. First is the obvious variability among assemblages
503 in geographic location, insect taxa, plant taxa, soil type, and climate. Second is uneven sampling across
504 time. No studies of herbivory for an entire flora have been published for the Jurassic, and within the
505 Cenozoic, the Paleogene is far better sampled than the Neogene or Quaternary. Third, especially for the

506 Cenozoic, assemblages are chosen for study based largely on their relevance to events such as the end-
507 Cretaceous extinction (Labandeira et al., 2002b; Wilf et al., 2006; Wappler et al., 2009; Donovan et al., 2017,
508 2018), the Paleocene/Eocene Thermal Maximum (Wilf and Labandeira, 1999; Wilf et al., 2001; Currano
509 et al., 2008; Currano, 2009; Currano et al., 2010), and the Early Eocene Climatic Optimum (Currano et al.,
510 2016, 2019). The assemblages for which we have data are disproportionately likely to come from the least
511 representative part of a geologic interval, such as the very earliest Eocene, because these time slices epitomize
512 the phenomena that underlie paleontologically interesting ecological and evolutionary questions. The uneven
513 temporal distribution of studied assemblages adds far more variability to reconstructed long-term trends
514 than one would expect if assemblages were chosen for study in an unbiased manner, and has the potential
515 to obscure these trends.

516 7 CONCLUSIONS

517 As the proliferation of free software packages and online databases continues, it becomes easier to run analyses
518 and generate graphs with insect herbivory data—regardless of whether the methods are appropriate for the
519 data and regardless of whether the datasets are sufficiently complete to meet the assumptions of the methods.
520 As shown here, the Chao1 estimator, a method that is particularly well-suited to handle the sparseness of
521 insect herbivory datasets, rarely provides estimates of damage type diversity that are both precise enough to
522 minimize the frequency of false negative results and accurate enough to contain the true, asymptotic value.
523 Moreover, asymptotic damage type diversity calculated through a combination of rarefaction and the
524 Chao1 estimator is formulated as an unbiased estimator but is severely biased by sample size, presumably
525 due to the exceptional sparsity of herbivory datasets. This finding highlights the need to regress estimators
526 of insect herbivory against the number of leaves and damage types in raw datasets to verify that they are
527 not biased by sampling effort.

528 Whereas size-based rarefaction and the Chao1 estimator present insurmountable limitations for insect
529 herbivory data, coverage-based rarefaction holds the most promise for generating estimates of damage type
530 diversity that are unbiased, and robust to leaf size and sample size. Combined with the herbivory index,
531 which measures the intensity of insect herbivory, estimates of damage type diversity can be leveraged as the
532 foundation of a theoretical ecospace. This ecospace holds the potential to address two overarching questions
533 that remained unresolved for two decades. First is how to distinguish the minor variation in insect herbivory
534 that inevitably occurs among assemblages from slightly different times or places (null hypothesized amount of
535 variation) from the changes in insect herbivory that accompany major abiotic events or innovations in insect
536 evolution (alternative hypothesis). Second is how to distinguish among abiotic causes (e.g., mass extinctions)

537 and biotic causes (insect and plant evolution) of major changes in the amount of insect herbivory.

538 There are two ways to build upon the many fossil herbivory datasets amassed over the past few
539 decades: collecting more data, and interrogating existing data to develop and refine analytical frameworks.
540 By advancing the latter, a theoretical ecospace for insect herbivory will hopefully underscore the
541 importance the former.

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Assemblage	Leaves	DTs	Period	Age	Citation
Birds River 111	7,106	35	Triassic	Carnian	Labandeira et al. (2018)
Greenvale 121	2,749	12	Triassic	Carnian	Labandeira et al. (2018)
Boesmanskhoek 112	1,058	2	Triassic	Carnian	Labandeira et al. (2018)
Cyphergat 111A	5,499	26	Triassic	Carnian	Labandeira et al. (2018)
Kannaskop 111	2,114	11	Triassic	Carnian	Labandeira et al. (2018)
Telemachus Spruit 111	5,860	14	Triassic	Carnian	Labandeira et al. (2018)
Kommandantskop 111	1,096	8	Triassic	Carnian	Labandeira et al. (2018)
Vineyard 111	2,146	10	Triassic	Carnian	Labandeira et al. (2018)
Elandspruit 111	1,054	10	Triassic	Carnian	Labandeira et al. (2018)
Kraai River 311	1,213	6	Triassic	Carnian	Labandeira et al. (2018)
Kraai River 111 ^a	1,780	8	Triassic	Carnian	Labandeira et al. (2018)
Lutherskop 311	5,180	22	Triassic	Carnian	Labandeira et al. (2018)
Waldeck 111	1,322	11	Triassic	Carnian	Labandeira et al. (2018)
Konings Kroon 222 ^a	1,679	13	Triassic	Carnian	Labandeira et al. (2018)
Konings Kroon 111A	1,073	5	Triassic	Carnian	Labandeira et al. (2018)
Peninsula 321	1,311	4	Triassic	Carnian	Labandeira et al. (2018)
Peninsula 311	1,665	12	Triassic	Carnian	Labandeira et al. (2018)
Peninsula 411	6,254	17	Triassic	Carnian	Labandeira et al. (2018)
Klein Hoek 111B	1,004	9	Triassic	Carnian	Labandeira et al. (2018)
Klein Hoek 111C	2,644	17	Triassic	Carnian	Labandeira et al. (2018)
Kappokraal 111	1,453	27	Triassic	Carnian	Labandeira et al. (2018)
Elandspruit 112A	1,131	5	Triassic	Carnian	Labandeira et al. (2018)
Nuwejaarspruit 111B	1,582	18	Triassic	Carnian	Labandeira et al. (2018)
Winnaarspruit 111	1,396	16	Triassic	Carnian	Labandeira et al. (2018)
Morija 111B	1,235	7	Triassic	Carnian	Labandeira et al. (2018)
Makoaneng 111	1,206	6	Triassic	Carnian	Labandeira et al. (2018)
Hlatimbe Valley 213	1,526	16	Triassic	Carnian	Labandeira et al. (2018)
Umkomaas 111	9,828	36	Triassic	Carnian	Labandeira et al. (2018)
Sani Pass 111	1,236	9	Triassic	Carnian	Labandeira et al. (2018)
Qachasnek 111 ^a	2,093	7	Triassic	Carnian	Labandeira et al. (2018)
Matatiele 111	3,667	23	Triassic	Carnian	Labandeira et al. (2018)
Golden Gate 111	1,208	16	Triassic	Carnian	Labandeira et al. (2018)
Little Switzerland 111	8,218	28	Triassic	Carnian	Labandeira et al. (2018)
Aasvoëlberg 111	2,941	11	Triassic	Carnian	Labandeira et al. (2018)
Aasvoëlberg 211	1,942	12	Triassic	Carnian	Labandeira et al. (2018)
Aasvoëlberg 311	10,809	18	Triassic	Carnian	Labandeira et al. (2018)
Aasvoëlberg 411	12,398	39	Triassic	Carnian	Labandeira et al. (2018)

Table 1: The assemblages with over 1,000 broadleaf specimens from [Labandeira et al. \(2018\)](#), collected from the Molteno formation by [Anderson and Anderson \(1983, 1985, 1989, 2003, 2008, 2017\)](#), that were analyzed in this study. ^aThese assemblages do not have sufficient sample coverage to be rarefied to a coverage of 0.8.

Assemblage	Leaves	DTs	Period	Age	Citation
Willershausen	7,932	85	Neogene	Piacenzian	Adroit et al. (2018)
Lincang ^a	1,103	36	Neogene	?Tort.–Mess.	Zhang et al. (2018)
Tröllatunga-Gautshamar	1,069	17	Neogene	Tortonian	Wappler and Grímsson (2016)
Brjónslækur-Seljá	1,613	29	Neogene	Serravallian	Wappler and Grímsson (2016)
Bílina–DSH	2,233	54	Neogene	Burdigalian	Knor et al. (2012)
Břešťany–LCH	1,181	28	Neogene	Burdigalian	Knor et al. (2012)
Rott	2,474	55	Paleogene	Chattian	Wappler (2010)
Enspel	1,622	39	Paleogene	Chattian	Gunkel and Wappler (2015)
MK-3	2,428	41	Paleogene	Priabonian	Deng et al. (2020)
Luckenau Clay Complex	1,455	27	Paleogene	Bart.–Pria.	Müller et al. (2017)
Eckfeld maar	6,748	78	Paleogene	Lutetian	Wappler et al. (2012)
Messel maar	9,334	73	Paleogene	Ypres.–Lutet.	Wappler et al. (2012)
Republic	1,019	34	Paleogene	Ypresian	Labandeira (2002)
Laguna del Hunco	3,599	56	Paleogene	Ypresian	Wilf et al. (2005)
Wind River Interior	1,755	40	Paleogene	Ypresian	Currano et al. (2019)
Fifteenmile Creek	1,822	49	Paleogene	Ypresian	Currano et al. (2010)
South Fork of Elk Creek	1,008	53	Paleogene	Ypresian	Currano et al. (2008)
Dead Platypus	1,016	28	Paleogene	Thanetian	Currano et al. (2010)
Lur'd Leaves	1,364	28	Paleogene	Thanetian	Wilf et al. (2006)
Kevin's Jerky	1,423	25	Paleogene	Selandian	Wilf et al. (2006)
Palacio de los Loros 2	1,137	50	Paleogene	Danian	Donovan et al. (2018)
Castle Rock	2,668	25	Paleogene	Danian	Wilf et al. (2006)
Mexican Hat	2,220	34	Paleogene	Danian	Donovan et al. (2014)
Palacio de los Loros 1	1,089	41	Paleogene	Danian	Donovan et al. (2018)
Somebody's Garden	1,528	32	Cretaceous	Maastrichtian	Labandeira et al. (2002b)
Kühwiesenkopf	1,075	36	Triassic	Anisian	Labandeira et al. (2016)
Kayitou	1,043	23	Permian	Changhsingian	Liu et al. (2020)
Bletterbach ^{a,b}	1,193	14	Permian	Wuch.–Chan.	Bernardi et al. (2017)
Clouston Farm	8,714	22	Permian	Wuch.–Chan.	Prevec et al. (2009)
Laguna Polina	1,531	37	Permian	?	Cariglino (2018)
Wuda ^{a,b}	~10,000	21	Permian	Asselian	Feng et al. (2020)
Williamson Drive	1,830	39	Permian	Asselian	Xu et al. (2018)
Kinney	1,992	9	Pennsylvanian	Kasimovian	Donovan and Lucas (2021)

Table 2: The published assemblages with over 1,000 broadleaf specimens from sources other than [Labandeira et al. \(2018\)](#) that were analyzed in this study. ^aThese assemblages are not included in the sensitivity analysis presented in Figure S2 because damage type data are not available for each individual specimen. ^bThese assemblages do not have sufficient sample coverage to be rarefied to a coverage of 0.8.

Assemblage	Leaves	DTs	Period	Age	Citation
Bernasso	535	40	Quaternary	Gelasian	Adroit et al. (2018)
Berga	534	25	Neogene	Piacenzian	Adroit et al. (2018)
Palo Pintado	856	34	Neogene	Mess.–Zanc.	Robledo et al. (2018)
Hreðavatn-Staffholt	678	23	Neogene	Messinian	Wappler and Grímsson (2016)
Skarðsströnd-Mókollsdalur	524	21	Neogene	Tortonian	Wappler and Grímsson (2016)
San José	384	9	Neogene	Serravallian	Robledo et al. (2018)
Selárdalur-Botn	256	10	Neogene	Langhian	Wappler and Grímsson (2016)
Punta Basílica	209	26	Neogene	Aqui.–Lang.	Gandolfo and Zamalao (2021)
Güvem	624	45	Neogene	Burdigalian	Adroit et al. (2021)
Hindon Maar	466	78	Neogene	Burdigalian	Möller et al. (2017)
Quegstein	404	13	Paleogene	Rupel.–Chat.	Wappler (2010)
MK-1	599	20	Paleogene	Rupelian	Deng et al. (2020)
Renardodden	413	18	Paleogene	Pria.–Rupel.	Wappler and Denk (2011)
Bonanza	894	26	Paleogene	Lutetian	Wilf and Labandeira (1999)
Wind River Edge	908	33	Paleogene	Ypresian	Currano et al. (2019)
PN	693	28	Paleogene	Ypresian	Currano et al. (2010)
Sourdough	792	23	Paleogene	Ypresian	Wilf et al. (2001)
Cool Period	491	25	Paleogene	Ypresian	Currano et al. (2010)
Level E	336	19	Paleogene	Ypresian	Azevedo Schmidt et al. (2019)
Hubble Bubble	994	39	Paleogene	Than./Ypres.	Currano et al. (2008)
Daiye Spa	843	33	Paleogene	Thanetian	Currano et al. (2008)
Clarkforkian	749	27	Paleogene	Thanetian	Wilf et al. (2001)
Cerrejón	507	28	Paleogene	Thanetian	Wing et al. (2009)
Level C	311	21	Paleogene	Thanetian	Azevedo Schmidt et al. (2019)
Skeleton Coast	840	21	Paleogene	Thanetian	Wilf et al. (2006)
Bogotá	955	90	Paleogene	Sela.–Than.	Giraldo et al. (2021)
Kolfjellet	357	18	Paleogene	Sela.–Than.	Wappler and Denk (2011)
Haz-Mat	757	18	Paleogene	Selandian	Wilf et al. (2006)
<i>Persites</i> Paradise	963	22	Paleogene	Selandian	Wilf et al. (2006)
Menat	938	39	Paleogene	Selandian	Wappler et al. (2009)
Las Flores	568	42	Paleogene	Danian	Donovan et al. (2018)
Pyramid Butte	655	17	Paleogene	Danian	Labandeira et al. (2002b)
Battleship	461	31	Cretaceous	Maastrichtian	Labandeira et al. (2002b)
Dean Street	709	32	Cretaceous	Maastrichtian	Labandeira et al. (2002b)
Lefipán East	607	45	Cretaceous	Maastrichtian	Donovan et al. (2018)
Luten's 4H Hadrosaur	426	26	Cretaceous	Maastrichtian	Labandeira et al. (2002b)
Camarena ^a	428	11	Jurassic	Aalenian	Santos et al. (2021)
Monte Agnello	646	19	Triassic	Ladinian	Labandeira et al. (2016)
Dos Hermanos ^a	359	16	Permian	Wuch.–Chan.	Cariglino (2018)
South Ash Pasture ^a	505	22	Permian	Road.–Capit.	Maccracken and Labandeira (2020)
Doña Ana Mts.	215	4	Permian	Artinskian	DiMichele et al. (2018)
Colwell Creek Pond	991	46	Permian	Artinskian	Schachat et al. (2014)
Laguna Lillo ^a	232	7	Permian	Assel.–Kung.	Cariglino (2018)
Beeman ^a	392	7	Pennsylvanian	Kasim.–Gzhel.	Lucas et al. (2021)

Table 3: The assemblages analyzed in this study with 200–1,000 broadleaf specimens. ^aThese assemblages do not have sufficient sample coverage to be rarefied to a coverage of 0.8.

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9 SUPPLEMENTAL MATERIAL

9.1 DIFFICULTIES IN ESTIMATING DAMAGE TYPE DIVERSITY

The differences in sampling intensity among Colwell Creek Pond and Mitchell Creek Flats pose an insurmountable obstacle to estimating true damage type diversity with raw or rarefied data. Rarefaction curves that include the dominant broadleaf plant taxa from both assemblages show that the estimate of higher damage type diversity for Colwell Creek Pond cannot be disentangled from the disparities in the amount of surface area examined (Figure S1A). If the damage type diversities tallied at each assemblage, 45 at Colwell Creek Pond and 19 at Mitchell Creek Flats, are interpreted at face value, the resulting conclusion that former contains greater damage type diversity is based entirely on an artifact of uneven sampling. However, even when the role of sampling completeness is taken into account, the overlapping confidence intervals complicate estimates of whether greater sampling would reveal significant differences in damage type diversity among the two assemblages (Knezevic, 2008).

The unreliability of diversity estimates derived from incompletely sampled assemblages is further underscored by the variability of estimated damage type diversity at Colwell Creek Pond when rarefaction curves are calculated with leaves from this locality that are randomly sampled only to the amount of surface area seen at Mitchell Creek Flats (Figure S1B–D). In other words, the difficulty of extrapolating damage type diversity at Mitchell Creek Flats for the amount of leaf area sampled from Colwell Creek Pond is due not only to the uncertainty surrounding damage type diversity beyond the amount of leaf area available, but also to the uncertainty surrounding the reliability of the damage type diversity estimated for Mitchell Creek Flats at the amount of leaf area that has been examined for this assemblage. Comparisons of damage type diversity, therefore, require a method such as rarefaction to control for differences in sampling completeness, and require fairly complete sampling of all localities—which can be difficult to achieve due to both the availability of fossil material and the investigator effort required.

The difficulty of estimating damage type diversity in light of sampling incompleteness is further compounded by the unavailability of surface area data for nearly all assemblages examined thus far (Schachat et al., 2018). Because leaf size varies so widely, two assemblages can have equivalent numbers of damage types per 400 cm² of leaf surface area while having widely different numbers of damage types per 400 leaves. Variable leaf surface area can complicate, if not invalidate, attempts to use damage type data to discern macroevolutionary and macroecological patterns. For example, the amount of surface area per leaf available to insect herbivores has changed through time because gymnosperms and angiosperms have different average leaf sizes. As another example, whereas heightened levels of insect herbivory in the tropics

1008 form the basis of various biogeographical theories (Dobzhansky, 1950; MacArthur, 1969; Janzen, 1970;
1009 Connell, 1971), the magnitude of the latitudinal variability in herbivory is not entirely understood and
1010 depends considerably on the metric used (Anstett et al., 2014; Andrew and Hughes, 2005; Adams et al.,
1011 2011, 2009; Zhang et al., 2011; Adams and Zhang, 2009; Adams et al., 2010; Salazar and Marquis, 2012;
1012 Moreira et al., 2015; Moles et al., 2011). And because plants in wet biomes tend to have larger leaves
1013 (Ackerly et al., 2002; Cunningham et al., 1999; Givnish, 1987), rarefaction curves that are scaled by
1014 number of leaves rather than amount of leaf surface area may confound increased damage type diversity in
1015 wet tropical forests with increased leaf size in these habitats.

1016 In addition to subsampling techniques such as rarefaction, a number of population estimators hold the
1017 potential to calculate true (asymptotic) damage type diversity and corresponding confidence intervals from
1018 datasets of varying levels of completeness (Chao, 1987; Chiu et al., 2014; O'hara, 2005; Palmer, 1990; Smith
1019 and van Belle, 1984). Of these, the Chao1 estimator is particularly well-suited to sparse datasets (Chao,
1020 1989). However, these estimators follow the same pattern seen in rarefaction curves of steeply increasing
1021 estimates of diversity at low levels of sampling completeness (Chao et al., 2009). One might hope that the
1022 Chao1 estimator would behave like estimates of the herbivory index, with confidence intervals that contain
1023 the true value even when sampling is very incomplete and then narrowing more closely around the true value
1024 as sampling becomes more complete. However, this often is not the case. An analysis of plant assemblages
1025 for which at least 1,000 broadleaf specimens have been examined for insect herbivory (Tables 1, 2) shows
1026 that confidence intervals are deceptively narrow when sampling is insufficiently complete to capture the true
1027 value of damage type diversity (Figure S2). As sampling increases, the confidence intervals are more likely
1028 to contain an accurate result but often widen so much that false negative results become inevitable (Figure
1029 S2).

1030 Whereas neither rarefaction nor the Chao1 estimator provide sufficient accuracy and precision to estimate
1031 damage type diversity, they have the potential to work far better in concert. The combination of rarefaction
1032 and Hill numbers (Chao et al., 2014), a family of metrics to which the Chao1 estimator belongs, was an
1033 extension of advances in the extrapolation of rarefaction curves (Colwell et al., 2012) and in rarefaction based
1034 on sampling completeness rather than sample size (Chao and Jost, 2012). The Chao1 estimator return an
1035 identical point estimates and confidence intervals for asymptotic diversity for a given frequency of damage
1036 type occurrences regardless of whether those damage type occurrences are spread out over 10 or 100,000
1037 leaves. But, unlike the Chao1 estimator, rarefaction has the advantage of taking sampling completeness into
1038 account while generating diversity estimates. Unlike rarefaction, the Chao1 estimator has the advantage
1039 of robustness to differences in leaf size and fragmentation because this metric estimates true, asymptotic
1040 diversity rather than diversity at a particular level of incomplete sampling. The method of Chao et al. (2014)

1041 possesses both of these advantages and is implemented in the R package `iNEXT` (Hsieh et al., 2016).

1042 However, the method of Chao et al. (2014) yields estimates of asymptotic damage type diversity that are
1043 biased by the number of leaves sampled. This is probably because of the extreme sparsity of damage type
1044 occurrence datasets combined with the large quantity of damage types that are only observed once. The
1045 sensitivity of this method to sampling completeness can be seen when comparing the rarefaction curves for
1046 assemblages with over 7,000 leaves examined to extrapolated rarefaction curves generated after subsampling
1047 each assemblage down to 1,000 or 2,000 leaves (Figure S3). 2,000 leaves are rarely sufficient to extrapolate
1048 the complete rarefaction curve, and 1,000 leaves are sufficient for only one assemblage, Willershausen (Adroit
1049 et al., 2018). With 1,000 leaves, these extrapolated rarefaction curves level off far too quickly—and even for
1050 Willershausen, the one assemblage that does not follow this pattern, the 95% confidence interval is much too
1051 wide to permit the detection of significant differences in damage type diversity among most assemblages.

1052 9.2 THE UNRELIABILITY OF EXTRAPOLATED DIVERSITY ESTIMATES

1053 GENERATED WITH COVERAGE-BASED RAREFACTION

1054 To evaluate whether rarefied damage type diversity at a sample coverage of 0.8 can be extrapolated from
1055 datasets that do not reach this level of coverage, we used an iterative subsampling procedure. Leaves were
1056 subsampled from the pre-angiosperm assemblage with the largest number of leaves examined, Aasvoëlberg
1057 411 (Labandeira et al., 2018), and the angiosperm assemblage with the largest number of leaves examined,
1058 Willershausen (Adroit et al., 2018). This procedure was repeated 3,000 times for each assemblage to ensure
1059 that a wide range of levels of sample coverage are represented among the subsampled datasets. For each
1060 subsampled dataset, we extrapolated rarefied damage type diversity to a sample coverage of 0.8. The results
1061 of this procedure (Figure S4) show that the accuracy of extrapolated estimates of damage type diversity
1062 does increase with sample coverage, but these extrapolated estimates are never accurate enough to inspire
1063 any confidence. Therefore, we do not recommend extrapolating estimates of damage type diversity for
1064 assemblages with a sample coverage below 0.79.

1065 9.3 EVALUATION OF OTHER METRICS

1066 The ecological literature contains a variety of complex metrics for the evaluation of trophic interactions.
1067 With paleontological data, and especially paleontological data from a clade with a history of biased collecting
1068 (Gunkel and Wappler, 2015), a fundamental question is whether the data at hand are sufficient to yield robust
1069 results when analyzed with more complex techniques.

1070 Both plants and their damage types follow a dominance-diversity distribution that approximates a

1071 lognormal or gamma distribution (Figure S5), as do nearly all other biotic communities (Diserud and
1072 Engen, 2000). In a typical floral assemblage, the majority of plant hosts and damage types are rare.

1073 Generating a standardized estimate of floral or damage type diversity at a given assemblage through
1074 rarefaction is a relatively simple matter of subsampling the more abundant taxa in a consistent fashion.
1075 Estimating relationships among plant hosts and damage types, however—for example, estimating the
1076 proportion of damage types on the second-most abundant plant host which also occur on the most
1077 abundant plant host—is a far more complex matter, and thus is far more sensitive to sample size. A single
1078 tree can grow thousands of leaves in a single year (White, 1993), but it is uncommon for thousands of
1079 leaves from a single fossil assemblage to be examined for insect herbivory (Tables 1 and 2). We are not sure
1080 if all fossil leaf material ever examined for the herbivory index contains more or less leaf surface area than a
1081 single elm tree.

1082 Therefore, whereas the amount of sampling typically seen in studies of fossil herbivory is likely sufficient to
1083 determine whether the most common damage type at an assemblage occurs on the most common plant host
1084 at the assemblage, this amount of sampling is probably not sufficient to determine whether the twentieth-
1085 most-common damage type occurs on the twentieth-most-common plant host. The absence of the twentieth-
1086 most-common damage type on the twentieth-most-common plant host may indicate that this particular
1087 interaction did not occur in the community represented by the fossil assemblage, but can just as easily
1088 indicate that sampling is not sufficient to document this interaction.

1089 To evaluate the reliability of complex metrics of insect herbivory calculated for fossil assemblages, we
1090 conducted two iterative subsampling routines with data from the pre-angiosperm assemblage with the largest
1091 number of leaves examined, Aasvoëlberg 411 (Labandeira et al., 2018), and the angiosperm assemblage with
1092 the largest number of leaves examined, Willershausen (Adroit et al., 2018). In both routines, we subsampled
1093 1,000 leaves per iteration and iterated this procedure 1,000 times. For the first routine, we calculated the
1094 proportion of damage types occurring on the second-most-abundant plant host that also occur on the most
1095 abundant plant host. (This is perhaps the simplest metric of the nestedness of damage type communities
1096 among the plant hosts at an assemblage.) For the second routine, we divided the number of leaves belonging
1097 to the second-most-abundant plant host by the number of leaves belonging to the most abundant plant host.
1098 (The simplicity of this metric is emblematic of the metrics commonly used in the study of fossil herbivory.)

1099 The results of these subsampling routines indicate that, whereas the dominance-diversity structure of
1100 the plant hosts within an assemblage is a valid and reliable metric (Figure S6B), the nestedness of damage
1101 type communities among individual plant hosts is not (Figure S6A). No sensitivity analyses are needed
1102 to demonstrate the impossibility of determining with any certainty whether the twentieth-most-common
1103 damage type at an assemblage is truly absent from the twentieth-most-common plant host—as opposed to

1104 evading detection due to the number of leaves examined. One could argue that this sort of failure of detection
1105 will be shared across all assemblages and therefore will not bias comparisons among assemblages. However,
1106 our sensitivity analysis demonstrates that even the very simplest measurement of damage type nestedness
1107 (the degree of nestedness among the two most abundant plant hosts) is hopelessly unreliable with 1,000
1108 leaves.

1109 Some analytical trends in paleobiology follow a boom–bust cycle in which a technique increases in
1110 popularity, perhaps due to positive connotations associated with its complexity, only to fall out of favor
1111 when its reliability, validity, and interpretability come into question (Smith et al., 1997). Our results
1112 indicate that the complexity of techniques that associate particular damage types with particular host
1113 plants, impressive as they may seem, require far more complete sampling than can be expected from
1114 studies of insect herbivory on fossil leaves.

1115 These results also have implications for quantifying host specificity. At present, it is customary to
1116 rank each damage type within an assemblage by its host specificity on a discrete scale of generalist (1),
1117 intermediate specificity (2), or specialized (3). To avoid misinterpreting artifacts of incomplete sampling as
1118 biological phenomena, these rankings are only assigned to damage types that occur on three or more plant
1119 specimens. However, the findings presented in this section raise the question of whether three occurrences
1120 of a damage type are sufficient to determine its host specificity, particularly if a damage type is classified as
1121 specialized because all three of its occurrences are on the most abundant plant host at the assemblage.

1122 9.4 RARE DAMAGE TYPES AND A COMPARISON OF ANGIOSPERM- AND 1123 NON-ANGIOSPERM-DOMINATED ASSEMBLAGES

1124 With size-based rarefaction, one could argue that higher damage type diversities in angiosperm-dominated
1125 floras might become apparent at higher sample sizes. An analogous argument for coverage-based rarefaction
1126 is that higher damage type diversities in angiosperm-dominated floras might become apparent at higher
1127 levels of sample coverage. To evaluate this possibility, we repeated all coverage-based rarefaction analyses
1128 by rarefying to a sample coverage of 0.9 instead of 0.8. This higher level of sample coverage incorporates
1129 a greater number of rare damage types in diversity estimates, thus constituting a sensitivity analysis of
1130 whether the similar damage type diversities estimated for Permian and Cenozoic assemblages in Figure 5
1131 are attributable to the level of sample coverage to which the assemblages were rarefied.

1132 For each assemblage we calculated the difference in damage type diversity when rarefied to a sampling
1133 coverage of 0.8 or 0.9 (Figure S7). We found that the differences are smallest for the Molteno assemblages
1134 listed in Table 1, which contain relatively low damage type diversities. Although far more

1135 angiosperm-dominated assemblages have been evaluated for insect herbivory than
1136 non-angiosperm-dominated assemblages outside of the Molteno Formation, the extent to which estimated
1137 damage type diversity differs with sample coverage is similar among these two categories. Of note, when
1138 sample coverage increases from 0.8 to 0.9, the four assemblages with the greatest increase in estimated
1139 damage type diversity are the Hindon Maar, Eckfeld, Messel, and Bogotá assemblages, all of which are
1140 angiosperm-dominated and contain high damage type diversities. To discern whether this great increase in
1141 damage type diversity with heightened sample coverage is truly unique to angiosperm-dominated
1142 assemblages, far more non-angiosperm-dominated assemblages will need to be evaluated for insect
1143 herbivory.

1144 Insect herbivory from these four assemblages was described within the last ten years (Wappler et al.,
1145 2012; Möller et al., 2017; Giraldo et al., 2021). Estimates of damage type diversity may be biased toward
1146 assemblages that have been described most recently under a scenario in which existing damage types are have
1147 increased in number (Dos Santos et al., 2020; Xiao et al., 2021). However, Giraldo et al. (2021) found that
1148 this is not the case: a more conservative approach toward splitting damage types does not yield a noticeable
1149 decrease in damage type diversity as standardized through rarefaction.

1150 9.5 THE NUMBER OF ITERATIONS NEEDED FOR RESAMPLING PROCEDURES

1151 To determine the number of iterations needed for the resampling procedures discussed in “Evaluating other
1152 potential dimensions of an ecospace for herbivory,” we iteratively performed our suggested resampling
1153 routine for quantifying uncertainty surrounding the offset between the prevalence of each plant host and
1154 the prevalence of insect damage on it. For this procedure we used data from the Colwell Creek Pond
1155 assemblage (Schachat et al., 2014) because surface area measurements are available. We found that the
1156 width of the 84% confidence interval varies minimally whether 100 or 100,000 iterations of the resampling
1157 routine are performed, but stabilizes around 5,000 iterations (Figure S8). Therefore, we recommend that
1158 future studies use 5,000 resampling iterations to generate confidence intervals.

1159 9.6 CRITERIA FOR INCLUSION OF LEAVES AND DAMAGE TYPES

1160 As noted in the main text, C.C. Labandeira employs a wide definition of “foliage” that includes needles,
1161 liverworts, phyllids, photosynthetic wings of seeds, and even flattened horsetail axes whereas S.R. Schachat
1162 employs a narrower definition restricted to multi-veined broad leaves and leaves with a defined midvein.
1163 (Workers who study angiosperm assemblages typically examine leaves that are at least 50% complete. No
1164 specimens were removed from the raw datasets for any of the angiosperm-dominated assemblages analyzed

1165 here.) When deciding which specimens from published datasets (Prevec et al., 2009; Cariglino, 2018;
1166 Labandeira et al., 2018; Liu et al., 2020; Bernardi et al., 2017) to include in the analyses presented here, we
1167 employed a compromise definition that excludes needles, liverworts, phyllids, photosynthetic wings of seeds,
1168 and flattened horsetail axes but includes scale leaves.

1169 The raw data for the Williamson Drive assemblage (Xu et al., 2018) preclude determinations of whether
1170 many of the individual specimens represent, or at least contain, broadleaf foliage. Therefore, the plant
1171 hosts from this assemblage included here are the five foliage types that were included in NMDS plot in the
1172 original publication: *Sigillariophyllum* leaves, *Pseudomariopteris cordato-ovata*, *Annularia carinata*, *Lilpopia*
1173 *raciborskii*, and *Macroneuropteris scheuchzeri*.

1174 The data for the Wuda flora (Feng et al., 2020) do not include assignments of damage types. Therefore,
1175 damage types were tallied based on the descriptions in the text of the article. An exact count of broadleaf
1176 specimens was unavailable due to the vast number of specimens examined. Because of the sparsity of insect
1177 damage at Wuda, the amount of sample coverage for this entire dataset is 0.675—well short of the threshold
1178 used here, 0.8. Our decisions about the Wuda plant taxa, therefore, are inconsequential, as this assemblage
1179 cannot be included in Figure 5.

1180 The description of herbivory at the Lincang assemblage (Zhang et al., 2018) includes a number of
1181 unrecognizable damage types. Because the abstract of Zhang et al. (2018) contains a count of damage
1182 types that is restricted to those with numbers assigned in the Damage Guide (Labandeira et al., 2007), we
1183 followed the authors' lead and included only those damage types in our analysis.

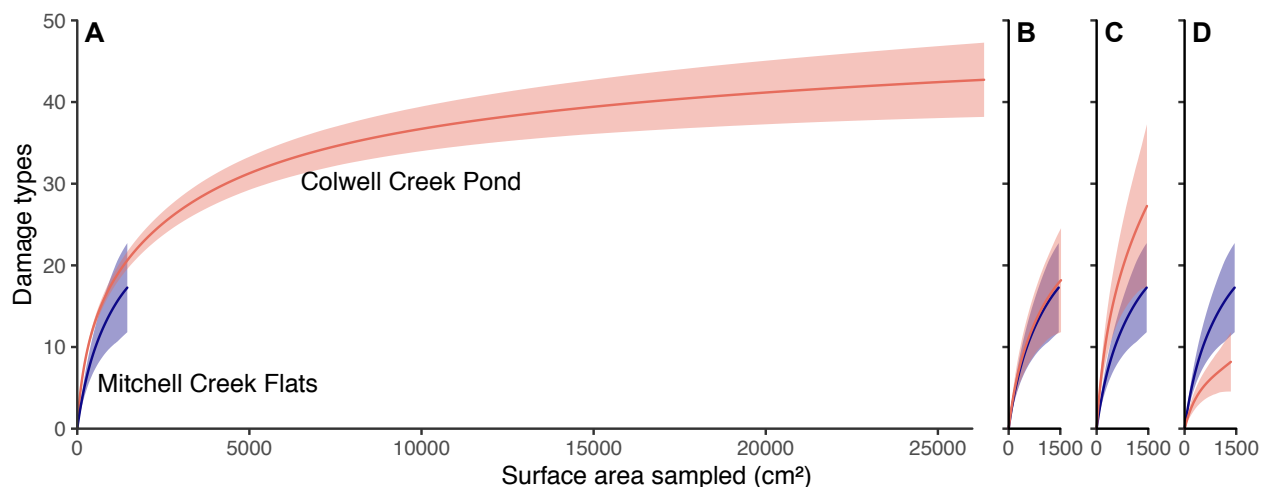


Figure S1: Rarefaction curves for two very similar assemblages: Colwell Creek Pond and Mitchell Creek Flats. **A** shows the rarefaction curves for all data from both assemblages. In **B** through **D**, the rarefaction curves for Colwell Creek Pond are calculated from a randomly subsampled set of leaves with nearly equal surface area to that measured at Mitchell Creek Flats.

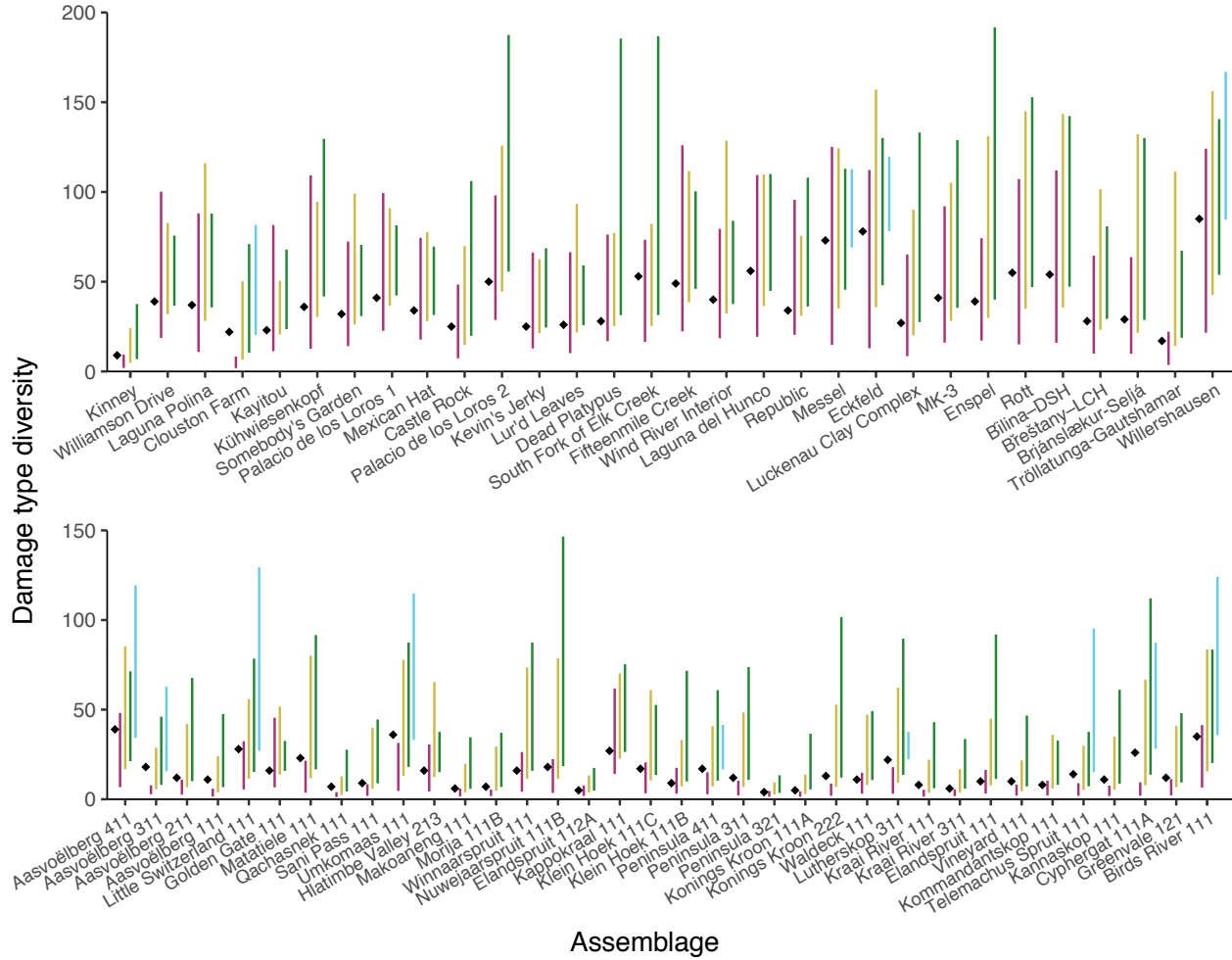


Figure S2: Precision and accuracy of the Chao1 estimator, examined by subsampling all assemblages listed in Tables 1 and 2. The black diamond denotes the raw damage type diversity observed at each assemblage and the lines denote the mean 95% confidence interval of the Chao1 estimator when used on a randomly sampled subset of leaves from the assemblage, as follows: magenta, 100 leaves; yellow, 500 leaves; green, 1,000 leaves; blue, 5,000 leaves. Not all assemblages contain sufficient material to subsample to 5,000 leaves. At low levels of sampling, the Chao1 estimator clearly underestimates damage type diversity. At higher levels of sampling, the Chao1 estimator becomes more accurate but far less precise.

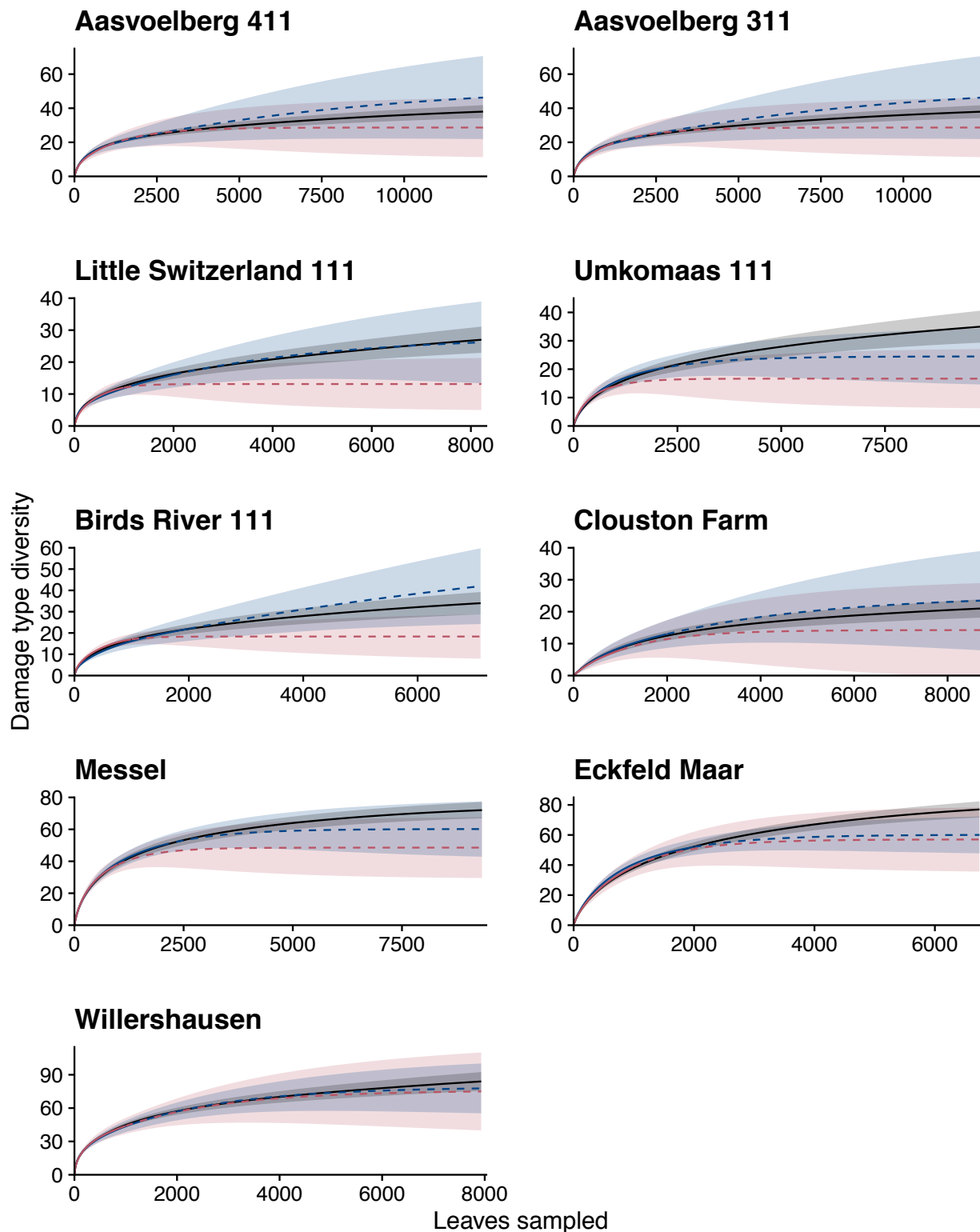


Figure S3: Precision and accuracy of estimates of damage type diversity generated with the method of Chao et al. (2014). Interpolated rarefaction curves are in solid lines and extrapolated curves are in dashed lines. The black lines represent the interpolated rarefaction curve for the raw dataset, the blue lines represent the raw dataset subsampled down to 2,000 leaves and then extrapolated to the number of leaves present in the raw dataset, and the red lines represent the raw dataset subsampled down to 1,000 leaves and then extrapolated to the number of leaves present in the raw dataset.

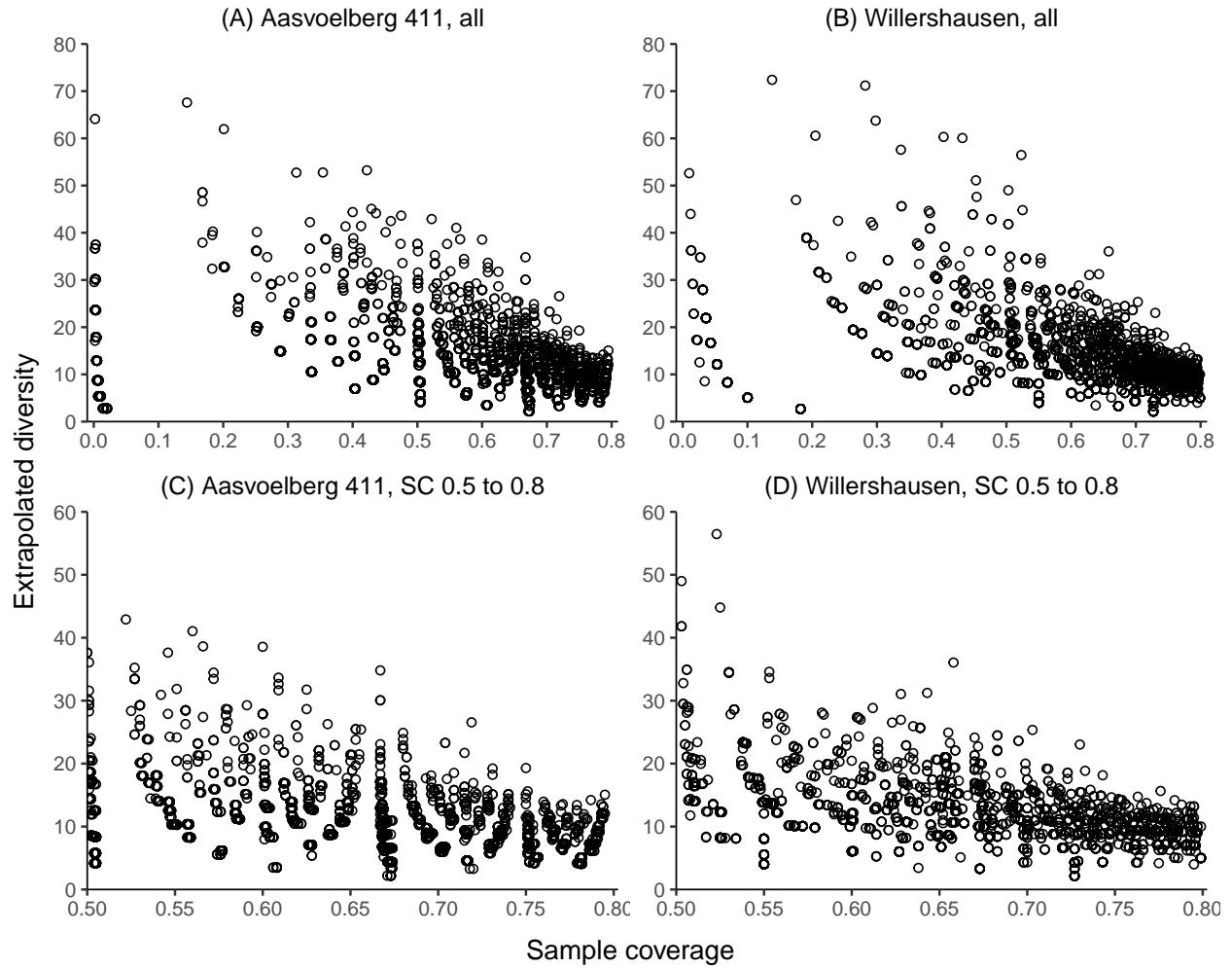


Figure S4: The variability of estimates of damage type diversity at a sample coverage of 0.8 when extrapolated from 1,000 subsampled leaves that yield a sample coverage below 0.8.

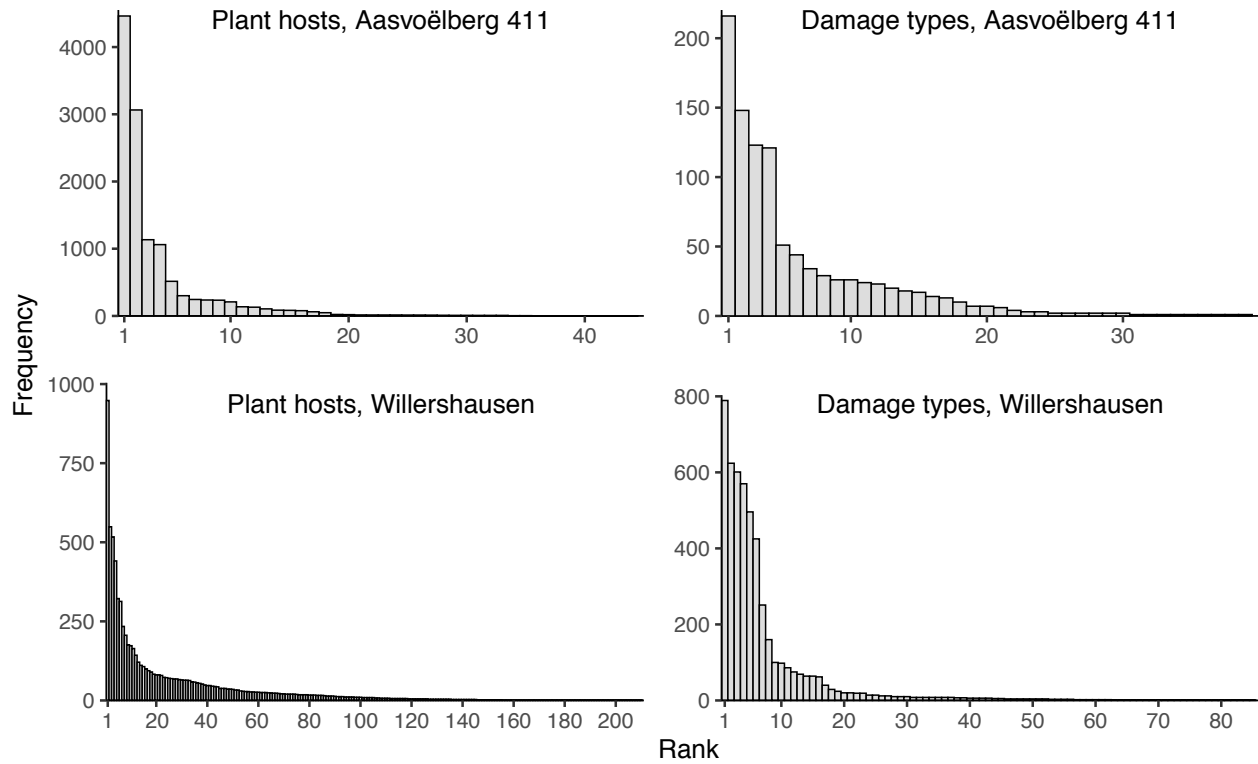


Figure S5: Dominance-diversity distributions for plant hosts and damage types at the pre-angiosperm assemblage with the largest number of leaves examined, Aasvoëlberg 411 (Labandeira et al., 2018), and the angiosperm assemblage with the largest number of leaves examined, Willershausen (Adroit et al., 2018).

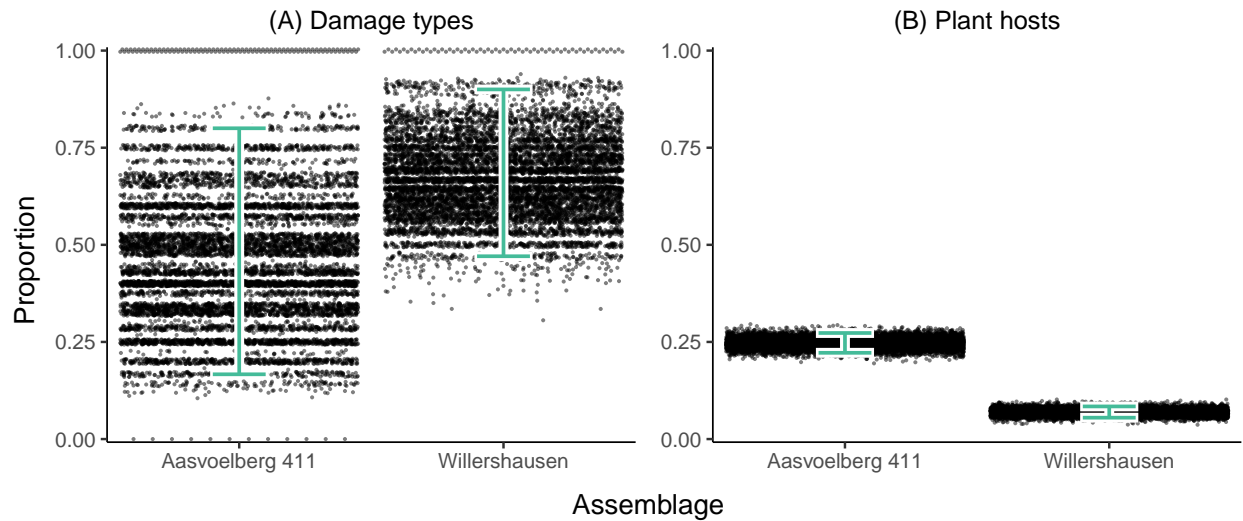


Figure S6: Measures of community structure at Aasvoëlberg 411 and Willershausen, the two assemblages featured in Figure S5, calculated by subsampling each assemblage to 1,000 leaves. Panel (a) shows the proportion of damage types on the second-most-abundant plant host that also occur on the most abundant plant host. Panel (b) shows the number of leaves belonging to the second-most-abundant plant host divided by the number of leaves belonging to the most abundant plant host. The lines denote 95% confidence intervals.

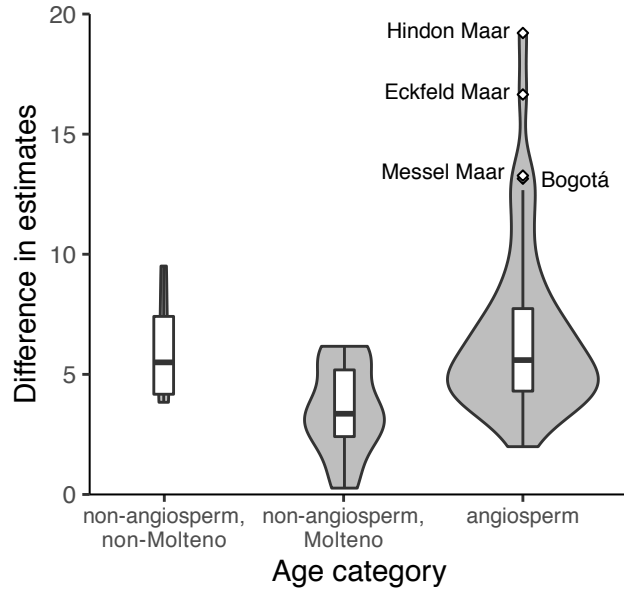


Figure S7: The difference in estimated damage type diversity when the sample coverage used in rarefaction increases from 0.8 to 0.9. The width of each violin represents the number of assemblages it contains. Boxplots are overlain atop each violin.

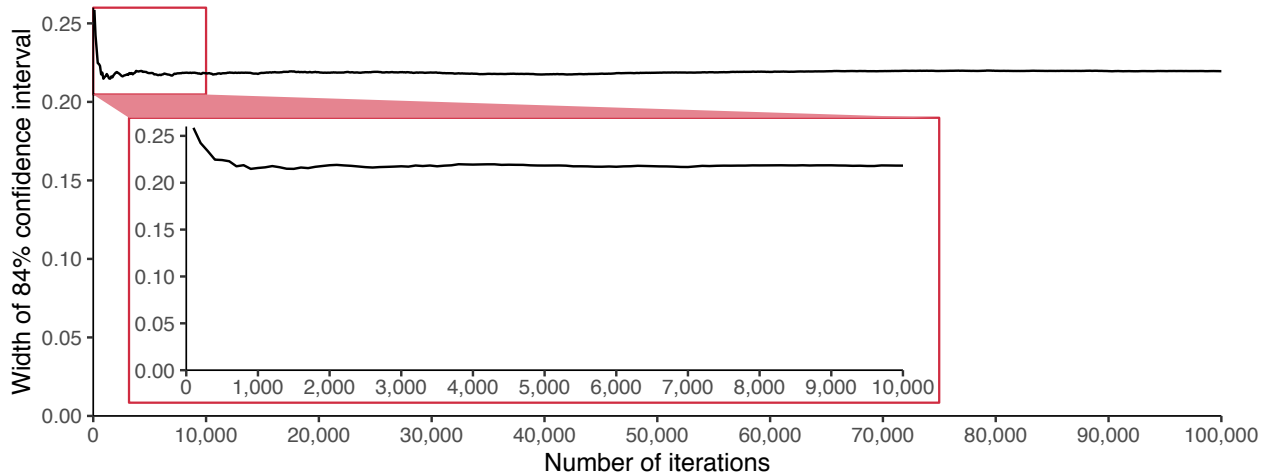


Figure S8: Changes in the width of the 84% confidence interval as the resampling procedure is further iterated.