

# The Role of Big Data Analytics on Improving Technical and Vocational Education Outcomes

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#### Abstract

This systematic literature review examines the potential of big data analytics to improve technical and vocational education and training (TVET) outcomes. A search of relevant databases yielded 22 studies that were included in the review. The findings suggest that big data analytics interventions can improve TVET outcomes, particularly in terms of student performance and engagement. However, the quality of the evidence is limited by the high risk of bias in many of the studies and the heterogeneity of the interventions and outcomes measured. Big data analytics interventions can provide real-time insights into student learning behaviors and performance, predict student outcomes, and offer personalized feedback and support. The positive impact of these interventions on TVET outcomes is consistent with previous research in other educational contexts. To maximize the potential of big data analytics in TVET, future research should use rigorous study designs, consistent outcome measures, and identify factors that contribute to the success or failure of these interventions. In conclusion, this review suggests that big data analytics can improve TVET outcomes, particularly where there is a strong culture of data use. Further research is needed to identify factors that contribute to the success or failure of big data analytics interventions in TVET, with the goal of improving TVET quality and enhancing graduate employability.

**Key words**: Big data analytics, Educational data mining, Learning outcomes, Student performance, Technical and vocational education and training (TVET)

# **I.Introduction**

The use of big data analytics in education has become increasingly popular in recent years, with the potential to improve teaching and learning outcomes. Technical and vocational education (TVET) is a critical sector that plays a vital role in preparing individuals for the workforce. However, the quality of TVET programs and the resulting outcomes have been widely debated, with concerns about the relevance, effectiveness, and efficiency of these programs.

The integration of big data analytics into TVET programs has been proposed as a potential solution to address these concerns. Big data analytics can provide insights into student performance, engagement, and behavior, as well as identify areas for improvement in teaching and learning processes. By leveraging these insights, TVET institutions can improve the quality of their programs and enhance the outcomes for their students.

Several studies have explored the role of big data analytics in education, but few have specifically focused on its impact on TVET outcomes. Therefore, the purpose of this systematic literature review is to examine the existing research on the role of big data analytics on improving technical and vocational education outcomes. Specifically, this review aims to answer the following research question:

What is the impact of big data analytics on TVET outcomes, and how can it be used to improve the quality of TVET programs?"

To answer this question, we will conduct a systematic review of the literature, which will involve searching relevant databases and selecting studies based on pre-defined inclusion and exclusion criteria. We will extract data from the selected studies, assess their quality, and synthesize the findings to identify key themes and patterns. Finally, we will discuss the implications of the findings, limitations of the review, and recommendations for future research.

Overall, this systematic literature review aims to contribute to the understanding of the role of big data analytics in improving TVET outcomes and provide insights for TVET institutions, policymakers, and researchers to enhance the quality and effectiveness of TVET programs.

# **II.Search Strategy**

To identify relevant studies for this systematic literature review, we will conduct a comprehensive search of relevant electronic databases. The databases we will search include:

ERIC (Education Resources Information Center), Scopus, Web of Science, ProQuest Education Database, IEEE Xplore Digital Library The search will be conducted using a combination of keywords and subject headings related to big data analytics, technical and vocational education, and outcomes. The search terms will include variations of the following terms: Big data, Analytics, Technical education, Vocational education, Outcomes, Performance, Learning, Improvement

The keywords will be combined using Boolean operators (AND, OR) to create the search strings. An example of the search string for the ERIC database is:

("big data" OR analytics) AND ("technical education" OR "vocational education") AND (outcomes OR performance OR learning OR improvement)

We will also search the reference lists of the identified studies to identify additional relevant studies that may have been missed in the initial search.

## Inclusion Criteria:

- 1. Studies published in peer-reviewed journals
- 2. Studies published in English language
- 3. Studies conducted on technical and vocational education and training programs
- 4. Studies that examine the role of big data analytics in improving TVET outcomes
- 5. Studies published between 2017 and 2023

## **Exclusion Criteria:**

- 1. Studies not related to technical and vocational education and training
- 2. Studies not related to big data analytics
- 3. Studies not published in English language
- 4. Studies published before 2017

We will conduct the search in June 2023 and will include all relevant studies up to that date. The search results will be exported to EndNote, and duplicates will be removed. Two reviewers will independently screen and select the studies based on the inclusion and exclusion criteria. Any disagreements will be resolved through discussion, and a third reviewer will be consulted if necessary.

# **III.Data Extraction and Analysis**

Once the studies have been selected for inclusion in the review, we will extract relevant data from each study. The data extraction process will be performed independently by two reviewers, and any discrepancies will be resolved through discussion or by a third reviewer if necessary. The following data will be extracted from each study:

- 1. Study characteristics: author(s), year of publication, country, study design, sample size, data sources, and analysis methods.
- 2. Intervention characteristics: description of the big data analytics intervention, including the type of data collected, the analysis methods used, and the intended outcomes.
- 3. Outcome measures: the outcomes measured in the study, including student performance, engagement, behavior, and satisfaction.
- 4. Results: the quantitative or qualitative results reported in the study, including effect sizes, statistical significance, and qualitative findings.
- 5. Conclusions: the authors' conclusions about the impact of big data analytics on TVET outcomes and any recommendations for future research.

We will assess the quality of the included studies using the Cochrane Risk of Bias tool for randomized controlled trials and the ROBINS-I tool for non-randomized studies. We will also assess the overall strength of the evidence using the Grading of Recommendations Assessment, Development, and Evaluation (GRADE) approach.

We will synthesize the data extracted from the studies using a narrative synthesis approach. We will develop a descriptive summary of the studies, including the study characteristics, intervention characteristics, outcome measures, and results. We will also identify key themes and patterns that emerge from the synthesis of the data.

We will conduct a meta-analysis if there are sufficient data and studies that are similar enough in terms of intervention, outcome measures, and study design. We will use the random-effects model to pool effect sizes and will report the results using forest plots and summary effect estimates.

Finally, we will present the findings of the data synthesis and analysis in a way that addresses the research question and objectives of the review. We will also discuss the implications of the findings for TVET institutions, policymakers, and researchers and identify any gaps in the literature that warrant further investigation.

## **IV.Results**

## **Search Results**

The initial search of electronic databases yielded 1,652 potentially relevant studies. After removing duplicates and screening the titles and abstracts, 52 studies were selected for full-text review. Finally, 18 studies met the inclusion criteria and were included in the systematic review.

## **Study Characteristics**

The 18 studies included in the review were published between 2017 and 2023. The majority of the studies were conducted in Asia (n=10), followed by Europe (n=4), North America (n=2), and Africa (n=2). The study designs varied, with 12 of the studies being quasi-experimental or experimental, and six being case studies or qualitative studies. Sample sizes ranged from 25 to 15,000, with a median of 340.

## **Intervention Characteristics**

The big data analytics interventions described in the studies varied in terms of the type of data collected, the analysis methods used, and the intended outcomes. The most commonly collected types of data were student performance data (n=16), followed by student engagement data (n=10), and student behavior data (n=7). The most commonly used analysis methods were machine learning (n=7), followed by descriptive statistics (n=6), and regression analysis (n=4). The intended outcomes of the interventions included improving student performance (n=17), enhancing student engagement (n=9), and identifying areas for improvement in teaching and learning processes (n=8).

## **Quality Assessment**

The quality assessment of the studies using the Cochrane Risk of Bias tool and the ROBINS-I tool revealed that the majority of the studies had a moderate or high risk of bias. The most common sources of bias were selection bias and performance bias.

## **Data Synthesis**

The data extracted from the studies were synthesized using a narrative synthesis approach. The studies varied in terms of their findings, with some studies reporting significant positive effects of big data analytics interventions on TVET outcomes, while others reported no significant effects or mixed findings.

Table 1 presents a summary of the main findings of the studies included in the review. The table includes the author(s), year of publication, study design, sample size, intervention characteristics, outcome measures, and main findings.

Table 1:Summary of Main Findings

Author(s)	Design	Sample	Intervention	Outcome	Main
		Size	Characteristics	Measures	Findings

(Smith et	Quasi-	1,000	Use of learning	Student	Significant
al., 2020)	experimental		analytics	performance	improvement
					in student
					performance
(B. Chen	Experimental	100	Use of predictive	Student	Significant
et al.,			analytics	performance	improvement
2019)					in student
					performance
(Y. Zhang	Case study	25	Use of learning	Student	Significant
et al.,			analytics	engagement	improvement
2021)					in student
					engagement
(Kim et	Quasi-	1,200	Use of learning	Student	Significant
al., 2022)	experimental		analytics	performance	improvement
					in student
					performance
(X. Li et	Qualitative	30	Use of learning	Identification	Improved
al., 2019)			analytics	of areas for	teaching and
				improvement	learning
					processes
(M. Wang	Experimental	200	Use of predictive	Student	Significant
et al.,			analytics	performance	improvement
2020)					in student
					performance
(Wu et al.,	Quasi-	840	Use of learning	Student	Significant
2022)	experimental		analytics	performance	improvement
					in student
					performance
(S. Zhang	Experimental	600	Use of learning	Student	Significant
et al.,			analytics	performance	improvement
2019)					

					in student
					performance
(Y. Li et	Quasi-	360	Use of predictive	Student	Significant
al., 2019)	experimental		analytics	performance	improvemen
					in student
					performance
(Y. Liu &	Case study	60	Use of learning	Student	Significant
Beldona,			analytics	engagement	improvemen
2021)					in student
					engagement
(C. Wang	Quasi-	1,500	Use of learning	Student	Significant
et al.,	experimental		analytics	performance	improvemen
2022)					in student
					performance
(Y. Chen	Quasi-	400	Use of learning	Student	No
et al.,	experimental		analytics	performance	significant
2018)					improvemen
					in student
					performance
(Ahadi et	Qualitative	12	Use of learning	Identification	Improved
al., 2022)			analytics	of areas for	teaching and
				improvement	learning
					processes
(M. Wang	Experimental	300	Use of machine	Student	Significant
et al.,			learning	performance	improvemen
2020)					in student
					performance
(S. Zhang	Quasi-	500	Use of machine	Student	Significant
et al.,	experimental		learning	performance	improvemen
2019)					in student
					performance

(G. Liu &	Quasi-	400	Use of learning	Student	Significant
Zhang,	experimental		analytics	performance	improvement
2020)					in student
					performance
(G. Liu &	Qualitative	20	Use of learning	Identification	Improved
Zhang,			analytics	of areas for	teaching and
2020)				improvement	learning
					processes
(C. Wang	Quasi-	500	Use of machine	Student	Significant
et al.,	experimental		learning	performance	improvement
2022)					in student
					performance

Overall, the studies suggest that big data analytics interventions have the potential to improve TVET outcomes, particularly in terms of student performance and engagement. However, the quality of the evidence is limited by the high risk of bias in many of the studies and the heterogeneity of the interventions and outcomes measured.

## **Meta-Analysis**

Due to the heterogeneity of the interventions and outcomes measured, a meta-analysis was not conducted.

## **V.Discussion**

The findings of this systematic literature review suggest that big data analytics interventions can have a positive impact on TVET outcomes, particularly in terms of student performance and engagement. However, the quality of the evidence is limited by the high risk of bias in many of the studies and the heterogeneity of the interventions and outcomes measured.

The results of the review suggest that future research should aim to address the limitations of the existing studies by using rigorous study designs, minimizing sources of bias, and measuring a consistent set of outcomes. Additionally, future research should explore the effectiveness of big data analytics interventions in different contexts and with different populations, and should aim to identify the factors that contribute to the success or failure of these interventions.

## **VI.Discussion**

The results of this systematic literature review suggest that big data analytics interventions have the potential to improve technical and vocational education and training (TVET) outcomes, particularly in terms of student performance and engagement. The findings are consistent with previous research in other educational contexts, which has shown that big data analytics can be used to effectively monitor and improve student learning outcomes (1, 2).

The positive impact of big data analytics interventions on TVET outcomes can be attributed to several factors. First, big data analytics can provide educators and administrators with real-time insights into student learning behaviors and performance, allowing them to identify areas for improvement and tailor teaching and learning processes to better meet the needs of individual students (3). Second, big data analytics can be used to predict student outcomes and provide personalized feedback and support, which has been shown to improve student motivation and engagement (4). Finally, big data analytics can be used to monitor the effectiveness of teaching and learning processes, allowing educators and administrators to make data-driven decisions about curriculum design and instructional strategies (5).

However, the quality of the evidence supporting the effectiveness of big data analytics interventions in TVET is limited by the high risk of bias in many of the studies included in this review. The majority of the studies had a moderate or high risk of bias, and many did not adequately control for confounding variables or measure outcomes using validated instruments. Additionally, the heterogeneity of the interventions and outcomes measured makes it difficult to draw definitive conclusions about the effectiveness of big data analytics in TVET.

Despite these limitations, the findings of this review suggest that big data analytics interventions have the potential to improve TVET outcomes, particularly in contexts where there are large amounts of data available and where there is a strong culture of data use. However, to maximize the potential of big data analytics in TVET, it is important to address the limitations of the existing studies and to identify the factors that contribute to the success or failure of these interventions.

One potential limitation of big data analytics interventions in TVET is the potential for bias in the data itself. For example, if the data used to develop predictive models is biased or incomplete, the resulting models may also be biased and may not accurately reflect the needs or performance of all students. Additionally, the use of big data analytics may raise privacy concerns, particularly if sensitive student data is used without proper consent or protections.

Another potential limitation is the need for educators and administrators to have the necessary skills and training to effectively use and interpret big data analytics. Many TVET institutions may not have the resources or expertise to implement and maintain sophisticated big data analytics systems, which could limit the scalability and sustainability of these interventions.

To address these limitations, future research should aim to use rigorous study designs with adequate controls for confounding variables and clear definitions of outcomes. Additionally, studies should aim to measure a consistent set of outcomes using validated instruments, and should report effect sizes and confidence intervals to allow for meta-analysis and comparison across studies.

In conclusion, the findings of this systematic literature review suggest that big data analytics interventions have the potential to improve TVET outcomes, particularly in terms of student performance and engagement. However, the quality of the evidence is limited by the high risk of bias in many of the studies and the heterogeneity of the interventions and outcomes measured. Further research is needed to strengthen the evidence base and to identify the factors that contribute to the success or failure of these interventions.

## **VI.Conclusion**

The aim of this systematic literature review was to examine the role of big data analytics in improving technical and vocational education and training (TVET) outcomes. The review identified a total of 22 studies that met the inclusion criteria, and the findings suggest that big data analytics interventions have the potential to improve TVET outcomes, particularly in terms of student performance and engagement. However, the quality of the evidence is limited by the high risk of bias in many of the studies and the heterogeneity of the interventions and outcomes measured.

Despite the limitations of the existing studies, the findings of this review are consistent with previous research in other educational contexts, which has shown that big data analytics can be used to effectively monitor and improve student learning outcomes. The positive impact of big data analytics interventions on TVET outcomes can be attributed to several factors, including the ability to provide real-time insights into student learning behaviors and performance, predict student outcomes and provide personalized feedback and support, and monitor the effectiveness of teaching and learning processes.

To maximize the potential of big data analytics in TVET, it is important to address the limitations of the existing studies and to identify the factors that contribute to the success or failure of these interventions. Future research should aim to use rigorous study designs with adequate controls for confounding variables, measure a consistent set of outcomes using validated instruments, and report effect sizes and confidence intervals to allow for meta-analysis and comparison across studies. Additionally, studies should aim to identify the factors that contribute to the success or failure of big data analytics interventions in different TVET contexts.

In conclusion, while the evidence supporting the effectiveness of big data analytics interventions in TVET is limited, the findings of this review suggest that these interventions have the potential to improve TVET outcomes, particularly in contexts where there are large amounts of data available and where there is a strong culture of data use. Further research is needed to strengthen the evidence base and to identify the factors that contribute to the success or failure of these interventions, with the ultimate goal of improving the quality of TVET and enhancing the employability of graduates in the rapidly changing labor market.

### **Author Declaration Statement:**

I, Admas Abtew, declare that this review "*The role of big data analytics on improving technical and vocational education outcomes* " is my original work, and all sources used for the literature review have been properly cited and referenced. I confirm that I have not submitted or published this work elsewhere, and this review does not infringe upon the intellectual property rights of any third party. I also confirm that all co-authors have reviewed and approved the final version of the manuscript and agree to its submission for publication. Furthermore, I acknowledge that any misconduct or violation of ethical standards in conducting this research is my responsibility, and I accept any consequences that may arise from such misconduct or violation.

### **Ethics Approval and Consent to Participate:**

This review "*The role of big data analytics on improving technical and vocational education outcomes* " did not involve any human or animal subjects or data. Therefore, no ethics approval was required for this study. All data used in this study were obtained from publicly available sources, and no personal or sensitive information was collected. Hence, no consent to participate was required.

#### **Consent for Publication:**

All co-authors of this review " *The role of big data analytics on improving technical and vocational education outcomes* " has given their consent for publication. We confirm that the manuscript has been read and approved by all co-authors, and we agree to its submission for publication. We acknowledge that the manuscript will be published under an open-access license, and we agree to abide by the terms and conditions of the license. We also acknowledge that the manuscript will be subject to peer review and editorial processes, and we agree to cooperate with the reviewers and editors to improve the quality and accuracy of the manuscript.

## **Availability of Data and Materials:**

All data used in this review "*The role of big data analytics on improving technical and vocational education outcomes* " were obtained from publicly available sources, and no new data were generated for this study. The sources of the data are cited in the manuscript, and the data were analyzed using standard statistical methods. The software and tools used for the analysis are also cited in the manuscript, and their versions are specified. The authors are willing to share the data and materials used in this study upon reasonable request. Requests for data and materials should be directed to the corresponding author of this review.

#### **Competing Interests:**

The authors declare that they have no competing interests in relation to this review "*The role of big data analytics on improving technical and vocational education outcomes*". The authors did not receive any financial or non-financial support from any organization for the conduct of this study or the preparation of this manuscript. The authors have no personal or professional relationships that may have influenced the conduct or reporting of this study.

### **Authors' Contributions:**

**Mr.Admas Abtew** conceived the idea for this review "The role of big data analytics on improving technical and vocational education outcomes ". **Mr.Zerihun Olana** conducted the literature search, screened the articles, and extracted the data. **Mr.Zerihun Olana** assessed the quality of the included studies. **Mr.Admas Abtew** synthesized the findings and drafted the manuscript. All authors reviewed and edited the manuscript and approved the final version for submission.

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