

Review Article

A Scoping Review of Text-Matching Software Used for Student Academic Integrity in Higher Education

K. Alix Hayden , **Sarah Elaine Eaton** , **Helen Pethrick** , **Katherine Crossman** ,
Bartłomiej A. Lenart , and **Lee-Ann Penaluna** 

University of Calgary, 2500 University Dr. NW, Calgary, Alberta T2N 1N4, Canada

Correspondence should be addressed to Sarah Elaine Eaton; seaton@ucalgary.ca

Received 13 July 2021; Revised 5 August 2021; Accepted 11 August 2021; Published 21 September 2021

Academic Editor: Balamuralithara Balakrishnan

Copyright © 2021 K. Alix Hayden et al. This is an open access article distributed under the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

Text-matching software has been used widely in higher education to reduce student plagiarism and support the development of students' writing skills. This scoping review provides insights into the extant literature relating to commercial text-matching software (TMS) (e.g., Turnitin) use in postsecondary institutions. Our primary research question was "How is text-matching software used in postsecondary contexts?" Using a scoping review method, we searched 14 databases to find peer-reviewed literature about the use of TMS among postsecondary students. In total, 129 articles were included in the final synthesis, which comprised of data extraction, quality appraisal, and the identification of exemplar articles. We highlight evidence about how TMS is used for teaching and learning purposes to support student success at the undergraduate and graduate levels.

1. Introduction

Plagiarism remains one of the most prominent types of academic misconduct across higher education. Commercially available text-matching software (TMS) has been adopted by many institutions in an effort to reduce violations and help students develop academic writing skills. Understanding the empirical evidence about how TMS is used to prevent violations of academic integrity and promote student learning can help institutions, educators, and policy makers make more informed decisions about how or if to use it.

2. Background: Text-Matching Software in Postsecondary Contexts

Although TMS is used in a variety of contexts including K-12 education, higher education, and scholarly publication, we delimited our study to examining the use of TMS in postsecondary contexts. We defined "postsecondary" as including universities and colleges, regardless of funding type (e.g., private or public) or the types of credentials they awarded (e.g., degrees, diplomas, and certificates). In the

subsections that follow, we elaborate on specific topics related to text-matching software in postsecondary contexts including (Section 2.1) plagiarism rates and the influence of the Internet; (Section 2.2) emergence of text-matching software; and (Section 2.3) previous reviews of TMS. This background shows gaps in the extant literature that led us to develop our research questions.

2.1. Plagiarism Rates and the Influence of the Internet. Plagiarism continues to be a pressing issue in higher education [1–3]. One particular aspect of academic integrity scholarship has centered around whether the Internet has resulted in an increase in plagiarism. Scholars are divided on the issue [4] with some arguing that there is a cause-and-effect relationship between advances in digital technologies and increases in academic misconduct [5, 6]. Others have found no empirical evidence to substantiate this claim [7, 8], instead showing that overall rates of plagiarism have decreased during the twenty-first century [2, 9]. Scholars agree that the emergence of sharing culture in the Internet era has led to changes in the ways in which information is shared, including how text can easily be copied and pasted from a

variety of online sources and file sharing is now commonplace among students [4, 10–12]. Hence, there is debate in the field about the ways in which the Internet has influenced plagiarism and other breaches of academic integrity, as well as interventions to address them. One such intervention is TMS.

2.2. Emergence of Text-Matching Software. TMS (also erroneously referred to as “anticheating,” “antiplagiarism,” “plagiarism detection,” or “plagiarism prevention” software) has existed in various forms for decades, though it started to become commercially available in the late 1990s with use expanding into the twenty-first century [5, 13]. The earliest mentions we found of TMS dated back to 1998 [14–16]. Since then, the use of TMS has proliferated and it is now used widely in some countries such as the UK and Australia, with use in other countries, such as Canada, being more limited [4, 17, 18]. TMS has garnered much attention in mainstream media, though stories in news media can sometimes demonstrate a particular viewpoint, either in favour or against the software [15, 16, 19–21]. In our study, we were concerned with what the scholarly literature presented.

2.3. Previous Reviews of TMS. Previous reviews of TMS include evaluations of the tools themselves [22, 23] and scholarly and systematic reviews [24]. Our study builds on previously published reviews by including studies published up to and including February 2021, thus offering the most current findings.

2.4. Review Questions (RQs). We began with a broad general research question, followed by two subquestions, both related to different aspects of the effectiveness of TMS software for (a) reducing academic misconduct and (b) increasing student learning as an educational intervention. The research team developed the research questions through collaborative and iterative team dialogue. The research questions were published in our protocol and remained constant through the duration of the project [25].

Primary RQ: how is text-matching software used in postsecondary contexts?

- (i) Sub-RQ1: what is the effectiveness of such software in reducing incidences of plagiarism?
- (ii) Sub-RQ2: what is the effectiveness of such software as an educational intervention?

We were particularly concerned with understanding the quality of the extant literature, as well as the contents of the studies themselves. This focus on quality is reflected in Section 3.

3. Methods

3.1. Design. Prior to commencing our review, we developed a detailed protocol, which was subsequently published [25]. The protocol guided our research, and any deviations are

reported. The review is reported in accordance with the PRISMA 2020 reporting guidelines [26].

We initially planned to conduct a systematic review; however, we discovered that the literature on TMS is diversified and heterogenous. Our research subquestions which focused on effectiveness could not be addressed because of this heterogeneity. Therefore, we decided to pivot to a scoping review approach. Scoping reviews are appropriate when a research team wants to “determine the scope or coverage of a body of literature on a given topic and give clear indication of the volume of literature and studies available as well as an overview (broad or detailed) of its focus” ([27], p. 2). Furthermore, the aim of scoping reviews is to map the literature, which responds to our primary research question.

The research team comprised six members including two academic librarians, one of whom had extensive experience with systematic and scoping reviews. Four academic integrity subject-matter experts collaborated with the librarians. The review took place from October 2018 to May 2021, with the protocol published in the first year of the project [25].

3.2. Eligibility: Inclusion and Exclusion Criteria. Our eligibility (i.e., inclusion and exclusion) criteria were informed by the Johanna Briggs Institute’s mnemonic PICO: Population, Phenomenon of Interest, Context, Outcomes [28]. The research team engaged in extensive and detailed dialogue to meticulously consider each and every inclusion and exclusion criterion.

We delimited our review to the population focused on postsecondary students including undergraduates and graduates (master’s and doctoral). Studies that focused *only* on faculty perspectives or experiences were excluded; however, articles that incorporated faculty activities to educate students or reduce plagiarism were included. Our rationale for this is that we were inherently interested in how TMS is used with respect to students. Therefore, we determined that studies would need to have at least some indication that the student experience was considered.

Studies focused on the K-12 population were excluded. The context was postsecondary education, which we defined as “universities, community college, trade, and vocational training centres” [29]. Due to the exploratory nature of scoping reviews, we did not identify specific outcomes associated with using TMS in a higher education environment.

The phenomenon of interest included commercially available TMS. As such, we excluded studies that focused on independent or individual coders who wrote their own software as an experiment or open access software, as these are not generally licensed at scale by educational institutions.

We did not exclude studies based on study or publication design. We included quantitative and qualitative studies, as well as text and opinion articles, legal overviews, and theoretical papers. However, we excluded popular media including blogs and social media. We also excluded promotional materials and conference abstracts. There were no restrictions on publication date or geographical location. All included articles needed to be available in English.

3.3. Sources and Search Strategy. Our search and information sources were exhaustive. The search was developed by two librarians (KAH and BL) and pilot tested against known relevant studies. For the education-related databases, only the concept of text matching software was searched. For all the other databases, the search focused on two main concepts: text matching software and postsecondary education. Each concept included both keywords and subject headings. Keywords were constant across all databases. Subject headings were determined by a database's controlled vocabulary. The search was developed in ERIC and then translated for other databases (see Table 1). All search strategies were completed as per PRISMA 2020 guidelines and saved on an open access database [30]. Searches are available in September 2019 and rerun again in February 2021 to update the results.

Both discipline-focused and interdisciplinary databases were searched. Databases to search were first identified by running a quick search of "text matching software" or "Turnitin" to determine whether the topic was covered by the database. We searched 14 relevant databases (Table 2). All were searched from the database inception until September 13/14, 2019, and updated on February 10, 2021.

Given the large number of studies included and the heterogeneity of the included studies, we did not search the grey literature, as originally planned nor did we complete a reference list scan or forward citation tracking because of feasibility concerns. These are considered limitations of our review.

3.4. Source Selection. Search results were imported into Covidence for screening and were automatically deduplicated. Covidence deduplicates based on title, year, and volume which need to be exact matches, and author names are similar [31].

Prior to screening, an interrater calibration exercise was conducted with all screeners. Randomly generated 100 titles/abstracts were screened independently by each reviewer, noting reasons for "No" and "Maybe" decisions. The screeners then met to clarify the inclusion/exclusion criteria and came to consensus on any disagreements. Screening occurred in two phases. The first phase included title/abstract screening. Records selected as potentially relevant were moved to the second phase, which involved reviewing the full-text articles. All screening was conducted independently, and all records and full-text articles were screened by two reviewers. Disagreements were resolved through discussion.

3.5. Data Extraction. All authors contributed to the data extraction process. Prior to data extraction, a calibration exercise was conducted with three studies to ensure that the data extraction template included all required elements and that each researcher was extracting data in the same way and understood the process. We extracted data from each included article under the following categories: purpose/research question, country where the study took place (if applicable), population (if applicable), text-matching software used, study design, methodology details, key findings,

themes (for qualitative study designs), and limitations. Data were also extracted with regard to the focus of the use of the text-matching software in the article (i.e., whether it was an educational intervention or used for punitive purposes).

3.6. Quality Appraisal. Four authors (BL, HP, KC, and LAP) completed quality appraisal. Due to the diversity of study designs in the field of educational research, we used two critical appraisal tools to assess the quality of the included studies. The Mixed Methods Appraisal Tool (MMAT) provided assessment criteria for empirical mixed methods and qualitative and quantitative study designs [32]. We used the Joanna Briggs Institute (JBI) Checklist for Text and Opinion to assess articles that were theoretical, program descriptions, or opinion pieces [33]. The MMAT produced a score out of seven items for each empirical article, and the JBI checklist produced a score out of six. Once scored, the articles were categorized as high, medium, or low quality. The MMAT guidelines do not specify a particular ranking system for article quality, but rather request that authors be transparent about how the scores were interpreted [32]. High-quality articles were scored five–seven with the MMAT tool or six–seven with the JBI checklist; medium-quality articles were scored three–four (both MMAT and JBI); and low-quality articles were scored zero–two (both MMAT and JBI). Articles were not excluded based on their quality scores.

3.7. Data Synthesis. Two authors (KAH and HP) compiled the data extraction and quality appraisal from the rest of the research team for data synthesis. We summarized the data in a series of descriptive tables and an accompanying narrative synthesis. Due to the large number of included articles and the heterogeneity of these studies, it was not possible to include every article in the narrative synthesis. Through group consensus, the authors identified a series of exemplar articles that scored high in the quality appraisal process and had notable methodologies or findings. These were the articles that were emphasized in the narrative synthesis.

4. Results

Figure 1 shows the flow of articles through the searching and screening process. The original search (September 2019) resulted in 7,352 records, of which 3,403 were duplicates. Titles and abstracts of 3,949 were screened, resulting in 363 full-text articles reviewed for eligibility, of which 116 met the inclusion criteria. The updated search (February 2021) identified an additional 1050 records, with 391 duplicates. After screening 659 titles/abstracts, 37 studies' full texts were reviewed and an additional 13 articles met the inclusion criteria. In total, 129 articles were included in the final analysis [5, 17, 18, 22, 33–155].

4.1. Study Characteristics. As shown in Table 3, most of the articles were from English-speaking countries, with 33 countries representing overall. The most common country where empirical studies took place was the United States of America ($n = 32$, 24.8%), followed by the United Kingdom

TABLE 1: ERIC (Ebsco) search strategy.

#	Query
S1	SU "Computer Software"
S2	SU "Computer Uses in Education"
S3	SU "Information Technology"
S4	S1 OR S2 OR S3
S5	SU "Plagiarism"
S6	SU "Cheating"
S7	S5 OR S6
S8	S4 AND S7
S9	(plagiarism or cheating) N3 detect*
S10	(plagiarism or cheating) N3 (software or tool* or program* or computer* or online or Internet or product*)
S11	anti-plagiarism N3 (software or tool* or program* or computer* or online or Internet or product*)
S12	antiplagiarism N3 (software or tool* or program* or computer* or online or Internet or product*)
S13	"text match*" N5 (software or tool* or program* or computer* or online or Internet or product*)
S14	(Turnitin* or iThenticate or SafeAssign or CrossCheck or Copyscape or CopyCatch or Urkund)
S15	S9 OR S10 OR S11 OR S12 OR S13 OR S14
S16	S8 OR S15

TABLE 2: Databases searched.

Database	Interface	Discipline
ERIC	EBSCO	Education
Education Research Complete	EBSCO	Education
Library and Information Science Source	EBSCO	Library science
Academic Search Complete	EBSCO	Interdisciplinary
CINAHL Plus with Full Text	EBSCO	Nursing and allied health
Business Source Complete	EBSCO	Business
International Bibliography of Social Sciences	ProQuest	Social sciences
Library and Information Science Abstracts	ProQuest	Library science
ABI/Business Premium Collection	ProQuest	Business
Compendex (Engineering Village)	Elsevier	Engineering/computer science
Scopus	Elsevier	Interdisciplinary
MEDLINE(R) and Epub Ahead of Print, In-Process, In-Data-review and Other Nonindexed Citations and Daily	OID	Medicine/health sciences
APA PscyInfo	OID	Psychology
Web of Science Core Collection	Clarivate	Interdisciplinary

($n = 19$, 14.7%) and Australia ($n = 11$, 8.5%). The publication dates of the articles ranged from 2004 to 2020 (Table 4), aligning with the release and uptake of commercially available text-matching software in higher education. The most common study designs (Table 5) were mixed methods ($n = 38$, 29.5%), quantitative descriptive ($n = 27$, 20.9%), and quantitative nonrandomized ($n = 25$).

Turnitin was the most frequently used text-matching software ($n = 111$, 86.0%) by far; the next most frequent software was SafeAssign ($n = 9$, 7.0%) and Urkund ($n = 3$, 2.3%). Other software used was Ephorus, EVE2, Grademark, JISC Plagiarism Detection Service, Scriptum, WCopyfind, iThenticate, MyDropBox, and Veriguide (all present once or twice in the included articles). Generic software or other programs accounted for 14 (10.9%) uses across the included articles. A total of 11 (8.5%) articles featured more than one type of software.

The study population is shown in Table 6. Some included articles ($n = 24$, 18.6%) did not have a study population because they were opinion articles. Of those articles with a study population, the majority included only postsecondary

students ($n = 84$, 65.1%). An additional 21 studies (16.3%) included both students and faculty/instructors. We were unable to report on students' year of study or status (undergraduate or graduate), as well as discipline, because many articles included multiple populations or the reporting was unclear.

Not surprisingly, articles focused specifically on student work were most frequent ($n = 47$, 36.4%). These articles utilized student papers, essays, or written assignments as the focus of their study. Articles in this category investigated ways to detect, compare, measure, deter, or minimize plagiarism; influence of TMS on student work; prevalence or incidence of plagiarism in student essays; positive and negative aspects of using TMS on student work; using TMS to improve student writing; and reliability, functionality, and accuracy of TMS.

Some studies ($n = 17$, 13.2%) investigated student perspectives of text-matching software. These articles explored reasons why students plagiarize, as well as students' understanding and awareness of, and attitudes towards TMS and plagiarism. Articles in this category also addressed

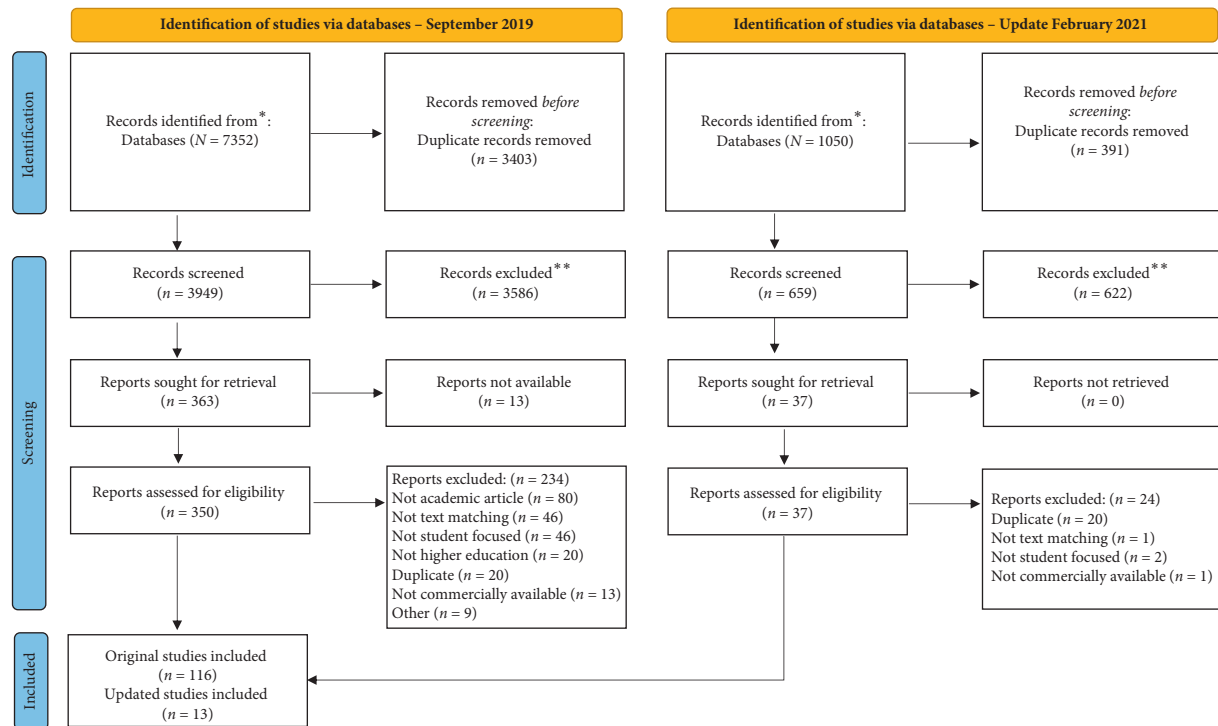


FIGURE 1: PRISMA Diagram. The flow of articles through the screening and searching process is shown, using the Preferred Reporting Items for Systematic Reviews and Meta-Analyses (PRISMA) 2020 guidelines [26].

TABLE 3: Study characteristics: country of study.

Country	n	%
Australia	11	8.5
Botswana	2	1.6
Canada	4	3.1
China	2	1.6
Croatia	1	0.8
Cyprus	1	0.8
Finland	1	0.8
Germany	1	0.8
India	2	1.6
Indonesia	1	0.8
Iraq	1	0.8
Ireland	1	0.8
Japan	2	1.6
Latvia	1	0.8
Malaysia	3	2.3
Mexico	1	0.8
New Zealand	5	3.9
Oman	1	0.8
Pakistan	2	1.6
Peru	1	0.8
Philippines	1	0.8
Qatar	1	0.8
Saudi Arabia	1	0.8
South Africa	1	0.8
Spain	2	1.6
Taiwan	1	0.8
Turkey	3	2.3
United Arab Emirates	1	0.8
United Kingdom	19	14.7
United States of America	32	24.8

TABLE 3: Continued.

Country	<i>n</i>	%
Vietnam	1	0.8
Not Stated	2	1.6
Not relevant (i.e., opinion article)	20	15.5
Total	129	100

TABLE 4: Study characteristics: year of publication.

Year	<i>n</i>	%
2004	3	2.3
2005	7	5.4
2006	5	3.9
2007	3	2.3
2008	8	6.2
2009	12	9.3
2010	11	8.5
2011	6	4.7
2012	5	3.9
2013	15	11.6
2014	6	4.7
2015	5	3.9
2016	8	6.2
2017	6	4.7
2018	9	7.0
2019	13	10.1
2020	7	5.4
Total	129	100

TABLE 5: Study characteristics: study design.

Study design	<i>n</i>	%
Mixed Methods	38	29.5
Opinion	18	14.0
Program Description	2	1.6
Qualitative	13	10.1
Quantitative Descriptive	27	20.9
Quantitative Nonrandomized	25	19.4
Quantitative randomized	2	1.6
Theoretical	4	3.1
Total	129	100

TABLE 6: Study population.

Population	Category	<i>n</i>	%
Students only	Student perspectives	17	13.2
	Student work	47	36.4
	Student work and student perspectives	20	15.5
Students and instructors	Student and instructor perspectives	15	11.6
	Student work and instructor perspectives	5	3.9
	Student work and student/instructor perspectives	1	0.8
Not relevant	Opinion articles	24	18.6
Total		129	100

students' perspectives of learning when taught about plagiarism and text-matching programs. Interestingly, some studies ($n = 20$, 15.5%) included both student work and student perspectives. Most often, these articles provided instructional sessions on academic integrity or TMS, assessed students' work with TMS, and additionally included student perspectives of TMS.

As mentioned, 21 studies included both students and instructors. The majority focused on the perspectives (detailed above) of both students and instructors with respect to TMS ($n = 15$, 11.6%). Instructor perspectives of student work assessed by TMS were addressed by a small number of studies ($n = 5$, 3.9%). These articles focused on problems and benefits of using TMS on student work, reducing instructor

workload/marking with TMS, and using TMS to enhance learning. One article investigated both student and instructor perspectives on student work which was analyzed by a TMS program.

Table 7 shows the uses of the text-matching software included as an educational intervention ($n = 50$, 38.8%), a punitive measure ($n = 51$, 39.5%), both or other/legal/ethical ($n = 17$, 13.2%), or was unclear or not relevant ($n = 11$, 8.5%). Educational interventions included studies focused on teaching students about plagiarism and/or text-matching software. Student work was often assessed by TMS, and then, feedback was provided to students on how to avoid plagiarism or strategies to write more effectively. Often, the TMS was presented in a positive light and as a support for student learning, and students were provided with the opportunity to revise work after submission. Punitive measures included those articles where no remediation or learning opportunities were provided to students. Often, these studies focused on detection of plagiarism through the use of TMS. Examples of articles where TMS was coded as “Other” included those that investigated advantages, disadvantages, or various uses of TMS, discussed legal or ethical implications of TMS use in postsecondary education, or questioned if TMS was punitive or educational.

4.2. Quality Appraisal. Table 8 shows the quality scores of the included articles, which varied from high ($n = 72$, 55.8%) to medium ($n = 30$, 23.3%) to low ($n = 27$, 20.9%). Of the different study designs, opinion articles had the highest proportion of high-quality scores ($n = 13$, 72.2%). There was a large number and proportion of mixed-method articles ($n = 11$, 28.9%) and quantitative descriptive articles ($n = 8$, 29.6%) with low-quality scores. The most common cause of low-quality scores was the lack of a clear research question, which was scored as “No” or “Can’t Tell” in 57 of the 106 empirical articles (i.e., mixed-method, qualitative, and quantitative articles).

4.3. Exemplars of Excellence. Using our quality score as a guide, we selected four studies, each using a different methodology, to highlight as an exemplar of excellence. The selection of the studies was led by one author (SEE) and confirmed by the others. None of the exemplars included all of the high-quality indicators for their study type, but nevertheless provide useful models for researchers with particular methodological expertise.

4.3.1. Exemplar #1: Quantitative Descriptive Study. The exemplar we selected in the category of quantitative descriptive studies was Dawson et al.’s study, “Can Software Improve Marker Accuracy at Detecting Contract Cheating? A Pilot Study of the Turnitin Authorship Investigate Alpha” [67]. In this study, “twenty-four experienced markers from five units of study were asked to make decisions about the presence of contract cheating in bundles of 20 student assignments, which included 14 legitimate assignments and six purchased from contract cheating sites” ([67], p. 473). The markers were representative of the disciplines of world

history ($n = 4$); cells and genes ($n = 5$); personality ($n = 5$); psychopathology ($n = 5$); and nutritional physiology ($n = 5$). Details were also provided about the number of students enrolled in each course, along with the number who participated in the study.

The method and analysis were explained in sufficient detail that we determined the study could be replicated either exactly or closely based on the published article. Although the paper did not include a section explicitly addressing the limitations of the study, there was some indirect discussion related to general limitations of detection of contract cheating.

4.3.2. Exemplar #2: Qualitative Study. We selected Rees and Emerson’s study “The impact that Turnitin® has had on text-based assessment practice” as our exemplar of excellence of a qualitative study [131]. Rees and Emerson conducted structured interviews with staff ($n = 9$) who had been trained to use Turnitin®, understood its capabilities, and were “high-volume” ([131], p. 4) users of the software. The authors analyzed case studies of courses at Massey University in New Zealand, where Turnitin® was available to all staff (including faculty) and its use was mandated in some programmes.

Rees and Emerson included a copy of their structured interview questions in their article and included participant quotations in their results. With regards to the two case studies of programmes, Rees and Emerson selected one program in nursing and another in communication in the sciences, presenting a robust discussion of the use of text-matching software in the selected programs.

The authors concluded by noting that although there were opportunities for faculty to adjust their assessment approaches using text-matching software to do so, they remained more likely to view the software as a tool for detecting plagiarism. The authors contemplated why faculty change regarding assessment has not occurred and called for further research.

Our evaluation of this study was that it presented a comprehensive institutional case study that others might be able to replicate. This was due, in part, to the level of detail provided in the methods and analysis sections of the study.

4.3.3. Exemplar #3: Mixed-Method Study. The exemplar of excellence we select for this category was the study of Kaktiš, “Does Turnitin support the development of international students’ academic integrity?” [96]. Kaktiš conducted a case study at a commercial educational institution offering higher education pathway programs for international students. Quantitative data were collected using a student survey, supplemented by two types of qualitative data collected using (a) a focus group with students and (b) interviews with teachers. Survey respondents ($n = 260$) represented an 89% response rate. The focus group included students ($n = 5$) from different countries to allow for representation of the student body at the school. Teachers ($n = 12$) were interviewed, with details included about their

TABLE 7: Use of TMS.

Use	<i>n</i>	%
Educational intervention	50	38.8
Punitive	51	39.5
Both and/or other/legal/ethical	17	13.2
Unclear/not relevant	11	8.5
Total	129	100

TABLE 8: Quality score by study design.

Study design	Quality score						Total
	<i>n</i> high	% high	<i>n</i> medium	% medium	<i>n</i> low	% low	
Mixed methods	20	52.6	7	18.4	11	28.9	38
Opinion	13	72.2	5	27.8	0	0.0	18
Program description	0	0.0	2	100.0	0	0.0	2
Qualitative	6	46.2	4	30.8	3	23.1	13
Quantitative descriptive	15	55.6	4	14.8	8	29.6	27
Quantitative nonrandomized	14	56.0	7	28.0	4	16.0	25
Quantitative randomized	0	0.0	1	50.0	1	50.0	2
Theoretical	4	100.0	0	0.0	0	0.0	4
Grand total	72	55.8	30	23.3	27	20.9	129

length of time teaching at the school and other demographic specifications.

The quality and depth of analysis, as well as the clear writing style of the author, allowed the reader to easily understand how the study was conducted and how results were analyzed. The discussion was balanced in that it explored both positive and negative aspects of using text-matching software. In addition, the author discussed limitations of the study and offered directions for future research. Kaktiņš concluded with a call for higher-education institutions:

To recognise such software as part of a suite of strategies within a broader context of developing academic integrity, in spirit as well as practice. To do otherwise would encourage a very narrow and potentially counterproductive attitude that could stunt the development of both students' critical thinking and their in-depth understanding of academic integrity ([96], p. 445).

Overall, this article presented an evidence-informed and balanced study that demonstrated methodological rigour.

4.3.4. Exemplar #4: Text and Opinion Study. Studies falling into this category include “text- and opinion-based evidence (which may also be referred to as nonresearch evidence) which is drawn from expert opinions, consensus, current discourse, comments, assumptions, or assertions that appear in various journals, magazines, monographs, and reports” [28]. We note that the JBI critical appraisal checklists were originally developed for use in medical and health sciences and as such, this definition is an imperfect fit with studies conducted in other disciplines such as the humanities, social sciences, or law. However, owing to the absence of any other protocol designed for use in other disciplines, the research team used the JBI definition to guide our work. Studies that fell into this category generally resulted in close scrutiny by

at least two researchers, often with extensive discussion regarding how to extract data and assess quality.

We selected “When Students Won’t Turnitin: An Examination of the Use of Plagiarism Prevention Services in Canada” [141] as our exemplar of excellence in this category. Though dated, this study presented a comprehensive analysis of a legal considerations related to the use of a commercial text-matching software. Strawczynski documented and analyzed a precedent-setting legal case in Canada (Rosenfeld v. McGill University) in which a student took the university to court over the use of text-matching software in the 2003-2004 academic year. Strawczynski provided an in-depth analysis of copyright law in Canada and considerations of the Canadian Charter of Rights and Freedoms that led the courts to find in favour of the student.

The case not only resulted in McGill University suspending its use of the software but also set legal precedent, ultimately resulting in more limited use of text-matching software across Canada compared with other countries. Strawczynski’s article was comprehensive in its treatment of the case and legal underpinnings that resulted in the court’s decision.

5. Discussion

5.1. Methodological Decision-Making Process regarding the Use of Systematic and Scoping Review Tools. Our study resulted in a large and nonhomogeneous dataset, necessitating the use of multiple tools to analyze the selected studies. It is unusual in a systematic review to use different tools to analyze the data, but the researchers determined that neither the Mixed Methods Appraisal Tool (MMAT) nor the Joanna Briggs Institute (JBI) checklist were appropriate for use on all the studies included. The methodological decision-making process to determine which systematic review tool would be most appropriate for each study was an ongoing

and iterative process throughout the project. We engaged in this decision-making process using both individual researcher evaluation of each study for possible fit with each tool, supplemented by collaborative dialogue at research team meetings.

5.2. Results As Related to the Review Questions. With regards to our review questions, we were able to answer the primary RQ: how is text-matching software used in post-secondary contexts? However, we were ultimately unable to arrive at satisfactory answers to our two subquestions:

- (i) Sub-RQ1: what is the effectiveness of such software in reducing incidences of plagiarism?

Based on the 129 studies we reviewed, the evidence about the effectiveness of text-matching tools to reduce incidences of plagiarism was inconclusive. There was insufficient evidence to support cause-and-effect claims that text-matching software by itself is effective in reducing plagiarism. We noted that several studies [50, 72, 96] advocated for the use of text-matching software as one tool used as part of a comprehensive instructional and institutional academic integrity strategy.

- (ii) Sub-RQ2: what is the effectiveness of such software as an educational intervention?

Numerous studies advocated for the use of text-matching software as an educational intervention [34, 35, 40, 42, 47, 48, 55], but with few exceptions [34, 55], and almost none of these studies included both control and experimental groups. As such, it was difficult to determine with certainty to what extent text-matching software is effective as an educational intervention. We can speculate that individual and institutional approaches to teaching and assessment may play a role in this regard, with pedagogical approaches varying among individual educators, in particular in jurisdictions where instructional staff have high levels of pedagogical autonomy and academic freedom with regards to teaching and assessment.

Another factor we considered was the sample size of the population under study. We noted that some studies that focused on using text-matching software as an educational intervention included author-identified limitations that included a small sample size [42, 54, 64, 75, 102, 119, 142]. The nature of the studies involving teaching and learning interventions varied in their scope and approach, and we found no evidence of educational intervention studies that had replicated previous studies, further limiting our ability to definitively determine the effectiveness of TMS as an educational intervention.

The research team consistently referred back to the review questions, as published in our protocol, making every effort to answer them. In the end, we recognized that we were unable to provide conclusive answers to these questions, and in the interest of research integrity, we determined

that it was important to address this. We contend that our inability to answer our original research questions was not due to deficiencies in the abilities of the research team, but rather to the nature of the body of literature that exists in the field. As discussed in the methods, we pivoted to a scoping review approach because of the lack of heterogeneity of the included studies. Our scoping review provides a map of the evidence and literature focused on the use of text matching software with the international postsecondary context.

5.3. Plagiarism Prevention versus Detection. Through this review we noted that TMS is used both for teaching and learning purposes (i.e., prevention) and as a plagiarism detection tool. When utilized for teaching and learning, TMS can be used as a mechanism to provide formative feedback to students to help them improve their academic writing skills, along with citing and referencing skills. We were not able to determine through this study if TMS is actually an effective intervention in terms of reducing academic misconduct. One reason for this is that the motivations for academic misconduct can be complex, varying from one individual and one case, to the next. Previous research has shown that it can be more appropriate to consider academic risk factors for students that might result in academic misconduct and consider the possibility of multiple compounded risk factors for some students [18, 157], as opposed to trying to establish a reductionist or linear cause-and-effect relationship between one factor (e.g., the use of TMS) and increases or decreases in academic misconduct. Our study supports the notion that TMS should not be used or viewed as a single solution to the prevention of plagiarism, but instead, it can be used as one aspect of a comprehensive institutional approach to establishing a culture of academic integrity.

5.4. Quality of TMS Research. Our study showed that just over half of the quantitative studies included in our review were determined to be high quality, and fewer than half of qualitative studies were identified as being high quality. This points to the need for more scholarly rigour in studies on TMS in order to better inform the field of academic integrity research, as well as policy and technology licensing decisions at higher education institutions. This is consistent with previous studies that have shown that academic integrity is a less mature field of study than other areas of educational research such as assessment [158].

It would be reasonable to expect that the quality of available research will continue to improve as the field of academic integrity research continues to mature.

5.5. Strengths and Limitations. Although the research team members held varying levels of proficiency in numerous languages other than English, we made the methodological decision to only include studies that were published in English. It is possible that we missed some studies published in other languages.

As with any scoping or systematic review, we tightly controlled our search terms in order to focus our review. We also made the methodological decision to exclude the grey literature. As such, our review, while comprehensive, may not be exhaustive of all available scholarships.

This was a methodologically complex study to undertake, as it spanned multiple disciplines, which necessitated the use of both the Mixed Methods Appraisal Tool (MMAT) and the JBI Checklist. The research team engaged in iterative and ongoing dialogue throughout the project, with quality assurance, internal peer review, and interrater reliability being consistent themes during the research process.

Although we began our project with the intention of undertaking a systematic review, we recognized through our analysis of the results that we were unable to adhere to the accepted standards of systematic reviews. We noted a couple of examples of published studies that have claimed to be systematic reviews of topics related to academic integrity [159, 160] but in our estimation, did not necessarily meet the rigorous criteria to be classified as such. It is not our intention to disparage the scholarly efforts of others who endeavour to contribute to research in this field of academic integrity, but rather to emphasize the need to pursue quality in research, above all else. To that end, and with the goal of upholding scholarly rigour and ensuring integrity in the reporting of our own results, we carefully considered our analysis and determined that, in the end, our study would more accurately be described as a scoping review, rather than a systematic one.

6. Recommendations for Future Research

As a result of this study, we recommend that independent scholarly research into text-matching software continue, as the field continues to evolve. We identified a general lack of longitudinal studies or those that replicated methods of previous studies, pointing to the need for more consistency in the ways in which studies about TMS are conducted.

We have further noted that numerous studies included small sample sizes, with studies often being conducted at a single university or in a single country. There is a need for multi-institutional, multicountry research, as well as longitudinal research and studies that replicate previous research. This would help to contribute to the overall rigour and quality of research related to TMS and academic integrity in general.

In recent years, companies that produce TMS have expanded their offerings to include complementary products, such as those designed to recognise writing that may be a result of contract cheating. It is likely that machine learning and artificial intelligence and, in particular, technologies such as the Generative Pretrained Transformer (GPT) 3 [4, 161, 162] will expand concerns about student writing and authorship, as well as the notion of what constitutes original work on the part of the student. Research into these new technologies will become an important and urgent question for academic integrity scholars in the coming years.

At the beginning of our study, there were few tools or resources available for knowledge synthesis (i.e., systematic or scoping) reviews outside of medical and health sciences.

During the course of our project, additional resources became available to support knowledge synthesis reviews in educational contexts [159, 163]. As such, we advocate for more systematic and scoping reviews to be conducted on topics related to academic integrity, with an ongoing emphasis on quality and methodological rigour.

7. Conclusions

Whether text-matching software is used as a tool to educate students, to detect misconduct, or a combination of both, the technology is here to stay. Our review showed methodological breadth in approaches to studying the use of such software including qualitative, quantitative, mixed methods, and other kinds of research design, with studies of varying quality across all research design types.

Although we were unable to provide conclusive advice regarding the efficacy of text-matching software as a tool to reduce plagiarism or as an educational intervention, we have shown how commercial text-matching software is used in postsecondary education as both a punitive and pedagogical tool at undergraduate and graduate levels.

This study contributes to the body of scholarship regarding text-matching software, offering evidence-informed insights as to methodological possibilities for research and the need to focus on rigour in scholarship to further develop academic integrity as an area of research. This study may also be useful to educational decision makers seeking advice on what to consider prior to licensing a commercial text-matching software and, in particular, the need to understand the limitations of both the software itself and the extant evidence to inform such decisions.

Data Availability

All data sources are cited in the manuscript.

Conflicts of Interest

The authors declare no conflicts of interest.

Acknowledgments

The authors are grateful to the University of Calgary for providing funding for this project. The authors extend our appreciation to their research assistant, Claire Rockliff.

References

- [1] I. K. Tindall and G. J. Curtis, "Negative emotionality predicts attitudes toward plagiarism," *Journal of Academic Ethics*, vol. 18, no. 1, pp. 89–102, 2020.
- [2] G. J. Curtis and K. Tremayne, "Is plagiarism really on the rise? Results from four 5-yearly surveys," *Studies in Higher Education*, vol. 46, no. 9, pp. 1816–1826, 2021.
- [3] K. Tremayne and G. J. Curtis, "Attitudes and understanding are only part of the story: self-control, age and self-imposed pressure predict plagiarism over and above perceptions of seriousness and understanding," *Assessment & Evaluation in Higher Education*, vol. 46, no. 2, pp. 208–219, 2020.

- [4] S. E. Eaton, *Plagiarism in Higher Education: Tackling Tough Topics in Academic Integrity*, Libraries Unlimited, Santa Barbara, CA, USA, 2021.
- [5] T. Batane, "Turning to Turnitin to fight plagiarism among university students," *Educational Technology & Society*, vol. 13, no. 2, pp. 1–12, 2010.
- [6] D. C. Ison, "The influence of the Internet on plagiarism among doctoral dissertations: an empirical study," *Journal of Academic Ethics*, vol. 13, no. 2, pp. 151–166, 2015.
- [7] L. J. P. Davies and R. M. Howard, "Plagiarism and the internet: fears, facts, and pedagogies," in *Handbook of Academic Integrity*, T. Bretag, Ed., Springer Singapore, Singapore, pp. 591–606, 2016.
- [8] R. M. Howard and L. J. Davies, "Plagiarism in the internet age," *Educational Leadership*, vol. 66, no. 6, pp. 64–67, 2009.
- [9] G. Curtis and L. Vardanega, "Is plagiarism changing over time? A 10-year time-lag study with three points of measurement," *Higher Education Research and Development*, vol. 35, no. 6, pp. 1–13, 2016.
- [10] S. D. Blum, *My Word! Plagiarism and Academic College Culture*, Cornell University Press, Ithaca, NY, USA, 2009.
- [11] S. D. Blum, "What it means to be a student today," in *Handbook of Academic Integrity*, T. Bretag, Ed., Springer Singapore, Singapore, pp. 383–406, 2016.
- [12] A. M. Rogerson and G. Basanta, "Peer-to-peer file sharing and academic integrity in the Internet age," in *Handbook of Academic Integrity*, T. Bretag, Ed., Springer Singapore, Singapore, pp. 273–285, 2016.
- [13] J. Badge and J. Scott, "Dealing with plagiarism in the digital age," 2009.
- [14] J. Hickman, *Cribbing Turns Tech as Students Tap into Net*. The Globe and Mail Inc., Toronto, ON, Canada, 1998.
- [15] A. MacGregor, *Outfoxing Online Cheaters: IntegriGuard Web Site Scans Papers Submitted by Students for Plagiarism*, Montreal Gazette, Montreal, Canada, 1998.
- [16] S. Schmidt, *Wired World Gives Cheating a New Face*, The Globe and Mail Inc., Toronto, ON, Canada, 1998.
- [17] S. Bruton and D. Childers, "The ethics and politics of policing plagiarism: a qualitative study of faculty views on student plagiarism and Turnitin," *Assessment & Evaluation in Higher Education*, vol. 41, no. 2, pp. 316–330, 2016.
- [18] S. E. Eaton, K. Crossman, L. Behjat, R. M. Yates, E. Fear, and M. Trifkovic, "An institutional self-study of text-matching software in a Canadian graduate-level engineering program," *Journal of Academic Ethics*, vol. 18, no. 3, pp. 263–282, 2020.
- [19] L. Brown, *McGill Teen Wins Battle over Online Cheat Check; System Smears Innocent Scholars, Student Says Turnitin.com Examines Essays for Plagiarism*, Toronto Star, Toronto, Canada, 2004.
- [20] L. Charbonneau, "The cheat checker: the use of Turnitin.com to combat plagiarism is raising passions on Canada's campuses," *University Affairs*, vol. 4, no. 16–20, p. 45, 2004.
- [21] M. Groark, D. G. Oblinger, and M. Choa, "Term paper mills, anti-plagiarism tools, and academic integrity," *Educause Review*, pp. 40–48.
- [22] R. K. Baker, B. Thornton, and M. Adams, "An evaluation of the effectiveness of Turnitin.com as a tool for reducing plagiarism in graduate student term papers," *College Teaching Methods & Styles Journal (CTMS)*, vol. 4, no. 9, pp. 1–4, 2008.
- [23] D. Weber-Wulff, "Plagiarism detection software: promises, pitfalls, and practices," in *Handbook of Academic Integrity*, T. Bretag, Ed., Springer Singapore, Singapore, pp. 625–638, 2016.
- [24] T. Foltýnek, N. Meuschke, and B. Gipp, "Academic plagiarism detection," *ACM Computing Surveys*, vol. 52, no. 6, pp. 1–42, 2020.
- [25] K. A. Hayden, S. E. Eaton, K. Crossman, L. A. Penaluna, and B. A. Lenart, "Text-matching software in post-secondary contexts: a systematic review protocol," *Canadian Perspectives on Academic Integrity*, vol. 3, no. 1, pp. 6–24, 2020.
- [26] M. Page, D. Moher, P. Bossuyt et al., "PRISMA 2020 explanation and elaboration: updated guidance and exemplars for reporting systematic reviews," *BMJ*, vol. 372, 2021.
- [27] Z. Munn, M. D. Peters, C. Stern, C. Tufanaru, A. McArthur, and E. Aromataris, "Systematic review or scoping review? Guidance for authors when choosing between a systematic or scoping review approach," *BMC Medical Research Methodology*, vol. 18, no. 1, pp. 1–7, 2018.
- [28] E. Aromataris and Z. Munn, *JBIC Manual for Evidence Synthesis*, JBI, Adelaide, Australia, 2020, <https://wiki.jbi.global/display/MANUAL>.
- [29] Statistics Canada, *Postsecondary Student Information System (PSIS): Detailed Information for 2016-2017*, Statistics Canada, Ottawa, Ontario, Canada, 2018, <http://www23.statcan.gc.ca/imdb/p2SV.pl?Function=getSurvey&SDDS=5017>.
- [30] K. A. Hayden, S. E. Eaton, H. Pethrick, K. Crossman, B. A. Lenart, and L. A. Penaluna, "Scoping review of text-matching software used for student academic integrity in higher education," 2021, <https://osf.io/5f8rd/>.
- [31] Viewing duplicates and errors (Covidence), 2019, <https://support.covidence.org/help/viewing-duplicate>.
- [32] Q. N. Hong, S. Fàbregues, G. Bartlett et al., "Mixed methods appraisal tool (MMAT)," 2018, http://mixedmethodsappraisaltoolpublic.pbworks.com/w/file/attach/127916259/MMAT_2018_criteria-manual_2018-08-01_ENG.pdf.
- [33] Joanna Briggs Institute, *Critical Appraisal Checklist for Text and Opinion*, Joanna Briggs Institute, Adelaide, Australia, 2017, https://jbi.global/sites/default/files/2019-05/JBI_Critical_Appraisal-Checklist_for_Text_and_Opinion2017_0.pdf.
- [34] Y. M. Abdullah and D. M. Krebt, "The effectiveness of using paraphrasing strategies on research writing: turnitin model," *Annals of the Faculty of Arts*, vol. 46, pp. 355–633, 2018.
- [35] E. Abrahamson and J. Mann, "For whom is the feedback intended? A student-focused critical analysis of Turnitin software as a tool for learning," *Journal of Pedagogical Research*, vol. 2, no. 3, pp. 146–166, 2018.
- [36] L. Adam, C. Bond, and A. Samaranayaka, "The score doesn't mean much": students' and staff understandings of text matching software," in *Rhetoric and Reality: Critical Perspectives on Educational Technology* Ascilite, Dunedin, New Zealand, 2014.
- [37] M. A. Alharbi and A. H. Al-Hoorie, "Turnitin peer feedback: controversial vs. non-controversial essays," *International Journal of Educational Technology in Higher Education*, vol. 17, no. 1, pp. 1–17, 2020.
- [38] H. I. Holi Ali, "Minimizing cyber-plagiarism through turnitin: faculty's & students' perspectives," *International Journal of Applied Linguistics & English Literature*, vol. 2, no. 2, pp. 33–42, 2013.
- [39] D. Atkinson and S. Yeoh, "Tudent and staff perceptions of the effectiveness of plagiarism detection software," *Australasian Journal of Educational Technology*, vol. 24, no. 2, pp. 222–240, 2008.
- [40] C. Bailey, R. Challen, C. Bailey, and R. Challen, "Student perceptions of the value of Turnitin text-matching software

- as a learning tool,” *Practitioner Research in Higher Education*, vol. 9, no. 1, pp. 38–51.
- [41] J. L. Badge, A. J. Cann, and J. Scott, “To cheat or not to cheat? A trial of the JISC plagiarism detection service with biological sciences students,” *Assessment & Evaluation in Higher Education*, vol. 32, no. 4, pp. 433–439, 2007.
- [42] S. Balbay and S. Kilis, “Perceived effectiveness of Turnitin® in detecting plagiarism in presentation slides,” *Contemporary Educational Technology*, vol. 10, no. 1, pp. 25–36, 2019.
- [43] I. B. Ballard, “The impact of an academic integrity module and Turnitin® on similarity index scores of undergraduate student papers,” *Research in the Schools*, vol. 20, no. 2, pp. 1–13, 2013.
- [44] S. Belli, C. L. Raventos, and T. Guarda, “Plagiarism detection in the classroom: honesty and trust through the Urkund and Turnitin software,” in *Proceedings of the International Conference on Information Technology & Systems*, Bogots, Colombia, February 2020.
- [45] E. R. Bensal, E. S. Miraflores, and N. C. C. Tan, “Plagiarism: shall we turn to turnitin?,” *Computer-Assisted Language Learning-Electronic Journal*, vol. 14, no. 2, pp. 2–22, 2013.
- [46] T. Bertram Gallant, M. Picciotto, G. Bozinovic, and E. Tour, “Plagiarism or not? investigation of Turnitin®-detected similarity hits in biology laboratory reports,” *Biochemistry and Molecular Biology Education*, vol. 47, no. 4, pp. 307–379, 2019.
- [47] L. R. Betts, L. R. Bostock, T. J. Elder, and M. Trueman, “Encouraging good writing practice in first-year psychology students writing: an intervention using Turnitin,” *Psychology Teaching Review*, vol. 18, no. 2, pp. 74–81, 2013.
- [48] J. Biggam and M. McCann, “A study of Turnitin as an educational tool in student dissertations,” *Interactive Technology and Smart Education*, vol. 7, no. 1, pp. 44–54, 2010.
- [49] L. Bilic-Zulle, J. Azman, V. Frkovic, and M. Petroveckii, “Is there an effective approach to deterring students from plagiarizing?,” *Science and Engineering Ethics*, vol. 14, no. 1, pp. 139–147, 2008.
- [50] T. Bretag and S. Mahmud, “A model for determining student plagiarism: electronic detection and academic judgement,” *Journal of University Teaching & Learning Practice*, vol. 6, no. 1, pp. 49–60, 2009.
- [51] B. Brinkman, “An analysis of student privacy rights in the use of plagiarism detection systems,” *Science and Engineering Ethics*, vol. 19, no. 3, pp. 1255–1266, 2013.
- [52] R. Brown, B. Fallon, J. Lott, E. Matthews, and E. Mintie, “Taking on turnitin: tutors advocating change,” *The Writing Center Journal*, vol. 27, no. 1, pp. 7–28, 2004.
- [53] V. Brown, R. Jordan, N. Rubin, and G. Arome, “Strengths and weaknesses of plagiarism detection software,” *The Journal of Literacy and Technology*, vol. 11, no. 1, pp. 110–131, 2010.
- [54] E. Buckley and L. Cowap, “An evaluation of the use of Turnitin for electronic submission and marking and as a formative feedback tool from an educator’s perspective,” *British Journal of Educational Technology*, vol. 44, no. 4, pp. 562–570, 2013.
- [55] C. A. Chao, W. J. Wilhelm, and B. D. Neureuther, “A study of electronic detection and pedagogical approaches for reducing plagiarism,” *Delta Pi Epsilon Journal*, vol. 51, no. 1, pp. 31–42, 2009.
- [56] C. H. Cheong, T. P. Lau, and I. King, “Policy and issues in deploying automated plagiarism detection systems in academic communities,” in *IT Policy and Ethics: Concepts, Methodologies, Tools, and Applications*, pp. 1425–1450, IGI Global, Hershey, PA, USA, 2013.
- [57] E. Chew and S. L. Ding, “Changing attitudes in learning and assessment: cast-off ‘plagiarism detection’ and cast-on self-service assessment for learning,” *Innovations in Education & Teaching International*, vol. 52, no. 5, pp. 454–463, 2015.
- [58] E. Chew and H. Blackey, “E-plagiarism detection at glarmorgan,” *World Academy of Science, Engineering and Technology*, vol. 69, pp. 195–199, 2010.
- [59] E. P. T. Chew, C. B. K. Rogers, N. Noblett, A. Martins, and I. Martins, “Turn it in or turn it off: confirming and disconfirming experience for innovative computer-aided assessment and feedback,” in *Proceedings of the 9th European Conference on eLearning*, Porto, Portugal, November 2010.
- [60] J. C. Clanton, “A moral case against certain uses of plagiarism detection services,” *International Journal of Applied Philosophy*, vol. 23, no. 1, pp. 17–26, 2009.
- [61] S. Coffey and C. Anyinam, “Trialing a contextual approach to academic honesty,” *Nurse Educator*, vol. 37, no. 2, pp. 62–66, 2012.
- [62] M. Cooper and L. Bullard, “Application of plagiarism screening software in the chemical engineering curriculum,” *Chemical Engineering Education*, vol. 48, no. 2, pp. 90–96, 2014.
- [63] R. F. da Silva, M. Y. Ramirez-Sanchez, I. Ilizaliturri-Flores et al., “The ethics in the practice of the research and academic work,” in *Proceedings of the 11th International Conference of Education, Research and Innovation*, Seville, Spain, November 2018.
- [64] S. Dahl, “Turnitin,” *Active Learning in Higher Education*, vol. 8, no. 2, pp. 173–191, 2007.
- [65] S. Daoud, H. Alrabaiah, and E. Zaitoun, “Technology for promoting academic integrity: the impact of using Turnitin on reducing plagiarism,” in *Proceedings of the 2019 International Arab Conference on Information Technology*, Al Ain, UAE, December 2019.
- [66] M. Davis and J. Carroll, “Formative feedback within plagiarism education: is there a role for text-matching software?,” *International Journal for Educational Integrity*, vol. 5, no. 2, pp. 58–70, 2009.
- [67] P. Dawson, W. Sutherland-Smith, and M. Ricksen, “Can software improve marker accuracy at detecting contract cheating? A pilot study of the Turnitin authorship investigate alpha,” *Assessment & Evaluation in Higher Education*, vol. 45, no. 4, pp. 473–482, 2020.
- [68] K. Do Ba, K. Do Ba, Q. D. Lam et al., “Student plagiarism in higher education in Vietnam: an empirical study,” *Higher Education Research and Development*, vol. 36, no. 5, pp. 934–946, 2017.
- [69] M. Dodigovic and X. Jiaotong, “The role of anti-plagiarism software in learning to paraphrase effectively,” *CALL-EJ*, vol. 14, no. 2, pp. 23–37, 2013.
- [70] L. Dreuth Zeman, J. A. Steen, and N. Metz Zeman, “Originality detection software in a graduate policy course: a mixed-methods evaluation of plagiarism,” *Journal of Teaching in Social Work*, vol. 31, no. 42, pp. 431–441, 2011.
- [71] L. Emerson, “Plagiarism, a Turnitin trial, and an experience of cultural disorientation,” in *Originality, Imitation, and Plagiarism: Teaching Writing in the Digital Age*, C. Eisner and M. Micinus, Eds., pp. 183–194, University of Michigan Press, Ann Arbor, MI, USA, 2008.
- [72] L. Emerson, M. T. Rees, and B. MacKay, “Scaffolding academic integrity: creating a learning context for teaching referencing skills,” *Journal of University Teaching & Learning Practice*, vol. 2, no. 3, 2005.
- [73] R. Evans, “Evaluating an electronic plagiarism detection service,” *Active Learning in Higher Education*, vol. 7, no. 1, pp. 87–99, 2006.

- [74] H. Ewing, A. Anast, and T. Roehling, "Addressing plagiarism in online programmes at a health sciences university: a case study," *Assessment & Evaluation in Higher Education*, vol. 41, no. 4, pp. 575–585, 2016.
- [75] P. Gannon-Leary, D. Trayhurn, and M. Home, "Good images, effective messages? Working with students and educators on academic practice understanding," *Journal of Further and Higher Education*, vol. 4, no. 33, pp. 435–448.
- [76] K. Gillis, S. Lang, M. Norris, and L. Palmer, "Electronic plagiarism checkers: barriers to developing academic voice," *The WAC Journal*, vol. 20, no. 1, pp. 51–62, 2009.
- [77] R. Goddard and R. Rudzki, "Using an electronic text-matching tool (turnitin) to detect plagiarism in a New Zealand university," *Journal of University Teaching and Learning Practice*, vol. 2, no. 3, 2005.
- [78] L. Graham-Matheson and S. Starr, "Is it cheating—or learning the craft of writing? Using Turnitin to help students avoid plagiarism," *Research in Learning Technology*, vol. 21, pp. 1–13, 2013.
- [79] D. Green, I. Lindemann, K. Marshall, and G. Wilkinson, "Student perceptions of a trial of electronic text matching software: a preliminary investigation," *Journal of University Teaching and Learning Practice*, vol. 2, no. 3, 2005.
- [80] B. Grohe, J. Schroeder, and S. R. Davis, "Using online resources to improve writing skills and attitudes about writing and plagiarism of criminal justice students," *Journal on Excellence in College Teaching*, vol. 24, no. 1, pp. 23–45, 2013.
- [81] M. N. Halgamuge, "The use and analysis of anti-plagiarism software: turnitin tool for formative assessment and feedback," *Computer Applications in Engineering Education*, vol. 25, no. 6, pp. 895–909, 2017.
- [82] M. Hamilton and J. Richardson, "Academic integrity compliance and education," in *Proceedings of the ASCILITE*, Melbourne, Burwood, Australia, November 2008.
- [83] A. Hapsari, M. I. Ghali, and M. H. Ammar, "The use of turnitin to teach academic integrity in essay writing coursework," in *Proceedings of the 2020 11th International Conference on E-Education, E-Business, E-Management, and E-Learning*, Osaka, Japan, January 2020.
- [84] J. Heather, "Turnitoff: identifying and fixing a hole in current plagiarism detection software," *Assessment & Evaluation in Higher Education*, vol. 35, no. 6, pp. 647–660, 2010.
- [85] N. C. Heckler, "Mitigating plagiarism in large introductory courses in higher education," [unpublished doctoral dissertation], [unpublished doctoral dissertation]2013.
- [86] N. C. Heckler, M. Rice, and C. Hobson Bryan, "Turnitin systems," *Journal of Research on Technology in Education*, vol. 45, no. 3, pp. 229–248, 2013.
- [87] J. D. Hill and E. F. Page, "An empirical research study of the efficacy of two plagiarism-detection applications," *Journal of Web Librarianship*, vol. 3, no. 3, pp. 169–181, 2009.
- [88] A. M. Houtman and S. Walker, "Decreasing plagiarism: what works and what doesn't," *Journal on Excellence in College Teaching*, vol. 21, no. 1, pp. 51–71, 2010.
- [89] J. Hunt and P. Tompkins, "A comparative analysis of SafeAssign and turnitin," *Inquiry*, vol. 19, no. 1, pp. 63–73, 2014.
- [90] D. C. Ison and K. J. Szathmary, "Assessing academic integrity using SafeAssign plagiarism detection software," *The Collegiate Aviation Review International*, vol. 34, no. 1, pp. 35–47, 2016.
- [91] C. L. Jocoy and D. DiBiase, "Plagiarism by adult learners online: a case study in detection and remediation," *International Review of Research in Open and Distance Learning*, vol. 7, no. 1, pp. 1–15, 2006.
- [92] F. Johari, M. H. Alias, A. A. Rahman, and M. F. Ibrahim, "The usage of 'turnitin' as an innovative educational tool: inculcating critical thinking in integrating naqli and aqli for subject of Malaysian economy," *Procedia-Social and Behavioral Sciences*, vol. 195, pp. 821–827, 2015.
- [93] K. O. Jones, "Practical issues for academics using the Turnitin plagiarism detection software," in *Proceedings of the 9th International Conference on Computer Systems and Technologies and Workshop for PhD Students in Computing*, Gabrovo, Bulgaria, June 2008.
- [94] K. O. Jones and T. A. Moore, "Turnitin is not the primary weapon in the campaign against plagiarism," in *Proceedings of the 11th International Conference on Computer Systems and Technologies and Workshop for PhD Students in Computing on International Conference on Computer Systems and Technologies*, Sofia, Bulgaria, June 2010.
- [95] T. Kakkonen and M. Mozgovoy, "An evaluation of web plagiarism detection systems for student essays," in *Proceedings of the 16th International Conference on Computers in Education*, Taipei, Taiwan, 2008.
- [96] L. Kaktiņš, "Does Turnitin support the development of international students' academic integrity?," *Ethics and Education*, vol. 14, no. 4, pp. 430–448, 2019.
- [97] S. T. Kale, "Use of Turnitin and Urkund anti-plagiarism software tools for plagiarism detection/similarity checks to the doctoral theses: Indian experience," *Cadernos BAD*, vol. 1, pp. 95–99, 2020.
- [98] P. P. Kiriakidis, "The effect of a policy of mandatory use of TurnItIn by graduate and postgraduate online students on reducing unoriginal writing," *Postmodern Openings*, vol. 4, no. 3, pp. 43–61, 2013.
- [99] J. Kirkpatrick, "Teaching acknowledgement practice using the internet-based plagiarism detection service," *Marketing Education Review*, vol. 16, no. 1, pp. 29–33, 2006.
- [100] K. E. Kohl, "Fostering academic competence or putting students under general suspicion? Voluntary plagiarism check of academic papers by means of a web-based plagiarism detection system," *Research in Learning Technology*, vol. 19, pp. 90–99, 2011.
- [101] A. Konstantinidis, D. Theodosiadou, and C. Pappos, "Plagiarism: examination of conceptual issues and evaluation of research findings on using detection services," *Contemporary Educational Technology*, vol. 4, no. 3, pp. 212–221, 2019.
- [102] O. Kose and A. Arikan, "Reducing plagiarism by using online software: an experimental study," *Contemporary Online Language Education Journal*, vol. 1, pp. 122–129, 2011.
- [103] I. Kostka and M. E. Ebsworth, "Using turnitin to support students' understanding of textual borrowing in academic writing," in *Scholarly Ethics and Publishing: Breakthroughs in Research and Practice*, pp. 269–297, IGI Global, Hershey, PA, USA, 2019.
- [104] I. Kostka and V. Maliborska, "Using turnitin to provide feedback on L2 writers' texts," *TESL-EJ*, vol. 20, no. 2, p. n2, 2016.
- [105] P. K. S. Kumar, "Similarity index of doctoral theses submitted to universities in Kerala: an investigation," *Library Philosophy & Practice*, pp. 1–10, 2019.
- [106] C. Kunschak, "Multiple uses of anti-plagiarism software," *The Asian Journal of Applied Linguistics*, vol. 5, no. 1, pp. 60–69, 2018.
- [107] A. Ledwith and A. Rísquez, "Using anti-plagiarism software to promote academic honesty in the context of peer reviewed

- assignments,” *Studies in Higher Education*, vol. 33, no. 4, pp. 371–384, 2008.
- [108] J. B. J. Lee and Y.-J. Chen, “Adapting anti-plagiarism tool into coursework in engineering program,” in *Proceedings of the American Society for Engineering Education*, Austin, TX, USA, June 2009.
- [109] J. Levine and V. Pazdernik, “Evaluation of a four-prong anti-plagiarism program and the incidence of plagiarism: a five-year retrospective study,” *Assessment & Evaluation in Higher Education*, vol. 43, no. 7, pp. 1094–1105, 2018.
- [110] M. Li and J. Li, “Online peer review using Turnitin in first-year writing classes,” *Computers and Composition*, vol. 46, pp. 21–38, 2017.
- [111] B. Marsh, “Turnitin.com and the scriptural enterprise of plagiarism detection,” *Computers and Composition*, vol. 21, no. 4, pp. 427–438, 2004.
- [112] D. F. Martin, “Plagiarism and technology: a tool for coping with plagiarism,” *The Journal of Education for Business*, vol. 80, no. 3, pp. 149–152, 2005.
- [113] G. McCarthy and A. Rogerson, “Links are not enough: using originality reports to improve academic standards, compliance and learning outcomes among postgraduate students,” *International Journal for Educational Integrity*, vol. 5, no. 2, pp. 47–57, 2009.
- [114] L. McKeever, “Online plagiarism detection services—saviour or scourge?,” *Assessment & Evaluation in Higher Education*, vol. 31, no. 2, pp. 155–165, 2006.
- [115] S. A. Meo and M. Talha, “Turnitin: is it a text matching or plagiarism detection tool?,” *Saudi Journal of Anaesthesia*, vol. 13, no. 5, pp. s48–s51, 2019.
- [116] E. Mohammadkarimi and M. Y. M. Amin, “ELT students’ attitudes toward the effectiveness the anti-plagiarism software, Turnitin,” *Applied Linguistics Research Journal*, vol. 3, no. 5, pp. 63–95, 2019.
- [117] A. Mphahlele and S. McKenna, “The use of turnitin in the higher education sector: decoding the myth,” *Assessment & Evaluation in Higher Education*, vol. 44, no. 7, pp. 1079–1089, 2019.
- [118] K. Oghigian, M. Rayner, and K. Chujo, “A quantitative evaluation of Turnitin from an L2 science and engineering perspective,” *CALL-EJ*, vol. 17, no. 1, pp. 1–18, 2015.
- [119] J. Orlando, J. Hanham, and J. Ullman, “Exploring intentional use of a technological proxy, Turnitin, to enhance student academic literacy practices,” *Australasian Journal of Educational Technology*, vol. 34, no. 4, pp. 44–56, 2018.
- [120] R. D. Pai and S. S. Parmar, “User awareness survey on Turnitin: an electronic plagiarism detection tool,” *International Journal of Information Dissemination and Technology*, vol. 5, no. 1, pp. 41–45, 2015.
- [121] C. Pardo-Garcia and M. Barac, “An experience in detection of plagiarism in an ordinary writing task,” in *Proceedings of the EDULEARN13*, Barcelona, Spain, July 2013.
- [122] R. Patton, D. Johnson, B. Bimber, K. Almeroth, and G. Michaels, “Technology and plagiarism in the university: brief report of a trial in detecting cheating,” *AACE Journal*, vol. 12, no. 3, pp. 281–299, 2004.
- [123] A. Pavelich, “The moral hazards of using turnitin as a learning tool,” *Teaching Ethics*, vol. 19, no. 2, pp. 195–206, 2019.
- [124] C. Penketh and C. Beaumont, “‘Turnitin said it wasn’t happy’: can the regulatory discourse of plagiarism detection operate as a change artefact for writing development?,” *Innovations in Education & Teaching International*, vol. 51, no. 1, pp. 95–104, 2014.
- [125] M. A. Pondayi and C. Rodreck, “The impact of the use of plagiarism detection technology (turnitin) on students’ assignments submitted online, a case study of botho college,” in *Proceedings of the NTED2012*, Francistown, Botswana, 2012.
- [126] J. Price and R. Price, “Finding the true incidence rate of plagiarism,” *International Education Journal*, vol. 6, no. 3, pp. 421–429, 2005.
- [127] J. P. Purdy, “Anxiety and the archive: understanding plagiarism detection services as digital archives,” *Computers and Composition*, vol. 26, no. 2, pp. 65–77, 2009.
- [128] J. P. Purdy, “Calling off the hounds: technology and the visibility of plagiarism,” *Pedagogy: Critical Approaches to Teaching Literature, Language, Composition, and Culture*, vol. 5, no. 2, pp. 275–296, 2005.
- [129] A. Rashid and A. Rashid, “Academic policing via top-down implementation of turnitin in Pakistan: students’ perspective and way forward,” *Pakistaniaat: A Journal of Pakistan Studies*, vol. 6, pp. 1–20, 2018.
- [130] S. A. Reed, “SafeAssign as a tool for student identification of potential plagiarism in an animal science writing course,” *Natural Sciences Education*, vol. 44, no. 1, pp. 95–100, 2015.
- [131] M. Rees and L. Emerson, “The impact that Turnitin® has had on text-based assessment practice,” *International Journal for Educational Integrity*, vol. 5, no. 1, pp. 20–29, 2009.
- [132] C. Rieber, “Using electronic plagiarism tools to teach university students to avoid plagiarism,” in *Proceedings of the Conference on Plagiarism across Europe and beyond 2017*, Brno, Czech Republic, May 2017.
- [133] V. Rolfe, “Can Turnitin be used to provide instant formative feedback?,” *British Journal of Educational Technology*, vol. 42, no. 4, pp. 701–710, 2011.
- [134] E. A. Sabapathy, “Realigning the focus of plagiarism detection using plagiarismdetect.com,” *English Language Teaching*, vol. 2, no. 4, pp. 210–214, 2009.
- [135] M. Sarwar, M. Moin, and M. Jabeen, “Role of plagiarism detecting software in reducing academic dishonesty at M. Phil level,” *Dialogue*, vol. 11, no. 4, pp. 414–426, 2016.
- [136] A. G. Scheg, “The impact of Turnitin to the student-teacher relationship,” *Journal of Interdisciplinary Studies in Education*, vol. 2, no. 1, pp. 29–38, 2013.
- [137] H.-F. Shang, “An investigation of plagiarism software use and awareness training on English as a foreign language (EFL) students,” *Journal of Computing in Higher Education*, vol. 31, no. 1, pp. 105–120, 2019.
- [138] J. Sheridan, R. Alany, and D. J. Brake, “Pharmacy students’ views and experiences of Turnitinw—an online tool for detecting academic dishonesty,” *Pharmacy Education*, vol. 5, no. 3, pp. 241–250, 2005.
- [139] V. Silvey, T. Snowball, and T. Do, “Bridge over troubled water: a literacy approach to using Turnitin,” *Journal of Academic Language and Learning*, vol. 10, no. 1, pp. a206–a222, 2016.
- [140] P. Stapleton, “Gauging the effectiveness of anti-plagiarism software: an empirical study of second language graduate writers,” *Journal of English for Academic Purposes*, vol. 11, no. 2, pp. 125–133, 2012.
- [141] J. Strawczynski, “When students won’t turnitin: an examination of the use of plagiarism prevention services in Canada,” *Education and Law Journal*, vol. 14, no. 2, pp. 167–190, 2004.
- [142] A. Thompsett and J. Ahluwalia, “Students turned off by Turnitin? Perception of plagiarism and collusion by

- undergraduate bioscience students,” *Bioscience Education*, vol. 16, no. 1, pp. 1–15, 2010.
- [143] P. Todd, “Plagiarism detection software: legal and pedagogical issues,” *The Law Teacher*, vol. 44, no. 2, pp. 137–148, 2010.
- [144] Z. Toprak and V. Yücel, “A peculiar practice of academic writing: epidemic writing in the Turkish graduate education,” *Cogent Education*, vol. 7, no. 1, 2020.
- [145] G. R. Townsend, “A study using plagiarism detection services to assess the effect of an APA formatting and plagiarism training lesson on the quality of student originality scores,” [unpublished doctoral dissertation], [unpublished doctoral dissertation]2018.
- [146] K. Tulley Pitchford, “Mouse-click plagiarism: can technology help to fight back?,” *Practitioner Research in Higher Education*, vol. 6, no. 2, pp. 58–68, 2012.
- [147] B. Vanacker, “Returning students’ right to access, choice and notice: a proposed code of ethics for instructors using Turnitin,” *Ethics and Information Technology*, vol. 13, no. 4, pp. 327–338, 2011.
- [148] S. Vie, “A pedagogy of resistance toward plagiarism detection technologies,” *Computers and Composition*, vol. 30, no. 1, pp. 3–15, 2019.
- [149] G. Villar-Mayuntupa, “Using Turnitin to boost the originality in the elaboration of argumentative essays among Peruvian university students,” in *Proceedings of the 2020 IEEE World Conference on Engineering Education*, Bogota, Colombia, 2020.
- [150] A. U. Waigand, “Using Turnitin to help students understand plagiarism,” *Learning and Teaching in Higher Education: Gulf Perspectives*, vol. 16, no. 1, 2019.
- [151] J. Walker, “Measuring plagiarism: researching what students do, not what they say they do,” *Studies in Higher Education*, vol. 35, no. 1, pp. 41–59, 2010.
- [152] J. Warn, “Plagiarism software: no magic bullet!” *Higher Education Research and Development*, vol. 25, no. 2, pp. 195–208, 2006.
- [153] R. J. Youmans, “Does the adoption of plagiarism-detection software in higher education reduce plagiarism?,” *Studies in Higher Education*, vol. 36, no. 7, pp. 749–761, 2011.
- [154] S. Zangenehmadar and T. B. Hoon, “ESL undergraduates’ patterns of plagiarism in academic essays based on print or internet sources,” *Pertanika Journal of Social Sciences and Humanities*, vol. 25, no. 1, pp. 205–220, 2017.
- [155] C. Zaza and A. McKenzie, “Turnitin® use at a Canadian university,” *Canadian Journal for the Scholarship of Teaching and Learning*, vol. 9, no. 2, p. n2, 2018.
- [156] Y. Zhang and D. K. Probst, “Thinking and understanding from writing,” in *Proceedings of the American Society for Engineering Education*, Columbus, OH, USA, June 2017.
- [157] T. Bertram Gallant, N. Binkin, and M. Donohue, “Students at risk for being reported for cheating,” *Journal of Academic Ethics*, vol. 13, no. 3, pp. 217–228, 2015.
- [158] B. Macfarlane, J. Zhang, and A. Pun, “Academic integrity: a review of the literature,” *Studies in Higher Education*, vol. 39, no. 2, pp. 339–358, 2014.
- [159] K. Ahsan and S. Akbar, “Contract cheating in higher education: a systematic literature review and future research agenda,” *Assessment & Evaluation in Higher Education*, Taylor and Francis Ltd., Abingdon, UK, pp. 1–17, 2021.
- [160] S. Awasthi, “Plagiarism and academic misconduct a systematic review,” *DESIDOC Journal of Library & Information Technology*, vol. 39, no. 2, pp. 94–100, 2019.
- [161] M. Mindzak, *What Happens when a Machine Can Write as Well as an Academic?*, University Affairs, Canada, 2020, <https://www.universityaffairs.ca/opinion/in-my-opinion/what-happens-when-a-machine-can-write-as-well-as-an-academic/>.
- [162] R. Morrison and M. Mindzak, *Exploring the Impacts of Text Generating Technologies on Academic Integrity. Webinar Presented at the Taylor Institute for Teaching and Learning Webinar Series: Academic Integrity—Urgent and Emerging Topics*, <https://taylorinstitute.ucalgary.ca/series-and-events/academic-integrity-urgent-emerging-topics>, 2020.
- [163] O. Zawacki-Richter, M. Kerres, S. Bedenlier, M. Bond, and K. Buntins, “Systematic reviews in educational research: methodology, perspectives and application,” 2021.