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9 **LLM-augmented taxonomy for >4500 palaeopalynology genera**

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19 **Abstract**

20 Large Language Models (LLMs), being text-based, are ideal types of artificial intelligence to
21 consider the complexities of palaeontological taxonomy because palaeontology depends on
22 published textual descriptions as the primary, authoritative record of a taxon. This paper
23 describes (1) the preparation of palynological (the study of organic-walled microfossils)
24 taxonomic text contained within the >4500 genera of the Jansonius and Hills
25 palaeopalynological catalogue (JHC) for an LLM-augmented taxonomy system (LATS), (2) the
26 efficiency and accuracy of the LATS, and (3) examples of possible further uses of the LATS
27 beyond aids to identification. The conversion of the JHC into a LATS is typical of the challenges
28 of making so called ‘long tail’ data suitable for AI development and can involve considerable
29 manual checking. Principles of development include (1) ‘inclusion’, that is, making sure that the
30 LATS as far as possible includes rather than excludes candidate genera; (2) the principle of
31 ‘assistance’ rather than supplantation so that the LATS is intended as an aid to taxonomy, not a
32 replacement for a human taxonomist; and (3) the principle of ‘non-intervention’ whereby no
33 alterations to original authoritative genus descriptions or diagnoses are applied. Training for the

34 dataset involved 500 Question/Answer pairs generated for the JHC by specialists, as well as
35 additional synthetic QA pairs which, combined, were used to supervised-fine tune the LLM.

36 The LATS functions through Retrieval Augmented Generation and returns candidate genera with
37 statistical measures of match against the prompt(s). Access to full descriptions of genera
38 extracted from the JHC and to scans of the original catalogue cards allow the taxonomist to use
39 their own judgment in final identification. The LATS produces generally good results but there
40 are two types of limitations or shortcomings: those that emanate from the JHC (and
41 palaeopalynological taxonomy itself), and those that emanate from the working of the LATS.
42 Limitations due to the JHC include (1) poor potential for discrimination between some genera
43 because of poor original descriptions which have not been subsequently emended, (2)
44 numerous candidate genus names that may be synonyms, and (3) invalid candidate genera, i.e.
45 that were illegitimately published. Limitations that emanate from the working of the LATS
46 include evidence of bias against finely described genera.

47 As well as providing a LATS to aid a palynologist through the stages of identification, the
48 information and ‘understanding’ that the underlying system has of a large area of
49 palaeopalynological taxonomy means that the system could be put to more general uses for
50 example in the identification of genus names that are likely synonyms, and investigation of the
51 distribution of genera (or taxa) in ‘morphological space’.

52 The development of the LATS described here has implications for other palaeontological groups
53 in terms of the text basis of their taxonomy (for example variable quality of descriptions and
54 inconsistency in terminology), and their suitability for development of other LLM-assisted
55 taxonomic aids.

56 Introduction

57 Palaeontological and palaeopalynological taxonomy aims to classify fossil organisms, provide
58 stable names for unambiguous communication, and organize diversity by placing fossils into

59 hierarchical groups (species, genera, families, etc.) that reflect evolutionary relationships. The
60 determination of a specimen to a taxon requires reference to literature, type specimens,
61 illustrations and photographs.

62 Early taxonomic keys (e.g. Lamarck 1778) were used as aids to determination and used
63 dichotomous processes that allow the user to choose between opposing pairs of morphological
64 characters; and keys continue to be used for example in botany and zoology (e.g. ZoologyTalks:
65 <https://www.zoologytalks.com/types-of-taxonomic-keys/>). Text-based palynological keys have
66 been published in archaeological studies (e.g. Hubbard 1992), and in palaeopalynology (e.g.
67 Steemans and Wellman 2018), and several modern pollen identification keys exist: for example,
68 Das et al. (2025) constructed a taxonomic key for Papilionoideae (Fabaceae) from Northeast
69 India. The use of artificial intelligence in identification keys is rare (Yu et al. 2024), an exception
70 being Liu et al. (1994) who constructed a dual-step identification tool with a Knowledge-Based
71 System (KBS) and an image analysis subsystem, which extracted graphic information from
72 foraminifera chambers and suture images in addition to previous shape analysis based on
73 Fourier analysis and edge detection.

74 Various forms of artificial intelligence have also been used to aid palynomorph identification
75 using image recognition (e.g. Zhang et al. 2004, Rodriguez-Damian et al. 2006, Mander et al.
76 2013, Kong et al. 2016, Niu et al. 2024, von Allmen et al. 2024) based on learning from very large
77 numbers of high-resolution images of specimens whose taxonomic status were validated by
78 experts, though not by direct reference to authoritative text-based diagnoses or descriptions of
79 taxa. Thus, the expert validations are placed between the user and the first-principle taxonomic
80 information. Though they may supply accurate determinations within that framework, their
81 taxonomic decisions are not visible to the user, and their methods do not offer pedagogic value
82 or insight into processes of taxonomic decision-making.

83 Large Language Models (LLMs), being text-based, are ideal types of artificial intelligence to
84 consider the complexities of palaeontological taxonomy because palaeontology depends on
85 published textual descriptions as the primary, authoritative record of a taxon (ICZN 1999;
86 Turland et al. 2025). An LLM-augmented taxonomic key was developed for a small dataset of
87 around 70 spore species from the Carboniferous-Permian of the Arabian Plate (Stephenson et
88 al. 2025) as a pilot project. This allowed users to make choices amongst a fixed set of options
89 and then allowed free text prompts to refine the search. Because of the small size and limited
90 applicability of the system, it was not released publicly.

91 Considering these developments, we investigated the possibility of constructing an LLM based
92 system that would respond to simple descriptive prompts rather than requiring a series of
93 dichotomous or polytomous choices. LLMs ‘learn’ from extensive text datasets, including
94 books, articles, and websites which helps them recognize patterns in language, grammar,
95 meaning, and context (e.g. Patil & Gudivada 2024). Transformers enable models to weight
96 different words and phrases based on their relationships within a sentence. Text is broken down
97 into smaller units (tokens), which the model processes to predict the most likely next words in a
98 sentence or generate full responses. LLMs generate text by predicting the most statistically
99 probable sequence of words based on their training (Bender and Koller 2020). When applied to
100 palynological identification scenarios, LLMs are also expected to bring several distinct
101 advantages:

- 102 1. Unlike many areas of modern biology that can rely on genetics, living specimens, or
103 high-resolution imaging, palaeontology depends on published textual descriptions as
104 the primary, authoritative record of a taxon (e.g. Gravendyck et al. 2021). Under the rules
105 of zoological and botanical nomenclature (ICZN 1999; Turland et al. 2025), a species
106 becomes valid only when it is described in text. Text provides the diagnosis (what makes
107 the taxon unique), the description (its full morphological characterisation), the type

108 designation (holotype, lectotype, etc.), and the comparisons with similar taxa. All these
109 elements must be written, published, and permanent. Even high-quality plates or
110 photographs cannot capture subtle morphological distinctions and variability, describe
111 internal structures not visible externally, or explain reasoning behind taxonomic
112 decisions. In this sense, text provides the conceptual framework that the images
113 support. Text allows reproducibility and comparison because future researchers must
114 be able to compare new specimens to the original description, evaluate whether a fossil
115 belongs to an existing species, and identify synonyms. Large Language Models (LLMs),
116 being text-based, are therefore ideal types of artificial intelligence to consider the
117 complexities of palaeontological taxonomy because palaeontology depends on
118 published textual descriptions as the primary, authoritative record of a taxon.

119 2. Human-like interaction and interpretability: unlike image-based identification apps for
120 botany (for example those on an iPhone such as PictureThis®) which function without
121 human interaction in decision making (i.e. they are black-boxes), LLMs can show
122 reasoning and generate human-readable explanations and responses, enabling more
123 interpretation of classification decisions. It also allows the interpretation of
124 morphological details that may not be discerned from a photograph. This capability
125 supports greater transparency and user trust in AI-assisted identification systems.

126 3. LLMs exhibit strong language comprehension abilities, allowing them to capture subtle
127 semantic distinctions (e.g. Rogers et al. 2021). This is particularly beneficial in
128 identification tasks where subtle differentiation between similar taxa is required,
129 thereby addressing persistent challenges such as ambiguous terminology, short or
130 inadequate descriptions, or continuous variability between taxa.

131 4. LLMs possess broad world knowledge, including foundational understanding in many
132 scientific domains allowing extraneous knowledge to be brought into taxonomic
133 decision making, for example translation of Latin-based terminology.

134 The purpose of this paper is to describe the preparation of palynological taxonomic text for an
135 LLM-augmented taxonomy system (LATS) of more than 4500 genera, the efficiency and
136 accuracy of the LATS, and potential further uses of the LATS beyond identification. The LATS is
137 intended for open and free use for all scientists and provides a guide to determination rather
138 than acting as an ultimate arbiter of genus determination.

139 Preparation of the data

140 Background

141 Much data generated by geoscientists is still not accessible to other geoscientists or to artificial
142 intelligence tools (Stephenson et al. 2020). This is a problem for geoscience and other
143 observational sciences which often rely on mixed qualitative and quantitative data that is part of
144 the so-called ‘long tail’ (e.g. Sinha et al. 2013) – the unstructured and heterogeneous datasets of
145 geological surveys, university research groups, and individual scientists. Palynological
146 taxonomic datasets are often similar: they may have been collected and compiled from the
147 peer-reviewed literature, developed separately from the peer-reviewed literature, or exist only as
148 separate records (diagnoses/descriptions) in the peer-reviewed literature, often behind a
149 paywall. This often hinders palynological taxonomic study and determination of species and
150 genera, because of the size of datasets, their heterogeneity in terms of language, terminology,
151 precision and comprehensibility - and in the case of paywalled information - their
152 inaccessibility.

153 An example of a large openly available palaeopalynological database is the Jansonius and Hills
154 Catalogue (JHC; 1976 and subsequent updates) of fossil spore and pollen genera. The JHC was
155 created as a series of physical cards - originally one card per genus – and contains more than
156 4500 spore (fungi, plants and algae) and pollen genera from the Phanerozoic Eon spanning the
157 last 540 million years. This database was created by Calgary palynologists J. Jansonius and L. V.
158 Hills and provides original authorial (i.e., sourced from the peer-reviewed literature)

159 descriptions and diagnoses for these genera, details of the source publications and authors,
160 descriptions of genus type species, and often subsequent genus descriptions, including formal
161 emendations. Taken as a whole, the catalogue is a useful resource which has no other
162 equivalent in palaeopalynology. The physical cards of the JHC were scanned into the form of a
163 PDF in the early 2000s, and are freely accessible online at the Calgary University Library website
164 (https://openlibrary.org/books/OL22388791M/Genera_file_of_fossil_spores_and_pollen).

165 Permission was granted from the Jansonius family and the University of Calgary to proceed with
166 the development. The first 78 pages of the PDF consists of an introduction, genus lists,
167 corrigenda and addenda, and p. 79-5676 are the genera files.

168 The AI assisted key of Stephenson et al. (2025) was developed for a small dataset, and ‘chain of
169 thought’ logical steps were embedded in the system through structuring of the learning material
170 (for example in providing sequenced choice stages linked to hierarchical aspects of taxonomy).

171 The approach to the development of a LATS for JHC was different because of the difficulties of
172 pre-structuring such a large dataset of very heterogeneous palynomorph types (fungal spores,
173 vascular plant spores, angiosperm and gymnosperm pollen, algal and other spores). For this
174 LATS, the Retrieval-Augmented Generation (RAG) technique was adapted to search for the best
175 matches based on users’ descriptions, incorporating, if necessary, iterative sequences of
176 prompts. (Lewis et al. 2020)

177 *Guiding principles during preparations of materials*

178 A number of principles were decided on before development, these included inclusion,
179 assistance, and non-intervention.

180 *Principle of inclusion*

181 Any taxonomic key or identification system attempts to narrow down possible ‘candidates’
182 using information provided by the user (the user’s ‘prompts’). Many early guides used
183 dichotomous keys, which are classification tools that allow the user to choose between
184 opposing pairs of morphological characters necessitating the rejection of groups of taxa at each

185 point of choice (e.g. Lamarck 1778). In the present system which uses a Retrieval,
186 Augmentation, and Generation (RAG) method, the quality of the initial prompt - in terms of
187 accuracy and detail -determines the numbers of candidate taxa, and the quality of the match
188 between the prompt (the description of the observed specimen) and the candidates. However,
189 both methods of input potentially exclude candidate taxa, if a mistake on the part of the user is
190 made early on. Thus, the system is designed as much as possible to allow a user to backtrack if
191 necessary, and to include rather than exclude candidate taxa; in other words, if in doubt, the
192 system includes rather than excludes candidate genera.

193 This principle is extended to taxa that are invalid (not published legally in line with ICBN rules).
194 Many of the genera in JHC are invalid in that for example, a holotype was not designated. The
195 system however includes any relevant invalid genera (identified by Jansonius and Hills or
196 identified subsequently) within its candidate lists following a prompt. The genus is noted as
197 invalid but is included for completeness. It is sometimes the case that invalid genera are later
198 validated (for example by designating a holotype) because they are considered useful. Thus, in
199 keeping with JHC, the system is inclusive in the spirit of helping to keep taxonomy transparent.

200 *Principle of assistance*

201 The LATS is intended as an aid to taxonomy, not a replacement for a human taxonomist. The
202 principle followed in the design is that the LATS provides relevant information to allow the
203 taxonomist to make an informed determination. The use of the LATS encourages transparency in
204 that the system provides information at every step; and through regular use, has pedagogic
205 value in that it illustrates how palynological taxonomy proceeds. This contrasts with image
206 recognition systems that essentially recognise visual patterns and do not consider the
207 authoritative text.

208 *Principle of non-intervention*

209 The database developed by Jansonius and Hills is unaltered in the LATS. Although it is possible
210 that improvements to diagnoses (for example making them clearer or more precise) would

211 improve the LATS's discrimination capability, this would involve decisions made by the creators
212 of the LATS, which would be interventionist. Further use of the LATS and the system that
213 underlies it could bring about improvement and rationalisations in palaeopalynological
214 taxonomy (for example by identifying possible synonyms, see later section) but this would have
215 to be done as a community, perhaps at some future date.

216 Challenges

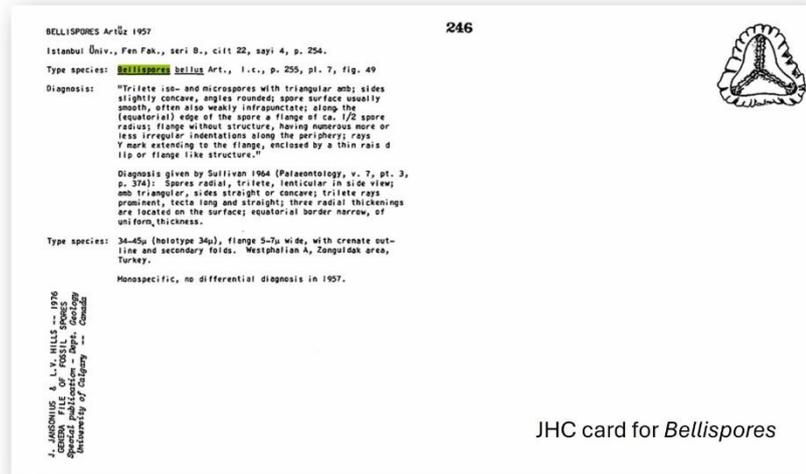
217 The first challenge is the quality and non-standard format of the PDF resulting from the scan of
218 the original JHC cards. The text was originally manually typed onto cards in the 1970s and 80s
219 and some of the typed letters are not clear which causes problems for machine recognition. An
220 example is that the machine does not distinguish always between the typed letter "O" and
221 number "0", with the result that the names of genera and their characteristics are sometimes
222 obscured. Similar mistakes occur with recognition of line breaks, non-standard symbols and
223 letters, or letters with umlauts, for example. Correction of these mistakes can be a long,
224 laborious process involving manual checking of OCR (Optical Character Recognition) results.

225 A second issue is that the quality of original descriptions and diagnoses of genera transferred
226 from the literature into JHC is very variable. There are some excellent comprehensive
227 descriptions and diagnoses of genera erected in the later years of the rapid expansion of
228 palaeopalynology in the 1960s and 1970s, particularly when the science was being taken up by
229 the oil industry. However, many early genus descriptions and diagnoses are very short and
230 sketchy, partly because at the time of their composition, palynology was a young science, and
231 few other genera had been recognised and low levels of detail in descriptions and diagnoses
232 were the norm. Jansonius and Hills also correctly recognised that some genera have been
233 invalidly published, so some names appearing in the JHC are superfluous. Jansonius and Hills
234 often dealt with descriptions and diagnoses in foreign languages by translating them (many are
235 translated from Russian). The variability in diagnosis/description leads to other difficulties of
236 bias and an inability of the system to distinguish genera. This is discussed in a later section.

237 Inconsistent use of terminology is a third issue. This inconsistency occurs in several forms, for
238 example synonyms for the same ornament element type (verrucae/warts; see e.g. Punt et al.
239 2007), or more commonly multiple slightly different versions of descriptive terms (spines or
240 spinae; bacula or baculae). Palynological terminology has also developed near parallel systems
241 (e.g. exoexine/intexine, nexine/sexine; cavate/camerate; e.g. Punt et al. 2007) often due to
242 different usage in the different international ‘schools’ of palynology, and different terminology
243 developed for different time periods (e.g. the Palaeozoic and Cenozoic; Stephenson 2016).
244 Elements of LLM training can alleviate these problems through the creation of ‘question-and-
245 answer pairs’ because they help the LLM to learn the meanings of terms, their interchangeability
246 and their variability, and how to generate relevant, structured, and context-aware responses
247 (e.g. Wei et al. 2022). For the JHC, two kinds of question-and-answer pairs were developed. The
248 first were structured QA pairs directly anchored in the original diagnosis or description as
249 supplied in the JHC; in other words, the answers to the questions can be found directly within
250 the text. A set of Q&A pairs for the spore genus *Bellisporos* is shown in Fig. 1. Most questions are
251 simple: ‘Question; What kind of germination mark does *Bellisporos* have?’, ‘Answer; *Bellisporos*
252 has a trilete mark’; and the answers can be found directly in the corresponding card/record (Fig.
253 1). QA pairs must also include similarly anchored compound questions (requiring integration of
254 multiple information), and counterintuitive questions (testing the model’s error-correction
255 ability, e.g. ‘Is *Bellisporos* monolete?’). The simple, compound and counterintuitive QA pairs are
256 used to ‘supervised fine tune’ the model to expand the word lists for the genera.

Q&A for *Bellisporos*

Q	A
What kind of germination mark does <i>Bellisporos</i> have?	<i>Bellisporos</i> is trilete
What kind of amb does <i>Bellisporos</i> have?	<i>Bellisporos</i> is triangular
What shape are the sides (between the apices) in <i>Bellisporos</i> ?	The sides (between the apices) in <i>Bellisporos</i> are slightly concave
What kind of ornament does <i>Bellisporos</i> have?	<i>Bellisporos</i> is smooth or weakly infrapunctate
Does <i>Bellisporos</i> have a zona or a flange?	Yes <i>Bellisporos</i> has a zona or a flange
How wide is the zona or flange in <i>Bellisporos</i> ?	The width of the zona or flange in <i>Bellisporos</i> is about half the spore radius
Does the zona or flange have any structure or features in <i>Bellisporos</i> ?	The flange of <i>Bellisporos</i> is unstructured but has indentations along the edge of the zona or flange
How long are the leaurae (rays of the Y mark) in <i>Bellisporos</i> ?	In <i>Bellisporos</i> the leaurae (rays of the Y mark) extend to the flange
Do the leaurae in <i>Bellisporos</i> have any structure?	The leaurae in <i>Bellisporos</i> have thin raised lips



JHC card for *Bellisporos*

257

258 Fig. 1. A series of QA pairs for the spore genus *Bellisporos*

259 The second type of QA pairs is known as 'incorporate' (e.g. Ouyang et al. 2022); these cover

260 areas of palaeopalynology, for example, basic principles and terminology outside the

261 immediate learning materials, so outside the original diagnosis or description as supplied in

262 JHC. They include questions like 'What are the characteristics of zonate spores?' and 'Describe

263 all the features of monolete spores.' The incorporate QA pairs are similarly used to supervised-

264 fine tune (SFT) the LLM in the broader knowledge of palynology, resulting in improved

265 understanding and interpretation of taxonomic terminology in the genera descriptions.

266 The embedding model was trained as follows: 1894 more incorporate QA pairs were generated

267 based on the 206 incorporate QA pairs provided by specialists with the GPT-5 model. These

268 generated QA pairs did not alter the meanings in the original question and answer pair but

269 changed the order and manner of the word usage. These QA pairs were then proofread by

270 specialists to make sure that no mistakes were made. For genus-related QA pairs, the LLM

271 learned the patterns of QA pairs created by specialists and generated 18 QA pairs in total for

272 each genus. (6 simple QA pairs, 6 compound QA pairs and 6 counterintuitive QA pairs) Some

273 genera in the JHC have no diagnosis (e.g. *nomina nuda*) and were removed so 69299 QA pairs

274 were used for training.

275 Among these 2100 incorporate and 69299 genus-related QA pairs, 1000 QA pairs were randomly
 276 selected as the test set and Mean Reciprocal Rank (MRR) and Recall were used as evaluation
 277 criteria. MRR indicates the quality of the position of the first relevant result in the retrieval
 278 ranking under a single query, which can be considered as a measure of the accuracy of the
 279 embedding model in semantic understanding. Recall, usually Top-K recall, measures the
 280 proportion of all relevant results that are retrieved and included in the Top -K ranking under a
 281 single query.

Criteria	MRR		Recall	
	Before SFT	After SFT	Before SFT	After SFT
Top 1	0.514	0.901	0.514	0.901
Top 10	0.625	0.935	0.846	0.984
Top 100	0.630	0.936	0.943	0.994

282 Table 1. embedding model performance on semantics of palynological knowledge and
 283 professional terminology before and after supervised fine tuning. MRR = Mean Reciprocal Rank;
 284 SFT = Supervised Fine Tuning.

285 Based on the results (Table 1), it is clear that the embedding model better understands the
 286 semantics of palynological knowledge and professional terminologies after supervised fine
 287 tuning. This improves the evaluation of semantic similarities between user input and diagnosis.

288 Model construction

289 There are three major steps in the LATS: (1) understand users' input (prompt) and provide
 290 reasonable matches; (2) provide a ranked list of candidates based on the calculated similarity
 291 scores between users' prompt and genera diagnoses; and (3) provide tips to improve users'
 292 prompts and a suggested refined query or prompt (Fig. 2.). Note that reasonable matches in

293 Step 1 are generated purely with the LLM and can be minorly varied given identical input. While
294 the list of candidates in Step 2 is provided based on statistical analysis, comparing the user
295 input only with diagnosis at genus level, the LATS would first transfer user input to vectors based
296 on the embedding model, then use cosine similarity to calculate the semantic similarity. (Salton
297 et al. 1975, Deerwester et al. 1990)

298
$$\cos\theta = \frac{\vec{A} \cdot \vec{B}}{\|\vec{A}\| \times \|\vec{B}\|}$$

299 where the numerator is the dot product of vectors \vec{A} and \vec{B} ; the denominator is the product of
300 the L2 norms of vectors \vec{A} and \vec{B} . Namely, the candidate list would remain the same given
301 identical user input. A threshold is then set to include as many similar taxa as possible, rather
302 than eliminating possibilities, and users can choose to show the top 5, top 10, or top 20 genera
303 only. At this point an ‘approximate string matching’ or ‘fuzzy’ search is also possible (Ukkonen
304 1985, Bast and Celikik 2013). This is a search method that finds results not identical to the
305 query but similar or closely related to it. It tolerates typos, misspellings, missing/extra
306 characters, or minor formatting differences—unlike exact search, which only returns perfect
307 matches.

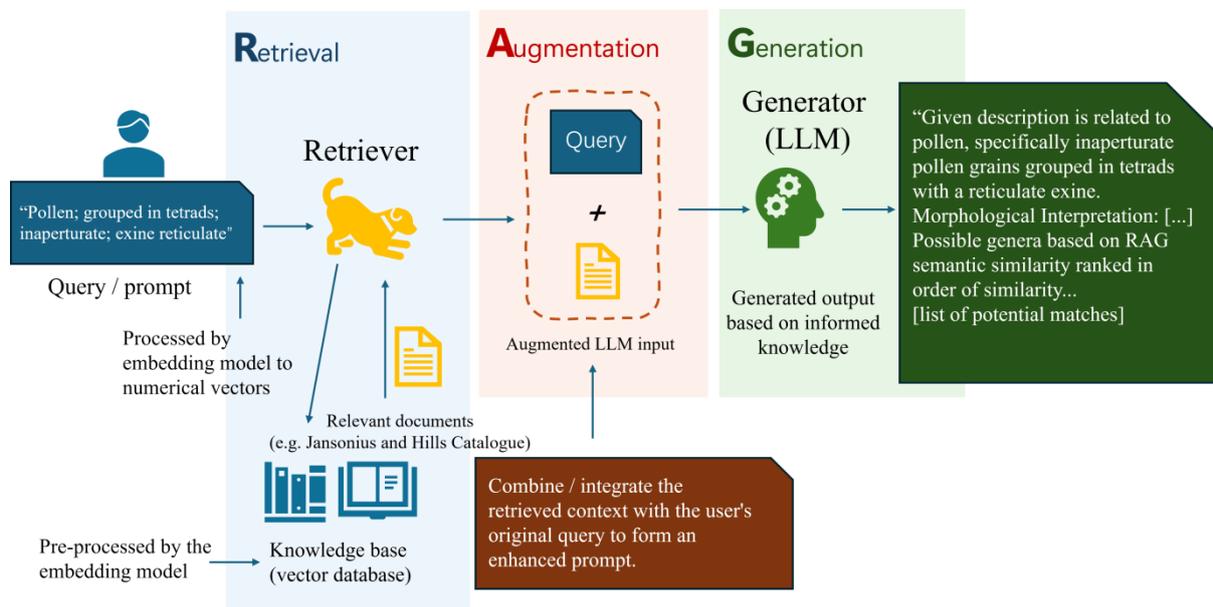
308 The LATS is built with the GeoGPT 72B model, which is trained on an open-source LLM Qwen-
309 72B from Alibaba Group (Yang et al. 2025). The training process for the GeoGPT models consists
310 of three key stages.:

- 311 1. Continual Pre-training (CPT) : The continual pre-training stage utilizes a variety of
312 datasets, including open access papers/books, and geoscience-related materials from
313 Wikipedia and Common Crawl that grant rights for non-commercial LLM modeling
314 training and model output sharing. This extensive corpus is used to ensure a solid
315 foundation in geoscience literature (Gururangan et al. 2020).

316 2. Multi-stage Supervised Fine-tuning (MSFT): The fine-tuning process is divided into three
 317 stages to refine the model's performance, using curated open-source QA pairs, the
 318 filtered Tulu-v3 dataset, geoscience-related QA data and long-context data. (Ouyang et
 319 al. 2022)

320 3. Direct Preference Optimization (DPO): The final stage, direct preference optimization,
 321 involves refining the model based on prompts paired with preferred answers. This
 322 ensures that the model's responses align more closely with user expectations and
 323 preferences. (Rafailov et al. 2023)

324 These processes ensure that the GeoGPT model is trained on a wide range of geoscience data,
 325 fine-tuned for specific tasks, and optimized for user preferences. All the data used for training
 326 comes from open-access publications with permission to train non-commercial LLM and share
 327 model output. (GeoGPT User Manual 2026).



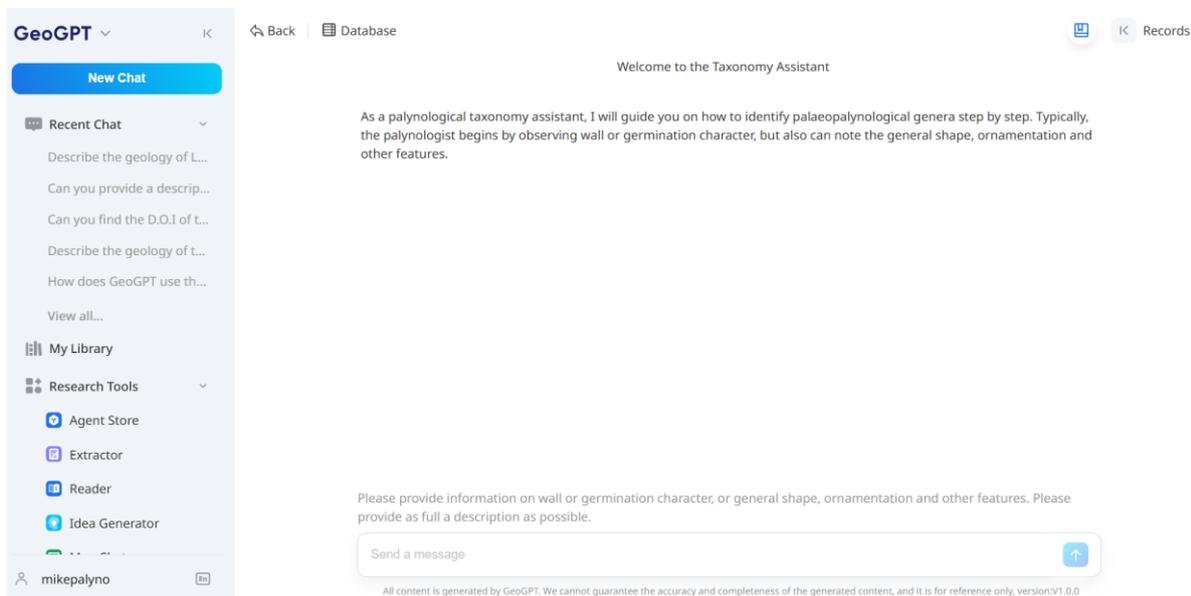
328
 329 Fig. 2. RAG workflow for LATS.

330 The essence of the embedding model is to convert discrete and unstructured text into fixed-
 331 dimensional continuous dense vectors (embedding vectors), enabling machines to achieve
 332 semantic-level understanding, matching, and interaction through vector similarity calculations.

333 (Reimers and Gurevych, 2019). The embedding model for this LATS is the BAAI General
334 Embedding large v1.5 (BGE-large-en v1.5) (Zhang et al. 2023, Chen et al. 2024). The BGE model
335 is a universal open-source embedding model that is commonly used in the open-source
336 community. The model adopts RetroMAE (Retrieval-oriented Language Models Via Masked
337 Auto-Encoder) pre-training technology and the large-scale contrastive learning training strategy,
338 and has been specially optimized for the problem of similarity distribution. (Xiao et al. 2022) In
339 version 1.5, it significantly improved the rationality of the similarity score distribution and
340 improved the ability to retrieve candidate genera. In the future development plan, the
341 embedding model will be trained using the QA pairs provided by palynological experts and
342 explanations of professional terms, enabling the LATS to possess palynological expertise,
343 thereby better understanding the user input and providing better matching of candidate genera.
344 Finally, a measure was taken to determine whether the prompt is related to spores, pollen, or
345 other palynomorphs (e.g. dinoflagellate cysts, chitinozoans, acritarchs). If irrelevant, the LATS
346 would clearly state that the description is not related to palynology and suggest how to describe
347 morphological features relevant to spores, pollen, or other palynomorphs (e.g. shape, size,
348 ornamentation, aperture type, etc.). Then immediately end the response. If there were multiple
349 rounds of input, the LATS would consider all user inputs and generate matching results based
350 on all historical inputs. Additionally, users are allowed to edit previous input and re-generate
351 answers from any historical round. After the re-generation, the system will retain all outputs
352 from previous rounds but clear the subsequent generation of the edited round and start
353 generating again based on the new input. Meanwhile, the system also provides a history record
354 module, through which users can access the historical records along with the corresponding
355 generated output.

356 **Operation of the system**

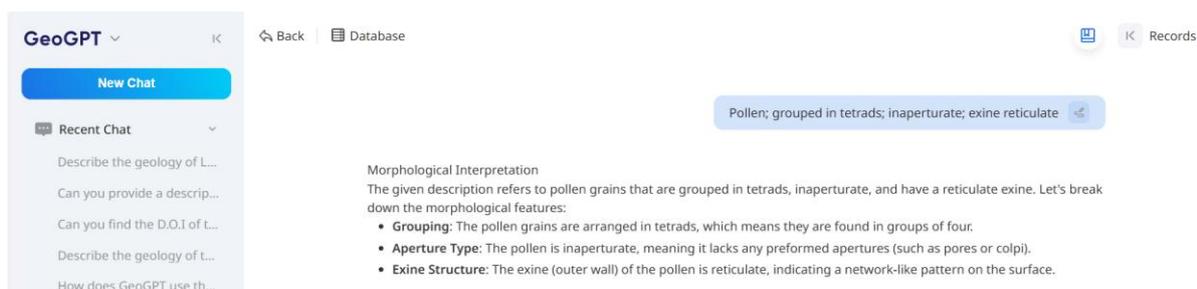
357 The LATS has a simple interface with a box for the user to type in a description as a prompt (Fig.
358 2).



359

360 *Fig. 3. Box for prompt*

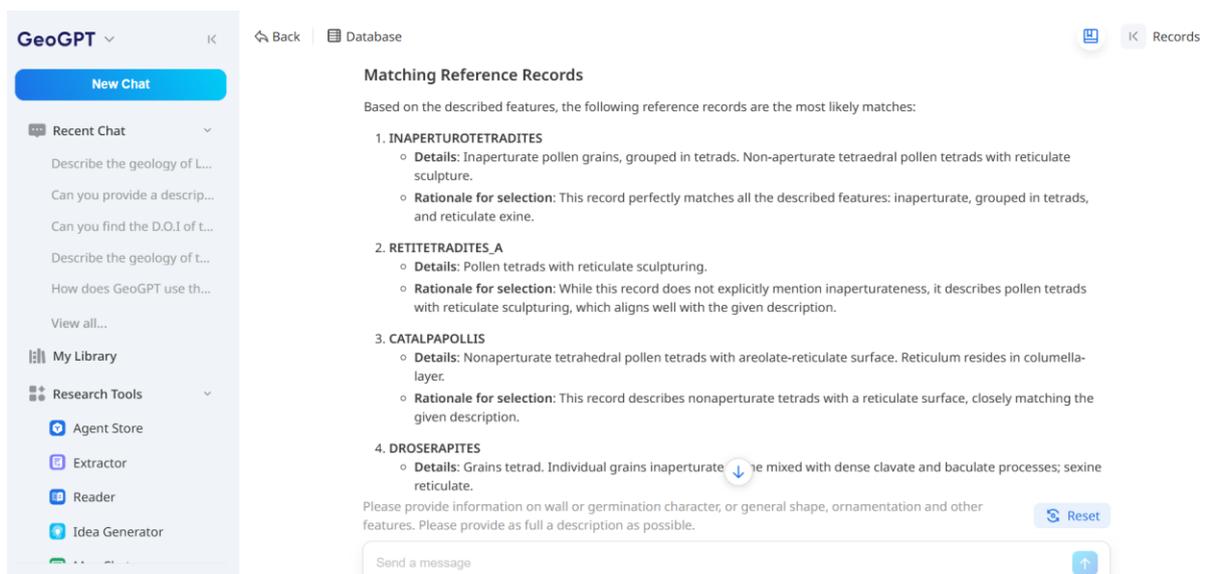
361 If a description such as ‘Pollen; grouped in tetrads; inaperturate; exine reticulate’ is entered, the
362 LATS provides first a morphological interpretation of the user’s description (Fig. 3). The purpose
363 of this section is for the user to check that the LATS understands the prompt, for example the
364 prompt mentions ‘grouped in tetrads’ and the LATS states that ‘the pollen grains are arranged in
365 tetrads which means they are found in groups of four’; similarly, the LATS confirms the meaning
366 of ‘inaperturate’. At this point, if the LATS does not understand the prompt, it will be clear to the
367 user who can backtrack and enter a slightly different prompt, perhaps with alternative terms.



368

369 *Fig. 4. Morphological interpretation of the user's description*

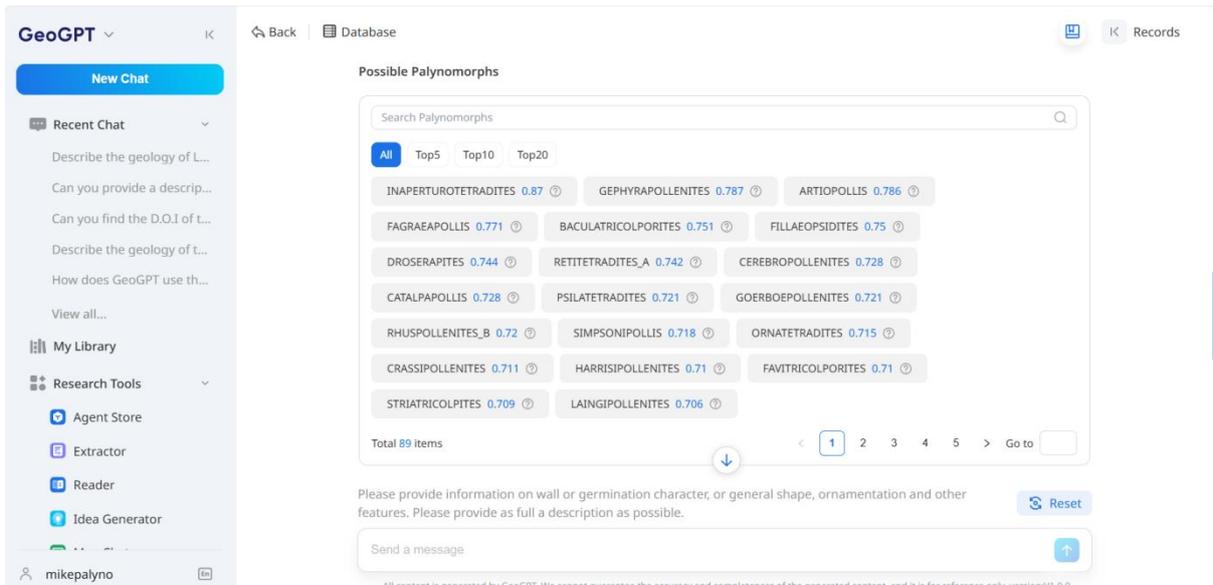
370 Following this stage the LATS produces the best five matches for the genera (Fig. 4) in relation to
371 the prompt description. These are not in order of level of fit. The LATS provides the name of the
372 candidate genus, details of the genus and the rationale for the selection.



373

374 *Fig. 5. Best five matches*

375 Below this, is a broader assessment of semantic similarity (Fig. 5), with candidate taxa ranked in
376 order of statistical similarity. This assessment provides the wider context of possible genera. At
377 either this stage or the previous stage the description/diagnoses of the candidate taxa can be
378 consulted, and the original scans of JHC can also be consulted through clicking the 'Database'
379 icon or the genus candidate names. Many JHC cards contain simple line drawings of the main
380 features of genera.



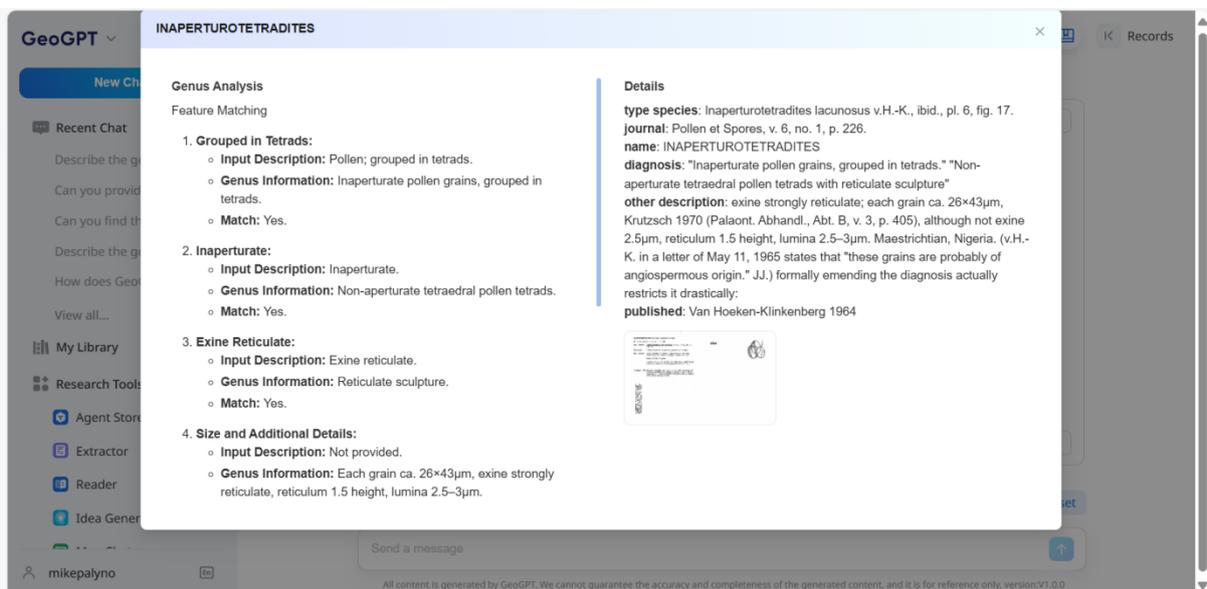
381

382 *Fig. 6. Broader assessment of semantic similarity*

383 For a detailed rationale for the match as well as areas where there is a mismatch, an

384 assessment is also available for each candidate genus (Fig. 6)

385

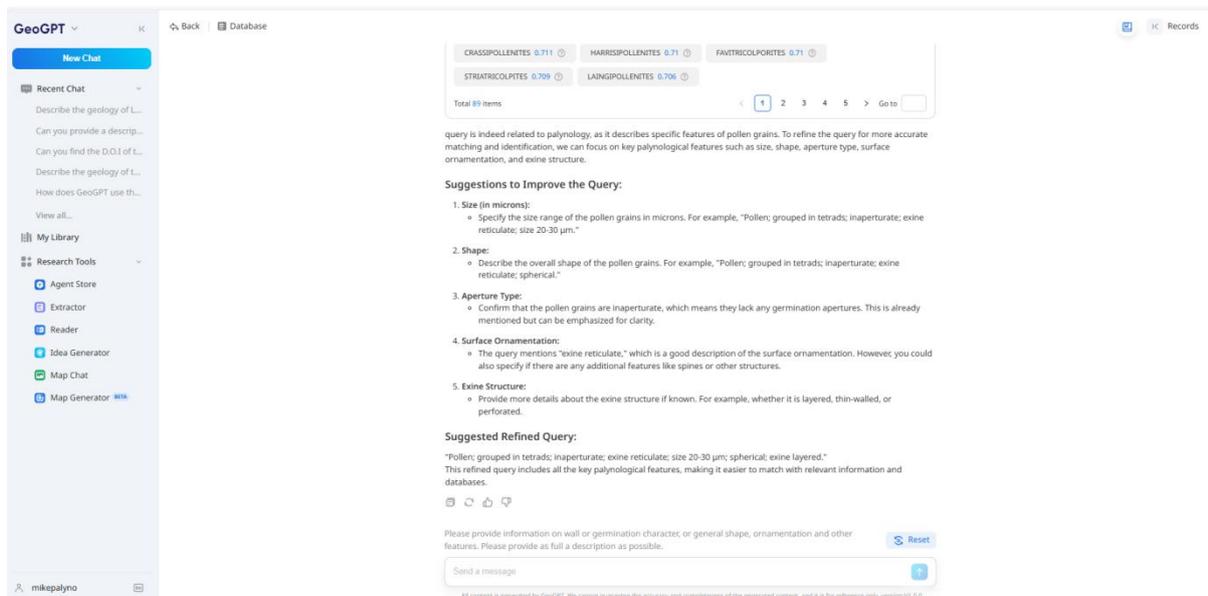


386

387 *Fig. 7. Detailed rationale including reasons for match and mismatch for each candidate genus*

388 The LATS also provides tips to improve the prompt and a suggested refined query or prompt (Fig.

389 7).



390

391 *Fig. 8. Tips to improve user's prompt and a suggested refined query or prompt*

392 Limitations of the system

393 There are two types of limitation or shortcoming: those the emanate from JHC itself, and those
 394 that emanate from the system.

395 Limitations due to JHC

396 The JHC contains genus names that are (1) poorly described, because they were originally
 397 poorly described in the relevant peer-reviewed paper or report, and their descriptions or
 398 diagnoses have never been improved, (2) invalid, (3) are described in translations into English
 399 from original diagnoses, introducing an element of interpretation on the part of the translator,
 400 (4) acknowledged and unacknowledged synonyms, and (5) homonyms. Jansonius and Hill's
 401 aims were to document as completely as possible the array of genera that a palaeopalynologist
 402 has at his/her disposal. Regular users of the PDF of JHC are aware of these characteristics, and
 403 aware that information in JHC is not sufficient to distinguish certain genera - or that certain
 404 useful genera are technically invalid. This often means that it is impossible using the
 405 information provided to make a satisfactory determination. This is not the fault of JHC, but of the
 406 taxonomy of palaeopalynology itself, which grew rapidly from a relatively small database in the

407 early 19th Century to a growing science in 1960s and 70s, particularly as its use in economic
408 biostratigraphy became important (e.g. McGowran 2005). Thus, the LATS does not provide any
409 better levels of discrimination between some genera but simply offers the same potential for
410 determination as study of the JHC. Here a few examples are examined and discussed in detail.

411 *Valid but 'redundant' genera*

412 Amongst the many genera of the JHC are those that are valid but rarely 'used'. This is often the
413 case with genera that have been described early in the growth of palaeopalynology as a
414 discipline. Their diagnosis or description may be rather short often because - to the original
415 describing palynologist - there appeared to be few similar genera to discriminate. An
416 explanation for their low usage maybe that later palaeopalynologists preferred to use genera
417 that seemed more appropriate with more circumscribed morphological variation. Another
418 explanation for this 'redundancy' relates to the siloed expertise of palaeopalynologists in a
419 certain Period (for example the Permian), and a particular group of genera become associated
420 with that system which become habitually used by 'Permian palynologists' even though suitable
421 genera may be available in the Mesozoic (for example the Triassic).

422 An example is the genus *Cingulatzonites* Mädler 1964, from the Upper Rhaetian of Germany
423 (Upper Triassic). Its diagnosis (translated from the German) is: '*Trilete miospores, amb*
424 *subcircular to rounded triangular; cingulum present, thinning wedge shaped towards the*
425 *equatorial margin, and grading into a narrow zona; cingulum smooth and structureless, which*
426 *differentiates the genus from Densosporites'* (Mädler 1964). A palaeopalynologist working in a
427 Carboniferous succession observing a spore with '*...cingulum present, thinning wedge shaped*
428 *towards the equatorial margin...*' might expect to find the genus *Cingulizonates* (first erected by
429 Dybova & Jachowicz 1957 from the Upper Carboniferous) among the higher-matching LATS
430 candidates but will also find *Cingulatzonites* which is never used in the Carboniferous, even
431 though its circumscription is suitable for many spores that might be assigned to *Cingulizonates*.

432 *Invalid genera*

433 As discussed above, the JHC includes genera that are invalid under ICBN rules, as well as
434 genera originally considered invalid but then made valid through, for example, the designation
435 of a holotype by a later author (see e.g. McNeil 2014). All this information, when available in
436 JHC, is also included in the LATS. Thus, a prompt will potentially provide these kinds of genera.
437 Though this could be considered a limitation, the LATS allows the palynologist to see a full range
438 of candidate genera without exclusion. As stated in the principle of ‘assistance’ above, the
439 purpose of the LATS is to assist the palynologist in his/her determination, rather than supplant
440 the palynologist’s role.

441 *Near-synonyms*

442 A related challenge is that as with JHC, the LATS, following a prompt, may return candidates
443 which are very close morphologically. As discussed later in this paper, conceptual
444 ‘morphological space’ appears to be crowded in some areas of palynological taxonomy,
445 particularly where small differences in morphology are considered important for
446 biostratigraphy, for example in the Palaeozoic of the Middle East and the Mesozoic of the North
447 Sea. Thus, prompts with low discriminative potential may supply many, very similar genera. In
448 these, the judgment of to the user is required to discriminate amongst the list of candidates.

449 *Limitations due to system operation*

450 *Bias*

451 The LATS considers similarity that is expressed through text and returns candidate genera based
452 on statistical matches. The most challenging aspect of this is that genera with short
453 descriptions or diagnoses, which encompass a wide theoretical range of morphology, are very
454 likely to appear in a candidate list. On the other hand, those that are described with detailed
455 textual terms that enable discrimination, will appear in candidate lists but not with a high level
456 of statistical similarity. In mathematical terms this can be explained simply: based on a one-
457 word input from a user, the statistical distance of a 10-word diagnosis is larger than that of a 2-

458 word diagnosis, although they are semantically similar, which leads to a higher similarity score
459 for the genus with a 2-word diagnosis. Thus, it could be said that the LATS is biased against
460 genera with a diagnosis/description with a high discriminative potential.

461 Without altering the detail of diagnosis/description (which would contravene the principle of
462 non-intervention above) the best solution to such bias is to allow the palynologist access to all
463 the candidate taxa as in the principle of inclusion above, and to allow swift access to the
464 original descriptions and diagnoses. This ensures that the palynologist makes the final
465 determination with all the relevant material at his/her disposal.

466 Further uses of the LATS

467 As well as the providing a LATS to aid a palynologist through the stages of identification, allowing
468 him/her to make the best judgements on determination with the best supporting information,
469 the information and 'understanding' that the underlying system has over a large area of
470 palaeopalynological taxonomy means that LATS could be put to more general uses. Two
471 examples follow.

472 Synonyms

473 In the early years of palaeopalynology many genera were named quickly, locally, sometimes
474 with limited access to global literature; and often without comparison to previously described
475 material (e.g. Fensome 1990, Traverse 1988). This led to a proliferation of synonyms; two (or
476 more) different names referring to the same taxon. Synonymy affects the accuracy of
477 biodiversity estimates, biostratigraphic correlation, macroevolutionary and palaeobiogeography
478 analyses, and the stability of scientific communication (e.g. Patterson et al. 2010). Because of
479 the 'understanding' that the LATS has of palaeopalynological taxonomy, it is capable of
480 detecting potential synonyms and indicating the priority of the name, the senior synonym,
481 which is the earliest validly published scientific name for a taxon.

482 An example is the Genus *Tethysispora* Vijaya and Tiwari (in Vijaya et al. 1988) whose diagnosis,
 483 when input as a prompt to the LATS produces candidate genera with high semantic similarity
 484 such as *Vallatisporites* Hacquebard 1957 and *Gondisporites* Bharadwaj 1962. If the user found
 485 that there were grounds for suspecting synonymy, then *Vallatisporites* Hacquebard 1957 would
 486 take priority.

487 Distribution of genera (or taxa) in ‘morphological space’

488 The morphological range of palynomorphs could be seen as occurring in multidimensional
 489 space with dimensions like width, height, density of ornaments, and size of ornaments.

490 By creating a feature table (Table 2), all JHC diagnoses can be reorganized or parsed so that the
 491 embedding model can convert the elements of descriptions into fixed-dimensional embedding
 492 vectors. Furthermore, an unsupervised learning approach, such as principal component
 493 analysis (PCA) or *K*-means clustering can be adapted to investigate genera in morphological
 494 space. (Jolliffe 2005, Hartigan and Wong 1979)

Features	Description of Feature
Wall layer number	This taxonomic character concerns the number of wall layers
Cameration or cavation	This taxonomic character concerns the presence and extent of space or cavity between wall layers
Ornament type	This taxonomic character concerns the type of ornament of the surface of the palynomorph or within its wall
Shape (amb)	This taxonomic character concerns the outline of a pollen grain or spore seen in polar view
Cingulum or zona	This taxonomic character concerns the presence of an equatorial extension of the palynomorph wall: thick (cingulum), thin (zona)

495 Table 2. Spore characters

496 As an example, a table of features of spores of this type enables *Tethysispora* to be categorized
 497 (Table 3).

498

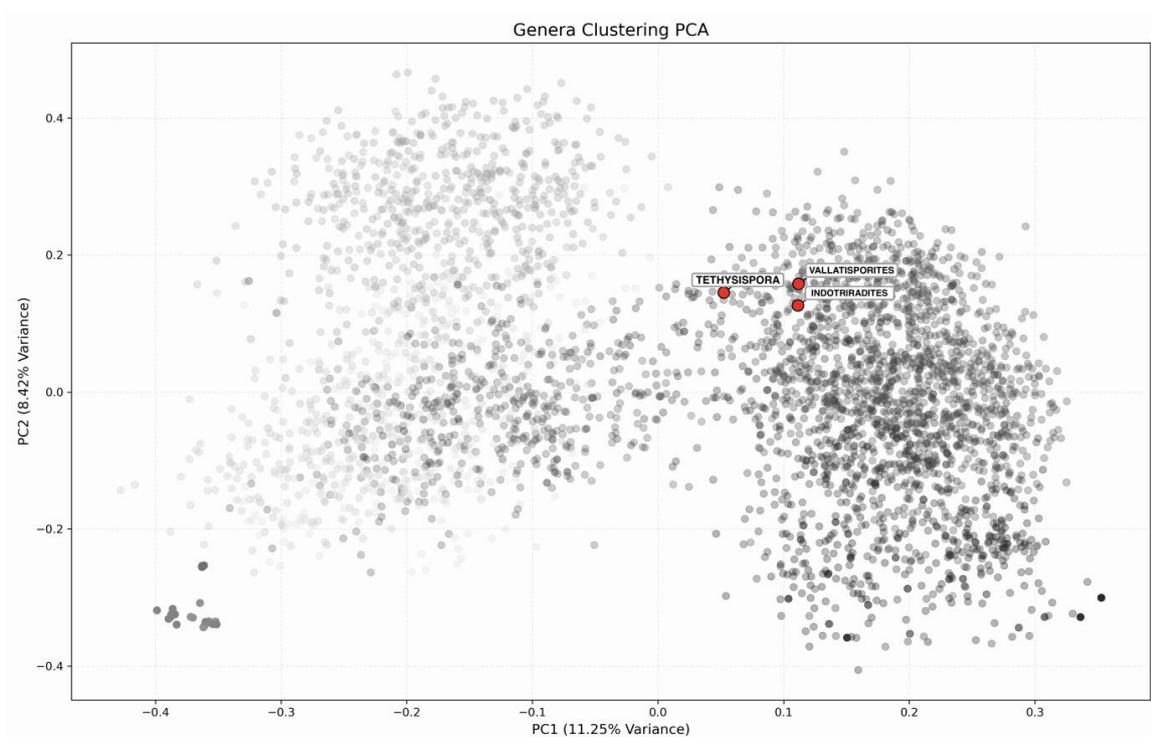
499

500

Genus name	Wall layer number and arrangement	Cameration or cavation	Ornament type	Shape (amb)	Cingulum or zona
<i>Tethysipora</i>	Exine two-layered	Cavity equatorially, distally and partly proximally within the central body, surrounding the inner body	Small, 1–2µm regular to irregular coni; broad-based to mammoid-shaped spines with elongated, curved detachable apices also seen in some processes; processes generally sparse on distal polar region but becoming bigger and denser towards body equator; sometimes mammoid process also present along the equator of body. Ornamentation crowded on the equatorial region of the body imparting a denser appearance to the zona-base and projecting out prominently below the zona	triangular to broadly subtriangular	zona

501 Table 3. Features of *Tethysipora* Vijaya and Tiwari (in Vijaya et al. 1988)

502 A 2D-visualization for all JHC genera using PCA is shown in Fig. 9, including the candidate
503 synonyms *Tethysipora*, *Indotriradites* and *Vallatisporites* indicating the small statistical
504 distances between the genera.



505
506 Fig. 9. 2D-visualized morphological variation amongst JHC diagnosis with marked genera

507 Implications for LATS in other parts of palaeontology

508 The use of artificial intelligence is likely to increase in palaeontological taxonomy. Perhaps the
509 most rapid development is in image recognition based on learning from very large numbers of
510 high-resolution images of specimens without direct reference to authoritative text-based
511 diagnoses or descriptions of taxa (e.g. Niu et al. 2024). Though valuable, this method places
512 expert validations between the user and the first-principle taxonomic information. Thus, a LATS
513 has a special role in providing direct access to authoritative text-based diagnoses or
514 descriptions of taxa and we believe should be encouraged as a useful aid to professional
515 taxonomists and as a pedagogic tool for educators.

516 Many of the challenges and opportunities illustrated in this study of the conversion of a
517 palaeopalynological dataset into LATS will be similar for other fossil groups. Much
518 palaeontological taxonomic information is disparate and varies in quality and detail or may be
519 unavailable to many as it is locked behind a paywall. In palaeontological groups with a long
520 history of research, for example the brachiopods, the original peer-reviewed literature also
521 contains variation in the meaning and use of morphological terms (e.g. Williams et al. 1965).
522 The Treatise of Invertebrate Paleontology, published from 1953 by the Geological Society of
523 America and the University of Kansas provides more consistent detail. It comprises 55 volumes,
524 written by more than 300 palaeontologists, and is subdivided into a number of sections each
525 dealing with a different invertebrate animal phylum. An example is the Treatise Section Part H
526 Brachiopoda (Revised) (Williams et al. 1999-2007), a revised version of earlier brachiopod
527 volumes (Williams et al. 1965), which is one of the most complete and recently updated
528 sections of the Treatise. The work of many brachiopod specialists allowed the provision of
529 condensed, rationalised descriptions and diagnoses of 4192 genera, and offer consistent
530 morphological terminology. However, complete original (authorial) diagnoses and descriptions
531 are not provided (though are referenced).

532 The methodology used in this study may be of value to other scientists working palaeontological
533 taxonomy.

534 Conclusions

535 Large Language Models (LLMs), being text-based, are ideal types of artificial intelligence to aid
536 in palaeontological taxonomic determination because published textual descriptions are the
537 ultimate, authority in taxonomy. This paper describes some of the preparation required to
538 develop a database for LLM development.

539 The Jansonius and Hills palaeopalynological catalogue (JHC) is a unique record of taxonomic
540 data in palaeopalynology in that it provides original authorial diagnoses and descriptions on
541 which to train the LLM augmented taxonomic key (LATS), thus providing authoritative ‘first
542 principle’ taxonomic information.

543 The LATS functions through Retrieval Augmented Generation (RAG) and returns candidate
544 genera with statistical measures of match against the prompt(s) or users’ descriptions.

545 The advantages of the use of the LATS include rapid access to statistically relevant candidate
546 genera, including the authorial diagnoses and descriptions, which allows the taxonomist to
547 make efficient and accurate determinations. The LATS also supplies information on the
548 structures and methods of taxonomy and thus has pedagogic value, as well as broader research
549 value in the theory and concepts of taxonomy.

550 The development of the LATS described here has implications for other palaeontological groups
551 in terms of the text basis of their taxonomy (for example variable quality of descriptions and
552 inconsistency in terminology), and their suitability for development of other LLM-assisted
553 taxonomic aids or keys.

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556 family in the development of this project.

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