# **Al Literacy in Healthcare**

8-Question Assessment (With Answer Guidance)

# Purpose:

Use this handout to help clinicians and leaders identify their current AI literacy level and pinpoint targeted next steps. Work through the questions in order. When you reach a question you can't confidently answer, note your level and review the guidance.

### How to use:

For each question, read the prompt and the "What a solid answer includes" guidance. If you can address most bullets with examples relevant to your setting, move to the next question.

- 1. Al vs. Traditional Automation
- 2. Predictive Al vs. Generative Al
- 3. Hallucinations in Al Tools
- 4. Data Bias and Performance Transfer
- 5. Evaluating Vendor Claims
- 6. "No-Regrets" Al Investments
- 7. Diagnosing Why Pilots Don't Scale
- 8. Real ROI vs. Hype in Healthcare AI

### 1. Al vs. Traditional Automation

# Prompt:

How would you explain the difference between "AI" and "traditional automation" to a colleague?

## What a solid answer includes:

#### • Traditional Automation:

Deterministic, rule-based workflows (if-then logic), does exactly what it's programmed to do, no learning.

#### Al:

Models learn patterns from data, handle ambiguity, make probabilistic predictions, improve with new data.



# Examples:

#### Rules-based:

EHR alert triggers when potassium > 6.0.

#### Al:

Model predicts risk of readmission using multi-variable patterns learned from historical data.

### 2. Predictive Al vs. Generative Al

# Prompt:

A colleague says, "We're using predictive AI to identify sepsis risk and generative AI to draft clinical notes." Why are these fundamentally different?

#### What a solid answer includes:

#### • Predictive AI:

estimates likelihood of an outcome (classification/regression).

- Inputs: structured/unstructured features
- *Output*: probabilities/scores (e.g., sepsis risk within 12–24 hours).

#### Generative AI:

creates new content (text/audio/images) conditioned on prompts and context (e.g., clinical note drafts).



# Different objectives, evaluation methods, and risks:

#### • Predictive:

measured by AUROC, calibration, sensitivity/specificity.

#### Generative:

measured by accuracy to source, coherence, factuality, style compliance; prone to hallucinations.

# 3. Hallucinations in Al Tools

# Prompt:

A documentation assistant claimed a symptom never mentioned during the visit. What likely went wrong, and why is this unique to Al tools?

### What a solid answer includes:

#### Hallucination:

Model outputs plausible but false information, not grounded in source data.

### How to prevent this:

#### • Causes:

Overgeneralization, weak grounding to transcripts/EHR, prompt or context errors, token truncation, model attempting to "fill gaps."

# Mitigations:

Retrieval-augmented generation (RAG) with strict citation to source, constrained templates, clinician-in-the-loop review, confidence flags, audit logs.

### 4. Data Bias and Performance Transfer

# Prompt:

A sepsis predictor trained at AMCs performs poorly at a rural hospital. What data issues could explain this?

#### What a solid answer includes:

#### Dataset shift:

differences in patient demographics, comorbidities, prevalence, care pathways, lab availability, documentation practices.

## Sampling Bias:

AMC data may overrepresent complex cases; rural data may differ in acuity/resource patterns.

#### Labeling Differences:

sepsis definitions/coding vary; gold standards may not match.

#### • Feature Distribution Drift:

lab timing/frequency, vital sign completeness, EHR vendor differences.

# How to prevent this:

#### Mitigations:

local validation, recalibration, transfer learning, federated/representative training data, ongoing monitoring.



# 5. Evaluating Vendor Claims

# Prompt:

Your organization is vetting a vendor's clinical decision support tool. What questions test credibility?

#### What a solid answer includes:

#### • Training Data:

Size, sources, representativeness, timeframes, inclusion/exclusion criteria, PHI handling.

#### Validation:

External, multi-site validation; prospective vs. retrospective; metrics (AUROC, PPV/NPV, calibration); subgroup performance (equity).

#### • Real-World Performance:

Workflow fit, measured clinical/operational outcomes, silent trial results, degradation over time, monitoring plan.

### • Explainability and Safety:

Model type, feature importance, error modes, human-in-the-loop, override, auditability.

### Governance and Compliance:

FDA status (if applicable), MDR/IVDR, HIPAA, SOC2/ISO 27001, bias testing, incident response.

### Implementation:

Integration (EHR APIs, FHIR), change management, training, total cost of ownership, ROI.

# 6. "No-Regrets" Al Investments

# Prompt:

What foundational investments pay off even if specific AI tools become obsolete?

### What a solid answer includes:

### • Data Quality and Interoperability:

Standardized data (FHIR, terminologies), metadata, lineage, de-duplication, robust ETL pipelines.

#### Governance:

Clear AI use policies, risk review boards, model lifecycle management, bias and safety monitoring, documentation standards.

#### Workforce Al Literacy:

Role-based training for clinicians, informatics, IT, leadership; safe-use practices; prompt hygiene; oversight.

#### Vendor Evaluation Processes:

Standardized due diligence checklist, pilot-to-scale criteria, post-deployment monitoring.





#### • Infrastructure:

Secure data platforms, MLOps, access controls, audit logging, API integration capabilities.

# 7. Diagnosing Why Pilots Don't Scale

# Prompt:

The CEO asks, "We keep piloting AI but nothing scales. What's our biggest bottleneck?" How do you diagnose it?

#### What a solid answer includes:

### • Data Infrastructure Maturity:

Reliability of feeds, interoperability, latency, environments for development/ testing/ production, monitoring.

#### Governance Structures:

Decision rights, risk approvals, clinical safety sign-off, model change control, accountability.

### • Change Management:

Workflow redesign, clinician adoption, training, incentives, measurement of impact, support channels.

#### Strategic Alignment and Business Case:

Clear problem statements, value metrics, funding, owner identified, scale plan, IT capacity.

### • Integration Readiness:

EHR/API integration, single sign-on, embedding in existing workflows, alert fatigue management.

# 8. Real ROI vs. Hype in Healthcare AI

# Prompt:

Where is AI creating real ROI today, and where is it mostly hype? Give specific examples.

# What a solid answer includes:

#### Proven ROI (examples):

- **Ambient Clinical Documentation:** reduced clinician burden, improved note quality, time savings.
- Operational Optimization: patient flow, staffing forecasts, denials management, supply chain.
- **Imaging Triage and Workflow Support:** prioritization for critical findings (site-dependent regulatory clearance).
- Contact Center / Administrative Automation: prior auth assistance, scheduling support.

#### Emerging/hype (examples to treat cautiously):

- Fully autonomous diagnosis without clinician oversight





- Broad, unsupervised generative summaries inserted into the EHR without grounding or review.
- Complex inpatient decision support without rigorous multi-site prospective validation.

#### • Evidence Checklist:

- Peer-reviewed studies, multi-site outcomes, regulatory status (where applicable), clinician satisfaction, sustained performance over time.

# Fluency Levels and Next Steps:

□ Al Novice (Q1–2):

Build foundational literacy; start with AI Fundamentals and Glossary.

 $\square$  Al Aware (Q3–4):

Deepen understanding of Al-specific risks/benefits; review Al Governance and risk slides.

 $\square$  Al Informed (Q5–6):

Focus on implementation and governance; use the FUMBLE framework and "no-regrets" investments.

☐ Healthcare AI Competent (Q7–8):

Lead strategy; apply the Al Maturity Roadmap and Use Case Library.

☐ Healthcare AI Fluent (All 8):

Maintain fluency; mentor peers; subscribe to AI Catalyst Pulse for updates.

# Quick Tips for Participants:

- ✓ Anchor AI tools to clinical/operational goals with measurable outcomes.
- ✓ Require local validation and ongoing monitoring before scaling.
- ✓ Keep humans in the loop; design for safety, transparency, and accountability.
- ✓ Invest in durable foundations (data, governance, literacy) to ensure ROI as tools evolve.

