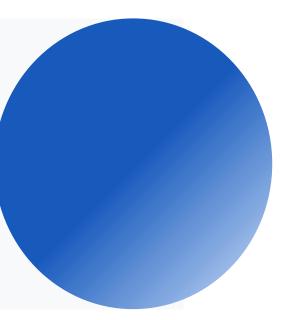
# Transforming cancer surveillance Emerging directions in multimodal health data analytics

### Dr. Pete Clardy, MD

Lead, Clinical Enterprise Team Google Health



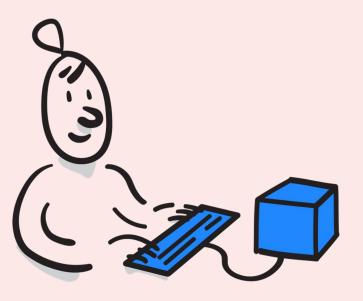
Google Health



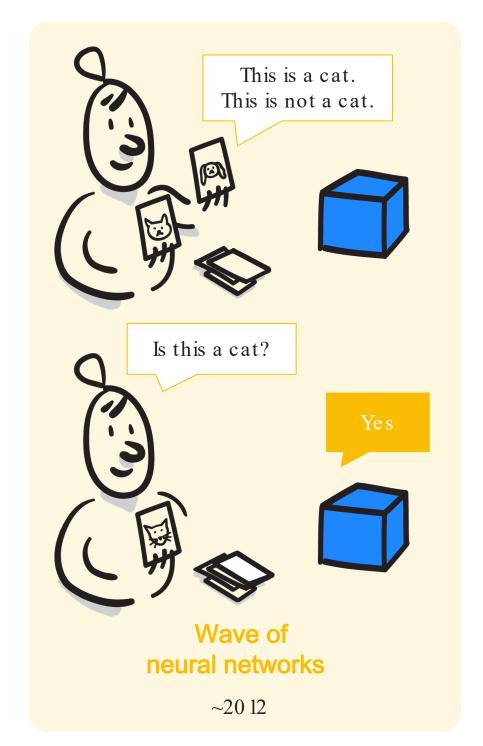
# We are in a new era of generative Al...

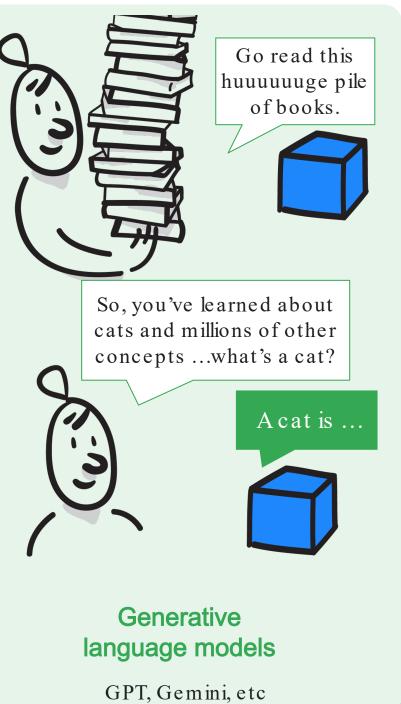
Cat: type:animal legs:4 ears: 2 fur: yes likes: yarn, catnip

(etc ...)



**Traditional** programming





Proprietary + Confidential

# We are in a new era of generative AI...

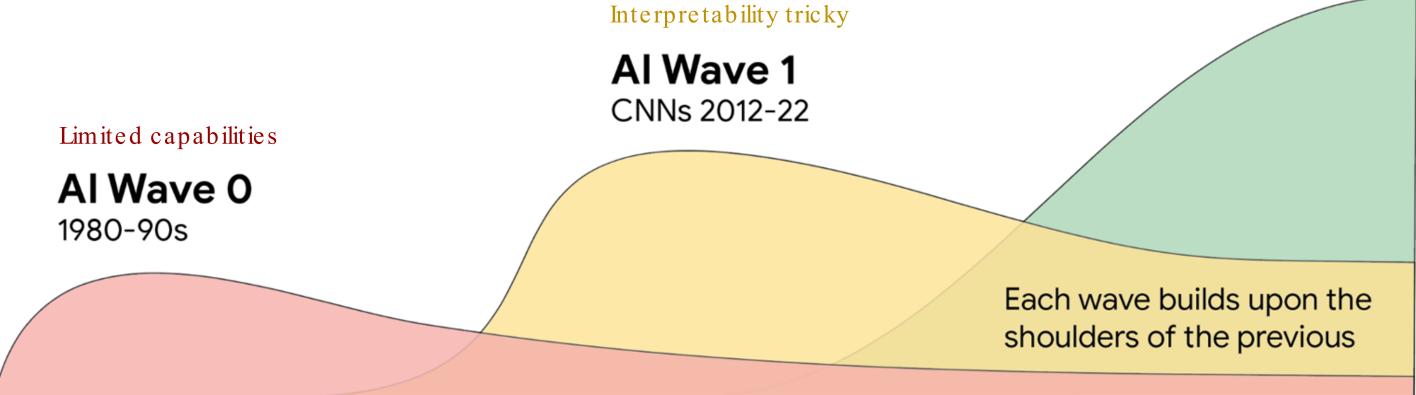
Successful at narrow tasks

Fragile

Expensive training data

Poor multimodal/sequential

Al Wave 1



Google Health

#### Generalization

Training using free text Multimodal and sequential Rich expressive outputs

### Al Wave 2 GenAl 2022+

#### JAMA 2024; 331:242

# Initial efforts focused on narrow applications:

- Highly constrained inputs
- Simple task
- High volume / lots of data
- Constrained outputs





Lung cancer detection

Diabetic retinopathy detection



Pathology classifiers / predictors



detection



AKIprediction

#### Google Health



#### Genomics

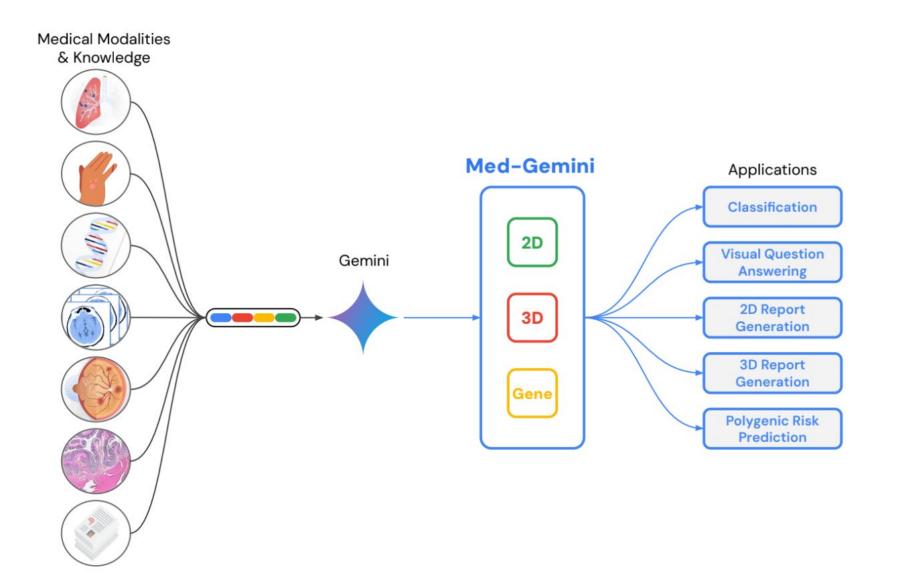




Dermatology

### ...and many others

# Attention has shifted to multimodal models:



#### Google DeepMind and Google Research

### Gemini

Google Research and Google DeepMind

2024

May

9

CV

CS.

arXiv:2405.03162v1

Many clinical tasks require an understanding of specialized data, such as medical images and genomics, which is not typically found in general-purpose large multimodal models. Building upon Gemini's multimodal models, we develop several models within the new Med-Gemini family that inherit core capabilities of Gemini and are optimized for medical use via fine-tuning with 2D and 3D radiology, histopathology, ophthalmology, dermatology and genomic data. Med-Gemini-2D sets a new standard for AI-based chest X-ray (CXR) report generation based on expert evaluation, exceeding previous best results across two separate datasets by an absolute margin of 1% and 12%, where 57% and 96% of AI reports on normal cases, and 43% and 65% on abnormal cases, are evaluated as "equivalent or better" than the original radiologists' reports. We demonstrate the first ever large multimodal model-based report generation for 3D computed tomography (CT) volumes using Med-Gemini-3D, with 53% of AI reports considered clinically acceptable, although additional research is needed to meet expert radiologist reporting quality. Beyond report generation, Med-Gemini-2D surpasses the previous best performance in CXR visual question answering (VQA) and performs well in CXR classification and radiology VQA, exceeding SoTA or baselines on 17 of 20 tasks. In histopathology, ophthalmology, and dermatology image classification, Med-Gemini-2D surpasses baselines across 18 out of 20 tasks and approaches task-specific model performance. Beyond imaging, Med-Gemini-Polygenic outperforms the standard linear polygenic risk score-based approach for disease risk prediction and generalizes to genetically correlated diseases for which it has never been trained. Although further development and evaluation are necessary in the safety-critical medical domain, our results highlight the potential of Med-Gemini across a wide range of medical tasks.

© 2024 Google. All rights reserved

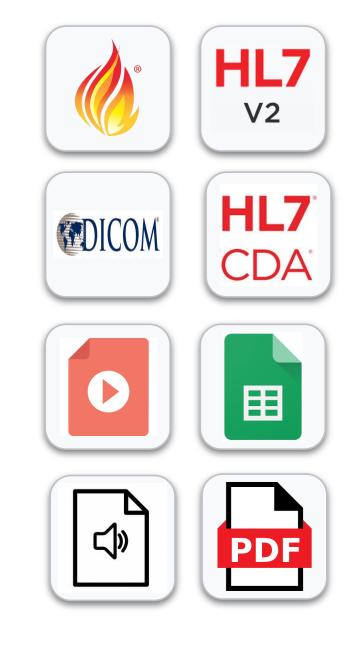
#### Google Health

#### Advancing Multimodal Medical Capabilities of

<sup>&</sup>lt;sup>†</sup> See Contributions and Acknowledgments section for full author list. Corresponding authors: {dangolden, shekazizi, kellych, roryp}@google.com.

# Problem statement:

- Longitudinal and comprehensive patient records are inherently multimodal and heterogeneous
- Questions we ask of healthcare data are often complex and require nuance
- Pre-processing data and questions increases cost, rigidity, and signal loss
- Interop and FHIR have significantly improved this dynamic, yet it still remains a challenge



# Implications for future clinical information systems design:

	Clear Box		
	Deterministic Search/SQL on FHIR Record	Hybrid RAG on FHIR+ Record	Cong Conte
Pre-Process Data	<ul> <li>High storage cost</li> <li>Higher map/merge burden</li> </ul>	<ul> <li>Higher storage cost</li> <li>+ Lower map/merge burden</li> </ul>	+ No map/r + Multimod
Process Query	<ul><li>+ Lower latency &amp; cost</li><li>- Requires explicit logic</li></ul>	<ul> <li>Requires concept expansion</li> <li>&amp; knowledge graphs</li> </ul>	- Higher lat + High flex
Post-Process Response	<ul><li>+ Accurate &amp; consistent</li><li>- Requires parsing</li></ul>	+ Grounding - Hallucination risk	+ Human ro - Hallucina
Use Cases	End Users / System to system Production	Human-in-the-loop Pilot	Profession Proof of Co

### Opaque Box

### Generative

text LLM on Raw Record

/merge burden dal ingest

atency & cost xibility w/ prompting

ready ation risk

nal-in-the-loop Concept