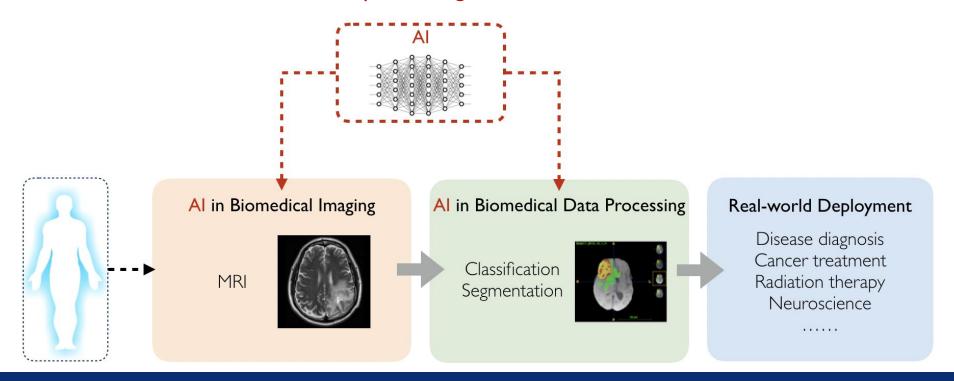
Lecture 6: Multimodal Foundation Models

In this class:

- Part I:Al in biomedical imaging
- Part II:Al in biomedical data processing



Today's agenda

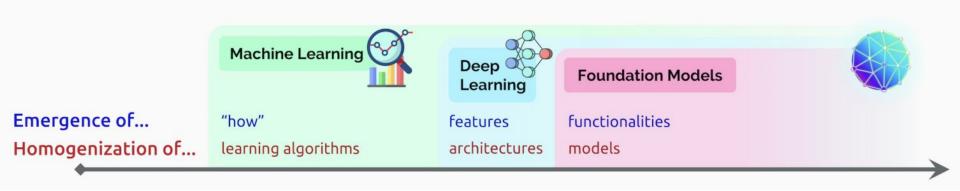
- Foundation model
- Language foundation model
- Vision foundation model
- Multimodal foundation model
- Generalist Medical Al (GMAI)

Today's agenda

- Foundation model
- Language foundation model
- Vision foundation model
- Multimodal foundation model
- Generalist Medical Al (GMAI)

Foundation model

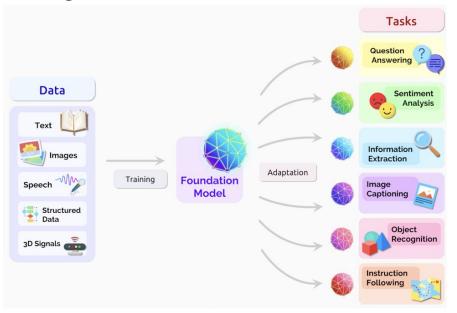
A new paradigm for building Al systems



Bommasani, et al., On the Opportunities and Risks of Foundation Models, arXiv 2022.

Foundation model

- A new paradigm for building Al systems
 - Train one model on a huge amount of multimodal data using self-supervision at scale
 - Can adapt to a wide range of downstream tasks



Bommasani, et al., On the Opportunities and Risks of Foundation Models, arXiv 2022.

Medical foundation model

Models trained on multiple medical data modalities can enable comprehensive clinical

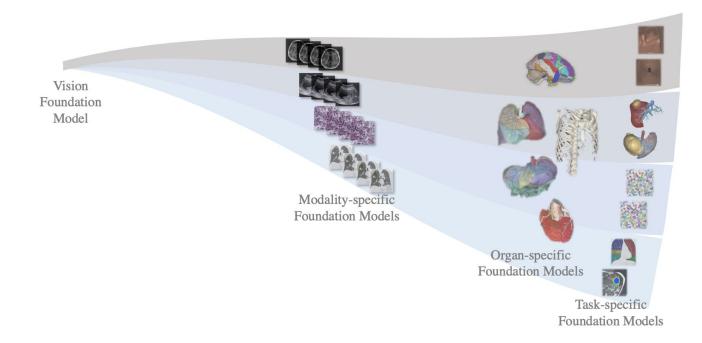
solutions



Zhang, et al., On the Challenges and Perspectives of Foundation Models for Medical Image Analysis, arXiv 2023.

Medical foundation model

Modality-specific and organ-specific foundation models



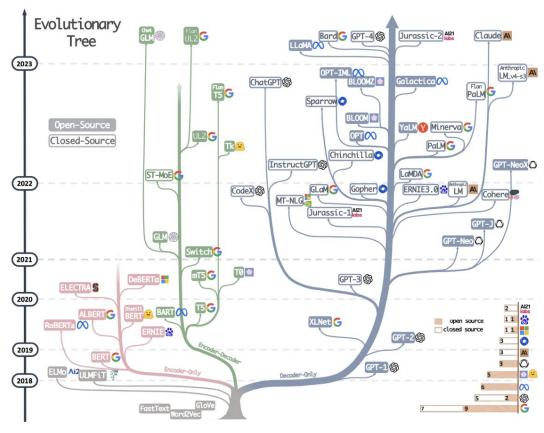
Zhang, et al., On the Challenges and Perspectives of Foundation Models for Medical Image Analysis, arXiv 2023.

Today's agenda

- Foundation model
- Language foundation model
- Vision foundation model
- Multimodal foundation model
- Generalist Medical AI (GMAI)

Language foundation model

- Large Language Models (LLMs)
 - Zero-shot generalization ability



Yang, et al., Harnessing the power of LLMs in practice: a survey on ChatGPT and beyond, arXiv 2023.

Language foundation model

- Large Language Models (LLMs)
 - BERT-style language models: encoder-decoder or encoder-only
 - o GPT-style language models: decoder-only

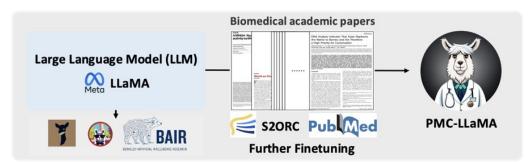
		Characteristic	LLMs			
Encoder-Decoder or Encoder-only (BERT-style)	Training: Model type: Pretrain task:	Masked Language Models Discriminative Predict masked words	ELMo [80], BERT [28], RoBERTa [65], DistilBERT [90], BioBERT [57], XLM [54], Xlnet [119], ALBERT [55], ELECTRA [24], T5 [84], GLM [123], XLM-E [20], ST-MoE [133], AlexaTM [95]			
Decoder-only (GPT-style)	Training Model type: Pretrain task:	Autoregressive Language Models Generative Predict next word	GPT-3 [16], OPT [126]. PaLM [22], BLOOM [92], MT-NLG [93], GLaM [32],Gopher [83], chinchilla [41], LaMDA [102], GPT-J [107], LLaMA [103], GPT-4 [76], BloombergGPT [117]			

Yang, et al., Harnessing the power of LLMs in practice: a survey on ChatGPT and beyond, arXiv 2023.

Medical language foundation model

Medical LLMs

- Train from scratch on medical language
 - BioGPT
 - BioMedLM (2.7-billion)
- Fine-tune from LLMs of natural language
 - PMC-LLaMA (7-billion)
 - MedPaLM (540-billion)



Flan-PaLM (540 B) +17% PubMed GPT MedQA (USMLE) 4 0 0 0 0 0 0 DRAGON 50.3 BioLinkBERT Galactica (120 B) 47.5 45.1 44.4 (100 M) **GPT-Neo** 38.1 35 (2.7 B) 33.3 Dec'20 July'20 Oct'22 Dec'22 March'22 Nov'22 Dec'22

Automated and Human Evaluation

Med-PaLM: Newborn jaundice is when a newborn baby's skin and eyes look yellow. It is caused by a buildup of a substance called bilirubin in the baby's blood. Bilirubin is a byproduct of the normal process of red blood cells breaking down. Newborn jaundice is common and typically harmless, but it can be a sign of a more serious condition. The jaundice typically goes away on its own within a few weeks. In some cases, it may last longer or require treatment. The time it takes for the jaundice to go away can vary depending on the cause and the severity of the jaundice. If the jaundice is severe or lasts longer than a few weeks, the doctor may recommend testing or treatment to

determine the cause and prevent complications.

Q: How long does it take for newborn jaundice to go away?

Med-PaLM performs encouragingly on consumer medical question answering

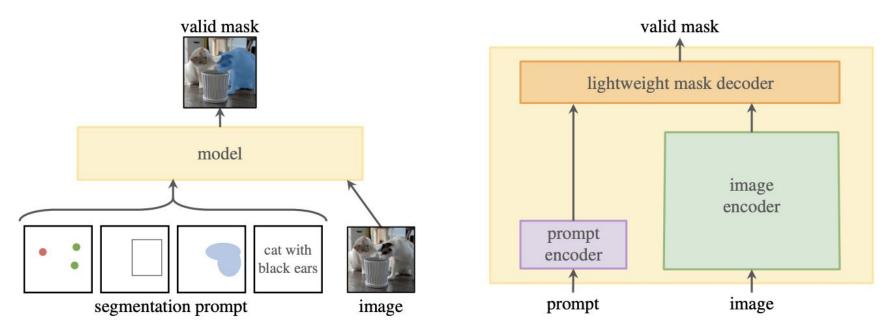
Luo, et al., BioGPT: generative pre-trained transformer for biomedical text generation and mining, Briefings in Bioinformatics 2022. Singhal, et al., Large language models encode clinical knowledge, arXiv 2022. Wu, et al., PMC-LLaMA: Further finetuning LLaMA on medical papers, arXiv 2023.

Today's agenda

- Foundation model
- Language foundation model
- Vision foundation model
- Multimodal foundation model
- Generalist Medical Al (GMAI)

Vision foundation model

- Segment Anything Model (SAM) for image segmentation
 - Enable zero-shot transfer to a range of tasks via prompt engineering



Kirillov et al. Segment Anything. arXiv 2023.

Medical vision foundation model

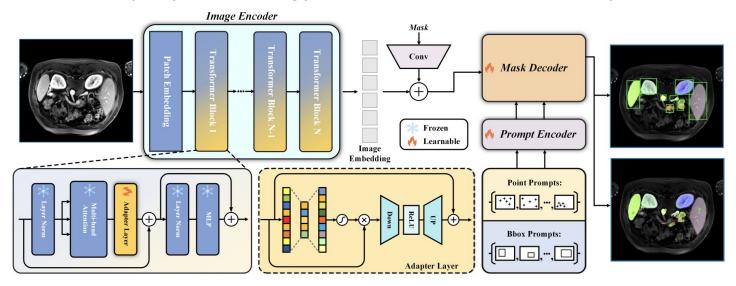
- Fine-tune SAM for medical images
 - Support multiple prompts

Model	Dataset (size)	Encoder	Pro	Decoder			
Model	Dataset (Size)	(or Adapter)	Point	t Bbox Mask		Decouel	
SAM-U [16]	6000 masks	X	×	~	X	X	
SAMed [17]	3779 masks	V	×	×	X	V	
AutoSAM [18]	ACDC [19]	×	×	×	X		
MedSAM [20]	\sim 1.1M masks	×	X	~	×	V	
MSA [21]	5 datasets	V		×	X		
SAM-Med2D (Ours)	\sim 19.7M masks	V	~	~	~	V	

Cheng et al. SAM-Med2D. arXiv 2023.

Medical vision foundation model

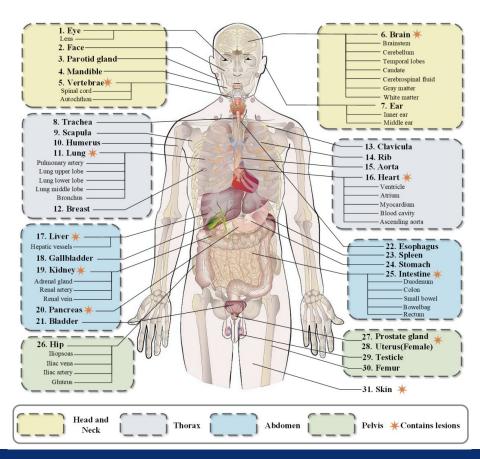
- Fine-tune SAM for medical images
 - Incorporate learnable adapter layers in each Transformer block to acquire domain-specific knowledge in the medical field
 - Fine-tune the prompt encoder using point, Bbox, and mask information, and update mask decoder



Cheng et al. SAM-Med2D. arXiv 2023.

Medical vision foundation model

- Fine-tune SAM for medical images
 - Datasets used cover 31 major organs



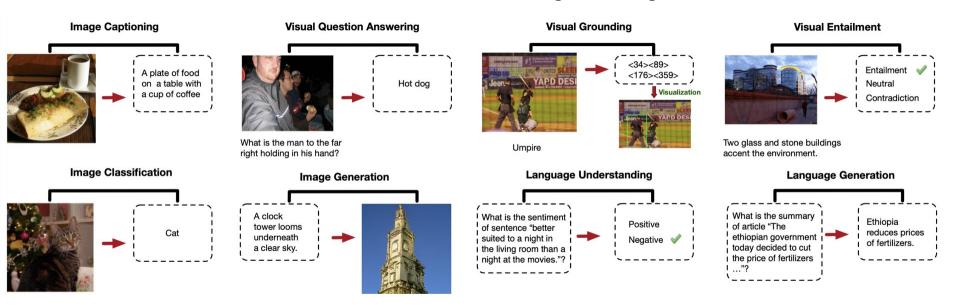
Cheng et al. SAM-Med2D. arXiv 2023.

Today's agenda

- Foundation model
- Language foundation model
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Multimodal foundation model

- Multimodal pre-training paradigm for general vision-language understanding
 - Learn transferable cross-modal representations from various self-supervised tasks
 - Generalize to advance various downstream tasks through fine-tuning



Wang, et al., OFA: Unifying architectures, tasks, and modalities through a simple sequence-to-sequence learning framework, ICML 2022.

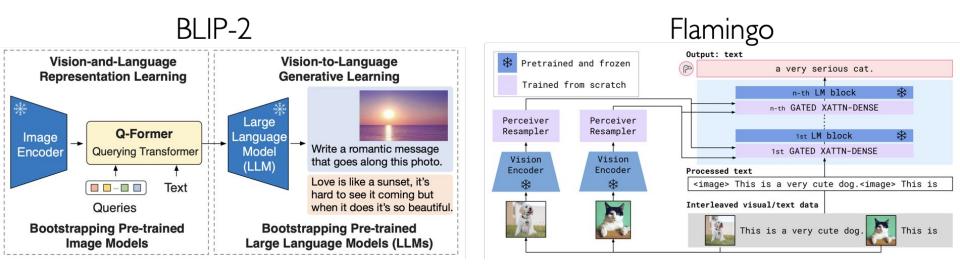
Multimodal foundation model

- Multimodal pre-training paradigm for general vision-language understanding
 - Learn transferable cross-modal representations from various self-supervised tasks
 - Generalize to advance various downstream tasks through fine-tuning

		Computer Visi	on		Natural	Langua	nge Processing	Image-Text			Video-Text			
Method	Image Cls.	Video Cls.	Det.	Seg.	Text Cls.	QA	Summarization	Retrieval	QA	Captioning	VG	Retrieval	QA	Captioning
BEiT-3	✓		✓	✓				✓	✓	✓				
EVA	✓	✓	✓	✓										
CLIP	✓							✓				✓		
ALBEF								✓	✓		✓			
BLIP								✓	✓	✓		✓	✓	
VATT	✓	\checkmark										✓		
Florence	✓	\checkmark	\checkmark					✓	✓			✓		
CoCa	✓	\checkmark						✓	✓	\checkmark		✓		
VideoCoCa		\checkmark												
Flamingo		\checkmark						✓	✓	✓			✓	✓
GIT2	✓				✓				✓	\checkmark			✓	✓
FLAVA	✓				✓	\checkmark	✓		\checkmark					
OFA	✓				✓	✓	\checkmark	✓	1	✓	✓			
OmniVL	✓	\checkmark						✓	\checkmark	✓		✓	✓	✓
mPLUG 2.0	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓

Xu, et al., mPLUG-2: A Modularized multi-modal foundation model across text, image and video, arXiv 2023.

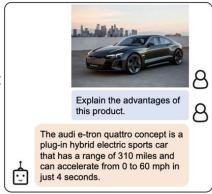
- Enhance vision-language understanding using LLM
 - Learn connection module between visual encoder and LLMs
 - o BLIP-2, Flamingo, FROMAGe, MiniGPT-4 ...



Alayrac, et al., Flamingo: a Visual Language Model for Few-Shot Learning, NeurIPS 2022. Li, et al., BLIP-2: Bootstrapping language-image pre-training with frozen image encoders and large language models, arXiv 2023.

BLIP-2:

Instructed zero-shot image-to-text generat



Is this photo unusual?

How could someone get

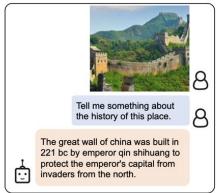
Yes, it's a house that looks like it's

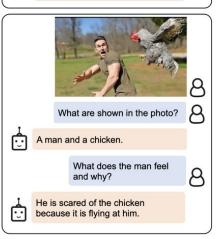
out of the house?

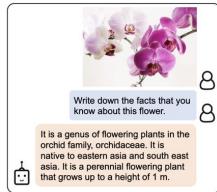
It has a slide on the side of the

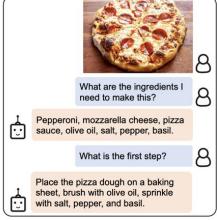
upside down.

house.





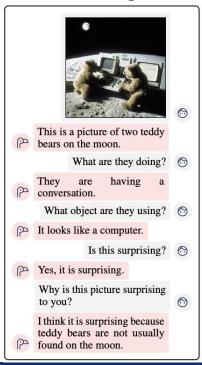


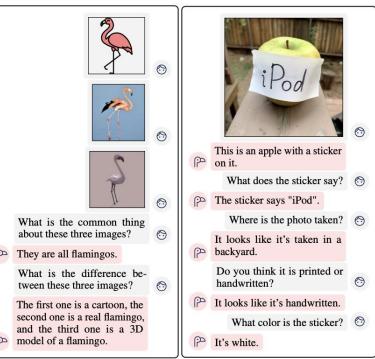


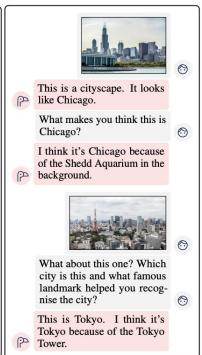
8

Flamingo:

Multi-image visual dialogue

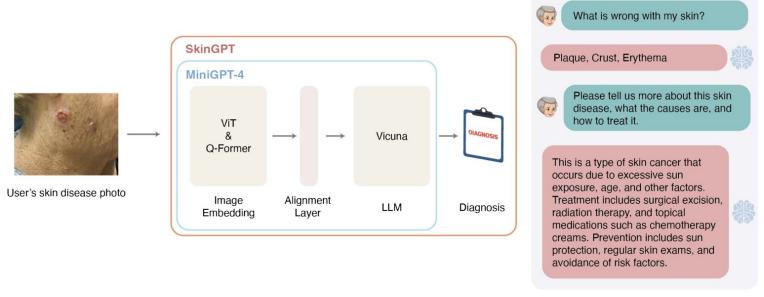






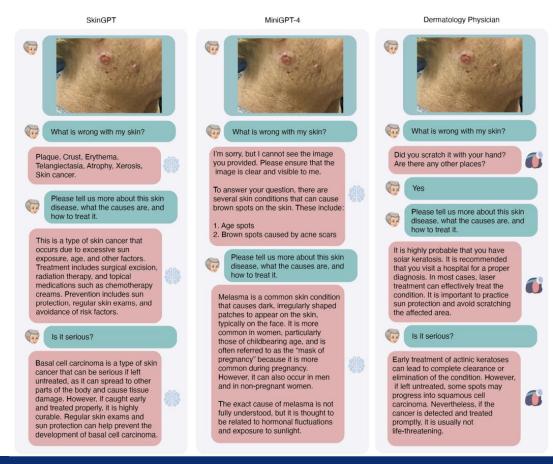
SkinGPT-4

- Fine-tune from MiniGPT-4 with skin disease images
- Enable interactive dialogue with multimodal inputs

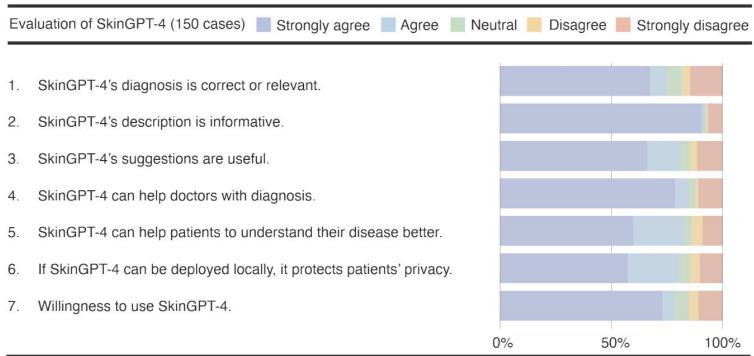


Zhou, et al., SkinGPT-4: An Interactive Dermatology Diagnostic System with Visual Large Language Model, arXiv 2023.

SkinGPT-4



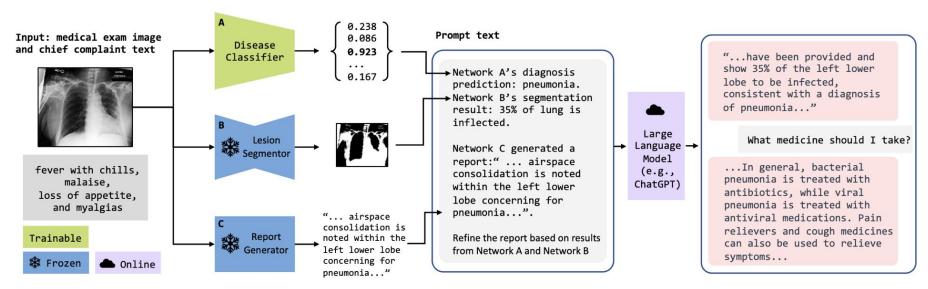
Clinical evaluation of SkinGPT-4 by certified offline and online dermatologists



Zhou, et al., SkinGPT-4: An Interactive Dermatology Diagnostic System with Visual Large Language Model, arXiv 2023.

ChatCAD

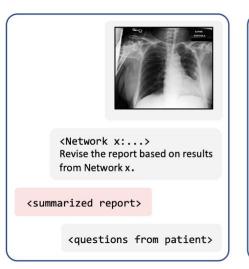
- Use LLMs to enhance the output of multiple CAD networks by summarizing and reorganizing information presented in natural language text format
- Leverage strengths of LLMs' medical domain knowledge and logical reasoning



Wang, et al., ChatCAD: Interactive computer-aided diagnosis on medical image using large language models, arXiv 2023.

ChatCAD

Interactive CAD with LLM



What medicine should I take?

For bacterial pneumonia, antibiotics are the most common form of treatment. Some commonly used antibiotics for pneumonia include penicillin, macrolides, fluoroquinolones, and beta-lactams.

If your pneumonia is caused by a virus, antibiotics are not effective. Instead, antiviral medications, such as oseltamivir (Tamiflu) or zanamivir (Relenza), if your pneumonia is caused by the flu.

What's airspace consolidation

Airspace consolidation may refer to the appearance of a particular area of the lung where there is a buildup of fluid, air, or other material that is causing the lung tissue to collapse. This can result in a decreased amount of air in the lung, which can make it difficult for the patient to breathe. Airspace consolidation can be a sign of various lung conditions, such as pneumonia, lung abscess...

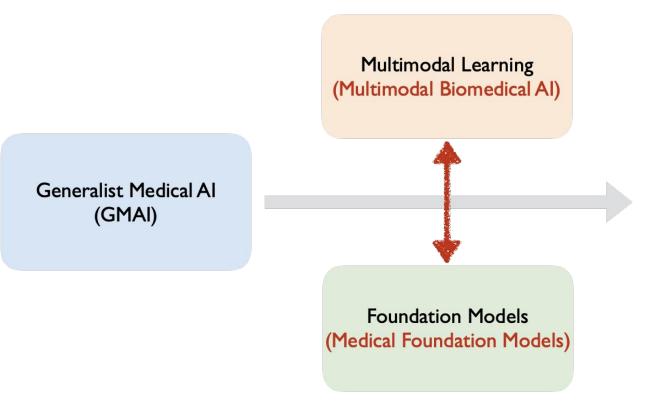
I lost smell for more than a week, why?

The loss of smell may be due to inflammation and swelling in the nasal passages, which can interfere with the ability to detect odors. Additionally, the infection can also spread to the olfactory nerve, which is responsible for transmitting smells from the nose to the brain.

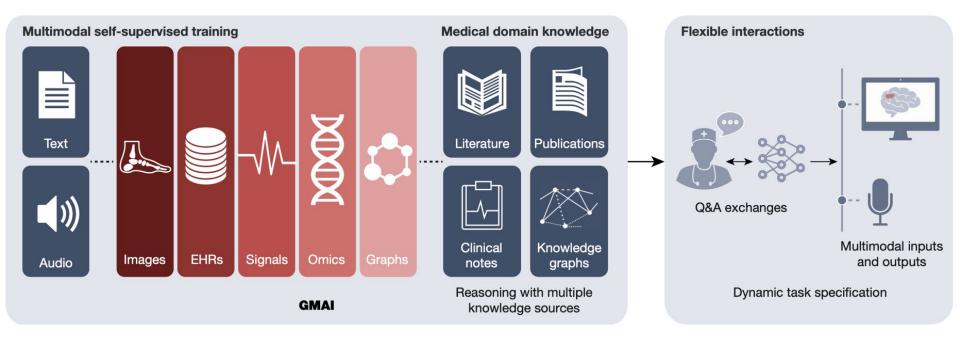
Wang, et al., ChatCAD: Interactive computer-aided diagnosis on medical image using large language models, arXiv 2023.

Today's agenda

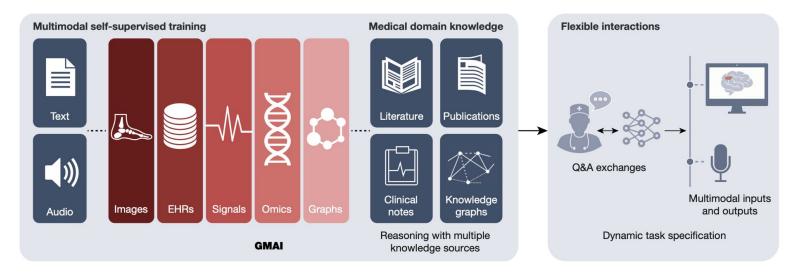
- Foundation model
- Language foundation model
- Vision foundation model
- Multimodal foundation model
- Generalist Medical AI (GMAI)

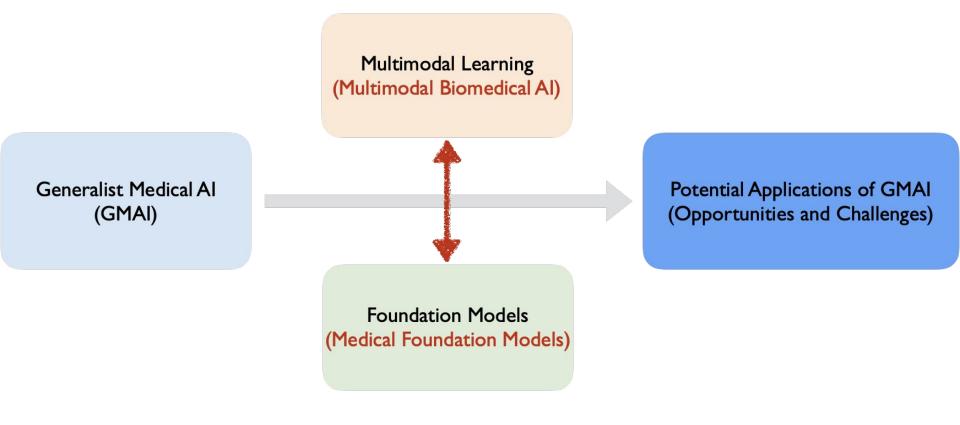


- Generalist paradigm: medical foundation models on multimodal architectures
 - Promise to solve more diverse and challenging tasks than current medical AI models



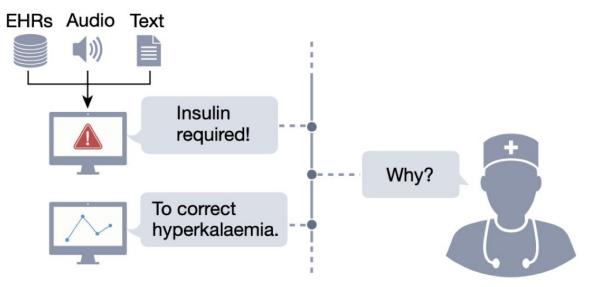
- Generalist paradigm: medical foundation models on multimodal architectures
 - Adapt to new tasks with dynamic task specification
 - Flexible combinations and interactions of data modalities
 - Represent medical domain knowledge for reasoning new tasks





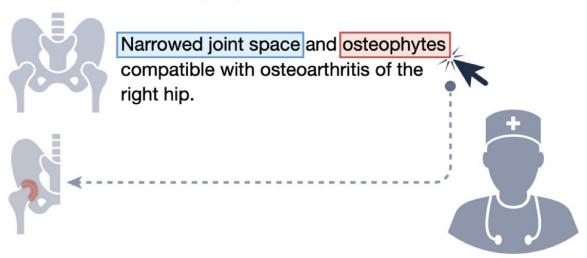
- Generalist paradigm: medical foundation models on multimodal architectures
 - Bedside decision support

a Bedside decision support

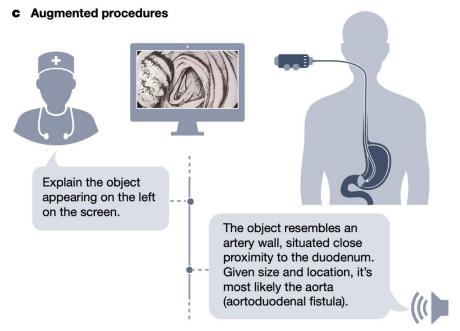


- Generalist paradigm: medical foundation models on multimodal architectures
 - Grounded radiology reports

b Grounded radiology reports



- Generalist paradigm: medical foundation models on multimodal architectures
 - Augmented procedures



- Generalist paradigm: medical foundation models on multimodal architectures
 - Bedside decision support
 - Grounded radiology reports
 - Augmented procedures
 - Interactive note-taking
 - Chatbots for patients
 - Text-to-protein generation
 - 0



patients





Augmented procedures



Grounded radiology reports



Text-to-protein generation

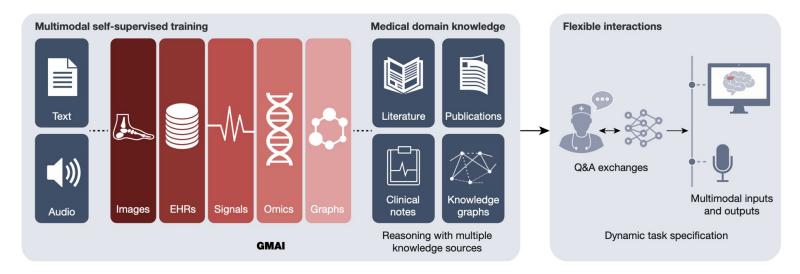


Bedside decision support

Challenges of GMAI

- Validation
- Verification
- Social biases
- Privacy
- Scale
-

- Generalist paradigm: medical foundation models on multimodal architectures
 - Adapt to new tasks with dynamic task specification
 - Flexible combinations and interactions of data modalities
 - Represent medical domain knowledge for reasoning new tasks



All lectures are done!

Any feedback and suggestions?

Date	Lecture #	Topic	Papers	Instructor / Presenter
Tue 8/29	1	Introduction and course overview		Liyue Shen
Thu 8/31	2	Biomedical imaging with deep learning [Fundamental]		Liyue Shen
Tue 9/5	3	Implicit neural representation learning [Advanced]		Liyue Shen
Thu 9/7	4	Generative diffusion models [Advanced]		Liyue Shen
Tue 9/12	5	Medical image analysis [Fundamental]		Liyue Shen
Thu 9/14	6	Multimodal foundation models [Advanced]		Liyue Shen
Mon 9/18		Drop/add deadline for full term classes		
Tue 9/19	7	Implicit neural representation learning		
Thu 9/21	8	Implicit neural representation learning		
Tue 9/26	9	Implicit neural representation learning		
Thu 9/28	10	Implicit neural representation learning		
Tue 10/3	11	Generative diffusion models		
Thu 10/5	12	Generative diffusion models		
Tue 10/10	13	Generative diffusion models		
Thu 10/12	14	Generative diffusion models		
Tue 10/17		No class (fall study break)		
Thu 10/19	15	Self-supervised learning		
Tue 10/24	16	Self-supervised learning		
Thu 10/26	17	Multimodal learning		
Tue 10/31	18	Multimodal learning		
Thu 11/2	19	Transformer and LLM		
Tue 11/7	20	Transformer and LLM		