

Deep Learning for Vision & Language

Referring Expression Comprehension (Visual Grounding),

Visual Question Answering, Explainable Heatmaps

RICE UNIVERSITY

Today

- Referring Expressions
 - Referring Expressions vs Image Captions
 - Generating Referring Expressions
 - Referring Expression Comprehension

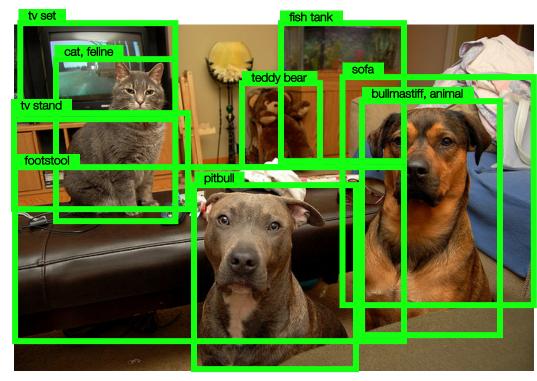
Computer Vision



Image tagging / Image classification

feline tv set teddy bear pitbull bullmastiff cat tv stand group of dogs fish tank room indoor man-made footstool furniture

Computer Vision



feline tv set teddy bear pitbull bullmastiff cat tv stand group of dogs fish tank room indoor man-made footstool furniture

Object Detection

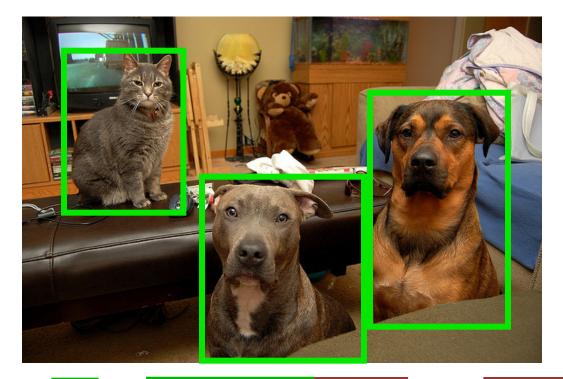
Computer Vision



Image Parsing / Image Segmentation

feline tv set teddy bear pitbull dog cat tv stand group of dogs fish tank room indoor man-made footstool furniture

How do we describe images?



Object Importance

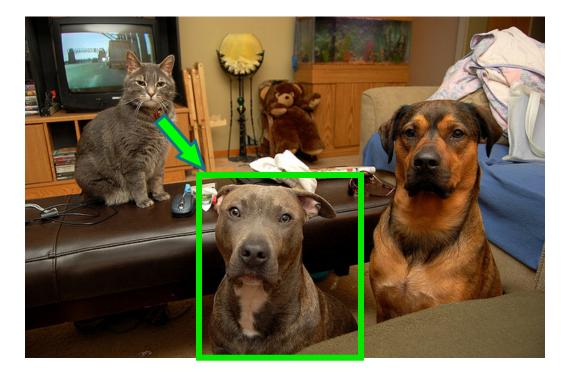
Attribute Importance

Action Importance

World knowledge

A cat and two big dogs staring at the camera

Referring to objects



The dog in the middle The gray dog in the middle

The gray dog

Work on Referring Expression



Size Corpus Mitchell et al 2011 [96 scenes]



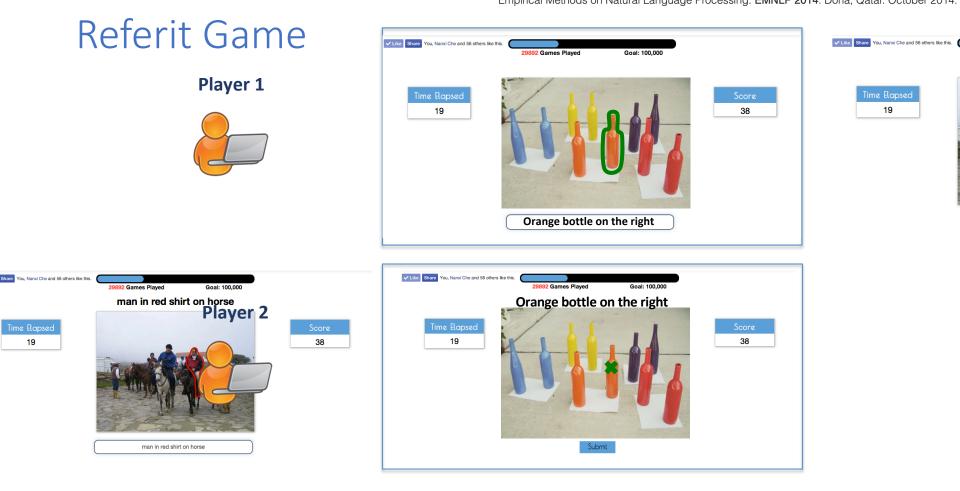
GenX Corpus FitzGerald et al 2013 [269 scenes]



Typicality Corpus Mitchell et al 2013 [35 scenes]



ReferItGame: Referring to Objects in Photographs of Natural Scenes Sahar Kazemzadeh, Vicente Ordonez, Mark Matten, Tamara L. Berg. Empirical Methods on Natural Language Processing. EMNLP 2014. Doha, Qatar. October 2014.



Referring Expressions for Natural Scenes

Diverse

Many real world objects

Complex

Many object instances

Big



IAPR TC-12 Segmented and Annotated Dataset. Escalante et. al. 2009

Referit Game Dataset



Blue shirt man

Blue guy

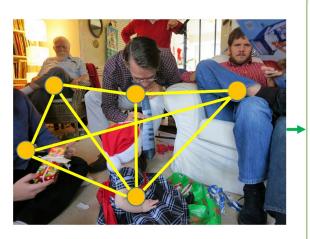
Second guy from left

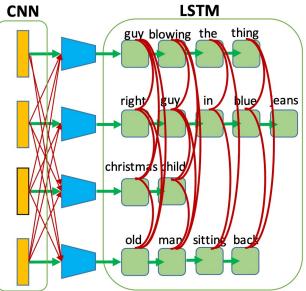
ReferItGame Dataset

130k Referring expressions for 90k Objects in 19k images

ReferItGame: Referring to Objects in Photographs of Natural Scenes Sahar Kazemzadeh, Vicente Ordonez, Mark Matten, Tamara L. Berg. Empirical Methods on Natural Language Processing. **EMNLP 2014**.

Deep Generation of Referring Expressions





Modeling Context in Referring Expressions

Licheng Yu, Patrick Poirson, Shan Yang, Alexander C. Berg, Tamara L. Berg

2016

Department of Computer Science, University of North Carolina at Chapel Hill {licheng,poirson,alexyang,aberg,tlberg}@cs.unc.edu

RefCOCO+ testA





Baseline: blue shirt Baseline: tennis player MMI: black shirt MMI: girl visdif: woman in white visdif: person in stripped shirt visdif+tie: arm with stripped shirt visdif+tie: tennis player

Baseline: man MMI: man visdif: man with glasses visdif+tie: man with glasses RefCOCO+ testB



Baseline: red jacket MMI: red jacket visdif: skier in white visdif+tie: man in white



Baseline: plant MMI: plant that is cut off visdif: tall plant visdif+tie: plant on screen side



Baseline: toilet MMI: toilet visdif: toilet with lid visdif+tie: toilet with lid



Baseline: donut at 3 MMI: glazed donut visdif: donut with hole



Baseline: car with red roof MMI: car visdif: car with headlights visdif+tie: donut with hole visdif+tie: car with headlights

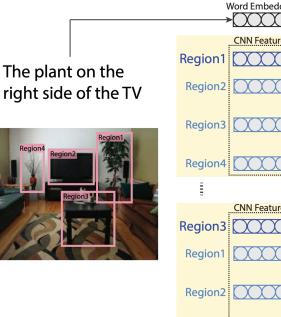
The plant on the right side of the TV

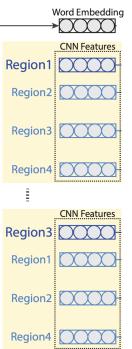


Modeling Context Between Objects for Referring Expression Understanding

Varun K. Nagaraja Vlad I. Morariu Larry S. Davis

University of Maryland, College Park, MD, USA. {varun,morariu,lsd}@umiacs.umd.edu





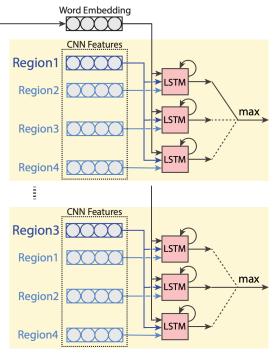
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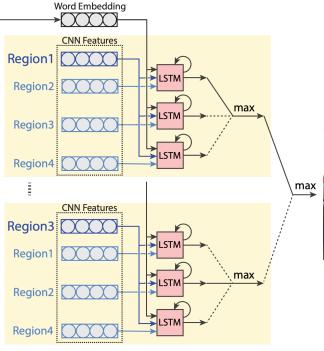
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2016

Other important work

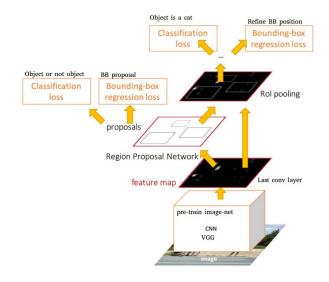
MattNet: Yu et al. https://arxiv.org/abs/1801.08186

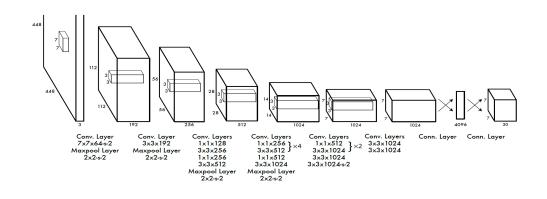
Mao et al. https://arxiv.org/abs/1511.02283

Rohrbach et al. https://arxiv.org/abs/1511.03745

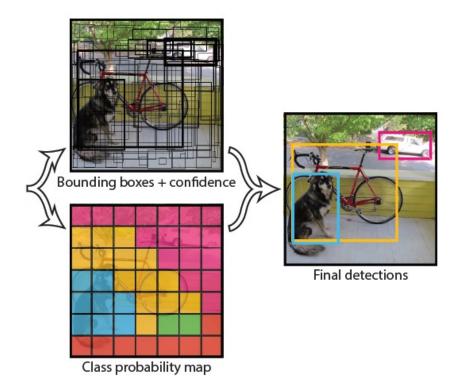
Detour: Recap on Object Detection

- Two-stage: Faster-RCNN, Mask-RCNN
- Single-stage: YOLO (You Only Look Once), SSD (Single Shot Detector)

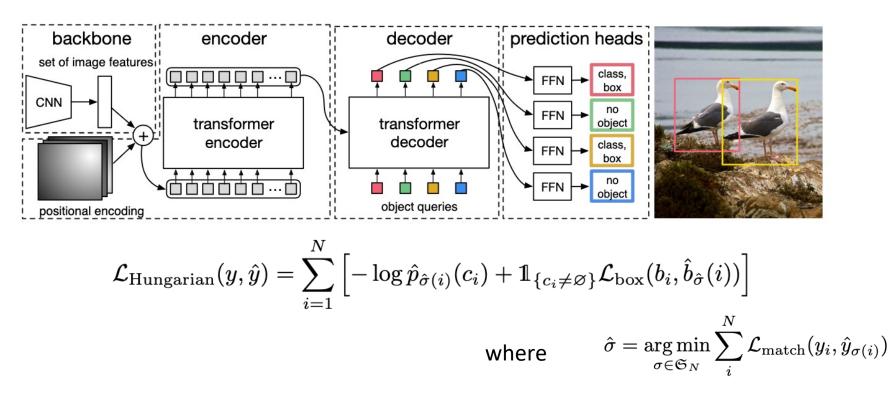




Post-processing: Non-Max Suppression



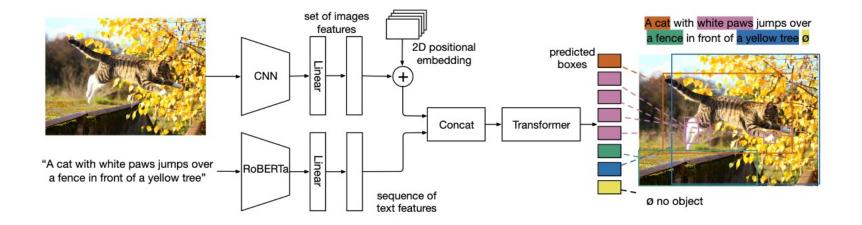
End-to-end Object Detection with Transfromers (DETR) (2020)



https://arxiv.org/abs/2005.12872

https://github.com/facebookresearch/detr

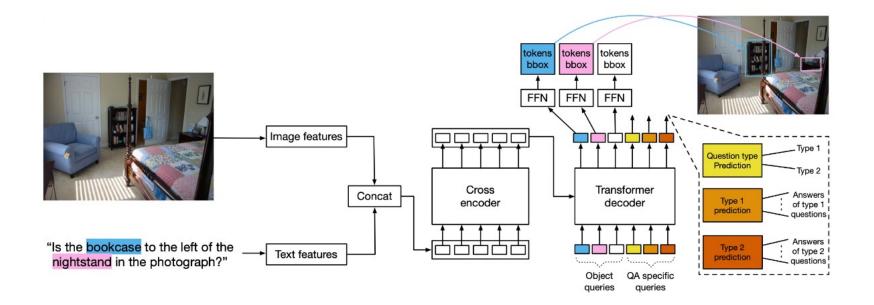
MDETR: Modulated Detection for Multimodal Understanding (2021)



https://arxiv.org/pdf/2104.12763.pdf

https://github.com/facebookresearch/detr

MDETR: For Question Answering



Visually Grounded Question Answering



How many slices of pizza are there? Is this a vegetarian pizza?

https://arxiv.org/pdf/1505.00468.pdf

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Visually Grounded Question Answering



How many slices of pizza are there? Is this a vegetarian pizza?

https://arxiv.org/pdf/1505.00468.pdf

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VQA: Visual Question Answering

Aishwarya Agrawal*, Jiasen Lu*, Stanislaw Antol*, Margaret Mitchell, C. Lawrence Zitnick, Dhruv Batra, Devi Parikh

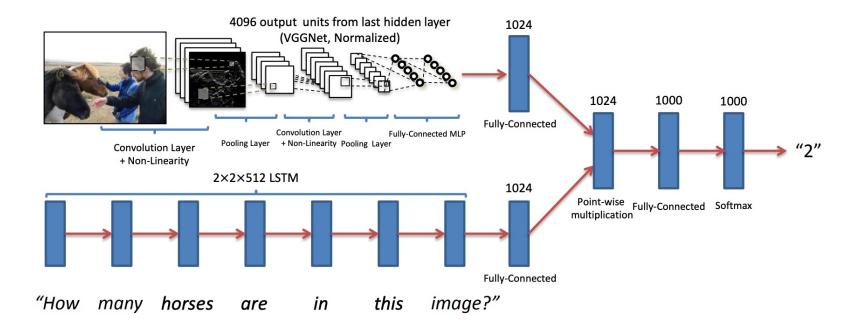


Is this person trying to hit a ball?	yes yes yes	yes yes yes
What is the person hitting the ball with?	frisbie racket round paddle	bat bat racket



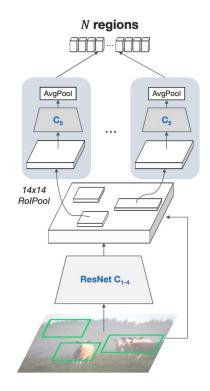
What is the guy	phone	reading
doing as he sits	taking picture	reading
on the bench?	taking picture with phone	smokes
What color are his shoes?	blue blue blue	black black brown

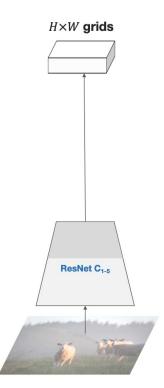
Visually Grounded Question Answering



https://arxiv.org/pdf/1505.00468.pdf

What Features to use as input visual features?

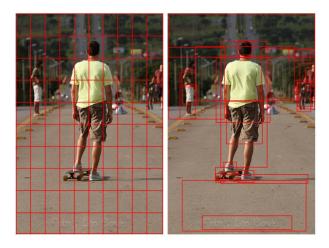






Bottom-Up and Top-Down Attention for Image Captioning and Visual Question Answering

Peter Anderson^{1*} Xiaodong He² Chris Buehler³ Damien Teney⁴ Mark Johnson⁵ Stephen Gould¹ Lei Zhang³ ¹Australian National University ²JD AI Research ³Microsoft Research ⁴University of Adelaide ⁵Macquarie University ¹firstname.lastname@anu.edu.au, ²xiaodong.he@jd.com, ³{chris.buehler,leizhang}@microsoft.com ⁴damien.teney@adelaide.edu.au, ⁵mark.johnson@mq.edu.au





Question: What room are they in? Answer: kitchen

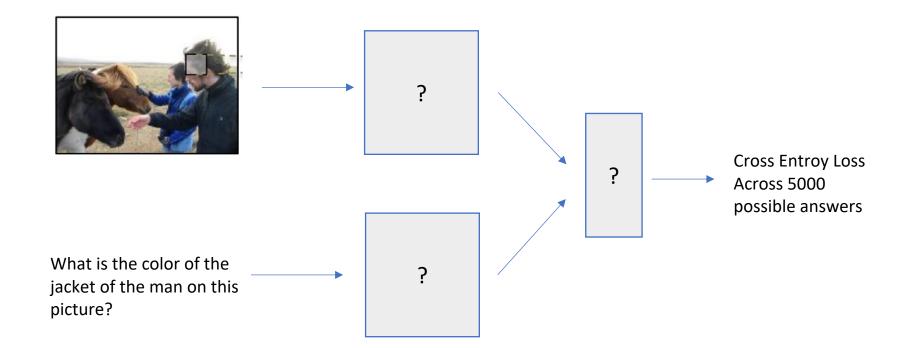
CVPR 2020

In Defense of Grid Features for Visual Question Answering

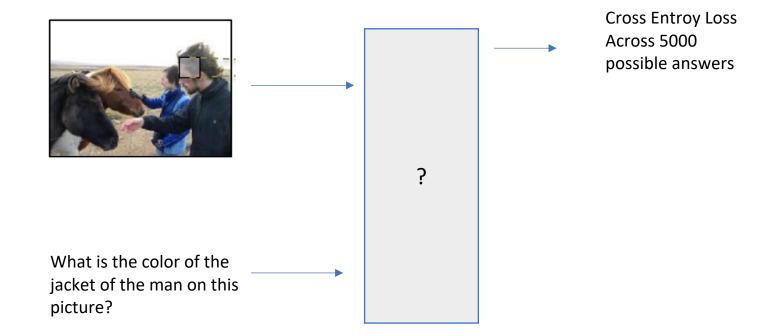
Huaizu Jiang^{1,2*}, Ishan Misra², Marcus Rohrbach², Erik Learned-Miller¹, and Xinlei Chen² ¹UMass Amherst, ²Facebook AI Research (FAIR)

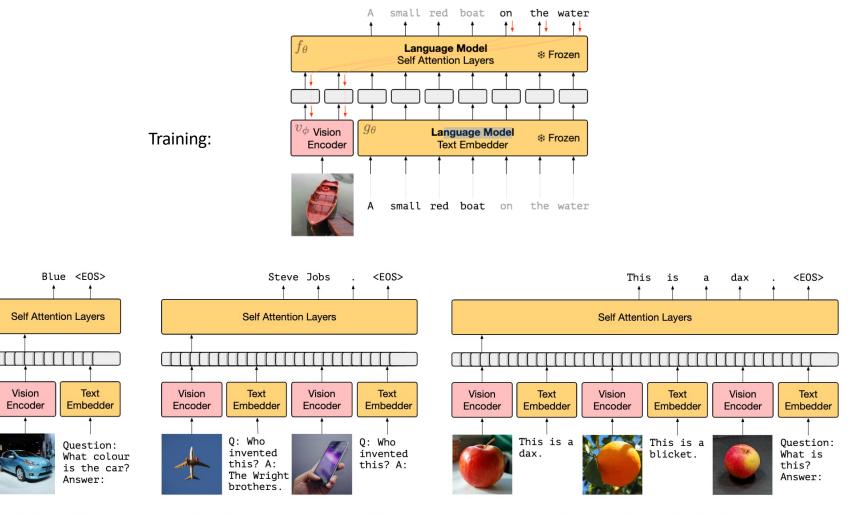
{hzjiang,elm}@cs.umass.edu, {imisra,mrf,xinleic}@fb.com

VQA Solution today?



VQA Solution today?





(a) 0-shot VQA

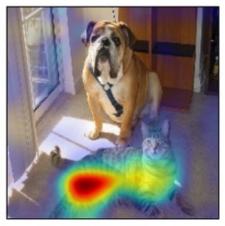
(b) 1-shot outside-knowledge VQA

(c) Few-shot image classification

Explainability: GradCAM



(a) Original Image



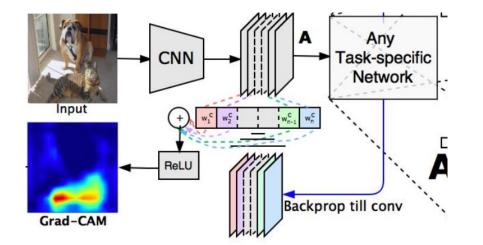
(c) Grad-CAM 'Cat'



(i) Grad-CAM 'Dog'

https://arxiv.org/abs/1610.02391

Explainability: GradCAM



 $\alpha_{k}^{c} = \underbrace{\frac{1}{Z}\sum_{i}\sum_{j}}_{j} \underbrace{\frac{\partial y^{c}}{\partial A_{ij}^{k}}}_{\text{gradients via backprop}}$ $L_{\text{Grad-CAM}}^{c} = ReLU\left(\sum_{k}\alpha_{k}^{c}A^{k}\right)$

https://arxiv.org/abs/1610.02391

linear combination

Questions?