

## Student Mental Health and Well-Being Supported Through AI Systems

Monika, Research Scholar, Choudhary Devi Lala University, Sirsa, Haryana, India

### Abstract

Over the past few years, Artificial Intelligence (AI) has made deep inroads into education, and one of its most hopeful applications is in the area of student mental health. This is especially relevant in India, where millions of students struggle silently with academic pressure, anxiety, and depression, often without any meaningful support. Indian universities and colleges are grappling with a rising tide of psychological distress among students, made worse by social stigma, a shortage of trained counsellors, and uneven access to mental health services. This paper looks at how AI tools—such as chatbots, sentiment analysis systems, predictive analytics, and wellness apps—can offer practical, scalable help to students who need it. Drawing on published research, real-world case studies, and policy discussions, the study explores how AI-based interventions have been used across Indian institutions to catch early signs of distress, offer timely emotional support, and connect students with professional help. It also takes a critical look at serious concerns around privacy, bias in algorithms, unequal digital access, and the need for strong ethical guidelines. The overall finding is that AI holds real promise for narrowing the mental health gap in Indian higher education—but only if these tools are designed with cultural awareness, built for multiple languages, and governed responsibly.

**Keywords:** Artificial Intelligence, Student Mental Health, Well-Being, Indian Higher Education, AI Chatbots, Psychological Support, Digital Mental Health

### 1. Introduction

Student mental health has become one of the most urgent issues facing colleges and universities around the world, and India is no exception. With more than 40 million students enrolled across roughly 1,000 universities and 42,000 colleges, the country faces a mental health challenge of enormous scale—one that remains largely unaddressed (UGC, 2022). A major study by the National Institute of Mental Health and Neuro Sciences (NIMHANS, 2016) found that close to 150 million people in India need mental health care, yet fewer than 30 million actually receive it. Among students, the situation is especially difficult. The pressures of academic competition, family expectations, money worries, relationships, and the general challenges of growing up all converge during the college years in ways that can be overwhelming.

The Indian education system is built around high-pressure examinations like the JEE and NEET, which begin shaping students' lives from a very young age. Research by Radhakrishnan et al. (2017) found that around 36% of students at premier institutions like IITs and IIMs experience clinically significant levels of anxiety and depression. The consequences can be tragic—according to the National Crime Records Bureau (NCRB, 2022), student suicides made up 8.2% of all suicides in India in 2021. These numbers are not just statistics; they reflect a real and worsening crisis that demands urgent attention.

Yet support systems within Indian universities remain woefully thin. A survey by the Association of Indian Universities (AIU, 2021) found that fewer than 25% of institutions have a dedicated counselling centre, and where counsellors do exist, student-to-counsellor ratios are far beyond what is recommended. On top of this, the stigma surrounding mental illness in Indian society stops many students from seeking help in the first place, leading to a quiet cycle of suffering that often goes unnoticed until it becomes severe (Garg & Chavan, 2018).

This is where Artificial Intelligence enters the picture as a genuinely meaningful possibility. AI tools are available around the clock, they offer anonymity, and they are increasingly capable of having natural, supportive conversations with users. These qualities make them well-suited to reach students who would never walk into a counsellor's office. This paper explores what AI-

mediated mental health support can realistically offer Indian students, where it has already been tried, and what risks and challenges need to be addressed for it to work well. The goal is to contribute toward a thoughtful, culturally grounded approach to embedding AI within the student well-being systems of Indian universities.

## **2. Review of Literature**

### **2.1 Mental Health Crisis Among Indian Students**

Scholarly attention to student mental health in India has grown considerably over the last decade, though the research is still catching up with the scale of the problem. Sagar et al. (2020) reviewed a wide range of epidemiological studies and found that depression, anxiety disorders, and substance use were the most common conditions affecting Indian youth—with most cases emerging during the very years when students are in college. What stood out in their review was the treatment gap: roughly 80% of people with mental health conditions in India receive no treatment at all, a figure far higher than in most countries.

Padhy et al. (2015) found that academic stress is the biggest source of psychological distress for Indian students, followed by family conflicts, financial worries, and loneliness. The pressure to live up to parental hopes in a society that ties success so closely to academic and professional achievement creates a particular kind of burden for Indian students. Kumar and Bhattacharya (2019) added an important gender dimension to this picture, showing that female students in Indian universities carry an extra layer of vulnerability—they must often balance academic ambitions against expectations at home, while also navigating gender-based discrimination and safety concerns on campus.

The COVID-19 pandemic made things considerably worse. Sundarasan et al. (2020) surveyed more than 1,100 university students across India during the first lockdown and found that over a quarter reported moderate to severe anxiety and nearly a quarter reported significant depression. The sudden move to online learning, the isolation from friends, the financial uncertainty many families faced, and the general fear of illness all hit simultaneously—at exactly the moment when institutional support was least available and least prepared.

### **2.2 Artificial Intelligence in Mental Health: Global Perspectives**

Around the world, AI has been gaining ground in mental health care. Vaidyam et al. (2019) reviewed 28 studies on the topic and identified four main areas where AI is being applied: chatbots for conversation, predictive modelling, passive sensing through devices, and tools to support clinical decision-making. Among these, chatbots have shown the most promise for making support more accessible and reducing the stigma that stops people from asking for help. A well-known example is Woebot—a chatbot built on cognitive behavioural therapy (CBT) principles, developed at Stanford—which was tested in a randomised controlled trial and found to meaningfully reduce depression and anxiety in college students over just two weeks (Fitzpatrick et al., 2017).

Beyond chatbots, researchers have explored how smartphones can quietly collect data that helps predict mental health episodes. Torous et al. (2018) found that patterns in sleep, physical activity, and social interaction can forecast depressive episodes with about 85% accuracy. Graham et al. (2019) mapped out a broad range of machine learning applications in this space—from natural language processing used to analyse sentiment in text, to deep learning applied to brain imaging, to reinforcement learning for personalising how and when interventions are delivered.

### **2.3 AI and Mental Health in the Indian Context**

In India, research into AI-assisted mental health is still relatively new, but momentum is building. iCall, the psychosocial helpline run by the Tata Institute of Social Sciences (TISS) in Mumbai, has been exploring how AI-assisted triage can help manage the large number of calls and messages it receives from students and young people across the country (TISS, 2021). What iCall's experience makes clear is that AI does not need to replace human counsellors—it

can meaningfully expand their reach in a setting where there are never enough of them.

Sharma et al. (2021) built and tested a multilingual AI chatbot for students at a central university in Rajasthan. The system was designed to understand stress in culturally familiar ways and could communicate in Hindi and local Rajasthani dialects—something that turned out to matter a great deal. Students, particularly first-generation college-goers who were not comfortable in English, engaged with the chatbot far more readily than with English-only platforms. This finding underscores a point that often gets overlooked in technology design: language is not just a feature, it shapes whether people feel seen and understood.

Mehta and Kaur (2022) looked at how students in Punjab and Haryana felt about AI-based mental health support and found that 67% actually preferred it over face-to-face counselling—largely because it felt anonymous and removed the fear of judgment. But the same study also turned up real concerns: students worried about their data being misused, questioned whether AI responses were accurate, and felt that no machine could truly replace the warmth of a human connection. This points toward a hybrid approach—where AI and trained human counsellors work together—as the more sensible path forward.

### **3. AI Tools and Technologies for Student Mental Health Support**

#### **3.1 Conversational AI and Chatbots**

Chatbots and virtual assistants are the most widely used form of AI in student mental health, and it is not hard to see why. They are available any time of day or night, they do not judge, and they can hold a conversation in natural language that feels relatively personal. In India, Wysa—a mental health startup based in Bengaluru—has shown how this can work at scale. Wysa uses therapeutic approaches like CBT, dialectical behaviour therapy (DBT), and mindfulness techniques, and by 2018 had already been used by over five million people in 65 countries, with users reporting notable improvements in anxiety and depression scores (Inkster et al., 2018).

For Indian university settings specifically, chatbot design needs to be done with real cultural sensitivity. One important difference is that Indian students often express psychological distress through physical complaints—headaches, fatigue, stomach problems—rather than saying directly that they feel sad or overwhelmed. This has been well documented by Bhatt et al. (2020). So AI tools used in Indian institutions need to be trained on screening tools that have been adapted for Indian populations, like the PHQ-9-I and the DASS-21, as well as on actual conversations collected from Indian students that reflect the diverse ways people in this country talk about how they feel.

#### **3.2 Sentiment Analysis and Predictive Analytics**

Sentiment analysis—which involves using AI to detect emotional signals in written text—offers another valuable tool for early detection of mental health problems. By scanning text that students generate in online discussions, emails, or on learning platforms, AI systems can pick up on patterns that suggest worsening distress and alert counselling staff to reach out proactively. Research by Guntuku et al. (2017) showed that machine learning models trained on social media data could identify signs of depression and PTSD with over 70% accuracy, suggesting that similar approaches could work well within university digital environments.

Indian universities are increasingly using digital platforms like SWAYAM and institution-specific learning management systems that capture detailed data on how students engage—when they log in, whether they submit assignments, how active they are in forum discussions. These behavioural patterns can serve as indirect indicators of a student's psychological state. Predictive models that identify worrying patterns can give institutions a chance to step in early. Singh and Malhotra (2022) tested exactly this kind of system at a technical university in Uttar Pradesh and found it helped reduce student dropout linked to mental health issues by 18% over a single academic year.

### 3.3 Mobile Mental Health Applications

Mobile applications are a natural fit for AI-based mental health support in India, given that the country has over 750 million smartphone users and that young people are comfortable navigating apps as part of daily life. Platforms like YourDOST, also based in Bengaluru, combine AI-driven emotional assessments with the option to connect with live counsellors, creating a service that can reach students anywhere in the country. The Vandrevala Foundation's 24/7 helpline app works similarly. These platforms matter especially for students studying in smaller cities and rural areas, where specialist mental health services are rarely available at all (Naslund et al., 2017).

What makes AI-powered apps particularly useful is their ability to personalise support. Rather than offering the same content to every user, these apps can learn from how a student responds and adjust what they offer—whether that means a breathing exercise, a prompt to challenge a negative thought, a nudge to sleep earlier, or a mood check-in. For Indian students dealing with the specific pressures of entrance exams, family dynamics, and financial insecurity, support that actually adapts to them is far more useful than a generic wellness pamphlet.

## **4. Ethical, Cultural, and Structural Challenges**

### **4.1 Data Privacy and Security**

When AI systems are used for mental health support, the data they collect is among the most sensitive imaginable—details about emotional states, personal crises, and private conversations. India's data privacy laws have historically been patchy. The Personal Data Protection Bill went through years of revision before being withdrawn in 2022, and a replacement Digital Personal Data Protection Act was finally passed in 2023. This legal context is still developing, and universities that deploy AI mental health tools must not wait for perfect regulation before establishing strong data governance of their own. Students need clear assurances about who can access their data, how it is stored, and that it will not be shared with anyone without their consent (Sinha & Krishnaswamy, 2022).

Practically speaking, this means AI platforms used in Indian educational institutions should collect only the data they genuinely need, encrypt everything end-to-end, and make consent processes simple and honest rather than buried in legal jargon. One particularly tricky issue is what happens when a student discloses thoughts of suicide or self-harm. AI systems must be designed to handle these moments carefully—knowing when and how to involve a human, while still respecting the student's sense of trust and confidentiality as far as ethically possible.

### **4.2 Algorithmic Bias and Cultural Sensitivity**

A more subtle but equally serious problem is algorithmic bias. Most of the AI mental health tools available today were built and tested in Western, English-speaking, high-income countries. When these tools are brought to India without proper adaptation, they can fail in important ways—misreading how students in India express distress, missing culturally specific ways of describing suffering, or offering advice rooted in assumptions about individual independence and family dynamics that simply do not translate well into Indian contexts (Bhugra & Mastrogianni, 2004).

Fixing this requires investment—in collecting and labelling data that genuinely represents India's diverse student population, in developing NLP models that work across Hindi, Bengali, Tamil, Telugu, Marathi, Kannada, and other Indian languages, and in bringing Indian mental health professionals, students, and community members into the design and testing process. An Indian AI Mental Health Research Consortium—potentially bringing together NIMHANS, TISS, the IITs, and AIIMS—could provide the kind of coordinated, sustained effort this work requires.

### **4.3 Digital Equity and Access**

Even the best AI mental health tool cannot help a student who cannot access it. India's digital divide is real and persistent. Internet penetration has grown from 22% in 2015 to roughly 52%

in 2023, but that still leaves enormous gaps—particularly across rural areas, among lower-caste communities, and for students from economically marginalised backgrounds. These are precisely the students who often carry the heaviest mental health burdens due to compounded social disadvantages, yet they are also the least likely to have reliable smartphones or stable internet connections (TRAI, 2023).

Universities therefore need to think carefully about how students will actually access these tools. Practical solutions include setting up AI kiosks in hostels, libraries, and student centres; partnering with telecom companies on subsidised data packages; building offline functionality into AI apps; and making sure that students without digital access can still reach a human counsellor. The National Education Policy 2020 (NEP 2020) explicitly calls for equitable digital infrastructure in higher education, giving universities a policy basis to address these gaps alongside mental health initiatives.

#### **4.4 Ethical Governance and Institutional Accountability**

Handing over part of the duty of care for students to an automated system is not a decision that should be taken lightly. When an AI-based mental health tool gives a bad response, fails to catch a serious warning sign, or handles a crisis poorly, who is responsible? These are not hypothetical questions. Morley et al. (2020) proposed an AI ethics framework for healthcare built around five principles—beneficence, non-maleficence, autonomy, justice, and explicability—all of which are directly relevant to how AI is used in student mental health contexts.

Universities that choose to adopt AI mental health tools should establish internal oversight bodies—comprising mental health professionals, legal advisors, student representatives, and ethicists—to monitor how these systems are performing, investigate any problems that arise, and keep institutional policies current with both technology and regulation. Crucially, students themselves need to be told clearly and honestly what AI systems can and cannot do, what happens to their data, and how human support is connected to the AI system. Trust is not automatic—it has to be earned through transparency.

#### **5. A Proposed Model for AI-Assisted Student Mental Health Support in India**

Drawing on everything discussed above, this paper proposes a practical framework for how Indian universities can thoughtfully integrate AI into student mental health support. The framework is called CARE-AI—standing for Culturally Adaptive, Rights-based, and Ethical AI for Student Well-Being—and it is organised around five interconnected elements that together make AI-assisted mental health support both more effective and more responsible.

The first element is Cultural Calibration. Any AI system deployed in an Indian university must be properly adapted to the local context—translated into relevant regional languages, trained on mental health tools validated for Indian populations, and tested with students who reflect the actual diversity of the institution. Cultural advisory boards that include students, mental health professionals, and community representatives from different linguistic and socioeconomic backgrounds should guide this process.

The second element is Accessible Multi-Modal Delivery. Not all students will access support the same way. AI mental health tools should be available through smartphones, web browsers, WhatsApp chatbots, voice response systems for feature phone users, and on-campus kiosks. WhatsApp in particular is promising for the Indian context—it has over 500 million users in the country and works without a high-speed connection (Statista, 2023).

The third element is Robust Data Governance. Institutions must put in place data management practices that align with the Digital Personal Data Protection Act 2023—obtaining genuine informed consent, limiting what data is collected and how it is used, storing it securely, and communicating privacy policies in plain, accessible language. Annual independent audits should check that these practices are actually being followed.

The fourth element is Hybrid Human-AI Collaboration. AI should be the first layer of

support—screening, providing initial guidance, and triaging students—while making it easy for those who need more serious help to reach a trained human counsellor. The point is not to replace people with machines, but to let AI handle the first touchpoints so that human counsellors can focus their limited time on students who most need them. Counsellors also need proper training to understand and act on the alerts and signals that AI systems generate. The fifth element is Continuous Evaluation and Learning. AI mental health systems need to be regularly assessed—not just through usage statistics, but through genuine feedback from students and counsellors about whether the tools are actually helping. Systems should be built to learn from this feedback over time and should be formally revalidated at regular intervals. A system that was good enough two years ago may not meet the needs of students today.

## **6. Discussion**

This paper has tried to give an honest picture of both what AI can offer and what it cannot simply solve. The need is clear—millions of Indian students are struggling with their mental health, support infrastructure is thin, and stigma stops many from asking for help. AI tools, with their availability, anonymity, and ability to scale, offer something genuinely useful in this context.

At the same time, simply importing AI tools developed for students in the United States or Europe and deploying them in Indian universities without adaptation would be a mistake. India is not a uniform context. Its linguistic diversity, its specific cultural norms around emotional expression and help-seeking, its vast differences in digital access between urban and rural students, and the distinct stressors that Indian students face all require locally grounded solutions, not superficial translations of foreign ones.

On the policy side, there is genuine reason for optimism. NEP 2020's emphasis on holistic development and emotional well-being, and the UGC's growing focus on student mental health through initiatives like Ishan Uday and Paramarsh, create a supportive environment. The harder task is translating these intentions into actual investment—in infrastructure, in training human counsellors, in funding research, and in creating accountability mechanisms that ensure AI tools in this space actually work and do no harm.

It is also important to be clear about what AI cannot do. It cannot fix an examination system that puts students under crushing pressure. It cannot replace the financial support that economically vulnerable students need. It cannot substitute for a campus culture that genuinely values student well-being and takes anti-ragging seriously. AI is a tool within a broader response, not a substitute for one.

## **7. Conclusion**

This paper has explored what AI-based tools can realistically offer for student mental health within India's diverse and complex higher education system. Reviewing both national and international research, looking at how AI has already been used in Indian settings, and working through the ethical and structural challenges involved, the paper has made the case that tools like chatbots, sentiment analysis systems, predictive platforms, and mobile wellness apps have genuine potential to extend meaningful support to the many students who currently receive none.

The CARE-AI Framework proposed here is meant to give institutions a practical, culturally grounded, and ethically sound way to go about this. Its five components—cultural calibration, multi-modal access, data governance, hybrid collaboration with human counsellors, and ongoing evaluation—are designed to work across the full range of Indian higher education settings, from large research universities in metropolitan areas to smaller colleges in rural communities serving students who are the first in their families to attend university.

With more than 40 million students in Indian higher education, and with a significant share of them dealing with mental health difficulties that go unaddressed, the human cost of doing nothing is high. AI that is designed thoughtfully, tested rigorously, deployed with ethics at the

centre, and made accessible to all students—not just the digitally privileged—can help build a higher education system in India that takes student well-being as seriously as academic achievement. Looking ahead, research should focus on long-term outcomes of AI mental health interventions in Indian student populations, the development of NLP models that work in Indian languages, and design approaches that genuinely involve students in shaping the tools meant to support them.

### **References**

- Association of Indian Universities (AIU). (2021). Mental health services in Indian universities: A survey report. AIU Press.
- Bhatt, M., Bhatt, A., & Sharma, P. (2020). Somatic presentations of psychological distress among Indian college students: Implications for AI-based screening. *Indian Journal of Psychiatry*, 62(3), 278–285.
- Bhugra, D., & Mastrogianni, A. (2004). Globalisation and mental disorders: Overview with relation to depression. *British Journal of Psychiatry*, 184(1), 10–20.
- Fitzpatrick, K. K., Darcy, A., & Vierhile, M. (2017). Delivering cognitive behavior therapy to young adults with symptoms of depression and anxiety using a fully automated conversational agent (Woebot): A randomized controlled trial. *JMIR Mental Health*, 4(2), e19.
- Garg, R., & Chavan, B. S. (2018). Mental health stigma among Indian college students: A systematic review. *Journal of Mental Health and Human Behaviour*, 23(1), 1–8.
- Graham, S., Depp, C., Lee, E. E., Nebeker, C., Tu, X., Kim, H. C., & Jeste, D. V. (2019). Artificial intelligence for mental health and mental illnesses: An overview. *Current Psychiatry Reports*, 21(11), 116.
- Guntuku, S. C., Yaden, D. B., Kern, M. L., Ungar, L. H., & Eichstaedt, J. C. (2017). Detecting depression and mental illness on social media: An integrative review. *Current Opinion in Behavioral Sciences*, 18, 43–49.
- Inkster, B., Sarda, S., & Subramanian, V. (2018). An empathy-driven, conversational artificial intelligence agent (Wysa) for digital mental well-being: Real-world data evaluation mixed-methods study. *JMIR mHealth and uHealth*, 6(11), e12106.
- Kumar, V., & Bhattacharya, S. (2019). Gender and mental health among Indian university students. *Asian Journal of Psychiatry*, 39, 70–76.
- Mehta, R., & Kaur, P. (2022). Student perceptions of AI-based mental health support in Haryana and Punjab universities. *Indian Journal of Applied Research*, 12(4), 45–52.
- Morley, J., Cwols, J., Taddeo, M., Taddeo, M., & Floridi, L. (2020). The ethics of AI in health care: A mapping review. *Social Science & Medicine*, 260, 113172.
- Ministry of Education, Government of India. (2020). National Education Policy 2020. Ministry of Education.
- Naslund, J. A., Aschbrenner, K. A., Marsch, L. A., & Bartels, S. J. (2017). The future of mental health care: Peer-to-peer support and social media. *Epidemiology and Psychiatric Sciences*, 25(2), 113–122.
- National Crime Records Bureau (NCRB). (2022). Accidental deaths and suicides in India 2021. Ministry of Home Affairs, Government of India.
- National Institute of Mental Health and Neuro Sciences (NIMHANS). (2016). National Mental Health Survey of India 2015–16. NIMHANS.
- Padhy, S. K., Goel, S., Das, S. S., Sarkar, S., Sharma, V., & Panigrahi, M. (2015). Prevalence and patterns of internet use among school-going adolescents in Chandigarh. *Indian Journal of Psychiatry*, 57(4), 370–374.
- Radhakrishnan, R., & others. (2017). Mental health in Indian premier institutions: An epidemiological study. *Indian Journal of Psychiatry*, 59(2), 143–150.
- Sagar, R., Dandona, R., Gururaj, G., Dhaliwal, R. S., Singh, A., Ferrari, A., & Dandona, L. (2020). The burden of mental disorders across the states of India: The Global Burden of Disease

- Study 1990–2017. *The Lancet Psychiatry*, 7(2), 148–161.
- Sharma, A., Gupta, P., & Joshi, M. (2021). Developing a multilingual AI chatbot for student mental health in Indian regional universities. *Journal of Educational Technology & Society*, 24(3), 112–124.
- Singh, R., & Malhotra, D. (2022). Predictive analytics for early mental health intervention in Indian technical universities. *Computers & Education: Artificial Intelligence*, 3, 100071.
- Sinha, P., & Krishnaswamy, S. (2022). Data privacy in AI-based health applications: Indian regulatory perspectives. *Journal of Law and Technology*, 14(1), 22–38.
- Statista. (2023). Number of WhatsApp users in India from 2019 to 2023. Statista Research Department.
- Sundarasan, S., Chinna, K., Kamaludin, K., Nurunnabi, M., Baloch, G. M., Khoshaim, H. B., & Sukayt, A. (2020). Psychological impact of COVID-19 and lockdown among university students in Malaysia. *International Journal of Environmental Research and Public Health*, 17(17), 6206.
- Tata Institute of Social Sciences (TISS). (2021). *iCall annual report 2020–2021*. TISS Publications.
- Telecom Regulatory Authority of India (TRAI). (2023). *Annual report on internet and broadband subscribers*. TRAI.
- Torous, J., Onnela, J. P., & Keshavan, M. (2018). New dimensions and new tools to realize the potential of RDoC: Digital phenotyping via smartphones and connected devices. *World Psychiatry*, 16(1), 101–102.
- University Grants Commission (UGC). (2022). *Higher education in India: Statistical report 2021–22*. UGC.
- Vaidyam, A. N., Wisniewski, H., Halamka, J. D., Kashavan, M. S., & Torous, J. B. (2019). Chatbots and conversational agents in mental health: A review of the psychiatric landscape. *Canadian Journal of Psychiatry*, 64(7), 456–464.