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**RESEARCH ARTICLE**

## Agent-Based Diagnostic Assistants for Streamlining Primary Care Workflows

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**ABSTRACT**

Primary care faces increasing pressure due to aging populations, rising chronic diseases, and growing demand for efficient services. This article explores a distributed network of AI-powered diagnostic assistants to support primary care physicians in their daily workflows. These intelligent agents operate within ethical boundaries under physician oversight, revolutionizing patient interactions by gathering pre-appointment information, suggesting diagnostic pathways based on symptoms and medical history, and aiding in test result interpretation. The system shows potential to improve efficiency, mitigate physician burnout, and enable faster, more accurate initial diagnoses. Critical implementation considerations include ethical implications, seamless integration with existing Electronic Health Records, and maintaining the physician's role as ultimate decision-maker. This agent-based approach represents a promising evolution in healthcare delivery that preserves human medical expertise while leveraging technological capabilities to address growing demands on primary care systems.

**KEYWORDS**

Artificial Intelligence, Diagnostic Assistants, Electronic Health Records, Primary Care, Workflow Optimization

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### 1. Introduction

Primary care physicians (PCPs) serve as the first point of contact for individuals seeking medical attention, playing a crucial role in disease prevention, health maintenance, and the management of acute and chronic conditions. According to the Health Resources and Services Administration's (HRSA) 2024 report, primary care accounts for more than 58% of all outpatient physician office visits in the United States, with over 480 million primary care visits annually [1]. This high volume of patient engagement reinforces the central role PCPs play in healthcare delivery. Despite this critical function, the demands placed on PCPs continue to escalate, with the average PCP now managing a panel of approximately 2,500 patients—representing a substantial increase over the past decade that contributes significantly to systemic strain [1].

This growing burden has led to concerning outcomes in the primary care landscape. The HRSA report highlights that approximately 45% of PCPs report symptoms of burnout, while average consultation times have decreased to just 15-20 minutes per patient encounter [1]. The diagnostic process has become increasingly complex, with PCPs required to consider numerous potential diagnoses for common presenting symptoms—a situation that contributes to diagnostic delays in approximately 5-15% of primary care cases according to recent studies [2]. The increasing volume of patient data within electronic health records, which often contain hundreds of data points per patient, coupled with the exponential growth of medical knowledge, creates significant cognitive challenges even for the most experienced clinicians [2].

Artificial intelligence (AI) presents a transformative opportunity to address these challenges and enhance the efficiency and effectiveness of primary care. While AI has shown promise in various medical domains, a recent systematic review published in *Diagnostics* indicates that AI systems have achieved diagnostic accuracy rates ranging from 76% to 93% for specific conditions

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commonly encountered in primary care settings [2]. The integration of intelligent agents directly into the primary care workflow holds even greater potential. This article posits that a distributed network of agent-based diagnostic assistants can act as a valuable extension of the PCP's capabilities, streamlining routine tasks and providing timely insights to support clinical decision-making. Recent research suggests that such systems could potentially reduce documentation time by up to 30% and improve diagnostic accuracy through comprehensive analysis of patient data that might otherwise be overlooked due to time constraints [2].

## 2. The Promise of Agent-Based Diagnostic Assistants in Primary Care

Imagine a scenario where, prior to a patient's appointment, an AI agent securely interacts with the patient through a user-friendly interface. This interaction could involve gathering detailed information about their current symptoms, reviewing their medical history, and even collecting relevant lifestyle data. Recent research on AI-enhanced patient engagement systems demonstrates that digital pre-appointment interactions can increase the comprehensiveness of collected clinical information by up to 34%, resulting in more productive consultations [3]. This proactively collected information would then be synthesized and presented to the PCP in a concise and organized manner before the consultation even begins, potentially saving valuable time during the actual encounter. AI-powered diagnostic assistants can analyze large volumes of health data to identify patterns and predict potential health risks, enabling PCPs to move from reactive to preventive care approaches. [11].

Furthermore, based on the reported symptoms and the patient's medical history, the agent could suggest potential diagnostic pathways, drawing upon vast medical knowledge and evidence-based guidelines. According to a recent study examining intelligent clinical decision support systems in primary care settings, AI-augmented diagnostic tools have demonstrated the ability to consider up to 29 differential diagnoses for complex presentations, compared to an average of 6-8 diagnoses typically considered by physicians working under time constraints [3]. This would not be a replacement for the physician's expertise but rather a tool to help them consider a broader range of possibilities and potentially expedite the diagnostic process. Studies indicate that collaborative human-AI diagnostic approaches can reduce the time required for comprehensive differential diagnosis formation by approximately 25.7% [4]. Beyond diagnostic support, AI agents can enhance medication management by identifying potential drug interactions, suggesting optimal dosing schedules, and monitoring patient adherence to treatment regimens. [12]. In the realm of initial test results, these agents could be trained to identify patterns and flag potentially concerning findings for the physician's review. For instance, an agent could highlight slightly elevated blood pressure readings or subtle anomalies in routine blood tests, prompting the PCP to investigate further. Current research into AI-based anomaly detection in clinical laboratory results shows promising results, with sensitivity rates reaching 87.3% and specificity of 82.6% for subtle deviations that might otherwise be overlooked in busy clinical environments [4].

The key advantages of such an agent-based system are multifaceted and supported by emerging evidence. Enhanced efficiency represents a significant benefit, with studies indicating that by automating pre-appointment data collection and providing preliminary diagnostic insights, intelligent systems can reduce administrative documentation workload by up to 40%, allowing physicians to dedicate more time to direct patient care [3]. Regarding physician burnout, implementation of AI-assisted workflow tools in pilot programs has been associated with measurable improvements in workplace satisfaction. A survey of PCPs utilizing such systems reported a 23% reduction in self-reported stress levels related to information management tasks, suggesting potential benefits for addressing the burnout epidemic affecting an estimated 42% of primary care providers [4].

The potential for faster and more accurate initial diagnoses represents another significant advantage. By providing timely suggestions and highlighting potential areas of concern, these agents could contribute to earlier and more accurate initial diagnoses, particularly for common ailments where pattern recognition plays a significant role. Evaluations of machine learning models for diagnostic support in primary care settings have demonstrated accuracy rates of 76-92% for common presentations, with particularly strong performance in identifying patterns suggestive of chronic conditions like diabetes, hypertension, and respiratory disorders [3]. Patient engagement also stands to benefit substantially, with research demonstrating that integrated digital interaction platforms lead to a 31% improvement in patient-reported preparedness for consultations and increased satisfaction with care experiences as measured by standardized patient experience metrics [4].

Metric	Traditional Approach	AI-Enhanced Approach
Differential diagnoses considered	6-8 diagnoses	29 diagnoses
Diagnostic time reduction	Baseline	25.7%
Clinical information comprehensiveness	Baseline	34% increase

Administrative workload reduction	Baseline	40% decrease
Physician stress reduction	Baseline	23% decrease
Patient preparedness improvement	Baseline	31% increase

Table 1. Efficiency and Accuracy Metrics of AI Diagnostic Assistants [3, 4]

**3. Streamlining Primary Care Workflows: A Distributed Network Approach**

The proposed system envisions a distributed network of specialized agents, each potentially focusing on specific aspects of the primary care workflow. This architectural approach leverages the concept of distributed intelligence, which has demonstrated significant advantages over monolithic AI systems in healthcare applications. Research indicates that modular, multi-agent frameworks can achieve 27.5% higher accuracy and 38.4% greater adaptability when processing complex healthcare data compared to single-agent approaches [3].

Pre-appointment information gathering agents would interact with patients to collect symptom details, medical history updates, and lifestyle information. Evaluation of such systems in real-world clinical environments shows they can reduce information collection time by approximately 11.3 minutes per patient while improving the completeness of gathered clinical data by 42% compared to traditional methods [4]. Differential diagnosis suggestion agents would analyze patient data and suggest potential diagnoses based on evidence-based guidelines and medical literature. Studies evaluating these capabilities have shown that natural language processing techniques can accurately extract relevant clinical concepts from patient-reported symptoms with precision rates of 83.7% and recall rates of 79.5% [3]. The distributed architecture of specialized AI agents mirrors the biological neural networks of human cognition, allowing for both specialized processing and holistic integration of information to support clinical decision-making. [13].

Initial test result interpretation agents would focus on identifying patterns and anomalies in routine test results for physician review. Current implementations of machine learning for laboratory result analysis have demonstrated the ability to flag clinically significant findings with an average reduction in review time of 36.2% while maintaining sensitivity of 88.9% for detecting abnormalities requiring intervention [4]. Referral guidance agents would assist physicians in identifying appropriate specialists for referrals based on diagnostic findings, with early implementations showing potential to reduce inappropriate referrals by 18.3% and decrease referral processing time by approximately 42% [3]. The implementation of AI-based workflow optimization has demonstrated potential to address administrative burdens that currently consume up to 50% of physician time, redirecting those resources toward direct patient care activities. [14]. This distributed architecture allows for modularity and scalability, enabling the system to adapt to the evolving needs of the primary care setting. Technical evaluations indicate that properly designed distributed systems can maintain response times below 2.1 seconds even while processing complex queries involving multiple data sources simultaneously, a crucial consideration for real-time clinical decision support [4]. The agents would communicate with each other and with the central EHR system, creating a seamless flow of information that reduces documentation redundancy and enhances care coordination across the healthcare continuum. Studies of interoperable healthcare systems indicate potential reductions in duplicate testing of 14-29% and improvements in care plan adherence of 17-23% through such enhanced information sharing [3].

Agent Type	Primary Function	Key Performance Metric	Value
Pre-appointment information gathering	Symptom collection	Information collection time reduction	11.3 minutes/patient
Differential diagnosis suggestion	Diagnostic analysis	Clinical concept extraction precision	83.7%
Test result interpretation	Anomaly detection	Review time reduction	36.2%
		Sensitivity for abnormalities	88.9%
Referral guidance	Specialist matching	Inappropriate referral reduction	18.3%
		Referral processing time reduction	42%

Table 2. Distributed Network Agent Performance Metrics [3, 4]

#### 4. The Indispensable Role of the Physician: Oversight and Ultimate Decision-Making

It is crucial to emphasize that the proposed agent-based system is intended to assist primary care physicians, not to replace them. Recent survey data examining physician perspectives on AI implementation indicates that 78.3% of primary care physicians believe AI should function primarily as a supportive tool rather than an autonomous decision-maker, with 83.6% expressing that maintaining clinical authority is essential for patient safety and quality of care [5]. The physician remains the ultimate decision-maker, responsible for integrating the information provided by the agents with their clinical judgment, patient context, and ethical considerations. Studies of clinical decision support implementations have found that hybrid approaches, where AI recommendations are reviewed and modified by physicians, demonstrate error reduction rates of 16.7% compared to AI-only approaches and improvement in clinical outcomes by 12.4% across multiple common conditions managed in primary care settings [5].

The implementation of agent-based systems must be guided by a physician-centered design philosophy that acknowledges the irreplaceable nature of clinical expertise. Analysis of AI integration in healthcare settings shows that systems designed with physician workflows in mind achieve adoption rates of 74.2% compared to just 31.8% for those developed without substantial clinician input [6]. This highlights the practical importance of preserving the physician's central role not just for clinical reasons but for successful technology implementation. Furthermore, patient trust considerations remain paramount, with survey data indicating that 67.3% of patients express concerns about diagnostic or treatment decisions made primarily by AI systems without significant physician interpretation and oversight [5]. While AI-based diagnostic assistants can enhance information processing capabilities, the physician's role remains essential for incorporating contextual factors such as patient preferences, social determinants of health, and unusual clinical presentations that may not be adequately captured in standardized datasets [15]. The agents should be designed as sophisticated tools that augment the physician's capabilities, providing them with timely and relevant information to support their decision-making process. Cognitive workload assessments conducted during clinical simulations found that well-designed AI assistants can reduce physicians' mental effort for information gathering tasks by up to 30.5% while increasing their consideration of relevant clinical variables by approximately 22.7% [6]. However, the same studies indicate that physicians still modified or adjusted AI-generated recommendations in 42.6% of cases, demonstrating the essential role of human judgment in contextualizing algorithmic outputs and adapting them to individual patient circumstances [6]. This integration of computational efficiency with clinical wisdom represents the optimal approach to leveraging technology while maintaining care quality.

The final diagnosis and treatment plan will always rest with the qualified medical professional. This principle is not merely philosophical but is supported by concrete evidence showing that physician-led collaborative diagnostic approaches outperform autonomous AI systems by 19.2% in terms of diagnostic accuracy for complex or atypical presentations common in primary care settings [5]. The concept of "augmented intelligence" rather than "artificial intelligence" better captures the complementary relationship between AI systems and physicians, emphasizing enhancement rather than replacement of human clinical judgment [16]. Additionally, implementation studies have found that physicians who maintain decisional authority while leveraging AI tools report 18.7% higher job satisfaction and 24.1% lower intention to leave practice compared to settings where they feel technological systems diminish their professional autonomy [6]. The complementary relationship between the physician and agent-based systems recognizes that while AI excels at certain analytical tasks, it cannot replicate the nuanced clinical reasoning, ethical judgment, and therapeutic relationship-building that characterize effective medical practice and that patients continue to value highly in their care experiences.

#### 5. Ethical Considerations: Navigating the Complexities of AI in Healthcare

The integration of AI into healthcare raises significant ethical considerations that must be carefully addressed. A comprehensive analysis of stakeholder perspectives revealed that ethical concerns rank among the top challenges for AI implementation in clinical settings, with survey data showing that 93.7% of healthcare professionals consider ethical governance essential for successful integration of AI technologies [7]. These ethical dimensions require systematic attention across multiple domains to ensure responsible deployment.

Patient privacy and data security represent primary concerns in the evolving healthcare technology landscape. Studies indicate that 82.4% of patients express significant concerns about the confidentiality of their information when used in AI-driven healthcare systems [7]. Robust security measures and strict adherence to privacy regulations (e.g., HIPAA in the United States) are paramount to ensure the confidentiality and integrity of patient data. Recent reviews of healthcare data protection protocols have found that despite technological advances, approximately 67% of healthcare institutions still lack comprehensive data protection frameworks specifically addressing AI applications [8]. Implementation of privacy-by-design principles during system development can substantially mitigate these risks, with healthcare organizations adopting such approaches reporting 57.3% fewer privacy-related incidents compared to those without structured protections [7]. Successful implementation of AI in healthcare requires a strategic

approach that addresses not just technological capabilities but also organizational readiness, clinical workflows, and ethical governance frameworks [17].

Algorithmic bias presents another critical ethical challenge in healthcare AI implementations. Current research demonstrates that diagnostic algorithms can reproduce or even amplify existing healthcare disparities when not properly designed and validated. Analysis of clinical decision support systems found that 43.2% showed statistically significant performance variations across different demographic groups, a concerning trend that could exacerbate existing healthcare inequities [8]. The training data used to develop AI agents must be carefully curated to avoid perpetuating such biases. Studies indicate that representative and diverse training datasets can reduce algorithmic performance disparities by 36.8% across population subgroups, highlighting the importance of inclusive approaches to algorithm development [7]. The onboarding and training of clinicians to effectively collaborate with AI systems represents a critical but often overlooked component of successful implementation, requiring attention to both technical competencies and adaptive challenges [18].

Transparency and explainability remain fundamental ethical requirements for clinical AI applications. Research shows that 78.5% of physicians express reluctance to follow AI-generated recommendations they cannot interpret or understand [8]. While the inner workings of complex AI algorithms can be opaque, efforts should be made to provide physicians with a degree of transparency and explainability regarding the agents' suggestions. Implementation studies have found that providing explanation features for AI outputs increases appropriate clinical utilization by 41.6% and physician confidence in the system by 53.7%, demonstrating the practical value of explainable AI approaches in healthcare settings [7].

Accountability and liability frameworks represent significant concerns for healthcare organizations implementing AI technologies. Survey data indicates that 73.9% of healthcare administrators cite unclear responsibility structures as a major barrier to AI implementation [7]. Clear lines of responsibility and accountability need to be established in the event of diagnostic errors or adverse outcomes involving the use of these AI agents. Legal and policy analyses suggest that developing formal governance structures with clearly delineated oversight responsibilities for AI systems can reduce institutional liability concerns by approximately 32.4% and improve stakeholder confidence in technology adoption [8].

Maintaining the human element in care remains essential despite technological advances. Patient experience studies consistently show that 89.3% of patients continue to value empathic communication and 91.2% prioritize trust in their healthcare providers above technological sophistication [7]. It is essential to ensure that the introduction of AI does not erode the crucial human connection and empathy that are fundamental to the patient-physician relationship. Comparative evaluations of AI-augmented clinical practices found that those maintaining an appropriate balance between technological assistance and interpersonal care achieved patient satisfaction scores 28.6% higher than technology-centered approaches [8].

Addressing these ethical considerations proactively through careful design, rigorous testing, and ongoing monitoring is crucial for the successful and responsible implementation of agent-based diagnostic assistants. Research on AI implementation in healthcare settings demonstrates that organizations with formal ethical review processes for AI applications report 46.8% fewer adverse events and 52.3% higher staff acceptance compared to those lacking structured ethical oversight [7]. Furthermore, systems developed with integrated ethical frameworks from the initial design stages achieve higher rates of sustainable implementation, with 61.7% remaining in active clinical use after two years compared to 37.4% of systems where ethical considerations were addressed retrospectively [8].

<b>Ethical Dimension</b>	<b>Stakeholder Concern</b>	<b>Impact of Addressing Concern</b>	<b>Percentage</b>
Ethical governance	Professional emphasis on importance	Professionals considering ethical oversight essential	93.7%
Patient privacy	Patient confidentiality concerns	Patients with concerns about AI data use	82.4%
Data security	Institutional readiness	Healthcare institutions lacking AI-specific data protection	67%
Algorithmic bias	Performance disparities	Decision support systems with demographic performance variations	43.2%

Transparency	Physician adoption barriers	Physicians reluctant to follow unexplainable AI recommendations	78.5%
Accountability	Implementation barriers	Administrators citing unclear responsibility as barrier	73.9%

Table 3. Ethical Dimensions of AI Implementation in Primary Care [7, 8]

### 6. Seamless Integration with Existing Electronic Health Records (EHRs): A Prerequisite for Success

The seamless integration of the agent-based system with existing EHRs is a critical prerequisite for its successful adoption in primary care settings. According to a comprehensive analysis of health information technology implementations, integration challenges account for approximately 32% of failed clinical AI deployments, underscoring the necessity of addressing this aspect from the outset [9]. The EHR serves as the central repository for patient information, with current systems generating between 80-250 structured data elements per patient encounter, and the AI agents must be able to securely access and interact with this vast repository to provide meaningful assistance [9]. Studies evaluating successful AI implementations in healthcare settings have found that systems with native EHR integration achieve 49.7% higher clinical utilization rates compared to standalone solutions requiring separate login or data transfer steps [10].

Challenges related to data interoperability, standardization, and security need to be overcome to ensure that the agents can effectively access and process the necessary information. Research indicates that despite progress in healthcare data standards, approximately 46% of primary care organizations still experience significant interoperability issues when implementing new technologies, with each organization maintaining an average of 4.7 distinct data systems that must exchange information seamlessly [9]. Implementation studies have shown that successful integration requires addressing these challenges systematically, with healthcare systems that establish formal data governance protocols achieving implementation success rates of 73.8% compared to 38.2% for those without structured approaches [10]. Additionally, standardization efforts that include comprehensive data dictionaries and clinical terminology mapping have demonstrated the potential to reduce ambiguity in clinical data interpretation by up to 64.8%, a critical consideration for ensuring the accuracy of AI-generated insights [9]. The development of standardized application programming interfaces (APIs) for healthcare data exchange represents a promising approach to addressing interoperability challenges between AI systems and diverse EHR platforms [19].

Collaboration between AI developers, EHR vendors, and healthcare providers is essential to achieve this seamless integration. Multistakeholder collaborative approaches to healthcare IT implementation have shown success rates 2.3 times higher than those with limited stakeholder involvement, with implementation timelines averaging 41% shorter duration when all relevant parties are engaged from the planning stages [10]. The complexity of this collaboration is substantial, with successful implementation typically requiring structured cooperation among at least six distinct organizational entities, including clinical, technical, administrative, and regulatory stakeholders [9]. Healthcare institutions that establish formal collaborative governance structures for technological implementations report 37.5% fewer post-implementation workflow disruptions and 52.3% higher satisfaction among both technical and clinical staff [10].

Integration Factor	Challenge	Success Metric	Value
Integration approach	Failed AI deployments due to integration issues	Percentage of AI failures attributed to integration	32%
Native EHR integration	Standalone vs. integrated solutions	Clinical utilization improvement with native integration	49.7%
Interoperability	Data exchange barriers	Primary care organizations with significant interoperability issues	46%
Data standardization	Clinical interpretation challenges	Reduction in data ambiguity with terminology mapping	64.8%
Formal data governance	Implementation success	Success rate with formal data governance	73.8%

Multi-stakeholder collaboration	Implementation efficiency	Implementation timeline reduction with full stakeholder engagement	41%
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Table 4. Critical Metrics for Successful AI-EHR Integration [9, 10]

**7. Future Research Directions and Potential Impact**

Further research is needed to fully realize the potential of agent-based diagnostic assistants in primary care. A systematic review of the current state of AI in healthcare identified several critical research gaps that must be addressed, with 56.3% of surveyed healthcare organizations citing the need for more rigorous real-world validation studies before widespread implementation can occur [9]. These research priorities span multiple domains that will shape the future evolution of these technologies in healthcare delivery.

Development and validation of agent algorithms represents a primary research focus. Current clinical validation studies for AI diagnostic tools in primary care typically include case volumes ranging from 1,500-7,000 patients, often lacking sufficient diversity in demographics, comorbidities, and presentation variants [10]. Rigorous research is required to develop and validate the accuracy and reliability of the algorithms used by these agents in various clinical scenarios. Studies of existing diagnostic algorithms have found performance variations of 14-31% between controlled testing environments and real-world clinical settings, highlighting the importance of pragmatic clinical trials that reflect actual practice conditions [9]. Furthermore, research indicates that diagnostic performance metrics can vary by up to 28.3% across different patient populations, emphasizing the need for inclusive validation approaches that encompass diverse demographic and clinical characteristics [10].

Usability and workflow integration studies represent another crucial research direction. Analysis of healthcare technology implementations indicates that approximately 42% of clinically sound technologies fail to achieve sustained adoption due to workflow integration challenges and user experience limitations [9]. Studies are needed to assess the usability of the system by physicians and to optimize its integration into existing primary care workflows. Current research suggests that healthcare technologies requiring more than 3.2 minutes of additional work per patient encounter face adoption rates below 25%, regardless of their potential clinical benefits [10]. Comprehensive time-motion studies across diverse clinical environments will be essential to identify optimal integration approaches that enhance rather than impede clinical efficiency, with particular attention to high-volume primary care settings where time constraints are especially pronounced [9]. The transformation of healthcare delivery through AI-based systems will require moving beyond narrow technical implementations toward holistic solutions that address the quadruple aim of improving patient experience, enhancing population health, reducing costs, and improving the work life of healthcare providers [20].

Impact on patient outcomes and costs requires longitudinal evaluation to establish definitive evidence. While preliminary studies suggest potential benefits, long-term research with follow-up periods of at least 18-24 months is necessary to determine sustained effects and account for learning curves and adaptation periods [10]. Long-term studies are necessary to evaluate the impact of these systems on patient outcomes, healthcare costs, and physician satisfaction. Current economic analyses suggest potential efficiency improvements valued at approximately £5.2-£7.8 per patient encounter in UK-based studies, though these projections require validation through prospective trials in diverse healthcare systems and payment models [9]. Additionally, comprehensive outcome evaluations must consider multiple dimensions, including clinical outcomes, safety metrics, patient experience indicators, and system-level impacts to provide a complete assessment of value [10].

Addressing edge cases and rare conditions presents unique research challenges that merit focused attention. Current diagnostic algorithms demonstrate performance decreases of 35-57% when confronting conditions with prevalence below 1% compared to common presentations, highlighting a significant limitation for comprehensive diagnostic support [9]. Further research is needed to explore how these agents can be effectively utilized in complex cases and for the diagnosis of rare conditions. Multi-institutional collaborations involving diverse clinical settings will likely be necessary to accumulate sufficient case volumes for algorithm training and validation for uncommon conditions, with preliminary studies suggesting that data from at least 8-10 distinct healthcare environments may be required to capture sufficient clinical variability [10]. Innovative research methods, including federated learning approaches and synthetic data augmentation, show promise for addressing these challenges while maintaining patient privacy and data security [9].

Despite these challenges, the potential impact of agent-based diagnostic assistants on primary care is significant. Implementation analyses project that well-designed systems could reduce documentation time by 25.7-38.2%, potentially saving 5.1-7.8 hours of clinician time per week that could be redirected to direct patient care or additional case volume [10]. By streamlining workflows, reducing physician burnout, and potentially improving the speed and accuracy of initial diagnoses, these systems could contribute to a more efficient, accessible, and patient-centered healthcare system. Research suggests potential improvements in diagnostic

workflow efficiency of 16.5–24.8% for common conditions and reductions in unnecessary investigations of 11.3–18.7%, representing substantial quality and resource utilization benefits [9]. Furthermore, simulation models indicate that optimal implementation could potentially expand primary care access by 12.8–19.3% without requiring additional clinicians, a critical consideration given projected workforce shortages in primary care across many healthcare systems [10].

## 8. Conclusion

The integration of agent-based diagnostic assistants into primary care workflows represents a promising paradigm shift in healthcare delivery. By leveraging AI to automate routine tasks, provide timely insights, and support clinical decision-making, these intelligent agents can significantly enhance the efficiency and effectiveness of primary care. Successful implementation requires careful consideration of ethical implications, seamless integration with existing EHR systems, and a clear understanding of the physician's continued role as the ultimate decision-maker. As this field continues to advance, agent-based diagnostic assistants hold the potential to transform primary care, ultimately leading to better patient outcomes and a more sustainable healthcare system that balances technological innovation with the irreplaceable human elements of medical care.

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