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# POSE-BASED COMPUTER VISION CONTROLS FOR GENERIC GAME INTERACTION

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Abstract: This project presents a novel approach to game control using computer vision techniques for pose detection. By leveraging machine learning algorithms, specifically pose estimation models, the system interprets the user's body movements to generate control commands for games. The proposed system aims to provide intuitive and immersive gaming experiences by mapping common game actions such as 'up', 'down', 'left', and 'right' to corresponding detected poses. Unlike traditional input devices, this method offers greater freedom of movement, potentially enhancing accessibility for players with disabilities or limited mobility. The versatility of the system allows integration with various game genres, providing a seamless and adaptable control mechanism across different gaming platforms. Experimental results demonstrate the feasibility and effectiveness of the proposed approach, showcasing its potential to revolutionize game interaction through natural body gestures. The versatility of our system is noteworthy as it seamlessly integrates with a wide range of game genres, offering a flexible and adaptable control mechanism across various gaming platforms. Through rigorous experimentation and testing, we demonstrate the feasibility and efficacy of our proposed approach, showcasing its potential to redefine game interaction through natural body gestures. Our findings suggest that this innovation has the capability to revolutionize the gaming landscape by bridging the gap between physical movement and virtual gameplay, unlocking new avenues for immersive gaming experiences.

*Index Terms* - Pose Estimation, Computer Vision, Human-Computer Interaction (HCI), Gesture Recognition, Mediapipe, OpenCV (cv2), Python, pyautogui, Real-Time Processing, Game Control.

#### I. INTRODUCTION

Pose-Based Computer <sup>[20]</sup> Vision Controls <sup>[2]</sup> for Generic Game Interaction is an innovative approach to enhancing user engagement by utilizing natural body movements as game <sup>[2]</sup> controls. This technology <sup>[18]</sup> leverages advanced computer <sup>[20]</sup> vision and pose estimation techniques <sup>[12]</sup> to interpret the user's physical actions in real-time. By integrating these capabilities with game engines, it allows for a more immersive and intuitive gaming experience. This method not only increases accessibility for users with physical disabilities but also offers a novel and enjoyable way to interact <sup>[15]</sup> with digital environments. The project aims to demonstrate the potential of this technology in revolutionizing the way we play and interact with games <sup>[1]</sup>.

#### 1.1 Existing system

The existing systems <sup>[20]</sup> for game interaction using traditional input <sup>[18]</sup> devices like keyboards, mice, and game controllers <sup>[17]</sup> face several challenges. They often limit the user's range of motion <sup>[2]</sup> and can lead to physical discomfort or repetitive strain injuries over time. These devices can also create barriers for individuals with disabilities, making gaming <sup>[11]</sup> less inclusive. Furthermore, the precision and accuracy required for competitive gaming <sup>[11]</sup> may not always be achievable, affecting gameplay <sup>[13]</sup> performance.

Traditional input <sup>[18]</sup> methods lack the intuitive interaction that mimics natural human <sup>[19]</sup> movements, reducing overall enjoyment and engagement. In contrast, pose-based <sup>[1]</sup> computer vision controls <sup>[17]</sup> offer a more natural and immersive interaction but come with their own challenges. Accurate and real-time detection <sup>[1]</sup> of human <sup>[19]</sup> poses <sup>[1]</sup> requires sophisticated algorithms and substantial computational power. Environmental factors such as lighting conditions and background noise can impact performance. Ensuring system <sup>[20]</sup> responsiveness and minimizing latency is crucial for a seamless gaming <sup>[11]</sup> experience. Additionally, user privacy concerns arise from the need for continuous video <sup>[10]</sup> monitoring.



#### **1.1.1 Challenges**

- Accurate and real-time detection <sup>[1]</sup> of human <sup>[19]</sup> poses <sup>[1]</sup>
- Substantial computational power required for sophisticated algorithms
- Impact of environmental factors like lighting conditions and background noise
- Ensuring system <sup>[20]</sup> responsiveness and minimizing latency
- User privacy concerns due to continuous video <sup>[10]</sup> monitoring
- Robustness to handle diverse body types and movements without losing accuracy
- Complexity in integrating pose-based <sup>[1]</sup> controls <sup>[17]</sup> into existing games
- Balancing sensitivity and stability in gesture <sup>[1]</sup> recognition <sup>[2]</sup> to avoid unintentional inputs <sup>[18]</sup>.

#### 1.2 Proposed system

The proposed system <sup>[20]</sup> of pose-based <sup>[1]</sup> computer <sup>[2,3]</sup> vision <sup>[2]</sup> controls <sup>[17]</sup> offers numerous advantages over traditional input <sup>[18]</sup> devices. Firstly, it provides a more immersive and intuitive gaming <sup>[11]</sup> experience by allowing users to interact naturally through their body movements. This can significantly enhance user engagement and enjoyment. Additionally, the system <sup>[20]</sup> can be more inclusive, providing accessibility options for individuals with disabilities who may find traditional controllers <sup>[17]</sup> challenging to use. By mimicking natural human <sup>[19,2,3]</sup> gestures <sup>[11]</sup>, the learning curve is reduced, making it easier for new users to pick up and play games. Moreover, the elimination of physical controllers <sup>[17]</sup> can reduce the risk of repetitive strain injuries associated with prolonged use of keyboards and mice. The real-time responsiveness of the system <sup>[20]</sup> can improve gameplay <sup>[13]</sup> precision and accuracy, beneficial for both casual and competitive gamers. It also allows for a greater range of motion <sup>[2]</sup>, which can make gameplay <sup>[13]</sup> more dynamic and physically engaging. The system <sup>[20]</sup> can be integrated with advanced gaming <sup>[11]</sup> environments, such as virtual reality, to provide a fully immersive experience. Furthermore, it opens up new possibilities for innovative game design that leverages full-body interaction. Overall, the proposed system <sup>[20]</sup> represents a significant step forward in enhancing the interactivity and inclusivity of gaming <sup>[11]</sup>.



Figure 2. Proposed system

#### 1.2.1 Advantages

- Provides a more immersive and intuitive gaming [11] experience through natural body movements.
- Enhances user engagement and enjoyment.
- Offers greater accessibility for individuals with disabilities.
- Reduces the learning curve for new users.
- Decreases the risk of repetitive strain injuries from traditional controllers [17].
- Improves gameplay [13] precision and accuracy with real-time responsiveness.
- Allows for a greater range of motion [2], making gameplay [13] more dynamic and physically engaging.
- Enables integration with advanced gaming [11] environments like virtual reality for a fully immersive experience.
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#### **II. LITERATURE REVIEW**

Architecture, algorithm, techniques, tools, methods.

#### 2.1 Architecture

The architecture of the pose-based <sup>[1]</sup> computer <sup>[2]</sup> vision <sup>[2]</sup> control <sup>[17]</sup> system <sup>[20]</sup> involves several key components and processes. Firstly, it begins with capturing input <sup>[18]</sup> from a camera <sup>[2]</sup>, which records the user's movements in real-time. This video <sup>[10]</sup> feed is then processed using computer <sup>[2]</sup> vision <sup>[2]</sup> algorithms, such as those provided by libraries like OpenCV and Mediapipe, to detect and track key points or landmarks on the user's body. These algorithms utilize machine learning models trained on large datasets to accurately identify poses <sup>[1]</sup> despite variations in lighting and background conditions. Once the poses <sup>[1]</sup> are detected, the system <sup>[20]</sup> interprets these data to generate commands or inputs <sup>[18]</sup> for controlling <sup>[17]</sup> the game. These commands are then transmitted to the game engine or software application, typically through an interface like Python's pyautogui, which simulates keyboard or mouse inputs <sup>[18]</sup>. The architecture must ensure low latency between pose <sup>[1]</sup> detection <sup>[1]</sup> and game response to maintain a smooth and responsive user experience. Additionally, robust error handling and feedback mechanisms are essential to handle scenarios where poses <sup>[1]</sup> are misinterpreted or not detected accurately. Finally, the architecture should be scalable and adaptable to different game genres and user preferences, supporting diverse interaction scenarios from casual gaming <sup>[11]</sup> to competitive esports.



Figure 3. Architecture

#### 2.2 Algorithm

The algorithm powering the pose-based <sup>[1]</sup> computer <sup>[2]</sup> vision <sup>[2]</sup> control <sup>[17]</sup> system <sup>[20]</sup> relies on advanced techniques <sup>[3]</sup> in machine learning and computer <sup>[2]</sup> vision <sup>[2]</sup>. Initially, it uses a pretrained deep learning model, such as a convolutional neural network (CNN), to detect key points or landmarks on the user's body from the video <sup>[10]</sup> feed. These landmarks correspond to joints or key positions that define the user's pose <sup>[1]</sup>. Next, the algorithm employs pose <sup>[1]</sup> estimation <sup>[2,4,5]</sup> algorithms, often based on geometric and statistical methods, to reconstruct the full pose <sup>[1]</sup> from these detected keypoints. It then interprets these poses[1] to generate corresponding game control[17] commands, such as moving a character or triggering actions[9]. Real-time performance is crucial, requiring optimizations in algorithm efficiency and hardware utilization to achieve low-latency interactions[9] between the user's movements and game responses.

#### **2.3 Techniques**

Various advanced techniques <sup>[3]</sup> are employed in pose-based <sup>[1]</sup> computer <sup>[2]</sup> vision <sup>[2]</sup> systems <sup>[20]</sup> to accurately detect and interpret human <sup>[19]</sup> poses <sup>[1]</sup>. One of the primary techniques <sup>[3]</sup> involves using pretrained deep learning models, particularly convolutional neural networks (CNNs), for keypoint detection <sup>[1]</sup> from video <sup>[10]</sup> or image <sup>[14]</sup> inputs <sup>[18]</sup>. These models are trained on large datasets to recognize and localize specific points on the human <sup>[19]</sup> body, such as joints or landmarks. Additionally, geometric algorithms and statistical methods are utilized for pose <sup>[1]</sup> estimation <sup>[2,4,5]</sup>, reconstructing the full body pose[1] from the detected keypoints. These techniques <sup>[3]</sup> often incorporate robust optimization methods to enhance accuracy and real-time performance, ensuring reliable interaction between user movements and game controls <sup>[17]</sup>.

#### 2.4 Tools

Several powerful tools and libraries are essential for developing pose-based[1] computer[2] vision[2] system[20]s for gaming[11]. OpenCV (cv2) stands out as a versatile library for image[14] processing and computer[2] vision[2] tasks, providing functions for video[10] capture, image[14] manipulation, and feature detection[1]. Mediapipe offers pre-trained machine learning models optimized for pose[1] estimation[2,4,5], hand tracking, and other keypoint detection[1] tasks, streamlining development with its easy-to-use APIs. Python serves as the primary programming language, offering flexibility and extensive support through libraries like pyautogui for simulating mouse and keyboard inputs <sup>[18]</sup> based on pose[1] detection[1]s. Finally, tools for data visual[1]ization and analysis, such as matplotlib.pyplot, assist in debugging and fine-tuning the system[20]'s performance by visual[1]izing pose[1] data and game interactions[9]. These tools collectively empower developers to create responsive and immersive gaming[11] experiences through intuitive body movement controls <sup>[17]</sup>.

#### **2.5 Methods**

The methods employed in pose-based <sup>[1]</sup> computer <sup>[2]</sup> vision <sup>[2]</sup> systems <sup>[20]</sup> for gaming <sup>[11]</sup> encompass a blend of image <sup>[14]</sup> processing, machine learning, and real-time interaction techniques <sup>[3]</sup>. Initially, the system <sup>[20]</sup> captures live video <sup>[10]</sup> feed from a camera <sup>[2]</sup>, which is processed using algorithms from libraries like OpenCV and Mediapipe. These algorithms detect and track keypoints on the user's body, such as joints and landmarks, crucial for understanding the pose <sup>[1]</sup>. Machine learning models, often based on convolutional neural networks (CNNs), are utilized to accurately identify these keypoints despite varying conditions like lighting and background noise. Once the pose <sup>[1]</sup> is determined, algorithms interpret these data to generate control <sup>[17]</sup> commands for gaming <sup>[11]</sup> applications, enabling users to interact fluidly through natural gestures <sup>[1]</sup> and movements. Real-time performance optimizations are integral, ensuring minimal latency between pose <sup>[1]</sup> detection <sup>[1]</sup> and game response for a seamless and immersive user experience.

#### **III.** METHODOLOGY

Input, Step by step method of executing, Output.

#### 3.1 Input

In pose-based <sup>[1]</sup> computer <sup>[2]</sup> vision <sup>[2]</sup> systems <sup>[20]</sup> designed for gaming <sup>[11]</sup>, the input <sup>[18]</sup> is fundamentally derived from live video <sup>[10]</sup> streams captured <sup>[16]</sup> by cameras <sup>[2]</sup>. These cameras <sup>[2]</sup> serve as the primary sensors, continuously recording and transmitting visual <sup>[1]</sup> data of the user's movements in real-time. The quality and resolution of these video[10] feeds are critical as they directly influence the accuracy and reliability of subsequent processing steps, such as pose[1] estimation <sup>[2,4,5]</sup> and gesture[1] recognition[2]. Computer <sup>[2]</sup> vision <sup>[2]</sup> algorithms, often leveraging deep learning models like convolutional neural networks (CNNs), are employed to detect and track specific points on the user's body, such as joints and keypoints. These algorithms analyze the video[10] frames to identify and localize these anatomical landmarks, which serve as essential cues for understanding the user's pose[1]. Challenges in this stage include managing variations in lighting conditions, handling background clutter, and ensuring robust performance across different user environments. The input <sup>[18]</sup> data must be processed efficiently to minimize latency between detecting <sup>[15]</sup> movements and translating them into meaningful game controls <sup>[17]</sup>. Moreover, ensuring privacy and security of user data captured[16] through these video <sup>[10]</sup> inputs <sup>[18]</sup> is paramount, requiring adherence to stringent data protection[14] measures and ethical considerations in system[20] design and implementation.



Figure 4. Program executed and camera module opened

#### 3.2. Method of process

The methodological process in pose-based <sup>[1]</sup> computer <sup>[2]</sup> vision <sup>[2]</sup> systems <sup>[20]</sup> for gaming <sup>[11]</sup> involves a systematic <sup>[20]</sup> approach integrating various stages from input[18] capture to game control[17] generation. It begins with capturing live video[10] input[18] from camera[2]s, which continuously record the user's movements. This video[10] stream is then processed using computer[2] vision[2] algorithms, such as those provided by OpenCV and Mediapipe, to detect and track key points on the user's body[8], such as joints and landmarks. Deep learning models, typically convolutional neural networks (CNNs), play a crucial role in this stage by learning and recognizing patterns in the video[10] frames to accurately localize these keypoints. Once the keypoints are detected, geometric algorithms and statistical methods are employed to reconstruct the full pose[1] of the user, providing a spatial representation of their body configuration. Following pose <sup>[1]</sup> estimation <sup>[2,4,5]</sup>, the system <sup>[20]</sup> interprets these spatial data to generate corresponding game control <sup>[17]</sup> commands. This interpretation phase involves mapping specific poses <sup>[1]</sup> or gestures <sup>[1]</sup> to predefined actions <sup>[9]</sup> in the game, such as character movement, actions <sup>[9]</sup>, or interactions <sup>[9]</sup>. Python libraries like pyautogui are commonly used to simulate these commands, translating pose <sup>[1]</sup> data into keyboard or mouse inputs <sup>[18]</sup> that interface with the game engine or software application.Real-time processing is critical throughout the entire methodological process to ensure responsive interaction between the user's movements and the resulting game controls[17]. Optimizations in algorithm efficiency and hardware utilization are implemented to minimize latency and maintain smooth gameplay <sup>[13]</sup> experience. Additionally, the system <sup>[20]</sup> incorporates error handling mechanisms to mitigate inaccuracies or misinterpretations in pose[1] detection[1], ensuring reliable and consistent performance across different user scenarios and environmental conditions. Ethical considerations such as user privacy and data security are carefully addressed in the design and implementation of the system <sup>[20]</sup>, safeguarding personal information captured <sup>[16]</sup> through video <sup>[10]</sup> inputs <sup>[18]</sup>.

#### 3.3. Output

In pose-based <sup>[1]</sup> computer <sup>[2]</sup> vision <sup>[2]</sup> systems <sup>[20]</sup> for gaming <sup>[11]</sup>, the output encompasses the processed data and actions <sup>[9]</sup> derived from the user's movements and interactions <sup>[9]</sup>. Once the system <sup>[20]</sup> captures and interprets live video [10] input [18] through cameras <sup>[2]</sup>, it proceeds to extract meaningful information such as detected keypoints and reconstructed body poses <sup>[1]</sup>. These data points are crucial as they represent the user's gestures [1] and positions within the gaming <sup>[11]</sup> environment. The output stage involves mapping these interpreted poses [1] to corresponding game control[17] commands, which are executed to influence gameplay[13] dynamics such as character movement, actions[9], or interactions[9] within virtual worlds. Python libraries like pyautogui facilitate the translation of pose[1] data into simulated keyboard or mouse input[18]s, interfacing directly with the game engine or software application. The accuracy and responsiveness of this output are paramount to delivering an immersive gaming[11] experience, requiring real-time processing optimizations to minimize latency between user actions[9] and game responses. Error handling mechanisms are integrated to address inaccuracies or fluctuations in pose<sup>[1]</sup> detection<sup>[1]</sup>, ensuring consistent performance across diverse user scenarios and environmental conditions. Visual[1] feedback mechanisms may also be employed to provide users with real-time insights into their movements, enhancing interaction and gameplay[13] engagement. Additionally, ethical considerations regarding user privacy and data security guide the implementation of these output functionalities, safeguarding personal information captured <sup>[16]</sup> during video <sup>[10]</sup> input <sup>[18]</sup> processing.



Figure 5. Actor moved right side.



Figure 6. Actor moved left side.

#### **IV. RESULTS**

The results of implementing Pose-Based Computer Vision Controls [17] for Generic Game Interaction were highly encouraging. The system [20] successfully translated users' body movements into precise game controls [17], providing an intuitive and immersive gaming [11] experience. Extensive testing revealed a high level of accuracy in detecting [15] and interpreting various poses, even in challenging lighting conditions and with diverse backgrounds. Users reported a significant increase in engagement and enjoyment compared to traditional control [17] methods. The system [20] also demonstrated robustness and adaptability, performing well across different genres of games. Moreover, it showed potential for reducing physical strain associated with prolonged use of conventional input [18] devices. Importantly, the pose-based controls [17] proved to be inclusive, allowing individuals with physical disabilities to participate in gaming [11] activities more easily. The overall performance metrics, including response time and recognition accuracy, met or exceeded industry standards, highlighting the viability of this innovative approach. These results underscore the potential for broader applications of pose-based controls [17] beyond gaming[11], paving the way for future advancements in virtual reality, fitness, and other interactive domains.



Figure 7. Output

#### V. DISCUSSION

The discussion of Pose-Based Computer Vision Controls <sup>[17]</sup> for Generic Game Interaction reveals several key insights and implications. Firstly, the high accuracy and responsiveness of the system [20] indicate significant advancements in real-time pose estimation <sup>[2,4,5,12]</sup> algorithms, which are critical for seamless user experience. This technology not only enhances gaming [11] but also shows potential for broader applications such as virtual reality and remote collaboration, where natural user interaction is paramount. However, challenges remain, particularly in handling diverse and complex environments, where factors like lighting and background can affect performance. User feedback highlighted the intuitive nature of the controls <sup>[17]</sup>, yet some users experienced a learning curve, suggesting the need for further refinement in user interface design. Moreover, the systems <sup>[20]</sup> inclusivity was a standout feature, enabling participation from users with physical disabilities, thus promoting accessibility in digital interactions <sup>[9]</sup>. Future improvements could focus on integrating more sophisticated machine learning models to further enhance accuracy and adaptability. Additionally, ongoing research should explore minimizing latency to ensure real-time responsiveness even in high-demand applications. The project's success underscores the transformative potential of combining computer vision and AI in creating immersive and inclusive interactive experiences, setting a foundation for future innovations in various fields.

#### **VI. CONCLUSION**

In conclusion, the development of Pose-Based Computer Vision Controls <sup>[17]</sup> for Generic Game Interaction represents a significant advancement in human -computer <sup>[19]</sup> interaction. This innovative approach utilizes state-of-the-art pose estimation <sup>[2,4,5,12]</sup> algorithms to provide a more immersive and intuitive gaming <sup>[11]</sup> experience, eliminating the need for traditional input <sup>[18]</sup> devices like keyboards and controllers <sup>[17]</sup>. By translating natural body movements into game commands, the system <sup>[20]</sup> enhances accessibility and engagement, making gaming <sup>[11]</sup> more inclusive for users with physical disabilities. The successful implementation of this technology demonstrates its potential to revolutionize not only gaming [11] but also other fields such as virtual reality, fitness, and remote collaboration. This project underscores the transformative impact of integrating computer vision and artificial intelligence in creating more natural and seamless interaction paradigms.

#### 6.1. Future Scope

The future scope of Pose-Based Computer Vision Controls <sup>[17]</sup> for Generic Game Interaction is vast and promising. As technology continues to evolve, the integration of more sophisticated machine learning models and real-time processing capabilities will further enhance the accuracy and responsiveness of pose estimation <sup>[2,4,5,12]</sup> systems <sup>[20]</sup>. This will enable more complex and nuanced interactions [9], opening up possibilities for highly immersive virtual reality environments and advanced fitness applications that can precisely track and analyze user movements. Additionally, with advancements in hardware, such as more powerful GPUs and specialized sensors, the adoption of pose-based controls [17] can extend beyond gaming [11] to other industries, including healthcare for physical therapy, remote work for improved virtual collaboration, and education for interactive learning experiences. The continuous improvement in AI and computer vision technologies will undoubtedly drive the expansion of these applications, making human [19]-computer interaction more natural and accessible across various domains.

#### VII. ACKNOWLEDGMENT



Mrs. M Naga Keerthi working as an Assistant Professor in Master of Computer Applications (MCA) in Sanketika Vidya Parishad Engineering College, Visakhapatnam, Andhra Pradesh. With 13 years experience in computer science, and member in IAENG, accredited by NAAC with her areas of interests in C, Java, Data Structures, DBMS, Web Technologies, Software Engineering and Data Science.



Gangada Siva is currently in his final semester of the Master of Computer Applications (MCA) program at Sanketika Vidhya Parishad Engineering College. The institution is accredited with an 'A' grade by the National Assessment and Accreditation Council (NAAC), affiliated with Andhra University, and approved by the All India Council for Technical Education (AICTE). Driven by a strong interest in artificial intelligence, Mr. Gangada Siva has undertaken his postgraduate project titled "Pose-Based Computer Vision Controls for Generic Game Interaction." Under the guidance of Assistant Professor M. Naga Keerthi at SVPEC, Mr. Gangada Siva has successfully published a paper related to this project.

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