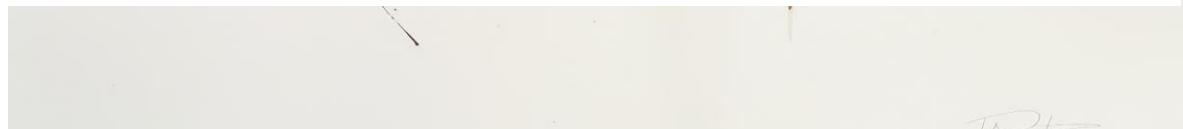




THE AGENTIC DIAGNOSTIC REVOLUTION: SOLVING THE "PAPER-TO-PATIENT" FRICTION WITH AGENTIC AI

BRIDGING THE ILLEGIBILITY GAP, DATA SILOS, AND LANGUAGE BARRIERS IN GLOBAL
HEALTHCARE



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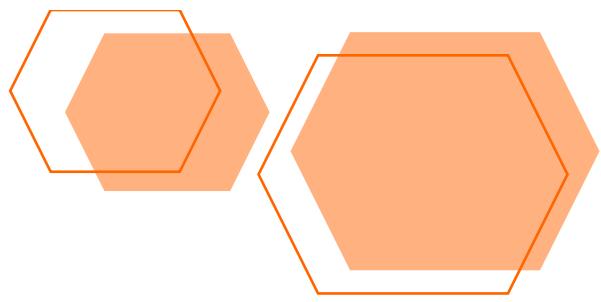
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EXECUTIVE SUMMARY



The global diagnostic industry is undergoing a "silent crisis." While imaging and lab equipment have reached sub-millimeter precision, the inputs (handwritten prescriptions and TRFs) and outputs (standardized reporting) remain tethered to manual, error-prone processes.

This whitepaper details how Agentic AI—a reasoning-based evolution of Artificial Intelligence, is solving these challenges. We present a roadmap from the successful automation of Test Requisition Forms (TRFs) and HIPAA authorizations to the "Holy Grail" of diagnostics: decoding handwritten physician prescriptions and providing real-time, multilingual clinical synthesis. The whitepaper also discusses on how the Agentic AI solution is not only a technical revolution but also a financial disruption.



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THE CRISIS OF UNSTRUCTURED DATA IN MODERN DIAGNOSTICS



THE "IL-LEGIBILITY TAX" IN INDIA AND THE US

Unstructured data (handwriting, scanned PDFs, and verbal notes) accounts for 80% of all healthcare information [Source: IDC, 2024].

- **The Error Rate:** Manual transcription of medical charts carries a **9.19% average error rate**. In critical variables like drug dosage, this spikes to 15% [Source: *Orthopedics* / *Healio*, 2020].
- **Workforce Exhaustion:** **58.3% of pathology staff** globally report burnout due to "administrative documentation" [Source: Bio-Reach, 2025].

CORE TECH: THE SHIFT TO AGENTIC AI ON GOOGLE VERTEX

Standard AI is a tool; **Agentic AI** is a teammate. Using the Google Vertex ecosystem, we have built a system that doesn't just "read" text—its "reasons" through clinical context.

THE TECH STACK

- **Med-Gemini & MedLM:** These models outperform general LLMs on medical benchmarks (MedQA) with a **91.1% accuracy** [Source: Google Research, 2024].
- **Vertex AI Document AI:** Specializes in high-fidelity OCR for non-standardized handwritten medical forms.
- **Agentic Orchestration:** The system uses "Function Calling" to query a **Medical Dictionary** in real-time, cross-referencing a handwritten "squiggle" against thousands of known drug brands and symptoms.

PRACTICAL APPLICATIONS: THE NEW STANDARD IN DIAGNOSTICS

1. DECODING "PATHETIC" DOCTOR PRESCRIPTIONS (INDIA)

The handwriting of Indian clinicians is a significant barrier to pharmacy and lab accuracy.

- **The Solution:** Our PoC uses Vertex AI to ingest the prescription. The "Agent" identifies a partial word (e.g., "Panto...") and queries a dictionary of the Indian pharmacopeia.
- **The Logic:** If the diagnosis is "Gastritis," the Agent confirms the word is *Pantocid*, not *Paracetamol*, achieving human-level accuracy in seconds.

2. "SCREEN-TO-REPORT" (RADIOLOGY/ULTRASOUND AUTOMATION)

For a leading diagnostic company, we have removed the transcription bottleneck for radiologists.



- **The Problem:** Radiologists manually type findings from ultrasound/X-ray screens into reports.
- **The Solution:** The Agent takes the handwritten readings or "on-screen" measurements and uses **MedSigLIP** (Google's medical vision encoder) to populate a standardized format.
- **Impact:** A **40% reduction** in Report Turnaround Time (TAT) [Source: Google Medical Imaging Suite Benchmarks].

3. MULTI-SOURCE PATHOLOGICAL DATA AGGREGATION

- **The Problem:** Patients have disparate reports from multiple diagnostic companies, often in different units and formats.
- **The Solution:** We aggregate these reports, normalize the data (e.g., converting \$mg/dL\$ to \$mmol/L\$), and provide a **Longitudinal Analysis**.
- **The Outcome:** The Agent identifies "silent trends"—such as a slow rise in glucose across three different labs—that a single snapshot report would miss.

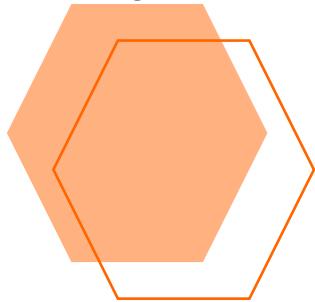


BREAKING THE LANGUAGE BARRIER: MARATHI AND TELUGU LOCALIZATION

Medicine is only effective if the patient understands it.

- **The Solution:** Leveraging Google's Neural Machine Translation (NMT) on Vertex AI, we translate complex findings into **Marathi, Telugu, and other regional languages**.

- **The Clinical Twist:** The Agent ensures the severity is translated correctly (e.g., "Critical" vs "Stable"), following their treatment plans accurately.



THE ECONOMIC BENEFITS OF IMPLEMENTING AGENTIC AI

THE ECONOMICS OF AUTOMATION — AGENTIC AI vs. LEGACY RPA+IDP

Historically, diagnostic labs used Robotic Process Automation (RPA) paired with Intelligent Document Processing (IDP). While effective for static forms, this stack fails in the "real world" of diagnostics for three reasons:

1. **Maintenance Overload:** RPA bots are "brittle." If a hospital changes its report template or a doctor uses a different prescription pad, the bot breaks. Maintenance costs for RPA can often reach **30–50% of the initial deployment cost** annually [Source: *Auxiliobits*, 2025].
2. **The "OCR Accuracy Gap":** Traditional IDP/OCR typically plateaus at **85–90% accuracy** for handwriting. The remaining 10-15% "exceptions" must be handled by expensive human medical coders, nullifying much of the ROI.
3. **Linear Scaling Costs:** To process more forms with RPA, you typically need more bot licenses. With Agentic AI on Vertex, you scale with **API consumption**, which is non-linear and significantly cheaper at high volumes.

WHY AGENTIC AI IS THE MORE COST-EFFECTIVE PATH

| Feature | Legacy RPA + IDP Add-on | Agentic AI |
|--|---|--|
| Handling "Pathetic" Handwriting | Requires custom "template" training for every variation. High failure rate. | Zero-Shot Reasoning: MedLM understands intent and context without needing a template. |

| | | |
|-----------------------------|---|--|
| Exception Handling | Breaks and requires human intervention (High OpEx). | Self-Healing: Agent queries dictionaries to "solve" the problem autonomously. |
| Implementation Speed | 3–6 months for complex medical forms. | Weeks: Pre-trained MedLM models are "medical-ready" on Day 1. |
| Long-term TCO | High (constant script updates/maintenance). | Low: The model improves with every document it reads. |