

# Lecture 05 - Contents

An overview of the parts in the medical model evaluation lecture.

## Part 1

Medical Benchmarks

## Part 2

Evaluation Metrics

## Part 3

Human Evaluation

## Hands-on

Evaluation Hands-on

This outline is for guidance. Navigate the slides with the left/right arrow keys.



Lecture 5:

# Evaluating Medical LLMs: Metrics That Matter

Evaluation Metrics and Medical Benchmarks

# Evaluation Framework Overview



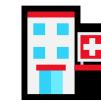
## Automated Evaluation

Metrics computed automatically on benchmark datasets



## Human Evaluation

Expert clinicians assess clinical relevance and safety



## Medical-Specific

Domain-specific metrics for healthcare applications

## Key Evaluation Categories

- Accuracy & Factuality
- Safety Assessment
- Consistency Measures
- Clinical Relevance
- Hallucination Detection
- Bias & Fairness

## Part 1/3:

# Medical Benchmark Datasets

- MedQA - USMLE-style medical questions
- USMLE Step Exams - Clinical competency assessment
- PubMedQA - Biomedical literature comprehension
- MedMCQA - Multi-specialty medical questions
- MMLU Medical - Multidisciplinary medical knowledge
- Clinical Case Benchmarks - Real-world scenarios

# MedQA Dataset Analysis

**11,450**

Total Questions

**4-Choice**

Answer Format

**2 Lang**

EN / ZH

**USMLE**

Style

## Key Features



Clinical vignette-based questions



Multi-step reasoning required



Bilingual support (English/Chinese)



Difficulty levels from easy to expert

Human Expert Performance: ~87% | GPT-4: ~78% | Random Baseline: 25%

# USMLE Step Exams

## Step 1

### Basic Sciences

Foundational medical knowledge,  
pathology, pharmacology

## Step 2

### Clinical Knowledge

Clinical reasoning, diagnosis,  
patient management

## Step 3

### Practice Skills

Independent practice, emergency  
care, longitudinal care

### Passing Threshold

Minimum score: ~60% correct | Human physicians: 85-95% pass rate



Med-PaLM 2: 86.5% on USMLE-style questions | 🧑 Expert consensus level achieved

# PubMedQA: Literature Comprehension

**273K**

Questions

**3-Way**

Yes/No/Maybe

**PubMed**

Source

**Expert**

Annotated

## Key Characteristics



Scientific literature understanding



Evidence-based reasoning



Abstract-based question answering



Requires nuanced interpretation

Challenges: Context understanding, Uncertain answers (Maybe), Clinical implications

# MedMCQA: Multi-Specialty Questions

**194K**

Questions

**21**

Subjects

**4-Choice**

MCQ Format

**India**

NEET/AIIMS

## Subject Coverage

Anatomy, Physiology, Pathology



Pharmacology, Biochemistry

Medicine, Surgery, Pediatrics



Microbiology, Forensics, Community Med

Detailed explanations provided | Difficulty histogram available | Cross-specialty evaluation

# MMLU Medical Subset

**1,089**

Questions

**6**

Medical Topics

**College**

Level

**4-Choice**

Format

## Medical Topics



Clinical Knowledge



Medical Genetics



Professional Medicine



College Medicine & Biology

Multidisciplinary approach | Zero-shot evaluation | Error analysis available

# Clinical Case Benchmarks

## Real-World Clinical Scenarios



Patient history and presentation



Diagnostic test ordering



Differential diagnosis generation



Treatment plan development

## Complexity Levels



Simple: Single diagnosis, clear presentation



Complex: Comorbidities, atypical presentation



Moderate: Multiple symptoms, common conditions



Expert: Rare diseases, multi-step reasoning

Evaluation: Diagnostic accuracy | Reasoning process | Treatment appropriateness

# Multilingual Medical Evaluation

10+

Languages

Global

Coverage

Cultural

Adaptation

Regional

Practices

## Evaluation Challenges



Translation quality and consistency



Cultural healthcare differences



Regional medical terminology



Performance variation across languages

Languages: English, Chinese, Spanish, French, German, Hindi, Arabic, Japanese, Korean, Portuguese

**Part 2/3:**

## **Medical-Specific Metrics**

- Clinical utility vs. accuracy tradeoffs
- Hallucination detection and factuality
- Consistency and uncertainty quantification
- Calibration and confidence scoring

# Accuracy vs Clinical Utility



## The Accuracy Paradox

- Misses rare but critical conditions
- Provides correct but irrelevant info
- Lacks actionable recommendations
- Ignores patient context



### Actionability

Clear, implementable recommendations



### Impact

Influences clinical decision-making



### Patient-Centered

Considers patient outcomes



### Treatment Quality

Improves care delivery

High Accuracy

95%

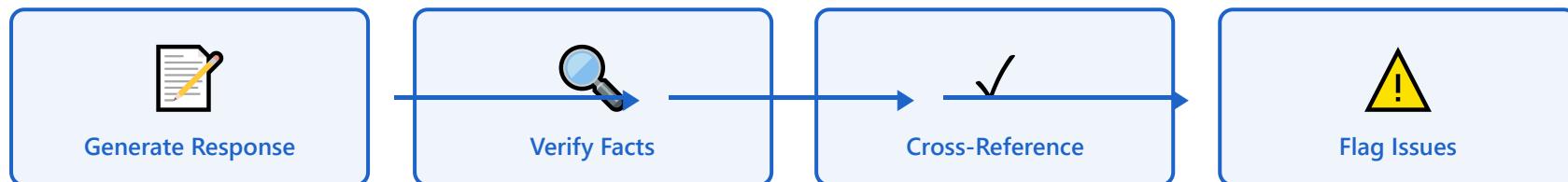


Clinical Utility

?

Key principle: Optimize for clinical impact, not just test scores

# Hallucination Detection



## Types of Medical Hallucinations



Fabricated medical facts



Incorrect statistics or dosages



Non-existent drugs or treatments



Contradictory information

CRITICAL

HIGH

MEDIUM

LOW

Detection Methods: Medical DB Cross-reference | Expert Validation | Citation Check | Automated Fact Verification

# Factuality Scoring

## Factuality Assessment Framework

 Evidence-based verification

 Citation traceability

 Source credibility ranking

 Internal consistency checks

Level 1

Peer-reviewed Research

Level 2

Clinical Guidelines

Level 3

Medical Textbooks

Level 4

Expert Opinion

0

2

5

7

10

Scoring: 0 (unverifiable) → 5 (single source) → 10 (multiple high-quality sources)

# Consistency Measures



Response over time



Paraphrase robustness



Multi-query consistency



Temperature sensitivity

## Testing Methodology

- 1 Ask same question 10+ times
- 2 Rephrase in different ways
- 3 Vary context and details
- 4 Measure agreement rate
- 5 Identify variation sources
- 6 Analyze error patterns

0%

50%

90%

100%



Target: >90% consistency for critical clinical decisions

# Uncertainty Quantification



## Aleatoric Uncertainty

Inherent randomness in the data that cannot be reduced with more training



## Epistemic Uncertainty

Model knowledge gaps that can be reduced with more data or better training

### Clinical Confidence Thresholds

High (>90%)

Proceed with recommendation

Medium (70-90%)

Flag for review

Low (<70%)

Require expert consultation

Very Low (<50%)

Decline to answer

"I don't know" is a valid and valuable medical response

# Calibration Metrics

**ECE**

Expected Calibration Error

**MCE**

Maximum Calibration Error

**Brier**

Brier Score

**T-Scale**

Temperature Scaling

## Calibration Process

Collect Predictions

Apply Scaling

Measure ECE

Validate

Perfect Calibration

Well-calibrated: When model says 80% confident, it's correct 80% of the time

## **Part 3/3:**

# **Human Expert Evaluation**

- Expert annotation protocols and guidelines
- Inter-rater agreement and quality control
- Clinical relevance and safety assessment
- Bias, fairness, and robustness testing
- Continuous monitoring and reporting

# Expert Annotation Protocols

1



## Training & Calibration

Experts complete mandatory training sessions

2



## Apply Clear Criteria

Use standardized evaluation rubrics

3



## Multi-Expert Review

Independent assessment by  $\geq 3$  experts

4



## Quality Validation

Regular audits and gold standard checks

## Quality Control Cycle

Calibration Phase



Regular Audits



Feedback Loops



Refinement

Target:  $\geq 3$  expert annotators per case with specialty-matched expertise

# Inter-Rater Agreement

## Cohen's $\kappa$

Two raters

## Fleiss' $\kappa$

Multiple raters

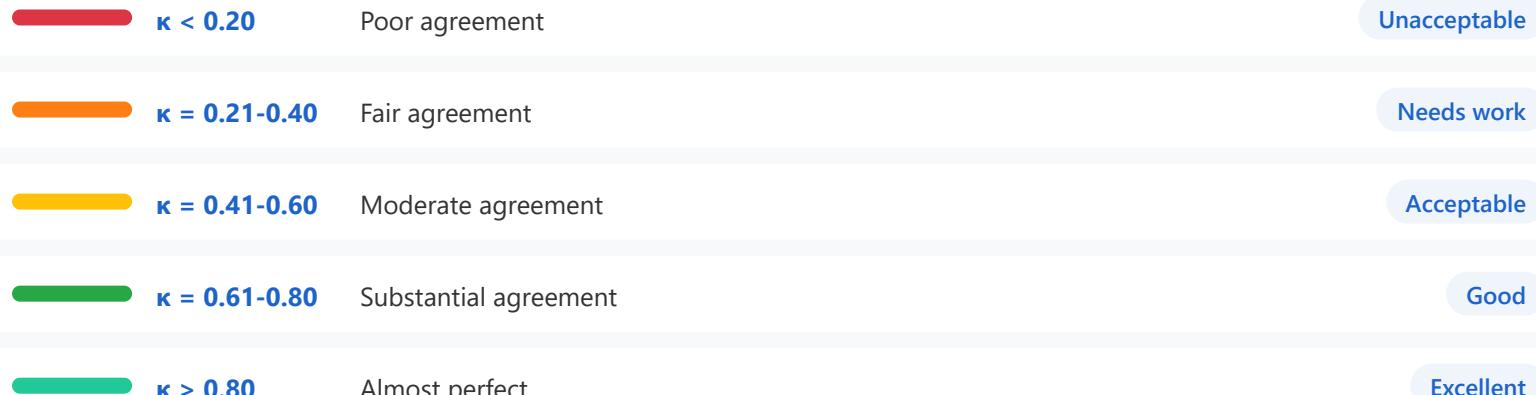
## ICC

Continuous scores

## Krippendorff's $\alpha$

Any data type

### Kappa Interpretation Scale



Disagreement Detected



Consensus Discussion



Third Expert

Disagreement resolution: Consensus discussion or third expert adjudication

# Blind Evaluation Setup



## Single-Blind

Evaluators unaware of model identity



## Double-Blind

Both evaluators and data collectors blinded

### Blind Evaluation Process



Randomize



Anonymize



Evaluate



Reveal & Analyze



#### Positive Controls

Known good responses



#### Negative Controls

Known poor responses



#### Baseline Comparison

Compare to human experts



#### Neutral Controls

Ambiguous cases

Minimize bias: Anonymize responses, randomize order, use control cases

# Clinical Relevance Scoring

## Relevance Dimensions



Diagnostic accuracy and reasoning



Urgency and triage accuracy



Treatment appropriateness



Patient-centered communication

### Score 1

Harmful

### Score 2-3

Not helpful

### Score 4-6

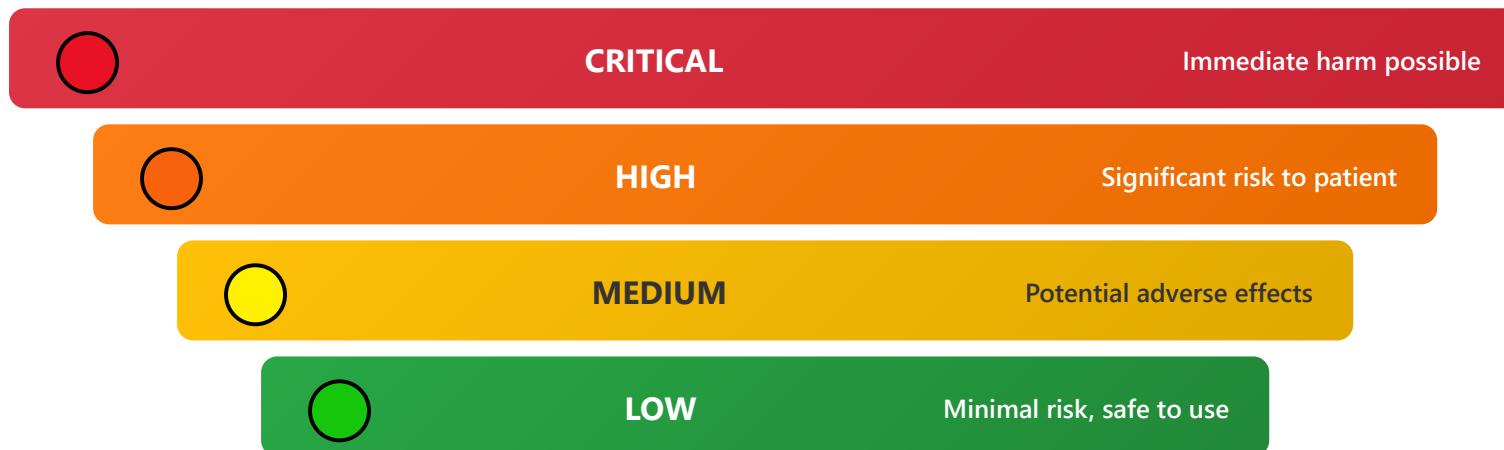
Somewhat useful

### Score 7-10

Highly valuable

Focus: Would this information improve patient care and outcomes?

# Safety Assessment Framework



## Safety Checklist

- ✓ Drug interactions check
- ✓ Diagnostic error screening
- ✓ Privacy & confidentiality

- ✓ Dosage accuracy verification
- ✓ Treatment appropriateness
- ✓ Contraindications review



**Zero tolerance: Any critical safety issue = immediate model review**



# Bias & Fairness Testing

## Bias Detection Areas



Demographic bias (age, sex, race)



Disability and accessibility bias



Socioeconomic and geographic bias



Language and cultural bias

## Fairness Metrics

- **Demographic parity:** Equal outcomes across groups
- **Equalized odds:** Equal TPR and FPR across groups
- **Calibration:** Predictions equally accurate for all groups
- **Individual fairness:** Similar individuals treated similarly

Test across diverse populations, especially vulnerable and underserved groups

# Robustness Evaluation



## Noisy Inputs

Test with incomplete or corrupted data



## Paraphrasing

Rephrase queries in different ways



## Domain Shift

Test on out-of-distribution data



## Adversarial Cases

Challenge with edge cases

### Robustness Metrics



#### Performance degradation:

% accuracy drop under noise



#### Input sensitivity:

Response stability to variations



#### Failure mode analysis:

When and why it breaks



#### Recovery ability:

How well it handles errors

0%

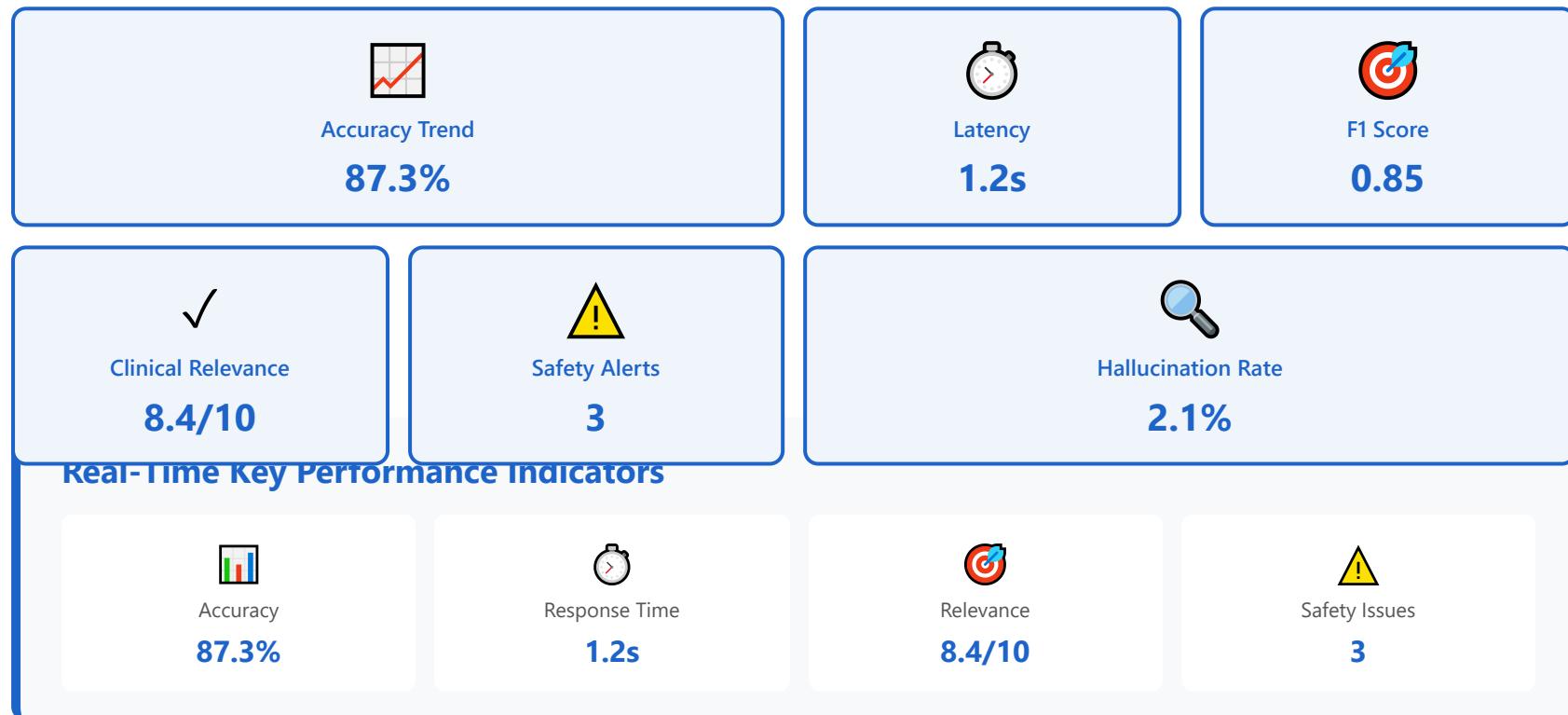


10%

20%+

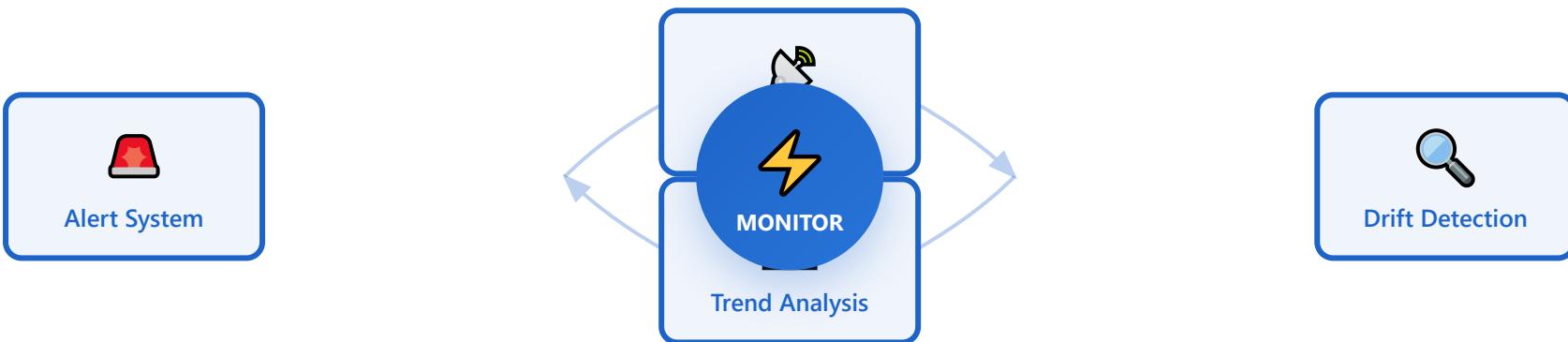
Target: <10% performance drop under typical real-world variations

# Performance Dashboards



Features: Real-time monitoring | Alert system | Trend analysis | Drill-down investigation

# Continuous Monitoring



## Intervention Triggers

**Accuracy drop >5%:** Investigate

**Drift detected:** Retrain model

**Safety incident:** Immediate review

**Unusual patterns:** Expert review

Regular audits: Weekly automated reports + Monthly expert review

# Hands-on: Evaluation Implementation

1

## Load Benchmark Dataset

→ MedQA, PubMedQA, or USMLE-style questions



2

## Run Model Inference

→ Generate predictions with confidence scores



3

## Calculate Metrics

→ Accuracy, F1, ECE, hallucination rate



4

## Visualize Results

→ Confusion matrices, calibration plots, error analysis



5

## Generate Report

→ Comprehensive evaluation summary

## Recommended Tools

[HuggingFace Evaluate](#)[TorchMetrics](#)[scikit-learn](#)[Weights & Biases](#)

# Reporting Standards

## Standard Reporting Guidelines



TRIPOD: Prediction models



STARD: Diagnostic accuracy



CONSORT-AI: Clinical trials



PRISMA: Systematic reviews

## Essential Documentation

- Model architecture and training details
- Datasets used (size, source, preprocessing)
- Evaluation metrics and benchmarks
- Performance results with confidence intervals
- Limitations and failure modes
- Ethical considerations and bias analysis

Goal: Transparency, reproducibility, and responsible AI deployment

# Thank you

## Key Takeaways

- ✓ Medical benchmarks: MedQA, USMLE, PubMedQA, MedMCQA, MMLU
  - ✓ Metrics: Accuracy, factuality, consistency, calibration, safety
- ✓ Human evaluation: Expert protocols, inter-rater agreement, clinical relevance
- ✓ Continuous monitoring: Dashboards, drift detection, reporting standards



Resources: [Papers with Code](#) | [Medical AI Benchmarks](#) | [TRIPOD Guidelines](#)