# Foundation Models in Healthcare: What You Need to Know



Foundation models are a breakthrough in AI development-built to scale quickly, broadly and efficiently. Trained on massive, diverse datasets, they allow you to go from building one algorithm to 100 in a fraction of the time, enabling full clinical coverage across pathologies without having to pick and choose use cases.

Traditional AI is like hiring a skilled craftsman to hand-make a single tool for one task. Foundation models, in contrast, are like a factory that can rapidly produce any tool you need—at speed, at scale and with the flexibility to meet a wide range of demands. This shift in approach transforms how AI is developed, deployed and expanded across the enterprise.







# **Training AI With Foundation Models** vs. Traditional Approaches



## Training Data

- Foundation Models: Trained on large, diverse datasets to learn generalizable patterns that allow them to adapt to a wide range of tasks with minimal fine-tuning.
- Traditional Al: Single-task models trained on smaller, narrowly focused datasets that require building and training a new model from scratch for each application.



# **Self-Supervised Learning**

- Foundation Models: Learn from huge, unlabeled multimodal datasets, which reduce reliance on labeled data - and improve AI scalability.
- Traditional AI: Depend on large, labeled datasets, making training more resource-intensive.



#### Generalizability

- Foundation Models: Can seamlessly handle multiple clinical tasks and imaging types (CT, MRI, X-ray) with minimal fine-tuning.
- Traditional Al: Designed for a single task on a single imaging type, and requiring de novo training for each function.



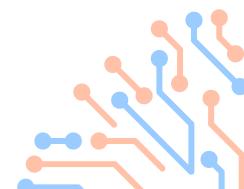
## Scalability

- Foundation Models: Can be quickly fine-tuned for new applications, accelerating Al adoption.
- Traditional AI: Requires lengthy development cycles due to task-specific training and regulatory approvals.



## **Multimodal Capabilities**

- Foundation Models: Trained to process both imaging and text (reports, clinical notes, lab results, etc.) in one system for holistic patient insights.
- Traditional Al: Analyzes one data type at a time (e.g., imaging-only or text-only), limiting cross-specialty insights.



# **Misconceptions of Foundation Models**

Despite their advantages, foundation models are often misunderstood.

"Foundation models can do everything perfectly without additional training."

**Reality:** While powerful, fine-tuning or prompt engineering is often required for optimal clinical performance.

"Foundation models replace the need for an Al platform."

**Reality:** A foundation model is one component of an AI platform—not a replacement. AI platforms ensure deployment, usability and workflow integration.

## **Challenges and Considerations**

While foundation models offer transformative potential, several barriers must be addressed:

#### 1. Compute Demands

High computational costs and energy demands raise sustainability concerns.

#### 2. Regulatory Hurdles

New tech. New rules. Foundation models demand regulatory pathways that either don't exist or are still emerging.

## 3. Workflow Integration

Due to broad coverage, clinicians may need new AI interfaces and workflow adjustments for seamless adoption.

#### 4. Data Scale and Trust

Truly robust foundation models require tens of millions of diverse cases. Models trained on smaller datasets may lack generalizability and should be evaluated critically.

## **Takeaways for Healthcare Leaders**

Foundation models mark a leap forward—enabling scalable AI development, faster deployment and broader clinical coverage with minimal fine-tuning.

## One Model, Multiple Applications

Foundation models can support multiple applications across care settings and service lines.

# Holistic Patient Analysis

Combines imaging, EHR and lab data for a more holistic clinical view—helping surface what individual tools might miss.

## Scalable, Rapid Al Deployment

Without a platform, foundation models can't be delivered, integrated or scaled in real-world clinical workflows.

Healthcare leaders must critically assess foundation models, ensuring they meet the demands of modern clinical workflows while addressing scalability and ensuring regulatory compliance. To learn more about how foundation models are shaping the future of clinical AI, visit www.aidoc.com.