A Visual Language Model for Few-Shot Learning

Few-shot in-context learning

Circulation revenue has increased by 5% in Finland. // Positive

Panostaja did not disclose the purchase price. // Neutral

Paying off the national debt will be extremely painful. // Negative

The company anticipated its operating profit to improve. //



Circulation revenue has increased by 5% in Finland. // Finance

They defeated ... in the NFC Championship Game. // Sports

Apple ... development of in-house chips. // Tech

The company anticipated its operating profit to improve. //



Multimodality



What is the title of this painting?
Answer: The Hallucinogenic Toreador.



Where is this painting displayed?
Answer: Louvres Museum, Paris.



What is the name of the city where this was painted?

Answer:



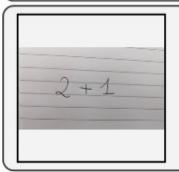
Output: "Underground"



Output: "Congress"



Output:

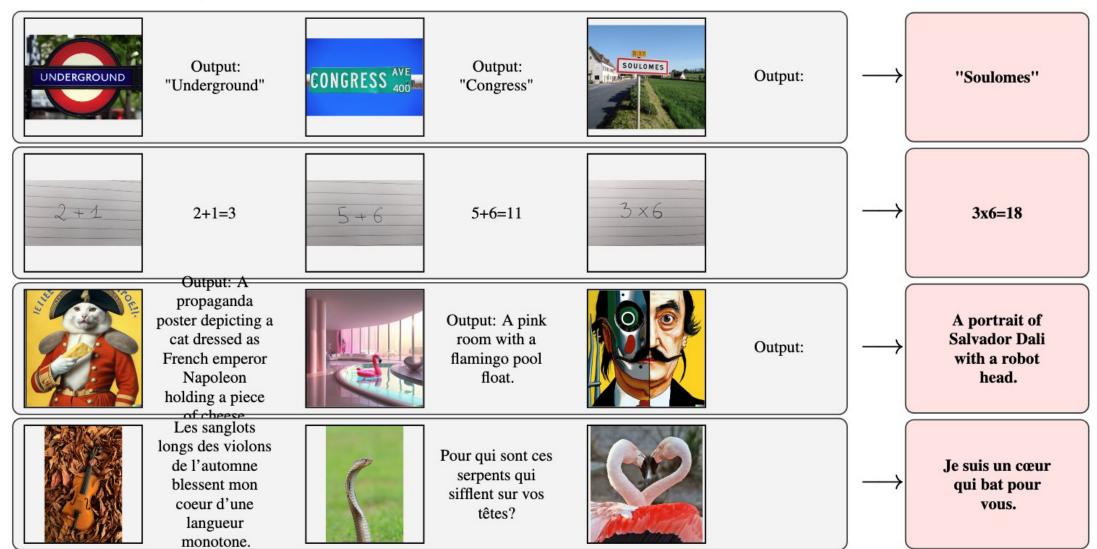


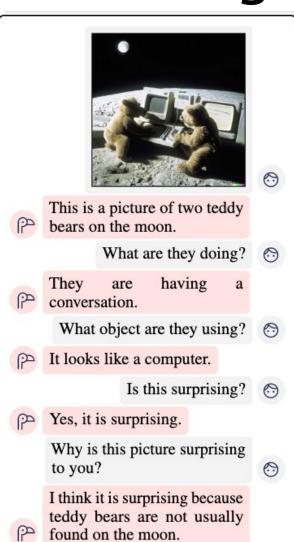
2+1=3

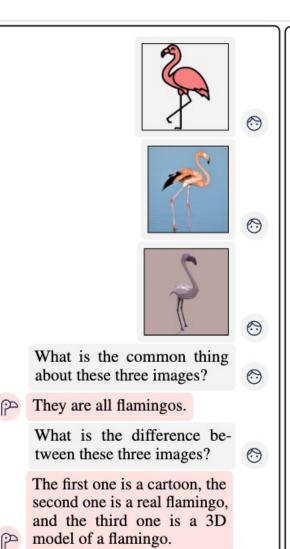


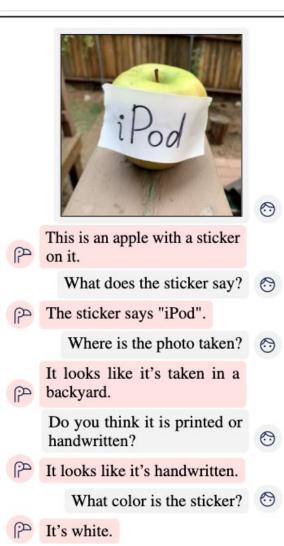
5+6=11

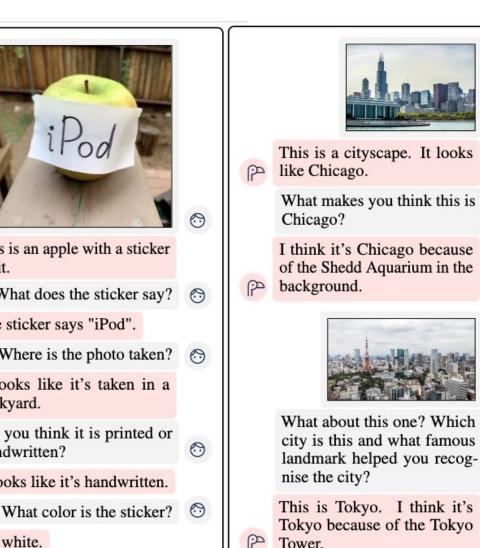












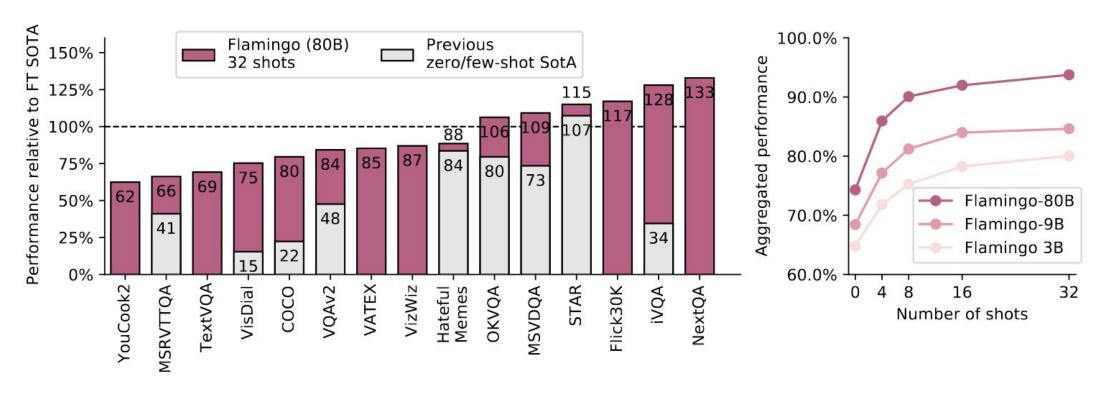
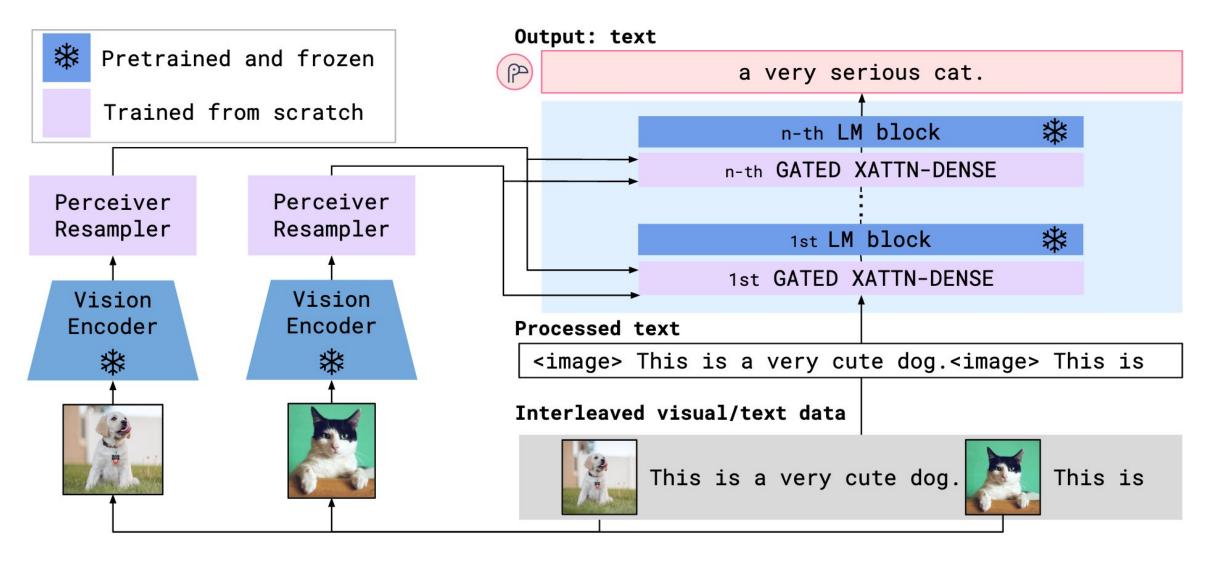


Figure 2: **Flamingo results overview.** *Left*: Our largest model, dubbed *Flamingo*, outperforms state-of-the-art fine-tuned models on 6 of the 16 tasks we consider with no fine-tuning. For the 9 tasks with published few-shot results, *Flamingo* sets the new few-shot state of the art. *Note:* We omit RareAct, our 16th benchmark, as it is a zero-shot benchmark with no available fine-tuned results to compare to. *Right*: Flamingo performance improves with model size and number of shots.

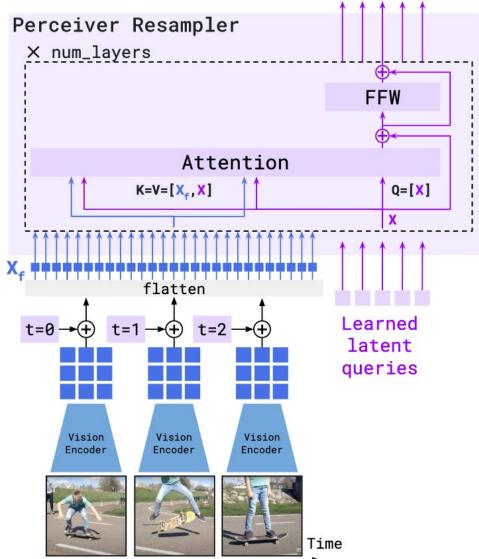
Problems

- -Add image processing into pretrained LM for *text*
- Get compact image embeddings with temporal encoding
- -Gather proper multimodal dataset

Architecture

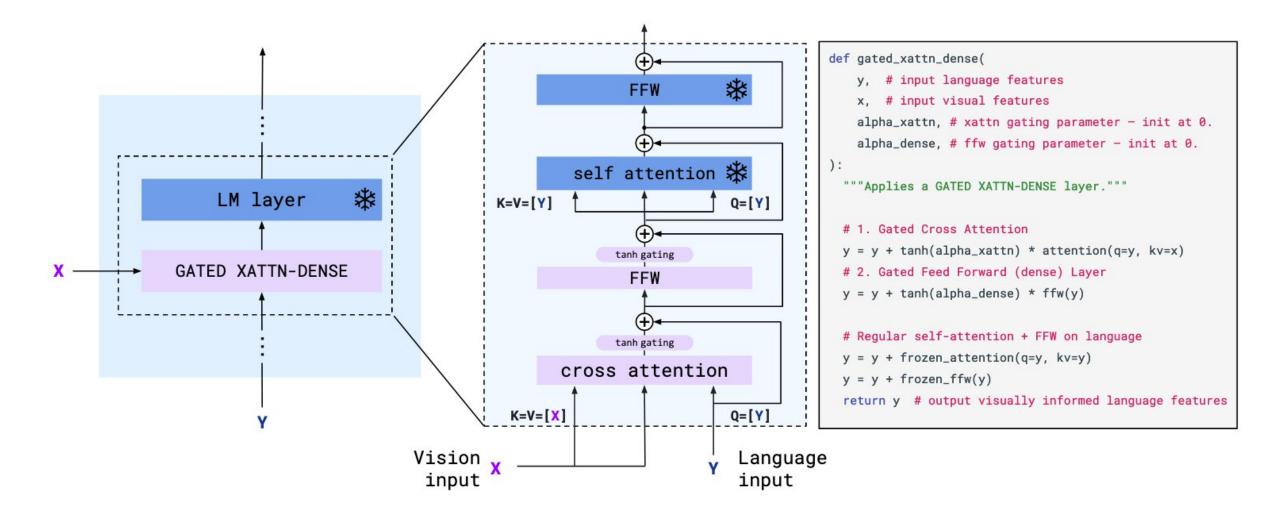


Perceiver Resampler



```
def perceiver_resampler(
   x_f, # The [T, S, d] visual features (T=time, S=space)
   time_embeddings, # The [T, 1, d] time pos embeddings.
   x, # R learned latents of shape [R, d]
   num_layers, # Number of layers
  """The Perceiver Resampler model."""
 # Add the time position embeddings and flatten.
 x_f = x_f + time_embeddings
 x_f = flatten(x_f) \# [T, S, d] \rightarrow [T * S, d]
 # Apply the Perceiver Resampler layers.
 for i in range(num_layers):
   # Attention.
   x = x + attention_i(q=x, kv=concat([x_f, x]))
   # Feed forward.
   x = x + ffw_i(x)
  return x
```

Gated CrossAttention



Dataset



Image-Text Pairs dataset [N=1, T=1, H, W, C]



A kid doing a kickflip.

Video-Text Pairs dataset [N=1, T>1, H, W, C]



This is a picture of my dog.



This is a picture of my cat.

Multi-Modal Massive Web (M3W) dataset [N>1, T=1, H, W, C]

Figure 9: Training datasets. Mixture of training datasets of different formats. N corresponds to the number of visual inputs for a single example. For paired image (or video) and text datasets, N=1. T is the number of video frames (T=1 for images). H,W, and C are height, width and color channels.

Dataset: RICES

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x_train // y_train

x_test // _____



Results: Benchmarks

Method	FT	Shot	OKVQA (I)	VQAv2 (I)	(I) 0000	MSVDQA (V)	VATEX (V)	VizWiz (I)	Flick30K (I)	MSRVTTQA (V)	iVQA (V)	YouCook2 (V)	STAR (V)	VisDial (I)	TextVQA (I)	NextQA (I)	HatefulMemes (I)	RareAct (V)
Zero/Few shot SOTA	X	(X)	[34] 43.3 (16)	[114] 38.2 (4)	[124] 32.2 (0)	[58] 35.2 (0)		=	=	[58] 19.2 (0)	[135] 12.2 (0)		[143] 39.4 (0)	[79] 11.6 (0)	-	G#	[85] 66.1 (0)	[85] 40.7 (0)
Flamingo-3B	X X	0 4 32	41.2 43.3 45.9	49.2 53.2 57.1	73.0 85.0 99.0	27.5 33.0 42.6	40.1 50.0 59.2	28.9 34.0 45.5	60.6 72.0 71.2	11.0 14.9 25.6	32.7 35.7 37.7	55.8 64.6 76.7	39.6 41.3 41.6	46.1 47.3 47.3	30.1 32.7 30.6	21.3 22.4 26.1	53.7 53.6 56.3	58.4
Flamingo-9B	X X	0 4 32	44.7 49.3 51.0	51.8 56.3 60.4	79.4 93.1 106.3	30.2 36.2 47.2	39.5 51.7 57.4	28.8 34.9 44.0	61.5 72.6 72.8	13.7 18.2 29.4	35.2 37.7 40.7	55.0 70.8 77.3	41.8 42.8 41.2	48.0 50.4 50.4	31.8 33.6 32.6	23.0 24.7 28.4	57.0 62.7 63.5	57.9 - -
Flamingo	X X	0 4 32	50.6 57.4 57.8	56.3 63.1 67.6	84.3 103.2 113.8	35.6 41.7 52.3	46.7 56.0 65.1	31.6 39.6 49.8	67.2 75.1 75.4	17.4 23.9 31.0	40.7 44.1 45.3	60.1 74.5 86.8	39.7 42.4 42.2	52.0 55.6 55.6	35.0 36.5 37.9	26.7 30.8 33.5	46.4 68.6 70.0	<u>60.8</u> - -
Pretrained FT SOTA	V	(X)	54.4 [34] (10K)	80.2 [140] (444K)	143.3 [124] (500K)	47.9 [28] (27K)	76.3 [153] (500K)	57.2 [65] (20K)	67.4 [150] (30K)	46.8 [51] (130K)	35.4 [135] (6K)	138.7 [132] (10K)	36.7 [128] (46K)	75.2 [79] (123K)	54.7 [137] (20K)	25.2 [129] (38K)	79.1 [62] (9K)	

Results: Classification

Model	Method	Prompt size	shots/class	ImageNet top 1	Kinetics700 avg top1/5	
SotA	Fine-tuned	-	full	90.9 [127]	89.0 [134]	
SotA	Contrastive	=	0	85.7 [82]	69.6 [<mark>85</mark>]	
NFNetF6	Our contrastive	_	0	77.9	62.9	
		8	1	70.9	55.9	
Flamingo-3B	RICES	16	1	71.0	56.9	
		16	5	72.7	58.3	
		8	1	71.2	58.0	
Flamingo-9B	RICES	16	1	71.7	59.4	
		16	5	75.2	60.9	
	Random	16	≤ 0.02	66.4	51.2	
		8	1	71.9	60.4	
Flamingo-80B	RICES	16	1	71.7	62.7	
		16	5	76.0	63.5	
<u></u>	RICES+ensembling	16	5	77.3	64.2	

Input Prompt



Question: What do you think the capacities of these are?

Answer:



Question: What is odd about this image? Explain why it is unusual. Answer:



Question: What country is this? Why do you think so?

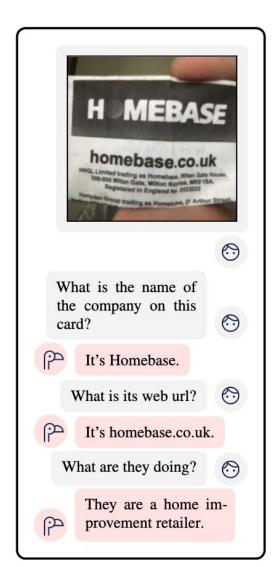
Answer:

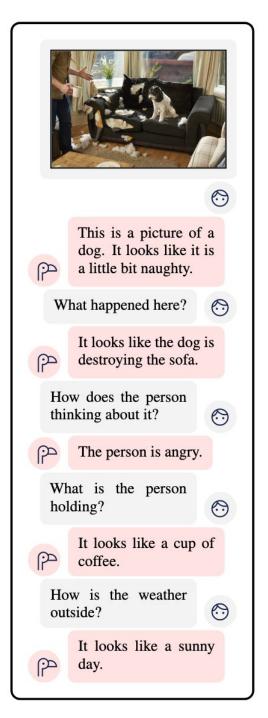
P Completion

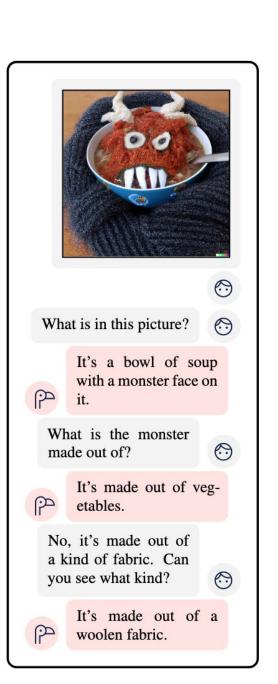
The floppy disk is 1.44MB and the CD is 700MB.

The image is odd because the elephant is in the back of the truck. It is unusual because elephants are not usually transported in the back of a truck.

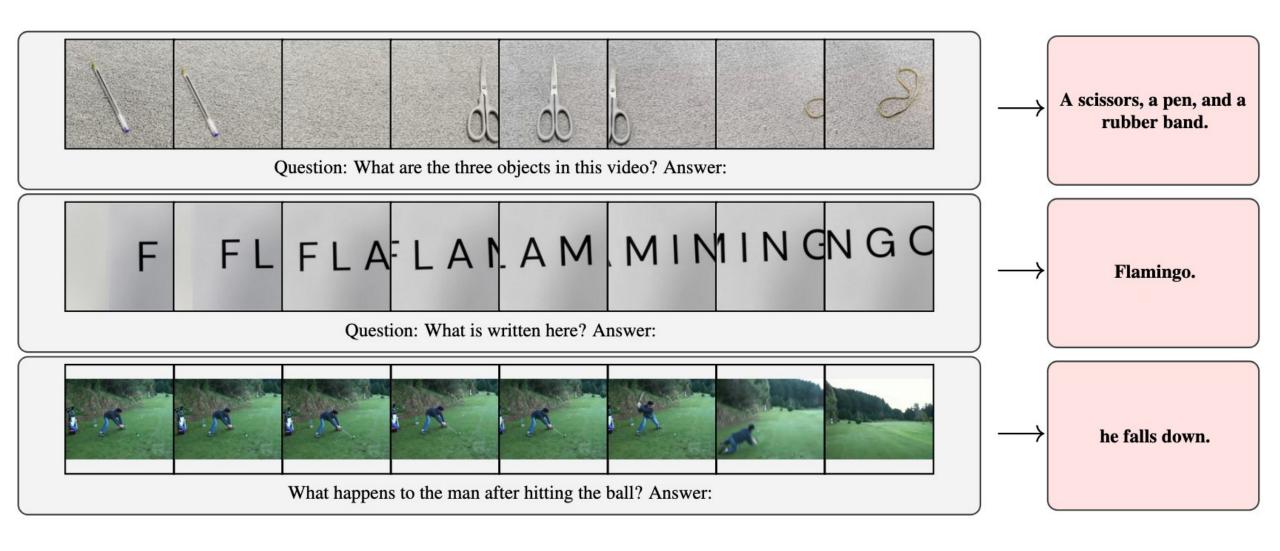
It is Canada. I think so because the flag is the Canadian flag.













Question: What is on the phone screen? Answer:



Question: What can you see out the window? Answer:



Question: Whom is the person texting? Answer:

P Output

A text message from a friend.

A parking lot.

The driver.