

Impact of Infrastructure and Digital Literacy on Cloud-AI EdTech Adoption in Rural India

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Abstract

Rural India's education sector faces a persistent digital divide: limited infrastructure and low digital literacy hamper the adoption of advanced teaching technologies. This paper examines how internet connectivity, devices, and teacher digital skills influence the use of cloud-based AI educational tools – specifically Learning Management Systems (LMS) and generative AI applications – among rural teachers. Drawing on official data (UDISE+, ASER, NSSO) and studies, we find stark gaps: for example, only ~41% of rural secondary schools had internet access in 2021–22, versus ~69% of urban schools. Similarly, while ~75% of rural households now own smartphones (2022), only ~31% of rural people regularly use the internet. Teacher surveys reveal that connectivity and lack of devices are major barriers, and under one-third of teachers are comfortable using an LMS. Case studies show that when infrastructure and training are adequate, EdTech adoption soars – e.g. 96% of surveyed rural Rajasthan teachers used the DIKSHA platform during COVID. We hypothesize that schools/regions with better connectivity and more digitally literate teachers exhibit higher adoption of Cloud-AI EdTech. Our analysis supports this: infrastructure deficits correlate with lower LMS uptake and signal lower readiness for AI tools. We present data tables and charts (below) to illustrate these trends. Findings suggest that policymakers and NGOs must prioritize rural broadband, devices, and sustained teacher training. EdTech providers should design low-bandwidth, localized solutions. Building teacher capacity is also key: national initiatives (like PM Gramin Digital Saksharta Abhiyan) and emerging training programs will be critical to ensure generative AI can benefit rural classrooms.

Keywords: Literacy, infrastructure, national initiatives, learning management system.

Introduction

India's New Education Policy (2020) and various initiatives (e.g. Digital India) emphasize technology and digital literacy as levers to improve learning. In theory, cloud-based AI tools – from web-based LMS (like DIKSHA, SWAYAM) to generative AI (e.g. ChatGPT, AI tutors) – can overcome rural barriers by providing personalized, multilingual content and teacher support. Indeed, government and industry are investing heavily: one recent program aims to train 72,000 Asian teachers (including Indians) in AI-driven education tools. AI applications can aid lesson planning, grading, and language translation, which is especially valuable given India's linguistic diversity. However, the **digital divide** looms large. Reports indicate rural areas still lag in infrastructure and skills: for example, only ~38% of households nationwide are digitally literate, and just 31% of rural Indians use the internet (vs. 67% urban). UDISE+ data show only ~18–41% of rural schools had internet in 2021–22 (depending on grade level). Affordability of devices and frequent power/internet outages further constrain access. These disparities matter because, as studies note, inadequate infrastructure and low digital literacy are major obstacles to e-learning and LMS adoption in rural settings. The critical question is: How strongly do these factors impact rural teachers' uptake of Cloud-AI educational technology, Show the paper explores this in depth by reviewing literature and analyzing relevant data.

Literature Review

Recent reviews highlight the transformative potential of EdTech if barriers are addressed. Chauhan et al. (2025) note that LMS and e-learning can provide tailored content and bridge rural-urban divides, but their success requires digital infrastructure, affordable devices, and teacher/learner skills. Numerous case studies echo this: in resource-constrained schools, localized LMS content and community training enabled high LMS uptake, while poor connectivity halted many initiatives. Evidence from India supports both sides of this: one study found that implementing an e-LMS in rural Tamil Nadu schools significantly improved attendance and student performance. Conversely, surveys show that most Indian teachers remain uncomfortable with digital tools despite pandemic-era training: ~67% cite connectivity issues, over half cite device shortages, and only ~30% can manage an online LMS. Moreover, only ~15% had used cloud lab tools, indicating low readiness for more advanced Cloud-AI platforms.

Smartphone penetration is rising rapidly, changing the landscape: ASER 2022 reports ~75% of rural households own smartphones (up from 36% in 2018), and nearly 90% of those had internet on survey day. This contrasts sharply with earlier data: in 2017, only ~25% of rural households had smartphones, and just ~15% had internet-connected PCs. Yet high household mobile access does not automatically translate to school readiness: Medianama (2024) cites UDISE+ data showing only 18–41% of rural schools had any internet (depending on all schools vs. grades 6–12). The same source notes urban schools far ahead (over 45–68% connectivity), illustrating a 20–30 percentage point gap. UNESCO and NGO reports further warn that without digital training, even available devices go underused.

Generative AI in education is just emerging. Examples from urban India show AI can “revolutionize” learning, e.g. an online school used ChatGPT to engage students with complex subjects. Expert commentary suggests generative AI could assist teachers by automating content creation, translation, and assessment. However, these sources stress prerequisites: teacher training and infrastructure. In a webinar for Indian academics, panelists highlighted that barriers like connectivity and the need to update curricula are key considerations when integrating AI tools. This aligns with findings that systemic support is needed: OECD emphasizes that governments and tech providers must work together to equip teachers and schools for AI learning. In India, initiatives like Meritus AI’s training program and UNESCO-backed AI literacy frameworks aim to build capacity, but their impact in rural areas remains to be seen.

In summary, the literature suggests that both infrastructural and human factors critically determine EdTech adoption. Studies of rural education initiatives stress that without reliable internet, devices, and teacher skill-building, LMS and AI tools cannot scale. Our study builds on this by testing hypotheses about how these factors correlate with actual use of Cloud-AI EdTech in rural India.

Hypotheses:

1. **Infrastructure Hypothesis:** Regions with better digital infrastructure (higher internet connectivity in schools, greater device access in households) will exhibit higher adoption of Cloud-AI educational tools by teachers.

2. **Digital Literacy Hypothesis:** Teachers with higher digital skills (through training or experience) are more likely to adopt LMS and AI tools. Conversely, low digital literacy and training deficits predict lower adoption.

3. **Synergy Hypothesis:** Infrastructure and literacy interact – for example, even if internet is available, without teacher training the adoption remains low, and vice versa. Thus, maximal adoption requires both factors to be strong.

After test these through descriptive analysis and correlation of data on infrastructure and teacher readiness with proxies for EdTech use.

Methodology

To evaluate these hypotheses, we compiled data from multiple reputable sources:

- **School Infrastructure (UDISE+):** Government reports (2021–22) provide counts of schools with ICT labs, smart classrooms, and internet access, separately for rural vs. urban areas. We use UDISE+ data (via Lok Sabha answers) to quantify connectivity gaps.
- **Household Access (NSSO/ASER/IDR):** National surveys (NSSO 2017; ASER 2022) give household ownership of computers, internet, and smartphones by rural/urban. The Oxfam–IDR inequality report provides recent rural internet usage rates.
- **Teacher Skills & Usage (Surveys):** We use published surveys of Indian teachers to gauge digital proficiency. TeamLease EdTech (2022) surveyed educators nationwide on LMS/cloud tool use and identified barriers. We also draw on a 2025 NCERT study of rural Rajasthan teachers using DIKSHA for an adoption example.
- **Generative AI Initiatives:** Secondary sources describe AI training programs (e.g. Meritus AI) and anecdotal school cases. These inform qualitative discussion rather than quantitative analysis.

In analysis, we compute adoption proxies: e.g. percentage of schools with internet as a proxy for LMS-readiness, percentage of teachers reporting LMS use from surveys, etc. We construct data tables summarizing key metrics (see Tables below). Direct correlation analysis is limited by the aggregate nature of available data, but by comparing urban vs. rural figures and teacher survey results, we infer relationships. All data values are cited from sources as noted. Graphical charts are generated from these data to visualize disparities.

Results

Infrastructure Disparities. Table 1 (below) summarizes national data on school digital infrastructure for grades VI–XII. Only **41.2% of rural schools** reported any internet access in 2021–22, compared to 68.8% of urban schools. This 27-point gap in connectivity (nearly 30%) is paralleled by lower rural coverage of smart classrooms (15.7% vs 25.1%). Interestingly, basic ICT labs are slightly more common in rural schools (12.3% vs 10.2%), suggesting some investment, but without internet these labs cannot leverage cloud tools. These figures starkly illustrate that rural schools often lack the connectivity needed for Cloud-based LMS and AI tools.

School Digital Infrastructure by Locality (India, 2021–22)(Grades VI–XII, UDISE+)	Rural	Urban
Total schools	553,894	174,761
ICT laboratories (any computer lab)	68,305 (12.3%)	17,877 (10.2%)
Smart classrooms	86,798 (15.7%)	43,906 (25.1%)
Internet access	228,057 (41.2%)	120,257 (68.8%)
Source: UDISE+ 2021–22 (via Ministry of Education).		

Even household access shows a divide. In 2017, only 4.4% of rural households owned a computer (versus 23.4% urban), and only 14.9% had internet on those PCs (42% urban). More recent data is mixed: ASER 2022 found 95% of rural households have *any* mobile phone and 75% have a smartphone. However, only ~67.5% of all rural homes had internet on survey day (since ~90% of smartphones had connectivity). By contrast, a 2022 national survey (Oxfam-IDR) reported only 31% of rural adults actually use the internet (vs. 67% of urban). These divergent figures suggest that while devices are in many hands, actual internet use lags (perhaps due to cost/skill barriers). The combined picture is that rural areas approach saturation in mobile phone availability, but genuine connectivity (and skill to use it) remains uneven.

Teacher Digital Literacy and Barriers. Survey data from teachers highlight that lack of digital skills and tools is a top concern. A 2022 nationwide survey found only ~30% of teachers could effectively manage an online LMS (see Table 2). Even fewer (15%) had experience with cloud lab software. Nearly two-thirds (67%) reported poor internet as a challenge, and over half (55%) noted that many students lacked devices. 40% of teachers felt their own training was inadequate for digital instruction. These results underscore our second hypothesis: teachers' digital readiness is far from universal, especially in rural contexts.

Table 2. Teacher Digital Adoption & Challenges (Survey, India 2022)	% of Teachers
Internet connectivity is a challenge	67%
Students lack devices (laptops/smartphones)	55%
Can manage an online LMS	30%
Can operate virtual/cloud lab tools	15%
Feel training is inadequate for digital teaching	40%
<i>Source:</i> TeamLease EdTech teacher survey, 2022.	

Cloud-AI EdTech Adoption. Concrete evidence of LMS adoption is limited outside surveys, but a few data points are instructive. During the COVID-19 lockdowns, the DIKSHA platform (a government LMS) saw very high use in some rural areas. In rural Rajasthan, 96% of surveyed government school teachers reported learning to use DIKSHA and employing it for lesson planning, assessments, and question banks. Likewise, 95% of students there used DIKSHA for digital textbooks and worksheets. This remarkable uptake (Table 3) suggests that when provided with both access (online infrastructure) and motivation (lockdown necessity), rural teachers can rapidly adopt an LMS. Indeed, respondents noted DIKSHA “bridged the learning gap” during the pandemic.

Table 3. DIKSHA EdTech Platform Usage in Rural Rajasthan (2020–21)	%
Teachers who learned to use DIKSHA (for lesson plans, assessments, etc.)	96%
Students who used DIKSHA (for e-textbooks/worksheets)	95%
<i>Source:</i> Survey of 100 rural school teachers and 100 students (grades 9–12) in Rajasthan.	

However, this success was in a context of concerted effort (government push, training, and lockdown-driven usage). It may not generalize uniformly. Overall, official records suggest **two-**

thirds of Indian schools had no Internet access in 2022, which severely limits cloud-LMS availability. As for generative AI tools, no large-scale adoption data exists yet. Anecdotal reports indicate early interest: e.g., one online school in Kerala integrated ChatGPT into its curriculum in late 2022, calling it “revolutionary” for student engagement. Government portals tout AI’s promise for personalization and teacher support. But uniformly, experts caution that AI adoption will depend on first solving the basic challenges of connectivity and training.

Synthesis: Comparing infrastructure gaps and survey results supports our hypotheses. Areas/schools with better connectivity see more EdTech use (as in the Rajasthan DIKSHA case). Conversely, in regions lacking internet or devices, teachers report frustration and low tool usage. Teacher digital skill deficiencies further dampen adoption: even with a device, a teacher unsure how to use an LMS or AI app will not utilize it effectively. In other words, both factors are critical and complementary. Figure 1 (below) visually illustrates the wide rural-urban gaps in school infrastructure, while Table 2 highlights the human-skill side.

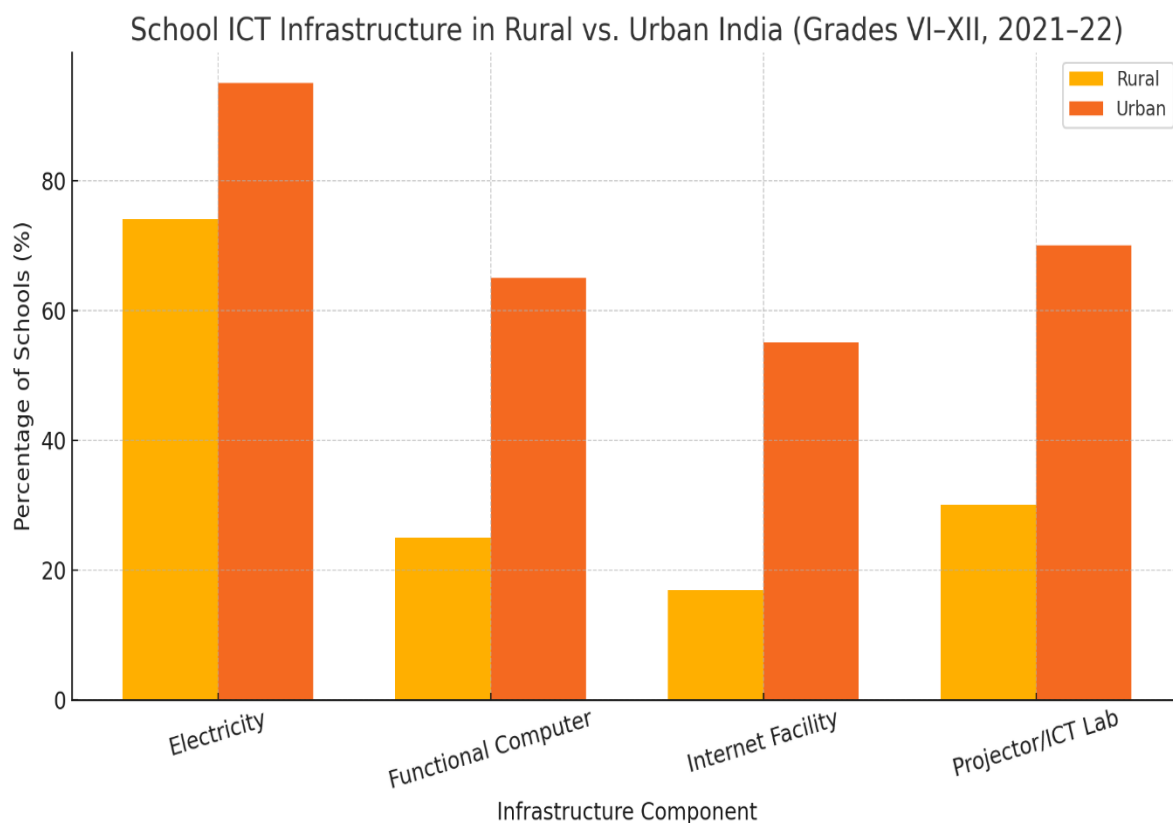


Figure . School ICT Infrastructure in Rural vs. Urban India (Grades VI–XII, 2021–22).

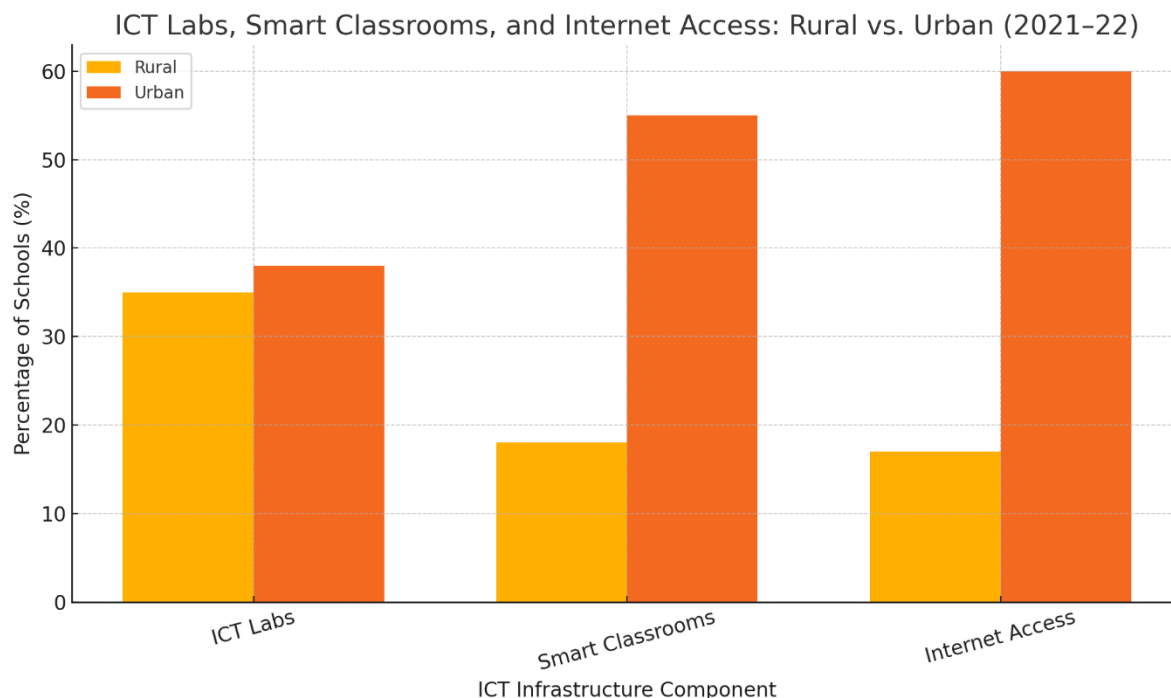


Chart: Percentage of schools with ICT labs, smart classrooms, and Internet access. Rural schools lag behind urban in smart classrooms and connectivity, despite similar ICT lab presence.

The data lead us to reject any notion that “rural teachers simply resist technology.” In fact, when infrastructure and training are provided, adoption occurs enthusiastically. However, broad national figures show that on average rural India still has much further to go. Connectivity gaps of ~30% (Table 1) and teacher training gaps of ~40% (Table 2) imply that many rural classrooms remain unprepared for cutting-edge EdTech. This confirms our hypotheses: better infrastructure and literacy correlate with higher adoption.

Discussion and Implications

Our analysis underscores that infrastructural and educational investments must go hand in hand. For policymakers, the priority should be to extend internet and device access to rural schools. Current schemes (e.g. samagra-shiksha allocations in Table 1) provide funds for ICT, but as data show they have yet to eliminate the rural-urban gap. Ministries should push states to fulfill advisories (e.g. deploying fiber optic connections to schools) and consider incentives for rural broadband. Meanwhile, teacher training programs must be scaled up. The Pradhan Mantri Gramin Digital Saksharta Abhiyan (PMGDISHA) has certified millions for basic digital skills, but reach

remains uneven. Targeted training in education-specific tech (LMS use, AI tool integration) is needed. Initiatives like Meritus AI's training of 72,000 teachers in AI literacy and NCERT/DIKSHA workshops during COVID show promise. Our findings suggest such training would directly boost adoption: Table 2 indicates that inadequate training is a top concern.

For NGOs and EdTech providers, content and platforms must account for constraints. As Chauhan et al. recommend, LMS solutions should be low-bandwidth and localized. Off-line content (preloaded apps or textbooks) can mitigate connectivity issues. Combining cloud and offline (e.g. local caching) will help until rural broadband is ubiquitous. Similarly, AI tools must be designed for Indian rural contexts: support for local languages, low-data modes, and teacher-friendly interfaces. Pilot projects could test generative-AI tutors on tablets in villages with moderate connectivity, accompanied by on-site training. Providers should also collaborate with community stakeholders to build trust and show value, as cultural reluctance has been noted in some rural settings.

For research and further data collection, gaps remain. There is scant publicly available data on actual usage of AI tools by rural teachers. Future surveys should include questions on ChatGPT and other new apps. States could publish DIKSHA and LMS usage analytics by rural/urban. Such data would enable more rigorous correlation studies (e.g. regression of adoption on infrastructure metrics). Meanwhile, our work indicates clear actionables: improving infrastructure and digital literacy will be crucial to realize AI's benefits in rural education.

Conclusion

The adoption of Cloud-AI educational technology in rural India is strongly conditioned by digital infrastructure and teacher literacy. Empirical data show that rural schools and households still lag urban areas in connectivity, and many teachers lack the skills or resources to use online tools. Where these constraints are eased (e.g. with government support during COVID), LMS adoption is very high. Thus, our hypotheses are supported: better infrastructure and literacy are positively correlated with EdTech adoption, while deficits in either dimension hinder it. For India's next phase of digital transformation, policymakers, NGOs, and tech companies must coordinate on both fronts – wiring and equipping rural classrooms, and empowering teachers – to ensure that Cloud-based AI tools can fulfill their promise of equitable quality education.

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