

A Mixed Reality Framework for Acupoint Therapy Teaching: Localization, Practice and Evaluation

Jiacheng Liu^{*†}
Southwest University
Peking University

Bohan Chen^{*}
Southwest University

Weichao Song^{*}
Southwest University

Fangfei Ye[‡]
Beijing University of Chinese Medicine

Haibin Ling[§]
Westlake University

Bingyao Huang[¶]
Southwest University

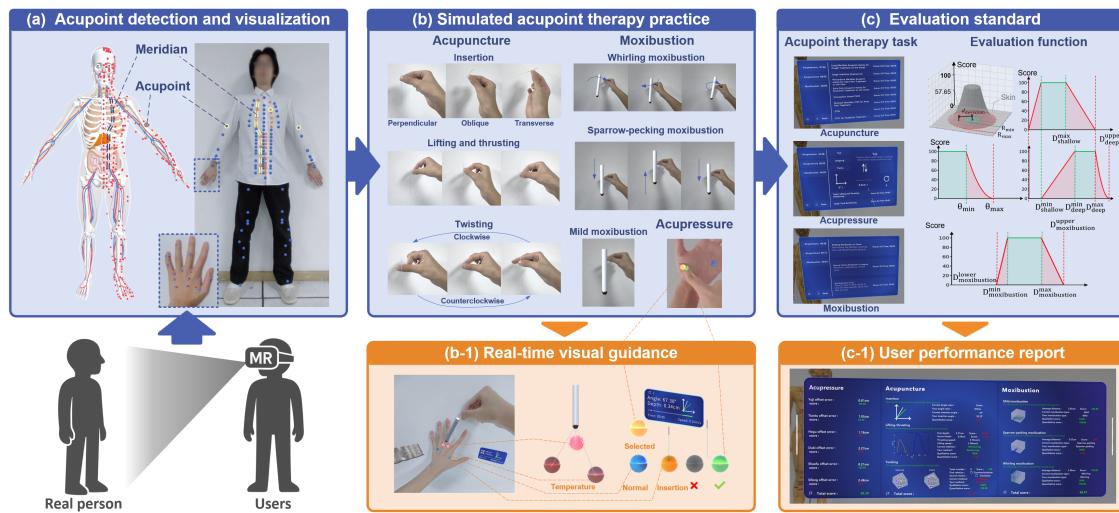


Figure 1: The proposed mixed reality framework consists of three main modules: (a) Acupoint detection and real-time visualization on real person in MR, (b) simulated acupoint therapy practice and real-time visual guidance in the HMD, and (c) a Traditional Chinese Medicine (TCM) theory-based evaluation standard with user performance report after simulated practice to improve user learning accuracy and efficiency.

ABSTRACT

Acupoint therapy is an essential therapeutic method in Traditional Chinese Medicine (TCM), and it requires precise acupoint localization and manipulation skills. Existing Mixed Reality (MR) systems often fall short in real-time acupoint detection and visualization, comprehensive simulated practice, and evaluation standards. To overcome these limitations, we introduce a framework for acupoint therapy teaching. It provides real-time full-body acupoint detection and visualization, guided simulated practice for complex acupuncture and moxibustion techniques, and a novel TCM theory-based standard for user performance evaluation. A user study confirms that our system improves users' 3D acupoint localization and procedural proficiency compared to traditional textbook-based methods.

Index Terms: MR-based teaching system, Acupoint therapy.

*e-mail: {yc2020, turou, swc4869}@email.swu.edu.cn

†e-mail: liujiacheng@stu.pku.edu.cn

‡e-mail: 20210121138@bucm.edu.cn

§e-mail: linghaibin@westlake.edu.cn

¶e-mail: bhuang@swu.edu.cn. Corresponding author.

1 INTRODUCTION

Acupoint therapy, including acupressure, acupuncture and moxibustion, is a fundamental part of TCM. Conventional teaching methods, such as textbooks and apprenticeships, often lack spatial immersion, learning efficiency, and safe interaction. Existing MR solutions are also limited, typically focusing only on facial acupoints or basic therapy techniques, while the real-time full-body acupoint detection, advanced simulation, and evaluation standard remain underdeveloped in MR environments.

To address these limitations, we propose a framework for acupoint therapy teaching, as shown in Figure 1. Users can learn acupoint locations and practice simulated therapeutic procedures on the detected acupoints, receiving immediate feedback and detailed performance reports. This makes our framework a more comprehensive and effective learning tool compared to previous approaches, as shown in Table 1.

2 METHODS

We combine TCM localization methods [1] (B-cun and F-cun) with human pose estimation algorithm (Mediapipe [3] and PICO hand-tracking API) to realize real-time **acupoint detection** and visualization on the real patient, as shown in Figure 2. Detected acupoints on hands, limbs, and torso are rendered as interactive virtual spheres, accurately overlaid onto the patient's body in the MR environment.

Our system supports **simulated practice** and real-time visual guidance for a full range of acupoint therapy tech-

Table 1: The comparison of different representative methods.

| Representative Methods | MR Supported | Hand Acupoint Visualization | Facial Acupoint Visualization | Limb Acupoint Visualization | Torso Acupoint Visualization | Real-time Visualization | Has Acupoint Model | Acupuncture Visualization | | | Evaluation |
|--------------------------|--------------|-----------------------------|-------------------------------|-----------------------------|------------------------------|-------------------------|--------------------|---------------------------|-----------------------|-----------|------------|
| | | | | | | | | Insertion | Lifting and thrusting | Twisting | |
| Acu. Training System [7] | Yes | N/A | N/A | N/A | N/A | N/A | Yes | N/A | N/A | N/A | Supported |
| E-faceatlasAR [9] | Yes | N/A | Supported | N/A | N/A | Supported | N/A | N/A | N/A | N/A | N/A |
| TCM Massage Robot [2] | No | N/A | N/A | Supported | Supported | Supported | N/A | N/A | N/A | N/A | N/A |
| RGB-D CNN [4] | No | Supported | N/A | N/A | N/A | Slow | N/A | N/A | N/A | N/A | N/A |
| Deep CNN [6] | No | N/A | N/A | Supported | N/A | Slow | N/A | N/A | N/A | N/A | N/A |
| Improved HRNet [8] | No | Supported | N/A | N/A | N/A | N/A | N/A | N/A | N/A | N/A | N/A |
| Ours | Yes | Supported | N/A | Supported | Supported | Supported | Yes | Supported | Supported | Supported | Supported |

niques—including acupressure, advanced acupuncture manipulations (insertion, lifting-thrusting and twisting), and moxibustion (mild, sparrow-pecking, whirling). Users interact using fingertip gestures, with the system providing immediate visual feedback to enhance skill acquisition and error correction.

We propose a comprehensive **evaluation** module. Based on TCM theory and clinical practice, we design quantitative and qualitative scoring functions to assess user accuracy and proficiency. After simulated practice, the system generates detailed user performance report to guide learning improvement. This integrated pipeline supports efficient, interactive, and standardized acupoint therapy education.

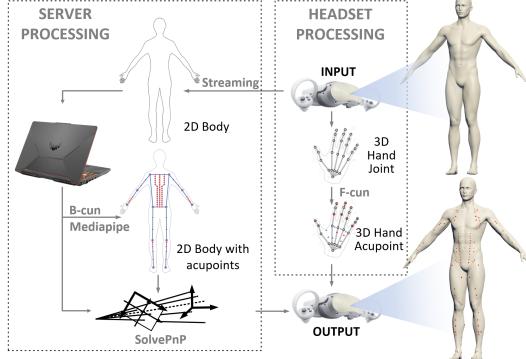


Figure 2: MR-based human acupoint detection using human pose estimation algorithm (Mediapipe and hand-tracking API) and traditional TCM acupoint localization methods (B-cun and F-cun).

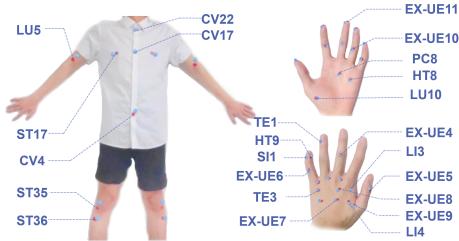


Figure 3: The locations of acupoints on the hands, limbs and torso detected by our system and those marked by the professional.

3 RESULTS

We first evaluate the system’s performance and then conduct a user study involving 42 participants divided into groups for controlled experiments.

Acupoint detection module achieves high detection accuracy on hands, limbs and torso, as shown in Figure 2 and Table 2. In static conditions, the average Absolute Error (AE) outperforms the General Criterion calculated from the 95% confidence interval of expert-marked acupoints in the previous study [5]. In terms of real-time performance, the system consistently maintains a stable frame rate of 90 FPS. Furthermore, professional physicians evaluate that

Table 2: Average offset error and evaluation criteria.

| State | Regions | RE | AE | General Criterion | | Precise Criterion |
|--------|---------|-------|----------|-------------------|---------|-------------------|
| | | | | Hands | Limbs | |
| Static | Hands | 0.065 | 0.322 cm | 1.13 cm | 0.56 cm | |
| | Limbs | 0.048 | 1.205 cm | 1.90 cm | 0.94 cm | |
| | Torso | 0.057 | 1.426 cm | 2.14 cm | 1.16 cm | |

the system’s performance is sufficient to meet educational requirements.

In controlled experiments, participants using our framework learned more accurate acupoint locations compared to those using 3D models (AG) or textbooks (TG), as shown in Figure 4 (a). Participants’ operation skills improved during simulated practice, as illustrated in Figure 4 (b). Most participants report positive feedback on system usability, immersion, and educational value.

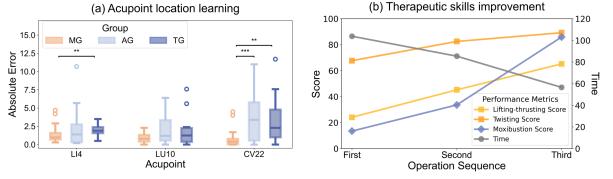


Figure 4: Learning outcomes of acupoint therapy.

4 CONCLUSION

We propose a mixed reality framework with real-time acupoint detection and evaluation standards for acupoint location and therapy learning, increasing teaching precision and efficiency.

REFERENCES

- [1] W. H. O. R. O. for the Western Pacific. *WHO standard acupuncture point locations in the Western Pacific Region*. 2008. 1
- [2] W. Hu, Q. Sheng, and X. Sheng. A novel realtime vision-based acupoint estimation for tcm massage robot. In *Proc. M2VIP*. IEEE, Shanghai, China, 2021. 2
- [3] C. Lugaresi, J. Tang, H. Nash, C. McClanahan, E. Uboweja, M. Hays, F. Zhang, C.-L. Chang, M. G. Yong, J. Lee, et al. Mediapipe: A framework for building perception pipelines. *arXiv preprint arXiv:1906.08172*, 2019. 1
- [4] D. Masood and J. Qi. 3d localization of hand acupoints using hand geometry and landmark points based on rgb-d cnn fusion. *Annals of Biomedical Engineering*, 50(9), jun 2022. 2
- [5] A. Molsberger, J. Manickavasagan, H. Abholz, W. Maixner, and H. Endres. Acupuncture points are large fields: the fuzziness of acupuncture point localization by doctors in practice. *European Journal of Pain*, 16(9):1264–1270, 2012. 2
- [6] L. Sun, S. Sun, Y. Fu, and X. Zhao. Acupoint detection based on deep convolutional neural network. In *Proc. CCC*. IEEE, Shenyang, China, 2020. 2
- [7] Q. Sun, J. Huang, H. Zhang, P. Craig, L. Yu, and E. G. Lim. Design and development of a mixed reality acupuncture training system. In *Proc. IEEE VR*. IEEE, Shanghai, China, 2023. 2
- [8] S. Sun, H. Xu, L. Sun, Y. Fu, Y. Zhang, and X. Zhao. Hand acupoint detection from images based on improved hrnet. In *Proc. IJCNN*. IEEE, Padua, Italy, 2022. 2
- [9] M. Zhang, J. P. Schulze, and D. Zhang. E-faceatlasar: extend atlas of facial acupuncture points with auricular maps in augmented reality for self-acupressure. *Virtual Real.*, 26(4), dec 2022. 2