

# Artificial Intelligence in Ophthalmology: From Innovation to Implementation



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All information presented reflects the **author's own analysis, interpretation, and professional judgment** based on publicly available evidence and recognized clinical standards

## About the Presenter

Jeffery Daigrepont, Senior Vice President at Coker, specializes in healthcare automation, system integration, cybersecurity, operations, and deployment of enterprise information systems for large integrated delivery networks and medical practices. His specific interests deal with data migration, vendor contracting, strategic IT planning and optimization, security, and compliance.

A popular national speaker, Jeffery is frequently engaged by highly respected organizations across the nation, including many non-profit trade associations and state medical societies. He has co-authored a top-selling book, Complete Guide and Toolkit to Successful EHR Adoption (2011 HIMSS) and is a contributor to The Healthcare Executive's Guide to ACO Strategy (2015 HCPro).

## Jeffery Daigrepont

Senior Vice President



# Learning Objectives

At the conclusion of this session, participants will be able to:

1. **Explain** the fundamental principles and terminology of artificial intelligence and machine learning relevant to ophthalmology.
2. **Analyze** the strengths, limitations, and validation methods of current AI-based diagnostic and decision-support tools.
3. **Differentiate** trustworthy AI applications from those lacking adequate evidence, transparency, or bias mitigation.
4. **Apply** ethical, regulatory, and clinical frameworks (e.g., FDA, NIST, AAO) to evaluate readiness for implementation.
5. **Integrate** responsible AI practices into clinical and operational workflows to enhance patient care and trust.









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The future of AI in ophthalmology provides a glimpse into technological transformation in vision science.

This presentation emphasizes clarity, evidence-based trends, and forward-looking perspectives



# Artificial Intelligence in Ophthalmology (The Fundamentals)



Photo Credit: Stock Adobe

## Key Concepts

- **Artificial Intelligence (AI):** Algorithms that mimic human decision-making to interpret clinical data.
- **Machine Learning (ML):** Systems that learn patterns from imaging and EHR data to improve prediction accuracy.
- **Deep Learning (DL):** Neural networks—especially **CNNs**—optimized for ophthalmic imaging (e.g., OCT, fundus).
- **Supervised vs. Unsupervised Learning:**
  - *Supervised* models learn from labeled pathology (DR, AMD, glaucoma).
  - *Unsupervised* models discover new phenotypes or clusters.
- **Key Metrics:** Sensitivity, specificity, calibration, generalizability.
- **Ophthalmology Relevance:** Image-heavy specialty makes it ideal for DL-enabled screening, segmentation, and progression modeling

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- **Strengths, Limitations & Validation of AI Diagnostic Tools**

- Strengths**

- High performance in image recognition tasks (DR, AMD, glaucoma).

- Scalable screening in primary care, retail health, and underserved settings.

- Consistent interpretation with reduced human variability.

- Limitations**

- Performance may drop with different imaging devices, demographics, or comorbidities.

- Limited transparency (“black box” models).

- Risk of overfitting if trained on narrow datasets.

- Validation Requirements**

- External validation** across sites, devices, and diverse populations.

- Prospective clinical trials** and real-world performance monitoring.

- Human factors testing** for workflow integration.

- Evaluation of **bias, false positives/negatives**, and **clinical impact**.



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- **Identifying Trustworthy vs. Untrustworthy AI**

### **Characteristics of Trustworthy AI**

**Evidence-backed:** Peer-reviewed studies, transparent methodology, strong external validation.

**Explainability:** Clear model rationale or visual heatmaps showing decision drivers.

**Bias Mitigation:** Training data reflects diverse age, race, imaging device, and disease stages.

**Robust Governance:** Data lineage, version control, monitoring, and risk assessments.

**Regulatory Alignment:** Meets UL 2933, NIST AI RMF, AAO guidelines.

### **Warning Signs of Untrustworthy AI**

No published validation or undisclosed datasets.

Claims that exceed the evidence (e.g., broad disease detection with small datasets).

Inconsistent performance across populations or devices.

Lack of transparency in data use, model updates, or error rates.

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- **Integrating Responsible AI into Clinical Workflows**

- Workflow Integration**

- Align AI tasks with existing imaging, triage, and decision-support pathways.

- Ensure seamless **EHR documentation**, clinical routing, and follow-up steps.

- Use AI for augmentation (screening, prioritization), not autonomous clinical judgment unless FDA-cleared.

- Operational Practices**

- Establish AI governance committees (clinical, IT, compliance, risk).

- Develop protocols for human oversight, escalations, and override scenarios.

- Continuously monitor model drift, error patterns, and user feedback.

- Patient-Centered Practices**

- Transparent communication: how AI is used and its benefits/limitations.

- Ensure equitable access and remove bias that could impact outcomes.

- Build patient trust through safe, consistent, evidence-based deployment.

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## Where is this going? Could Agentic AI replace the EHR entirely??? Many speculate it will!

- Software as a Service will soon become AI Agents as a Service
- The new UI is no UI (Smart Sensors)
- No Keyboard or Monitors

### Why the Theory Makes Sense:

- 1 **Frictionless Data Capture** – Voice + sensors (wearables, IoT devices, computer vision) could eliminate the need for manual typing, dramatically improving clinician efficiency and patient interaction time.
- 2 **Contextual Understanding** – Agent AI could interpret conversations, lab results, imaging, vitals, and patient history in real time, then auto-populate structured records without human data entry.
- 3 **Interoperability Leapfrog** – Instead of building on top of rigid EMR databases, Agent AI could work across systems as an orchestration layer, bypassing legacy user interfaces entirely.



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## My prediction on this...

**Most Likely Path** - I don't see Agent AI *wholly* replacing EMR vendors in the near term.

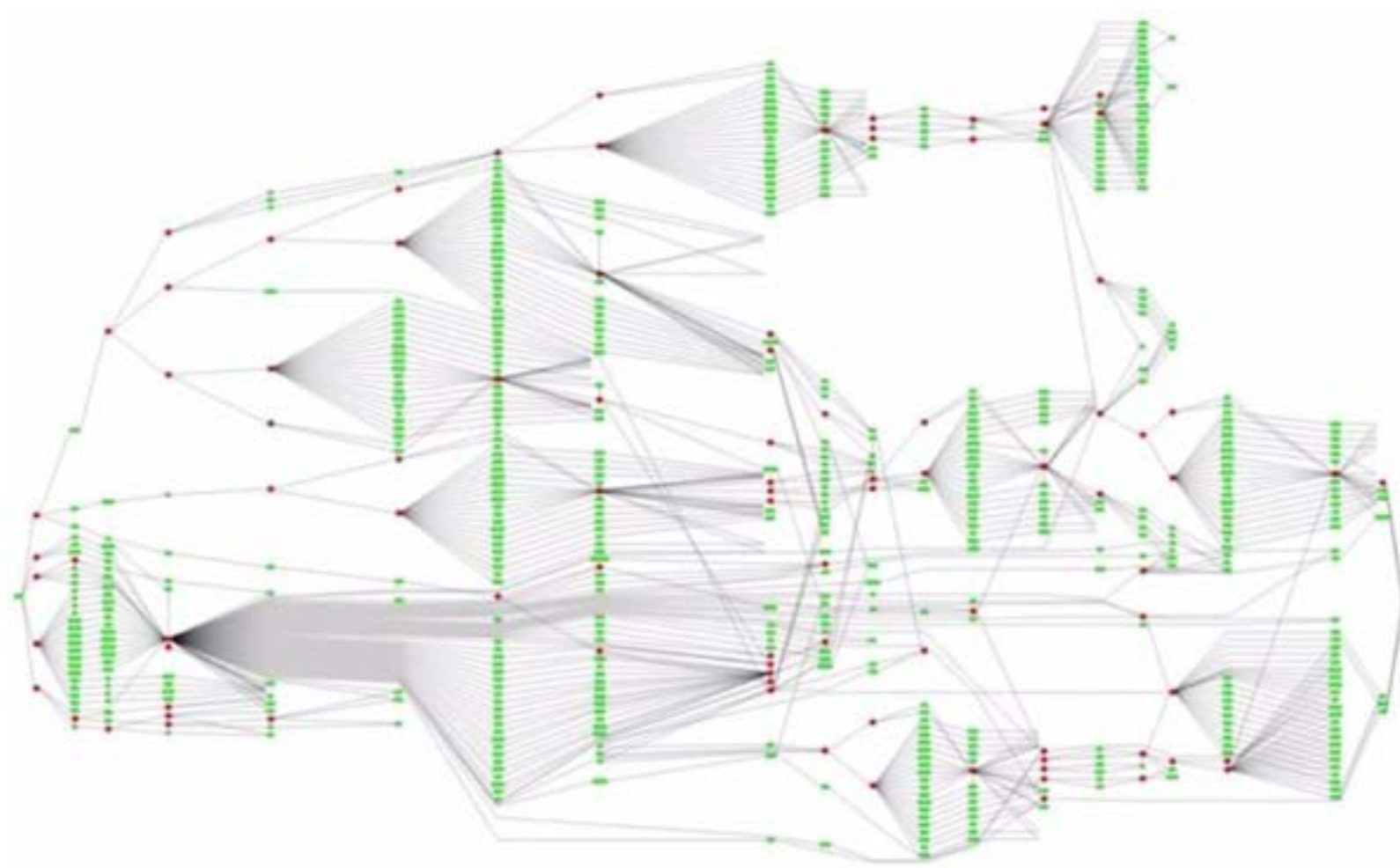
Instead:

**Short Term (1–3 years)** – Agent AI will act as a *co-pilot* to existing EMRs, sitting in on patient encounters, transcribing, coding, and entering data into existing fields.

**Medium Term (3–7 years)** – Some specialty-specific or ambulatory care settings could adopt “EMR-lite” models where the AI agent becomes the primary interface, with a minimal underlying system for regulatory compliance.

**Long Term (7–10+ years)** – We could see a complete inversion, where the EMR is reduced to a compliant back-end database, and Agent AI is the *front door* for all interaction—clinicians rarely touch the traditional UI.

# AI Logic



## Happening Now..

### Transforming Practice Management with AI

- **Operations:** Automates scheduling, EHR data entry, and resource allocation to boost efficiency.
- **Patient Access:** Enhances engagement with AI chatbots, telehealth integration, and accessible screening tools.
- **Compliance:** Ensures accurate coding, tracks regulations, and supports audits to minimize errors.
- **Security:** Protects patient data with AI-driven cybersecurity and fraud detection.
- **Revenue Cycle Management (RCM):** Streamlines insurance verification, reduces claim denials, and optimizes cash flow.

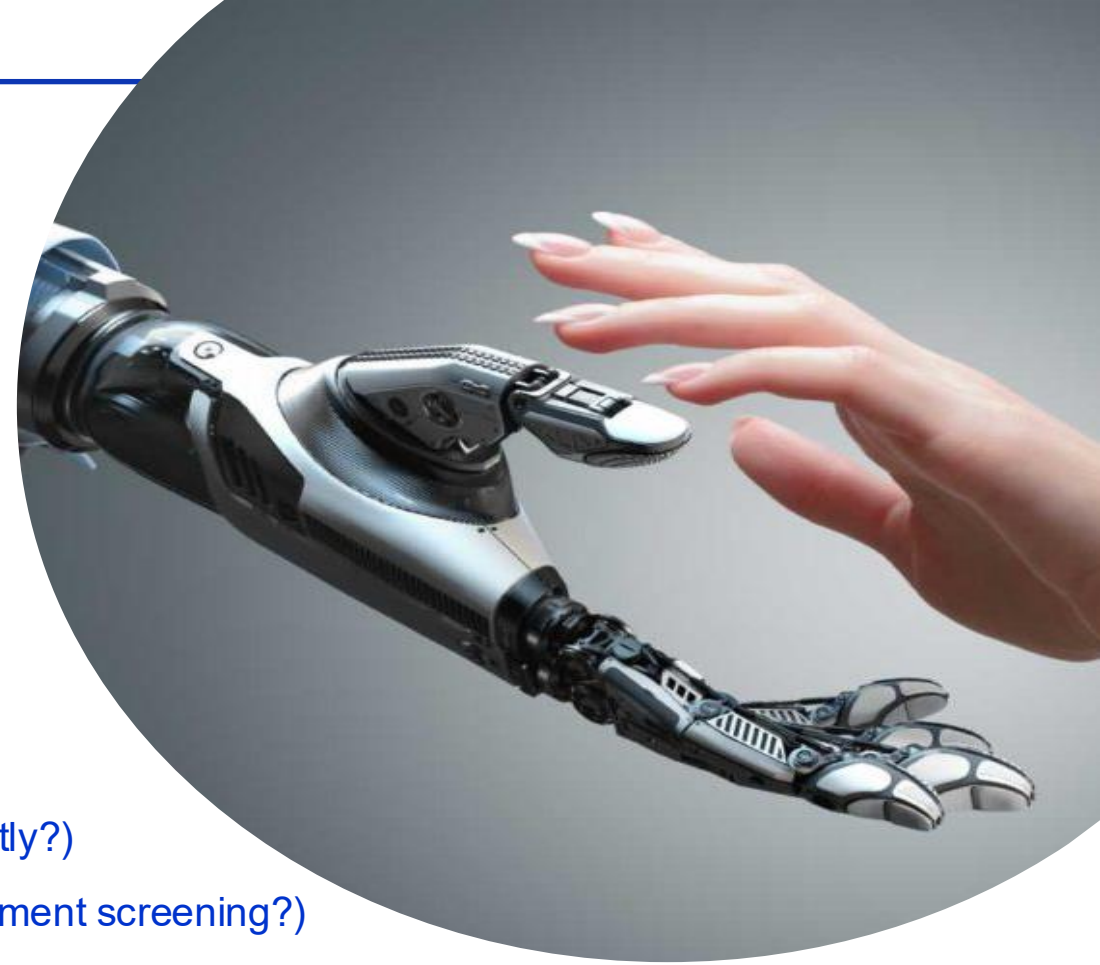


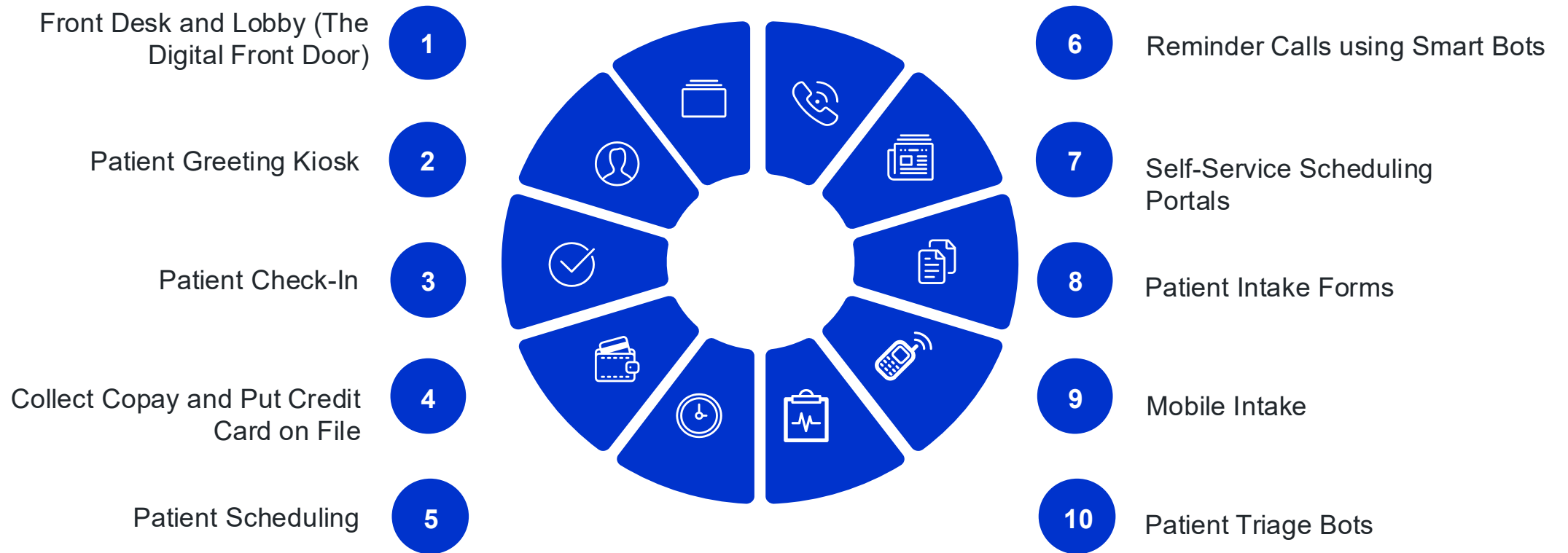
Photo Credit: AI Generated



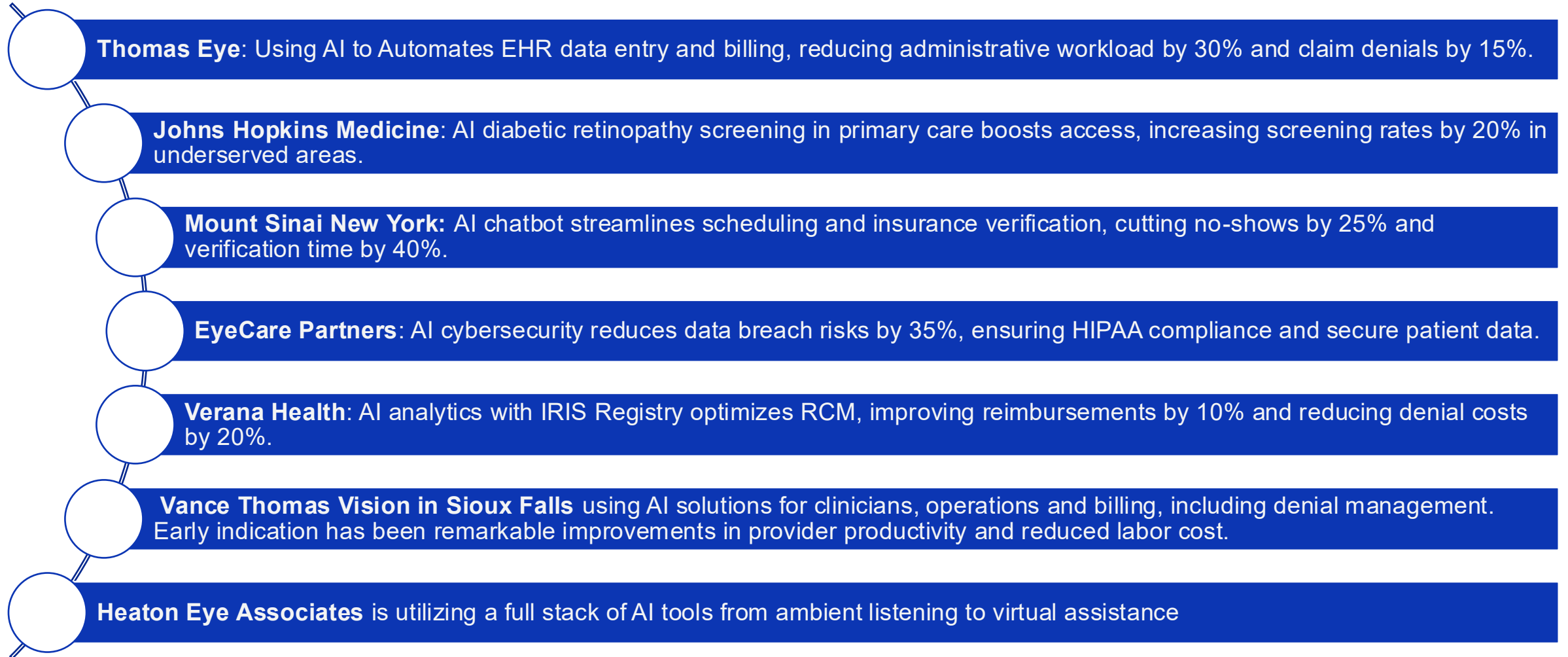
# AI Concerns and Risks

- AI discrimination such as bias or racism (see: Microsoft Tay)
- Inequality (how do we distribute the wealth created by machines?)
- Evil AI (e.g., cures cancer by destroying humanity)
- AI Rights (can non-humans have rights? A: See Corporations)
- Humanity (how do machines affect our behavior and interaction?)
- IP developed by AI (who owns it?)
- Causing harm (who is at fault?)
- Privacy and consent (facial recognition, tagging)
- AI can now predict the propensity to pay (are these patients treated differently?)
- AI can now predict the propensity to commit crimes (should it be for employment screening?)
- Security (how do we keep AI safe from adversaries?) (threat monitoring)
- Limitation (requires access to the cloud and enormous amounts of data)
- AAO Guidance:
  - AI must complement, not replace clinician judgement
  - Prioritize safety, equity and continuous monitoring





## Real World Examples of AI Happening Now...



Examples are actual health systems  
using AI



# AI and Tech Trends in Healthcare



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- Precision Medicine
  - Digital Health
  - Care Anywhere
  - Consumerism
  - Integration of individual healthcare delivery into population/public health functions
  - Data Analytics and Artificial Intelligence
  - Interoperability
  - Cloud Computing
  - Centralized models of clinical decision support
  - 21 Century Cures Act

# AI in Cybersecurity

Vulnerability Scans

3

2

Insider Threat Monitoring

4

Chat-Bot  
Managed Call Centers

1

Self-Health Networks

IT Support

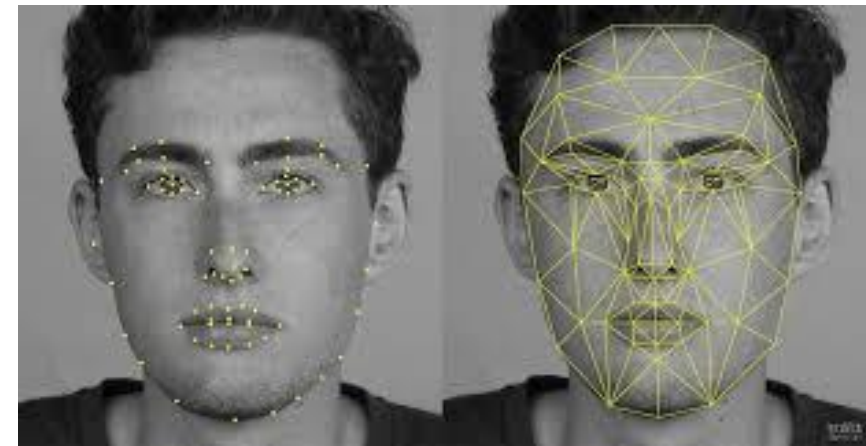
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Self-Service Help  
Desk and Ticketing System



# Facial Recognition

- Detection of non-employees interacting with PHI or entering into secure areas
- Advanced knowledge of patients entering the building who may require special services (Wheelchair, translator)
- Triggers check in processes
- Alerting the campus of a hostile/threatening patient
- Active shooter/terrorist in the area





# Leveraging AI Electronic Patient Communications





Negotiate an IT  
Contract Like a Pro

## Strength in Numbers



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# Knowledge Sharing

## Vendor Demonstration Tools

- Score Cards
- Demo Scenarios
- Scribe my exam

## Specialty-specific RFPs

## Reference Check Tools

## Site Visit Tools

## Tools for Comparing Cost, Including Recurring Cost

## HCIT User Conferences/Summits/Forums

## HCIT Workshops in Conjunction with Annual Meeting

## Formal Strategies

## Displacement Guidance

## Hospital Alignment/Community Connect



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## Ways to Negotiate

- Initial costs
- Hardware cost
- Software cost
- Communications cost
- Installation cost
- Ongoing support cost
- Implementation cost
- Support cost
- Technical support cost
- Integration costs
- Interface cost
- Entitlement to new releases/bug fixes
- Cost of tailoring
- Future upgrades and releases (this should always be at no additional cost)



# Modifying the Contract



- Source code
- Acceptance period (hardware and software)
- Implementation caveats
- No front loading of support fees
- No front-loading the purchase terms
- Assignment
- Future upgrades and new releases
- Copyright infringements
- Warranties
- Termination
- Future providers and fees (recurring cost)

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## Emerging Trends



Expansion from detection to prediction (e.g. risk modeling for disease progression)



Integration with wearable and mobile diagnostic tools



Increased use in screening programs and teleophthalmology

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## Personalized medicine



AI enables individualized treatment plans by analyzing multimodal data



Combines genetic, imaging, and clinical history for personalized insights



Potential to optimize treatment efficacy and reduce adverse outcomes



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## Global impact and accessibility



AI facilitates eye care delivery in remote and underserved regions



Portable AI devices allow frontline workers to screen for vision-threatening diseases



Democratization of care through scalable AI platforms

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## Integration with other technologies



Synergy with robotics, augmented reality (AR), and virtual reality (VR)



Real-time intraoperative decision support using AI



Fusion of AI with EHRs for holistic patient care management

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## Challenges and considerations



Ethical concerns: bias, transparency, and data privacy



Regulatory barriers to clinical deployment and approval



Need for robust validation and clinician oversight

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## Research directions



Development of explainable AI (XAI) for clinical interpretability



Training models on large, diverse, and labeled datasets



Exploring AI's role in predicting systemic disease via ocular imaging



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## Conclusion

The future of AI in ophthalmology is promising but requires careful guidance

Interdisciplinary collaboration is vital to ensure safety, efficacy, and equity

AI is not replacing clinicians—it is enhancing their capabilities



# Final Thoughts

# THE BLACK SWAN

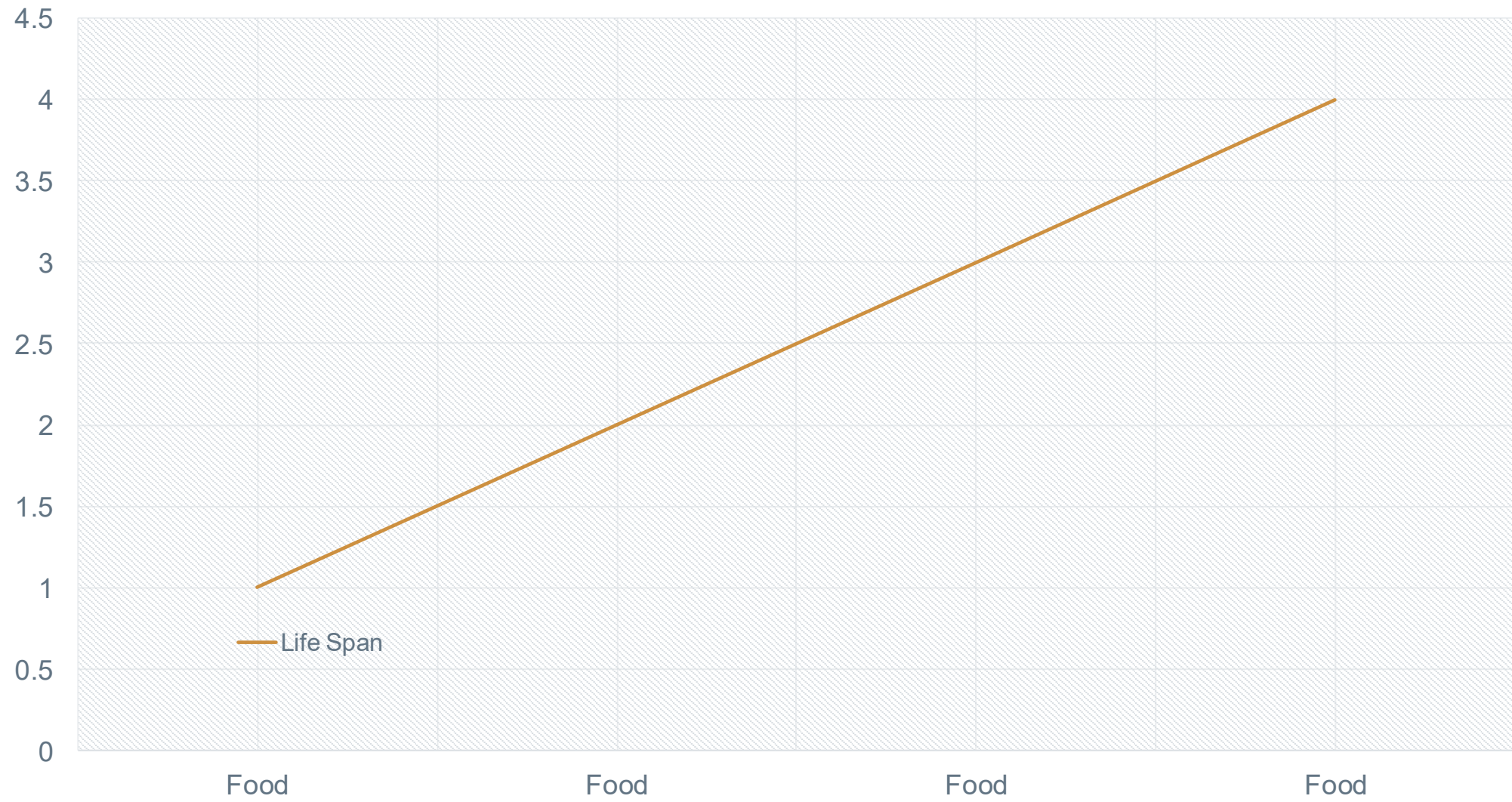


The Impact of the  
HIGHLY IMPROBABLE

Nassim Nicholas Taleb



## Life Span







# Questions?





Thank You