# Comparing General-Purpose Vision-Language

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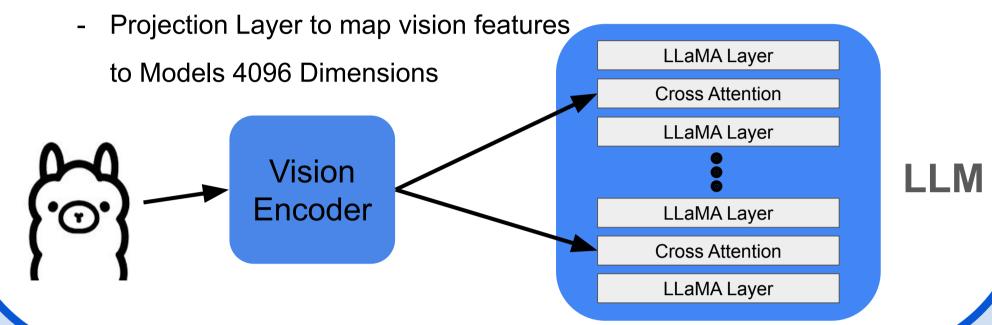


# Models for Medical Diagnostic Tasks

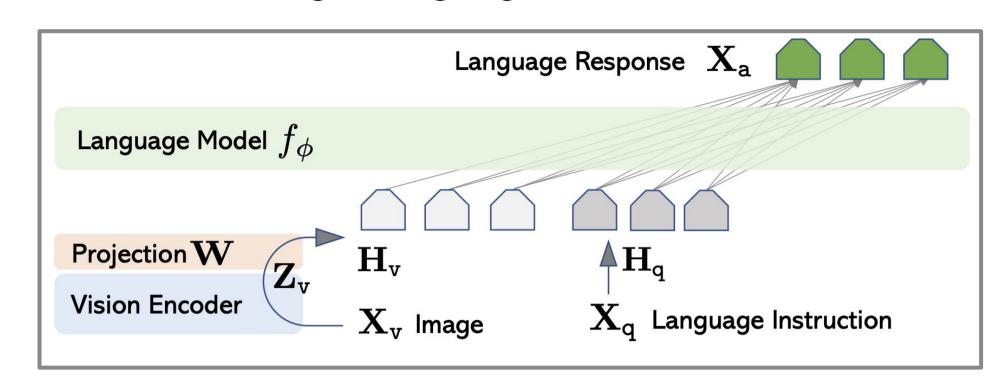


# Llama 3.2 Vision: A Large Vision Language Model by Meta

- 11B parameter (also available with 70B)
- Built on top of Llama 3.1 (LLM)
- Separately trained vision adapter
  - Cross-attention layers (every fifth) that feed image encoder representations into the core LLM
  - 32 Layer Transformer, preserving intermediate representations, concatenated to 8 Layer Global encoder



# LLaVA: Large Language and Vision Assistant



- Vicuna as LLM (LLaMA 2 fine-tuned on following instructions)
- Pretrained CLIP vision encoder (ViT/L 14)
- ViT outputs are projected into embedding space of language tokens
- Language tokens are appended to image tokens
- Conditional generation based on the whole token sequence

### Results on medical image tasks

#### Chest x-rays

#### 1. Classification:

 Accuracy: 0.240 • F1-Score: 0.387

All results classified as unhealthy

#### 2. Bounding Boxes:

- Not successful in returning bounding boxes
- 'I cant help you with that. Is there anything else I can help you with?'

#### **Brain MRI Slices**

- 1. Generate Medical Description:
  - Evaluated against ground truth with LLM Llama3
  - 3 Correct 22 Incorrect

#### 2. Bounding Boxes:

- Not successful in returning bounding boxes
- 'I cant help you with that. Is there anything else I can help you with?'
- 3. Disease Diagnosis (Clinical History+Image Findings):
  - Evaluated against ground truth with LLM Llama3
  - 11 Correct 14 Incorrect

# Results on medical image tasks

#### Chest x-rays

#### 1. Classification:

Accuracy: 64%

• F1-Score: 0.18

#### 2. Abnormality Grounding:

• mAP < 0.001

Predicted bounding box values always between 0 and 1

Scaled bounding boxes by image size

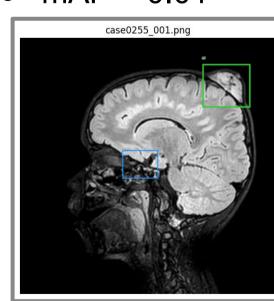
#### **Brain MRI**

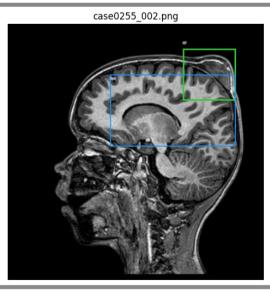
### 1. Description Generation:

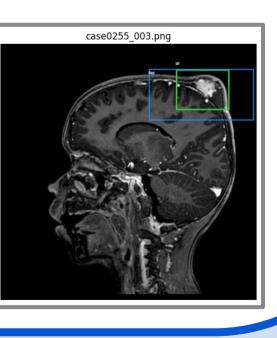
- Very similar more general descriptions for all images
- E.g.: [...] displays a close-up of a human brain, focusing on the cranial area. [...]
- Best BLEU-1 score: 0.2247

#### 2. Abnormality Detection:

• mAP = 0.01







### Conclusion LLaMA 3.2 Vision

- Not able to consistently return bounding boxes as instructed
- Weak performance in medical image classification, or interpretation
- Nearly impossible to get consistent output format

# Comparison

- LLaVA performed significantly better at creating bounding boxes
- LLaVA was better able to return output in a consistent format
- LLaVA performed better at direct classification tasks

#### Conclusion LLaVA

- Fulfilled tasks with little prompt engineering
- Relatively easy to get a consistent output format
- Poor performance on the medical tasks which is not surprising

