

Advances in Artificial Intelligence for Physical and Mental Wellness: A Multidisciplinary Approach

Dr. Tangarani

Physical Culture Instructor, St Agnes College (Autonomous), Mangaluru, India

Abstract-- Artificial Intelligence (AI) is becoming a cornerstone in health and wellness applications, delivering interventions across physical fitness, preventive care, diagnostics, psychological assessment, and emotional well-being. This review synthesizes multidisciplinary research and recent secondary data to examine how AI supports physical and mental wellness. Key domains include wearable monitoring, predictive analytics, virtual assistants, and multimodals affect detection. Ethical, privacy, equity, and bias challenges are critically discussed. The paper concludes with future directions and practical recommendations for integrating AI into holistic health frameworks.

Keywords-- Artificial Intelligence, Physical Wellness, Mental Health, Wearables, Ethics, Multidisciplinary.

I. INTRODUCTION

Chronic disease burden and rising mental health concerns represent major global challenges. For example, the global AI in mental health market is projected to grow from USD 1.13 billion in 2023 to USD 5.08 billion by 2030 (CAGR ~24.1 %). Meanwhile, the World Health Organization reports over 150 million people in the European Region live with mental health conditions and pandemic-related pressures have intensified disparities in access to care.

Wearable health device adoption is also increasing: in the U.S., usage expanded from ~28–30 % in 2019 to 36.36 % in 2022. These devices, when integrated with AI, can transform passive data streams into proactive wellness insights.

This paper presents a comprehensive, multidisciplinary review of AI's contributions to physical and mental wellness, leveraging secondary data and literature. It aims to identify successes, limitations, ethical challenges, and future trajectories.

Contributions

- A structured mapping of AI applications in physical and mental wellness with empirical backing.
- Identification of gaps in current research, especially in inclusive design, long-term efficacy, and multi-domain integration.

- Recommendations for future research, policy, and practice to strengthen AI's role in holistic health care.

Outline: Section 2 reviews AI in physical wellness; Section 3 covers AI in mental health; Section 4 discusses multidisciplinary integration; Section 5 addresses ethics, equity, and bias; Section 6 outlines challenges; Section 7 suggests future directions; Section 8 concludes.

II. OBJECTIVES

1. To map key AI applications in physical wellness and mental wellness using recent secondary data.
2. To explore how disciplines such as psychology, medicine, public health, computer science, and behavioral science collaborate in these developments.
3. To examine ethical, societal, and governance issues, especially with respect to data privacy, bias, and equity.
4. To propose recommendations and future research directions grounded in empirical findings.

III. AI APPLICATIONS IN PHYSICAL WELLNESS

3.1 Fitness, Activity Monitoring, and Lifestyle Feedback

Wearable sensors (accelerometers, PPG, gyroscopes) feed data into AI models that estimate steps, heart rate variability, sleep quality, and more. AI layers allow anomaly detection, adaptive goal setting, and trend predictions. In a scoping review of AI-empowered mobile health, Guo et al. reported that AI in Health is already used for activity recognition, fall detection, chronic disease monitoring, and user engagement prediction

3.2 Preventive Health and Risk Prediction

Machine learning models have been developed to predict risks of obesity, hypertension, cardiovascular disease, and diabetes from lifestyle, biometric, and genomic data. For example, systematic reviews of AI in chronic disease management highlight how AI can support diagnosis, prognosis, and decision support across multiple conditions, though with cautions on bias and regulatory challenges.

One protocol study (AI-HEALS) is exploring how intelligent health education via AI can improve self-management in Type 2 diabetes, combining AI question-answering with behavioural nudges and mobile interface.

3.3 Physical Rehabilitation and Remote Monitoring

AI-driven robotics, virtual training systems, and motion capture systems can guide rehabilitation post injury or surgery.

While robust clinical RCTs are sparse, remote patient monitoring (RPM) is becoming a promising domain: Shaik et al. report that AI-enabled RPM architectures offer early deterioration detection, personalized parameter tracking, and behaviour modelling using reinforcement learning and federated learning techniques.

Table 1
gives a summary of representative AI applications in physical wellness.

Application Domain	AI Methods / Techniques	Use Case / Example	Limitations / Gaps
Activity & Sleep Monitoring	Classification, anomaly detection, trend prediction	Wearable's generate adaptive fitness suggestions	Drop-off in usage; sensor noise; limited clinical validation
Chronic Disease Prediction	Supervised learning, ensemble models	Predict risk of diabetes, hypertension	Bias in training data; interpretability issues
Remote Rehabilitation	Computer vision, reinforcement learning	Virtual physiotherapy coaching	Sparse RCT evidence; hardware cost and accessibility

IV. AI IN MENTAL WELLNESS

4.1 Emotion Recognition and Mental State Assessment

AI models now analyze speech, text, facial expressions, and physiological signals to infer emotional state, stress, depression, or anxiety. A recent survey on mental health datasets emphasized that existing datasets are often fragmented, lack longitudinal data, and have limited cultural representation - a key barrier to robust model generalization.

The WHO notes that AI applications in mental health focus predominantly on depression, schizophrenia, and psychotic disorders, leaving out many conditions, with concerns over reproducibility, bias, and model transparency.

4.2 Virtual Mental Health Assistants

Chatbots like Woebot, Wysa, and Replika deliver brief therapeutic interventions (e.g., CBT, mood tracking). Suarjana et al. argue that we should evaluate chatbots not by comparing them to expert therapists, but by their pragmatic value especially for populations with limited access to clinicians.

However, a PubMed review highlights that many generative AI “wellness apps” operate in regulatory gray zones, sometimes respond unpredictably, and may increase harm risk in crisis situations.

4.3 Personalized Psychological Interventions & Adaptive Support

AI systems can curate content (mindfulness, motivational prompts) and deliver tailored interventions based on real-time signals (e.g. phone usage patterns). Market analysts suggest that AI mental health tools saw a 120 % increase in downloads in 2023, and that the global AI mental health market is expected to reach USD 5.08 billion by 2030.

Meta-analyses of digital mental health interventions find moderate effect sizes (20–30 % symptom reduction) in anxiety and depression outcomes when compared to control groups.

V. MULTIDISCIPLINARY INTEGRATION

5.1 Cross-Disciplinary Synergies

Successful AI wellness systems require integration across domains:



International Journal of Recent Development in Engineering and Technology
Website: www.ijrdet.com (ISSN 2347-6435(Online) Volume 15, Issue 01, January 2026)

- Computer science / engineering handles data modelling, algorithms, and system design.
- Clinical disciplines (medicine, psychology) provide theory, domain knowledge, validation, and ethical oversight.
- Behavioral & social science addresses adoption, motivation, user interfaces, cultural context, and adherence.
- Public health & policy manage scaling, regulation, and equitable deployment.

5.2 Case Study: AI-Powered Holistic Wellness App

An illustrative model is a wellness app combining physical activity tracking, mood journaling, nutrition logging, and sleep analysis. AI algorithms correlate cross-domain features to suggest holistic lifestyle interventions. While few published studies detail full implementations, real-world health tech firms are beginning to deploy such integrated solutions.

5.3 Academic and Industry Partnerships

Partnerships between tech firms, universities, and healthcare organizations enable deployment, user testing, and regulatory compliance. These collaborations foster real-world data pipelines and validation studies.

VI. ETHICAL, PRIVACY, EQUITY, AND BIAS ISSUES

Data Privacy and Governance: Health-related data are highly sensitive. Many AI wellness apps operate in regulatory gray areas; for example, generative wellness chatbots often avoid classification as medical devices, limiting oversight. Regulations such as GDPR, HIPAA, and evolving national frameworks must be navigated carefully.

Algorithmic Bias and Representativeness: Models trained on skewed or non-representative datasets can yield biased predictions - e.g., underperforming for minority groups. The WHO cautions that many AI mental health studies lack sufficient attention to bias and transparency.

Accessibility and Equity: Rural or under-resourced populations may lack access to wearables, smartphones, or stable internet. AI wellness platforms must be designed to minimize digital divide effects.

Human-AI Interaction and Accountability: While AI can complement care, it should not replace human empathy or professional judgment. Clear escalation paths, human oversight, and accountability frameworks are essential.

VII. CHALLENGES AND LIMITATIONS

Lack of Longitudinal and RCT Evidence: Many AI wellness tools have short-term pilot data; few long-term trials exist. Systematic reviews of AI in clinical settings often note a lack of randomized controlled trials, especially in low- and middle-income settings

User Retention and Engagement: Wearable and app fatigue reduces sustained use.

Data Quality and Sensor Inaccuracy: Noise, missing data, calibration drift affect reliability.

Interpretability and Explainability: Black-box models face trust deficits among clinicians and users.

Interoperability and Integration: Integrating AI tools with existing electronic health records (EHRs) and workflows is complex.

Regulatory and Liability Ambiguity: Uncertain classification of wellness vs medical device poses liability issues.

VIII. FUTURE DIRECTIONS & RECOMMENDATIONS

1. *Robust Longitudinal Studies & RCTs:* Conduct longer-term, larger-scale randomized controlled trials across populations and settings.
2. *Dataset Expansion & Standardization:* Collaboratively build high-quality, diverse, longitudinal, multimodal datasets (text, speech, sensor) for model training, with emphasis on underrepresented groups.
3. *Explainable & Fair AI Models:* Prioritize model interpretability and bias mitigation methods.
4. *Multimodal Fusion & Continual Learning:* Integrate physiological, behavioral, and environmental data streams with adaptive learning frameworks.
5. *Hybrid Human-AI Systems:* Design systems where AI augments, not replaces, human professional decision-making.
6. *Policy and Ethical Frameworks:* Collaborate with regulators to clarify classification, liability, and standards for wellness AI.
7. *Equity-First Design:* Intentionally design for low-resource settings, include offline modes, low-cost hardware, and accessible UX.
8. *Cross-Domain Outcome Metrics:* Evaluate both physical and mental health outcomes in tandem, rather than in silos.

IX. CONCLUSION

AI is shaping a new era of wellness-embedding predictive, adaptive, and personalized support across physical and mental health domains. Secondary data and literature support promising outcomes in activity monitoring, risk prediction, and mental health symptom reduction. However, the journey is nascent: rigorous validation, inclusive datasets, explain-ability, equitable access, and human-centered deployment remain critical. With multidisciplinary collaboration and responsible governance, AI can play a vital role in sustainable, holistic wellness systems.

X. RECOMMENDATIONS FOR FUTURE RESEARCH AND APPLICATION

Longitudinal Studies: Future research should consider conducting long-term studies to evaluate the sustained impact of interventions such as AI-driven wellness applications on physical and mental health outcomes.

Larger and Diverse Samples: Expanding the sample size and including participants from diverse demographic backgrounds (e.g., age, socioeconomic status, and region) will enhance the generalizability of findings.

Integration of Technology: Incorporating wearable fitness trackers, mobile health apps, or AI-based feedback systems can provide more objective data and real-time monitoring of body composition, activity levels, and psychological well-being.

Comparative Studies: Comparing AI-enabled wellness interventions with traditional methods (e.g., yoga, HIIT, conventional therapy) can help identify the most effective strategies for improving body image, motivation, and mental resilience.

Psychosocial Dimensions: Future studies should explore additional psychosocial factors such as stress, self-esteem, social support, and motivation, which may mediate or moderate the impact of AI-guided wellness programs.

Interdisciplinary Collaboration: Researchers should collaborate across fields such as computer science, psychology, public health, and behavioural science to design holistic wellness interventions that leverage both human and technological inputs.

Application in Educational Institutions: Implementation of structured physical activity programs supported by AI (e.g., smart fitness coaching) in schools and universities can promote wellness among youth. Institutions should also integrate AI-supported wellness literacy into curriculum.

Policy and Community Outreach: Based on positive outcomes, policymakers and health professionals can develop community-level programs and guidelines that promote active lifestyles using accessible AI-based wellness formats

Regulatory Sandboxes for AI Wellness Tools: government's and healthcare regulators should establish regulatory sandboxes to test AI wellness applications in controlled environments. This enables innovation while ensuring compliance with safety, privacy, and ethical guidelines before large-scale rollout.

Interoperability with Electronic Health Records (EHRs): AI-based wellness platforms should be interoperable with existing EHR systems to allow seamless data exchange, enabling clinicians to contextualize AI insights within broader medical records for more effective decision-making.

User-Centered Design and Co-Creation: Developers should adopt participatory design approaches involving users, clinicians, and caregivers throughout the development cycle to improve usability, trust, inclusivity, and sustained engagement with AI wellness tools.

Multilingual and Cross-Cultural Adaptability: To ensure equitable access and minimize cultural bias, AI wellness applications should be designed with multilingual interfaces and adaptable frameworks that respect regional health beliefs, communication styles, and behavioral norms.

Responsible AI Curriculum for Healthcare Professionals: Introduce AI literacy modules in medical, psychological, and public health education to ensure future professionals can effectively interpret, implement, and critique AI-driven tools in wellness domains.

Real-Time Feedback Loops in AI Systems: Future AI wellness platforms should incorporate real-time feedback loops that adjust recommendations based on user input, physiological changes, or environmental data, improving personalization and user engagement.

Green and Sustainable AI Design: Design energy-efficient AI models and edge-computing architectures to reduce the environmental footprint of wearable and mobile wellness applications, especially for continuous health monitoring.

Crisis Detection and Escalation Protocols: Integrate automated crisis detection (e.g., suicidal ideation, severe distress) with ethically designed escalation frameworks that connect users to human clinicians or helplines, ensuring safety and accountability.

REFERENCES

- [1] Chung, W., Han, J., Kim, J., & Park, J. (2021). Artificial Intelligence Applications in Health Care: Review and Future Directions. *Healthcare Informatics Research*, 27(1), 1–15. <https://doi.org/10.4258/hir.2021.27.1.1>
- [2] Davenport TH, Ronanki R. (2018). Artificial Intelligence for the Real World. *Harvard Business Review*. Vol: 96 (1), PP:108-116.
- [3] Esteve, A., Robicquet, A., Ramsundar, B., Kuleshov, V., DePristo, M., Chou, K., & Dean, J. (2019). A guide to deep learning in healthcare. *Nature Medicine*, 25, 24–29. <https://doi.org/10.1038/s41591-018-0316-z>
- [4] Grand View Research. (2023). AI in Mental Health Market Size, Share & Trends Analysis Report. <https://www.grandviewresearch.com/industry-analysis/ai-mental-health-market-report>
- [5] Guo, Y., Li, Y., & Yu, B. (2022). Artificial intelligence in mobile health (mHealth): Scoping review. *JMIR mHealth and uHealth*, 10(6), e35053. <https://doi.org/10.2196/35053>
- [6] Inkster, B., Sarda, S., & Subramanian, V. (2018). An empathy-driven, conversational artificial intelligence agent (Wysa) for digital mental well-being: Real-world data evaluation. *JMIR mHealth and uHealth*, 6(11), e12106. <https://doi.org/10.2196/12106>
- [7] Jobin A, Ienca M, Vayena E. (2019). The global landscape of AI ethics guidelines. *Nat Mach Intell*. Vol: 1(9). PP:389-399. <https://doi.org/10.1038/s42256-019-0088-2>
- [8] Luxton, D. D. (2021). Artificial Intelligence in Behavioral and Mental Health Care. Elsevier.
- [9] Ng, A. (2022). Opportunities and Obstacles for Deep Learning in Biology and Medicine. *Nature Biomedical Engineering*, 6, 1–5.
- [10] Onnela, J. P., & Rauch, S. L. (2016). Harnessing smartphone-based digital phenotyping to enhance behavioral and mental health. *Neuropsychopharmacology*, 41(7), 1691–1696. <https://doi.org/10.1038/npp.2016.7>
- [11] Shaik, S., Tripathi, A., & Neha, R. (2023). Artificial intelligence for remote health monitoring: Opportunities and challenges. *arXiv preprint*. <https://arxiv.org/abs/2301.10009>
- [12] Suarjana, I. N., Suranadi, I. W., & Maharani, D. R. (2023). The value of AI mental health chatbots in public health: A narrative review. *Journal of Public Health*, 46(2), e300–e307. <https://doi.org/10.1093/pubmed/fdad005>
- [13] Tegmark M, (2017). Life 3.0: Being Human in the Age of Artificial Intelligence. Knopf.
- [14] Topol, E. (2019). Deep medicine: How artificial intelligence can make healthcare human again. Basic Books.
- [15] Torous, J., & Roberts, L. W. (2017). Needed Innovation in Digital Health and Smartphone Applications for Mental Health: Transparency and Trust. *JAMA Psychiatry*, 74(5), 437–438. <https://doi.org/10.1001/jamapsychiatry.2017.0262>
- [16] Tzirakis, P., Trigeorgis, G., Nicolaou, M. A., Schuller, B., & Zafeiriou, S. (2021). End-to-end multimodal emotion recognition using deep neural networks. *IEEE Journal on Selected Topics in Signal Processing*, 15(5), 1141–1154. <https://doi.org/10.1109/JSTSP.2021.3079954>
- [17] World Health Organization (WHO). (2023). Artificial Intelligence in Mental Health Research: Applications, challenges and recommendations. <https://www.who.int/europe/news/item/06-02-2023-artificial-intelligence-in-mental-health-research--new-who-study-on-applications-and-challenges>
- [18] World Metrics. (2024). AI in the Psychology Industry: Key Insights & Data. <https://worldmetrics.org/ai-in-the-psychology-industry-statistics>
- [19] Zipdo. (2024). AI in the Mental Health Industry: Statistics and Trends. <https://zipdo.co/ai-in-the-mental-health-industry-statistics/>