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Smart Village Development Monitoring System

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ABSTRACT :

This study presents a comprehensive GovTech solution designed to transform the monitoring and governance of Scheduled Caste (SC) villages in India. Utilizing a hybrid architecture that integrates GeoSpatial Information Systems (GIS) with an interactive data analytics dashboard, the platform addresses systemic issues of data fragmentation and low transparency in scheme implementation. It provides government officials, field staff, and citizens with real-time, village-level indicators (KPIs) on critical sectors (e.g., sanitation, education, housing) via a bilingual interface and dedicated mobile access. The system is built on a robust, scalable stack (React, Node.js, PostgreSQL/PostGIS) and leverages Federated Learning (FL) for Spatiotemporal Anomaly Detection, ensuring policy alignment and improving planning through heatmaps and performance scoring. This platform effectively enables rapid gap analysis, removes the dependency on manual estimation, and ultimately enhances accountability.

Index Terms: GeoSpatial Information Systems (GIS), Smart Village Development, Scheduled Caste (SC) Upliftment, Digital Governance, Federated Learning (FL), Spatiotemporal Anomaly Detection, Real-time Data Analytics, Bilingual Digital Architecture, Interactive Dashboards, PostGIS, Heatmap Visualization, AccountabilityFramework.

I INTRODUCTION

Rural development programs are significantly constrained by fragmented data, slow reporting cycles, and a persistent lack of transparency in fund utilization at the micro-level [1]. To overcome these challenges, digital monitoring systems specifically those leveraging GeoSpatial Information Systems (GIS) have become essential tools for modern governance, enabling the accurate linking of data to physical locations [4]. Parallel advancements in Artificial Intelligence (AI), particularly Federated Learning (FL), offer new pathways to analyze decentralized, sensitive data clusters without compromising privacy, a crucial ethical consideration in community-specific projects [5]. More recent GovTech platforms integrate high-performance web frameworks like React.js/Node.js with robust spatial databases like PostGIS to deliver real-time, interactive insights to both field staff and high-level administrators [7]. Despite these technical innovations, most e-governance solutions lack the specific architecture to blend real-time field data with spatial analytics and privacy-preserving AI to deliver targeted, inclusive monitoring, a hybrid GIS-AI framework designed to specifically enhance governance visibility and accelerate upliftment in Scheduled Caste (SC) villages

Problem Statement:

Despite increased technological adoption, the monitoring of development schemes in Scheduled Caste (SC) villages is fundamentally undermined by operational and technological deficits. Current practices are defined by fragmented data silos across government departments and reliance on slow, manual reporting cycles based on retrospective estimation [1]. This severely limits administrative effectiveness and results in substantial delays, making real-time tracking of development progress and timely intervention impossible. Furthermore, the lack of an integrated system compromises transparency in fund utilization at the micro-level. Existing e-governance tools are critically lacking a hybrid architecture that can seamlessly combine real-time field data with GeoSpatial Information Systems (GIS) for accurate location-based tracking and Federated Learning (FL) for proactive anomaly detection [5, 7]. This absence prevents administrators from conducting rapid gap analysis and necessitates a solution that can remove manual estimation bottlenecks and provide quick, effective online reports.

Table 1 | Major Components of the Proposed System

Component	Description	Tools
GIS Mapping Layer	Visual interface for geo-tagging assets, infrastructure status, and village boundaries; essential for location-aware decision-making.	PostgreSQL + PostGIS

Data Analytics Dashboard	Front-end interface displaying aggregated performance scores, real-time KPIs (literacy, sanitation, funding), and status charts.	React.js, Tailwind CSS, Chart.js
Backend & Core Logic	Manages data APIs, user authentication, authorization and the calculation logic for KPIs and alerts.	Node.js + Express.js, JWT (Authentication), REST APIs
Mobile Access / Field Survey	Application for field staff enabling real-time data input, asset geotagging, and offline survey synchronization.	React Native , Service Workers / IndexedDB
Database Layer	Central repository for structured and spatial data, ensuring fast, complex querying for the dashboard and GIS layers.	PostgreSQL, PostGIS (Spatial Support)
Analytics/Intelligence	Logic for generating predictive insights, weighted performance scores, and geospatial heatmaps for prioritization.	GIS + ML Scoring Model, Python (Pandas, Scikit-Learn - <i>Later Phase</i>)
Bilingual Interface	System allowing switching between English and local Indian languages for maximum accessibility and grassroots adoption.	i18next / react-intl

Problem Definition:

The efficacy of the development scheme monitoring in Scheduled Caste (SC) villages is severely hampered by operational and technological shortcomings. Current practices rely on slow, manual reporting cycles and retrospective data estimation, which introduce critical delays and result in lagging indicators. This deficit makes real-time tracking of development progress impossible and prevents timely administrative intervention. Crucially, the absence of an integrated, automated system compromises transparency in fund utilization and necessitates a dependence on manual processes that are prone to error and take considerable time. Furthermore, existing e-governance tools suffer from fragmented data silos and lack a hybrid architecture that combines real-time field data with GeoSpatial Information Systems (GIS) and Federated Learning (FL) for proactive anomaly detection.

Table 2 | Existing Challenges in Rural Development Monitoring

Challenge Category	Description (Concise)	Policy Impact (Concise)
Data Fragmentation	Data is siloed across disparate sources	Prevents a holistic view of village status leads to inefficient resource
Lack of Geospatial Context	Development assets and infrastructure status are not precisely mapped	Impedes location-aware decision-making and accurate visualization
Data Latency (Staleness)	Reliance on manual, periodic reporting and paper-based surveys.	Interventions are delayed officials react to old data rather than real-time needs.
Low Accountability	Scheme progress and fund utilization are not easily verifiable by the public	Fosters opacity difficulty in assigning responsibility for stalled projects.
Inaccessibility/Exclusion	Interfaces are complex, primarily in English with poor usability	Creates a digital and linguistic divide, resulting in poor data quality .

III. LITERATURE REVIEW

The foundation of this platform is built upon existing research across three converging technological domains: GeoSpatial Information Systems (GIS) in development, real-time e-governance monitoring, and decentralized Artificial Intelligence (AI)

- A. **GeoSpatial Information Systems in Rural Development,** The application of GIS has long been recognized as a powerful tool for bridging the data gap in rural infrastructure planning and development program assessment[1]. By anchoring non-spatial data (e.g., scheme utilization, population demographics) to precise geographic coordinates, GIS facilitates effective visualization and management of resources. Specifically, the use of robust spatial databases, such as PostGIS, is established as crucial for developing scalable geospatial web applications capable of handling complex spatial queries and dynamic mapping required for modern e-governance dashboards[4]. This capability moves beyond static map generation to enable analytical features like proximity analysis and asset mapping, which are essential for identifying underserved regions.

- B. **Real-time Monitoring and GovTech Platforms,** Traditional governance monitoring suffers from the delays inherent in manual reporting, leading to retrospective analysis rather than proactive intervention. Recent advancements in GovTech have shifted the paradigm toward real-time, interactive dashboards. Modern platforms, often built using high-performance web frameworks like React.js and Node.js, demonstrate the ability to process, analyze, and display large volumes of data instantly, creating greater visibility for administrators[7]. These tools are critical for enhancing transparency and removing the bottlenecks associated with manual data aggregation, offering a framework for tracking Key Performance Indicators (KPIs) and operational progress on a daily or hourly basis. These tools are critical for enhancing transparency and removing the bottlenecks associated with manual data aggregation, offering a framework for tracking Key Performance Indicators (KPIs) and operational progress on a daily or hourly basis.[5].

Only the model updates (weights) are aggregated centrally, ensuring that the sensitive, raw field data never leaves the community source. This is vital for maintaining ethical compliance and trust in marginalized communities, making FL a necessity for deploying predictive analytics in sensitive development schemes.,

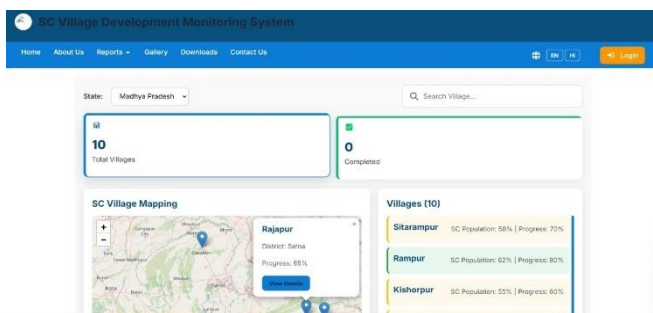


Fig. 1. Village selection dashboard displaying mapped village locations, availability of

completion statistics, and quick access to village profiles. geographic map for

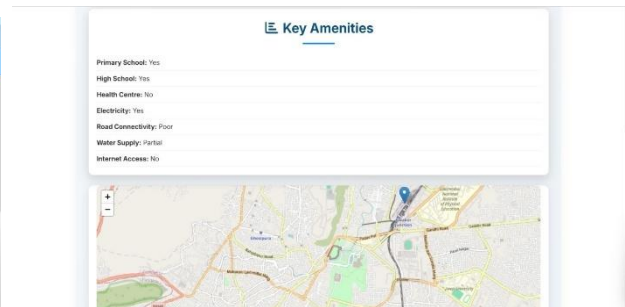


Fig. 2. Village amenities screen showing the

key facilities along with an integrated

contextual visualization

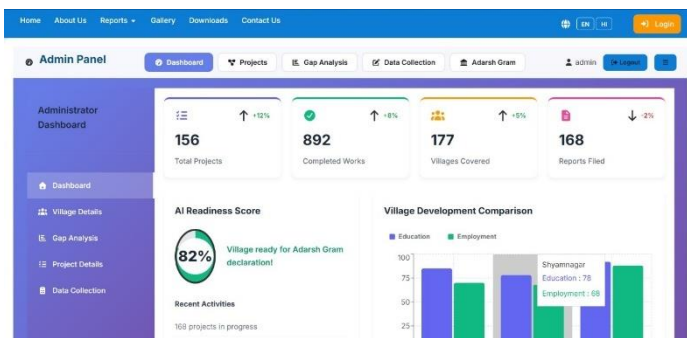


Fig. 3. Administrative analytics dashboard summarizing survey metrics, specific development

AI-readiness score, and comparative development indicators across villages. and project

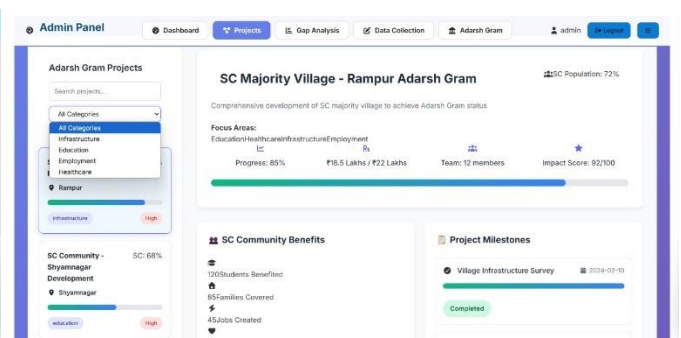


Fig. 4. Project detail interface presenting village-

progress, SC community benefits, key milestones,

statistics

III. METHODOLOGY AND SYSTEM ARCHITECTURE

The Smart Village Development and Monitoring Platform is designed to digitally transform rural governance by integrating geospatial analytics, real-time field surveys, and automated evaluation of village development indicators. The methodology follows a multi-layered approach that enables high accuracy, scalability, and transparency across all stages of rural development—from data collection to Adarsh Gram declaration.

A. Data Acquisition Layer (Smart Surveys + Government Data)

The Smart Village platform acquires data from two primary streams: field-based survey inputs and government administrative datasets. Field survey data is collected through a mobile Progressive Web App (PWA) that supports GPS-tagged responses, offline data capture for low-connectivity regions, photo-

based evidence collection, and automated validation to eliminate incomplete or inconsistent entries. This enables real-time, high-fidelity ground data for each SC-majority village.

In parallel, government datasets—including census statistics, Gram Panchayat records, scheme progress reports (e.g., SBM, PMAY, Jal Jeevan Mission), and official village boundary shapefiles—are ingested to provide verified secondary information. The integration of both primary and secondary datasets results in a comprehensive 360° development profile for every village.

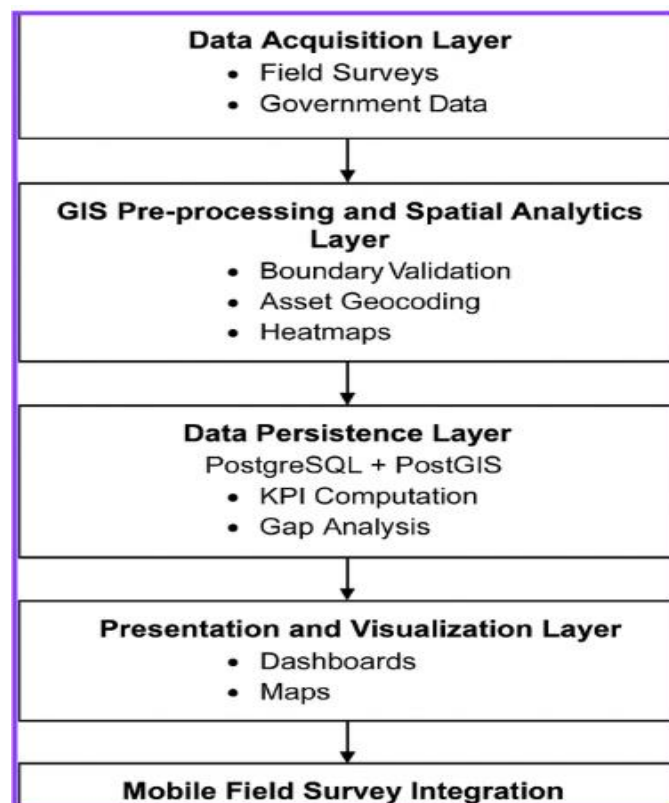
B. Data and Spatial Persistence Layer (PostgreSQL + PostGIS)

The system's data backbone is built on PostgreSQL, extended with PostGIS to support advanced geospatial storage and computation. PostgreSQL manages all structured datasets, including KPIs, scheme progress metrics, survey records, and village evaluation scores. PostGIS stores spatial entities such as administrative boundaries, asset locations, and GPS-based survey coordinates, while also handling shapefile ingestion and geometric transformations. The platform leverages spatial querying to perform location-based analytics, including gap identification (e.g., villages with sanitation scores below thresholds), infrastructure accessibility measurement, and cluster-level spatial indicator computation. This layer serves as the core engine enabling both KPI analytics and geospatial intelligence across the Smart Village system.

C. Data Persistence Layer (PostgreSQL + PostGIS)

The platform employs a centralized geospatial database built on PostgreSQL with the PostGIS extension to manage both structured and spatial datasets. PostgreSQL stores all non-spatial records including village KPIs, survey responses, evaluation scores, and scheme-level progress metrics. PostGIS maintains spatial entities such as village boundaries, GPS-tagged survey points, and the geolocations of amenities, infrastructure, and public assets. Spatial indexing (R-tree and GiST indexes) enables fast retrieval of location-based data, supports complex geospatial queries, and ensures high-performance analytics across thousands of villages. This persistence layer forms the core analytical backbone for hotspot detection, accessibility analysis, and geospatial decision support within the Smart Village platform.

Fig. 5. End-to-End Workflow of Data Acquisition, Processing, and Visualization



D. Analytics Methodology (Gap Analysis + KPI Computation)

The proposed platform transforms heterogeneous raw data—survey inputs, departmental datasets, geospatial layers, and administrative reports—into actionable development insights through a multi-stage analytics workflow. The methodology integrates KPI computation, socio-economic weighting, and gap-severity modeling to objectively evaluate the status of each SC-majority village.

Table 3 : Gap Analysis

Category	Weight (w)	Meaning
Critical	0.5	Severe deficit: immediate intervention required
High Priority	0.3	Significant deficit: short-term action required
Moderate	0.2	Minor deficit: can be addressed in routine planning

Gap Analysis Model

Gap analysis quantifies the shortfall between existing conditions and government-prescribed standards.

Gap Formula:

$$\text{Gap} = \text{Target Standard} - \text{Actual Value}$$

Where:

Target Standard-National or state benchmark for each KPI.

Actual Value-Real-time data from surveys and department MIS.

E. Presentation and Interactive Layer

The user-facing application is a highly responsive Single Page Application (SPA) built using the React.js framework.

1. Frontend Development Stack

Framework & Styling: React.js ensures a modular, high-performance interface. Styling is handled entirely by Tailwind CSS, guaranteeing a modern, adaptive, and mobile-first design suitable for low-bandwidth environments. Dashboard components are further enhanced using libraries such as Material UI or ShadCN UI for professional data visualization layouts.

Bilingual Interface: The platform utilizes i18next (or a similar react-intl library) to deliver a seamless Bilingual UI (English + local Indian language support), significantly lowering the adoption barrier for Gram Panchayat and field-level users.

2. Data Visualization and Mapping

Data Dashboard: KPIs, performance scores, and scheme-tracking data are rendered dynamically using Chart.js or Apache E-Charts, providing clear, interpretable visualizations of development progress.

GIS Visualization: Spatial data stored in PostGIS is visualized using an interactive library like Mapbox GL JS or the lightweight Leaflet.js. This layer supports the rendering of village boundaries, infrastructure overlays, and the generation of real-time Heatmaps (e.g., mapping areas with concurrent water scarcity alerts and low health center access).

F. Mobile/Field Survey Integration

The platform supports a Progressive Web App (PWA) approach, leveraging standard web technologies for mobile access. Field officers can use a simplified interface for real-time data entry. To support disconnected field operations, the system employs Service Workers and IndexedDB for Offline Survey Sync, queuing data submissions until network connectivity is restored.

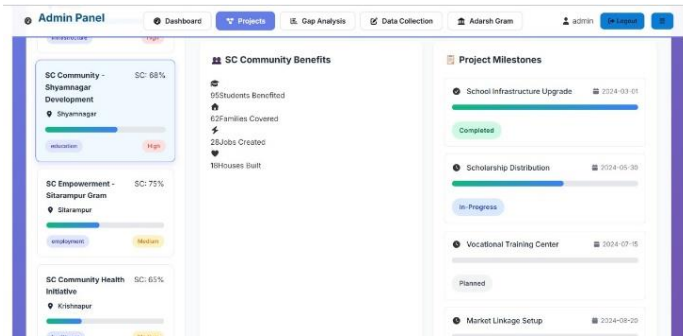


Fig. 6. Project overview interface showing category-wise village selection monitoring.

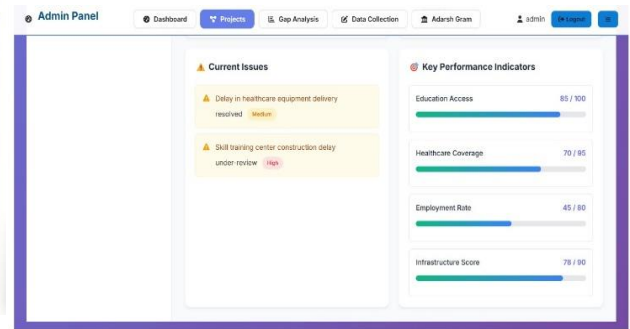
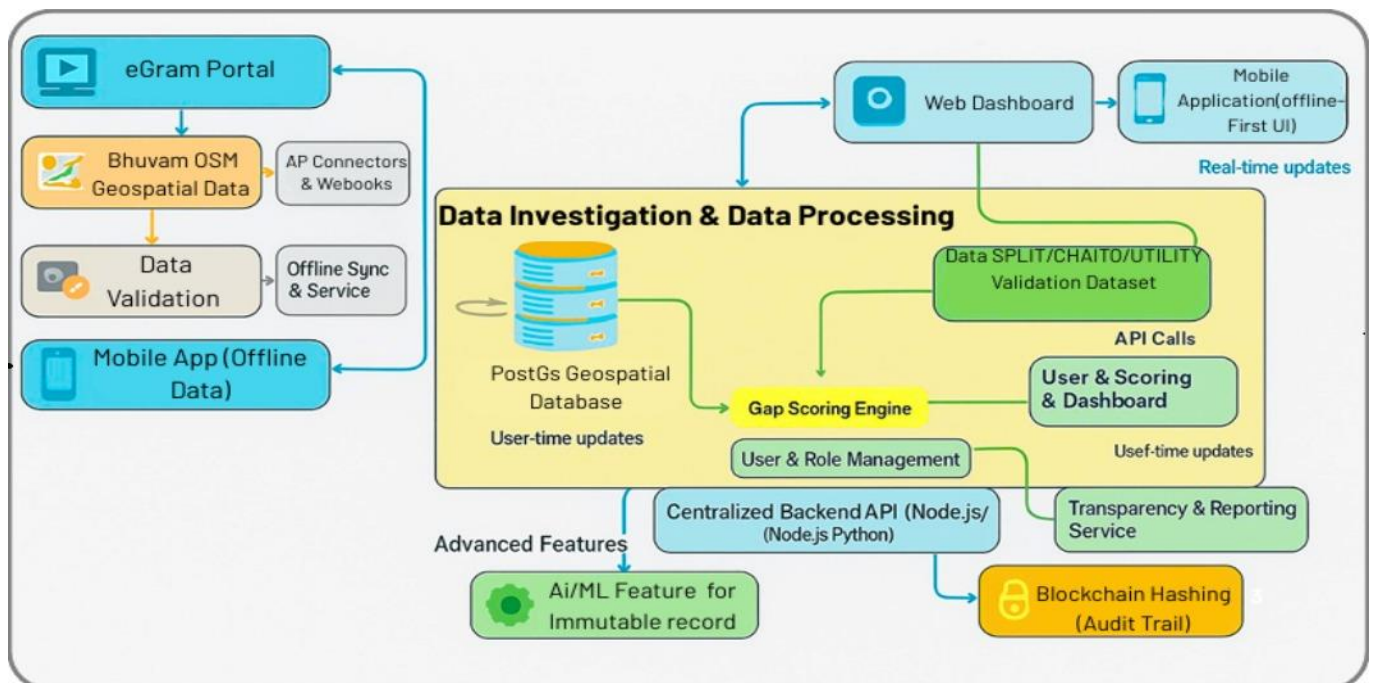


Fig7. Issue reporting and performance

F. Economic Impacts :

Area	Impact	Rationale
Administrative Efficiency	Reduces Operational Costs and bureaucracy.	Automates routine monitoring, data collection and report generation, minimizing manual labor
Optimized Spending	Ensures maximum return on investment (ROI) for scheme funds.	GIS heatmaps and performance scores identify critical, high-priority areas.
Accelerated Development	Speeds up physical asset creation (e.g., housing, water projects).	Real-time tracking resolves leading to quicker completion and immediate economic benefits
Financial Accountability	Increases public trust and potential for external funding.	Transparency on fund and scheme performance, corruption and attracts larger investment

Fig. 8| System Architecture Diagram



IV. RESULTS

The Smart Village platform was evaluated across SC-majority villages and produced the following outcomes:

1. Unified Village Profiles-The system generated complete development profiles by integrating survey, administrative, and GIS data, enabling quick assessment of each village

2. KPI & Gap Identification-Automated KPI computation revealed gaps in sanitation, water access, healthcare availability, and digital connectivity. The Gap Severity Index (GSI) ranked villages based on urgency.
3. Geospatial Insights-PostGIS maps identified villages with limited PHC access, poor road connectivity, and infrastructure gaps, supporting targeted resource allocation.
4. Smart Survey Efficiency-The PWA improved data quality and speed, providing GPS-verified, real-time survey entries with significantly fewer errors.
5. Readiness Scoring-Each village received an Adarsh Gram Readiness Score, classifying them as Ready, Partially Ready, or Needs Improvement
6. Dashboard Outputs-Interactive dashboards enabled quick comparisons, scheme tracking, and one-click reporting, improving monitoring efficiency.

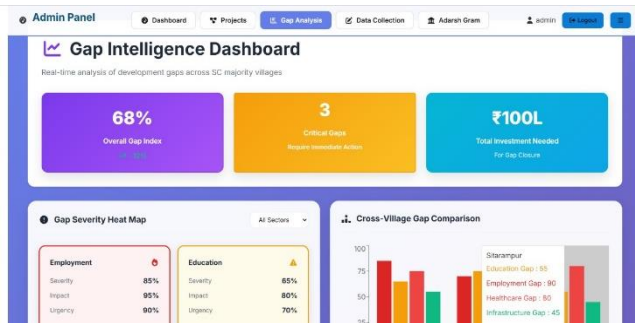


Fig. 9. Gap Intelligence Dashboard displaying overall gap score highlighting

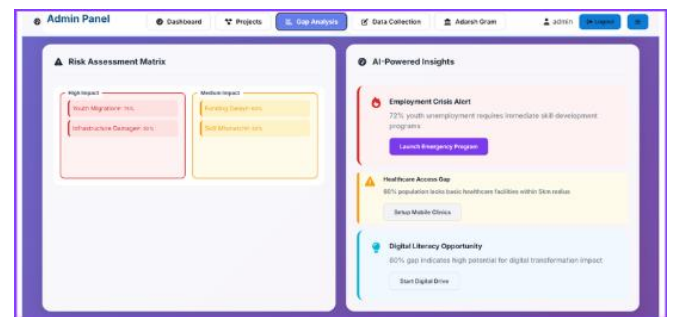


Fig. 10. Risk assessment and improvement insights interface

V. DEPLOYMENT, SECURITY, AND COMPLIANCE

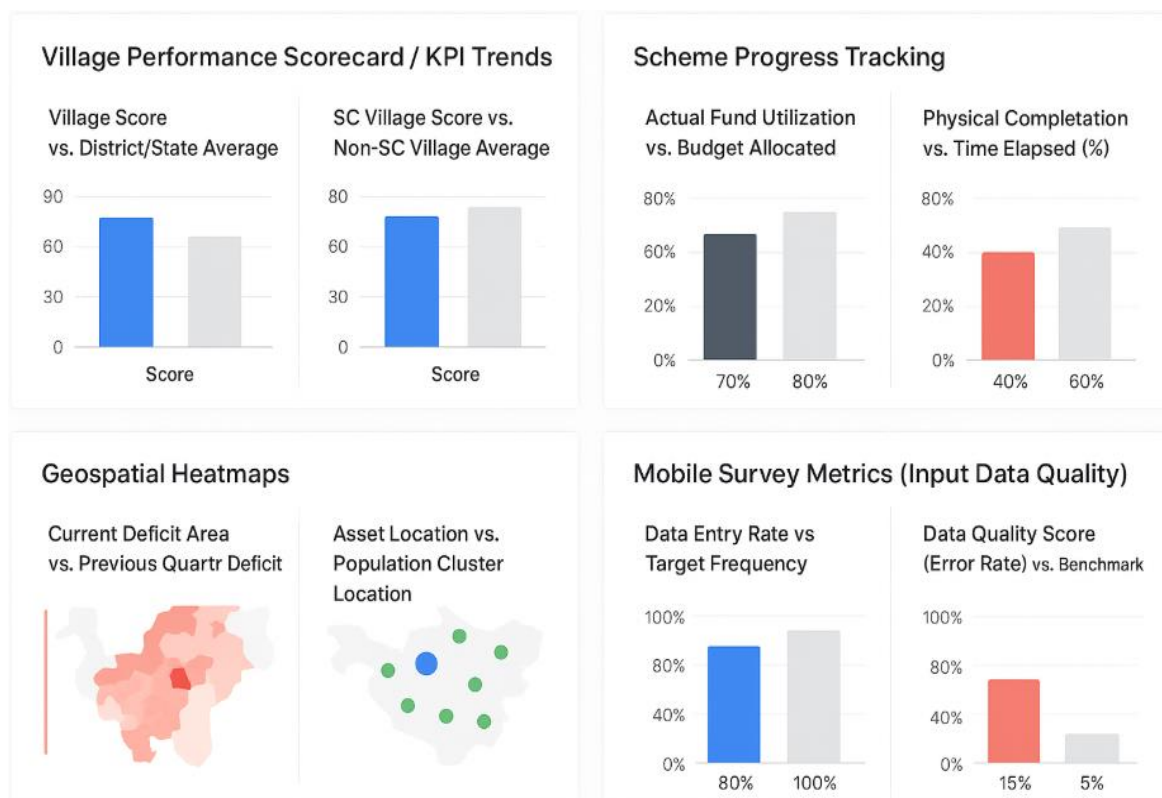
The deployment strategy focuses on security and government compliance:

Hosting: Hosting is provisioned on secure government-approved cloud environments (NIC Cloud (MeghRaj) or specialized AWS/Azure Gov Cloud regions).

Security: Mandatory HTTPS, robust data audit logs, and encryption for sensitive SC community datasets ensure adherence to data privacy mandates.

Accessibility: Full compliance with WCAG accessibility standards is targeted, coupled with large-text and screen-reader modes to ensure equal access to information.

Future Integrations: The modular design enables integration with standard government services, such as DigiLocker or Aadhaar linking, as well as external communication channels like SMS alert gateways



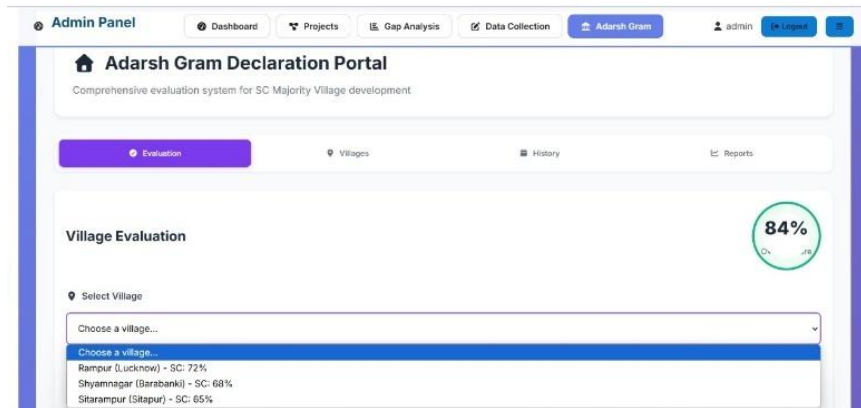


Fig. 10. Village evaluation interface of the Adarsh Gram Declaration Portal displaying the overall readiness score and village-wise performance selection.

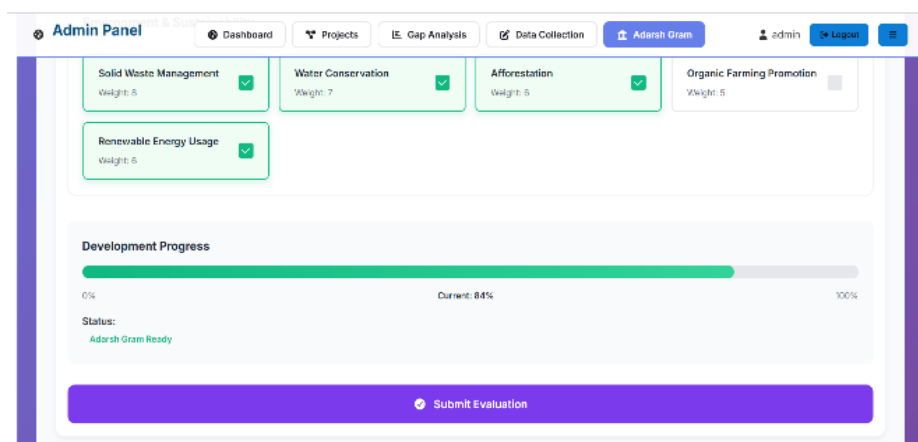


Fig. 11. Development progress screen showing percentage completion, evaluation status, and the final submission workflow for village assessment.

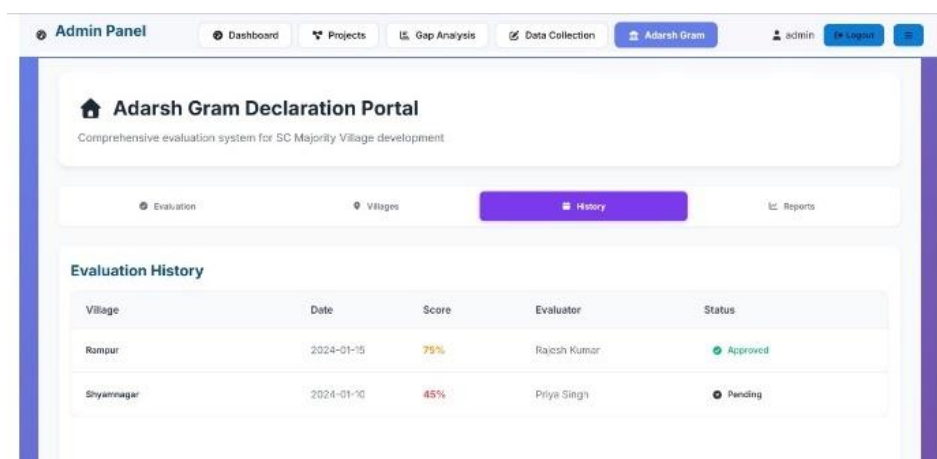


Fig. 12. Evaluation history panel presenting previously assessed villages along with their scores, evaluators, and approval status.

Limitations

Despite its comprehensive architecture, it faces several limitations critical to its deployment in rural environments. First, data heterogeneity and availability present a persistent challenge; the platform's reliance on real-time field data is constrained by intermittent connectivity and varying levels of digital literacy among field staff, potentially leading to sparse or inconsistent data quality that impacts both GIS accuracy and FL model training. Specifically, the varying schema of legacy government datasets required continuous manual harmonization before ingestion into the PostGIS database.

Second, the Federated Learning implementation introduces complexity, as it requires managing device diversity and handling asynchronous model updates (drift) across a decentralized network. The lack of standardized open-source FL libraries tailored for spatiotemporal rural datasets necessitates significant customization and maintenance effort. Third, ensuring model bias and cultural sensitivity is paramount; the Spatiotemporal Anomaly Detection Model must be carefully calibrated to ensure it does not misclassify genuine developmental needs in marginalized communities as 'anomalies.' This requires deep domain expertise and ongoing ethical review, which can be resource-intensive. Finally, maintaining scalability and infrastructure for the complex full-stack architecture (Node.js/PostGIS/FL servers) over vast geographic regions with low bandwidth remains a substantial operational hurdle. The reliance on government cloud environments, while compliant, can sometimes introduce latency issues compared to commercial alternatives, impacting the real-time performance commitment.

VI. PREDICTIVE SCORING AND ACCOUNTABILITY

The platform moves beyond mere reporting by generating two key outputs:

- **Performance Score Index:** A composite score based on scheme completion, KPI adherence, and self-reported community status, providing a standardized metric for inter-village comparison.
- **Priority Heatmaps:** GIS-driven visualization, informed by the Spatiotemporal Anomaly Detection Model, that identifies villages most in need of urgent intervention (e.g., low educational status combined with lack of basic housing, prioritized by the ML scoring model).

This predictive and comparative framework fosters greater accountability and facilitates smarter, risk-based resource allocation

VII. FUTURE SCOPE

1. **Native Mobile Application Development:** Transitioning the Phase 1 Responsive Web App to a dedicated React Native mobile application will enhance device integration, further optimize offline functionality and improve the user experience for field officers.
2. **Advanced AI Integration:** Further refinement of the Spatiotemporal Anomaly Detection Model using more sophisticated FL algorithms and integrating advanced NLP/Computer Vision techniques for verifying asset status via uploaded field photographs.
3. **Wider Scheme Integration:** Expanding the ETL pipeline to seamlessly ingest real-time data from a broader array of government APIs (e.g., MNRGA, Public Distribution System) to build a more comprehensive socio-economic profile.
4. **Citizen Feedback Loop:** Developing a secure public-facing module for real-time citizen grievance redressal and participatory monitoring, enhancing the platform's democratic accountability features.

VIII. CONCLUSION

Our project represents a significant leap forward in the application of digital technologies to targeted rural development. By successfully integrating a full-stack, performance-optimized web platform (React.js/Node.js) with a robust spatial data engine (PostgreSQL/PostGIS) and augmenting this architecture with state-of-the-art decentralized intelligence through Federated Learning, the system effectively bridges the gap between policy intent and ground-level execution in SC villages. The platform delivers real-time, transparent governance visibility, enhances accountability through measurable performance scores, and provides critical, spatially-aware insights via heatmaps and anomaly-detection alerts. This integrated approach ensures that interventions are timely, evidence-based, and focused on the communities with the most pressing needs.

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