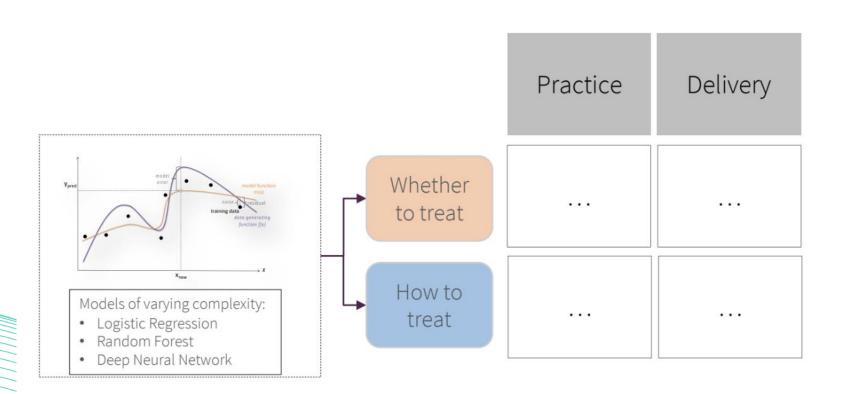




Building Responsible Al that Serves All of Us

Predictive AI helps us risk stratify and make recommendations to improve the delivery of care with responsive action



- Predicting long term outcomes after pulmonary embolism using imaging and EHR data
- Multimodal models for predicting recurrence risk in surgically resectable colorectal cancer
- Opportunistic ASCVD risk estimation, using CT images and EMR data



- Predicting mortality to improve advance care planning
- Classifying presence of undiagnosed disease
 - Familial hypercholesterolemia to order sequencing
 - Peripheral artery disease to order ABI measurement
- · Classifying ischemic vs. hemorrhagic stroke for prioritizing air ambulance transport
- Predicting no-shows for providing transportation support
- Predicting length of stay, readmissions, bed-demand etc. ...





And we are seeing the Generative Al Landscape evolving rapidly to meet demand from health systems, payors, & life science companies

Where Generative Al Meets Healthcare: Updating The Healthcare Al Landscape Life Sciences, \$6.5B Raised **Clinical Trials** Medicine & Patient-Facing. Navigation Services Base Genomics Syapse. Discovery S MEDIDATA \$2.3B Raised **babylon** Curai Health AETION emoneert **TEMPUS** INCEPTIVE DEEP 6 A uxa Health м monogram health ConcertA Atomwise Remote Patient biofourmis Care Woebot Health Admin, \$2.7B Raised **@tyto**care Monitoring Al Suite Operations Analytics & IT, **C**are.coach Recruiting Athelas current health Artisight Hippocratic A \$2.7B Raised notable Ölér Health IntelyCare ClosedLoop.ai MEMORA HEALTH Reverence winnow. Ferrum Clinician-Facing, Decision \$6.0B Raised Support regard Prior Auth pieces cohere **kahun N** navina Robotics INTUITIVE W BUDDI AI * patient |Q evidation myndshft 🔮 vion ARTERYS aidoc FATHOM **Clarify** banj prognos **APIXIO** © GLASS Develop Health ibM Watson Health **CODAMETRIX** innovaccer **Rhyme** I VERB **PTT** - CARTA ethermed **ด**ั Diagnostics hume PathAl PathAl Arintra Notetaking Knowtex **KINTSUGI Butterfly** Security & Gentem DIGITAL eleos ROBIN Center Compliance SmarterD Operations Ambience PRIVATE AI **∴** PROSCIA © CODOXO Tapero elements Syntrillo ... **₩ Deep**Scribe

Centaur Labs

Segmed

Workflow

SIRONA RamSoft

Care Studio

Syllable

INFINITUS

Birch AI

meMR

ScienceIO

📚 cascade

UNLEARN • Clinithink

RIALTIC

candidhealth

Anomaly

Adonis

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NOTE

AUGMEDIX

NABLA

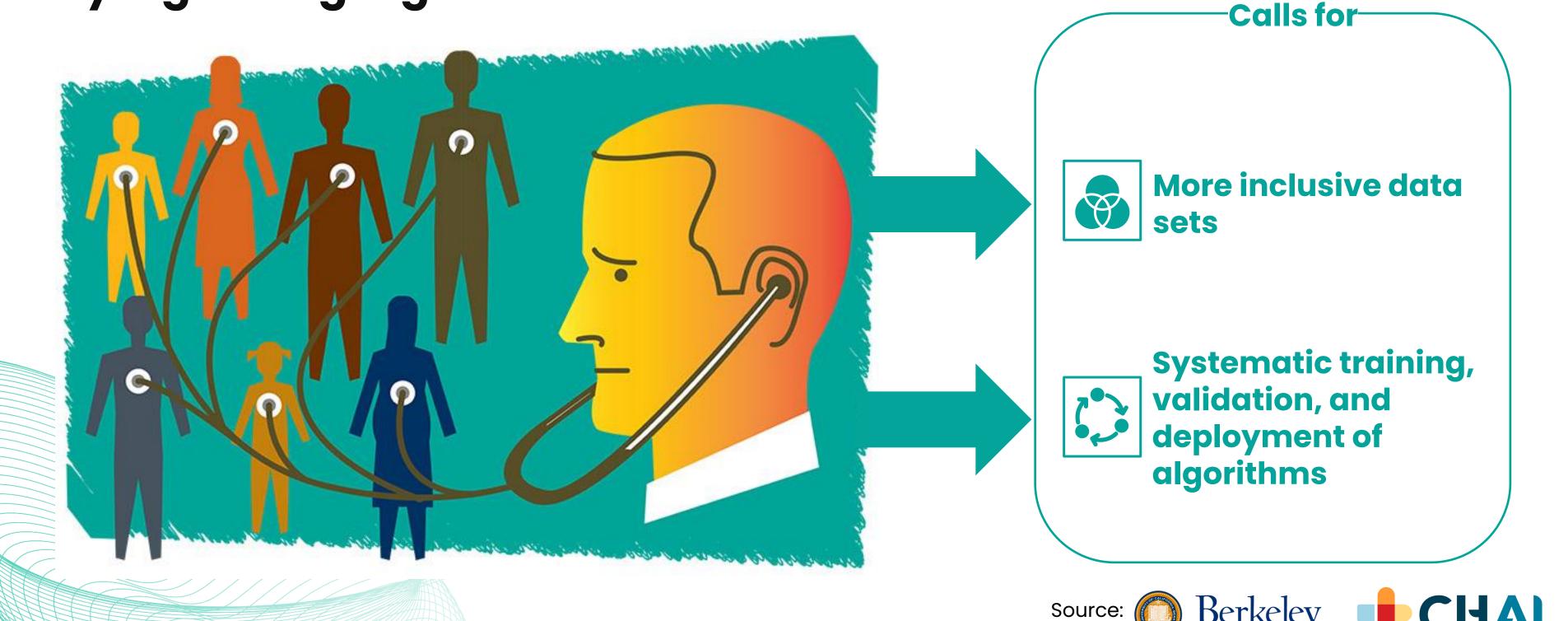
NUANCE

Research

ATROPOSHEALTH

scite_

The track record of new technologies in healthcare is one marked by a growing digital divide



CHAI is setting the best practice frameworks for Al safety and reliability in Health

CHAI, the Coalition for Health AI, is committed to setting the best practice frameworks for the trustworthy and responsible implementation of AI in health.

By bringing together a coalition of patient-community advocates, technology companies start-ups, public sector organizations, medical device manufacturers, payors and healthcare organizations, we will develop a consensus set of frameworks for Responsible Health Al.

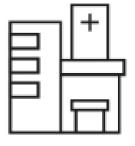


CHAI is bringing together a diverse community of creators, users and beneficiaries to develop consensus-driven products





+







Creators of Al

Health Systems

Medical Device manufacturers
Pharmaceutical Companies
Software Creators for health
Hardware Creators for health
& more

Users of Al

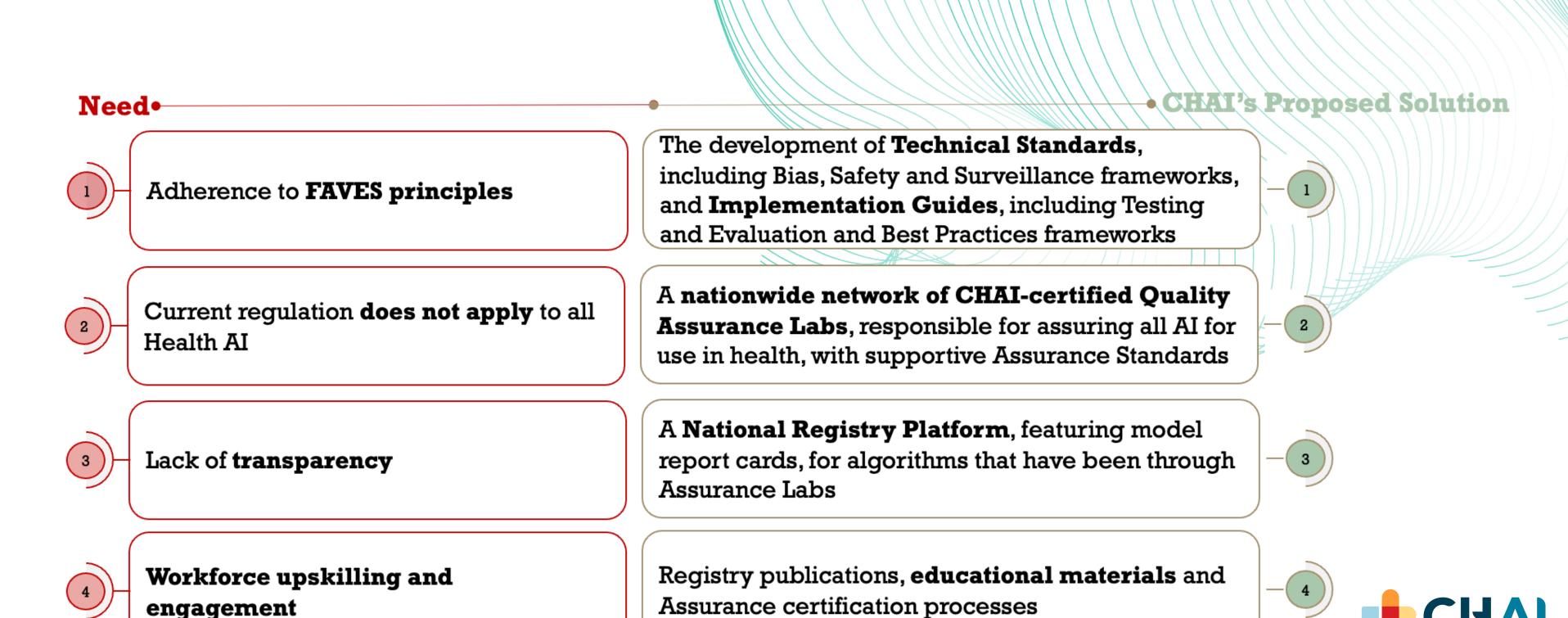
Health Systems
Life Science
Companies
Physicians
Payors
CMS

Patient Consumers of Al

Patient Advocacy Groups



We are using consensus to develop the gold-standard for trustworthy and Responsible Health Al that Serves All



THE AI LIFECYCLE

The Al lifecycle is central to understanding and implementing CHAl's Assurance Standards in healthcare. The six-step lifecycle outlines the essential stages and processes involved in developing, deploying, and maintaining Al systems.

By systematically addressing each phase of the lifecycle the framework ensures that AI systems adhere to the highest standards of safety, efficacy, fairness, transparency, and security. This structured approach supports risk mitigation, managing biases, and promotes accountability and trustworthiness in AI applications.





- Engage stakeholders to define the problem & perform root-cause analysis
- Identify solution & plan future state
- Gather business requirements
- Assess feasibility, potential for impact, & prioritization
- Make procure/build/ partner decision

- 2
- Select/understand model task & architecture
 Capture design &
 - technical requirements or determine best solution to meet business requirements
 - Design solution application & system workflow according to human-centered design principles
 - Design deployment strategy with end users
 - Design risk management, monitoring & reporting plan



- Access data
 Prepare data
- Develop data
 management plan
 Train 6 tuna madel
- Train & tune model
- 4
- Conduct installation qualification (when applicable)
- Validate local system performance (when applicable)
- Execute prospective, silent evaluation
- Establish risk
- management plan

 Train end users
- Test usefulness
- Ensure compliance with applicable healthcare regulations & standards



- Implement small-scale pilot to assess real-world impact
- Execute and update risk management plan
- Educate & train users on Al application reporting



- Deploy at a larger scale on a general population
- Audit Al system to inform whether to maintain, refine or sunset
- Conduct ongoing risk management

Generative Al Work Group Goal

Goal: Focusing on a Generative Al Notes Summarization Use Case, we seek to identify a digestible set of criteria and metrics that demonstrate principles of responsible and trustworthy Al, which can be utilized or referenced by an independent Al solution evaluator.









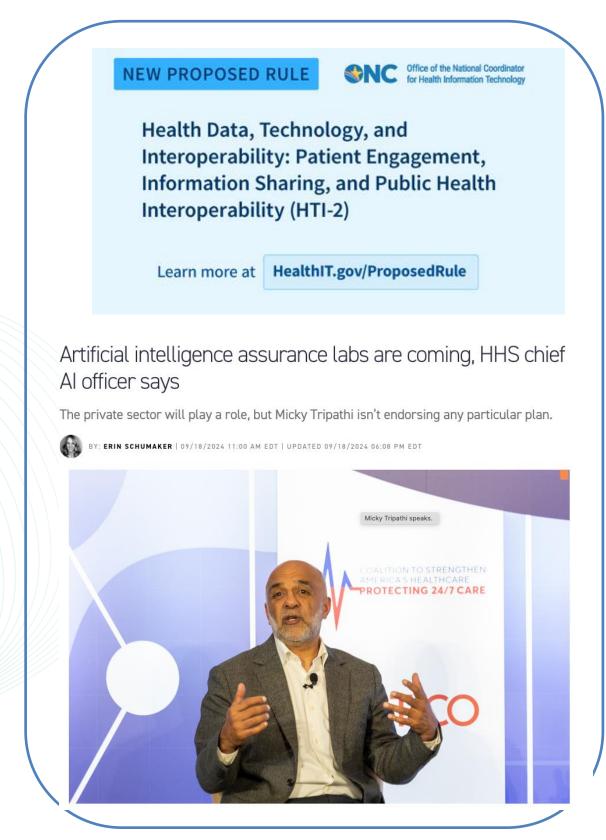


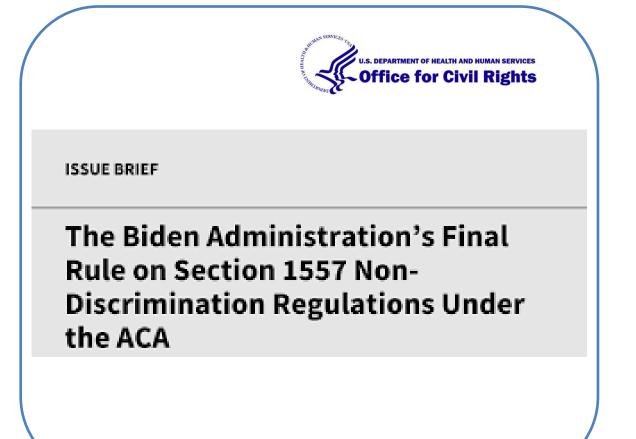
Transparency

Meanwhile, the regulatory landscape is similarly changing

rapidly

FDA Issues Final Guidance on Clinical
Decision Support Software
and Software as a Medical
Device









CHAI is supporting the development of anationwide network of Assurance Labs with its certification framework





A Federated Network of Labs - All Models are Local













Assurance Labs need to be trusted by both model developers, their customers and by patients



Access to data



Required disclosures on impartiality – ISO standards



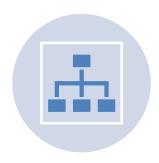
Security Certification
SOC II Certification



Data de-identification or Privacy Preserving Approaches



Infrastructure



Established and well-documented processes



Demonstrated competence in validation methods



Use of CHAI Metrics



Post-deployment Monitoring



An Al "Nutrition Label": CHAI Consensus Model Cards



Model Card - Title

Model Details

- Developers
- Model Date, Version & Type
- Training algorithms
- Resources, Citation, License

Evaluation Data

- Details on data used for quantitative analysis
- Datasets, Motivation, Preprocessing

Training Data

- Same detail as evaluation data if possible (privacy constraints)
- Details of distribution over factors

Intended Use

- Primary intended uses & users
- Out of scope use cases

Quantitative Analysis

Unitary & intersectional results





Factors

- Groups, Environments, Instrumentation
- Relevant factors & evaluation factors

Metrics

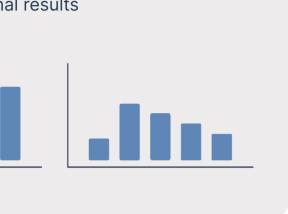
- Model performance measures
- Decision thresholds
- Variation approaches

Ethical Considerations

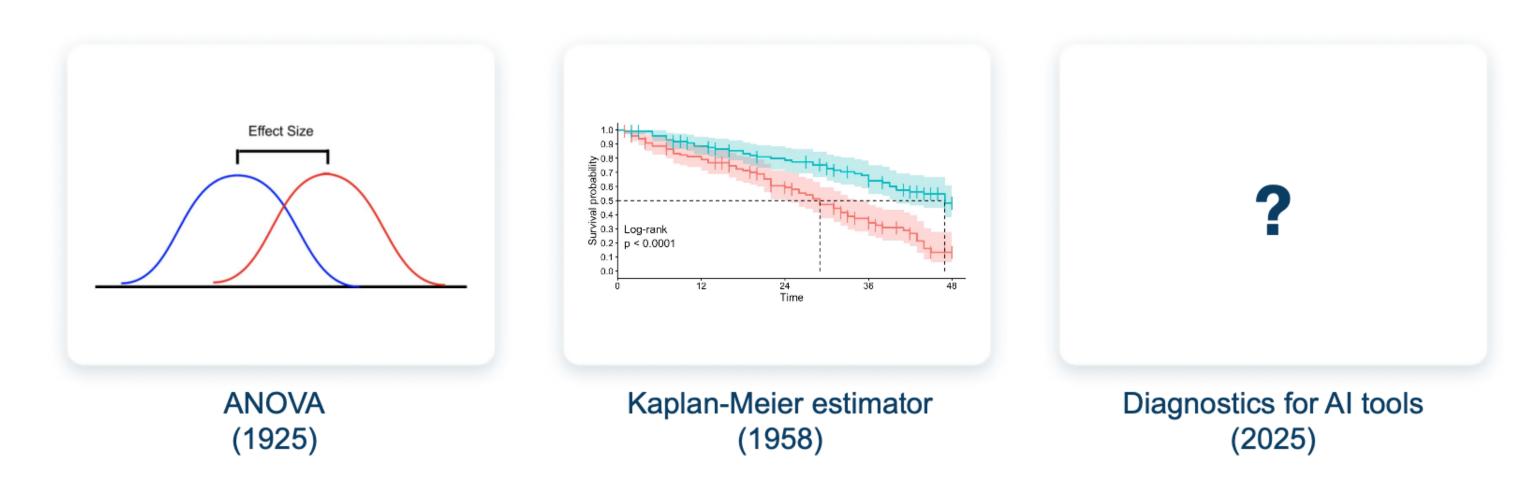
- Bias, fairness, ethical considerations
- Mitigation efforts

Caveats, Recommendations

- Concerns not already covered
- Usage information
- · Limitations, risks, trade-offs



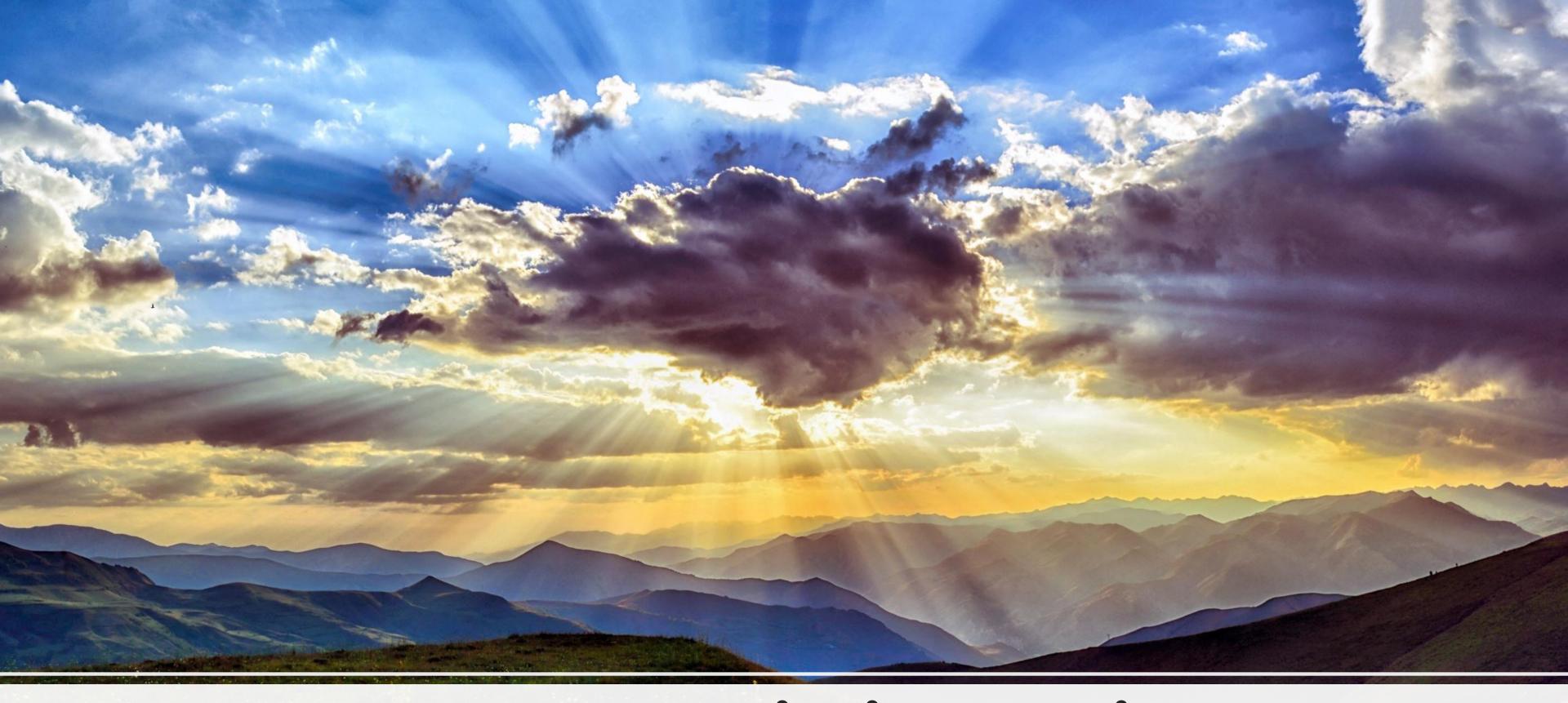
Need for Advancing Evaluation and Regulatory Science in Al



What will the the next generation of statistical tools applied to medicine look like?







Is Human Flourishing Possible?

The story of the humble sewing machine





An Urgent Need to Develop Consensus Guidelines for GenAl