

# Lecture 01 - Contents

An overview of the main sections in this lecture.

## Part 1

Medical AI Revolution

## Part 2

State-of-the-Art Models

## Part 3

Real-World Deployments

## Hands-on

Environment Setup and Assignment

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



# **Advanced Medical LLMs: Transforming Healthcare with AI**

20-Lecture Comprehensive Course

Introduction to Biomedical Data Science

**Instructor Name**

Medical AI Research Center

Fall 2025

# Course Overview and Prerequisites

## Learning Objectives

- Understand medical LLM architectures
- Master clinical NLP techniques
- Deploy healthcare AI systems
- Ensure HIPAA compliance
- Evaluate model performance

## Prerequisites

- Python programming
- Machine learning basics
- Medical terminology
- Neural networks fundamentals

PyTorch

Hugging Face

FHIR

Docker

## Assessment Breakdown

**40%**

Project

**30%**

Assignments

**30%**

Exam

## Course Structure

- 20 comprehensive lectures
- Hands-on coding sessions
- Real-world case studies
- Industry expert guest lectures

**Part 1/3:**

# **The Medical AI Revolution**

1. Evolution from Rule-Based Systems to LLMs
2. GPT-4, Claude, and Gemini in Healthcare
3. Medical vs General-Purpose LLMs

# Evolution From Rule-Based Systems to LLMs

## Rule-Based

1970s - 1990s



### Technology:

Expert systems  
Decision trees  
If-then rules

### Limitation:

*Cannot handle  
uncertainty*

## Statistical ML

2000s - 2010s



### Technology:

SVM, Random  
Forest, Naive  
Bayes

### Limitation:

*Feature engineering  
required*

## Deep Learning

2010s - 2020



### Technology:

CNN, RNN  
LSTM, Attention  
mechanisms

### Limitation:

*Task-specific  
Large labeled data*

## Large LLMs

2020s - Present

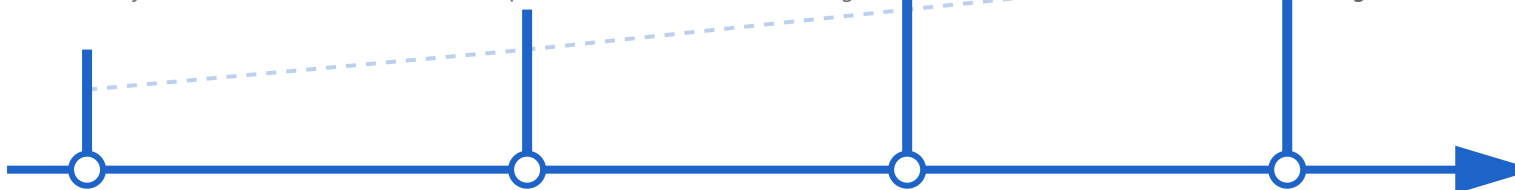


### Technology:

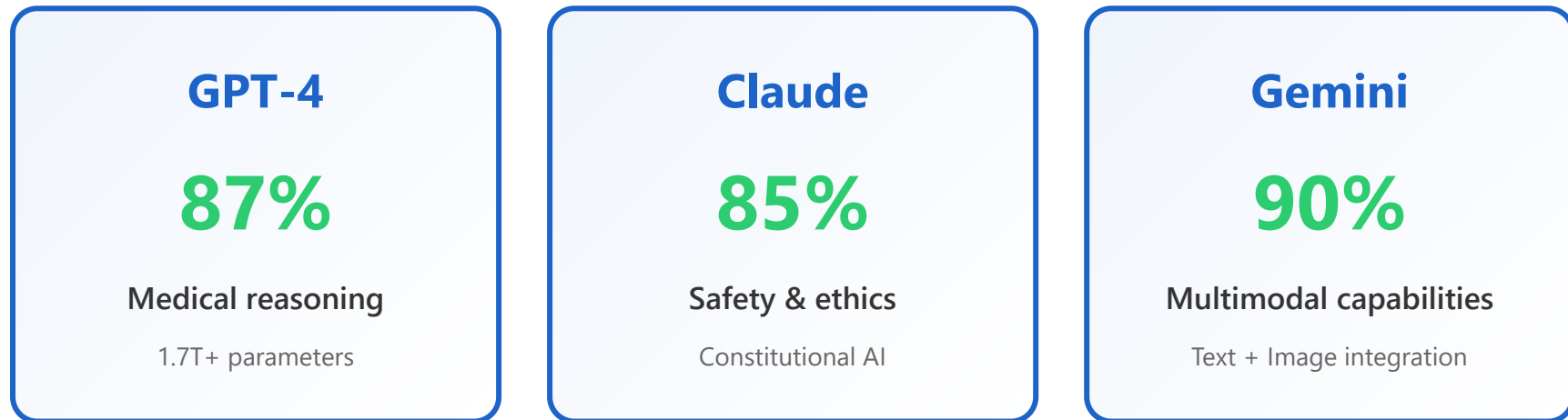
Transformers  
Self-attention  
Pre-train + Fine-tune

### Strength:

*Few-shot learning  
Broad knowledge*



## GPT-4, Claude, and Gemini in Healthcare



# Medical vs General-Purpose LLMs



## Medical LLMs

- ✓ Domain-specific knowledge
- ✓ Clinical safety protocols
- ✓ HIPAA/regulatory compliance
- ✓ Medical terminology precision
- ✓ Evidence-based responses
- ✓ Rare disease understanding



## General-Purpose LLMs

- ✓ Broad knowledge coverage
- ✓ Creative problem-solving
- ✓ Multi-domain versatility
- ✓ General reasoning ability
- ✓ Larger training datasets
- ✓ Conversational flexibility



## Common Ground

Language understanding • Contextual reasoning • Pattern recognition • Natural language generation



# Key Medical AI Challenges



## Data Privacy

HIPAA compliance, patient data protection, secure data handling and storage

Critical Risk



## Medical Errors

Hallucinations, incorrect diagnoses, liability concerns, patient safety

High Risk



## Regulatory Approval

FDA clearance process, clinical validation, compliance documentation

High Complexity



## Explainability

Model interpretability, decision transparency, clinical reasoning clarity

Medium Risk



## Bias & Fairness

Demographic bias, training data representation, equitable healthcare access

High Risk



## Implementation Cost

Infrastructure investment, training costs, maintenance and updates

Medium Complexity

# HIPAA Privacy and Data Security



## 18 PHI Identifiers



Names



SSN



Addresses



Medical Records



Dates



Phone Numbers



Email



IP Address



Biometric IDs



Photos



## Security Requirements

- ✓ End-to-end encryption (AES-256)
- ✓ Access control & authentication
- ✓ Audit logging & monitoring
- ✓ Data de-identification
- ✓ Secure data transmission (TLS)
- ✓ Regular security assessments



## Compliance Checklist



Privacy Notice



Data Backup



Incident Response



Staff Training

Business Associate Agreements

Risk Analysis

# Clinical NLP Task Taxonomy



## Named Entity Recognition

EXTRACTION

Identify diseases, medications, symptoms, procedures



Patient has diabetes and takes metformin



## Relation Extraction

RELATION

Find relationships between medical entities



Drug X → causes → Side Effect Y



## Temporal Expression

TEMPORAL

Normalize time expressions and event sequences



"2 weeks ago" → 2024-10-30



## Negation Detection

LOGIC

Identify negated medical concepts



"No signs of infection" ≠ "Infection present"



## Medical Code Mapping

MAPPING

Map clinical text to ICD-10, CPT, SNOMED codes



"Hypertension" → ICD-10: I10



## Clinical Text Summarization

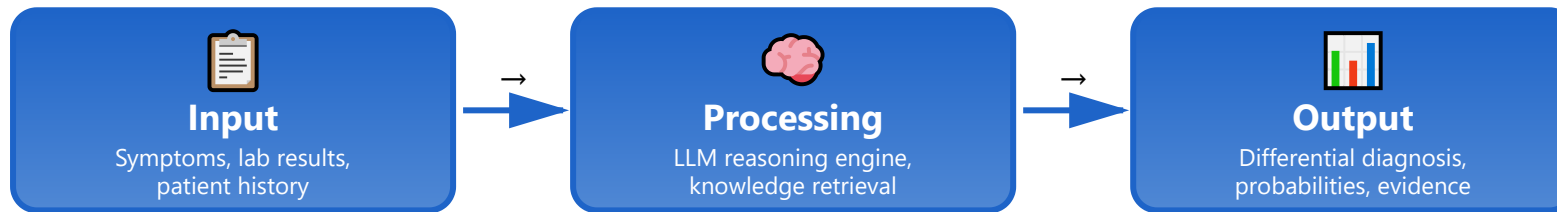
GENERATION

Generate concise summaries from clinical notes



Long note → Discharge summary

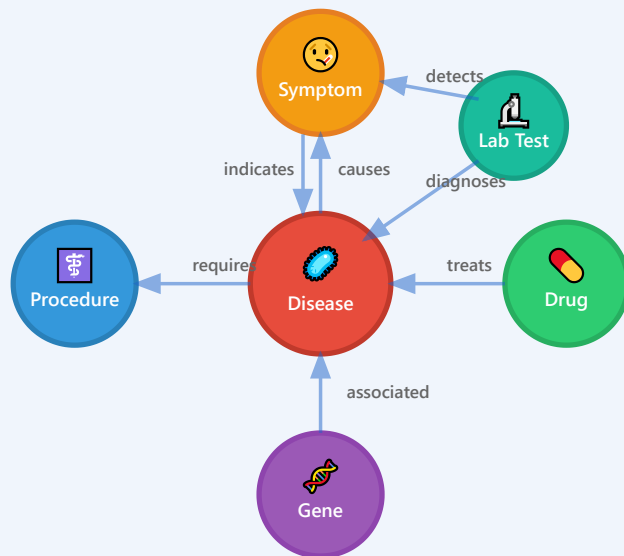
# Diagnosis Generation Systems



## Example Output: Differential Diagnosis

- Community-Acquired Pneumonia**  
Probability: 78% | Evidence: Fever, cough, chest X-ray findings
- Acute Bronchitis**  
Probability: 15% | Evidence: Cough, no infiltrates on imaging
- Viral Upper Respiratory Infection**  
Probability: 7% | Evidence: Mild symptoms, no bacterial markers

# Medical Knowledge Graph Integration



Triple Structure: Entity → Relation → Entity

**2.8M**

UMLS Concepts

**15M+**

Relationships

## Graph Embedding

- TransE, TransR for relations
- Node2Vec for embeddings
- Graph Attention Networks
- Knowledge Graph Embeddings

## LLM Integration Methods

Retrieval-Augmented Generation (RAG) • Knowledge-enhanced prompting • Graph-guided reasoning • Entity linking and normalization • Hybrid knowledge retrieval

Part 2/3:

# State-of-the-Art Medical Models



MedPaLM 2



BioGPT



RadBERT



Multimodal Systems



GatorTron



PubMedGPT



PathLLM

# MedPaLM 2 Architecture



540B parameters



Flan-PaLM based



86.5% on MedQA



Instruction fine-tuning on medical tasks



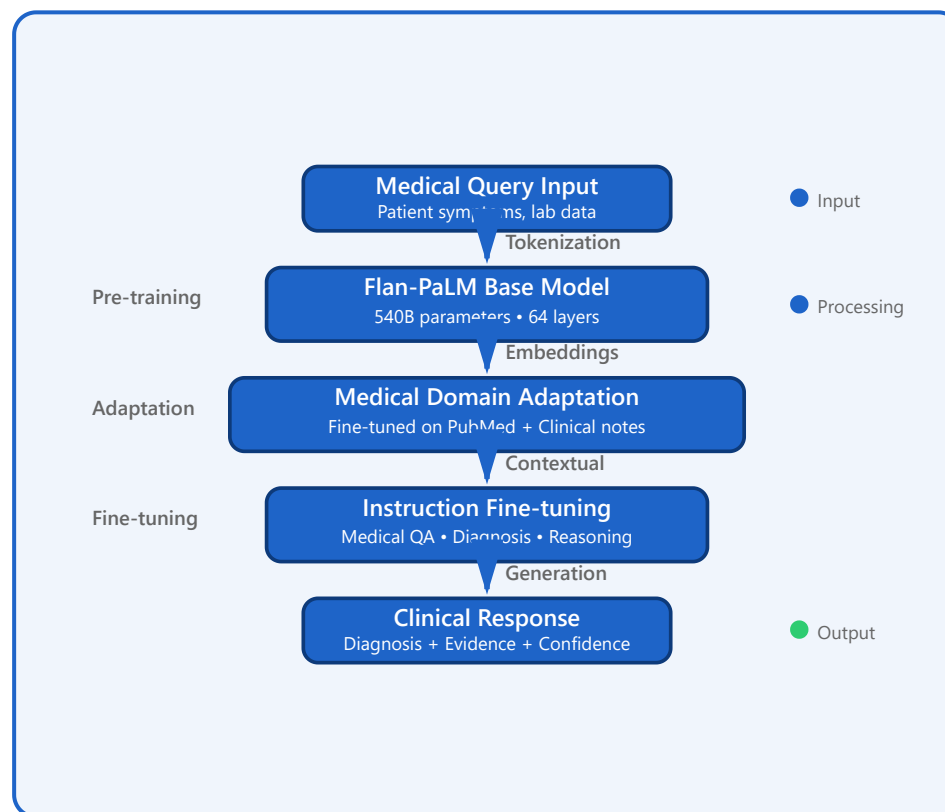
Medical domain adaptation with specialized corpus



Few-shot clinical reasoning capabilities



Safety-focused design with alignment





# GatorTron - Clinical BERT

## 8.9B Parameter Clinical Language Model

Trained on 90 billion words from UF Health clinical notes

### NER Performance

**96%**

F1 Score

### Relation Extract

**94%**

F1 Score

### Parameters

**8.9B**

Largest clinical

# BioGPT vs PubMedGPT Comparison

## BioGPT

1.5B Parameters

- Focus:** General biomedical text generation
- Training:** PubMed abstracts (15M documents)
- Strengths:** Question answering, summarization
- Use Case:** Research assistance, literature review
- Speed:** Faster inference time

## PubMedGPT

2.7B Parameters

- Focus:** Medical literature specialization
- Training:** Full PubMed papers (3M+ full text)
- Strengths:** Scientific writing, detailed analysis
- Use Case:** Literature review, paper generation
- Depth:** More comprehensive knowledge



## Performance Comparison



## **Reserved Slot (L01\_16)**

추후 내용이 추가될 자리입니다. 강의 흐름의 연속성을 위해 번호를 보존합니다.

# Multimodal Medical Models



## Text + Image

Radiology reports with X-rays, CT, MRI



## Text + Signal

ECG, EEG, waveform analysis



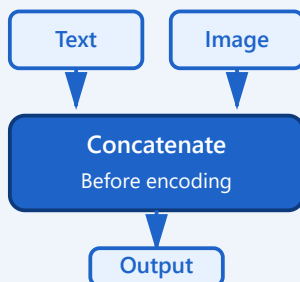
## Text + Video

Surgical procedures, endoscopy

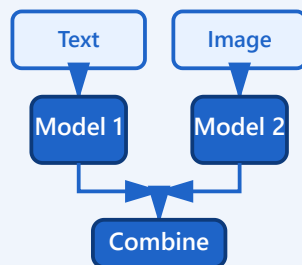


## Fusion Strategies

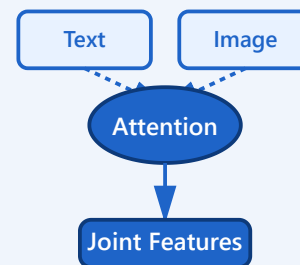
### 🌀 Early Fusion



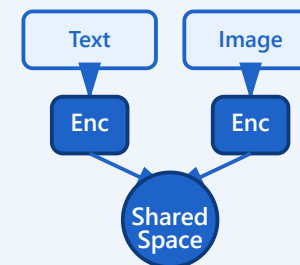
### 🕒 Late Fusion



### 🧠 Cross-Attention



### ⚡ Joint Space





# RadBERT

Radiology Applications



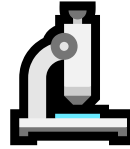
4.7M radiology reports

✓ Chest X-ray abnormality detection: 94% AUC

✓ Automated report generation

✓ Finding classification & localization

✓ Report quality assessment



# PathLLM

Digital Pathology



Whole Slide Image (WSI) processing

✓ Cancer grading & classification

✓ Cell counting & detection

✓ Tissue segmentation

✓ Histopathology report generation

Part 3/3:

# Real-World Clinical Deployments



Hospital Case Studies



Emergency Department Applications



Clinical Decision Support



FDA Approved Systems



ROI Analysis



## Mayo Clinic Case Study



**2021**

Deployment Year

**3M annually**

Scale

**15%  
diagnosis  
accuracy ↑**

Key Result



### Key Achievements

✓ 30% physician time saved

✓ 3-year payback period



✓ Enhanced clinical workflows and patient outcomes



# Stanford Healthcare Case Study



**2022**

Deployment Year

**Epic EHR  
integration**

System

**47%  
medication  
error  
reduction**

Key Result

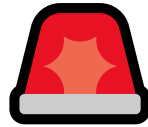


## Key Achievements

✓ Real-time clinical decision support

✓ Improved patient safety

✓ Enhanced clinical workflows and patient outcomes



## Emergency Department Triage

23 min average wait time reduction

92% priority accuracy

ESI level prediction

Real-time risk assessment



## Clinical Decision Support Systems

Drug interaction alerts

Evidence-based guidelines

Lab test optimization

Risk score calculation



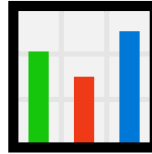
## FDA Approved AI Systems

520+ FDA cleared devices

IDx-DR (diabetic retinopathy)

Caption Health (echocardiography)

Rapid regulatory growth



## Performance Metrics & Benchmarks

MedQA: 87% accuracy

PubMedQA: 78% F1

MMLU-Medical: 91%

AUROC, Sensitivity, Specificity



## Cost-Benefit Analysis

Initial: \$2M investment

Operating: \$500K/year

Savings: \$3M/year (error reduction)

ROI: 250% over 3 years





## Future Directions & Opportunities



### Personalized Medicine

Genomic data integration, tailored treatment plans



### Real-time Monitoring

Wearable integration, continuous health tracking



### Drug Discovery

Accelerated compound identification, clinical trials



### Precision Surgery

AI-assisted robotics, real-time guidance



### Preventive Medicine

Early disease detection, risk prediction



### Global Health Access

Telemedicine, underserved populations



## Hands-On Environment Setup



### Required Software

✓ Python 3.9+

✓ PyTorch 2.0

✓ Transformers 4.30

✓ Docker Desktop

✓ CUDA 11.8+ (GPU)

✓ Jupyter Notebook

```
# Installation Commands pip install torch transformers datasets pip install huggingface_hub  
accelerate pip install pandas numpy scikit-learn pip install jupyter notebook # Verify  
Installation python -c "import torch; print(torch.cuda.is_available())"
```



## Assignment: Medical QA System



**Task:** Build a medical question-answering system



**Dataset:** MedQA or PubMedQA (provided)



**Deadline:** 2 weeks



**Evaluation:** Accuracy, code quality, documentation

# Thank You!

Questions? Contact your instructor



[instructor@university.edu](mailto:instructor@university.edu)

Next Lecture: Fine-tuning Medical LLMs