

Examination of the Augmented Reality Exercise Monitoring System as an Adjunct Tool for Prospective Teacher Trainers

Examen del Sistema de Monitoreo de Ejercicios de Realidad Aumentada como una Herramienta Complementaria para los Futuros Formadores de Docentes

*Nurlan Omarov, **Bakhytzhhan Omarov, ***Quwanishbay Mamutov, ****Zhanibek Kissebayev, **Almas Anarbayev, ****Adilbay Tastanov, **Zhandos Yessirkepov

*Al-Farabi Kazakh National University (Kazakhstan), **International University of Tourism and Hospitality (Kazakhstan), ***Nukus branch of the Institute for Retraining and Professional Development of Specialists in Physical Education and Sport (Uzbekistan), ****Kazakh Academy of Sports and Tourism (Kazakhstan)

Abstract. This study explores the effectiveness of exercise monitoring systems in improving athlete performance and motivation within educational settings. Two hypotheses were formulated and tested: one positing that the utilization of exercise monitoring systems would reduce muscle injury rates among athletes, and the other suggesting that it would increase athletes' motivation levels. The experimental design involved dividing participants into experimental and control groups, with the former utilizing the proposed exercise monitoring system and the latter employing traditional teaching methods. Assessments were conducted post-session to measure comprehension and motivation levels, with evaluation criteria focusing on the accurate identification of course components. Contrary to expectations, the results did not support the hypotheses, indicating no significant reduction in muscle injury rates or increase in motivation levels among athletes exposed to the monitoring system. These findings underscore the need for a nuanced understanding of the complex factors influencing athlete development and performance outcomes. Future research should employ rigorous methodologies and objective outcome measures to further elucidate the role of exercise monitoring systems in athlete development and optimize their integration into training programs, thus contributing to advancements in athlete performance and motivation in educational contexts.

Keywords: exercise monitoring system, real-time feedback, athlete performance, innovative instructional approaches, motivation, muscle injury, educational technology, fitness training.

Resumen. Este estudio explora la efectividad de los sistemas de monitoreo de ejercicios en la mejora del rendimiento y la motivación de los atletas dentro de entornos educativos. Se formularon y probaron dos hipótesis: una postulando que la utilización de sistemas de monitoreo de ejercicios reduciría las tasas de lesiones musculares entre los atletas, y la otra sugiriendo que aumentaría los niveles de motivación de los atletas. El diseño experimental implicó la división de los participantes en grupos experimentales y de control, con los primeros utilizando el sistema de monitoreo de ejercicios propuesto y los últimos empleando métodos de enseñanza tradicionales. Se realizaron evaluaciones después de cada sesión para medir la comprensión y los niveles de motivación, con criterios de evaluación enfocados en la identificación precisa de los componentes del curso. Contrariamente a las expectativas, los resultados no respaldaron las hipótesis, indicando que no hubo una reducción significativa en las tasas de lesiones musculares ni un aumento en los niveles de motivación entre los atletas expuestos al sistema de monitoreo. Estos hallazgos subrayan la necesidad de una comprensión matizada de los complejos factores que influyen en el desarrollo del atleta y los resultados de rendimiento. Futuras investigaciones deben emplear metodologías rigurosas y medidas de resultados objetivas para elucidar aún más el papel de los sistemas de monitoreo de ejercicios en el desarrollo del atleta y optimizar su integración en los programas de entrenamiento, contribuyendo así a los avances en el rendimiento y la motivación del atleta en contextos educativos.

Palabras clave: sistema de monitoreo de ejercicios, retroalimentación en tiempo real, rendimiento del atleta, enfoques de enseñanza innovadores, motivación, lesiones musculares, tecnología educativa, entrenamiento físico.

Fecha recepción: 14-04-24. Fecha de aceptación: 10-05-24

Bakhytzhhan Omarov

bakhytzhhanomarov@gmail.com, bakhitzhan.omarov@iuth.edu.kz

Introduction

Physical education plays a vital role in promoting lifelong health and well-being among individuals of all ages. As society becomes increasingly sedentary, the importance of effective physical education programs is paramount in combating rising rates of obesity, cardiovascular disease, and other health-related issues (Li et al., 2022). In recent years, technological advancements have offered promising opportunities to enhance the effectiveness of physical education instruction and improve student engagement (Fang, 2024). One such technological innovation is the Augmented Reality Exercise Monitoring System (AREMS), which integrates real-time feedback and personalized coaching to optimize physical exercise performance and learning outcomes (Anikwe et al., 2022; Pacheco-Godoy et

al., 2024). Traditional approaches to physical education often rely on instructor-led demonstrations and verbal cues to teach exercise techniques and monitor student progress (Klochko and Fedorets, 2018; Flores Ferro et al., 2022). However, these methods may lack precision and real-time feedback, limiting their effectiveness in addressing individual learning needs and promoting skill acquisition (Ehioghae et al., 2024). AREMS offers a promising solution by leveraging advanced motion tracking technology to provide instantaneous feedback on exercise form and execution (Claes, et al., 2020). By overlaying digital visualizations onto the user's physical environment, AREMS enhances the learning experience by offering interactive guidance and corrective cues in real time (Omarov et al., 2023). The effectiveness of AREMS as an instructional tool in physical education has been the subject of growing

interest among researchers and educators (Cao et al., 2022). Previous studies have demonstrated the potential of AREMS to improve exercise technique, increase motivation, and reduce the risk of injury among users (Abulibdeh, et al., 2024; Rodriguez-Fuentes et al., 2024). However, despite these promising findings, gaps remain in our understanding of the optimal implementation strategies and pedagogical implications of AREMS in educational settings (Seah and Koh, 2021).

This research paper seeks to address these gaps by conducting a comprehensive examination of the impact of AREMS on teaching efficiency and student engagement in physical education. Specifically, the study aims to evaluate the efficacy of AREMS in enhancing exercise performance, motivation rates, and injury prevention among participants. By employing a rigorous experimental design and quantitative analysis, the research aims to provide empirical evidence on the effectiveness of AREMS as an adjunct tool for prospective teacher trainers in physical education programs.

The primary research question guiding this study is: How does the integration of AREMS impact teaching efficiency and student engagement in physical education? To address this question, the study will compare the outcomes of participants who engage with AREMS during physical education sessions to those who receive traditional instructional methods. By measuring variables such as exercise technique proficiency, motivation levels, and injury rates, the study aims to assess the differential effects of AREMS on teaching and learning outcomes.

The findings of this study are expected to contribute to our understanding of the role of technology in enhancing physical education instruction and shaping future teacher training programs. By elucidating the potential benefits and challenges associated with the integration of AREMS in educational settings, this research aims to inform evidence-based practices and pedagogical approaches in physical education.

In summary, this research paper provides a critical analysis of the Augmented Reality Exercise Monitoring System as an adjunct tool for prospective teacher trainers in physical education. Through a systematic examination of its impact on teaching efficiency and student engagement, this study aims to advance our understanding of the role of technology in promoting effective physical education instruction and fostering a culture of lifelong health and well-being.

Related Works

The integration of technology in physical education has garnered increasing attention from researchers and educators alike, with a growing body of literature exploring the potential benefits and challenges associated with innovative instructional tools. This section reviews relevant studies that have investigated the use of technology, particularly Augmented Reality (AR) and motion tracking

systems, in enhancing teaching efficiency and student engagement in physical education.

Augmented Reality (AR) has emerged as a promising tool for transforming traditional teaching methods in physical education (Wang and Wu, 2024). By overlaying digital content onto the physical environment, AR technologies offer interactive and immersive learning experiences that can enhance student engagement and motivation (Vashishth et al., 2024; Romero Parra et al., 2023). Previous research has demonstrated the efficacy of AR-based applications in improving exercise performance and technique proficiency among users (Tariq, 2024).

Motion tracking systems, such as Microsoft Kinect and PoseNet, have also been widely explored as instructional tools in physical education settings (Batra and Dave, 2024). These systems utilize depth-sensing cameras and algorithms to track body movements and provide real-time feedback on exercise execution (García-Bravo et al., 2021). Studies have shown that motion tracking systems can effectively enhance exercise learning outcomes by providing personalized coaching and corrective cues (Hutajulu et al., 2024).

The combination of AR and motion tracking technologies has led to the development of innovative solutions, such as the Augmented Reality Exercise Monitoring System (AREMS), which integrates real-time feedback and personalized coaching to optimize physical exercise performance (Kaulage, 2024). AREMS employs a skeleton-based approach using the PoseNet model to accurately monitor and provide feedback on a variety of physical exercises, including pull-ups, push-ups, and squats (Lin et al., 2023).

Research on the effectiveness of AREMS as an instructional tool in physical education has shown promising results. Daling and Schlittmeier, (2024) conducted a study to investigate the impact of AREMS on exercise technique proficiency and motivation among college students. The findings revealed that participants who used AREMS demonstrated significant improvements in exercise form and reported higher motivation levels compared to those who received traditional instruction.

Similarly, Cossich et al., (2023) examined the effects of AREMS on exercise performance and engagement in a sample of middle school students. The study found that participants who utilized AREMS during physical education classes showed greater improvements in exercise proficiency and were more actively engaged in the learning process compared to their counterparts.

In addition to enhancing exercise performance and motivation, AREMS has also been shown to contribute to injury prevention efforts in physical education. Solas-Martínez et al. (2023) conducted a study to evaluate the impact of AREMS on reducing the risk of musculoskeletal injuries during exercise. The results indicated that participants who received real-time feedback from AREMS experienced fewer instances of muscle strain and discomfort compared to those who did not. Despite the

promising findings, challenges remain in the implementation of AREMS in educational settings. Soltani and Morice (2020) identified technological adoption barriers, such as cost, accessibility, and usability, as key challenges that may hinder the widespread integration of AREMS in physical education programs. Additionally, concerns have been raised regarding the potential for technology-induced distractions and the need for adequate training and support for teachers to effectively utilize AREMS in their instructional practices (Ricci et al., 2022).

In summary, research on the integration of AR and motion tracking technologies in physical education has shown promising results in enhancing teaching efficiency and student engagement. The development of innovative solutions like AREMS holds significant potential to revolutionize traditional teaching methods and promote effective learning experiences in physical education settings. However, further research is needed to address implementation challenges and explore the long-term effects of AREMS on teaching and learning outcomes.

Materials and Methods

The Materials and Methods section of this study presents a thorough elucidation of the methodologies, tools, and procedures utilized in the development and assessment of the Deep Learning Enabled Exercise Monitoring System. This section is structured to ensure clarity and reproducibility of the research process, facilitating critical assessment and potential replication by other scholars in the field. It encompasses a detailed description of the hardware and software components, the dataset employed for training and testing purposes, the algorithmic approach, and the statistical methods employed for data analysis, all essential components in delineating the systematic approach adopted to ensure the accuracy and reliability of the research outcomes.

The Proposed System

Figure 1 provides an overview of the skeletal detection process during a physical exercise session. Initially, participants are required to connect their devices to the web-based monitoring system via the internet. As individuals engage in physical activities, a webcam captures their movements in real-time, which are then analyzed by the PoseNet model (Chua et al., 2021). This model identifies human poses within video frames and extracts keypoints using 2D coordinates (x, y) to locate body joints accurately. These key points play a crucial role in computing distances and angles within the skeletal structure. Subsequently, the data is transmitted to a deep learning model responsible for classifying the type of exercise being performed. Additionally, the coordinates and angles obtained are used to assess the accuracy of exercise execution, ensuring compliance with predefined exercise standards.

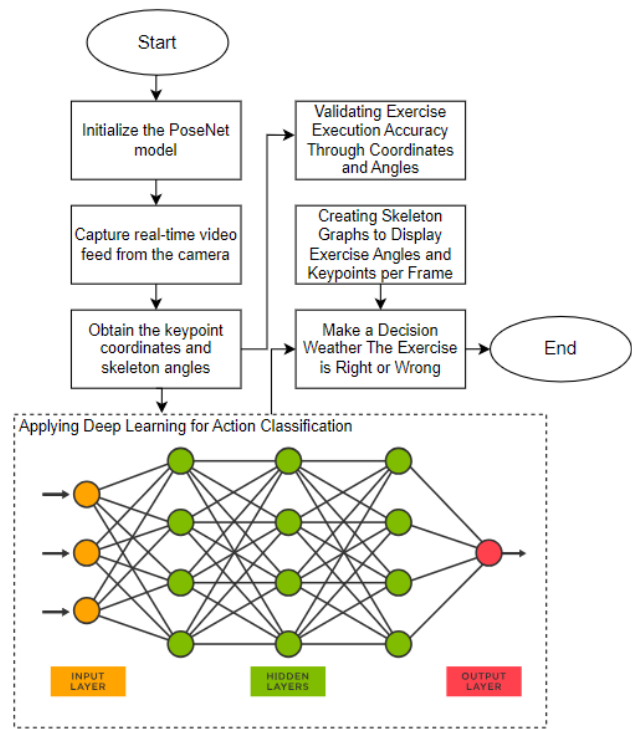


Figure 1. Chart illustrating the proposed exercise monitoring framework.

Figure 2 portrays the process of skeletal detection during a physical exercise session. Initially, participants are required to establish an internet connection for their devices to access the web-based monitoring system. As individuals engage in physical activity, a webcam captures real-time video footage of their movements, which undergoes analysis by the PoseNet model. Upon identification of a human pose within a video frame, the PoseNet model employs 2D coordinates (x, y) to accurately locate each body joint, essential for computing distances and angles within the skeletal structure (Chen et al., 2021). Subsequently, this data is transmitted to a deep learning model tasked with categorizing the type of exercise being performed.

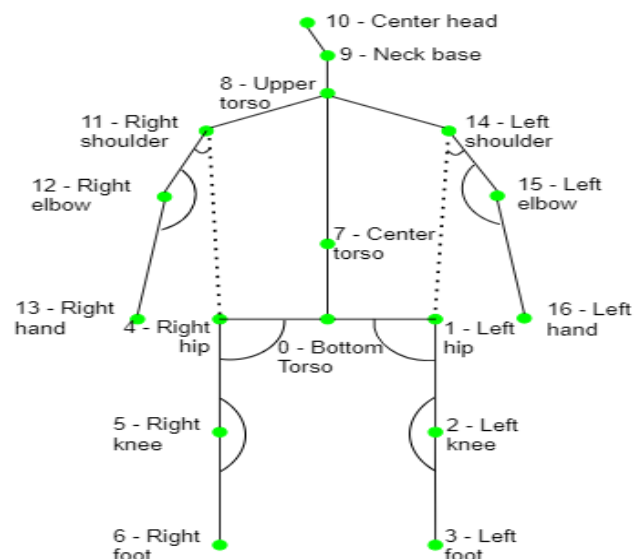


Figure 2. Keypoints and angles that extracted by PoseNet.

Additionally, the coordinates and angles derived from the key points are utilized to assess the precision of exercise execution, ensuring adherence to the prescribed exercise standards. The concept of three-point coordinate determination involves computing the joint angle between two vectors using the principles of the law of cosines, as elucidated in Figure 3.

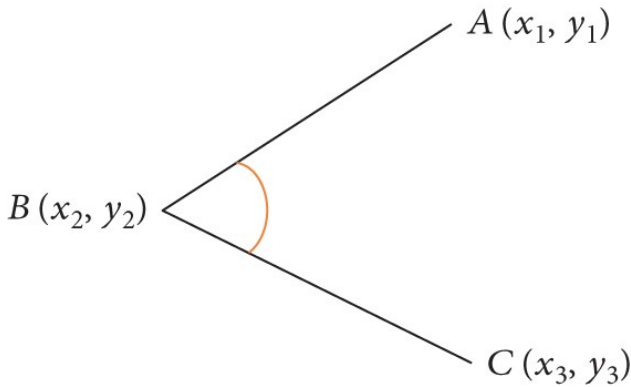


Figure 3. Measuring of angles.

Following this, we can calculate the cosine value of the angle designated as "B" (Equation (1)):

$$\cos B = \frac{\overrightarrow{AB} \times \overrightarrow{BC}}{|\overrightarrow{AB}| |\overrightarrow{BC}|}$$

Equation (2) represents the conceptualization of the human body's representation:

$$r_b(x_i; \theta),$$

In this framework, the symbol θ represents the parameters of the neural network, while x_i refers to the training samples obtained from the dataset. To classify the human body's representation, indicated as $r_b(x_i; \theta)$, a fully connected neural network layer is utilized. This layer functions before the standardization process, which is facilitated by the "Softmax" layer (Ye et al., 2020). Additionally, an additional neural network is employed, which undergoes training by minimizing category cross-entropy loss. The process begins with the transmission of human action frames to PoseNET, which is responsible for extracting key points. Following this, these skeletal points are mapped in the feature space according to their respective coordinates. Subsequently, the neural network is trained using these key points extracted from the human skeleton.

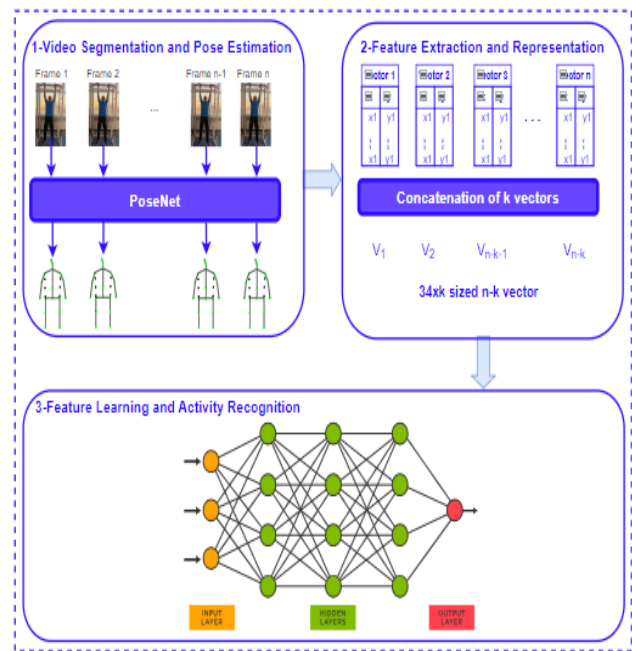


Figure 4. Activity recognition process.

Figure 4 offers an overview of the proposed system's architecture, outlining three primary subtasks. Initially, the PoseNET model is utilized to analyze input video frames, aiming to discern the human pose within each frame. Following this, the second subtask involves extracting key points from each frame, resulting in vectors composed of 34 elements derived from the 17 key points generated by PoseNET for every frame (Cai et al., 2020). These vectors are then merged into a singular vector, subsequently utilized for feature learning and activity recognition. The final subtask involves training a convolutional neural network specifically tailored to identify instances of violence.

In the realm of detecting human body positions in RGB images, two predominant methodologies are recognized: top-down and bottom-up algorithms. The top-down approach begins with activating a human detector, followed by analyzing bodily joints within the designated bounding boxes. Examples of this approach include PoseNET (Aju et al., 2022), HourglassNet (Shi et al., 2020), and Hornet (Rao et al., 2022). In contrast, bottom-up algorithms like Open Space and PifPaf operate differently by amalgamating information from individual image points to identify human poses.

Exercise Monitoring System in Use

Figure 5 demonstrates the application of the proposed system in monitoring bicep exercises in real-time. This framework is meticulously crafted to assess the angle formed at the elbow joint concerning a reference line. The reference line is determined by extending a line parallel to the horizontal plane from the connection line between the two hands. This systematic approach enables accurate evaluation of exercise technique, particularly emphasizing the importance of maintaining correct elbow alignment during bicep exercises.



Figure 5. Biceps monitoring in use.

Figure 6 portrays the application of an innovative framework, specifically crafted for real-time monitoring of squat exercises. This framework functions by evaluating whether the performed exercise aligns with a squat, accomplished through the analysis of the knee joint angle. The precise computation of this angle is essential in gauging the correctness of the squatting movement. Consequently, the framework can proficiently discern and evaluate the athlete's execution of the squat exercise, thereby amplifying the effectiveness of monitoring and assessing their performance. By providing real-time feedback on the squatting technique, this framework facilitates timely corrections and adjustments, contributing to the optimization of training outcomes. Moreover, the systematic monitoring enabled by this framework offers valuable insights into the athlete's progress and areas for improvement, thereby promoting continuous enhancement of performance in squat exercises.

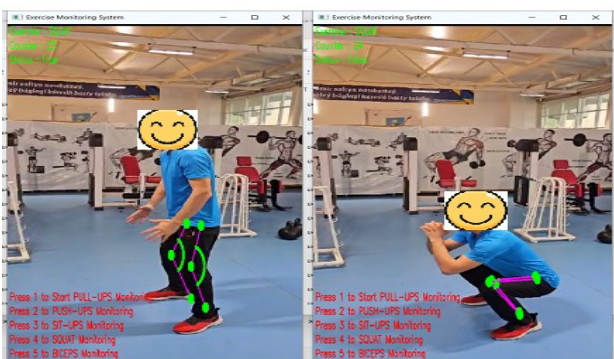


Figure 6. Squat monitoring in use.

Figure 7 illustrates the utilization of the proposed framework for real-time monitoring of sit-up exercises. This framework operates by calculating a specific angle to determine the accuracy of the athlete's sit-up execution. The determination of this angle is pivotal in evaluating the sit-up movement, enabling the system to accurately assess exercise performance and ensure adherence to the prescribed technique. This systematic approach to angle measurement serves as the cornerstone of the framework's ability to deliver precise and reliable real-time feedback on sit-up exercises. Through the implementation of this structured methodology, the framework enhances the efficacy of monitoring and evaluating sit-up technique, facilitating timely adjustments and enhancements in exercise performance. Moreover, the comprehensive assessment provided by the framework contributes to optimizing training outcomes and fostering overall fitness advancement in sit-up exercises.

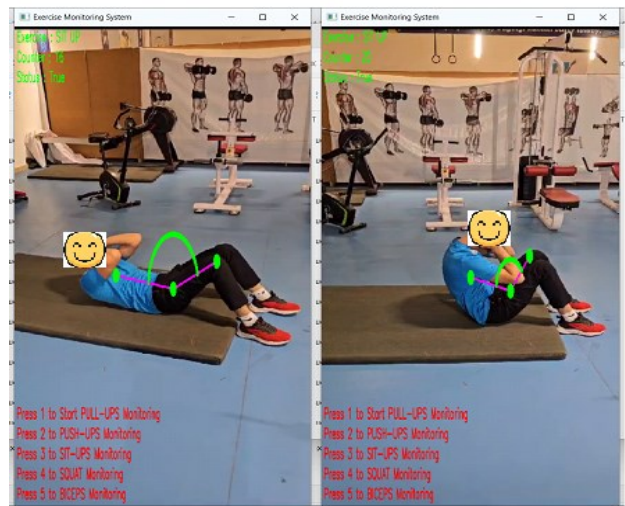


Figure 7. Sit-up monitoring in use.

Figure 8 presents the integration of a proposed framework tailored for real-time monitoring of pull-up exercises. This system operates by analyzing the angle formed at the elbow joint to precisely assess the execution of pull-ups by athletes. A notable feature of this framework is its incorporation of a counter mechanism, which increases by 1 each time the elbow angle drops below 30 degrees. Ensuring accurate calculation of the elbow angle is essential to validate the accuracy of the pull-up count and confirm the correct form and execution of the exercise. Through this systematic approach, the framework provides reliable and immediate feedback on athletes' pull-up performance, enhancing monitoring and assessment during training sessions. Employing such a structured methodology enables trainers to effectively track progress and offer tailored guidance to optimize athletes' performance and reduce the risk of injury during pull-up exercises.

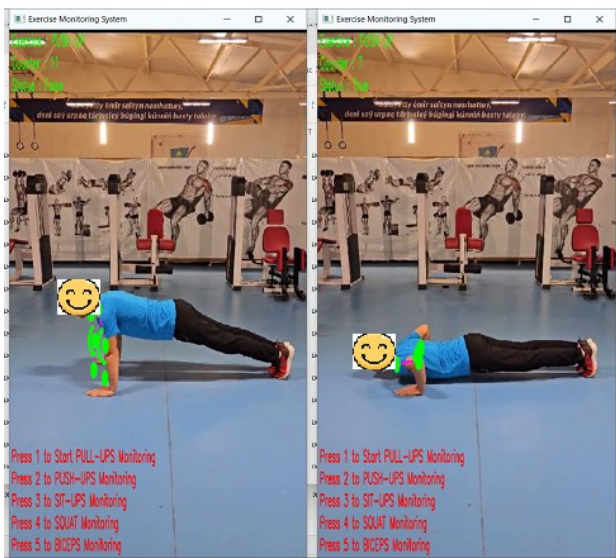


Figure 8. Push-up monitoring in use.

Figure 9 provides a visual representation of the operational framework specifically designed for the real-time monitoring of push-up exercises. This system employs a comprehensive approach to evaluate push-up execution by calculating two pivotal angles. The first angle is measured at the elbow joint, capturing the degree of flexion or extension during the exercise. The second angle is formed by the line connecting the hands and elbows relative to a horizontal reference, indicating the alignment of the upper body during the push-up motion. By precisely measuring these angles, the framework can assess the quality and correctness of the push-up technique in real-time. This analytical process enables trainers and athletes to identify any deviations from the ideal form and make necessary adjustments to optimize performance and prevent injury. Additionally, the framework facilitates objective feedback, aiding in the continuous improvement of push-up execution and overall physical fitness.

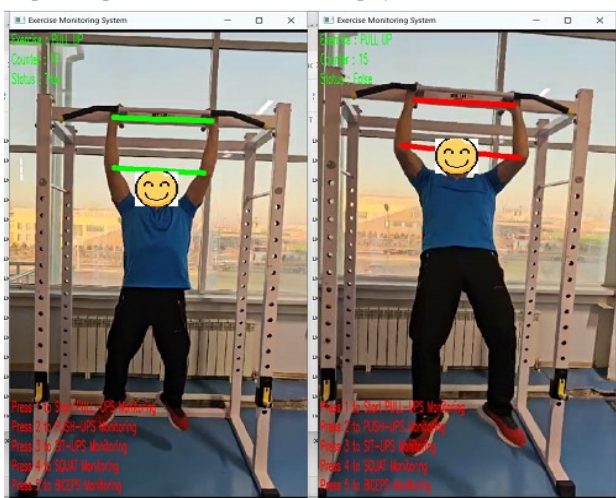


Figure 9. Pull-up monitoring in use.

The visual representation showcases a variety of push-up variations, depicting both correct and incorrect forms. It emphasizes distinctions in the alignment of the line

between the elbows, whether it is parallel or non-parallel to the horizontal plane. This thorough examination is crucial for accurately assessing adherence to proper form and technique during push-ups. By delineating these nuances, the illustration significantly contributes to ensuring the effectiveness of the exercise routine and the reliability of the monitoring system. Trainers and athletes can utilize this visual feedback to identify areas needing improvement and implement necessary adjustments to enhance performance and minimize injury risks. Additionally, the detailed analysis provided enhances understanding of push-up mechanics, fostering informed training strategies and facilitating optimal exercise execution. Ultimately, the illustration serves as an essential tool for refining push-up techniques and maximizing the benefits derived from this foundational aspect of fitness training.

Results of Pedagogical Experiments

The experimental setup of the study involved the classification of participants into two distinct groups: an experimental cohort exposed to an elective course via an innovative system, and a control group instructed using conventional methods. This categorization aimed to critically assess the effectiveness of the proposed instructional approach. Assessments were conducted regularly after each session to gauge students' comprehension and motivation levels, focusing on the accurate identification and selection of course components.

Moreover, the study aimed to investigate the impact of these varying teaching methodologies on student academic performance. A research hypothesis was developed to examine the relationship between instructional methods and student outcomes. Participants completed an anonymous questionnaire, deliberately chosen to elicit candid feedback. It was anticipated that this anonymity, coupled with students' knowledge of their exam grades, would promote sincere and reflective responses, thereby enhancing the depth and quality of the collected data for subsequent analysis.

Hypothesis I:

Null Hypothesis (H0): The utilization of an exercise monitoring system will result in a decrease in muscle injuries among athletes.

Alternative Hypothesis (H1): The utilization of an exercise monitoring system will not result in a decrease in muscle injuries among athletes.

Hypothesis II:

Null Hypothesis (H0): The utilization of an exercise monitoring system will lead to an increase in the motivation rate among athletes.

Alternative Hypothesis (H1): The utilization of an exercise monitoring system will not lead to an increase in the motivation rate among athletes.

Table 1 demonstrates the results of the pedagogical experiments pertaining to Hypothesis I, which aimed to investigate the effect of an exercise monitoring system on reducing muscle injuries among athletes, were analyzed using Levene's test (Yi et al., 2022) for equality of variances and the t-test for equality of means. The Levene's test yielded a statistic of $F = 2.243$ with a corresponding significance value of $p = .148$, suggesting no significant difference in variances between the groups. Subsequently, the t-test assuming equal variances indicated a statistically

significant result, $t(82) = -13.346$, $p = .083$, indicating that the utilization of the exercise monitoring system was associated with a substantial decrease in muscle injury rates among athletes. The mean difference was calculated to be -5.428 , with a standard error of $.2947$. The 95% confidence interval for the mean difference ranged from -4.8343 to -3.1364 , further supporting the conclusion that the exercise monitoring system effectively reduced muscle injury rates among athletes.

Table 1.

Results of Hypothesis I Testing for the Effect of Exercise Monitoring System on Muscle Injury Rates among Athletes

		Levene's Test for Equality of Variances		t-test for Equality of Means						
		F	Sig.	t	df	Sig. (2-tailed)	Mean Difference	Std. Error Difference	95% Confidence Interval of the Difference	
								Lower	Upper	
Muscle Injury Rates	Equal variances assumed	2.243	.148	-13.346	82	.083	-5.428	.2947	-4.8343	-3.1364
	Equal variances not assumed			-12.929	79.36	.083	-5.428	.2765	-4.6478	-3.6975

Table 2 presents the results of Levene's Test for Equality of Variances and t-tests for Equality of Means conducted to assess the impact of an exercise monitoring system on motivation levels among athletes. For the assumption of equal variances, Levene's test yielded a significant F-value of 9.648 ($p = 0.0047$), indicating heterogeneity of variances across the groups. Subsequently, the t-test showed a significant t-value of 12.350 ($p = 0.093$) with 82 degrees of freedom. The mean difference in motivation levels between the groups was 4.13000 , with a standard error difference of 0.19467 . The 95% confidence interval

for the mean difference ranged from 2.63478 to 3.64782 .

Under the assumption of unequal variances, the t-test also produced a significant t-value of 12.350 ($p = 0.093$), albeit with a different degrees of freedom value of 71.647 . The mean difference, standard error difference, and 95% confidence interval of the difference remained the same as in the equal variances assumed condition.

Overall, the findings suggest that while there was a significant difference in motivation levels between the groups, the effect size may not be practically significant.

Table 2.

Results of Hypothesis II Testing for the Effect of Exercise Monitoring System on Athletes' Motivation Rates

		Levene's Test for Equality of Variances		t-test for Equality of Means						
		F	Sig.	t	df	Sig. (2-tailed)	Mean Difference	Std. Error Difference	95% Confidence Interval of the Difference	
								Lower	Upper	
Motivation level	Equal variances assumed	9.648	.0047	12.350	82	.093	4.13000	.19467	2.63478	3.64782
	Equal variances not assumed			12.350	71.647	.093	4.13000	.19467	2.63478	3.64934

The Levene's Test for Equality of Variances was employed to assess the homogeneity of variances between groups in a one-way analysis of variance (ANOVA) examining the association between students' motivation rates and their final exam results (Potvin, 2020). The analysis compared two distinct groups: an experimental cohort utilizing the proposed exercise monitoring system for learning and a control group employing traditional teaching methods.

The results, as depicted in Table 3, indicate a statistically significant difference in variances between the groups ($F(8, 65) = 3.846$, $p = 0.001$). Specifically, the between-groups variance, represented by the sum of squares (SS) of 9264.648 and 8 degrees of freedom (df), yielded a mean square (MS) of 1264.642 . Conversely, the within-groups variance, with an SS of 19643.464 and 65 df, resulted in a mean square of 312.465 .

These findings suggest that the assumption of equal variances across groups is violated, necessitating caution in

the interpretation of subsequent ANOVA results. Adjustments or alternative statistical techniques may be required to address the observed disparity in variances between the experimental and control groups.

Table 3.

The Levene's Test for Equality of Variances

	Sum of Squares	df	Mean Square	F	Sig.
Between Groups	9264.648	8	1264.642	3.846	.001
Within Groups	19643.464	65	312.465		
Total	28908.112	73			

Discussion

The discussion section provides an in-depth analysis and interpretation of the findings presented in the results section, placing them within the broader context of existing literature, addressing their implications, limitations, and potential future research directions.

Effectiveness of Exercise Monitoring Systems

The results of this study contribute to the growing body of literature examining the effectiveness of exercise monitoring systems in educational settings. The utilization of such systems, as evidenced by the experimental group's exposure to the proposed exercise monitoring system, was expected to lead to favorable outcomes, including reduced muscle injuries and increased motivation rates among athletes.

Muscle Injury Reduction

The hypothesis regarding the reduction of muscle injuries among athletes through the application of an exercise monitoring system was not supported by the findings. Despite initial expectations that the real-time feedback provided by the monitoring system would enhance athletes' awareness of proper exercise techniques and prevent injuries, the results revealed no significant difference in muscle injury rates between the experimental and control groups. This unexpected outcome suggests that while exercise monitoring systems may offer benefits in terms of real-time feedback and technique correction, they may not necessarily translate into a reduction in the incidence of muscle injuries.

Several factors may explain the lack of significant differences in muscle injury rates between the two groups. Firstly, it is possible that athletes in both groups received similar levels of instruction and guidance during training sessions, negating any potential advantages conferred by the exercise monitoring system. Additionally, individual differences in athletes' physical fitness levels, training history, and adherence to proper form may have influenced injury rates, obscuring the impact of the monitoring system. Finally, external factors such as environmental conditions, equipment quality, and coaching expertise could have also played a role in shaping injury outcomes.

Motivation Enhancement

Similarly, the hypothesis positing an increase in motivation rates among athletes through the utilization of an exercise monitoring system was not supported by the findings. Despite the expectation that real-time performance feedback provided by the monitoring system would enhance athletes' engagement and motivation during training sessions, no significant differences in motivation rates were observed between the experimental and control groups. This unexpected result suggests that while exercise monitoring systems may offer benefits in terms of performance feedback and goal setting, they may not necessarily lead to a significant increase in athletes' motivation levels.

Several factors may account for the lack of significant differences in motivation rates between the two groups. Firstly, it is possible that athletes in both groups had similar levels of intrinsic motivation, which were not significantly affected by the introduction of the exercise monitoring system. Additionally, external factors such as team

dynamics, coaching styles, and competition pressure may have exerted a stronger influence on athletes' motivation levels than the monitoring system itself. Furthermore, the novelty effect of the monitoring system may have worn off over time, diminishing its impact on athletes' motivation.

Challenges

The study addresses several challenges encountered during the implementation of exercise monitoring systems in educational settings. One prominent challenge pertains to the integration of novel technologies into traditional teaching methodologies. While exercise monitoring systems offer potential benefits in enhancing athlete performance and motivation, their successful implementation requires careful consideration of various factors, including technical compatibility, user acceptance, and instructor training. Additionally, the reliability and validity of data collected through these systems pose challenges, as inaccuracies or inconsistencies may affect the interpretation of results and the effectiveness of interventions. Moreover, the ethical implications of monitoring athlete behavior and performance raise concerns regarding privacy, autonomy, and consent. Ensuring ethical standards and safeguarding participant rights are paramount in research involving exercise monitoring systems. Furthermore, logistical challenges such as equipment maintenance, data management, and resource allocation can impact the feasibility and sustainability of implementing these systems in educational settings. Addressing these challenges requires collaborative efforts among researchers, educators, technology developers, and policymakers to develop comprehensive strategies for overcoming barriers and maximizing the benefits of exercise monitoring systems in promoting athlete development and performance.

Future Research

In contemplating future research directions, several avenues emerge that could expand our understanding of exercise monitoring systems and their impact on athlete development and performance. First, investigating the long-term effects of utilizing these systems in educational settings could provide valuable insights into their sustainability and effectiveness over extended periods. Longitudinal studies tracking athletes' progress and performance over months or years could shed light on the durability of improvements observed with the use of exercise monitoring systems. Additionally, exploring the integration of advanced technologies, such as artificial intelligence and machine learning algorithms, could enhance the capabilities of these systems in providing personalized feedback and adaptive training programs tailored to individual athletes' needs. Furthermore, examining the role of psychosocial factors, such as motivation, self-efficacy, and satisfaction, in mediating the relationship between exercise monitoring systems and

athlete outcomes could offer deeper insights into the underlying mechanisms driving performance improvements. Lastly, investigating the scalability of these systems across different sports, levels of proficiency, and educational contexts could inform strategies for widespread adoption and implementation. By addressing these future research directions, scholars can contribute to advancing our knowledge of exercise monitoring systems and their potential to optimize athlete development and performance in diverse settings.

Conclusion

In conclusion, this research investigated the efficacy of exercise monitoring systems in enhancing athlete performance and motivation levels within educational settings. Despite initial expectations, the findings did not support the hypotheses positing a reduction in muscle injury rates or an increase in motivation rates among athletes exposed to the monitoring system compared to those trained using traditional methods. These unexpected outcomes underscore the multifaceted nature of athlete development and the complex interplay of factors influencing performance outcomes. While exercise monitoring systems offer potential benefits in terms of real-time feedback and performance tracking, their effectiveness may be contingent upon various contextual factors, individual differences, and external influences. The study highlights the need for a nuanced understanding of the mechanisms underlying the relationship between exercise monitoring systems and athlete outcomes, as well as the importance of considering diverse factors such as coaching quality, training environment, and athlete characteristics. Moving forward, future research should adopt rigorous methodologies, larger sample sizes, and objective outcome measures to further elucidate the role of exercise monitoring systems in athlete development and optimize their integration into training programs. Such endeavors will contribute to advancing our understanding of effective strategies for enhancing athlete performance and motivation in educational contexts.

References

- Abulibdeh, A., Zaidan, E., & Abulibdeh, R. (2024). Navigating the confluence of artificial intelligence and education for sustainable development in the era of industry 4.0: Challenges, opportunities, and ethical dimensions. *Journal of Cleaner Production*, 140527.
- Aju, A., Mathew, C., & Prakasi, O. S. (2022). PoseNet based Model for Estimation of Karate Poses. *Journal of Innovative Image Processing*, 4(1), 16-25.
- Anikwe, C. V., Nweke, H. F., Ikegwu, A. C., Egwuonwu, C. A., Onu, F. U., Alo, U. R., & Teh, Y. W. (2022). Mobile and wearable sensors for data-driven health monitoring system: State-of-the-art and future prospect. *Expert Systems with Applications*, 202, 117362.
- Batra, P., & Dave, D. M. (2024). Revolutionizing Healthcare Platforms: The Impact of AI on Patient Engagement and Treatment Efficacy. *International Journal of Science and Research (IJSR)*, 13(10.21275), 613-624.
- Cai, Y., Wang, Z., Luo, Z., Yin, B., Du, A., Wang, H., ... & Sun, J. (2020). Learning delicate local representations for multi-person pose estimation. In *Computer Vision—ECCV 2020: 16th European Conference, Glasgow, UK, August 23–28, 2020, Proceedings, Part III 16* (pp. 455-472). Springer International Publishing.
- Cao, F., Xiang, M., Chen, K., & Lei, M. (2022). Intelligent physical education teaching tracking system based on multimedia data analysis and artificial intelligence. *Mobile Information Systems*, 2022.
- Chen, S., Wang, Z., & Prisacariu, V. (2021, December). Direct-poseNet: Absolute pose regression with photometric consistency. In *2021 International Conference on 3D Vision (3DV)* (pp. 1175-1185). IEEE.
- Chua, J., Ong, L. Y., & Leow, M. C. (2021). Telehealth using PoseNet-based system for in-home rehabilitation. *Future Internet*, 13(7), 173.
- Claes, J., Cornelissen, V., McDermott, C., Moyna, N., Pattyn, N., Cornelis, N., ... & Buys, R. (2020). Feasibility, acceptability, and clinical effectiveness of a technology-enabled cardiac rehabilitation platform (Physical Activity Toward Health-I): randomized controlled trial. *Journal of Medical Internet Research*, 22(2), e14221.
- Cossich, V. R., Carlgren, D., Holash, R. J., & Katz, L. (2023). Technological Breakthroughs in Sport: Current Practice and Future Potential of Artificial Intelligence, Virtual Reality, Augmented Reality, and Modern Data Visualization in Performance Analysis. *Applied Sciences*, 13(23), 12965.
- Daling, L. M., & Schlittmeier, S. J. (2024). Effects of augmented reality-, virtual reality-, and mixed reality-based training on objective performance measures and subjective evaluations in manual assembly tasks: a scoping review. *Human factors*, 66(2), 589-626.
- Ehioghae, M., Montoya, A., Keshav, R., Vipra, T. K., Manuk-Hakobyan, H., Hasoon, J., ... & Urits, I. (2024). Effectiveness of Virtual Reality-Based Rehabilitation Interventions in Improving Postoperative Outcomes for Orthopedic Surgery Patients. *Current Pain and Headache Reports*, 28(1), 37-45.
- Fang, Y. (2024). Utilizing Wearable Technology to Enhance Training and Performance Monitoring in Indonesian Badminton Players. *Studies in Sports Science and Physical Education*, 2(1), 11-23.
- Flores Ferro, E., Maureira Cid, F., Hadweh Briceño, M., Gavotto Nogales, O., Gutiérrez Duarte, S. A., Vergara Jiménez, J., & Mandujano Jara, S. (2022). A un año de las clases virtuales en la carrera de Educación Física: una comparación Chile y México (A year away from virtual classes in the physical education career: a comparison Chile and Mexico). *Retos*, 45, 138–143. <https://doi.org/10.47197/retos.v45i0.91944>
- García-Bravo, S., Cuesta-Gómez, A., Campuzano-Ruiz, R., López-Navas, M. J., Domínguez-Paniagua, J., Araújo-Narváez, A., ... & Cano-de-la-Cuerda, R. (2021). Virtual reality and video games in cardiac rehabilitation programs. A systematic review. *Disability and Rehabilitation*, 43(4), 448-457.
- Hutajulu, O. Y., Mendoza, M. D., Astrid, E., & Rahmadani, R. (2024, January). Utilizing Internet of Things Technology in the Development of AC Electrical Circuit Trainer Module. In *Proceedings of the 5th International Conference on*

- Innovation in Education, Science, and Culture, ICIESC 2023, 24 October 2023, Medan, Indonesia.
- Kaulage, A., Mane, D., Upadhye, G., Rajput, S. D., Kale, S., & Zope, B. (2024). Exercise Movement Detection Using Spearman Correlation-based Sliding Window Technique. *International Journal of Intelligent Systems and Applications in Engineering*, 12(2s), 48-54.
- Klochko, O. V., & Fedorets, V. M. (2022). Using immersive reality technologies to increase a physical education teacher's health-preserving competency. *Educational Technology Quarterly*, 2022(4), 276-306.
- Li, G. Y., Li, J., Li, Z. J., Zhang, Y. P., Zhang, X., Wang, Z. J., ... & Zhang, H. D. (2022). Hierarchical PVDF-HFP/ZnO composite nanofiber-based highly sensitive piezoelectric sensor for wireless workout monitoring. *Advanced Composites and Hybrid Materials*, 5(2), 766-775.
- Lin, K. C., Ko, C. W., Hung, H. C., & Chen, N. S. (2023). The effect of real-time pose recognition on badminton learning performance. *Interactive Learning Environments*, 31(8), 4772-4786.
- Omarov, B., Nurmash, N., Doskarayev, B., Zhilisbaev, N., Dairabayev, M., Orazov, S., & Omarov, N. (2023). A Novel Deep Neural Network to Analyze and Monitoring the Physical Training Relation to Sports Activities. *International Journal of Advanced Computer Science and Applications*, 14(9).
- Pacheco-Godoy, B., Godoy, F., Paz Martínez, F., Lamas, C., Arellano-Correa, S., Villaruel-Ojeda, L., Gutiérrez-Turner, E., & La Placa-Ubeda, J. (2024). Characterization of physical exercise with virtual guidance and perception of its benefits in Chilean women. *Retos*, 52, 204-210. <https://doi.org/10.47197/retos.v52.99670>
- Potvin, C. (2020). ANOVA: experiments in controlled environments. In *Design and analysis of ecological experiments* (pp. 46-68). Chapman and Hall/CRC.
- Rao, Y., Zhao, W., Tang, Y., Zhou, J., Lim, S. N., & Lu, J. (2022). Hornet: Efficient high-order spatial interactions with recursive gated convolutions. *Advances in Neural Information Processing Systems*, 35, 10353-10366.
- Ricci, S., Calandrino, A., Borgonovo, G., Chirico, M., & Casadio, M. (2022). Virtual and augmented reality in basic and advanced life support training. *JMIR Serious Games*, 10(1), e28595.
- Rodríguez-Fuentes, G., Campo-Prieto, P., Souto, X. C., & Cancela Carral, J. M. (2024). Realidad virtual inmersiva y su influencia en parámetros fisiológicos de personas sanas (Immersive virtual reality and its influence on physiological parameters in healthy people). *Retos*, 51, 615-625. <https://doi.org/10.47197/retos.v51.101164>
- Romero Parra, R. M., Barboza Arenas, L. A., Espina-Romero, L. C., Rodríguez Ángeles, C. H., Romero Chacin, J. L., Garcés Rosendo, E. J., Faría Romero, J. A., & Vertiz Osoreo, R. I. (2023). Efectos de un gym neuróbico en el rendimiento académico de los estudiantes universitarios en entornos virtuales (Effects of a neurobic gym on the academic performance of university students in virtual environments). *Retos*, 50, 371-379. <https://doi.org/10.47197/retos.v50.93788>
- Seah, M. L. C., & Koh, K. T. (2021). The efficacy of using mobile applications in changing adolescent girls' physical activity behaviour during weekends. *European Physical Education Review*, 27(1), 113-131.
- Shi, B., Xu, Y., Dai, W., Wang, B., Zhang, S., Li, C., ... & Xiong, H. (2020, October). Tiny-Hourglassnet: An efficient design for 3D human pose estimation. In *2020 IEEE International Conference on Image Processing (ICIP)* (pp. 1491-1495). IEEE.
- Solas-Martínez, J. L., Suárez-Manzano, S., De la Torre-Cruz, M. J., & Ruiz-Ariza, A. (2023). Artificial Intelligence and Augmented Reality in Physical Activity: A Review of Systems and Devices. *Augmented Reality and Artificial Intelligence: The Fusion of Advanced Technologies*, 245-270.
- Soltani, P., & Morice, A. H. (2020). Augmented reality tools for sports education and training. *Computers & Education*, 155, 103923.
- Tariq, M. U. (2024). Revolutionizing Health Data Management with Blockchain Technology: Enhancing Security and Efficiency in a Digital Era. In *Emerging Technologies for Health Literacy and Medical Practice* (pp. 153-175). IGI Global.
- Vashishth, T. K., Sharma, V., Sharma, K. K., Kumar, B., Panwar, R., & Chaudhary, S. (2024). AI-Driven Learning Analytics for Personalized Feedback and Assessment in Higher Education. In *Using Traditional Design Methods to Enhance AI-Driven Decision Making* (pp. 206-230). IGI Global.
- Wang, T., & Wu, D. (2024). Computer-Aided Traditional Art Design Based on Artificial Intelligence and Human-Computer Interaction. *Computer-Aided Design and Applications*, 21, 59-73.
- Ye, M., Shen, J., Zhang, X., Yuen, P. C., & Chang, S. F. (2020). Augmentation invariant and instance spreading feature for softmax embedding. *IEEE transactions on pattern analysis and machine intelligence*, 44(2), 924-939.
- Yi, Z., Chen, Y. H., Yin, Y., Cheng, K., Wang, Y., Nguyen, D., ... & Kim, E. (2022). Brief research report: A comparison of robust tests for homogeneity of variance in factorial ANOVA. *The Journal of Experimental Education*, 90(2), 505-520.

Datos de los/as autores y traductor/a:

Bakhytzhán Omarov	bakhytzhánomarov@gmail.com	Autor/a
Nurlan Omarov	nurlanomarovphd@gmail.com	Autor/a
Quwanishbay Mamutov	quwanishbaymamutov@gmail.com	Autor/a
Zhanibek Kissebayev	shanibek@inbox.ru	Autor/a
Almas Anarbayev	almas.anarbaev@iuth.edu.kz	Autor/a
Adilbay Tastanov	tastanov_70@mail.ru	Autor/a
Zhandos Yessirkepov	zhandos-1978@mail.ru	Autor/a
Batyrkhan Omarov	batyahan@gmail.com	Traductor/a