

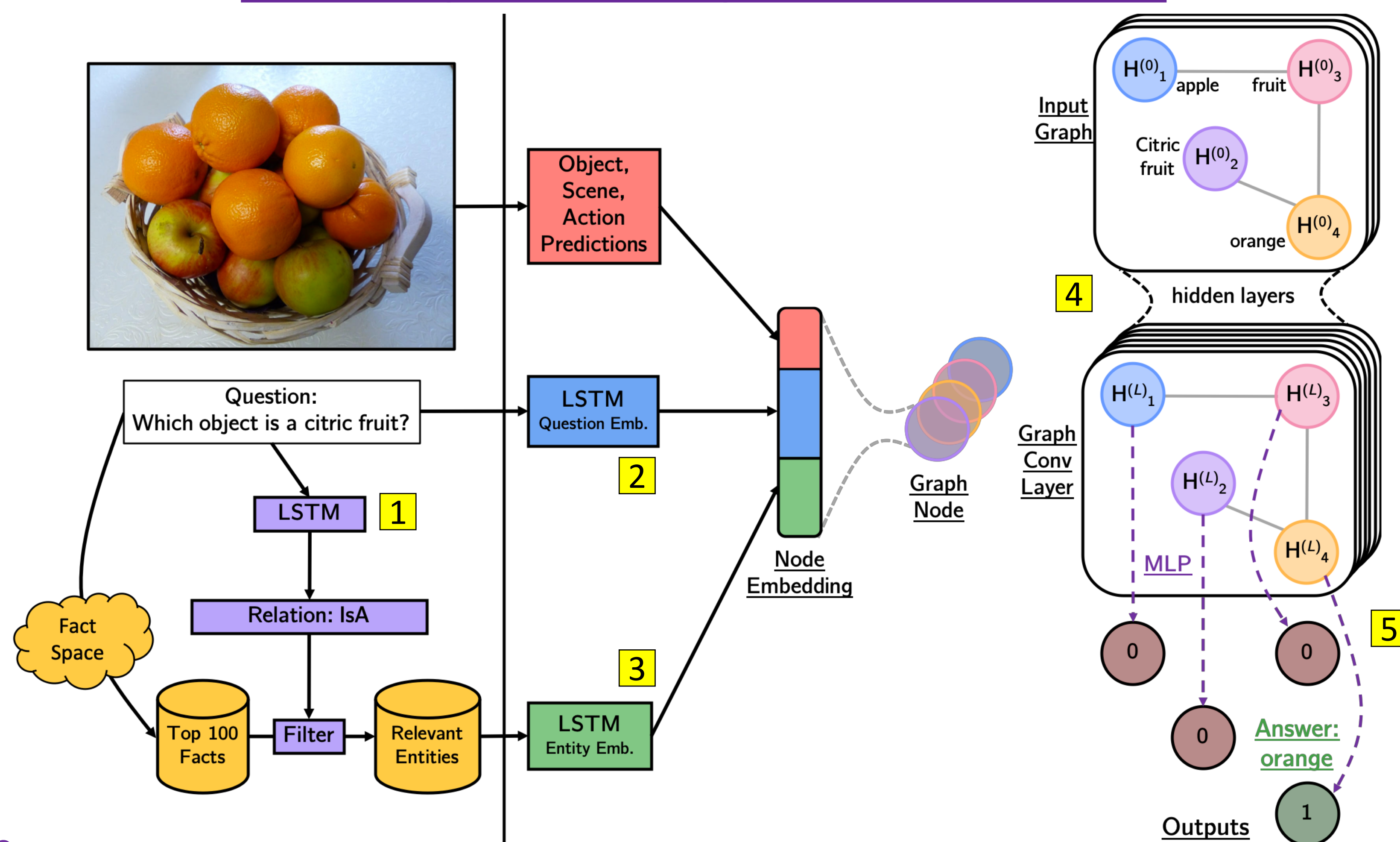
Overview

- Objective:** To answer open ended questions about an image using facts from an external knowledge base.
- We use the **FVQA Dataset** containing image – question pairs and the corresponding **FVQA Knowledge Base** of facts. [1]
- We develop a model that **reasons using message passing** across multiple **relevant facts** before arriving at an answer.



Question: Which object in the image is more similar to a tiger?
Fact: (Cat, RelatedTo, Tiger)
Answer: Cat

Learning Knowledge Base Retrieval



Inference

1. Retrieval of Relevant Facts

- Fact consists of (visual concept, relation, attribute), e.g., (Orange, IsA, Fruit)
- 100 relevant facts retrieved based on GloVe similarity of the fact with the question and visual concepts in image
- One relation out of 13 possible is obtained from the question by using an LSTM [1], proposed in [2]
- Top 100 facts further reduced by filtering according to the predicted relation, e.g., IsA
- Entity Embedding.** Each entity, (visual concept, attribute) in the fact is embedded using an LSTM [3]

2. Question and Visual Concept Embedding

- Question:** Embedding of dimension 100 learned using an LSTM [2]
- Visual Concepts:** Objects, scenes, and actions detected using pre-trained models

3. Node Embedding and Graph Construction

- The visual concept, question, and entity embeddings are concatenated to form an embedding of a node
- The nodes of the graph are connected based on the relations connecting the entities

4. Answer Prediction from the Graph

- A 2-layer graph convolution network (GCN) performs a joint assessment of the nodes in the graph
- Each hidden layer of the GCN is a non-linear function given by,

$$H^{(l)} = f(H^{(l-1)}, A) = \sigma(\tilde{D}^{-1/2} \tilde{A} \tilde{D}^{-1/2} H^{(l-1)} W^{(l-1)}) \quad \forall l \in \{1, \dots, L\}$$

- The output of the GCN is passed through an MLP which predicts the answer

Learning

1. Relation Prediction

- The LSTM [1] is trained using ground truth question-relation pairs and standard cross-entropy loss

2. Answer Prediction

- The answer predictor's parameters consist of the question and entity embedding, the layers of the GCN and MLP
- The LSTMs [2] and [3], the GCN [4], and the MLP [5] are trained end-to-end using the ground truth answer and binary cross-entropy loss

Quantitative and Qualitative Results

Method	Accuracy	
	@1	@3
FVQA [1]	56.91	64.65
FVQA Ensemble [1]	58.76	—
STTF [2]	62.20	75.60
Ours (1 layer GCN)	57.89	65.14
Ours (3 layer GCN)	60.78	68.65
Ours (2 layer GCN)	69.35	80.25
Human	77.99	-

Method	Synonyms (in FVQA)	Synonyms (Generated)	Homographs (in FVQA)
FVQA [1]	78	61	66.3
STTF [2]	91.6	89	79.4
Ours	95.38	91	81.16

Answer Prediction Accuracy on Question-Fact pairs with Synonyms and Homographs

Answer Prediction Results

Correctly Answered Questions



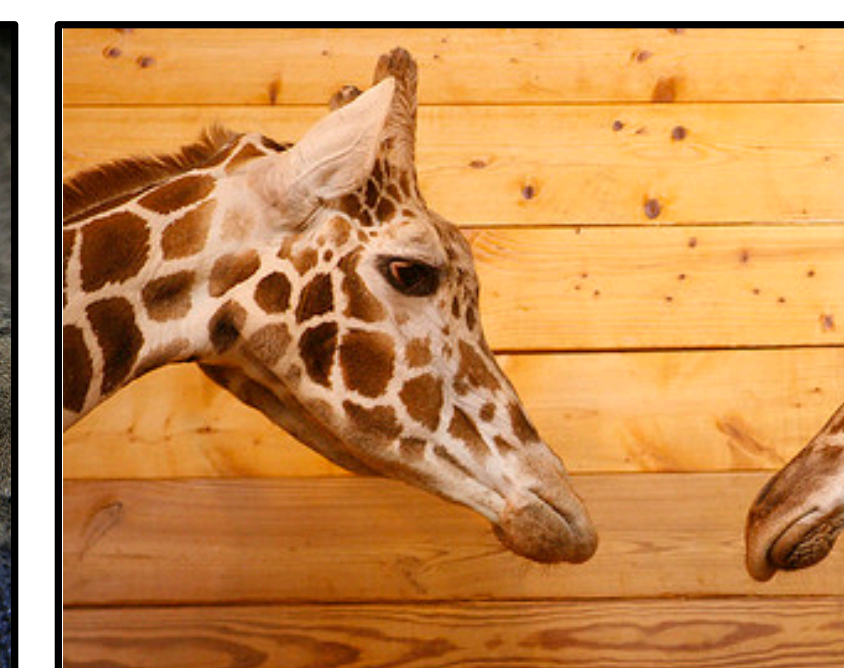
Question: Which vehicle shown here can float?
Pred. Relation: CapableOf
Pred. Visual Concept: Boat (object)
Supporting Fact: (Boat, CapableOf, Sailing)
Pred./GT Answer: Boat



Question: What is the place in this image used for?
Pred. Relation: UsedFor
Pred. Visual Concept: Kitchen (scene)
Supporting Fact: (Kitchen, UsedFor, Cooking)
Pred./GT Answer: Kitchen



Question: What does the animal in the image like to chase?
Pred. Relation: CapableOf
Pred. Visual Concept: Cat (object)
Supporting Fact: (Cat, CapableOf, Hunting mice)
Pred./GT Answer: Cat



Question: What is the plant-eating animal shown here?
Pred. Relation: Category
Pred. Visual Concept: Giraffe (object)
Supporting Fact: (Giraffe, Category, Herbivore)
Pred./GT Answer: Giraffe



Question: What is the area in the image used for?
Pred. Relation: UsedFor
Pred. Visual Concept: Field (Scene)
Supporting Fact: (Field, UsedFor, Grazing Animals)
Pred./GT Answer: Grazing Animals



Question: What in this image is made by baking?
Pred. Relation: Category
Pred. Visual Concept: Donut (object)
Supporting Fact: (Donut, Category, Cooking)
Pred./GT Answer: Donut



Question: What object in this image is spiky?
Pred. Relation: RelatedTo
Pred. Visual Concept: Pineapple (object)
Supporting Fact: (Pineapple, RelatedTo, Spiky)
Pred./GT Answer: Pineapple



Question: Which object in this image is venomous?
Pred. Relation: HasProperty
Pred. Visual Concept: Snake (object)
Supporting Fact: (Snake, HasProperty, Venomous)
Pred./GT Answer: Snake

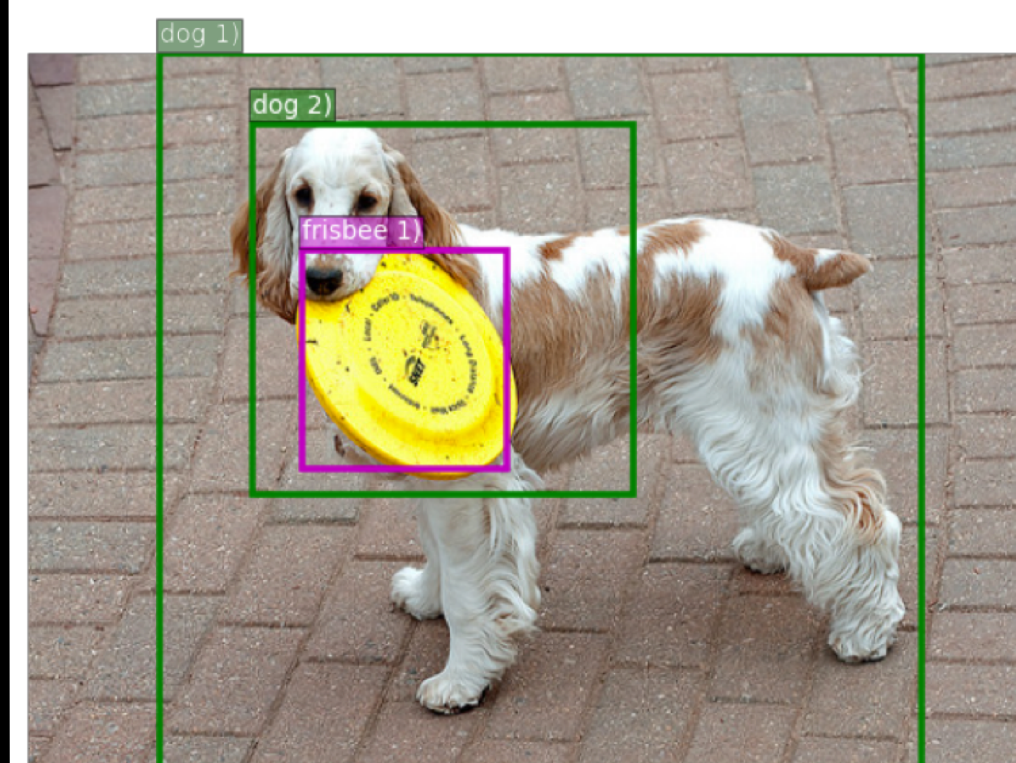


Question: Which action shown here is faster than walking?
Pred. Relation: Comparative (faster)
Pred. Visual Concept: Cycling (action)
Supporting Fact: (Cycling, Faster, Walking)
Pred./GT Answer: Cycling

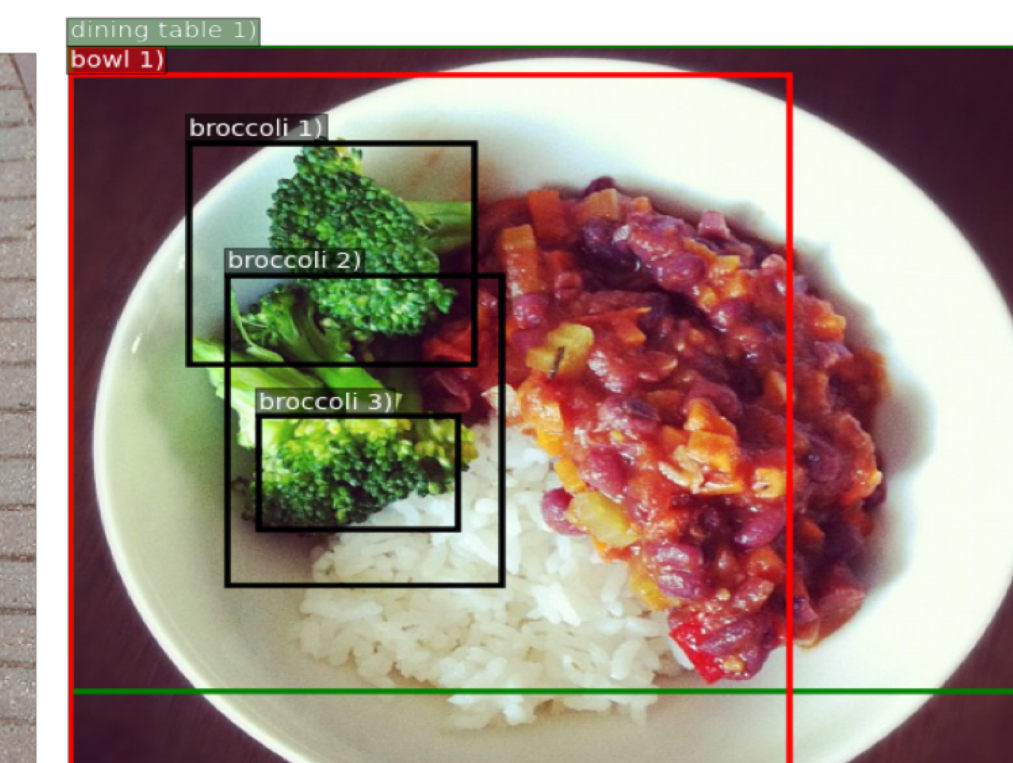


Question: What is on the ground in this image?
Pred. Relation: AtLocation
Pred. Visual Concept: Beach (Scene)
Supporting Fact: (Sand, AtLocation, Beach)
Pred./GT Answer: Sand

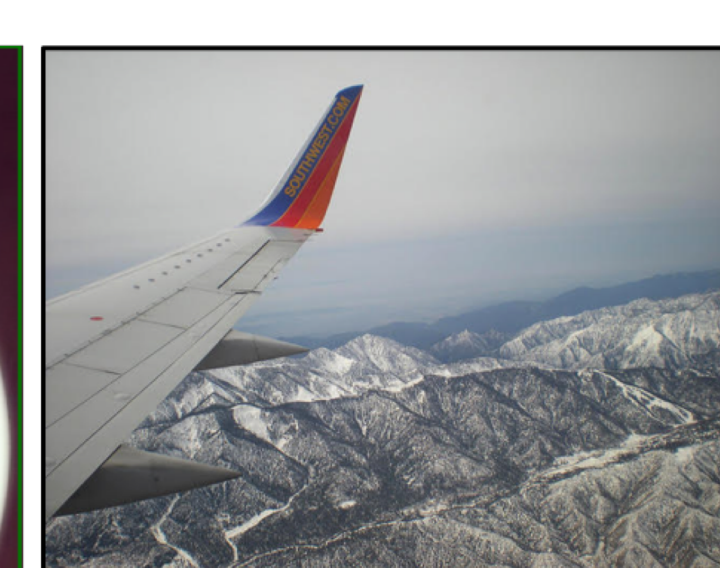
Visual Concepts Prediction



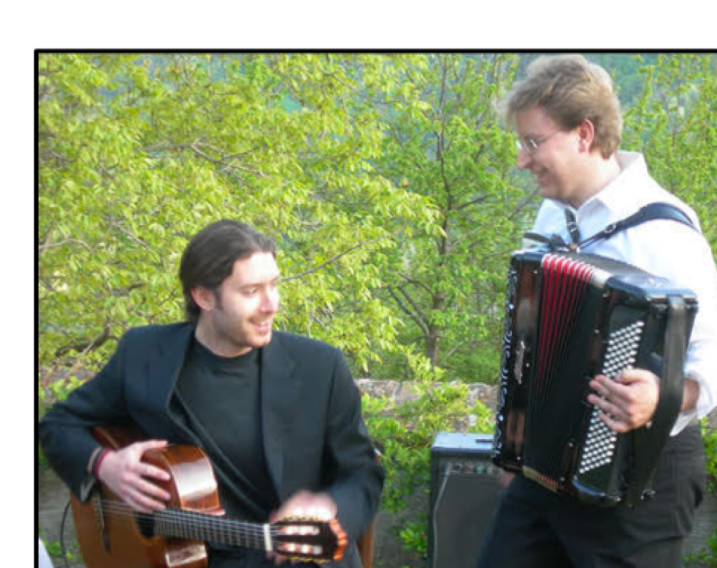
Question: What object in this image can fly?
Relevant Object: Frisbee
Predicted/GT Relation: CapableOf
Supporting Fact: (Frisbee, CapableOf, Flying)
Predicted/GT Answer: Frisbee



Question: What are the greens shown in this image?
Relevant Object: Broccoli
Predicted/GT Relation: IsA
Supporting Fact: (Broccoli, IsA, Green Vegetable)
Predicted/GT Answer: Broccoli



Question: What is the object that the picture is taken from used for?
Pred. Relation: UsedFor
GT Supporting Fact: (Airplane, UsedFor, Flying)
Pred. Answer: Printing pictures
GT Answer: Flying
Error: GT Fact not retrieved in Top-100.



Question: What object in this image is used to play polka music?
Pred. Relation: UsedFor
GT Relation: ReceivesAction
GT Supporting Fact: (Accordion, ReceivesAction, Polka Music)
Pred. Answer: Guitar
GT Answer: Accordion
Error: Incorrect annotation / Wrong relation predicted.



Question: What object in this image is used for entering data?
Pred. Relation: UsedFor
GT Supporting Fact: (Keyboard, UsedFor, Data entry)
Pred. Answer: Laptop
GT Answer: Keyboard
Error: GCN predicted the wrong node.

References

- [1] Wang P, Wu Q, Shen C, Dick A, van den Hengel A. Fvqa: Fact-based visual question answering. *IEEE TPAMI*, 2018.
- [2] Narasimhan M, Schwing AG. Straight to the Facts: Learning Knowledge Base Retrieval for Factual Visual Question Answering. In *ECCV*, 2018.