

Comparing General and Scientometrics-Specific Academic Word Lists for Improving Reading Comprehension in Library and Information Science Students

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Abstract

This study addresses a key gap in English for Specific Purposes (ESP) pedagogy within Library and Information Science (LIS) by developing and evaluating a Scientometrics-Specific Academic Word List (SSAWL) to enhance reading comprehension. General Academic Word Lists (AWLs), such as Coxhead's (2000), often overlook domain-specific terminology essential for scientometrics (e.g., "bibliometrics," "h-index"). A specialized 8.5-million-word corpus from leading scientometrics journals produced a validated SSAWL of 1,105 high-frequency word families. Using a quasi-experimental pretest-posttest design, 60 intermediate-to-advanced EFL undergraduate LIS students were assigned to an experimental group receiving 12 weeks of explicit SSAWL instruction via Research-based Academic Vocabulary Education (RAVE) or a control group receiving equivalent general AWL training. Assessments included multiple-choice, short-answer, and cloze reading tasks using authentic scientometrics passages, an adapted Vocabulary Levels Test, surveys, and focus groups. Results showed that the SSAWL group achieved significantly greater gains in scientometrics text comprehension (accuracy and speed) and in domain-specific vocabulary knowledge, with large effect sizes (Cohen's $d = 1.67-3.85$; partial $\eta^2 = 0.39-0.79$). Qualitative findings indicated improved confidence and contextual vocabulary use. This study demonstrates the superiority of corpus-derived, domain-specific word lists over general AWLs for specialized reading in LIS and offers a validated resource for ESP curriculum design.

1. INTRODUCTION

Reading comprehension is a cornerstone of academic success, particularly in specialized fields such as Library and Information Science (LIS), where students must navigate complex, domain-specific texts to engage with disciplinary knowledge effectively (Arboleda et al., 2022; Coxhead,

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2000). Within LIS, the subfield of scientometrics, which focuses on quantitative study of science, scholarly communication, and research evaluation, presents unique lexical challenges as it relies on specialized terminology such as "bibliometrics," "h-index," and "citation analysis" (Hood & Wilson, 2001; Inastrilla et al., 2024; Sedighi, 2016). These terms, often underrepresented in general Academic Word Lists (AWLs) like Coxhead's (2000) seminal list, are critical for comprehending scientometric literature and participating in professional discourse. The inadequacy of general AWLs in addressing domain-specific lexical demands creates a significant barrier for English as a Foreign Language (EFL) LIS students, hindering their ability to process research articles efficiently and develop disciplinary expertise (Hyland & Tse, 2007; Tangsakul, 2024).

The critical role of vocabulary knowledge in reading comprehension is well-established in the literature, with both breadth (vocabulary size) and depth (understanding of word meanings and usage) serving as strong predictors of text understanding (Melani et al., 2021; Steensel et al., 2014; Tunmer & Chapman, 2012). General AWLs, designed to capture high-frequency academic vocabulary across disciplines, have been instrumental in English for Specific Purposes (ESP) pedagogy (Coxhead, 2000; Schmitt et al., 2011). However, empirical studies highlight that such lists often fail to account for the specialized lexis required in fields like scientometrics, leading to reduced lexical coverage and comprehension difficulties when students encounter domain-specific texts (Hyland & Tse, 2007; Lesaux et al., 2010; Vuković-Stamatović, 2024). This misalignment underscores the need for discipline-specific lexical resources, such as Scientometrics-Specific Academic Word Lists (SSAWLs), to bridge the gap between general academic vocabulary and the technical terminology essential for LIS education (Arboleda et al., 2022; Zhang, 2013).

Corpus-based methodologies have emerged as a robust approach to developing tailored word lists, enabling researchers to identify high-frequency, domain-relevant word families through systematic analysis of specialized texts (It-ngam & Phoocharoensil, 2019; Uehara et al., 2022). Prior studies in disciplines such as applied linguistics, physics, engineering, and medicine have demonstrated that specific academic word lists achieve superior lexical coverage and pedagogical relevance compared to general AWLs, supporting enhanced reading comprehension and disciplinary literacy (Chang, 2023; Khani & Tazik, 2013; Tazik et al., 2022; Yotimart, 2021). In scientometrics, where texts are characterized by dense technical jargon and methodological constructs, a corpus-derived Scientometrics-Specific Academic Word List (SSAWL) offers a promising solution to address the lexical demands of the field, aligning instructional materials with authentic reading tasks (Inastrilla et al., 2024; Luo et al., 2024).

Despite the recognized importance of vocabulary in reading comprehension, there is a paucity of empirically validated, domain-specific lexical resources for scientometrics within LIS education. General AWLs, while useful for broad academic contexts, often fail to capture the specialized terminology critical for understanding scientometric texts, potentially leading to inefficient vocabulary instruction and persistent comprehension challenges for EFL LIS students (Arboleda et al., 2022; Lesaux et al., 2010; Tangsakul, 2024). The lack of tailored lexical resources limits students' ability to engage with core scientometric literature, conduct literature reviews, and develop professional competencies, particularly in English-medium instruction (EMI) or ESP settings where lexical proficiency is paramount (Green et al., 2023; Rogobete et al., 2023). This gap necessitates the development and evaluation of a SSAWL to enhance reading comprehension and vocabulary knowledge in LIS programs.

The present study aims to develop and empirically evaluate the comparative efficacy of a SSAWL versus a general AWL in enhancing reading comprehension and domain-specific vocabulary knowledge among intermediate-to-advanced EFL undergraduate LIS students. Specific objectives include:

1. To compile a validated SSAWL, and compare the effectiveness of explicit SSAWL instruction, delivered through Research-based Academic Vocabulary Education (RAVE) methods, against equivalent general AWL instruction in a 12-week quasi-experimental intervention.
2. To assess improvements in reading comprehension and vocabulary knowledge using multi-method assessments, including authentic scientometrics passages, an adapted Vocabulary Levels Test, and qualitative data from surveys and focus groups.
3. To provide empirical evidence and a validated lexical resource to inform ESP curriculum design and instructional practices in LIS education.

This research addresses a critical gap in ESP pedagogy by providing a corpus-derived SSAWL tailored to the lexical demands of scientometrics, offering a replicable model for developing discipline-specific lexical resources. The findings contribute robust empirical evidence on the superiority of domain-specific vocabulary instruction, with implications for enhancing reading comprehension, vocabulary retention, and learner confidence in LIS programs. By aligning instructional materials with the authentic linguistic demands of scientometric texts, this research supports curriculum designers and educators in fostering disciplinary literacy and professional competence among EFL LIS students, particularly in EMI contexts (Chen et al., 2024; Xodabande et al., 2022). Additionally, the mixed-methods approach enriches the pedagogical discourse by integrating quantitative outcomes with qualitative insights into learners' experiences, informing future innovations in ESP vocabulary instruction (Hajiyeva, 2023).

In sum, this study advances the discourse on ESP vocabulary instruction by providing a validated SSAWL and empirical evidence for its efficacy in LIS education. By addressing the limitations of general AWLs and aligning vocabulary instruction with the lexical demands of scientometrics, the research offers a practical and theoretically grounded contribution to curriculum design, instructional practice, and disciplinary literacy development in specialized academic contexts.

2. LITERATURE REVIEW

Vocabulary Knowledge and Reading Comprehension

Vocabulary knowledge is widely recognized as a cornerstone of reading comprehension, serving as both a predictor and facilitator of academic literacy across proficiency levels and educational contexts. Empirical studies demonstrate that vocabulary breadth (the number of words known) and depth (semantic, morphological, and collocational knowledge) together explain a substantial proportion of variance in reading comprehension outcomes (Nation, 2006; Qian, 2002; Zhang & Yang, 2016). The lexical threshold hypothesis posits that readers must know approximately 95% of the words in a text for minimal comprehension and 98% for independent understanding (Laufer, 2013; Nation, 2006). In ESP, where texts are laden with technical and discipline-specific terms, lexical mastery becomes indispensable for comprehension and academic success (Chang, 2023; Masrai & Milton, 2018).

Limitations of General Academic Word Lists

Coxhead's (2000) Academic Word List, consisting of 570 word families, has been foundational for EAP pedagogy by identifying high-frequency academic vocabulary. However, subsequent research has revealed that general AWLs provide uneven coverage across disciplines and frequently omit critical domain-specific terms (Dang et al., 2017; Durrant, 2016; Hyland & Tse, 2007; Whitaker, 2024). For instance, scientometric concepts such as bibliometrics, h-index, and

citation analysis, central to Library and Information Science, are absent or underrepresented in the AWL (Durrant, 2016; Gaffas, 2024; Sedighi, 2016). This gap underscores the need for discipline-specific lexical resources to better support learners engaging with specialized texts.

Development of Domain-Specific Word Lists

Corpus-based methodologies have emerged as the most reliable approach to developing domain-specific AWLs. By analyzing large, representative corpora, researchers can identify high-frequency items, collocations, and lexical bundles that define disciplinary discourse (Argyroulis, 2022; Vuković-Stamatović & Živković, 2022). Notable examples include the Applied Linguistics AWL (Khani & Tazik, 2013), Zoology AWL (Kruawong & Phoocharoensil, 2020), Forestry Word List (Luo et al., 2024), and Medical AWL (Tazik et al., 2022). These domain-specific lists consistently demonstrate superior lexical coverage and pedagogical relevance compared to Coxhead's AWL. For scientometrics, corpus-driven development of a SSAWL ensures the inclusion of technical vocabulary absent from general AWLs while reflecting the evolving nature of LIS discourse (Durrant, 2016; Sedighi, 2016; Whitaker, 2024).

Pedagogical Applications and Instructional Frameworks

Domain-specific vocabulary instruction has been shown to significantly enhance comprehension accuracy, speed, and retention compared to instruction based on general AWLs (Lawrence et al., 2021; McKeown et al., 2018). Effective frameworks, such as RAVE, integrate retrieval, automaticity, elaboration, and orthography, enabling learners to consolidate specialized vocabulary knowledge (Wolf et al., 2000, 2009). Studies highlight additional benefits, including improved learner confidence, engagement, and disciplinary literacy (Whitaker, 2024; Wright & Cervetti, 2016).

Technology-enhanced approaches further expand instructional possibilities. Digital tools such as concordancers, mobile applications, and game-based platforms provide authentic, contextualized exposure to specialized lexis, improving motivation and long-term retention (Xodabande et al., 2022). Integrating authentic scientometric texts, such as journal articles and bibliometric reports, into instruction strengthens the relevance and applicability of specialized vocabulary learning (Durrant, 2016; Sedighi, 2016).

Assessment and Methodological Considerations

Assessment of domain-specific AWL interventions requires multiple measures to capture both vocabulary growth and reading comprehension gains. Standardized instruments such as the Vocabulary Levels Test and comprehension tasks (e.g., cloze, multiple-choice, short-answer) are commonly used (Beglar, 2009; Schmitt et al., 2011). Pretest–posttest quasi-experimental designs consistently demonstrate moderate to very large effect sizes for domain-specific AWL-based interventions (Lawrence et al., 2021; McKeown et al., 2018). Complementary qualitative evidence also suggests affective benefits, such as increased confidence and reduced anxiety, when learners engage with domain-specific vocabulary instruction (Binhomran & Altalhab, 2023).

Challenges and Research Gap in LIS

Despite promising results in other disciplines, LIS and scientometrics remain underrepresented in the development and validation of domain-specific academic word lists (Luo et al., 2024; Nekrasova-Beker, 2020). Existing instruction often relies on general AWLs, which inadequately address the specialized lexical demands of scientometrics. Moreover, challenges such as resource-

intensive corpus compilation, the need for teacher training, and the rapid evolution of scientometric terminology pose barriers to effective implementation (Jay & Sato, 2018).

Conclusion and Rationale for the Present Study

The literature converges on four key points: (1) vocabulary knowledge is critical for academic reading comprehension, (2) general AWLs, while useful, lack coverage of specialized terminology, (3) corpus-driven domain-specific AWLs provide superior lexical coverage and learning outcomes, and (4) explicit, technology-supported instruction maximizes their pedagogical impact. However, a significant gap remains in LIS, where no empirically validated SSAWL exists for scientometrics. The present study addresses this gap by developing a Scientometrics-Specific Academic Word List (SSAWL) and empirically evaluating its efficacy against a general AWL in enhancing LIS students' reading comprehension.

3. METHODOLOGY AND DESIGN

This study employed a mixed-methods approach to develop and evaluate the Scientometrics-Specific Academic Word List and its impact on reading comprehension among Library and Information Science students. The methodology was structured in two primary phases: the development of the SSAWL and the implementation of a quasi-experimental intervention. Data collection and analysis integrated quantitative and qualitative techniques to provide comprehensive insights into the efficacy of the SSAWL compared to a general Academic Word List.

Design

A quasi-experimental pretest-posttest design with a control group was utilized to assess the comparative effectiveness of SSAWL instruction versus general AWL training. This design facilitated causal inferences in a naturalistic educational setting, where full randomization was limited by existing class structures. The unit of assignment was intact classes (i.e., pre-existing student groups rather than individual students), with two classes randomly assigned to either the experimental or control condition after stratification to balance English proficiency (based on Oxford Placement Test scores) and prior LIS exposure (self-reported semesters completed). Stratification involved dividing students into strata based on these variables and ensuring proportional representation in each group. To minimize contamination risk, classes were scheduled in different time slots within the same term and department, with no shared activities between groups. Both groups were taught by the same instructor, who followed standardized lesson plans, materials, and pacing to ensure consistency across conditions. The experimental group received targeted SSAWL instruction, while the control group underwent equivalent training with a general AWL (Coxhead, 2000). The intervention spanned 12 weeks, allowing for pre- and post-intervention assessments to measure changes in reading comprehension and vocabulary knowledge.

Participants

A convenience sample of 60 undergraduate LIS students was recruited from Tehran University in an EFL context, with participants demonstrating intermediate to advanced English proficiency (equivalent to CEFR B2-C1 levels, as determined by a preliminary Oxford Placement Test). The sample size was determined through a priori power analysis using G*Power software, yielding sufficient power (>0.80) to detect medium effect sizes (Cohen's $d = 0.50$) at $\alpha = 0.05$, consistent

with prior vocabulary intervention studies (Xodabande et al., 2022). Inclusion criteria required enrollment in LIS courses with scientometrics content, while exclusion criteria eliminated students with native-level English proficiency or prior specialized scientometrics training. Demographic balancing was achieved through stratification, resulting in comparable groups in terms of age ($M = 21.4$ years, $SD = 1.8$), gender (52% female), and prior LIS coursework exposure. Participants were evenly divided between the experimental ($n = 30$) and control ($n = 30$) groups.

Phase 1: SSAWL Development

The SSAWL was developed using a corpus-based methodology to ensure empirical validity and domain specificity. A specialized corpus comprising 8.5 million words was compiled from 1,800 full-text articles sourced from six highest impact factor scientometrics journals: *Scientometrics*, *Journal of Informetrics*, *Journal of the Association for Information Science and Technology (JASIST)*, *Journal of Information Science*, *Research Evaluation*, and *COLLNET Journal of Scientometrics and Information Management*. Articles were selected from publications spanning 2000–2024 to capture contemporary terminology (except for *COLLNET Journal of Scientometrics and Information Management*, which was first published in 2007), with texts processed to remove non-linguistic elements (e.g., references, tables) using AntConc (Anthony, 2024) software.

Tokenization and lemmatization were performed using AntWordProfiler's default settings, which reduce words to their base forms (e.g., "citations" to "citation"). Word families were defined following Bauer and Nation's (1993) Level 6 criteria, including the headword, inflections, and common derivations (e.g., "cite" family includes "citation," "cited," "citing"). Proper nouns (e.g., journal names) and abbreviations (e.g., "JASIST") were excluded unless they functioned as common lexical items (e.g., "h-index" retained as a hyphenated single token). Hyphenated terms were treated as unified tokens if domain-specific (e.g., "co-citation"). Word families were identified and selected solely based on their frequency, range, and dispersion in the scientometrics corpus using AntWordProfiler (Anthony, 2024b). Families from the GSL and AWL were included in the SSAWL if they met the domain-specific selection criteria (token frequency >28 per million words, range $>50\%$, dispersion Juilland's $D >0.80$). This inclusive approach defines the SSAWL as a comprehensive domain list (not an add-on excluding general words) that captures all high-frequency lexis relevant to scientometrics, whether general or specialized; overlaps with GSL/AWL are explicitly marked in Appendix B. Frequency analysis was conducted with AntWordProfiler (Anthony, 2024b) to extract word families, prioritizing metrics such as token frequency (>28 occurrences per million words), range (appearance in $>50\%$ of texts), and dispersion (Juilland's $D >0.80$ for even distribution). AWL and the GSL (West, 1953) families were not excluded from SSAWL if they met the word selection criteria. This yielded an initial list of 1,309 word families, which was refined to the top 1,105 through iterative filtering. Validation involved expert review by three LIS faculty members specializing in scientometrics, who rated items for relevance and pedagogical utility on a 5-point Likert scale (inter-rater reliability: $\kappa = 0.85$). Adjustments were made based on feedback, resulting in a finalized SSAWL.



Figure 1: Word cloud of 50 most frequent word families in SSAWL

Coverage analysis showed that the SSAWL provided an additional 11.2% lexical coverage in independent scientometric test texts (e.g., recent articles not in the corpus) beyond the GSL + AWL, compared to 4.5% for the general AWL alone. See [Appendix](#) for detailed software settings (e.g., AntConc stopwords lists, AntWordProfiler range/dispersion parameters), full word family criteria, and the complete wordlist items.

Phase 2: Intervention

The 12-week intervention was delivered in-person during regular class sessions, with each session lasting 2 hours to maintain consistency across groups. Instruction for both groups was facilitated by the same trained ESP instructor to minimize instructor bias, and materials were standardized in duration and intensity.

The experimental group focused on explicit SSAWL instruction, introducing approximately 25 word families per week through RAVE strategies ([Lesaux et al., 2010](#)). Activities included:

- Morphological analysis (e.g., breaking down prefixes and suffixes in terms like "co-citation" or "altmetrics") and definitional instruction using bilingual glossaries.
- Contextual embedding via authentic scientometrics readings (e.g., journal abstracts and excerpts), supplemented by interactive exercises such as digital flashcards for spaced repetition (Anki software) and sentence-completion tasks.
- Collaborative activities, including group discussions and role-plays simulating scientometric analyses, to promote incidental learning and deeper retention through multiple exposures.

To address overlap with the general AWL (see [Appendix B](#) for shared items marked as italicized), instruction and materials for the experimental group highlighted domain-specific applications of overlapping terms within scientometrics contexts, while introducing the unique SSAWL terms not present in the general AWL. The control group received instruction exclusively on the general AWL items without scientometrics-specific contextualization. This design allowed evaluation of the added value of domain-specific focus and unique terms.

The control group received parallel instruction using Coxhead's (2000) general AWL, with equivalent activities adapted to non-domain-specific academic vocabulary (e.g., "analyze," "evaluate"). Fidelity was ensured through session observations and instructor logs, with attendance rates exceeding 95% in both groups.

Data Collection

A multi-method data collection strategy was implemented to triangulate findings, encompassing quantitative assessments and qualitative insights.

Pretest and posttest measures were administered to both groups at Weeks 1 and 13, respectively:

- Reading comprehension was evaluated using three authentic scientometrics passages (500–800 words each, sourced from the corpus journals), followed by 20 items per passage (10 multiple-choice, 5 short-answer, 5 cloze with SSAWL deletions to assess inference). Speed was operationally defined as words per minute (WPM), calculated as total words across passages divided by total time in minutes; participants read silently, with timing starting when they began the first passage and stopping upon completing the last item, using a digital stopwatch for precision. Short-answer items were scored using a 2-point rubric (0 = incorrect/irrelevant, 1 = partially correct with key elements, 2 = fully correct and complete), with two independent raters achieving inter-rater reliability (intraclass correlation coefficient, ICC = 0.92); discrepancies were resolved through consensus. Cloze items targeted SSAWL words in novel contexts not used in instruction, measuring vocabulary integration via contextual inference rather than direct recall; acceptable synonyms or semantically equivalent responses were scored correct to avoid teaching-to-the-test bias. Scoring emphasized accuracy (percentage correct) and speed (time to completion), with internal consistency (Cronbach's $\alpha = 0.87$).
- Vocabulary knowledge was assessed via an adapted Vocabulary Levels Test (Schmitt et al., 2001), targeting SSAWL items through definition-matching (receptive knowledge), sentence-completion (productive recall), and recognition tasks (30 items total; $\alpha = 0.82$). Supplementary data included:
 - Self-reported surveys (Likert-scale, 15 items) administered post-intervention to gauge perceived comprehension efficacy and vocabulary confidence ($\alpha = 0.89$).
 - Focus group interviews (semi-structured, 45 minutes each) with 15 participants (three focus group interviews, 5 participants each, randomly selected from the experimental group) explored qualitative insights into experiences, challenges, benefits, and perceived improvements in applying SSAWL terms to scientometric texts, audio-recorded and transcribed for analysis.

All instruments were piloted with a similar cohort ($n = 20$) to refine clarity and reliability, with counterbalancing applied to mitigate order effects.

Data Analysis

Quantitative data were analyzed using SPSS version 28.0. Descriptive statistics summarized participant characteristics and score distributions. Within-group changes were examined via paired-samples t-tests to detect pre-to-post improvements. Between-group differences were assessed with independent t-tests, ANOVA, or ANCOVA (covarying pretest scores and baseline proficiency). Effect sizes were calculated using Cohen's d (for t-tests) and partial η^2 (for ANOVA), with interpretations following conventions (e.g., $d \geq 0.50$ indicating moderate effects). Gain-score Cohen's d was computed to illustrate the magnitude of within-group changes, justified here as it

highlights practical improvements in a pre-post design, while ANCOVA partial η^2 and adjusted mean differences provide between-group comparisons controlling for baselines (Plonsky & Oswald, 2014). The primary (headline) analysis for between-group differences was ANCOVA on posttest scores covarying pretest scores; this was selected because it directly adjusts for baseline differences in this quasi-experimental pretest-posttest design, provides clear partial η^2 effect sizes, and is the recommended approach for such studies to maximize interpretability and statistical power. Exploratory repeated-measures ANOVA (Time \times Group) and regression models yielded congruent results and are not reported in detail here to maintain a clear primary analysis path and avoid any impression of method shopping.

Qualitative data from surveys (open-ended responses) and focus groups were analyzed thematically using NVivo software. An inductive coding approach identified emergent themes (e.g., "increased confidence in contextual application"), with inter-coder reliability ($\kappa = 0.82$) achieved through double-coding of 20% of transcripts. Triangulation integrated quantitative results with qualitative patterns to provide nuanced interpretations of the SSAWL's efficacy.

Ethical Considerations

The research strictly complied with ethical standards for studies involving human participants, prioritizing their well-being and the integrity of the data. Participation was voluntary, with informed consent obtained from all individuals before their involvement. Participants were fully informed about the study's objectives, procedures, potential benefits, and their right to withdraw at any time without consequences. To protect privacy, personal identifiers were replaced with unique codes to anonymize data, and confidentiality was upheld throughout the entire research process, from data collection to reporting.

4. RESULTS

Quantitative Findings

Descriptive statistics for pre- and post-intervention measures are presented in [Table 1](#). Both groups showed improvements across all measures, but the experimental group (SSAWL instruction) exhibited substantially larger gains.

Table 1: Descriptive Statistics for Vocabulary, Reading Accuracy, and Speed by Group and Time (N = 60)

Measure	Time	Control Group (n=30)		Experimental Group (n=30)		Mean Difference (Post-Pre)	
		Mean	SD	Mean	SD	Control	Experimental
Vocabulary	Pre	14.44	2.65	15.04	2.93		
	Post	17.25	3.11	24.00	3.20	+2.81	+8.96
Accuracy (%)	Pre	59.05	10.07	60.66	10.35		
	Post	70.44	9.98	84.99	8.49	+11.39	+24.33
Speed	Pre	147.13	22.78	155.64	15.28		
	Post	168.20	25.43	192.54	18.63	+21.07	+36.90

Note: Vocabulary scored 0-30; Accuracy 0-100%; Speed in words per minute (WPM). Higher speed values indicate faster reading. Standard deviations are in parentheses.

Paired-samples t-tests revealed significant within-group improvements for both groups on all measures (all $p < .001$). For the control group: vocabulary ($t(29) = 11.05$), accuracy ($t(29) = 13.17$), and speed ($t(29) = 12.56$). For the experimental group: vocabulary ($t(29) = 27.00$), accuracy ($t(29) = 17.72$), and speed ($t(29) = 20.04$).

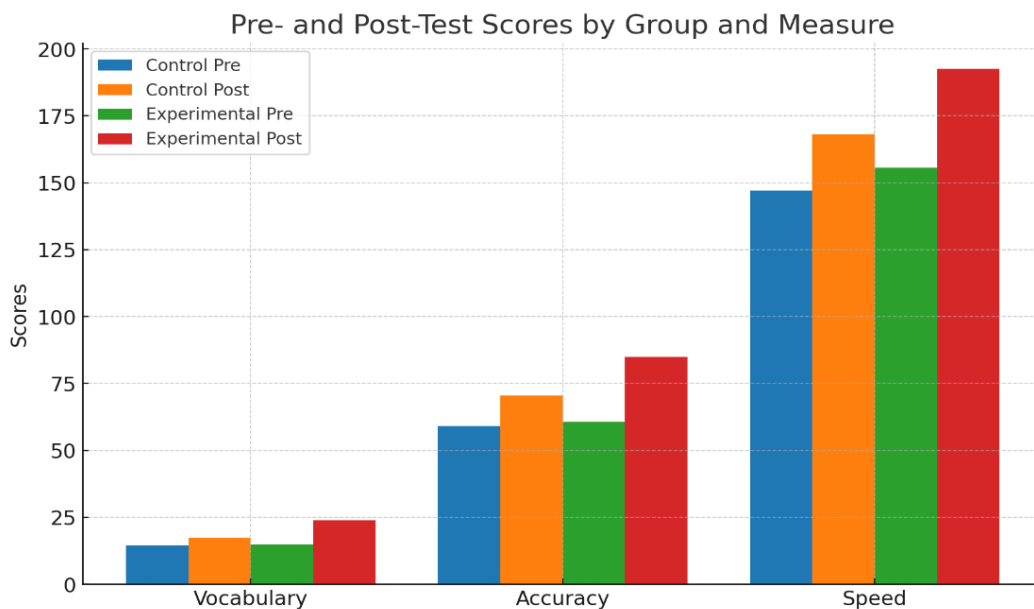


Figure 2: Pre- and Posttest Scores by Group and Measure

To assess between-group differences, ANCOVA was conducted on posttest scores, covarying pretest scores. Results are summarized in Table 2. Significant group effects were found for vocabulary ($F(1,57) = 210.59, p < .001, \eta_p^2 = 0.79$), accuracy ($F(1,57) = 83.06, p < .001, \eta_p^2 = 0.59$), and speed ($F(1,57) = 36.47, p < .001, \eta_p^2 = 0.39$), indicating superior outcomes for the

SSAWL group after controlling for baseline performance. Pretest scores were significant covariates in all models (all $p < .001$). Adjusted posttest means (controlling for pretest) were: vocabulary (experimental: 26.47, control: 17.53), accuracy (experimental: 89.12%, control: 64.88%), and speed (experimental: 184.03 WPM, control: 147.97 WPM).

Table 2: ANCOVA Results for Posttest Scores, Covarying Pretest Scores

Measure	Source	SS	df	F	p-value	ηp^2
Vocabulary	Group	561.92	1	210.59	<0.001	0.79
	Pretest	445.39	1	166.92	<0.001	
Accuracy	Group	2667.01	1	83.06	<0.001	0.59
	Pretest	3320.81	1	103.42	<0.001	
Speed	Group	3431.91	1	36.47	<0.001	0.39
	Pretest	24444.20	1	259.78	<0.001	

Cohen’s d for gain scores further quantified the magnitude of differences: vocabulary ($d = 3.85$), accuracy ($d = 2.09$), and speed ($d = 1.67$), all representing large effect sizes.

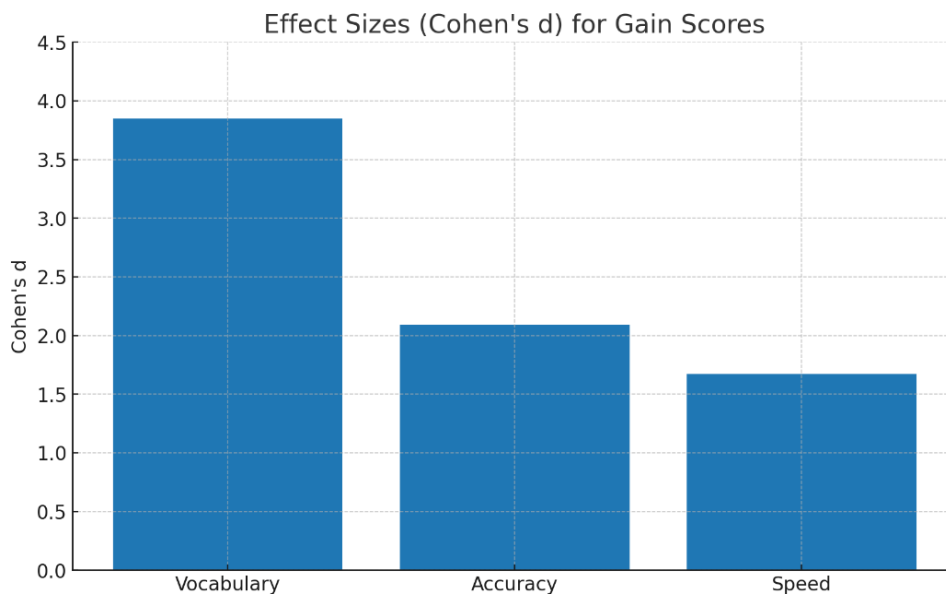


Figure 3: Effect sizes (Cohen’s d) for gain scores

Qualitative Findings

Thematic analysis of focus group interviews ($n = 15$ from the experimental group) and open-ended survey responses identified three key themes illustrating the perceived benefits of SSAWL instruction:

1. Enhanced Domain-Specific Understanding: Participants reported that learning SSAWL terms like "h-index" and "citation analysis" facilitated deeper comprehension of scientometrics texts. One student stated, "*Before, words like 'bibliometrics' confused me, but now I understand the concepts behind them.*"

2. Increased Confidence in Reading Specialized Texts: Many noted greater self-assurance when engaging with authentic materials. A common sentiment was, "*I feel more confident reading journal articles because I know the key vocabulary.*"

3. Improved Application in Contextual Tasks: Students highlighted the ability to apply new vocabulary in discussions and assignments. For instance, "*Using terms like 'altmetrics' in my essays makes my analysis more professional.*"

These themes were prevalent across responses, with inter-coder reliability $\kappa = 0.82$.

Table 3: Summary of qualitative themes (SSAWL focus groups and open responses, n = 15)

Theme	Frequency (of 15 participants)	Description	Exemplary (anonymized) quote
Enhanced Domain-Specific Understanding	12/15	Learners described better ability to map SSAWL items to real research tasks (e.g., identifying h-index usage).	"When we saw 'co-citation', I could connect it to what we learned and explain why it mattered in the study."
Increased Confidence in Reading Specialized Texts	10/15	Participants reported feeling more able to approach abstracts, methods, and results sections, and to infer meaning from context.	"I could read an abstract and understand it without constantly checking the glossary; I felt more confident in class discussions."
Improved Application in Contextual Tasks	7/15	Several students reported using SSAWL items in assignments and in oral presentations, indicating productive uptake.	"I used 'altmetrics' and 'citation network' in my presentation; the instructor commented on the precision of my language."

Integrated Findings

The quantitative results demonstrate the superior efficacy of SSAWL over general AWL in enhancing vocabulary knowledge, reading accuracy, and speed, with large effect sizes. Qualitative insights complement these findings by elucidating how domain-specific vocabulary fostered confidence and practical application, aligning with the principles of ESP pedagogy. Together, the data provide strong evidence for the value of tailored lexical resources in LIS education.

5. DISCUSSION AND CONCLUSION

The findings of this study provide compelling evidence for the superior efficacy of a Scientometrics-Specific Academic Word List over a general Academic Word List in enhancing reading comprehension and domain-specific vocabulary knowledge among EFL undergraduate Library and Information Science students. The experimental group, which received explicit instruction in the SSAWL through RAVE methods, demonstrated significantly greater improvements in vocabulary knowledge (M gain = +8.96), reading accuracy (M gain = +24.33%), and reading speed (M gain = +36.90 words per minute) compared to the control group, with large

effect sizes (Cohen's d ranging from 1.67 to 3.85, partial η^2 ranging from 0.39 to 0.79). These quantitative outcomes were corroborated by qualitative insights, where participants reported enhanced domain-specific understanding, increased confidence in engaging with specialized texts, and improved ability to apply SSAWL terms in contextual tasks. Such results affirm the study's primary hypothesis that domain-specific lexical resources better address the unique terminological demands of scientometrics, thereby facilitating more effective comprehension of disciplinary texts.

These results align closely with established literature on the limitations of general AWLs and the advantages of corpus-derived, discipline-specific word lists. As noted by Hyland and Tse (2007) and Dang et al. (2017), general AWLs like Coxhead's (2000) often fail to capture the specialized lexis essential for fields such as scientometrics, leading to inadequate coverage and comprehension barriers. The SSAWL's focus on high-frequency terms absent from general lists, such as "bibliometrics," "h-index," and "citation analysis", directly mitigated these gaps, resulting in superior lexical coverage and pedagogical relevance, consistent with findings from domain-specific lists in applied linguistics (Khani & Tazik, 2013), medicine (Tazik et al., 2022), and forestry (Luo et al., 2024). The large effect sizes observed here are consistent with those reported in similar interventions (e.g., Lawrence et al., 2021; McKeown et al., 2018), underscoring the value of tailoring vocabulary instruction to the dense technical jargon characteristic of scientometric discourse (Hood & Wilson, 2001; Inastrilla et al., 2024; Sedighi, 2016). Furthermore, the integration of RAVE strategies, which emphasize morphological analysis, contextual embedding, and spaced repetition, enhanced retention and transfer, echoing the benefits of explicit, multi-exposure approaches documented by Lesaux et al. (2010) and Whitaker (2024).

Qualitatively, the emergent themes of enhanced understanding, confidence, and application resonate with prior research highlighting the affective and practical dimensions of domain-specific vocabulary learning. For instance, participants' reports of reduced confusion with terms like "co-citation" and "altmetrics" mirror the improved disciplinary literacy and engagement noted in ESP studies (Wright & Cervetti, 2016). This increased self-assurance aligns with the lexical threshold hypothesis (Laufer, 2013; Nation, 2006), as greater familiarity with specialized lexis enabled learners to surpass the 95–98% coverage needed for independent comprehension, thereby reducing anxiety and fostering productive use in assignments and discussions (Binhomran & Altalhab, 2023). The triangulation of quantitative gains with these qualitative insights strengthens the argument that SSAWL instruction not only boosts measurable outcomes but also cultivates a more holistic disciplinary competence, particularly in English-medium instruction contexts (Chen et al., 2024; Xodabande et al., 2022).

The implications of these findings are multifaceted for ESP pedagogy and LIS curriculum design. Practically, the validated SSAWL offers a ready-to-use lexical resource that educators can integrate into courses, potentially through digital tools like Anki for personalized learning or concordancers for authentic exposure. This study advocates for a shift from reliance on general AWLs to customized interventions, providing a replicable model for other LIS subfields or disciplines with similar lexical demands (Argyroulis, 2022; Vuković-Stamatović & Živković, 2022). Theoretically, it contributes to the discourse on vocabulary's role in reading comprehension by demonstrating how domain-specific lists bridge the gap between breadth and depth of knowledge (Nation, 2006; Qian, 2002), thus informing broader frameworks for specialized literacy development (Chang, 2023; Masrai & Milton, 2018).

Despite these strengths, several limitations of the present study should be acknowledged. First, the sample was a convenience cohort of 60 undergraduate LIS students drawn from a single university in an EFL setting (Tehran), all at intermediate-to-advanced proficiency (CEFR B2–C1). Although stratified assignment and an a priori power analysis strengthened internal validity, the

demographic and institutional homogeneity of the sample constrains the generalizability of the findings to learners from different educational contexts, L1 backgrounds, proficiency levels, or institutions. Relatedly, the intervention was delivered within regular class time by a single trained instructor. While using the same instructor reduced variability between conditions, it raises the possibility of instructor-related or expectancy effects; subtle differences in enthusiasm, rapport, or delivery could have contributed to the observed advantage for the SSAWL group despite efforts to standardize materials and observe fidelity. Second, the study examined outcomes immediately after a 12-week intervention and therefore cannot speak to the durability of the observed gains. Large posttest improvements in vocabulary, accuracy, and speed are promising, but without delayed post testing (e.g., at three or six months), the extent to which these gains persist, consolidate, or decay remains unknown. Future research incorporating longitudinal follow-ups would clarify retention and the need for continued reinforcement to maintain disciplinary lexical competence. Third, although the SSAWL was developed using a large, carefully curated corpus and validated by expert raters, the decisions involved in list construction, journal selection (high-impact scientometrics titles, 2000–2024), frequency thresholds, range and dispersion cutoffs, and the exclusion of general service items, reflect a set of methodological choices that shape the final list. Different corpus parameters or inclusion criteria might yield slightly different item sets; thus, the SSAWL presented here should be understood as one empirically defensible operationalization of scientometrics-relevant lexis rather than the sole or definitive list. Similarly, the comprehension assessment relied on three authentic passages drawn from the same or similar journals; while this choice enhances ecological validity, it also raises the possibility of some teaching-to-test alignment between instructional texts and assessment materials. Fourth, the qualitative component, focus groups and open responses, provided valuable insight into learners' experiences but was limited to participants from the experimental group. Including comparative qualitative data from the control group would have offered a fuller account of differential affective and metacognitive responses to the two instructional approaches. Finally, although class fidelity was high (attendance >95%) and instruments were piloted, the controlled, instructor-led nature of the intervention may not mirror implementation at scale or in asynchronous, resource-constrained settings. Practical constraints such as teacher training, availability of corpus-informed materials, and institutional curriculum requirements could influence outcomes in wider applications. These limitations do not invalidate the study's central findings but delimit their scope. The strengths of the design—mixed methods, rigorous corpus development, validated instruments, and robust effect sizes—support the conclusion that SSAWL instruction can substantially improve scientometrics reading performance in the studied context. Nonetheless, replication with more diverse samples, delayed retention testing, comparative qualitative sampling, and alternative corpus construction strategies will be important next steps to confirm robustness and guide scalable implementation.

In conclusion, this mixed-methods quasi-experiment provides strong empirical support for the pedagogical value of a corpus-derived SSAWL taught through RAVE-consistent methods. Compared with matched general AWL instruction, SSAWL learners achieved substantially larger gains in domain-specific vocabulary knowledge, reading accuracy, and reading speed, and reported heightened confidence and productive use of scientometrics terms. While limitations, especially related to generalizability and long-term retention, suggest caution, the study offers a clear proof of concept: aligning lexical targets to disciplinary text demands and teaching them explicitly using evidence-based procedures meaningfully enhances disciplinary reading comprehension. ESP practitioners, curriculum designers, and LIS educators should consider integrating domain-specific word lists and corresponding instructional designs into courses that prepare students to read, interpret, and produce scientometric research.

Declaration

During the preparation of this work, the authors used ChatGPT to increase the readability of the manuscript and improve its writing quality. After using this tool, the authors reviewed and edited the content as needed and take full responsibility for the publication's content.

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Appendices

Appendix A: SSAWL Development Details

- *Software Settings*: AntConc (v4.3.1) used for corpus cleaning with default tokenization (whitespace-separated); AntWordProfiler (v2.2.1) with frequency threshold >28/million, range >50%, dispersion D>0.80, lemma list enabled.
 - *Word Family Criteria*: Based on Bauer & Nation Level 6 (headword + inflections/derivations, e.g., cite: citation, cited, citing, citable).

Appendix B: SSAWL (Word families ranked by frequency)

Words shared with the AWL are *italicized*, those in the first 1,000 GSL are in regular font, those in the second 1,000 GSL are **bold**, and non-GSL/non-AWL words are underlined.

1. <i>research</i>	11. <i>publication</i>	21. set	31. quality
2. <i>citation</i>	12. science	22. year	32. <i>academic</i>
3. <i>journal</i>	13. study	23. model	33. <i>evaluation</i>
4. number	14. <i>analysis</i>	24. high	34. different
5. <i>data</i>	15. result	25. social	35. example
6. paper	16. <i>index</i>	26. average	36. <i>network</i>
7. article	17. <u>web</u>	27. <i>indicator</i>	37. <u>collaboration</u>
8. <i>impact</i>	18. field	28. found	38. use
9. information	19. knowledge	29. country	39. value
10. <i>author</i>	20. level	30. total	40. university

- | | | | |
|--------------------------|---------------------|-----------------------|-------------------------|
| 41. term | 88. general | 135. overall | 182. activity |
| 42. document | 89. possible | 136. respectively | 183. question |
| 43. important | 90. main | 137. obtain | 184. support |
| 44. distribution | 91. influence | 138. communication | 185. collection |
| 45. case | 92. several | 139. policy | 186. engineering |
| 46. performance | 93. area | 140. part | 187. multiple |
| 47. method | 94. <u>scholar</u> | 141. percentage | 188. criteria |
| 48. related | 95. effect | 142. <u>keyword</u> | 189. learn |
| 49. review | 96. mean | 143. full | 190. point |
| 50. figure | 97. <u>peer</u> | 144. problem | 191. still |
| 51. <u>reference</u> | 98. sample | 145. pattern | 192. form |
| 52. factor | 99. rate | 146. assessment | 193. corresponding |
| 53. process | 100. development | 147. <u>cluster</u> | 194. associated |
| 54. subject | 101. list | 148. need | 195. organizations |
| 55. system | 102. classification | 149. page | 196. range |
| 56. literature | 103. link | 150. fact | 197. conference |
| 57. group | 104. low | 151. project | 198. evidence |
| 58. approach | 105. single | 152. <u>algorithm</u> | 199. production |
| 59. international | 106. increase | 153. context | 200. records |
| 60. similar | 107. available | 154. common | 201. instance |
| 61. significant | 108. particular | 155. <u>nodes</u> | 202. especial |
| 62. topic | 109. various | 156. future | 203. report |
| 63. period | 110. present | 157. domain | 204. <u>patents</u> |
| 64. discipline | 111. open | 158. positive | 205. frequency |
| 65. order | 112. provide | 159. calculate | 206. statistical |
| 66. <u>bibliometric</u> | 113. applied | 160. recent | 207. students |
| 67. institution | 114. library | 161. test | 208. nature |
| 68. individual | 115. include | 162. <u>dataset</u> | 209. <u>physics</u> |
| 69. since | 116. receive | 163. role | 210. items |
| 70. section | 117. growth | 164. core | 211. extent |
| 71. <u>database</u> | 118. productivity | 165. function | 212. sets |
| 72. funding | 119. size | 166. observe | 213. mention |
| 73. according | 120. previous | 167. current | 214. standard |
| 74. measure | 121. selected | 168. focus | 215. members |
| 75. specific | 122. degree | 169. feature | 216. concept |
| 76. self | 123. content | 170. public | 217. certain |
| 77. consider | 124. define | 171. expect | 218. conduct |
| 78. ranking | 125. source | 172. language | 219. innovation |
| 79. output | 126. world | 173. <u>query</u> | 220. <u>usage</u> |
| 80. count | 127. variable | 174. major | 221. contribution |
| 81. access | 128. share | 175. characteristic | 222. title |
| 82. <u>correlation</u> | 129. finding | 176. computer | 223. hence |
| 83. text | 130. <u>metric</u> | 177. potential | 224. <u>humanities</u> |
| 84. technology | 131. structure | 178. education | 225. interest |
| 85. compare | 132. identify | 179. account | 226. coverage |
| 86. type | 133. <u>score</u> | 180. change | 227. long |
| 87. category | 134. manage | 181. <u>online</u> | 228. law |

229. *aspect*
 230. *issue*
 231. *age*
 232. **key**
 233. **probability**
 234. *strong*
 235. *negative*
 236. *describe*
 237. *bibliographic*
 238. *known*
 239. *global*
 240. *traditional*
 241. *empirical*
 242. *book*
 243. *basic*
 244. *must*
 245. *framework*
 246. **discuss**
 247. *shape*
 248. *view*
 249. *proportion*
 250. **scale**
 251. *basis*
 252. *retrieve*
 253. *position*
 254. *normalize*
 255. *ratio*
 256. *theory*
 257. *limited*
 258. *represent*
 259. *decision*
 260. *name*
 261. *software*
 262. *take*
 263. *assign*
 264. *participant*
 265. *human*
 266. *survey*
 267. **original**
 268. *clear*
 269. *techniques*
 270. *life*
 271. *now*
 272. *electronic*
 273. *understanding*
 274. *seems*
 275. *become*
 276. *reason*
277. *equal*
 278. *task*
 279. *note*
 280. *people*
 281. *next*
 282. **weight**
 283. *usually*
 284. **practice**
 285. *help*
 286. *early*
 287. *regression*
 288. **attention**
 289. *independent*
 290. **medicine**
 291. *local*
 292. **frequently**
 293. *call*
 294. *matrix*
 295. *relevance*
 296. *furthermore*
 297. *difficult*
 298. *institute*
 299. *class*
 300. *determine*
 301. **health**
 302. **examine**
 303. **tend**
 304. *faculty*
 305. *power*
 306. *produce*
 307. *together*
 308. *step*
 309. *existing*
 310. *digital*
 311. *centrality*
 312. *involve*
 313. *graph*
 314. *quantitative*
 315. *around*
 316. *short*
 317. *regarding*
 318. *almost*
 319. *appear*
 320. *respondents*
321. *direct*
 322. *accuracy*
 323. *control*
 324. *address*
 325. *business*
 326. *able*
 327. *scientometrics*
 328. *necessary*
 329. *instead*
 330. *suggest*
 331. *trend*
 332. *industry*
 333. **tools**
 334. *internet*
 335. *purpose*
 336. *half*
 337. **behavior**
 338. *simple*
 339. *random*
 340. *team*
 341. *lack*
 342. *window*
 343. *bibliometrics*
 344. *abstract*
 345. *advantage*
 346. *state*
 347. *maximum*
 348. *perspective*
 349. *theoretical*
 350. *bias*
 351. *appropriate*
 352. *classified*
 353. *parameter*
 354. *complex*
 355. *external*
 356. *line*
 357. *unique*
 358. *contain*
 359. *mathematics*
 360. *end*
 361. *establish*
362. *gender*
 363. **reflect**
 364. *already*
 365. *quite*
 366. *yet*
 367. *contrast*
 368. *extract*
 369. *media*
 370. *technical*
 371. *biology*
 372. *complete*
 373. *career*
 374. *far*
 375. *later*
 376. *design*
 377. *real*
 378. *always*
 379. *expert*
 380. *length*
 381. *cross*
 382. *respect*
 383. **map**
 384. *effective*
 385. *economic*
 386. *space*
 387. *amount*
 388. *towards*
 389. *actual*
 390. *initial*
 391. **improve**
 392. *equation*
 393. *procedure*
 394. *wide*
 395. *site*
 396. *dimension*
 397. *lead*
 398. *environment*
 399. *strategy*
 400. *majority*
 401. *element*
 402. *diversity*

403. pair	444. reader	484. <i>physical</i>	525. <i>distinct</i>
404. training	445. <i>affect</i>	485. choice	526. <i>reveal</i>
405. distance	446. staff	486. popular	527. <i>underlying</i>
406. <u>multi</u>	447. aim	487. course	528. practical
407. <i>hypothesis</i>	448. <i>grant</i>	488. <i>qualitative</i>	529. <u>cumulative</u>
408. units	449. true	489. population	530. <u>prestige</u>
409. path	450. divide	490. <i>scope</i>	531. <u>recall</u>
410. place	451. choose	491. matching	532. <u>professor</u>
411. <i>generate</i>	452. <i>emerging</i>	492. property	533. beyond
412. introduce	453. <u>vector</u>	493. <i>consistent</i>	534. kind
413. <i>target</i>	454. society	494. <u>altmetrics</u>	535. remaining
414. <i>alternative</i>	455. <u>intellectual</u>	495. fit	536. rules
415. <i>proceeding</i>	456. detail	496. let	537. perhaps
416. personal	457. idea	497. <i>precision</i>	538. <i>phenomenon</i>
417. <i>editor</i>	458. <u>multidisciplinary</u>	498. strength	539. presence
418. history	y	499. formal	540. <u>affiliation</u>
419. <i>require</i>	459. <i>status</i>	500. proposal	541. foreign
420. write	460. <i>visibility</i>	501. entire	542. <i>primary</i>
421. <u>semantic</u>	461. sense	502. easily	543. <u>column</u>
422. <i>investigate</i>	462. along	503. except	544. sentences
423. <u>metadata</u>	463. carry	504. <i>oriented</i>	545. <i>regional</i>
424. <i>outcome</i>	464. companies	505. <u>cognitive</u>	546. differ
425. program	465. united	506. log	547. critical
426. little	466. material	507. <i>cooperation</i>	548. engines
427. <i>create</i>	467. <i>prior</i>	508. <i>professional</i>	549. <i>submit</i>
428. latter	468. experience	509. <i>prediction</i>	550. <i>credit</i>
429. stage	469. <i>sector</i>	510. <i>internal</i>	551. <i>conclusion</i>
430. <i>input</i>	470. <i>derive</i>	511. <u>percentile</u>	552. <i>achieve</i>
431. allow	471. <i>component</i>	512. explain	553. conditions
432. <i>evolution</i>	472. department	513. <i>series</i>	554. <i>domestic</i>
433. broad	473. success	514. effort	555. typically
434. variety	474. <i>version</i>	515. explore	556. free
435. service	475. <i>interaction</i>	516. reputation	557. <i>assumption</i>
436. <i>significance</i>	476. <i>comprehensive</i>	517. special	558. <u>mobility</u>
437. <i>despite</i>	477. close	518. <i>annual</i>	559. <u>repository</u>
438. government	478. dependent	519. slightly	560. <u>corpus</u>
439. combine	479. <i>entity</i>	520. <u>threshold</u>	561. accept
440. applicant	480. <i>scheme</i>	521. situation	562. <i>approximately</i>
441. <u>median</u>	481. date	522. <i>validity</i>	563. few
442. <i>objective</i>	482. <u>linear</u>	523. employ	564. news
443. indeed	483. experiment	524. <i>perceive</i>	565. connect

566. concern
567. play
568. post
569. substantial
570. agreement
571. right
572. actors
573. efficiency
574. *error*
575. essential
576. *consequent*
577. fraction
578. *locate*
579. *volume*
580. downloads
581. *goal*
582. *benefits*
583. *minimum*
584. answer
585. *occurrence*
586. *appendix*
587. *assume*
588. decrease
589. *estimate*
590. diffusion
591. profile
592. *panel*
593. *detection*
594. machine
595. adopted
596. *reliability*
597. *challenges*
598. side
599. *interpretation*
600. enough
601. *subsequent*
602. *response*
603. *colleagues*
604. mining
605. *summary*
606. solution
607. market
608. *exclude*
609. agency
610. geographical
611. *consist*
612. female
613. cost
614. hybrid
615. *transfer*
616. *formula*
617. *constant*
618. *phase*
619. *psychology*
620. arts
621. belong
622. turn
623. *generation*
624. curve
625. inclusion
626. teaching
627. ask
628. *dynamic*
629. come
630. private
631. *format*
632. risk
633. *mechanism*
634. excellence
635. temporal
636. versus
637. *overlap*
638. *considerable*
639. south
640. price
641. million
642. *implication*
643. interview
644. *integration*
645. *promotion*
646. former
647. description
648. expressed
649. firms
650. look
651. *financial*
652. *demonstrate*
653. *construct*
654. rest
655. *interval*
656. say
657. left
658. believe
659. print
660. candidate
661. gold
662. objects
663. background
664. confidence
665. gain
666. doctoral
667. matter
668. *fundamental*
669. separate
670. skewed
671. *stable*
672. *chemical*
673. *comment*
674. nearly
675. suitable
676. comparative
677. political
678. sentiment
679. *manually*
680. cell
681. *conclude*
682. density
683. *file*
684. *visualization*
685. extend
686. clinical
687. interface
688. *allocation*
689. overview
690. argue
691. reduce
692. recommendation
693. *expertise*
694. missing
695. biomedical
696. letter
697. put
698. *somewhat*
699. *obvious*
700. *implement*
701. *imply*
702. gap
703. joint
704. month
705. event
706. robust
707. skills
708. *attributes*
709. build
710. *seek*
711. attempt
712. capture
713. *insights*
714. manuscript
715. *analytical*
716. *sufficient*
717. avoid
718. beginning
719. *automatic*
720. *academia*
721. trust
722. *medium*
723. *ensure*
724. flow
725. tags
726. *partners*
727. depth
728. recognize
729. advanced

730. <u>disseminate</u>	771. opinion	812. <u>informetrics</u>	853. false
731. <i>energy</i>	772. family	813. <i>confirm</i>	854. <u>filter</u>
732. draw	773. hard	814. <i>capacity</i>	855. square
733. fixed	774. <i>aware</i>	815. direction	856. <u>genetic</u>
734. offer	775. <u>discourse</u>	816. <u>quartile</u>	857. intend
735. discovery	776. extensive	817. <u>readership</u>	858. face
736. <u>questionnaire</u>	777. move	818. progress	859. <i>utility</i>
737. <u>novel</u>	778. <i>participation</i>	819. <u>sociology</u>	860. run
738. <i>motivation</i>	779. rise	820. <u>biochemistry</u>	861. today
739. besides	780. male	821. <i>assistance</i>	862. <i>expansion</i>
740. <i>thesis</i>	781. distinguish	822. <i>principle</i>	863. <u>senior</u>
741. absolute	782. <u>plagiarism</u>	823. <u>prominent</u>	864. exercise
742. delay	783. <u>download</u>	824. <i>deviation</i>	865. <i>code</i>
743. young	784. <i>sequence</i>	825. edges	866. <i>legal</i>
744. board	785. capital	826. <i>shift</i>	867. <i>innovative</i>
745. weak	786. characterize	827. <i>crucial</i>	868. <i>classic</i>
746. commercial	787. <u>baseline</u>	828. <i>decline</i>	869. modern
747. <i>reject</i>	788. <i>hierarchical</i>	829. behind	870. <i>image</i>
748. correct	789. numerous	830. check	871. plan
749. exchange	790. alone	831. deal	872. balance
750. <i>decade</i>	791. <i>principal</i>	832. committee	873. rapid
751. <i>explicit</i>	792. attract	833. <u>phrases</u>	874. <u>plot</u>
752. picture	793. council	834. think	875. <u>feedback</u>
753. <i>theme</i>	794. light	835. <i>enhance</i>	876. translation
754. back	795. <i>denotes</i>	836. <u>subjective</u>	877. <u>latent</u>
755. opportunity	796. directed	837. <i>emphasis</i>	878. <u>credibility</u>
756. serve	797. reach	838. exception	879. regular
757. body	798. ties	839. manner	880. <u>subscription</u>
758. old	799. cause	840. <u>biological</u>	881. <u>axis</u>
759. <i>culture</i>	800. <u>molecule</u>	841. <i>edition</i>	882. <i>label</i>
760. <i>aggregate</i>	801. <i>evident</i>	842. <i>equivalent</i>	883. <i>automated</i>
761. <i>facilitate</i>	802. worth	843. <u>proxy</u>	884. <u>macro</u>
762. home	803. <i>incentives</i>	844. extreme	885. <u>prolific</u>
763. informal	804. <u>competitive</u>	845. <i>concentrate</i>	886. century
764. <i>illustrate</i>	805. continuous	846. <i>mode</i>	887. <i>enable</i>
765. <i>restricted</i>	806. <u>exponential</u>	847. <u>retract</u>	888. block
766. indirect	807. none	848. <i>identical</i>	889. <i>conventional</i>
767. <i>couple</i>	808. agriculture	849. instrument	890. <u>span</u>
768. <i>foundation</i>	809. <i>channel</i>	850. school	891. <i>validation</i>
769. <u>meta</u>	810. <i>remove</i>	851. <i>consistency</i>	892. return
770. middle	811. scarce	852. <i>consequence</i>	893. <i>notion</i>

894. referee
895. upper
 896. utilize
 897. union
 898. submission
 899. cancer
 900. college
 901. driven
 902. consensus
 903. blog
 904. operation
 905. venue
 906. ontology
 907. optimal
 908. session
 909. demand
 910. email
 911. heterogeneous
 912. chance
 913. dissertation
914. inventor
 915. keep
 916. spread
917. bottom
 918. platform
 919. duplicate
 920. fast
 921. plus
922. pure
923. exact
924. solve
 925. surprising
 926. neither
 927. earth
 928. sensitive
929. tail
 930. things
931. responsible
932. intention
 933. proximity
 934. wise
935. *intelligence*
 936. peak
937. root
 938. *adjusted*
 939. city
940. origin
 941. *specify*
942. net
943. stem
 944. *guideline*
945. prefer
 946. *apparent*
 947. *constitute*
 948. *contrary*
 949. micro
950. treatment
 951. *initiatives*
 952. claim
 953. proper
 954. *option*
 955. neural
956. abroad
957. artificial
 958. *intensity*
959. lot
 960. *job*
961. mixed
 962. *authority*
963. row
 964. award
965. double
966. boundary
 967. *dominant*
 968. binary
 969. entries
 970. meet
 971. *capabilities*
 972. engagement
 973. north
 974. press
 975. *infrastructure*
- 976. absence**
 977. *display*
 978. graduate
 979. judgment
 980. *debate*
 981. white
982. competition
 983. embedded
 984. biotechnology
 985. *minor*
 986. elite
 987. roughly
 988. *modified*
989. opposite
 990. barriers
 991. copyright
 992. day
 993. try
994. forward
 995. *chapter*
 996. *intrinsic*
 997. dependence
 998. *practitioners*
 999. really
 1000. fair
 1001. hyperlink
 1002. *parallel*
 1003. astronomy
 1004. *priority*
 1005. *style*
 1006. webometrics
 1007. *complementary*
 1008. favor
 1009. *highlight*
 1010. decide
 1011. *investment*
 1012. loss
1013. rarely
 1014. hot
1015. narrow
 1016. postdoc
1017. yield
1018. track
1019. overcome
 1020. pressure
 1021. entropy
 1022. simultaneous
 1023. *income*
 1024. prove
 1025. speed
 1026. *transform*
 1027. *undertaken*
 1028. fall
 1029. mind
1030. moderate
 1031. *virtual*
 1032. laboratory
1033. quickly
 1034. quantify
 1035. marked
 1036. headings
 1037. care
1038. ease
1039. repeat
 1040. outliers
1041. reward
 1042. cloud
1043. host
1044. compose
 1045. front
 1046. literacy
 1047. applicable
 1048. *coherence*
 1049. extension
 1050. geometric
 1051. color
1052. firm
 1053. *multidimensional*
 1054. predatory
 1055. supervise
 1056. *mutual*
 1057. verify

1058. <u>vocabulary</u>	1071. <u>fuzzy</u>	1084. disease	1097. master
1059. audience	1072. <i>norms</i>	1085. paid	1098. bring
1060. <i>maintain</i>	1073. <i>logic</i>	1086. <i>exclusive</i>	1099. <i>registered</i>
1061. <u>nanotechnology</u>	1074. apart	1087. <i>exhibit</i>	1100. ever
1062. <u>propensity</u>	1075. <u>tenure</u>	1088. formation	1101. <u>seminal</u>
1063. rich	1076. <i>transition</i>	1089. <i>cycle</i>	1102. <i>paradigm</i>
1064. <i>adequate</i>	1077. <i>corporate</i>	1090. <u>monograph</u>	1103. <u>webometric</u>
1065. <u>generic</u>	1078. <i>implicit</i>	1091. rights	1104. <u>mission</u>
1066. <u>huge</u>	1079. string	1092. <u>taxonomy</u>	1105. <u>sophisticated</u>
1067. numerical	1080. <i>simulation</i>	1093. division	
1068. promising	1081. <u>logistic</u>	1094. <i>emphasize</i>	
1069. <i>attitude</i>	1082. <u>magnitude</u>	1095. <i>inherent</i>	
1070. gathered	1083. <i>preliminary</i>	1096. <u>null</u>	