



Leveraging Integrated Master Data Management for Enhanced Healthcare Decision-Making

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Article Info:

DOI: 10.22399/ijcesen.5114

Received : 05 February 2026

Revised : 29 March 2026

Accepted : 01 April 2026

Keywords

Master Data Management,
Healthcare Integration,
Patient Identity Management,
Clinical Protocol Management,
Digital Healthcare Transformation

Abstract:

Healthcare organizations are experiencing a fundamental shift in data management through the implementation of Master Data Management (MDM) solutions. The integration of advanced technologies, including artificial intelligence and machine learning, has revolutionized how healthcare providers handle patient information, clinical protocols, and treatment strategies. From streamlining clinical trial processes to enhancing patient identification accuracy, MDM systems serve as the cornerstone for modern healthcare delivery. The adoption of standardized protocols and integration frameworks has significantly improved data exchange capabilities while maintaining security and compliance. Through the implementation of sophisticated matching algorithms and validation processes, healthcare organizations have achieved remarkable improvements in patient record management, care coordination, and treatment outcomes. The emergence of AI-driven platforms has particularly transformed oncology care, enabling precise mutation detection and personalized treatment approaches.

1. Introduction

Healthcare organizations are undergoing a transformative shift in their data management approaches, with Master Data Management (MDM) solutions emerging as a cornerstone technology for handling complex patient-centric data. According to recent market analysis, the global MDM market is experiencing unprecedented growth, with projections indicating an expansion from USD 16.71 billion in 2022 to USD 48.26 billion by 2032, demonstrating a robust CAGR of 11.2% during this period [1]. This remarkable growth trajectory underscores the healthcare sector's increasing reliance on sophisticated data management solutions, as organizations now manage an average of 8.41 petabytes of data per institution.

Modern healthcare infrastructure has evolved to incorporate multiple MDM components, creating a comprehensive data management ecosystem. Patient MDM systems now manage an average of 2.5 million unique patient records per mid-sized healthcare system, while Payor MDM coordinates with over 1,000 insurance providers and manages more than 150,000 unique billing codes. The Customer MDM component tracks relationships with healthcare providers and specialists, while

Patient Accession Master processes thousands of laboratory orders daily. Test Master systems maintain extensive records for diagnostic tests, contributing to a robust framework that has demonstrated significant improvements in healthcare delivery efficiency [1].

Recent research in healthcare data quality measurement has revealed remarkable improvements through integrated MDM implementations. A comprehensive multi-center study utilizing the Observational Medical Outcomes Partnership Common Data Model (OMOP CDM) demonstrated that healthcare organizations implementing integrated MDM systems achieved a 72.3% improvement in data accuracy and a 56.8% reduction in duplicate patient records. The study, which examined data quality across multiple healthcare centers, found that standardized MDM implementations led to a 41.5% increase in successful first-time insurance claim submissions and a 63.7% improvement in patient identification accuracy [2].

The impact of MDM systems extends beyond operational efficiency into crucial research initiatives. In the realm of genetic research and disease tracking, these systems now process genomic data from hundreds of thousands of patient

samples annually. The integration of MDM components has proven particularly valuable in communicable disease surveillance, enabling healthcare organizations to track and analyze millions of test results daily. This capability has become increasingly critical in the context of global health challenges, where rapid data processing and analysis can significantly impact response times and treatment outcomes [2].

2. The Challenge of Clinical Protocol Management

The management of clinical study protocols presents significant challenges in modern medical research, with industry perspectives highlighting critical inefficiencies in current practices. According to comprehensive industry analysis, clinical data management (CDM) systems face substantial obstacles in protocol handling, with traditional paper-based and electronic data capture (EDC) systems showing considerable limitations. Research indicates that approximately 80% of clinical trials continue to rely on hybrid data collection methods, combining paper-based and electronic systems, leading to increased complexity in data management and quality control processes [3].

The persistent reliance on hybrid systems has created significant operational challenges for research organizations. Data reconciliation between paper and electronic formats requires extensive manual intervention, increasing both time requirements and error potential. Industry experts have noted that this dual-system approach often results in data inconsistencies, with reconciliation processes consuming up to 35% of clinical data managers' time. Moreover, the maintenance of parallel systems has led to increased storage costs and complicated audit trails, making regulatory compliance more challenging and resource-intensive [3].

The evolution of clinical trial complexity has significantly impacted protocol management efficiency. A recent cross-sectional study examining multicenter clinical trial protocols revealed that only 61.7% of protocols adequately reported primary outcome measures, while just 56.7% properly documented sample size calculations. The study, which analyzed protocols across multiple research centers, found that merely 48.3% of protocols included clear descriptions of randomization procedures, highlighting significant gaps in protocol standardization and documentation [4]. These documentation gaps have far-reaching implications for trial validity and reproducibility. The inadequate reporting of primary outcome

measures can lead to selective outcome reporting and potential bias in trial results. Similarly, insufficient documentation of sample size calculations may compromise the statistical power of studies and their ability to detect meaningful clinical effects. The lack of standardization in randomization procedure documentation raises concerns about allocation concealment and potential selection bias [4].

The implications of these documentation challenges extend throughout the clinical research process. Industry analyses show that data entry and verification processes consume approximately 30% of clinical trial budgets, with error rates in manual data entry ranging from 2% to 8%. The implementation of risk-based quality management approaches has become crucial, as traditional 100% source data verification (SDV) approaches prove increasingly unsustainable in modern clinical trials. Organizations implementing risk-based monitoring strategies have reported cost reductions of up to 25% in data verification processes while maintaining data quality standards [3].

Recent assessment of multicenter clinical trial protocols has uncovered specific areas requiring immediate attention. Only 25% of protocols effectively detailed adverse event reporting procedures, while just 35% provided comprehensive recruitment strategies. Furthermore, the study identified that only 43.3% of protocols included adequate descriptions of statistical methods, and a mere 40% properly documented data monitoring procedures. These deficiencies in protocol documentation directly impact trial execution, regulatory compliance, and the ability to generate reliable evidence for clinical decision-making [4].

The standardization challenges in protocol development extend to international collaborative trials, where variations in regulatory requirements and reporting standards across jurisdictions add another layer of complexity. Cross-border trials often face additional challenges in maintaining consistent documentation practices while complying with multiple regulatory frameworks. The study found that protocols involving multiple countries showed even lower rates of complete documentation, with only 38.2% meeting all essential reporting requirements [4].

3. An Innovative Solution: Integrated Oncology Data Platform

3.1 Core Components

The evolution of integrated oncology platforms represents a significant advancement in healthcare

data management, with MDM serving as its foundational architecture. Research on integrated clinical and genomic data platforms has demonstrated remarkable capabilities in supporting translational research and precision medicine. The OncDRS platform implementation has shown successful integration of diverse data types, including clinical, genomic, and imaging data, while maintaining HIPAA compliance and data security. These platforms have demonstrated the ability to handle complex queries across multiple data domains, enabling researchers to identify cohorts based on specific molecular and clinical characteristics for targeted therapeutic approaches [5].

Modern healthcare systems leverage this integrated approach to process vast amounts of patient data across different modalities. The platform architecture supports real-time integration of clinical trial matching algorithms with genomic profiling results, enabling rapid identification of suitable clinical trials for patients. Studies have shown that such integrated systems can effectively manage comprehensive patient profiles, including detailed molecular characterization, treatment histories, and outcomes data, while maintaining data integrity and accessibility across multiple departments [5].

3.2 Technical Architecture

The implementation of health information technology through integrated platforms has demonstrated significant improvements in patient safety and care quality. Research indicates that electronic health record systems with integrated decision support capabilities can reduce medication errors by 50%, with computerized physician order entry systems showing a 13% to 86% decrease in medication errors across different healthcare settings. Moreover, bar code medication administration systems have resulted in a 54% to 87% reduction in medication administration errors [6].

The technical framework's impact extends beyond error reduction into comprehensive care improvement. Studies have shown that healthcare organizations implementing integrated health information technology solutions have experienced significant improvements in preventive care delivery, with vaccination rates increasing by 12% to 18% and appropriate testing adherence improving by 3% to 33%. Furthermore, the implementation of clinical decision support systems has led to improved adherence to guideline-based care, with studies reporting 12% to 72% improvements in clinical practice. The integration

of these systems has also resulted in more efficient resource utilization, with a 24% reduction in redundant laboratory tests and a 14% decrease in redundant radiology tests [6].

4. Data Consolidation Strategy

4.1 Systems Integration

The integration of laboratory information systems (LIS) represents a fundamental challenge in modern healthcare infrastructure, particularly as healthcare organizations move toward digital transformation. Recent systematic reviews of LIS integration technologies have identified several key architectural patterns and integration approaches that form the backbone of modern healthcare data management. Service-Oriented Architecture (SOA), Enterprise Service Bus (ESB), and Application Programming Interface (API) implementation have emerged as primary methodologies for achieving seamless data integration. The systematic review indicates that organizations implementing these architectures experience significant improvements in data exchange efficiency and system interoperability [7].

Integration frameworks utilizing HL7 standards, particularly FHIR (Fast Healthcare Interoperability Resources), have revolutionized healthcare data exchange capabilities. The review highlights that FHIR implementations demonstrate superior flexibility in handling diverse data types, from basic patient demographics to complex genomic information. Modern LIS implementations increasingly leverage cloud-based solutions and microservices architecture, enabling healthcare organizations to achieve unprecedented scalability and adaptability in their data management strategies. This architectural approach has proven particularly effective in managing the growing volume of healthcare data while maintaining system performance and reliability [7].

The systematic review further reveals that successful LIS implementations commonly incorporate multiple standardized protocols, including ASTM, DICOM, and XML-based messaging systems. These protocols form a comprehensive communication framework that enables seamless interaction between various laboratory instruments and information systems. The integration frameworks have demonstrated remarkable effectiveness in managing diverse data types, with particular success in handling both routine clinical chemistry results and complex genomic data. The review emphasizes that modern integration approaches must balance the need for efficient data exchange with stringent security

requirements and regulatory compliance standards [7].

4.2 Patient Data Management

The management of patient identity across diverse data sources presents unique challenges in healthcare systems, requiring sophisticated solutions for accurate patient matching and record maintenance. Research indicates that patient matching errors occur in approximately 8% of cases across institutions, with this rate doubling to 16% when matching records between different organizations. These error rates highlight the critical importance of implementing robust patient identity management systems, particularly in increasingly complex healthcare networks [8].

Healthcare organizations have developed sophisticated approaches to patient identity management, incorporating advanced matching algorithms and standardized data formats. The implementation of probabilistic matching algorithms has shown superior accuracy compared to traditional deterministic matching methods. Studies demonstrate that the use of standardized demographic data elements, combined with robust validation processes, can reduce duplicate record creation by up to 24%. This improvement is particularly significant in large healthcare networks where accurate patient identification is crucial for care coordination [8].

Modern patient identity management systems have evolved to incorporate multiple layers of validation and verification. Healthcare organizations implementing comprehensive patient identity management strategies report significant improvements in their ability to maintain longitudinal patient records, with some systems achieving match rates exceeding 95% when utilizing enhanced demographic data sets and standardized matching algorithms [14]. These systems typically employ a combination of demographic elements, including name variants, date of birth, gender, address components, and additional identifiers to ensure accurate patient matching across different care settings [8].

The effectiveness of patient identity management systems is further enhanced through the implementation of standardized data quality processes. Organizations that maintain rigorous data quality standards and regular validation procedures report higher success rates in patient matching and record maintenance. The research emphasizes the importance of ongoing system monitoring and performance assessment to ensure consistent accuracy in patient identification and

record matching across diverse healthcare settings [8].

5. Future Impact on Clinical Research

The integration of artificial intelligence (AI) and big data analytics in oncology has fundamentally transformed precision medicine approaches. Deep learning algorithms have achieved unprecedented accuracy rates of up to 95% in identifying cancer-specific mutations, revolutionizing how clinicians approach cancer diagnosis and treatment. Machine learning algorithms integrated into clinical decision support systems enable real-time analysis of complex molecular patterns, facilitating rapid patient stratification and treatment selection. These AI systems have proven particularly valuable in rare cancer research, demonstrating exceptional capabilities in identifying novel biomarkers and potential therapeutic targets. The implementation of these advanced systems has reduced diagnostic timeframes from weeks to hours, enabling faster treatment initiation and improved patient outcomes [9].

The scope of big data analytics extends well beyond individual patient care to population-level insights and predictive modeling. AI-powered platforms can simultaneously process and analyze data from multiple sources, including electronic health records, genomic databases, and clinical trial repositories, creating comprehensive patient profiles that were previously impossible to achieve. These integrated systems have successfully identified new cancer subtypes and molecular patterns, leading to more targeted therapeutic strategies. The adoption of AI-driven precision oncology has revolutionized treatment planning, with predictive models achieving accuracy rates comparable to expert oncologists while processing information at significantly faster rates. Furthermore, these systems have demonstrated the ability to predict treatment responses with higher precision, enabling clinicians to adjust therapeutic approaches proactively rather than reactively [9].

Digital transformation in healthcare settings has yielded substantial improvements in both service delivery and patient outcomes. Healthcare organizations implementing comprehensive digital solutions have reported EHR adoption rates reaching 96%, marking a significant milestone in healthcare digitization. This high adoption rate has facilitated enhanced inter-provider communication and reduced medical errors through improved information sharing and standardization. The implementation of digital platforms has also enabled real-time access to patient information across different healthcare settings, leading to more

coordinated care delivery and improved patient safety protocols [10].

The success of digital technology implementation in healthcare settings varies based on several critical factors, including user acceptance and system design. Healthcare professionals' acceptance of new technologies is strongly influenced by perceived usefulness (PU) and perceived ease of use (PEOU), with correlation coefficients of 0.523 and 0.498 respectively. These metrics highlight the importance of developing user-friendly systems that demonstrate clear clinical value. Organizations implementing comprehensive digital transformation strategies have reported patient satisfaction improvements of up to 27%, attributed to faster service delivery, reduced wait times, and more personalized care experiences. Additionally, digital platforms have enabled better patient engagement through improved access to health information and treatment plans, leading to higher treatment adherence rates and better health outcomes [10].

The future of clinical research is being shaped by the convergence of AI, big data analytics, and digital transformation. Healthcare organizations are increasingly adopting hybrid approaches that combine traditional clinical expertise with AI-driven insights. These integrated systems support more sophisticated trial design, patient recruitment, and outcome monitoring processes. The implementation of AI-powered predictive models has improved protocol adherence and reduced adverse events by identifying potential risks before they materialize. Furthermore, these advanced systems facilitate more efficient resource allocation and better strategic planning in clinical research settings [9].

The ongoing evolution of digital healthcare technologies continues to reveal new possibilities for improving patient care and research outcomes. Machine learning algorithms are becoming increasingly sophisticated in their ability to analyze complex medical data and identify subtle patterns that might escape human observation. The integration of these technologies with traditional healthcare practices is creating a more robust and efficient healthcare ecosystem, capable of delivering more personalized and effective patient care while advancing medical knowledge through improved research capabilities [10].

6. Limitations of the Study

While this study presents a comprehensive review of integrated Master Data Management strategies in healthcare, several limitations must be acknowledged to ensure appropriate interpretation of the findings.

The generalizability of the reported outcomes represents a primary concern. The performance improvements documented across MDM implementations are predominantly drawn from large, well-resourced healthcare systems, and the findings may not translate directly to smaller organizations, rural providers, or healthcare systems in low- and middle-income countries where infrastructure and workforce capacity differ substantially. The market projections and institutional data volumes cited [1] further reflect a landscape skewed toward mature digital environments, which may overstate the readiness of the broader healthcare ecosystem[11].

A second limitation concerns the heterogeneity in methodologies and definitions across the referenced studies. The wide variance in improvement ranges reported for health information technology outcomes [6] reflects differences in clinical environments, patient populations, and implementation approaches rather than a uniform effect. This inconsistency limits the comparability of results and makes it difficult to establish standardized benchmarks applicable across diverse healthcare settings.

The temporal currency of certain evidence also warrants attention. Some foundational studies informing the patient identity management and clinical data management discussions [3, 8] were conducted over a decade ago. The healthcare technology landscape has evolved considerably since then, with the introduction of cloud-native architectures and updated FHIR standards, meaning that some of the cited error rates and adoption barriers may no longer accurately reflect current practice.

As a narrative review, this study is further subject to publication bias. Research documenting positive MDM and digital health outcomes is more likely to be published and cited, while studies reporting implementation failures or neutral results may be underrepresented [10]. This could lead to an overestimation of the benefits associated with MDM adoption. Finally, the review does not comprehensively address the total cost of ownership associated with MDM implementations, nor does it sufficiently explore the organizational and change-management challenges that frequently accompany large-scale health IT deployments [6].

7. Future Research Directions

The findings of this review open several important avenues for future investigation. As MDM systems mature and AI becomes more deeply embedded in clinical workflows, research must evolve to address

emerging technical, ethical, and operational challenges[12].

One of the most pressing priorities is the development of ethical frameworks governing AI deployment within healthcare data management. Future studies should examine how algorithmic bias affects clinical decision-making across diverse patient populations, particularly in underserved communities. Research into explainable AI methodologies would help clinicians understand and audit model outputs, which is especially critical in high-stakes domains such as oncology where AI-driven platforms are already influencing treatment decisions [9]. Governance frameworks defining accountability for AI-generated clinical recommendations represent a significant gap requiring both empirical investigation and regulatory guidance.

Data privacy and consent management present another important research frontier. As MDM platforms consolidate genomic, clinical, and behavioral data at scale [2], future research should explore dynamic consent systems that give patients granular control over data use across research, clinical, and commercial contexts. Comparative analyses of existing regulatory frameworks, alongside investigations into privacy-preserving computation techniques such as federated learning and differential privacy, would support the

development of globally harmonized data governance standards[13].

Scalability remains an underexplored dimension of MDM research. While current evidence demonstrates effectiveness in large institutions [1, 5], there is a need to develop implementation models suited to smaller providers and resource-limited settings. Studies examining lightweight, FHIR-compliant architectures and the performance of patient matching algorithms [8] under real-world data load conditions would significantly advance the field.

Longitudinal evaluation of MDM outcomes is another critical gap. Most existing studies measure performance over short timeframes, and multi-year investigations would help assess sustained impact on patient outcomes, data quality, and research productivity using standardized models such as the OMOP CDM [2]. Research into human factors, workforce training, and change management should also be prioritized, given that technology acceptance remains a determinant of implementation success [10]. Finally, future work should examine how MDM architectures can be extended to incorporate patient-generated data from wearables and mobile health platforms, enabling more representative population health analyses and supporting real-world evidence generation alongside traditional clinical and laboratory data [7].

Table 1. Clinical Protocol Documentation Completeness Analysis [3, 4].

Protocol Management Aspect	Percentage (%)
Hybrid Data Collection Usage	80
Primary Outcome Documentation	61.7
Sample Size Calculation Documentation	56.7
Randomization Procedure Documentation	48.3
Clinical Trial Budget for Data Entry	30
Recruitment Strategy Documentation	35
Statistical Methods Documentation	43.3
Data Monitoring Documentation	40
Adverse Event Reporting Documentation	25
Manual Data Entry Error Rate (Upper Range)	8
Manual Data Entry Error Rate (Lower Range)	2

Table 2. Impact of Health Information Technology on Healthcare Metrics [5, 6].

Improvement Category	Minimum Improvement (%)	Maximum Improvement (%)
Medication Error Reduction (EHR)	50	50
Medication Error Reduction (CPOE)	13	86
Medication Administration Error Reduction	54	87
Vaccination Rate Increase	12	18
Testing Adherence Improvement	3	33

Guideline-Based Care Adherence	12	72
Redundant Laboratory Test Reduction	24	24
Redundant Radiology Test Reduction	14	14

Table 3. Healthcare Data Matching Accuracy Analysis [7, 8].

Metric Category	Performance Rate (%)
Single Institution Patient Matching Error Rate	8
Cross-Organization Patient Matching Error Rate	16
Duplicate Record Reduction Rate	24
Enhanced System Match Rate	95

Table 4. Healthcare Digital Transformation Success Indicators [9, 10].

Metric Category	Success Rate (%)
AI Mutation Detection Accuracy	95
EHR System Adoption Rate	96
Patient Satisfaction Improvement	27
Technology Perceived Usefulness (PU)	52.3
Technology Perceived Ease of Use (PEOU)	49.8

8. Conclusions

The evolution of integrated MDM strategies in healthcare represents a pivotal advancement in patient care delivery and medical innovation. By unifying diverse data sources and implementing sophisticated management systems, healthcare organizations have enhanced their ability to deliver personalized care, streamline operations, and advance medical discoveries. The convergence of traditional healthcare processes with modern digital solutions has created a robust framework for future healthcare delivery, particularly in complex fields such as oncology. The successful implementation of these systems demonstrates the transformative potential of integrated data management in shaping the future of healthcare delivery and patient outcomes.

Author Statements:

- **Ethical approval:** The conducted research is not related to either human or animal use.
- **Conflict of interest:** The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper
- **Acknowledgement:** The authors declare that they have nobody or no-company to acknowledge.
- **Author contributions:** The authors declare that they have equal right on this paper.

- **Funding information:** The authors declare that there is no funding to be acknowledged.
- **Data availability statement:** The data that support the findings of this study are available on request from the corresponding author. The data are not publicly available due to privacy or ethical restrictions.
- **Use of AI Tools:** The author(s) declare that no generative AI or AI-assisted technologies were used in the writing process of this manuscript.

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